

Last lecture

Bayes formula (Ch 2.10)

- Examples


$$P(B|A) = \frac{P(AB)}{P(A)}$$

Independent Events/ RVs (Ch 2.4)

- Definition
- Motivation
- Examples and Facts

$$P(\underline{B|A}) = P(B)$$

$$\underline{P(AB)} = P(A)P(B)$$

Agenda

Independent Events/ RVs (Ch 2.4)

- Examples and Facts

Distributions (Ch 2.4)

- Bernoulli
- Binomial

Geometric distribution (Ch 2.5)

- Definition

Terms and Facts

- A, B, C are pairwise independent if $(A, B), (B, C), (A, C)$ are mutually independent

- Toss a fair coin twice

- $A \triangleq \{\text{First coin is Head}\}$

- $B \triangleq \{\text{second coin is Head}\}$

- $C \triangleq \{\text{toss results are the same}\}$

Does not imply
 A, B, C are all independent

$$P(ABC) = P(A)P(B)P(C)$$

$$\Omega_A = \{(H, T), (H, H)\} \quad \|\Omega\| = 4 \quad P(A) = P(B) = P(C)$$

$$\Omega_B = \{(T, H), (H, H)\} \quad = \frac{1}{2}$$

$$\Omega_C = \{(H, H), (T, T)\}$$

$$P(A|B) = \frac{P(AB)}{P(B)} = \frac{\frac{1}{4}}{\frac{1}{2}} = \frac{1}{2} = P(A).$$

$$P(B|C) = \frac{1}{2}, \quad P(A|C) = \frac{1}{2}.$$

$$P(ABC) = \frac{1}{4} \neq P(A)P(B)P(C)$$

Terms and Facts

$F(B, C)$

if ✓

- A, B, C are independent they are pairwise independent and $P(ABC) = P(A)P(B)P(C)$

Proof $P(A \cap (B \cup C))$

$$= P(A) P(B \cup C)$$

- $A_1, A_2, \dots, A_i$ are independent if
$$P(A_{i_1} A_{i_2} \dots A_{i_k}) = P(A_{i_1}) P(A_{i_2}) \dots P(A_{i_k})$$

Distributions



$\{X, P_X\}$

of commonly seen exp.

Bernoulli Distribution

X is Bernoulli distribution with parameter p if

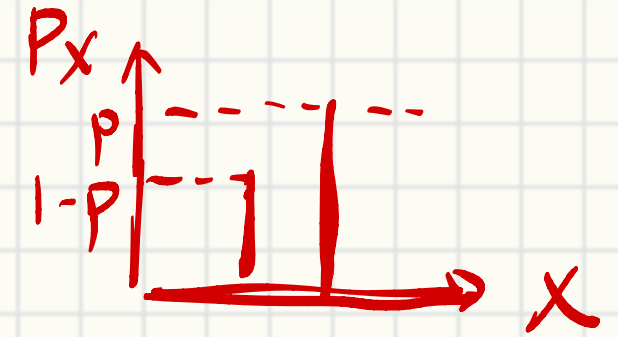
- $P\{X = 1\} = p$ and $P\{X = 0\} = 1 - p$
- “Toss a (unfair) coin with p probability Head”
- Only **2** possible outcomes, pmf contains **2** bins

$$E[X] = 1 \times P_X(1) + 0 \times P_X(0) = p$$

$$E[X^2] = p$$

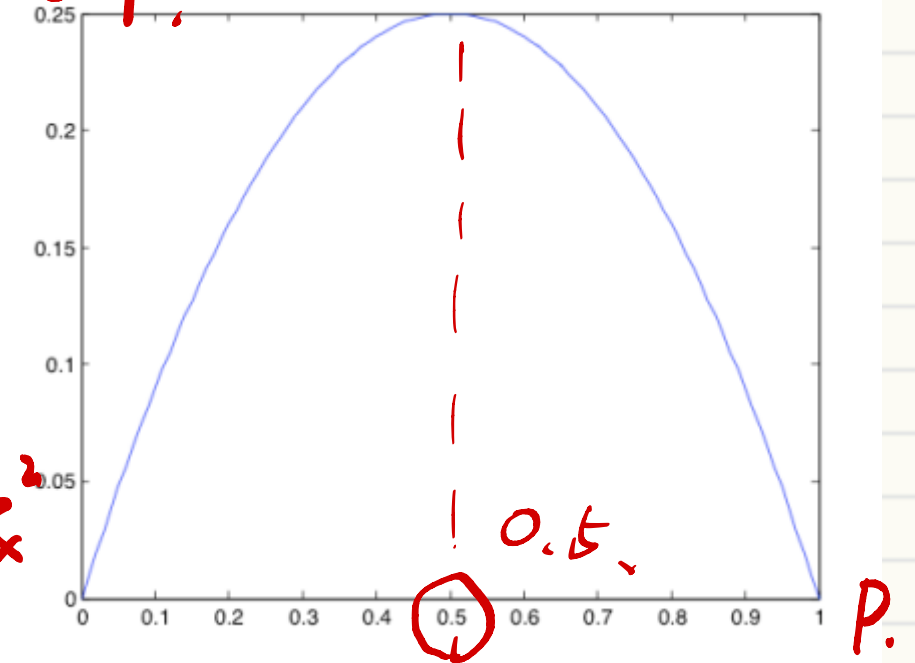
$$\sigma_x^2 = E[X^2] - (E[X])^2$$

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



first
momentum

second
momentum



Binomial Distribution

X is binomial distribution with parameter (n, p) if $\Rightarrow n+1$ bins
 $[0, n]$

- X is sum of n Bernoulli trials with parameter p
- Draw the unfair coin n times and count the Head

$$(n, p) = (24, \frac{1}{3})$$

Head

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

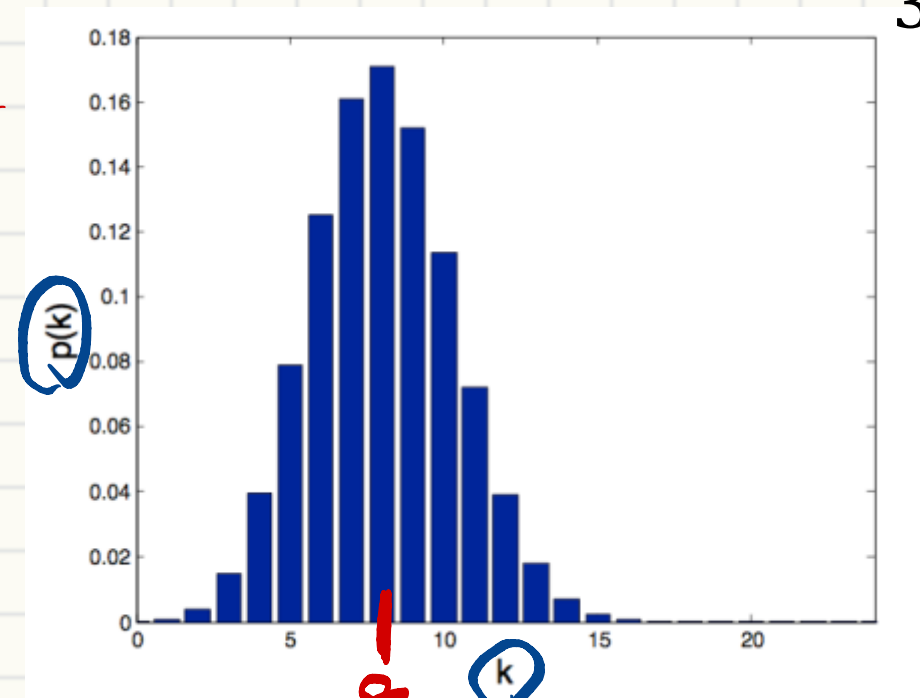
$p_X(2)$ for $(n, p) = (4, 0.5)$

$\binom{4}{2}$

H	H	T	T
H	T	H	T
H	T	T	H
T	T	H	H
T	H	T	H
T	H	H	T

$\frac{1}{2} \times \frac{1}{2} \times \frac{1}{2} \times \frac{1}{2}$

$p^2 (1-p)^{4-2}$



Binomial Distribution

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

Proof - $\sum_{k=0}^n p_X(k) = 1$

Exam-safe

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

$$x^n + \binom{n}{1} x^{n-1} \dots$$

$$\left(1 + \frac{p}{1-p}\right)^n = \sum \binom{n}{k} \left(\frac{p}{1-p}\right)^k$$

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

$$\left(\frac{1}{1-p}\right)^n = \sum \binom{n}{k} p^k \left(\frac{1}{1-p}\right)^k \times (1-p)^n$$

Binomial mean

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

Exam-Safe

$$\begin{aligned} E[X] &= \sum_{k=0}^n k \binom{n}{k} p^k (1-p)^{n-k} \\ &= \sum_{k=1}^n k \frac{n!}{(n-k)!k!} p^k (1-p)^{n-k} \\ &= np \sum_{k=1}^n \frac{(n-1)!}{(n-k)!(k-1)!} p^{k-1} (1-p)^{n-k} \\ &= np \sum_{m=0}^{n-1} \frac{n-1!}{(n-1-m)!m!} p^m (1-p)^{n-1-m} \\ &= np \sum_{m=0}^j \binom{j}{m} p^m (1-p)^{j-m} = np, \end{aligned}$$

Binomial Properties

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

- Mean $\underline{E[X]} = np$
- Variance $\underline{Var(X)} = np(1-p) = \underline{n} \times Var(\text{Bernoulli}(p))$
 - why? n tosses are independent.
- Shape of the pmf- what is the most likely k ?
 - $k^* = \lfloor (n+1)p \rfloor$

$$\text{Var}(X_1 + X_2) = \text{Var}(X_1) + \text{Var}(X_2)$$

↓ if X_1, X_2 are independent

$P_{X_1, X_2}(j, k)$

$$\underline{E[(X_1 + X_2)^2]} - \underline{(E[X_1 + X_2])^2}$$

$$= E[X_1^2] + E[2X_1X_2] + E[X_2^2] - (\mu_1 + \mu_2)^2$$

$$= \underbrace{E[X_1^2] - \mu_1^2}_{\text{independ}} + \underbrace{E[X_2^2] - \mu_2^2}_{\text{independ}} + \underbrace{E[2X_1X_2] - 2\mu_1\mu_2}_{\text{independ}}$$

$$= \text{Var}(X_1) + \text{Var}(X_2)$$

$$\begin{aligned} & \sum_j \sum_k jk P_{X_1}(j) P_{X_2}(k) \\ & \left[\sum_j j P_{X_1}(j) \right] \left[\sum_k k P_{X_2}(k) \right] \end{aligned}$$

μ_1 ← μ_2

Slido



#5018375

Back to early 1900's, there's a mail fraud

- I mail N people that I can predict a series of 50-50 games
- I predict A wins in $\frac{N}{2}$ mail, B wins for the other
- Stop mailing a person after 2 wrong guess
- Say people will subscribe to my prediction if I only make at most 1 mistake in 5 guesses 4 correct 5 correct
- What should be N if I want to get 300 subs?

1600

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

$$P_X(5) = \binom{5}{5} \left(\frac{1}{2}\right)^5 \left(\frac{1}{2}\right)^{5-5}!$$
$$= \frac{1}{32}$$

$$+) P_X(4) = \binom{5}{4} \left(\frac{1}{2}\right)^4 \left(\frac{1}{2}\right)^{5-4}$$
$$= 5 \times \frac{1}{32}$$

$$\frac{6}{32}$$

$$= \frac{3}{16}$$