# Topic 1: Probability STAT 211 - 509 9/4/2018

# Sample spaces and events

- Outcome: possible result of an experiment.
- Sample space (S): set of all possible outcomes of an experiment.
- **Event**: any collection (subset) of outcomes contained in the sample space S.

## Example

Toss a fair coin twice. List the sample space and define the event *A* as "the first toss is heads"

- Sample space:  $S = \{HH, HT, TH, TT\}$
- Event:  $A = \{HH, HT\}$

## Set operations

- The **union** of two events *A* and *B*, denoted *A* ∪ *B* is the event consisting of all outcomes that are either in A, in B, or in both
- The intersection of two events A and B, denoted  $A \cap B$  is the consisting of all outcomes in both A and B. A and B are **mutually exclusive** if they do not have any outcomes in common.
- The complement of an event A, denoted A', is the set of all outcomes in S that are not contained in A

# **Probability**

- Probability theory is the study of randomness and uncertainty
- Given an experiment and a sample space S, want to assign each A
  the number P(A), which is the measure of the chance that A will
  occur
- $\bullet\,$  The assignments  $P(\cdot)$  should satisfy axioms that accord with intuitive notions of probability

- 1. For any event A,  $P(A) \ge 0$
- 2. P(S) = 1
- 3. If  $A_1, A_2, \ldots, A_k$  is a collection of mutually exclusive events, then

$$P(A_1 \cup A_2 \cup \cdots \cup A_k) = \sum_{i=1}^k P(A_i)$$

*Implications* 

- For the empty set  $\emptyset$ ,  $P(\emptyset) = 0$
- For any event *A*:

$$-0 \le P(A) \le 1$$
  
 $-P(A') = 1 - P(A)$ 

• For any events *A* and *B*:

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

Example

Toss a fair coin twice. What is the probability that at least one toss is heads? First, define relevant events

- Sample space:  $S = \{HH, HT, TH, TT\}$
- Let  $A = \{ \text{at least one toss is heads} \} = HH \cup HT \cup TH$ What is P(A)?
- Union:

$$P(A) = P(HH \cup HT \cup TH) = \frac{1}{4} + \frac{1}{4} + \frac{1}{4} = \frac{3}{4}$$

• Complement:

$$A' = \{ \text{no toss is heads} \} = TT$$
 
$$P(A') = \frac{1}{4}$$
 
$$P(A) = 1 - P(A') = \frac{3}{4}$$

• Another:

Let  $H_1 = \{\text{heads on first toss}\}\$ and  $H_2 = \{\text{heads on second toss}\}\$ Then  $A = \{\text{at least one toss is heads}\} = H_1 \cup H_2$ 

$$P(A) = P(H_1 \cup H_2)$$

$$= P(H_1) + P(H_2) - P(H_1 \cap H_2)$$

$$= \frac{1}{2} + \frac{1}{2} - \frac{1}{4} = \frac{3}{4}$$

What is the probability exactly one toss is heads?  $H_1 = \{\text{heads on first toss}\}\$ and  $H_2 = \{\text{heads on second toss}\}\$  $A = \{\text{exactly one toss is heads}\} = (H_1 \cap H_2') \cup (H_1' \cap H_2)$ 

# Uniform probability

Suppose we have a finite sample space S with N outcomes, integer  $N \ge 1$ . Each outcome is equally likely. If there are m outcomes in event A, then

$$P(A) = \frac{m}{N}$$

#### Example

Roll two fair dice, each outcome in sample space has equal probability 1/36

$$S = \{ (1,1), (1,2), (1,3), (1,4), (1,5), (1,6), (2,1), (2,2), (2,3), (2,4), (2,5), (2,6), (3,1), (3,2), (3,3), (3,4), (3,5), (3,6), (4,1), (4,2), (4,3), (4,4), (4,5), (4,6), (5,1), (5,2), (5,3), (5,4), (5,5), (5,6), (6,1), (6,2), (6,3), (6,4), (6,5), (6,6) \}$$

What is the probability that the sum of the two faces is 7?  $A = \{\text{sum of the two faces is 7}\} = \{(1,6), (2,5), (3,4), (4,3), (5,2), (6,1)\}$ 6 outcomes in A, 36 possible outcomes:  $P(A) = \frac{6}{36} = \frac{1}{6}$ 

#### Counting

Suppose we toss a coin 10 times.

- How many outcomes are there?
- How many would have exactly 4 heads?
- What is the probability of exactly 4 heads?

### Counting rules

- Rule of sum: if there are a ways to do thing one and b ways to do thing two, and you can't do both things, then there are a + b things to do in total
- Rule of product: if there are *a* ways to do thing one, *b* ways to do thing two, then there are *ab* ways to do both

# Binomial coefficient

- **Binomial coefficients** are family of positive integers that occur as coefficients in the binomial theorem
- Coefficient of the  $x^k$  term in the polynomial expansion of  $(1+x)^n$
- Indexed by two nonnegative integers *n* and *k*

•

$$\binom{n}{k} = \frac{n!}{k!(n-k)!}$$

• The number of distinct ways of choosing k objects out of n

## Example

We toss a coin 10 times

- How many possible outcomes are there?  $2^{10}$
- How many of these outcomes have exactly 4 heads?  $\binom{10}{4}$
- What is the probability of exactly 4 heads?

$$P(X = 4) = \frac{\binom{10}{4}}{2^{10}} = \frac{210}{1024} = 0.205078125$$

# Independence

• Two events *A* and *B* are **independent** if and only if

$$P(A \cap B) = P(A)P(B)$$

• Intuitively, probability of *A* does not depend on whether or not *B* occurred, and vice versa.

- Many statistical methods assume independence between all pairs of observations in dataset
- Independence of two events does not imply they are mutually exclusive
- Examples of dependence:
  - Multiple measurements of same individual over time
  - Measurements located near each other

#### NFL example

Are the outcomes of two games independent of one another?

- Suppose previous game was demoralizing loss. Is the team extra motivated for next?
- Suppose the team won previous game, guaranteeing entrance into playoffs. Is the team *less* motivated for next?
- Suppose team has won 5 games in a row. Do they have momentum?

#### Random variables and distributions

- For a sample space *S* of some experiment, a **random variable (rv)** is any rule that associates a number with each outcome in *S*.
  - variable because different numerical values possible
  - random because observed value depends on experimental out-
  - a function that maps from sample space to real numbers: *X* :  $S \to \mathbb{R}$

## Types of random variables

- A random variable is **discrete** if its possible values are from a finite set, or can be listed in an infinite sequence with first, second, etc. elements
- A random variable is **continuous** if its possible values are from an entire interval of the real line
- The random variables  $X_1, X_2, ..., X_n$  are independent and identically distributed (iid) if they are mutually independent and all follow the same distribution

• The **probability mass function (pmf)** of a discrete rv is defined for every number *x* by

$$f(x) = P(X = x) = P(s \in S : X(s) = x)$$

P(X = x) is the probability that the rv X assumes the value x

• The **cumulative distribution function (cdf)** F(x) of a discrete rv X with pmf f(x) is defined for every number x by

$$F(x) = P(X \le x) = \sum_{y:y \le x} f(y)$$

For any number x, F(x) is the probability that the observed value of X will be at most x

## Example

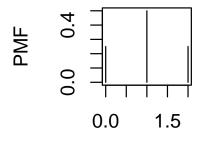
Flip a fair coin twice. Derive and plot the pmf and cdf. Let X be the random variable that equals the number of occurences of heads. The possible values of X are 0,1,2.

• pmf:

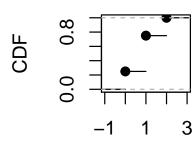
$$P(X = x) = \begin{cases} 1/4, & x = 0 \\ 1/2, & x = 1 \\ 1/4, & x = 2 \\ 0, & x \notin \{0, 1, 2\} \end{cases}$$

• cdf:

$$P(X \le x) = \begin{cases} 0, & x < 0 \\ 1/4, & 0 \le x < 1 \\ 3/4, & 1 \le x < 2 \\ 1, & x \ge 2 \end{cases}$$



X



X

#### The binomial distribution

#### Bernoulli distribution

- Consider experiment with single success/failure trial, with probability of success  $p \in [0,1]$ . The Bernoulli random variable X associated with this experiment is 0 if trial is a failure, 1 if it is a success.
- pmf of *X*:

$$f(x) = \begin{cases} p^x (1-p)^{1-x} & x \in \{0,1\} \\ 0 & \text{otherwise} \end{cases}$$

• p is **parameter** of the distribution, must know to evaluate f(x)

#### Binomial distribution

- Consider an experiment consisting of *n* independent Bernoulli trails. The binomial random variable *X* associated with this experiment is defined as the number of successes out of the *n* trials
- pmf of *X*:

$$f(x) = \begin{cases} \binom{n}{x} p^x (1-p)^{1-x} & x \in 0, 1, \dots, n \\ 0 & \text{otherwise} \end{cases}$$

- *n* and *p* are parameters of the distribution
- We write  $X \sim Binomial(n, p)$

#### Example

Suppose we toss a coin 10 times.  $X \sim Binomial(10, 0.5)$ . What is the probability of exactly 4 heads?

$$P(X = 4) = f(4) = {10 \choose 4} 0.5^4 0.5^{10-4} = 0.205078125$$

### NFL example

- Let X be the number of wins out of a season's 16 games. If we assume that the outcome of each game is independent of all others, we can say  $X \sim Binomial(16, p)$ , where p is the probability that the Texans win any given game.
- Suppose that p = 0.5. What is the probability the Texans win 9 or more games in a season?

- $P(X \ge 9) = \sum_{x=9}^{16} f(x) = 0.4018097$
- Would not be unusual at all for an "average" team to win 9 or more games in a season, just by chance.

#### Other discrete distributions

Discrete uniform distribution: rv *X* with integer parameters *a*, *b*,
 a ≤ b, with pmf:

$$f(x) = \frac{1}{b-a+1}$$
  $x = a, a+1, ..., b-1, b$ 

• **Geometric distribution**: rv X with parameter p > 0, with pmf:

$$f(x) = (1-p)^x p$$
  $x = 0, 1, 2, ...$ 

Describes the number of failures required until the first success in a series of Bernoulli trials

• **Poisson distribution**: rv with parameter  $\lambda > 0$ , with pmf:

$$f(x) = \frac{e^{-\lambda}\lambda^x}{x!} \quad x = 0, 1, 2, \dots$$

Describes counts of events, like number of 911 calls on Friday night

#### Binomial distribution in R

```
• X \sim Binomial(10, .3)
```

```
• P(X = 3):
```

n <- 10

p < -0.3

dbinom(3, n, p)

## [1] 0.2668279

•  $P(X \ge 3)$ :

1 - **pbinom**(3, n, p)

## [1] 0.3503893

pbinom(3, n, p, lower.tail = FALSE)

## [1] 0.3503893