

浙大, 2023.

### Problem ①:

- (1)  $X_1, \dots, X_n \sim F$ , derive the delta ~~est~~ method estimation of Variance
- (2) Write the definition of influence function of  $F$ .
- (3) Derive the asymptotic distribution of  $\sqrt{n}(T(\hat{F}_n) - T(F))$ ,  $T(F)$  is linear functional.

### Problem ②:

- (1) Write the definition of kernel density estimator.
- (2)  $\hat{f}_n^h(x)$  is the kernel density estimator, write the bias and variance of  $\hat{f}_n^h(x)$
- (3) Give a practical way to pick out an  $h$  in  $\hat{f}_n^h(x)$ , ~~in~~ to minimize  $E \int (\hat{f}_n^h(x) - f(x))^2 dx$

### Problem ③:

- (1) Write the definition of linear smoother.
- (2) Write the effective degree of freedom of linear smoother, explain why it is a reasonable measure of a degree of freedom.
- (3) Write the definition of Nadaraya-Watson kernel estimator, explain why it is called as local constant estimator.
- (4) Write the definition of natural spline. ~~When~~ When does natural spline come into nonparametric regression, why we need natural spline?
- (5)  $Y_i = r(X_i) + \sigma(X_i)\varepsilon_i$ , explain how to derive  $\hat{\sigma}^2$  here.

### Problem ④:

- $Z_i = \theta_i + \sigma \varepsilon_i$ ,  $\varepsilon_i \sim N(0, 1)$ ;  $\theta \in \{\theta: \sum_{i=1}^n \theta_i^2 \leq C^2\}$ .
- (1) How does normal-means problem related to nonparametric regression?
  - (2) Write the definition of J-S estimator.
  - (3) Derive the J-S estimator. (hint: use SURE or Bayes).
  - (4) Explain why J-S estimator is good (hint: use Pinsker's theorem).
  - (5) ~~Explain~~ Explain why the assumption  $\theta \in \{\theta: \sum_{i=1}^n \theta_i^2 \leq C^2\}$  is reasonable.

### Problem ⑤:

- (1) Write the definition of Fourier transform.
- (2) Explain one limitation of Fourier transform in detail.
- (3) Explain why wavelet can solve the limitation if you explained in (2), explain in detail.
- (4) Let  ~~$f_{jk}(x) = 2^{j/2} f(2^j x - k)$~~   $f_{jk}(x) = 2^{j/2} f(2^j x - k)$   $\phi(x) = I(0 \leq x \leq 1)$   $\psi(x) = -I(0 \leq x \leq \frac{1}{2}) + I(\frac{1}{2} < x \leq 1)$   
and  $W_j = \{\psi_{jk}(x), k=0, 1, \dots, 2^j-1\}$ .  
Prove:  $\{\phi, w_0, w_1, \dots, w_j\}$  is an orthonormal basis.
- (5) When using wavelet in nonparametric regression, how will you estimate the coefficient? explain in detail.