

## Conditional Probability Models

sample space  
event space  
probability measure

- Recall that, for a general probability space  $(\Omega, \mathcal{E}, \mathbb{P})$ , we defined the conditional probability of event  $A \in \mathcal{E}$  occurring given that event  $B \in \mathcal{E}$  occurs as

$$\mathbb{P}[A | B] = \frac{\mathbb{P}[A \cap B]}{\mathbb{P}[B]} \quad (\text{assuming } \mathbb{P}[B] > 0)$$

- Let  $X$  be a discrete random variable with PMF

$$P_x(x) = \mathbb{P}[\{\omega \in \Omega : X(\omega) = x\}] = \mathbb{P}[X = x]$$

- The conditional PMF  $P_{x|B}(x)$  of  $X$  given event  $\{X \in B\}$  is

$$P_{x|B}(x) = \mathbb{P}[\{\omega \in \Omega : X(\omega) = x\} | \{X \in B\}]$$

$$\Rightarrow P_{x|B}(x) = \begin{cases} \frac{P_x(x)}{\mathbb{P}[\{X \in B\}]} & x \in B \\ 0 & x \notin B \end{cases} = \begin{cases} \frac{P_x(x)}{\sum_{x \in B} P_x(x)} & x \in B \\ 0 & x \notin B \end{cases} \quad (\text{assuming } \mathbb{P}[\{X \in B\}] > 0)$$

• Why?

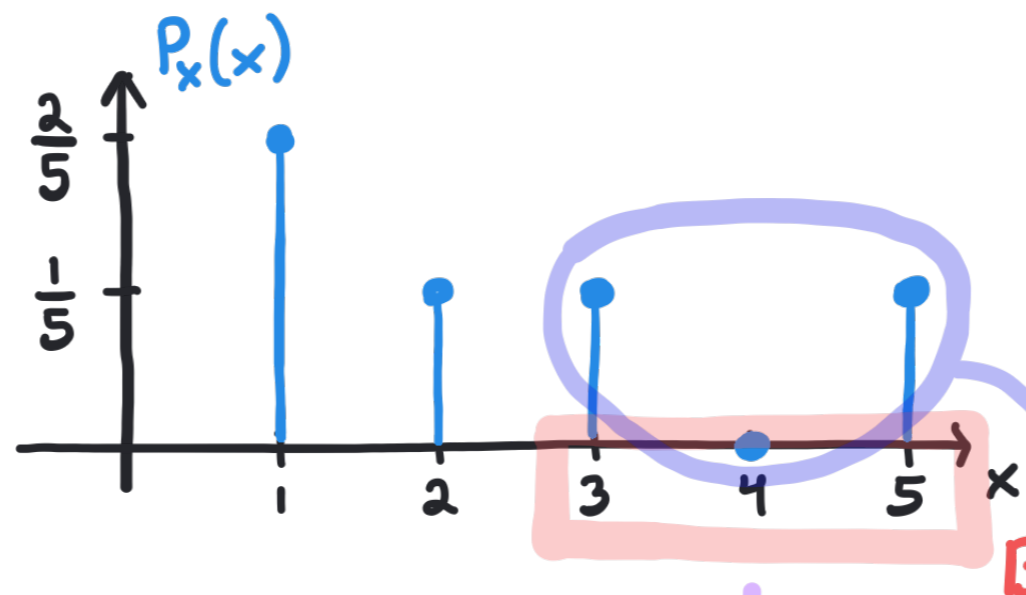
$$\begin{aligned} P_{X|B}(x) &= \mathbb{P}[\underbrace{\{\omega \in \Omega : X(\omega) = x\}}_{\text{event A}} \mid \underbrace{\{X \in B\}}_{\text{event C}}] \\ &= \mathbb{P}[A \mid C] \\ &= \frac{\mathbb{P}[A \cap C]}{\mathbb{P}[C]} \quad (\text{assuming } \mathbb{P}[C] > 0, \text{ otherwise undefined}) \\ &= \frac{\mathbb{P}[\{\omega \in \Omega : X(\omega) = x\} \cap \{X \in B\}]}{\mathbb{P}[\{X \in B\}]} \end{aligned}$$

only outcomes in B occur

Definition of PMF

$$= \begin{cases} \frac{\mathbb{P}[\{\omega \in \Omega : X(\omega) = x\}]}{\mathbb{P}[\{X \in B\}]} & x \in B \\ 0 & x \notin B \end{cases} \xrightarrow{\text{additivity}} \begin{cases} \frac{P_x(x)}{\sum_{x \in B} P_x(x)} & x \in B \\ 0 & x \notin B \end{cases}$$

• Example:



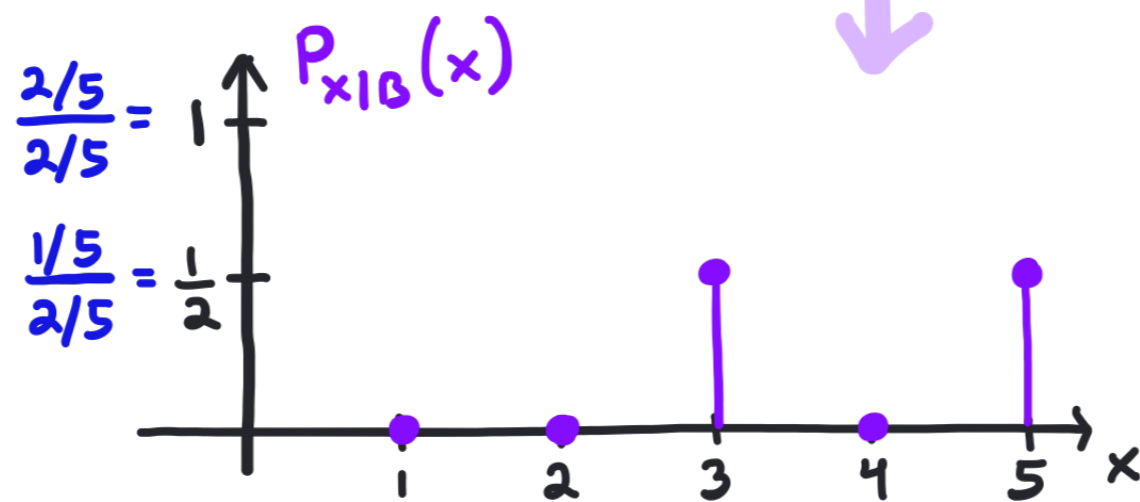
$$P_x(x) = \begin{cases} \frac{2}{5} & x=1 \\ \frac{1}{5} & x=2,3,5 \end{cases}$$

$$\begin{aligned} IP[\{x \in B\}] &= \sum_{x \in B} P_x(x) \\ &= P_x(3) + P_x(4) + P_x(5) \\ &= \frac{1}{5} + 0 + \frac{1}{5} \\ &= \frac{2}{5} \end{aligned}$$

Let  $B = \{3, 4, 5\}$ .

Find  $P_{x|B}(x)$ .

Rescale by dividing by  $IP[\{x \in B\}]$



$$\begin{aligned} P_{x|B}(x) &= \begin{cases} \frac{P_x(x)}{IP[\{x \in B\}]} & x \in B \\ 0 & x \notin B \end{cases} \\ &= \begin{cases} \frac{1/5}{2/5} & x=3,5 \\ 0 & \text{otherwise} \end{cases} \\ &= \begin{cases} \frac{1}{2} & x=3,5 \\ 0 & \text{otherwise} \end{cases} \end{aligned}$$

→ Intuition:  $P_{x|B}(x)$  is just  $P_x(x)$  restricted to the values in  $B$  and rescaled so these remaining values sum to 1.

- Conditional PMF Properties:

- $P_{x|B}(x) \geq 0$  (Non-negativity)

- $\sum_{x \in B} P_{x|B}(x) = 1$  (Normalization)

- For any event  $C \subset R_x$ , the conditional probability that  $X$  falls into  $C$  given that  $X$  falls into event  $B$  is

$$P_{x|B}[C] = P[\{x \in C\} | \{x \in B\}] = \sum_{x \in C} P_{x|B}(x)$$

(Additivity)

- We can also develop versions of the multiplication rule, the law of total probability, and Bayes' rule. (See lecture notes for details.)

- Given that the event  $B$  occurs, what is the average value of  $X$ ? We need to generalize our notion of expectation to allow for conditioning.

- The conditional expected value  $\mathbb{E}[X|B]$  of  $X$  given event  $B$  is

$$\mathbb{E}[X|B] = \sum_{x \in B} x P_{X|B}(x)$$

- The conditional expected value  $\mathbb{E}[g(X)|B]$  of a function  $g(X)$  given event  $B$  is

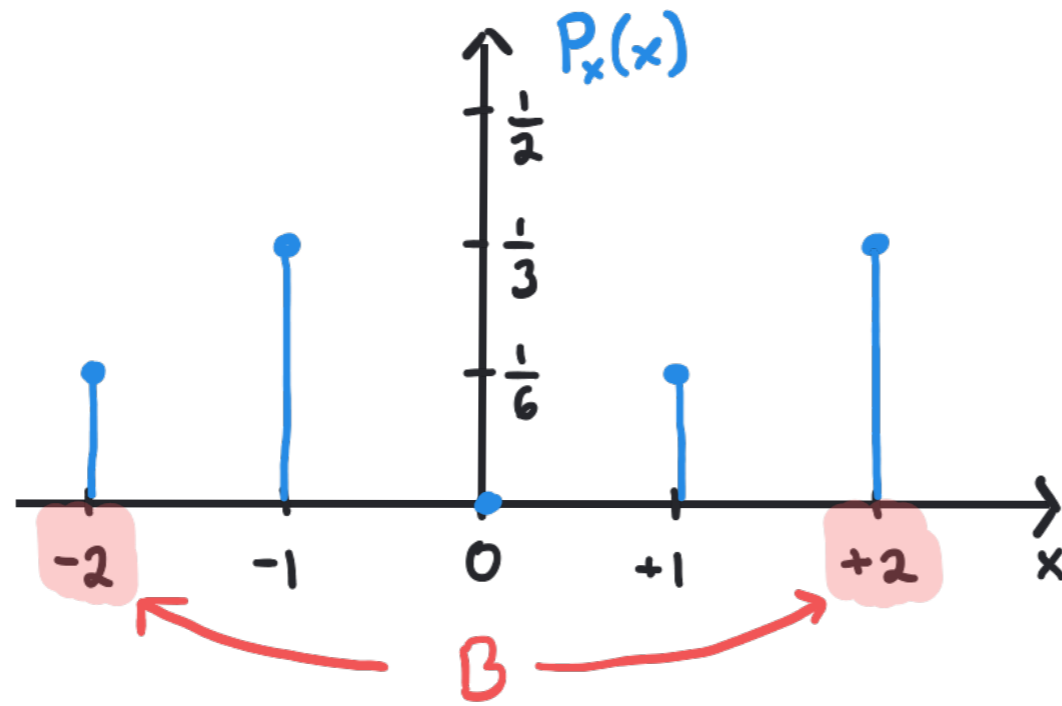
$$\mathbb{E}[g(X)|B] = \sum_{x \in B} g(x) P_{X|B}(x)$$

→ Linearity of Expectation:  $\mathbb{E}[aX+b|B] = a\mathbb{E}[X|B] + b$

- The conditional variance  $\text{Var}[X|B]$  of  $X$  given event  $B$  is

$$\text{Var}[X|B] = \mathbb{E}[(X - \mathbb{E}[X|B])^2 | B] = \mathbb{E}[X^2 | B] - (\mathbb{E}[X|B])^2$$

• Example:



$$P_x(x) = \begin{cases} \frac{1}{3} & x = -1, +2 \\ \frac{1}{6} & x = -2, +1 \\ 0 & \text{otherwise} \end{cases}$$

→ Condition on the event that  $|X| > 1$ .

$$\{|X| > 1\} = \{X \in B\}$$

$$B = \{-2, +2\}$$

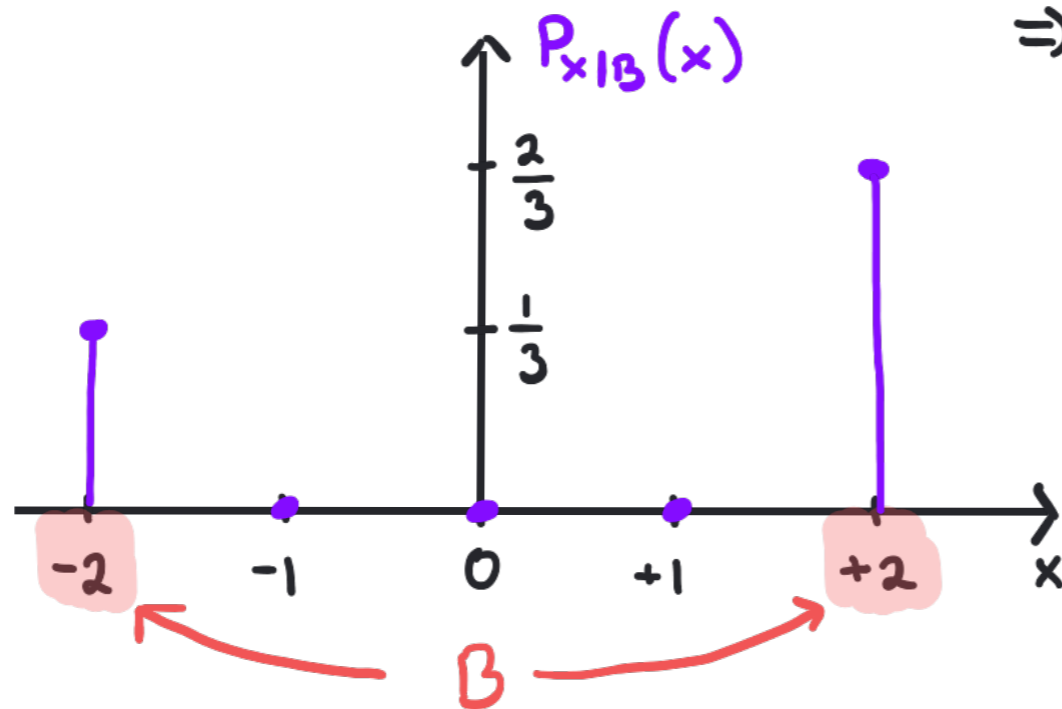
→ Determine the conditional PMF.

$$P_{x|B}(x) = \begin{cases} \frac{P_x(x)}{IP[\{X \in B\}]} & x \in B \\ 0 & x \notin B \end{cases}$$

$$IP[\{X \in B\}] = \sum_{x \in B} P_x(x)$$

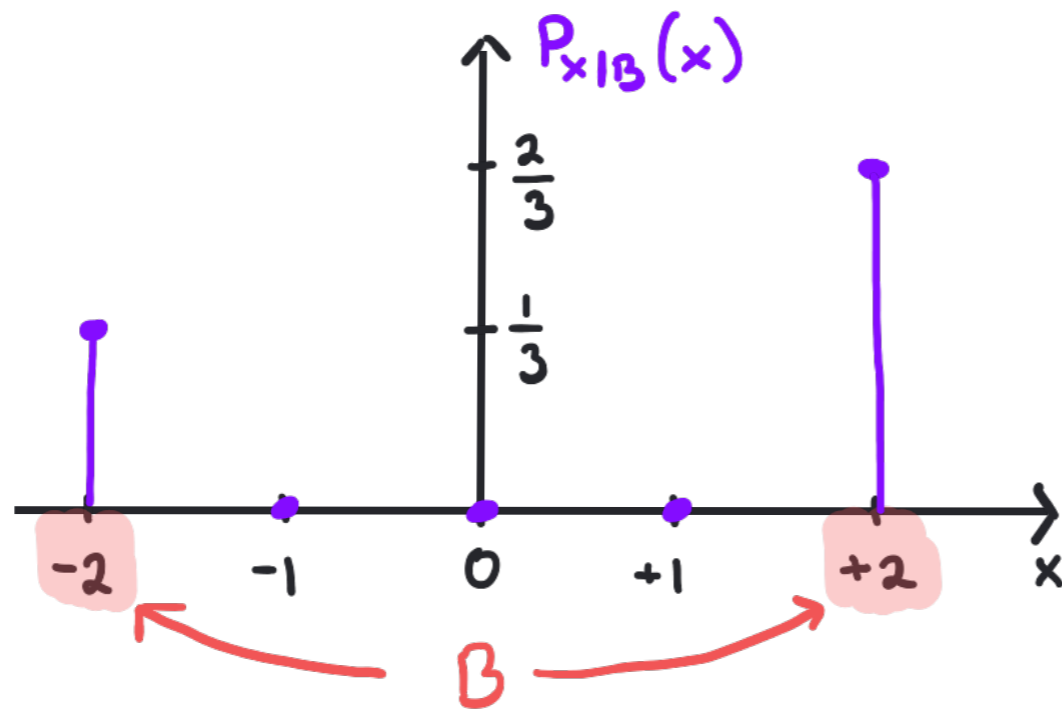
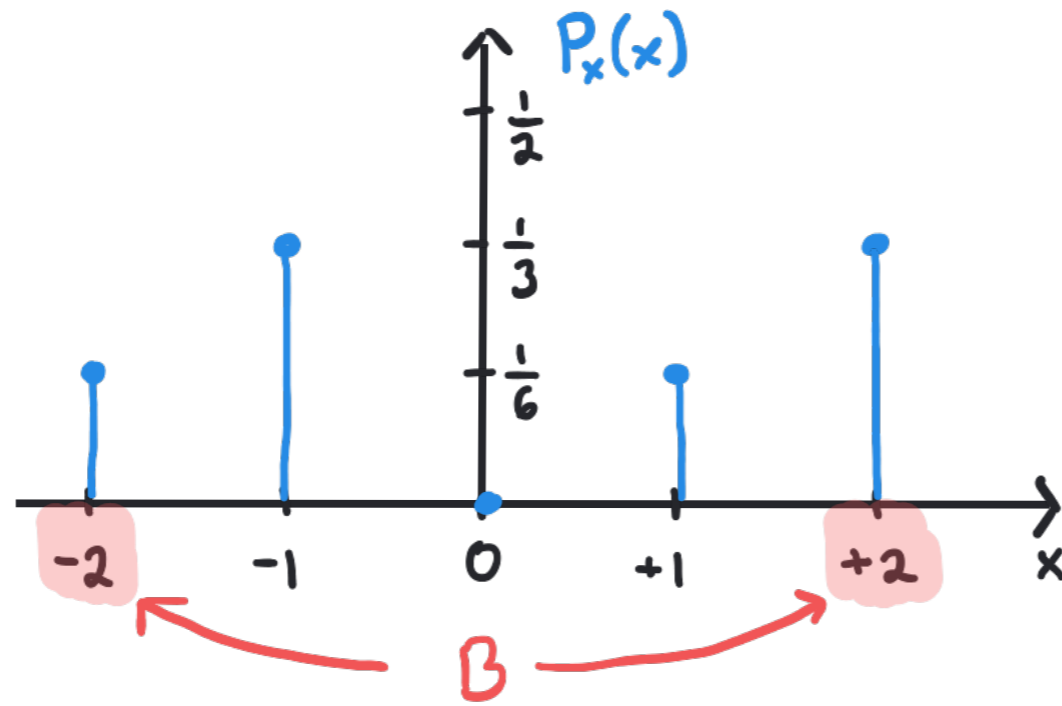
$$= P_x(-2) + P_x(+2) = \frac{1}{6} + \frac{1}{3} = \frac{1}{2}$$

$$\Rightarrow P_{x|B}(x) = \begin{cases} \frac{2}{3} & x = +2 \\ \frac{1}{3} & x = -2 \\ 0 & \text{otherwise} \end{cases}$$



PMF is **restricted** to  $B$  and **rescaled** so it sums to 1.

• Example:



$$P_x(x) = \begin{cases} \frac{1}{3} & x = -1, +2 \\ \frac{1}{6} & x = -2, +1 \\ 0 & \text{otherwise} \end{cases}$$

→ Condition on the event that  $|X| > 1$ .

$$\{|X| > 1\} = \{X \in B\}$$

$$B = \{-2, +2\}$$

$$P_{x|B}(x) = \begin{cases} \frac{2}{3} & x = +2 \\ \frac{1}{3} & x = -2 \\ 0 & \text{otherwise} \end{cases}$$

→ Calculate  $E[X]$ ,  $E[X|B]$ .

$$E[X] = \sum_{x \in R_x} x P_x(x) = (-2) \cdot \frac{1}{6} + (-1) \cdot \frac{1}{3} + (+1) \cdot \frac{1}{6} + (+2) \cdot \frac{1}{3} = \frac{1}{6}$$

$$E[X|B] = \sum_{x \in B} x P_{x|B}(x) = (-2) \cdot \frac{1}{3} + (+2) \cdot \frac{2}{3} = \frac{2}{3}$$

• Example: Wait for delivery of a missing part. Arrival time  $X$  is uniformly distributed over  $\{1, 2, \dots, 20\}$  days.

→ Assume part has not yet arrived after 6 days. Determine the conditional PMF, expected value, and variance.

$$B = \{7, 8, \dots, 20\} \quad \sum_{x \in B} P_X(x) = \sum_{x=7}^{20} P_X(x) = \sum_{x=7}^{20} \frac{1}{20} = \frac{14}{20}$$

$$P_{X|B}(x) = \begin{cases} \frac{P_X(x)}{\sum_{x \in B} P_X(x)} & x \in B \\ 0 & x \notin B \end{cases} = \begin{cases} \frac{1/20}{14/20} = \frac{1}{14} & x = 7, 8, \dots, 20 \\ 0 & \text{otherwise} \end{cases}$$

$$E[X|B] = \sum_{x \in B} x P_{X|B}(x) = \sum_{x=7}^{20} x \cdot \frac{1}{14} = \frac{7+20}{2} = \frac{27}{2}$$

→ *Simplifying observation*:  $P_{X|B}(x)$  is PMF for Discrete Uniform(7, 20)

$$\Rightarrow \text{Var}[X|B] = \frac{(20-7+2)(20-7)}{12} = \frac{15 \cdot 13}{12} = \frac{65}{4}$$



• Example:  $X$  is Binomial  $(5, \frac{1}{3})$ . Let  $B = \{0, 1, 2\}$ .

Determine the conditional PMF of  $X$  given  $\{X \in B\}$

and the conditional expected value.

$$P_X(x) = \begin{cases} \binom{5}{x} \left(\frac{1}{3}\right)^x \left(\frac{2}{3}\right)^{5-x} & x = 0, 1, \dots, 5 \\ 0 & \text{otherwise} \end{cases}$$

$$\begin{aligned} \sum_{x \in B} P_X(x) &= P_X(0) + P_X(1) + P_X(2) \\ &= \binom{5}{0} \left(\frac{1}{3}\right)^0 \left(\frac{2}{3}\right)^5 + \binom{5}{1} \left(\frac{1}{3}\right)^1 \left(\frac{2}{3}\right)^4 + \binom{5}{2} \left(\frac{1}{3}\right)^2 \left(\frac{2}{3}\right)^3 \\ &= \frac{32}{243} + \frac{80}{243} + \frac{80}{243} = \frac{192}{243} \end{aligned}$$

$\underbrace{\frac{32}{243}}_{P_X(0)} \quad \underbrace{\frac{80}{243}}_{P_X(1)} \quad \underbrace{\frac{80}{243}}_{P_X(2)}$

$$P_{X|B}(x) = \begin{cases} \frac{P_X(x)}{\sum_{x \in B} P_X(x)} & x \in B \\ 0 & x \notin B \end{cases} = \begin{cases} \frac{32/243}{192/243} & x = 0 \\ \frac{80/243}{192/243} & x = 1, 2 \\ 0 & \text{otherwise} \end{cases} = \begin{cases} \frac{1}{6} & x = 0 \\ \frac{5}{12} & x = 1, 2 \\ 0 & \text{otherwise} \end{cases}$$

$$E[X|B] = \sum_{x \in B} x P_{X|B}(x) = 0 \cdot \frac{1}{6} + 1 \cdot \frac{5}{12} + 2 \cdot \frac{5}{12} = \frac{15}{12} = \frac{5}{4}$$