

마르코프 : $u(x) > 0, E(u(X)) < \infty, P[u(X) \geq c] \leq \frac{E[u(X)]}{c}$

체비셰프 : $X: r.v, \text{분산 } \sigma^2 < \infty, E(X) = \mu, \text{ 모든 } K > 0 \text{ 에 대해 } P(|X - \mu| \geq K\sigma) \leq \frac{1}{K^2}$

Convergence in probability : $X_n \xrightarrow{P} X \Leftrightarrow \forall \epsilon > 0, \lim_{n \rightarrow \infty} P(|X_n - X| \geq \epsilon) = 0$

$\bar{X}_n = \frac{1}{n} \sum X_i \Rightarrow \bar{X}_n \xrightarrow{P} \mu$ 증명 : $P(|\bar{X}_n - \mu| \geq \epsilon) = P(|\bar{X}_n - \mu| \geq \frac{(\frac{\sqrt{n}\epsilon}{\sigma}) \cdot (\frac{\sigma}{\sqrt{n}})}) \leq \left(\frac{\sigma^2}{n\epsilon^2}\right) \xrightarrow{n \rightarrow \infty} 0$

Convergence in distribution : $X_n \xrightarrow{D} X \Leftrightarrow \lim_{n \rightarrow \infty} F_{X_n}(x) = F_X(x) \quad \forall x \in C(F_X)$
 $\xrightarrow{F_X \text{가 연속한 모든 점의 집합}}$

Slutsky th. : $X_n, X, A_n, B_n: r.v, a, b: \text{st}, X_n \xrightarrow{D} X, A_n \xrightarrow{P} a, B_n \xrightarrow{P} b$
 $\Rightarrow A_n + B_n X_n \xrightarrow{D} a + bX$

Delta method : $\sqrt{n}(X_n - \theta) \xrightarrow{D} N(0, \sigma^2), g(x): \theta \text{에서 미분가능}, g'(\theta) \neq 0$
 $\Rightarrow \sqrt{n}(g(X_n) - g(\theta)) \xrightarrow{D} N(0, \sigma^2 [g'(\theta)]^2) \quad / \quad \frac{X_n - \theta}{\sigma/\sqrt{n}} \xrightarrow{D} N(0, 1) \Rightarrow \frac{g(X_n) - g(\theta)}{\frac{\sigma}{\sqrt{n}} g'(\theta)} \xrightarrow{D} N(0, 1)$

$g'(\theta) = 0, g''(\theta) \neq 0 \Rightarrow n(g(X_n) - g(\theta)) \xrightarrow{D} \sigma^2 \frac{g''(\theta)}{2} \chi_1^2 = \Gamma(\frac{1}{2}, \sigma^2 g''(\theta))$

증명 idea : $g(X_n) = g(\theta) + g'(\theta)(X_n - \theta) + g''(\theta) \frac{(X_n - \theta)^2}{2} + o_p(|X_n - \theta|^2)$

CLT : $X_1, \dots, X_n: \text{id r.v}, \mu, \sigma^2, Y_n = \frac{\sum X_i - n\mu}{\sqrt{n}\sigma} = \frac{\bar{X}_n - \mu}{\sigma/\sqrt{n}} \xrightarrow{D} N(0, 1)$

증명 : $Z_i = \frac{X_i - \mu}{\sigma}, \sum Z_i = \frac{\sum X_i - n\mu}{\sigma} \quad Y_n = \sum Z_i \cdot \frac{1}{\sqrt{n}}, M_{Y_n}(t) = \frac{1}{\sqrt{n}} \sum M_{Z_i}(\frac{t}{\sqrt{n}}) \stackrel{\text{iid}}{=} \left(M_Z(\frac{t}{\sqrt{n}})\right)^n$
 $= \left(\sum_{k=0}^{\infty} \frac{M_Z^{(k)}(\mu)}{k!} \left(\frac{t}{\sqrt{n}}\right)^k\right)^n \xrightarrow{n \rightarrow \infty} e^{-\frac{t^2}{2}}$

이항분포 : $p(x) = \binom{n}{x} p^x (1-p)^{n-x}, M(t) = [(1-p) + pe^t]^n, \mu = np, \sigma^2 = np(1-p)$

음이항분포 : $p(x) = \binom{x+r-1}{r-1} p^{r-1} (1-p)^x, M(t) = p^r (1 - (1-p)e^t)^{-r}, \mu = r \frac{1-p}{p}, \sigma^2 = r \frac{1-p}{p^2}$

다항분포 : $p(x_1, \dots, x_k) = \frac{n!}{x_1! \dots x_k!} p_1^{x_1} \dots p_k^{x_k}, M(t_1, \dots, t_k) = (p_1 e^{t_1} + p_2 e^{t_2} + \dots + p_k e^{t_k} + p_{k+1})^n$

초기하분포 : $p(x) = \frac{\binom{N-D}{n-x} \binom{D}{x}}{\binom{N}{n}}, \mu = \frac{Dn}{N}, \sigma^2 = n \cdot \frac{D}{N} \cdot \frac{N-D}{N} \cdot \frac{N-n}{N-1}$

포아송분포 : $p(x) = \frac{\lambda^x e^{-\lambda}}{x!}, M(t) = e^{\lambda(e^t - 1)}, \mu = \lambda, \sigma^2 = \lambda$

- 정규분포
- $P(A) \geq 0 \quad \forall A \in \mathcal{F}$
 - $P(\Omega) = 1$
 - $A_n, A_n \text{ 독립} \Rightarrow P(\bigcap_{n=1}^{\infty} A_n) = \prod_{n=1}^{\infty} P(A_n)$

정규분포 : $f(x) = \frac{1}{\sqrt{2\pi}\sigma} \exp(-\frac{(x-\mu)^2}{2\sigma^2}), M(t) = \exp(\mu t + \frac{1}{2}\sigma^2 t^2)$

$\Gamma(x) = \int_0^{\infty} y^{x-1} e^{-y} dy, \Gamma(1) = 1, \Gamma(x) = (x-1)\Gamma(x-1) = (x-1)!, \Gamma(\frac{1}{2}) = \sqrt{\pi}$

감마분포 : $\Gamma(\alpha, \beta), f(x) = \frac{1}{\Gamma(\alpha)\beta^\alpha} x^{\alpha-1} e^{-\frac{x}{\beta}}, M(t) = (1 - \beta t)^{-\alpha}, \mu = \alpha\beta, \sigma^2 = \alpha\beta^2$

$X_i \sim \Gamma(\alpha_i, \beta) \text{ 이면 } \sum X_i \sim \Gamma(\sum \alpha_i, \beta) \text{ 이고, } kX \sim \Gamma(\alpha, k\beta) \text{ 이다}$

$W_k: \text{Poi}(\lambda) \text{에서 } k\text{번째 사건이 일어나기까지 걸린 시간}, W_k \sim \Gamma(k, \frac{1}{\lambda})$

지수분포 : $\text{Exp}(\lambda), \Gamma(1, \frac{1}{\lambda}), f(x) = \lambda e^{-\lambda x}, F(x) = 1 - e^{-\lambda x}, \mu = \frac{1}{\lambda}, \sigma^2 = \frac{1}{\lambda^2}$

카이제곱분포 : $\chi^2(r) = \Gamma(\frac{r}{2}, \frac{1}{2}), f(x) = \frac{1}{\Gamma(\frac{r}{2}) 2^{\frac{r}{2}}} x^{\frac{r}{2}-1} e^{-\frac{x}{2}}, M(t) = (1 - 2t)^{-\frac{r}{2}} \quad (t < \frac{1}{2})$
 $\mu = r, \sigma^2 = 2r \quad / \quad X \sim N(\mu, \sigma^2) \Rightarrow V = \frac{(X-\mu)^2}{\sigma^2} \sim \chi^2(1)$