Exercise

June 14, 2016

1 Introduction

A linear least square problem has the form

$$minimize ||Fx - f|| (1)$$

where

$$F \in \mathcal{R}^{m \times n}$$
 and $f \in \mathcal{R}^m$

Typically, then

$$m \gg n$$
.

Sometimes the solution to (1) has very large norm i.e. ||x|| is large. This may be undesirable and therefore one may solve

minimize
$$||Fx - f|| + \sum_{j=1}^{n} \lambda_j |x_j|$$
 (2)

instead where $\lambda_j > 0$ is positive parameters. Clearly, the additional term has a regularizing effect. Think about what happens as λ goes to infinity.

See http://www.swissquant.com/data/docs/en/1901/Mathematical-Challenge-March-2014.pdf for a motivation.

Data for the problem can be obtained as follows

```
import lassodata
(F,f) = lassodata.get(1000,100,279)
```

where lassodata.py is on github page. Use $\lambda = 0.1$.

2 Questions

- 1. Formulate (2) as a conic quadratic optimization problem.
 - (a) Is the problem feasible?
 - (b) Is the interior of the feasible solutions nonempty.
 - (c) How many nonzeros are there approximately in the most dense row and column of the constraint matrix.

- 2. Implement the formulated problem in MOSEK Fusion.
- 3. State the dual problem to (2).
 - (a) Is the dual problem feasible?
 - (b) Is the interior of the feasible solutions to the dual nonempty.
 - (c) How many nonzeros are there approximately in the most dense row and column of the constraint matrix of the dual.
- 4. Implement the dual problem of (2) in MOSEK Fusion.