



ACADGILD

Mastering Data  
Science



# Statistics



## Session 15 - Introduction to Statistics



# Agenda

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1 Statistics

2 Introduction to Basic Terms

3 Variables

4 Mean, Median and Mode

5 Measure of Dispersion

6 Range

7 Sample Variance

8 Standard Deviation

9 Population Vs. Sample

10 Chebysheff's Theorem

11 Law of Expected Values and Variance

12 Probability Density Function

- Statistics is the science of collecting, organizing, presenting, analyzing, and interpreting data to help in making more effective decisions.
- Statistical Analysis is implemented to manipulate, summarize and investigate data, so that useful decision-making information results are obtained.

- Descriptive Statistics is a method of organizing, summarizing, and presenting data in an informative way.
- Inferential Statistics is a method which is used in determining something about a population on the basis of a sample.
  - Population - The entire set of individuals or objects of interest or the measurements obtained from all individuals or objects of interest.
  - Sample - A portion, or part, of the population of interest.

- Population - A collection/set of individuals/objects/events whose properties are to be analyzed. There are two kinds:
  - Finite
  - Infinite
- Sample - A population subset.

- Variable - A **characteristic** about each individual element of a population/sample.
- Data (singular) - A **value** of the associated variable with one element of a population/sample. This value may be a number, a word, or a symbol.
- Data (plural) - A **set of values** collected for the variable from each of the elements belonging to the sample.
- Experiment - A **planned activity** whose results yield a set of data.
- Parameter - A **numerical value** which summarizes the entire population data.
- Statistics - A **numerical value** which summarizes the sample data.

## Qualitative, or Attribute, or Categorical, Variable

- A variable that categorizes or describes a population element.

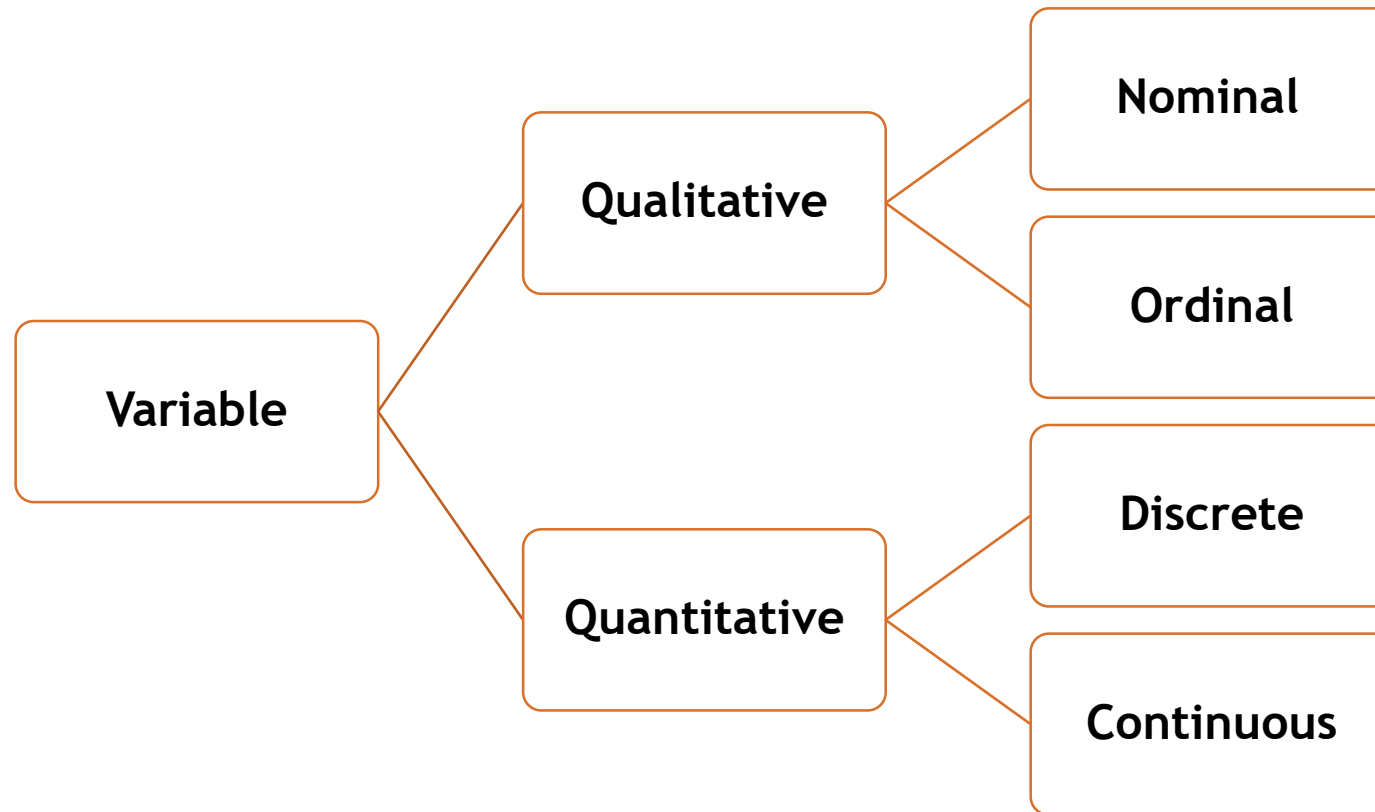
**Note:** Arithmetic operations such as addition and averaging, are not meaningful for data resulting from a qualitative variable.

## Quantitative, or Numerical, Variable

- A variable that quantifies a population element.

**Note:** Arithmetic operations such as addition and averaging, are meaningful for data resulting from a quantitative variable.





# Two Kinds of Variables



- Nominal Variable - A **qualitative variable** that categorizes (or describes, or names) a population element.
- Ordinal Variable - A **qualitative variable** that incorporates an ordered position or ranking.
- Discrete Variable - A **quantitative variable** that can assume a countable number of values.
  - This can assume values corresponding to the isolated points along a line interval.
  - There is a gap between any two values
- Continuous Variable - A **quantitative variable** that can assume an uncountable number of values.
  - This can assume any value along a line interval
  - Including every possible value between any two values

- Let  $x_1, x_2, x_3, \dots, x_n$  be the realized values of a random variable 'X', from a sample of size 'n'.

The **sample arithmetic mean** is defined as:

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

## Example

➤ The systolic blood pressure of seven middle aged men were as follows:

151, 124, 132, 170, 146, 124 and 113.

$$\begin{aligned}\text{The Mean is } \bar{x} &= \frac{(151 + 124 + 132 + 170 + 146 + 124 + 113)}{7} \\ &= 137.14\end{aligned}$$

- The median for the sample data arranged in an increasing order is defined as:
  - i. If “n” is an odd number - Middle value
  - ii. If “n” is an even number - Midway between the two middle values
- The mode is the most commonly occurring value.

## Example - n is odd

The re-ordered systolic blood pressure data seen earlier are:

113, 124, 124, 132, 146, 151, and 170.

- The **Median** is the middle value of the ordered data, i.e. 132.
- Two individuals have systolic blood pressure = 124 mm Hg, so the **Mode** is 124.

## Example - n is even

Six men with high cholesterol participated in a study to investigate the effects of diet on cholesterol level. At the beginning of the study, their cholesterol levels (mg/dL) were as follows:

366, 327, 274, 292, 274 and 230

Rearrange the data in numerical order as follows:

230, 274, 274, 292, 327 and 366.

- The **Median** is **half way between the middle two readings**, i.e.  $(274+292) / 2 = 283$ .
- The **mode** between the **two men having the same cholesterol level** = 274.

- If the histogram of the data is **right-skewed** then large sample values tend to inflate the mean.
- If the distribution is **skewed** then the median is not influenced by large sample values and is a better measure of centrality.

**Note** - If **mean = median = mode** then the data are said to be symmetrical.

For example,

- In the CK measurement study, the sample mean = 98.28.
- The median = 94.5, i.e. mean is larger than median indicating that mean is inflated by two large data values 201 and 203.



```
import numpy as np
from random import randint

x = [randint(1, 10) for p in range(0, 10)]
print(x)
mean = np.mean(x)
print("The mean is {:.2f}".format(mean))
median = np.median(x)
print("The median is {:.2f}".format(median))
mode = max(set(x), key=x.count)
print("The mode is {:.2f}".format(mode))
```

```
[3, 9, 5, 10, 10, 1, 7, 8, 5, 5]
The mean is 6.30
The median is 6.00
The mode is 5.00
```

- The concept **Measures of Dispersion** characterize how to spread out the distribution, i.e., how variable the data are.
- The commonly used dispersion measures include:
  - Range
  - Variance and Standard Deviation

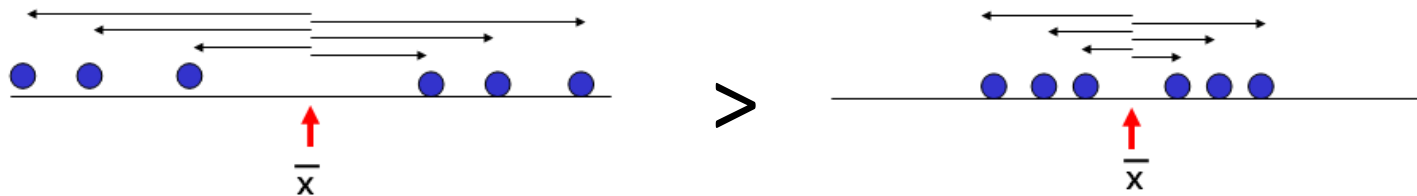
- The **Range** is the difference between the largest and the smallest observations in the sample.
- For example, the minimum and maximum blood pressure is 113 and 170 respectively. Hence the **range** is 57 mmHg
  - Easy to calculate;
  - Implemented for both “best” or “worst” case scenarios
  - Too sensitive for extreme values

# Sample Variance



- The sample variance,  $s^2$ , is the **arithmetic mean** of the squared deviations from the sample mean:

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



- The sample standard deviation (s) is the **square-root of the variance**.

$$s = \sqrt{\frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n-1}}$$

- The sample standard deviation has an advantage of being in the same units as the **original variable (x)**.

```
import numpy as np
from random import randint

x = [randint(1, 10) for p in range(0, 10)]
print(x)
variance = np.var(x)
print("The variance is {:.2f}".format(variance))
std = np.std(x)
print("The standard deviation is {:.2f}".format(std))
rng = max(x) - min(x)
print("The range is {:.2f}".format(rng))
```

```
[1, 10, 2, 9, 1, 4, 4, 2, 4, 6]
The variance is 9.01
The standard deviation is 3.00
The range is 9.00
```

[Practice Code](#)

$$\mu = \frac{\sum_{i=1}^N x_i}{N}$$

*Population Mean*

**Vs.**

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

*Sample Mean*



	Population	Sample
Size	N	n
Mean		



# Population Vs. Sample



	Population	Sample
Size	N	n
Mean		
Variance		

# Population Vs. Sample



- The variance of a population is:

$$\sigma^2 = \frac{\sum_{i=1}^N (x_i - \mu)^2}{N}$$

Population Mean

Population Size

The diagram shows the formula for population variance. A blue arrow points from the text 'Population Mean' to the Greek letter mu in the numerator. Another blue arrow points from the text 'Population Size' to the letter N in the denominator.

- The variance of a sample is:

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

Sample Mean

The diagram shows the formula for sample variance. A blue arrow points from the text 'Sample Mean' to the x-bar symbol in the numerator. Another blue arrow points from the text 'Note! the denominator is sample size (n) minus one !' to the 'n-1' in the denominator.

Note! the denominator is sample size (n) minus one !

➤ The **square root of the variance** is termed as the Standard Deviation, thus:

- The population Standard Deviation =  $\sigma = \sqrt{\sigma^2}$

- The Sample Standard deviation =  $s = \sqrt{s^2}$

- A more general interpretation of the standard deviation is derived from **Chebysheff's Theorem**, which applies to all shapes of histograms (except bell shaped).
- The proportion of observations in any sample that lie within k standard deviations of the mean is at least:

$$1 - \frac{1}{k^2} \text{ for } k > 1$$

For k=2 (say), the theorem states that at least 3/4 of all observations lie within 2 standard deviations of the mean. This is a “lower bound” compared to Empirical Rule's approximation (95%).

# Two Types of Random Variables



## Discrete Random Variable

- Takes on a **countable number** of values
- For example, values on the roll of dice: 2, 3, 4, ..., 12

## Continuous Random Variable

- Values are not **discrete**, not **countable**
- For example, time (30.1 minutes? 30.10000001 minutes?)

## Analogy

- Integers are **discrete**, while Real Numbers are **Continuous**

➤  $E(C) = C$

- The expected Value of a Constant is just the value of the constant.

➤  $E(X + C) = E(X) + C$

➤  $E(CX) = cE(X)$

- We can “pull” a constant out of the expected value expression (either as part of a sum with a random variable  $X$  or as a coefficient of random variable  $X$ ).

➤  $V(c) = 0$

- The Variance of constant (c) is zero.

➤  $V(X + c) = V(X)$

- The Variance of random variable and a constant is just the variance of the random variable (per 1 above).

➤  $V(cX) = c^2 V(X)$

- The Variance of a random variable and a constant co-efficient is the co-efficient squared times in the variance of the random variable.

# Probability Density Functions



Unlike a discrete random variable, a continuous random variable is one that can assume an uncountable number of values.

- We cannot list the possible values because there is an infinite number of them.
- The probability of each individual value is virtually 0 as there is an infinite number of values



# Point Probabilities are Zero



If the probability of each individual value is virtually 0 then there is an infinite number of values.

Thus, we can determine the probability of a **range of values** only.

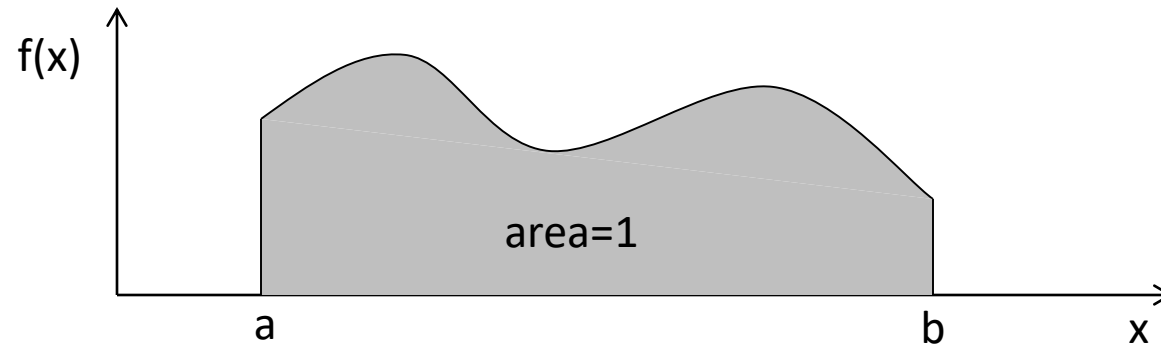
- For example, with a **discrete** random variable like tossing a die, it is meaningful to talk about  $P(X=5)$
- In a **continuous** setting (e.g. with time as a random variable), the probability the random variable of interest say task length, takes exactly 5 minutes is infinitely small, hence  $P(X=5) = 0$ .

# Probability Density Function



A function  $f(x)$  is called a Probability Density Function over the range  $a \leq x \leq b$  if it meets the following requirements:

1.  $f(x) \geq 0$  for all  $x$  between  $a$  and  $b$ , and



2. The total area under the curve between  $a$  and  $b$  is 1.0



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