

## Homework 7

MA430

As always, give full justification for your answers!

1. Read Chapter 1, Chapter 6, and anything else I handed out (e.g., section I to III of the paper I wrote with Deborah Walter).
2. Consider a one-dimensional version of the problem of locating radio frequency (RF) sources from the paper I wrote with Dr. Walter. Specifically, suppose there are emitters scattered in the interval  $[0, 1]$  on the real line. Possible locations for the emitters are  $x_j = (j - 0.5)/N$  for  $1 \leq j \leq N$  (equispaced points in the interval). We have RF sensors at altitude  $h$  (in the  $y$  direction) and with  $x$  coordinates  $x = a_i$ ,  $1 \leq i \leq M$  (so there are  $M$  sensors). The distance  $r_{ij}$  from the  $i$ th sensor to the  $j$ th potential emitter location is then

$$r_{ij} = \sqrt{(a_i - x_j)^2 + h^2}.$$

Assume the emitter at position  $x_j$  has power  $p_j \geq 0$ . Then the total received signal strength by sensor  $i$  is

$$d_i = \sum_{j=1}^N \frac{p_j}{r_{ij}^2}$$

for  $1 \leq i \leq M$ , if the signal attenuates as  $1/r^2$  and normalizing constants are set to 1. We can write this as

$$\mathbf{A}\mathbf{p} = \mathbf{d}$$

where  $\mathbf{p} = [p_1, \dots, p_N]^T$ ,  $\mathbf{d} = [d_1, \dots, d_M]^T$ , and  $\mathbf{A}$  has entries

$$A_{ij} = \frac{1}{r_{ij}^2} = \frac{1}{(a_i - x_j)^2 + h^2}.$$

By the way, in this problem you can play around with the various tolerances and iteration caps—maybe you'll find a better combination than I suggest below.

- (a) Let  $N = 50$ , and consider the case  $M = 10$  with sensors at  $x = a_i$  where the  $a_i$  are given by  $a_i = (i - 0.5)/10$ ,  $1 \leq i \leq 10$  (equispaced) and the sensors are at altitude  $h = 0.2$ . In Matlab, set up the sensing matrix. Then let  $\mathbf{p}^*$  be the “true” (sparse) power vector with nonzero entries  $p_{17}^* = 1.2$  and  $p_{36}^* = 0.5$ . Compute the data  $\mathbf{d} = \mathbf{A}\mathbf{p}^*$ , then try to recover  $\mathbf{p}^*$  using both OMP and BP (my OMP routine, and the BP routine in the  $L^1$  magic package). Cap the iterations for OMP at 10 with a tolerance of  $10^{-4}$ . You can let BP have 100 iterations, same tolerance. Plot the results (try a stem plot).

What is the coherence of the sensing matrix?

(b) Repeat part (a), but now add a bit of noise to the data vector  $\mathbf{d}$ , as

```
e = 0.1 * norm(d)/sqrt(M) * randn(size(d));  
d = d + e;
```

This adds noise to  $\mathbf{d}$  at about a 10 percent level, in that  $\|\mathbf{e}\|_2 \approx 0.1\|\mathbf{d}\|_2$ , on average.

(c) Repeat part (b) but now with a “noise aware” version of basis pursuit, or with OMP but using a much looser tolerance. Specifically, in OMP use tolerance 10.0 (this is about 1.5 times the noise level).

For basis pursuit recovery use the  $L^1$  magic BPDN routine

```
xp = l1qc_logbarrier(x0, A, [], d, epsilon, 1e - 3);
```

where “epsilon” is  $0.15\|\mathbf{d}\|_2$  the noise level, in the Euclidean norm, comparable to what we used for OMP). Here  $\mathbf{x}_0$  is an initial guess, say  $\mathbf{x}_0 = \text{minnorm}(\mathbf{A}, \mathbf{d})$ ; (the minimum norm solution).

3. Suppose I have data that falls into 3 classes. I collect 2 feature vectors for each class, amalgamate them into a dictionary  $\mathbf{D}$  given by

$$\mathbf{D} = \begin{bmatrix} 1.1 & 3.02 & 0.07 & 0.03 & -1.01 & -2.00 \\ 1.05 & 2.96 & 2.02 & -3.01 & 1.03 & 1.98 \end{bmatrix}.$$

Columns 1 and 2 are class 1, 3 and 4 are class 2, columns 5 and 6 are class 3. I then collect feature vector  $\mathbf{x} = [-2.1, 2.4]^T$ . Find a sparse solution  $\boldsymbol{\alpha}$  to  $\mathbf{D}\boldsymbol{\alpha} = \mathbf{x}$  using OMP, then classify  $\mathbf{x}$  using the method of the text.