



# Machine Learning for Everyone



By Aaron Ma



# I am Aaron Ma



- 11 years old
- Youngest TensorFlow contributor
- Robotics, machine learning, algorithms, self driving cars
- AI, ML, TensorFlow, Python, C++, HTML, CSS, TypeScript, JavaScript, Angular, Swift, ROS, AWS, Google Cloud, and more
- Tremendous passion in Computer Science
- Hard-core software and hardware
- Youngest graduated from Udacity Self Driving Car Engineer Nanodegree and more!
- Twitter **@aaronhma**



VERIFIED CERTIFICATE OF COMPLETION

June 18, 2019



Aaron Ma

Has successfully completed the

Self-Driving Car Engineer

NANO DEGREE PROGRAM



Sebastian Thrun  
Founder, Udacity

Co-Created with

Mercedes Nvidia Uber ATG Didi BMW McLaren

Udacity has confirmed the participation of this individual in this program.  
Confirm program completion at [confirm.udacity.com/C597KLNL](https://confirm.udacity.com/C597KLNL).

@aaronhma



IBM

07/06/2019

**Aaron Ma**

has successfully completed the online, non-credit Specialization

## IBM Data Science Professional Certificate

In this Professional Certificate learners developed and honed hands-on skills in Data Science and Machine Learning. Learners started with an orientation of Data Science and its Methodology, became familiar with a variety of machine learning techniques, learned Python and SQL, performed Data Visualization and Analysis, and created Machine Learning models. In the process they completed several labs and assignments on the cloud including a Capstone Project at the end to apply and demonstrate their knowledge and skills.

Joseph Stanisicaglio  
Senior Data Scientist  
IBM

Peihong Lin  
Data Scientist, IBM

Roy Achup  
AI & Data Science Program Director,  
IBM Skills Network

Jasel A.  
Sr. Data Scientist  
IBM

Verify this certificate at:  
[coursera.org/verify/specification/N22YJCFB0D5](https://coursera.org/verify/specification/N22YJCFB0D5)

deeplearning.ai

04/19/2019

**Aaron Ma**

has successfully completed the online, non-credit Specialization

## Deep Learning

The Deep Learning Specialization is designed to prepare learners to participate in the development of cutting-edge AI technology, and to understand the capability, the challenges, and the consequences of the rise of deep learning. Through five interconnected courses, learners will develop a practical knowledge of the hottest AI algorithms, machine learning from its foundations (neural networks) to its industry applications (Computer Vision, Natural Language Processing, Speech Recognition, etc.).

Andrew Ng  
Adjunct Professor  
Computer Science

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[coursera.org/verify/specification/Z79UC36UQ4Z](https://coursera.org/verify/specification/Z79UC36UQ4Z)

Google Cloud

07/07/2019

**Aaron Ma**

has successfully completed the online, non-credit Specialization

## Machine Learning with TensorFlow on Google Cloud Platform

This five-course online specialization teaches course participants how to write distributed machine learning models that scale in TensorFlow, scale out the training of these models, and offer high-performance predictions. Also featured is the conversion of raw data to features in a way that allows ML to learn important characteristics of the data, and how to use human insights on the data to also teach the model how to increase the right mix of parameters that yields accurate, generalized models and knowledge of the theory to solve specific types of ML problems. Course participants experimented with end-to-end ML, starting from basic ML-focused theory and progressing into model training, optimization, and productionalization with hands-on labs using Google Cloud Platform.

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[coursera.org/verify/specification/PSENGC3ND3](https://coursera.org/verify/specification/PSENGC3ND3)

Google Cloud

04/04/2019

**Aaron Ma**

has successfully completed the online, non-credit Specialization

## Advanced Machine Learning with TensorFlow on Google Cloud Platform

This specialization focuses on advanced machine learning topics using Google Cloud's machine learning services. It covers applying, deploying, and scaling production ML models to various types of hands-on jobs. This specialization picks up where "Machine Learning on GCP" left off and teaches you how to build scalable, accurate, and production-ready models for structured image data, time-series, and natural language text. If you've taken our "Machine Learning with TensorFlow on GCP" specialization, it's recommended that you take the courses in exactly this order.

Ian Hjelmfelt  
Associate Professor, Aerospace Studies  
Director, Toronto Robotics and Artificial Intelligence Laboratory

Jonathan Kelly  
Associate Professor, Aerospace Studies  
Director, Space & Terrestrial Autonomous Robotic Systems Laboratory

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[coursera.org/verify/specification/VVUJ5TQ48SP5](https://coursera.org/verify/specification/VVUJ5TQ48SP5)

UNIVERSITY OF TORONTO

04/19/2019

**Aaron Ma**

has successfully completed

## Introduction to Self-Driving Cars

An online non-credit course authorized by University of Toronto and offered through Coursera

Steve Wardlaw  
Associate Professor, Aerospace Studies  
Director, Toronto Robotics and Artificial Intelligence Laboratory

Jonathan Kelly  
Associate Professor, Aerospace Studies  
Director, Space & Terrestrial Autonomous Robotic Systems Laboratory

Verify at [coursera.org/verify/YKE13CR0J8Za](https://coursera.org/verify/YKE13CR0J8Za)  
Coursera has confirmed the identity of this individual and their participation in the course.

NYU TANDON SCHOOL OF ENGINEERING

06/21/2019

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## Reinforcement Learning in Finance

An online non-credit course authorized by New York University Tandon School of Engineering and offered through Coursera

Igor Halperin  
Research Professor of Financial Machine Learning  
NYU Tandon School of Engineering

Verify at [coursera.org/verify/AZEx49D77s5](https://coursera.org/verify/AZEx49D77s5)  
Coursera has confirmed the identity of this individual and their participation in the course.



# YOU









**WELCOME BACK!**



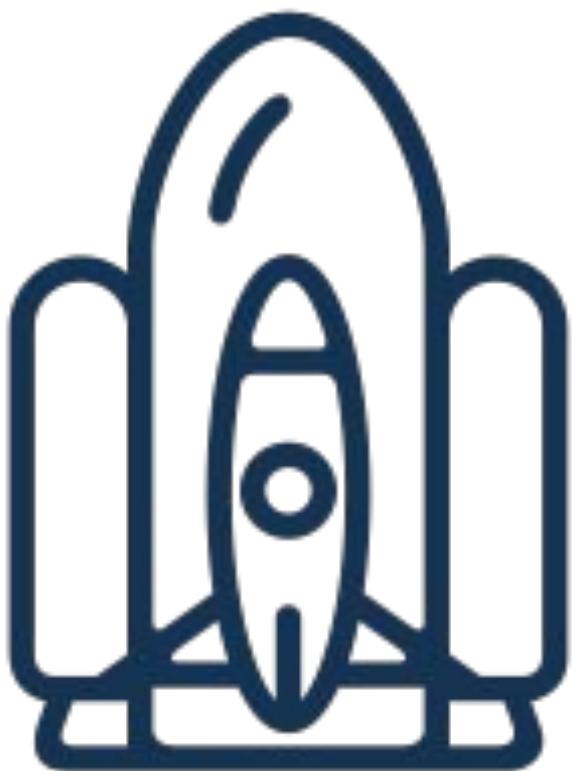


OMG!

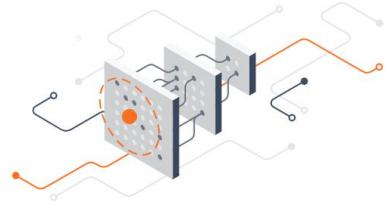


# Journey to the Moon (maybe coming back!?!)





***BLAST OFF!***



# ML Core Concepts





A close-up of a yellow cartoon character's face. The character has large, white, oval-shaped eyes with black pupils. Its mouth is a simple black line. A speech bubble originates from its mouth, containing the text "UNACCEPTABLE!" in bold, white, sans-serif capital letters with a black outline.

**UNACCEPTABLE!**

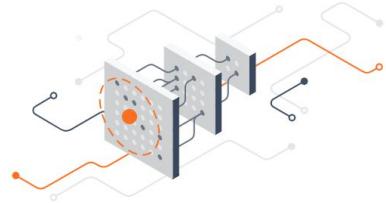
**CN**





# What is ML?

- Complex algorithm and statistical models





# What is ML?



- Complex algorithm and statistical models
- Ability to learn and improve from experience without being explicitly programmed



# What is ML?



- Complex algorithm and statistical models
- Ability to learn and improve from experience without being explicitly programmed
- Allow computers to learn automatically without human intervention or assistance and adjust actions accordingly



# What is ML?



- Complex algorithm and statistical models
- Ability to learn and improve from experience without being explicitly programmed
- Allow computers to learn automatically without human intervention or assistance and adjust actions accordingly
- The goal is to find the features parameters (e.g. weights and biases) by tweaking input hyperparameters



Machine Learning is

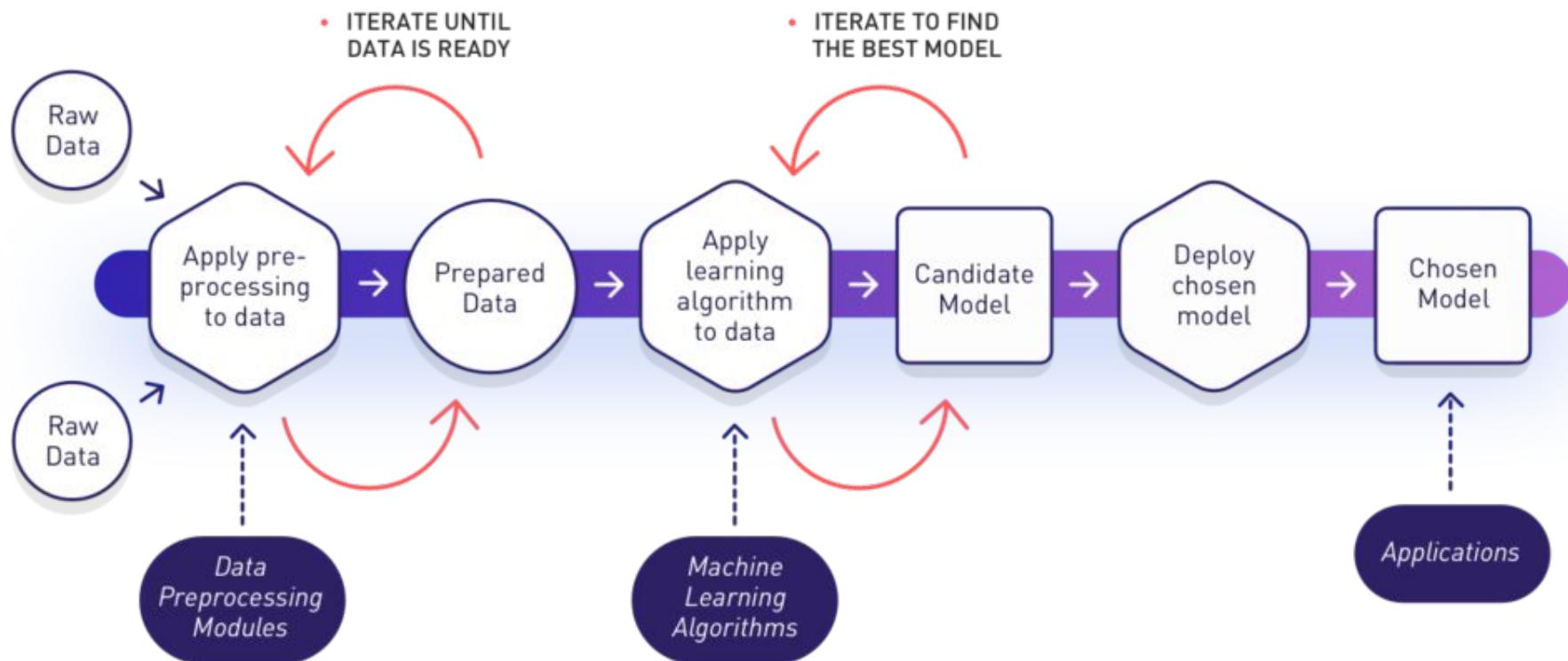
**NOT Magic.**

Rather tools and technology to **answer questions with data!**



## Using Data (for Training)

## Answer Questions (Prediction or Inference)

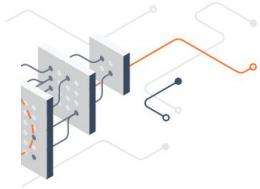




# Human Friendly Way



## Let's meet John



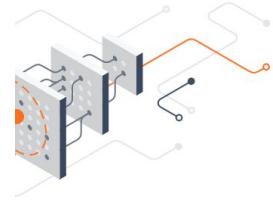
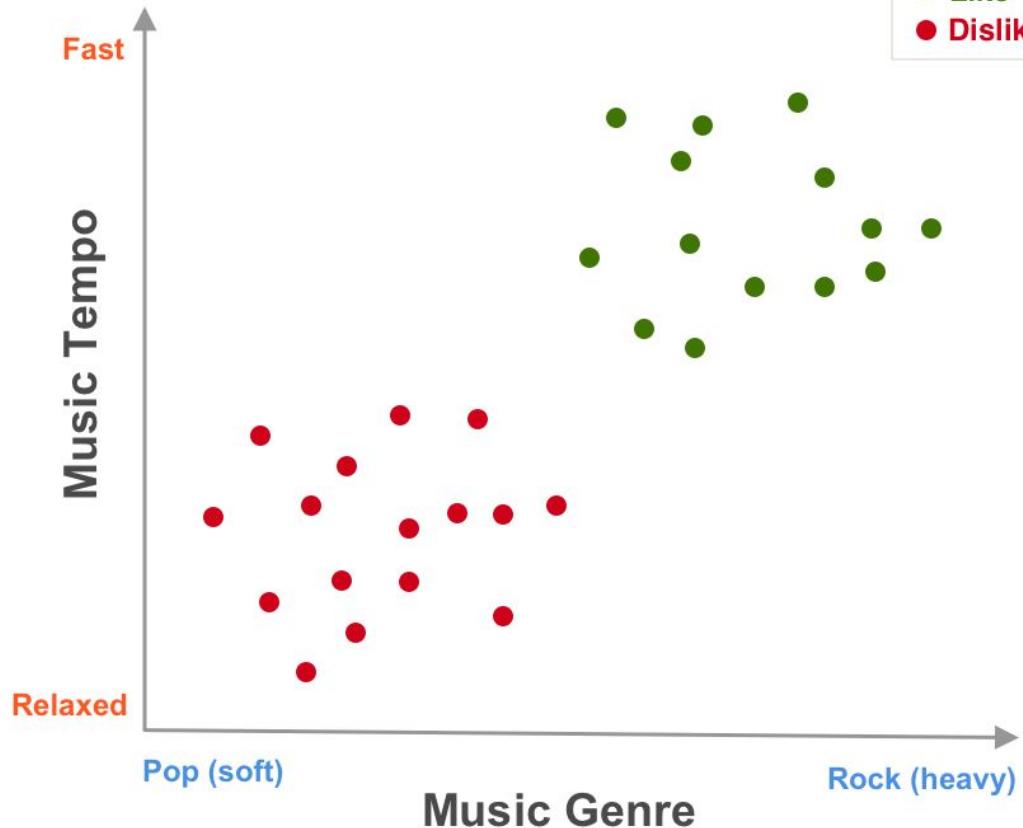
- Fast tempo
- Rock



- Slow tempo
- Pop

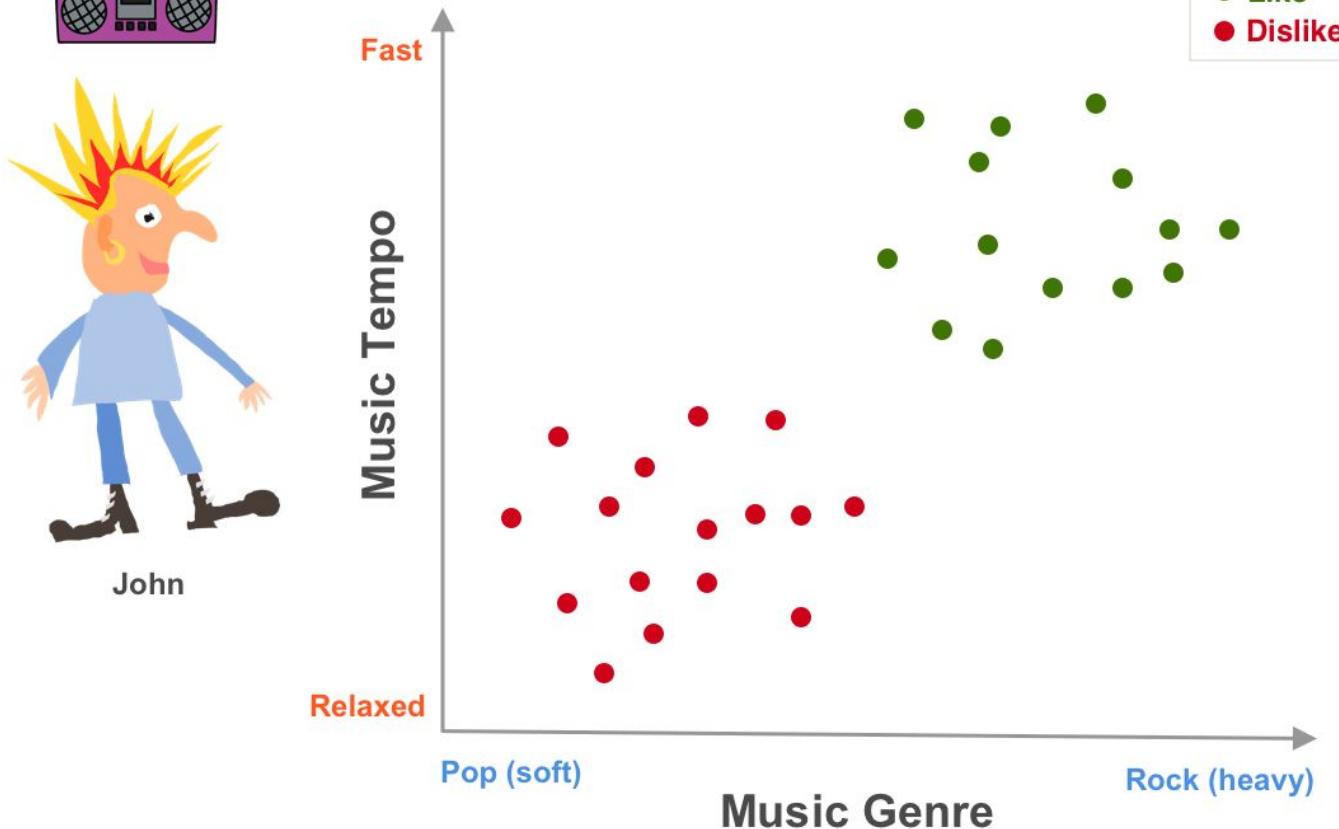
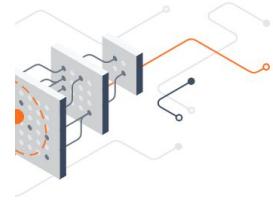


John





Listening a new song:  
**Metallica - Master of Puppets**

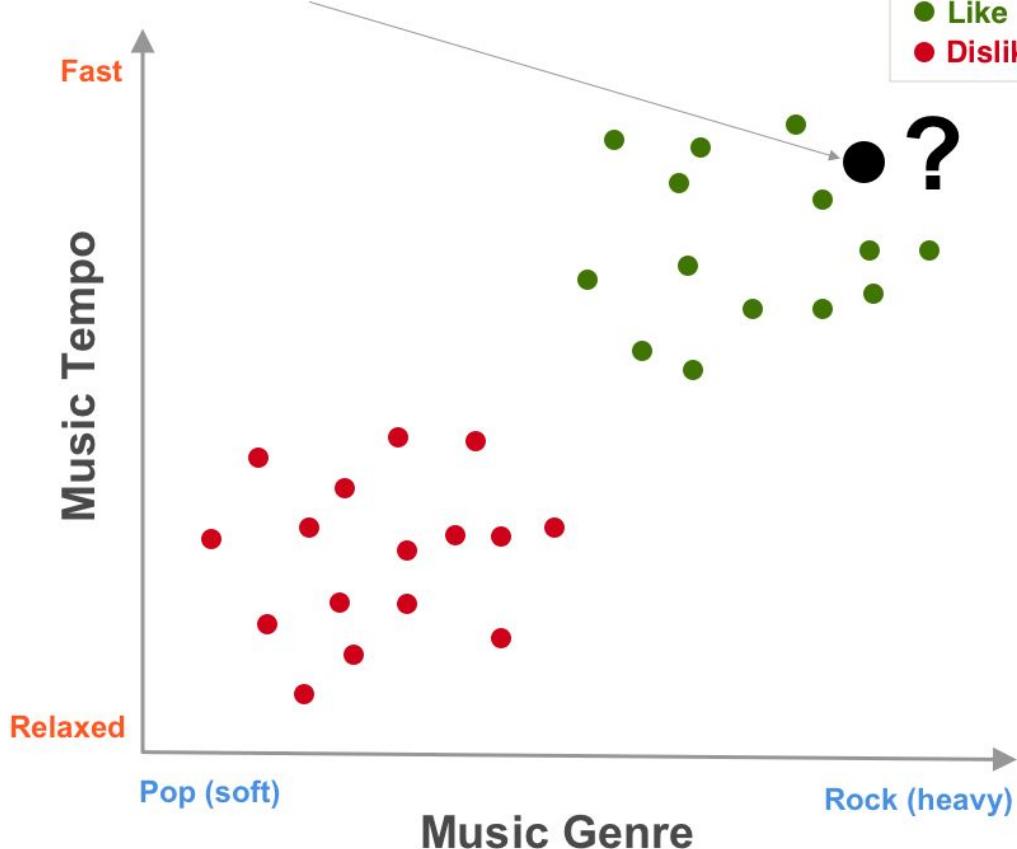




John

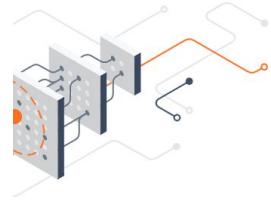


Listening a new song:  
Metallica - Master of Puppets

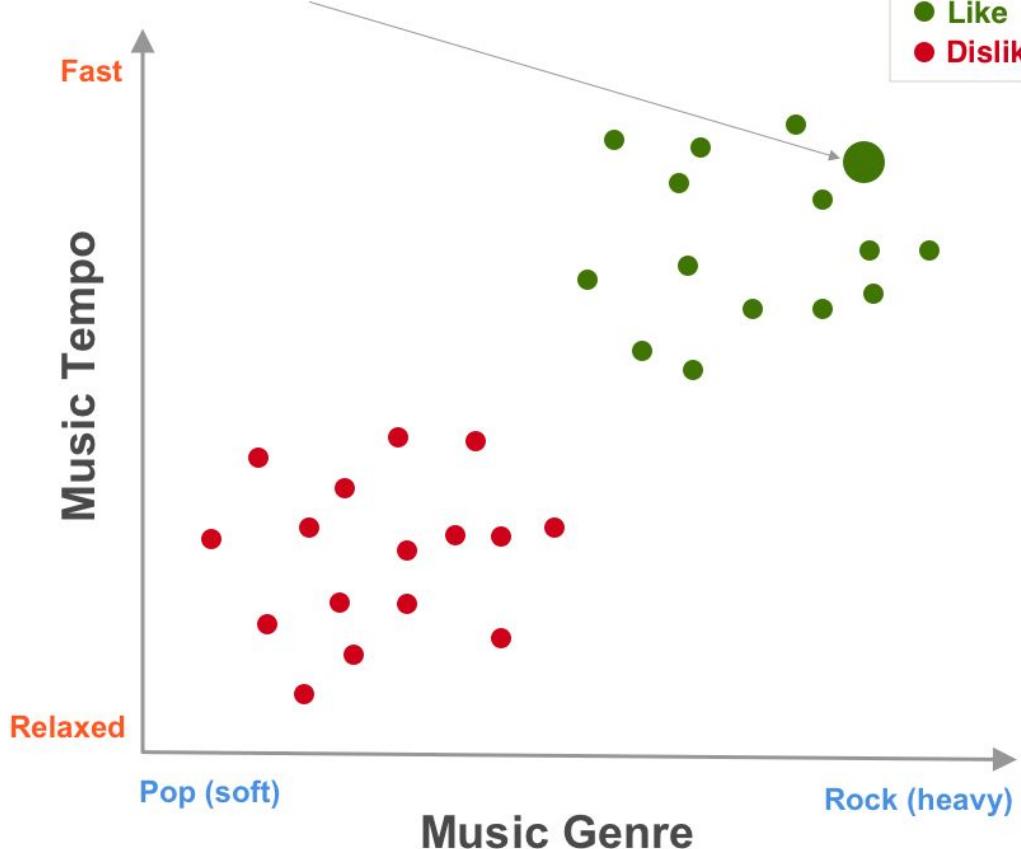




## Listening a new song: **Metallica - Master of Puppets**



John

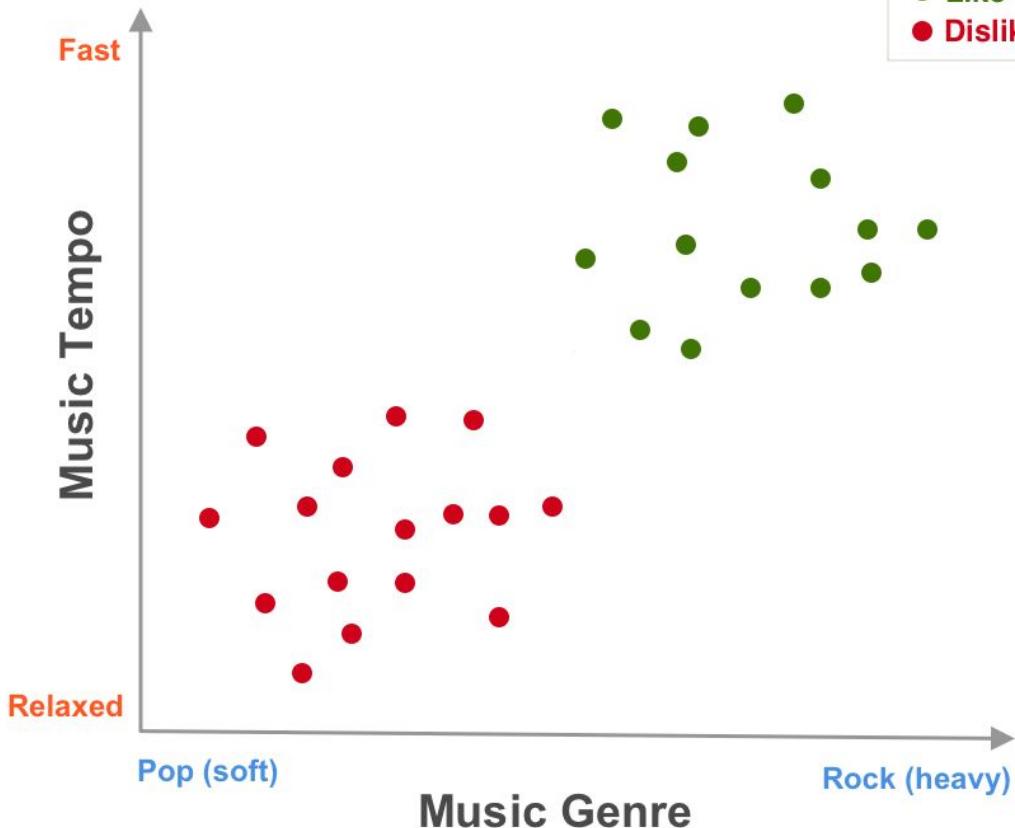




John

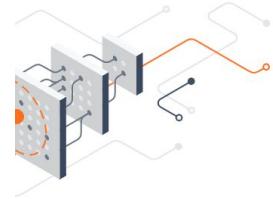
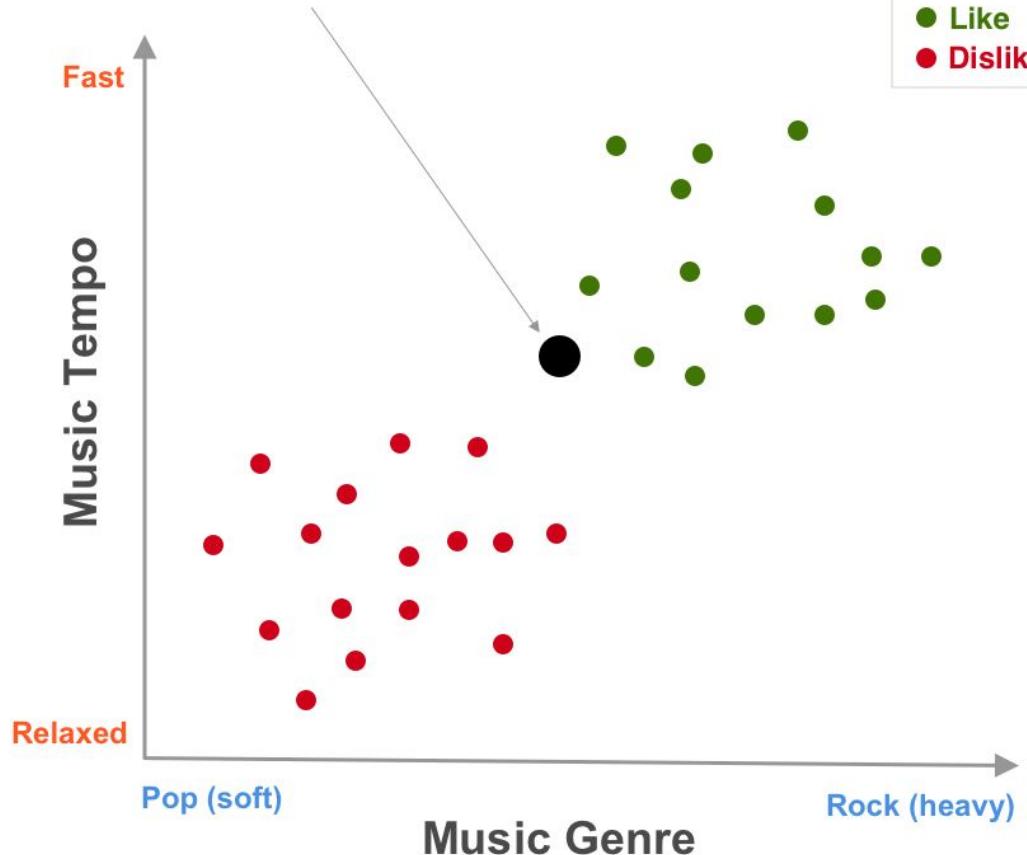


Listening another new song:  
Billie Eilish - Bad Guy





Listening another new song:  
**Billie Eilish - Bad Guy**

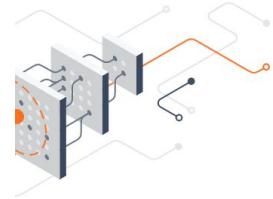
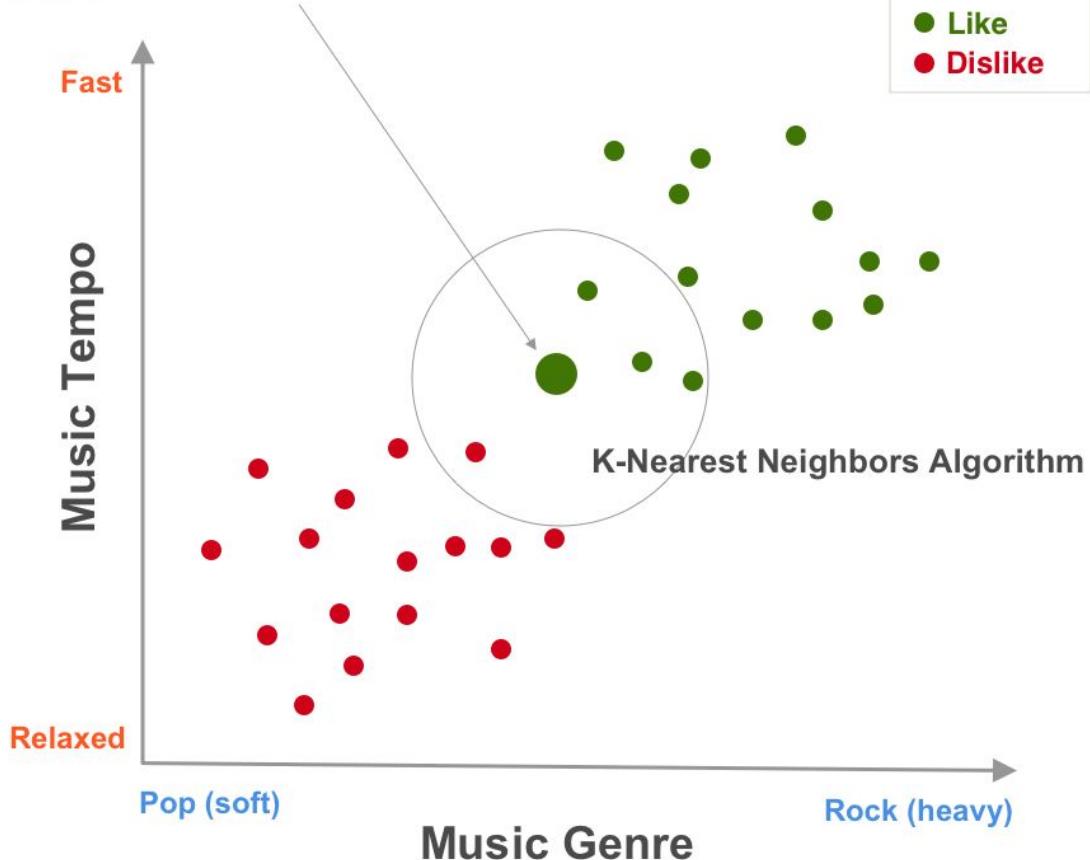




Listening another new song:  
**Billie Eilish - Bad Guy**

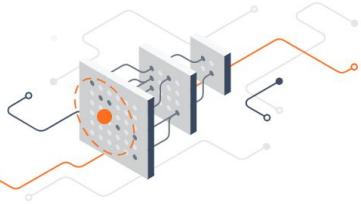
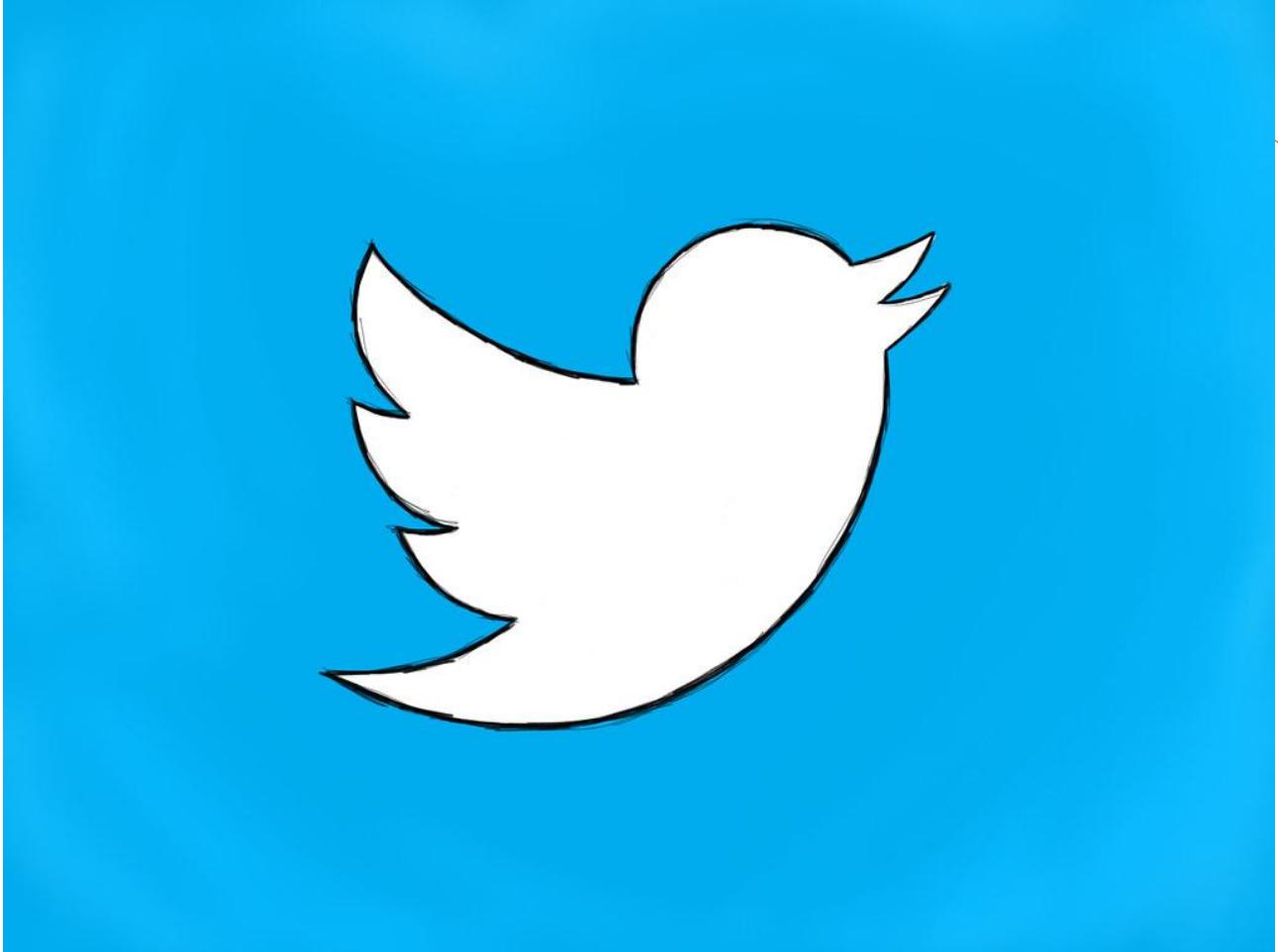


John





# Scenarios of ML



@aaronhma



The image shows a composite of two visual elements. On the left is a screenshot of a mobile application interface titled "My TweetBook". The interface includes a profile picture placeholder, a progress bar, and a section with three colored squares (green, pink, blue). On the right is a diagram of a neural network layer, featuring a grid of neurons with orange arrows indicating data flow between them.

My TweetBook

38 124 79 32

Tweets Tweets&replies Media

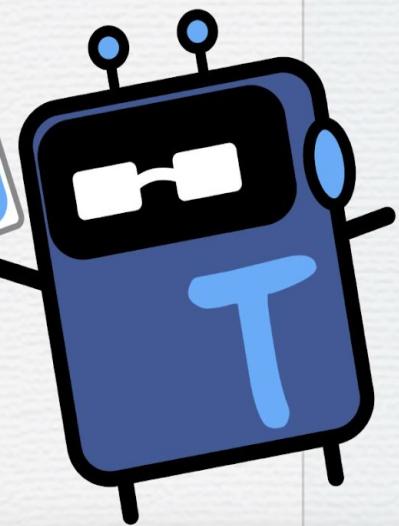
	Tweets	Tweets&replies	Media
1	38	124	79
2	38	124	79
3	38	124	79
4	38	124	79



## My TweetBook

38      124      79      32

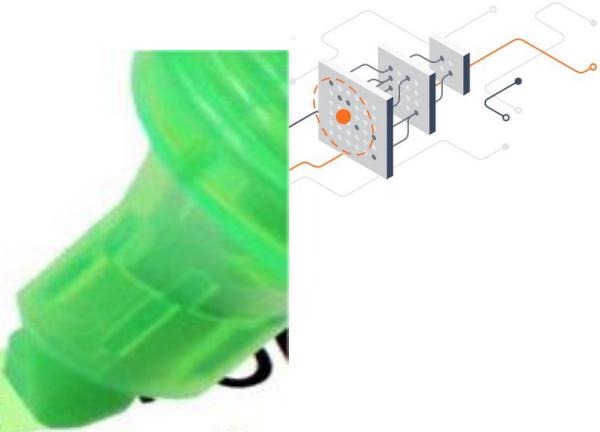
Tweets    Tweets&replies    Media

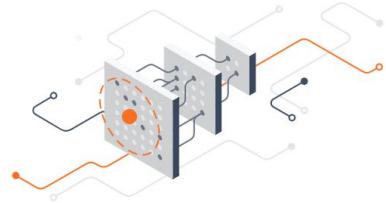




# Content

Content is the  
Content can be de-  
quantify the infor-  
information prov-  
shining orig

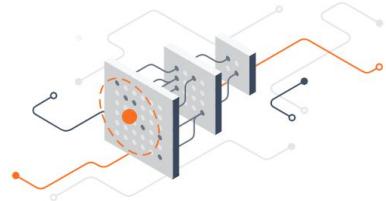


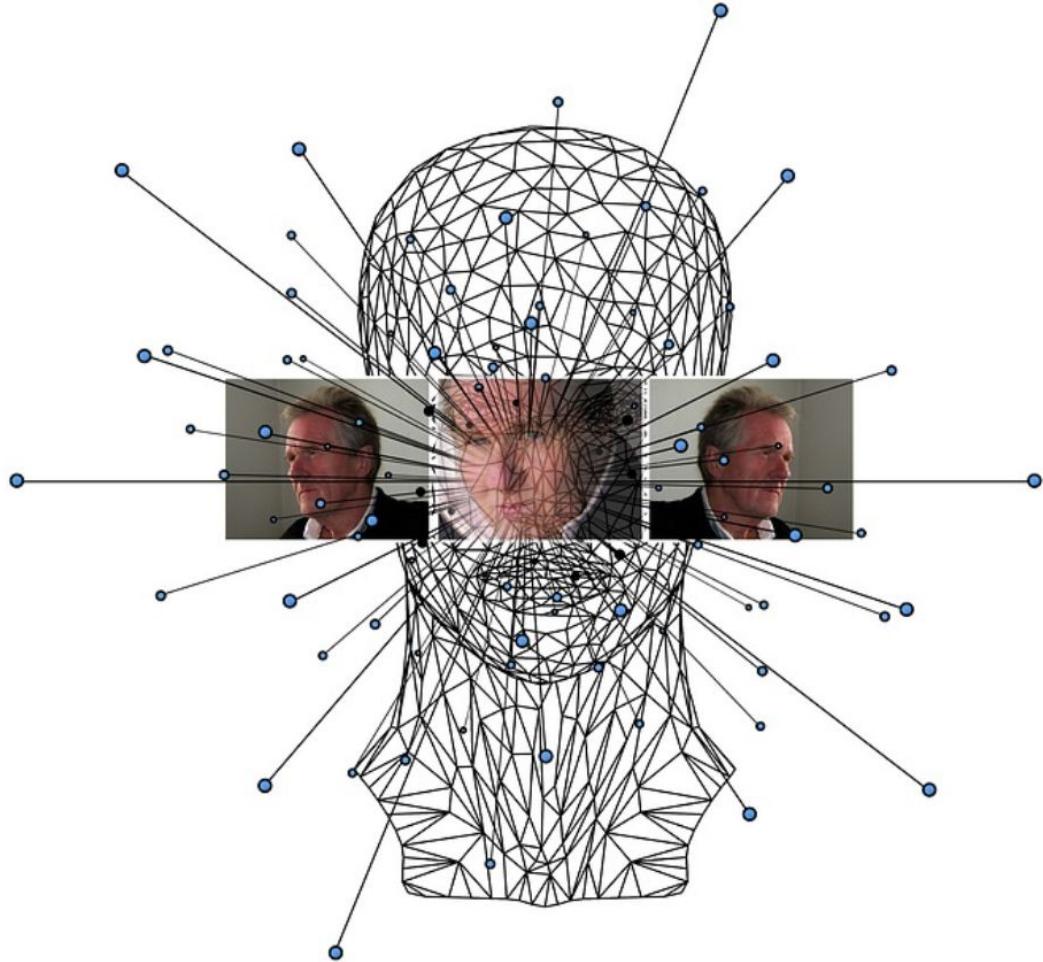




# Differential pricing in real time based on:

- Demand (predictive model for area of high demand)
- Number of cars available
- Bad weather
- Rush hours / Real time traffics
- Travel distances
- Popular Destinations
- Special events (e.g Super Bowl, Concerts, etc)

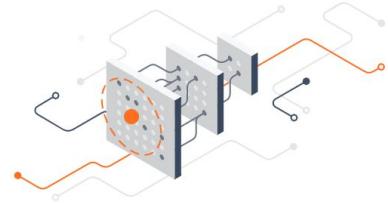
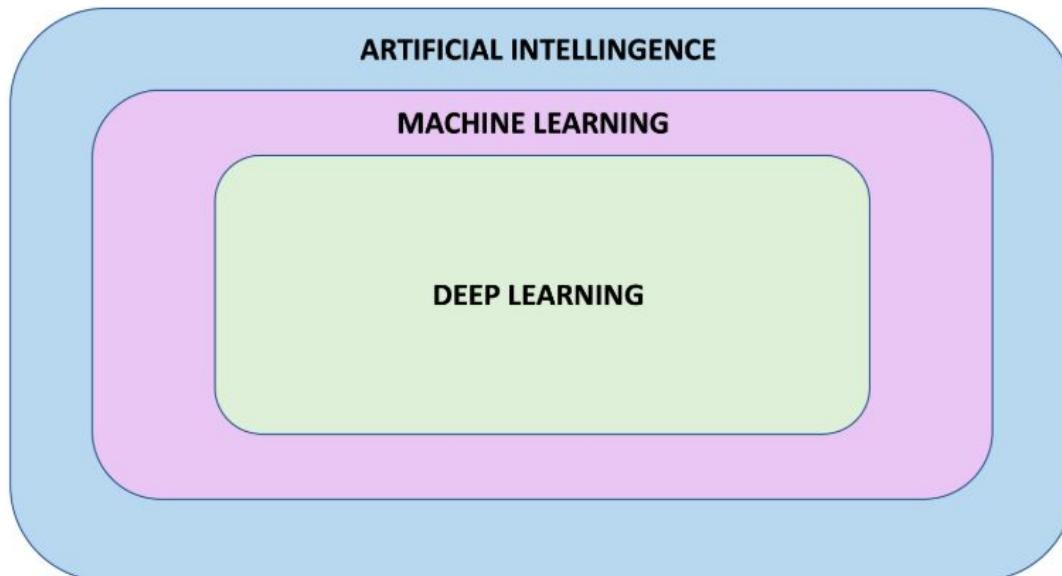




@aaronhma



# Fields of Study in Machine Learning



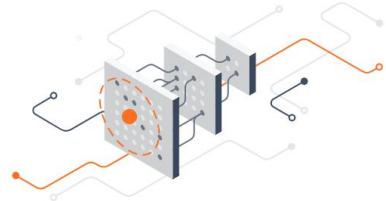
**AI:** The broad discipline of creating intelligent machines.

**ML:** A system that can learn from experience.

**DL:** A system that learns from experience on large data sets.



# Types of Machine Learning



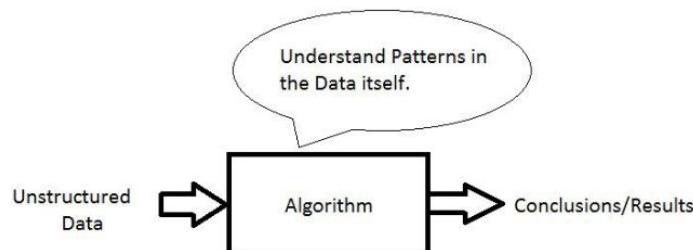
## Supervised Learning

- **Task driven**
- Data -> input
- Label -> targets
- Iteration of training, algorithm finds desired output
- Regression and Classification problems



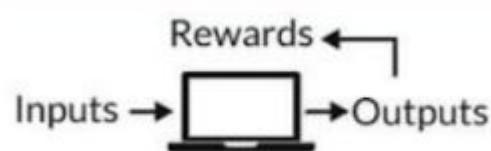
## Unsupervised Learning

- **Data driven**
- No labels
- Not structured data
- Clustering problems  
(Group similar things)



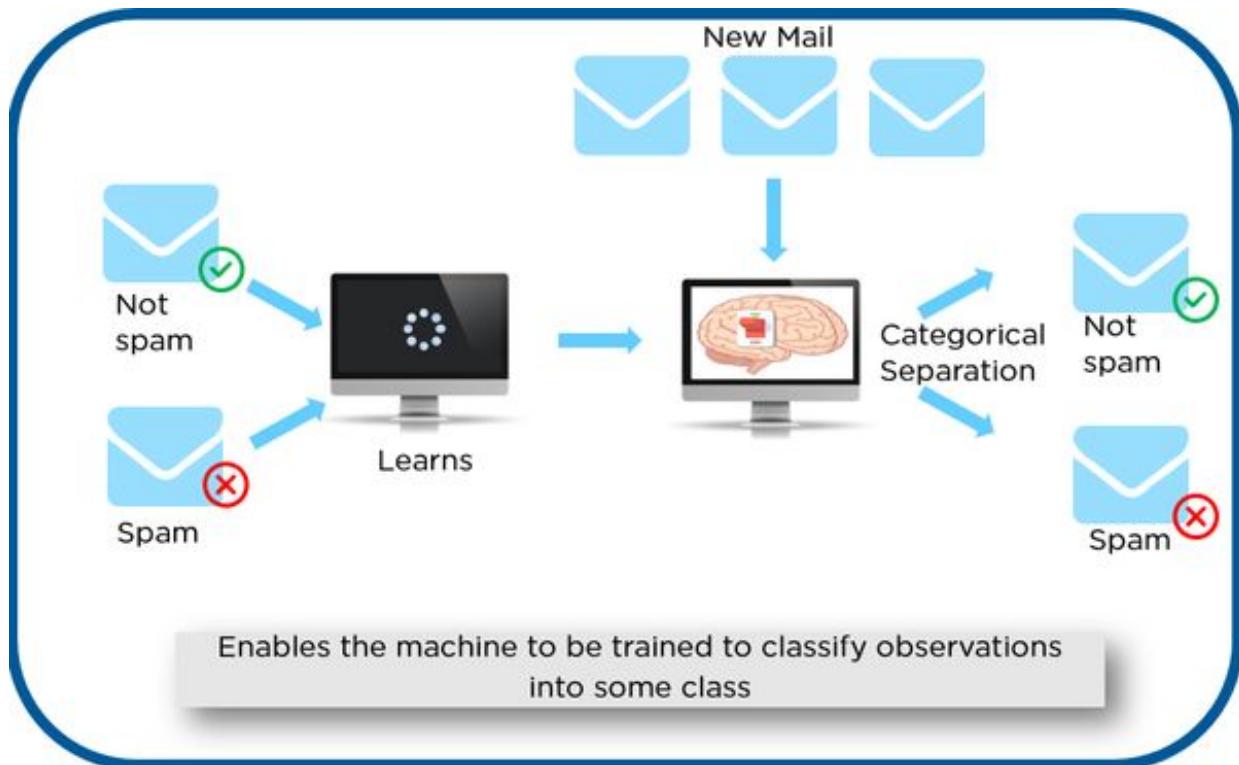
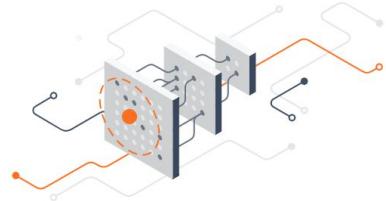
## Reinforcement Learning

- **Algorithm driven**
- Reward-based learning by mapping state-action pairs to expected rewards
- learn from its experience
- Incentivize / Penalized



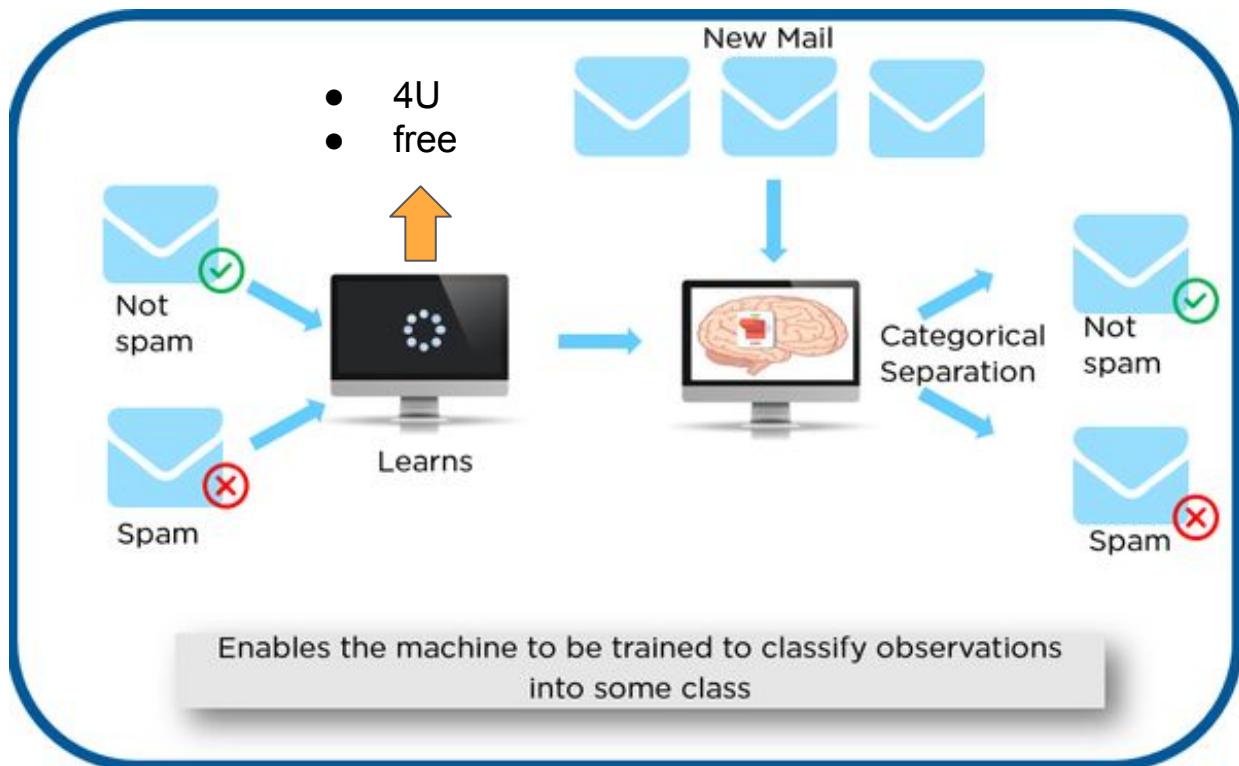


# Supervised Learning: Classification



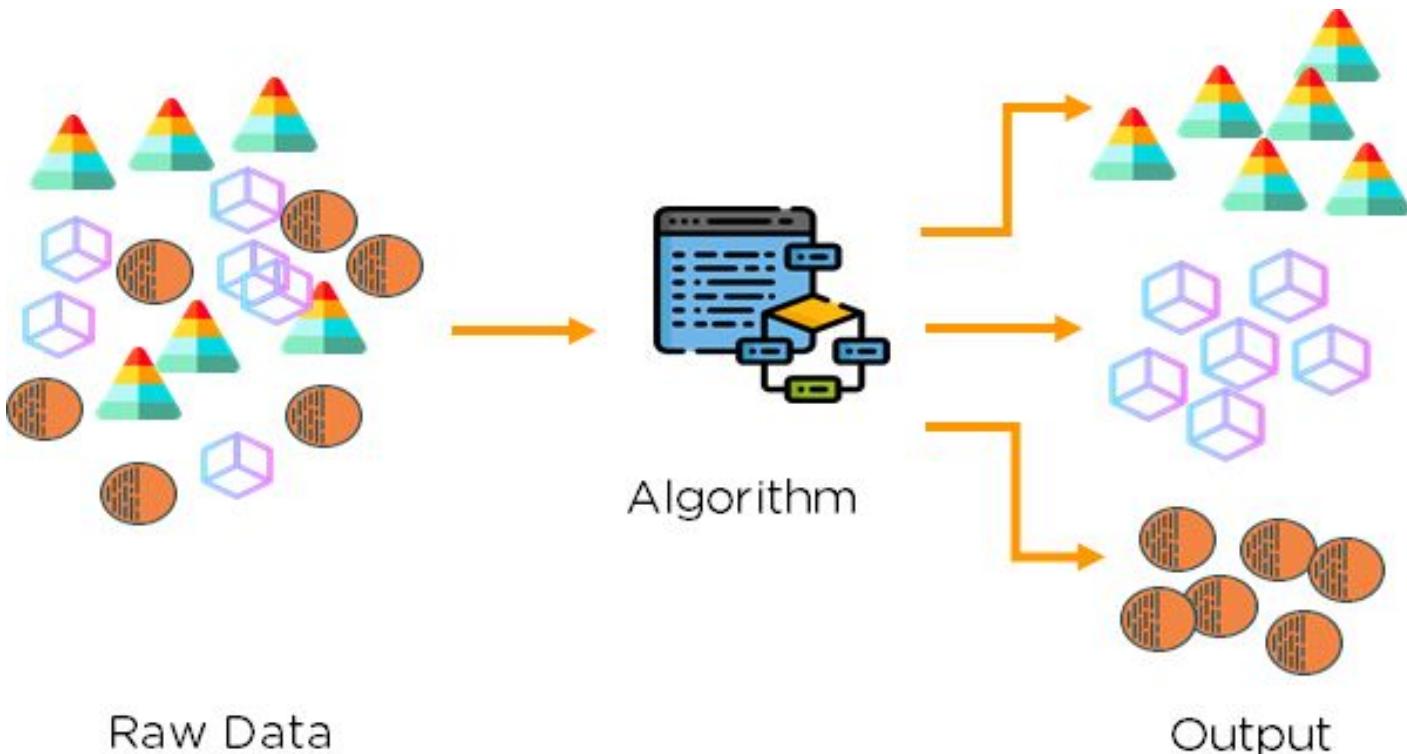
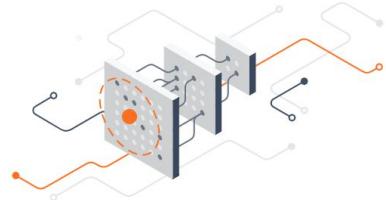


# Supervised Learning: Classification



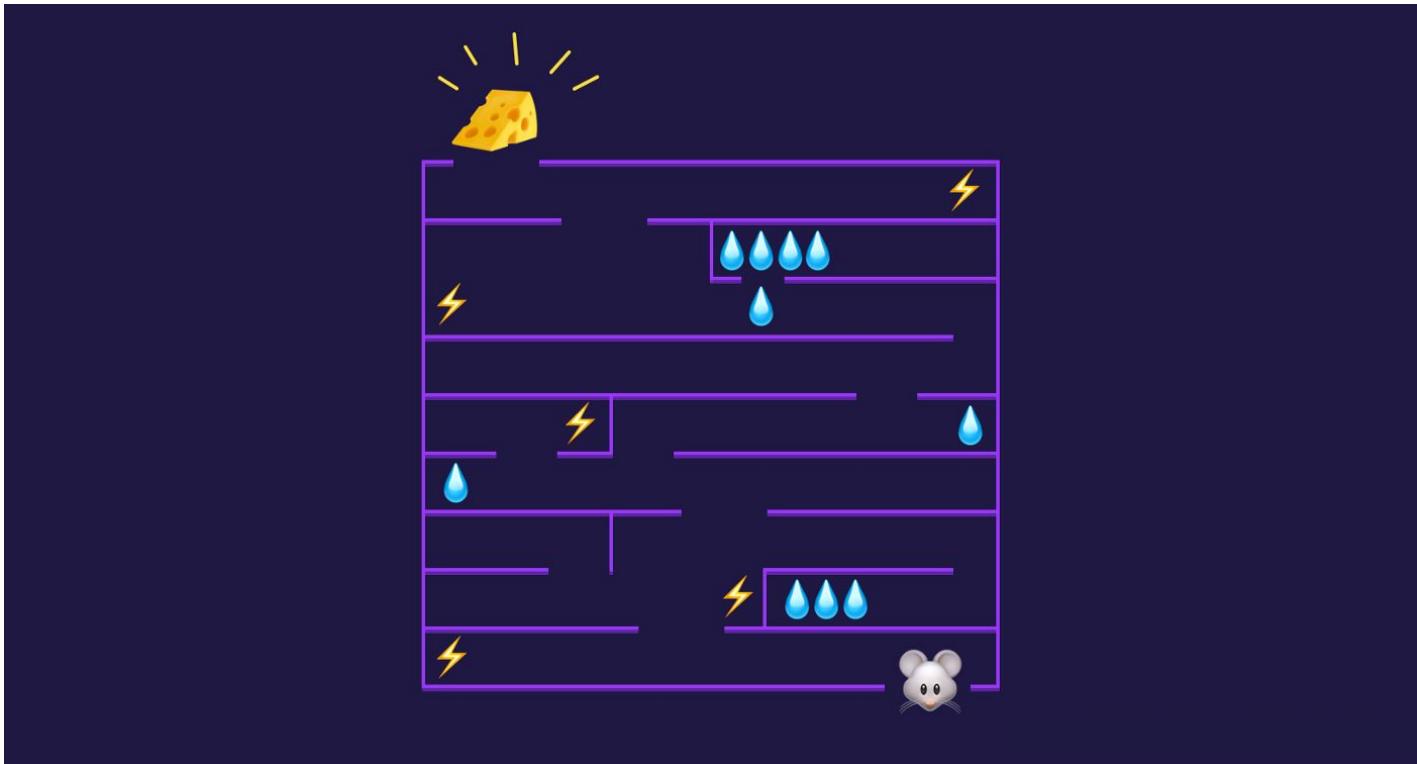


# Unsupervised Learning: Clustering



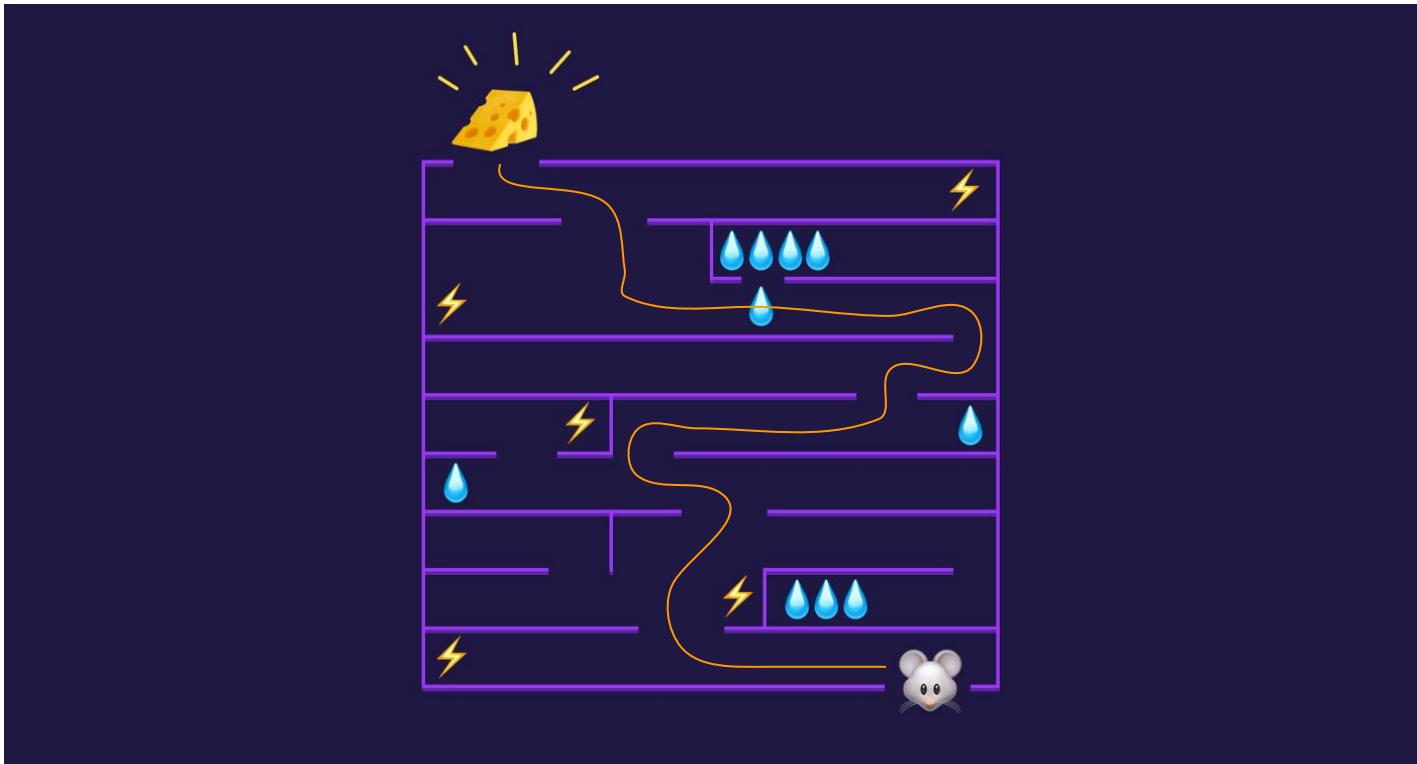


# Reinforcement Learning



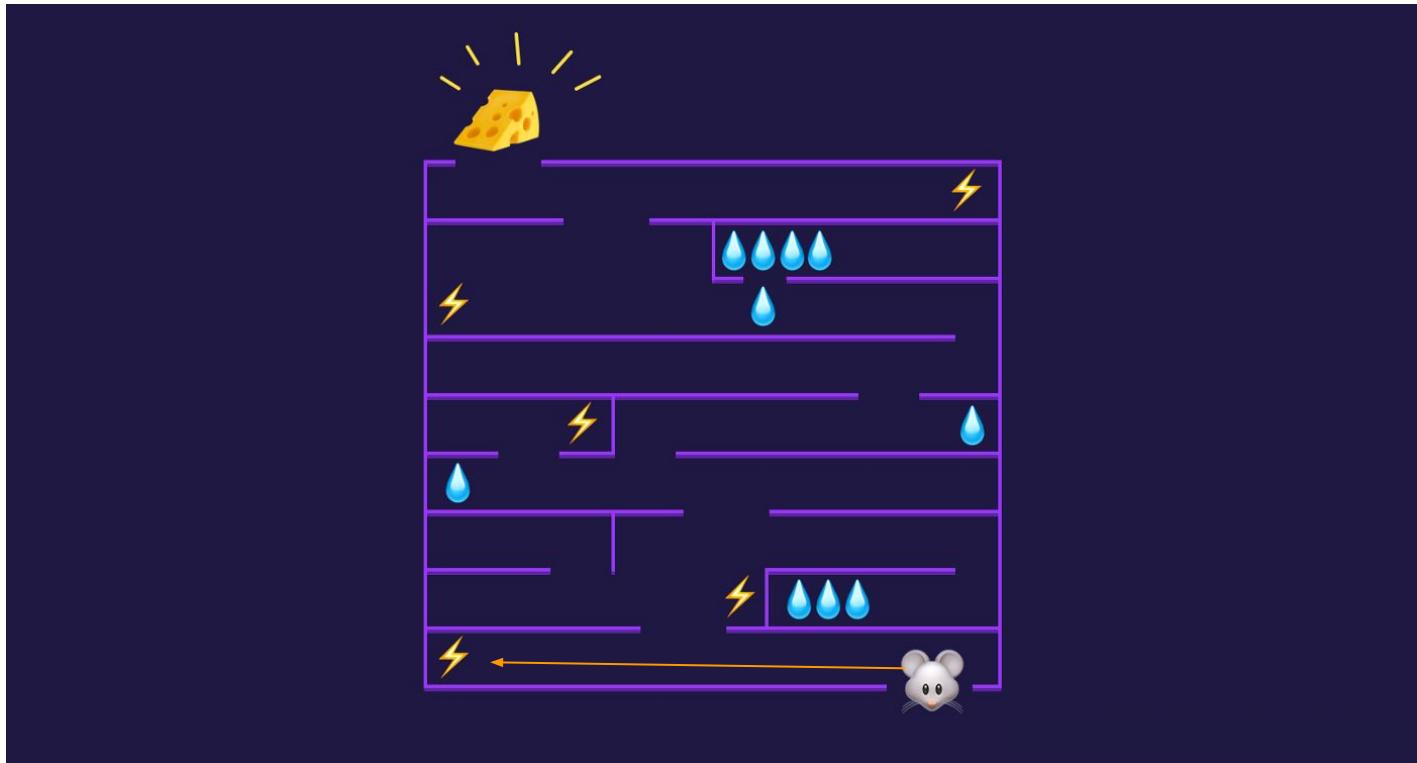
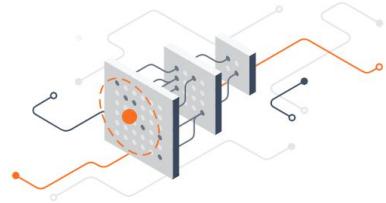


# Reinforcement Learning



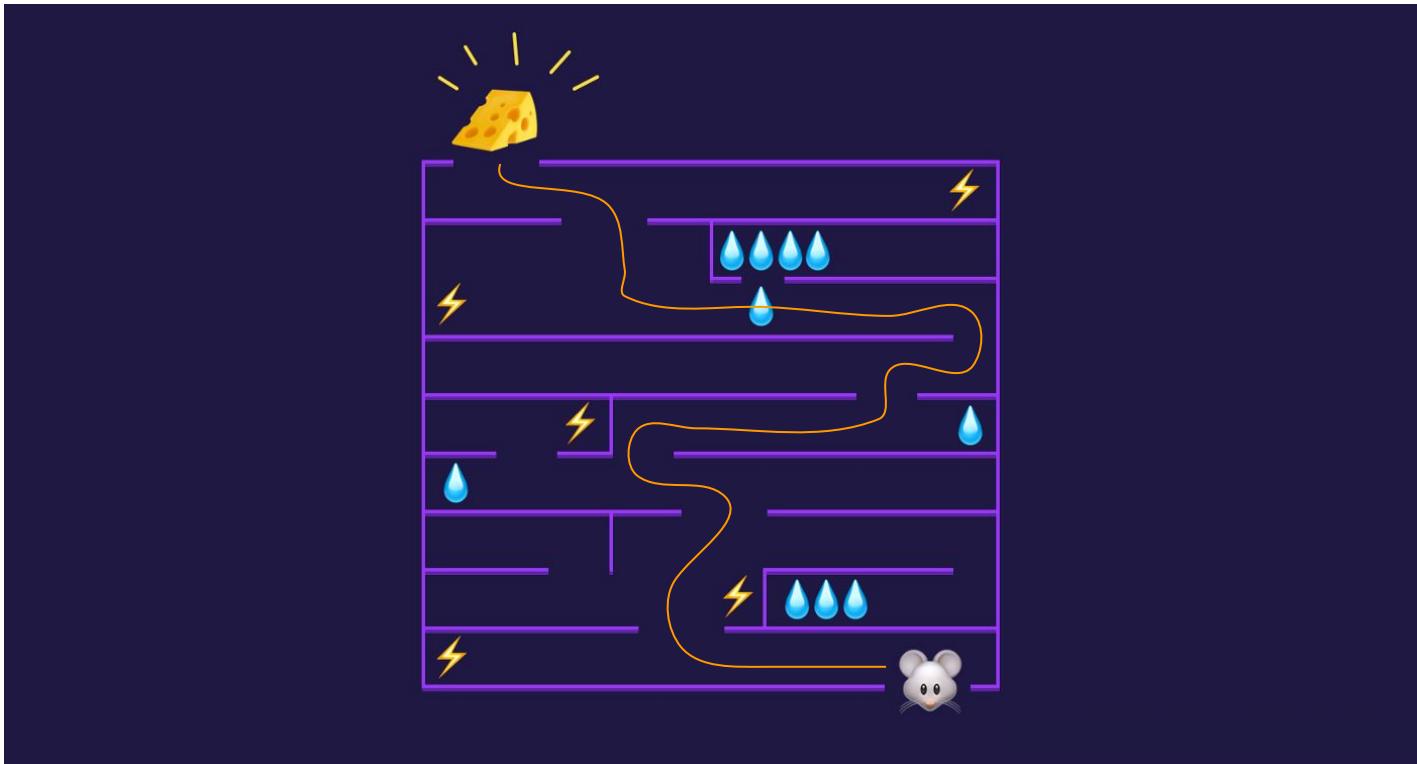


# Reinforcement Learning





# Reinforcement Learning

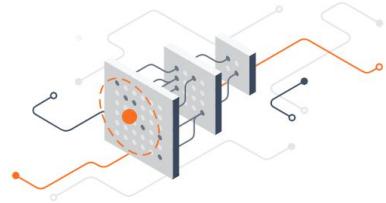


# Neural Network



# What's a Neural Network?

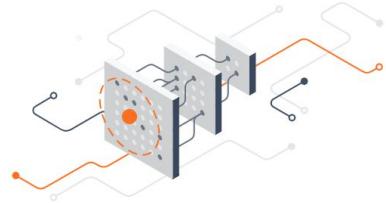
- Sort of computer software, inspired by the brain



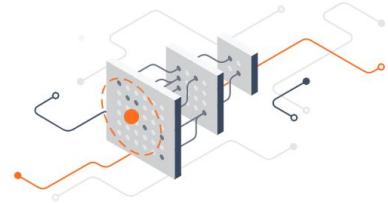
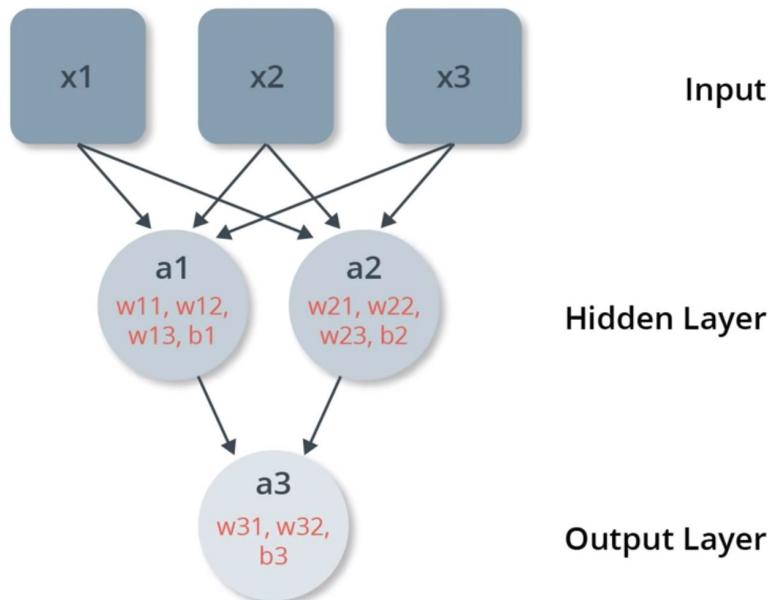


# What's a Neural Network?

- Sort of computer software, inspired by the brain
- Made up of units(aka neurons) to solve a problem together



# A Neural Network



**NN:** a stack of layers.

**Layer:** is made up of units.

**Unit:** also called neurons, like the nerve cells in the biological brain.

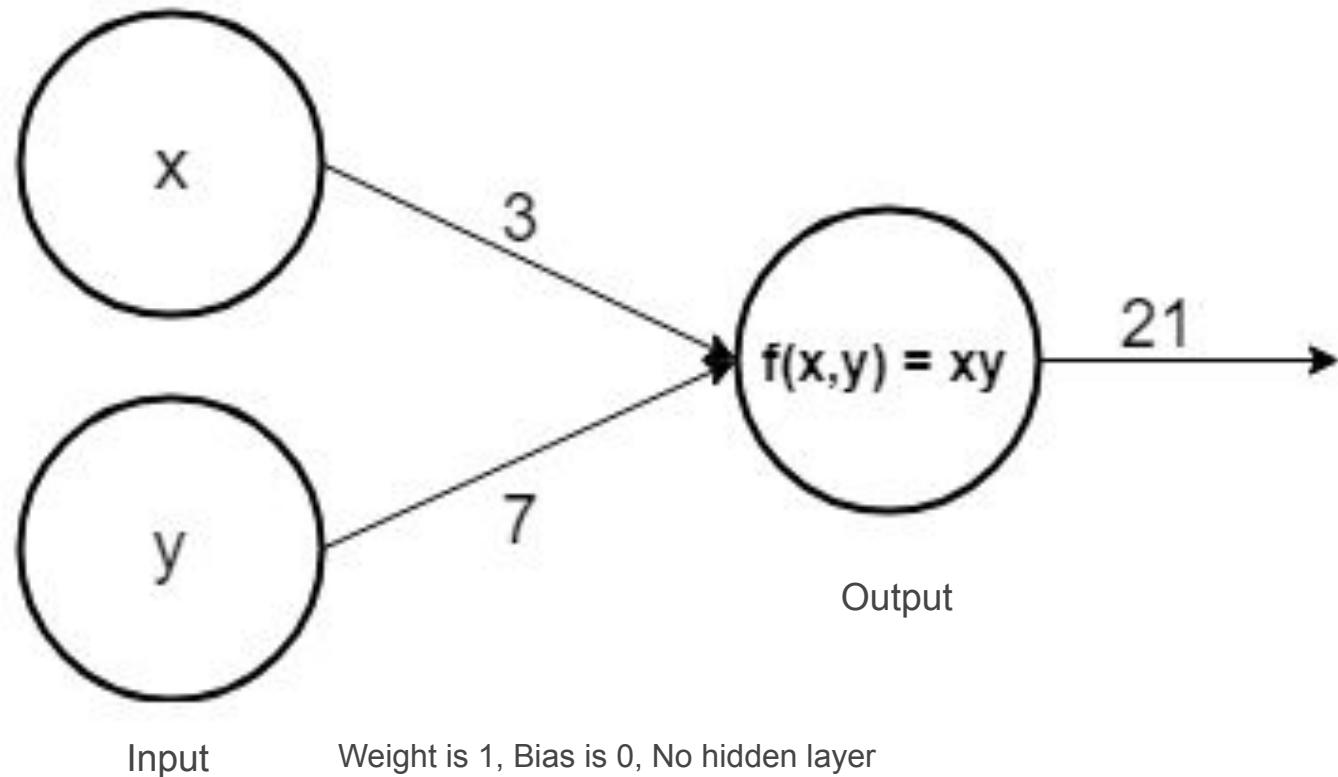
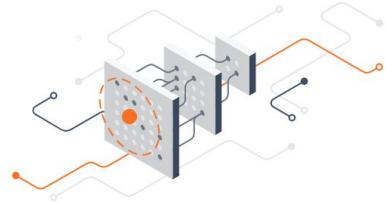
**Dense Layer:** stack of layers are fully connected

**Weight & Biases:** internal variables of model which are updated during training

**Model:** the representation of neural network

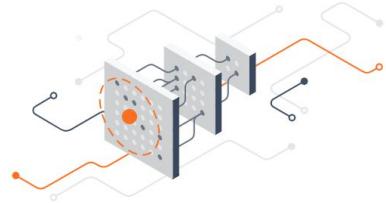


# World's Simplest Neural Network





# World's Simplest Neural Network in Python



```
def neural_network (input, weight):
    return input * weight

inputs = [8.5, 9.5, 10, 9]
weight = 0.1
prediction = neural_network(inputs[0], weight)
print (prediction)    # 0.85
```

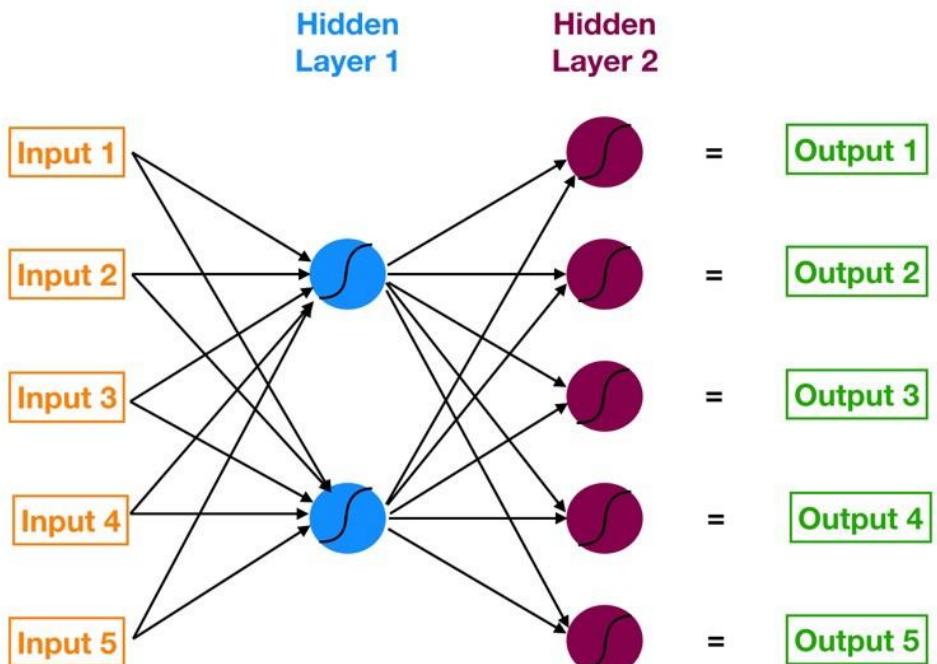
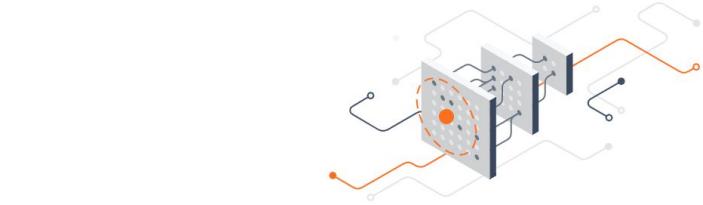
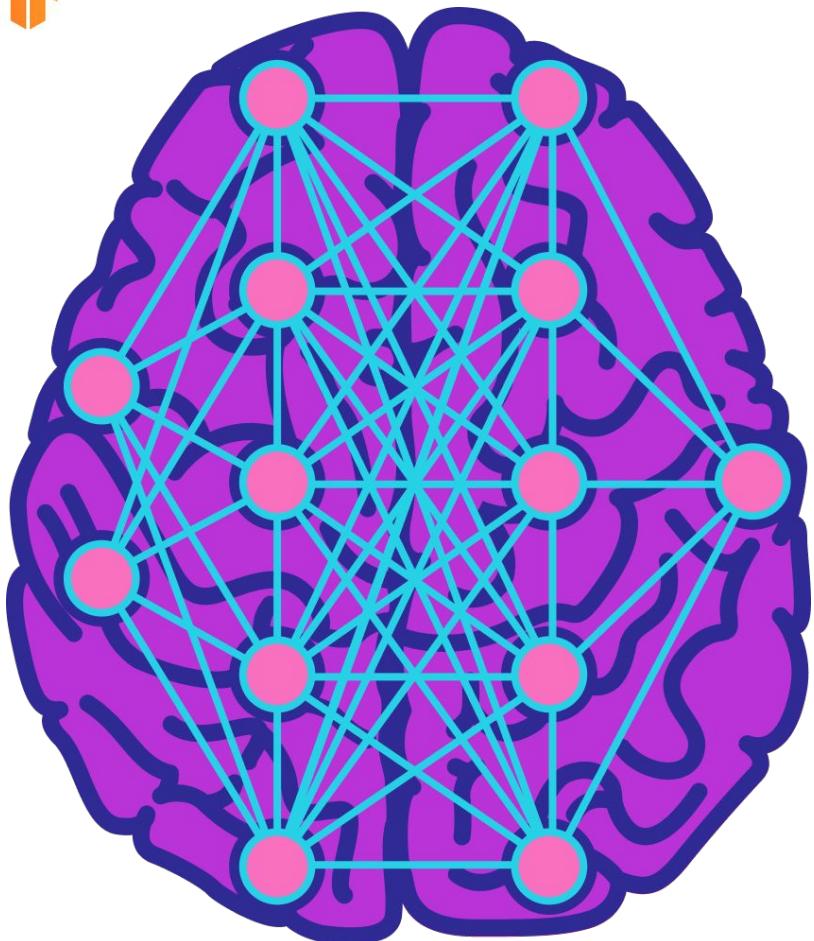
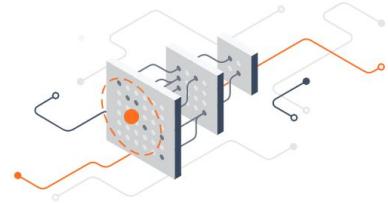
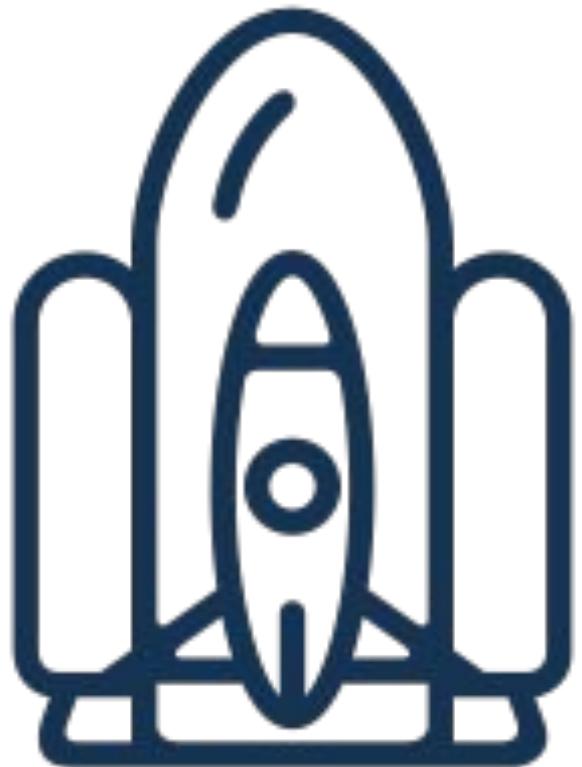


Photo copyright: Towards data science

@aaronhma



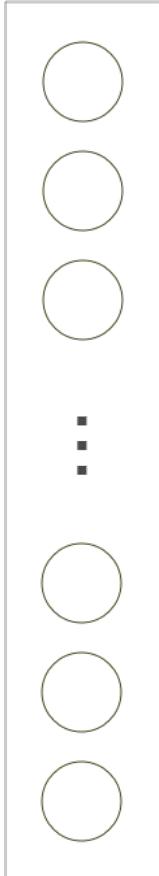
1ST STAGE SEPARATION!



# Neural Network Deep Dive



Input layer



Hidden layers

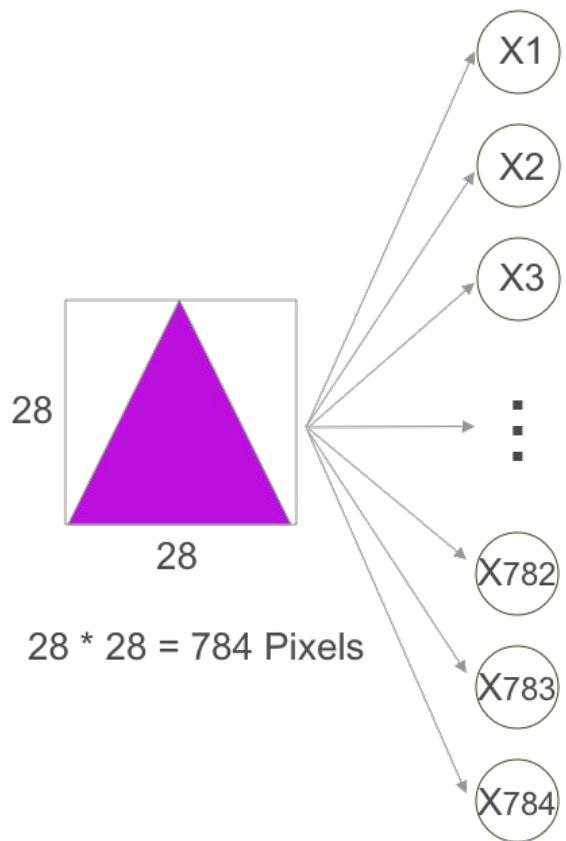


Output layer

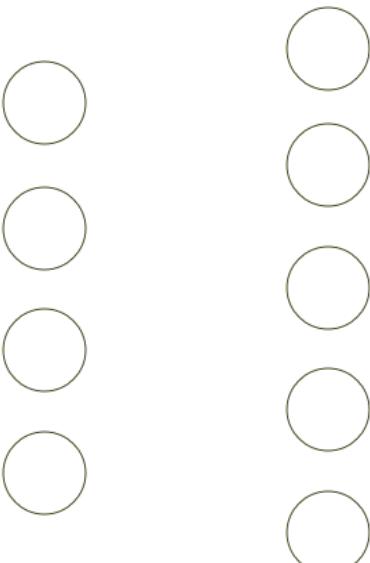




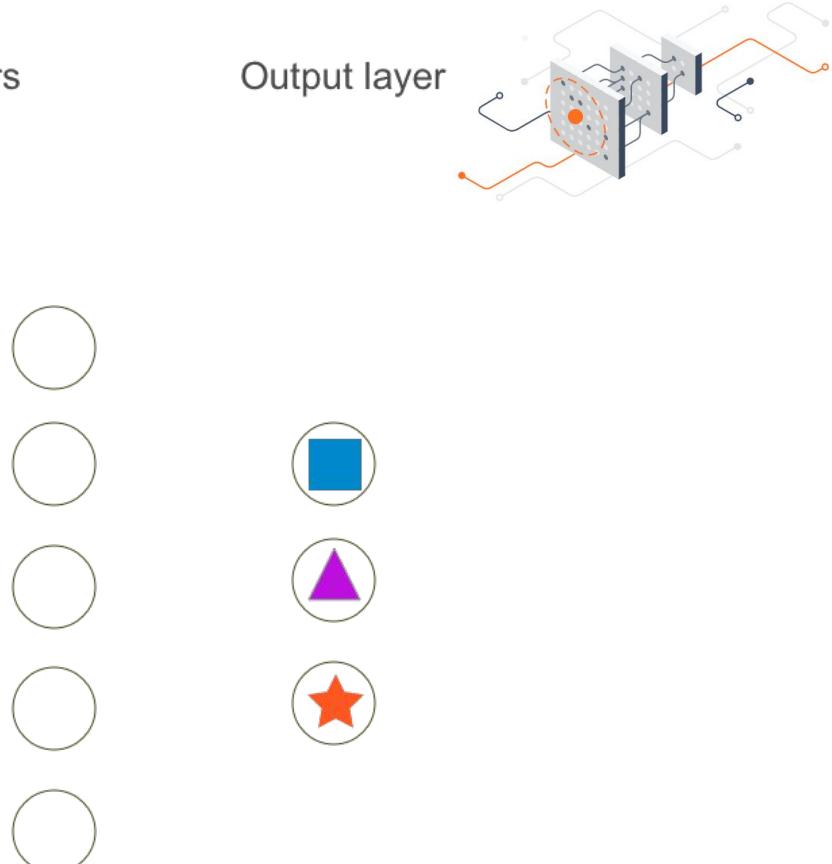
Input layer



Hidden layers



Output layer





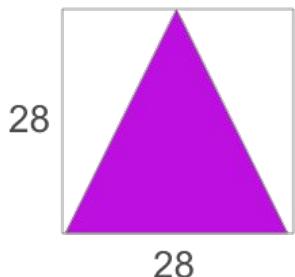
$$(X_1 * 0.5 + X_2 * 1.2 + X_3 * -2.3 + \dots + X_{782} * 3.8 + X_{783} * 0.5 + X_{784} * 4) + 9.8$$



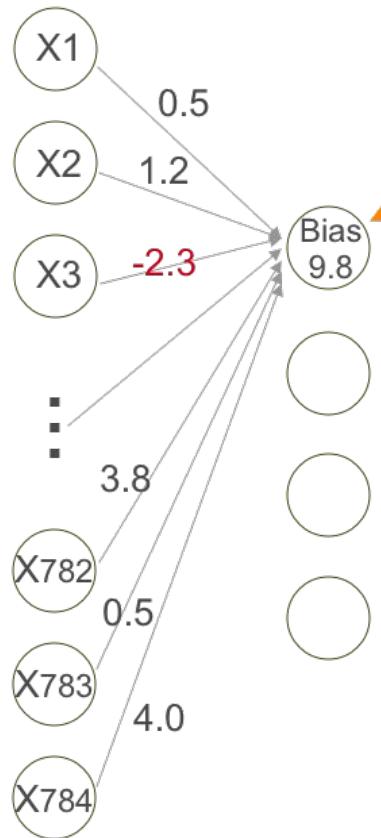
Activation Function  
Sigmoid Fn [0, 1]



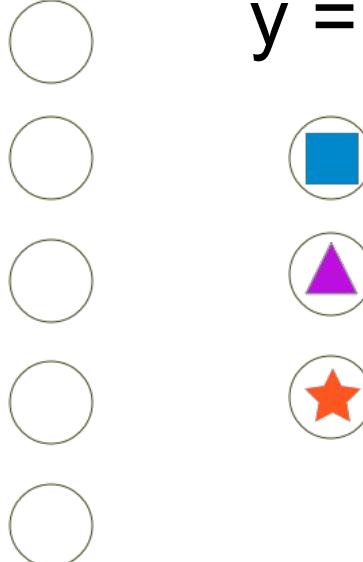
0.875



$28 * 28 = 784$  Pixels

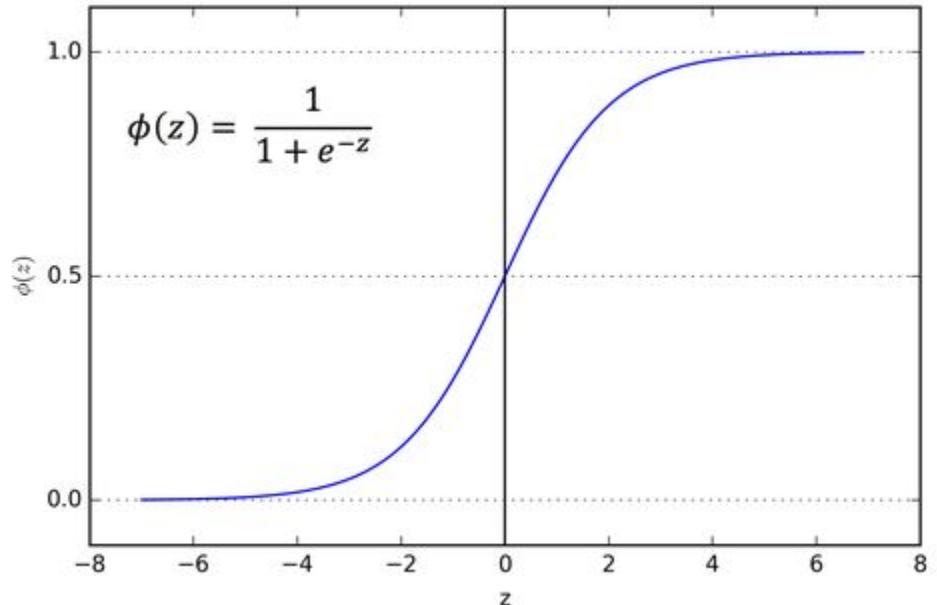


$$y = wx + b$$

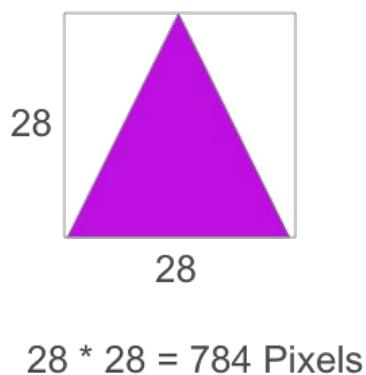




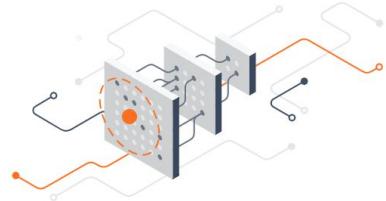
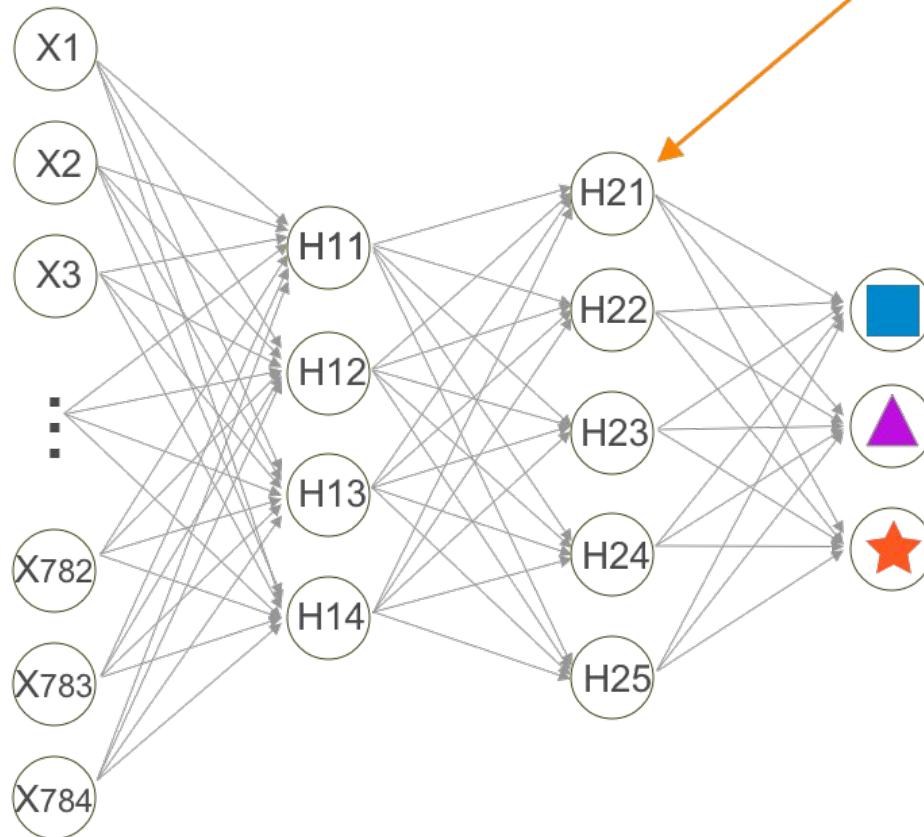
# Sigmoid Function

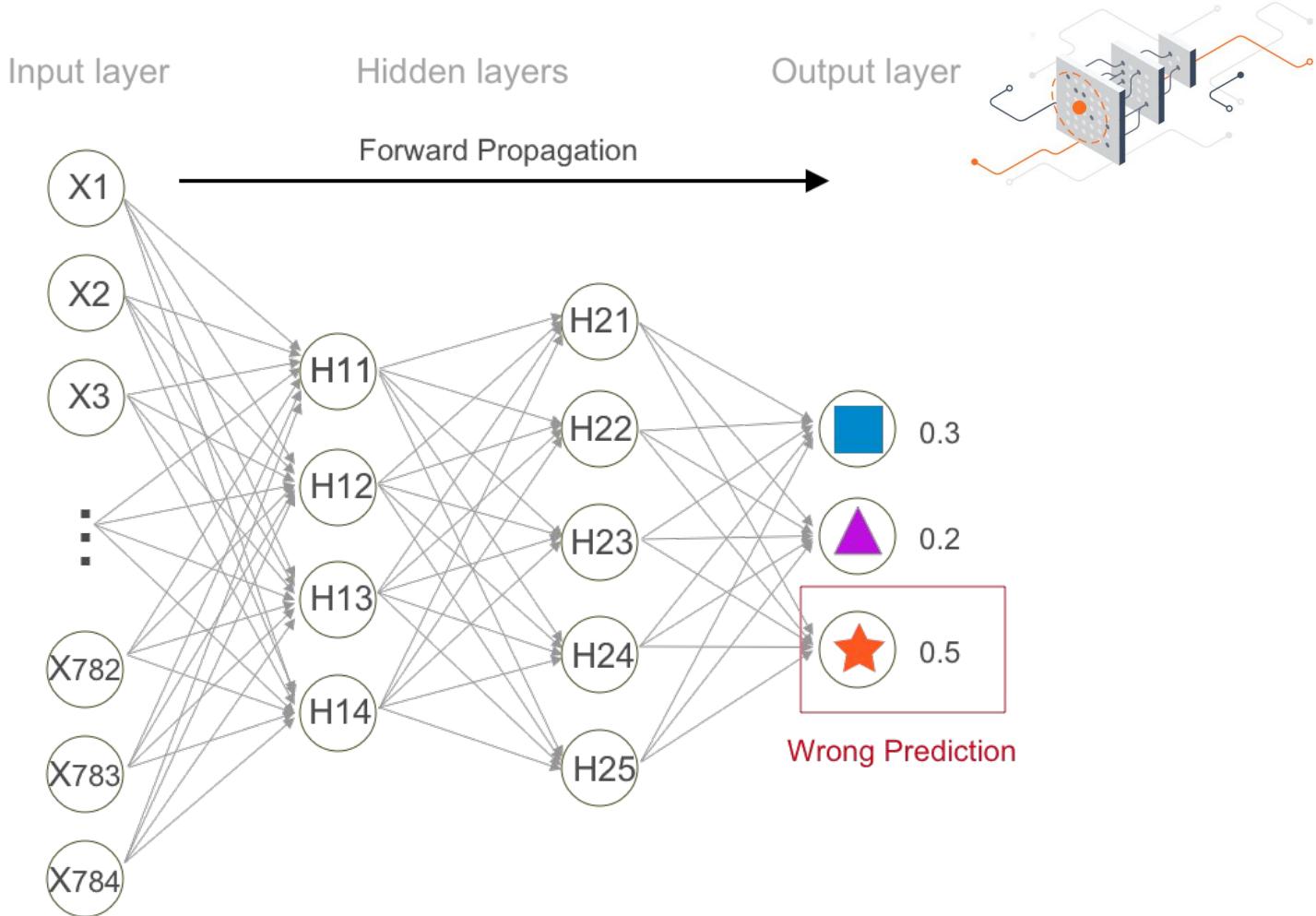
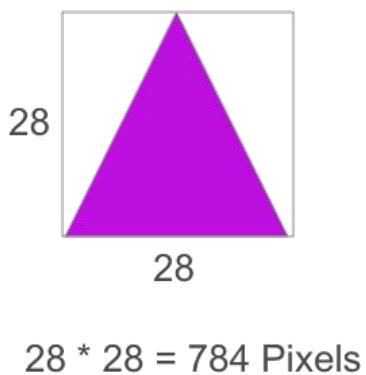


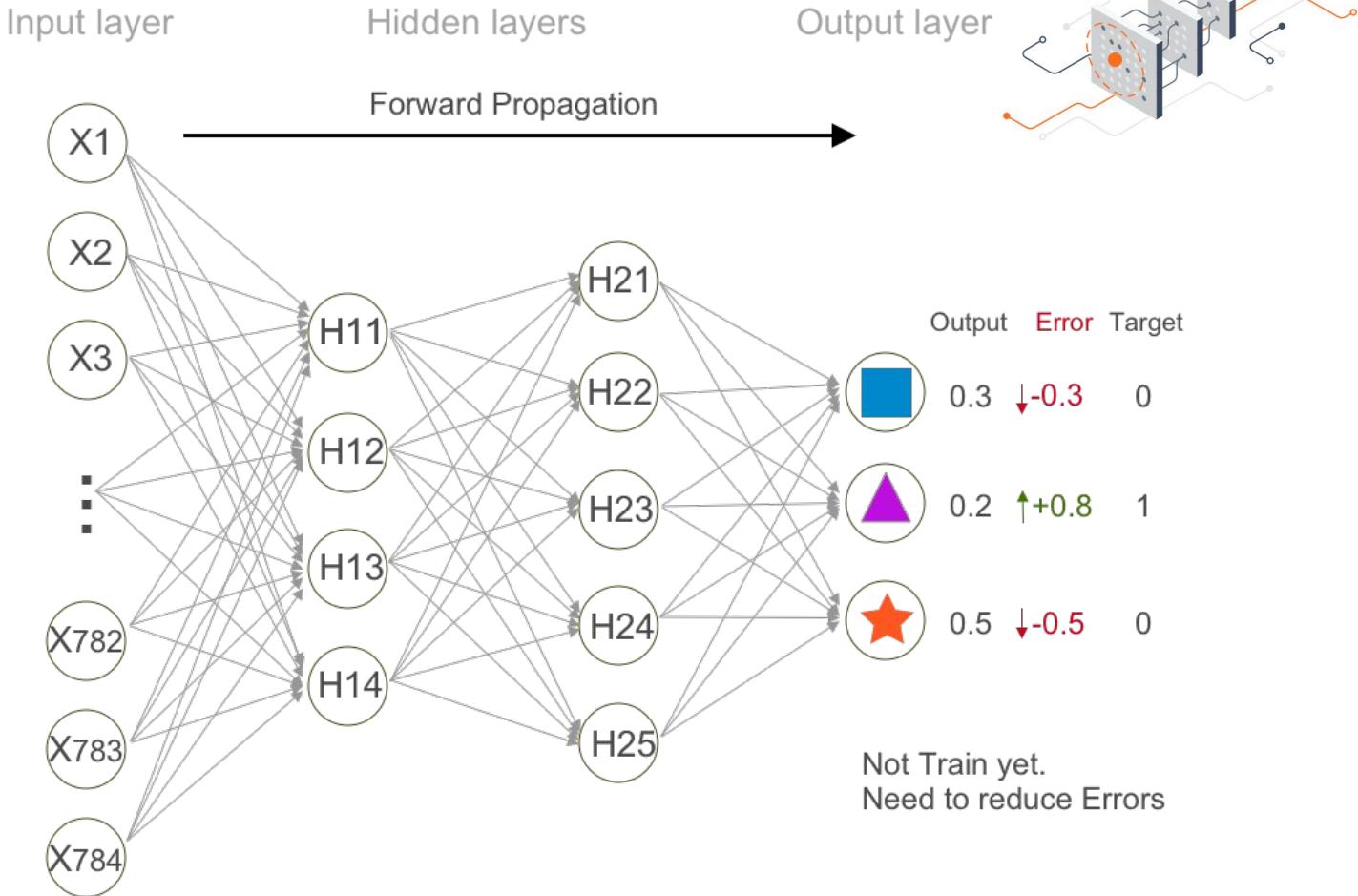
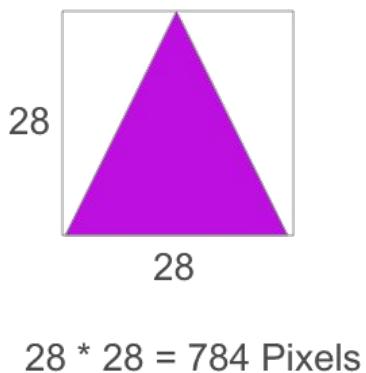
- Value between 0 - 1
- Useful for predicting probabilities as the output

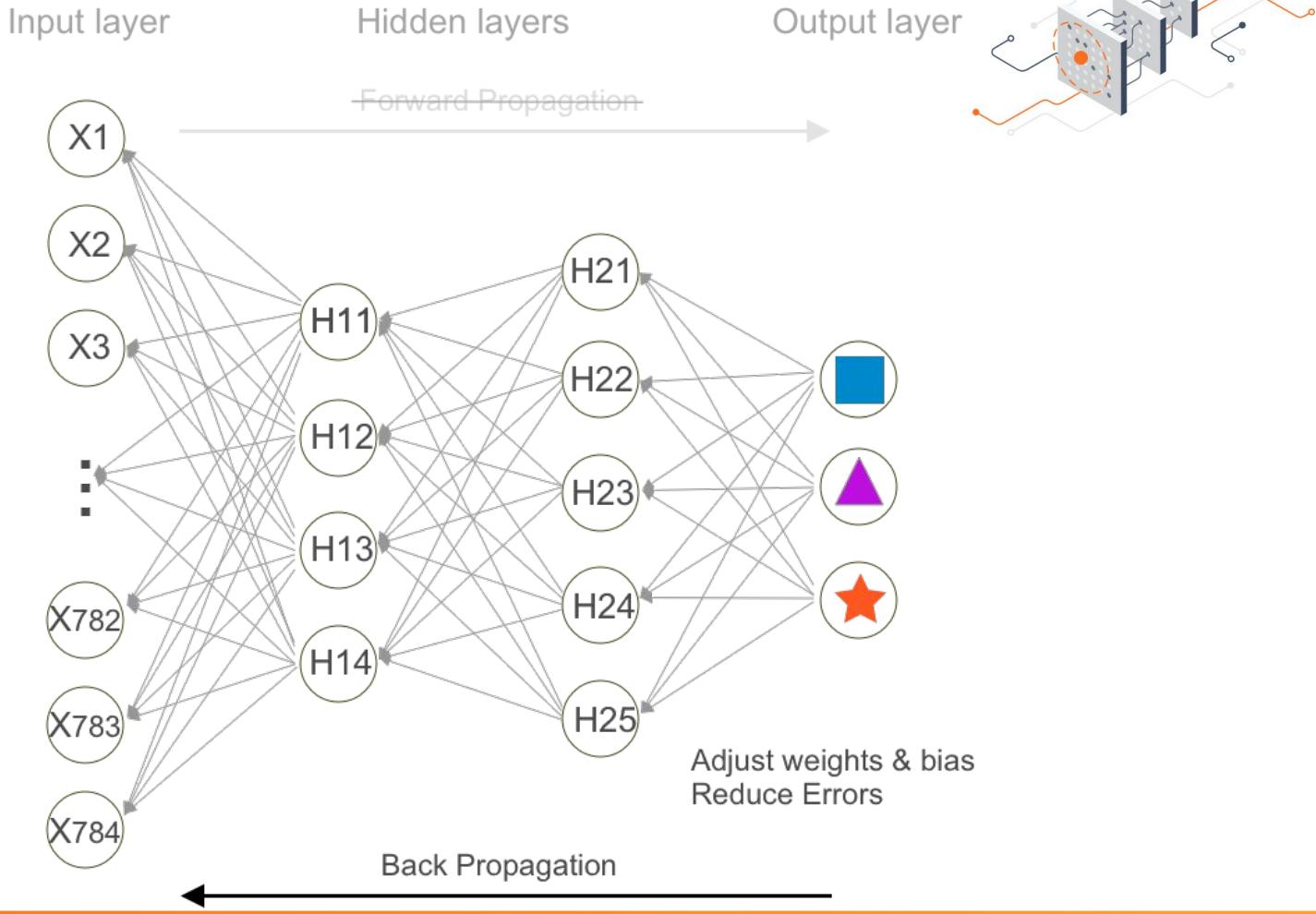
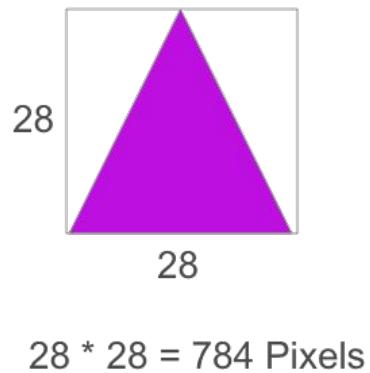


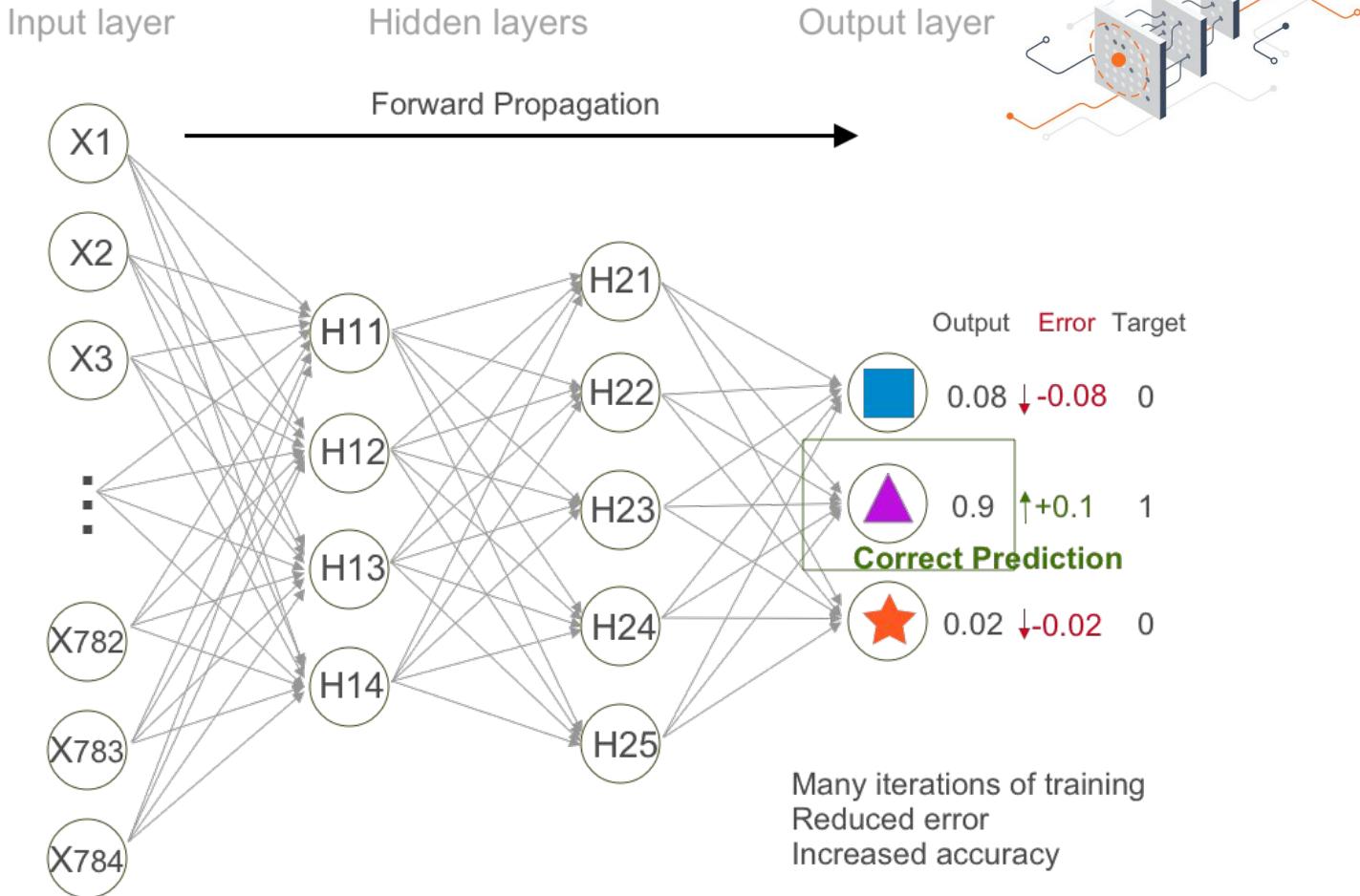
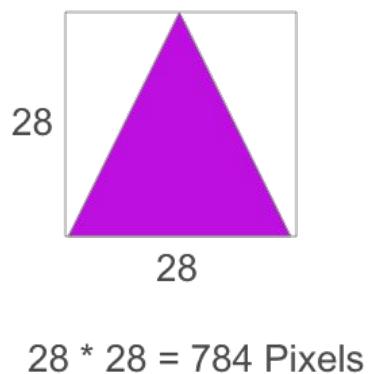
$(H11 * 2.5 + H12 * 4.2 +$   
 $H13 * 1.3 + H14 * -9.2) + 12.3 \rightarrow$  Activation Function  
Sigmoid Fn [0, 1]  $\rightarrow 0.34$





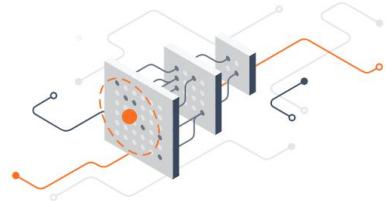




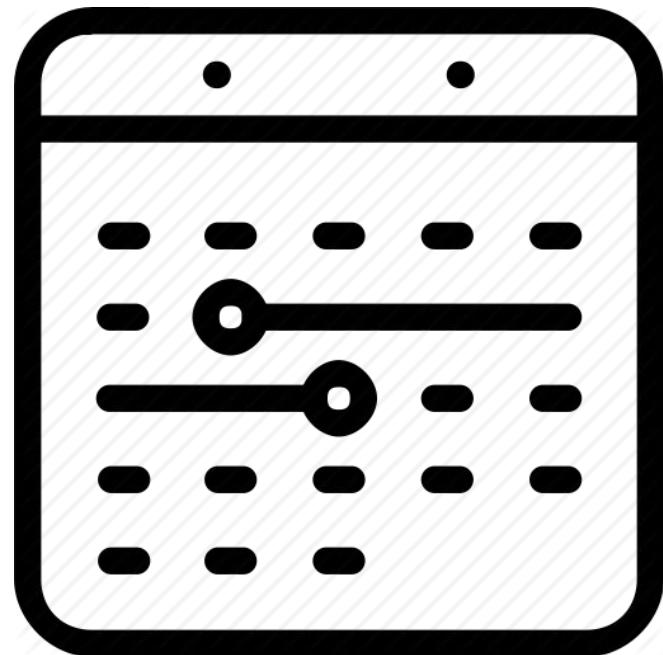




# How long does the training take?



Hours



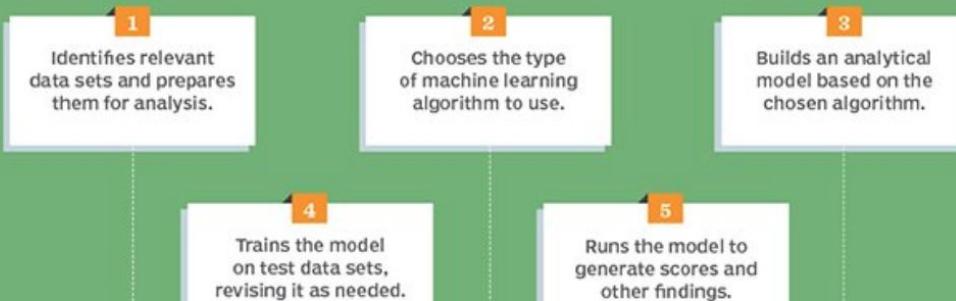
# ML Process

(5 simple steps)



# The machine learning process

In machine learning applications, a data scientist or other analyst:





# Step 1: Get Data



- Machine Learning depends largely on training data.
- Data preparation is a crucial step to make it suitable for ML.
- A large amount of data is generally required for the most common forms of ML.



# Step 1: Data selection

- Google Dataset Search <https://datasetsearch.research.google.com/>
- Kaggle Datasets
- UCI ML Repository
- HackerEarth ML Datasets
- Amazon Datasets
- Microsoft Datasets
- Other sources (e.g. Government Datasets)

Google



Dataset Search

🔍

Try [boston education data](#) or [weather site:noaa.gov](#)

[Learn more](#) about including your datasets in Dataset Search.



## Step 2: Preprocess Data

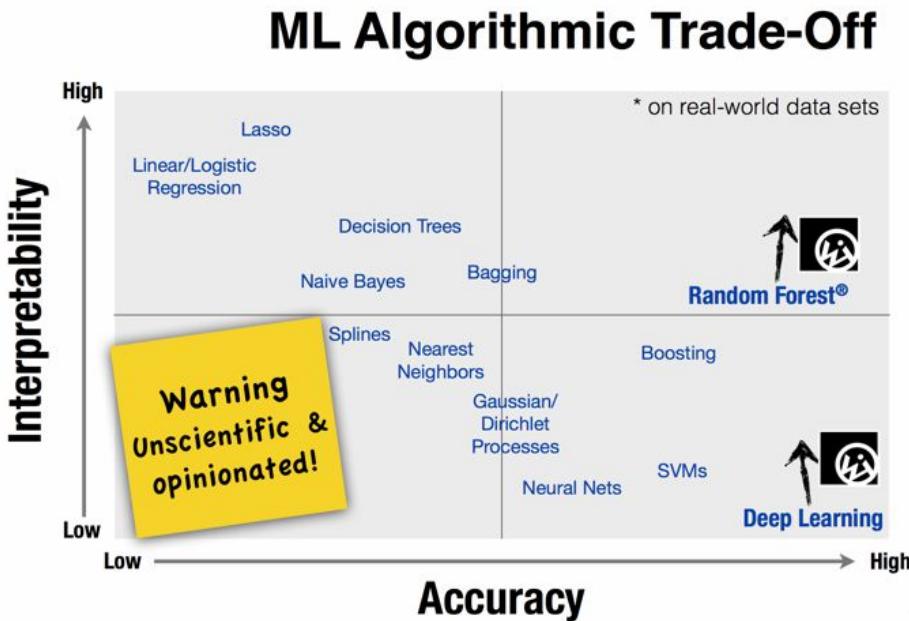
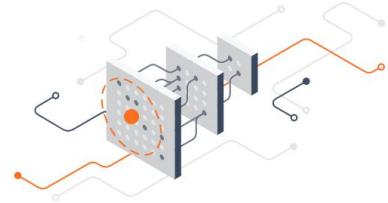


- Data preparation involves data selection, filtering, transformation, etc
- Formatting data to make it suitable for ML (structured format)
- Cleaning data to remove incomplete variables
- Sampling data further to reduce running times for algorithms and memory requirements.
- Separating data into training dataset, validation dataset and testing dataset
- Visualize Data using plots
- Data transformation methods: Scaling, Aggregation, etc



# Step 3: Choose an Algorithm

- The algorithm used to train your model
- Choose wisely
- Popular libraries like Scikit-Learn or Keras
- Commonly used ML algorithms: Linear regression, Logistic regression, Decision trees, K-means, etc





# Step 4: Build & Compile Model



- Build a model layer by layer
- Each layer has weights that correspond to the layer it follows
- Activation function for layer. It allows models to take into account nonlinear relationships
- Compiling the model takes parameters: optimizer, loss function, etc.

```
model.add (  
    keras.layers.Dense(10, activation='relu')  
    ...  
)
```

```
model.compile (  
    optimizer, loss, metrics, weights, ...  
)
```



## Step 5: Train the model



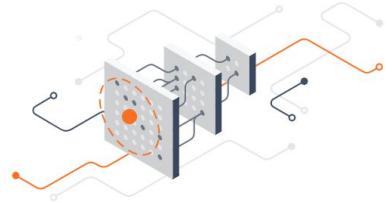
- Making a prediction based on the current state of the model
- Calculating how incorrect the prediction is
- Updating the weights or parameters of the network to minimize this error and make the model predict better
- Repeatedly iterating over the entire dataset for a given number of "epochs" until our model has converged(or define your own callback!)

```
model.fit (   
    x, y, epochs=EPOCHS,verbose=VERBOSE,  
    callback=CALLBACK  
)
```



## Step 6: Test the model

- See how well your model has learned the dataset during training
- Prediction result

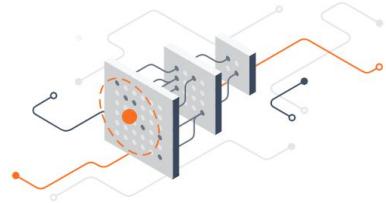


```
model.predict (   
    predict values go here  
)
```



---

REPEAT



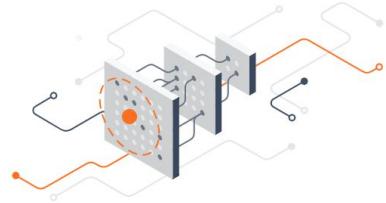


# Human Friendly Way



# Who Use ML?

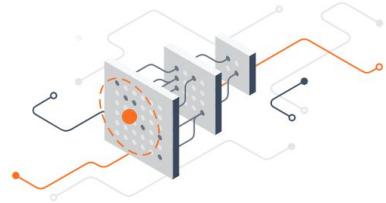
## 1. Researchers





# Who Use ML?

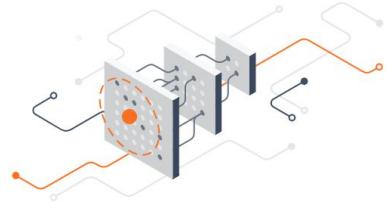
1. Researchers
2. Data scientists





# Who Use ML?

1. Researchers
2. Data scientists
3. ML engineers

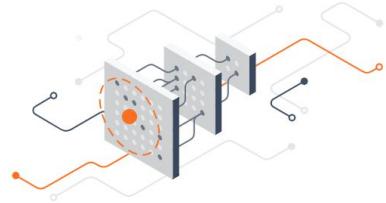






# Who Use ML?

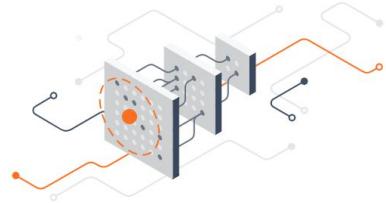
1. Researchers
2. Data scientists
3. ML engineers
4. Developers

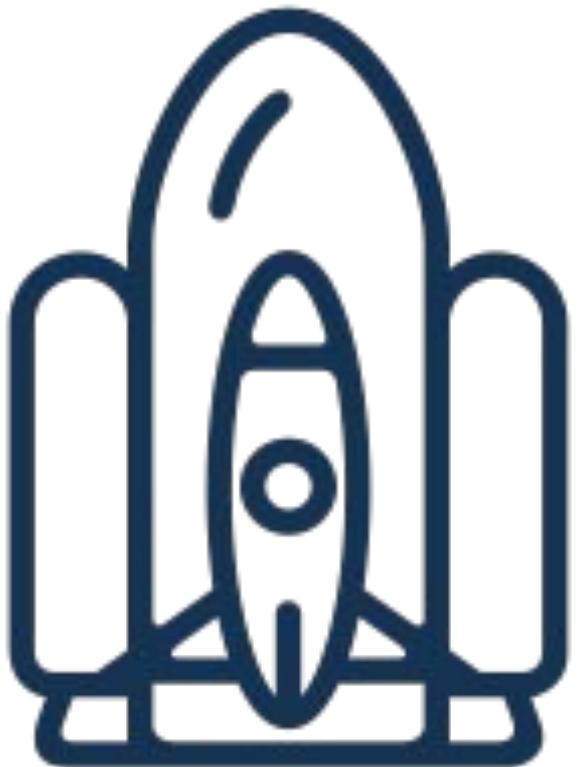




# Who Use ML?

1. Researchers
2. Data scientists
3. ML engineers
4. Developers
5. Me





# 2ND STAGE SEPARATION!

# Traditional Learning

vs

# Machine Learning





@aaronhma



GOOD  
JOB



# Traditional Software Development VS ML

// Traditional Software Development

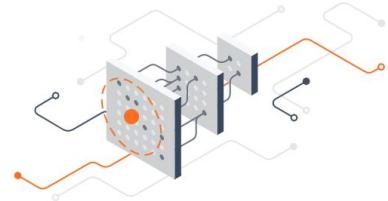
The input and the algorithm is known, and you write a function to produce an output

- Input data
- Apply logic to it
- Which produces a result
- Focused on code

// Machine Learning

You know the input and the output, but you do not know the algorithm that creates the output given the input.

- Take pairs of input and output data
- Figure out the algorithm
- Focused on data





# Traditional Software Development VS ML



// Traditional Software Development

```
def cToF(celsius):  
    f = celsius * 1.8 + 32  
    return f
```

```
cToF(0); // return fahrenheit value 32.  
cToF(38); // return fahrenheit value 100.4.
```



# Traditional Software Development VS ML



## // Traditional Software Development

```
def cToF(celcius):  
    f = celcius * 1.8 + 32  
    return f
```



```
cToF(0); // return fahrenheit value 32.  
cToF(38); // return fahrenheit value 100.4.
```

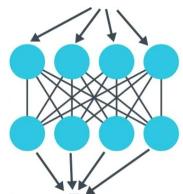


# Traditional Software Development VS ML

// Machine Learning

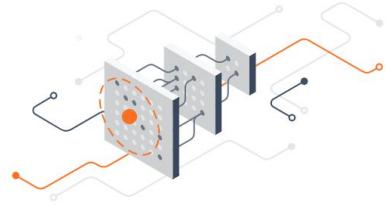
```
inputs = [0, 8, 15, 38]
outputs = [32, 46.4, 59, 100.4]
```

```
def cToF(inputs):
```



```
    return outputs
```

// A model. Algorithm:  $celsius * 1.8 + 32$



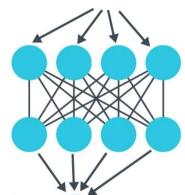


# Traditional Software Development VS ML

// Machine Learning

```
inputs = [0, 8, 15, 38] ← input  
outputs = [32, 46.4, 59, 100.4] ← output
```

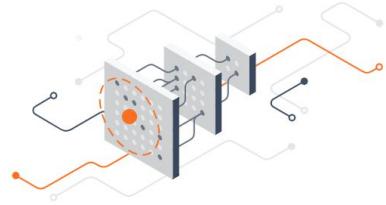
```
def cToF(inputs):
```



```
← Iterations of training to find algorithm
```

```
    return outputs
```

```
// A model. Algorithm: celsius * 1.8 + 32
```

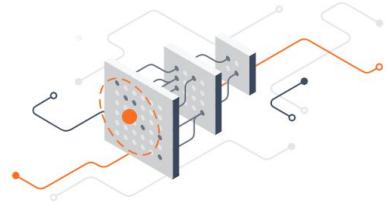




# Simple Scenario



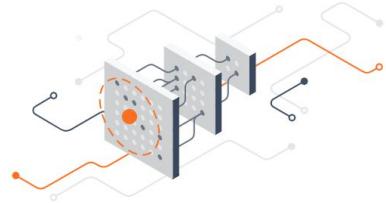
# Scenario: Different language to say Hello?



ALOHA HEI ZDRAVO SALUT  
TIENS A HI DIA DUIT MERHABA  
ALOHA Cześć HEJ HEJSAN  
SVEIKI SERVUS Cześć  
Bună ziua HEI  
SELAM HEJSAN  
Dzień dobry SZIA  
GUTEN TAG HEJ  
BONJOUR SZERVUSZ  
CIAO BONGHJORN  
DAR FIA PRONTO  
VERWELKOMMING  
MERHABA  
Hylö HALLO  
Olá SALVE  
Allô Allô



Hello in English

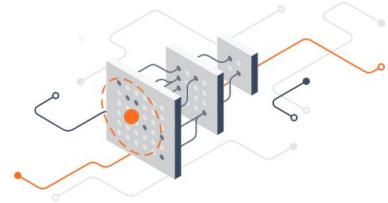


# Hello

```
if (text == 'hello') {  
    return language = 'English';  
}  
}
```



Hello in French

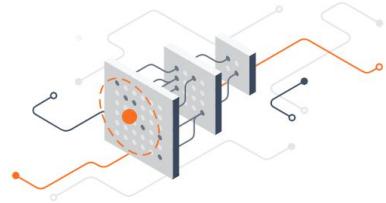


# Bonjour

```
if (text == 'hello') {  
    return language = 'English';  
} else {  
    return language = 'French';  
}
```



Hello in Spanish



Hola

```
if (text == 'hello') {  
    return language = 'English';  
} else if (text == 'bonjour') {  
    return language = 'French';  
} else {  
    return language = 'Spanish';  
}
```



Which language is it? Hello in ?



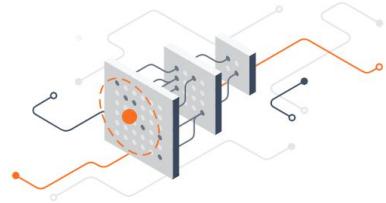
你好

? ! ! ?





# Traditional Programming VS ML







# Now what?



# Machine Learning





```
counter = 1
start_time = time.time()

if self.load(self.checkpoint_dir):
    print(" [*] Load SUCCESS")
else:
    print(" [!] Load failed...")

for epoch in xrange(config.epoch):
    if config.dataset == 'mnist':
        batch_idxs = min(len(data_X), config.train_size) // config.batch_size
    else:
        data = glob(os.path.join("./data", config.dataset, "*.jpg"))
        batch_idxs = min(len(data), config.train_size) // config.batch_size

    for idx in xrange(0, batch_idxs):
        if config.dataset == 'mnist':
            batch_images = data_X[idx*config.batch_size:(idx+1)*config.batch_size]
            batch_labels = data_y[idx*config.batch_size:(idx+1)*config.batch_size]
        else:
            batch_files = data[idx*config.batch_size:(idx+1)*config.batch_size]
            batch_images = np.array([imread(batch_file, self.image_size, keep_ratio=False).astype(np.float32) for batch_file in batch_files])
```

```
(training_data, training_label), (test_data, test_label) = source.load()

model = keras.models.Sequential([
    keras.layers.Dense(128, activation=nn.relu),
    ...
]);
model.compile(
    optimizer='adam',
    loss='sparse_categorical_crossentropy',
    metrics='accuracy'
)

model.fit( training_data, training_label, epochs = 30 )
model.evaluate( test_data, test_label )
```

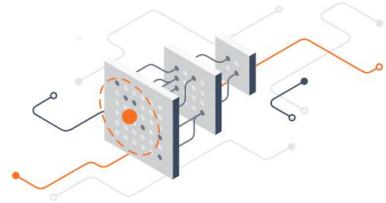


Which language is it? Hello in ?



你好

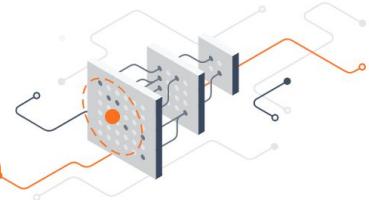
language = 'Chinese';



# Much Better!

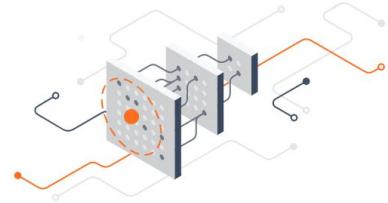
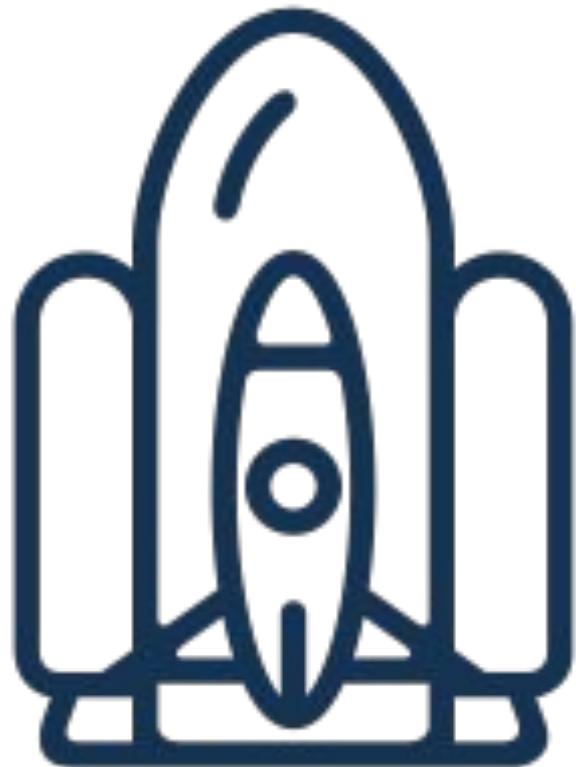


# BERT (Bidirectional Encoder Representations from Transformers)



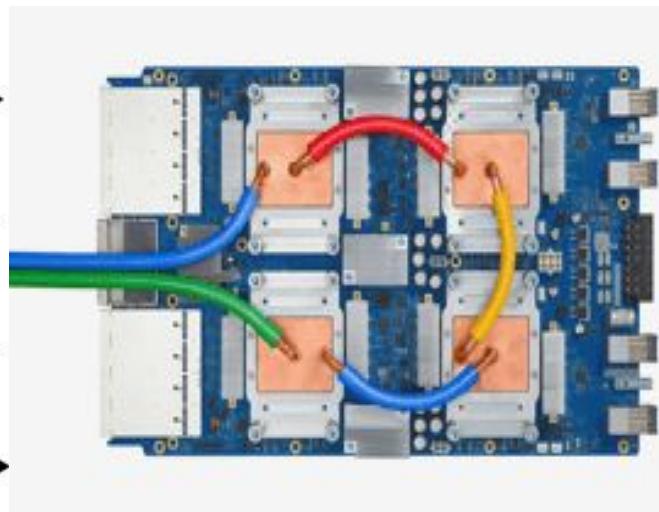
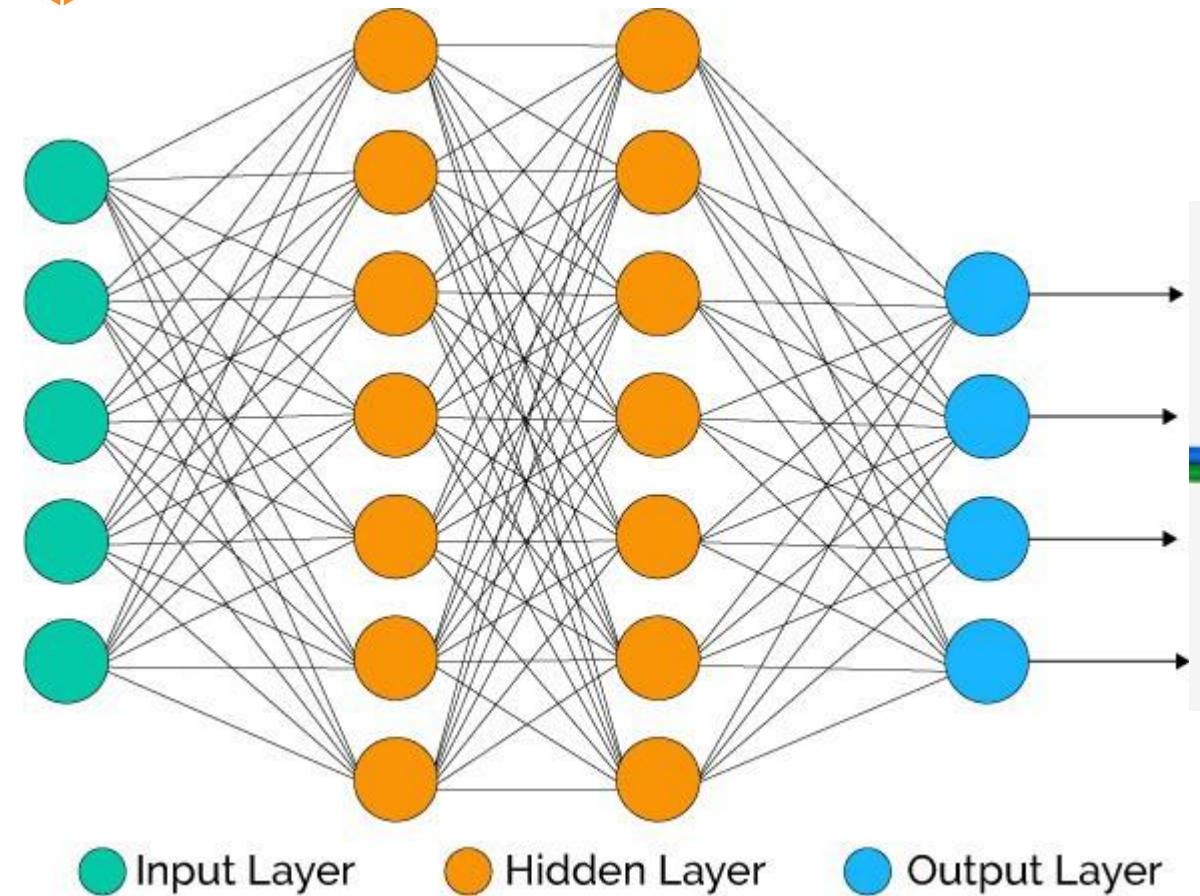
- State of the art language model for NLP
- Apply the bidirectional training of transformer, a popular attention model, to language modeling
- An attention mechanism that learns contextual relations between words/sub-words in a text
- A project by Google AI



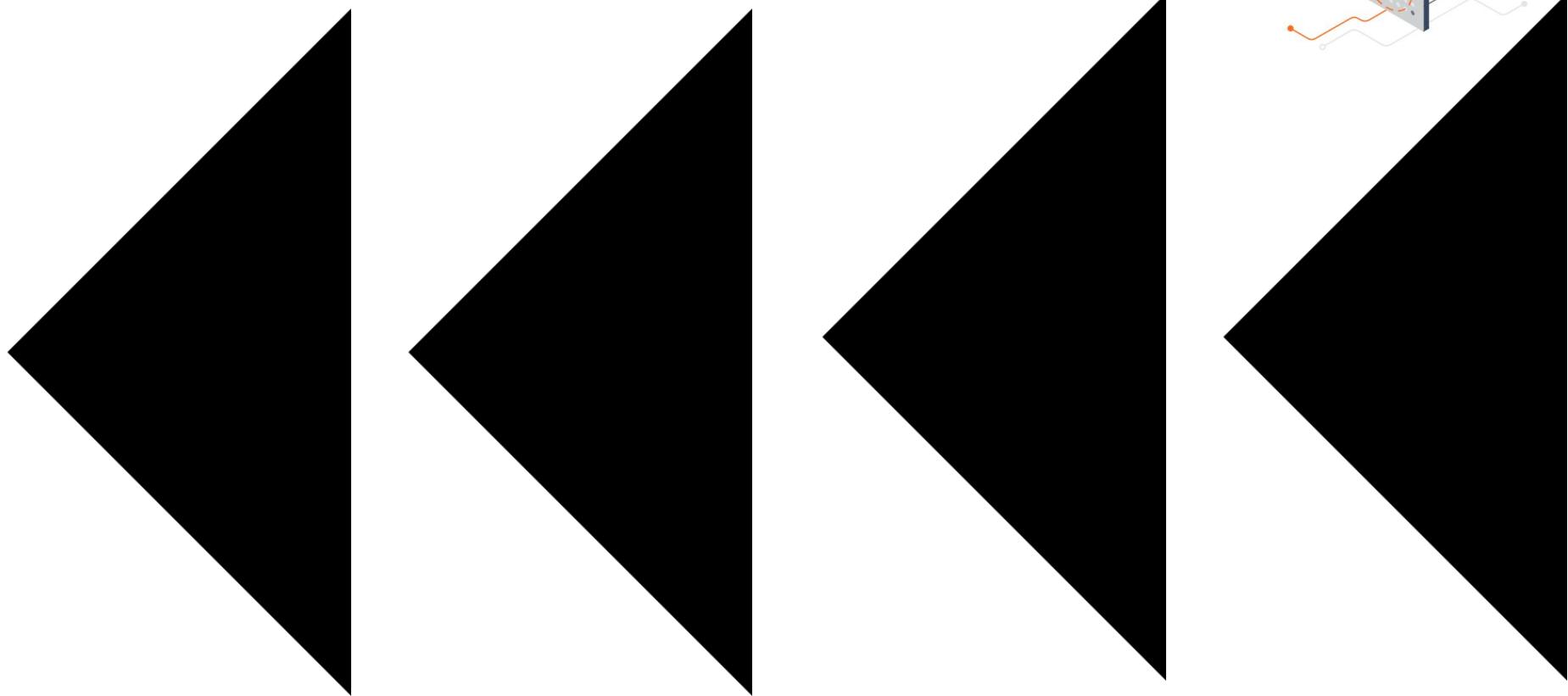


# 3RD STAGE IGNITION

# The History of ML









# A Brief History of ML



- 1940 - The idea of ML was introduced
- 1951 - First neural network by Marvin Minsky from MIT
- 1952 - First RL Agent by Arthur Samuel
- 1974 - First ML 'winter'
- 1996 - IBM's Deep Blue defeats chess champion, Garry Kasparov
- 2000 - Backpropagation for NN is introduced
- 2014 - DeepMind is formed
- 2016 - DeepMind's AlphaGo beats the best Go players
- 2030 - AI Robots taking over our jobs and daily tasks

uh-oh.





I can fix that.



# TensorFlow



# TensorFlow



- An open-source ML framework for everyone
- Built with C++
- 141K Github stars & >2400 contributors
- Powers nearly every single ML project @ Google
- No secret sauce
- Large framework
- Latest Release: 2.1 as of today



# The History of TensorFlow



- Originally titled “DistBelief” and Google Brain team started developing it in 2011
- Open-sourced in 2015
- TensorFlow 1.0 announced in 2017
- The most hottest ML library in 2017
- TensorFlow 1.3 announced in 2017 which included Keras!
- TensorFlow 2.0 announced in September of 2019
- TensorFlow 2.1 announced in January of 2020



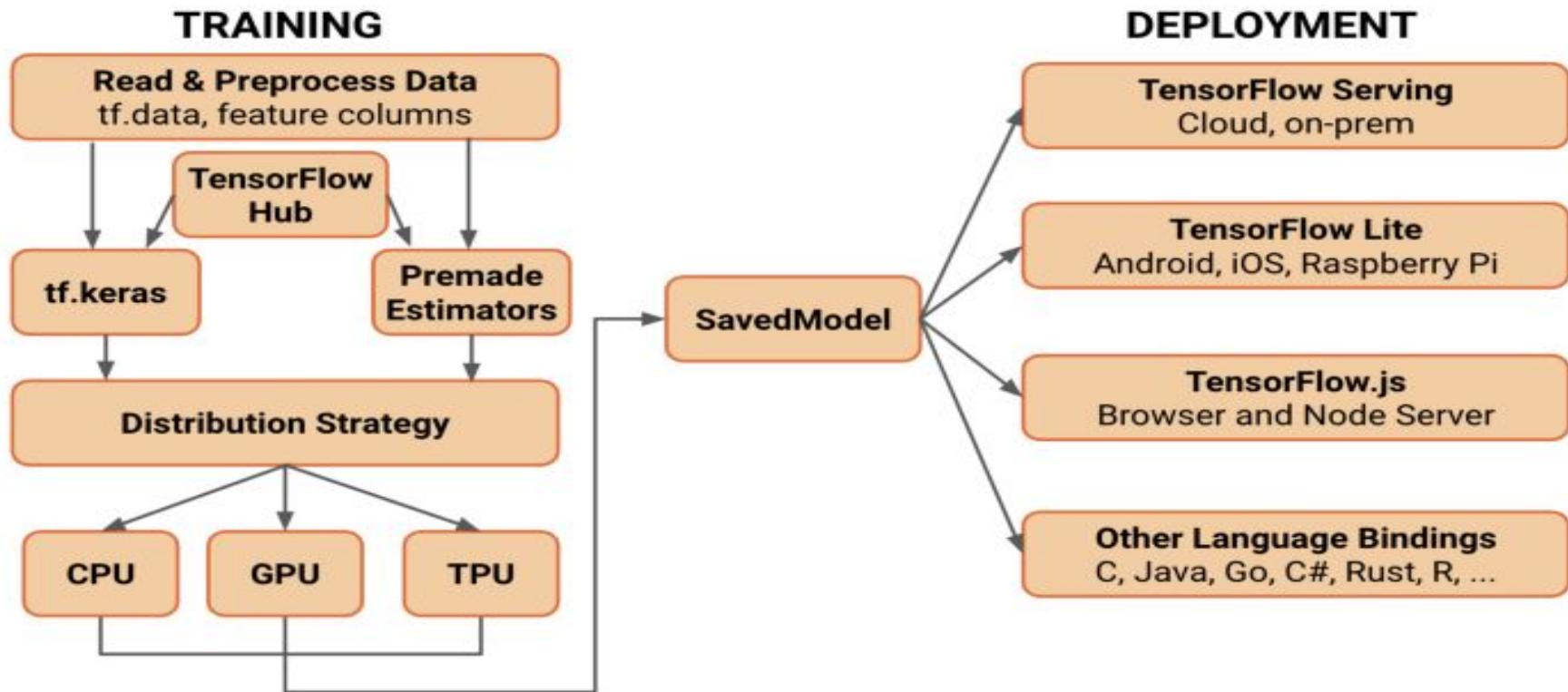
# What is new in TensorFlow 2.x?



- Eager execution by default (Pythonic way)
- Keras as the high level API (`tf.keras`)
- `tf.data` simplified API to read training data using input pipelines
- `tf.function` decorator translates Python programs into TensorFlow graphs. (No more `tf.Session()`)
- Still run TensorFlow 1.x code with 2.0 release (Pro tips: Do not do it!)



# TensorFlow 2.x Architecture



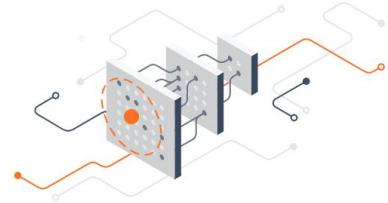
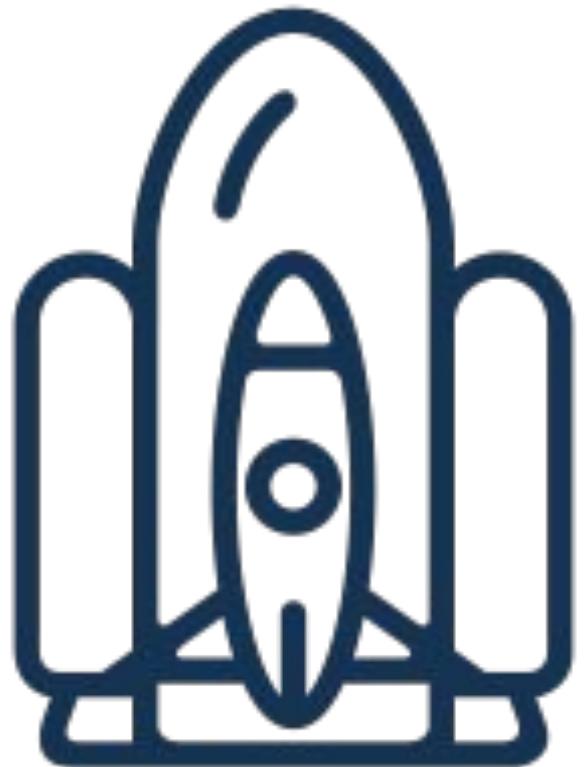




# Everyone

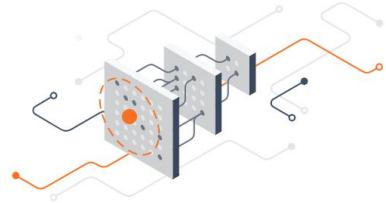


# TensorFlow!



# 3RD STAGE BURNOUT

# Other tools for ML



# Jupyter Notebook

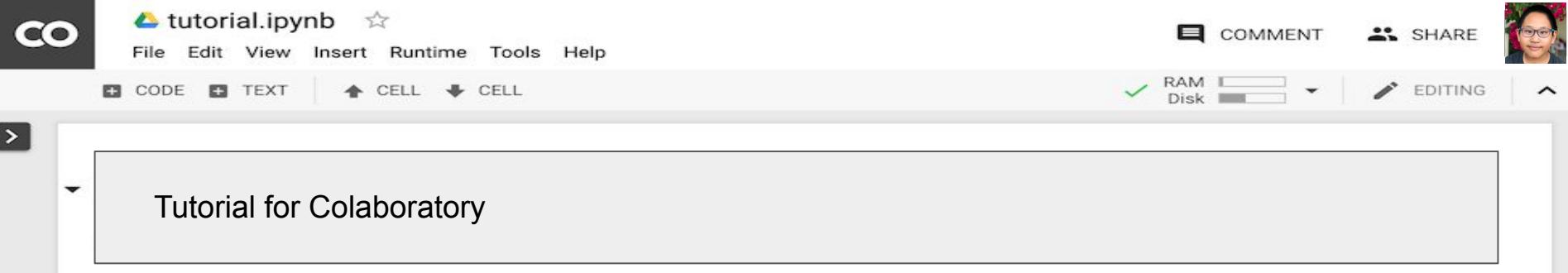


A screenshot of a Jupyter Notebook interface titled "Jupyter DataFrames". The notebook has a blue header bar with the title "Jupyter DataFrames" and a menu bar with "File", "Edit", "View", "Insert", "Cell", "Kernel", "Widgets", and "Help". The main content area shows a code cell with the following Python code:

```
#DataFrames in Python
In [ ] : This is pandas DataFrame|
In [ 6 ] : import pandas as pd
           import numpy as np
           df = pd.DataFrame(data=np.array([[1,2,3],[4,5,6]]), columns=['A','B','C'])
           df
```



# Google Colab

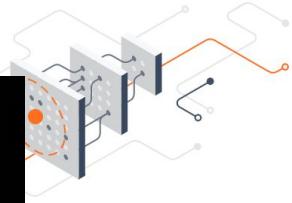


Ha, ha, ha! I've hidden the  
rest of the Colab!



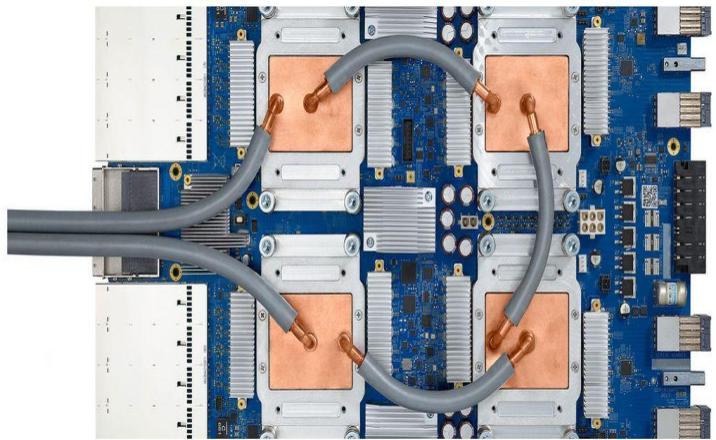
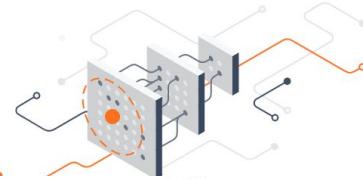


Oh Yes! IT'S  
**FREE**



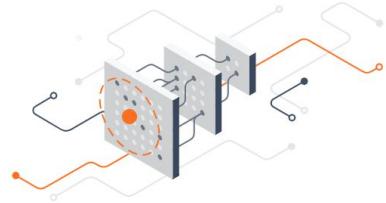


free  
stuff





# NumPy





# What is Numpy?



- Numerical Python
- Open source library that can be used for scientific and numerical computing
- Combination of C and Python
- Efficient computation on arrays
- Can process a homogeneous multi-dimensional array



# Pandas



# What is Pandas?



- High-performance data manipulation and analysis tool
- Load, prepare, manipulate, model, analyze, clean, transform (on Data)

```
In [20]: url = 'http://46.101.230.157/dilan/pandas_tutorial_read.csv'  
column_names = ['my_datetime', 'event', 'country', 'user_id', 'source', 'topic']  
pd.read_csv(url, delimiter=';', names = column_names)
```

```
Out[20]:
```

	my_datetime	event	country	user_id	source	topic
0	2018-01-01 00:01:01	read	country_7	2458151261	SEO	North America
1	2018-01-01 00:03:20	read	country_7	2458151262	SEO	South America
2	2018-01-01 00:04:01	read	country_7	2458151263	AdWords	Africa
3	2018-01-01 00:04:02	read	country_7	2458151264	AdWords	Europe
4	2018-01-01 00:05:03	read	country_8	2458151265	Reddit	North America
5	2018-01-01 00:05:42	read	country_6	2458151266	Reddit	North America
6	2018-01-01 00:06:06	read	country_2	2458151267	Reddit	Europe
7	2018-01-01 00:06:15	read	country_6	2458151268	AdWords	Europe
8	2018-01-01 00:07:21	read	country_7	2458151269	AdWords	North America
9	2018-01-01 00:07:29	read	country_5	2458151270	Reddit	North America
10	2018-01-01 00:07:57	read	country_5	2458151271	AdWords	Asia



# Human Friendly Way











```
[In]: df.drop(to_drop, inplace=True, axis=1)
```





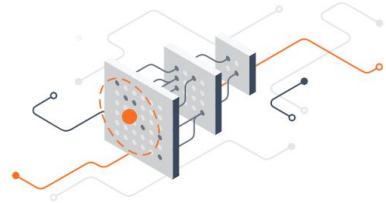


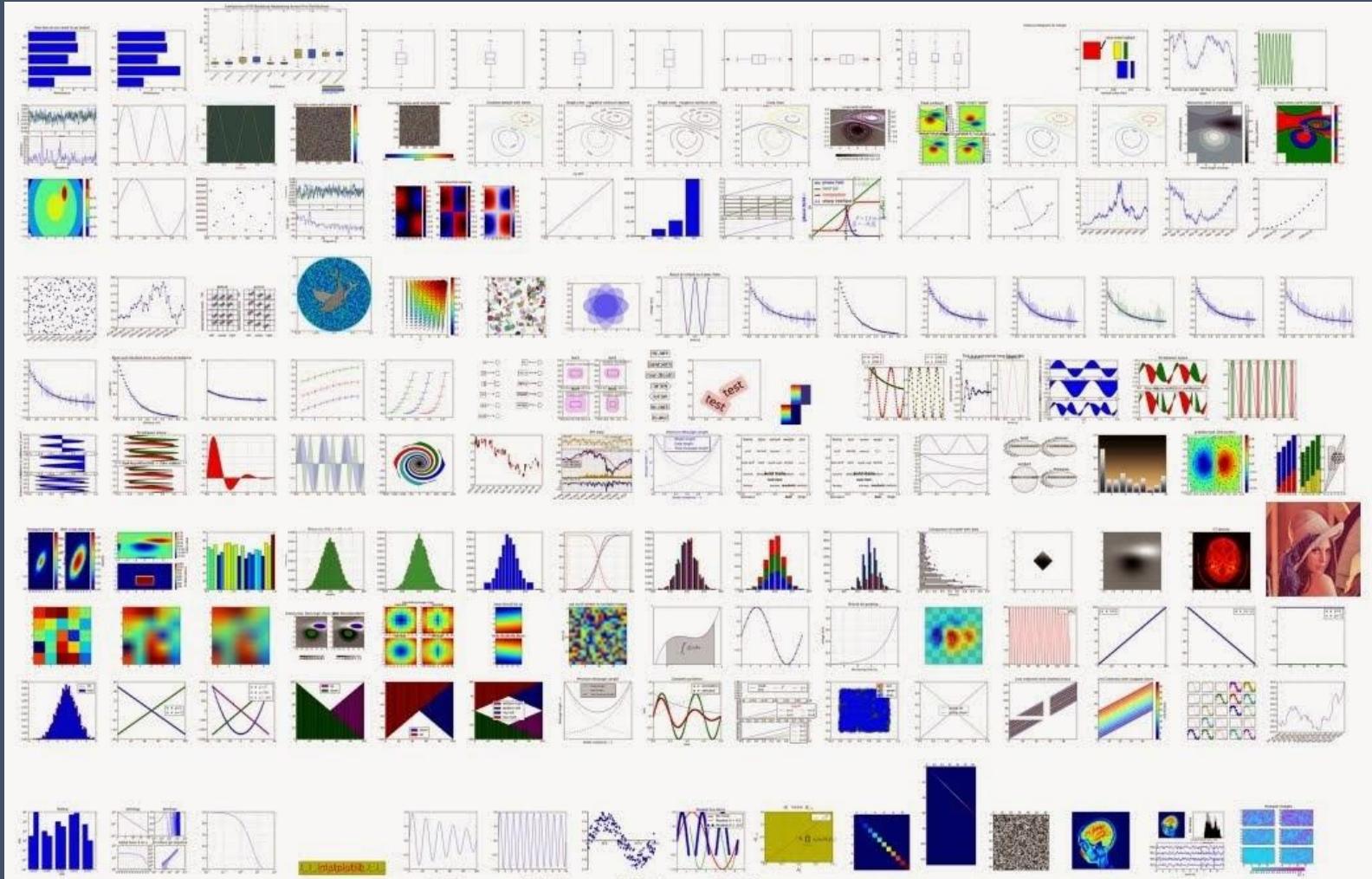
# Matplotlib



# What is Matplotlib?

- Multi-platform data visualization library for 2D plots of arrays
- Create high-quality graphs, charts, and figures



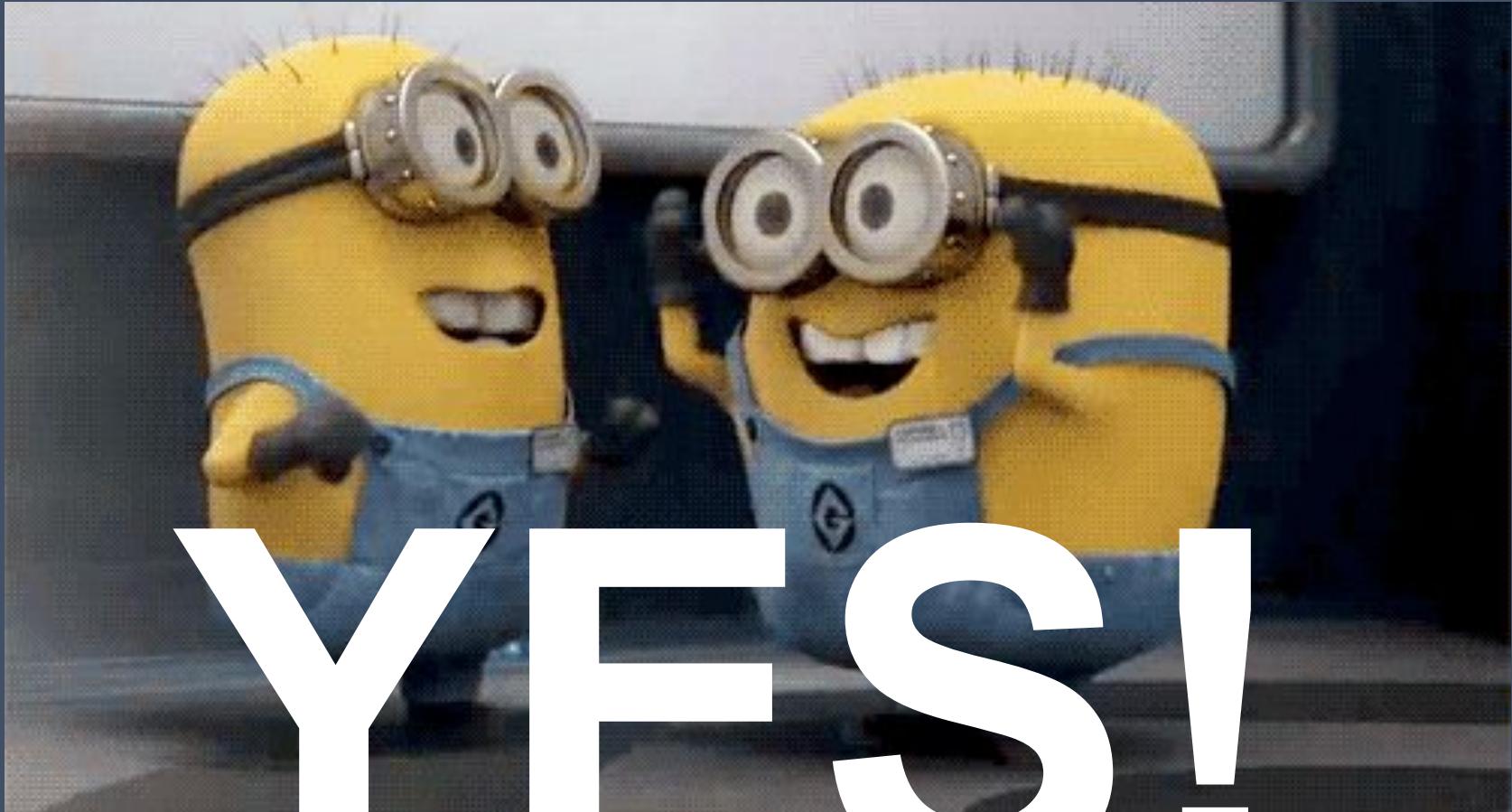


# Project 1: Fashion MNIST





# Coding?!?



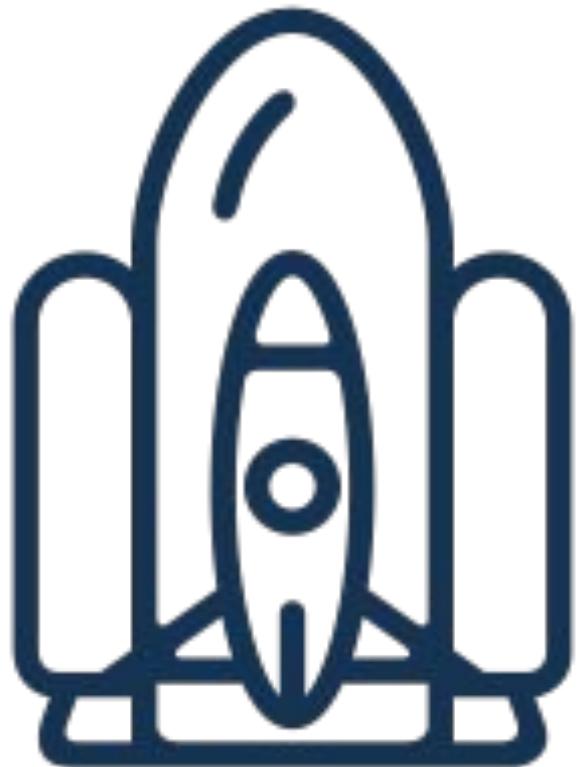
YES!



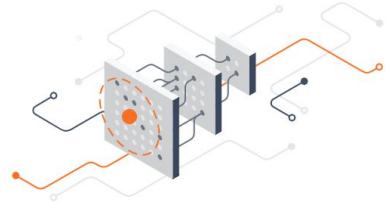
Demo

<https://bit.ly/pycon20-fm>





MOON LANDING!



# Grand Finale





# What If



# What If

# WHAT?????!!!!



# What If

## No Code required to create ML Model?



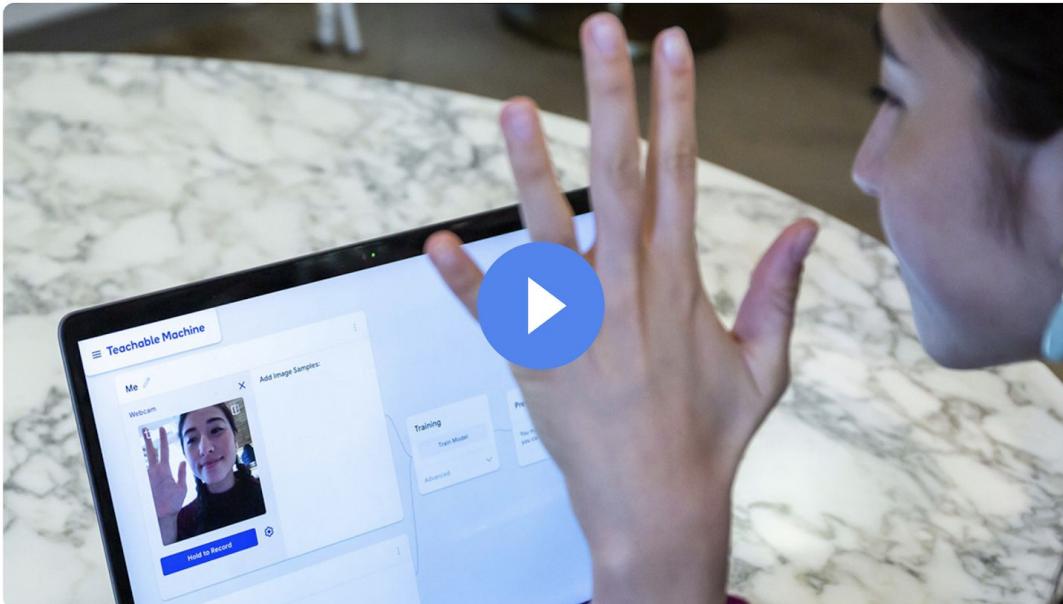
# What If

No Code required to create ML Model?  
No ML experience required?





## Teachable Machine



Teachable Machine is a web-based tool that makes creating machine learning models fast, easy, and accessible to everyone.

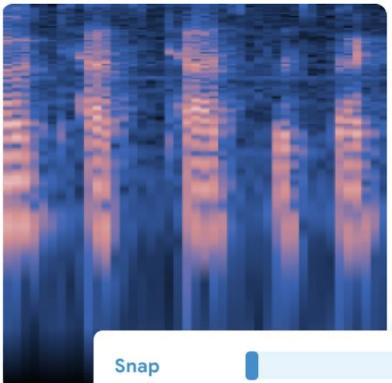


Me + Dog <3

Just Me

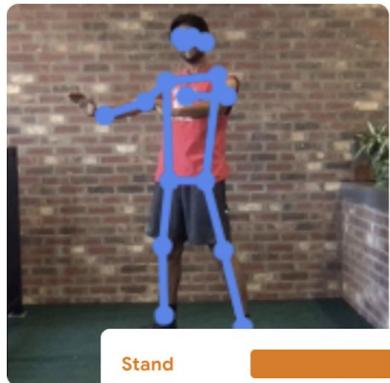
## Images

Teach a model to classify images using files or your webcam.



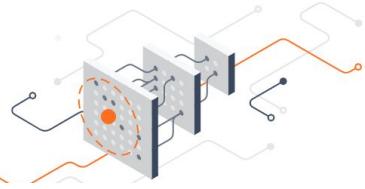
## Sounds

Teach a model to classify audio by recording short sound samples.  
(WAV/MP3/etc file support coming soon.)



## Poses

Teach a model to classify body positions using files or striking poses in your webcam.





Class 1



Class 2



## 1 Gather

Gather and group your examples into classes, or categories, that you want the computer to learn.

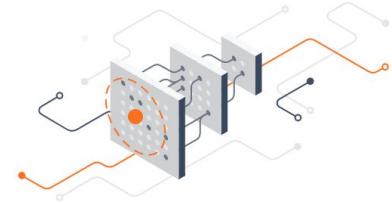
## 2 Train

Train your model, then instantly test it out to see whether it can correctly classify new examples.



## 3 Export

Export your model for your projects: sites, apps, and more. You can download your model or host it online for free.

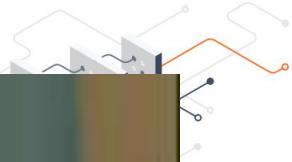


Demo

<https://bit.ly/pycon20-3>

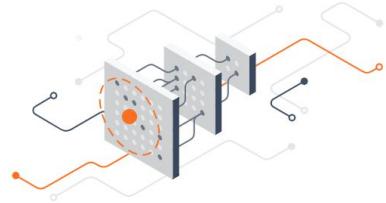


@aaronhma



# TAKEAWAYS

TAKEAWAYS EVERYWHERE



# Machine Learning is **Everywhere**



# Machine Learning is for Everyone

# WE WANT YOU





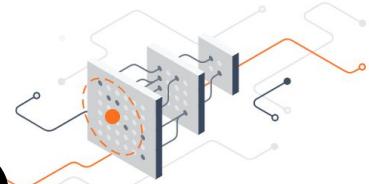
With a little help  
from:

Google

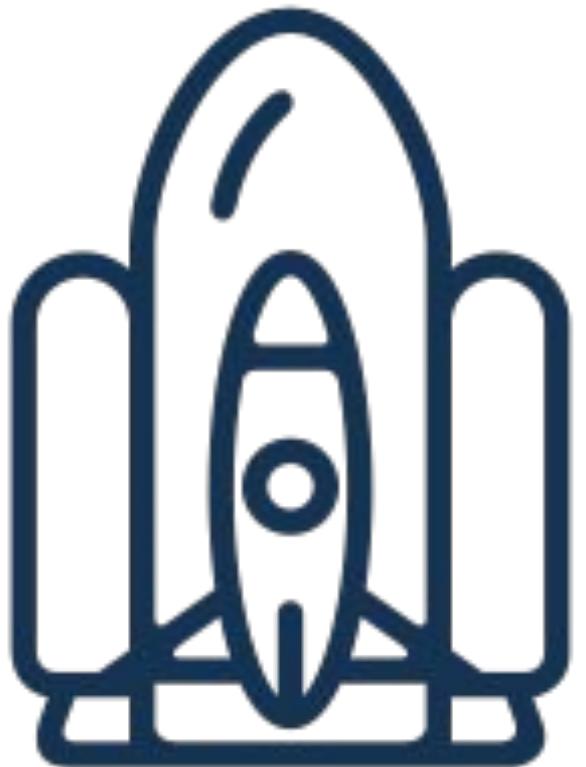




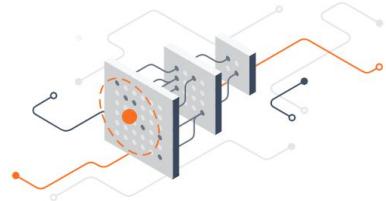
With a little help  
from:



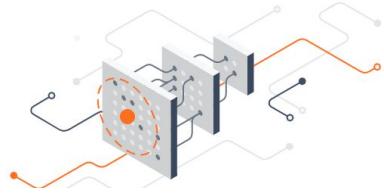
 TensorFlow



WE'RE BACK ON EARTH! YOU SURVIVED!



#PartyTime



**AI is akin to building a rocket ship. You need a huge engine and a lot of fuel. The rocket engine is the learning algorithms but the fuel is the huge amounts of data we can feed to these algorithms.**

**Andrew Ng**

*The Inevitable: Understanding the 12 Technological Forces That Will Shape Our Future*



# Next steps



- Official Tensorflow site: <https://www.tensorflow.org>
- Machine Learning Crash Course:  
<https://developers.google.com/machine-learning/crash-course/>
- Awesome Tensorflow: <https://github.com/jtoy/awesome-tensorflow>
- TensorFlow Playground: <https://playground.tensorflow.org>
- Teachable Machine: <https://teachablemachine.withgoogle.com/>
- ML Zero to Hero (YouTube):  
<https://www.youtube.com/playlist?list=PLQY2H8rRoyvwLbzbnKJ59NkZvQAW9wLbx>
- Towards Data Science (blog, commonly referred to as TDS):  
<https://towardsdatascience.com/machine-learning/home>

**NOW!** Give Yourself a round of applause!



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