

# Deep Learning Introduction



**California Science  
and Technology**  
UNIVERSITY

# About Me

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 Author and Technologist

 Worked for TI, Magma, Apache, Cadence, Paripath and now AITS.

 20 years in EDA/CAD/ML industry

 [www.linkedin.com/in/srohit0/](https://www.linkedin.com/in/srohit0/)

 [medium.com/@srohit0](https://medium.com/@srohit0)

 [qr.ae/TWGS9](https://qr.ae/TWGS9)

 [github.com/srohit0](https://github.com/srohit0)

 [twitter.com/srohit](https://twitter.com/srohit)



# Course Overview

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- ❑ Pre-requisites
  - ❑ basic computer science principles and skills
  - ❑ Probability theory
  - ❑ Multivariable calculus and Linear algebra
- ❑ Applied course with emphasis on real life projects
- ❑ Math and programming makes it fun and challenging
- ❑ Make friends for study groups for projects.
- ❑ Reference Book :
  - ❑ Machine Intelligence, Rohit Sharma, 2018.
  - ❑ [srohit0.github.io/mida/](https://srohit0.github.io/mida/)

Homework	Quiz	Midterm Project	Final Project	Final Exam	Participation	Total
5%	15%	20%	25%	30%	5%	100%

# Material

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- ❑ Text Book:

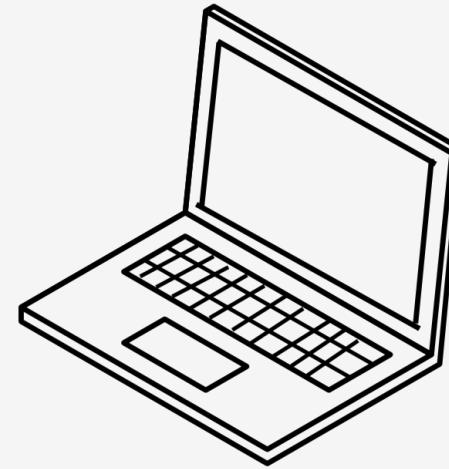
- ❑ [Machine Intelligence](#) by Rohit Sharma

- ❑ References Material:

- ❑ [Python Machine Learning, 2nd Edition](#), by Sebastian Raschka
  - ❑ [Deep Learning](#), by Ian Goodfellow

- ❑ Software

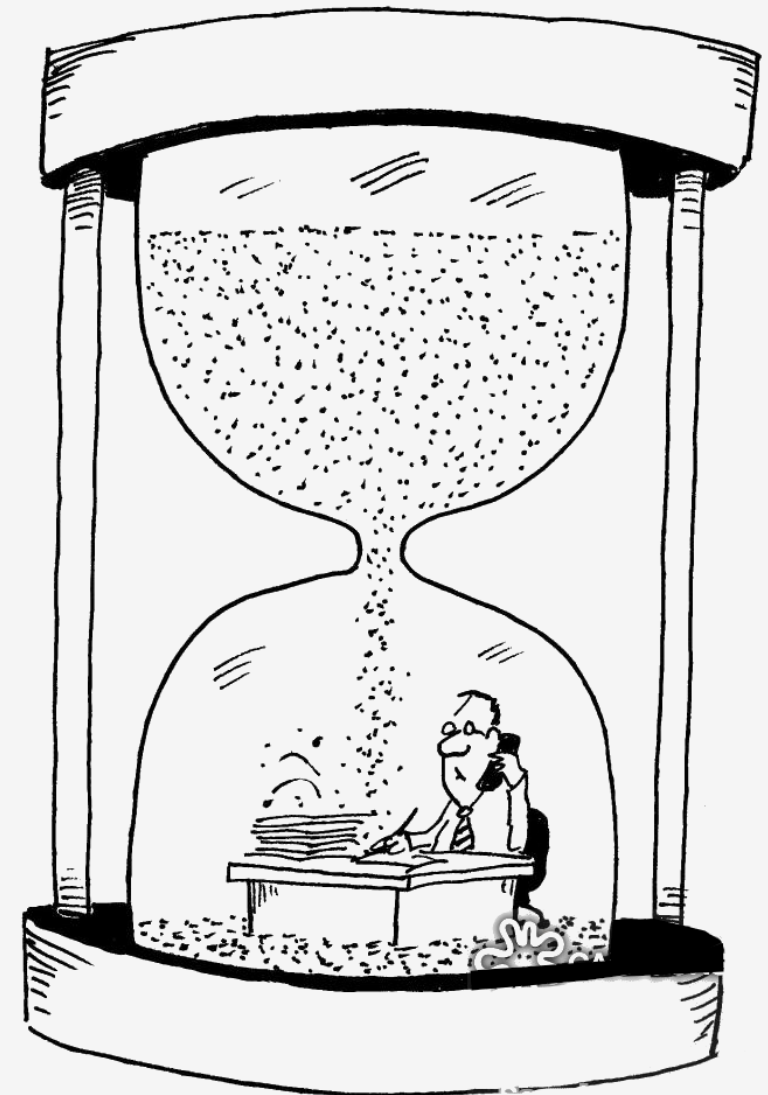
- ❑ Python
  - ❑ Google Colab or Jupyter notebook



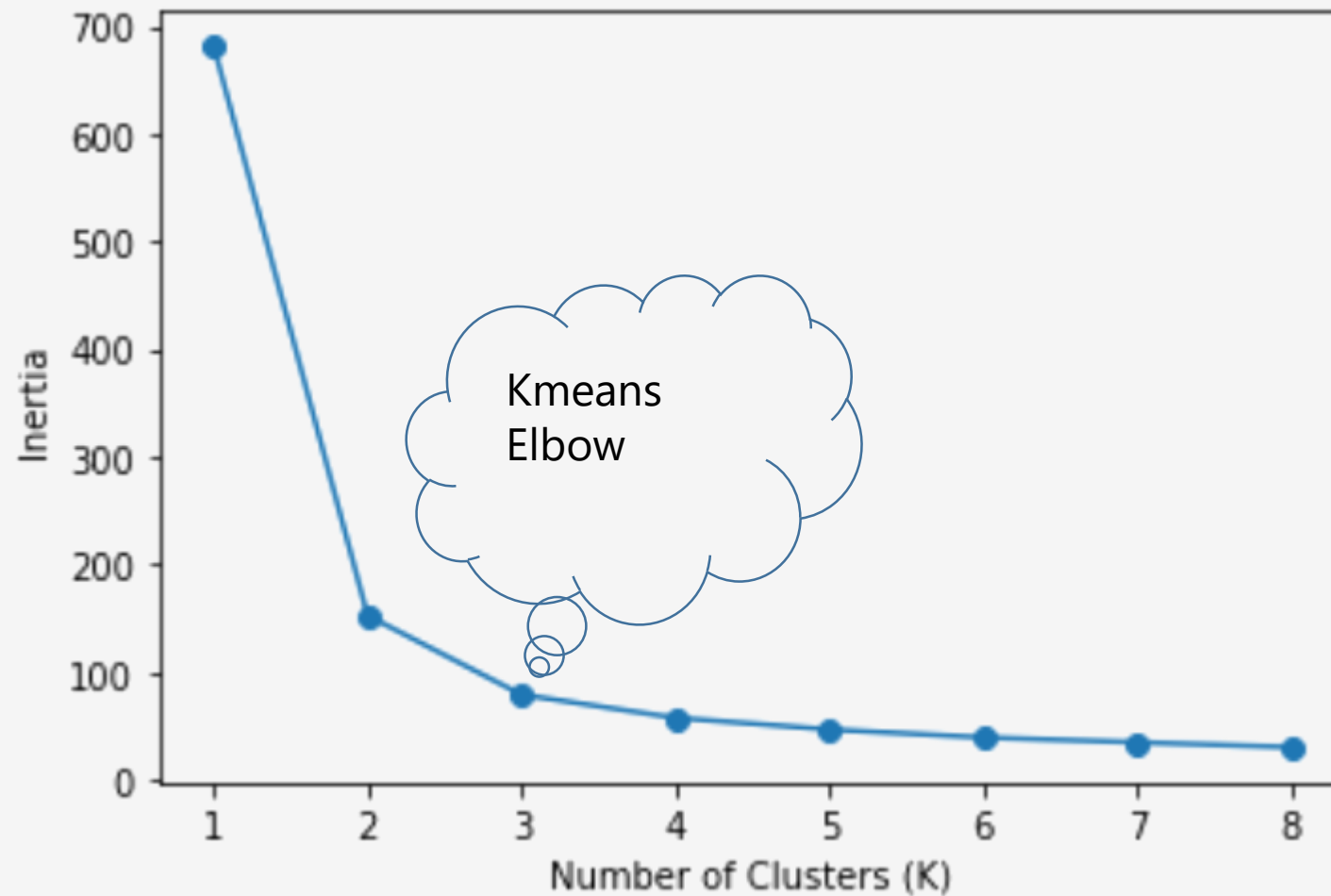
# Mid Term Projects Reminder

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Mid Tern Project Submission Date – Nov 29<sup>th</sup>, 2019



# Elbow Method for optimal value of k in KMeans



**Credit:** Maggie

# Homework

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## Question

Find Eigenvalues and Eigenvectors of the matrix

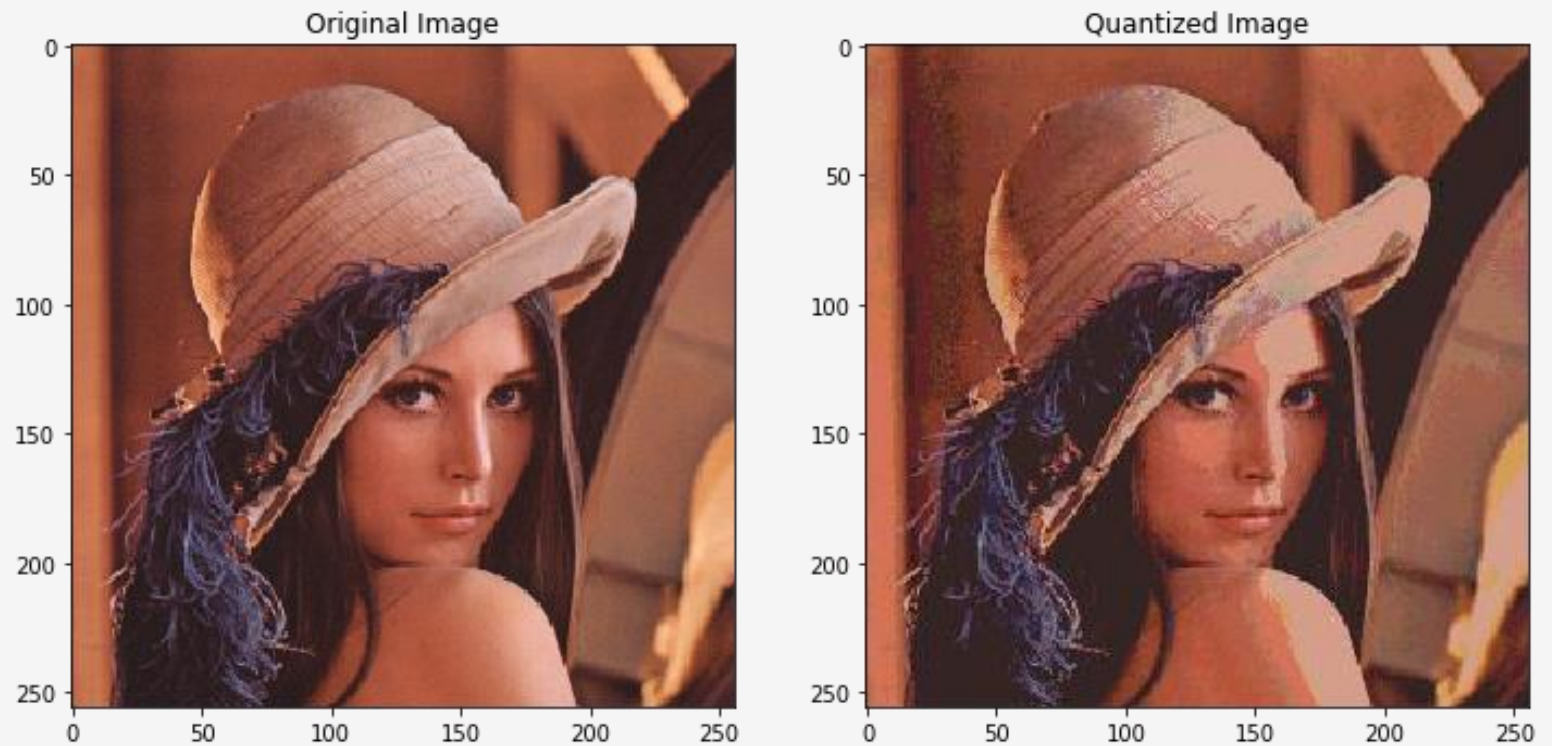
$$\begin{bmatrix} 0 & 1 & 2 \\ 3 & 4 & 5 \\ 6 & 7 & 8 \end{bmatrix}$$

## Answer

Eigenvalue	Eigenvector
$\lambda = 0$	$\begin{pmatrix} 1 \\ -2 \\ 1 \end{pmatrix}$
$\lambda = 3(2 + \sqrt{6})$	$\begin{pmatrix} 6\sqrt{6} - 11 \\ 4 + 3\sqrt{6} \\ 19 \end{pmatrix}$
$\lambda = 3(2 - \sqrt{6})$	$\begin{pmatrix} -11 - 6\sqrt{6} \\ -3\sqrt{6} + 4 \\ 19 \end{pmatrix}$

# Image Color Quantization

- ❑ Color Quantization is a lossy compression technique achieved by compressing a range of values to a single quantum value.
- ❑ Color quantization reduces the number of colors used in an image;
- ❑ Most bitmap editors and many operating systems have built-in support for color quantization.
- ❑ Median cut algorithm is typically used in practice.
- ❑ In this example, we use Kmeans algorithm for familiarity reasons.

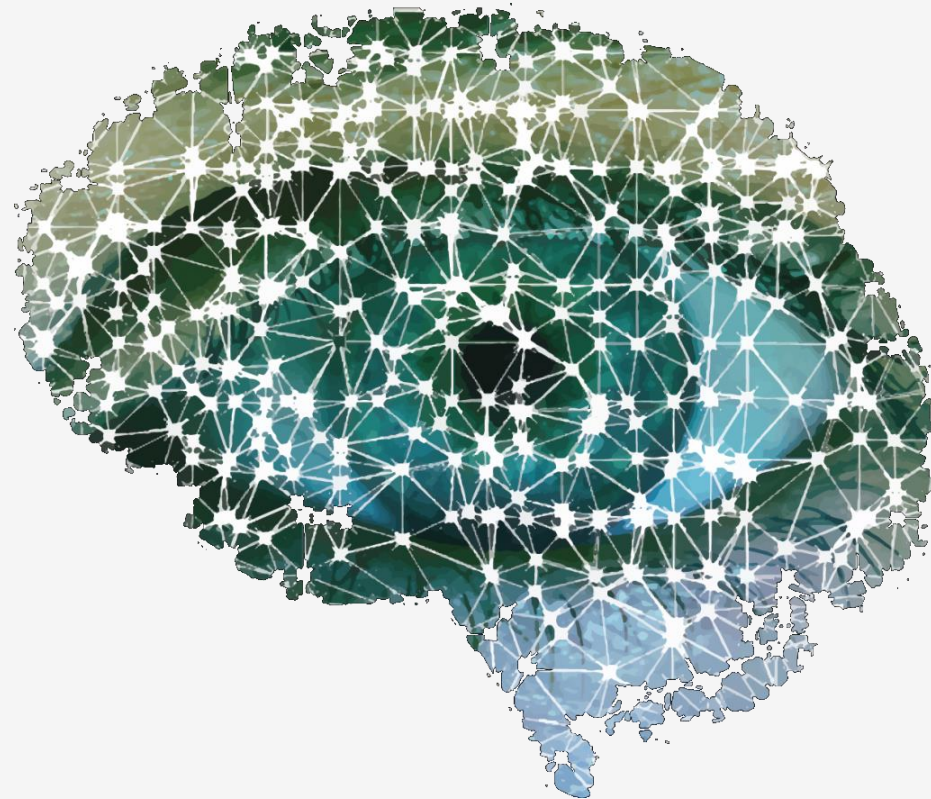


Notebook: <https://colab.research.google.com/drive/1VSVmCMj-HPAMRo4t11bVqcn-l4DE9ecg>



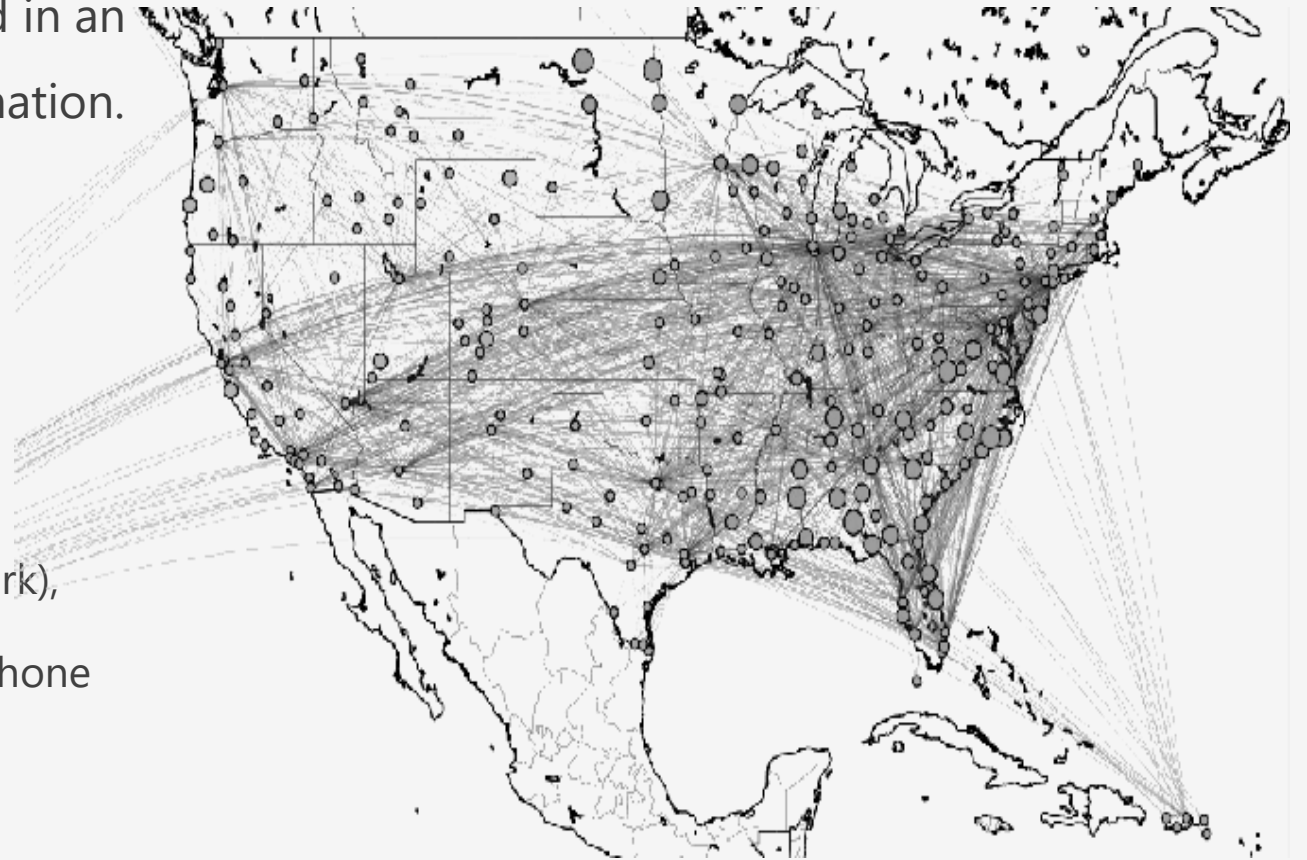
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# Deep Learning Introduction



# What are networks?

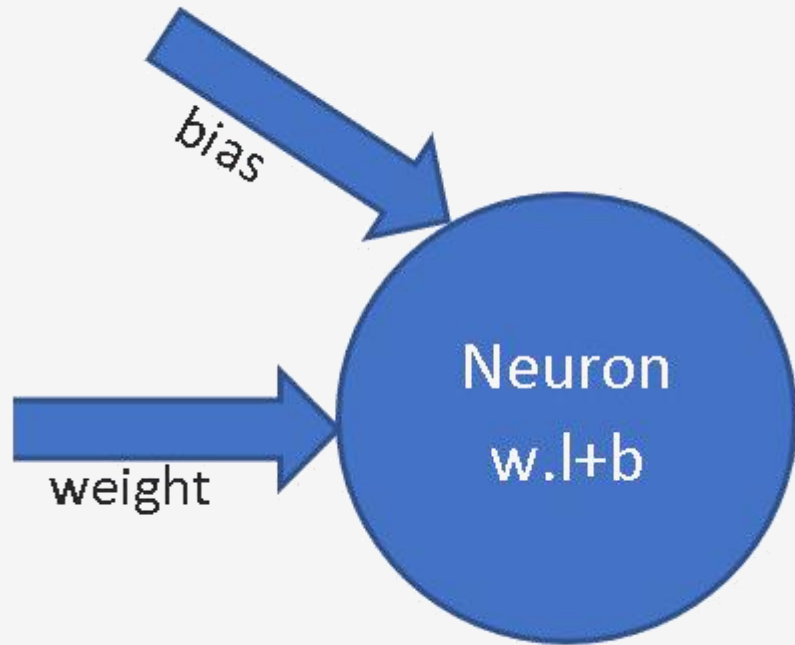
- ❑ Network is a group of entities interconnected in an arbitrarily ordered fashion to exchange information.
- ❑ Popular examples
  - ❑ Social networks (e.g. Facebook),
  - ❑ Professional networks (e.g. LinkedIn),
  - ❑ Road, Railway and port networks,
  - ❑ Electrical circuit networks (e.g. parasitic network),
  - ❑ Communication networks (e.g. switched telephone network),
  - ❑ Computer networks (e.g. the internet)



# What are neural networks?

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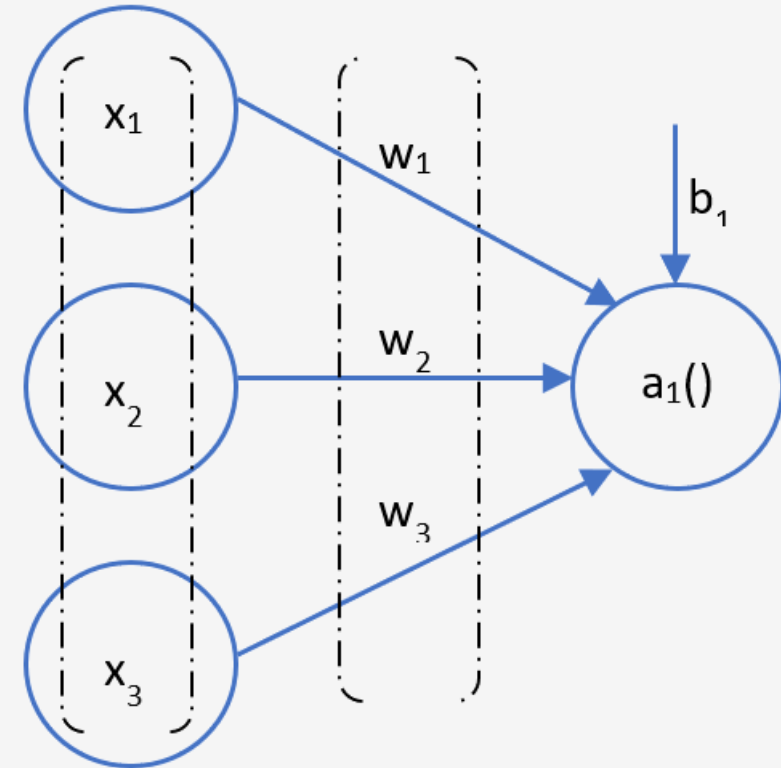
❑ When Entities in a network is represented with neurons, it is called neural network.



- ❑ Neuron is a fancy term for a computational node that performs set of mathematical operations with an offset on the information it is fed as input.
- ❑ Arbitrary combination of multiplication and addition has the ability to generate any expression, So  $\mathbf{a=w*x+b}$  is used for generality.

# What is deep learning?

- ❑ Hypothesis is created using a compute graph aka neural network.
- ❑ Relies on the structure of network to compute and propagate values forward.
- ❑ Optimization is non-trivial, since error gradient isn't radially available to find next step.

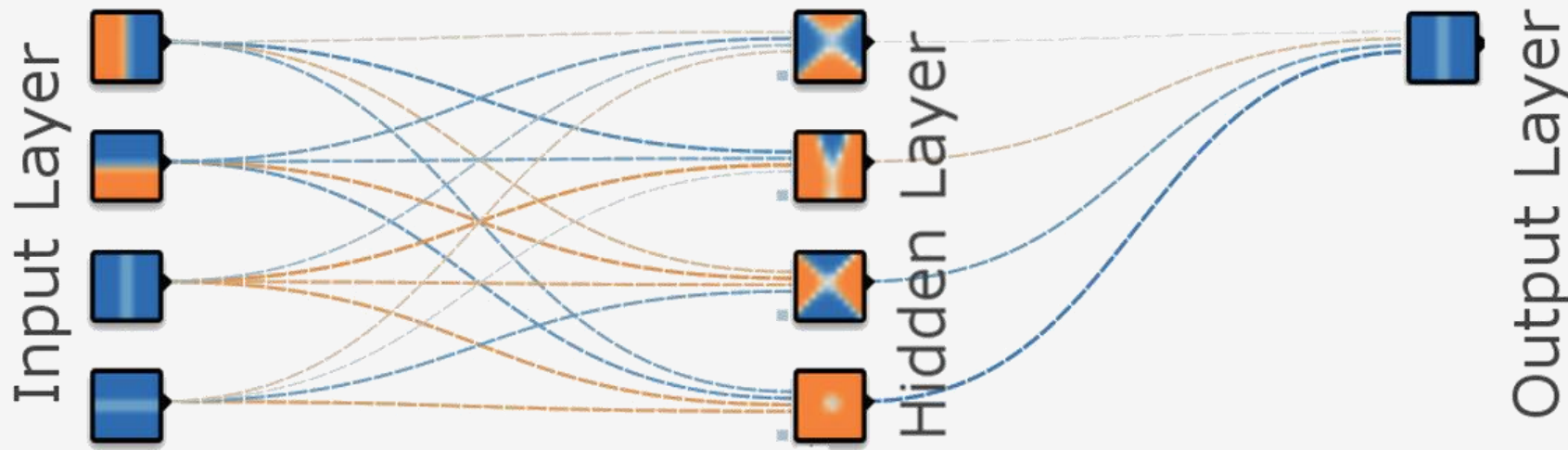


$$a = w * x + b$$

# Deep Learning Terminology?

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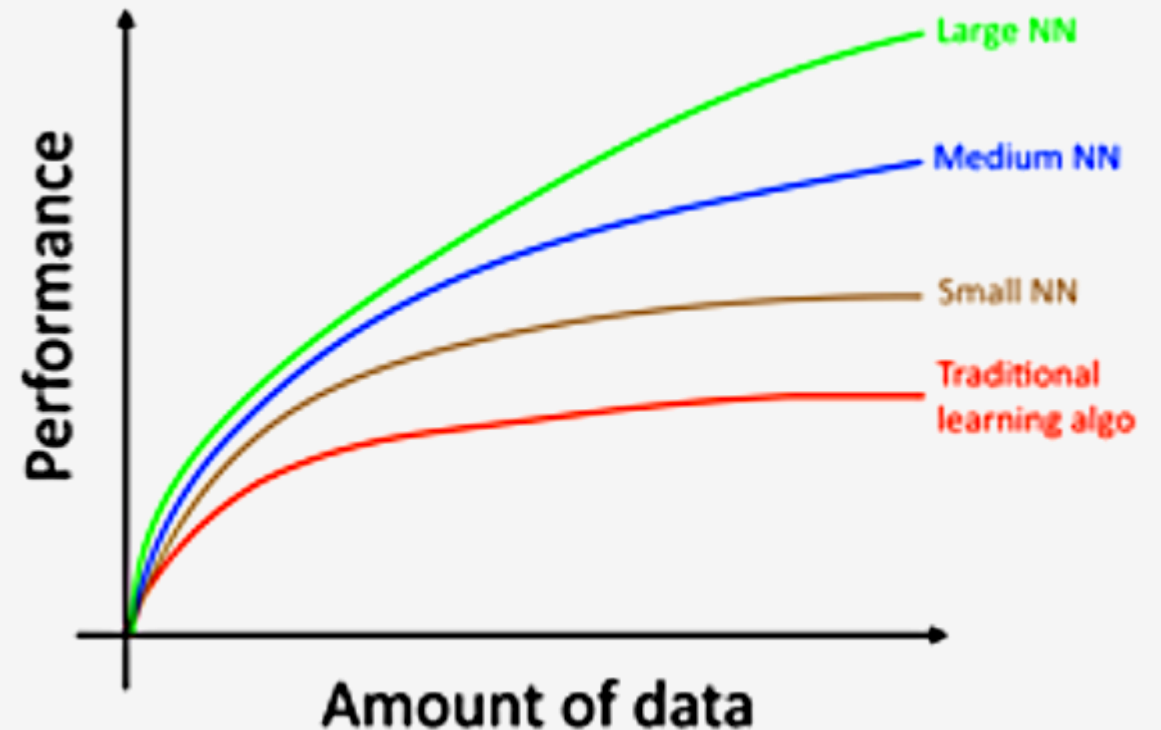
- ❑ Neurons
- ❑ Weights & Biases
- ❑ Model Parameters
- ❑ Perceptron
- ❑ Compute Graph
- ❑ Layer aka operator
- ❑ Activation functions
- ❑ Forward Propagation
- ❑ Backward Propagation
- ❑ Error Gradient
- ❑ Vanishing Gradients
- ❑ Exploding Gradients



# Why Deep Learning?

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- ❑ Why Now?
  - ❑ Data
  - ❑ Compute Power
  - ❑ Algorithms
- ❑ Performance superiority



# Types and Application

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## ☐ Neural Network Types

- ☐ Feed Forward

- ☐ Recurrent Neural Network

- ☐ Self Organizing Neural Network

## ☐ Applications

- ☐ Prediction

- ☐ Classification

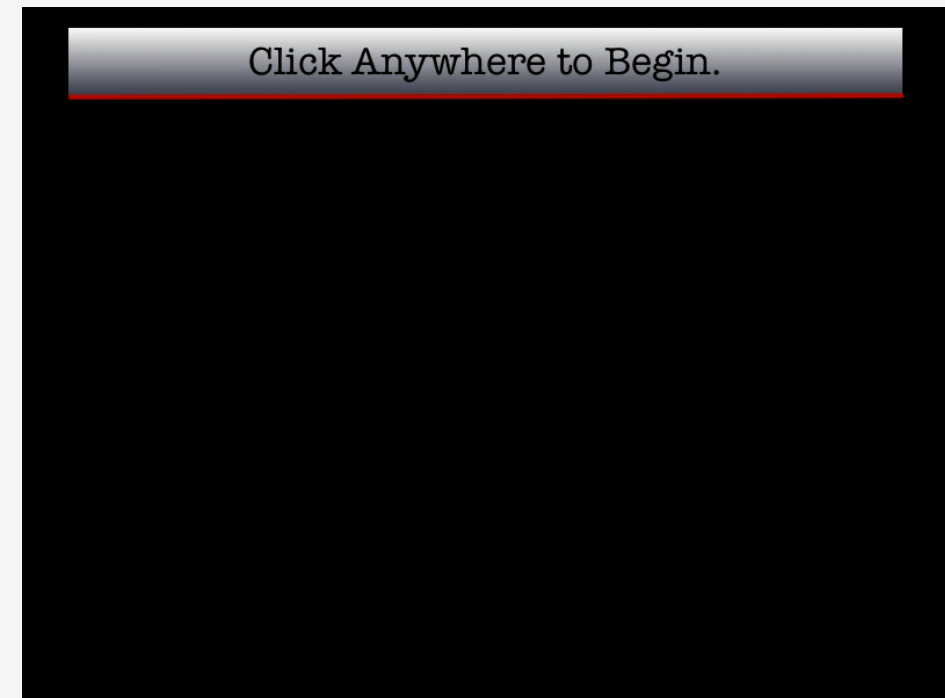
- ☐ Clustering

- ☐ Optimization

# Human Vision vs Deep Vision

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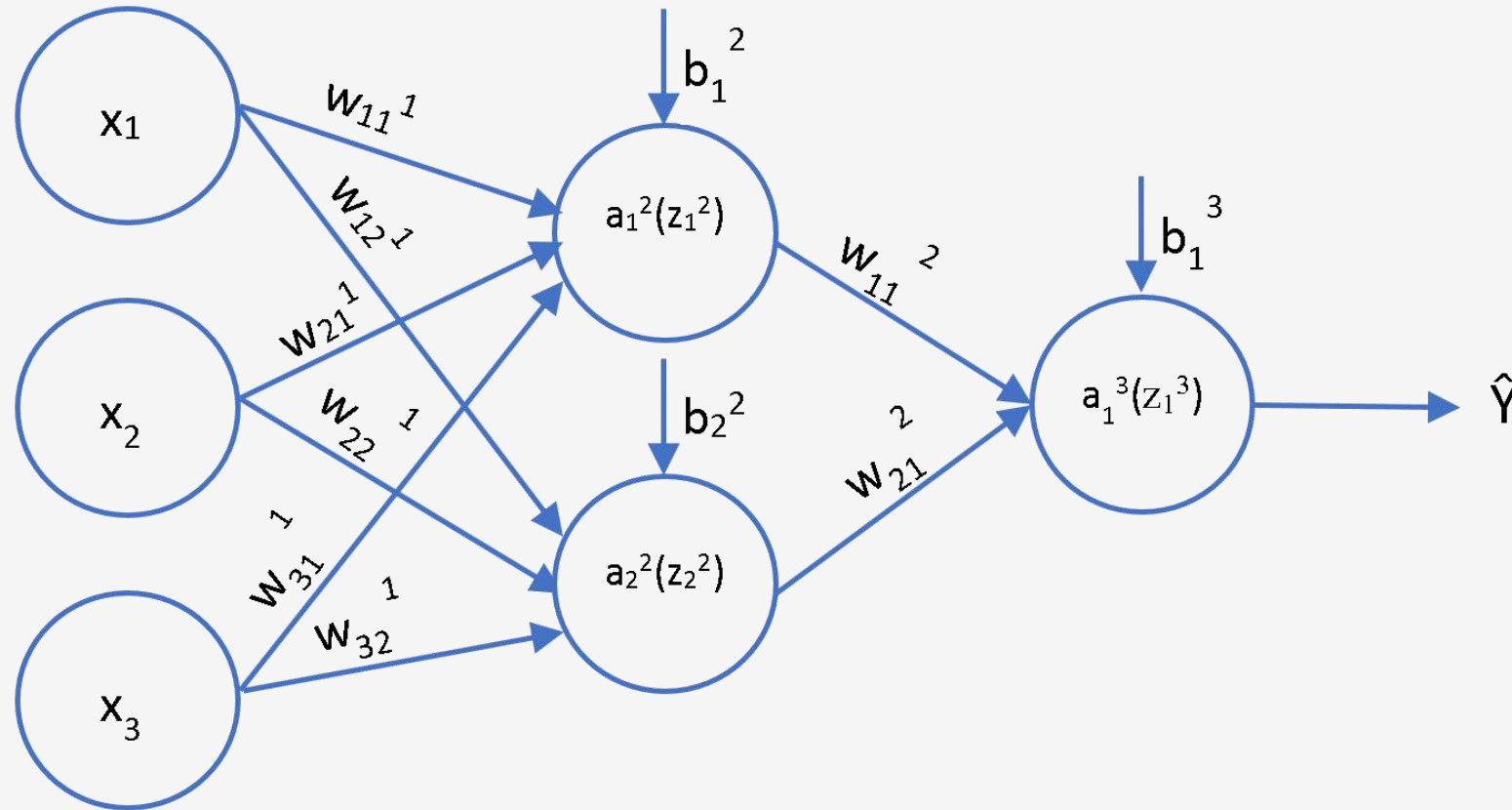
- ❑ Video on the right is an effort to convince you that deep learning vision is superior to human vision in many ways.
- ❑ Human vision system is continuously under training.





# Linear Feed Forward Neural Network

Compute Function Y for the network shown in the figure.



# Linear Feed Forward Neural Network

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Compute Function Y for the network shown in the figure.

$$a^1 = x$$

$$z^2 = (w^1)^T \cdot a^1 + b^1$$

$$a^2 = \text{sigmoid}(z^2)$$

$$z^3 = (w^2)^T \cdot a^2 + b^2$$

$$a^3 = \text{sigmoid}(z^3)$$

$$\hat{Y} = a^3$$

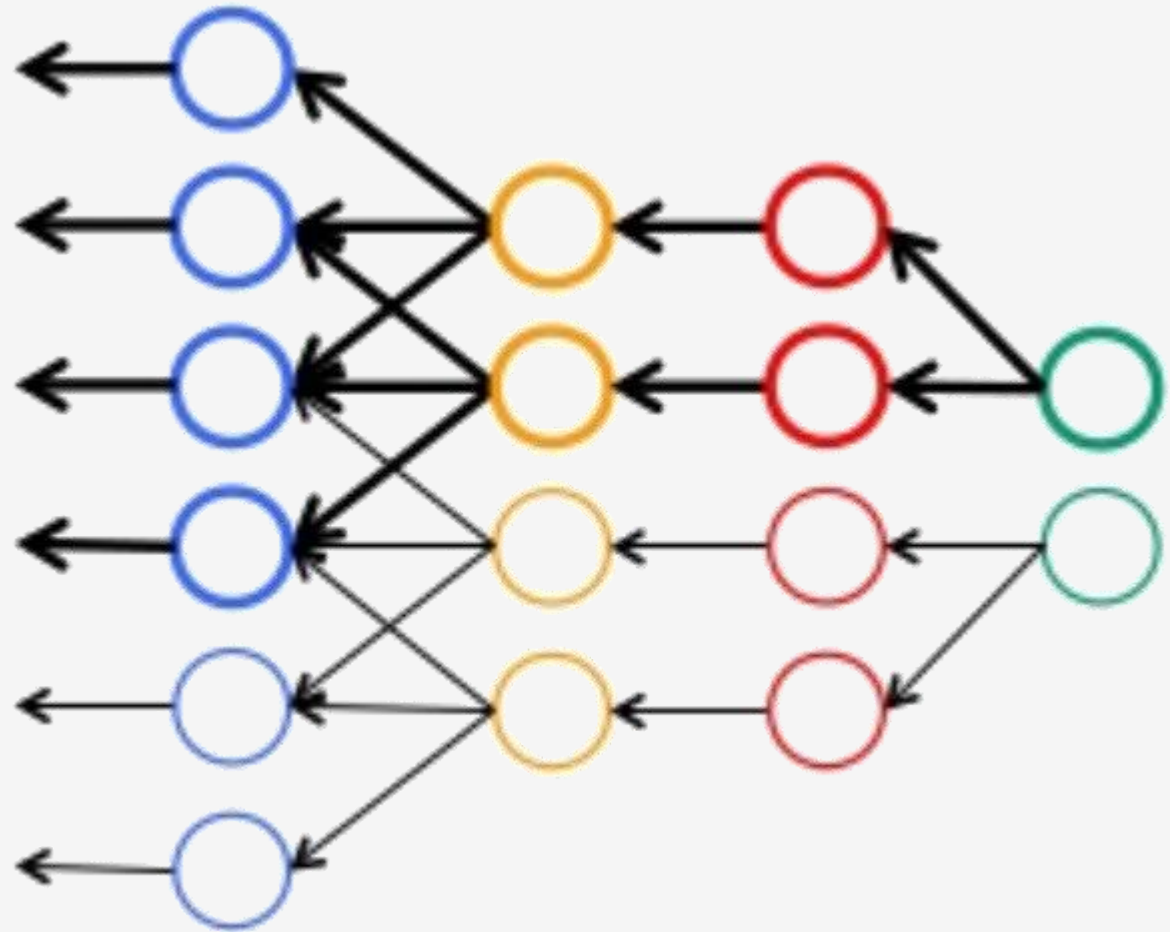
Here, superscript T denotes transpose operator and sigmoid is given as:  $\frac{1}{1+e^{-z}}$

<https://playground.tensorflow.org/>

# Backward Propagation

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- ❑ Chain Rule
- ❑ Computing Error Gradients
- ❑ Backprop algorithm
- ❑ Vanishing Gradients
- ❑ Exploding Gradients
- ❑ Workarounds
  - ❑ Gradient Clipping
  - ❑ Regularization





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