

## Chapter 5

Markov's Inequality

Chebyshev's Inequality

Weak Law of Large Numbers

Continuous Mapping Theorem : 어떤 확률변수들이 확률수렴하면, 확률변수들의 선형결합도 확률수렴한다.

Delta Method

## Chapter 6

Functional Invariance Property of MLE : 임의의 확률변수의 분포와 MLE를 알면, 어떤 분포와 관련된 추정치는 반드시 MLE의 함수이다.

Jensen's Inequality :  $\phi(E(X)) \leq E(\phi(X))$

Rao-Cramer Lower Bound

Efficiency :  $\frac{(nI''(\theta))^{-1}}{\text{Var}(\hat{\theta})}$

Asymptotical Normality of MLE :  $\sqrt{n}(\hat{\theta} - \theta) \xrightarrow{D} N(0, I''(\theta))$

Asymptotic Efficiency

## Chapter 7

Sufficient Statistic : (i)  $Y$  is a statistic (ii)  $P_\theta[(X_1, \dots, X_n) \in A | T(X) = y]$  does not depend on  $\theta$  for each  $y$  and all  $A$ .

Factorization Theorem : exponential family가 아닐 경우 sufficient statistic을 알려주고, exponential family일 경우 Complete Sufficient Statistic

Rao-Blackwell Theorem : Rao-Blackwell에 의해, 좋은 함수는 CSS의 함수이어야 하는데, CSS는 유일하다.

Complete Sufficient Statistic

Minimum Variance Unbiased Estimator

Lehmann-Scheffe Theorem :  $E(\phi(T)) = \theta \Rightarrow \phi(T)$  is the MVUE

## Chapter 8

Most Powerful Test

Uniformly Most Powerful Test

Neyman-Pearson Theorem :  $\Delta$ 가 1보다 작을수록 기각하고, 1에 가까울수록 기각하지 않겠다는 의미

Monotone Likelihood Ratio : MLR Property를 가지면 UMP가 존재하고, MP와 동일하다.

Likelihood Ratio Test :  $-2(\ell(\hat{\theta}_0) - \ell(\hat{\theta})) > k$

Wald Test :  $(\sqrt{n}I(\hat{\theta})(\hat{\theta} - \theta_0))^2 > k$

Score Test :  $\left(\frac{\ell'(\hat{\theta}_0)}{\sqrt{n}I(\hat{\theta}_0)}\right)^2 > k$

Asymptotic Likelihood Ratio Test :  $-2\log \Delta = -2(\ell(\hat{\theta}_0) - \ell(\hat{\theta})) \xrightarrow{D} \chi^2(1)$