

Moderately Clipped Lasso

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1 Poisson distribution

1.1 likelihood function

$$L(\boldsymbol{\beta}) = \sum_{i=1}^n \{y_i \mathbf{x}_i^T \boldsymbol{\beta} - \exp(\mathbf{x}_i^T \boldsymbol{\beta}) - \log(y_i!)\}$$

1.2 gradient

$$\nabla L(\boldsymbol{\beta}) = \sum_{i=1}^n \{\mathbf{x}_i y_i - \mathbf{x}_i \exp(\mathbf{x}_i^T \boldsymbol{\beta})\} = \sum_{i=1}^n \{\mathbf{x}_i (y_i - \exp(\mathbf{x}_i^T \boldsymbol{\beta}))\}$$

1.3 hessian

$$\nabla^2 L(\boldsymbol{\beta}) = \sum_{i=1}^n \{O_{p \times 1} - \mathbf{x}_i \exp(\mathbf{x}_i^T \boldsymbol{\beta}) \mathbf{x}_i^T\} = - \sum_{i=1}^n \{\mathbf{x}_i \exp(\mathbf{x}_i^T \boldsymbol{\beta}) \mathbf{x}_i^T\}$$