

# Histogram

- Most commonly used tool in descriptive statistics.
- Histogram for discrete data:
  - Determine the frequency and relative frequency of each  $x$  value.
  - Mark possible  $x$  values on a horizontal scale.
  - Above each value, draw a rectangle whose height is the relative frequency (or the frequency) of that value.
- Histogram for continuous data:
  - Divide the range of the data into classes (5-10) of *equal width*. (It can also be unequal.)
  - Determine the frequency and relative frequency for each class.
  - Mark the class boundaries on a horizontal measurement axis.
  - Above each class interval, draw a rectangle whose height is the corresponding relative frequency (or frequency).

# Constructing histogram

- **Example:** The maximum daily temperature in degrees Fahrenheit measured from May to September 1973 at La Guardia Airport. (154 observations)

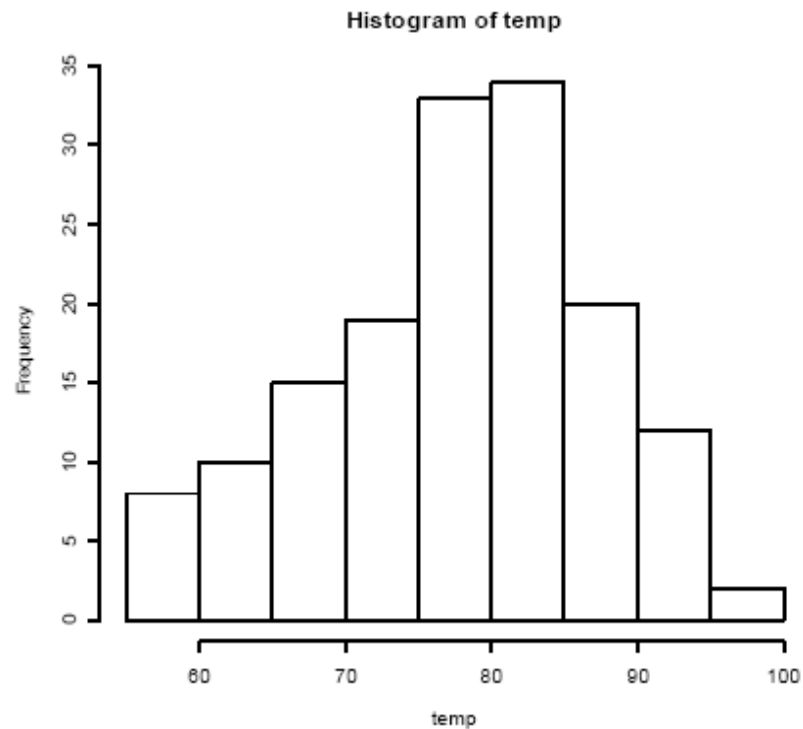
Data

{67 72 74 62 56 66 65 59 61 69 74 69 66 68 58 64 66 57 68 62 59 73 61 61 57  
58 57 67 81 79 76 78 74 67 84 85 79 82 87 90 87 93 92 82 80 79 77 72 65 73  
76 77 76 76 76 75 78 73 80 77 83 84 85 81 84 83 83 88 92 92 89 82 73 81 91  
80 81 82 84 87 85 74 81 82 86 85 82 86 88 86 83 81 81 81 82 86 85 87 89 90  
90 92 86 86 82 80 79 77 79 76 78 78 77 72 75 79 81 86 88 97 94 96 94 91 92  
93 93 87 84 80 78 75 73 81 76 77 71 71 78 67 76 68 82 64 71 81 69 63 70 77  
75 76 68}

Draw a histogram.

## Example cont.

| Class   | Count | Percent |
|---------|-------|---------|
| 55-59.9 | 8     | 5.2     |
| 60-64.9 | 10    | 6.5     |
| 65-69.9 | 15    | 9.8     |
| 65-74.9 | 19    | 12.4    |
| 75-79.9 | 33    | 21.6    |
| 80-84.9 | 34    | 22.2    |
| 85-89.9 | 20    | 13.1    |
| 90-94.9 | 12    | 7.9     |
| 95-99.9 | 2     | 1.3     |



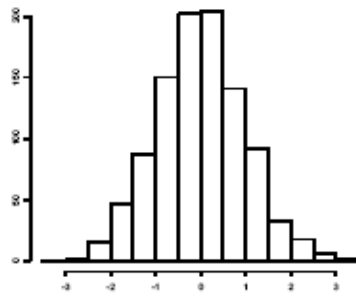
- R demo. `>hist(x)` (option: `breaks=...`)

# Examining distributions

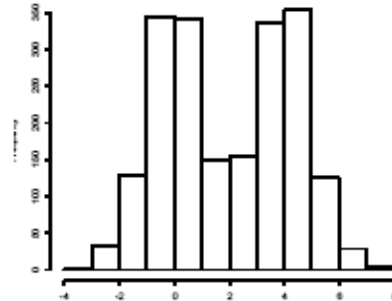
- When examining a distribution, look at its **shape**, **center** and **spread**. Look for clear deviations from the overall shape.
- We are interested in whether it is symmetric or skewed, as well as the number of modes.
- **Outliers** are observations that lie outside of the overall pattern of a distribution.

# Examining distributions

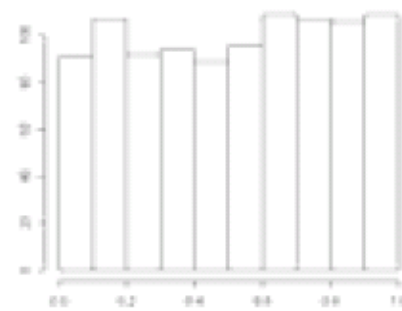
(a) Symmetric, unimodal



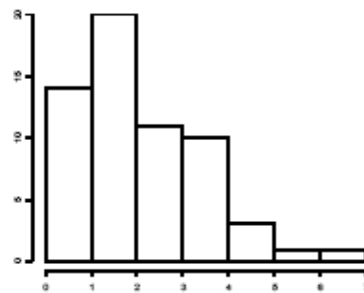
(b) bimodal



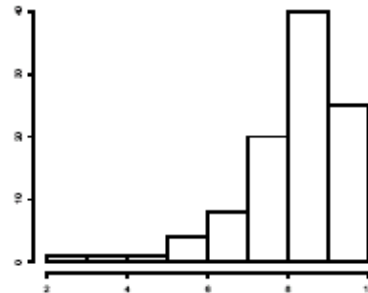
(c) Uniform



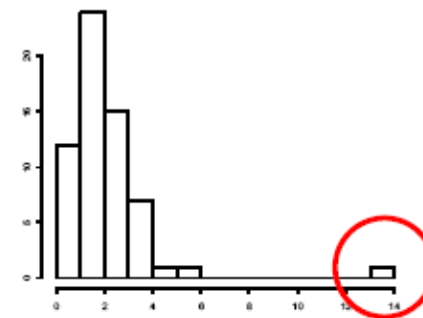
(d) right skewed



(e) left skewed



(f) Outlier



# Examining a new data set

1. Examine each variable by itself.
2. Study the relationship between variables.

For both steps 1 and 2 we want to:

- Display the data graphically.
- Summarize the data numerically (Statistics).
- Construct a mathematical model.

# Describing distributions numerically

- For single variables, We are interested in summaries that provide information about the **center** and **spread** of the distribution.
- A **statistic** is a numerical summary of data.
- The two most common measures of center are the **mean** and **median**.
- “generous” vs. “selfish”.

# Mean

- If we have  $n$  ,observations, their **mean** is defined by,

$$\bar{x} = \frac{x_1 + x_2 + x_3 + \cdots + x_n}{n}$$

or

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$$

Ex. Calculate the mean of the data set: {1,2,3,4,5}.

$$\bar{x} = \frac{1 + 2 + 3 + 4 + 5}{5} = \frac{15}{5} = 3$$

Ex. Calculate the mean of the data set: {1,2,3,4,30}.

$$\bar{x} = \frac{1 + 2 + 3 + 4 + 30}{5} = \frac{40}{5} = 8$$



# Mean cont.

- The mean is **non-resistant**, meaning that it is influenced by very large or very small data points that are extreme values for the data set.



# Median

The **median**, written as  $M$ , is defined as the middle value of a data set.

1. List all  $n$  observations in order of size.
2. If  $n$  is odd, the median is the center value of the ordered list.
3. If  $n$  is even, the median is the average of the two center observations.

# Median Cont.

Ex. Calculate the median of {6,2,5,19,12,10}.

2 5 6 | 10 12 19

M is the average of 6 and 10, hence  $M=8$ .

Ex. Calculate the median of {1,2,3,4,5} and {1,2,3,4,30}.

1 2 3 | 4 5       $M=3$ .

1 2 3 | 4 30       $M=3$ .

# Median cont.

- The median is **resistant** (**robust**) to the extremes in the data set. Extremely large or small values do NOT influence the median.

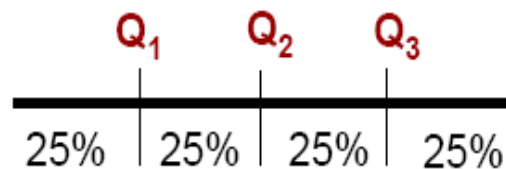


# Measures of variability

- Mean and median provide measures of **location** (**center**).
- One also needs some measures of **variability** to further describe the **spread** of the data set.
- Commonly used numerical values that can summarize the spread of a distribution.
  - Range
  - Interquartile Range (IQR)
  - Standard deviation

# Quartiles

- The median divides the data into two groups of equal size.
- The quartiles divide the data into four groups of equal size.



# Quartiles cont.

To find the quartiles:

1. Find the median.
2. Find the first quartile (Q1, or the *lower fourth*) by finding the median of the lower half of the data.
3. Find the third quartile (Q3, or the *upper fourth*) by finding the median of the upper half of the data.

*(When  $n$  is odd include the median in both halves in steps 2 and 3.)*

Ex. Find the quartiles for the data set {2,4,6,8,12,14,18,19,41}.

2 4 6 8 **12** 14 18 19 41

2 4 **6** 8 12

12 14 **18** 19 41

M = 12   Q1 = 6   Q3 = 18

# IQR

- The Interquartile Range, IQR, is the distance between the first and third quartiles,

$$\text{IQR} = Q3 - Q1.$$

- The IQR measures the spread of the middle 50% of the data.
- An observation is a suspected **outlier** if it falls more than  $1.5 \times \text{IQR}$  from the closest fourth. An outlier is **extreme** if it is more than  $3 \times \text{IQR}$  from the nearest fourth, and it is **mild** otherwise.

Ex. Can any of the observations in the data set  $\{2, 4, 6, 8, 12, 14, 18, 19, 41\}$  be considered outliers?

Recall we had  $M = 12$ ,  $Q1 = 6$ ,  $Q3 = 18$ . Therefore,  $\text{IQR} = 18 - 6 = 12$ .

$1.5 \times \text{IQR} = 1.5 \times 12 = 18$ .  $Q3 + 18 = 36$ ,  $Q1 - 18 = -12$ . Since  $41 > 36$ , 41 is classified as a potential outlier.



# 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)`

# 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  $x_1, x_2, x_3, \dots, x_n$ .
- 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

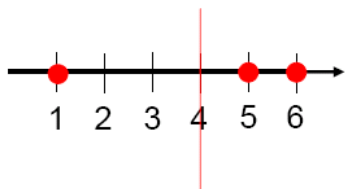
- If we have  $n$  observations  $x_1, x_2, x_3, \dots, x_n$ . The **variance** is defined as

$$s^2 = \frac{(x_1 - \bar{x})^2 + (x_2 - \bar{x})^2 + \dots + (x_n - \bar{x})^2}{n-1} = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

- The **standard deviation**,  $s$ , is the square root of the variance.
  1.  $s$  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

# Standard deviation

Ex. Let  $x_1 = 1, x_2 = 5, x_3 = 6$



$$\bar{x} = 4$$

Calculate the deviations:

$$(x_1 - \bar{x}) = -3$$

$$(x_2 - \bar{x}) = 1$$

$$(x_3 - \bar{x}) = 2$$

**Note that the  
deviations sum to 0.**

Calculate the squared deviations:

$$(x_1 - \bar{x})^2 = 9$$

$$(x_2 - \bar{x})^2 = 1$$

$$(x_3 - \bar{x})^2 = 4$$

Calculate the 'average' squared deviation:

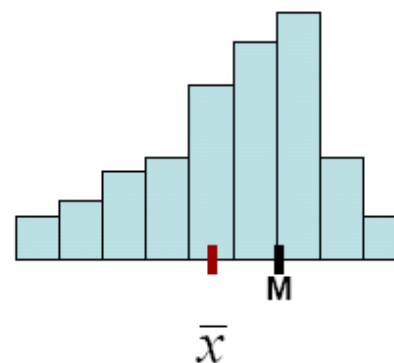
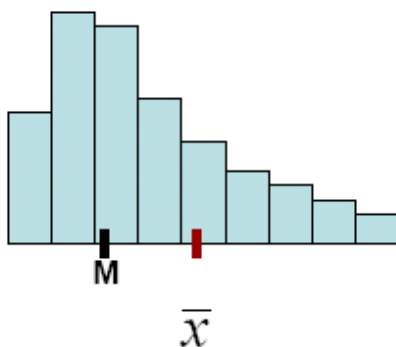
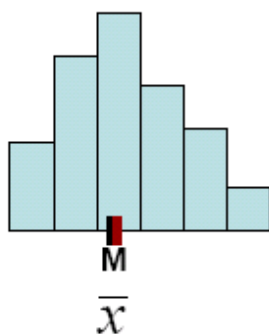
$$\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2 = \frac{1}{2} (9 + 1 + 4) = \frac{14}{2} = 7$$

# Degree of freedom

- As the sum of the deviations are always zero, the last deviation can be found once we know the other  $n-1$ .
- Only  $n-1$  of the squared deviations can vary freely, so we average by dividing the total by  $n-1$ .
- $n-1$  are the **degrees of freedom** of the variance and standard deviation.

# Measures of center and spread

- If the distribution is:
  1. symmetric, then  $\bar{x} = M$  and both are located exactly in the middle of the distribution.
  2. skewed right, then  $\bar{x} > M$ .
  3. skewed left, then  $\bar{x} < M$ .



- As a **rule of thumb**: if a data set is reasonably symmetric use the mean and standard deviation, if it is highly skewed use the five-number summary.

# 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.  $S=\{x: 0 \leq x < \infty\}$ .



# 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.  $B=\{HH\}$ .

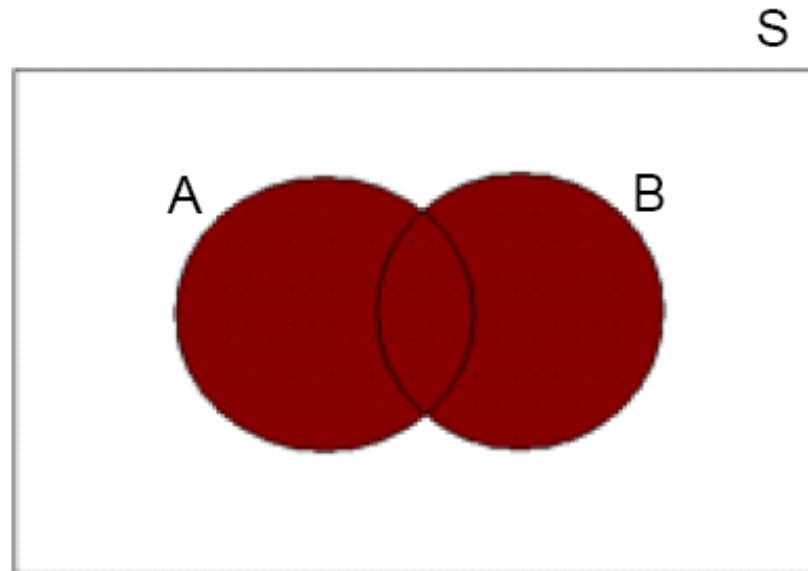
Ex. Battery life example. The event that the battery lasts less than 3 hours is denoted as  $A=\{x: 0 \leq x < 3\}$ .

# 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.

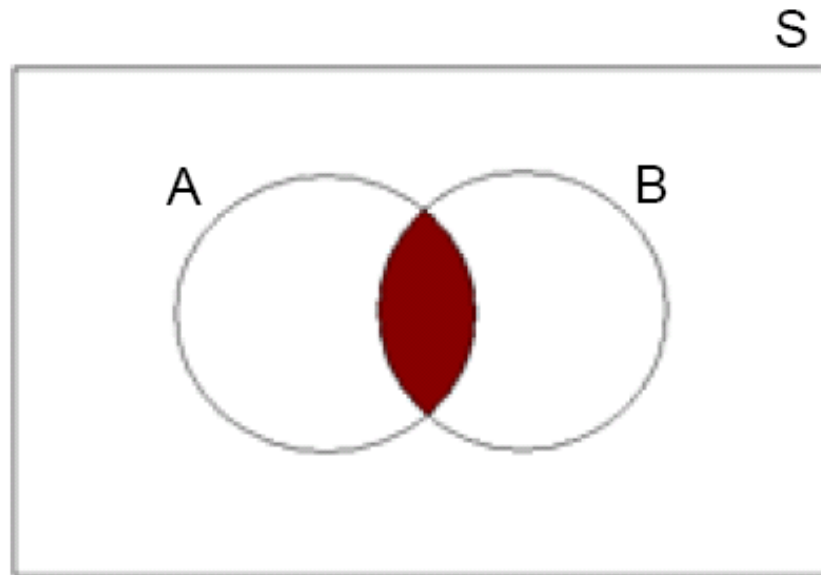
# Union

- The **union** of A and B, written as  $A \cup B$  and read “A or B”, is the set of outcomes that belong to either A or B or both.



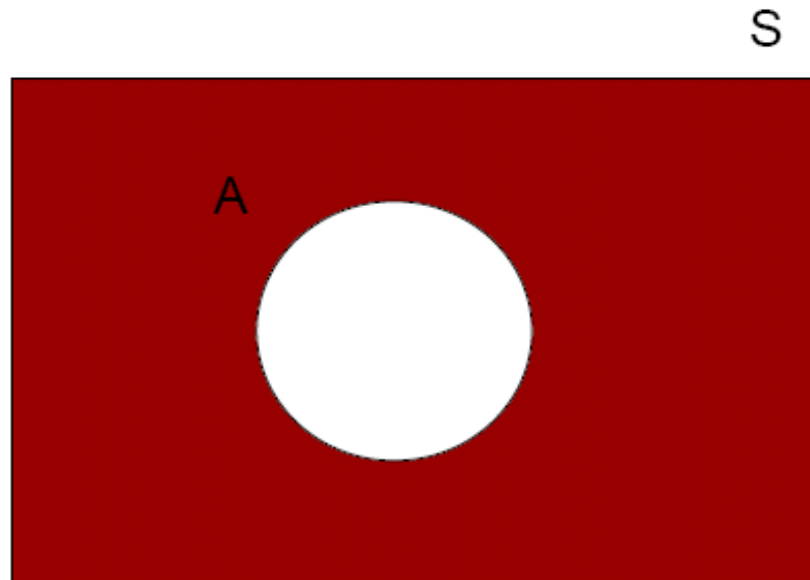
# Intersection

- The **intersection** of A and B, written as  $A \cap B$ , 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  $A^c$ , 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 (Cl), diamonds (D), hearts (H) or spades (Sp).

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

Let:  $A = \{\text{Cl}, D\}$ ,  $B = \{D, H, \text{Sp}\}$  and  $C = \{H\}$ .

$$A \cup B = \{\text{Cl}, D, H, \text{Sp}\} = S$$

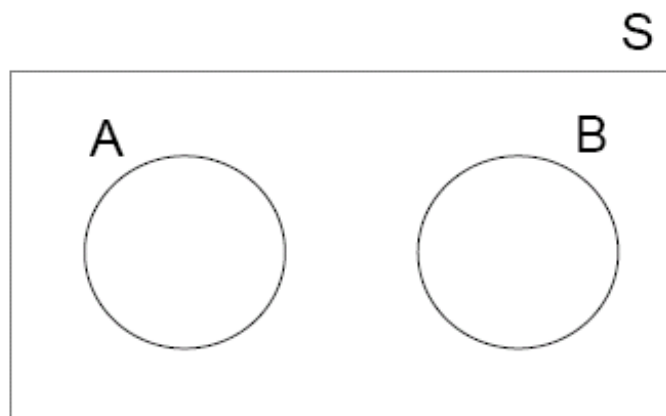
$$A \cap B = \{D\}$$

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

$$A \cap C = \emptyset \text{ (null event – event consisting of no outcomes)}$$

# Disjoint events

- If  $A \cap B = \emptyset$  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.
  1. For any event A,  $P(A) \geq 0$ .
  2.  $P(S) = 1$ .
  3. If  $A_1, A_2, A_3, \dots$  is an infinite (finite) collection of disjoint events, then
$$P(A_1 \cup A_2 \cup A_3 \cup \dots) = \sum P(A_i)$$

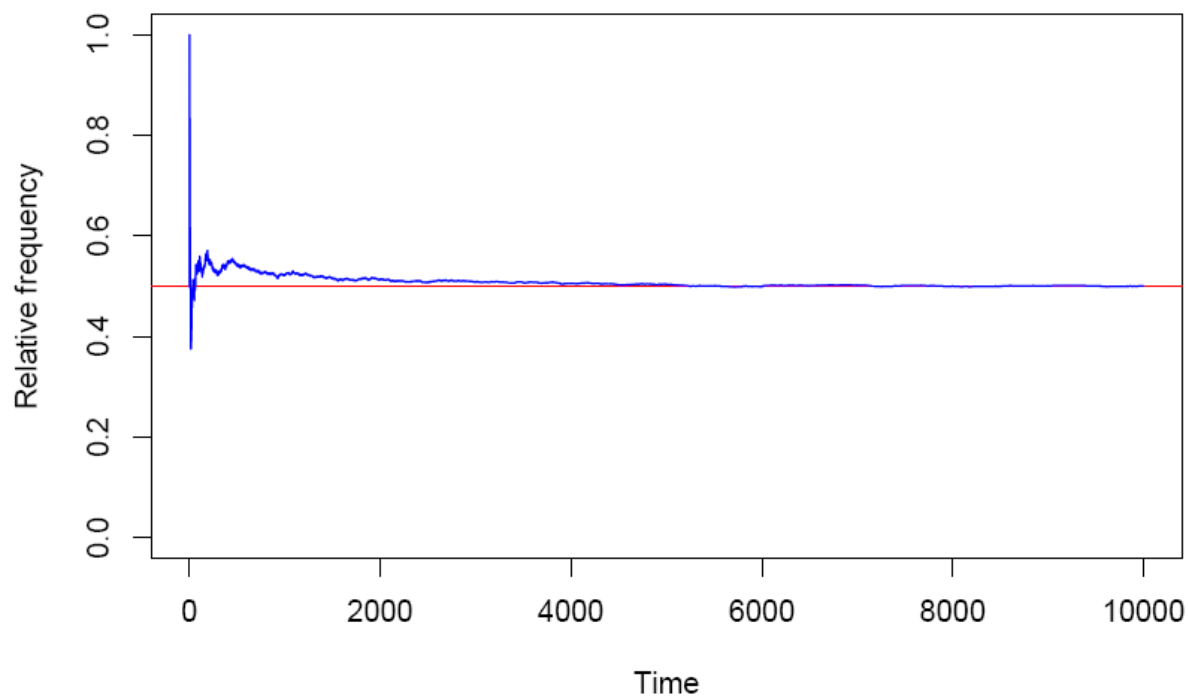


# Interpreting Probability

- 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.

$$\frac{\text{\# of occurrence of event A}}{\text{\# of trials } (n)} \xrightarrow{n \rightarrow \infty} P(A)$$

# 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.

$P(\text{roll a } 1) = 1/6$ .

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

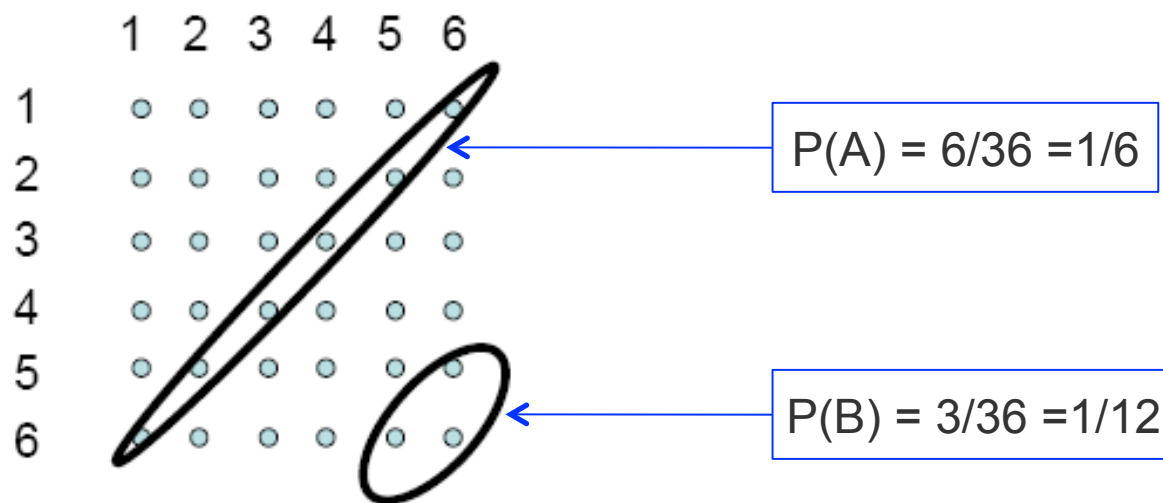
Let  $B = \text{roll an even number} = \{2,4,6\}$ .  $P(B) = 3/6$ .

# Example

Ex. Roll two fair dice.

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

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

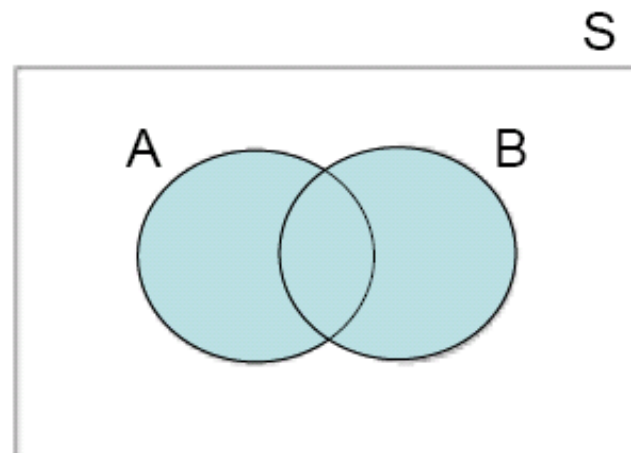
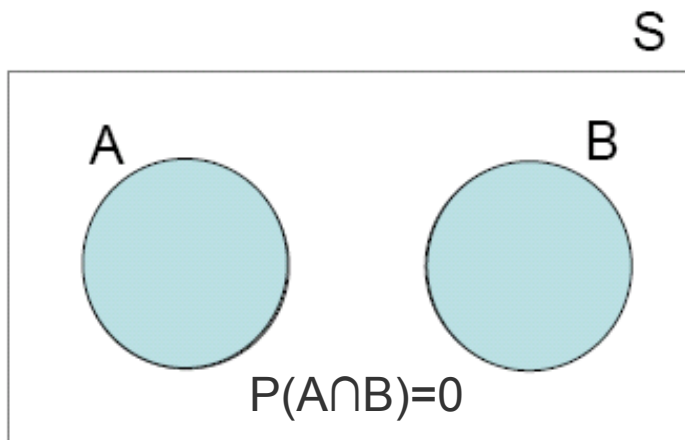


# Counting Techniques

- In the previous example, we used brute force to calculate the probability for event A and event B.
- In this class, quite often, we need better ways to count how many outcomes there are in a particular event.
- Permutation:  $P_{k,n} = n!/(n-k)!$
- Combination:  $C_{k,n} = P_{k,n}/k!$  Also denoted as  $\binom{n}{k}$ .

# 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:
  - $0 \leq P(A) \leq 1$ .
  - $P(S)=1$ .
  - $P(A^c)=1-P(A)$ .
  - If  $A$  and  $B$  are disjoint, then  $P(A \cup B)=P(A)+P(B)$ .
  - For any two events  $A$  and  $B$ ,  $P(A \cup B)=P(A)+P(B)-P(A \cap B)$ .



# 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

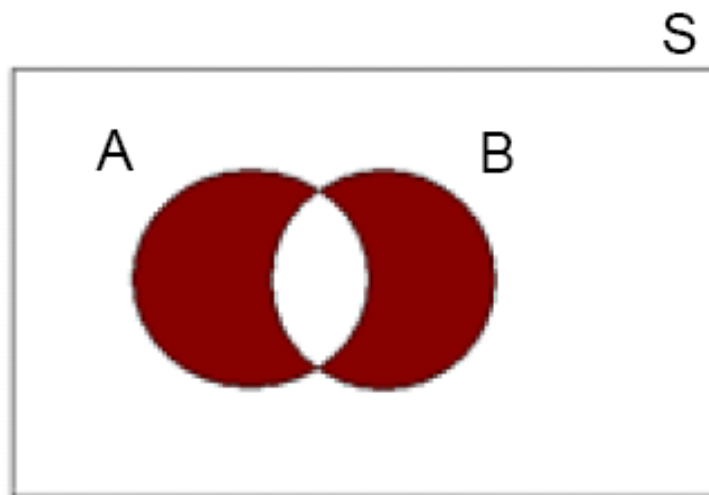
B = customers has Mastercard

$$\begin{aligned} P(A \cup B) &= P(A) + P(B) - P(A \cap B) \\ &= 0.5 + 0.3 - 0.1 = 0.7 \end{aligned}$$



## Example cont.

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



$$\begin{aligned} P(A \text{ or } B \text{ but not both}) &= P(A) + P(B) - 2P(A \cap B) \\ &= 0.5 + 0.3 - 0.2 = 0.6 \end{aligned}$$

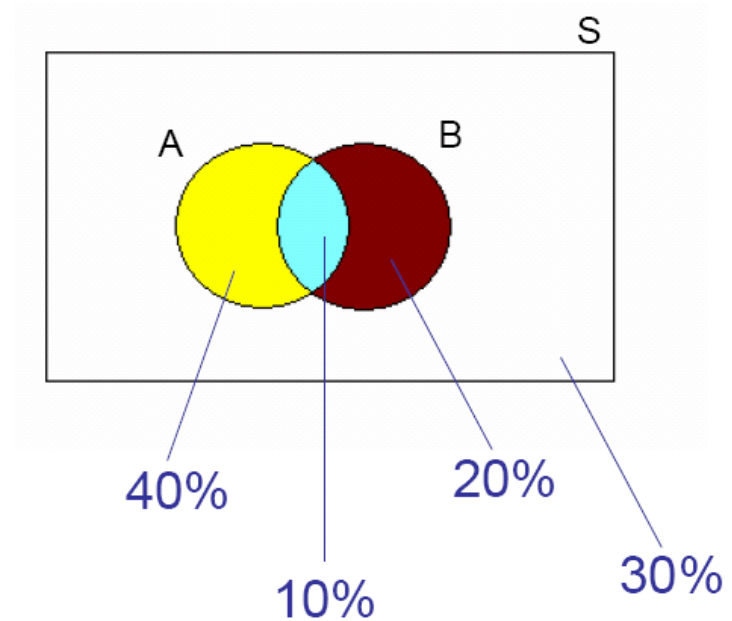
## Example Cont.

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

$$\begin{aligned}P(\text{A but not both}) &= P(A) - P(A \cap B) \\ &= 0.5 - 0.1 = 0.4\end{aligned}$$

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

$$\begin{aligned}P(\text{B but not both}) &= P(B) - P(A \cap B) \\ &= 0.3 - 0.1 = 0.2\end{aligned}$$



# Three Events

- For any three events A, B and C,

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

