

Probability Theory

1. Basic concepts
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Probability Theory

Basic Concepts

Statistical Thinking

- 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

- 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

- 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

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

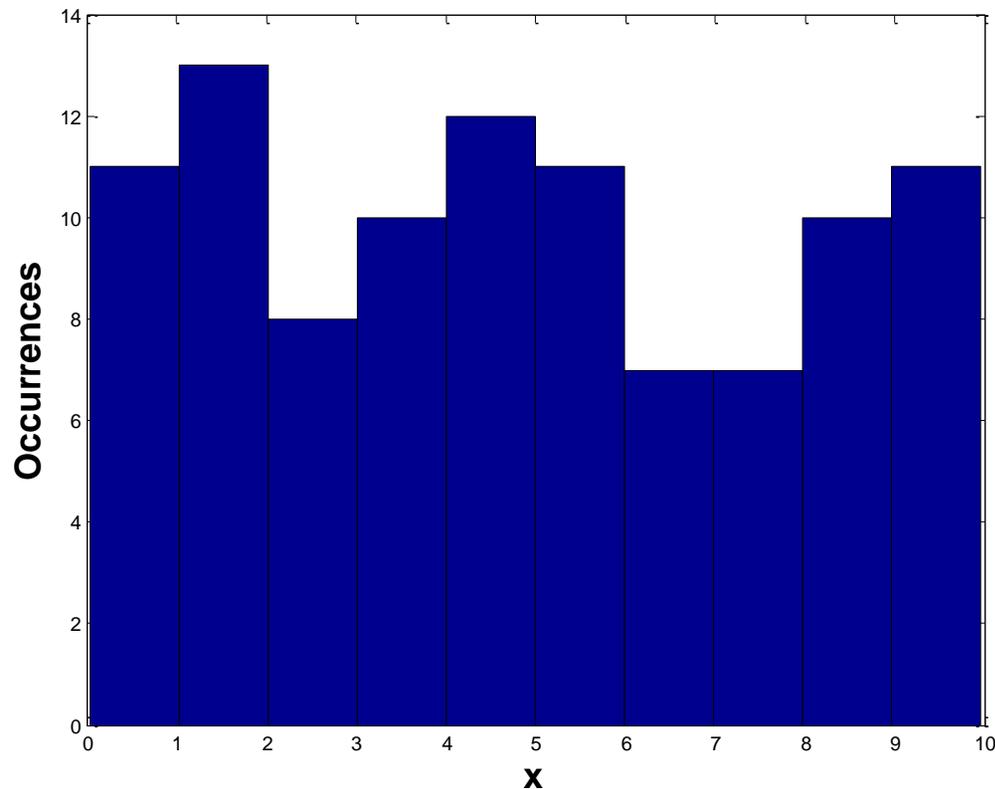
$$\bar{x} = \frac{1}{7} \sum_{j=1}^7 x_j = 1$$

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

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

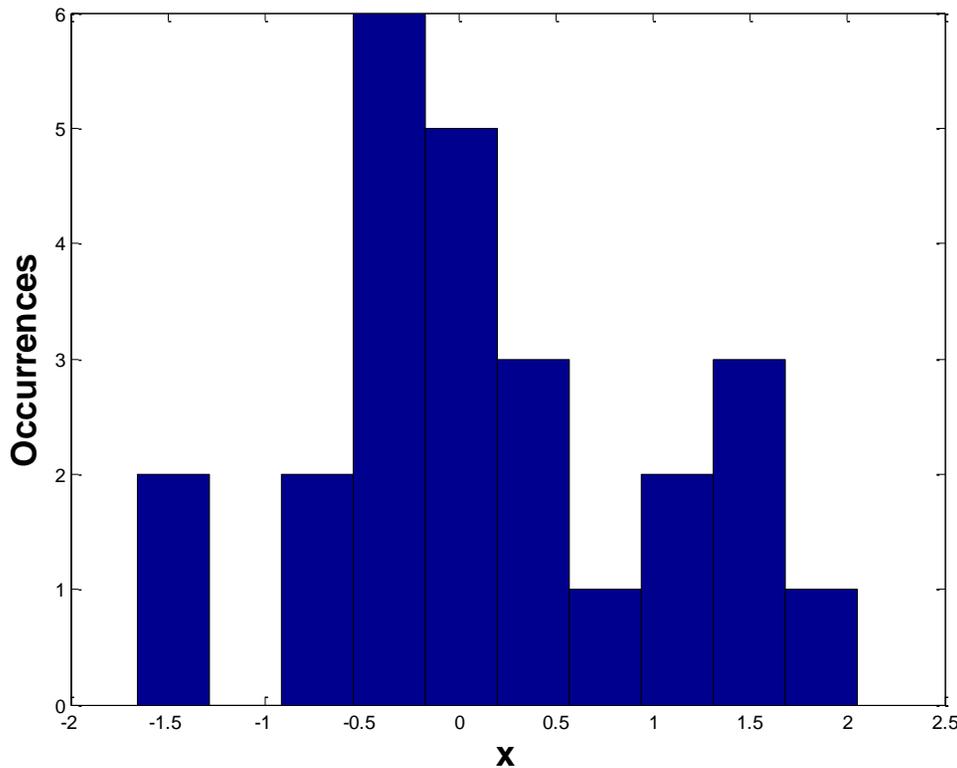
Histograms

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



Effect of Data Set Size

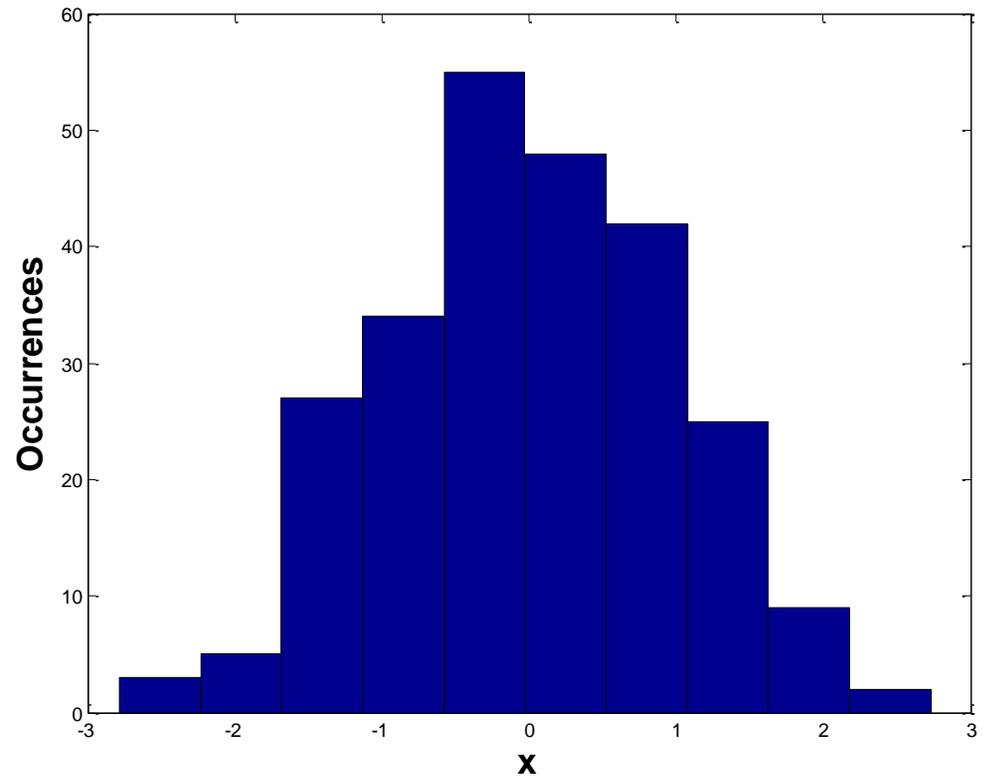
- Estimation of statistical properties is improved as the number of samples increases



25 samples

Mean = 0.16

Variance = 0.82



250 samples

Mean = -0.003

Variance = 0.92

Probability Theory

Probability

Probability Theory

- 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

- 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

- 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

- 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:

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

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

Basic Theorems of Probability

- Complementation: $P(A_j^c) = 1 - P(A_j)$

- Addition rule for mutually exclusive events

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

- 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

- 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:

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

Probability Theory

Permutations and Combinations

Permutations

- 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

- 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

- 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

- 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

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

- 500 light bulbs taken 5 at a time

- » Repetitions not possible

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

Probability Theory

In-class Exercise