

Discrete and continuous probability distributions

Discrete

Bernoulli

$X \sim \text{Bernoulli}(p)$, probability of success in one Bernoulli trial.

$$\begin{aligned}P[X = k] &= p[X = 1] + (1 - p)[X = 0] \\E[X] &= 1 \times p + 0 \times (1 - p) = p \\ \text{Var}(X) &= p(1 - p)\end{aligned}$$

Binomial

$X \sim \text{Binomial}(n, p)$, number of success in n independent Bernoulli trials with probability p .

$$P[X = k] = \binom{n}{k} p^k (1 - p)^{n-k} \quad \forall k \in \{0, 1, \dots, n\}$$

$$\begin{aligned}E[X] &= \sum_{k=0}^n k \binom{n}{k} p^k (1 - p)^{n-k} \\&= n \sum_{k=1}^n \binom{n-1}{k-1} p^k (1 - p)^{n-k} \\&= np \sum_{k=1}^n \binom{n-1}{k-1} p^{k-1} (1 - p)^{n-k} \\&= np \sum_{k=0}^{n-1} \binom{n-1}{k} p^k (1 - p)^{n-k-1} \\&= np \\ \text{Var}(X) &= np(1 - p)\end{aligned}$$

Geometric

$X \sim \text{Geometric}(p)$, number of Bernoulli trials until one success (counting the one that succeed). p must be nonzero.

$$P[X = k] = p(1 - p)^{k-1} \quad \forall k \in \{1, 2, \dots\}$$

$$\begin{aligned}
E[X] &= \sum_{k=1}^{\infty} kp(1-p)^{k-1} \\
&= p \sum_{k=0}^{\infty} k(1-p)^{k-1} \\
&= p \sum_{k=0}^{\infty} -\frac{d}{dp}(1-p)^k \\
&= -p \times \frac{d}{dp} \sum_{k=0}^{\infty} (1-p)^k \\
&= -p \times \frac{d}{dp} \frac{1}{p} \\
&= p \times \frac{1}{p^2} = \frac{1}{p} \\
\text{Var}(X) &= \frac{1-p}{p^2}
\end{aligned}$$

Poisson

$$X \sim \text{Poisson}(\lambda),$$

$$P[X = k] = \frac{\lambda^k e^{-\lambda}}{k!} \quad \forall k \in \{0, 1, \dots\}$$

$$\begin{aligned}
E[X] &= \sum_{k=0}^{\infty} \frac{\lambda^k e^{-\lambda}}{k!} \\
&= e^{-\lambda} \sum_{k=1}^{\infty} k \frac{\lambda^k}{k!} \\
&= e^{-\lambda} \lambda \sum_{k=1}^{\infty} \frac{\lambda^{k-1}}{(k-1)!} \\
&= e^{-\lambda} \lambda \sum_{k=0}^{\infty} \frac{\lambda^k}{k!} \\
&= \lambda
\end{aligned}$$

$$\text{Var}(X) = \lambda$$

Continuous

Uniform

$X \sim \text{Uniform}(a, b),$

$$f(x) = \frac{1}{b-a} \quad \forall x \in [a, b]$$

$$\begin{aligned} E[X] &= \int_a^b \frac{x}{b-a} dx \\ &= \frac{b^2 - a^2}{2(b-a)} \\ &= \frac{b+a}{2} \end{aligned}$$

$$\text{Var}(X) = \frac{(b-a)^2}{12}$$

Exponential

$X \sim \text{Exponential}(\lambda),$

$$f(x) = \lambda e^{-\lambda x} \quad \forall x \in [0, +\infty)$$

$$\begin{aligned} E[X] &= \int_0^{+\infty} x \lambda e^{-\lambda x} dx \\ &= -x e^{-\lambda x} \Big|_0^{+\infty} + \int_0^{+\infty} e^{-\lambda x} dx \\ &= 0 + \frac{-1}{\lambda} (0 - 1) \\ &= \frac{1}{\lambda} \end{aligned}$$

$$\text{Var}(X) = \frac{1}{\lambda^2}$$

Normal

Theorem 1.

$$\iint_A f(x, y) dx dy = \iint_S f(g(u, v), h(u, v)) \left| \frac{\partial(x, y)}{\partial(u, v)} \right| du dv$$

Lemma 1.

$$\int_{-\infty}^{+\infty} e^{-x^2} dx = \sqrt{\pi}$$

Proof. Let $I = \int_{-\infty}^{+\infty} e^{-x^2} dx$, then

$$\begin{aligned}
 I^2 &= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} e^{-x^2} e^{-y^2} dx dy \\
 &= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} e^{-(x^2+y^2)} dx dy \\
 &= \int_0^{2\pi} \int_0^{+\infty} e^{-r^2} r dr d\theta && \text{by Theorem 1} \\
 &= 2\pi \int_0^{+\infty} e^{-r^2} r dr \\
 &= \pi \int_0^{+\infty} e^{-u} du \\
 &= \pi
 \end{aligned}$$

□

$X \sim \text{Normal}(\sigma, \mu)$,

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left(-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2\right) \quad \forall x \in (-\infty, +\infty)$$

$$\begin{aligned}
 E[X] &= \int_{-\infty}^{+\infty} x f(x) dx \\
 &= \int_{-\infty}^{+\infty} \frac{u}{\sigma\sqrt{2\pi}} \exp\left(-\frac{1}{2}\left(\frac{u}{\sigma}\right)^2\right) du + \int_{-\infty}^{+\infty} \frac{\mu}{\sigma\sqrt{2\pi}} \exp\left(-\frac{1}{2}\left(\frac{u}{\sigma}\right)^2\right) du \\
 &= 0 + \mu \int_{-\infty}^{+\infty} \frac{1}{\sqrt{\pi}} e^{-v^2} dv && \text{odd function} \\
 &= \mu && \text{by Lemma 1}
 \end{aligned}$$

$$\text{Var}(X) = \sigma^2$$