

# The Georeg Regularizer

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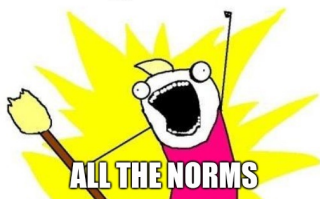
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## Abstract

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.

COMBINE



## 1 Introduction

We define the  $p$ -norm of a vector  $x \in \mathbf{R}^n$  to be<sup>1</sup>

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

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

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

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  $i = 1, \dots, m$ .

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

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

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<sup>1</sup>This is well known, but we needed to make the paper longer.

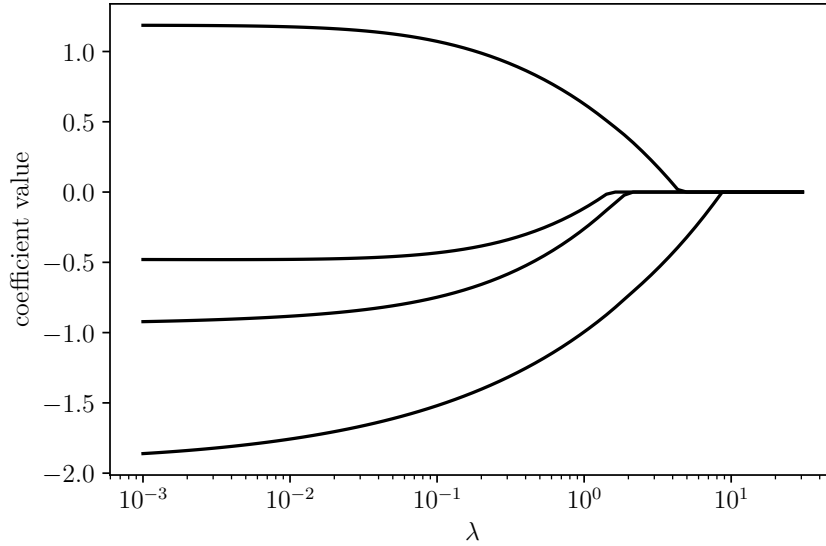


Figure 1: Regularization path.

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

$$\begin{aligned}
 & \text{minimize} && f(x) + \mathbf{1}^T r - n, \\
 & \text{subject to} && \|(2, r_i - 1 + t_i/\alpha)\|_2 \leq r_i + 1 - t_i/\alpha, \quad i = 1, \dots, n \\
 & && -t \leq x \leq t.
 \end{aligned} \tag{1}$$

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

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

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.

## References

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