# W1211 Introduction to Statistics Lecture 3

Wei Wang

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# Measure of Variability

- Two samples that have same mean or median might have drastically different spreads.
- So to numerically summarize the sample, we need a measure of variability.

#### Standard deviation

- The variance and standard deviation are measures of spread that indicate how far values in the data set are from the mean, on average.
- Consider the observations  $x_1, x_2, x_3, \ldots, x_n$ .
- The deviations  $(x_i \bar{x})$  display the spread of  $x_i$  about their mean  $\bar{x}$ .
- The sum of the deviations is always 0, as some of the deviations are positive and others are negative.
- Squaring the deviations makes them all positive. Observations far from the mean will have large positive squared deviations.
- The variance is the 'average' squared deviation.

#### Standard deviation

• If we have *n* observations  $x_1, x_2, x_3, \dots, x_n$ . The variance is defined as

$$s^{2} = \frac{(x_{1} - \overline{x})^{2} + (x_{2} - \overline{x})^{2} + \dots + (x_{n} - \overline{x})^{2}}{n - 1} = \frac{1}{n - 1} \sum_{i=1}^{n} (x_{i} - \overline{x})^{2}$$

- The standard deviation, s, is the square root of the variance.
  - s is a measure of spread about the mean and should be used when the mean is used as the measure of center.
  - 2. If s=0, then all the values in the data set are exactly the same (no spread). Why?
  - 3. The more spread out the data, the greater the standard deviation.
  - 4. s is always positive.
  - 5. s has the same unit of measurement as the original data

## A short cut formula for $s^2$

#### Theorem

An alternative expression for variance  $s^2$  is

$$s^{2} = \frac{1}{n-1} \left( \sum_{i=1}^{n} x_{i}^{2} - n(\bar{x})^{2} \right)$$

#### Proof.

Do some algebra on the numerator.

$$\sum_{i=1}^{n} (x_i - \bar{x})^2 = \sum_{i=1}^{n} (x_i^2 - 2\bar{x} \cdot x_i + (\bar{x})^2)$$

$$= \sum_{i=1}^{n} x_i^2 - 2\bar{x} \sum_{i=1}^{n} x_i + \sum_{i=1}^{n} (\bar{x})^2$$

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# Fourth Spread

- ▶ Fourth Spread is a robust measure of variability based on quartiles.
- ▶ It is defined as

 $f_S$  = upper fourth – lower fourth

#### **Boxplot**

- A five number summary lists, in order, the minimum, Q1, the median, Q3, and the maximum.
- A boxplot is a graphical representation using a five number summary.
  - 1. Draw a vertical (horizontal) measurement scale.
  - 2. Place a rectangle to the right of (above) this axis; the lower (left) edge of the rectangle is at the lower fourth, and the upper (right) edge is at the upper fourth.
  - 3. Place a horizontal (vertical) line segment inside the rectangle at the location of the median.
  - 4. Draw "whiskers" out from either end of the rectangle to the smallest and largest observations that are NOT outliers.
  - 5. Using dots to represent outliers.
- R demo. >boxplot(x)

# **Probability**

#### What is randomness?

- The world is full of random events that we seek to understand.
- An event is random if we know what outcomes could occur, but not the particular values that will happen.
- The outcome of these events is uncertain, but they follow a regular pattern.
- Deterministic models vs. Random models.
- Probability theory is the mathematical representation of random phenomena.

#### **Notation**

- An experiment is any action or process whose outcome is subject to uncertainty.
   e.g. tossing a coin once or several times; selecting a card or cards from a deck; weighing a loaf of bread; etc.
- The sample space of an experiment, denoted by S, is the set of all possible outcomes of that experiment.

Ex. Flip a coin. Two possible outcomes: Heads (H) or Tails (T). S={H,T}.

Ex. Battery life.  $S=\{x: 0 \le x < \infty\}$ .

#### **Notation**

- An event is any collection of possible outcomes, that is, any subset of S
  (including S itself). An event is simple if it consists of exactly one outcome and
  compound if it consists of more than one outcome.
- If the outcome of a random phenomenon is contained in an event A, then we say that A has occurred.
- Ex. Flip a coin twice. Four possible outcomes, S={HH, HT, TH, TT}. Let A be the event that we obtain at least one H in the two flips. A={HH, HT, TH}. Let B be the event that we obtain two H's in the two flips. B={HH}.
- Ex. Battery life example. The event that the battery lasts less than 3 hours is denoted as  $A=\{x: 0 \le x < 3\}$ .

#### **Set Operations**

 Given any two events (or sets) A and B, we have the following elementary set operations:

The union

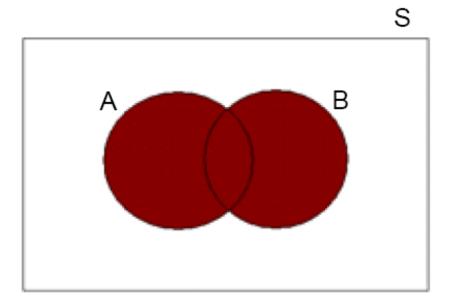
The intersection

The complement

Venn diagrams are often used to illustrate relationships between sets.

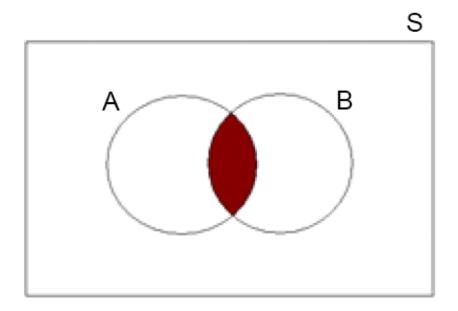
#### Union

• The union of A and B, written as AUB and read "A or B", is the set of outcomes that belong to either A or B or both.



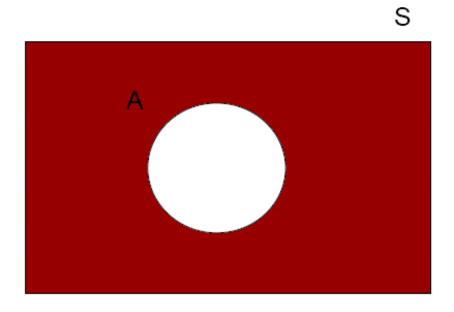
#### Intersection

 The intersection of A and B, written as A∩B, read "A and B", is the set of outcomes that belong to both A and B.



# Complement

 The complement of A, written as A' or A<sup>c</sup>, is the set of all outcomes in S that are not in A.



#### **Example**

Ex. Select a card at random from a standard deck of cards, and note its suit: clubs (CI), diamonds (D), hearts (H) or spades (Sp).

The sample space is S={Cl, D, H, Sp}.

Let: A={CI, D}, B={D, H, Sp} and C={H}.

 $AUB=\{CI, D, H, Sp\}=S$ 

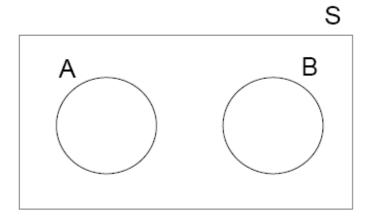
 $A \cap B = \{D\}$ 

 $A^c=\{H, Sp\}$ 

A∩C= ∅ (null event – event consisting of no outcomes)

# **Disjoint events**

If A∩B= ∅ then A and B are said to be mutually exclusive or disjoint events.



Any event and its complement are disjoint!

### **Probability models**

- A probability model consists of a sample space and the assignment of probabilities to each possible outcome.
- Probability that event A occurs is written as P(A), which will give a precise measure of the chance that A will occur.
- To ensure the probability assignments will be consistent with our intuitive notions of probability, all assignments should satisfy the following axioms (basic properties) of probability.
  - For any event A, P(A)≥0.
  - 2. P(S)=1.
  - If  $A_1$ ,  $A_2$ ,  $A_3$ , ... is an infinite (finite) collection of disjoint events, then

$$P(A_1 \cup A_2 \cup A_3 \cup ...) = \sum P(A_i)$$

# **Propositions**

- ▶ For any event A,  $0 \le P(A) \le 1$ .
- $P(A) + P(A^{c}) = 1.$
- ▶ If event A is contained in event B, in the sense that every outcome in A is also in B, then

$$P(A) \leq P(B)$$

 $P(\emptyset) = 0.$ 

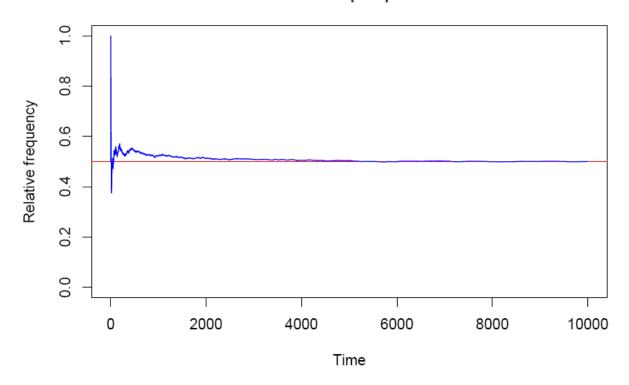
### **Interpreting Probability**

 What does it mean when we say we have 50% chance of having a head when flipping a coin? Or what does it mean when we put P(H)=0.5?

 Probability is often treated as the long-term relative frequency or the limiting relative frequency.

# **Interpreting Probability**

Ex. Flip a fair coin *n* times and calculate the proportion of heads.



R demo. (Function: sample(x, size); rbinom(x, size, prob))

#### Law of Large Numbers

 The law of large numbers says that the long-run relative frequency of repeated independent events gets closer and closer to the true relative frequency as the number of trials increases.

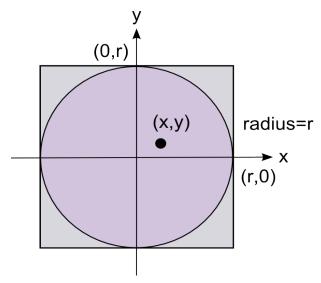
# of occurrence of event A

# of trials 
$$(n)$$
 $n \geq \infty$ 

P(A)

#### How to calculate Pi

- An interesting application of Law of Large Numbers is to calculate Pithrough simulations.
- ▶ If we spread a large quantity of seeds randomly but evenly on this square, what percentage of the seeds will lie inside the circle?



- R Demo.
- This type of simulation-based methods has a fancy name: Monte Carlo methods.

### **Assigning Probabilities**

- The assignment of probabilities can often be derived from the physical set-up of an experiment.
- Suppose we have N outcomes in our sample space, each equally likely to occur.
   The each has a probability of 1/N, and the probability of any event A is,

$$P(A) = \frac{\text{number of outcomes in A}}{N}$$

Ex. Roll a fair die. S={1,2,3,4,5,6}. Our sample space consists of 6 points, each of which is equally likely to occur.

P(roll a 1) = 1/6.

Let A = roll a 4 or less =  $\{1,2,3,4\}$ . P(A) = 4/6.

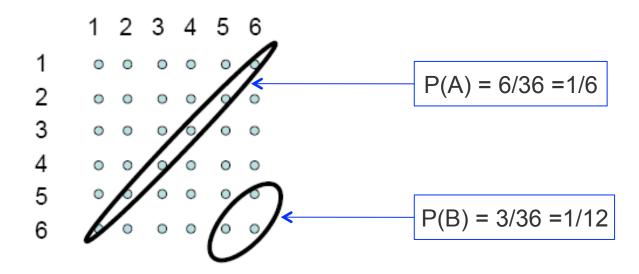
Let B = roll an even number =  $\{2,4,6\}$ . P(B) = 3/6.

#### **Example**

Ex. Roll two fair dice.

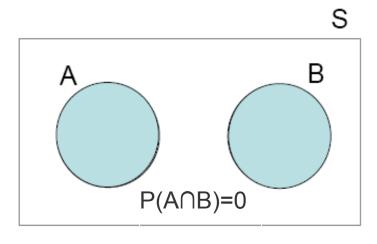
There are 36 possible outcomes:  $\{(1,1),(1,2),(1,3),...,(6,5),(6,6)\}$ .

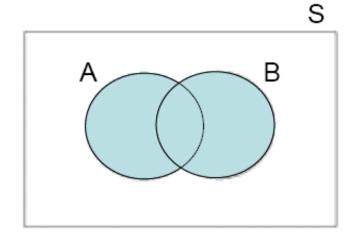
Let A = sum of two rolls is 7; B = sum of two rolls is 11 or more. What are P(A) and P(B)?



#### **More Probability Properties**

- Consider an experiment whose sample space is S. For each event A (B) in S, we assume that a number P(A) is defined and satisfies the following rules:
  - 1.  $0 \le P(A) \le 1$ .
  - 2. P(S)=1.
  - 3.  $P(A^c)=1-P(A)$ .
  - 4. If A and B are disjoint, then P(AUB)=P(A)+P(B).
  - 5. For any two events A and B,  $P(AUB)=P(A)+P(B)-P(A\cap B)$ .





#### **Example**

Ex. A store accepts either VISA or Mastercard. 50% of the stores customers have VISA, 30% have Mastercard and 10% have both. What is the probability that a customer has a credit card the store accepts?

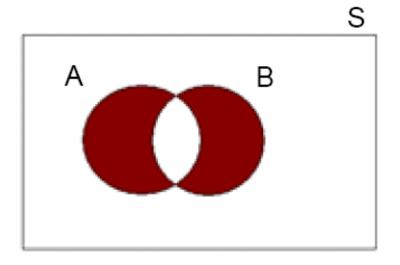
A = customers has VISA

B = customers has Mastercard

$$P(AUB) = P(A) + P(B) - P(A \cap B)$$
  
= 0.5 + 0.3 - 0.1 = 0.7

#### **Example cont.**

What is the probability that a customer has either a VISA or MC, but not both?



P(A or B but not both) = P(A) + P(B) - 2P(A \cap B)  
= 
$$0.5 + 0.3 - 0.2 = 0.6$$

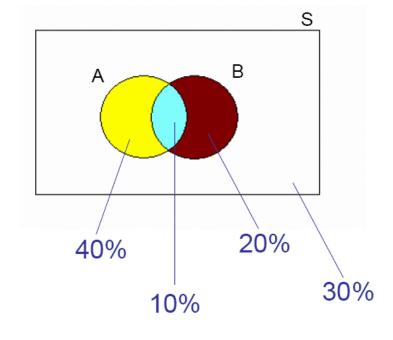
#### **Example Cont.**

What is the probability that a customer has a VISA but no MC?

P(A but not both) = P(A) – P(A
$$\cap$$
B)  
= 0.5 – 0.1 = 0.4

What is the probability that a customer has a MC but no VISA?

P(B but not both) = P(B) - P(A
$$\cap$$
B)  
= 0.3 - 0.1 = 0.2



#### **Three Events**

For any three events A, B and C,

$$P(AUBUC) = P(A) + P(B) + P(C) - P(A \cap B) - P(A \cap C)$$
$$- P(B \cap C) + P(A \cap B \cap C)$$

