Chapter 7 - Estimating CDF

Q1: prove Theorem 7.3

$$E\left(\hat{F_n}(x)\right) = F(x)$$

$$\hat{F_n} = \frac{1}{n} \sum I(Xi\langle x) \Rightarrow E(\hat{F_n}) = \frac{1}{n} \sum E(I(Xi\langle x))$$

$$E(I(X; \leq x)) = P(X; \leq x) = F(x)$$

$$\Rightarrow E(\hat{F}_n) = F(x)$$

$$V\left(\stackrel{\wedge}{F_n}(x)\right) = \frac{1}{n}F(x)\left(1-F(x)\right)$$

$$V(\hat{F_n}) = \frac{1}{n^2} \sum V(I(Xi \leqslant x))$$

$$V(I(Xi(x)) = E(I(Xi(x)^{2}) - E(I(Xi(x))^{2}) = F(x) - F(x)^{2}$$

$$\Rightarrow V(\hat{F}_{n}) = \frac{1}{n} F(x) (1 - F(x))$$

$$MSE = \frac{1}{n} F_{(x)} (1 - F_{(x)}) \rightarrow 0$$

$$MSE = bias(\hat{F})^2 + V(\hat{F}) = \frac{1}{n} F(x) (1 - F(x))$$

$$\hat{F}_n \xrightarrow{P} F(x)$$

$$MSE \Rightarrow 0 \Rightarrow \stackrel{\widehat{F}}{F_n} \stackrel{qm}{\Rightarrow} F(x) \Rightarrow \stackrel{\widehat{F}}{F_n} \stackrel{P}{\Rightarrow} F(x)$$

Q2: X1,..., Xn ~ Bernoulli(p) y,..., y, ~ Bernoulli(q) find estimator & se for p & p-q find 90% conf. interval for ps p-q Estimator for ρ : $\hat{\rho} = \overline{X_n}$, $\hat{Se}(\hat{\rho}) = \frac{\delta^2}{2} = \frac{1}{2} \hat{\rho}(1-\hat{\rho})$ Estimator for p-q: $\hat{p} - \hat{q} = X_n - \overline{Y}_n$, $\hat{se}^2(\hat{p} - \hat{q}) = \prod_{n} [\hat{p}(1-\hat{p}) + \hat{q}(1-\hat{q})]$ 90% conf. interval: p + Zoil se $Z_{0.05} = 1.65 \Rightarrow 90\% CI \Rightarrow \overline{\chi}_{n} \pm 1.65 \hat{Se}(\hat{p})$ | xn-yn + 1.65 se(p-q) Q3: computer experiment find solution in Jupyter note book. Q4: X1, ..., Xn~F8 empirical dist. F use CLT to find limiting dist. of F $F_n = \frac{1}{n} \sum I(X_i \langle x \rangle)$, assume x fixed, $Z_i = I(X_i \langle x \rangle)$ Zi are IID => Using CLT we have: $\hat{F}_n = Z \longrightarrow \mathcal{N}(\mu, \frac{\sigma^2}{n})$ where $\int \mu = E(Z_i) = F(x_i)$ 02 / F(x) (1-F(x))

Q5:
$$Cov(\hat{F}_n(x), \hat{F}_n(y)) = ?$$

 $Cov(\hat{F}(x), \hat{F}(y)) = E(\hat{F}_x \hat{F}_y) - E(\hat{F}_x) E(\hat{F}_y)$
 $= E(\hat{F}_x \hat{F}_y) - E(\hat{F}_x \hat{F}_y) = E(\hat{F}_x) E(\hat{F}_y)$

$$E\left(\hat{F}_{x}\hat{F}_{y}\right) = \frac{1}{n^{2}}E\left[\sum_{i}I(X_{i}\leqslant x)\sum_{j}I(X_{j}\leqslant y)\right]$$

$$= \frac{1}{n^{2}}E\left[\sum_{i=j}I(X_{i}\leqslant x)I(X_{j}\leqslant y) + \sum_{i\neq j}I(X_{i}\leqslant x)I(X_{j}\leqslant y)\right]$$

$$= \frac{1}{n^{2}}\left[nE(I(X_{i}\leqslant x)I(X_{i}\leqslant y)) + n(n-1)E(I(X_{i}\leqslant x)I(X_{2}\leqslant y))\right]$$

$$F(min(x,y))$$

$$F(x)F(y)$$

Q6:
$$X_1, \dots, X_n \sim F$$
, empirical \hat{F}_n
let $a < b$, $\theta = T(F) = F(b) - F(a)$

$$\hat{\theta} = T(\hat{F}) = \hat{F}(b) - \hat{F}(a)$$

$$\hat{Se}(\hat{\theta}) = ?$$
 1-a conf. interval for θ ?

$$V(\hat{\theta}) = V(\hat{F}(b)) + V(\hat{F}(a)) - 2Cov(\hat{F}(b),\hat{F}(a))$$

=>
$$\hat{Se}(\hat{\theta}) = \frac{1}{n} (\hat{F}(b) - \hat{F}(a)) (1 - \hat{F}(b) + \hat{F}(a))$$

1- α Confidence interval: $\hat{\theta} \pm Z_{\frac{1}{2}} \hat{Se}(\hat{\theta})$

Q7: Computer experiment

find solutions in Jupyter notebook.

Q8: Computer experiment

find solutions in Jupyter notebook.

$$\theta = P_1 - P_2 \implies \hat{\theta} = ?$$
 $\hat{Se}(\hat{\theta}) = ?$ 80%, 95% CI

Same as question 2:
$$\hat{\theta} = \hat{\rho}_1 - \hat{\rho}_2 = 0.9 - 0.85 = 0.05$$

80% C.I.
$$\Rightarrow Z_{\frac{1}{2}} = 1.28$$
 C.I. $= 0.05 \pm Z_{\frac{1}{2}} \times 0.047$
95% C.I. $\Rightarrow Z_{\frac{1}{2}} = 1.96$

Q10: Computer experiment

find solutions in Jupyter notebook.