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Nineth exercice: non-Cartesian MR image reconstruction
         In this tutorial we will reconstruct an MRI image from radial undersampled kspace measurements. Let us denote \Omega the undersampling mask, the under-
         sampled Fourier transform now reads F_{\rm Q}.
         Import neuroimaging data
         We use the toy datasets available in pysap, more specifically a 2D brain slice and the radial under-sampling scheme. We compare zero-order image
         reconstruction with Compressed sensing reconstructions (analysis vs synthesis formulation) using the FISTA algorithm for the synthesis formulation and the
         Condat-Vu algorithm for the analysis formulation.
         We remind that the synthesis formulation reads (minimization in the sparsifying domain):
                                                       \hat{z} = \arg \min_{z \in C_{\Psi}^{n}} \frac{1}{2} \|y - F_{\Omega} \Psi^* z\|_{2}^{2} + \lambda \|z\|_{1}
         and the image solution is given by \hat{x} = \Psi * \hat{z}. For an orthonormal wavelet transform, we have n_{\Psi} = n while for a frame we may have n_{\Psi} > n.
         while the analysis formulation consists in minimizing the following cost function (min. in the image domain):
                                                       \hat{x} = \arg\min_{x \in C^n} \frac{1}{2} \|y - F_{\Omega} x\|_2^2 + \lambda \|\Psi x\|_1.

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           • Target: ATSI MSc students, Paris-Saclay University
 In [9]: # Package import
         #from mri.numerics.fourier import NFFT
         #from mri.numerics.reconstruct import sparse_rec_fista
         #from mri.numerics.utils import generate_operators
         #from mri.numerics.utils import convert_locations_to_mask
         #from mri.parallel_mri.extract_sensitivity_maps import \
         # gridded_inverse_fourier_transform_nd
         from mri.operators import NonCartesianFFT, WaveletN, WaveletUD2
         from mri.operators.utils import convert_locations_to_mask, \
             gridded_inverse_fourier_transform_nd
         from mri.reconstructors import SingleChannelReconstructor
         import pysap
         from pysap.data import get_sample_data
         # Third party import
         from modopt.math.metrics import ssim
         from modopt.opt.linear import Identity
         from modopt.opt.proximity import SparseThreshold
         import numpy as np
         import matplotlib.pyplot as plt
         Loading input data
In [16]: image = get_sample_data('2d-mri')
         radial_mask = get_sample_data("mri-radial-samples")
         kspace_loc = radial_mask.data
         mask = pysap.Image(data=convert_locations_to_mask(kspace_loc, image.shape))
         plt.figure()
         plt.imshow(image, cmap='gray')
         plt.figure()
         plt.imshow(mask, cmap='gray')
         plt.show()
          <pysap.base.image.Image object at 0x7ff26d2d6750>
           100
           200
           300
           400
                 100 200 300 400 500
                 100 200 300 400 500
         Generate the kspace
         From the 2D brain slice and the acquisition mask, we retrospectively undersample the k-space using a cartesian acquisition mask We then reconstruct the
         zero order solution as a baseline
         Get the locations of the kspace samples
In [7]: #fourier_op = NFFT(samples=kspace_loc, shape=image.shape)
          #kspace_obs = fourier_op.op(image.data)
          fourier_op = NonCartesianFFT(samples=kspace_loc, shape=image.shape,
                                        implementation='cpu')
          kspace_obs = fourier_op.op(image.data)
          -----
                                                     Traceback (most recent call last)
         <ipython-input-7-fa4814d2c13c> in <module>
               2 #kspace_obs = fourier_op.op(image.data)
               3 fourier_op = NonCartesianFFT(samples=kspace_loc, shape=image.shape,
                                                implementation='cpu')
                5 kspace_obs = fourier_op.op(image.data)
         ~/work/code/git/pysap-mri/mri/operators/fourier/non_cartesian.py in __init__(self, samples, shape, implementation, n_
         coils, **kwargs)
                          if implementation == 'cpu':
             550
             551
                              self.implementation = NFFT(samples=samples, shape=shape,
          --> 552
                                                          n_coils=self.n_coils)
             553
                          elif implementation == 'cuda' or implementation == 'opencl':
                              self.implementation = NUFFT(samples=samples, shape=shape,
             554
          ~/work/code/git/pysap-mri/mri/operators/fourier/non_cartesian.py in __init__(self, samples, shape, n_coils)
                          # TODO Parallelize this if possible
             105
                          self.nb_coils = n_coils
             106
                          self.plan = pynfft.NFFT(N=shape, M=len(samples))
          --> 107
             108
                          self.plan.x = self.samples
                          self.plan.precompute()
         NameError: name 'pynfft' is not defined
         Gridded solution
 In [ ]: grid_space = np.linspace(-0.5, 0.5, num=image.shape[0])
          grid2D = np.meshgrid(grid_space, grid_space)
         grid_soln = gridded_inverse_fourier_transform_nd(kspace_loc, kspace_obs,
                                                            tuple(grid2D), 'linear')
         plt.imshow(np.abs(grid_soln), cmap='gray')
         # Calculate SSIM
         base_ssim = ssim(grid_soln, image)
         plt.title('Gridded Solution\nSSIM = ' + str(base_ssim))
         plt.show()
         FISTA optimization
         We now want to refine the zero order solution using a FISTA optimization. The cost function is set to Proximity Cost + Gradient Cost
In [11]: linear_op = WaveletN(wavelet_name="sym8", nb_scales=4)
         regularizer_op = SparseThreshold(Identity(), 6 * 1e-7, thresh_type="soft")
         Generate operators
In [12]: reconstructor = SingleChannelReconstructor(
              fourier_op=fourier_op,
             linear_op=linear_op,
             regularizer_op=regularizer_op,
             gradient_formulation='synthesis',
              verbose=1,
          ______
                                                     Traceback (most recent call last)
          <ipython-input-12-8eee53257f67> in <module>
               1 reconstructor = SingleChannelReconstructor(
          ----> 2 fourier_op=fourier_op,
               3 linear_op=linear_op,
               4 regularizer_op=regularizer_op,
                5 gradient_formulation='synthesis',
         NameError: name 'fourier_op' is not defined
         Synthesis formulation: FISTA optimization
         We now want to refine the zero order solution using a FISTA optimization. The cost function is set to Proximity Cost + Gradient Cost
 In [ ]: x_final, costs, metrics = reconstructor.reconstruct(
             kspace_data=kspace_obs,
             optimization_alg='fista',
             num_iterations=200,
          image_rec = pysap.Image(data=np.abs(x_final))
         recon_ssim = ssim(image_rec, image)
         plt.imshow(np.abs(image_rec), cmap='gray')
         recon_ssim = ssim(image_rec, image)
         plt.title('FISTA Reconstruction\nSSIM = ' + str(recon_ssim))
         plt.show()
         Analysis formulation: Condat-Vu reconstruction
In [14]: linear_op = WaveletUD2(
              wavelet_id=24,
              nb_scale=4,
In [15]: reconstructor = SingleChannelReconstructor(
              fourier_op=fourier_op,
             linear_op=linear_op,
             regularizer_op=regularizer_op,
              gradient_formulation='analysis',
              verbose=1,
         NameError
                                                     Traceback (most recent call last)
         <ipython-input-15-6c81cdac1fa0> in <module>
               1 reconstructor = SingleChannelReconstructor(
         ----> 2 fourier_op=fourier_op,
               3 linear_op=linear_op,
               4 regularizer_op=regularizer_op,
                     gradient_formulation='analysis',
         NameError: name 'fourier_op' is not defined
 In [ ]: x_final, costs, metrics = reconstructor.reconstruct(
              kspace_data=kspace_obs,
              optimization_alg='condatvu',
             num_iterations=200,
          image_rec = pysap.Image(data=np.abs(x_final))
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plt.imshow(np.abs(image_rec), cmap='gray')

plt.title('Condat-Vu Reconstruction\nSSIM = ' + str(recon_ssim))

recon_ssim = ssim(image_rec, image)

plt.show()