

# NEURAL NETWORKS

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## **PROGRESSING IN YOUR DATA SCIENCE CAREER**

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# **LEARNING OBJECTIVES**

- Understand various types of neural networks
- Applications of neural networks
- Apply a neural network model for regression
- Apply a neural network model for classification

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**COURSE**

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**PRE-WORK**

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## **PRE-WORK REVIEW**

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- Understand Logistic Regression and link functions
- Be familiar with training and testing classifiers and regressors

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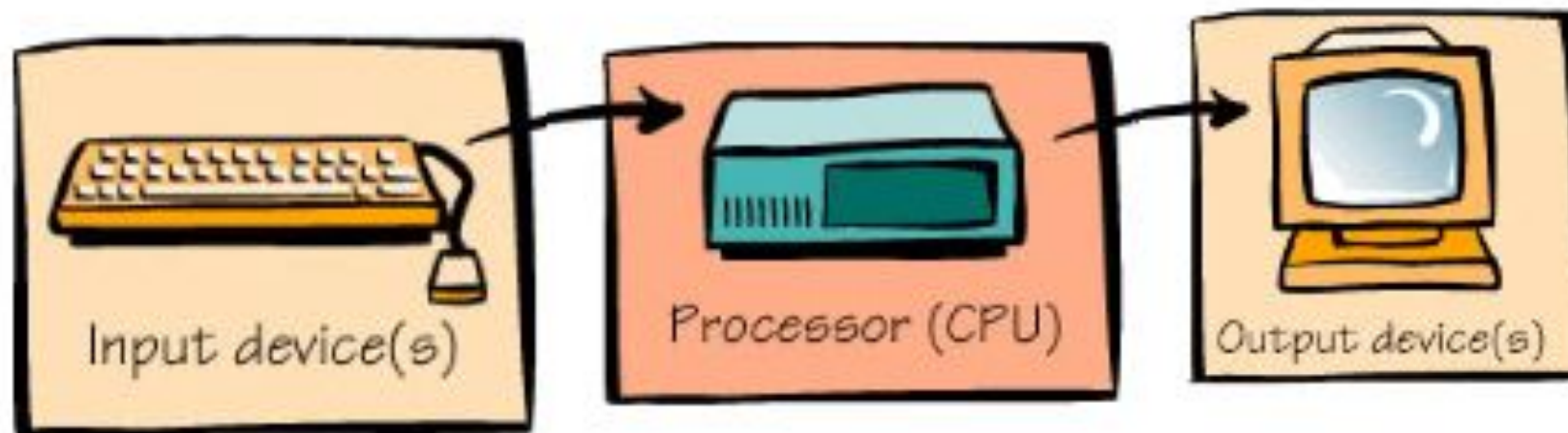
**OPENING**

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# ARTIFICIAL NEURAL NETWORKS

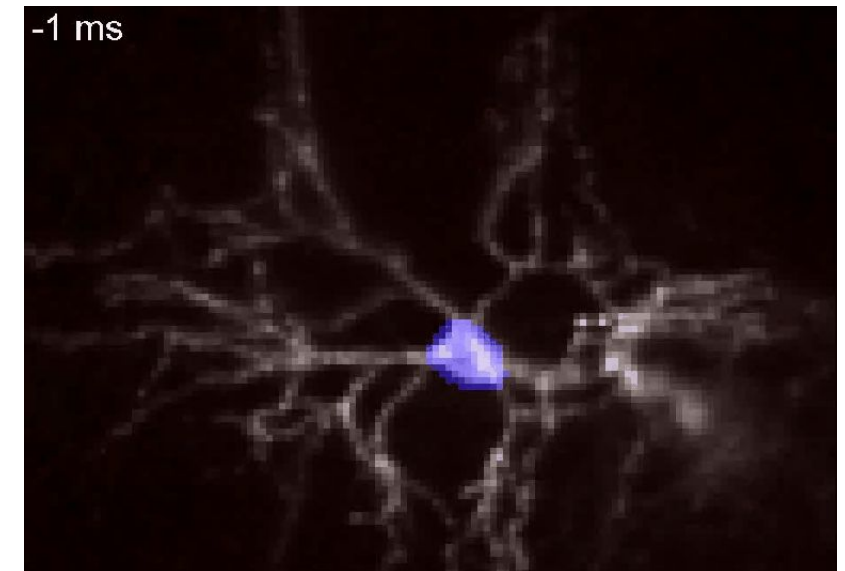
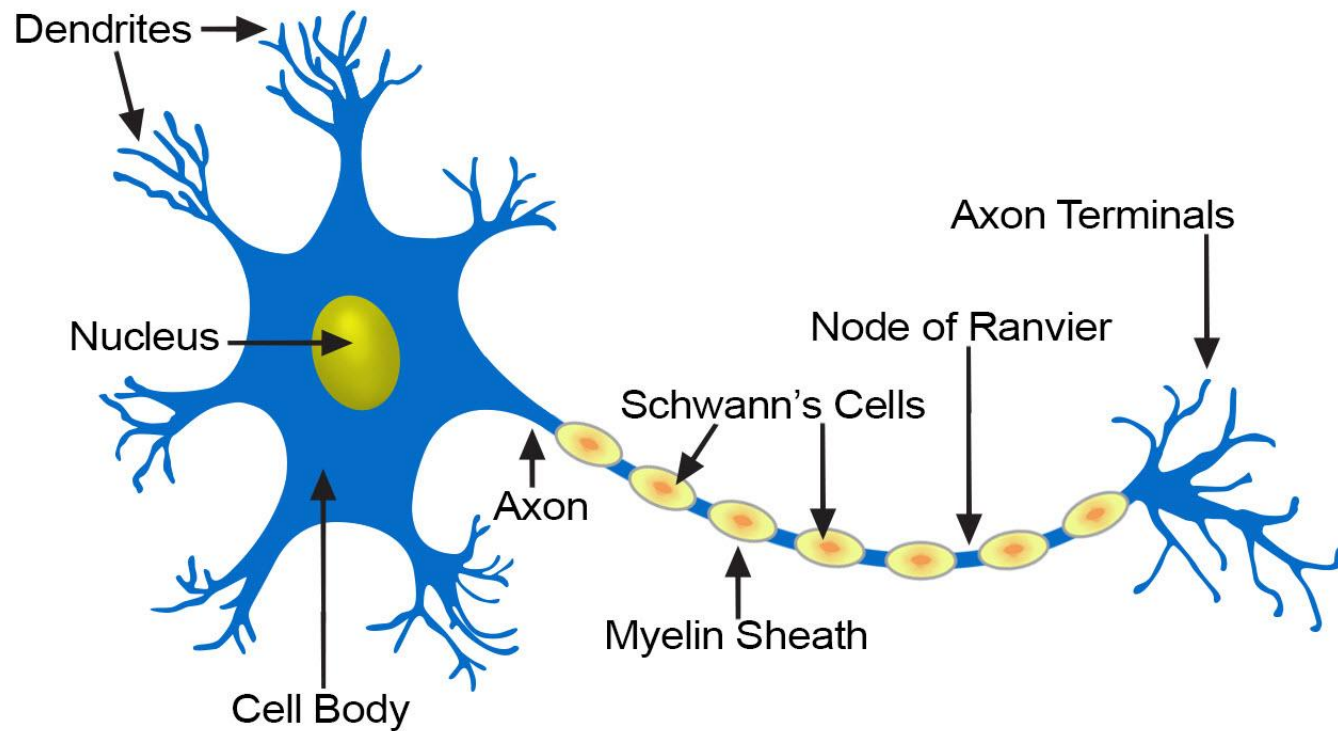
# WHAT COMPUTERS DO

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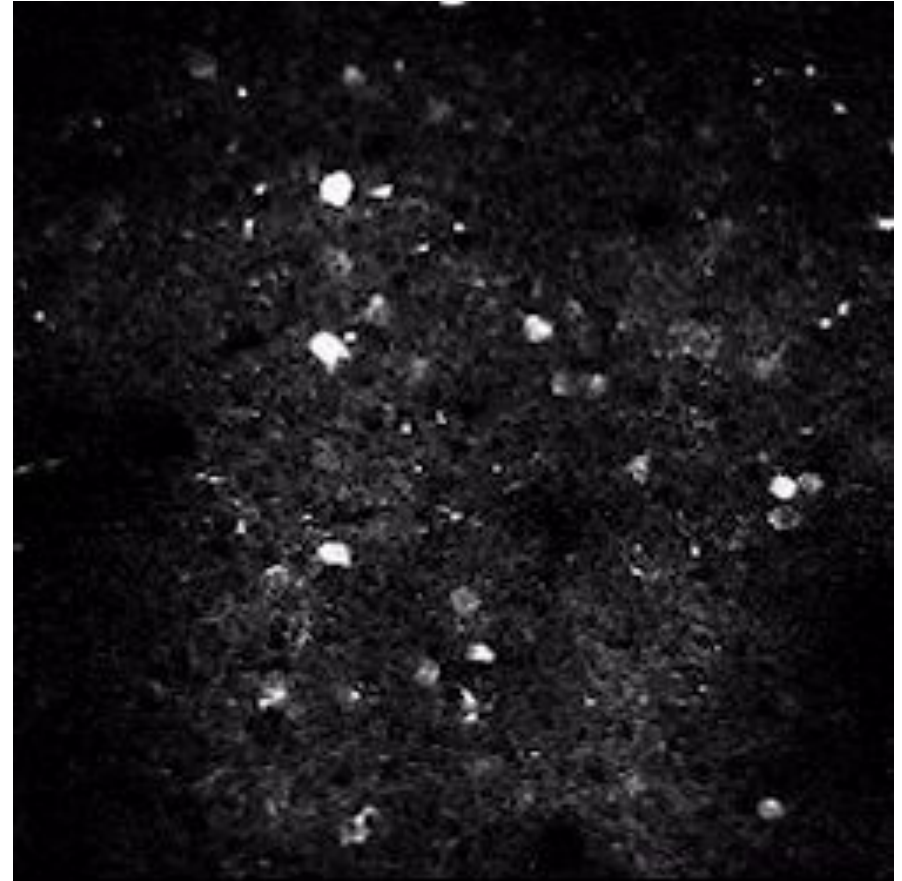
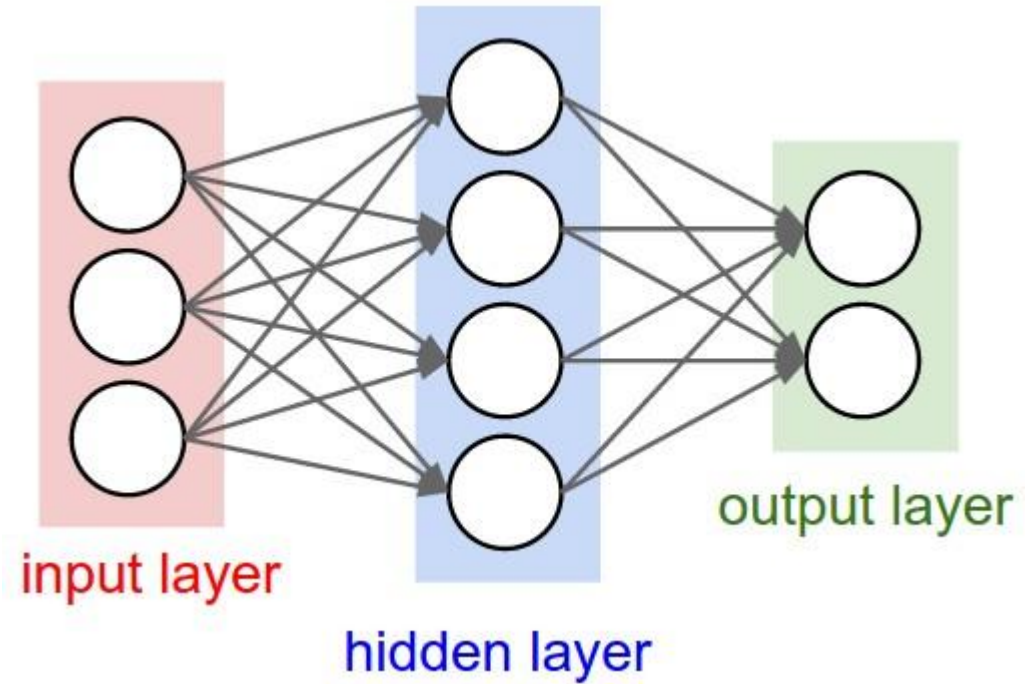


# ARTIFICIAL NEURAL NETWORKS

- The idea is to emulate the behavior of [neurons](#)



# ARTIFICIAL NEURAL NETWORKS

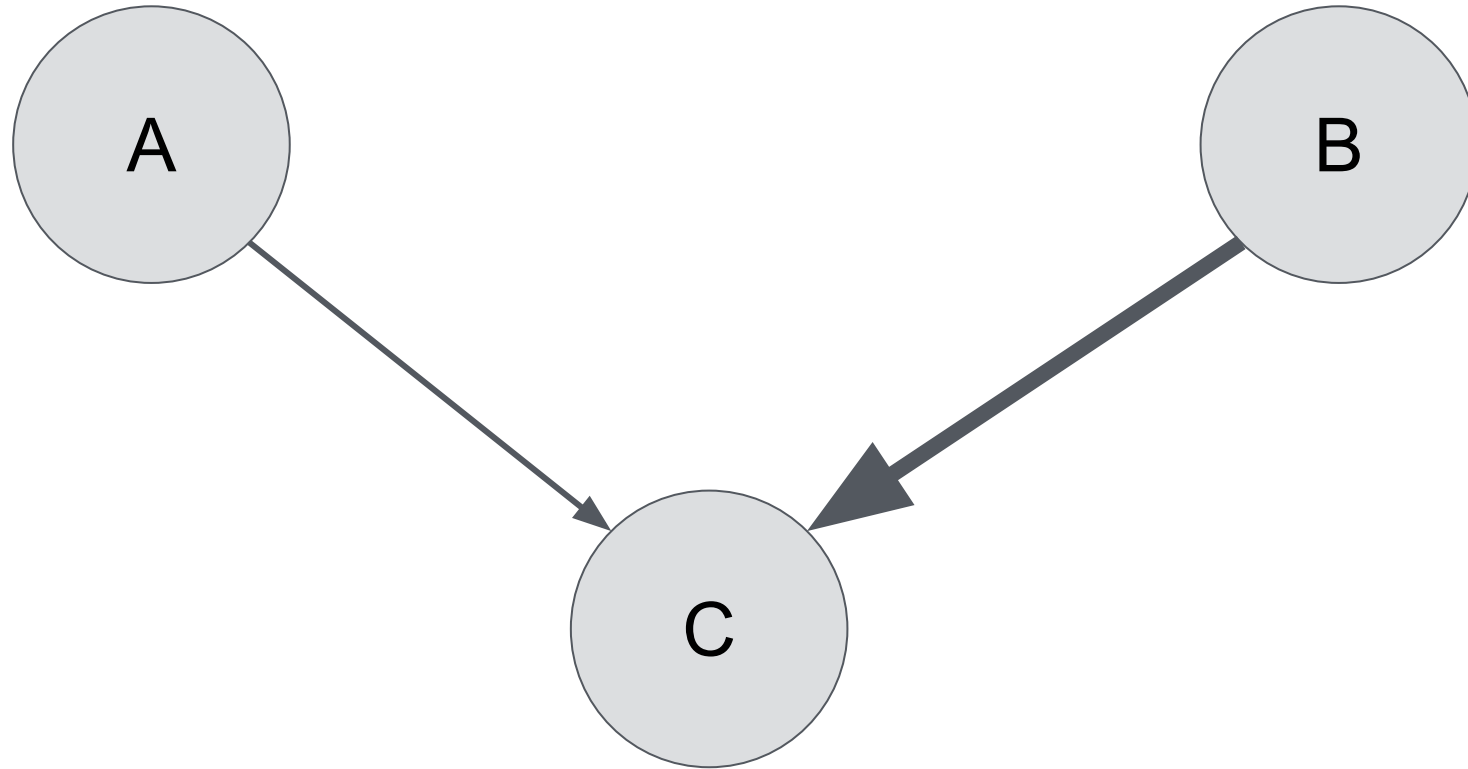




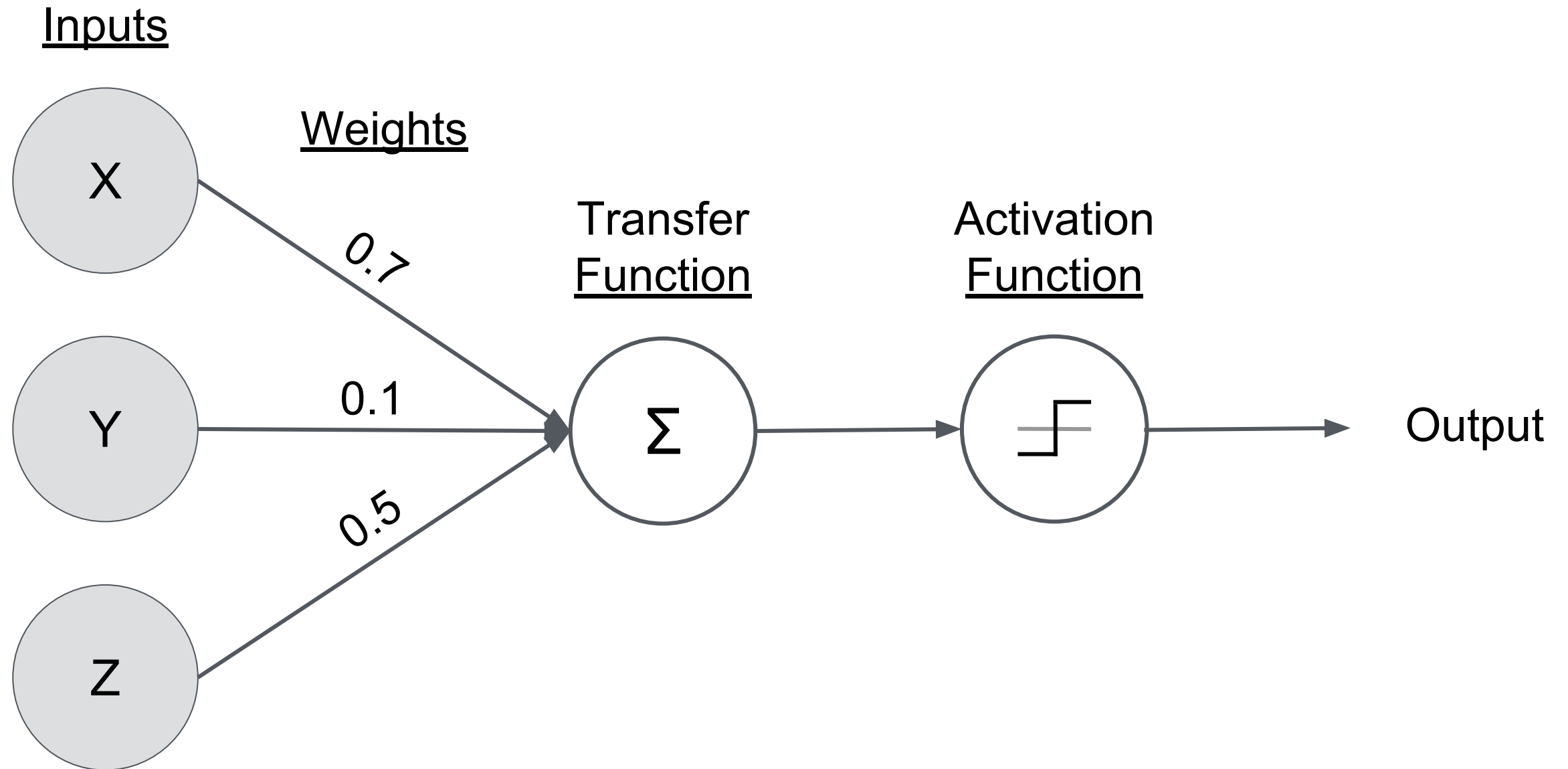
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# ARTIFICIAL NEURAL NETWORKS

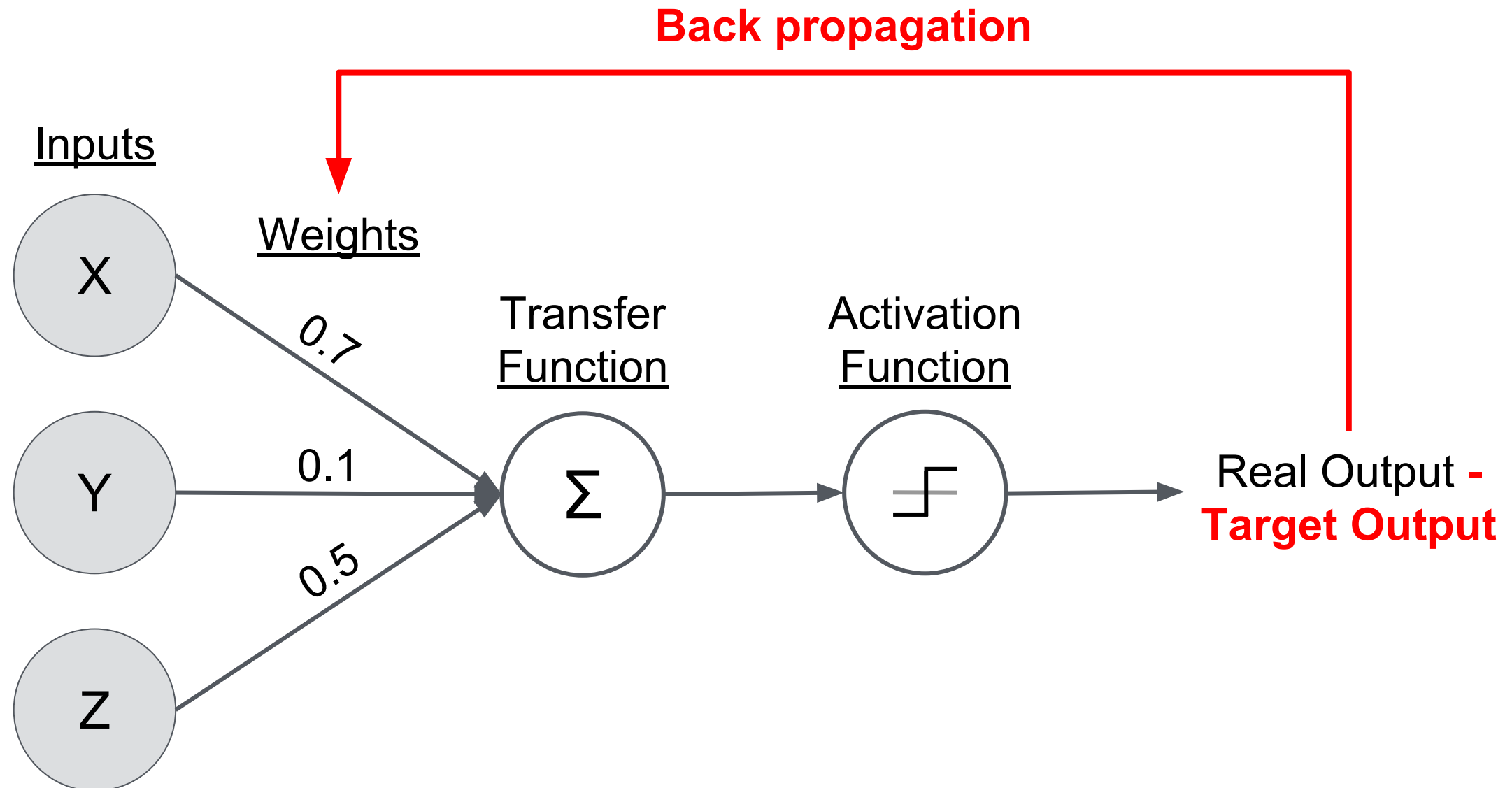
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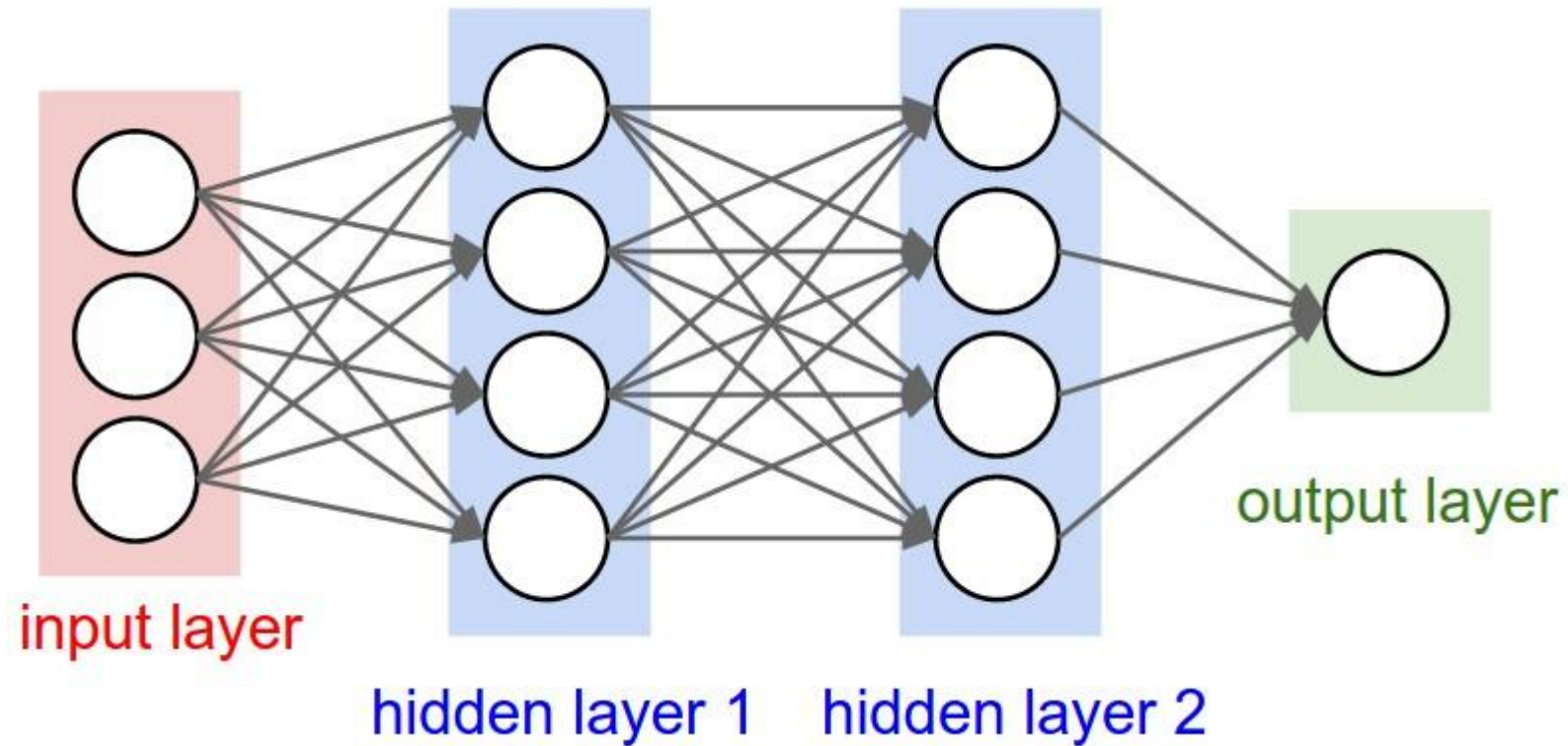
# ARTIFICIAL NEURAL NETWORKS



# ARTIFICIAL NEURAL NETWORKS

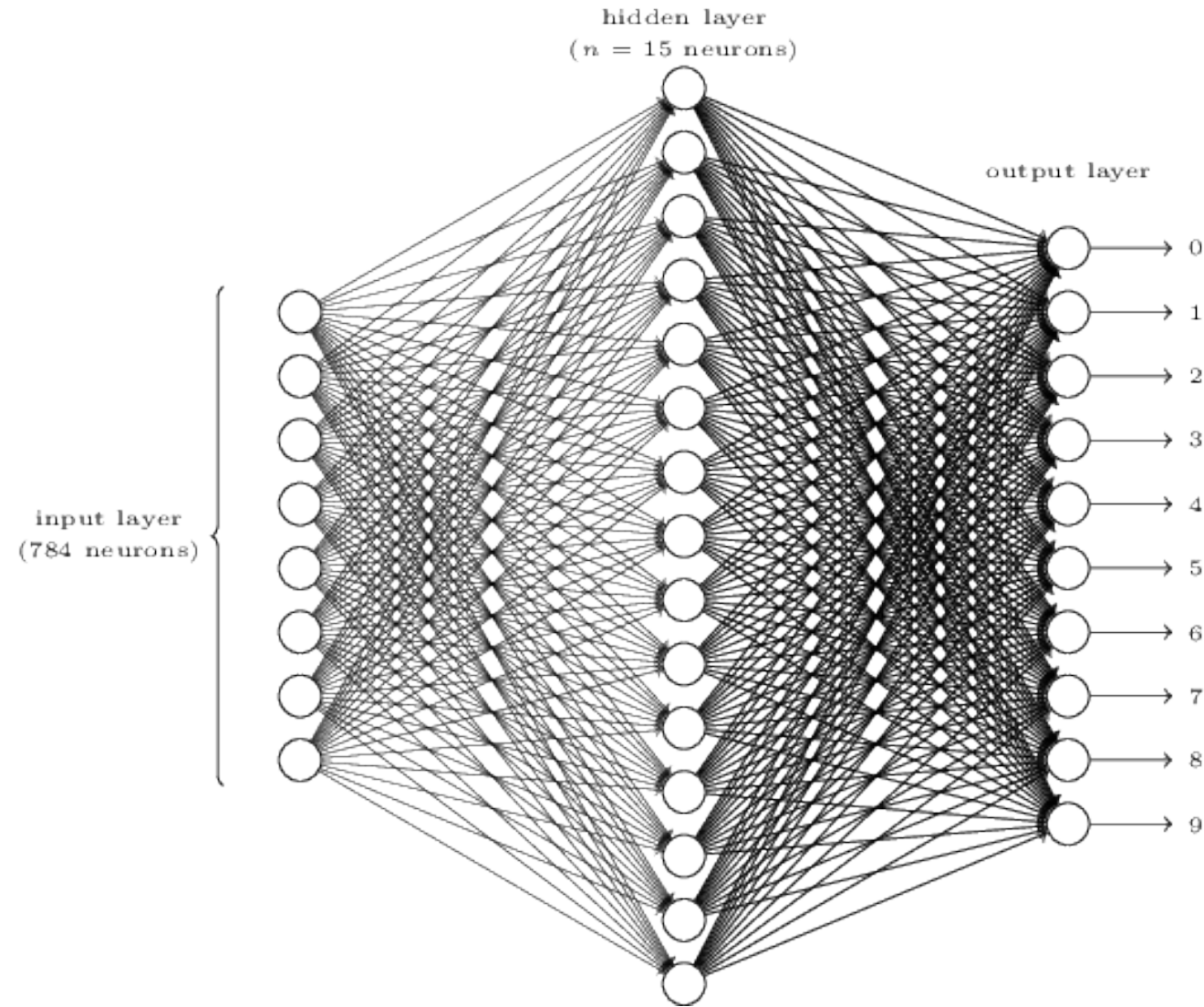


# ARTIFICIAL NEURAL NETWORKS



[http://cs231n.github.io/assets/nn1/neural\\_net2.jpeg](http://cs231n.github.io/assets/nn1/neural_net2.jpeg)

# ARTIFICIAL NEURAL NETWORKS



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# ARTIFICIAL NEURAL NETWORKS

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General use case	Industry
<b>Sound</b>	
Voice recognition	UX/UI, Automotive, Security, IoT
Voice search	Handset maker, Telecoms
Sentiment analysis	CRM
Flaw detection (engine noise)	Automotive, Aviation
Fraud detection (latent audio artifacts)	Finance, Credit Cards
<b>Time Series</b>	
Log analysis/Risk detection	Data centers, Security, Finance
Enterprise resource planning	Manufacturing, Auto., Supply chain
Predictive analysis using sensor data	IoT, Smart home, Hardware manufact.
Business and Economic analytics	Finance, Accounting, Government
Recommendation engine	E-commerce, Media, Social Networks
<b>Text</b>	
Sentiment Analysis	CRM, Social media, Reputation mgt.
Augmented search, Theme detection	Finance
Threat detection	Social media, Govt.
Fraud detection	Insurance, Finance
<b>Image</b>	
Facial recognition	
Image search	Social media
Machine vision	Automotive, aviation
Photo clustering	Telecom, Handset makers
<b>Video</b>	
Motion detection	Gaming, UX, UI
Real-time threat detection	Security, Airports

[https://deeplearning4j.org/use\\_cases.html](https://deeplearning4j.org/use_cases.html)

# ARTIFICIAL NEURAL NETWORKS

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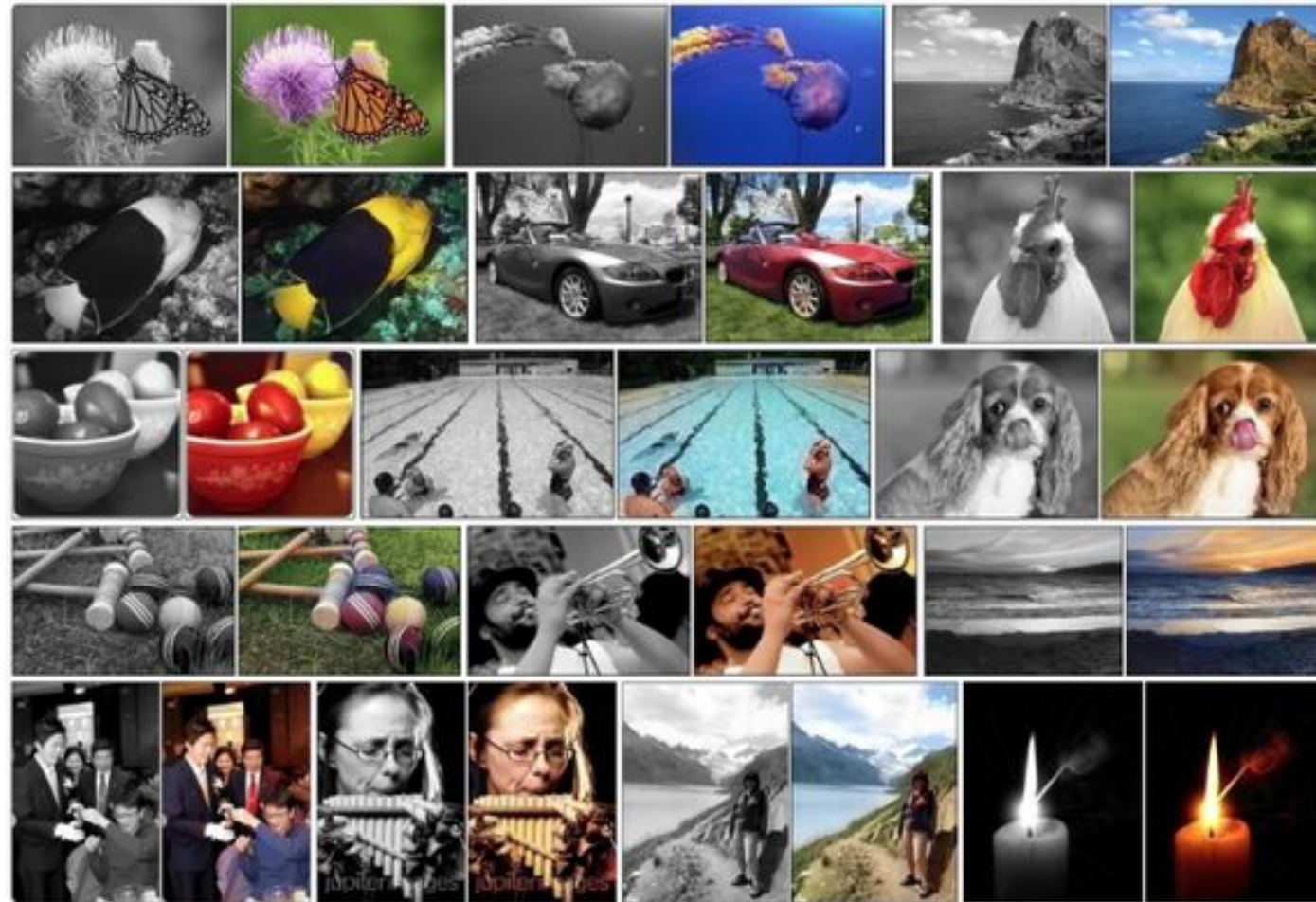
- [Medical images](#)





# ARTIFICIAL NEURAL NETWORKS

## ► Automatic Colorization with CNN



<http://tinyclouds.org/colorize/>



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**GUIDED PRACTICE**

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# NEURAL NETWORKS IN PYTHON

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# NN IN PYTHON

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- There are many NN libraries for python and other languages
- Python
  - Theano
  - Keras
  - Lasagne
  - TensorFlow
  - Py-torch
  - Scikit Learn support for NN coming soon
- Some of these libraries utilize **GPUs** for (much) faster training

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# NN IN PYTHON

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- Next, we'll look at some examples in Keras
  - Regression
  - Classification

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# ACTIVITY: KNOWLEDGE CHECK

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## ANSWER THE FOLLOWING QUESTIONS



### EXERCISE

1. Let's practice using [neural networks for classification](#). For each of the four datasets, experiment with the number of layers and neurons to find the best model
2. Also take a look at this [visualization](#)

## DELIVERABLE

Answers to the above questions

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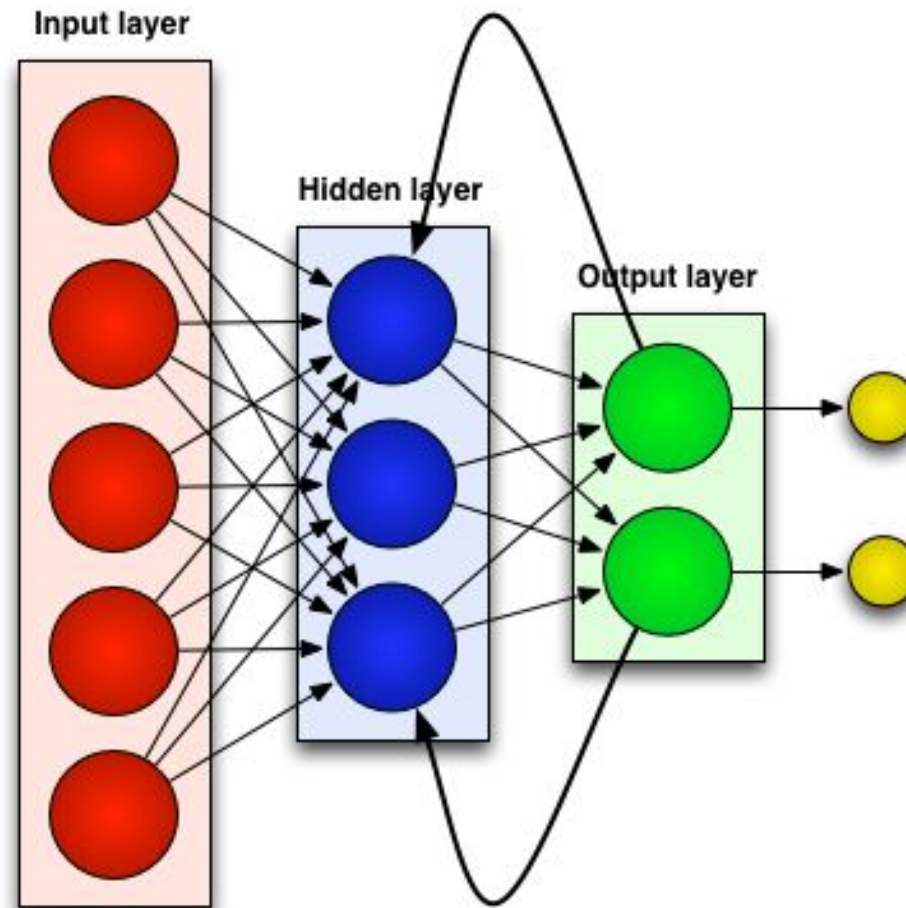
**RECURRENT NN**

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# RECURRENT NEURAL NETWORKS

# RECURRENT NEURAL NETWORKS

- Recurrent Neural Networks contain loops ([source](#))



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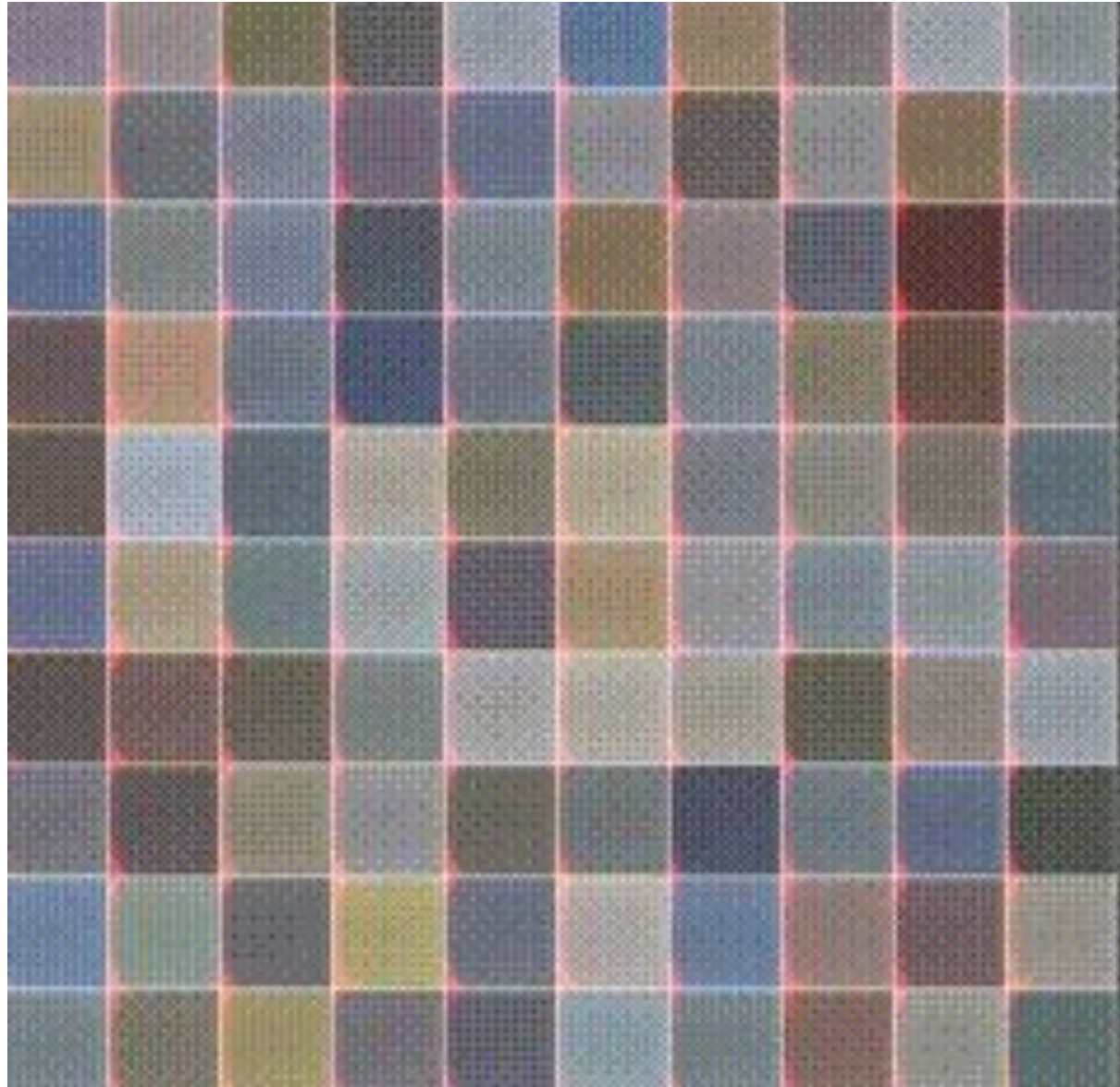
# RECURRENT NEURAL NETWORKS

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- Recurrent Neural Networks contain loops
- This implements feedback and gives neural networks “memory” or context
- Particularly good for predicting sequences, translating text, recognizing objects in images, speech translation
- Commonly referred to as **deep learning**, involving both feature extraction and modeling
- [Nice intro here](#)

# RECURRENT NEURAL NETWORKS

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**CONCLUSION**

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# TOPIC REVIEW

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## **CONCLUSION: Neural Networks**

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### Pros:

- Flexible
- Good for a variety of tasks
- Good for many types of data

### Cons:

- Can require a lot of data
- Training may be slow
- Many parameters to tune
- Many layer types and activations
- Black Box model

**COURSE**

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**BEFORE NEXT CLASS**

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## **BEFORE NEXT CLASS**

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# **DUE DATE**

- Project: Final Project, Part 5!!

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**LESSON**

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**Q & A**

## **LESSON**

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# **EXIT TICKET**

**DON'T FORGET TO FILL OUT YOUR EXIT TICKET**