

6. Conditional PMFs and Independent Random Variables

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Conditional PMF

Let X be a random variable and \mathcal{A} be an **event**.

The **conditional PMF** of X given \mathcal{A} is

$$P(X = x \mid \mathcal{A}) = \frac{P(X = x \text{ and } \mathcal{A})}{P(\mathcal{A})}$$

What is the PMF of a 6-sided die roll given that the outcome is even?

You flip 3 coins. What is the PMF of number of heads given that there is **at least one**?

Conditioning on a random variable

The **conditional PMF** of X given Y is

$$\mathbf{P}(X = x \mid Y = y) = \frac{\mathbf{P}(X = x \text{ and } Y = y)}{\mathbf{P}(Y = y)}$$

$$p_{X|Y}(x \mid y) = \frac{p_{XY}(x, y)}{p_Y(y)}$$

For fixed y , $p_{X|Y}$ is a PMF as a function of x .

Roll two 3-sided dice. What is the PMF of the sum given the first roll?

Roll two 3-sided dice. What is the PMF of the first roll given the sum?

Conditional Expectation

The **conditional expectation** of X given event \mathcal{A} is

$$\mathbf{E}[X \mid \mathcal{A}] = \sum_x x \mathbf{P}(X = x \mid \mathcal{A})$$

The **conditional expectation** of X given $Y = y$ is

$$\mathbf{E}[X \mid Y = y] = \sum_x x \mathbf{P}(X = x \mid Y = y)$$

You flip 3 coins. What is the **expected** number of heads given that there is at least one?

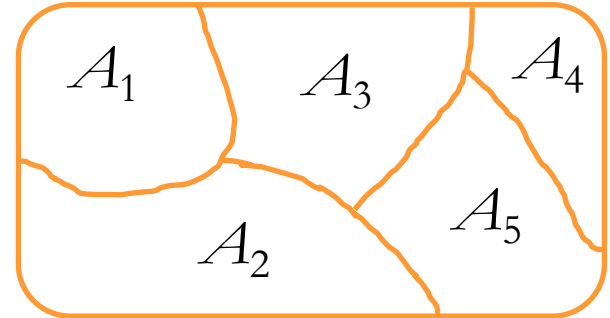
Total Expectation Theorem

$$\mathbf{E}[X] = \mathbf{E}[X | \mathcal{A}] \mathbf{P}(\mathcal{A}) + \mathbf{E}[X | \mathcal{A}^c] \mathbf{P}(\mathcal{A}^c)$$

Proof

Total Expectation Theorem (general form)

If A_1, \dots, A_n **partition** Ω
then



$$\mathbf{E}[X] = \mathbf{E}[X | A_1] \mathbf{P}(A_1) + \dots + \mathbf{E}[X | A_n] \mathbf{P}(A_n)$$

In particular

$$\mathbf{E}[X] = \sum_y \mathbf{E}[X | Y = y] \mathbf{P}(Y = y)$$

type



average time
on facebook

30 min

60 min

10 min

% of visitors

60%

30%

10%

average visitor time =

You play 10 rounds of roulette. You invest \$100 and bet 10% of your balance on red in every round.

What is your average balance after 10 rounds?

You flip 3 coins. What is the **expected** number of heads given that there is at least one?

Mean of the Geometric

$X = \text{Geometric}(p)$ random variable

$\mathbf{E}[X] =$

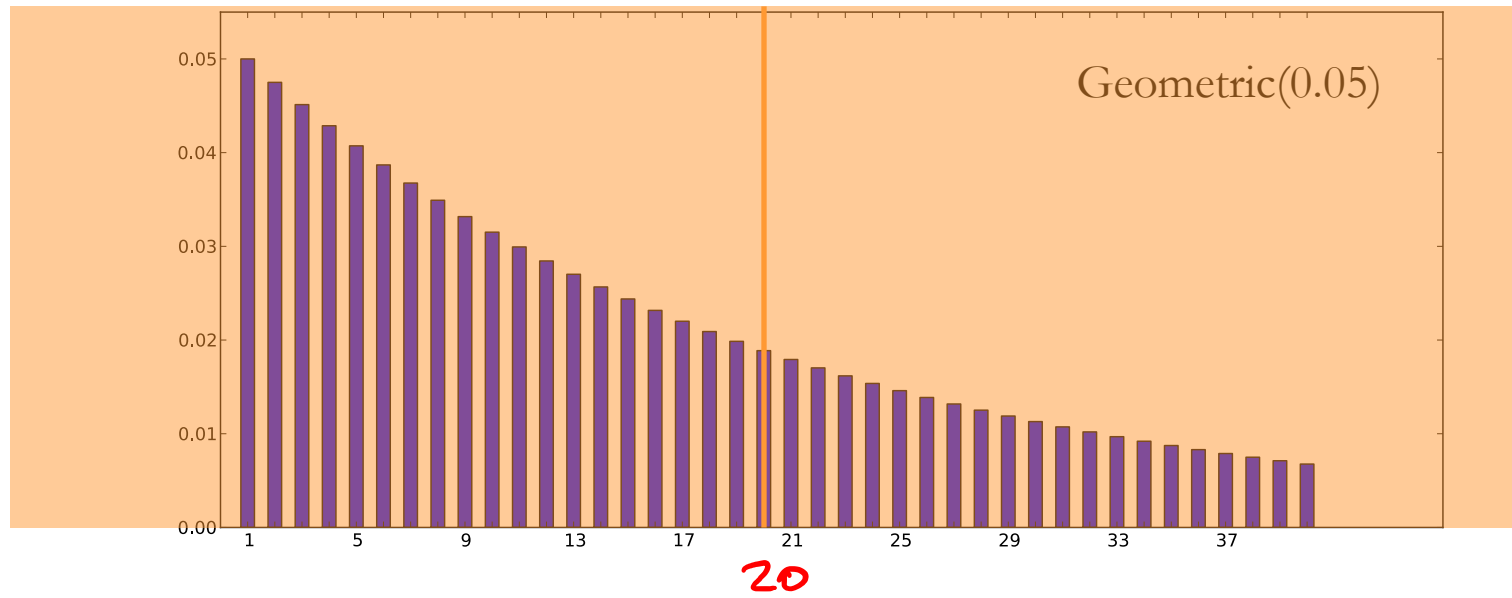
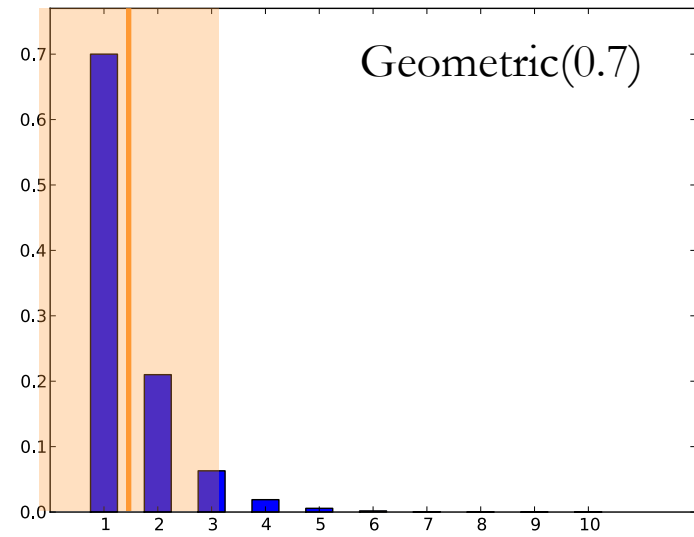
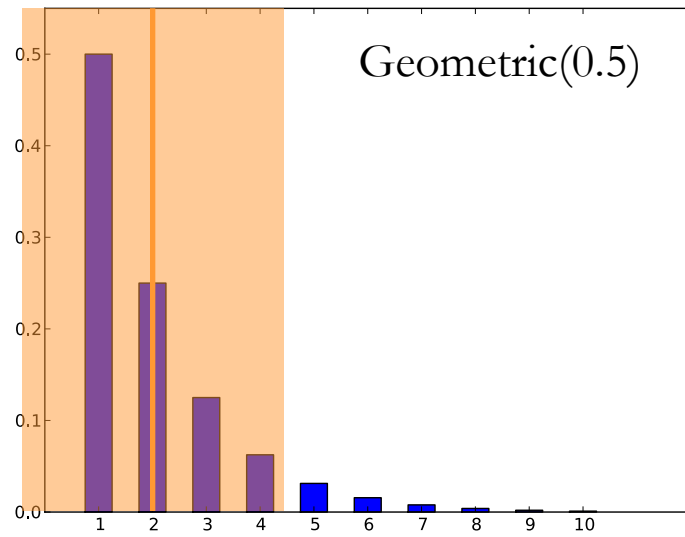
Variance of the Geometric

$X = \text{Geometric}(p)$ random variable

$$\text{Var}[X] = \frac{1-p}{p^2} \quad \sigma = \frac{\sqrt{1-p}}{p} < \frac{1}{p}.$$

Proof (optional):

$$\begin{aligned} \text{Var}[X] &= E[(X - 1/p)^2] \\ &= E[(X - 1/p)^2 | W] \cdot p + E[(X - 1/p)^2 | W^c] \cdot (1-p) \\ &= (1 - 1/p)^2 \cdot p + E[(1 + Y - 1/p)^2 | W^c] \cdot (1-p) \\ &= (1 - 1/p)^2 \cdot p + E[1 + 2(Y - 1/p) + (Y - 1/p)^2 | W^c] \cdot (1-p) \\ &= (1 - 1/p)^2 \cdot p + \underbrace{(1 + 2E[Y - 1/p | W^c])}_0 + \underbrace{E[(Y - 1/p)^2 | W^c]}_{\text{Var}[X]} \cdot (1-p) \\ &= (1 - 1/p)^2 \cdot p + (1 + \text{Var}[X]) \cdot (1-p) \end{aligned}$$



Independent random variables

Let X and Y be **discrete** random variables.

X and Y are **independent** if

$$\mathbf{P}(X = x, Y = y) = \mathbf{P}(X = x) \mathbf{P}(Y = y)$$

for all possible values of x and y .

In PMF notation, $p_{XY}(x, y) = p_X(x) p_Y(y)$ for all x, y .

Independent random variables

X and Y are **independent** if

$$\mathbf{P}(X = x \mid Y = y) = \mathbf{P}(X = x)$$

for all x and y such that $\mathbf{P}(Y = y) > 0$.

In PMF notation, $p_{X|Y}(x \mid y) = p_X(x)$ if $p_Y(y) > 0$.

Alice tosses three coins and so does Bob. Alice gets \$1 per head and Bob gets \$1 per tail.

Are their earnings independent?

Now they toss **the same coin** three times. Are their earnings independent?

Expectation and independence

X and Y are independent if and only if

$$\mathbf{E}[f(X)g(Y)] = \mathbf{E}[f(X)] \mathbf{E}[g(Y)]$$

for all real valued functions f and g .

Expectation and independence

In particular, if X and Y are independent then

$$E[XY] = E[X] E[Y]$$

Not true in general!

Variance of a sum

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

$$\text{Var}[X + Y] =$$

Variance of a sum

$$\mathbf{Var}[X_1 + \dots + X_n] = \mathbf{Var}[X_1] + \dots + \mathbf{Var}[X_n]$$

if every pair X_i, X_j is independent.

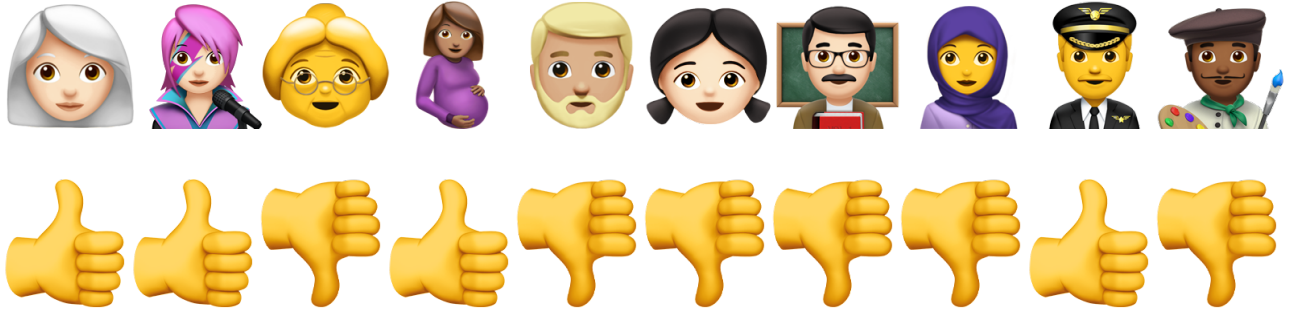
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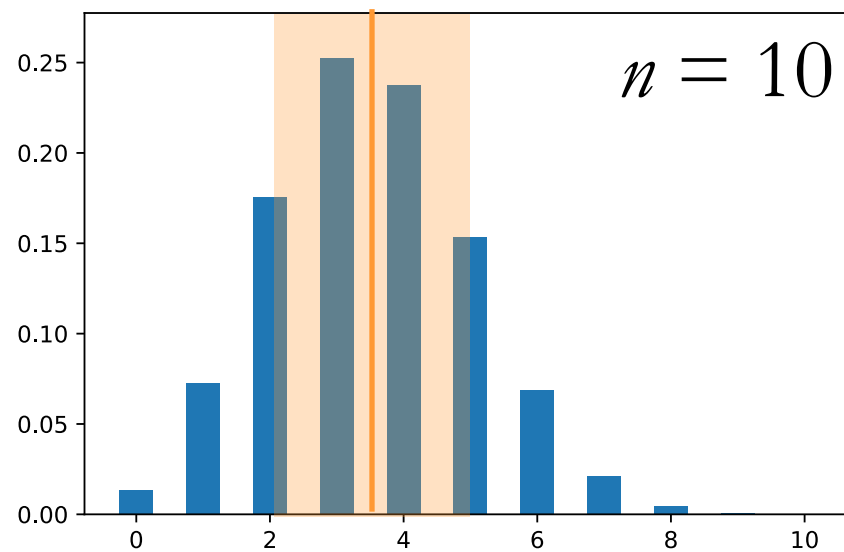
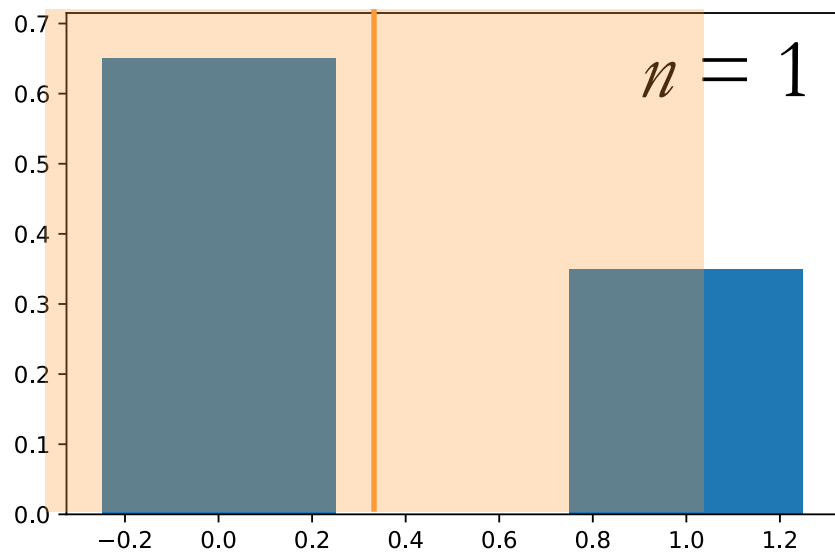
Keep tossing a fair coin until you get ten heads.
On average, how many times T will you toss?

What is the standard deviation of T ?

Variance of the Binomial

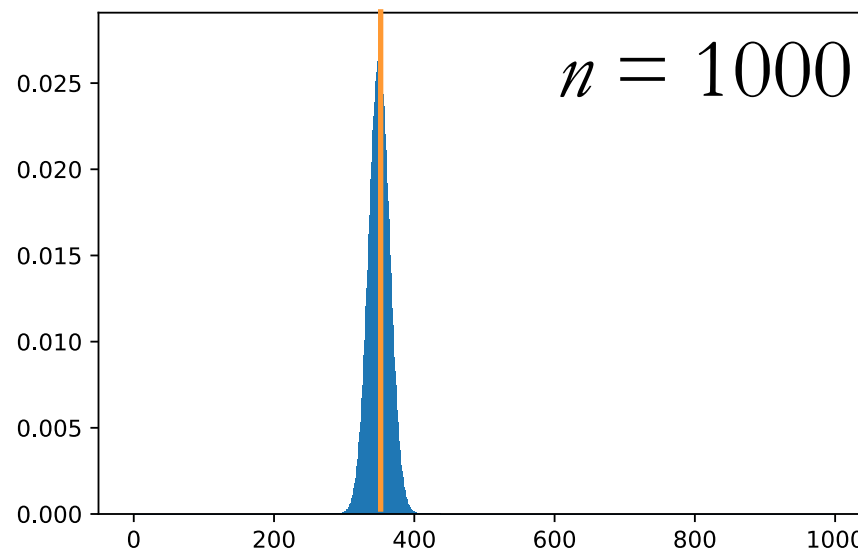
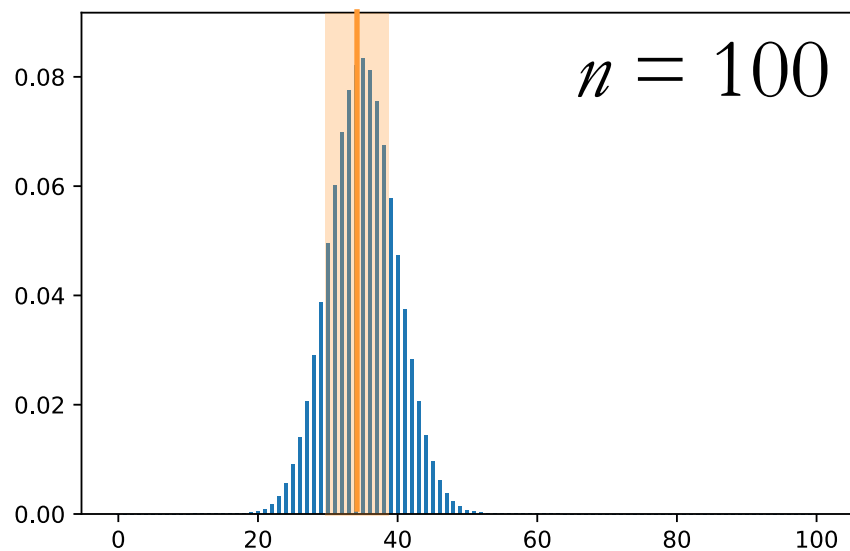
Sample mean





$p = 0.35$

$\mu - \sigma \leftarrow \mu \rightarrow \mu + \sigma$



Variance of the Poisson

Poisson(λ) **approximates** Binomial($n, \lambda/n$) **for large** n

$$p(k) = e^{-\lambda} \lambda^k / k!$$

$$k = 0, 1, 2, 3, \dots$$

Independence of multiple random variables

X, Y, Z **independent** if

$$\mathbf{P}(X = x, Y = y, Z = z) = \mathbf{P}(X = x) \mathbf{P}(Y = y) \mathbf{P}(Z = z)$$

for all possible values of x, y, z .

X, Y, Z independent if and only if

$$\mathbf{E}[f(X)g(Y)h(Z)] = \mathbf{E}[f(X)] \mathbf{E}[g(Y)] \mathbf{E}[h(Z)]$$

for all f, g, h .

Usual warnings apply.