

## Review: Random Variables

Sample space and events:

- Sample space  $S$  consists of the set of all possible outcomes of an experiment
- An event  $E$  is a subset of  $S$ ,  $E \subset S$
- $P(E)$  is probability of event  $E$ .

A random variable  $X(\omega)$  maps from outcomes  $\omega$  in the sample space  $S$  to a real number.

- Often  $\omega$  is dropped and just write  $X$ , leaving the  $\omega$  as understood.
- The set of outcomes for which  $X = x$  is  $E_x = \{\omega | X(\omega) = x, \omega \in S\}$
- $P(X = x)$  is probability that random variable  $X$  takes value  $x$ , probability mass function (PMF).
- $P(X = x)$  is probability that event  $E_x$  occurs:  $P(X = x) = P(E_x)$ .
- $F(a) = P(X \leq a)$  is the cumulative distribution function (CDF)
- Indicator random variable  $I_E$ ,  $I_E = 1$  when event  $E$  occurs and 0 otherwise.  $P(I_E = 1) = P(E)$

## Example

Roll a die:

- Sample space  $S = \{1, 2, 3, 4, 5, 6\}$
- $E = \{1, 2\}$  is event that a 1 or a 2 is observed.
- $P(E)$  is probability of event  $E$ .
- Random variable  $X(\omega) = 1$  when  $\omega = 1$  or  $\omega = 2$  and  $X(\omega) = 0$  when  $\omega = 3, 4, 5, 6$ .
- $E_1 = \{\omega | X(\omega) = 1, \omega \in S\} = \{1, 2\} = E$ ,  
 $E_0 = \{3, 4, 5, 6\} = E_1^c = E^c$ .
- $P(X = 1) = P(E_1) = P(E) = \frac{2}{6}$ .
- $P(X = 0) = P(E_0) = P(E^c) = 1 - P(E) = \frac{4}{6}$

## Review: Random Variables

All of the rules for probabilities of events carry over to random variables using the fact that  $P(X = x) = P(E_x)$

For two discrete random variables  $X$  and  $Y$  on same sample space  $S$ :

- $E_x = \{\omega \in S : X(\omega) = x\}$  is set of outcomes for which  $X = x$ ,  
 $E_y = \{\omega \in S : Y(\omega) = y\}$  is set of outcomes for which  $Y = y$ .  
 $P(X = x) = P(E_x)$ ,  $P(Y = y) = P(E_y)$
- $P(X = x \text{ and } Y = y) = P(E_x \cap E_y)$ .
- $P(X = x \text{ and } Y = y)$  is joint probability mass function of  $X$  and  $Y$

Conditional probability:

- $P(X = x | Y = y) = \frac{P(X=x \text{ and } Y=y)}{P(Y=y)} = \frac{P(E_x \cap E_y)}{P(E_y)} = P(E_x | E_y)$

## Example

Roll a die (again):

- Sample space  $S = \{1, 2, 3, 4, 5, 6\}$
- $E = \{1, 2\}$  is event that a 1 or a 2 is observed.  $F = \{2, 3\}$  that a 2 or 3 is observed.
- Random variable  $X = 1$  on event  $E$  and 0 otherwise. Random variable  $Y = 1$  on event  $F$  and 0 otherwise.
- $P(X = 1 \text{ and } Y = 1) = P(E \cap F) = P(\{2\}) = \frac{1}{6}$
- $P(X = 1 \text{ and } Y = 0) = P(E \cap F^c) = P(\{1, 2\} \cap \{1, 4, 5, 6\}) = P(\{1\}) = \frac{1}{6}$
- $P(X = 0 \text{ and } Y = 1) = P(E^c \cap F) = P(\{3, 4, 5, 6\} \cap \{2, 3\}) = P(\{3\}) = \frac{1}{6}$
- $P(X = 0 \text{ and } Y = 0) = P(E^c \cap F^c) = P(\{4, 5, 6\}) = \frac{3}{6}$
- $P(Y = 0) = P(F^c) = P(\{1, 4, 5, 6\}) = \frac{4}{6}$ .  $P(Y = 1) = \frac{2}{6}$
- $P(X = 0|Y = 0) = \frac{P(X=0 \text{ and } Y=0)}{P(Y=0)} = \frac{\frac{3}{6}}{\frac{4}{6}} = \frac{3}{4}$ .  $P(X = 1|Y = 0) = \frac{1}{4}$

## Review: Random Variables

Chain rule:  $P(X = x \text{ and } Y = y) = P(X = x|Y = y)P(Y = y)$ .

Consequences of chain rule:

- Marginalisation:

Suppose RV  $Y$  takes values in  $\{y_1, y_2, \dots, y_n\}$ . Then

$$\begin{aligned}P(X = x) &= P(X = x \text{ and } Y = y_1) + \dots + P(X = x \text{ and } Y = y_n) \\ &= \sum_{i=1}^n P(X = x|Y = y_i)P(Y = y_i)\end{aligned}$$

Example: roll a die,  $X$  and  $Y$  as before.

- $P(X = 0) = P(X = 0|Y = 0)P(Y = 0) + P(X = 0|Y = 1)P(Y = 1) = \frac{3}{4} \times \frac{4}{6} + \frac{1}{2} \times \frac{2}{6} = \frac{2}{3}$
- Double check:  $P(X = 0) = P(\{3, 4, 5, 6\}) = \frac{4}{6} = \frac{2}{3}$ .

## Review: Random Variables

Consequences of chain rule (cont):

- Bayes rule:  $P(X = x|Y = y) = \frac{P(Y=y|X=x)P(X=x)}{P(Y=y)}$ .
- $P(X = x) = \sum_{i=1}^m P(X = x \text{ and } Y = y_i)$  when RV  $Y$  takes values in  $\{y_1, y_1, \dots, y_m\}$

Independence:

- Discrete random variables  $X$  and  $Y$  are independent if  $P(X = x \text{ and } Y = y) = P(X = x)P(Y = y)$  for all  $x$  and  $y$

Example: roll a die,  $X$  and  $Y$  as before.

- Are  $X$  and  $Y$  independent? Both depend on outcome 2, so not independent. Check:  $P(X = 0 \text{ and } Y = 0) = \frac{3}{6} = 0.5$  and  $P(X = 0)P(Y = 0) = \frac{4}{6} \times \frac{4}{6} \approx 0.444$

## Expected Value

The Expected Value of discrete random variable  $X$  taking values in  $\{x_1, x_2, \dots, x_n\}$  is:

$$E[X] = \sum_{i=1}^n x_i P(X = x_i)$$

- Linearity:  $E[X + Y] = E[X] + E[Y]$ ,  $E[aX + b] = aE[X] + b$ .
- For **independent** random variables  $X$  and  $Y$  then  $E[XY] = E[X]E[Y]$

Example: roll a die,  $X$  and  $Y$  as before.

- $E[X] = 1 \times P(X = 1) + 0 \times P(X = 0) = 1 \times \frac{2}{6} + 0 \times \frac{4}{6} = \frac{2}{6}$
- $E[3X + 1] = (3 \times 1 + 1)P(X = 1) + (3 \times 0 + 1)P(X = 0) = 4 \times \frac{2}{6} + 1 \times \frac{4}{6} = \frac{12}{6} = 2$
- Double check:  $3E[X] + 1 = 3 \times \frac{2}{6} + 1 = 2$

## Expected Value

Conditional expectation of  $X$  given  $Y = y$  is:

$$E[X|Y = y] = \sum_x xP(X = x|Y = y)$$

- Linearity:  $E[\sum_i Y_i|X = x] = \sum_i E[Y_i|X = x]$
- $E[X] = \sum_y E[X|Y = y]P(Y = y)$

Example: roll a die,  $X$  and  $Y$  as before.

- $E[X|Y = 0] = 1 \times P(X = 1|Y = 0) + 0 \times P(X = 0|Y = 0) = 1 \times \frac{1}{4} + 0 \times \frac{3}{4} = \frac{1}{4}$
- $E[X|Y = 1] = 1 \times P(X = 1|Y = 1) + 0 \times P(X = 0|Y = 1) = 1 \times \frac{1}{6} + 0 \times \frac{5}{6} = \frac{1}{6}$
- $E[X] = E[X|Y = 0]P(Y = 0) + E[X|Y = 1]P(Y = 1) = \frac{1}{4} \times \frac{4}{6} + \frac{1}{6} \times \frac{2}{6} = \frac{2}{6}$



## Variance

The variance of  $X$  taking values in  $D = \{x_1, x_2, \dots, x_n\}$  is:

$$\text{Var}(X) = \sum_{i=1}^n (x_i - \mu)^2 p(x_i) = E[X^2] - (E[X])^2$$

with  $\mu = E[X] = \sum_{i=1}^n x_i p(x_i)$

- $\text{Var}(X) \geq 0$
- Standard deviation is square root of variance  $\sqrt{\text{Var}(X)}$ .
- $\text{Var}(aX + b) = a^2 \text{Var}(X)$
- For **independent** random variables  $X$  and  $Y$  then  $\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$

## Covariance

The covariance of  $X$  and  $Y$  is :

$$\text{Cov}(X, Y) = E[(X - \mu_X)(Y - \mu_Y)] = E[XY] - E[X]E[Y]$$

- $\text{Cov}(X, X) = \text{Var}(X)$ .
- When  $X$  and  $Y$  are independent then  $E[XY] = E[X]E[Y]$  and  $\text{Cov}(X, Y) = 0$ .

The correlation between  $X$  and  $Y$  is:

$$\text{Corr}(X, Y) = \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X)\text{Var}(Y)}}$$

- Takes values between -1 and 1.

# Summary Statistics

Expected value, variance, covariance and correlation are all examples of summary statistics.

- Expected value  $E[X]$  indicates the overall outcome of many repetitions of an experiment (its a sort of prediction)
- Variance  $Var(X)$  indicates the spread of  $X$
- Covariance  $Cov(X, Y)$  is positive if  $X$  and  $Y$  tend to increase together, and negative if an increase in one tends to correspond to a decrease in the other.
- Correlation  $Corr(X, Y)$  indicates the strength of a linear relationship between  $X$  and  $Y$ .

## Bernoulli Random Variable

Suppose an experiment results in Success or Failure.

- $X$  is a random indicator variable,  $X = 1$  on success,  $X = 0$  on failure
- $P(X = 1) = p$
- $P(X = 0) = 1 - p$
- $X$  is a **Bernoulli** random variable.
- $E[X] = p$ ,  $\text{Var}(X) = E[X^2] - E[X]^2 = p - p^2 = p(1 - p)$ .
- Sometimes write  $X \sim \text{Ber}(p)$ .

Examples:

- Coin flip
- Random binary digit
- Packet erasure in a wireless network

## Binomial Random Variable

Consider  $n$  independent trials of a  $Ber(p)$  random variable

- $X$  is the number of successes in  $n$  trials
- $X$  is the sum of  $n$  Bernoulli random variables,  $X = X_1 + X_2 + \dots + X_n$ , where random variable  $X_i \sim Ber(p)$  is 1 if success in trial  $i$  and 0 otherwise.
- $X$  is a **Binomial** random variable:  $X \sim Bin(n, p)$

$$P(X = i) = \binom{n}{i} p^i (1 - p)^{n-i}, \quad i = 0, 1, \dots, n$$

(recall  $\binom{n}{i}$  is the number of outcomes with exactly  $i$  successes and  $n - i$  failures)

- $E[X] = np$ ,  $Var(X) = np(1 - p)$  ( $X$  is the sum of  $n$  independent Bernoulli RVs)

Examples:

- number of heads in  $n$  coin flips
- number of 1's in randomly generated bit string of length  $n$
- number of packets erased out of a file of  $n$  packets

## Example

A shopper wants to compare bottles of wine. From a shelf with 6 bottles labelled A-F, each different, he selects 3 independently and uniformly at random. What is the probability that he picks bottle B ?

- Let indicator random variable  $X = 1$  if pick bottle B and 0 otherwise. For  $X = 1$  there are three cases to consider:
  - Picks B first. Happens with probability  $\frac{1}{6}$ .
  - Does not pick B first but picks B second. Happens with probability  $(1 - \frac{1}{6})\frac{1}{5}$ .
  - Does not pick B first or second but picks B third. Happens with probability  $(1 - \frac{1}{6})(1 - \frac{1}{5})\frac{1}{4}$ .
- So  $P(X = 1) = \frac{1}{6} + (1 - \frac{1}{6})\frac{1}{5} + (1 - \frac{1}{6})(1 - \frac{1}{5})\frac{1}{4} = 0.5$

## Example

Joe Lucky plays the lottery on any given week with probability  $p$ , independently of other weeks. Each time he plays he has probability  $q$  of winning. During a period of  $n$  weeks, let  $X$  be the number of times that he played the lottery and  $Y$  the number of times that he won.

- What is the probability that he played the lottery in a week given that he did not win anything that week ?
- Let  $E$  be the event that he played and  $F$  the event that he did not win. Use Bayes Rule.

$$\begin{aligned}P(E|F) &= \frac{P(F|E)P(E)}{P(F)} = \frac{P(F|E)P(E)}{P(F|E)P(E) + P(F|E^c)P(E^c)} \\ &= \frac{(1-q)p}{(1-q)p + 1 \times (1-p)} = \frac{p-pq}{1-pq}\end{aligned}$$

## Example (cont)

Recall  $X$  is the number of times that he played the lottery and  $Y$  the number of times that he won.

- What is the conditional PMF  $P(Y = y|X = x)$  ?
- Its Binomial:

$$P(Y = y|X = x) = \begin{cases} \binom{x}{y} q^y (1 - q)^{x-y} & 0 \leq y \leq x \\ 0 & \text{otherwise} \end{cases}$$



## Example

Since there is no direct flight from Dublin (D) to Atlanta (A) you need to travel via either Chicago (C) or New York (N). Flights from D to C and from C to A independently are delayed by 2 hours with probability  $p$ . Flights from D to N and from N to A independently are delayed by 1 hour with probability  $q$ . Each time you fly you choose to fly via C or N with equal probability.

- What is the average delay from D to A ?
- Let random variable  $X_{DC}$  be the delay D to C (either 0 or 2), similarly  $X_{CA}$ ,  $X_{DN}$  and  $X_{NA}$ .

$$\begin{aligned}E[\text{delay}] &= E[\text{delay}|C]P(C) + E[\text{delay}|N]P(N) \\&= E[X_{DC} + X_{CA}]P(C) + E[X_{DN} + X_{NA}]P(N) \\&= E[X_{DC}]P(C) + E[X_{CA}]P(C) + E[X_{DN}]P(N) + E[X_{NA}]P(N) \\&= 2p\frac{1}{2} + 2p\frac{1}{2} + q\frac{1}{2} + q\frac{1}{2} = 2p + q\end{aligned}$$

## Example (cont)

Suppose you arrive with delay 2 hours. What is the probability that you travelled via New York ?

- Let  $E$  be the event that travelled via New York and  $F$  be the event that delayed 2 hours. Use Bayes:

$$\begin{aligned} P(E|F) &= \frac{P(F|E)P(E)}{P(F)} = \frac{P(F|E)P(E)}{P(F|E)P(E) + P(F|E^c)P(E^c)} \\ &= \frac{q^2 \frac{1}{2}}{q^2 \frac{1}{2} + 2p(1-p) \frac{1}{2}} = \frac{q^2}{q^2 + 2p(1-p)} \end{aligned}$$