

# Review of Key Concepts

Demetris Athienitis



# Review of key statistical concepts

Number of concepts which will come up quite frequently and it is important to understand them before we begin:

- Random variables
  - Probability mass function
  - Expected value
  - Variance
  - Rules for means and variances
- Conditional probability

# Random variables

## Definition (Random Variable)

A random variable is a function that assigns a numerical value to each outcome of an experiment. It is a measurable function from a probability space into a measurable space known as the state space.

## Example

- Flipping a coin 10 times and counting how many heads turn up
- Rolling two dice and observing the sum
- The range is the set of possible values it can take
  - If we flip a coin 10 times then the range is  $\{0, 1, 2, \dots, 10\}$
  - Sum of two dice  $\{2, 3, \dots, 12\}$
- Seek to understand the probability that the random variable takes

# Probability mass function

- Let  $Y$  be a categorical random variable of interest
- The probability mass function (PMF) is defined as  $P(Y = y)$  for any  $y \in \text{Range}(Y)$
- Describes the probability behavior of the random variable
- We will generally denote random variables with capital letters and potential values the random variable can take with lower-case letters

# Expected value

- The expected value is denoted by  $E(Y) = \sum_{\forall y} yP(Y = y)$
- $E(Y)$  is the mean of the random variable
- If we could re-run the experiment that generates  $Y$  an infinitely large number of times,  $E(Y)$  would be the average value of the random variable.
- Represents the average value one would expect to see
  - Note that  $E(Y)$  does not need to be in the range of  $Y$

- The variance of a random variable  $Y$  is defined as

$$V(Y) = E\left[(Y - \mu)^2\right]$$

where  $\mu = E(Y)$

- Describes the average squared deviation from the mean
- Provides a measure of how uncertain we are about the value of  $Y$
- $SD(Y) = \sqrt{V(Y)}$

# Rules for means and variances

Let  $X$  and  $Y$  be random variables throughout and let  $a, b$  be known constants

- $E(aX + b) = aE(X) + b$
- $E(aX + bY) = aE(X) + bE(Y)$
- $V(aX + b) = a^2 V(X)$
- $V(aX + bY) = a^2 V(X) + b^2 V(Y) + 2ab\text{Cov}(X, Y)$

# Conditional probability

Let  $X$  and  $Y$  be random variables

- $P(Y = y|X = x)$  is the probability that the random variable  $Y$  takes value  $y$  conditional on the fact that  $X = x$
- Two key properties of conditional probability:

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

$$P(Y = y|X = x) = \frac{P(X = x|Y = y)P(Y = y)}{P(X = x)}$$



- A statistic is simply a function of the observed data
- Statistics, before a value is observed is are random variables
- The standard deviation of a statistic is referred to as its *standard error*
- Estimate unknown parameters  $\theta$  and denote estimates as  $\hat{\theta}$
- $\hat{\theta}$  will be a function of the observed data

# We learned

- Key concepts about (discrete) random variables
- Conditional probabilities