
Two Luis Miguel fans walk into a bar in Nagoya yada yada yada → an ITK-implementation of a popular patch-based denoising filter

Release 0.00

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August 27, 2016

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Abstract

This document describes an ITK implementation of the adaptive patch-based image denoising algorithm described in [1]. This offering consists of the templated class `itkPatchBasedDenoisingImageFilter` interfaced through the `DenoiseImage` program which includes its own interface program. We include a couple of examples.

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1 Introduction

Image filtering techniques are usually applied to increase the quality of MR images. Most of these techniques assume an equal noise distribution across the image. When this assumption is not met, the resulting filtering becomes suboptimal. This is the case of MR images with spatially varying noise levels, such as those obtained by parallel imaging (sensitivity-encoded), intensity inhomogeneity-corrected images or surface coil based acquisitions. In this work, we have adapted a recently proposed filter so-called Non-Local Means (NL-means) to deal with MR images with spatially varying noise levels (for both Gaussian and Rician distributed noise). With this new method, information regarding the local image noise level is used to adjust the amount of denoising strength of the filter. Such information is automatically obtained from the images by using the local noise estimation method described in [1].

This method has been validated and compared with the standard NL-means filter on simulated and real MR imaging data. This noise-adaptive NL-means method was demonstrated to outperform the standard NL-means based filter when spatially varying noise is present in the images. Although source code is available from Prof. Manjón Herrera's website (<http://personales.upv.es/jmanjon/denoising/arnlm.html>). This particular algorithm is currently not available in the Insight Toolkit which provides good performance with

minimal parameter tuning. In this Insight Journal submission, we provide a brief overview of the ITK implementation and provide a couple examples.

2 ITK Implementation

The following files define the adaptive NL-means algorithm and is written in ITK coding style:

- `itkAdaptiveNonLocalMeansDenoisingImageFilter.h`
- `itkAdaptiveNonLocalMeansDenoisingImageFilter.hxx`

The following helper class

- `itkVarianceImageFilter.h`
- `itkVarianceImageFilter.hxx`

is the variance analog to the `itk::MeanImageFilter` which “Computes an image where a given pixel is the [variance] value of the the pixels in a neighborhood about the corresponding input pixel.” Both the adaptive and variance filters are multi-threaded.

Additionally, we wrote an easy-to-use interface program, `DenoiseImage.cxx`, with an ANTs-based command line interface facilitated by the following helper classes:

- `antsCommandLineOption.h`
- `antsCommandLineOption.cxx`
- `antsCommandLineParser.h`
- `antsCommandLineParser.cxx`

These latter files are not essential for other applications or for inclusion into ITK. They are included simply to facilitate the generation of a stand-alone application that people can use. One can invoke the help menu via `$ DenoiseImage --help 1` at the command line:

```
$ DenoiseImage --help 1
```

COMMAND:

```
DenoiseImage
    Denoise an image using a spatially adaptive filter originally described in J. V.
    Manjon, P. Coupe, Luis Marti-Bonmati, D. L. Collins, and M. Robles. Adaptive
    Non-Local Means Denoising of MR Images With Spatially Varying Noise Levels,
    Journal of Magnetic Resonance Imaging, 31:192-203, June 2010.
```

OPTIONS:

```
-d, --image-dimensionality 2/3/4
    This option forces the image to be treated as a specified-dimensional image. If
    not specified, the program tries to infer the dimensionality from the input
    image.
```

```

-i, --input-image inputImageFilename
    A scalar image is expected as input for noise correction.

-n, --noise-model Rician/(Gaussian)
    Employ a Rician or Gaussian noise model.

-x, --mask-image maskImageFilename
    If a mask image is specified, denoising is only performed in the mask region.

-s, --shrink-factor (1)/2/3/...
    Running noise correction on large images can be time consuming. To lessen
    computation time, the input image can be resampled. The shrink factor, specified
    as a single integer, describes this resampling. Shrink factor = 1 is the
    default.

-p, --patch-radius 1
                    1x1x1
    Patch radius. Default = 1x1x1

-r, --search-radius 3
                    3x3x3
    Search radius. Default = 3x3x3.

-o, --output correctedImage
    [correctedImage,<noiseImage>]
    The output consists of the noise corrected version of the input image.
    Optionally, one can also output the estimated noise image.

-v, --verbose (0)/1
    Verbose output.

-h
    Print the help menu (short version).

--help
    Print the help menu.

```

which should provide minimal detail regarding each parameter. The default parameters should provide adequate performance for a variety of cases thus requiring only the simple call (for a 3-D image)

```
$ DenoiseImage -d 3 -i ${inputImage} -o ${outputImage}
```

As mentioned, this filter is multi-threaded so one might want to set the environment variable to enable more than one thread, e.g.,

```
$ export ITK_GLOBAL_DEFAULT_NUMBER_OF_THREADS=4
```

To see the progress of the filter, one also needs to set the verbose flag, i.e., `-v 1`.

3 Examples

We included a 2-D brain slice and a 3-D hyperpolarized gas volume each with a corresponding mask. Below we include the call and resulting output for these two sets of data.

3.1 T1-weighted MRI brain slice

```
$ DenoiseImage -d 2 -i t81slice.nii.gz -x t81mask.nii.gz -o t81denoised.nii.gz -v 1
```

Running for 2-dimensional images.

```
*****10*****20*****30*****40*****50*****60*****70*****80*****90*****
```

```
AdaptiveNonLocalMeansDenoisingImageFilter (0x7f9733997080)
  RTTI typeinfo:   itk::AdaptiveNonLocalMeansDenoisingImageFilter<itk::Image<float, 2u>, itk::Image<float, 2u>>
  Reference Count: 1
  Modified Time: 517
  Debug: Off
  Object Name:
  Observers:
    ProgressEvent (Command)
  Inputs:
    Primary: (0x7f9733996c10) *
    _1: (0x7f97339988f0)
  Indexed Inputs:
    0: Primary (0x7f9733996c10)
    1: _1 (0x7f97339988f0)
  Required Input Names: Primary
  NumberOfRequiredInputs: 1
  Outputs:
    Primary: (0x7f9733996a50)
  Indexed Outputs:
    0: Primary (0x7f9733996a50)
  NumberOfRequiredOutputs: 1
  Number Of Threads: 1
  ReleaseDataFlag: Off
  ReleaseDataBeforeUpdateFlag: Off
  AbortGenerateData: Off
  Progress: 1
  Multithreader:
    RTTI typeinfo:   itk::MultiThreader
    Reference Count: 1
    Modified Time: 299
    Debug: Off
    Object Name:
    Observers:
      none
    Thread Count: 1
    Global Maximum Number Of Threads: 128
    Global Default Number Of Threads: 1
  CoordinateTolerance: 1e-06
  DirectionTolerance: 1e-06
```

```

Using Gaussian noise model.
Epsilon = 1e-05
Mean threshold = 0.95
Variance threshold = 0.5
Smoothing variance = 2
Neighborhood radius for local mean and variance = [1, 1]
Neighborhood search radius = [3, 3]
Neighborhood block radius = [1, 1]
Elapsed time: 0.0831451

```

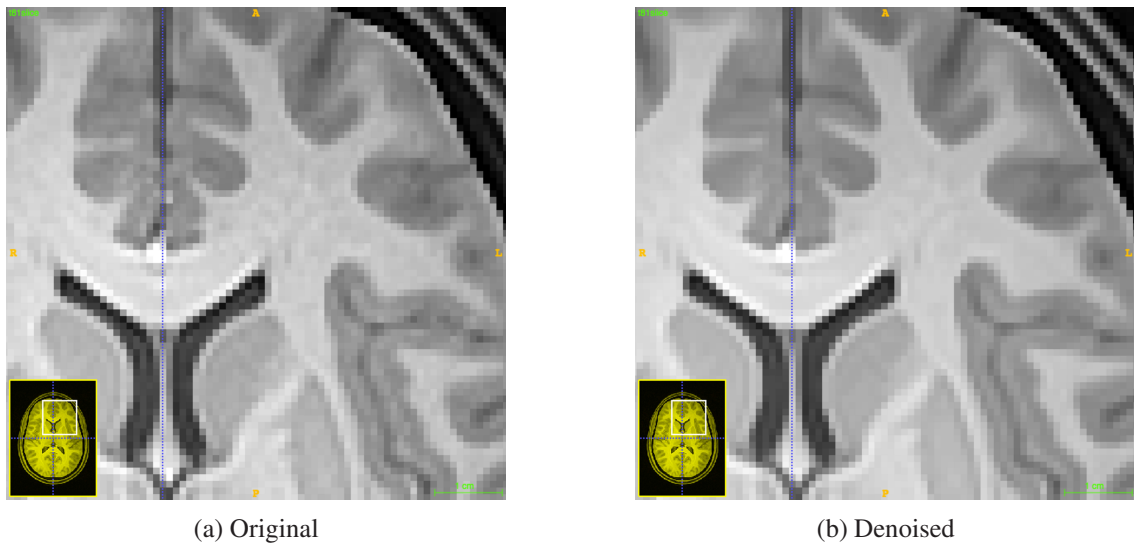


Figure 1: Effects of the NL-means filter on a 2-D T1-weighted MRI brain slice.

3.2 Hyperpolarized He3 Lung MRI volume

```
$ DenoiseImage -d 3 -i he3volume.nii.gz -x he3mask.nii.gz -o he3denoised.nii.gz -v 1
```

Running for 3-dimensional images.

```
*****10*****20*****30*****40*****50*****60*****70*****80*****90*****
```

```

AdaptiveNonLocalMeansDenoisingImageFilter (0x7ff279e0a440)
RTTI typeid: itk::AdaptiveNonLocalMeansDenoisingImageFilter<itk::Image<float, 3u>, itk::Image
Reference Count: 1
Modified Time: 506
Debug: Off
Object Name:
Observers:
  ProgressEvent(Command)
Inputs:
  Primary: (0x7ff27a97ea20) *
  _1: (0x7ff27a97f040)
Indexed Inputs:
  0: Primary (0x7ff27a97ea20)
  1: _1 (0x7ff27a97f040)

```

```
Required Input Names: Primary
NumberOfRequiredInputs: 1
Outputs:
  Primary: (0x7ff279e0a710)
Indexed Outputs:
  0: Primary (0x7ff279e0a710)
NumberOfRequiredOutputs: 1
Number Of Threads: 4
ReleaseDataFlag: Off
ReleaseDataBeforeUpdateFlag: Off
AbortGenerateData: Off
Progress: 1
Multithreader:
  RTTI typeid: itk::MultiThreader
  Reference Count: 1
  Modified Time: 294
  Debug: Off
  Object Name:
  Observers:
    none
  Thread Count: 4
  Global Maximum Number Of Threads: 128
  Global Default Number Of Threads: 4
CoordinateTolerance: 1e-06
DirectionTolerance: 1e-06
Using Gaussian noise model.
Epsilon = 1e-05
Mean threshold = 0.95
Variance threshold = 0.5
Smoothing variance = 2
Neighborhood radius for local mean and variance = [1, 1, 1]
Neighborhood search radius = [3, 3, 3]
Neighborhood block radius = [1, 1, 1]
Elapsed time: 1.7134
```

References

- [1] José V Manjón, Pierrick Coupé, Luis Martí-Bonmatí, D Louis Collins, and Montserrat Robles. Adaptive non-local means denoising of mr images with spatially varying noise levels. *J Magn Reson Imaging*, 31(1):192–203, Jan 2010. ([document](#)), 1

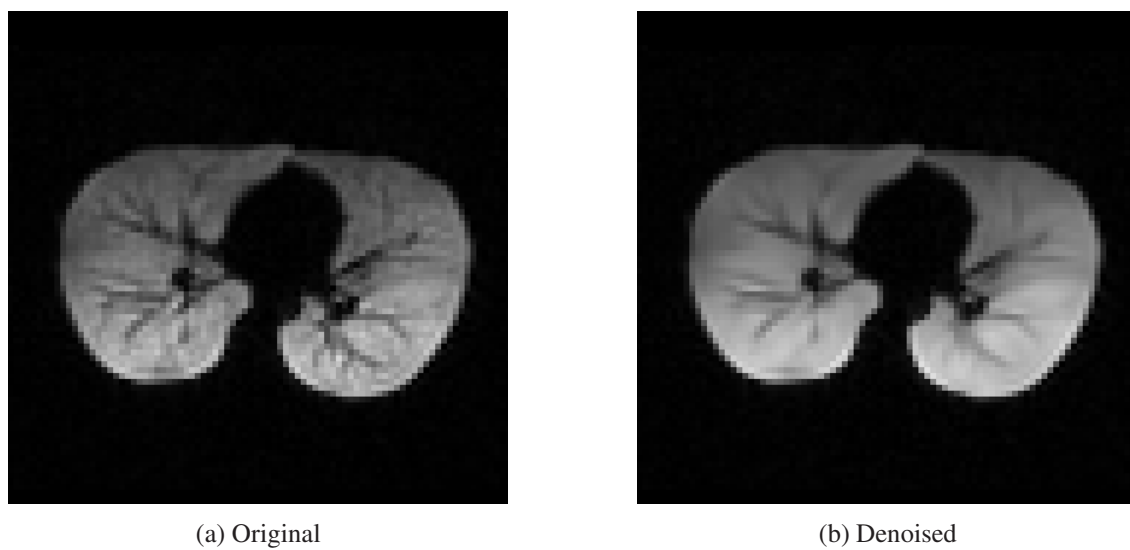


Figure 2: Effects of the NL-means filter on a 3-D hyperpolarized gas MRI lung volume.