

Proposal of Double Descent Phenomenon in SURE

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1 SURE

Stein's unbiased risk estimator for $y \sim \mathbb{N}(\theta, \sigma^2 I_n)$ is to minimize the following risk

$$\text{SURE}(\hat{\theta}) = \|\hat{\theta} - y\|^2 - \sigma^2 n + 2\sigma^2 \sum_{j=1}^n \frac{\partial \hat{\theta}_j(y)}{\partial y_j}.$$

We know that James-Stein estimator is a shrinkage estimator. It is natural that SURE is a kind of regularization when applied on linear model.

2 SURE in Linear Regression

In linear regression $y \sim X\beta + \epsilon$, $X \in \mathbb{R}^{n \times p}$, $y \sim \mathbb{N}(\theta, \sigma^2 I_n)$. Assuming σ^2 is known and X is full rank, then

$$\begin{aligned}\hat{\theta} &= \hat{y}, \\ &= X\hat{\beta}, \\ &= X(X^\top X)^\dagger X^\top y, \\ &= Hy. \\ \sum_{j=1}^n \frac{\partial \hat{\theta}_j}{\partial y_j} &= \sum_{j=1}^n H_{jj}, \\ &= \text{tr}(H).\end{aligned}$$

2.1 Low Dimension

When $p < n$, $\text{rank}(X) = p$, then

$$\begin{aligned}H &= X(X^\top X)^{-1}X^\top, \\ \text{tr}(H) &= \text{tr}\left(X^\top X(X^\top X)^{-1}\right), \\ &= \text{tr}(I_p), \\ &= p. \\ \Rightarrow \text{SURE}(\hat{\theta}) &= \|y - X\hat{\beta}\|^2 - n\sigma^2 + 2p\sigma^2.\end{aligned}$$

We can see now n is bonus and p is penalty, which is consistent with regularization we are familiar with.(e.g. AIC)

2.2 High Dimension

If $p > n$, $\text{rank}(X) = n$, we can use rank-retaining factorization to solve the pseudo inverse

$$(X^\top X)^\dagger = X^\top (XX^\top)^{-2} X.$$

Since $XX^\top \in \mathbb{R}^{n \times n}$ is full rank, its matrix inverse exists. Then we can compute the trace of hat matrix

$$\begin{aligned}\text{tr}(H) &= \text{tr} \left(XX^\top (XX^\top)^{-2} XX^\top \right), \\ &= \text{tr}(I_n), \\ &= n. \\ \Rightarrow \text{SURE}(\hat{\theta}) &= \|\hat{\theta} - y\|^2 - n\sigma^2 + 2n\sigma^2 \\ &= \|y - X\hat{\beta}\|^2 + n\sigma^2\end{aligned}$$

Surprisingly, n becomes penalty in the $p > n$ regime.

So when $p < n$, as $\frac{p}{n}$ increasing, the risk follows classical U-shape curve; when $p > n$, as n decreasing i.e. $\frac{p}{n}$ increasing, the risk will decrease, which is very similar to double descent phenomenon.

3 Linear System Estimation

What if just choose $\hat{\beta} = X^\dagger y$ as estimator?

$$\begin{aligned}\hat{\theta} &= \hat{y}, \\ &= X\hat{\beta}, \\ &= XX^\dagger y, \\ &= Hy. \\ X^\dagger &= \begin{cases} (X^\top X)^{-1} X^\top & \text{if } p < n \\ X^\top (XX^\top)^{-1} & \text{if } n > p \end{cases} \\ \text{tr}(H) &= \begin{cases} \text{tr}(I_p) = p & \text{if } p < n \\ \text{tr}(I_n) = n & \text{if } n > p \end{cases}\end{aligned}$$

So the result is consistent with above.

4 Experimental Results

4.1 Using SURE Package

There is a rough experiment with figures of MSE $\|\hat{\theta} - y\|^2$ and SURE risk $\|\hat{\theta} - y\|^2 - \sigma^2 n + 2\sigma^2 \sum_{j=1}^n \frac{\partial \hat{\theta}_j(y)}{\partial y_j}$. The SURE estimator and SURE risk are computed by R package asus. Set $p = 1000$, n from 200 to 10000.

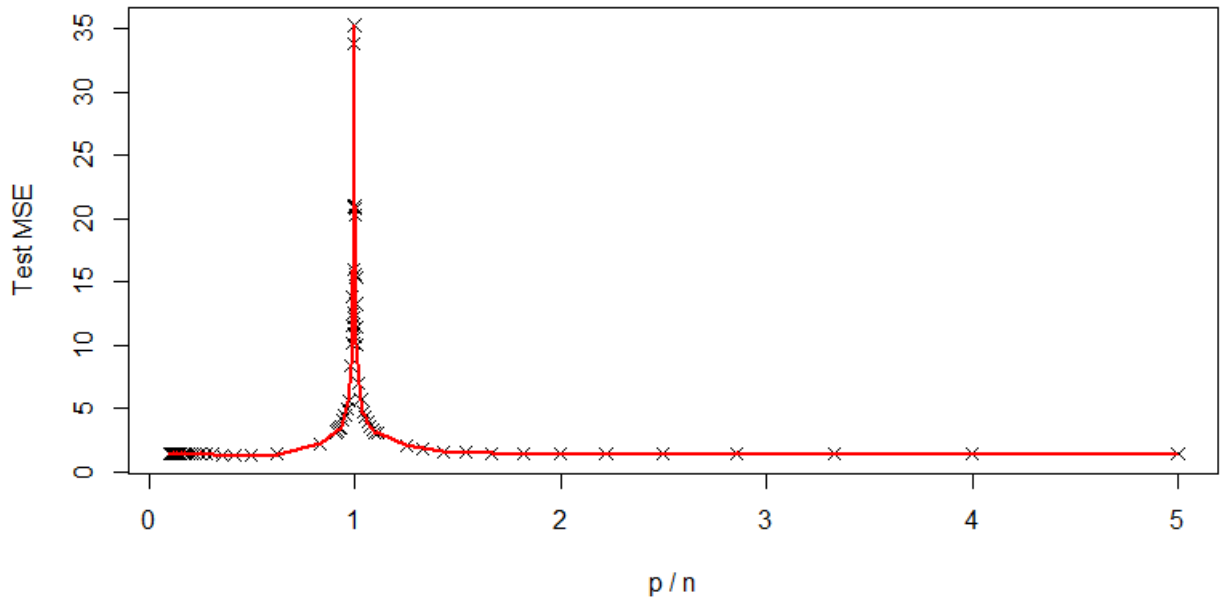


Figure 1: MSE

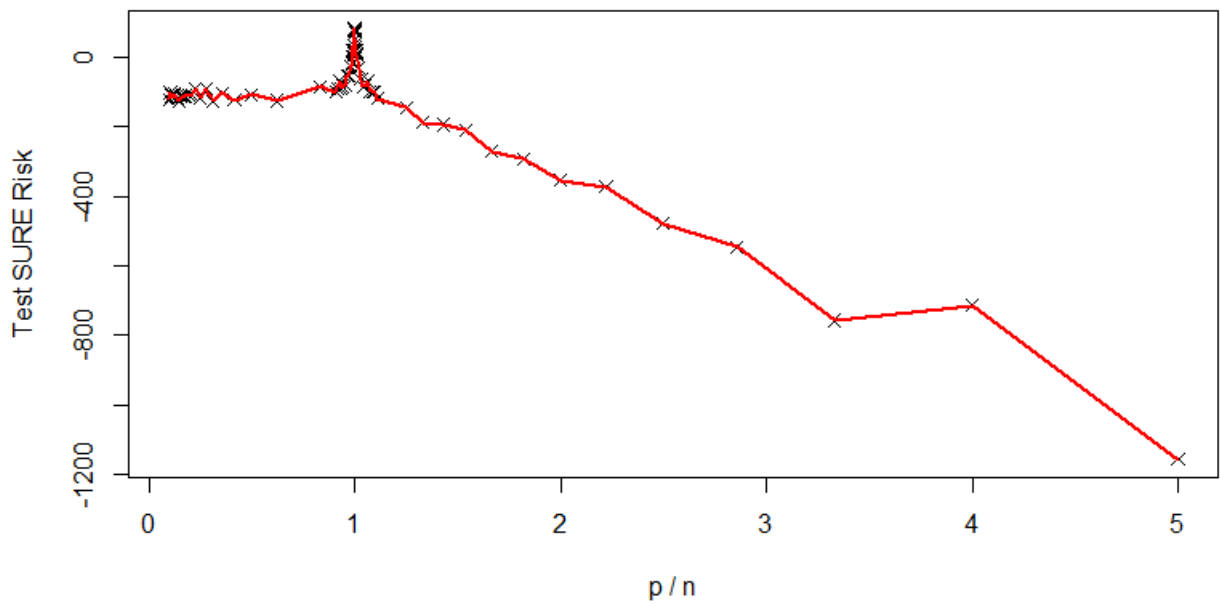


Figure 2: SURE Risk

Both figures indicate the double descent phenomenon. However, the SURE risks are negative at many points, which is unusual even considering the $-\sigma^2 n$ term in the SURE risk. Therefore, I decided to solve the optimization problem directly.

4.2 Using Optimization

Use R package nloptr to solve:

$$\min_{\hat{\beta} \in \mathbb{R}^p} \text{SURE}(\hat{\beta}) = \|y - X\hat{\beta}\|^2 - n\sigma^2 + 2\sigma^2 \text{tr}(H)$$

Since the computation of a large matrix (with $p = 1000$ and n ranging from 200 to 10,000) is quite expensive, I changed the settings to $p = 100$ and n ranging from 50 to 150. The drawback is that points around $\frac{p}{n} = 1$ will be sparse.

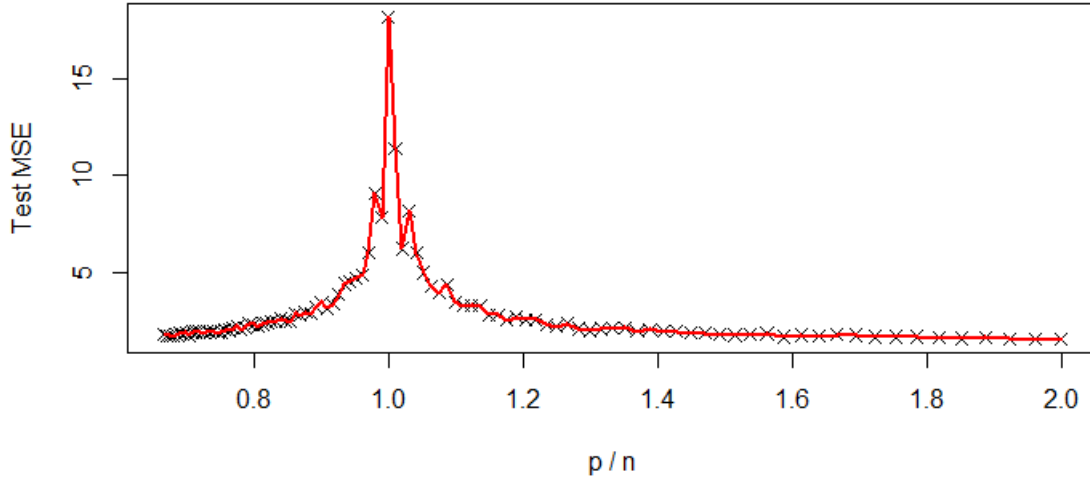


Figure 3: MSE with respect to $\frac{p}{n}$

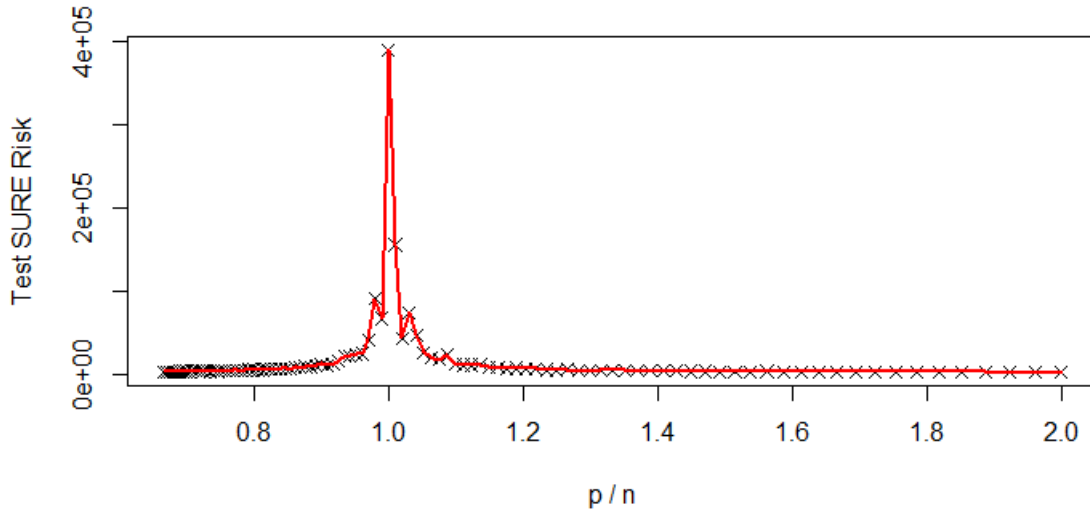


Figure 4: SURE Risk with respect to $\frac{p}{n}$

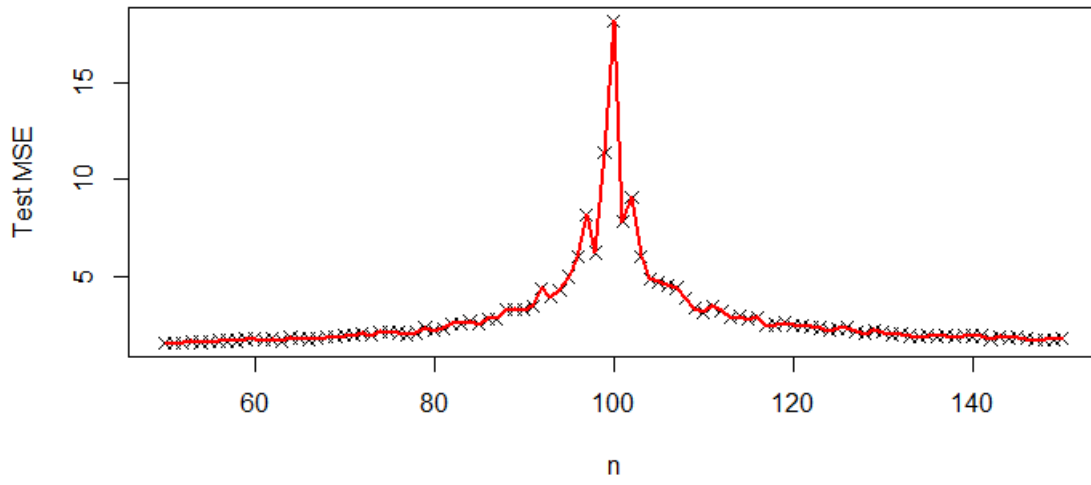


Figure 5: MSE with respect to n

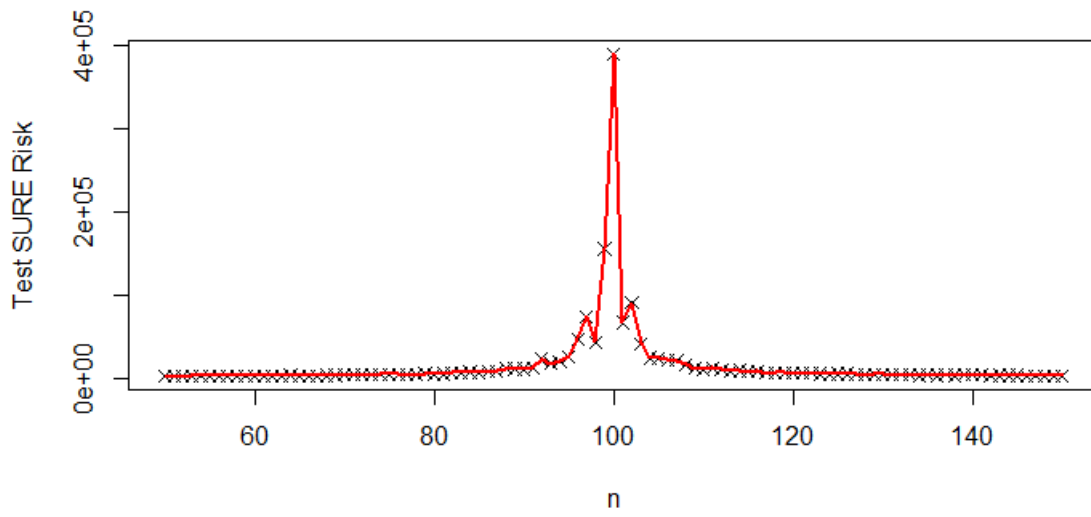


Figure 6: SURE Risk with respect to n

We can see obvious double descent phenomenon in above figures.