

Announcements

- HW 4: Typo in v_{boot} (now fixed)
- Quiz 6 (11/22) will be optional
- Cheating: HW must be written alone
Quizzes no talking, looking at other people, etc.

Parameter Inference

$$\mathcal{F} = \{f_\theta(x) : \theta \in \Theta\} \quad \text{parameter model}$$

$$\text{Ex. } \mathcal{F} = \left\{ \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right), (\mu, \sigma) \in \mathbb{R} \times \mathbb{R}_{>0} \right\}$$

$$\theta = (\mu, \sigma) \quad \Theta = \mathbb{R} \times \mathbb{R}_{>0}$$

Goal: learn θ from data

Def. A parameter of interest is some quantity (depend on θ) that we care about.

$$\text{Ex. } \mu = T(\theta)$$

$$\text{Ex. } \tau = 1 - \Phi\left(\frac{1-\mu}{\sigma}\right), \quad \Phi(x) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{z^2}{2}\right) dz$$

Def. The j -th moment of a RV X is $\mathbb{E}[X^j]$.

Def. The j -th moment of a distribution F_θ is

$$\alpha_j = \alpha_j(\theta) = \int X^j dF_\theta(x) = \mathbb{E}[Z^j]$$

$\underbrace{\hspace{10em}}_{Z \sim F}$

Def. The j -th sample moment of iid data

$$X_1, \dots, X_n \text{ is } \hat{\alpha}_j = \frac{1}{n} \sum_{i=1}^n X_i^j$$

Idea: find the value of θ that gives the same moments as what are observed in the data.

Def. Let $\mathcal{F} = \{F_\theta(x) : \theta \in \Theta\}$, where $\Theta \in \mathbb{R}^k$.

The method of moments estimator is

the parameter $\hat{\theta}_n \in \mathbb{R}^k$ such that
 $\searrow (\hat{\theta}_{n,1}, \hat{\theta}_{n,2}, \dots, \hat{\theta}_{n,k})$

$$\hat{\alpha}_1 = \alpha_1(\hat{\theta}_n) = \alpha_1(\hat{\theta}_{n,1}, \dots, \hat{\theta}_{n,k})$$

$$\hat{\alpha}_2 = \alpha_2(\hat{\theta}_n) = \alpha_2(\hat{\theta}_{n,1}, \dots, \hat{\theta}_{n,k})$$

\vdots

$$\hat{\alpha}_k = \alpha_k(\hat{\theta}_n) = \alpha_k(\hat{\theta}_{n,1}, \dots, \hat{\theta}_{n,k})$$

This is k equations and k unknowns.

Ex. $\mathcal{F} = \{ \text{Bernoulli with param } p \}$

$$\theta = p, \quad \mathcal{H} = (0, 1)$$

$$\hat{\alpha}_1 = \frac{1}{n} \sum_{i=1}^n X_i, \quad \alpha_1(\theta) = \mathbb{E}[X] = 1 \cdot p + 0 \cdot (1-p) = p$$

$X \sim \text{Ber}(p)$

Solve: $\hat{p}_n = \frac{1}{n} \sum_{i=1}^n X_i$ for \hat{p}

Ex. $\mathcal{F} = \{ F_{\mu, \sigma^2} : \mu = \int x dF(x), \sigma^2 = \int (x-\mu)^2 dF(x), \mu \in \mathbb{R}, \sigma^2 > 0 \}$

$$\hat{\alpha}_1 = \frac{1}{n} \sum_{i=1}^n X_i$$

$$\alpha_1 = \mu$$

$$\hat{\alpha}_2 = \frac{1}{n} \sum_{i=1}^n X_i^2$$

$$\alpha_2 = \mathbb{E}[X^2] = \text{V}[X] + \mathbb{E}[X]^2 = \sigma^2 + \mu^2$$

$X \sim F_\theta$

Solve: $\hat{\mu}_n = \frac{1}{n} \sum_{i=1}^n X_i$

$$\hat{\mu}_n^2 + \hat{\sigma}_n^2 = \frac{1}{n} \sum_{i=1}^n X_i^2$$

$$\Rightarrow \hat{\mu}_n = \frac{1}{n} \sum_{i=1}^n X_i$$

$$\hat{\sigma}_n^2 = \frac{1}{n} \sum_{i=1}^n X_i^2 - \left(\frac{1}{n} \sum_{i=1}^n X_i \right)^2 = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X}_n)^2$$

Theorem

Let $\hat{\theta}_n$ be the method of moments estimator. Under appropriate conditions:

1. $P[\hat{\theta}_n \text{ exists}] \rightarrow 1$

2. $\hat{\theta}_n \xrightarrow{P} \theta$ (consistent)

3. $\sqrt{n}(\hat{\theta}_n - \theta) \rightarrow N(0, \Sigma)$

where $\Sigma = g \mathbb{E}[y y^T] g^T$,

$$y = (x, x^2, \dots, x^k)^T, \quad g = (g_1, \dots, g_k)$$

$$g_j = \frac{\partial \alpha_j^{-1}(\theta)}{\partial \theta}$$