
nighres Documentation

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1	Installing Nighres	3
2	Nighres usage examples	7
3	Brain	31
4	Cortex	41
5	Data	45
6	Filtering	47
7	Intensity	53
8	Input/Output	63
9	Laminar	67
10	Microscopy	73
11	Registration	77
12	Segmentation	87
13	Shape	91
14	Statistics	95
15	Surface	97
16	Data handling and formats	105
17	Saving outputs	107
18	Levelsets	109
19	Overview	111
20	Setting up	113

21 Wrapping an existing CBS Tools class	115
22 Adding a new Python function	117
23 Writing examples	119
24 Adapting the docs	121
25 Making a Pull Request	123
Index	125

Nighres is a Python package for processing of high-resolution neuroimaging data. It developed out of [CBS High-Res Brain Processing Tools](#) and aims to make those tools easier to install, use and extend. Nighres now includes new functions from the [IMCN imaging toolkit](#).

Please see *Troubleshooting* if you run into errors during installation.

1.1 Requirements

To build Nighres you need:

- Python 3.5 or higher
- Java JDK 1.7 or higher
- JCC 3.0 or higher

The following Python packages are automatically installed with Nighres * `numpy` * `nibabel` * `psutils`

For further dependencies of specific interfaces see *Optional dependencies*.

1.2 From PyPI

You can download the latest stable release of Nighres from [PyPI](#).

Because parts of the package have to be built locally it is currently not possible to use `pip install` directly from PyPI. Instead, please download and unpack the tarball to *Build Nighres*. (Or use the *Docker image*)

1.3 From Github

You can also get the latest version from Github

```
git clone https://github.com/nighres/nighres
```

Or download and unpack the zip file from Github under **Clone and download** -> **Download ZIP**

1.4 Build Nighres

1. Make sure you have Java JDK and JCC installed and set up. You will likely need to point the JCC_JDK variable to your Java JDK installation, e.g on a Debian/Ubuntu amd64 system:

```
sudo apt-get install openjdk-8-jdk
export JCC_JDK=/usr/lib/jvm/java-8-openjdk-amd64
python3 -m pip install jcc
```

2. Navigate to the Nighres directory you downloaded and unpacked, and run the build script:

```
./build.sh
```

3. Install the Python package:

```
python3 -m pip install .
```

1.5 Installation to a custom directory (e.g., servers and module-based systems)

This is generally only useful for administrators supporting server installs and/or where it is necessary to retain support for multiple versions of Nighres.

Complete 1. and 2. to build Nighres as described above.

3. Create an empty directory within your desired installation directory to satisfy Setuptools. This example will install to `/opt/quarantine/nighres/install/` and use Python3.7:

```
mkdir -p /opt/quarantine/nighres/install/lib/python3.7/site-packages/
```

4. Update your Python path environment variable and install to your custom directory:

```
export PYTHONPATH=/opt/quarantine/nighres/install/lib/python3.7/site-packages/:
↪$PYTHONPATH
python3 setup.py install --prefix /opt/quarantine/nighres/install/
```

5. Update PYTHONPATH for all users to point to Nighres:

```
PYTHONPATH=/opt/quarantine/nighres/install/lib/python3.7/site-packages/:
↪$PYTHONPATH
```

1.6 Testing the installation

You can often catch installation problems by simply import Nighres in Python. Make sure to navigate out of the directory from which you installed to make sure Nighres has actually been installed correctly and can be accessed from any location

```
python3 -c "import nighres"
```

If that works, you can try running one of the examples. You can find them inside the unpacked Nighres directory, in the subdirectory *examples*. Alternatively, you can also download the *examples* from the online documentation.

1.7 Docker

To quickly try out nighres in a preset, batteries-included environment, you can use the included Dockerfile, which includes Ubuntu 14 Trusty, openJDK-8, nighres, and Jupyter Notebook. The only thing you need to install is [Docker](#), a lightweight container platform that runs on Linux, Windows and Mac OS X.

To build the Docker image, do the following:

```
git clone https://github.com/nighres/nighres
cd nighres
docker build . -t nighres
```

To run the Docker container:

```
docker run --rm -p 8888:8888 nighres
```

Now go with your browser to <http://localhost:8888> to start a notebook. You should be able to import nighres by entering:

```
import nighres
```

into the first cell of your notebook.

Usually you also want to have access to some data when you run nighres. You can grant the Docker container access to a data folder on your host OS by using the `-v` tag when you start the container:

```
docker run --rm -v /home/me/my_data:/data -p 8888:8888 nighres
```

Now, in your notebook you will be able to access your data on the path `/data`

1.8 Optional dependencies

Working with surface mesh files

- [pandas](#)

Using the registration tools

- [nipy](#)
- [ANTs](#)

Plotting in the examples

- [Nilearn](#) and its dependencies, if Nilearn is not installed, plotting in the examples will be skipped and you can view the results in any other nifti viewer

Using the docker image

- [Docker](#)

Building the documentation

- [sphinx](#)
- [sphinx-gallery](#)
- [matplotlib](#)
- [sphinx-rtd-theme](#) (`pip install sphinx-rtd-theme`)

- pillow (pip install pillow)
- mock

1.9 Troubleshooting

If you experience errors not listed here, please help us by reporting them through neurostars.org using the tag **nighres**, or on [github](https://github.com). Or if you solve them yourself help others by contributing your solution here (see *Developers guide*)

1.9.1 Missing Java libraries

If you get errors regarding missing java libraries (such as `ljvm/libjvm` or `ljava/libjava`), although you install Java JDK, it means that JCC does not find the libraries. It can help to search for the “missing” library and make a symbolic link to it like this:

```
sudo find / -type f -name libjvm.so
>> /usr/lib/jvm/java-11-openjdk-amd64/lib/server/libjvm.so
sudo ln -s /usr/lib/jvm/java-11-openjdk-amd64/lib/server/libjvm.so /usr/lib/libjvm.so
```

1.9.2 Missing Python packages

If you get errors about Python packages not being installed, it might be that you are trying to run a function that requires *Optional dependencies*. If packages are reported missing that you think you have installed, make sure that they are installed under the same python installation as nighres. They should be listed when you run:

```
python3 -m pip list
```

If they aren't, install them using:

```
python3 -m pip install <package_name>
```

If there is still confusion, make sure nighres is installed in the same directory that your `python3 -m pip` command points to. These two commands should give the same base directory:

```
python3 -m pip
python3 -c 'import nighres; print(nighres.__file__)'
```

2.1 Vascular Reconstruction

This example shows how to perform vascular reconstruction based on MP2RAGE data by performing the following steps:

1. **Downloading three open MP2RAGE datasets using** `nighres.data.download_7T_TRT()`¹
2. **Remove the skull and create a brain mask using** `nighres.brain.mp2rage_skullstripping()`
3. Vasculature reconstruction using `nighres.filtering.multiscale_vessel_filter()`²

Important note: this example extracts arteries as bright vessels in a T1-weighted 7T scan, instead of the veins extracted from QSM in², because these data sets could not be made openly available. Processing of QSM images would however follow the same pipeline

2.1.1 Import and download

First we import `nighres` and the `os` module to set the output directory. Make sure to run this file in a directory you have write access to, or change the `out_dir` variable below.

```
import nighres
import os
import nibabel as nb

in_dir = os.path.join(os.getcwd(), 'nighres_examples/data_sets')
out_dir = os.path.join(os.getcwd(), 'nighres_examples/vascular_reconstruction')
```

We also try to import Nilearn plotting functions. If Nilearn is not installed, plotting will be skipped.

¹ Gorgolewski et al (2015). A high resolution 7-Tesla resting-state fMRI test-retest dataset with cognitive and physiological measures. DOI: 10.1038/sdata.2014.54

² Huck et al. (2019) High resolution atlas of the venous brain vasculature from 7 T quantitative susceptibility maps. DOI: 10.1007/s00429-019-01919-4

```
skip_plots = False
try:
    from nilearn import plotting
except ImportError:
    skip_plots = True
    print('Nilearn could not be imported, plotting will be skipped')
```

Now we download an example MP2RAGE dataset. It is the structural scan of the first subject, first session of the 7T Test-Retest dataset published by Gorgolewski et al (2015)¹.

```
dataset = nighres.data.download_7T_TRT(in_dir, subject_id='sub001_sess1')
```

2.1.2 Skull stripping

First we perform skull stripping. Only the second inversion image is required to calculate the brain mask. But if we input the T1map and T1w image as well, they will be masked for us. We also save the outputs in the `out_dir` specified above and use a subject ID as the base `file_name`.

```
skullstripping_results = nighres.brain.mp2rage_skullstripping(
    second_inversion=dataset['inv2'],
    t1_weighted=dataset['t1w'],
    t1_map=dataset['t1map'],
    save_data=True,
    file_name='sub001_sess1',
    output_dir=out_dir)
```

Tip: in Nighres functions that have several outputs return a dictionary storing the different outputs. You can find the keys in the docstring by typing `nighres.brain.mp2rage_skullstripping?` or list them with `skullstripping_results.keys()`

To check if the skull stripping worked well we plot the brain mask on top of the original image. You can also open the images stored in `out_dir` in your favourite interactive viewer and scroll through the volume.

Like Nilearn, we use Nibabel SpatialImage objects to pass data internally. Therefore, we can directly plot the outputs using Nilearn plotting functions .

```
if not skip_plots:
    plotting.plot_roi(skullstripping_results['brain_mask'], dataset['t1map'],
                     annotate=False, black_bg=False, draw_cross=False,
                     cmap='autumn')
```

2.1.3 Vessel reconstruction

Next, we use the vessel filter to estimate the vasculature from the QSM data

```
vessel_result = nighres.filtering.multiscale_vessel_filter(
    input_image=skullstripping_results['t1w_masked'],
    scales=2,
    save_data=True, file_name="sub001_sess1",
    output_dir=out_dir)
```

Now we look at the topology-constrained segmentation MGDM created

```

if not skip_plots:
    plotting.plot_img(vessel_result['pv'],
                     vmin=0, vmax=1, cmap='cubehelix', colorbar=True,
                     annotate=False, draw_cross=False)
    plotting.plot_img(vessel_result['diameter'],
                     vmin=0, vmax=4, cmap='cubehelix', colorbar=True,
                     annotate=False, draw_cross=False)

```

If the example is not run in a jupyter notebook, render the plots:

```

if not skip_plots:
    plotting.show()

```

2.1.4 References

Total running time of the script: (0 minutes 0.000 seconds)

2.2 Tissue classification from MP2RAGE data

This example shows how to obtain a tissue classification from MP2RAGE data by performing the following steps:

1. Downloading open MP2RAGE dataset using `nighres.data.download_7T_TRT()`
2. Remove the skull and create a brain mask using `nighres.brain.mp2rage_skullstripping()`
3. Atlas-guided tissue classification using MGDM `nighres.brain.mgdm_segmentation()`¹

2.2.1 Import and download

First we import `nighres` and the `os` module to set the output directory. Make sure to run this file in a directory you have write access to, or change the `out_dir` variable below.

```

import nighres
import os

in_dir = os.path.join(os.getcwd(), 'nighres_examples/data_sets')
out_dir = os.path.join(os.getcwd(), 'nighres_examples/tissue_classification')

```

We also try to import Nilearn plotting functions. If Nilearn is not installed, plotting will be skipped.

```

skip_plots = False
try:
    from nilearn import plotting
except ImportError:
    skip_plots = True
    print('Nilearn could not be imported, plotting will be skipped')

```

Now we download an example MP2RAGE dataset. It is the structural scan of the first subject, first session of the 7T Test-Retest dataset published by Gorgolewski et al (2015)².

¹ Bogovic, Prince and Bazin (2013). A multiple object geometric deformable model for image segmentation. DOI: 10.1016/j.cviu.2012.10.006.A

² Gorgolewski et al (2015). A high resolution 7-Tesla resting-state fMRI test-retest dataset with cognitive and physiological measures. DOI: 10.1038/sdata.2014.54

```
dataset = nighres.data.download_7T_TRT(in_dir, subject_id='sub001_sess1')
```

2.2.2 Skull stripping

First we perform skull stripping. Only the second inversion image is required to calculate the brain mask. But if we input the T1map and T1w image as well, they will be masked for us. We also save the outputs in the `out_dir` specified above and use a subject ID as the base `file_name`.

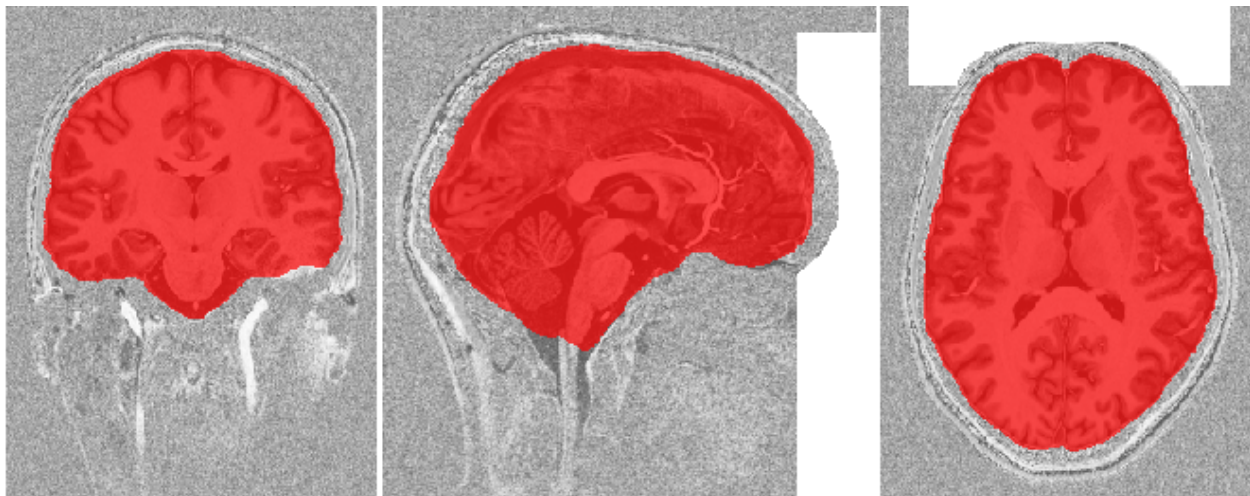
```
skullstripping_results = nighres.brain.mp2rage_skullstripping(
    second_inversion=dataset['inv2'],
    t1_weighted=dataset['t1w'],
    t1_map=dataset['t1map'],
    save_data=True,
    file_name='sub001_sess1',
    output_dir=out_dir)
```

Tip: in Nighres functions that have several outputs return a dictionary storing the different outputs. You can find the keys in the docstring by typing `nighres.brain.mp2rage_skullstripping?` or list them with `skullstripping_results.keys()`

To check if the skull stripping worked well we plot the brain mask on top of the original image. You can also open the images stored in `out_dir` in your favourite interactive viewer and scroll through the volume.

Like Nilearn, we use Nibabel SpatialImage objects to pass data internally. Therefore, we can directly plot the outputs using Nilearn plotting functions .

```
if not skip_plots:
    plotting.plot_roi(skullstripping_results['brain_mask'], dataset['t1w'],
                     annotate=False, black_bg=False, draw_cross=False,
                     cmap='autumn')
```



2.2.3 MGDM classification

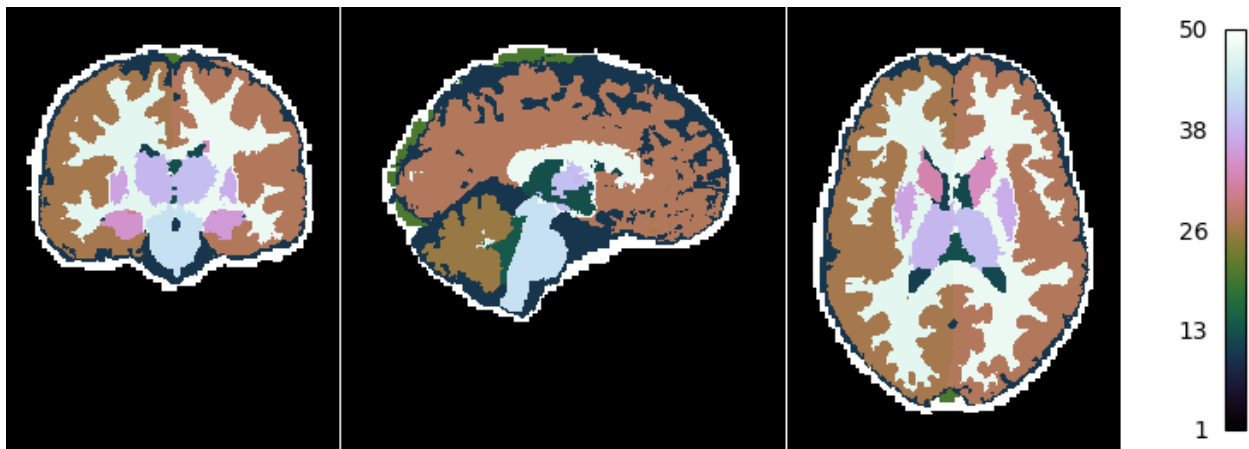
Next, we use the masked data as input for tissue classification with the MGDM algorithm. MGDM works with a single contrast, but can be improved with additional contrasts. In this case we use the T1-weighted image as well as

the quantitative T1map.

```
mgdm_results = nighres.brain.mgdm_segmentation(
    contrast_image1=skullstripping_results['t1w_masked'],
    contrast_type1="Mp2rage7T",
    contrast_image2=skullstripping_results['t1map_masked'],
    contrast_type2="T1map7T",
    save_data=True, file_name="sub001_sess1",
    output_dir=out_dir)
```

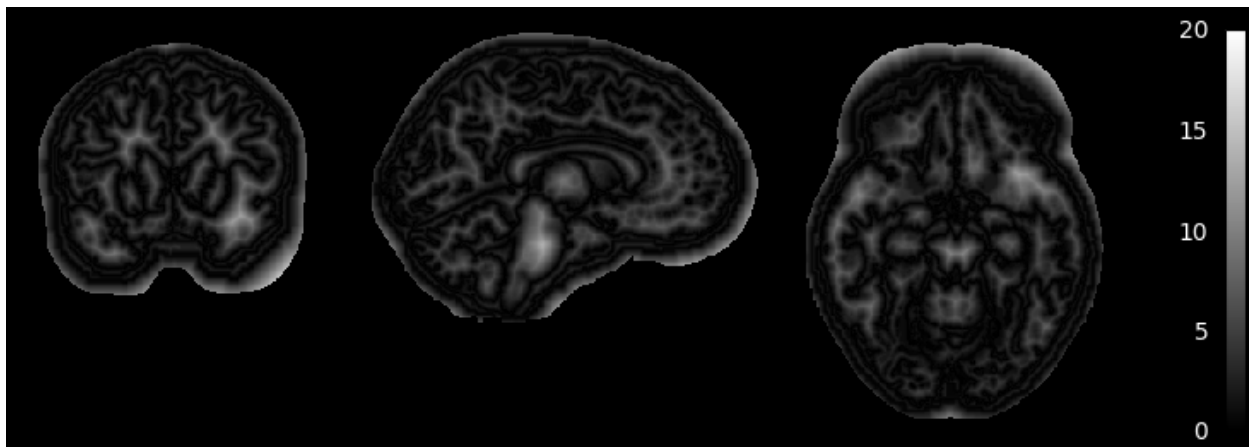
Now we look at the topology-constrained segmentation MGDM created

```
if not skip_plots:
    plotting.plot_img(mgdm_results['segmentation'],
                     vmin=1, vmax=50, cmap='cubehelix', colorbar=True,
                     annotate=False, draw_cross=False)
```



MGDM also creates an image which represents for each voxel the distance to its nearest border. It is useful to assess where partial volume effects may occur

```
if not skip_plots:
    plotting.plot_anat(mgdm_results['distance'], vmin=0, vmax=20,
                      annotate=False, draw_cross=False, colorbar=True)
```



If the example is not run in a jupyter notebook, render the plots:

```
if not skip_plots:
    plotting.show()
```

2.2.4 References

Total running time of the script: (0 minutes 0.000 seconds)

2.3 Subcortex Parcellation

This example shows how to perform multi-contrast subcortical parcellation with the MASSP algorithm on MP2RAGEME data by performing the following steps:

1. **Downloading an open MP2RAGE datasets using** `nighres.data.download_MP2RAGEME_sample()`¹
2. **Downloading the open AHEAD template using** `nighres.data.download_AHEAD_template()`²
3. **Register the data to the AHEAD brain template** `nighres.registration.embedded_antsreg()`³
4. Subcortex parcellation with MASSP `nighres.parcellation.massp()`⁴

Note: MASSP labels are listed inside the corresponding module and can be accessed with `nighres.parcellation.massp_17structures_label()`

2.3.1 Import and download

First we import `nighres` and the `os` module to set the output directory. Make sure to run this file in a directory you have write access to, or change the `out_dir` variable below.

```
import nighres
import os
import nibabel as nb

in_dir = os.path.join(os.getcwd(), 'nighres_examples/data_sets')
out_dir = os.path.join(os.getcwd(), 'nighres_examples/massp_parcellation')
```

We also try to import Nilearn plotting functions. If Nilearn is not installed, plotting will be skipped.

```
skip_plots = False
try:
    from nilearn import plotting
except ImportError:
    skip_plots = True
    print('Nilearn could not be imported, plotting will be skipped')
```

Now we download an example MP2RAGEME dataset, including a quantitative R1 map, a quantitative R2* map, and a QSM, all skull-stripped.

¹ Caan et al. (2018) MP2RAGEME: T1, T2*, and QSM mapping in one sequence at 7 tesla. DOI: 10.1002/hbm.24490

² Alkemada et al (under review). The Amsterdam Ultra-high field adult lifespan database (AHEAD): A freely available multimodal 7 Tesla submillimeter magnetic resonance imaging database.

³ Avants et al (2008). Symmetric diffeomorphic image registration with cross-correlation: evaluating automated labeling of elderly and neurodegenerative brain. DOI: 10.1016/j.media.2007.06.004

⁴ Bazin et al. (in prep) Multi-contrast Anatomical Subcortical Structures Parcellation


```
dataset = nighres.data.download_MP2RAGEME_sample(data_dir=in_dir)
```

Now we download the AHEAD template for coregistration to the atlas space.

```
template = nighres.data.download_AHEAD_template()
```

Co-registration First we co-register the subject to the AHEAD template, and save the transformation mappings in the `out_dir` specified above, using a subject ID as the base `file_name`.

```
ants = nighres.registration.embedded_antsreg_multi(
    source_images=[dataset['qr1'], dataset['qr2s'], dataset['qsm']],
    target_images=[template['qr1'], template['qr2s'], template['qsm']
↪']],
    run_rigid=True, run_affine=True, run_syn=True,
    rigid_iterations=10000,
    affine_iterations=2000,
    coarse_iterations=180,
    medium_iterations=60, fine_iterations=30,
    cost_function='MutualInformation',
    interpolation='NearestNeighbor',
    regularization='High',
    ignore_affine=True,
    save_data=True, file_name="sample-subject",
    output_dir=out_dir)
```

Tip: in Nighres functions that have several outputs return a dictionary storing the different outputs. You can find the keys in the docstring by typing `nighres.brain.mp2rage_skullstripping?` or list them with `skullstripping_results.keys()`

To check if the skull stripping worked well we plot the brain mask on top of the original image. You can also open the images stored in `out_dir` in your favourite interactive viewer and scroll through the volume.

Like Nilearn, we use Nibabel SpatialImage objects to pass data internally. Therefore, we can directly plot the outputs using Nilearn plotting functions .

```
if not skip_plots:
    plotting.plot_anat(ants['transformed_source'], cut_coords=[0.0, 0.0, 0.0],
                       annotate=False, draw_cross=False)
    plotting.plot_anat(template['qr1'], cut_coords=[0.0, 0.0, 0.0],
                       annotate=False, draw_cross=False)
```

2.3.2 MASSP Parcellation

Finally, we use the MASSP algorithm to parcellate the subcortex

```
massp = nighres.parcellation.massp(target_images=[dataset['qr1'], dataset['qr2s'],
↪ dataset['qsm']],
    map_to_target=ants['inverse'],
    max_iterations=120, max_difference=0.1,
    save_data=True, file_name="sample-subject",
    output_dir=out_dir, overwrite=False)
```

Now we look at the topology-constrained segmentation MGDM created

```

if not skip_plots:
    plotting.plot_roi(massp['max_label'], dataset['qrl'],
                    annotate=False, black_bg=False, draw_cross=False,
                    cmap='cubehelix')
    plotting.plot_img(massp['max_proba'],
                    vmin=0, vmax=1, cmap='gray', colorbar=True,
                    annotate=False, draw_cross=False)

```

If the example is not run in a jupyter notebook, render the plots:

```

if not skip_plots:
    plotting.show()

```

2.3.3 References

Total running time of the script: (0 minutes 0.000 seconds)

2.4 Brain co-registration from MP2RAGE data

This example shows how to co-register MP2RAGE data by performing the following steps:

1. **Downloading two open MP2RAGE datasets using** `nighres.data.download_7T_TRT()`
2. **Remove the skull and create a brain mask using** `nighres.brain.mp2rage_skullstripping()`
3. **Co-register non-linearly the brains using** `nighres.registration.embedded_antsreg()`¹
4. **Deform additional contrasts using** `nighres.registration.apply_deformation()`

2.4.1 Import and download

First we import `nighres` and the `os` module to set the output directory. Make sure to run this file in a directory you have write access to, or change the `out_dir` variable below.

```

import nighres
import os

in_dir = os.path.join(os.getcwd(), 'nighres_examples/data_sets')
out_dir = os.path.join(os.getcwd(), 'nighres_examples/brain_registration')

```

We also try to import Nilearn plotting functions. If Nilearn is not installed, plotting will be skipped.

```

skip_plots = False
try:
    from nilearn import plotting
except ImportError:
    skip_plots = True
    print('Nilearn could not be imported, plotting will be skipped')

```

¹ Avants et al (2008). Symmetric diffeomorphic image registration with cross-correlation: evaluating automated labeling of elderly and neurodegenerative brain. DOI: 10.1016/j.media.2007.06.004

Now we download an example MP2RAGE dataset. It is the structural scan of the first subject, first session of the 7T Test-Retest dataset published by Gorgolewski et al (2015)².

```
dataset1 = nighres.data.download_7T_TRT(in_dir, subject_id='sub001_sess1')
dataset2 = nighres.data.download_7T_TRT(in_dir, subject_id='sub002_sess1')
```

2.4.2 Skull stripping

First we perform skull stripping. Only the second inversion image is required to calculate the brain mask. But if we input the T1map and T1w image as well, they will be masked for us. We also save the outputs in the `out_dir` specified above and use a subject ID as the base `file_name`.

```
skullstripping_results1 = nighres.brain.mp2rage_skullstripping(
    second_inversion=dataset1['inv2'],
    t1_weighted=dataset1['t1w'],
    t1_map=dataset1['t1map'],
    save_data=True,
    file_name='sub001_sess1',
    output_dir=out_dir, overwrite=False)

skullstripping_results2 = nighres.brain.mp2rage_skullstripping(
    second_inversion=dataset2['inv2'],
    t1_weighted=dataset2['t1w'],
    t1_map=dataset2['t1map'],
    save_data=True,
    file_name='sub002_sess1',
    output_dir=out_dir, overwrite=False)
```

Tip: in Nighres functions that have several outputs return a dictionary storing the different outputs. You can find the keys in the docstring by typing `nighres.brain.mp2rage_skullstripping?` or list them with `skullstripping_results.keys()`

To check if the skull stripping worked well we plot the brain mask on top of the original image. You can also open the images stored in `out_dir` in your favourite interactive viewer and scroll through the volume.

Like Nilearn, we use Nibabel SpatialImage objects to pass data internally. Therefore, we can directly plot the outputs using Nilearn plotting functions .

```
if not skip_plots:
    plotting.plot_roi(skullstripping_results1['brain_mask'], dataset1['t1map'],
                     annotate=False, black_bg=False, draw_cross=False,
                     cmap='autumn')
    plotting.plot_roi(skullstripping_results2['brain_mask'], dataset2['t1w'],
                     annotate=False, black_bg=False, draw_cross=False,
                     cmap='autumn')
```

2.4.3 SyN co-registration

Next, we use the masked data as input for co-registration. The T1 maps are used here as they are supposed to be more similar

² Gorgolewski et al (2015). A high resolution 7-Tesla resting-state fMRI test-retest dataset with cognitive and physiological measures. DOI: 10.1038/sdata.2014.54

```
syn_results = nighres.registration.embedded_antsreg(  
    source_image=skullstripping_results1['t1map_masked'],  
    target_image=skullstripping_results2['t1map_masked'],  
    run_rigid=True, run_syn=True,  
    rigid_iterations=1000, coarse_iterations=40,  
    medium_iterations=0, fine_iterations=0,  
    cost_function='MutualInformation',  
    interpolation='NearestNeighbor',  
    save_data=True, file_name="sub001_sess1",  
    output_dir=out_dir, overwrite=False)
```

Now we look at the coregistered image that SyN created

```
if not skip_plots:  
    plotting.plot_img(syn_results['transformed_source'],  
                     annotate=False, draw_cross=False)
```

2.4.4 Apply deformations to images

Finally, we use the computed deformation to transform other associated images

```
deformed = nighres.registration.apply_coordinate_mappings(  
    image=dataset1['t1map'],  
    mapping1=syn_results['mapping'],  
    save_data=True, file_name="sub001_sess1_t1map",  
    output_dir=out_dir, overwrite=False)  
  
inverse = nighres.registration.apply_coordinate_mappings(  
    image=dataset2['t1w'],  
    mapping1=syn_results['inverse'],  
    save_data=True, file_name="sub002_sess1_t1w",  
    output_dir=out_dir, overwrite=False)
```

Now we look at the coregistered images from applying the deformation

```
if not skip_plots:  
    plotting.plot_img(deformed['result'],  
                     annotate=False, draw_cross=False)  
    plotting.plot_img(inverse['result'],  
                     annotate=False, draw_cross=False)
```

If the example is not run in a jupyter notebook, render the plots:

```
if not skip_plots:  
    plotting.show()
```

2.4.5 References

Total running time of the script: (0 minutes 0.000 seconds)

2.5 Cortical depth estimation from MGDM segmentation

This example shows how to obtain a cortical laminar depth representation from an MGDM segmentation result with the following steps:

1. Get a segmentation result from the *tissue_classification* example
2. Extract the cortex of the left hemisphere with `nighres.brain.extract_brain_region()`
3. Cortical reconstruction with CRUISE `nighres.cortex.cruise_cortex_extraction()`¹
4. Anatomical depth estimation through `nighres.laminar.volumetric_layering()`²

Important note: this example assumes you have run the tissue classification example first (example_tissue_classification.py)

2.5.1 Import and point to previous example

First we import `nighres` and the `os` module to set the output directory. Make sure to run this file in a directory you have write access to, or change the `out_dir` variable below.

```
import nighres
import os

in_dir = os.path.join(os.getcwd(), 'nighres_examples/tissue_classification')
out_dir = os.path.join(os.getcwd(),
                      'nighres_examples/cortical_depth_estimation')
```

We also try to import Nilearn plotting functions. If Nilearn is not installed, plotting will be skipped.

```
skip_plots = False
try:
    from nilearn import plotting
except ImportError:
    skip_plots = True
    print('Nilearn could not be imported, plotting will be skipped')
```

Now we pull the MGDM results from previous example

```
segmentation = os.path.join(in_dir, 'sub001_sess1_mgdm-seg.nii.gz')
boundary_dist = os.path.join(in_dir, 'sub001_sess1_mgdm-dist.nii.gz')
max_labels = os.path.join(in_dir, 'sub001_sess1_mgdm-lbls.nii.gz')
max_probab = os.path.join(in_dir, 'sub001_sess1_mgdm-mems.nii.gz')

if not (os.path.isfile(segmentation) and os.path.isfile(boundary_dist)
        and os.path.isfile(max_labels) and os.path.isfile(max_probab)) :
    print('This example builds upon the example_tissue_segmentation.py one')
    print('Please run it first')
    exit()
```

2.5.2 Region Extraction

Here we pull from the MGDM output the needed regions for cortical reconstruction: the GM cortex ('region'), the underlying WM (with filled subcortex and ventricles, 'inside') and the surrounding CSF (with masked regions, 'back-

¹ Han et al (2004) CRUISE: Cortical Reconstruction Using Implicit Surface Evolution, NeuroImage, vol. 23, pp. 997–1012.

² Waehnert et al (2014) Anatomically motivated modeling of cortical laminae. DOI: 10.1016/j.neuroimage.2013.03.078

ground')

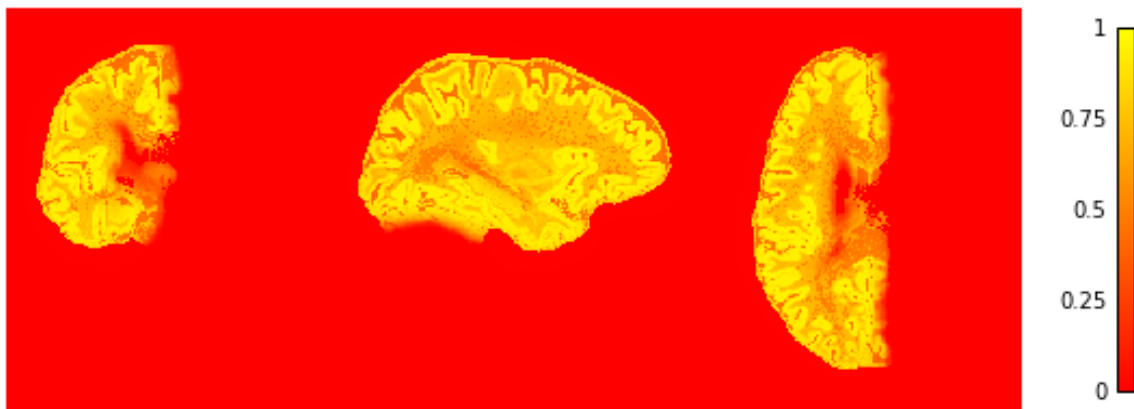
```
cortex = nighres.brain.extract_brain_region(segmentation=segmentation,
                                           levelset_boundary=boundary_dist,
                                           maximum_membership=max_probab,
                                           maximum_label=max_labels,
                                           extracted_region='left_cerebrum',
                                           save_data=True,
                                           file_name='sub001_sess1_left_cerebrum',
                                           output_dir=out_dir)
```

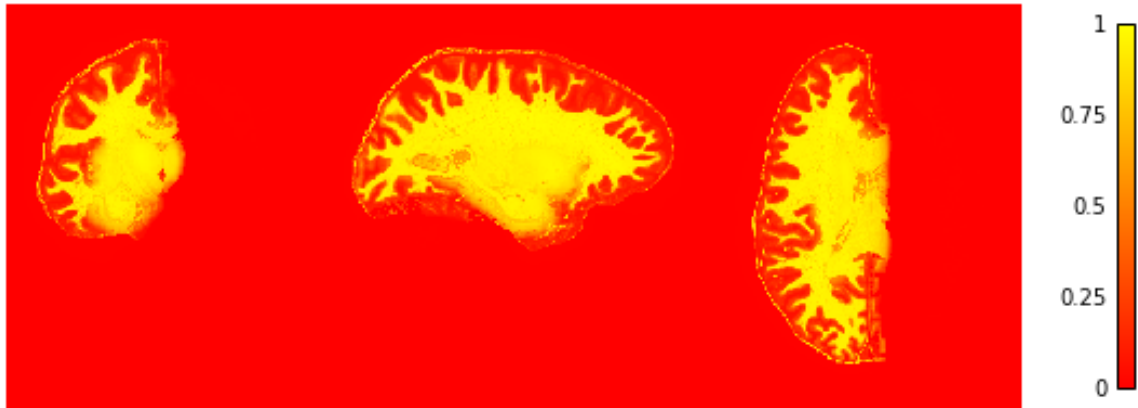
Tip: in Nighres functions that have several outputs return a dictionary storing the different outputs. You can find the keys in the docstring by typing `nighres.brain.mp2rage_extract_brain_region?` or list them with `cortex.keys()`

To check if the extraction worked well we plot the GM and WM probabilities. You can also open the images stored in `out_dir` in your favourite interactive viewer and scroll through the volume.

Like Nilearn, we use Nibabel SpatialImage objects to pass data internally. Therefore, we can directly plot the outputs using Nilearn plotting functions .

```
if not skip_plots:
    plotting.plot_img(cortex['region_proba'],
                     vmin=0, vmax=1, cmap='autumn', colorbar=True,
                     annotate=False, draw_cross=False)
    plotting.plot_img(cortex['inside_proba'],
                     vmin=0, vmax=1, cmap='autumn', colorbar=True,
                     annotate=False, draw_cross=False)
```





2.5.3 CRUISE cortical reconstruction

Next, we use the extracted data as input for cortex reconstruction with the CRUISE algorithm. CRUISE works with the membership functions as a guide and the WM inside mask as a (topologically spherical) starting point to grow a refined GM/WM boundary and CSF/GM boundary

```

cruise = nighres.cortex.cruise_cortex_extraction(
    init_image=cortex['inside_mask'],
    wm_image=cortex['inside_proba'],
    gm_image=cortex['region_proba'],
    csf_image=cortex['background_proba'],
    normalize_probabilities=True,
    save_data=True,
    file_name="sub001_sess1_left_cerebrum",
    output_dir=out_dir)

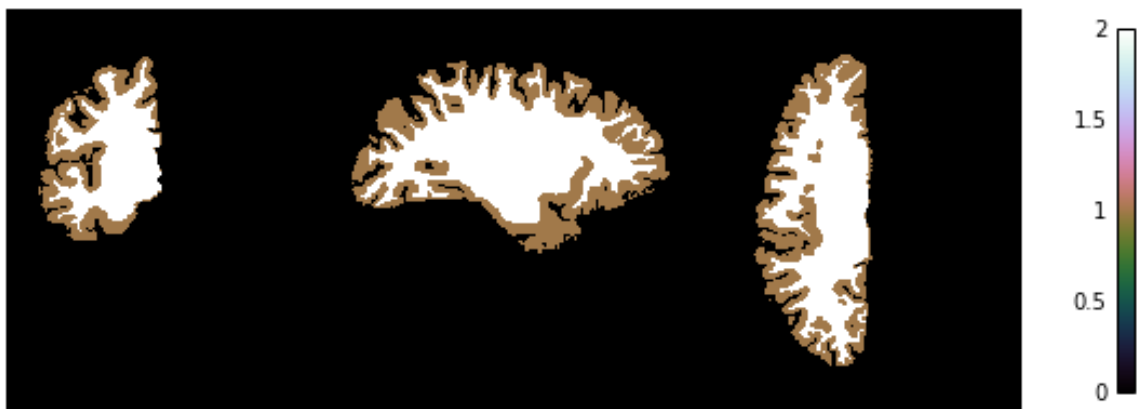
```

Now we look at the topology-constrained segmentation CRUISE created

```

if not skip_plots:
    plotting.plot_img(cruise['cortex'],
                     vmin=0, vmax=2, cmap='cubehelix', colorbar=True,
                     annotate=False, draw_cross=False)

```



2.5.4 Cortical surface inflation

For display purposes, we create a surface mesh from the average cortical CRUISE surface, which we then inflate

```
cortical_surface = nighres.surface.levelset_to_mesh(
    levelset_image=cruise['avg'],
    save_data=True,
    file_name="sub001_sess1_left_cerebrum.vtk",
    output_dir=out_dir)

inflated_surface = nighres.surface.surface_inflation(
    surface_mesh=cortical_surface['result'],
    save_data=True,
    file_name="sub001_sess1_left_cerebrum.vtk",
    output_dir=out_dir)
```

2.5.5 Volumetric layering

Finally, we use the GM/WM boundary (GWB) and CSF/GM boundary (CGB) from CRUISE to compute cortical depth with a volume-preserving technique

```
depth = nighres.laminar.volumetric_layering(
    inner_levelset=cruise['gwb'],
    outer_levelset=cruise['cgb'],
    n_layers=4,
    save_data=True,
    file_name="sub001_sess1_left_cerebrum",
    output_dir=out_dir)
```

Now we look at the laminar depth estimates

```
if not skip_plots:
    plotting.plot_img(depth['depth'],
        vmin=0, vmax=1, cmap='autumn', colorbar=True,
        annotate=False, draw_cross=False)
```



If the example is not run in a jupyter notebook, render the plots:

```
if not skip_plots:
    plotting.show()
```


2.5.6 References

Total running time of the script: (0 minutes 0.000 seconds)

2.6 DOTS white matter segmentation

This example shows how to perform white matter segmentation using Diffusion Oriented Tract Segmentation (DOTS) algorithm¹:

1. **Downloading DOTS atlas prior information** using `nighres.data.download_DOTS_atlas()`
2. **Downloading an example DTI data set** using `nighres.data.download_DTI_2mm()`
3. **Segmenting white matter in DTI images into major tracts** using `nighres.brain.dots_segmentation()`
4. Visualizing the results using `matplotlib`

2.6.1 Import and download

First we import `nighres` and `os` to set the output directory. Make sure to run this file in a directory you have write access to, or change the `out_dir` variable below. We can download the DOTS atlas priors and an example DTI dataset using the following command. The registration step of the DOTS function relies on `dipy`, so make sure you have installed it (<https://nipy.org/dipy/>).

```
import os
import dipy
import nighres

in_dir = os.path.join(os.getcwd(), 'nighres_examples/data_sets')
out_dir = os.path.join(os.getcwd(), 'nighres_examples/dots_segmentation')
atlas_dir = os.path.join(os.getcwd(), 'nighres_examples')
nighres.data.download_DOTS_atlas(atlas_dir)
dataset = nighres.data.download_DTI_2mm(in_dir)
```

We also try to import Nilearn plotting functions. If Nilearn is not installed, plotting will be skipped.

```
skip_plots = False
try:
    from nilearn import plotting
except ImportError:
    skip_plots = True
    print('Nilearn could not be imported, plotting will be skipped')
```

2.6.2 White matter segmentation

The DOTS segmentation can be run as follows. By default, the algorithm uses the tract atlas consisting of 23 tracts specified in². This can be changed to the full atlas by changing the value of the parameter `'wm_atlas'` to 2. Please see documentation for details.

¹ Bazin, Pierre-Louis, et al. "Direct segmentation of the major white matter tracts in diffusion tensor images." *Neuroimage* (2011) doi: <https://doi.org/10.1016/j.neuroimage.2011.06.020>

² Bazin, Pierre-Louis, et al. "Efficient MRF segmentation of DTI white matter tracts using an overlapping fiber model." *Proceedings of the International Workshop on Diffusion Modelling and Fiber Cup* (2009)

```
dots_results = nighres.brain.dots_segmentation(tensor_image=dataset['dti'],
                                              mask=dataset['mask'],
                                              atlas_dir=atlas_dir,
                                              save_data=True,
                                              output_dir=out_dir,
                                              file_name='example')

segmentation = nighres.io.load_volume(dots_results['segmentation']).get_data()
posterior = nighres.io.load_volume(dots_results['posterior']).get_data()
```

Tip: The parameter values of the DOTS algorithm can have a significant effect on segmentation results. Experiment with changing their values to obtain optimal results.

2.6.3 Interpretation of results

The integers in the segmentation array and the fourth dimension of the posterior array correspond to the tracts specified in `atlas_labels_1` (in case of using `wm_atlas 1`) which can be imported as follows

```
from nighres.brain.dots_segmentation import atlas_labels_1
```

2.6.4 Visualization of results

We can visualize the segmented tracts overlaid on top of a fractional anisotropy map. Let's first import the necessary modules and define a colormap. Then, we calculate the FA map and show the tracts. Let's also show the posterior probability distribution of an individual tract.

```
import numpy as np
import nibabel as nb
import matplotlib.pyplot as plt
from matplotlib import cm
from matplotlib.colors import ListedColormap

# This defines the following colormap
# transparent = isotropic
# semi-transparent red = unclassified white matter
# opaque colours = individual tracts
# white = overlapping tracts

N_t = 23 # 41 if using atlas 2
N_o = 50 # 185 is using atlas 2
newcolors = np.zeros((N_t + N_o, 4))
newcolors[0,:] = np.array([.2, .2, .2, 0])
newcolors[1,:] = np.array([1, 0, 0, .25])
rainbow = cm.get_cmap('rainbow', N_t - 2)
newcolors[2:N_t,:] = rainbow(np.linspace(0, 1, N_t - 2))
newcolors[N_t::,:] = np.ones(newcolors[N_t:::].shape)
newcmp = ListedColormap(newcolors)

# Calculate FA
tensor_img = nb.load(os.path.join(in_dir, 'DTI_2mm/DTI_2mm.nii.gz'))
tensor_volume = tensor_img.get_data()
xs, ys, zs, _ = tensor_volume.shape
```

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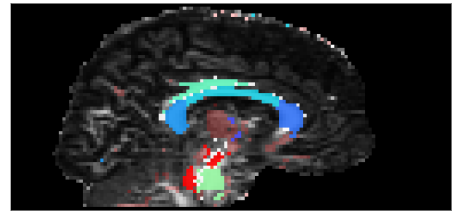
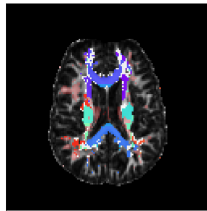
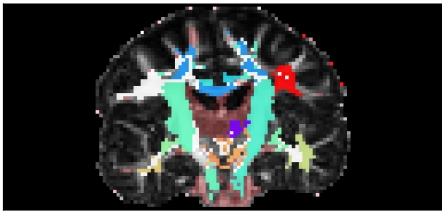
(continued from previous page)

```

tenfit = np.zeros((xs, ys, zs, 3, 3))
tenfit[:, :, :, 0, 0] = tensor_volume[:, :, :, 0]
tenfit[:, :, :, 1, 1] = tensor_volume[:, :, :, 1]
tenfit[:, :, :, 2, 2] = tensor_volume[:, :, :, 2]
tenfit[:, :, :, 0, 1] = tensor_volume[:, :, :, 3]
tenfit[:, :, :, 1, 0] = tensor_volume[:, :, :, 3]
tenfit[:, :, :, 0, 2] = tensor_volume[:, :, :, 4]
tenfit[:, :, :, 2, 0] = tensor_volume[:, :, :, 4]
tenfit[:, :, :, 1, 2] = tensor_volume[:, :, :, 5]
tenfit[:, :, :, 2, 1] = tensor_volume[:, :, :, 5]
tenfit[np.isnan(tenfit)] = 0
evals, evecs = np.linalg.eig(tenfit)
R = tenfit / np.trace(tenfit, axis1=3, axis2=4)[:, :, :, np.newaxis, np.newaxis]
FA = np.sqrt(0.5 * (3 - 1/(np.trace(np.matmul(R,R), axis1=3, axis2=4))))
FA[np.isnan(FA)] = 0
# save for convenience
nighres.io.save_volume(os.path.join(out_dir, 'FA.nii.gz'),
                      nb.NiftiImage(FA, tensor_img.affine, tensor_img.header))

# Show segmentation
fig, ax = plt.subplots(1, 3, figsize=(28,5))
ax[0].imshow(np.rot90(FA[:,60,:]), cmap='gray', vmin=0, vmax=1)
ax[0].imshow(np.rot90(segmentation[:,60,:]), cmap=newcmp, alpha=.9)
ax[1].imshow(np.rot90(FA[:, :, 30]), cmap='gray', vmin=0, vmax=1)
ax[1].imshow(np.rot90(segmentation[:, :, 30]), cmap=newcmp, alpha=.9)
ax[2].imshow(np.rot90(FA[60, :, :]), cmap='gray', vmin=0, vmax=1)
ax[2].imshow(np.rot90(segmentation[60, :, :]), cmap=newcmp, alpha=.9)
for i in range(3):
    ax[i].set_xticks([])
    ax[i].set_yticks([])
fig.tight_layout()
#fig.savefig('segmentation.png')
plt.show()

```



```

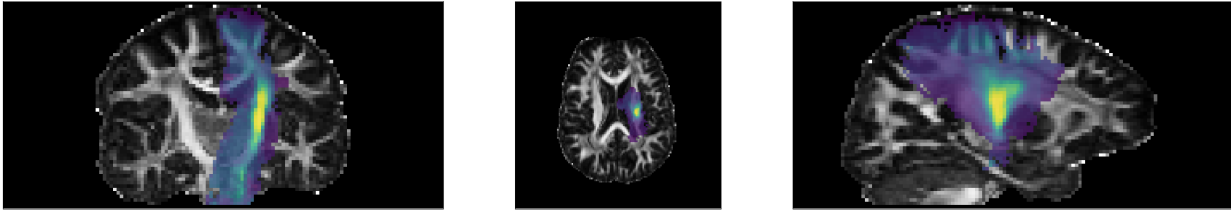
# Show posterior probability of the left corticospinal tract
tract_idx = 9
fig, ax = plt.subplots(1, 3, figsize=(28,5))
ax[0].imshow(np.rot90(FA[:,60,:]), cmap='gray', vmin=0, vmax=1)
ax[0].imshow(np.rot90(posterior[:,60,:,tract_idx]), alpha=.75)
ax[1].imshow(np.rot90(FA[:, :, 30]), cmap='gray', vmin=0, vmax=1)
ax[1].imshow(np.rot90(posterior[:, :, 30, tract_idx]), alpha=.75)
ax[2].imshow(np.rot90(FA[75, :, :]), cmap='gray', vmin=0, vmax=1)
ax[2].imshow(np.rot90(posterior[75, :, :, tract_idx]), alpha=.75)
for i in range(3):
    ax[i].set_xticks([])
    ax[i].set_yticks([])
fig.tight_layout()
#fig.savefig('CST_posterior.png')

```

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```
plt.show()
```



2.6.5 References

Total running time of the script: (0 minutes 0.000 seconds)

2.7 Multiatlas Segmentation

This example shows how to perform multi-atlas segmentation based on MP2RAGE data by performing the following steps:

1. **Downloading three open MP2RAGE datasets using** `nighres.data.download_7T_TRT()`
2. **Remove the skull and create a brain mask using** `nighres.brain.mp2rage_skullstripping()`
3. **Atlas-guided tissue classification using MGDM for first two subjects to be used as an atlas using** `nighres.brain.mgdm_segmentation()`¹
4. **Co-register non-linearly the atlas brains the the third subject using** `nighres.registration.embedded_syn()`²
5. **Deform segmentation labels using** `nighres.registration.apply_deformation()`
6. **Turn individual labels into levelset surfaces using** `nighres.surface.probability_to_levelset()`
7. **Build a final shape average using**

```
func nighres.shape.levelset_fusion
```

Important note: this example is both computationally expensive (recomputing everything from basic inputs) and practically pointless (a direct MGDM segmentation or a multi-atlas approach with manually defined labels and more subjects would both be meaningful). This example is only meant as illustration.

2.7.1 Import and download

First we import `nighres` and the `os` module to set the output directory. Make sure to run this file in a directory you have write access to, or change the `out_dir` variable below.

```
import nighres
import os
import nibabel as nb
```

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¹ Bogovic, Prince and Bazin (2013). A multiple object geometric deformable model for image segmentation. DOI: 10.1016/j.cviu.2012.10.006.A

² Avants et al (2008). Symmetric diffeomorphic image registration with cross-correlation: evaluating automated labeling of elderly and neurodegenerative brain. DOI: 10.1016/j.media.2007.06.004

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```
in_dir = os.path.join(os.getcwd(), 'nighres_examples/data_sets')
out_dir = os.path.join(os.getcwd(), 'nighres_examples/multiatlas_segmentation')
```

We also try to import Nilearn plotting functions. If Nilearn is not installed, plotting will be skipped.

```
skip_plots = False
try:
    from nilearn import plotting
except ImportError:
    skip_plots = True
    print('Nilearn could not be imported, plotting will be skipped')
```

Now we download an example MP2RAGE dataset. It is the structural scan of the first subject, first session of the 7T Test-Retest dataset published by Gorgolewski et al (2015)³.

```
dataset1 = nighres.data.download_7T_TRT(in_dir, subject_id='sub001_sess1')
dataset2 = nighres.data.download_7T_TRT(in_dir, subject_id='sub002_sess1')
dataset3 = nighres.data.download_7T_TRT(in_dir, subject_id='sub003_sess1')
```

2.7.2 Skull stripping

First we perform skull stripping. Only the second inversion image is required to calculate the brain mask. But if we input the T1map and T1w image as well, they will be masked for us. We also save the outputs in the `out_dir` specified above and use a subject ID as the base `file_name`.

```
skullstripping_results1 = nighres.brain.mp2rage_skullstripping(
    second_inversion=dataset1['inv2'],
    t1_weighted=dataset1['t1w'],
    t1_map=dataset1['t1map'],
    save_data=True,
    file_name='sub001_sess1',
    output_dir=out_dir)

skullstripping_results2 = nighres.brain.mp2rage_skullstripping(
    second_inversion=dataset2['inv2'],
    t1_weighted=dataset2['t1w'],
    t1_map=dataset2['t1map'],
    save_data=True,
    file_name='sub002_sess1',
    output_dir=out_dir)

skullstripping_results3 = nighres.brain.mp2rage_skullstripping(
    second_inversion=dataset3['inv2'],
    t1_weighted=dataset3['t1w'],
    t1_map=dataset3['t1map'],
    save_data=True,
    file_name='sub003_sess1',
    output_dir=out_dir)
```

Tip: in Nighres functions that have several outputs return a dictionary storing the different outputs. You can find the keys in the docstring by typing `nighres.brain.mp2rage_skullstripping?` or list them with

³ Gorgolewski et al (2015). A high resolution 7-Tesla resting-state fMRI test-retest dataset with cognitive and physiological measures. DOI: 10.1038/sdata.2014.54

```
skullstripping_results.keys()
```

To check if the skull stripping worked well we plot the brain mask on top of the original image. You can also open the images stored in `out_dir` in your favourite interactive viewer and scroll through the volume.

Like Nilearn, we use Nibabel SpatialImage objects to pass data internally. Therefore, we can directly plot the outputs using Nilearn plotting functions .

```
if not skip_plots:
    plotting.plot_roi(skullstripping_results1['brain_mask'], dataset1['t1map'],
                     annotate=False, black_bg=False, draw_cross=False,
                     cmap='autumn')
    plotting.plot_roi(skullstripping_results2['brain_mask'], dataset2['t1w'],
                     annotate=False, black_bg=False, draw_cross=False,
                     cmap='autumn')
    plotting.plot_roi(skullstripping_results3['brain_mask'], dataset3['t1w'],
                     annotate=False, black_bg=False, draw_cross=False,
                     cmap='autumn')
```

2.7.3 MGDM classification

Next, we use MGDM to estimate anatomical labels from subjects 1 and 2

```
mgdm_results1 = nighres.brain.mgdm_segmentation(
    contrast_image1=skullstripping_results1['t1w_masked'],
    contrast_type1="Mp2rage7T",
    contrast_image2=skullstripping_results1['t1map_masked'],
    contrast_type2="T1map7T",
    save_data=True, file_name="sub001_sess1",
    output_dir=out_dir)

mgdm_results2 = nighres.brain.mgdm_segmentation(
    contrast_image1=skullstripping_results2['t1w_masked'],
    contrast_type1="Mp2rage7T",
    contrast_image2=skullstripping_results2['t1map_masked'],
    contrast_type2="T1map7T",
    save_data=True, file_name="sub002_sess1",
    output_dir=out_dir)
```

Now we look at the topology-constrained segmentation MGDM created

```
if not skip_plots:
    plotting.plot_img(mgdm_results1['segmentation'],
                     vmin=1, vmax=50, cmap='cubehelix', colorbar=True,
                     annotate=False, draw_cross=False)
    plotting.plot_img(mgdm_results2['segmentation'],
                     vmin=1, vmax=50, cmap='cubehelix', colorbar=True,
                     annotate=False, draw_cross=False)
```

2.7.4 SyN co-registration

Next, we use the masked data as input for co-registration. The T1 maps are used here as they are supposed to be more similar

```

ants_results1 = nighres.registration.embedded_antsreg(
    source_image=skullstripping_results1['t1map_masked'],
    target_image=skullstripping_results3['t1map_masked'],
    run_rigid=True, run_affine=False, run_syn=False,
    coarse_iterations=40,
    medium_iterations=0, fine_iterations=0,
    cost_function='MutualInformation',
    interpolation='NearestNeighbor',
    ignore_affine=True,
    save_data=True, file_name="sub001_sess1",
    output_dir=out_dir)

ants_results2 = nighres.registration.embedded_antsreg(
    source_image=skullstripping_results2['t1map_masked'],
    target_image=skullstripping_results3['t1map_masked'],
    run_rigid=True, run_affine=False, run_syn=False,
    coarse_iterations=40,
    medium_iterations=0, fine_iterations=0,
    cost_function='MutualInformation',
    interpolation='NearestNeighbor',
    ignore_affine=True,
    save_data=True, file_name="sub002_sess1",
    output_dir=out_dir)

```

Now we look at the coregistered image that SyN created

```

if not skip_plots:
    plotting.plot_img(ants_results1['transformed_source'],
                      annotate=False, draw_cross=False)
    plotting.plot_img(ants_results2['transformed_source'],
                      annotate=False, draw_cross=False)

```

2.7.5 Apply deformations to segmentations

We use the computed deformation to transform MGDM segmentations

```

deformed1 = nighres.registration.apply_coordinate_mappings(
    image=mgdm_results1['segmentation'],
    mapping1=ants_results1['mapping'],
    save_data=True, file_name="sub001_sess1_seg",
    output_dir=out_dir)

deformed2 = nighres.registration.apply_coordinate_mappings(
    image=mgdm_results2['segmentation'],
    mapping1=ants_results2['mapping'],
    save_data=True, file_name="sub002_sess1_seg",
    output_dir=out_dir)

```

Now we look at the segmentations deformed by SyN

```

if not skip_plots:
    plotting.plot_img(deformed1['result'],
                      annotate=False, draw_cross=False)
    plotting.plot_img(deformed2['result'],
                      annotate=False, draw_cross=False)

```

2.7.6 Transform a selected label into levelset representation

We use the deformed MGDM segmentations

```
# label 32 = left caudate
img1 = nighres.io.load_volume(deformed1['result'])
struct1 = nb.Nifti1Image((img1.get_data()==32).astype(float),
                        img1.affine, img1.header)

img2 = nighres.io.load_volume(deformed2['result'])
struct2 = nb.Nifti1Image((img2.get_data()==32).astype(float),
                        img2.affine, img2.header)

levelset1 = nighres.surface.probability_to_levelset(
    probability_image=struct1,
    save_data=True, file_name="sub001_sess1_struct",
    output_dir=out_dir)

levelset2 = nighres.surface.probability_to_levelset(
    probability_image=struct2,
    save_data=True, file_name="sub002_sess1_struct",
    output_dir=out_dir)

final_seg = nighres.shape.levelset_fusion(levelset_images=[levelset1['result'],
    levelset2['result']],
    correct_topology=True,
    save_data=True, file_name="sub003_sess1_struct_seg",
    output_dir=out_dir)
```

Now we look at the final segmentation from shape fusion

```
if not skip_plots:
    img = nighres.io.load_volume(levelset1['result'])
    mask = nb.Nifti1Image((img.get_data()<0).astype(bool),
                        img.affine, img.header)
    plotting.plot_roi(mask, dataset3['t1map'],
                    annotate=False, black_bg=False, draw_cross=False,
                    cmap='autumn')

    img = nighres.io.load_volume(levelset2['result'])
    mask = nb.Nifti1Image((img.get_data()<0).astype(bool),
                        img.affine, img.header)
    plotting.plot_roi(mask, dataset3['t1map'],
                    annotate=False, black_bg=False, draw_cross=False,
                    cmap='autumn')

    img = nighres.io.load_volume(final_seg['result'])
    mask = nb.Nifti1Image((img.get_data()<0).astype(bool),
                        img.affine, img.header)
    plotting.plot_roi(mask, dataset3['t1map'],
                    annotate=False, black_bg=False, draw_cross=False,
                    cmap='autumn')
```

If the example is not run in a jupyter notebook, render the plots:

```
if not skip_plots:
    plotting.show()
```


2.7.7 References

Total running time of the script: (0 minutes 0.000 seconds)

3.1 dots_segmentation

nighres.brain.dots_segmentation(*tensor_image*, *mask*, *atlas_dir*, *wm_atlas=1*, *max_iter=25*, *convergence_threshold=0.005*, *s_I=0.023809523809523808*, *c_O=0.5*, *max_angle=67.5*, *save_data=False*, *overwrite=False*, *output_dir=None*, *file_name=None*)

DOTS segmentation

Segment major white matter tracts in diffusion tensor images using Diffusion Oriented Tract Segmentation (DOTS) algorithm.

Parameters

- **tensor_image** (*nimg*) – Input image containing the diffusion tensor coefficients in the following order: volumes 0-5: D11, D22, D33, D12, D13, D23
- **mask** (*nimg*) – Binary brain mask image which limits computation to the defined volume.
- **atlas_dir** (*str*) – Path to directory where the DOTS atlas information is stored. The atlas information should be stored in a subdirectory called ‘DOTS_atlas’ as generated by `nighres.data.download_DOTS_atlas()`.
- **wm_atlas** (*int*, *optional*) – Define which white matter atlas to use. Option 1 for 23 tracts² and option 2 for 39 tracts¹. (default is 1)
- **max_iter** (*int*, *optional*) – Maximum number of iterations in the conditional modes algorithm. (default is 20)
- **convergence_threshold** (*float*, *optional*) – Threshold for when the iterated conditional modes algorithm is considered to have converged. Defined as the fraction of labels that change during one step of the algorithm. (default is 0.002)

² Bazin, Pierre-Louis, et al. “Efficient MRF segmentation of DTI white matter tracts using an overlapping fiber model.” Proceedings of the International Workshop on Diffusion Modelling and Fiber Cup (2009)

¹ Bazin, Pierre-Louis, et al. “Direct segmentation of the major white matter tracts in diffusion tensor images.” Neuroimage (2011) doi: <https://doi.org/10.1016/j.neuroimage.2011.06.020>

- **s_I** (*float, optional*) – Parameter controlling how isotropic label energies propagate to their neighborhood. (default is 1/42)
- **c_O** (*float, optional*) – Weight parameter for unclassified white matter atlas prior. (default is 1/2)
- **max_angle** (*float, optional*) – Maximum angle (in degrees) between principal tensor directions before connectivity coefficient *c* becomes negative. Possible values between 0 and 90. (default is 67.5)
- **save_data** (*bool, optional*) – Save output data to file. (default is False)
- **overwrite** (*bool, optional*) – Overwrite existing results. (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist.
- **file_name** (*str, optional*) – Desired base name for output files without file extension, suffixes will be added.

Returns

Dictionary collecting outputs under the following keys (type of output files in brackets)

- **segmentation** (array_like): Hard segmentation of white matter.
- **posterior** (array_like): POsterior probabilities of tracts.

Return type dict

Notes

Algorithm details can be found in the references below.

References

3.1.1 Examples using `nighres.brain.dots_segmentation`

- *DOTS white matter segmentation*

3.2 `extract_brain_region`

`nighres.brain.extract_brain_region` (*segmentation, levelset_boundary, maximum_membership, maximum_label, extracted_region, atlas_file=None, normalize_probabilities=False, estimate_tissue_densities=False, partial_volume_distance=1.0, save_data=False, overwrite=False, output_dir=None, file_name=None*)

Extract Brain Region

Extracts masks, probability maps and levelset surfaces for specific brain regions and regions from a Multiple Object Geometric Deformable Model (MGDM) segmentation result.

Parameters

- **segmentation** (*nimg*) – Segmentation result from MGDM.
- **levelset_boundary** (*nimg*) – Levelset boundary from MGDM.

- **maximum_membership** (*niimg*) – 4D image of the maximum membership values from MGDM.
- **maximum_label** (*niimg*) – 4D image of the maximum labels from MGDM.
- **atlas_file** (*str, optional*) – Path to plain text atlas file (default is stored in DEFAULT_ATLAS). or atlas name to be searched in ATLAS_DIR
- **extracted_region** (*{'left_cerebrum', 'right_cerebrum', 'cerebrum', 'cerebellum', 'cerebellum_brainstem', 'subcortex', 'tissues(anat)', 'tissues(func)', 'brain_mask'}*) Region to be extracted from the MGDM segmentation.
- **normalize_probabilities** (*bool*) – Whether to normalize the output probabilities to sum to 1 (default is False).
- **estimate_tissue_densities** (*bool*) – Whether to recompute partial volume densities from the probabilities (slow, default is False).
- **partial_volume_distance** (*float*) – Distance in mm to use for tissues densities, if recomputed (default is 1mm).
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets, # stands for shorthand names of the different extracted regions, respectively: rcr, lcr, cr, cb, cbs, sub, an, fn)

- **region_mask** (*niimg*): Hard segmentation mask of the (GM) region of interest (*_xmask-#gm*)
- **inside_mask** (*niimg*): Hard segmentation mask of the (WM) inside of the region of interest (*_xmask-#wm*)
- **background_mask** (*niimg*): Hard segmentation mask of the (CSF) region background (*_xmask-#bg*)
- **region_proba** (*niimg*): Probability map of the (GM) region of interest (*_xproba-#gm*)
- **inside_proba** (*niimg*): Probability map of the (WM) inside of the region of interest (*_xproba-#wm*)
- **background_proba** (*niimg*): Probability map of the (CSF) region background (*_xproba-#bg*)
- **region_lvl** (*niimg*): Levelset surface of the (GM) region of interest (*_xlvl-#gm*)
- **inside_lvl** (*niimg*): Levelset surface of the (WM) inside of the region of interest (*_xlvl-#wm*)
- **background_lvl** (*niimg*): Levelset surface of the (CSF) region background (*_xlvl-#bg*)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin.

3.2.1 Examples using `nighres.brain.extract_brain_region`

- *Cortical depth estimation from MGDM segmentation*

3.3 filter_stacking

```
nighres.brain.filter_stacking(dura_img=None, pvcsf_img=None, arteries_img=None,  
                             save_data=False, overwrite=False, output_dir=None,  
                             file_name=None)
```

Filter stacking

A small utility to combine multiple priors derived from filtering of the CSF partial voluming, arteries, and/or remaining dura mater. The filter priors (in [0,1]) are arranged in a specific way (in increments of 2) that is expected by the Filters contrast type in MGDM

Parameters

- **dura_img** (*niimg, optional*) – Prior for the location of remaining dura mater after skull stripping. At least one prior image is required
- **pvcsf_img** (*niimg, optional*) – Prior for the location of CSF partial voluming (mostly in sulcal regions). At least one prior image is required
- **arteries_img** (*niimg, optional*) – Prior for the location of arteries, visible e.g. in MP2RAGE images. At least one prior image is required
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix in brackets)

- **result** (*niimg*): Combined image, where only the strongest priors are kept (*_bfs-img*)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin.

3.4 intensity_based_skullstripping

```
nighres.brain.intensity_based_skullstripping(main_image, extra_image=None,
                                             noise_model='exponential',
                                             skip_zero_values=True, iterate=False,
                                             dilate_mask=0, dynamic_range=0.8,
                                             topology_lut_dir=None, save_data=False,
                                             overwrite=False, output_dir=None,
                                             file_name=None)
```

Intensity-based skull stripping

Estimate a brain mask for a dataset with good brain/background intensity separation (e.g. PD-weighted). An extra image can be used to ensure high intensities are preserved (e.g. T1 map, T2-weighted data or a probability map for a ROI).

Parameters

- **main_image** (*niimg*) – Main Intensity Image
- **extra_image** (*niimg*, *optional*) – Extra image with high intensity at brain boundary
- **noise_model** (*{'exponential', 'half-normal', 'exp+log-normal', 'half+log-normal'}*) – Background noise model (default is 'exponential')
- **skip_zero_values** (*bool*) – Ignores voxels with zero value (default is True)
- **iterate** (*bool*) – Whether to iterate the estimation (may be unstable in some cases, default is False)
- **dilate_mask** (*int*) – Additional dilation (or erosion, if negative) of the brain mask (default is 0)
- **dynamic_range** (*float*) – Dynamic range for the foreground / background differences in [0,1] (default is 0.8)
- **topology_lut_dir** (*str*, *optional*) – Path to directory in which topology files are stored (default is stored in TOPOLOGY_LUT_DIR)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **brain_mask** (*niimg*): Binary brain mask (*_istrip-mask*)
- **brain_proba** (*niimg*): Probability brain map (*_istrip-proba*)
- **main_masked** (*niimg*): Masked main image (*_istrip-main*)
- **extra_masked** (*niimg*): Masked extra map (*_istrip-extra*)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin. Details on the algorithm can be found in¹

References

3.5 mgdm_segmentation

```
nighres.brain.mgdm_segmentation(contrast_image1, contrast_type1, contrast_image2=None,
                                contrast_type2=None, contrast_image3=None, contrast_type3=None,
                                contrast_image4=None, contrast_type4=None, n_steps=5,
                                max_iterations=800, topology='wcs', atlas_file=None,
                                topology_lut_dir=None, adjust_intensity_priors=False,
                                normalize_qmaps=True, compute_posterior=False,
                                posterior_scale=5.0, diffuse_probabilities=False,
                                save_data=False, overwrite=False, output_dir=None,
                                file_name=None)
```

MGDM segmentation

Estimates brain structures from an atlas for MRI data using a Multiple Object Geometric Deformable Model (MGDM)

Parameters

- **contrast_image1** (*niimg*) – First input image to perform segmentation on
- **contrast_type1** (*str*) – Contrast type of first input image, must be listed as a prior in used atlas (specified in *atlas_file*). Possible inputs by default are DWIFA3T, DWIMD3T, T1map9T, Mp2rage9T, T1map7T, Mp2rage7T, PV, Filters, T1pv, Mprage3T, T1map3T, Mp2rage3T, HCPT1w, HCPT2w, NormMPRAGE.
- **contrast_image2** (*niimg, optional*) – Additional input image to inform segmentation, must be in the same space as *contrast_image1*, requires *contrast_type2*
- **contrast_type2** (*str, optional*) – Contrast type of second input image, must be listed as a prior in used atlas (specified in *atlas_file*). Possible inputs by default are the same as with parameter *contrast_type1* (see above).
- **contrast_image3** (*niimg, optional*) – Additional input image to inform segmentation, must be in the same space as *contrast_image1*, requires *contrast_type3*
- **contrast_type3** (*str, optional*) – Contrast type of third input image, must be listed as a prior in used atlas (specified in *atlas_file*). Possible inputs by default are the same as with parameter *contrast_type1* (see above).
- **contrast_image4** (*niimg, optional*) – Additional input image to inform segmentation, must be in the same space as *contrast_image1*, requires *contrast_type4*
- **contrast_type4** (*str, optional*) – Contrast type of fourth input image, must be listed as a prior in used atlas (specified in *atlas_file*). Possible inputs by default are the same as with parameter *contrast_type1* (see above).
- **n_steps** (*int, optional*) – Number of steps for MGDM (default is 5, set to 0 for quick testing of registration of priors, which does not perform true segmentation)

¹Bazin et al. (2014). A computational framework for ultra-high resolution cortical segmentation at 7 Tesla. DOI: 10.1016/j.neuroimage.2013.03.077

- **max_iterations** (*int, optional*) – Maximum number of iterations per step for MGDm (default is 800, set to 1 for quick testing of registration of priors, which does not perform true segmentation)
- **topology** (*{'wcs', 'no'}, optional*) – Topology setting, choose ‘wcs’ (well-composed surfaces) for strongest topology constraint, ‘no’ for no topology constraint (default is ‘wcs’)
- **atlas_file** (*str, optional*) – Path to plain text atlas file (default is stored in DEFAULT_ATLAS) or atlas name to be searched in ATLAS_DIR
- **topology_lut_dir** (*str, optional*) – Path to directory in which topology files are stored (default is stored in TOPOLOGY_LUT_DIR)
- **normalize_qmaps** (*bool*) – Normalize quantitative maps into [0,1] (default is False)
- **adjust_intensity_priors** (*bool*) – Adjust intensity priors based on dataset (default is False)
- **normalize_qmaps** – Normalize quantitative maps in [0,1] (default in True, change this if using one of the -quant atlas text files in ATLAS_DIR)
- **compute_posterior** (*bool*) – Compute posterior probabilities for segmented structures (default is False)
- **posterior_scale** (*float*) – Posterior distance scale from segmented structures to compute posteriors (default is 5.0 mm)
- **diffuse_probabilities** (*bool*) – Regularize probability distribution with a non-linear diffusion scheme (default is False)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn’t exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- segmentation (niimg): Hard brain segmentation with topological constraints (if chosen) (`_mgdm_seg`)
- labels (niimg): Maximum tissue probability labels (`_mgdm_lbls`)
- memberships (niimg): Maximum tissue probability values, 4D image where the first dimension shows each voxel’s highest probability to belong to a specific tissue, the second dimension shows the second highest probability to belong to another tissue etc. (`_mgdm_mems`)
- distance (niimg): Minimum distance to a segmentation boundary (`_mgdm_dist`)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin. Algorithm details can be found in¹ and²

References

3.5.1 Examples using `nighres.brain.mgdm_segmentation`

- *Tissue classification from MP2RAGE data*
- *Multiatlas Segmentation*

3.6 `mp2rage_dura_estimation`

```
nighres.brain.mp2rage_dura_estimation(second_inversion, skullstrip_mask, back-  
ground_distance=5.0, output_type='dura_region',  
save_data=False, overwrite=False, out-  
put_dir=None, file_name=None)
```

MP2RAGE dura estimation

Filters a MP2RAGE brain image to obtain a probability map of dura matter.

Parameters

- **second_inversion** (*niimg*) – Second inversion image derived from MP2RAGE sequence
- **skullstrip_mask** (*niimg*) – Skullstripping mask defining the approximate region including the brain
- **background_distance** (*float*) – Maximum distance within the mask for dura (default is 5.0 mm)
- **output_type** (*{'dura_region', 'boundary', 'dura_prior', 'bg_prior',}*) – ‘intens_prior’} Type of output result (default is ‘dura_region’)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn’t exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)
- **return_filename** (*bool, optional*) – Return filename instead of object

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **result** (*niimg*): Dura probability image (`_dura-proba`)

Return type `dict`

¹ Bazin et al. (2014). A computational framework for ultra-high resolution cortical segmentation at 7 Tesla. doi:10.1016/j.neuroimage.2013.03.077

² Bogovic et al. (2013). A multiple object geometric deformable model for image segmentation. doi:10.1016/j.cviu.2012.10.006.A

Notes

Original Java module by Pierre-Louis Bazin. Details on the algorithm can be found in¹ and a presentation of the MP2RAGE sequence in²

References

3.7 mp2rage_skullstripping

`nighres.brain.mp2rage_skullstripping` (*second_inversion*, *t1_weighted=None*, *t1_map=None*, *skip_zero_values=True*, *topology_lut_dir=None*, *save_data=False*, *overwrite=False*, *output_dir=None*, *file_name=None*)

MP2RAGE skull stripping

Estimates a brain mask from MRI data acquired with the MP2RAGE sequence. At least a T1-weighted or a T1 map image is required

Parameters

- **second_inversion** (*niimg*) – Second inversion image derived from MP2RAGE sequence
- **t1_weighted** (*niimg*) – T1-weighted image derived from MP2RAGE sequence (also referred to as “uniform” image) At least one of `t1_weighted` and `t1_map` is required
- **t1_map** (*niimg*) – Quantitative T1 map image derived from MP2RAGE sequence At least one of `t1_weighted` and `t1_map` is required
- **skip_zero_values** (*bool*) – Ignores voxels with zero value (default is True)
- **topology_lut_dir** (*str*, *optional*) – Path to directory in which topology files are stored (default is stored in `TOPOLOGY_LUT_DIR`)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn’t exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- `brain_mask` (*niimg*): Binary brain mask (`_strip-mask`)
- `inv2_masked` (*niimg*): Masked second inversion image (`_strip-inv2`)
- `t1w_masked` (*niimg*): Masked T1-weighted image (`_strip-t1w`)
- `t1map_masked` (*niimg*): Masked T1 map (`_strip-t1map`)

Return type `dict`

¹ Bazin et al. (2014). A computational framework for ultra-high resolution cortical segmentation at 7 Tesla. DOI: 10.1016/j.neuroimage.2013.03.077

² Marques et al. (2010). MP2RAGE, a self bias-field corrected sequence for improved segmentation and T1-mapping at high field. DOI: 10.1016/j.neuroimage.2009.10.002

Notes

Original Java module by Pierre-Louis Bazin. Details on the algorithm can be found in¹ and a presentation of the MP2RAGE sequence in²

References

3.7.1 Examples using `nighres.brain.mp2rage_skullstripping`

- *Vascular Reconstruction*
- *Tissue classification from MP2RAGE data*
- *Brain co-registration from MP2RAGE data*
- *Multiatlas Segmentation*

¹ Bazin et al. (2014). A computational framework for ultra-high resolution cortical segmentation at 7 Tesla. DOI: 10.1016/j.neuroimage.2013.03.077

² Marques et al. (2010). MP2RAGE, a self bias-field corrected sequence for improved segmentation and T1-mapping at high field. DOI: 10.1016/j.neuroimage.2009.10.002

4.1 `cruise_cortex_extraction`

```
nighres.cortex.cruise_cortex_extraction(init_image, wm_image, gm_image, csf_image,
                                         vd_image=None, data_weight=0.4, regulariza-
                                         tion_weight=0.1, max_iterations=500, normal-
                                         ize_probabilities=False, correct_wm_pv=True,
                                         wm_dropoff_dist=1.0, topology='wcs', topol-
                                         ogy_lut_dir=None, save_data=False, over-
                                         write=False, output_dir=None, file_name=None)
```

CRUISE cortex extraction

Segments the cortex from a whole brain segmented data set with the CRUISE method (includes customized partial voluming corrections and the Anatomically-Consistent Enhancement (ACE) of sulcal fundi).

Note that the main input images are generated by the nighres module `nighres.brain.extract_brain_region()`.

Parameters

- **`init_image`** (*niimg*) – Initial white matter (WM) segmentation mask (binary mask >0 inside WM)
- **`wm_image`** (*niimg*) – Filled WM probability map (values in [0,1], including subcortical GM and ventricles)
- **`gm_image`** (*niimg*) – Cortical gray matter (GM) probability map (values in [0,1], highest inside the cortex)
- **`csf_image`** (*niimg*) – Sulcal cerebro-spinal fluid (CSf) and background probability map (values in [0,1], highest in CSf and masked regions)
- **`vd_image`** (*niimg*, *optional*) – Additional probability map of vessels and dura mater to be excluded
- **`data_weight`** (*float*) – Weighting of probability-based balloon forces in CRUISE (default 0.4, sum of {`data_weight`, `regularization_weight`} should be below or equal to 1)

- **regularization_weight** (*float*) – Weighting of curvature regularization forces in CRUISE (default 0.1, sum of {data_weight, regularization_weight} should be below or equal to 1)
- **max_iterations** (*int*) – Maximum number of iterations in CRUISE (default is 500)
- **normalize_probabilities** (*bool*) – Whether to normalize the wm, gm, and csf probabilities (default is False)
- **correct_wm_pv** (*bool*) – Whether to correct for WM partial voluming in gyral crowns (default is True)
- **wm_dropoff_dist** (*float*) – Distance parameter to lower WM probabilities away from current segmentation (default is 1.0 voxel)
- **topology** ({'wcs', 'no'}) – Topology setting, choose 'wcs' (well-composed surfaces) for strongest topology constraint, 'no' for no topology constraint (default is 'wcs')
- **topology_lut_dir** (*str*) – Path to directory in which topology files are stored (default is stored in TOPOLOGY_LUT_DIR)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- cortex (niimg): Hard segmentation of the cortex with labels background=0, gm=1, and wm=2 (`_cruise_cortex`)
- gwb (niimg): Gray-White matter Boundary (GWB) level set function (`_cruise_gwb`)
- cgb (niimg): CSF-Gray matter Boundary (CGB) level set function (`_cruise_cgb`)
- avg (niimg): Central level set function, obtained as geometric average of GWB and CGB (*not* the middle depth of the cortex, use `volumetric_layering` if you want accurate depth measures) (`_cruise-avg`)
- thickness (niimg): Simple cortical thickness estimate: distance to the GWB and CGB surfaces, in mm (`_cruise-thick`)
- pwm (niimg): Optimized WM probability, including partial volume and distant values correction (`_cruise-pwm`)
- pgm (niimg): Optimized GM probability, including CSF sulcal ridges correction (`_cruise_pgm`)
- pcsf (niimg): Optimized CSF probability, including sulcal ridges and vessel/dura correction (`_cruise-pwm`)

Return type `dict`

Notes

Original algorithm by Xiao Han. Java module by Pierre-Louis Bazin. Algorithm details can be found in¹

References

4.1.1 Examples using `nighres.cortex.cruise_cortex_extraction`

- *Cortical depth estimation from MGDM segmentation*

¹ X. Han, D.L. Pham, D. Tosun, M.E. Rettmann, C. Xu, and J. L. Prince, CRUISE: Cortical Reconstruction Using Implicit Surface Evolution, NeuroImage, vol. 23, pp. 997–1012, 2004

5.1 download_data

`nighres.data.download_7T_TRT` (*data_dir*, *overwrite=False*, *subject_id='sub001_sess1'*)

Downloads the MP2RAGE data from the 7T Test-Retest dataset published by Gorgolewski et al (2015) [1].

Parameters

- **data_dir** (*str*) – Writeable directory in which downloaded files should be stored. A subdirectory called ‘7T_TRT’ will be created in this location.
- **overwrite** (*bool*) – Overwrite existing files in the same exact path (default is False)
- **subject_id** (*'sub001_sess1', 'sub002_sess1', 'sub003_sess1'*) – Which dataset to download (default is ‘sub001_sess1’)

Returns

Dictionary with keys pointing to the location of the downloaded files

- `inv2` : path to second inversion image
- `t1w` : path to T1-weighted (uniform) image
- `t1map` : path to quantitative T1 image

Return type `dict`

Notes

The full dataset is available at http://openscience.cbs.mpg.de/7t_trt/

References

5.1.1 Examples using `nighres.data.download_7T_TRT`

- *Vascular Reconstruction*
- *Tissue classification from MP2RAGE data*
- *Brain co-registration from MP2RAGE data*
- *Multiatlas Segmentation*

`nighres.data.download_DTI_2mm` (*data_dir*, *overwrite=False*)

Downloads an example DTI data set

Parameters

- **data_dir** (*str*) – Writeable directory in which downloaded files should be stored. A subdirectory called ‘DTI_2mm’ will be created in this location.
- **overwrite** (*bool*) – Overwrite existing files in the same exact path (default is False)

Returns

Dictionary with keys pointing to the location of the downloaded files

- `dti` : path to DTI image
- `mask` : path to binary brain mask

Return type `dict`

5.1.2 Examples using `nighres.data.download_DTI_2mm`

- *DOTS white matter segmentation*

`nighres.data.download_DOTS_atlas` (*data_dir=None*, *overwrite=False*)

Downloads the statistical atlas presented in [\[1\]](#)

Parameters

- **data_dir** (*str*) – Writeable directory in which downloaded atlas files should be stored. A subdirectory called ‘DOTS_atlas’ will be created in this location.
- **overwrite** (*bool*) – Overwrite existing files in the same exact path (default is False)

Returns

Dictionary with keys pointing to the location of the downloaded files

- `fiber_p` : path to atlas probability image
- `fiber_dir` : path to atlas direction image

Return type `dict`

References

5.1.3 Examples using `nighres.data.download_DOTS_atlas`

- *DOTS white matter segmentation*

6.1 filter_ridge_structures

```
nighres.filtering.filter_ridge_structures(input_image, structure_intensity='bright',
                                         output_type='probability',
                                         use_strict_min_max_filter=True,
                                         save_data=False, overwrite=False, out-
                                         put_dir=None, file_name=None)
```

Filter Ridge Structures

Uses an image filter to make a probabilistic image of ridge structures.

Parameters

- **input_image** (*nimg*) – Image containing structure-of-interest
- **structure_intensity** (*{'bright', 'dark', 'both'}*) – Image intensity of structure-of-interest
- **output_type** (*{'probability', 'intensity'}*) – Whether the image should be normalized to reflect probabilities
- **use_strict_min_max_filter** (*bool, optional*) – Choose between the more specific recursive ridge filter or a more sensitive bidirectional filter (default is True)
- **save_data** (*bool, optional*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- `ridge_structure_image`: Image that reflects the presence of ridges in the image (`_rdg_img`)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin.

6.2 multiscale_vessel_filter

`nighres.filtering.multiscale_vessel_filter` (*input_image*, *structure_intensity*='bright', *filterType*='RRF', *propagationtype*='diffusion', *threshold*=0.5, *factor*=0.5, *max_diff*=0.001, *max_itr*=100, *scale_step*=1.0, *scales*=4, *prior_image*=None, *invert_prior*=False, *save_data*=False, *overwrite*=False, *output_dir*=None, *file_name*=None)

Vessel filter with prior

Uses an image filter to make a probabilistic image of ridge structures.

Parameters

- **input_image** (*niimg*) – Image containing structure-of-interest
- **structure_intensity** (*str*) – Image intensity of structure-of-interest ‘bright’, ‘dark’, or ‘both’ (default is ‘bright’).
- **filterType** (*str*) – Decide for a filter type: either RRF or Hessian (default is ‘RRF’)
- **propagationtype** (*str*) – Set the diffusion model of the filter: either ‘diffusion’ or ‘belief’ propagation model (default is ‘diffusion’)
- **threshold** (*float*) – Set the propability treshold to decide at what probability the detected structure should be seen as a vessel (default is 0.5)
- **factor** (*float*) – Diffusion factor between 0 and 100 (default is 0.5)
- **max_diff** (*float*) – maximal difference for stopping (default is 0.001)
- **max_itr** (*int*) – maximale iteration number (default is 100)
- **scale_step** (*float*) – Scaling step between diameters (default is 1)
- **scales** (*int*) – Number of scales to use (default is 4)
- **prior_image** (*niimg (opt)*) – Image prior for the region to include (positive) or exclude (negative)
- **invert_prior** (*boolean, optional (default is False)*) – In case there is a prior, the prior can be considered as negative prior (False) or as positive prior (True)
- **save_data** (*bool, optional*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn’t exist

- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **segmentation**: segmented vessel centerlines (`_mvf-seg`)
- **filtered**: result of the vessel filtering step (`_mvf-filter`)
- **probability**: probability score of segmented centerlines (`_mvf-proba`)
- **scale**: discrete scale at which the centerlines are detected (`_mvf-scale`)
- **diameter**: estimated vessel diameter (`_mvf-dia`)
- **length**: length of continuous vessel segments (`_mvf-length`)
- **pv**: partial volume estimate of vessels (`_mvf-pv`)
- **label**: labeling of individual vessel segments (`_mvf-label`)
- **direction**: estimated vessel direction (`_mvf-dir`)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin and Julia Huck.

6.2.1 Examples using `nighres.filtering.multiscale_vessel_filter`

- *Vascular Reconstruction*

6.3 recursive_ridge_diffusion

`nighres.filtering.recursive_ridge_diffusion` (*input_image*, *ridge_intensities*, *ridge_filter*, *surface_levelset=None*, *orientation='undefined'*, *loc_prior=None*, *min_scale=0*, *max_scale=3*, *diffusion_factor=1.0*, *similarity_scale=0.1*, *max_iter=100*, *max_diff=0.001*, *threshold=0.5*, *save_data=False*, *overwrite=False*, *output_dir=None*, *file_name=None*)

Recursive Ridge Diffusion

Extracts planar of tubular structures across multiple scales, with an optional directional bias.

Parameters

- **input_image** (*nimg*) – Input image
- **ridge_intensities** (*{'bright', 'dark', 'both'}*) – Which intensities to consider for the filtering
- **ridge_filter** (*{'2D', '1D', '0D'}*) – Whether to filter for 2D ridges, 1D vessels, or 0D holes

- **surface_levelset** (*niimg*, *optional*) – Level set surface to restrict the orientation of the detected features
- **orientation** (`{'undefined', 'parallel', 'orthogonal'}`) – The orientation of features to keep with regard to the surface or its normal
- **loc_prior** (*niimg*, *optional*) – Location prior image to restrict the search for features
- **min_scale** (*int*) – Minimum scale (in voxels) to look for features (default is 0)
- **max_scale** (*int*) – Maximum scale (in voxels) to look for features (default is 3)
- **diffusion_factor** (*float*) – Scaling factor for the diffusion weighting in [0,1] (default is 1.0)
- **similarity_scale** (*float*) – Scaling of the similarity function as a factor of intensity range
- **max_iter** (*int*) – Maximum number of diffusion iterations
- **max_diff** (*int*) – Maximum difference to stop the diffusion
- **threshold** (*float*) – Detection threshold for the structures to keep (default is 0.5)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **filter** (*niimg*): raw filter response (`_rrd-filter`)
- **propagation** (*niimg*): propagated probabilistic response after diffusion (`_rrd-propag`)
- **scale** (*niimg*): scale of the detection filter (`_rrd-scale`)
- **ridge_dir** (*niimg*): estimated local ridge direction (`_rrd-dir`)
- **ridge_pv** (*niimg*): ridge partial volume map, taking size into account (`_rrd-pv`)
- **ridge_size** (*niimg*): estimated size of each detected component (`rrd-size`)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin. Extension of the recursive ridge filter in¹.

¹ Bazin et al (2016), Vessel segmentation from quantitative susceptibility maps for local oxygenation venography, Proc ISBI.

References

6.4 total_variation_filtering

`nighres.filtering.total_variation_filtering` (*image*, *mask=None*, *lambda_scale=0.05*,
tau_step=0.125, *max_dist=0.0001*,
max_iter=500, *save_data=False*,
overwrite=False, *output_dir=None*,
file_name=None)

Total Variation Filtering

Total variation filtering.

Parameters

- **image** (*niimg*) – Input image to filter
- **mask** (*niimg*, *optional*) – Data mask for processing
- **lambda_scale** (*float*, *optional*) – Relative intensity scale for total variation smoothing (default is 0.5)
- **tau_step** (*float*, *optional*) – Internal step parameter (default is 0.125)
- **max_dist** (*float*, *optional*) – Maximum distance for convergence (default is 1e-4)
- **max_iter** (*int*, *optional*) – Maximum number of iterations (default is 500)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **filtered** (*niimg*): The filtered image (`_tv-img`)
- **residual** (*niimg*): The image residuals (`_tv-res`)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin. Algorithm adapted from¹

References

¹ Chambolle (2004). An Algorithm for Total Variation Minimization and Applications. doi:10.1023/B:JMIV.0000011325.36760.1e

7.1 background_estimation

`nighres.intensity.background_estimation` (*image*, *distribution*='exponential', *ratio*=0.001, *skip_zero*=True, *iterate*=True, *dilate*=0, *threshold*=0.5, *save_data*=False, *overwrite*=False, *output_dir*=None, *file_name*=None)

Background Estimation

Estimates image background by robustly fitting various distribution models

Parameters

- **image** (*nimg*) – Input image
- **distribution** (*{ 'exponential', 'half-normal' }*) – Distribution model to use for the background noise
- **ratio** (*float, optional*) – Robustness ratio for estimating image intensities
- **skip_zero** (*bool, optional*) – Whether to consider or skip zero values
- **iterate** (*bool, optional*) – Whether to run an iterative estimation (preferred, but sometimes unstable)
- **dilate** (*int, optional*) – Number of voxels to dilate or erode in the final mask
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- `masked (niimg)`: The background-masked input image
- `proba (niimg)`: The probability map of the foreground
- `mask (niimg)`: The mask of the foreground

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin.

7.2 flash_t2s_fitting

```
nighres.intensity.flash_t2s_fitting(image_list, te_list, r2s_threshold=None,
                                    save_data=False, overwrite=False, output_dir=None,
                                    file_name=None)
```

FLASH T2* fitting

Estimate T2*/R2* by linear least squares fitting in log space.

Parameters

- `image_list` (`[niimg]`) – List of input images to fit the T2* curve
- `te_list` (`[float]`) – List of input echo times (TE)
- `r2s_threshold` (`float`) – Threshold of R2* values to reduce the echoes used in fitting (optional, default is None)
- `save_data` (`bool, optional`) – Save output data to file (default is False)
- `overwrite` (`bool, optional`) – Overwrite existing results (default is False)
- `output_dir` (`str, optional`) – Path to desired output directory, will be created if it doesn't exist
- `file_name` (`str, optional`) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- `t2s (niimg)`: Map of estimated T2* times (`_qt2fit-t2s`)
- `r2s (niimg)`: Map of estimated R2* relaxation rate (`_qt2fit-r2s`)
- `s0 (niimg)`: Estimated PD weighted image at TE=0 (`_qt2fit-s0`)
- `residuals (niimg)`: Estimated residuals between input and estimated echoes (`_qt2fit-err`)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin.

7.3 intensity_propagation

`nighres.intensity.intensity_propagation` (*image*, *mask=None*, *combine='mean'*, *distance_mm=5.0*, *target='zero'*, *scaling=1.0*, *save_data=False*, *overwrite=False*, *output_dir=None*, *file_name=None*)

Intensity Propagation

Propagates the values inside the mask (or non-zero) into the neighboring voxels

Parameters

- **image** (*niimg*) – Input image
- **mask** (*niimg*, *optional*) – Data mask to specify acceptable seeding regions
- **combine** (*{'min', 'mean', 'max'}*, *optional*) – Propagate using the mean (default), max or min data from neighboring voxels
- **distance_mm** (*float*, *optional*) – Distance for the propagation (note: this algorithm will be slow for large distances)
- **target** (*{'zero', 'mask', 'lower', 'higher'}*, *optional*) – Propagate into zero (default), masked out, lower or higher neighboring voxels
- **scaling** (*float*, *optional*) – Multiply the propagated values by a factor ≤ 1 (default is 1)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **result** (*niimg*): The propagated intensity image

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin.

7.4 lcat_denoising

`nighres.intensity.lcat_denoising` (*image_list*, *image_mask*, *phase_list=None*, *ngb_size=3*, *ngb_time=3*, *stdev_cutoff=1.05*, *min_dimension=0*, *max_dimension=-1*, *save_data=False*, *overwrite=False*, *output_dir=None*, *file_names=None*)

LCaT denoising

Denoise multi-contrast time series data with a local PCA-based method

Parameters

- **image_list** (*[niimg]*) – List of input 4D magnitude images to denoise
- **image_mask** (*niimg*) – 3D mask for the input images
- **phase_list** (*[niimg]*) – List of input 4D phase images to denoise (optional)
- **ngb_size** (*int, optional*) – Size of the local PCA neighborhood, to be increased with number of inputs (default is 3)
- **ngb_time** (*int, optional*) – Size of the time window to use (default is 3)
- **stdev_cutoff** (*float, optional*) – Factor of local noise level to remove PCA components. Higher values remove more components (default is 1.05)
- **min_dimension** (*int, optional*) – Minimum number of kept PCA components (default is 0)
- **max_dimension** (*int, optional*) – Maximum number of kept PCA components (default is -1 for all components)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_names** (*[str], optional*) – Desired base names for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **denoised** (*[niimg]*): The list of denoised input images (`_lcat_den`)
- **dimensions** (*[niimg]*): Map of the estimated local dimensions (`_lcat_dim`)
- **residuals** (*[niimg]*): Estimated residuals between input and denoised images (`_lcat_err`)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin. Algorithm inspired by¹ with a different approach to set the adaptive noise threshold and additional processing to handle the time series properties.

References

7.5 lcpc_a_denoising

```
nighres.intensity.lcpc_a_denoising(image_list, phase_list=None, ngb_size=4,
                                   stdev_cutoff=1.05, min_dimension=0, max_dimension=-1,
                                   unwrap=True, process_2d=False, use_rmt=False,
                                   save_data=False, overwrite=False, output_dir=None,
                                   file_names=None)
```

LCPCA denoising

¹ Manjon, Coupe, Concha, Buades, Collins, Robles (2013). Diffusion Weighted Image Denoising Using Overcomplete Local PCA doi:10.1371/journal.pone.0073021

Denoise multi-contrast data with a local complex-valued PCA-based method

Parameters

- **image_list** (*[niimg]*) – List of input images to denoise
- **phase_list** (*[niimg]*, *optional*) – List of input phase to denoise (order must match that of `image_list`)
- **ngb_size** (*int*, *optional*) – Size of the local PCA neighborhood, to be increased with number of inputs (default is 4)
- **stdev_cutoff** (*float*, *optional*) – Factor of local noise level to remove PCA components. Higher values remove more components (default is 1.05)
- **min_dimension** (*int*, *optional*) – Minimum number of kept PCA components (default is 0)
- **max_dimension** (*int*, *optional*) – Maximum number of kept PCA components (default is -1 for all components)
- **unwrap** (*bool*, *optional*) – Whether to unwrap the phase data of keep it as is, assuming radians (default is True)
- **process_2d** (*bool*, *optional*) – Whether to denoise in 2D, for instance when acquiring a thin slab of data (default is False)
- **use_rmt** (*bool*, *optional*) – Whether to use random matrix theory rather than noise fitting to estimate the noise threshold (default is False)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_names** (*[str]*, *optional*) – Desired base names for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- `denoised` (`[niimg]`): The list of denoised input images (`_lcpca_den`)
- `dimensions` (`[niimg]`): Map of the estimated local dimensions (`_lcpca_dim`)
- `residuals` (`[niimg]`): Estimated residuals between input and denoised images (`_lcpca_err`)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin. Algorithm adapted from¹ with a different approach to set the adaptive noise threshold and additional processing to handle the phase data.

¹ Manjon, Coupe, Concha, Buades, Collins, Robles (2013). Diffusion Weighted Image Denoising Using Overcomplete Local PCA doi:10.1371/journal.pone.0073021

References

7.6 mp2rage_t1_mapping

`nighres.intensity.mp2rage_t1_mapping` (*first_inversion*, *second_inversion*, *inversion_times*, *flip_angles*, *inversion_TR*, *excitation_TR*, *N_excitations*, *efficiency=0.96*, *correct_B1=False*, *B1_map=None*, *scale_phase=True*, *save_data=False*, *overwrite=False*, *output_dir=None*, *file_name=None*)

MP2RAGE T1 mapping

Estimate T1/R1 by a look-up table method adapted from¹

Parameters

- **first_inversion** (*[niimg]*) – List of {magnitude, phase} images for the first inversion
- **second_inversion** (*[niimg]*) – List of {magnitude, phase} images for the second inversion
- **inversion_times** (*[float]*) – List of {first, second} inversion times, in seconds
- **flip_angles** (*[float]*) – List of {first, second} flip angles, in degrees
- **inversion_TR** (*float*) – Inversion repetition time, in seconds
- **excitation_TR** (*[float]*) – List of {first,second} repetition times,in seconds
- **N_excitations** (*int*) – Number of excitations
- **efficiency** (*float*) – Inversion efficiency (default is 0.96)
- **correct_B1** (*bool*) – Whether to correct for B1 inhomogeneities (default is False)
- **B1_map** (*niimg*) – Computed B1 map
- **scale_phase** (*bool*) – Whether to rescale the phase image in [0,2PI] or to assume it is already in radians
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **t1** (*niimg*): Map of estimated T1 times (`_qt1map-t1`)
- **r1** (*niimg*): Map of estimated R1 relaxation rate (`_qt1map-r1`)
- **uni** (*niimg*): Estimated PD weighted image at TE=0 (`_qt1map-uni`)

Return type dict

¹ Marques, Kober, Krueger, van der Zwaag, Van de Moortele, Gruetter (2010) MP2RAGE, a self bias-field corrected sequence for improved segmentation and T1-mapping at high field. doi: 10.1016/j.neuroimage.2009.10.002.

Notes

Original Java module by Pierre-Louis Bazin.

References

7.7 mp2rageme_pd_mapping

`nighres.intensity.mp2rageme_pd_mapping` (*first_inversion*, *second_inversion*, *t1map*, *r2smap*, *echo_times*, *inversion_times*, *flip_angles*, *inversion_TR*, *excitation_TR*, *N_excitations*, *efficiency=0.96*, *b1map=None*, *save_data=False*, *overwrite=False*, *output_dir=None*, *file_name=None*)

MP2RAGEME PD mapping

Estimate PD maps from MP2RAGEME data, combining T1 and R2* estimates with the MPRAGE model of¹.

Parameters

- **first_inversion** (*[nimg]*) – List of {magnitude, phase} images for the first inversion
- **second_inversion** (*[nimg]*) – List of {magnitude, phase} images for the second inversion
- **t1map** (*nimg*) – Quantitative T1 map image, in milliseconds
- **r2smap** (*nimg*) – Quantitative R2* map image, in kHz
- **echo_times** (*[float]*) – List of {te1, te2, te3, te4, te5} echo times, in seconds
- **inversion_times** (*[float]*) – List of {first, second} inversion times, in seconds
- **flip_angles** (*[float]*) – List of {first, second} flip angles, in degrees
- **inversion_TR** (*float*) – Inversion repetition time, in seconds
- **excitation_TR** (*[float]*) – List of {first,second} repetition times,in seconds
- **N_excitations** (*int*) – Number of excitations
- **efficiency** (*float*) – Inversion efficiency (default is 0.96)
- **correct_B1** (*bool*) – Whether to correct for B1 inhomogeneities (default is False)
- **b1map** (*nimg*) – Computed B1 map (optional)
- **scale_phase** (*bool*) – Whether to rescale the phase image in [0,2PI] or to assume it is already in radians
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

¹ Marques, Kober, Krueger, van der Zwaag, Van de Moortele, Gruetter (2010) MP2RAGE, a self bias-field corrected sequence for improved segmentation and T1-mapping at high field. doi: 10.1016/j.neuroimage.2009.10.002.

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- `pd (niimg)`: Map of estimated proton density ratio (`_qpd-map`)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin.

References

7.8 phase_unwrapping

`nighres.intensity.phase_unwrapping` (*image*, *mask=None*, *nquadrants=3*, *tv_flattening=False*, *tv_scale=0.5*, *save_data=False*, *overwrite=False*, *output_dir=None*, *file_name=None*)

Fast marching phase unwrapping

Fast marching method for unwrapping phase images, based on `_[1]`

Parameters

- **image** (*niimg*) – Input phase image to unwrap
- **mask** (*niimg*, *optional*) – Data mask to specify acceptable seeding regions
- **nquadrants** (*int*, *optional*) – Number of image quadrants to use (default is 3)
- **tv_flattening** (*bool*, *optional*) – Whether or not to run a post-processing step to remove background phase variations with a total variation filter (default is False)
- **tv_scale** (*float*, *optional*) – Relative intensity scale for the TV filter (default is 0.5)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- `result (niimg)`: The unwrapped image rescaled in radians

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin. Algorithm adapted from¹ with additional seeding in multiple image quadrants to reduce the effects of possible phase singularities

References

7.9 t2s_optimal_combination

`nighres.intensity.t2s_optimal_combination` (*image_list*, *te_list*, *depth=None*, *save_data=False*, *overwrite=False*, *output_dir=None*, *file_name=None*)

T2*-Optimal Combination

Estimate T2*/R2* and use it to combine multi-echo data

Parameters

- **image_list** (*[niimg]*) – List of 4D input images, one per echo time
- **te_list** (*[float]*) – List of input echo times (TE)
- **depth** (*[int]*) – List of echo depth to keep for input time points
- **save_data** (*bool, optional*) – Save output data to file (default is False)
- **overwrite** (*bool, optional*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **combined** (niimg): Optimally combined 4D image
- **t2s** (niimg): Map of estimated T2* times (`_qt2fit-t2s`)
- **r2s** (niimg): Map of estimated R2* relaxation rate (`_qt2fit-r2s`)
- **s0** (niimg): Estimated PD weighted image at TE=0 (`_qt2fit-s0`)
- **residuals** (niimg): Estimated residuals between input and estimated echoes (`_qt2fit-err`)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin.

¹ Abdul-Rahman, Gdeisat, Burton and Lalor. Fast three-dimensional phase-unwrapping algorithm based on sorting by reliability following a non-continuous path. doi: 10.1117/12.611415

8.1 io_volume

`nighres.io.load_volume` (*volume*)

Load volumetric data into a [Nibabel SpatialImage](#)

Parameters `volume` (*niimg*) – Volumetric data to be loaded, can be a path to a file that nibabel can load, or a [Nibabel SpatialImage](#)

Returns `image`

Return type [Nibabel SpatialImage](#)

Notes

Originally created as part of [Laminar Python \[1\]](#) .

References

8.1.1 Examples using `nighres.io.load_volume`

- [DOTS white matter segmentation](#)
- [Multiatlas Segmentation](#)

`nighres.io.save_volume` (*filename, volume, dtype='float32', overwrite_file=True*)

Save volumetric data that is a [Nibabel SpatialImage](#) to a file

Parameters

- **filename** (*str*) – Full path and filename under which volume should be saved. The extension determines the file format (must be supported by [Nibabel](#))
- **volume** (*Nibabel SpatialImage*) – Volumetric data to be saved

- **dtype**(*str*, *optional*) – Datatype in which volumetric data should be stored (default is float32)
- **overwrite_file**(*bool*, *optional*) – Overwrite existing files (default is True)

Notes

Originally created as part of Laminar Python [1]_ .

References

8.1.2 Examples using `nighres.io.save_volume`

- *DOTS white matter segmentation*

8.2 `io_mesh`

`nighres.io.load_mesh_geometry` (*surf_mesh*)

Load a mesh geometry into a dictionary with entries “points” and “faces”

Parameters **surf_mesh** – Mesh geometry to be loaded, can be a path to a file (currently supported formats are freesurfer geometry formats, gii and ASCII-coded vtk, ply or obj) or a dictionary with the keys “points” and “faces”

Returns Dictionary with a numpy array with key “points” for a Numpy array of the x-y-z coordinates of the mesh vertices and key “faces” for a Numpy array of the the indices (into points) of the mesh faces

Return type `dict`

Notes

Originally created as part of Laminar Python [1]_

References

`nighres.io.save_mesh_geometry` (*filename*, *surf_dict*)

Saves surface mesh geometry to file

Parameters

- **filename** (*str*) – Full path and filename under which surfaces data should be saved. The extension determines the file format. Currently supported are freesurfer geometry formats, gii and ASCII-coded vtk, obj, ply’
- **surf_dict** (*dict*) – Surface mesh geometry to be saved. Dictionary with a numpy array with key “points” for a Numpy array of the x-y-z coordinates of the mesh vertices and key “faces2 for a Numpy array of the the indices (into points) of the mesh faces

Notes

Originally created as part of Laminar Python [1]_

References

`nighres.io.load_mesh_data` (*surf_data*, *gii_darray=None*)

Loads mesh data into a Numpy array

Parameters

- **surf_data** – Mesh data to be loaded, can be a Numpy array or a path to a file. Currently supported formats are freesurfer data formats (mgz, curv, sulc, thickness, annot, label), nii, gii, ASCII-coded vtk and txt
- **gii_darray** (*int*, *optional*) – Index of gii data array to load (default is to load all)

Returns Numpy array containing the data

Return type np.ndarray

Notes

Originally created as part of Laminar Python [\[1\]](#)

References

`nighres.io.save_mesh_data` (*filename*, *surf_data*)

Saves surface data that is a Numpy array to file

Parameters

- **filename** (*str*) – Full path and filename under which surfaces data should be saved. The extension determines the file format. Currently supported are freesurfer formats curv, thickness, sulc and ASCII-coded txt
- **surf_data** (*np.ndarray*) – Surface data to be saved

Notes

Originally created as part of Laminar Python [\[1\]](#)

References

9.1 laminar_iterative_smoothing

`nighres.laminar.laminar_iterative_smoothing` (*profile_surface_image*, *intensity_image*,
fwhm_mm, *roi_mask_image=None*,
save_data=False, *overwrite=False*, *output_dir=None*, *file_name=None*)

Smoothing data on multiple intracortical layers

Parameters

- **data_image** (*niimg*) – Image from which data should be sampled
- **profile_surface_image** (*niimg*) – 4D image containing levelset representations of different intracortical surfaces on which data should be sampled
- **fwhm_mm** (*float*) – Full width half maximum distance to use in smoothing (in mm)
- **roi_mask_image** (*niimg*, *optional*) – Mask image defining a region of interest to restrict the smoothing
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **result** (*niimg*): smoothed intensity image (*_lis-smooth*)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin

Important: this method assumes isotropic voxels

9.2 laminar_regional_approximation

`nighres.laminar.laminar_regional_approximation` (*profile_surface_image*, *intensity_image*, *roi_image*, *save_data=False*, *overwrite=False*, *output_dir=None*, *file_name=None*)

Sampling data on multiple intracortical layers

Parameters

- **profile_surface_image** (*niimg*) – 4D image containing levelset representations of different intracortical surfaces on which data should be sampled
- **intensity_image** (*niimg*) – Image from which data should be sampled
- **roi_image** (*niimg*) – Label image of the region to sample
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **weights** (*niimg*): weight image, representing the weighting of profiles in the estimation (`_lra-weight`)
- **degree** (*niimg*): degree image, representing the degree of profiles in the estimation (`_lra-deg`)
- **residuals** (*niimg*): residuals image, representing the residual error in the estimation (`_lra-res`)
- **median** (*[float]*): the estimated median profile (`_lra-med`)
- **perc25** (*[float]*): the estimated 25 percentile profile (`_lra-p25`)
- **perc75** (*[float]*): the estimated 75 percentile profile (`_lra-p75`)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin

9.3 profile_averaging

`nighres.laminar.profile_averaging` (*profile_surface_image*, *intensity_image*, *roi_image*,
save_data=False, *overwrite=False*, *output_dir=None*,
file_name=None)

Sampling data on multiple intracortical layers

Parameters

- **profile_surface_image** (*niimg*) – 4D image containing levelset representations of different intracortical surfaces on which data should be sampled
- **intensity_image** (*niimg*) – Image from which data should be sampled
- **roi_image** (*niimg*) – Label image of the region to sample
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **weights** (*niimg*): weight image, representing the weighting of profiles in the estimation (*_lpa-weight*)
- **best** (*[float]*): the estimated best profile
- **median** (*[float]*): the estimated median profile
- **iqr** (*[float]*): the estimated IQR profile

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin

9.4 profile_meshing

`nighres.laminar.profile_meshing` (*profile_surface_image*, *starting_surface_mesh*,
save_data=False, *overwrite=False*, *output_dir=None*,
file_name=None)

Profile meshing

Creates a point-by-point matched set of 3D meshes, one for each of layer of the layer boundary surface, with the mesh topology of the starting mesh. The starting mesh should be inside the cortex, for instance the mid-layer surface from volumetric layering.

Parameters

- **profile_surface_image** (*niimg*) – 4D image containing levelset representations of different intracortical surfaces on which data should be sampled

- **starting_surface_mesh** (*mesh*) – Mesh model of the surface
- **save_data** (*bool, optional*) – Save output data to file (default is False)
- **overwrite** (*bool, optional*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **profile** ([*mesh*]): Collection of surface mesh dictionary of “points” and “faces” (*_mesh-p#*)

Return type `dict`

Notes

Ported from original Java module by Pierre-Louis Bazin

9.5 profile_sampling

`nighres.laminar.profile_sampling` (*profile_surface_image, intensity_image, save_data=False, overwrite=False, output_dir=None, file_name=None*)

Sampling data on multiple intracortical layers

Parameters

- **profile_surface_image** (*niimg*) – 4D image containing levelset representations of different intracortical surfaces on which data should be sampled
- **intensity_image** (*niimg*) – Image from which data should be sampled
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **result** (*niimg*): 4D profile image, where the 4th dimension represents the profile for each voxel (*_lps-data*)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin and Juliane Dinse

9.6 volumetric_layering

`nighres.laminar.volumetric_layering` (*inner_levelset*, *outer_levelset*, *n_layers=4*, *topology_lut_dir=None*, *method='volume-preserving'*, *layer_dir='outward'*, *curv_scale=3*, *save_data=False*, *overwrite=False*, *output_dir=None*, *file_name=None*)

Equivolometric layering of the cortical sheet.

Parameters

- **inner_levelset** (*niimg*) – Levelset representation of the inner surface, typically GM/WM surface
- **outer_levelset** (*niimg*) – Levelset representation of the outer surface, typically GM/CSF surface
- **n_layers** (*int*, *optional*) – Number of layers to be created (default is 10)
- **topology_lut_dir** (*str*, *optional*) – Path to directory in which topology files are stored (default is stored in TOPOLOGY_LUT_DIR)
- **method** (*str*) – Which model to use, either “volume-preserving” or “distance-preserving” (default is “volume-preserving”)
- **layer_dir** (*str*) – Direction to perform the layering, either “outward” or “inward” (default is “outward”)
- **curv_scale** (*int*) – Scale of the curvature approximation window in voxels (default is 3, computation may become very slow for higher values)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **depth** (*niimg*): Continuous depth from 0 (inner surface) to 1 (outer surface) (`_layering-depth`)
- **layers** (*niimg*): Discrete layers from 1 (bordering inner surface) to `n_layers` (bordering outer surface) (`_layering-layers`)
- **boundaries** (*niimg*): Levelset representations of boundaries between all layers in 4D (`_layering-boundaries`)

Return type `dict`

Notes

Original Java module by Miriam Waehnert, Pierre-Louis Bazin and Juliane Dinse. Algorithm details can be found in¹

¹ Waehnert et al (2014) Anatomically motivated modeling of cortical laminae. DOI: 10.1016/j.neuroimage.2013.03.078

References

9.6.1 Examples using `nighres.laminar.volumetric_layering`

- *Cortical depth estimation from MGDM segmentation*

10.1 mgdm_cells

```
nighres.microscopy.mgdm_cells(contrast_image1, contrast_type1, contrast_image2=None,
                              contrast_type2=None, contrast_image3=None, contrast_type3=None,
                              stack_dimension='2D', force_weight=0.6, curvature_weight=0.3,
                              cell_threshold=0.1, max_iterations=200, min_change=0.0001,
                              topology='wcs', topology_lut_dir=None, save_data=False,
                              overwrite=False, output_dir=None, file_name=None)
```

MGDM cell segmentation

Estimates cell structures using a Multiple Object Geometric Deformable Model (MGDM)

Parameters

- **contrast_image1** (*niimg*) – First input image to perform segmentation on
- **contrast_type1** (*{"centroid-proba", "local-maxima", "foreground-proba",}*) – “image-intensities”} Contrast type of first input image
- **contrast_image2** (*niimg, optional*) – Additional input image to inform segmentation, must be in the same space as `contrast_image1`, requires `contrast_type2`
- **contrast_type2** (*str, {"centroid-proba", "local-maxima", "foreground-proba",}*) – “image-intensities”}, optional Contrast type of second input image
- **contrast_image3** (*niimg, optional*) – Additional input image to inform segmentation, must be in the same space as `contrast_image1`, requires `contrast_type3`
- **contrast_type3** (*{"centroid-proba", "local-maxima", "foreground-proba",}*) – “image-intensities”}, optional Contrast type of third input image
- **stack_dimension** (*{'2D', '3D'}, optional*) – Dimension of the data for processing, either a stack of independent 2D slices or a fully 3D stack

- **max_iterations** (*int, optional*) – Maximum number of iterations per step for MGDM (default is 800, set to 1 for quick testing of registration of priors, which does not perform true segmentation)
- **min_change** (*float, optional*) – Minimum amount of change in the segmentation for MGDM to stop (default is 0.001)
- **force_weight** (*float, optional*) – Forces to drive MGDM to cell boundaries (default is 0.5)
- **curvature_weight** (*float, optional*) – Curvature regularization forces (default is 0.1)
- **cell_threshold** (*float, optional*) – Ratio of lower intensities from the local maximum to be included in a given cell (default is 0.1)
- **topology** (*{'wcs', 'no'}, optional*) – Topology setting, choose 'wcs' (well-composed surfaces) for strongest topology constraint, 'no' for no topology constraint (default is 'wcs')
- **topology_lut_dir** (*str, optional*) – Path to directory in which topology files are stored (default is stored in TOPOLOGY_LUT_DIR)
- **save_data** (*bool, optional*) – Save output data to file (default is False)
- **overwrite** (*bool, optional*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- segmentation (niimg): Hard brain segmentation with topological constraints (if chosen) (_mgdmc-seg)
- distance (niimg): Minimum distance to a segmentation boundary (_mgdmc-dist)

Return type dict

Notes

Original Java module by Pierre-Louis Bazin. Algorithm details can be found in¹ and²

References

10.2 stack_intensity_regularisation

nighres.microscopy.**stack_intensity_regularisation** (*image, ratio=50, save_data=False, overwrite=False, output_dir=None, file_name=None*)

Stack intensity regularisation

¹ Bogovic, Prince and Bazin (2013). A multiple object geometric deformable model for image segmentation. doi:10.1016/j.cviu.2012.10.006.A

² Fan, Bazin and Prince (2008). A multi-compartment segmentation framework with homeomorphic level sets. DOI: 10.1109/CVPR.2008.4587475

Estimates an image-to-image linear intensity scaling for a stack of 2D images

Parameters

- **image** (*niimg*) – Input 2D images, stacked in the Z dimension
- **ratio** (*float, optional*) – Ratio of image differences to keep (default is 50%)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **result** (*niimg*): The intensity regularised input

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin.

11.1 apply_coordinate_mappings

```
nighres.registration.apply_coordinate_mappings (image, mapping1, mapping2=None, mapping3=None, mapping4=None, interpolation='nearest', padding='closest', save_data=False, overwrite=False, output_dir=None, file_name=None)
```

Apply a coordinate mapping (or a succession of coordinate mappings) to a 3D or 4D image.

Parameters

- **image** (*niimg*) – Image to deform
- **mapping1** (*niimg*) – First coordinate mapping to apply
- **mapping2** (*niimg*, *optional*) – Second coordinate mapping to apply
- **mapping3** (*niimg*, *optional*) – Third coordinate mapping to apply
- **mapping4** (*niimg*, *optional*) – Fourth coordinate mapping to apply
- **interpolation** (*{'nearest', 'linear'}*) – Interpolation method (default is 'nearest')
- **padding** (*{'closest', 'zero', 'max'}*) – Image padding method (default is 'closest')
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- `result (niimg)`: Result image (`_def-img`)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin

11.1.1 Examples using `nighres.registration.apply_coordinate_mappings`

- *Brain co-registration from MP2RAGE data*
- *Multiatlas Segmentation*

11.2 `apply_coordinate_mappings_2d`

`nighres.registration.apply_coordinate_mappings_2d` (*image*, *mapping1*, *mapping2=None*, *mapping3=None*, *mapping4=None*, *interpolation='nearest'*, *padding='closest'*, *save_data=False*, *overwrite=False*, *output_dir=None*, *file_name=None*)

Apply a 2D coordinate mapping (or a succession of coordinate mappings) to a 2D or 3D image.

Parameters

- **image** (*niimg*) – Image to deform
- **mapping1** (*niimg*) – First coordinate mapping to apply
- **mapping2** (*niimg*, *optional*) – Second coordinate mapping to apply
- **mapping3** (*niimg*, *optional*) – Third coordinate mapping to apply
- **mapping4** (*niimg*, *optional*) – Fourth coordinate mapping to apply
- **interpolation** (`{'nearest', 'linear'}`) – Interpolation method (default is 'nearest')
- **padding** (`{'closest', 'zero', 'max'}`) – Image padding method (default is 'closest')
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- `result (niimg)`: Result image (`_def-img`)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin

11.3 embedded_antsreg

```
nighres.registration.embedded_antsreg(source_image, target_image,
                                       run_rigid=False, rigid_iterations=1000,
                                       run_affine=False, affine_iterations=1000,
                                       run_syn=True, coarse_iterations=40,
                                       medium_iterations=50, fine_iterations=40,
                                       cost_function='MutualInformation', interpolation='NearestNeighbor',
                                       regularization='High', convergence=1e-06,
                                       mask_zero=False, ignore_affine=False,
                                       ignore_header=False, save_data=False,
                                       overwrite=False, output_dir=None,
                                       file_name=None)
```

Embedded ANTS Registration

Runs the rigid and/or Symmetric Normalization (SyN) algorithm of ANTs and formats the output deformations into voxel coordinate mappings as used in CBSTools registration and transformation routines.

Parameters

- **`source_image`** (*niimg*) – Image to register
- **`target_image`** (*niimg*) – Reference image to match
- **`run_rigid`** (*bool*) – Whether or not to run a rigid registration first (default is False)
- **`rigid_iterations`** (*float*) – Number of iterations in the rigid step (default is 1000)
- **`run_affine`** (*bool*) – Whether or not to run an affine registration first (default is False)
- **`affine_iterations`** (*float*) – Number of iterations in the affine step (default is 1000)
- **`run_syn`** (*bool*) – Whether or not to run a SyN registration (default is True)
- **`coarse_iterations`** (*float*) – Number of iterations at the coarse level (default is 40)
- **`medium_iterations`** (*float*) – Number of iterations at the medium level (default is 50)
- **`fine_iterations`** (*float*) – Number of iterations at the fine level (default is 40)
- **`cost_function`** (*{'CrossCorrelation', 'MutualInformation'}*) – Cost function for the registration (default is 'MutualInformation')
- **`interpolation`** (*{'NearestNeighbor', 'Linear'}*) – Interpolation for the registration result (default is 'NearestNeighbor')
- **`regularization`** (*{'Low', 'Medium', 'High'}*) – Regularization preset for the SyN deformation (default is 'Medium')
- **`convergence`** (*float*) – Threshold for convergence, can make the algorithm very slow (default is convergence)

- **mask_zero** (*bool*) – Mask regions with zero value (default is False)
- **ignore_affine** (*bool*) – Ignore the affine matrix information extracted from the image header (default is False)
- **ignore_header** (*bool*) – Ignore the orientation information and affine matrix information extracted from the image header (default is False)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **transformed_source** (niimg): Deformed source image (`_ants-def`)
- **mapping** (niimg): Coordinate mapping from source to target (`_ants-map`)
- **inverse** (niimg): Inverse coordinate mapping from target to source (`_ants-invmap`)

Return type `dict`

Notes

Port of the CBSTools Java module by Pierre-Louis Bazin. The main algorithm is part of the ANTs software by Brian Avants and colleagues¹. The interfacing with ANTs is performed through Nipype². Parameters have been set to values commonly found in neuroimaging scripts online, but not necessarily optimal.

References

11.3.1 Examples using `nighres.registration.embedded_antsreg`

- *Subcortex Parcellation*
- *Brain co-registration from MP2RAGE data*
- *Multiatlas Segmentation*

¹ Avants et al (2008), Symmetric diffeomorphic image registration with cross-correlation: evaluating automated labeling of elderly and neurodegenerative brain, *Med Image Anal.* 12(1):26-41

² Gorgolewski et al (2011) Nipype: a flexible, lightweight and extensible neuroimaging data processing framework in python. *Front Neuroinform* 5. doi:10.3389/fninf.2011.00013

11.4 embedded_antsreg_2d

```
nighres.registration.embedded_antsreg_2d(source_image, target_image, run_rigid=False,
                                         rigid_iterations=1000,      run_affine=False,
                                         affine_iterations=1000,      run_syn=True,
                                         coarse_iterations=40,      medium_iterations=50,
                                         fine_iterations=40,         scaling_factor=32,
                                         cost_function='MutualInformation', interpolation='NearestNeighbor',
                                         regularization='High',
                                         convergence=1e-06,      mask_zero=False, ignore_affine=False,
                                         ignore_header=False,
                                         save_data=False,      overwrite=False,      output_dir=None,
                                         file_name=None)
```

Embedded ANTS Registration 2D

Runs the rigid and/or Symmetric Normalization (SyN) algorithm of ANTs and formats the output deformations into voxel coordinate mappings as used in CBSTools registration and transformation routines.

Parameters

- **source_image** (*nimg*) – Image to register
- **target_image** (*nimg*) – Reference image to match
- **run_rigid** (*bool*) – Whether or not to run a rigid registration first (default is False)
- **rigid_iterations** (*float*) – Number of iterations in the rigid step (default is 1000)
- **run_affine** (*bool*) – Whether or not to run an affine registration first (default is False)
- **affine_iterations** (*float*) – Number of iterations in the affine step (default is 1000)
- **run_syn** (*bool*) – Whether or not to run a SyN registration (default is True)
- **coarse_iterations** (*float*) – Number of iterations at the coarse level (default is 40)
- **medium_iterations** (*float*) – Number of iterations at the medium level (default is 50)
- **fine_iterations** (*float*) – Number of iterations at the fine level (default is 40)
- **cost_function** (*{'CrossCorrelation', 'MutualInformation'}*) – Cost function for the registration (default is 'MutualInformation')
- **interpolation** (*{'NearestNeighbor', 'Linear'}*) – Interpolation for the registration result (default is 'NearestNeighbor')
- **convergence** (*float*) – Threshold for convergence, can make the algorithm very slow (default is convergence)
- **ignore_affine** (*bool*) – Ignore the affine matrix information extracted from the image header (default is False)
- **ignore_header** (*bool*) – Ignore the orientation information and affine matrix information extracted from the image header (default is False)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist

- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **transformed_source** (*niimg*): Deformed source image (*_ants-def*)
- **mapping** (*niimg*): Coordinate mapping from source to target (*_ants-map*)
- **inverse** (*niimg*): Inverse coordinate mapping from target to source (*_ants-invmap*)

Return type `dict`

Notes

Port of the CBSTools Java module by Pierre-Louis Bazin. The main algorithm is part of the ANTs software by Brian Avants and colleagues¹. Parameters have been set to values commonly found in neuroimaging scripts online, but not necessarily optimal.

References

11.5 embedded_antsreg_multi

```
nighres.registration.embedded_antsreg_multi(source_images, target_images,  
                                           run_rigid=True, rigid_iterations=1000,  
                                           run_affine=False, affine_iterations=1000,  
                                           run_syn=True, coarse_iterations=40,  
                                           medium_iterations=50, fine_iterations=40,  
                                           cost_function='MutualInformation', interpolation='NearestNeighbor',  
                                           regularization='High', convergence=1e-06,  
                                           mask_zero=False, ignore_affine=False,  
                                           ignore_header=False, save_data=False,  
                                           overwrite=False, output_dir=None,  
                                           file_name=None)
```

Embedded ANTS Registration Multi-contrasts

Runs the rigid and/or Symmetric Normalization (SyN) algorithm of ANTs and formats the output deformations into voxel coordinate mappings as used in CBSTools registration and transformation routines. Uses all input contrasts with equal weights.

Parameters

- **source_images** (*[niimg]*) – Image list to register
- **target_images** (*[niimg]*) – Reference image list to match
- **run_rigid** (*bool*) – Whether or not to run a rigid registration first (default is False)
- **rigid_iterations** (*float*) – Number of iterations in the rigid step (default is 1000)
- **run_affine** (*bool*) – Whether or not to run a affine registration first (default is False)
- **affine_iterations** (*float*) – Number of iterations in the affine step (default is 1000)

¹ Avants et al (2008), Symmetric diffeomorphic image registration with cross-correlation: evaluating automated labeling of elderly and neurodegenerative brain, Med Image Anal. 12(1):26-41

- **run_syn** (*bool*) – Whether or not to run a SyN registration (default is True)
- **coarse_iterations** (*float*) – Number of iterations at the coarse level (default is 40)
- **medium_iterations** (*float*) – Number of iterations at the medium level (default is 50)
- **fine_iterations** (*float*) – Number of iterations at the fine level (default is 40)
- **cost_function** (*{'CrossCorrelation', 'MutualInformation'}*) – Cost function for the registration (default is 'MutualInformation')
- **interpolation** (*{'NearestNeighbor', 'Linear'}*) – Interpolation for the registration result (default is 'NearestNeighbor')
- **regularization** (*{'Low', 'Medium', 'High'}*) – Regularization preset for the SyN deformation (default is 'Medium')
- **convergence** (*float*) – Threshold for convergence, can make the algorithm very slow (default is convergence)
- **mask_zero** (*bool*) – Mask regions with zero value (default is False)
- **ignore_affine** (*bool*) – Ignore the affine matrix information extracted from the image header (default is False)
- **ignore_header** (*bool*) – Ignore the orientation information and affine matrix information extracted from the image header (default is False)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **transformed_source** ([*niimg*]): Deformed source image list (*_ants_def0,1,...*)
- **mapping** (*niimg*): Coordinate mapping from source to target (*_ants_map*)
- **inverse** (*niimg*): Inverse coordinate mapping from target to source (*_ants_invmap*)

Return type `dict`

Notes

Port of the CBSTools Java module by Pierre-Louis Bazin. The main algorithm is part of the ANTs software by Brian Avants and colleagues¹. Parameters have been set to values commonly found in neuroimaging scripts online, but not necessarily optimal.

¹ Avants et al (2008), Symmetric diffeomorphic image registration with cross-correlation: evaluating automated labeling of elderly and neurodegenerative brain, *Med Image Anal.* 12(1):26-41

References

11.5.1 Examples using `nighres.registration.embedded_antsreg_multi`

- *Subcortex Parcellation*

11.6 `generate_coordinate_mapping`

`nighres.registration.generate_coordinate_mapping` (*reference_image*,
source_image=None, *transform_matrix=None*, *invert_matrix=False*, *save_data=False*,
overwrite=False, *output_dir=None*,
file_name=None)

Generate coordinate mapping

Generate a coordinate mapping for the image(s) and linear transformation as used in CBSTools registration and transformation routines.

Parameters

- **reference_image** (*niimg*) – Image to generate a coordinate mapping from, listing its X,Y,Z coordinates
- **source_image** (*niimg*, *optional*) – In case the mapping is from a source and target in different coordinate spaces, this image represents the source space
- **transform_matrix** (*niimg*, *optional*) – Whether to use a MIPAV formatted transformation matrix to define the mapping
- **invert_matrix** (*bool*) – Whether or not to invert the transformation, if given
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **result** (*niimg*): Coordinate mapping of the reference image (`_coord-map`)

Return type `dict`

Notes

Port of the CBSTools Java module by Pierre-Louis Bazin. Currently the transformation matrix follows the MIPAV conventions.

11.7 simple_align

```
nighres.registration.simple_align(source_image, target_image, copy_header=False,
                                  align_center=False, rescale=False, data_type='intensity',
                                  ignore_affine=False, ignore_header=False,
                                  save_data=False, overwrite=False, output_dir=None,
                                  file_name=None)
```

Simple alignment routines

Simple routines to align image headers (image data is unchanged)

Parameters

- **source_image** (*niimg*) – Image to align
- **target_image** (*niimg*) – Reference image to match
- **copy_header** (*bool*) – To copy the target header to the source (default is False)
- **align_center** (*bool*) – To align the source center of mass to the target (default is False)
- **rescale** (*bool*) – To rescale the source to the volume of the target (default is False)
- **data_type** (*{'intensity', 'nonzero', 'boundingbox'}*) – The type of data to consider for alignment (default is 'intensity')
- **ignore_affine** (*bool*) – Ignore the affine matrix information extracted from the image header (default is False)
- **ignore_header** (*bool*) – Ignore the orientation information and affine matrix information extracted from the image header (default is False)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **result** (*niimg*): Aligned source image (*_al-*img**)

Return type `dict`

Notes

This uses the ANTs/ITK conventions with regard to Nifti headers, for better or for worse. Note that Nibabel conventions, of course, are different.

12.1 conditional_shape

```
nighres.segmentation.conditional_shape(target_images, structures, contrasts, back-
ground=1, shape_atlas_probab=None,
shape_atlas_labels=None, inten-
sity_atlas_hist=None, skeleton_atlas_probab=None,
skeleton_atlas_labels=None, map_to_atlas=None,
map_to_target=None, max_iterations=80,
max_difference=0.1, ngb_size=4, save_data=False,
overwrite=False, output_dir=None,
file_name=None)
```

Conditionanl Shape Parcellation

Estimates subcortical structures based on a multi-atlas approach on shape

Parameters

- **target_images** (*[nimg]*) – Input images to perform the parcellation from
- **structures** (*int*) – Number of structures to parcellate
- **contrasts** (*int*) – Number of image intensity contrasts
- **background** (*int*) – Number of background tissue classes (default is 1)
- **shape_atlas_probab** (*nimg*) – Pre-computed shape atlas from the shape levelsets (replacing them)
- **shape_atlas_labels** (*nimg*) – Pre-computed shape atlas from the shape levelsets (replacing them)
- **intensity_atlas_hist** (*nimg*) – Pre-computed intensity atlas from the contrast images (replacing them)
- **skeleton_atlas_probab** (*nimg*) – Pre-computed skeleton atlas from the shape levelsets (replacing them)

- **skeleton_atlas_labels** (*niimg*) – Pre-computed skeleton atlas from the shape levelsets (replacing them)
- **map_to_atlas** (*niimg*) – Coordinate mapping from the target to the atlas (opt)
- **map_to_target** (*niimg*) – Coordinate mapping from the atlas to the target (opt)
- **max_iterations** (*int*) – Maximum number of diffusion iterations to perform
- **max_difference** (*float*) – Maximum difference between diffusion steps
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **max_spatial_proba** (*niimg*): Maximum spatial probability map (`_cspmax-sproba`)
- **max_spatial_label** (*niimg*): Maximum spatial probability labels (`_cspmax-slabel`)
- **max_combined_proba** (*niimg*): Maximum spatial and intensity combined probability map (`_cspmax-cproba`)
- **max_combined_label** (*niimg*): Maximum spatial and intensity combined probability labels (`_cspmax-clabel`)
- **max_proba** (*niimg*): Maximum probability map (`_cspmax-proba`)
- **max_label** (*niimg*): Maximum probability labels (`_cspmax-label`)
- **neighbors** (*niimg*): Local neighborhood maps (`_cspmax-ngb`)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin.

12.2 fuzzy_cmeans

```
nighres.segmentation.fuzzy_cmeans(image, clusters=3, max_iterations=50,  
                                   max_difference=0.01, smoothing=0.1, fuzziness=2.0,  
                                   mask_zero=True, save_data=False, overwrite=False,  
                                   output_dir=None, file_name=None)
```

Fuzzy C-means image segmentation

Estimates intensity clusters with spatial smoothness and partial voluming. Based on the RFCM algorithm of (Pham, 2001).

Parameters

- **image** (*niimg*) – Input image to segment
- **clusters** (*int*) – Number of clusters to estimate (default is 3)

- **max_iterations** (*int*) – Maximum number of iterations to perform (default is 50)
- **max_difference** (*float*) – Maximum difference between steps for stopping (default is 0.01)
- **smoothing** (*float*) – Ratio of spatial smoothness to impose on the clusters (default is 0.1)
- **fuzziness** (*float*) – Scaling of the C-means measure, in [1.0 - 3.0] (default is 2.0)
- **mask_zero** (*bool*) – Whether to ignore zero values (default is true)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **memberships** [niimg]: List of membership functions for each cluster in [0,1] (`_rfcm-mem#cluster`)
- **classification** (niimg): Hard classification of most likely cluster per voxel (`_rfcm-class`)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin.

13.1 levelset_fusion

`nighres.shape.levelset_fusion` (*levelset_images*, *correct_topology=True*, *topology_lut_dir=None*,
save_data=False, *overwrite=False*, *output_dir=None*,
file_name=None)

Levelset fusion

Creates an average levelset surface representations from a collection of levelset surfaces, with same average volume and (optionally) spherical topology

Parameters

- **levelset_images** (*niimg*) – List of levelset images to combine.
- **correct_topology** (*bool*, *optional*) – Corrects the average shape to ensure correct topology (default is True)
- **topology_lut_dir** (*str*, *optional*) – Path to directory in which topology files are stored (default is stored in TOPOLOGY_LUT_DIR)
- **save_data** (*bool*, *optional*) – Save output data to file (default is False)
- **overwrite** (*bool*, *optional*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **result** (*niimg*): Levelset representation of combined surface (*_lsf-avg*)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin

13.1.1 Examples using `nighres.shape.levelset_fusion`

- *Multiatlas Segmentation*

13.2 simple_skeleton

```
nighres.shape.simple_skeleton(input_image, shape_image_type='signed_distance',
                              boundary_threshold=0.0, skeleton_threshold=2.0, topology_lut_dir=None, save_data=False, overwrite=False, output_dir=None, file_name=None)
```

Simple Skeleton

Create a skeleton for a levelset surface or a probability map (loosely adapted from Bouix et al., 2006)

Parameters

- **input_image** (*niimg*) – Image containing structure-of-interest
- **shape_image_type** (*str*) – Shape of the input image: either ‘signed_distance’ or ‘probability_map’.
- **boundary_threshold** (*float*) – Boundary threshold (>0: inside, <0: outside)
- **skeleton_threshold** (*float*) – Skeleton threshold (>0: inside, <0: outside)
- **topology_lut_dir** (*str*) – Directory of LUT topology
- **save_data** (*bool, optional*) – Save output data to file (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn’t exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- medial (niimg): A 2D medial surface extracted from the shape (_ssk-med)
- skeleton (niimg): The 1D skeleton extracted from the shape (_ssk-skel)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin.

13.3 topology_correction

```
nighres.shape.topology_correction(image, shape_type, connectivity='wcs',
                                  propagation='object->background',
                                  minimum_distance=1e-05, topology_lut_dir=None,
                                  save_data=False, overwrite=False, output_dir=None,
                                  file_name=None)
```

Topology correction

Corrects the topology of a binary image, a probability map or a levelset surface to spherical with a fast marching technique¹.

Parameters

- **image** (*niimg*) – Image representing the shape of interest
- **shape_type** (*{'binary_object', 'probability_map', 'signed_distance_function'}*) – Which type of image is used as input
- **connectivity** (*{'wcs', '6/18', '6/26', '18/6', '26/6'}*) – What connectivity type to use (default is wcs: well-composed surfaces)
- **propagation** (*{'object->background', 'background->object'}*) – Which direction to use to enforce topology changes (default is 'object->background')
- **minimum_distance** (*float, optional*) – Minimum distance to impose between successive voxels (default is 1e-5)
- **topology_lut_dir** (*str, optional*) – Path to directory in which topology files are stored (default is stored in TOPOLOGY_LUT_DIR)
- **save_data** (*bool, optional*) – Save output data to file (default is False)
- **overwrite** (*bool, optional*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **corrected** (*niimg*): Corrected image (output file suffix `_tpc-img`)
- **object** (*niimg*): Corrected binary object (output file suffix `_tpc-obj`)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin

References

¹ Bazin and Pham (2007). Topology correction of segmented medical images using a fast marching algorithm doi:10.1016/j.cmpb.2007.08.006

14.1 segmentation_statistics

```
nighres.statistics.segmentation_statistics (segmentation, intensity=None, template=None, statistics=None, output_csv=None, atlas=None, skip_first=True, ignore_zero=True, save_data=False, overwrite=False, output_dir=None, file_name=None)
```

Segmentation Statistics

Compute various statistics of image segmentations

Parameters

- **segmentation** (*niimg*) – Input segmentation image
- **intensity** (*niimg*, *optional*) – Input intensity image for intensity-based statistics
- **template** (*niimg*, *optional*) – Input template segmentation for comparisons
- **statistics** (*[str]*) – Statistics to compute. Available options include: “Voxels”, “Volume”, “Center_of_mass”, “Mean_intensity”, “Std_intensity”, “10_intensity”, “25_intensity”, “50_intensity”, “75_intensity”, “90_intensity”, “Median_intensity”, “IQR_intensity”, “SNR_intensity”, “rSNR_intensity”, “Volumes”, “Dice_overlap”, “Jaccard_overlap”, “Volume_difference”, “False_positives”, “False_negatives”, “Dilated_Dice_overlap”, “Dilated_false_positive”, “Dilated_false_negative”, “Dilated_false_negative_volume”, “Dilated_false_positive_volume”, “Center_distance”, “Detected_clusters”, “False_detections”, “Cluster_numbers”, “Mean_cluster_sizes”, “Cluster_maps”, “Average_surface_distance”, “Average_surface_difference”, “Average_squared_surface_distance”, “Hausdorff_distance”
- **output_csv** (*str*) – File name of the statistics file to generate or expand
- **atlas** (*str*, *optional*) – File name of an atlas file defining the segmentation labels

- **skip_first** (*bool, optional*) – Whether to skip the first segmentation label (usually representing the background, default is True)
- **ignore_zero** (*bool, optional*) – Whether to ignore zero intensity values in the intensity image (default is True)
- **save_data** (*bool, optional*) – Save output data to file (default is False)
- **overwrite** (*bool, optional*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **csv** (str): The csv statistics file
- **map** (niimg): Map of the estimated statistic, if relevant (stat-map)

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin.

15.1 levelset_to_mesh

`nighres.surface.levelset_to_mesh` (*levelset_image*, *connectivity*='18/6', *level*=0.0, *inclusive*=True, *save_data*=False, *overwrite*=False, *output_dir*=None, *file_name*=None)

Levelset to mesh

Creates a triangulated mesh from the distance to a levelset surface representation using a connectivity-consistent marching cube algorithm.

Parameters

- **levelset_image** (*nimg*) – Levelset image to be turned into a mesh
- **connectivity** (*{ "6/18", "6/26", "18/6", "26/6" }*, *optional*) – Choice of digital connectivity to build the mesh (default is 18/6)
- **level** (*float*, *optional*) – Value of the levelset function to use as isosurface (default is 0)
- **inclusive** (*bool*, *optional*) – Whether voxels at the exact ‘level’ value are inside the isosurface (default is True)
- **save_data** (*bool*, *optional*) – Save output data to file (default is False)
- **overwrite** (*bool*, *optional*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn’t exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **result** (mesh): Surface mesh dictionary of “points” and “faces” (_l2m-mesh)

Return type dict

Notes

Ported from original Java module by Pierre-Louis Bazin. Original algorithm from¹ and adapted from².

References

15.1.1 Examples using `nighres.surface.levelset_to_mesh`

- *Cortical depth estimation from MGDM segmentation*

15.2 levelset_to_probability

`nighres.surface.levelset_to_probability` (*levelset_image*, *distance_mm=5*, *save_data=False*,
overwrite=False, *output_dir=None*,
file_name=None)

Levelset to probability

Creates a probability map from the distance to a levelset surface representation.

Parameters

- **levelset_image** (*niimg*) – Levelset image to be turned into probabilities
- **distance_mm** (*float*, *optional*) – Distance used for the range of probabilities in]0,1[
- **save_data** (*bool*, *optional*) – Save output data to file (default is False)
- **overwrite** (*bool*, *optional*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **result** (*niimg*): Probability map (output file suffix `_l2p-proba`)

Return type dict

Notes

Ported from original Java module by Pierre-Louis Bazin

¹ Han et al (2003). A Topology Preserving Level Set Method for Geometric Deformable Models doi:

² Lucas et al (2010). The Java Image Science Toolkit (JIST) for Rapid Prototyping and Publishing of Neuroimaging Software doi:

15.3 mesh_to_levelset

`nighres.surface.mesh_to_levelset` (*surface_mesh*, *reference_image*, *save_data=False*, *overwrite=False*, *output_dir=None*, *file_name=None*)

Mesh to levelset

Creates a signed distance function from a triangulated mesh using pseudonormals.

Parameters

- **surface_mesh** (*mesh*) – Mesh model of the surface
- **reference_image** (*niimg*) – Image of the dimensions and resolutions corresponding to the mesh
- **save_data** (*bool*, *optional*) – Save output data to file (default is False)
- **overwrite** (*bool*, *optional*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **result** (*niimg*): Levelset function representing the mesh (`_m2l-lvl`)

Return type `dict`

Notes

Ported from original Java module by Christine Tardif and Pierre-Louis Bazin.

References

15.4 probability_to_levelset

`nighres.surface.probability_to_levelset` (*probability_image*, *mask_image=None*, *save_data=False*, *overwrite=False*, *output_dir=None*, *file_name=None*)

Levelset from probability map

Creates a levelset surface representations from a probabilistic map or a mask. The levelset indicates each voxel's distance to the closest boundary. It takes negative values inside and positive values outside of the object.

Parameters

- **probability_image** (*niimg*) – Probability image to be turned into levelset. Values should be in $[0, 1]$, either a binary mask or defining the boundary at 0.5.
- **mask_image** (*niimg*, *optional*) – Mask image defining the region in which to compute the levelset. Values equal to zero are set to maximum distance.
- **save_data** (*bool*, *optional*) – Save output data to file (default is False)
- **overwrite** (*bool*, *optional*) – Overwrite existing results (default is False)

- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- result (niimg): Levelset representation of surface (_p2l-surf)

Return type dict

Notes

Original Java module by Pierre-Louis Bazin

15.4.1 Examples using `nighres.surface.probability_to_levelset`

- *Multiatlas Segmentation*

15.5 surface_inflation

`nighres.surface.surface_inflation` (*surface_mesh, step_size=0.75, max_iter=2000, max_curv=10.0, save_data=False, overwrite=False, output_dir=None, file_name=None*)

Surface inflation

Inflate a surface with the method of Tosun et al [1].

Parameters

- **surface_mesh** (*mesh*) – Mesh model of the surface
- **step_size** (*float*) – Relaxation rate in [0, 1]: values closer to 1 are more stable but slower (default is 0.75)
- **max_iter** (*int*) – Maximum number of iterations (default is 2000)
- **max_curv** (*float*) – Desired maximum curvature (default is 10.0)
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str, optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str, optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- result (mesh): Surface mesh dictionary of “points”, “faces” and “data” showing the SOM coordinates on the mesh

Return type dict

Notes

Original Java module by Pierre-Louis Bazin

15.5.1 Examples using `nighres.surface.surface_inflation`

- *Cortical depth estimation from MGDM segmentation*

15.6 surface_mesh_mapping

`nighres.surface.surface_mesh_mapping` (*intensity_image*, *surface_mesh*, *inflated_mesh=None*, *mapping_method='closest_point'*, *save_data=False*, *overwrite=False*, *output_dir=None*, *file_name=None*)

Surface mesh mapping

Maps volumetric data onto a surface mesh. A second mesh with the same graph topology (e.g. an inflated cortical surface) can also be mapped with the same data.

Parameters

- **intensity_image** (*nimg*) – Intensity image to map onto the surface mesh
- **surface_mesh** (*mesh*) – Mesh model of the surface
- **inflated_mesh** (*mesh*, *optional*) – Mesh model of the inflated surface
- **mapping_method** (*{'closest_point', 'linear_interp', 'highest_value'}*, *optional*) – Choice of mapping method
- **save_data** (*bool*, *optional*) – Save output data to file (default is False)
- **overwrite** (*bool*, *optional*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- original (mesh): Surface mesh dictionary of “points” and “faces” (_map-orig)
- inflated (mesh): Surface mesh dictionary of “points” and “faces” (_map-inf)

Return type `dict`

Notes

Ported from original Java module by Pierre-Louis Bazin

15.7 surface_som_mapping

`nighres.surface.surface_som_mapping` (*surface_mesh*, *mask_zeros=False*, *som_size=100*,
learning_time=100000, *total_time=500000*,
save_data=False, *overwrite=False*, *output_dir=None*,
file_name=None)

Surface SOM mapping

Maps surface coordinates onto a self-organizing map.

Parameters

- **surface_mesh** (*mesh*) – Mesh model of the surface
- **mask_zeros** (*bool*) – Whether to mask out zero values (default is False)
- **som_size** (*int*) – Size of the 2D SOM to generate
- **learning_time** (*int*) – Time for the learning stage iterations
- **total_time** (*int*) – Total number of iterations
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **original** (mesh): Surface mesh dictionary of “points”, “faces” and “data” showing the SOM coordinates on the mesh
- **som** (mesh): SOM mesh dictionary of “points”, “faces” and “data” generated from the SOM grid itself

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin

15.8 volume_som_mapping

`nighres.surface.volume_som_mapping` (*proba_image*, *som_size=100*, *learning_time=100000*, *total_time=500000*, *save_data=False*, *overwrite=False*, *output_dir=None*, *file_name=None*)

Surface SOM mapping

Maps surface coordinates onto a self-organizing map.

Parameters

- **proba_image** (*nii*) – Probabilistic representation of the surface

- **som_size** (*int*) – Size of the 2D SOM to generate
- **learning_time** (*int*) – Time for the learning stage iterations
- **total_time** (*int*) – Total number of iterations
- **save_data** (*bool*) – Save output data to file (default is False)
- **overwrite** (*bool*) – Overwrite existing results (default is False)
- **output_dir** (*str*, *optional*) – Path to desired output directory, will be created if it doesn't exist
- **file_name** (*str*, *optional*) – Desired base name for output files with file extension (suffixes will be added)

Returns

Dictionary collecting outputs under the following keys (suffix of output files in brackets)

- **map** (nii): Volumetric image showing the SOM coordinates map
- **som** (mesh): SOM mesh dictionary of “points”, “faces” and “data” generated from the SOM grid itself

Return type `dict`

Notes

Original Java module by Pierre-Louis Bazin

CHAPTER 16

Data handling and formats

Nighres represents data internally using `NibabelSpatialImage` objects which are referred to as `niiimg`.

Much of the input and output functionality has been adopted or inspired from `Nilearn`'s [conventions for data handling](#)

Todo: Explanation why this is useful and little example how it works, also mention dictionary outputs

Saving outputs

Each Nighres processing interface allows you to save the outputs by setting the `save_data` parameter to `True`. If this parameter is not specified, it defaults to `False` and the data is returned as a data object (see *Data handling and formats*) but not saved to disk.

If `save_data` is set to `True`, Nighres applies the following logic:

Output directory

1. If `output_dir` is specified, the data is saved there. In case `output_dir` doesn't exist, it is created
2. If `output_dir` is not specified, Nighres tries to use the location of an input file as the location for saving. This only works if the input is a file name and not a data object
3. Otherwise, the data is saved in the current working directory

File names

1. If `file_name` is specified, this name is used as a base to create the output names. A suffix is added to each output (you can see in the docstrings which suffix refers to which output). The extension of `file_name` specifies the format in which the output will be saved. If `file_name` has no extension, Nighres defaults to `nii.gz`
2. If `file_name` is not specified, Nighres tries to use the name of an input file as a base name for saving. This only works if the input is indeed a file name and not a data object

CHAPTER 18

Levelsets

Todo: What is a levelset and why is Nighres using it, example of creating levelset using `nighres.surface.probability_to_levelset()`

You want to contribute to Nighres? That's fantastic! A few words to get started:

You don't have to be an expert!

Contributing to open source software is a great way of learning by doing, and others in the Nighres community will be happy to help you out if you have questions or get stuck.

What to do

Every contribution helps – whether it's fixing a typo in the documentation or adding an entire new feature. Often it's a good idea to start by fixing a smaller issue to get familiar with the process. If you don't know where to start, you can look at the [issue tracker](#).

Git/Github

To contribute you will need to know some basics about git and Github, which are not covered in this guide. There are lots of tutorials online (e.g. <https://www.atlassian.com/git/tutorials>, especially “Getting started” and “Collaborating”)

Questions?

For specifics about code you want to contribute *Making a Pull Request* is a great way to start a discussion and get input. Otherwise, you can post a question on neurostars.org. Please use the tag **nighres** for your question.

Ready to go?

The next pages will take you through the steps you need from cloning the Nighres repository, to making a pull request.

Note: Help us improve this developer's guide! If you find anything unclear or not covered, open an [issue on github](#) or add what you were missing (see *Adapting the docs*)

1. Fork and clone the Nighres github repository

You can find a good description [here](#). Make sure that you set up your local clone to track both your fork (typically as *origin*) as well as the original Nighres repo (typically as *upstream*).

2. Make a new branch to work on

```
git checkout -b <branch_name>
```

This is important. When you make a pull request later, we will likely ask you to [rebase](#) because the Nighres master might have moved on since you forked it. You need a clean master branch for that.

Pick a descriptive branch name. For example, when you fix a bug in the function `load_volume` a good name is `fix/load_volume`. When you write a new interface called `laminar_connectivity` (that would be cool!) a good name is `enh/laminar_connectivity`.

Make one branch for each new feature or fix. That way independent changes can be handled in different *pull requests* later.

3. Install in editable mode (optional)

```
pip install -e <path_to_nighres_directory>
```

This way, when you import Nighres functions in Python, they will always come from the current version of the code in your Nighres directory. This is convenient for testing your code while developing.

(Alternatively, you can stay inside your Nighres directory and import functions directly from there without installing at all)

4. Let the coding begin!

If you want to work on an existing function, you can most likely just make your changes, check if the *Examples* still run and the move on to *Adapting the docs* and *Making a Pull Request*.

If you want to add new functions be sure to check out our instructions on *Wrapping an existing CBS Tools class* or *Adding a new Python function*

Important: Please adhere to [PEP8 style conventions](#). This is easiest by using a Python linter which is available in most editors.

Wrapping an existing CBS Tools class

You want to wrap an existing CBS Tools module into Nighres? Great!

21.1 How does it work?

The CBS Tools modules have been first created for JIST. Most of them are however not directly dependent on the JIST and MIPAV libraries, which is why we can turn them into a JCC-compatible, lightweight library.

JCC requires that all data is passed as numbers, strings, and 1D arrays, so this is what the CBS Tools modules need to manipulate as input and output, and also why we created a second python layer for structured data handling. Once a CBS Tools module has been formatted to the adequate i/o structure, JCC wraps it and you can call it directly from python.

21.2 Standard procedures

Each JCC-wrapped CBS Tools module has three important functions: `.setSomeParameter()`, `.getSomeResult()`, and `.execute()`. You need to set the inputs with meaningful values, execute the module, and retrieve the results. Note that default parameters are always assumed when possible, but a good wrapper should expose all parameters.

21.3 General wrapper content

1 Start the Java Virtual Machine (JVM)

```
cbstools.initVM(initialheap='6000m', maxheap='6000m')
```

Note that the initial and maximum amount of allocated memory may be adjusted or set up as a global parameter.

2 Create an instance of the module

```
mymodule = cbstools.SomeCoolModule()
```

3 Set all the parameters and data arrays

```
my_module.setThisImportantParameter(some_value)          my_module.  
setInputImage(cbstools.JArray('float')((my_image_data.flatten('F')).  
astype(float)))
```

4 Run the module

```
my_module.execute()
```

5 Retrieve the outputs

```
my_result_data = np.reshape(np.array(my_module.getCoolResultImage(),  
dtype=np.float32), dimensions, 'F')
```

Note that because you are passing simple 1D arrays, you need to keep a record of image dimensions, resolutions, headers, etc.

21.4 Convenience python layer

We strongly advise (at least for inclusion into Nighres) to handle the inputs and outputs of python structures as we do, i.e. loading images from files or passing them as Nifti1Image, and allowing to save all outputs automatically.

We also adhere to a somewhat strict formatting of the outputs with two suffixes, first for the module name, and second for the specific output, e.g. “_mgdm_seg” for the segmentation obtained from the mgdm_brain_segmentation module.

21.5 What if my favorite CBS Tools module is not ready to wrap?

Because preparing modules for wrapping requires some reformatting, only a few modules so far are ready. You can accelerate the process in two ways: 1) send a request to [the CBS Tools developers](#) or 2) write a reformatted module for CBS Tools yourself and make a pull request. All the currently wrapped CBS Tools modules follow the same formatting procedure, so transforming a JIST module definition into a core module (which then can be called either from JIST or from Nighres) should be reasonably easy in most cases.

Adding a new Python function

Great, you want to contribute a Python function or module to Nighres!

22.1 Before you begin

Please take a look at the code that is [under development](#) or [discussion](#). If you see something related get in touch with the developer(s) directly to combine your efforts.

Take a moment to familiarise yourself with the documentation and functions, it is likely that some helper functions that you will want to use (e.g., [I/O](#)) have already been coded and tested.

22.2 Let's get into it

1. Follow the steps described in [Setting up](#)
2. Decide where to put your code

If the functionality that you are adding falls within the scope of a pre-existing submodule, edit or create files within the submodule. If it is entirely new (e.g., [statistics](#)), create a new submodule with its own dedicated subdirectory.

3. Coding If you are creating a submodule from scratch, an easy way to start is to copy the `__init.py__` and initial import statements from an existing module.

Please code [PEP8 compliant](#), best to use a Python linter in your editor.

Please keep within our [documentation guidelines](#).

Leave plenty of comments in your code so that others can understand and possibly adapt your code later.

Use the standard [I/O interfaces](#) wherever possible but also feel free to add additional I/O functionality as necessary.

Test you code internally. We aim to add unittests in the future, feel free to make a start on that.

4. *Write an example* showcasing your new function
5. To get your code into the Nighres master follow *Adapting the docs* and *Making a Pull Request*
6. Pat yourself on the back, you have now joined the Nighres developer community!

A word on dependencies

One of Python's strengths are the many different packages with specific functionalities. We try to keep the dependencies for Nighres as slim as possible, so please keep that in mind while you are coding. For example, if you need to perform a correlation and would normally use `scipy.stats.pearsonr` maybe you can get by with `numpy.corrcoef` (numpy is already a dependency, scipy isn't). If additional packages are required, they should be pip installable and potentially relevant to other functionality that may be coded in the future.

CHAPTER 23

Writing examples

Nighres uses [Sphinx](#) which makes it easy to automatically pull the docstrings of functions into the online documentation. This means once a function has a good docstring, the work is essentially done. It also means that writing great docstrings is important.

24.1 If you changed an existing function

Make sure to check if your changes also require changes in the docstring. Adapting an existing docstring is usually straightforward. If you are curious how your changes will look in the online docs, you can try out [Building the docs locally](#).

24.2 If you added a new function

Make sure to write a comprehensive docstring. We use [NumPy/SciPy docstring](#) conventions. But the easiest is to just copy another function's docstring and adapt. Then you need to add a few things to the `nighres/doc` directory:

1. Go to the subdirectory which refers to the submodule you added a function to, e.g. `nighres/doc/laminar/`.
2. Create a new file named after your new function, e.g. `laminar_connectivity.rst`
3. Add the name of this file in the `index.rst` file of the submodule
4. Copy the content of another function's `rst` file inside of your new file
5. Adapt the title and the two mentions of the function name in your new file to match your new function
6. Submit the changes to the docs along with your PR

Again, you can check how the changes you made will look in the online documentation by [Building the docs locally](#).

24.3 If you added a new submodule

This is going to be a rare case. But if indeed added a new submodule in your PR, say its called *nighres.fancypants*:

1. In the doc directory, add a new subdirectory for your new module, e.g. `nighres/doc/fancypants`
2. In that subdirectory make a file called *index.rst*. Best to just copy the `index.rst` from another submodule. Then adapt the title in the file to the name of your new submodule, and remove all function names in the toctree.
3. Follow steps 3 to 5 from *If you added a new function* to add all functions of your new submodule
4. Add your submodule in the *index.rst* file in the `nighres/doc` main directory under the “Modules and Functions” toctree. In our case we would add the line *fancypants/index*
5. Submit the changes to the docs along with your PR

24.4 More than docstrings

You can also make changes to parts of the documentation that are not function docstrings. For example to help improving this guide!

The easiest way is to browse to the page in the [online documentation](#) that you want to change and click on **Edit on Github** to find the location of the file you have to change. For example, if you want to make a change to this page, the link would send you to <https://github.com/nighres/nighres/blob/master/doc/developers/docs.rst>, which tells you that in your local copy of the repo you should edit the file *nighres/doc/developers/docs.rst*.

Sphinx uses reStructuredText (reST) syntax. For formatting take a look at this [Sphinx Cheatsheet](#).

You can also add whole new parts to the online documentation. This might require learning a bit more about [Sphinx](#). You can also open an [issue](#) and get help.

When you are done, you can check your changes by *Building the docs locally*.

24.5 Building the docs locally

If you make changes to the documentation, they will only appear online at <http://nighres.readthedocs.io/en/latest/> once your PR has been merged and a new release was made. You can, however, build the docs locally for a preview:

1. Make sure to have all *additional dependencies* installed that you need to build the documentation
2. Navigate into your local copy of `nighres/doc`
3. Type `make clean` and then `make html`
4. Open an Internet browser and point it to the local html pages that were just built for you, e.g. `file:///home/nighres/doc/_build/html/index.html`

You can make changes and repeat this process until you are satisfied.

Making a Pull Request

If you are working on changes to the Nighres code, let others know by opening a [Pull Request \(PR\)](#) .

You don't have to wait until your code is perfect, opening a PR early is a great way of letting other members of the community know that you are working on this. That avoids redundant work and gets you input at an early stage.

To open a PR:

1. Go to your fork of Nighres in your Github account
2. In the **Branch** drop-down menu, switch to the branch that you have worked on
3. Click on **New pull request**. You will see the changes that will be included in your PR.
4. Enter a descriptive title and small comment why these changes are necessary. If you are addressing a specific [issue](#), refer to the issue number.
5. Click **Create pull request** and wait for responses (by default you will be informed about activity on your PR per e-mail)

After your PR is reviewed, you might be asked to make some changes. This is normal, almost no PR is merged without revisions. Pushing changes to your own branch automatically updates the code in the PR.

When all comments are addressed we will likely ask you to [rebase](#), and then merge your code into the Nighres repository.

Reference

Huntenburg, Steele & Bazin (2018). Nighres: processing tools for high-resolution neuroimaging. *GigaScience*, 7(7). <https://doi.org/10.1093/gigascience/giy082>

Make sure to also cite the references indicated for the particular functions you use!

Credit

Nighres is a community-developed project made possible by its [contributors](#). The project was born and continues to evolve at [brainhack](#). We thank the [Google Summer of Code 2017](#) and [INCF](#) as a mentoring organization, for supporting the initial development phase of Nighres. See also the [development blog](#).

-
- A**
- `apply_coordinate_mappings()` (in module *nighres.registration*), 77
 - `apply_coordinate_mappings_2d()` (in module *nighres.registration*), 78
- B**
- `background_estimation()` (in module *nighres.intensity*), 53
- C**
- `conditional_shape()` (in module *nighres.segmentation*), 87
 - `cruise_cortex_extraction()` (in module *nighres.cortex*), 41
- D**
- `dots_segmentation()` (in module *nighres.brain*), 31
 - `download_7T_TRT()` (in module *nighres.data*), 45
 - `download_DOTS_atlas()` (in module *nighres.data*), 46
 - `download_DTI_2mm()` (in module *nighres.data*), 46
- E**
- `embedded_antsreg()` (in module *nighres.registration*), 79
 - `embedded_antsreg_2d()` (in module *nighres.registration*), 81
 - `embedded_antsreg_multi()` (in module *nighres.registration*), 82
 - `extract_brain_region()` (in module *nighres.brain*), 32
- F**
- `filter_ridge_structures()` (in module *nighres.filtering*), 47
 - `filter_stacking()` (in module *nighres.brain*), 34
- G**
- `flash_t2s_fitting()` (in module *nighres.intensity*), 54
 - `fuzzy_cmeans()` (in module *nighres.segmentation*), 88
- I**
- `intensity_based_skullstripping()` (in module *nighres.brain*), 35
 - `intensity_propagation()` (in module *nighres.intensity*), 55
- L**
- `laminar_iterative_smoothing()` (in module *nighres.laminar*), 67
 - `laminar_regional_approximation()` (in module *nighres.laminar*), 68
 - `lcat_denoising()` (in module *nighres.intensity*), 55
 - `lcpca_denoising()` (in module *nighres.intensity*), 56
 - `levelset_fusion()` (in module *nighres.shape*), 91
 - `levelset_to_mesh()` (in module *nighres.surface*), 97
 - `levelset_to_probability()` (in module *nighres.surface*), 98
 - `load_mesh_data()` (in module *nighres.io*), 65
 - `load_mesh_geometry()` (in module *nighres.io*), 64
 - `load_volume()` (in module *nighres.io*), 63
- M**
- `mesh_to_levelset()` (in module *nighres.surface*), 99
 - `mgdm_cells()` (in module *nighres.microscopy*), 73
 - `mgdm_segmentation()` (in module *nighres.brain*), 36
 - `mp2rage_dura_estimation()` (in module *nighres.brain*), 38
-

mp2rage_skullstripping() (in module *nighres.brain*), 39
mp2rage_t1_mapping() (in module *nighres.intensity*), 58
mp2rageme_pd_mapping() (in module *nighres.intensity*), 59
multiscale_vessel_filter() (in module *nighres.filtering*), 48
volumetric_layering() (in module *nighres.laminar*), 71

P

phase_unwrapping() (in module *nighres.intensity*), 60
probability_to_levelset() (in module *nighres.surface*), 99
profile_averaging() (in module *nighres.laminar*), 69
profile_meshing() (in module *nighres.laminar*), 69
profile_sampling() (in module *nighres.laminar*), 70

R

recursive_ridge_diffusion() (in module *nighres.filtering*), 49

S

save_mesh_data() (in module *nighres.io*), 65
save_mesh_geometry() (in module *nighres.io*), 64
save_volume() (in module *nighres.io*), 63
segmentation_statistics() (in module *nighres.statistics*), 95
simple_align() (in module *nighres.registration*), 85
simple_skeleton() (in module *nighres.shape*), 92
stack_intensity_regularisation() (in module *nighres.microscopy*), 74
surface_inflation() (in module *nighres.surface*), 100
surface_mesh_mapping() (in module *nighres.surface*), 101
surface_som_mapping() (in module *nighres.surface*), 102

T

t2s_optimal_combination() (in module *nighres.intensity*), 61
topology_correction() (in module *nighres.shape*), 93
total_variation_filtering() (in module *nighres.filtering*), 51

V

volume_som_mapping() (in module *nighres.surface*), 102