Introduction to Data Science

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UET, Lahore

(Week 1; January 15 - 19, 2024)

Instructor

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Course Description

This course introduces students to the basics of Data Science including programming in R or Python, statistical inference, exploratory data analysis, basic machine learning algorithms, feature generation and feature selection. The foundation is laid for big data analytics ranging from social networks to business informatics.

Measurable Student Learning Outcomes

CLOs	Description	PLOs	Domain	Domain Level
CLO1	Explain basic statistical modeling and analysis.	PLO-01	Cognitive	2. Understand
CLO2	Use Exploratory Data Analysis (EDA) on different types of data.	PLO-02	Cognitive	3. Apply
CLO3	Compare tools and techniques to mine big datasets and graphs.	PLO-04	Cognitive	4. Analyze

Text Books

 Cathy O'Neil and Rachel Schutt. Doing Data Science, Straight Talk from The Frontline. Edition: 1st, Year: 2014, Publisher: O'Reilly Media

2

Grading Policy

Quizzes	20%
Presentations	10%
Mid Term	30%
Final Term	40%

Week	Topics	CLO(s)
1	What is data science? Big Data hype, Skill set needed for a data scientist.	1
2	Basic Statistics, Statistical Inference, Probability Distributions,	1
3	Introduction to Python, Basic concepts	1
4	Model fitting in Python, Data handling in Python	1
5	Exploratory Data Analysis, EDA in data science, Basic tools (graphs, summary).	2
6	Case studies on Exploratory Data Analysis, Social, Medical and Business Data Exploration	2
7	Basic Machine Learning Algorithms (Linear Regression, k-nearest neighbors, k-means)	3
8	Naïve Bayes, Spam Filtering	3

9	Feature Generation and Selection, Filters; Wrappers; Decision Trees; Random Forests.	3
10	Recommendation Systems, Dimensionality Reduction, Singular Value Decomposition, Principal Component Analysis	3
11	Basics of Graph Theory, Big real-world Graphs, Handling graphs in Python	1
12	Social Networks as Graphs, Characteristics of Social Graphs, Social Network Analysis	2
13	Bipartite Graphs, Business Graphs,	2
14	Data Visualization, Different tools to visualize data.	3
15	Data Science and Ethical Issues, Discussions on privacy, security, ethics, A look back at Data Science, Next-generation data scientists	2, 3
16	Revision	1,2,3

Why Data Science?

- One of the topmost professions
- New driving force behind industries is Data.
- Data Science is the Career of Tomorrow.

Skill Set Needed

- Statistics
- Programming skills
- Multivariable Calculus & Linear Algebra

Statistics

• In plural form, it refers to set of numerical data.

• In singular form, it is an academic discipline.

Data

- Facts and statistics collected for reference or analysis.
- Data are units of information, often numeric, that are collected through observation.
- Data is a collection of facts, such as numbers, words, measurements, observations or just descriptions of things.

What is Statistics

• Statistics is a branch of mathematics that deals with the scientific collection, organization, presentation, analysis, and interpretation of numerical data in order to obtain useful and meaningful information.

Descriptive Statistics

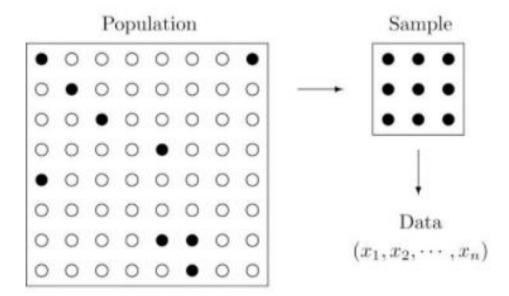
• A statistical method concerned with the collection, organization, presentation and description of sample data.

Inferential Statistics

• Inferential Statistics concerned with the analysis of a sample data leading to **prediction**, **inferences**, **interpretation**, **decision** or **conclusion** about the entire population

Population vs. Sample

- Population: The totality of all the elements or persons for which one has an interest at a particular time.
 - Students of 2018 session of CS-KSK
- Sample: It is a subset of a population
 - Students with CGPA > 3.0



Parameter vs. Statistic

• A parameter is a number describing a whole population.

• A <u>statistic</u> is a number describing a sample.

• With inferential statistics, we use <u>sample statistics</u> to make educated guesses about <u>population parameters</u>.

Quantitative vs. Qualitative Data

- Quantitative: These are numerical information obtained from counting or measuring that which can be manipulated by any fundamental operation.
 - Age, Weight, Height
- Qualitative: These are descriptive attributes and characterized by categorical responses.
 - Gender, Weather, Attitude

Variable

- A variable is any characteristics, number, or quantity that can be measured or counted.
- Independent variables: Variables you manipulate in order to affect the outcome of an experiment, e.g., Age
- Dependent variables: Variables that represent the outcome of the experiment, e.g., Salary

Descriptive vs. Inferential Statistics

• Descriptive: concerned with the <u>collection</u>, <u>organization</u>, <u>presentation and description</u> of sample data.

• Inferential: concerned with the analysis of a sample data leading to prediction, inferences, interpretation, decision or conclusion about the entire population

Inferential Statistics

- Inferential statistics takes data from a sample and makes inferences about the larger population from which the sample was drawn.
- Because the goal of inferential statistics is to draw conclusions from a sample and generalize them to a population, we need to have confidence that our sample accurately reflects the population.
 - Define the population we are studying.
 - Draw a representative sample from that population.
 - Use analyses that incorporate the sampling error.

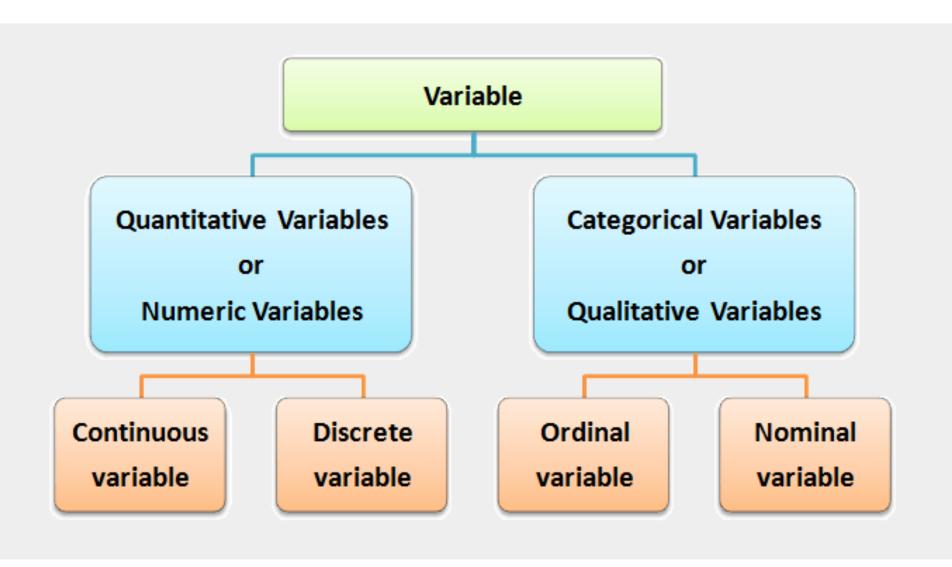
Probability Distributions

- A probability distribution is the mathematical function that gives the probabilities of occurrence of different possible outcomes of an experiment.
 - Tossing a coin
 - throwing a fair die
- Probability distributions are typically defined in terms of the probability distribution functions.

Probability Distribution Functions

Probability Mass Function (PMF) for Discrete Data

Probability Density Function (PDF) for Continuous Data Cumulative Distribution Function (CDF)



Discrete vs. Continuous Variable

- A **discrete variable** is a variable that takes on distinct, countable values. In theory, you should always be able to <u>count</u> the values of a discrete variable.
- A **continuous variable** is a variable that can take on any value within a range. Because the possible values for a continuous variable are infinite, we <u>measure</u> continuous variables (rather than count),

Probability Density Functions (PDFs)

• For a discrete random variable X that takes on a finite or countably infinite number of possible values, we determine P(X=x) for all the possible values of X, and call it the **probability mass function** (pmf)

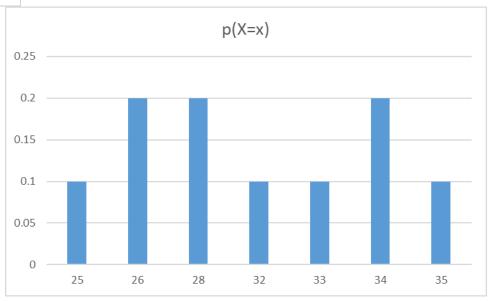
• For continuous random variables, the probability that X takes on any particular value x is 0. That is, finding P(X=x) for a continuous random variable is not going to work. Instead, we'll need to find the probability that falls in some interval (a,b), that is, we'll need to find P(a < X < b). We'll do that using a **probability density function** (pdf).

Probability Mass Function

Day	Travel Time (min)	pms	Χ	p(X=x)
1	25	0.1	25	0.1
2	26	0.2	26	0.2
3	26	0.2	28	0.2
4	28	0.2	32	0.1
5	28	0.2	33	0.1
6	32	0.1	34	0.2
7	33	0.1	35	0.1
8	34	0.2		
9	34	0.2		
10	35	0.1		

$$0 \le P(X = x) \le 1$$
 for any x

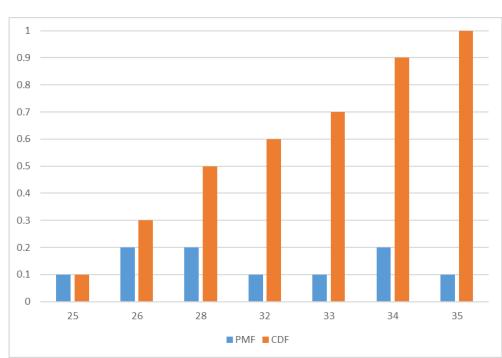
$$\sum_{x} P(X = x) = 1$$



Cumulative Distribution Function of PMF

Day	Travel Time (min)	pms	Χ	PMF	CDF
1	25	0.1	25	0.1	0.1
2	26	0.2	26	0.2	0.3
3	26	0.2	28	0.2	0.5
4	28	0.2	32	0.1	0.6
5	28	0.2	33	0.1	0.7
6	32	0.1	34	0.2	0.9
7	33	0.1	35	0.1	1
8	34	0.2			
9	34	0.2			
10	35	0.1			

$$F(x) = P(X \le x)$$



The **probability density function** (pdf), denoted f, of a continuous random variable X satisfies the following:

- 1. $f(x) \geq 0$, for all $x \in \mathbb{R}$
- 2. *f* is piecewise continuous

3.
$$\int\limits_{-\infty}^{\infty}f(x)\,dx=1$$

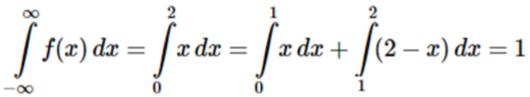
4.
$$P(a \leq X \leq b) = \int\limits_{b}^{a} f(x) \, dx$$

Let the random variable X denote the time a person waits for an elevator to arrive. Suppose the longest one would need to wait for the elevator is 2 minutes, so that the possible values of X (in minutes) are given by the interval [0,2].

A possible pdf for X is given by:

$$f(x) = egin{cases} x, & ext{for } 0 \leq x \leq 1 \ 2-x, & ext{for } 1 < x \leq 2 \ 0, & ext{otherwise} \end{cases}$$

$$f(x) = \left\{ egin{array}{ll} x, & ext{for } 0 \leq x \leq 1 \ 2-x, & ext{for } 1 < x \leq 2 \ 0, & ext{otherwise} \end{array}
ight.$$



probability that a person waits less than 30 seconds (or 0.5 minutes).

$$P(0 \leq X \leq 0.5) = \int\limits_{0}^{0.5} f(x) \, dx = \int\limits_{0}^{0.5} x \, dx = 0.125$$

Integral Formula
$$\int x^n dx = \frac{x^{n+1}}{(n+1)} + c$$

Continuous random variables have **zero point probabilities**, i.e., the probability that a continuous random variable equals a single value is always given by 0.

$$P(X=a)=P(a\leq X\leq a)=\int\limits_{a}^{a}f(x)\,dx=0.$$

Probability for a continuous random variable is given by areas under pdf's.

$$P(a \leq X \leq b) = P(a < X < b) = P(a \leq X < b) = P(a < X \leq b) = \int_{a}^{b} f(x) \, dx$$

Cumulative Distribution Function of PDF

Let X have pdf f, then the cdf F is given by

$$F(x)=P(X\leq x)=\int\limits_{-\infty}^x f(t)\,dt,\quad ext{for }x\in\mathbb{R}.$$

Let X be a continuous random variable with pdf f and cdf F.

• By definition, the cdf is found by *integrating* the pdf:

$$F(x) = \int\limits_{-\infty}^x \! f(t) \, dt$$

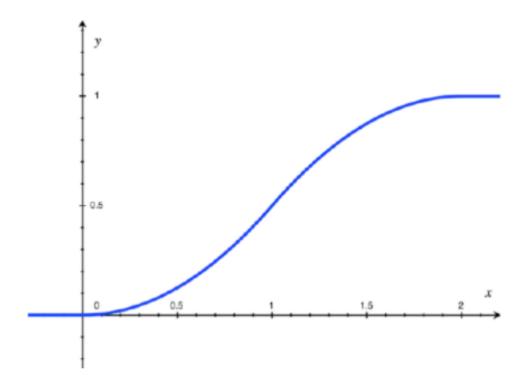
By the Fundamental Theorem of Calculus, the pdf can be found by differentiating the cdf:

$$f(x) = \frac{d}{dx}[F(x)]$$

Cumulative Distribution Function of PDF

PDF to CDF

$$F(x) = egin{cases} 0, & ext{for } x < 0 \ rac{x^2}{2}, & ext{for } 0 \leq x \leq 1 \ 2x - rac{x^2}{2}\,, & ext{for } 1 < x \leq 2 \ 1, & ext{for } x > 2 \end{cases}$$



Normal Distribution Formula

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}}e^{\frac{-(x-\mu)^2}{2\sigma^2}}$$

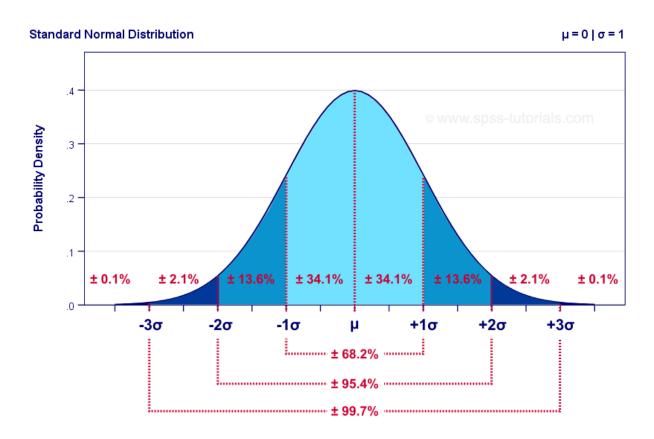
$$\mu = \text{mean of } x$$

$$\sigma$$
 = standard deviation of x

$$\pi \approx 3.14159 \dots$$

$$e \approx 2.71828 \dots$$

- The mean, median and mode are exactly the same.
- The distribution is **symmetric about the mean**—half the values fall below the mean and half above the mean.
- The distribution can be described by two values: the mean and the standard deviation.



Day	Time	11	28.24
1	32.14	12	29.10
2	31.30	13	28.34
3	29.17	14	28.50
4	28.15	15	29.26
5	30.30	16	28.29
6	30.41	17	25.36
7	32.37	17	25.50
8	33.19	18	27.18
9	31.19	19	30.29
10	30.37	20	27.15

Normal Distribution Formula

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}}e^{\frac{-(x-\mu)^2}{2\sigma^2}}$$

 $\mu = \text{mean of } x$

 σ = standard deviation of x

 $\pi \approx 3.14159 ...$

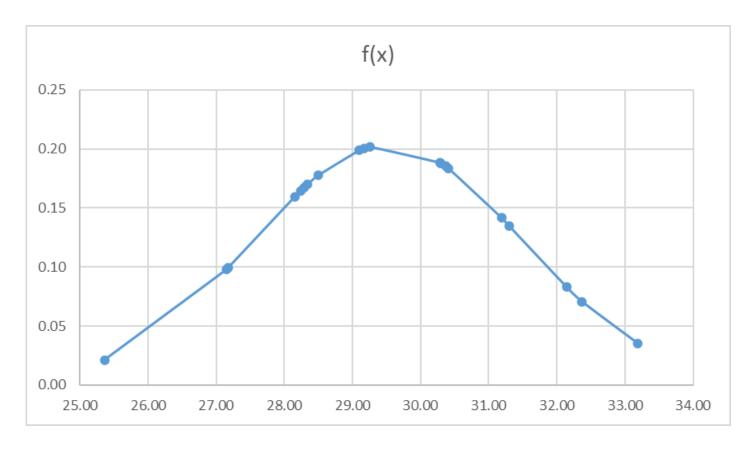
 $e \approx 2.71828 ...$

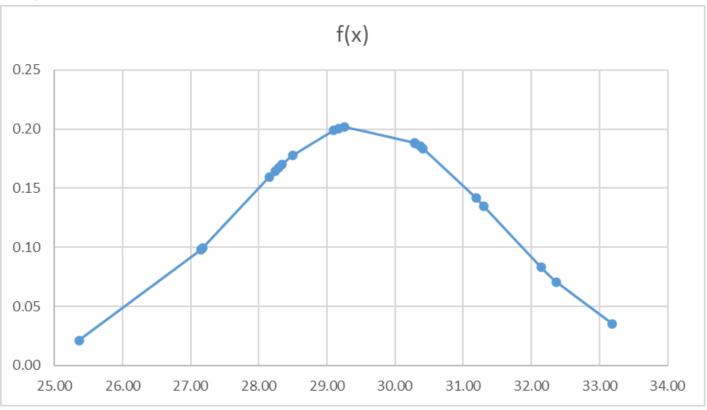
Standard Deviation Formula				
Population	Sample			
$\sigma = \sqrt{\frac{\sum (X - \mu)^2}{N}}$	$s = \sqrt{\frac{\sum (X - \bar{x})^2}{n - 1}}$			
X – The Value in the data distribution μ – The population Mean N – Total Number of Observations	X — The Value in the data distribution \bar{x} — The Sample Mean n - Total Number of Observations			

Day		Time	f(x)
	L	32.14	0.08
	2	31.30	0.13
3	3	29.17	0.20
4	1	28.15	0.16
Ĺ	5	30.30	0.19
(ĵ	30.41	0.18
-	7	32.37	0.07
8	3	33.19	0.04
g)	31.19	0.14
10)	30.37	0.19
13	L	28.24	0.16
12	2	29.10	0.20
13	3	28.34	0.17
14	1	28.50	0.18
15	5	29.26	0.20
16	5	28.29	0.17
17	7	25.36	0.02
18	3	27.18	0.10
19)	30.29	0.19
20)	27.15	0.10

Mean	29.52
St. Dev	1.96

Time	f(x)
25.36	0.02
27.15	0.10
27.18	0.10
28.15	0.16
28.24	0.16
28.29	0.17
28.34	0.17
28.50	0.18
29.10	0.20
29.17	0.20
29.26	0.20
30.29	0.19
30.30	0.19
30.37	0.19
30.41	0.18
31.19	0.14
31.30	0.13
32.14	0.08
32.37	0.07
33.19	0.04

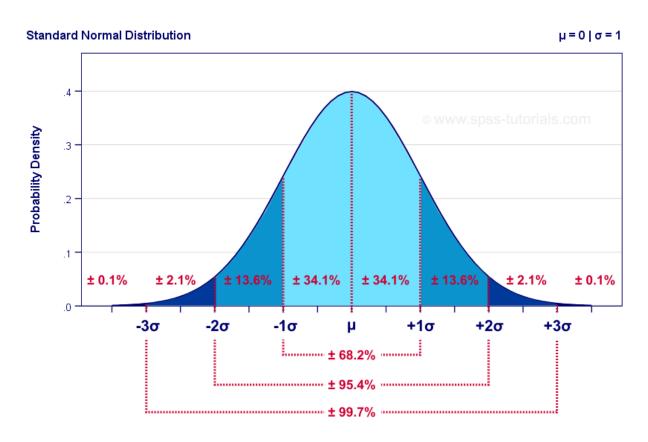


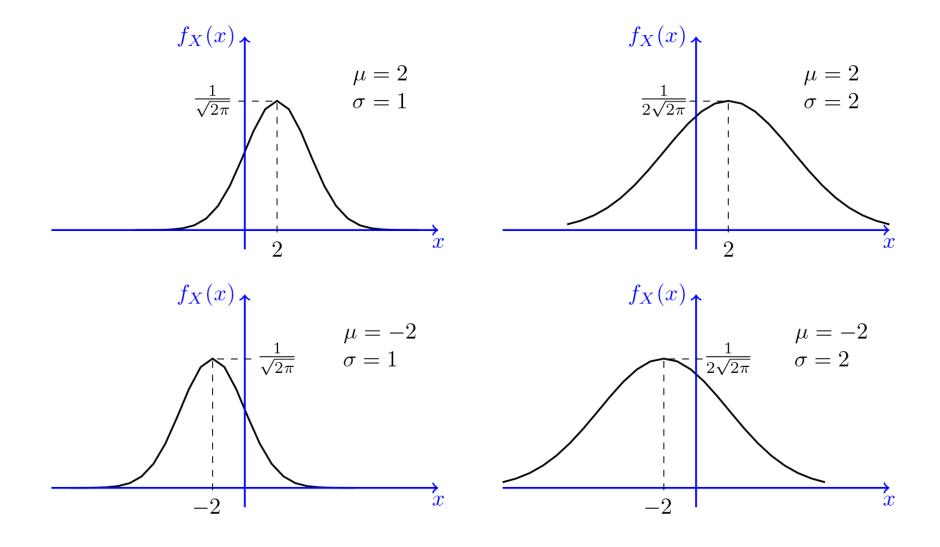


Mean	29.52
St. Dev	1.96

M+SD	31.48
M-SD	27.55

- The mean, median and mode are **exactly the same**.
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- The distribution can be described by two values: the mean and the standard deviation.



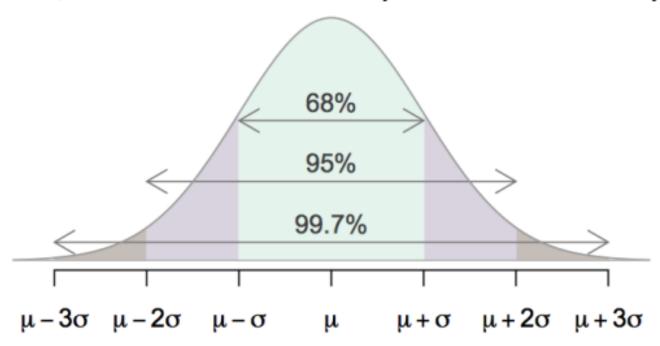


68-95-99.7 Rule

For nearly normally distributed data,

- about 68% falls within 1 SD of the mean,
- about 95% falls within 2 SD of the mean,
- about 99.7% falls within 3 SD of the mean.

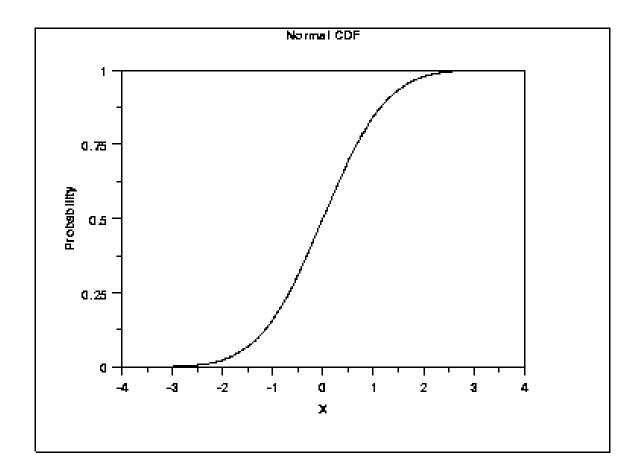
It is possible for observations to fall 4, 5, or more standard deviations away from the mean, but these occurrences are very rare if the data are nearly normal.



Activista IA/i

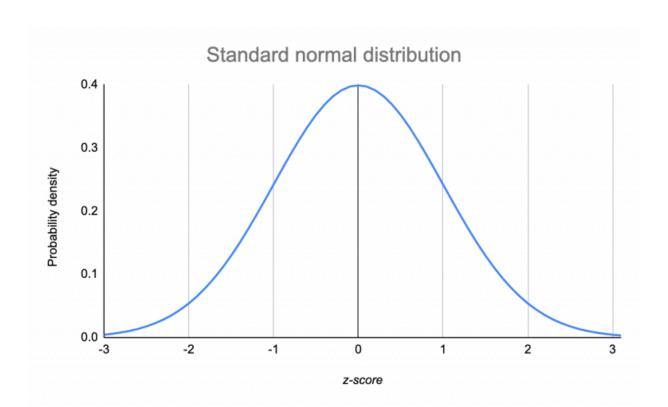
CDF of Normal Distribution

- The cumulative distribution function (cdf) is the probability that the variable X takes a value less than or equal to x.
- (Here in the figure below, Mean=0, SD=1)

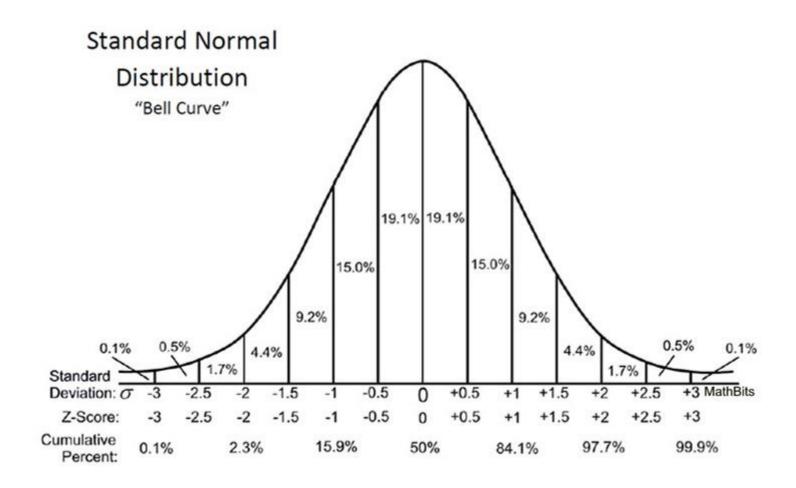


Z-Distribution

- The standard normal distribution, also called the z-distribution, is a special normal distribution where the mean is 0 and the standard deviation is 1.
- Z-scores tell you how many standard deviations away from the mean each value lies.



Z-Distribution



Z-Score

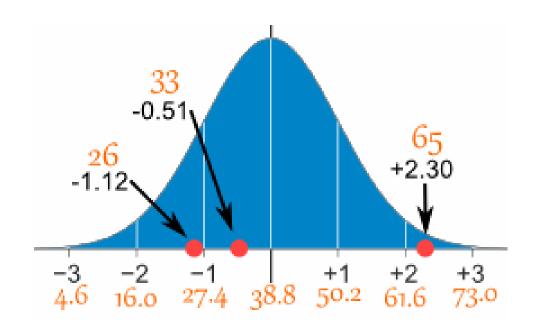
$$z = \frac{x - \mu}{\sigma}$$
 $x = raw score$
 $\mu = mean$
 $\sigma = standard deviation (std)$

As the formula shows, the **z-score** is simply the raw score minus the <u>population</u> mean, divided by the <u>population</u> standard deviation.

Z-Distribution

		1961 1
Day		Time
	1	26
	2	33
	3	65
	4	28
	5	34
	6	55
	7	25
	8	44
	9	50
	10	36
	11	26
	12	37
	13	43
	14	62
	15	35
	16	38
	17	45
	18	32
	19	28
	20	34

Mean is 38.8 minutes Standard Deviation is 11.4 minutes



Summary

• Introduction to Data Science