

# Probability Distributions

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September 29, 2018

## 1 Bernoulli Distribution

A *Bernoulli random variable*,  $X$ , takes the value 1 with probability  $p$  and 0 otherwise. Thus,

$$\begin{aligned} E(X) &= p \\ V(X) &= p(1 - p). \end{aligned} \tag{1}$$

## 2 Binomial Distribution

A *Binomial random variable*,  $X$ , is a sum of  $n$  iid. Bernoulli random variables. We write  $X \sim B(n, p)$ . Trivially,

$$P(X = k) = \binom{n}{k} p^k (1 - p)^{n-k}. \tag{2}$$

From (1) it follows that

$$\begin{aligned} E(X) &= np \\ V(X) &= np(1 - p). \end{aligned} \tag{3}$$

## 3 Poisson Distribution

A *Poisson random variable*,  $X$ , is given by ... LOTS OF EXPLANATION HERE.

$$P(X = k) = e^{-\lambda} \frac{\lambda^k}{k!}, \quad k \in \mathbb{N}_0 \tag{4}$$

$$\begin{aligned} E(X) &= \lambda \\ V(X) &= \lambda \end{aligned} \tag{5}$$

To prove (5) note that

$$E(X) = e^{-\lambda} \sum_{k=0}^{\infty} k \frac{\lambda^k}{k!} = e^{-\lambda} \lambda \sum_{k=1}^{\infty} \frac{\lambda^{k-1}}{(k-1)!} = \lambda.$$

Similarly,

$$E(X^2) = e^{-\lambda} \sum_{k=0}^{\infty} k^2 \frac{\lambda^k}{k!} = e^{-\lambda} \lambda \sum_{k=1}^{\infty} ((k-1) + 1) \frac{\lambda^{k-1}}{(k-1)!} = \lambda^2 + \lambda.$$

## 4 Exponential Distribution

The exponential distribution describes the time of first-time occurrence for an event with constant chance to occur in each infinitesimal time interval.

A random variable is exponentially distributed,  $X \sim \text{Exp}(\lambda)$ , if its density function is

$$f(x; \lambda) = \lambda e^{-\lambda x}, \quad x \geq 0.$$

Equivalently,  $P(X \geq h) = e^{-\lambda h}$ . We note that

$$\begin{aligned} E(X) &= \frac{1}{\lambda} \\ V(X) &= \frac{1}{\lambda^2} \end{aligned} \tag{6}$$

### 4.1 Memoryless

For  $s, t \geq 0$ ,

$$P(X \geq s + t \mid X \geq s) = \frac{e^{-\lambda(s+t)}}{e^{-\lambda s}} = P(X \geq t).$$

\*constant hazard function\*