

This section deals with finding the *population mean* — $\mu = E(x)$ — of various probability distributions

Binomial Distribution

If $X \sim \text{Bin}(n, p)$, then $\mu = E(X) = np$

Proof:

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

$$E(x) = \sum_{x=0}^n x \cdot f(x)$$

$$= \sum_{x=0}^n x \cdot \binom{n}{x} p^x (1-p)^{n-x}$$

\vdots

$$= np$$

Alternate proof: Let X_i be a Bernoulli random variable.

$$X_i = \begin{cases} 1, & \text{with probability } p \\ 0, & \text{with probability } 1-p \end{cases} \quad \begin{matrix} \text{results of} \\ \text{trials} \end{matrix}$$

If $X \sim \text{Bin}(n, p)$, then

$$X = x_1 + x_2 + \dots + x_n$$

$$\Rightarrow E(X_i) = 1(p) + 0(1-p) = p \quad \begin{matrix} \text{each trial occurs with} \\ \text{probability } p \end{matrix}$$

$$\begin{aligned} \Rightarrow E(X) &= E(x_1 + x_2 + \dots + x_n) \\ &= p + \dots + p \\ &= np \end{aligned}$$

Poisson Distribution

Let $X \sim \text{Pois}(\lambda t)$

$$\lambda t = \mu$$

λ - positive samples per interval t

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

$$\begin{aligned} \Rightarrow E(X) &= \sum_{\forall x} x \cdot \frac{e^{-\lambda t} (\lambda t)^x}{x!} \\ &= \lambda t \text{ eventually} \end{aligned}$$

Variability

Some potential measures for variability:

- $E(X - \mu)$
 - $x - \mu$ for all x — how far from the mean?
 - By linearity, $E(X - \mu) = E(X) - \mu = \mu - \mu = 0$. So this sucks.
- $E(|X - \mu|)$ — not a bad idea, but absolute values don't have *nice* mathematical properties
- Better one:

$$\text{Variance: } \sigma^2 = \text{Var}(X) = E[(X - \mu)^2]$$

summed over all x

But these are in squared units

$$\text{standard deviation: } \sqrt{E[(X - \mu)^2]}$$

Alternate form:

$$\text{Var}(X) = E[(X - \mu)^2] = E(X^2 - 2\mu X + \mu^2)$$

$$= E(X^2) - 2\mu \underbrace{E(X)}_{\mu} + \underbrace{\mu^2}_{\text{constant}}$$

$$= E(X^2) - \mu^2$$

More variance \rightarrow data moves away from the mean

The variance of a binomial distribution is $np(1-p)$

Proof: $\text{Var}(X) = E[X(X-1)] + \mu - \mu^2$

$$E[X(X-1)] = \sum_{x=0}^n \underbrace{(x)(x-1)}_{g(x)} \underbrace{\binom{n}{x} p^x (1-p)^{n-x}}_{f(x)}$$

$$= n(n-1)p^2$$

$$\Rightarrow n(n-1)p^2 + np - n^2p^2$$

$$= np(1-p)$$

The variance of a Poisson distribution is μ

3. A person plays a game in which a fair coin is tossed until the first tail occurs. The person wins $\$2^x$ if x tosses are needed for $x = 1, 2, 3, 4, 5$ but loses $\$256$ if $x > 5$.

(a) Determine the expected winnings.

(b) Determine the variance of the winnings.

$$(a) P(X \leq 5) = \frac{1}{2} + \frac{1}{4} + \frac{1}{8} + \frac{1}{16} + \frac{1}{32}$$

$$= \frac{16+8+4+2+1}{32} = \frac{31}{32}$$

$$\Rightarrow P(X > 5) = 1 - \frac{31}{32} = \frac{1}{32}$$

$$\Rightarrow \text{Expected winnings: } \left(\frac{1}{2}\right) \cdot 2 + \left(\frac{1}{4}\right) \cdot 2^2 + \left(\frac{1}{8}\right) \cdot 2^3 \dots + \frac{1}{32} \cdot (-256)$$

$$= 5 - \frac{256}{32}$$

$$= -3???$$