

Topic 14: Regularization Methods in Thresholded Parameter Space

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Key points: The connections and differences of all regularization methods and some interesting phase transition phenomena.

Disclaimer: The note is built on Prof. *Jinchi Lv*'s lectures of the course at USC, DSO 607, High-Dimensional Statistics and Big Data Problems.

14.1 Model Setup

Now, consider a generalized linear model (GLM) linking a p -dimensional predictor \mathbf{x} to a scalar response Y . With canonical link, the conditional distribution of Y given \mathbf{x} has density

$$f(y; \theta, \phi) = \exp [y\theta - b(\theta) + c(y, \phi)]$$

where $\theta = \mathbf{x}'\boldsymbol{\beta}$ with $\boldsymbol{\beta}$ a p -dimensional regression coefficient vector, $b(\cdot)$ and $c(\cdot, \cdot)$ are known functions and ϕ is dispersion parameter. Again, $\boldsymbol{\beta} = (\beta_{0,1}, \dots, \beta_{0,p})'$ is sparse with many zero components, and $\log p = O(n^a)$ for some $0 < a < 1$.

The penalized negative log-likelihood is

$$Q_n(\boldsymbol{\beta}) = -n^{-1} [\mathbf{y}'\mathbf{X}\boldsymbol{\beta} - \mathbf{1}'\mathbf{b}(\mathbf{X}\boldsymbol{\beta})] + \|p_\lambda(\boldsymbol{\beta})\|_1$$

where

- $\mathbf{y} = (y_1, \dots, y_n)'$, $\mathbf{X} = (\mathbf{x}_1, \dots, \mathbf{x}_n)'$, each column of \mathbf{X} is rescaled to have L_2 -norm \sqrt{n}
- $\mathbf{b}(\boldsymbol{\theta}) = (b(\theta_1), \dots, b(\theta_n))'$ with $\boldsymbol{\theta} = (\theta_1, \dots, \theta_n)'$
- $\|p_\lambda(\boldsymbol{\beta})\|_1 = \sum_{j=1}^p p_\lambda(|\beta_j|)$