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# Random Variables and Their Distributions Week 10

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#### **Review of Week 9**

- Probability (Chapter 6.1.)
  - Definitions
  - Axioms
  - Permutations
  - Combinations
- Conditional Probability (Chapter 6.2.)
  - Conditional probability
  - Joint probability
  - Independence

#### **Example 1: Write Down Equivalence/Definition and Plug in Values**

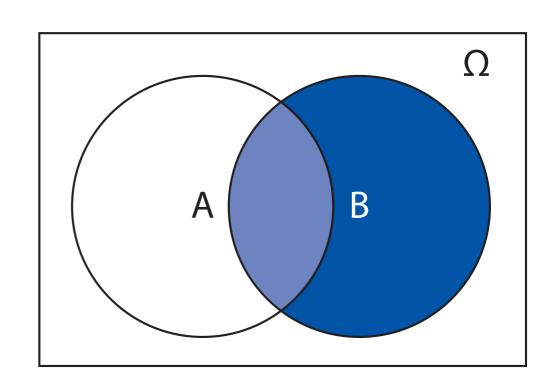
- Rolling a dice once
  - A: a multiple of 2; B: a multiple of 3
    - $\mathbf{P} \Omega =$
    - $\triangleright$  as a set: A =
    - $\blacktriangleright$  as a set:  $A^c =$
    - $\triangleright$  P(A) =
    - $ightharpoonup P(A^c) =$
    - $\triangleright$  P(A and B) =
    - $\triangleright$  P(A|B) =
    - $\triangleright$  P(B|A) =
    - Arr P(A) = P(A and B) + P(A and Bc)

$$B =$$

$$B^c =$$

$$P(B) =$$

$$P(B^c) =$$



#### **Example 2: Write Down Equivalence/Definition and Plug in Values**

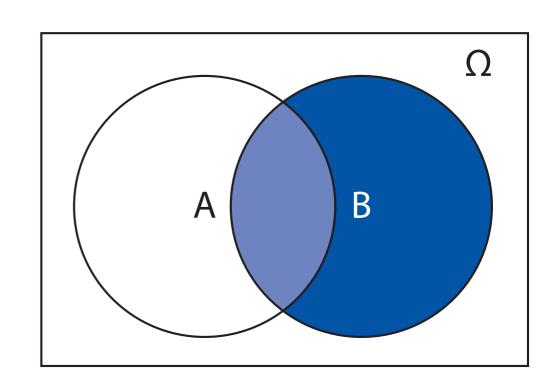
- Flipping coin 3 times
  - A: first flip head; B: second flip tail
    - $\mathbf{P} \Omega =$
    - $\triangleright$  as a set: A =
    - $\blacktriangleright$  as a set:  $A^c =$
    - $\triangleright$  P(A) =
    - $ightharpoonup P(A^c) =$
    - $\triangleright$  P(A and B) =
    - $\triangleright$  P(A|B) =
    - $\triangleright$  P(B|A) =
    - Arr P(A) = P(A and B) + P(A and Bc)

$$B =$$

$$B^c =$$

$$P(B) =$$

$$P(B^c) =$$



## **Example 3: Permutations and Combinations**

- You are a producer in a large entertainment company
  - You want to form a 5-member unit from <u>IZ\*ONE</u>
    - Assume that you are selecting 5 completely at random
    - ► What is the chance that you select 1 JP member and 4 KR member?
    - What is the chance that you select at least 2 JP members?
  - ► You are selecting 7 members from IZ\*ONE to cover a <u>BTS</u> song
    - Each will be assigned to a different role (e.g. V, Jin, RM ...)
      - How many potential scenarios are there?

# Bayes' Rule: Check Textbook 6.2.4

Bayes' rule (obvious if you know definitions / axioms)

$$\underbrace{\Pr(A \mid B)}_{conditional \ probability} = \underbrace{\frac{\Pr(A \text{ and } B)}{\Pr(B)}}_{conditional \ probability} = \underbrace{\frac{\Pr(B \mid A) \Pr(A)}{\Pr(B \mid A) \Pr(A)}}_{pr(B \mid A) \Pr(A) + \Pr(B \mid \text{not } A) \Pr(\text{not } A)}$$

- ► What does Pr(A|B) mean?
  - ► The chance of A being true given B is observed (i.e. true)
- Why is above the most important result in probability theory?
  - e.g. A is an event happened before B; only B is observed
    - Using Bayes' rule: you can infer about A given B
- Bayesian updating: prior belief Pr(A) -> posterior belief Pr(A|B)

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- Posterior probability Pr(A|B): belief of A after observing evidence B
- Prior probability (belief of A happening w/o evidence): P(A)
- Prior probability (belief of A not happening w/o evidence): P(not A)
- Examples (A: prior event; B: data/evidence)

A (prior event)	I voted for A	having a disease	studied hard
B (data)	A won	positive on medical test	high grade in exam

Time order is not necessary but good for examples

## **Example: Monty Hall Problem**

- ► The most famous probability problem
  - http://www.youtube.com/watch?v=mhlc7peGlGg



Even a great mathematician failed!

## **Example: Monty Hall Problem**

- You pick door A. Monty opens door C that has a goat.
  - Should you switch to door B?
- Prior beliefs: P(A) = P(B) = P(C) = 1/3 (e.g. A: the car is behind door A)
- Data: Monty reveals door C (i.e. MC)
- ▶ Posterior belief (inferential goals):  $P(A \mid MC)$  and  $P(B \mid MC)$
- ▶ Question:  $P(A \mid MC) < P(B \mid MC)$
- What do we need? Bayes' rule!
  - Key: Monty's behavior is constrained to open a goat door

▶ Take home message: Do not believe your intuition, rely on logic.

#### **Contents**

- Bayes' rule (Chapter 6.2.3)
- Random Variables and Probability Distributions (Chapter 6.3)
  - Overview
  - Bernoulli and uniform distributions
  - Binomial distribution
  - Uniform distribution
  - Normal (or Gaussian) distribution
  - Expectation

## Random Variables and Probability Distribution

- Random variable assigns a numeric value to each event of the experiment.
  - Coin flip side: head = 1; tail =0
  - #secs took for commuting: any value greater than 0.
  - These values represent mutually exclusive and exhaustive events, together forming the entire sample space  $\Omega$ .
  - ► A discrete RV takes a finite or at most countably infinite #distinct values.
    - coin flip; race; number of years of education
  - A continuous RV assumes an uncountably infinite number of values.
    - height, distance from earth, gross domestic product
  - Probability distribution: Probability that a random variable takes a certain value or range of values.
    - P(side): P(side=1) = 0.5; P(side=0) = 0.5
    - P(#secs): P(0<#secs<1000) = 0.3; P(1000<#secs<2000) = 0.4 ....

# See you next week.