

# Consistency of the PMF

- ▶ PMF:  $p_X(x) = \mathbb{P}(\{\omega \in \Omega : X(\omega) = x\})$  for  $x \in \Omega'$ .
- ▶ How do you check if  $p_X$  is legitimate PMF?
- ▶  $\sum_{x \in \Omega'} p_X(x) = 1$ . Can you prove this?

$$\begin{aligned} \sum_{x \in \Omega'} p_X(x) &= \sum_{x \in \Omega'} \mathbb{P}(\{\omega \in \Omega : X(\omega) = x\}) \\ &= \mathbb{P}(\cup_{x \in \Omega'} \{\omega \in \Omega : X(\omega) = x\}) \\ &= \mathbb{P}(\Omega) \quad \square \end{aligned}$$

# Linearity of Expectation

- ▶ Recall that  $E[X] = \sum_{x \in \Omega'} xp_X(x)$ .
- ▶ Functions of random variables are random variables.
- ▶ Furthermore,  $E[g(X)] := \sum_{x \in \Omega'} g(x)p_X(x)$
- ▶ For  $Y = aX + b$ , what is  $E[Y]$ ?

$$\begin{aligned} E[Y] &= \sum_{x \in \Omega'} (ax + b)p_X(x) \\ &= a \sum_{x \in \Omega'} xp_X(x) + b \\ &= aE[X] + b. \end{aligned}$$

- ▶ What is the PMF of  $Y$ ?

## PMF of $Y$ where $Y = aX + b$ .

- ▶ Suppose the range of  $X$  is  $\Omega' = \{x_1, x_2, \dots, x_n\}$ . Then what is the range  $\Omega''$  of  $Y$  ?
- ▶  $\Omega'' = \{y_1, \dots, y_n\}$  where  $y_i = ax_i + b$  for  $i \in \{1, 2, \dots, n\}$ .
- ▶ It is easy to see that,  $p_Y(y_i) = p_X(x_i)$  for  $i \in \{1, 2, \dots, n\}$ .

$$\begin{aligned} E[Y] &= \sum_{y \in \Omega''} yp_Y(y) \\ &= \sum_{x \in \Omega'} (ax + b)p_y(ax + b) \\ &= \sum_{x \in \Omega'} (ax + b)p_x(x) \\ &= aE[X] + b. \end{aligned}$$

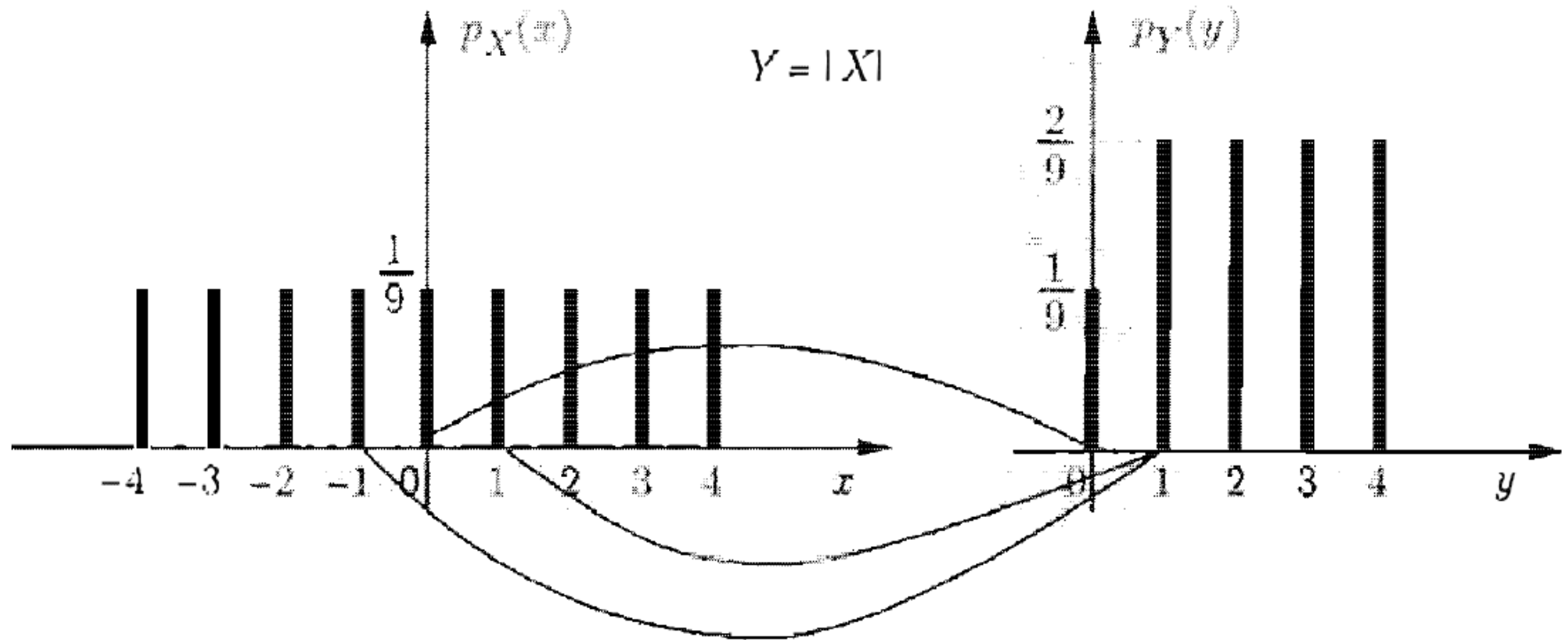
- ▶ What if  $Y = g(X)$  where the function  $g(\cdot)$  is many to one? What is the PMF of  $Y$  then ?

# Function of random variables

- ▶ Consider  $Y = |X|$  where  $X$  is the outcome of an experiment where an integer is chosen uniformly from  $-4$  to  $4$ .
- ▶  $p_X(x) = \frac{1}{9}$  for  $x \in \{-4, -3, \dots, 3, 4\}$ .
- ▶ What is the range  $\Omega'$  for  $Y$ ?  $\Omega' = \{0, \dots, 4\}$ .
- ▶ What is  $p_Y(2)$ ?
- ▶  $p_Y(2) = \sum_{\{x: |x|=2\}} p_X(x) = p_X(-2) + p_X(2) = \frac{2}{9}$ .

# Function of random variables

►  $p_Y(2) = \sum_{\{x: |x|=2\}} p_X(x) = p_X(-2) + p_X(2) = \frac{2}{9}.$



Suppose  $Y = g(X)$  and  $X$  is discrete with pmf  $p_X(\cdot)$ . Then  $p_Y(y) = \sum_{\{x: g(x)=y\}} p_X(x)$ . (Proof is HW)

# $E[g(X)]$

Theorem: Suppose  $Y = g(X)$  and  $X$  is discrete with pmf  $p_X(\cdot)$ . Then,  $E[Y] = \sum_x g(x)p_X(x)$

Proof

$$\begin{aligned} E[Y] &= \sum_y yp_Y(y) \\ &= \sum_y \sum_{\{x:g(x)=y\}} g(x)p_X(x) \\ &= \sum_x g(x)p_X(x). \end{aligned}$$

□

[https://en.wikipedia.org/wiki/Law\\_of\\_the\\_unconscious\\_statistician](https://en.wikipedia.org/wiki/Law_of_the_unconscious_statistician)

## Towards Variance ..

- ▶ Recall  $E[X] = \sum_{x \in \Omega'} x p_X(x)$ .
- ▶ Furthermore,  $E[X^n] = \sum_{x \in \Omega'} x^n p_X(x)$ .
- ▶ In general,  $E[g(X)] := \sum_{x \in \Omega'} g(x) p_X(x)$
- ▶ Now consider  $g(X) = (X - E[X])^2$ .  $g(X)$  quantifies the square of the deviation of  $X$  from the mean.
- ▶ Note  $g(X)$  cannot track if the deviation is positive or negative!
- ▶  $E[g(X)]$  would then tell us the mean of the square of the deviation.
- ▶ In fact,  $\sqrt{E(g(X))}$  quantifies the deviation.

# Variance

- ▶  $E[g(X)] = E[(X - E[X])^2]$  is called as the variance of random variable  $X$ .
- ▶  $Var(X) := E[(X - E[X])^2]$
- ▶ HW: Prove that  $E[(X - E[X])^2] = E[X^2] - E[X]^2$
- ▶  $\sigma_X = \sqrt{Var(X)}$  is called as the standard deviation of  $X$ .
- ▶ For a fair coin toss, instead of  $\Omega' = \{1, -1\}$ , what if we use  $\{+100, -100\}$  ? The latter has more variance!
- ▶ HW: What is  $Var(Y)$  where  $Y = aX + b$ ?



# Examples of discrete random variables

# Indicator random variable

- ▶ Indicator random variable  $1_A(\omega) = \begin{cases} 1, & \text{If } \omega \in A \subseteq \Omega \\ 0, & \text{otherwise.} \end{cases}$
- ▶ Its PMF is  $p_{1_A}(x) = \begin{cases} \mathbb{P}(A), & \text{when } x = 1 \\ 1 - \mathbb{P}(A), & \text{when } x = 0. \end{cases}$
- ▶ This is a discrete random variable even though  $\Omega$  could be continuous.
- ▶ For example, Event  $A$  could be that the number picked uniformly on the real line is positive.
- ▶ What is its CDF and mean denoted by  $E[1_A]$ ?
- ▶ What about its mean variance and moments?

# Bernoulli random variable

- ▶ Bernoulli random variable  $X = \begin{cases} 1, & \text{with probability } p \\ 0, & \text{otherwise.} \end{cases}$
- ▶ This is same as an indicator variable but here we do not specify  $A$ .
- ▶ As a matter of convenience, we will start ignoring  $\Omega$  from now on.
- ▶ These random variables are used in Binary classification in ML.  $X = 1$  if image has a cat.
- ▶ Basic models of Multi-arm bandit problem assume Bernoulli Bandits.
- ▶  $E[X] = p, E[X^n] = p.$

# Binomial $B(n, p)$ random variable.

- ▶ Consider a biased coin (head with probability  $p$ ) and toss it  $n$  times.
- ▶ Denote head by 1 and tail by 0.
- ▶ Let random variable  $N$  denote the number of heads in  $n$  tosses.
- ▶ PMF of  $N$ ?.  $p_N(k) = \binom{n}{k} p^k (1 - p)^{n-k}$ .
- ▶ HW: What is  $E[N]$ ,  $E[N^2]$ ,  $Var(X)$ ?