Model Selection And Sharp Asymptotic Minimaxity

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Problem description

- **Multivariate normal mean problem:**

$$y_i = \theta_i + \sigma_n z_i,\ z_i \overset{i.i.d.}{\sim} N(0,1), i = 1, ..., n.$$

- **Sparse parameter space:** $\theta \in \Theta_{n,p}(\eta_n)$, $\Theta_{n,p}(\eta_n)$ is one of

$$l_0 \text{ ball} : \{\theta \in \mathbb{R}^n : \|\theta\|_0 \leq \eta_n n\}$$

$$l_p \text{ ball} : \left\{\theta \in \mathbb{R}^n : \sum_{i=1}^{n} |\theta_i|^p \leq \eta_n^p n, 0 < p < 2\right\}$$

$$m_p \text{ ball} : \left\{\theta \in \mathbb{R}^n : |\theta|_{[k]} \leq \eta_n \left(\frac{n}{k}\right)^{1/p}, 0 < p < 2, k = 1, ..., n\right\}$$

where $\eta_n \in \left[n^{-1}\log^{\gamma} n, b_2 n^{-b_3}\right]$, $\gamma > 4.5$.

- **Goal:** to estimate $\theta = (\theta_1, ..., \theta_n)$.

Main theorems

- Penalized estimation: $\hat{\theta} = \underset{\|\mu\|_0 \le n/\log n}{\arg\min} \left[\|y-\mu\|_2^2 + \mathrm{Pen}\left(\|\mu\|_0\right) \right]$.
- Main theorems: Let $\|\mu\|_0 = k$, $\mathrm{Pen}(k) = \sum_{i=1}^{k} u_i^2$, $R_n\left(\Theta_{n,p}\left(\eta_n\right)\right)$ be the minimax risk,

Theorem

For some $0 < \varepsilon < 1$, *and any constant* c', *let*
$2\log\frac{n}{i} - (1-\varepsilon)\log\log\frac{n}{i} \le u_i^2 \le 2\log\frac{n}{i} + c'\log\log n$, *then*
$\sup_{\theta\in\Theta_{n,p}(\eta_n)} \mathbb{E}\left\|\hat{\theta}-\theta\right\|_2^2 \sim R_n\left(\Theta_{n,p}\left(\eta_n\right)\right)$.

Theorem

Let $u_i^2 = c_n \log\frac{n}{i}$, $c_n \to c > 2$, *then*
$\sup_{\theta\in\Theta_{n,p}(\eta_n)} \mathbb{E}\left\|\hat{\theta}-\theta\right\|_2^2 \sim c^* R_n\left(\Theta_{n,p}\left(\eta_n\right)\right)$, *where* $c^* = \frac{c}{2}$ *for* l_0 *ball,*
$c^* = \left(\frac{c}{2}\right)^{1-p/2}$ *for* l_p *ball and* m_p *ball.*

Interpretation and significance

Interpretation

- L0 norm leads to minimaxity if $\mathrm{Pen}(k)$ is in a sharp range of $2k \log \frac{n}{k}$.
- L0 norm leads to non-minimaxity if $\mathrm{Pen}(k)$ is not dominated by $2k \log \frac{n}{k}$.

Significance

- Minimax estimation procedure for $\mathrm{Pen}(k)$
  - Foster and Stine (1999): $2\sum_{i=1}^{k} \log \frac{n}{i}$.
  - George and Foster (2000): $2\sum_{i=1}^{k} \log\left(\frac{n+1}{i} - 1\right)$.
  - Berge and Massart (2001): $2k \log \frac{n}{k}$.
  - ABDJ (2006): $\sum_{i=1}^{k} \left(\tilde{\Phi}^{-1}\left(\frac{q_n i}{2n}\right)\right)^2$.
- Non-minimax estimation procedure for $\mathrm{Pen}(k)$
  - Tibshirani and Knight (1999): $\sum_{i=1}^{k} \log \frac{n}{i}$.
  - Abramovich, Grinshtein, Pensky (2007): $ck \log \frac{n}{k}$, some $c > 2$.