The Georeg Regularizer

Guillermo Angeris

Shane Barratt

Jonathan Tuck*†

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Abstract

In this paper, we introduce the *georeg* regularizer as a natural generalization of the elastic net regularizer by combining *all the norms*. We present a simple example showing some interesting properties of the regularizer and introduce a new cvxpy atom, geo_reg, which implements the georeg regularizer.

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1 Introduction

We define the *p*-norm of a vector $x \in \mathbf{R}^n$ to be¹

$$||x||_p = \left(\sum_{i=1}^n |x_i|^p\right)^{1/p}.$$

Note that any p-norm (and its pth power) is always convex [1]. We will define the mth georeg regularizer as the following convex combination of powers of norms:

$$g_m(x,\alpha) = \left(\sum_{k=1}^m \alpha^{-k}\right)^{-1} \sum_{k=1}^m \frac{\|x\|_k^k}{\alpha^k} = \left(\frac{1-\alpha^{m+1}}{1-\alpha^{-1}} - 1\right)^{-1} \left(\frac{1-\left(\frac{|x_i|}{\alpha}\right)^{m+1}}{1-\frac{|x_i|}{\alpha}} - 1\right), \quad \alpha \neq 1.$$

We note (a) that $g_m(\cdot, \alpha)$ is convex in its domain for any $\alpha > 0$ as it is a sum of convex functions and (b) that the georeg regularizer reduces to the elastic net regularizer if we choose m = 2. Additionally, the series is absolutely convergent as $m \uparrow \infty$ whenever $|x_i| < \alpha$ and $\alpha > 1$ for all $i = 1, \ldots, m$.

In this paper, we focus on the $m=\infty$ case which reduces to

$$g_{\infty}(x,\alpha) = (\alpha - 1) \left(\frac{1}{1 - \frac{|x_i|}{\alpha}} - 1 \right),$$

^{*}Authors listed in alphabetical order.

^{†{}angeris, sbarratt, jonathantuck}@stanford.edu

¹This is well known, but we needed to make the paper longer.

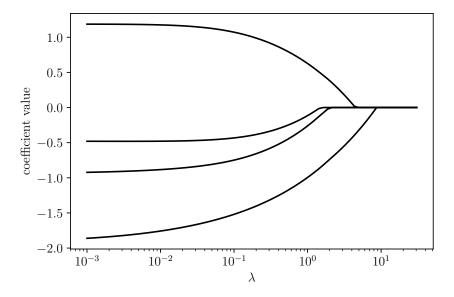


Figure 1: Regularization path.

and is easily representable by a set of second-order cone constraints (SOCs). Let $f : \mathbf{R}^n \to \mathbf{R}$ be a loss function, then the learning problem for x can be written as the following problem with SOCs and variables $x, r, t \in \mathbf{R}^n$:

minimize
$$f(x) + \mathbf{1}^T r - n$$
,
subject to $\|(2, r_i - 1 + t_i/\alpha)\|_2 \le r_i + 1 - t_i/\alpha$, $i = 1, \dots, n$ (1)
 $-t \le x \le t$.

If f is convex and easily representable in conic form, the problem can be written as a conic program, which is often efficiently solvable [1].

2 Example

As an example, we solve the problem

minimize
$$||Ax - b||_2^2 + \lambda g_{\infty}(x, \alpha),$$
 (2)

where $x \in \mathbf{R}^4$ is the optimization variable, and $A \in \mathbf{R}^{100 \times 4}$ and $b \in \mathbf{R}^{100}$ are problem data. We randomly generate A and b and plot the values of the optimal x for varying values of $\lambda > 0$ in figure 1.

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References

[1] S. Boyd and L. Vandenberghe. Convex Optimization. Cambridge University Press, 2004.