

$$E[Y] = E[E[Y|X]]$$

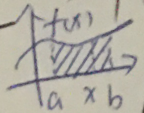
$$= \sum_{k=1}^b E[Y|X=k] \times P(X=k)$$

cdf $F(x) = P(X \leq x) \leq 1$
 distribution $P(X=x)=0$ continuous random variable.

$$F_X(x) = \int_{-\infty}^x f_X(u) du \rightarrow \text{pdf}$$

$$E[X] = \int_{-\infty}^{\infty} u f_X(u) du \quad E[X^2] = \int_{-\infty}^{\infty} u^2 f_X(u) du$$

$$P(a \leq X \leq b) = \int_a^b f_X(x) dx$$



$$f_X(x) = F'_X(x) = \frac{d}{dx} F_X(x)$$

$$\min\{T_1, T_2\} = P(Y=y) = P(Y \leq y \cup T_2 \leq y)$$

$$P(X \leq x \leq x+h) = \int_x^{x+h} f_X(x) dx \approx h f_X(x)$$

$$\max\{T_1, T_2\} = P(Y=y) = P(T_1 \leq y \cap T_2 \leq y)$$

$X \sim \text{Geometric}(p)$ is the probability that more than n times.

$$P(X > n) = (1-p)^n$$

$$\therefore g(x) = g(f_X(x))$$

is distribution, the pdf, $g(f_X(x))$

$$\therefore E[g(X)] = \int_{-\infty}^{\infty} g(u) f_X(u) du$$

density $X \sim \text{Uniform}[a, b]$
 pdf $\int_a^b f_X(x) dx = \int_a^b \frac{1}{b-a} dx$

$$E[X] = \frac{a+b}{2} \quad \text{Var}(X) = \frac{(b-a)^2}{12}$$

$X \sim \text{exp}(\lambda)$
 pdf $f_X(x) = \lambda e^{-\lambda x} \quad x \geq 0$
 cdf $F_X(x) = 1 - e^{-\lambda x} \quad x \geq 0$

$$E[X] = \frac{1}{\lambda} \quad \text{Var}(X) = \frac{1}{\lambda^2} \quad P(X > x+y | X > y) = P(X > x)$$

$$\int u(x) v'(x) dx = u(x) v(x) - \int v(x) u'(x) dx$$

eg: $\int_0^{\infty} x \cdot \lambda e^{-\lambda x} dx$, $[u(x)=x, v'(x)=-e^{-\lambda x}]$

$$= x \cdot (-e^{-\lambda x}) \Big|_0^{\infty} - \int_0^{\infty} (-e^{-\lambda x}) \cdot 1 dx = x \cdot (-e^{-\lambda x}) \Big|_0^{\infty} + \left(-\frac{e^{-\lambda x}}{\lambda} \right) \Big|_0^{\infty} = \frac{1}{\lambda}$$

Standard Normal Distribution

$$X \sim \text{Normal}(0, 1^2) \quad E[X] = 0 \quad \text{Var}(X) = 1^2$$

$$\text{pdf } f(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} \quad x \in \mathbb{R} \quad E[X^2] = \text{Var}(X) + (E[X])^2 = 1 + 0 = 1$$

查表 $P(X \leq x) = 1 - P(X > x) \quad X \sim \text{Normal}(\mu, \sigma^2)$

$$P(X > x) = P(X < -x)$$

$$P(X \leq -x) = P(X > x) \quad \text{pdf } f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$$P(-x \leq X \leq x) = 1 - 2P$$

μ ~ mean, expectation

σ^2 ~ variance

σ ~ standard deviation

$$X \sim \text{Normal}(0, 4)$$

$$\therefore P(X \leq 3.92)$$

$$\hat{Z} = \frac{X - \mu}{\sigma} \quad Z \sim \text{Normal}(0, 1)$$

$$P(Z \leq \frac{3.92 - 0}{\sqrt{4}}) = P(Z \leq 1.96)$$

$$= 1 - P(Z > 1.96) = 0.975 \text{ (查表)}$$

$$\frac{X - \mu}{\sigma} \sim \text{Normal}(0, 1)$$