Continuous Distributi- Chi-Square Distribution

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Uniform Distribution

$$f(k) = \frac{1}{b-a+1}.$$

$$M(t) = \frac{1}{b-a+1} \frac{e^{at}(1-e^{b-a+1})}{1-e^t}.$$

$$\mu = \frac{a+b}{2}, \quad \sigma^2 = \frac{(b-a+1)^2 - 1}{12}.$$

Discrete Distributions

$$\mu = \frac{a+b}{2}$$
, $\sigma^2 = \frac{(b-a+1)^2 - 12}{12}$

Geometric Distribtuion

$$f(k) = (1 - p)^{k-1} p.$$

Hypergeometric Distributi-

$$f(k) = \frac{\binom{K}{k}\binom{N-K}{n-k}}{\binom{N}{n}}.$$

where N is the population size, K is the number of success states, n is the number of draws, *k* is the number of observed successes.

$$\mu=n\frac{K}{N},\quad \sigma^2=n\frac{K}{N}\frac{N-K}{N}\frac{N-n}{N-1}.$$

Binomial Distribution

$$f(k) = \binom{n}{k} p^k (1-p)^{n-k}.$$

$$\mu = np$$
, $\sigma^2 = np(1-p)$.

Negative Binomial Distribu-

$$f(k) = {\binom{k-1}{r-1}} p^n (1-p)^{k-r}.$$

When n = 1, it is a geometric distributi-

$$M(t) = \left(\frac{pe^t}{1 - (1 - p)e^t}\right)^r.$$

$$M'(t) = r(pe^t)^r [1 - (1-p)e^t]^{-r-1}.$$

$$\mu=\frac{r}{p}, \quad \sigma^2=\frac{r(1-p)}{p^2}.$$

Poisson Distribution

$$f(k) = \frac{\lambda^x e^{-\lambda}}{x!}$$
.

Uniform Distribution

$$f(x) = \frac{1}{b-a}, \quad F(x) = \frac{x-a}{b-a}.$$

$$M(t) = \frac{e^{tb} - e^{ta}}{t(b-a)}.$$

$$M'(t) = \frac{(bt-1)e^{bt} - (at-1)e^{at}}{t^2(b-a)}.$$

$$M\boxtimes(t) = \frac{[(bt-1)^2 - 1]e^{bt} - [(at-1)^2 - 1]e^{at}}{t^3(b-a)}. \qquad M(t) = \exp\left(\mu t + \frac{\sigma^2 t^2}{2}\right).$$

$$\mu = \frac{a+b}{2}, \quad \sigma^2 = \frac{(b-a)^2}{12}$$

Expotential Distribution

$$\theta = \frac{1}{\lambda}$$
.

where λ is the mean number of occurrences in the unit time interval so θ is the meaning waiting time for the first occur-

$$f(x) = \frac{1}{\Theta}e^{-x/\theta}, \quad F(x) = 1 - e^{-x/\theta}.$$

$$M(t) = \frac{1}{1 - \theta t}.$$

$$M'(t) = \frac{\theta}{(1-\theta t)^2}, \quad M\boxtimes (t) = \frac{2\theta^2}{(1-\theta t)^3}.$$

$$\mu = \theta$$
, $\sigma^2 = \theta^2$.

The forgetfulness property

$$P(X > b|X > a) = P(X > b - a).$$

Gamma Distribution

 α is the number of the total occurrences.

$$\Gamma(t) = \int_0^\infty x^{t-1} e^{-x} dx, \quad \Gamma(n) = (n-1)!.$$

$$f(x) = \frac{1}{\Gamma(\alpha)\theta^{\alpha}} x^{\alpha-1} e^{-x/\theta}.$$

$$M(t) = \frac{1}{(1 - \theta t)^{\alpha}}.$$

$$\mu = \alpha \theta$$
, $\sigma^2 = \alpha \theta^2$.

A special case of the gamma distribution

$$\theta = 2$$
, $\alpha = \frac{r}{2}$.

r is the number of degrees of freedom.

$$\mu=r$$
, $\sigma^2=2r$.

NormalDistribution

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left[-\frac{(x-\mu)^2}{2\sigma^2}\right].$$

$$M(t) = \exp\left(\mu t + \frac{\sigma^2 t^2}{2}\right)$$

$$\mu = \frac{a+b}{2}, \quad \sigma^2 = \frac{(b-a)^2}{12}.$$
 $M'(t) = (\mu + \sigma^2 t) \exp\left(\mu t + \frac{\sigma^2 t^2}{2}\right).$

$$M\boxtimes(t) = \left[(\mu + \sigma^2 t)^2 + \sigma^2 \right] \exp\left(\mu t + \frac{\sigma^2 t^2}{2}\right).$$

If X is $N(\mu, \sigma^2)$, then $Z = (X - \mu)/\sigma$ is N(0, 1). If X is $N(\mu, \sigma^2)$, then $V = (X - \mu)^2$ μ)²/ σ ² = Z² is χ ²(1).