

Statistics for Computing

Lecture 1B

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(Brought over from last lecture)

- ① Contingency Tables
- ② Conditional Probability: Worked Examples
- ③ Joint Probability Tables
- ④ The Multiplication Rule
- ⑤ Law of Total Probability
- ⑥ Bayes' Theorem
- ⑦ Exam standard Probability Question
- ⑧ Sampling (Samples and Populations)
- ⑨ Sampling with and without Replacement

(Later in Class : A look at Descriptive Statistic)

Multiplication Rule

The multiplication rule is a result used to determine the probability that two events, A and B , both occur. The multiplication rule follows from the definition of conditional probability.

The result is often written as follows, using set notation:

$$P(A|B) \times P(B) = P(B|A) \times P(A) \quad (= P(A \cap B))$$

Recall that for independent events, that is events which have no influence on one another, the rule simplifies to:

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

Multiplication Rule

From the first year intake example, check that

$$P(E|F) \times P(F) = P(F|E) \times P(E)$$

- $P(E|F) \times P(F) = 0.58 \times 0.38 = 0.22$
- $P(F|E) \times P(E) = 0.55 \times 0.40 = 0.22$

Law of Total Probability

The law of total probability is a fundamental rule relating marginal probabilities to conditional probabilities. The result is often written as follows, using set notation:

$$P(A) = P(A \cap B) + P(A \cap B^c)$$

where $P(A \cap B^c)$ is probability that event A occurs and B does not.

Using the multiplication rule, this can be expressed as

$$P(A) = P(A|B) \times P(B) + P(A|B^c) \times P(B^c)$$

Law of Total Probability

From the first year intake example , check that

$$P(E) = P(E \cap M) + P(E \cap F)$$

with $P(E) = 0.40$, $P(E \cap M) = 0.18$ and $P(E \cap F) = 0.22$

$$0.40 = 0.18 + 0.22$$

Remark: M and F are complement events.

Bayes' Theorem

Bayes' Theorem is a result that allows new information to be used to update the conditional probability of an event.

Recall the definition of conditional probability:

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

Using the multiplication rule, gives Bayes' Theorem in its simplest form:

$$P(A|B) = \frac{P(B|A) \times P(A)}{P(B)}$$

Probability: Worked Example

An electronics assembly subcontractor receives resistors from two suppliers: Deltatech provides 70% of the subcontractors's resistors while another company, Echelon, supplies the remainder.

1% of the resistors provided by Deltatech fail the quality control test, while 2% of the resistors from Echelon also fail the quality control test.

- 1 What is the probability that a resistor will fail the quality control test?
- 2 What is the probability that a resistor that fails the quality control test was supplied by Echelon?

Probability: Worked Example

Firstly, let's assign names to each event.

- D : a randomly chosen resistor comes from Deltatech.
- E : a randomly chosen resistor comes from Echelon.
- F : a randomly chosen resistor fails the quality control test.
- P : a randomly chosen resistor passes the quality control test.

We are given (or can deduce) the following probabilities:

- $P(D) = 0.70$,
- $P(E) = 0.30$.

Probability: Worked Example

We are given two more important pieces of information:

- The probability that a randomly chosen resistor fails the quality control test, given that it comes from Deltatech: $P(F|D) = 0.01$.
- The probability that a randomly chosen resistor fails the quality control test, given that it comes from Echelon: $P(F|E) = 0.02$.

Probability: Worked Example

The first question asks us to compute the probability that a randomly chosen resistor fails the quality control test. i.e. $P(F)$.

All resistors come from either Deltatech or Echelon. So, using the *law of total probability*, we can express $P(F)$ as follows:

$$P(F) = P(F \cap D) + P(F \cap E)$$

Probability: Worked Example

Using the *multiplication rule* i.e. $P(A \cap B) = P(A|B) \times P(B)$, we can re-express the formula as follows

$$P(F) = P(F|D) \times P(D) + P(F|E) \times P(E)$$

We have all the necessary probabilities to solve this.

$$P(F) = 0.01 \times 0.70 + 0.02 \times 0.30 = 0.007 + 0.006 = 0.013$$

Probability: Worked Example

- The second question asks us to compute probability that a resistor that fails the quality control test was supplied by Echelon.
- In other words; of the resistors that did fail the quality test only, what is the probability that a randomly selected resistor was supplied by Echelon?
- We can express this mathematically as $P(E|F)$.
- We can use *Bayes' theorem* to compute the answer.

Probability: Worked Example

Recall Bayes' theorem

$$P(A|B) = \frac{P(B|A) \times P(A)}{P(B)}$$

$$P(E|F) = \frac{P(F|E) \times P(E)}{P(F)} = \frac{0.02 \times 0.30}{0.013} = 0.46$$

Sampling

The major use of statistics is to use information from a *sample* to infer something about a *population*.

- A *population* is a collection of data whose properties are analyzed. The population is the complete collection to be studied, it contains all subjects of interest.
- A *sample* is a part of the population of interest, a sub-collection selected from a population.
- A *parameter* is a numerical measurement that describes a characteristic of a population, while a *sample statistic* is a numerical measurement that describes a characteristic of a sample.
- In general, we will use a statistic to infer something about a parameter.

Sampling without replacement

- Sampling is said to be “without replacement” when a unit is selected at random from the population and it is not returned to the main lot.
- The first unit is selected out of a population of size N and the second unit is selected out of the remaining population of $N - 1$ units and so on.
- For example, if you draw one card out of a deck of 52, there are only 51 cards left to draw from if you are selecting a second card.

Sampling without replacement

A lot of 100 semiconductor chips contains 20 that are defective. Two chips are selected at random, without replacement from the lot.

- What is the probability that the first one is defective?
(Answer : $20/100$, i.e 0.20)
- What is the probability that the second one is defective given that the first one was defective?
(Answer: $19/99$)
- What is the probability that the second one is defective given that the first one was not defective?
(Answer: $20/99$)

Sampling With Replacement

Sampling is called “with replacement” when a unit selected at random from the population is returned to the population and then a second element is selected at random. Whenever a unit is selected, the population contains all the same units.

- What is the probability of guessing a PIN number for an ATM card at the first attempt.
- Importantly a digit can be used twice, or more, in PIN codes.
- For example 1337 is a valid pin number, where 3 appears twice.
- We have a one-in-ten chance of picking the first digit correctly, a one-in-ten chance of the guessing the second, and so on.
- All of these events are independent, so the probability of guess the correct PIN is $0.1 \times 0.1 \times 0.1 \times 0.1 = 0.0001$

Descriptive Statistics

We will digress from Probability for a while, and look at **Descriptive Statistics**.

- Sample Mean
- Sample Median
- Measures of dispersion

Descriptive Statistics

- Measures of Centrality
 - Mean
 - Median
- Measures of Dispersion
 - Range
 - Variance
 - Standard Deviation

Measures of Centrality

- Measures of centrality give one representative number for the location of the centre of the distribution of data.
- The most common measures are the *mean* and the *median* .
- We must make a distinction between a sample mean and a population mean: The sample mean is simply the average of all the items in a sample.
- The population mean (often represented by the Greek letter μ) is simply the average of all the items in a population.
- Because a population is usually very large, the population mean is usually an unknown constant.
- We will return to the matter of population means in due course. For now, we will look at sample means.

Sample Mean

- The sample mean is an estimator available for estimating the population mean . It is a measure of location, commonly called the average, often denoted \bar{x} , where x is the data set.
- Its value depends equally on all of the data which may include outliers. It may not appear representative of the central region for skewed data sets.
- It is especially useful as being representative of the whole sample for use in subsequent calculations.
- The sample mean of a data set is defined as :

$$\bar{x} = \frac{\sum x_i}{n}$$

- $\sum x_i$ is the summation of all the elements of x , and n is the sample size.

Computing the sample mean

Suppose we roll a die 8 times and get the following scores:

$$x = \{5, 2, 1, 6, 3, 5, 3, 1\}$$

What is the sample mean of the scores \bar{x} ?

$$\bar{x} = \frac{5 + 2 + 1 + 6 + 3 + 5 + 3 + 1}{8} = \frac{26}{8} = 3.25$$

Using R to compute mean (and median)

When implementing this in R, we would use the following code

```
> # create the "vector" x with the required values
> x=c(5, 2, 1, 6, 3, 5, 3, 1)
>
> mean(x)
[1] 3.25
>
> # See next slides first.
> sort(x)
[1] 1 1 2 3 3 5 5 6
> median(x)
[1] 3
```


Median

- The other commonly used measure of centrality is the median.
- The median is the value halfway through the ordered data set, below and above which there lies an equal number of data values.
- For an odd sized data set, the median is the middle element of the **ordered** data set.
- For an even sized data set, the median is the average of the middle pair of elements of an **ordered** data set.
- It is generally a good descriptive measure of the location which works well for *skewed data*, or data with *outliers*.
- For later, the median is the 0.5 quantile, and the second quartile Q_2 .

Computing the median

Example:

With an odd number of data values, for example nine, we have:

- Data : $\{96, 48, 27, 72, 39, 70, 7, 68, 99\}$
- Ordered Data : $\{7, 27, 39, 48, 68, 70, 72, 96, 99\}$
- Median : 68, leaving four values below and four values above

With an even number of data values, for example 8, we have:

- Data : $\{96, 48, 27, 72, 39, 70, 7, 68\}$
- Ordered Data : $\{7, 27, 39, 48, 68, 70, 72, 96\}$
- Median : Halfway between the two 'middle' data points - in this case halfway between 48 and 68, and so the median is 58

Using R to compute mean (and median)

When implementing this in R, we would use the following code

```
> x1=c(96, 48, 27, 72, 39, 70, 7, 68, 99 )
> sort(x1)
[1] 7 27 39 48 68 70 72 96 99
> median(x1)
[1] 68
>
> x2=c(96, 48 ,27 ,72, 39, 70, 7, 68)
> sort(x2)
[1] 7 27 39 48 68 70 72 96
> median(x2)
[1] 58
```

Dispersion

- The data values in a sample are not all the same. This variation between values is called *dispersion*.
- When the dispersion is large, the values are widely scattered; when it is small they are tightly clustered.
- There are several measures of dispersion, the most common being the variance and standard deviation. These measures indicate to what degree the individual observations of a data set are dispersed or 'spread out' around their mean.
- In engineering and science, high precision is associated with low dispersion.

Range

- The range of a sample (or a data set) is a measure of the spread or the dispersion of the observations.
- It is the difference between the largest and the smallest observed value of some quantitative characteristic and is very easy to calculate.
- A great deal of information is ignored when computing the range since only the largest and the smallest data values are considered; the remaining data are ignored.
- The range value of a data set is greatly influenced by the presence of just one unusually large or small value in the sample (outlier).

Example

The range of $\{65, 73, 89, 56, 73, 52, 47\}$ is $89 - 47 = 42$.

Introducing Variance

Consider the three data sets X , Y and Z

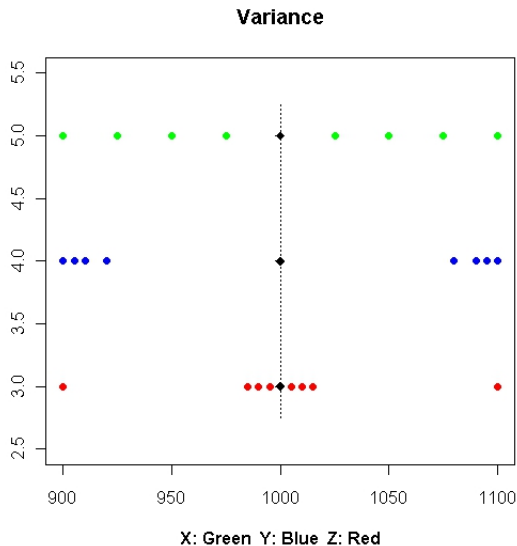
- $X = \{900, 925, 950, 975, 1025, 1050, 1075, 1100\}$
- $Y = \{900, 905, 910, 920, 1080, 1090, 1095, 1100\}$
- $Z = \{900, 985, 990, 995, 1005, 1010, 1015, 1100\}$

For each of the data sets, the following statements can be verified

- The mean of each data set is 1000
- There are 8 elements in each data set
- The minima and maxima are 900 and 1100 for each set
- The range is 200.

From the plot on the next slide, notice how different the three data sets are in terms of dispersion around the mean value.

Introducing Variance



Variance

- The (population) variance of a random variable is a non-negative number which gives an idea of how widely spread the values are likely to be; the larger the variance, the more scattered the observations on average.
- Stating the variance gives an impression of how closely concentrated round the expected value the distribution is; it is a measure of the 'spread' of a distribution about its average value.
- We distinguish between population variance (denoted σ^2) and sample variance (denoted s^2). For now, we will look only at sample variance.

Sample Variance

- Sample variance is a measure of the spread of or dispersion within a set of sample data.
- The sample variance is the sum of the squared deviations from their mean divided by one less than the number of observations in the data set.
- For example, for n observations $x_1, x_2, x_3, \dots, x_n$ with sample mean \bar{x} , the sample variance is given by

$$s^2 = \frac{\sum (x - \bar{x})^2}{n - 1}$$

Sample Standard Deviation

- Standard deviation is the square root of variance
- Standard deviation is commonly used in preference to variance because it is denominated in the same units as the mean.
- For example, if dealing with time units, we could have a variance of something like 25 *square minutes* , whereas the equivalent standard deviation is 5 minutes.
- Population standard deviation is denoted σ .
- Sample standard deviation is denoted s .

Using R

Using R to compute standard deviation and variance for these data sets.

```
> X=c(900,925,950,975,1025,1050,1075,1100)
> Y=c(900,905,910,920,1080,1090,1095,1100)
> Z=c(900,985,990,995,1005,1010,1015,1100)
>
> sd(X);sd(Y);sd(Z)
[1] 73.19251
[1] 97.87018
[1] 54.37962
>
>var(X);var(Y);var(Z)
[1] 5357.143
[1] 9578.571
[1] 2957.143
```