

## Homework Assignment VI

### Computer Assignment:

You can find the code package on the course webpage which implements a few popular sparse-recovery algorithms. This exercise helps you to investigate and compare their accuracy as well as robustness in the recovery of sparse signals with various sparsity level  $S$  using real images. You can choose two: your favorite greedy algorithm and your favorite  $ell_1$ -minimization algorithm in the assignment.

The sensing matrices in comparison are: Random Gaussian, Random Subsampling, and Structurally Random Matrices. Consider the following SRM construction – a block-diagonal matrix where each block on the diagonal of size  $B$  is a scaled product of 3 matrices:  $\mathbf{RFD}$ , where  $\mathbf{D}$  is either a diagonal matrix of i.i.d Bernoulli random variables or a matrix of uniform random permutation;  $\mathbf{F}$  is either an  $B \times B$  Hadamard matrix; and  $\mathbf{R}$  is a random subset of rows of the  $B \times B$  identity matrix.

Consider the following three images: Phantom (synthetic), Brain (real) and Boat (real), all available on the course web page in the same code package. This time, devise your own stopping criterion and try to fine-tune other parameter(s) for all of the algorithms. The two sparsifying matrices that you should consider are: DCT and Wavelet. The main.m file in the package helps you to set up the Compressed Sensing problem for these images. For DCT, you would like to use a small patch size, say  $8 \times 8$ . However, for Wavelet, you would like to set the patch size to be as large as the image itself. Compute the distortion based on the peak signal-to-noise ratio, often abbreviated PSNR, defined as follows

$$PSNR = 10 \log_{10} \frac{MAX^2}{MSE} \text{ where } MSE = \frac{1}{N^2} \|\hat{\mathbf{x}} - \mathbf{x}\|_2^2.$$

For our three test images,  $MAX = 255$  (the maximum dynamic range) and  $N$  is the image dimension. Use the psnr.m file in the package to plot the PSNR between the recovered images and the original with respect to the number of measurements  $M$ . What are your observations on how to obtain the best recovery performance?

Due date: **Thurs, April 6** in lecture or via email