

ILT 3 - Mathematics for Machine Learning

Ground Rules

Observe the following rules to ensure a supportive, inclusive, and engaging classes



Give full attention
in class



Mute your microphone
when you're not talking



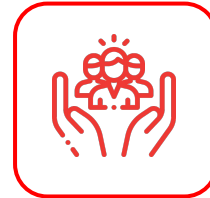
Keep your
camera on



Turn on the CC Feature
on Meet



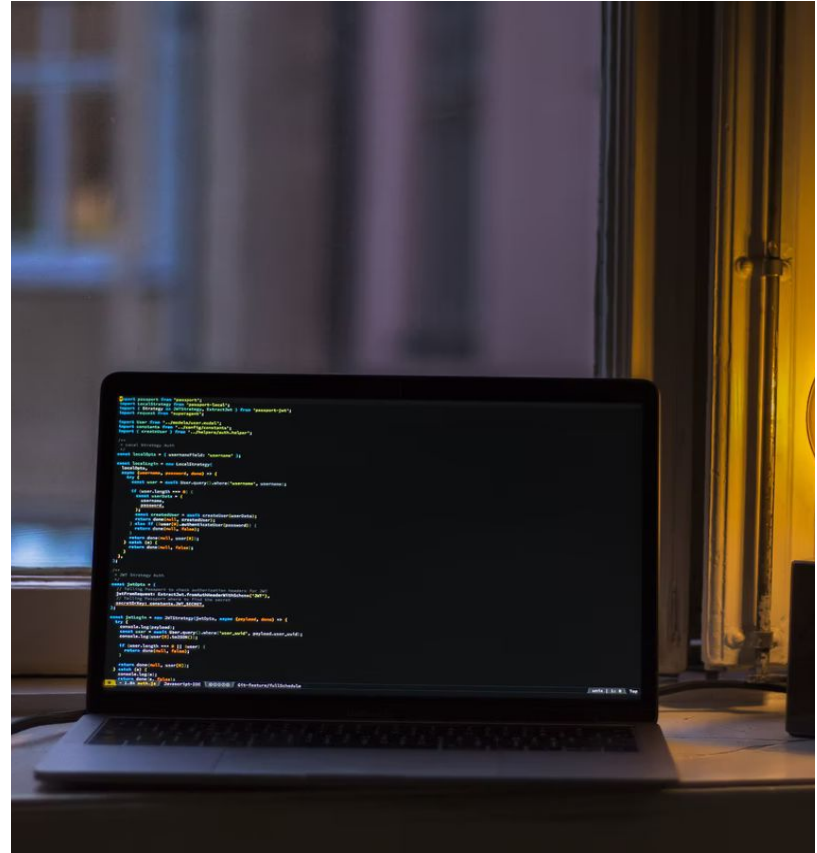
Use raise hand or chat
to ask questions



Make this room a safe place
to learn and share

Outline **Session**

- Introduction to **Linear Algebra** & **Calculus**
- Basic **Statistics** of Data
- **PCA**



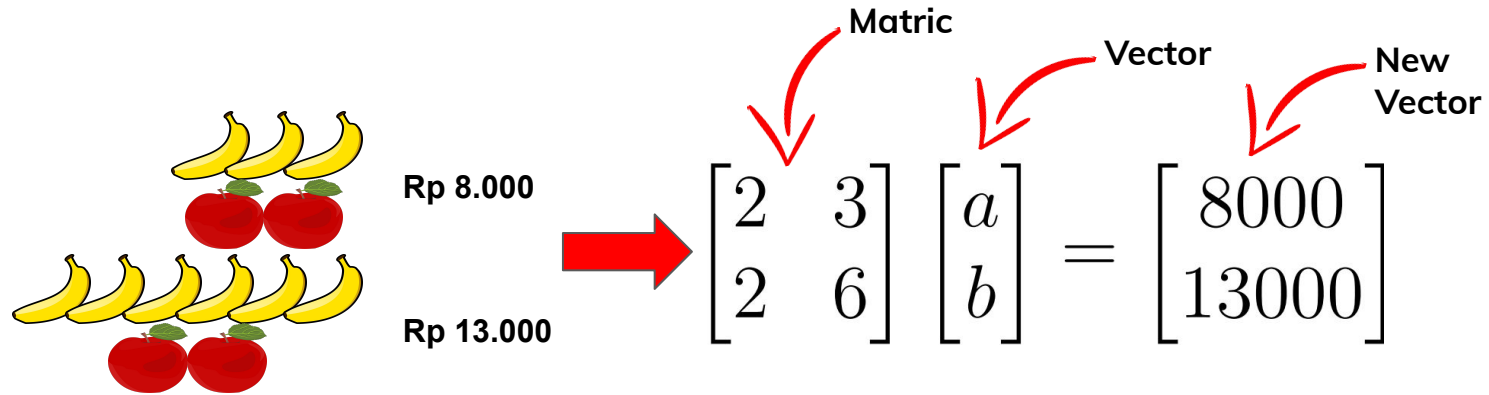
Introduction to Linear Algebra

What is **Linear Algebra**?

Linear algebra is a mathematical system for **manipulating vectors** in the spaces described by **vectors**.



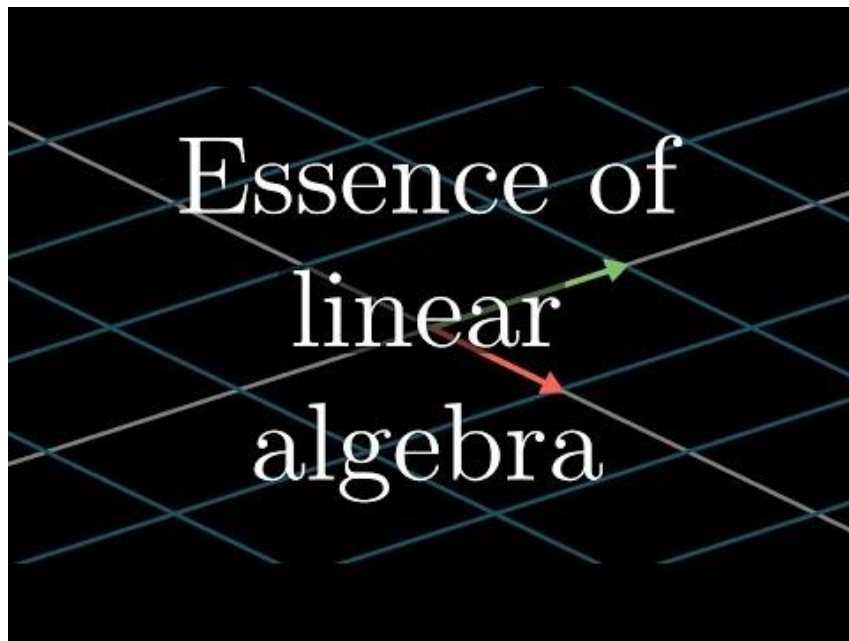
How to **Manipulate** a Vector?



Matrix is an object that will **transform** a vector into a **new vector**.

Why **Linear Algebra**?

- Linear algebra is about **vectors** and **matrices**
- Linear algebra is essentially the **mathematics of data**



Important Concepts

Mathematics for Machine Learning: Linear Algebra

Formula Sheet

Vector operations

$$\mathbf{r} + \mathbf{s} = \mathbf{s} + \mathbf{r}$$

$$2\mathbf{r} = \mathbf{r} + \mathbf{r}$$

$$\|\mathbf{r}\|^2 = \sum_i r_i^2$$

- dot or inner product:

$$\mathbf{r} \cdot \mathbf{s} = \sum_i r_i s_i$$

commutative $\mathbf{r} \cdot \mathbf{s} = \mathbf{s} \cdot \mathbf{r}$

distributive $\mathbf{r} \cdot (\mathbf{s} + \mathbf{t}) = \mathbf{r} \cdot \mathbf{s} + \mathbf{r} \cdot \mathbf{t}$

associative $\mathbf{r} \cdot (a\mathbf{s}) = a(\mathbf{r} \cdot \mathbf{s})$

$$\mathbf{r} \cdot \mathbf{r} = \|\mathbf{r}\|^2$$

$$\mathbf{r} \cdot \mathbf{s} = \|\mathbf{r}\| \|\mathbf{s}\| \cos \theta$$

- scalar and vector projection:

scalar projection: $\frac{\mathbf{r} \cdot \mathbf{s}}{\|\mathbf{r}\|}$

vector projection: $\frac{\mathbf{r} \cdot \mathbf{s}}{\mathbf{r} \cdot \mathbf{r}} \mathbf{r}$

Change of basis

Change from an original basis to a new, primed basis. The columns of the transformation matrix B are the new basis vectors in the original coordinate system. So

$$B\mathbf{r}' = \mathbf{r}$$

where \mathbf{r}' is the vector in the B -basis, and \mathbf{r} is the vector in the original basis. Or;

$$\mathbf{r}' = B^{-1}\mathbf{r}$$

If a matrix A is *orthonormal* (all the columns are of unit size and orthogonal to each other) then:

$$A^T = A^{-1}$$

Gram-Schmidt process for constructing an orthonormal basis

Start with n linearly independent basis vectors $\mathbf{v} = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$. Then

$$\mathbf{e}_1 = \frac{\mathbf{v}_1}{\|\mathbf{v}_1\|}$$



$$\mathbf{u}_2 = \mathbf{v}_2 - (\mathbf{v}_2 \cdot \mathbf{e}_1) \mathbf{e}_1 \quad \text{so} \quad \mathbf{e}_2 = \frac{\mathbf{u}_2}{\|\mathbf{u}_2\|}$$


... and so on for \mathbf{u}_3 being the remnant part of \mathbf{v}_3 not composed of the preceding \mathbf{e} -vectors, etc. ...

Eigenstuff


An eigenstuff (**eigenvector** & **eigenvalue**) is a set of tools for finding the **characteristic properties** of a matrix.

$$\mathbf{A}\mathbf{x} = \lambda\mathbf{x}$$

matrix  eigenvector 

eigenvalue 

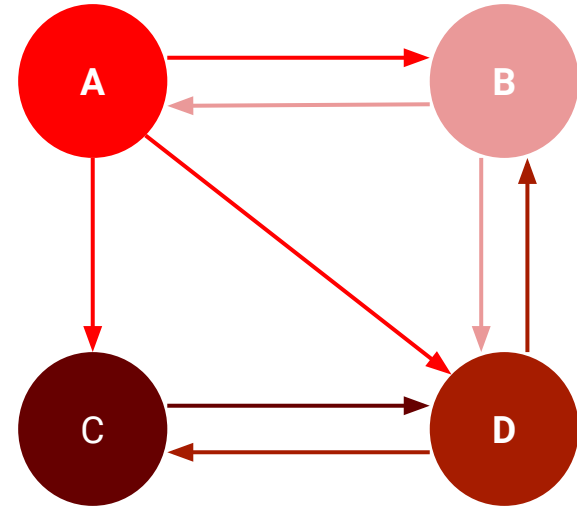
$$(\mathbf{A} - \lambda\mathbf{I})\mathbf{x} = 0$$

 identity matrix

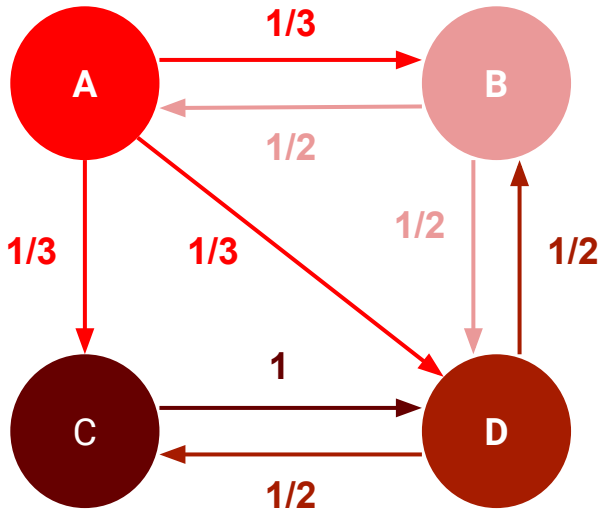
What are the **examples of Linear Algebra** in Machine Learning?

PageRank Algorithm

- The PageRank algorithm was published by Larry Page (Google founder) and colleagues in 1998.
- Used by Google to help them decide which order to display their websites when they returned from the search.



How PageRank Algorithm Works?



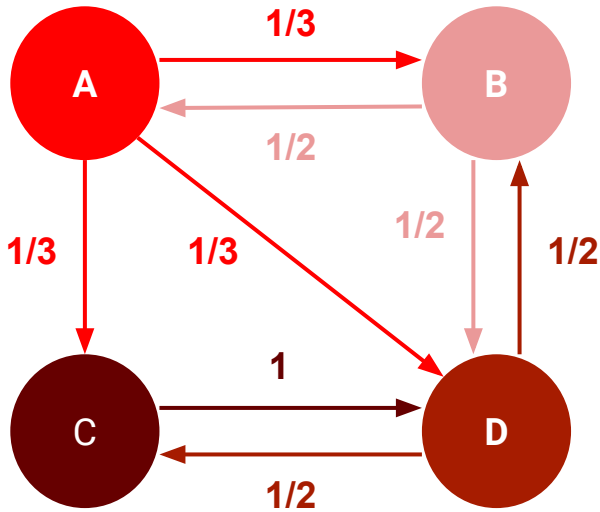
$$L_a = [0 \ \frac{1}{3} \ \frac{1}{3} \ \frac{1}{3}]$$

$$L_b = [\frac{1}{2} \ 0 \ 0 \ \frac{1}{2}]$$

$$L_c = [0 \ 0 \ 0 \ 1]$$

$$L_d = [0 \ \frac{1}{2} \ \frac{1}{2} \ 0]$$

How PageRank Algorithm Works?

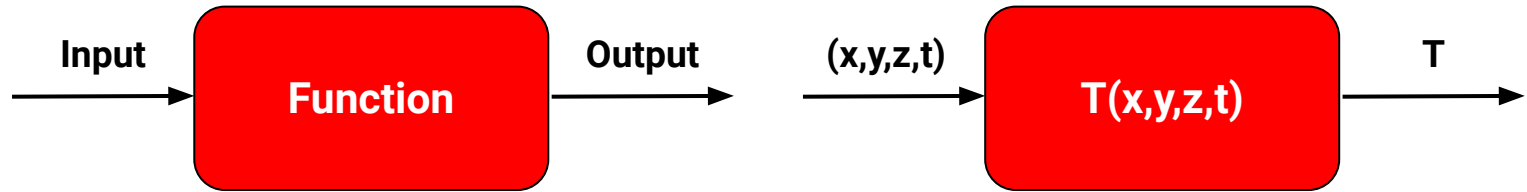


$$L = \begin{bmatrix} 0 & 1/2 & 0 & 0 \\ 1/3 & 0 & 0 & 1/2 \\ 1/3 & 0 & 0 & 1/2 \\ 1/3 & 1/2 & 1 & 0 \end{bmatrix}$$

$$r_{i+1} = Lr_i$$

Introduction to Calculus

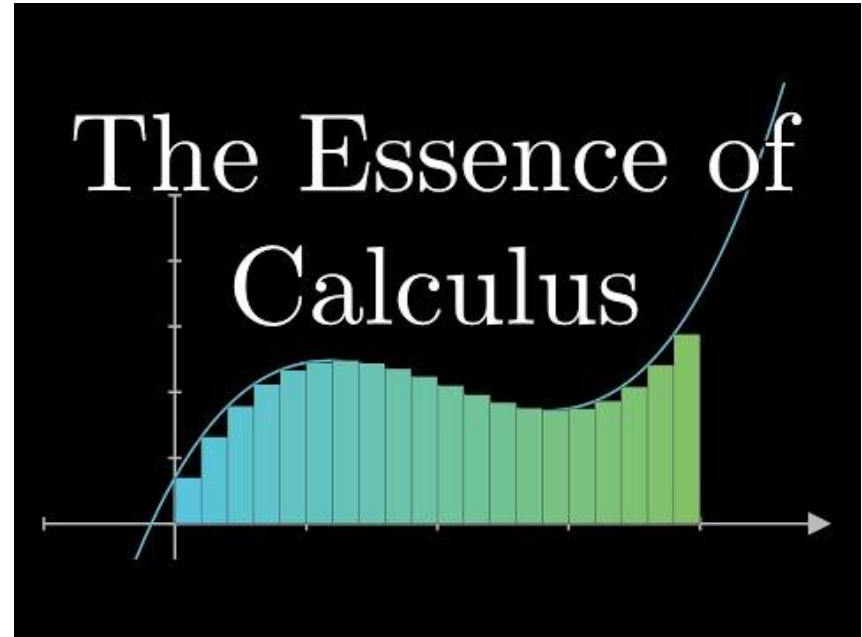
Concept of **Function**



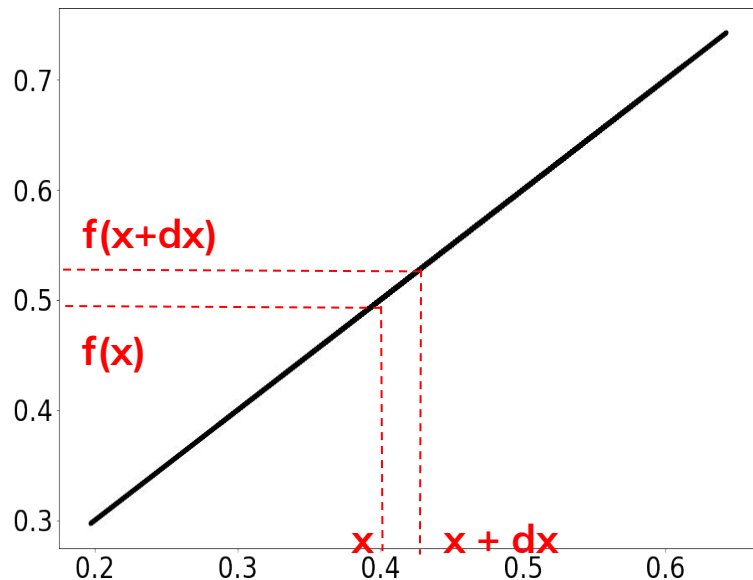
A function is a **relationship** between some **inputs** and an **output**.

Concept of **Calculus**

- Calculus is the study of how these **functions change with respect to their input variables.**
- Many important machine learning approaches **have calculus at their core**



Definition of Derivative



$$\text{gradient} = \frac{f(x + dx) - f(x)}{dx}$$

$$f'(x) = \lim_{dx \rightarrow 0} \frac{f(x + dx) - f(x)}{dx}$$

Important Concepts

Mathematics for Machine Learning

Multivariate Calculus

Formula sheet

Dr Samuel J. Cooper

Prof. David Dye

Dr A. Freddie Page

Definition of a derivative

$$f'(x) = \frac{df(x)}{dx} = \lim_{\Delta x \rightarrow 0} \left(\frac{f(x + \Delta x) - f(x)}{\Delta x} \right)$$

Time saving rules

- *Sum Rule:*

$$\frac{d}{dx} (f(x) + g(x)) = \frac{d}{dx} (f(x)) + \frac{d}{dx} (g(x))$$

- *Power Rule:*

$$\begin{aligned} \text{Given } f(x) &= ax^b, \\ \text{then } f'(x) &= abx^{(b-1)} \end{aligned}$$

Derivatives of named functions

$$\frac{d}{dx} \left(\frac{1}{x} \right) = -\frac{1}{x^2}$$

$$\frac{d}{dx} (\sin(x)) = \cos(x)$$

$$\frac{d}{dx} (\cos(x)) = -\sin(x)$$

$$\frac{d}{dx} (\exp(x)) = \exp(x)$$

Derivative structures

$$\text{Given } f = f(x, y, z)$$

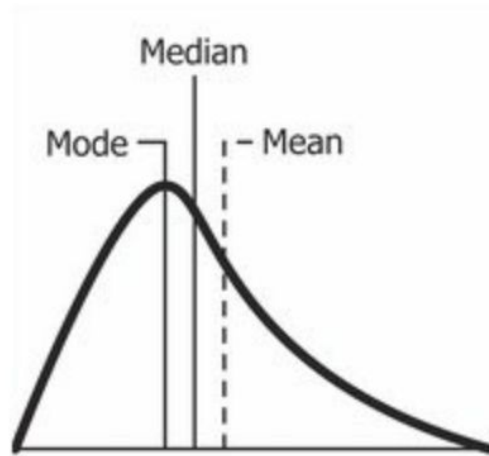
Basic Statistics of Data

Analyzing Quantitative Data

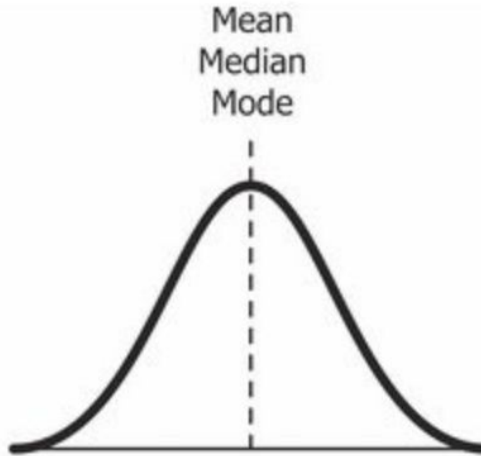
There are 4 main aspects of analyzing quantitative data

- **Shape** describes the distribution of data.
- **Spread** describes the variation of data.
Two measures of spread are variance and standard deviation.
- **Outliers** is an observation that lies outside the overall pattern of a distribution
- **Center**
The measure of center can give us an idea of the central position of the data set.
 - Mode
 - Median
 - Mean

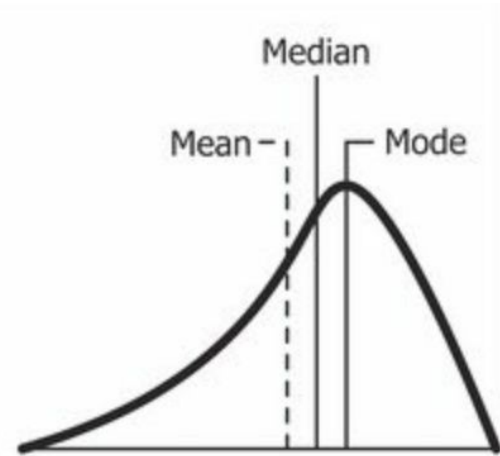
Data Distribution



Right Skewed
Distribution



Normal
Distribution



Left Skewed
Distribution

Mean

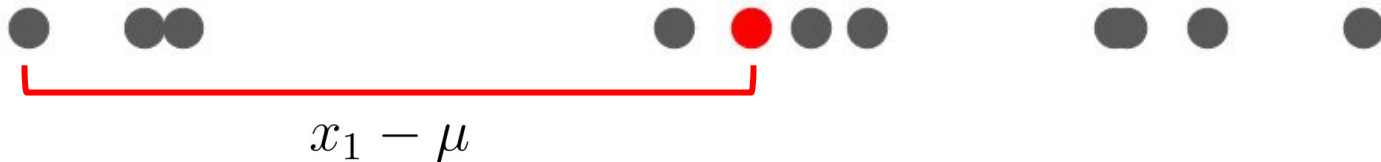


The mean of a data set describes **the average data point**.

$$D = \{x_1, x_2, x_3, \dots, x_N\}$$

$$E[D] = \frac{1}{N} \sum_{n=1}^N x_n$$

Variance



Variance is used to characterize the **variability or spread of data points** in a dataset.

$$D = \{x_1, x_2, x_3, \dots, x_N\}$$

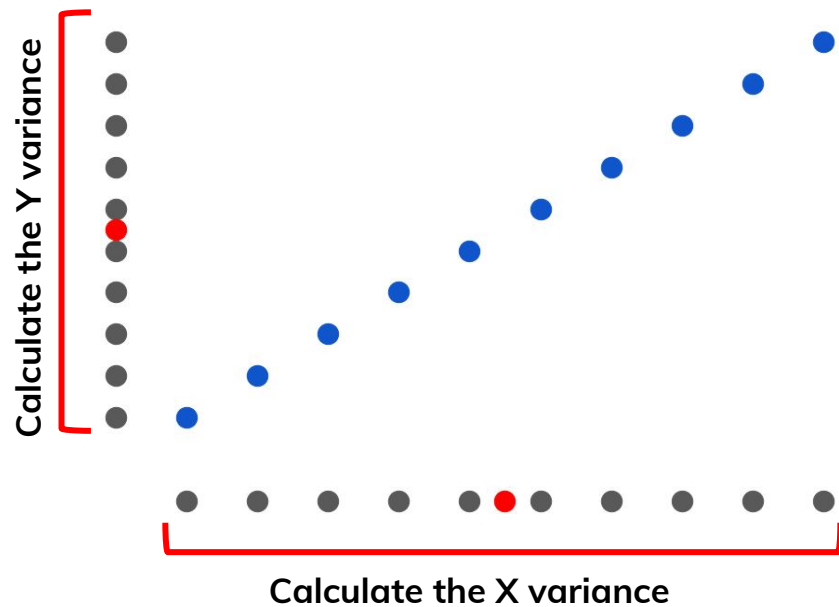
$$Var[D] = \frac{1}{N} \sum_{n=1}^N (x_n - \mu)^2 \quad \mu = E[D]$$

Why Variance is Important?

- Variance can describe how well the mean or median represents the data
- Variance can tell how much do we trust conclusions based on the mean and median.

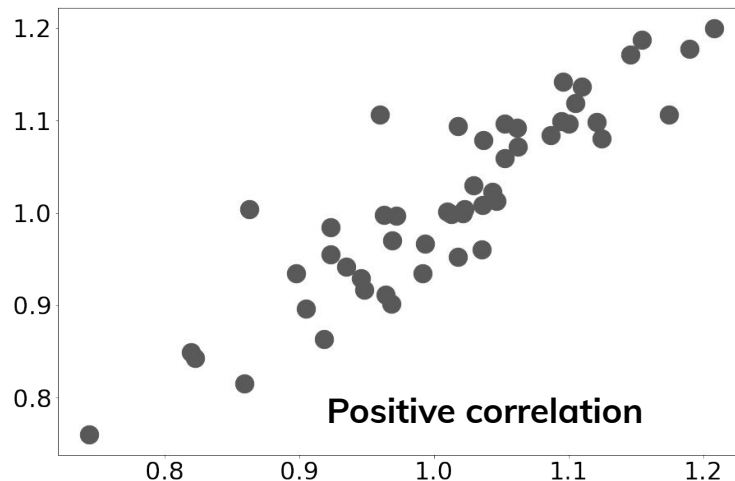


Variance 2D



Covariance

- The correlation between two variable can be captured by extending the notion of the variance to what is called the **covariance** of the data.
- There 3 types of covariance:
 - **Positive** correlation.
 - **Negative** correlation.
 - **Zero** correlation.

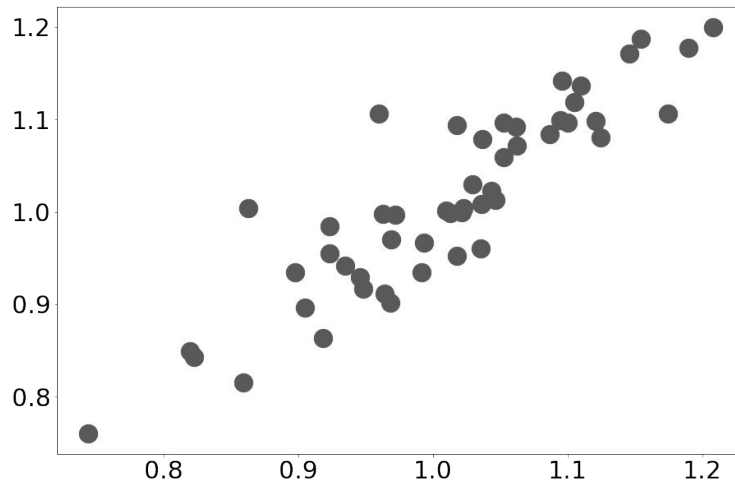


$$\text{cov}[x, y] = E[(x - \mu_x)(y - \mu_y)]$$

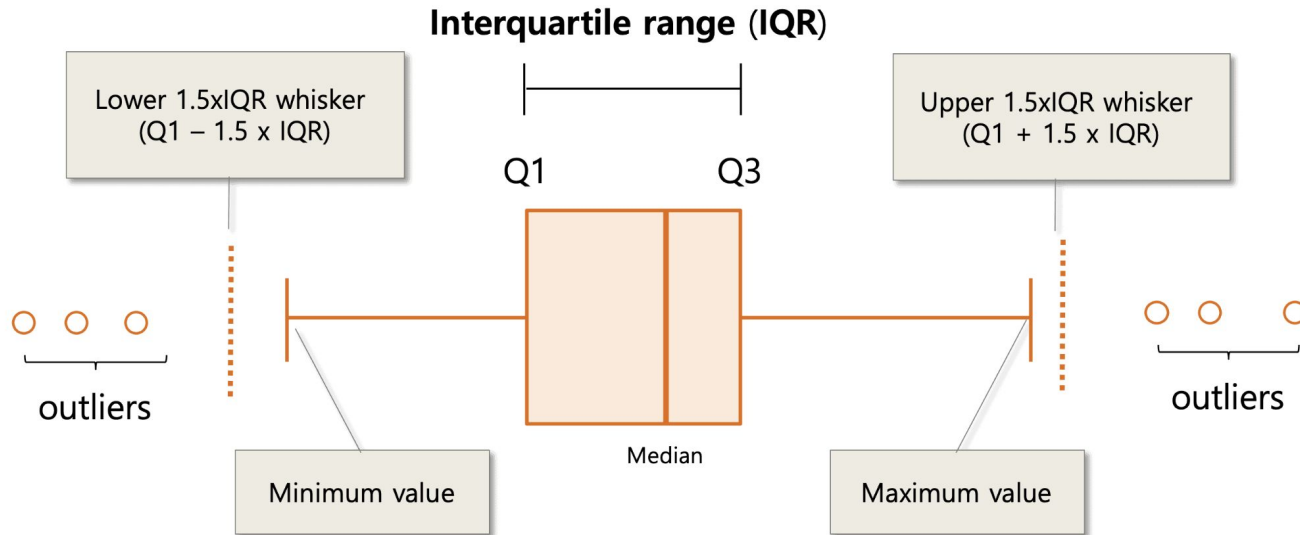
Covariance Matrix

- The variance of X
- The variance of Y
- The covariance of X,Y
- The covariance of Y,X

$$\begin{bmatrix} \text{var}[x] & \text{cov}[x, y] \\ \text{cov}[y, x] & \text{var}[y] \end{bmatrix}$$



Outliers





PCA

Principal Component Analysis

Data in Real Life

- Data in real life is often **high dimensional**.
- Working with high dimensional data comes with some difficulties.
- **Dimensionality reduction** allows us to work with a more compact representation of the data.



Bedrooms

Bathrooms

Size

Sale Price

Crime Rate

Condition

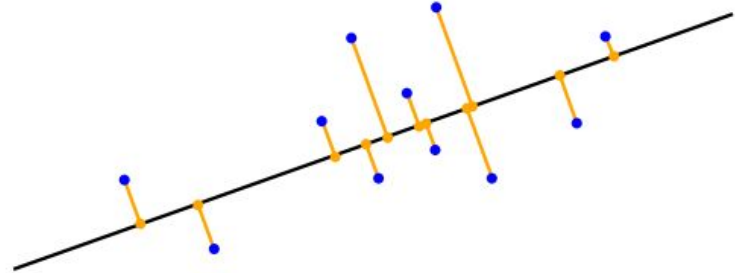
Type of House

Year Built

etc..

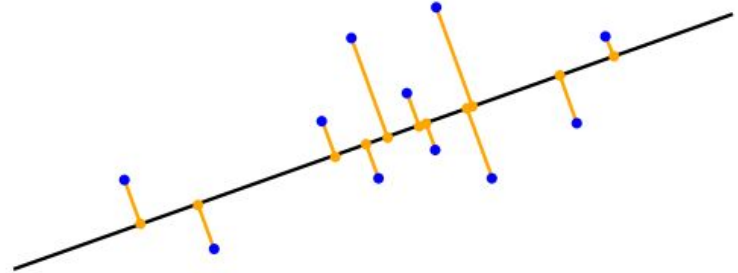
What is PCA?

- **Principal Component Analysis** (PCA) is a classical algorithm for linear dimensionality reduction
- PCA reduced the dimensionality of the data by projecting them into a **lower-dimensional subspace**
- In PCA we'll use some concepts such as **eigenstuff**, **variance**, and **covariance**

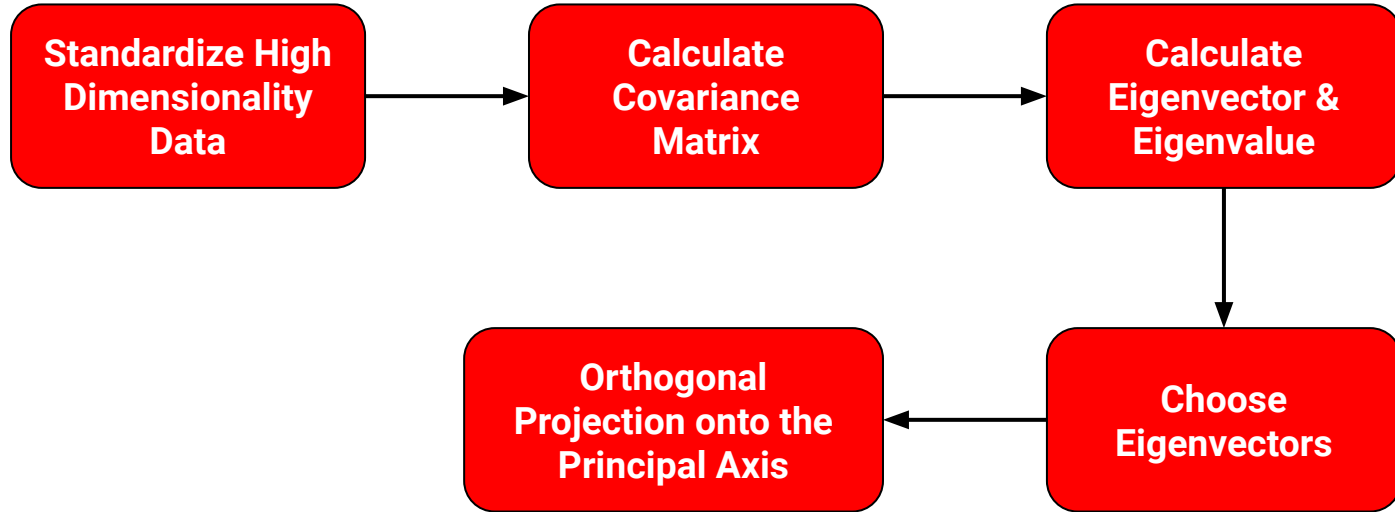


When Should We use PCA?

- If you want to reduce the dimensionality of the data but **can't identify which variables** to completely remove from consideration
- If you are comfortable making your data & model **less interpretable**
- If you want to ensure your **variables are independent** of one another

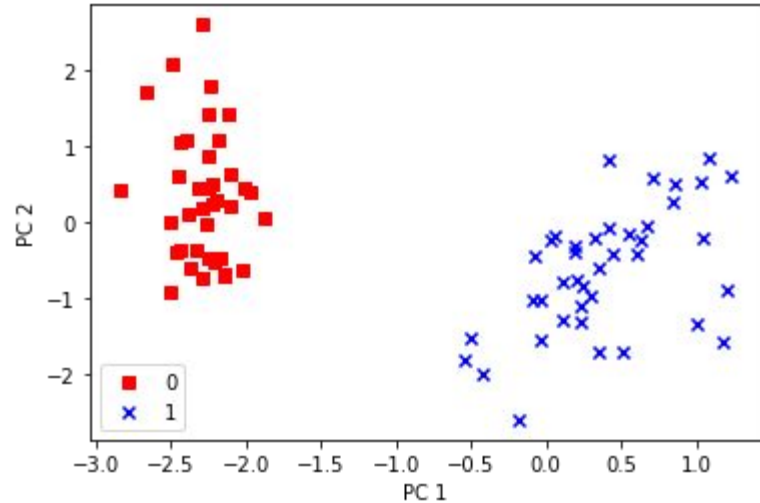


Key Step of **PCA**



How PCA Impact the **Training Process**?

- **Speed up** the training process by reducing the dimensionality of the data
- **Improve the accuracy** of the classification model



Demo Link

Demo PCA from scratch & use scikit-learn:

https://colab.research.google.com/drive/1tqCvpEcDDB9_kZ0PihRLDg8ZWmyJCDcg?usp=sharing

Demo Improve the Accuracy of the Classification Model using PCA :

https://colab.research.google.com/drive/1qKSszYc3TMTEP7xsHbq0X-i_AluJqNtw?usp=sharing

Sharing Session

Discussions

Quiz

Thank You