



# MACHINE LEARNING PROGRAM

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# CONTENT

- **Basics of Python**
  - Introduction To Variables
  - Loops and Lists\_Tuples
  - Pandas, Numpy
  - Functions And Strings
  - Data Aggregations using Pandas
- Introduction to Data Science
- Introduction to Machine Learning
  - Supervised Learning
  - Un-Supervised Learning
- Introduction to NLP
- Project

# PYTHON BASIC, NUMPY, PANDAS

- Lets see it on Jupiter Notebook

# ALGEBRA

Broad parts of mathematics, together with number theory, geometry and analysis.



# WHAT IS LINEAR ALGEBRA?

- Linear algebra is a sub-field of mathematics concerned with vectors, matrices, and operations on these data structures.
- In this we mostly work with Matrices and Vectors.

Scalar

1

Vector

$$\begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

Matrix

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

Tensor

$$\begin{bmatrix} \begin{bmatrix} 1 & 2 \end{bmatrix} & \begin{bmatrix} 3 & 2 \end{bmatrix} \\ \begin{bmatrix} 1 & 7 \end{bmatrix} & \begin{bmatrix} 5 & 4 \end{bmatrix} \end{bmatrix}$$

# WHY IS LINEAR ALGEBRA IMPORTANT TO MACHINE LEARNING?

- Linear Algebra is elementary unit for Machine Learning.
- Its concepts are a crucial prerequisite for understanding the theory behind all Machine Learning Algorithms.
- Linear algebra is used in data preprocessing, data transformation, and model evaluation.
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# LINEAR ALGEBRA INCLUDES

- Vectors
- Matrices
- Transpose of a matrix
- Inverse of a matrix
- Determinant of a matrix
- Dot product
- Eigenvalues
- Eigenvectors

# LINEAR ALGEBRA – KEY CONCEPT

- Linear Algebra Notation
- Linear Algebra Arithmetic
- Linear Algebra for Statistics
- Matrix Factorization
- Linear Least Squares

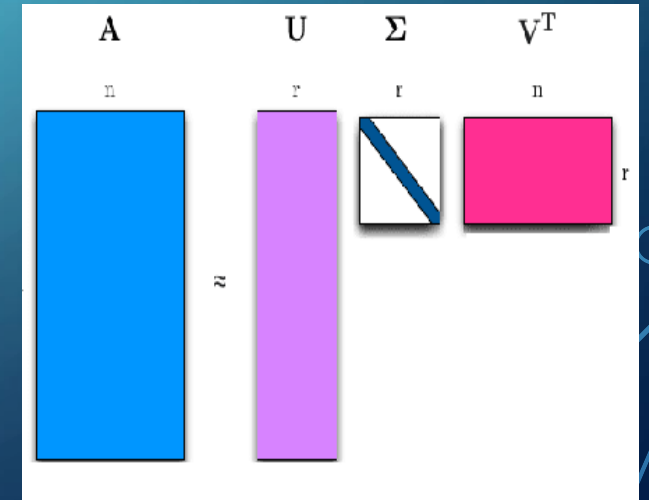
$$\text{RSS} := \sum_{i=1}^N (y_i - \hat{y}_i)^2$$

Addition	+
Subtraction	-
Multiplication	×
Division	÷

Matrix **A**:  $\mathbf{A} = \begin{pmatrix} 1 & 7 \\ 3 & 11 \end{pmatrix}$

Transpose of **A**:  $\mathbf{A}^T = \begin{pmatrix} 1 & 3 \\ 7 & 11 \end{pmatrix}$

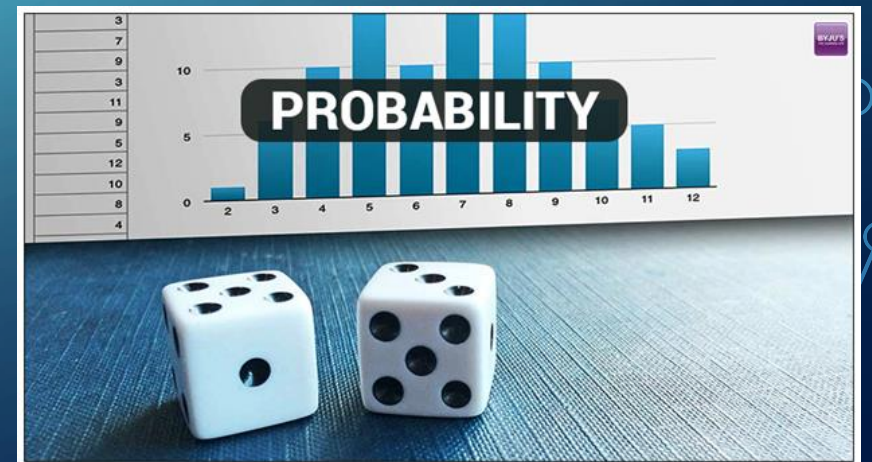
Row Vector **a**:  $\mathbf{a} = (5 \quad 0 \quad 4)$





# PROBABILITY

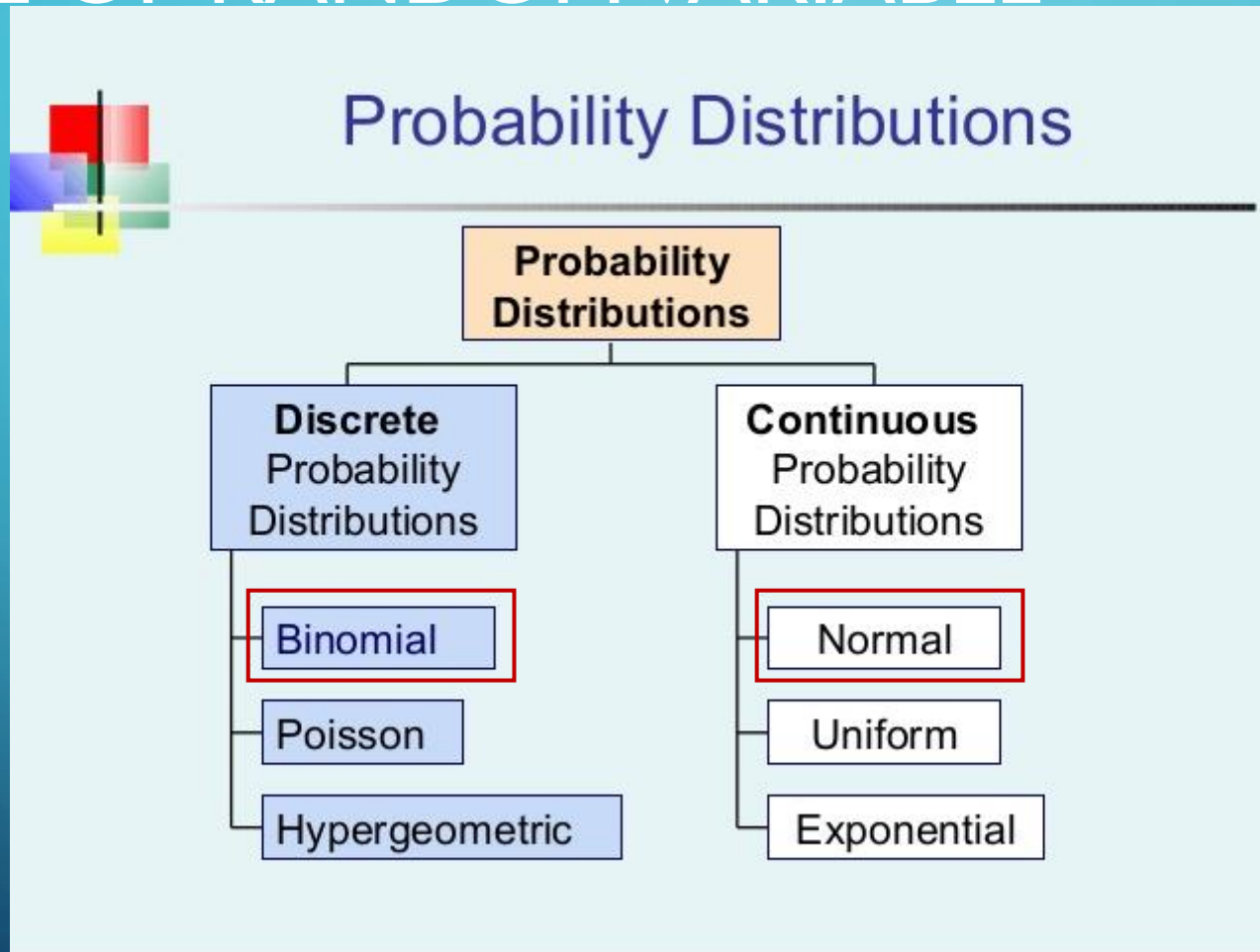
- Measure quantifying the likelihood that events will occur.



# PROBABILITY TERMINOLOGY

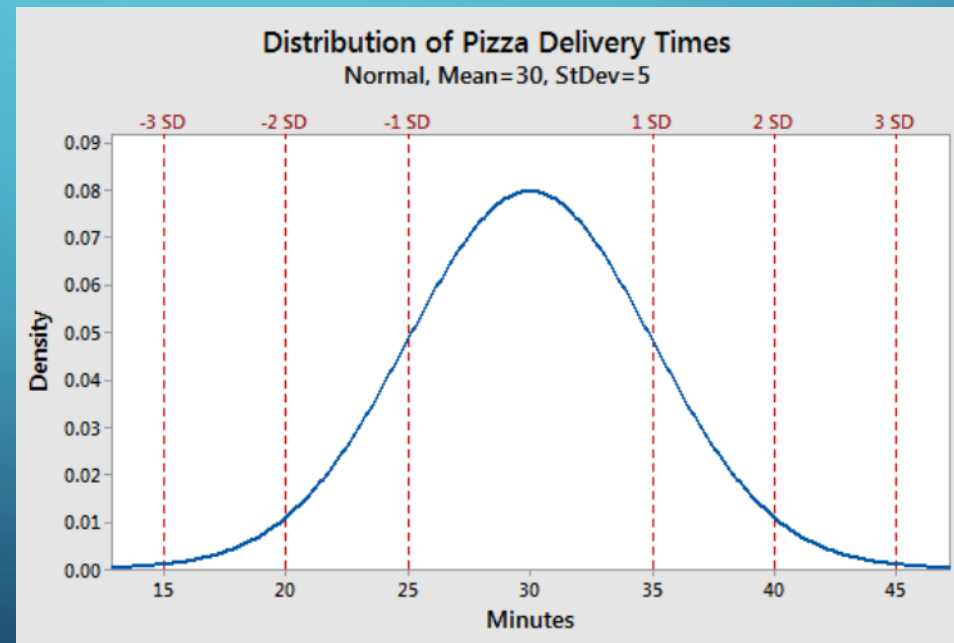
- **Experiment:** An occurrence with an uncertain outcome that we can observe. For example, rolling a die.
- **Outcome:** The result of an experiment; one particular state of the world. What Laplace calls a "case." For example: 4.
- **Sample Space:** The set of all possible outcomes for the experiment. For example,  $\{1, 2, 3, 4, 5, 6\}$ .
- **Event:** A subset of possible outcomes that together have some property we are interested in. For example, the event "even die roll" is the set of outcomes  $\{2, 4, 6\}$ .

# TYPE OF RANDOM VARIABLE



# PROBABILITY DISTRIBUTION

- A probability distribution tells you what is the probability of an event happening is.
- It can show **simple events**, like tossing a coin or picking a card or **complex events**, like the probability of a certain drug successfully treating cancer.



# BINOMIAL DISTRIBUTION

- The word “**Binomial**” literally means “**two numbers**”
- A binomial distribution for a random variable  $X$  is one in which there are only two possible outcomes either success or failure.
- **Success** and **Failure**, must be mutually exclusive

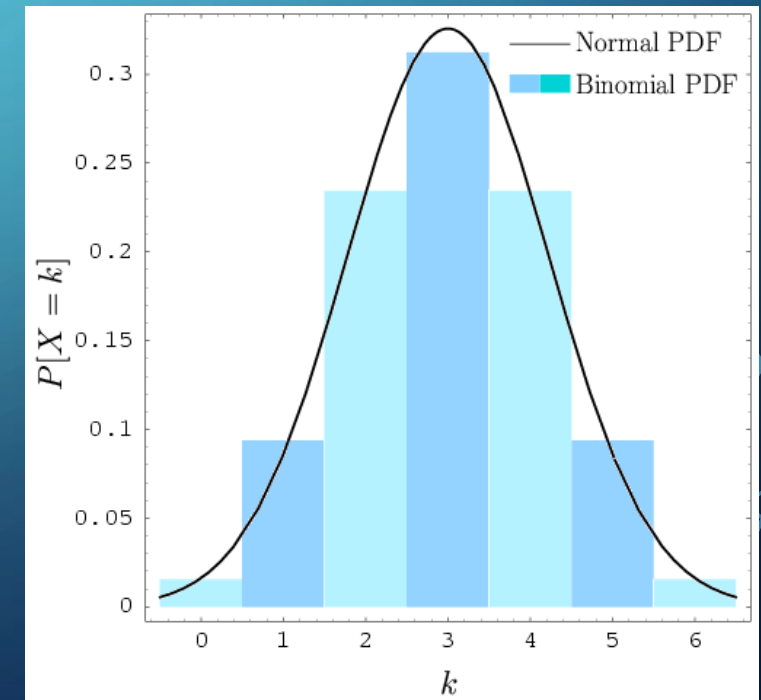
## Example:

- Let's assume we are flipping a coin 6 times.
  - Success: “the coin lands heads”  $\{p = 0.5\}$
  - Failure “the coin lands tails.”  $\{q = 0.5\}$

## Possible Outcomes

HHHHHT  
HHHHTH  
HHHTHH  
HHTHHH  
HTHHHH  
THHHHH

As the number of trials increases,  
the binomial distribution will approach the normal distribution.



# NORMAL DISTRIBUTION

- The normal distribution is a continuous distribution or a function that can take on values anywhere on the real line.
- The normal distribution is parameterized by two parameters: **the mean of the distribution  $\mu$  and the variance  $\sigma^2$ .**

## Example:

- Suppose that we measure and record the heights of 1000 girls selected at random.

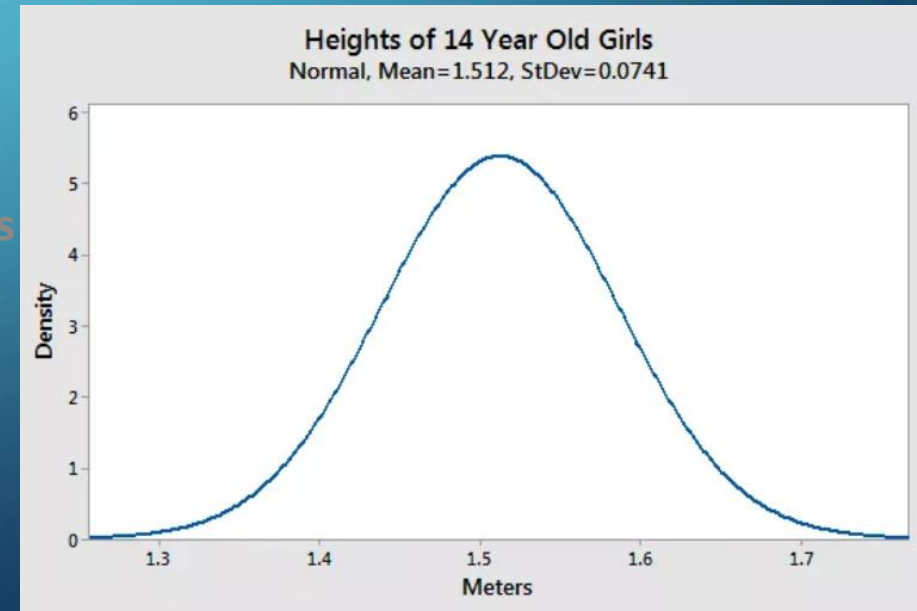
{1.35, 1.43, 1.57, 1.66, 1.72....}

- **Probability of occurrence of (1.35) =**
  - **No. of times 1.35 has occurred / total no. of entries**

$$P(1.35) = 67/1000 = 0.067$$

$$P(1.66) = 205/1000 = 0.205$$

$$P(1.72) = 445/1000 = 0.445$$





# REFERENCES

- <https://medium.com/@rathi.ankit/linear-algebra-for-data-science-a9648b9daee0>
- <https://machinelearningmastery.com/why-learn-linear-algebra-for-machine-learning/>
- <https://machinelearningmastery.com/examples-of-linear-algebra-in-machine-learning/>



Thank you.  
Questions?