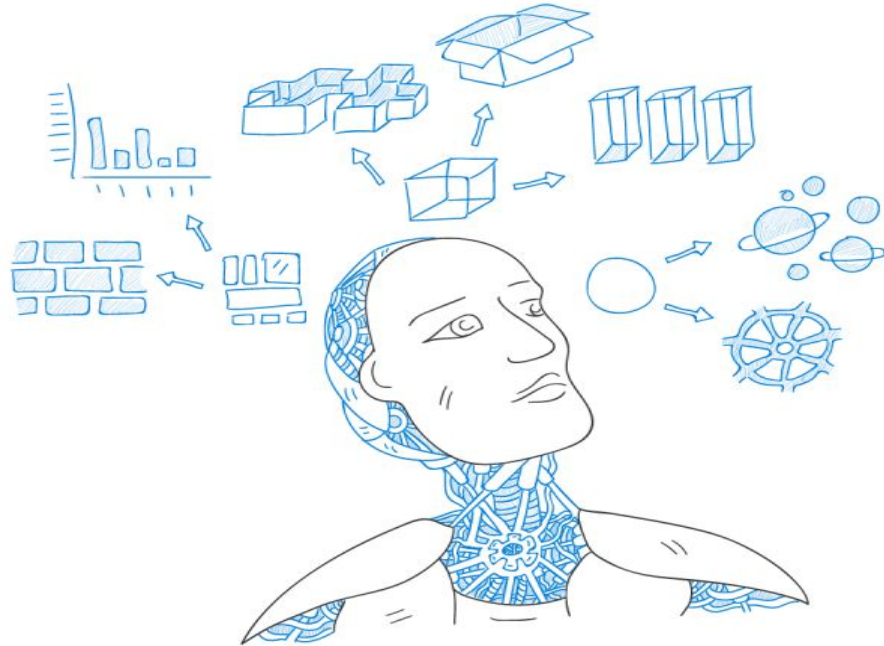
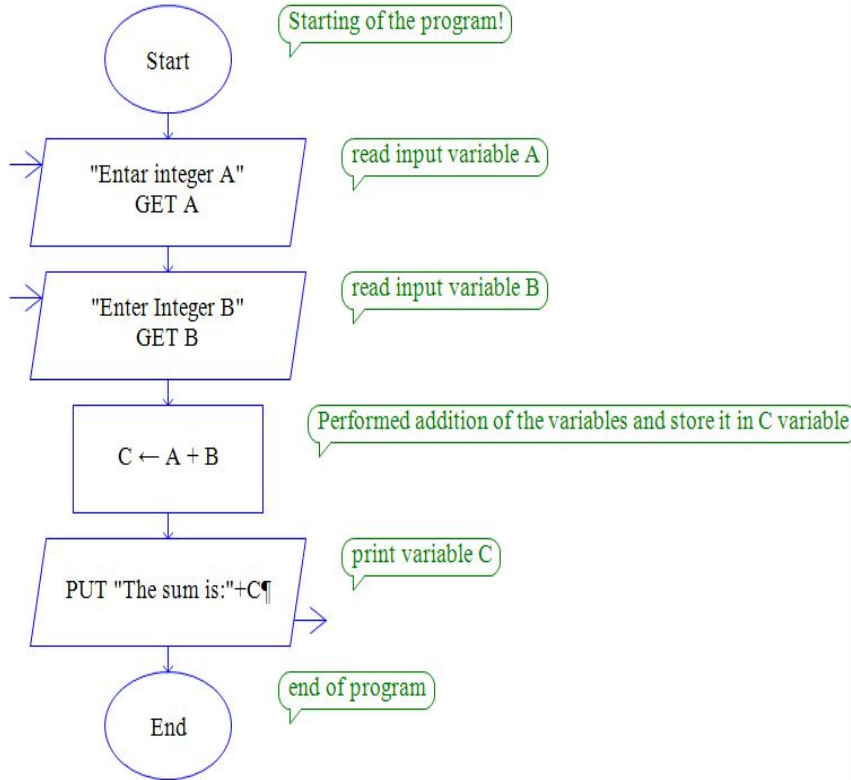


Introduction to Machine Learning



By-Ashish Vishwakarma
Mtech CSE IITBhilai



Write a program to classify between dog and a cat ?

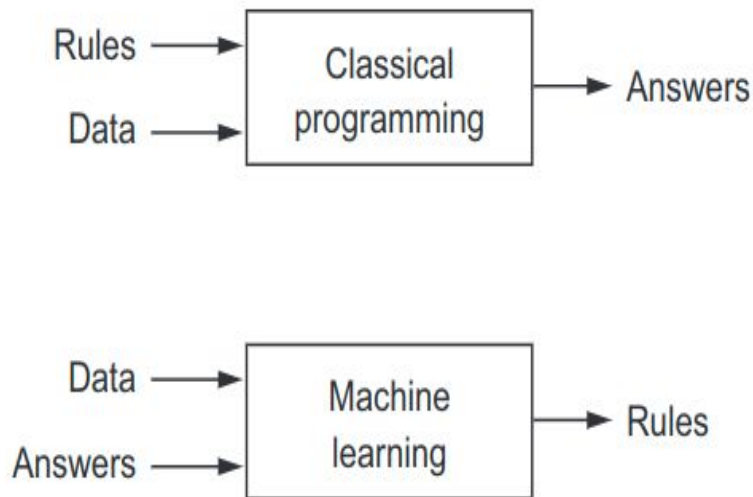


or



Machine learning: a new programming paradigm

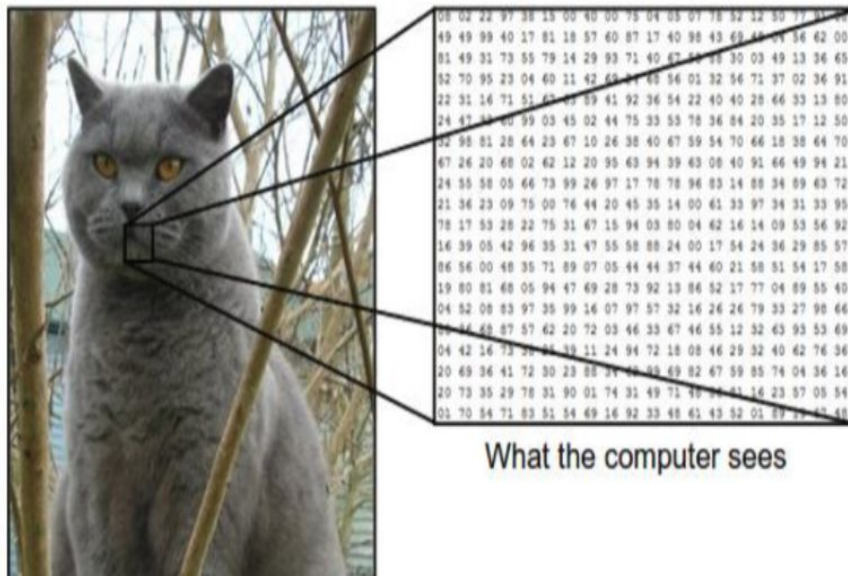
A computer program is said to learn from experience \mathcal{E} with respect to some class of tasks \mathcal{T} and performance measure \mathcal{P} , if its performance at tasks in \mathcal{T} , as measured by \mathcal{P} , improves with experience \mathcal{E}



When to Use Machine Learning ?

Can't code the rule: When rules depend on too many factors and many of these **rules overlap** or need to be tuned very finely, it soon becomes **difficult for a human to accurately code the rules**. You can use ML to effectively solve this problem.

Images are Numbers



What the computer sees

Can't scale : You might be able to manually recognize a few hundred images and decide whether they are dog or cat. However, this task becomes tedious for millions of emails. ML solutions are effective at handling large-scale problems.



You don't need ML if you can determine a target value by **using simple rules**, **computations**, or **predetermined steps** that can be programmed without needing any data-driven learning.

Examples:

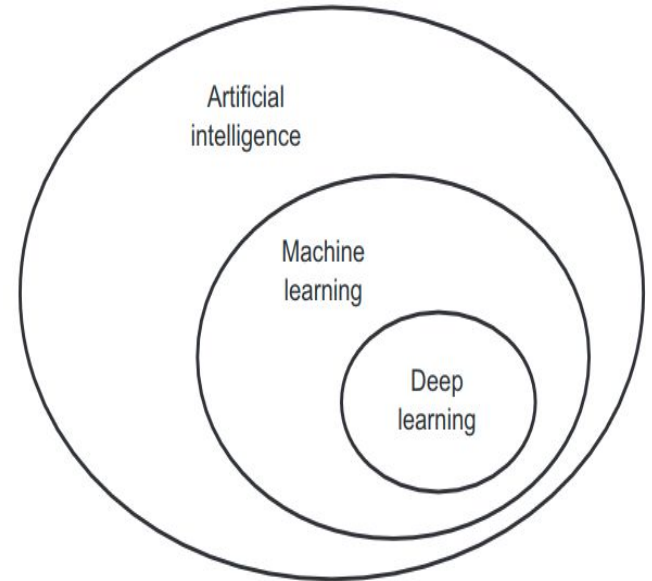
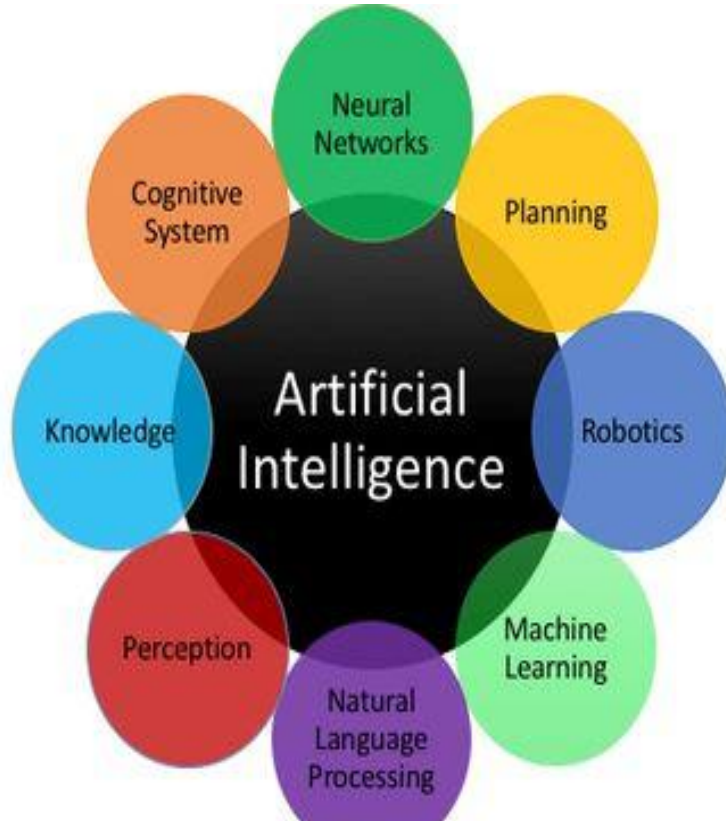
- Arithmetic operation ,eg : ADD,SUB,MUL,...
- Logical operation,eg: AND ,OR ,NOT,....
- Sorting numbers eg: Increasing ,Decreasing.
- Rule based eg: Direct Formulas($V=IR$),factorial,..

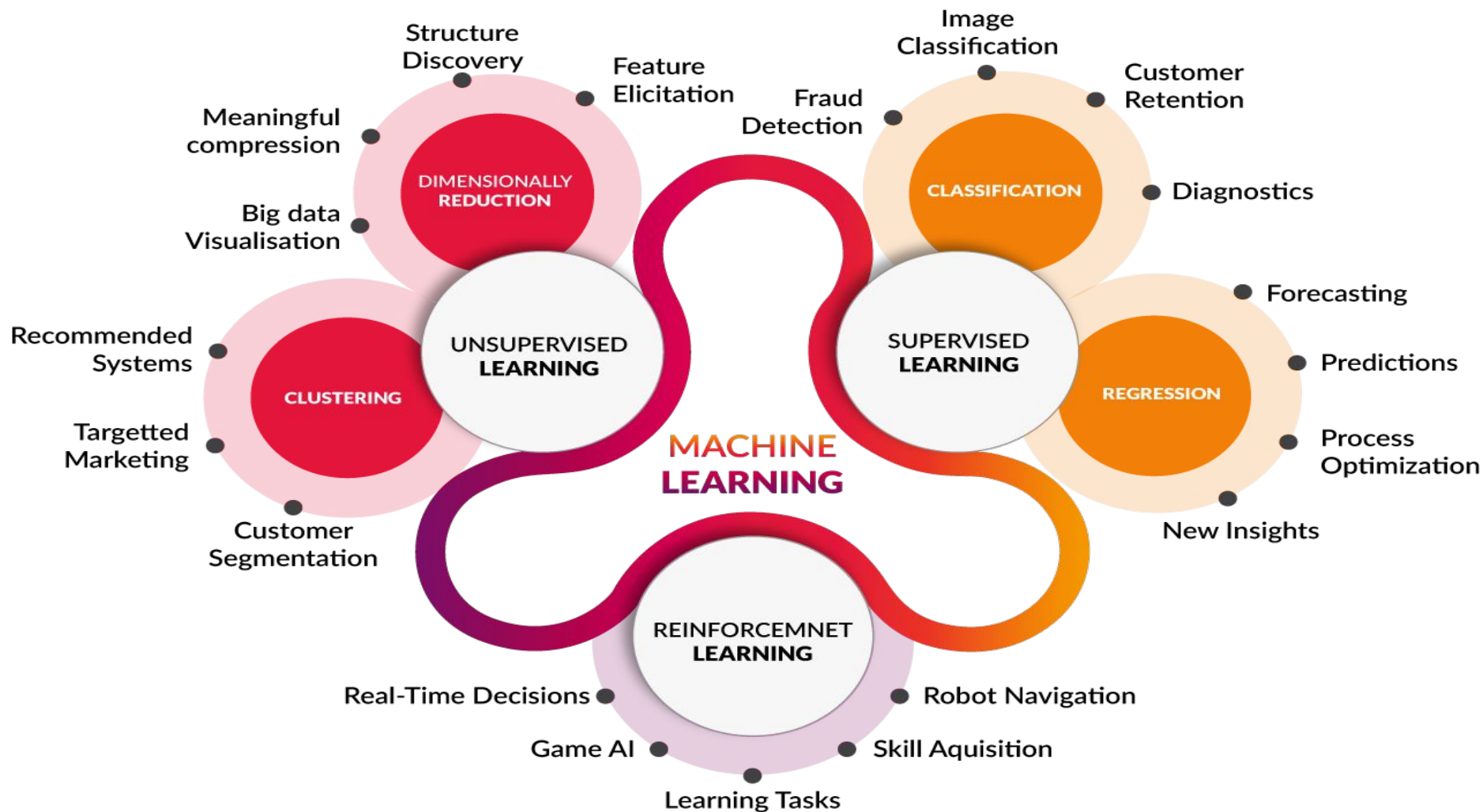
From where idea of ML come's from ?

*“AI began with an ancient wish to
forge the gods.”*

- Pamela McCorduck, *Machines Who Think*, 1979

An effort to automate intellectual tasks normally performed by humans.





Some achievement's:

- Near-human-level image classification.
- Near-human-level speech recognition.
- Near-human-level handwriting transcription.
- Improved machine translation.
- Improved text-to-speech conversion.
- Digital assistants such as Google Now and Amazon Alexa.
- Near-human-level autonomous driving.
- Improved ad targeting, as used by Google, Baidu, and Bing.
- Improved search results on the web.
- Ability to answer natural-language questions.
- Superhuman Go playing.

ML in basic sciences

ML In Condensed Matter Physics:

“Understanding the wavefunction of a many-particle quantum system with relevant accuracy which can pave the way for the designing of new quantum materials and devices.”

Researchers are **using ML algorithms to understand the phases of matter**, which could lead to theoretical breakthroughs in quantum bit or even silicon

<https://www.quora.com/profile/Juan-Felipe-Carrasquilla>

Deep Learning for Tracking in High Energy Physics:

<https://pdfs.semanticscholar.org/c2b3/951c35159a08d48c3624921927e98fd32302.pdf>

The researchers deployed an LSTM approach and proposed an **end-to-end solution for the HL-LHC track pattern recognition challenge**

<https://arxiv.org/pdf/1402.4735.pdf>

deep learning techniques improved the classification metric by 8 percent

ML In Astrophysics:

processing the wide amount of astronomical data generated which are often heterogeneous in nature.

gain data-driven insights and visualize it further

<https://arxiv.org/pdf/1411.5039.pdf>

References:

PY 895 Machine Learning for Physicists. Fall 2018 <http://physics.bu.edu/~pankajm/PY895-ML.html>

A high-bias, low-variance introduction to Machine Learning for physicists

<https://arxiv.org/pdf/1803.08823.pdf>

<https://www.quora.com/What-are-the-connections-between-machine-learning-and-physics>

Junction Tree Variational Autoencoder for Molecular Graph Generation:

automate the design of molecules based on specific chemical properties

<https://arxiv.org/abs/1802.04364>

Using Deep Reinforcement Learning to Generate Rationales for Molecules:

Learn to identify molecular substructures – rationales – that are associated with the target chemical property.

<http://chem.csail.mit.edu/deep-reinforcement-learning.pdf>

Predicting Organic Reaction Outcomes with Weisfeiler-Lehman Network

Prediction of Organic Reaction Outcomes Using Machine Learning

Convolutional Embedding of Attributed Molecular Graphs for Physical Property Prediction

Automating drug discovery

Computational Protein Design with Deep Learning Neural Networks

References:

<http://chem.csail.mit.edu/papers>

Representation, optimization and generalization properties of deep neural networks

Robustness in unsupervised and supervised machine

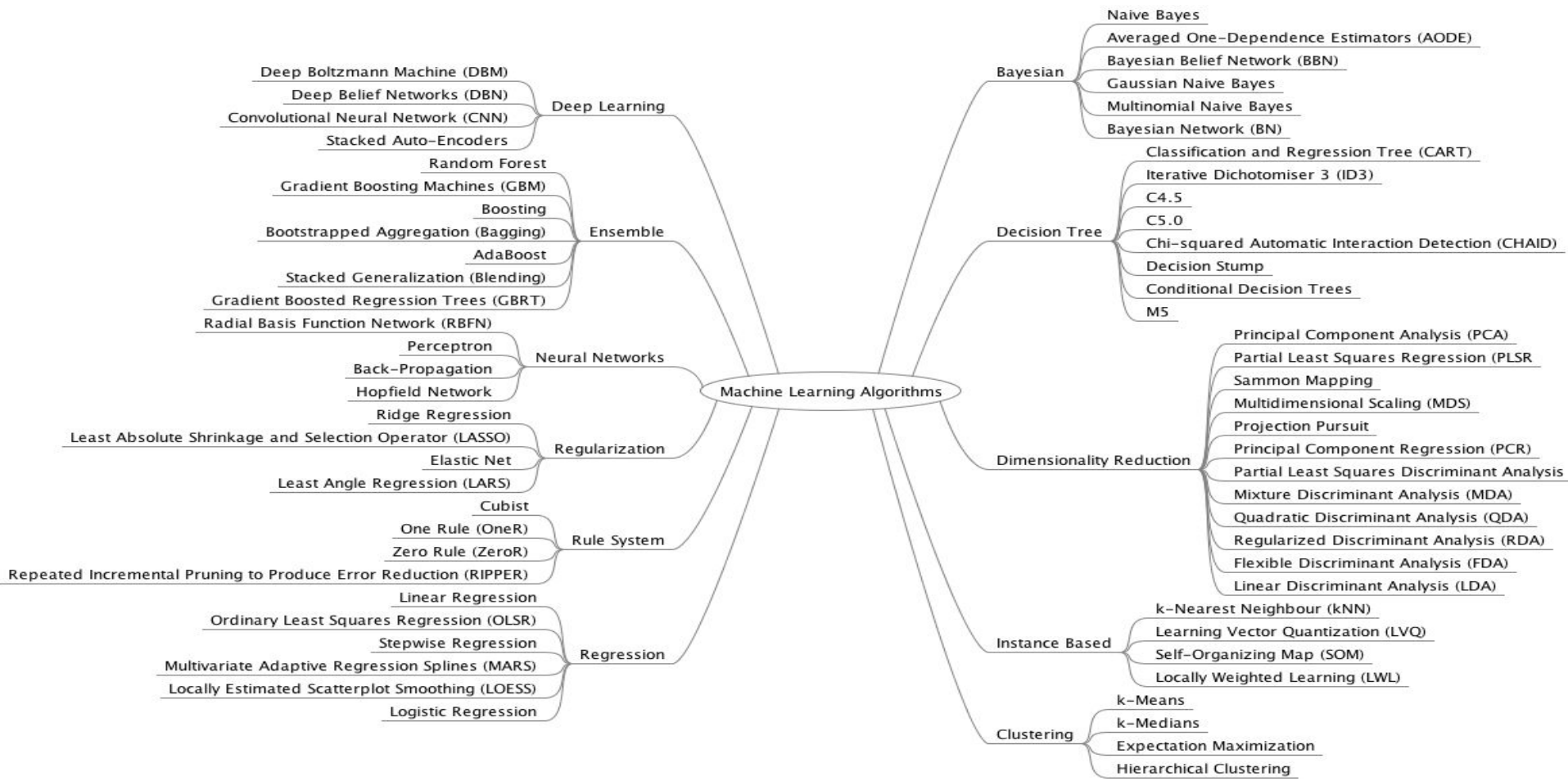
A modular analysis of adaptive online (non-)convex optimization

References:

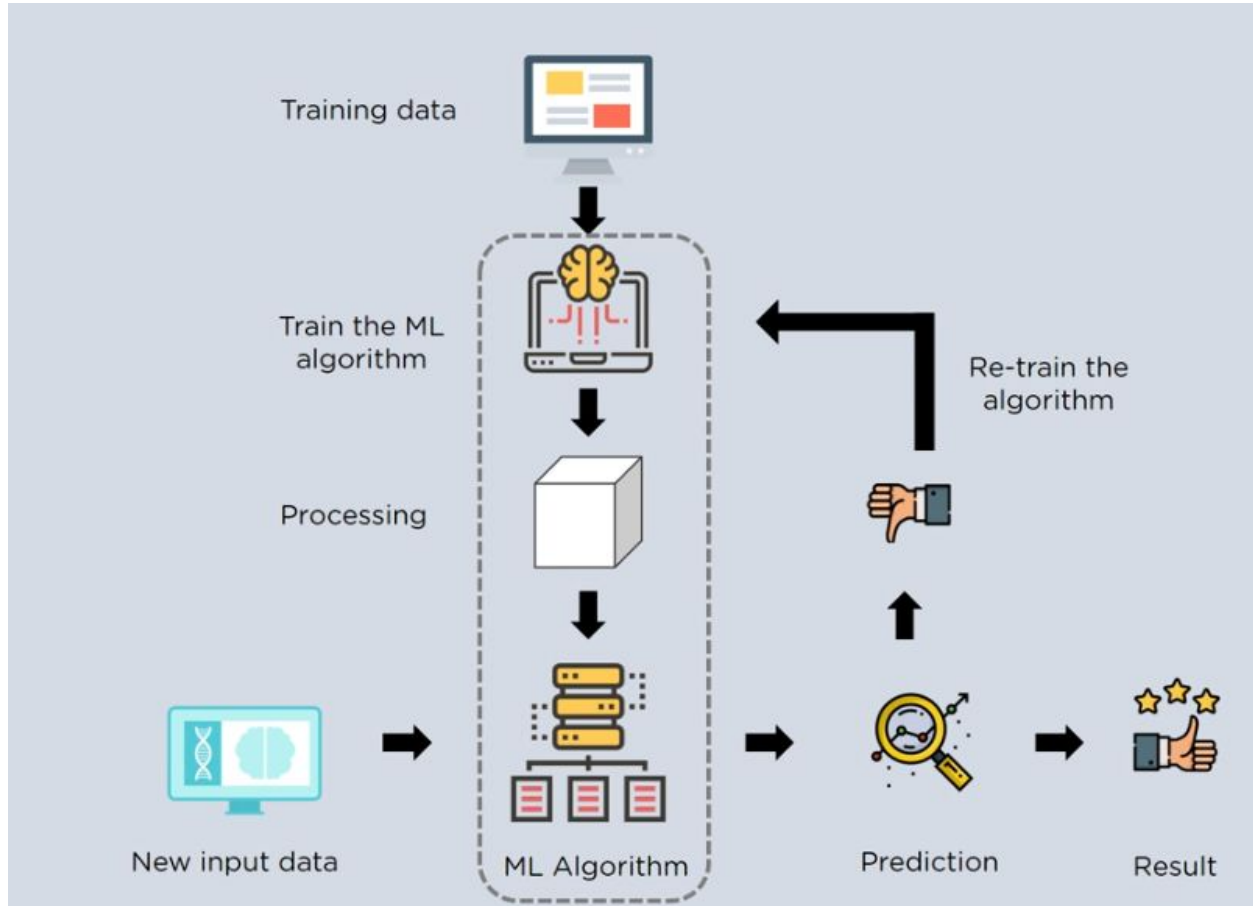
http://www.crm.umontreal.ca/2018/Modern18/horaire_e.html

Some Insight!

Algorithms:

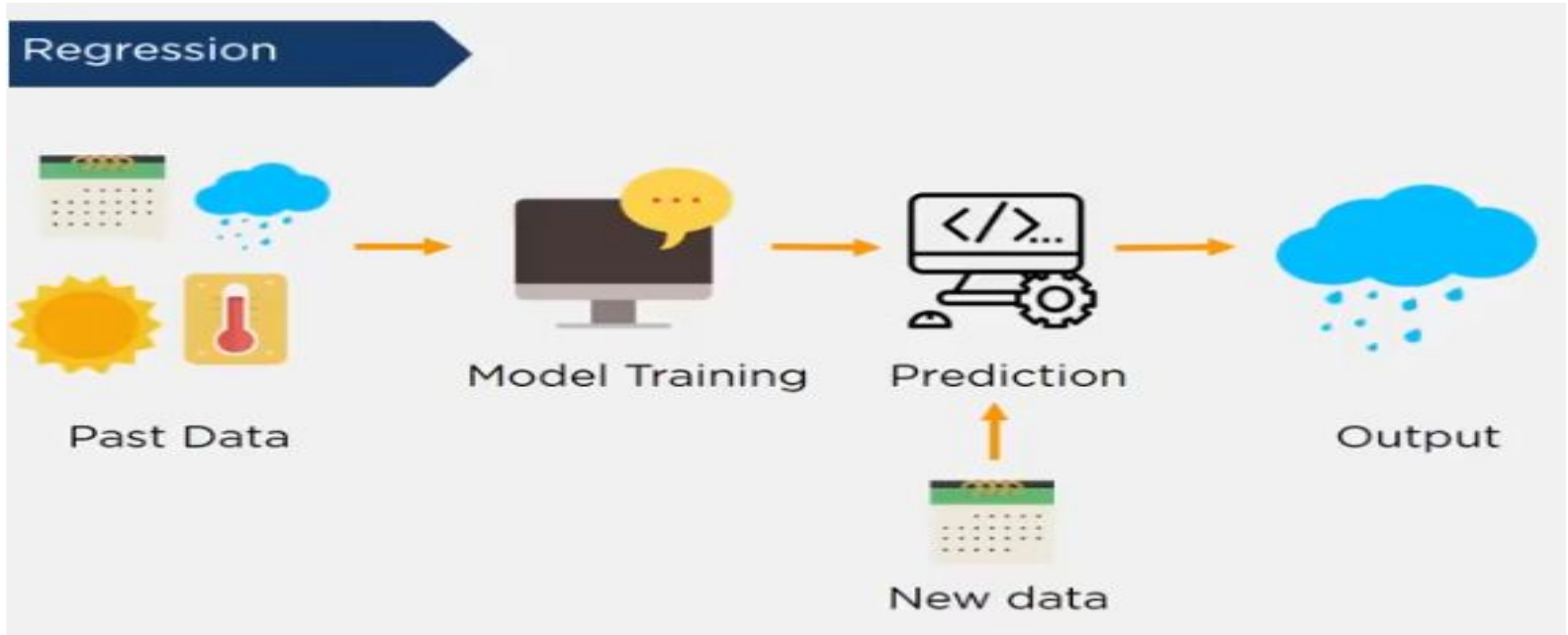


How does Machine Learning works?



Supervised Learning

Machine learning model learns from the past input data and makes future prediction as output



Price of a house



\$70,000

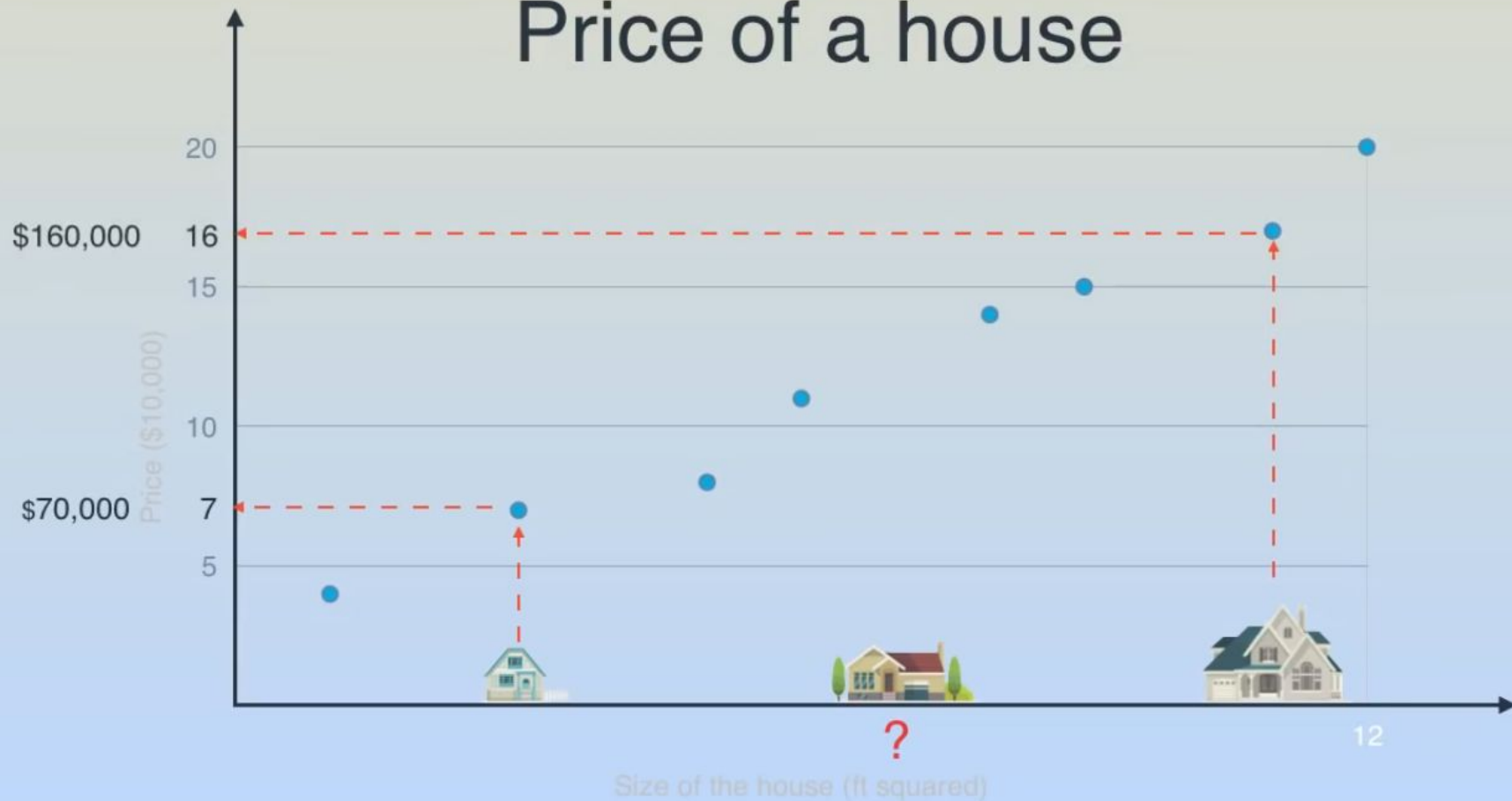


?

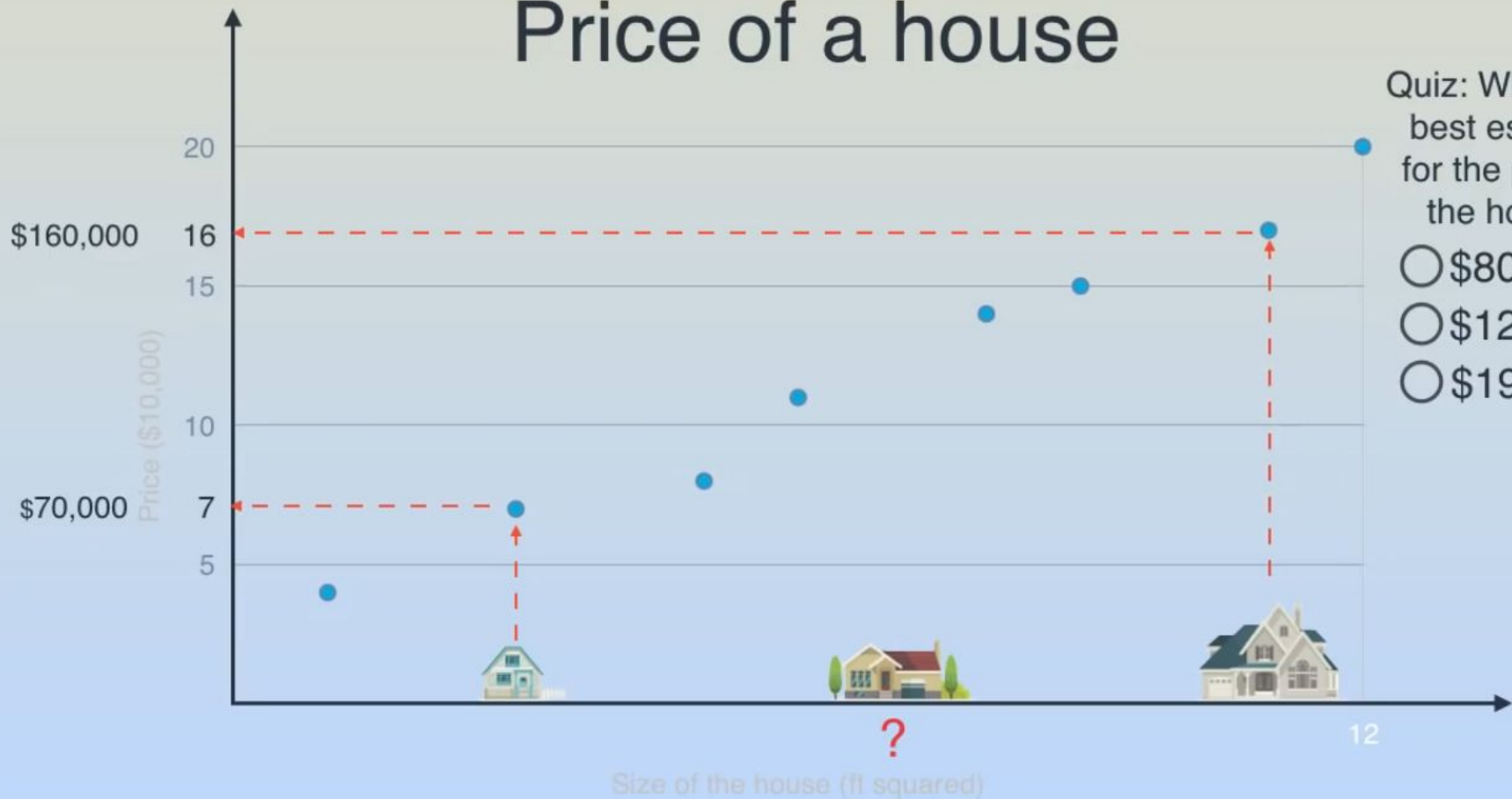


\$160,000

Price of a house



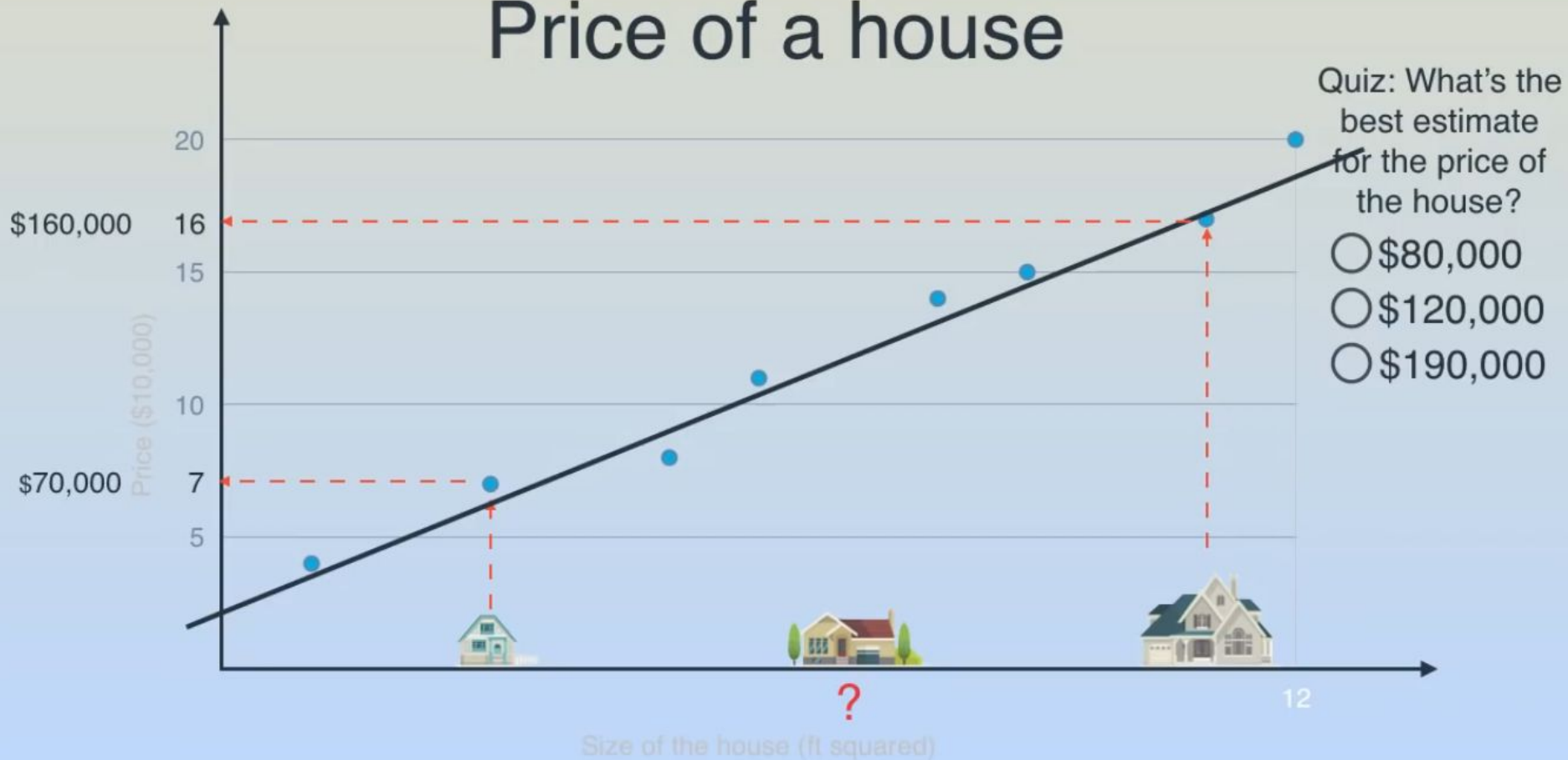
Price of a house



Quiz: What's the best estimate for the price of the house?

- ☐ \$80,000
- ☐ \$120,000
- ☐ \$190,000

Price of a house



Price of a house



Price of a House

This Method is Called Linear Regression

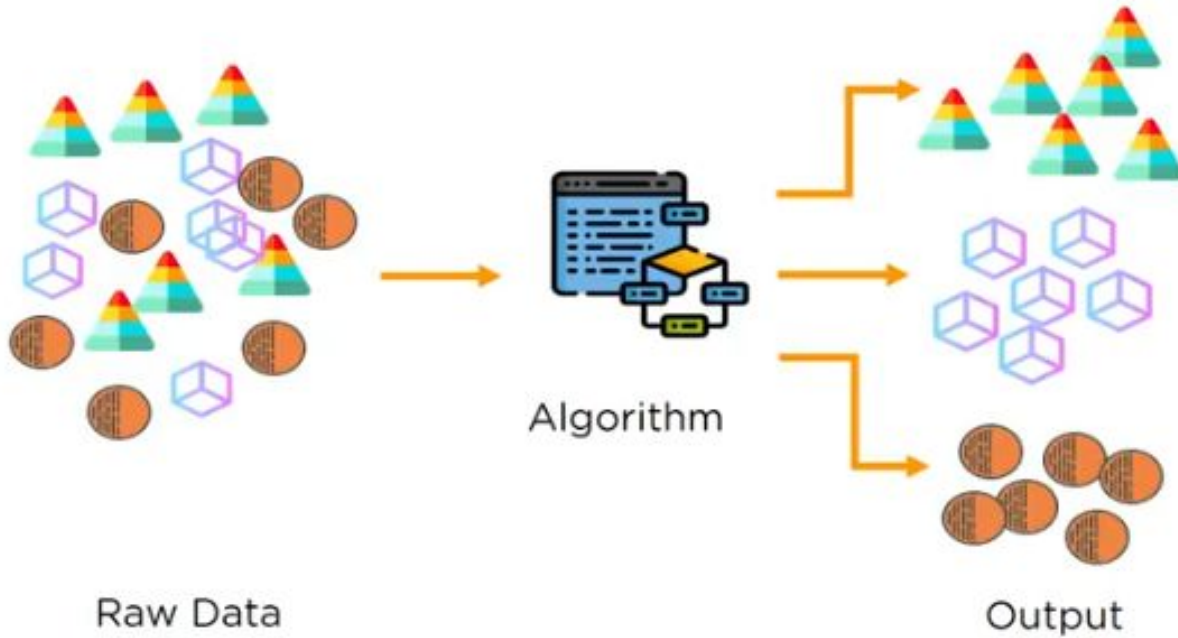


How You Find This Line ?

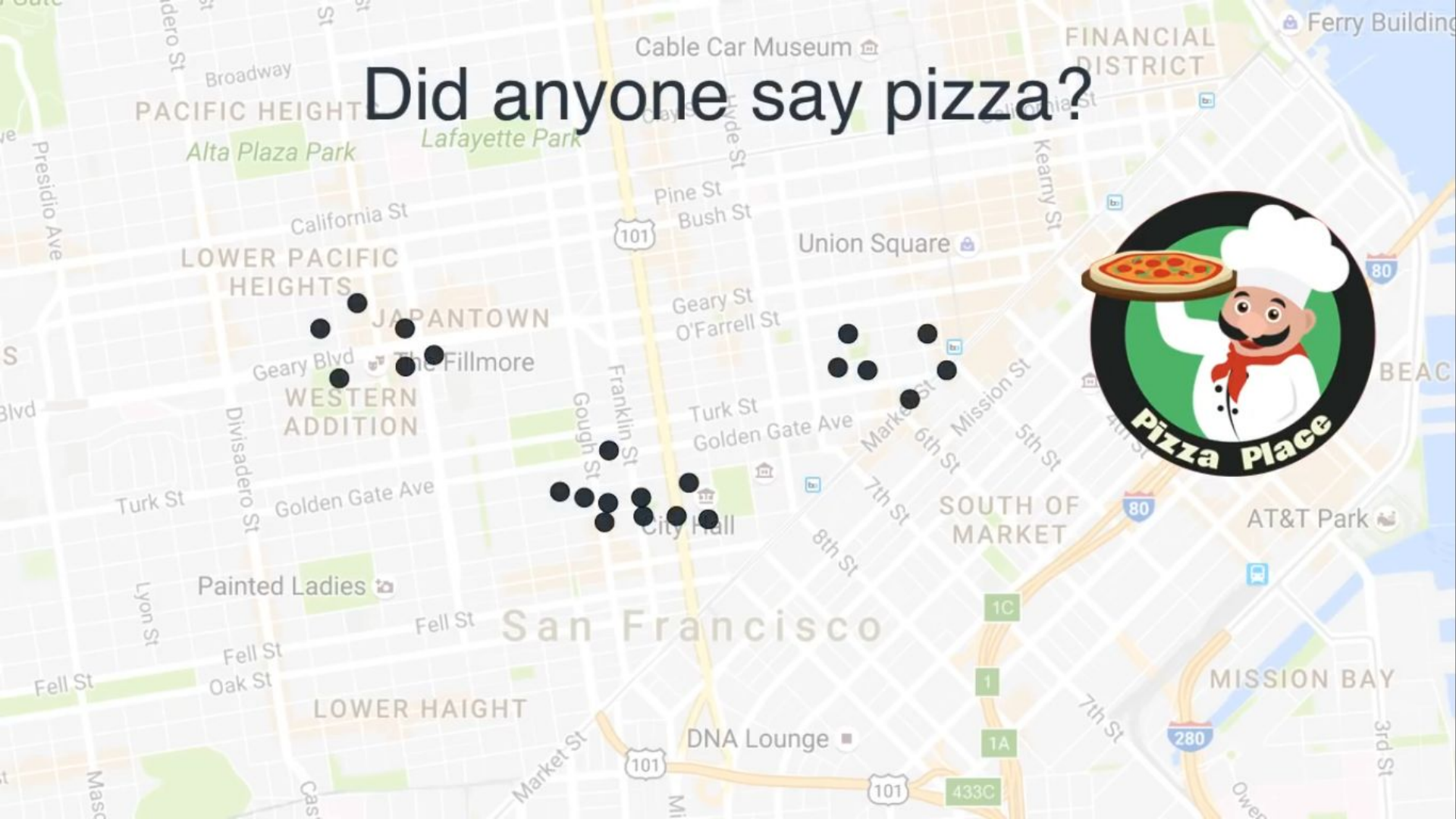
Answer : Gradient Descent Method

Un-Supervised Learning

Machine Learning Model uses unlabelled input data and allows the algorithm to act on that information without guidance



Did anyone say pizza?



Did anyone say pizza?



Did anyone say pizza?



Did anyone say pizza?

The map shows San Francisco with various neighborhoods labeled: PACIFIC HEIGHTS, LOWER PACIFIC HEIGHTS, JAPANTOWN, WESTERN ADDITION, SOUTH OF MARKET, SOUTH BEACH, MISSION BAY, and LOWER HAIGHT. Key streets like Broadway, Geary Blvd, Divisadero St, Golden Gate Ave, Market St, and Mission St are visible. Landmarks such as the Cable Car Museum, Union Square, City Hall, and AT&T Park are marked. The map is overlaid with colored dots representing different pizza styles: red dots in Japantown and Western Addition, blue dots in South of Market, orange dots in South of Market, and a yellow star in South of Market. A red star is located in Western Addition. The text 'Did anyone say pizza?' is overlaid on the map.

Did anyone say pizza?



Did anyone say pizza?



Did anyone say pizza?



Did anyone say pizza?



Did anyone say pizza?



Did anyone say pizza?



Did anyone say pizza?

Did anyone say pizza?

The map shows San Francisco with the following neighborhoods and landmarks labeled:

- PACIFIC HEIGHTS
- Alta Plaza Park
- Lafayette Park
- LOWER PACIFIC HEIGHTS
- JAPANTOWN
- WESTERN ADDITION
- Geary Blvd
- The Fillmore
- Franklin St
- Golden Gate Ave
- Turk St
- Divisadero St
- Painted Ladies
- Fell St
- LOWER HAIGHT
- City Hall
- Union Square
- South Market
- South Beach
- AT&T Park
- Mission Bay
- DNA Lounge
- Financial District
- Ferry Building

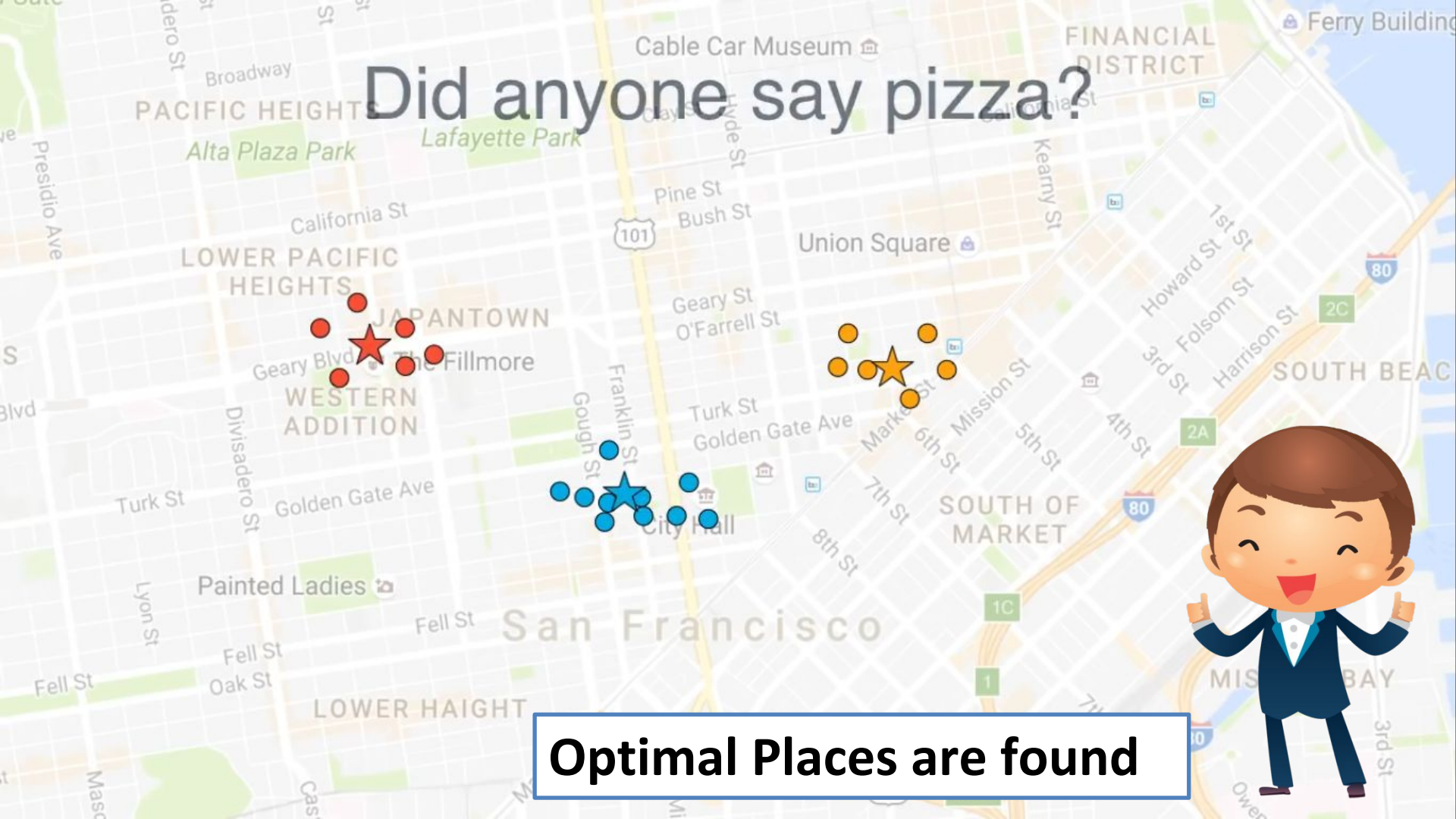
Colored dots and stars are placed on the map to indicate pizza locations:

- Red dots: Located in the Western Addition and Japantown areas.
- Blue dots: Located near City Hall.
- Orange dots: Located in the South Market area.
- Red star: Located in the Western Addition.
- Blue star: Located near City Hall.
- Orange star: Located in the South Market area.

Did anyone say pizza?

The map shows San Francisco with the following neighborhoods labeled: PACIFIC HEIGHTS, LOWER PACIFIC HEIGHTS, JAPANTOWN, WESTERN ADDITION, SOUTH OF MARKET, SOUTH BEACH, MISSION BAY, and LOWER HAIGHT. Key landmarks include Alta Plaza Park, Lafayette Park, Union Square, City Hall, and AT&T Park. Major streets like Broadway, Geary Blvd, Golden Gate Ave, and Market St are visible. The text 'Did anyone say pizza?' is prominently displayed at the top.

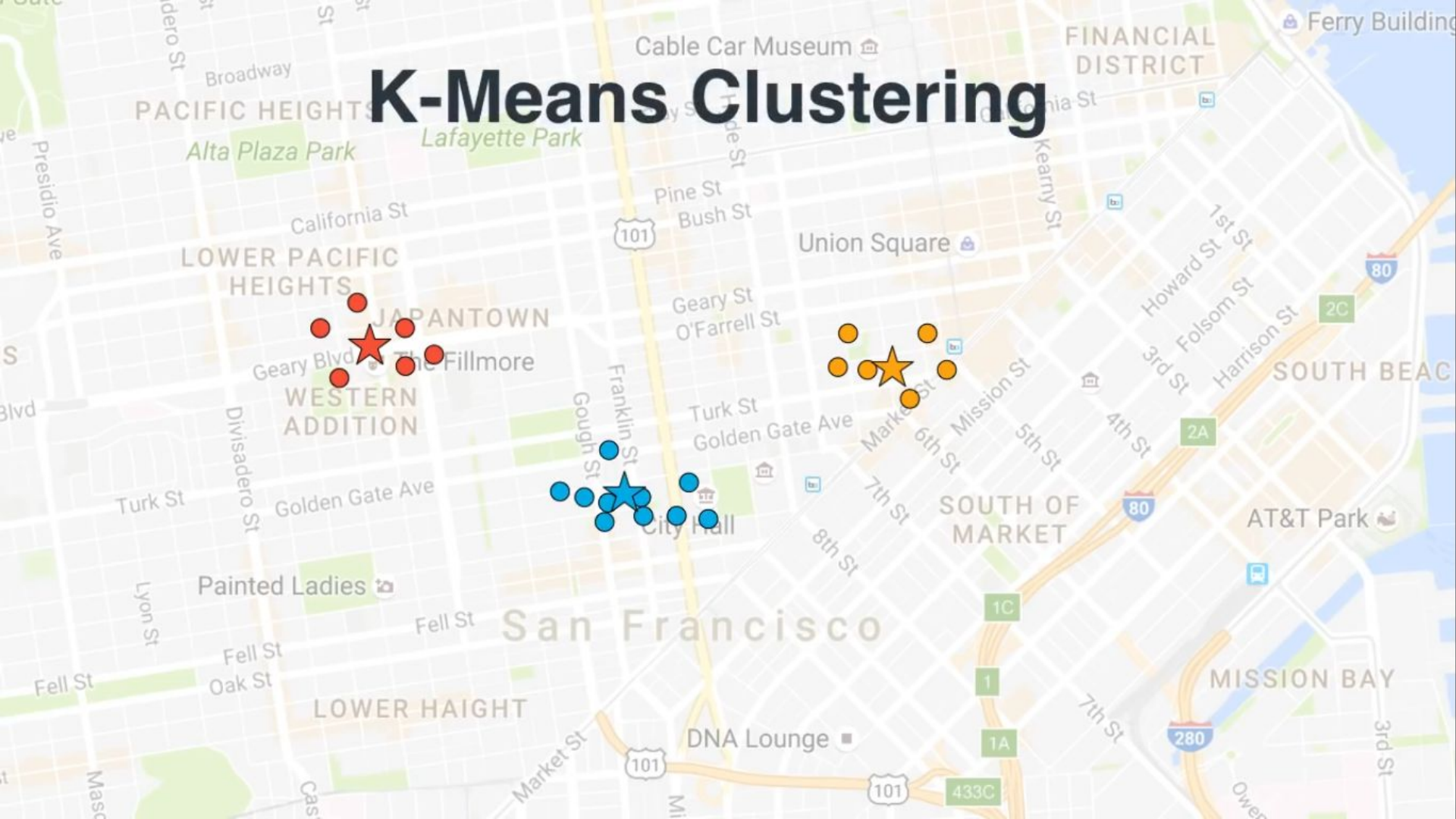
Did anyone say pizza?



Optimal Places are found

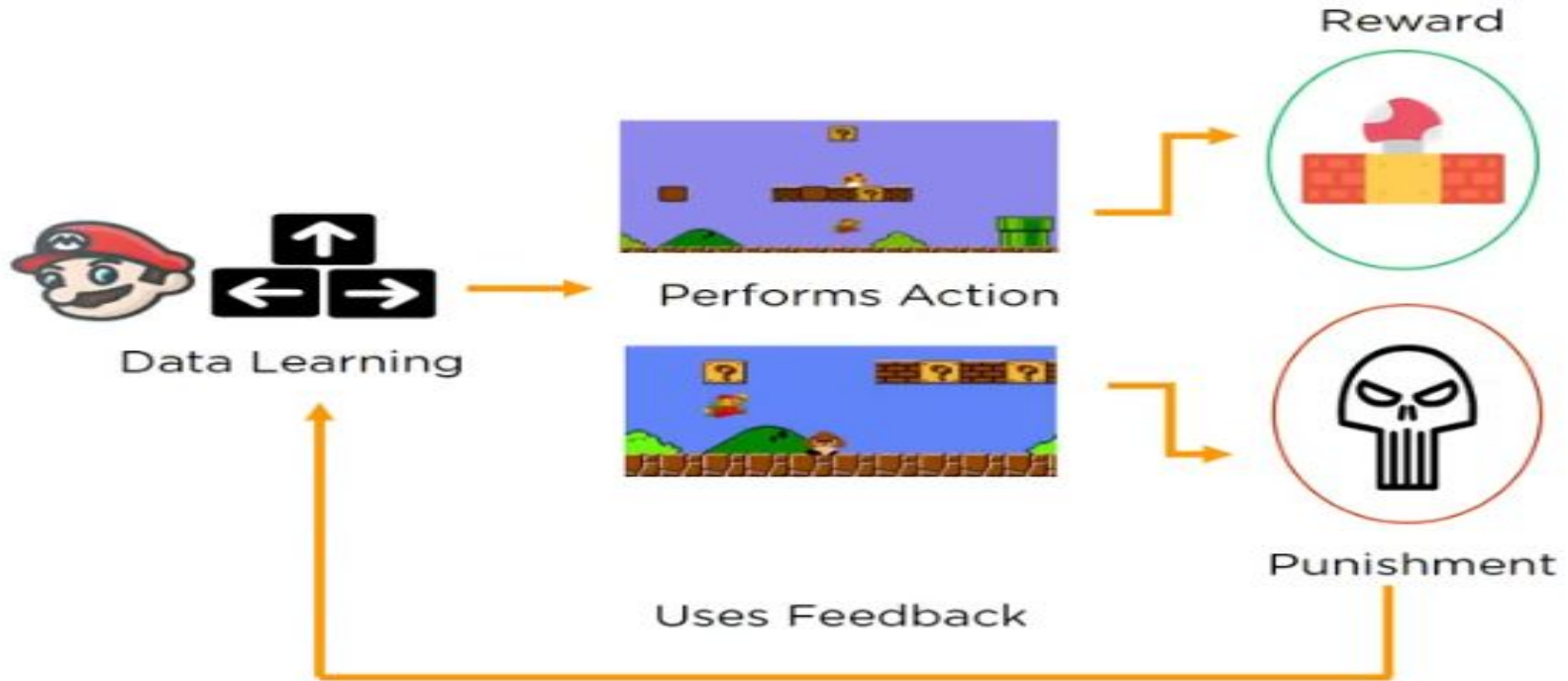


K-Means Clustering



Reinforcement Learning

Reinforcement Learning involves teaching the machine to think for itself based on its past action reward.



Special Thanks!



Mr Vishal Sathwane



Mr. Shivaram (Prof . Vashikaran !)

Good Starting point...

<https://codelabs.developers.google.com/codelabs/cloud-tensorflow-mnist/#0>

<https://www.coursera.org/learn/machine-learning>

<https://www.kaggle.com/learn/machine-learning>

<https://mlcourse.ai/prerequisites>

<https://www.fast.ai/>

<https://deeplearning.mit.edu/>

Coding!
time