

Stat 13, Thu 4/19/12.

0. Hand in HW2!

1. Resistance.

2. $n-1$ in sample sd formula, and parameters and statistics.

3. Probability basic terminology.

4. Probability axioms.

5. Multiplication rules.

6. Probability trees.

7. Combinations.

Read chapter 3.

Hw3 is due Thur, 4/26, and Midterm 1 is Thur, 4/26.

You can use calculators, pen or pencil, and one 8.5x11 page of notes, double sided, for the exam.

1. Resistance to outliers.

Sample mean, sd, variance, range, and CV → sensitive to outliers.

Sample median and IQR → resistant to outliers.

{1,2,1,3,5,0}. mean is 2, median is 1.5, sd ~ 1.79, IQR = 2.

{1,2,1,3,5,0, 1000}. mean ~ 145, median is 2, sd ~ 377, IQR = 3.

2. n-1, parameters, and statistics.

In the formula for the sample sd, $s = \sqrt{\sum [(x_i - \bar{x})^2 / (n-1)]}$

If we replace the n-1 by n, then s is the *RMS* (root-mean-square) of deviations from \bar{x}

In other words, s is the RMS of these deviations times a correction factor $\sqrt{\frac{n}{n-1}}$

Why this correction factor?

Parameters are properties of the population. Typically unknown. Represented by Greek letters (like μ , or σ).

Statistics (also called *random variables* or *estimates*) are properties of the sample. Represented by Roman letters (like \bar{x} or s).

Typically, you're interested in a value of a parameter. But you can't know it. So you *estimate* it with a statistic, based on the sample.

There are two means and two standard deviations!

The sample mean \bar{x} and sample std deviation s are statistics.

Define the population average μ as the sum of all values in the population \div the number of subjects in the population. (parameter).

It turns out \bar{x} is an unbiased estimate of μ .

That is, \bar{x} is neither higher nor lower, on average, than μ .

Define the population std deviation σ as the RMS of the deviations in the whole population. (a parameter.)

It turns out that if you estimate σ with the RMS of the sample deviations, the estimate you get tends to be a little SMALLER on average than σ .

We multiply by the correction factor $\sqrt{\frac{n}{n-1}}$, so that the estimate s is unbiased.

Note that this correction factor is very nearly 1 for large n , so it doesn't matter much when the sample size n is large.

3. Basic terminology of probability.

- a) Notation. $P(\text{event}) = \#$. $P(X = 5)$ means *the probability that X is 5*.
- b) Conditional probability. $P(A | B)$ means the probability of A GIVEN B.
- c) Disjoint. Events A and B are *disjoint* if $P(A \text{ and } B) = 0$.
- d) Independent. A and B are independent if knowing A doesn't effect the probability that B will happen, and vice versa.
That is, if $P(A | B) = P(A)$, and $P(B | A) = P(B)$.
e.g. A = event first die roll is a 5, B = event second die is a 5.
- e) Or. In math, A OR B always means one or the other or both! If you mean but not both, must say ``but not both”.
- f) E^c means not E.
- g) $P(AB)$ means $P(A \text{ and } B)$.

4. Probability axioms.

- (i) For any event E, $P(E) \geq 0$.
- (ii) For any event E, $P(E) + P(E^c) = 1$.
- (iii) For any disjoint events E_1, E_2, \dots, E_n ,
$$P(E_1 \text{ or } E_2 \text{ or } \dots \text{ or } E_n) = P(E_1) + P(E_2) + \dots P(E_n).$$

(iii) is sometimes called the addition rule for disjoint events.

Note connection with Venn diagrams.

For *any* events E_1 and E_2 , $P(E_1 \text{ or } E_2) = P(E_1) + P(E_2) - P(E_1 E_2)$.

5. Multiplication rules.

We define $P(A | B)$ as $P(AB) / P(B)$.

Independence means $P(A | B) = P(A)$, which means $P(AB)/P(B) = P(A)$, i.e.

$$P(AB) = P(A) P(B).$$

If this true, then $P(B | A) = P(AB)/P(A) = P(A)P(B)/P(A) = P(B)$, as well.

This is sometimes called the multiplication rule, because it means if A and B are independent, then $P(AB) = P(A) \times P(B)$.

Outcomes of different die rolls, spins of a spinner, flips of a coin, etc. can be assumed independent. Similarly, observations sampled *with replacement* are independent, or without replacement from large population are nearly independent.

e.g. roll 2 dice. $P(\text{add up to } 12) = P(1^{\text{st}} \text{ is } 6 \text{ and } 2^{\text{nd}} \text{ is } 6) = P(1^{\text{st}} \text{ is } 6)P(2^{\text{nd}} \text{ is } 6) = 1/6 \times 1/6$.

$P(\text{at least one } 6) = 1 - P(1^{\text{st}} \text{ isn't } 6 \text{ and } 2^{\text{nd}} \text{ isn't } 6) = 1 - (5/6 \times 5/6) = 1 - 25/36 = 11/36$.

In general, if A and B might not be independent, $P(AB) = P(A) \times P(B | A)$.

because the right side = $P(A) P(AB)/P(A) = P(AB)$.

e.g. pick a card. A = the event that it's a king. B = the event that it's a spade.

$$P(AB) = P(\text{it's the king of spades}) = 1/52.$$

$P(A) = 4/52$, $P(B) = 1/4$, so here $P(AB) = P(A)P(B)$, so A and B are independent.

6. Probability trees.

Sometimes you can use probability trees for multiplication rule problems.

e.g. deal two cards. $P(\text{two red kings}) = P(\text{K of hearts} \rightarrow \text{K of diamonds or K of diamonds} \rightarrow \text{K of hearts})$
 $= P(\text{K of hearts} \rightarrow \text{K of diamonds}) + P(\text{K of diamonds} \rightarrow \text{K of hearts})$
 $= 1/52 \times 1/51 + 1/52 \times 1/51.$

Suppose 1% of the population has a disease, and a test is 80% accurate at detection, i.e. $P(+ | \text{you have it}) = 80\%$, and $P(- | \text{you don't have it}) = 80\%$. Choose someone at random. What is $P(\text{test } +)$?

$$\begin{aligned} P(+) &= P(\text{has it and test } +) + P(\text{doesn't have it and test } +) \\ &= P(\text{has it}) P(\text{test } + | \text{has it}) + P(\text{doesn't}) P(+ | \text{doesn't}) \\ &= 1\% (80\%) + 99\% (20\%) \\ &= 20.6\%. \end{aligned}$$

What is $P(\text{has it} | \text{test } +)$?

$$\begin{aligned} P(\text{has it} | +) &= P(\text{has it and } +) / P(+) \\ &= 1\% (80\%) / 20.6\% \\ &\sim 38.8\%. \end{aligned}$$

7. Combinations.

The number of ways of choosing k distinct objects from a group of n different objects, where the order doesn't matter, is $C(n,k) = n!/[k! (n-k)!]$, with the convention that $0! = 1$. Your book writes this ${}_nC_k$.
 $k!$ means $1 \times 2 \times \dots \times k$.

These are called combinations.

For instance, pick 2 cards from a deck. $P(\text{Ace and King, in either order})$?

There are $n = 52$ cards in the deck.

Thus there are $C(52,2) = 52! / (2! 50!) = 1326$ different possible combinations. If each combination is equally likely, then each combination has a probability of $1/1326$.

How many combinations are AK? 4 aces, and for each choice of ace, there are 4 kings to go with it, so there are $4 \times 4 = 16$ such combinations. Each has a probability of $1/1326$, so $P(AK) = 16/1326$.

8. Expected value.

The *expected value* or *mean* of a random variable is the long-term average, if we observe it over and over.

e.g. Mean of a die roll. $1/6 (1) + 1/6 (2) + \dots + 1/6 (6) = 3.5$.