

Robust Sparse Principal Component Regression

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Neural Information Processing Systems (NIPS)
Lake Tahoe, Nevada, 2013

Poster number: Fri36

Advantages of Simple PCR over LSE

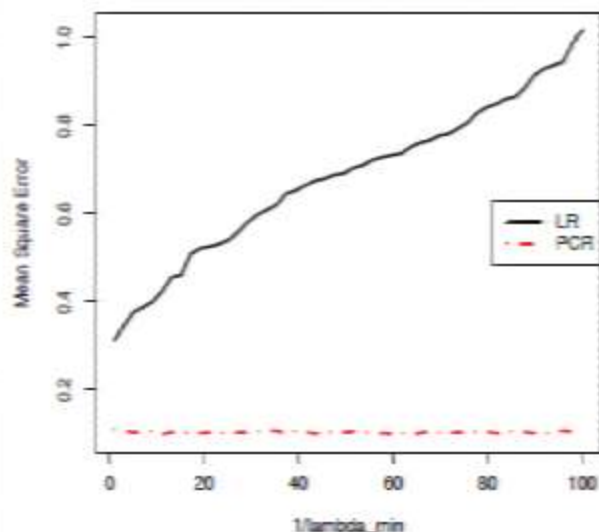
- Principal Component Regression (PCR) model is a **subset** of the linear regression model:

(Classical linear regression model) $Y = \mathbf{X}\beta + \epsilon;$

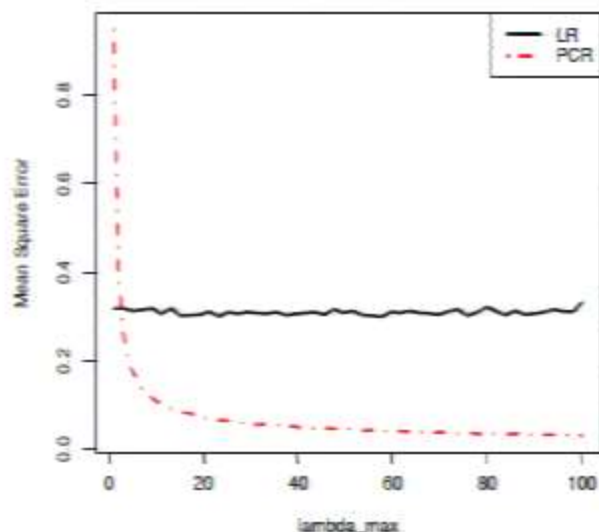
(Simple PCR Model) $Y = \alpha \mathbf{X}u_1 + \epsilon.$

- Three advantages of Simple PCR over LSE:

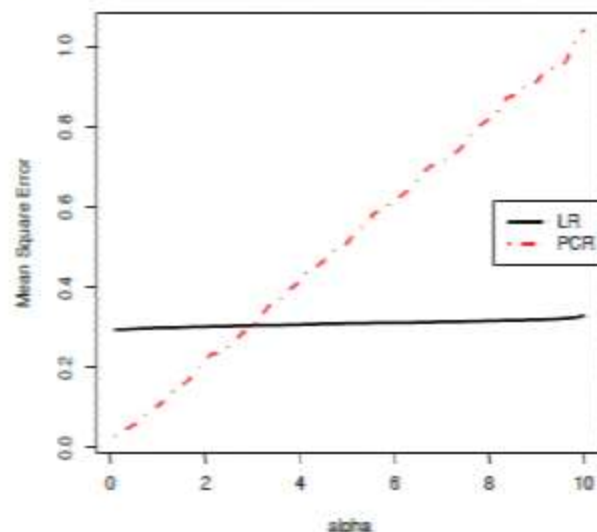
Collinearity ($1/\sigma_{\min}$)



Effective Rank ($\text{tr}(\Sigma)/\sigma_{\max}$)



coefficient α



Robust Sparse PCR

$$\mathbf{Y}_{n \times 1} = \underbrace{\mathbf{X}_{n \times d}}_{\sim (\mathbf{0}, \Sigma)} \underbrace{\mathbf{U}_{d \times k}}_{\text{eigenspace of } \Sigma} \boldsymbol{\alpha}_{k \times 1} + \boldsymbol{\epsilon}_{n \times 1}, \quad d \gg n.$$

Heavy tailed models on \mathbf{X} and $\boldsymbol{\epsilon}$: \mathbf{X} is elliptically distributed (stretched Gaussian: $\boldsymbol{\mu} + \xi \cdot N(\mathbf{0}, \mathbf{S})$), $\boldsymbol{\epsilon}$ has finite fourth moments.

Sparsity on \mathbf{U} and $\boldsymbol{\alpha}$: model constrained for identifiability:

$\|\mathbf{U}\|_0 + \|\boldsymbol{\alpha}\|_0 = s + k \leq n$. Procedure should induce sparsity on \mathbf{U} .

Robust Sparse PCR: Estimating $\boldsymbol{\beta} = \mathbf{U}\boldsymbol{\alpha}$ in optimal (parametric) rate, robust to potential outliers.

Robust Sparse PCR

1. **Model:** Random design matrix, drawn from elliptical distribution. (Light \Rightarrow heavy tails)
2. **Procedure 1:** Conduct sparse PCA on the pairwise self-normalized data $(\mathbf{x}_i - \mathbf{x}_j)/\|\mathbf{x}_i - \mathbf{x}_j\|$. (Elliptical: self-normalization preserves the eigenspace)
3. **Procedure 2:** Project the data to the eigenspace learnt in the second step, then conduct LSE to estimate α .
4. **Theory:** $\|\hat{\beta} - \beta\| = O_P(\sqrt{s \log d/n})$. (optimal rate)

