

# Probability Theory

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1. Basic concepts
2. Probability
3. Permutations and combinations
4. In-class exercise

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# Probability Theory

## Basic Concepts

# Statistical Thinking

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- Experimental measurements can be viewed as variables that are subject to random variations
- This randomness cannot be predicted in a deterministic sense
- Such variables are known as random variables
- The chance of a random variable having a particular value is governed by probability theory
- Data can be thought of as random samples from an underlying, unknown probability distribution
- A goal of statistics is to extract information about the probability distribution from a usually small number of samples

# Mean and Variance

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- Data: multiple measurements of the same quantity

$$\{x_1 \quad x_2 \quad \cdots \quad x_{n-1} \quad x_n\}$$

- Definitions

» Range:  $R = x_{\max} - x_{\min}$

» Median: middle value when values are ordered according to their magnitudes

» Important properties:

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

$$\text{Variance} \quad s^2 = \frac{1}{n-1} \sum_{j=1}^n (x_j - \bar{x})^2 \quad \text{Standard deviation} \quad s = \sqrt{\frac{1}{n-1} \sum_{j=1}^n (x_j - \bar{x})^2}$$

» The mean represents the average value

» The variance is a measure of variability

# Mean and Variance Example

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- Data set 1

$$x = \{0.25 \quad 0.50 \quad 0.75 \quad 1 \quad 1.25 \quad 1.50 \quad 1.75\}$$

$$\bar{x} = \frac{1}{7} \sum_{j=1}^7 x_j = 1$$

$$s^2 = \frac{1}{7-1} \sum_{j=1}^7 (x_j - \bar{x})^2 = 0.29$$

- Data set 2

$$x = \{-2 \quad -1 \quad 0 \quad 1 \quad 2 \quad 3 \quad 4\}$$

$$\bar{x} = \frac{1}{7} \sum_{j=1}^7 x_j = 1$$

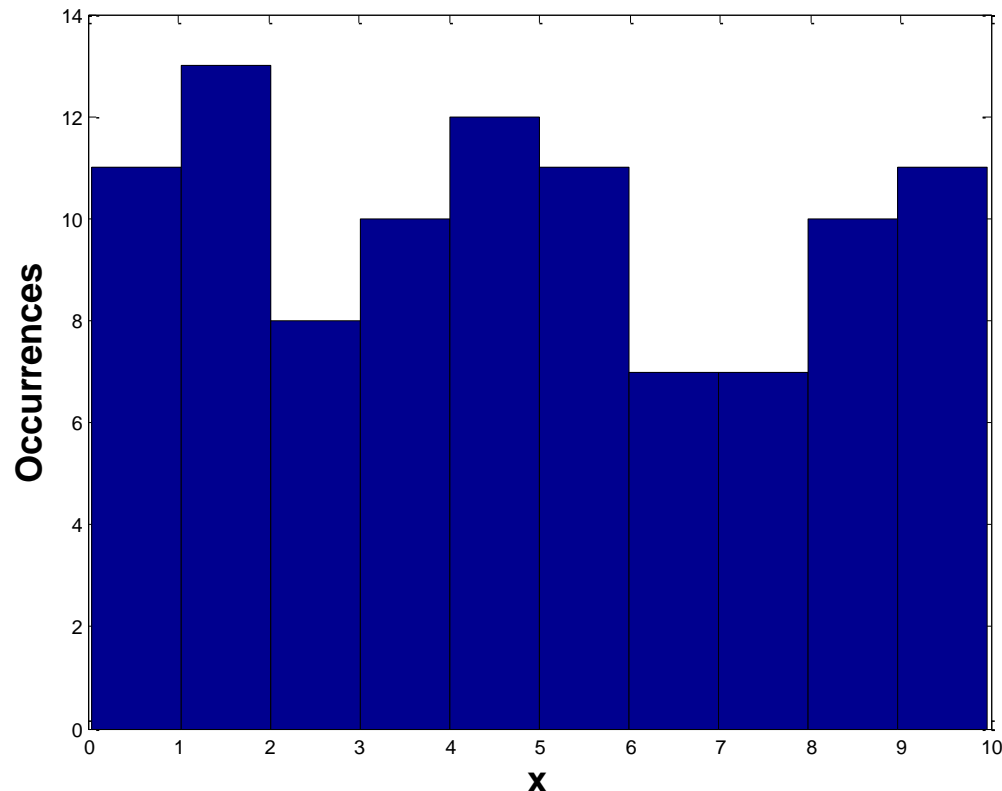
$$s^2 = \frac{1}{7-1} \sum_{j=1}^7 (x_j - \bar{x})^2 = 4.67$$

- These two data sets are very different despite having the same mean

# Histograms

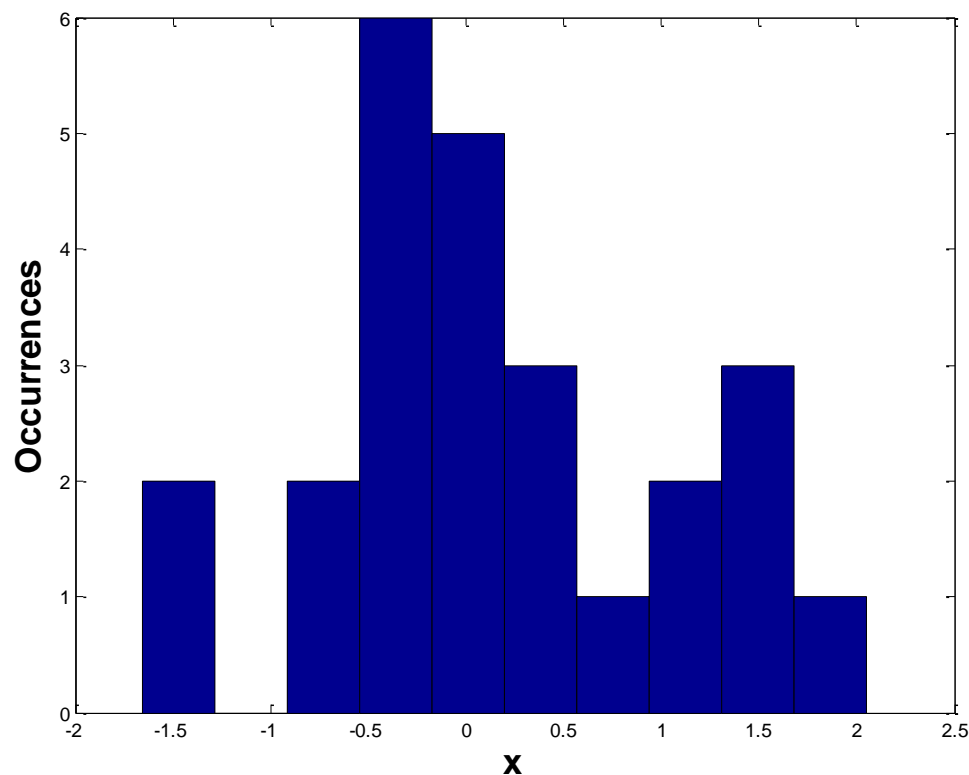
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- Large data sets are conveniently represented graphically in histograms
- Data are binned into intervals based on their values
- The number of occurrences is plotted verses the bin values to represent the variability



# Effect of Data Set Size

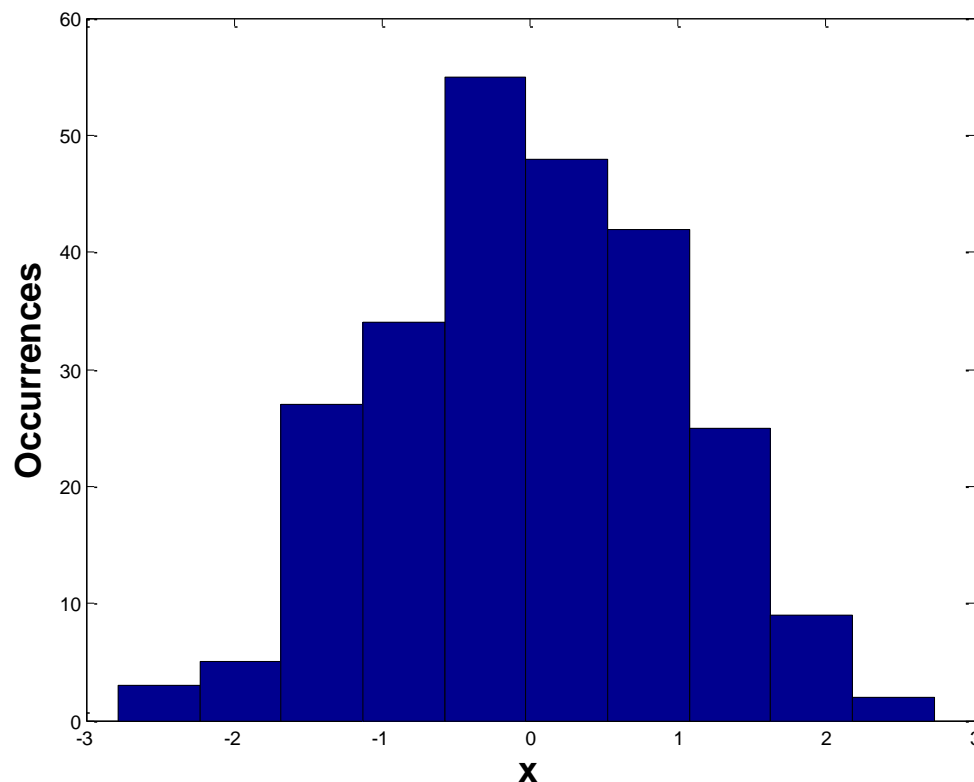
- Estimation of statistical properties is improved as the number of samples increases



25 samples

Mean = 0.16

Variance = 0.82



250 samples

Mean = -0.003

Variance = 0.92

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# Probability Theory

Probability



# Probability Theory

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- Experiment: process of measurement or observation
- Trial: performance of a single experiment
- Outcome: the result of a trial (also called a sample)
- Sample size  $n$ : the number of trials performed
- Sample space  $S$ : set of all possible outcomes
- Events: subsets of the sample space
- Key idea: each sample represents a value of the random variable

# Probability Theory Example

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- Experiment: measurement of polymer thin film thickness
- Trial: performance of a thickness measurement on a single thin film
- Outcome: the thickness measurement
- Sample size  $n$ : the number of thin films measured
- Sample space  $S = \{\text{too thin, acceptable, too thick}\}$
- Events: too thin, acceptable, too thick
- But cannot realistically measure the thickness of every thin film manufactured

# Events

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- Sample space divided into events ( $A_1, A_2, A_3, \dots$ )
- Union and intersection of events

Union  $A_1 \cup A_2 =$  all points in either set

Intersection  $A_1 \cap A_2 =$  all points in both sets

$$S = A_1 \cup A_2 \cup A_3 \cup \dots$$

- Disjoint and complement events

Disjoint events  $A_1 \cap A_2 = \emptyset$

Complement events  $A_1 \cup A_1^c = S$   $A_1 \cap A_1^c = \emptyset$

- Thin film example
  - » Three events: too thin, acceptable, too thick
  - » The union of the three events is the sample space
  - » The events have no intersection and are therefore disjoint
  - » Each event is the complement of the other two events

# Definition of Probability

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- Simple definition for finitely many equally likely outcomes:

$$P(A_j) = \frac{\text{number of points in } A_j}{\text{number of points in } S} \quad P(S) = 1$$

- Relative frequency for disjoint events:

$$f_{rel}(A_j) = \frac{\text{number of times } A_j \text{ occurs}}{n} \quad 0 \leq f_{rel}(A_j) \leq 1$$
$$f_{rel}(S) = 1 \quad f_{rel}(A_1 \cup A_2) = f_{rel}(A_1) + f_{rel}(A_2) \quad f_{rel}(A_1 \cap A_2) = 0$$

- $P(A_j)$  satisfies the following axioms of probability:

$$0 \leq P(A_j) \leq 1 \quad P(S) = 1$$

$$P(A_1 \cup A_2 \cup \cdots) = P(A_1) \cup P(A_2) \cup \cdots \quad P(A_1 \cap A_2 \cap \cdots) = P(A_1) \cap P(A_2) \cap \cdots = \emptyset$$

# Basic Theorems of Probability

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- Complementation:  $P(A_j^c) = 1 - P(A_j)$
- Addition rule for mutually exclusive events

$$P(A_1 \cup A_2 \cup \cdots \cup A_m) = P(A_1) + P(A_2) + \cdots + P(A_m)$$

- Addition rule for arbitrary events

$$P(A_1 \cup A_2) = P(A_1) + P(A_2) - P(A_1 \cap A_2)$$

- Conditional probability of  $A_2$  given  $A_1$

$$P(A_2 | A_1) = \frac{P(A_1 \cap A_2)}{P(A_1)}$$

$$P(A_1 \cap A_2) = P(A_1)P(A_2 | A_1) = P(A_2)P(A_1 | A_2)$$

- Independent events

$$P(A_1 \cap A_2) = P(A_1)P(A_2) \quad P(A_2 | A_1) = P(A_2) \quad P(A_1 | A_2) = P(A_1)$$

# Probability Examples

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- Probability that at least one coin will turn heads up from five tossed coins
  - » Number of outcomes:  $2^5 = 32$
  - » Probability of each outcome:  $1/32$
  - » Probability of no heads:  $P(A^C) = 1/32$
  - » Probability at least one head:  $P(A) = 1 - P(A^C) = 31/32$
- Probability of getting an odd number or a number less than 4 from a single dice toss
  - » Probability of odd number:  $P(A) = 3/6$
  - » Probability of number less than 4:  $P(B) = 3/6$
  - » Probability of both:  $P(A \cap B) = 2/6$
  - » Probability of either:

$$P(A \cup B) = P(A) + P(B) - P(A \cap B) = 3/6 + 3/6 - 2/6 = 2/3$$

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# Probability Theory

## Permutations and Combinations

# Permutations

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- Permutation – arrangement of objects in a particular order
- The number of permutations of  $n$  different objects taken all at a time is:  $n! = 1 \cdot 2 \cdot 3 \cdots n$
- The number of permutations of  $n$  objects divided into  $c$  different classes taken all at a time is:

$$\frac{n!}{n_1! n_2! \cdots n_c!} \quad n_1 + n_2 + \cdots + n_c = n$$

- Number of permutations of  $n$  different objects taken  $k$  at a time is:

Without repetitions  $n(n-1)(n-2)\cdots(n-k+1) = \frac{n!}{(n-k)!}$

With repetitions

$$n^k$$



# Permutation Examples

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- Box containing 6 red and 4 blue balls
  - » Compute probability that all red balls and then all blue balls will be removed
  - »  $n_1 = 6, n_2 = 4$
  - » Probability

$$P = \frac{1}{n! / n_1! n_2!} = \frac{6! 4!}{10!} \approx 0.005 = 0.5\%$$

- Coded telegram
  - » Letters arranged in five-letter words:  $n = 26, k = 5$
  - » Total number of different words:  $n^k = 26^5 = 11,881,376$
  - » Total number of different words containing each letter no more than once:

$$\frac{n!}{(n-k)!} = \frac{26!}{(26-5)!} = 7,893,600$$

# Combinations

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- Combination – selection of objects without regard to order
- Binomial coefficients

Real number  $a$  
$$\binom{a}{k} \equiv \frac{a(a-1)(a-2)\cdots(a-k+1)}{k!}$$

Integer  $0 \leq k \leq n$  
$$\binom{n}{k} \equiv \frac{n(n-1)(n-2)\cdots(n-k+1)}{k!} = \frac{n!}{k!(n-k)!}$$

- Number of combinations of  $n$  different objects taken  $k$  at a time is:

Without repetitions 
$$\binom{n}{k} = \frac{n!}{k!(n-k)!}$$

With repetitions 
$$\binom{n+k-1}{k} = \frac{(n+k-1)!}{k!(n-1)!}$$

# Combination Examples

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- Effect of repetitions

- » Three letters  $a, b, c$  taken two at a time ( $n = 3, k = 2$ )

- » Combinations without repetition

$$\binom{n}{k} = \binom{3}{2} = \frac{3!}{2!(3-2)!} = 3 \quad ab \quad ac \quad bc$$

- » Combinations with repetitions

$$\binom{n+k-1}{k} = \binom{4}{2} = \frac{4!}{2!(4-2)!} = 6 \quad \begin{matrix} ab & ac & bc \\ aa & bb & cc \end{matrix}$$

- 500 light bulbs taken 5 at a time

- » Repetitions not possible

- » Combinations  $\binom{n}{k} = \binom{500}{5} = \frac{500!}{5!(500-5)!} = 255,244,686,600$

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# Probability Theory

In-class Exercise