

# Introduction to Machine Learning

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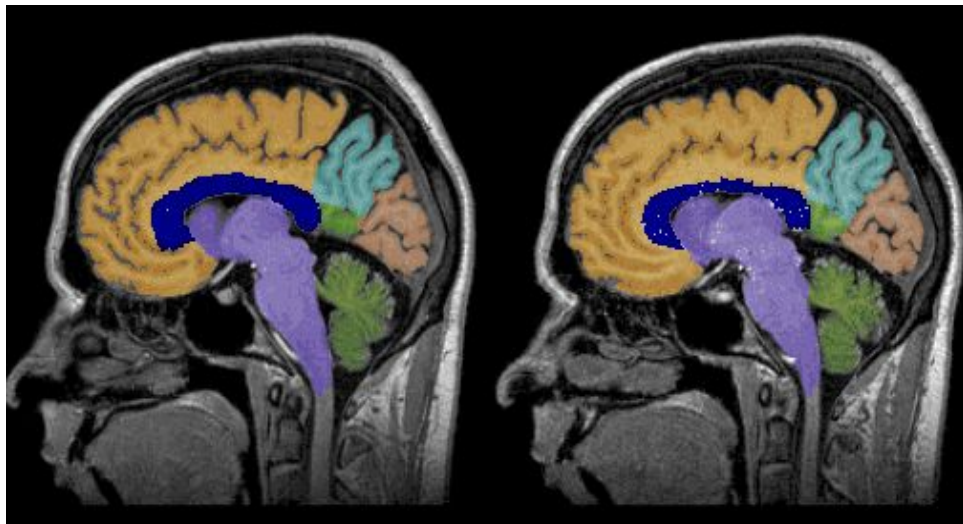


# Quick introduction

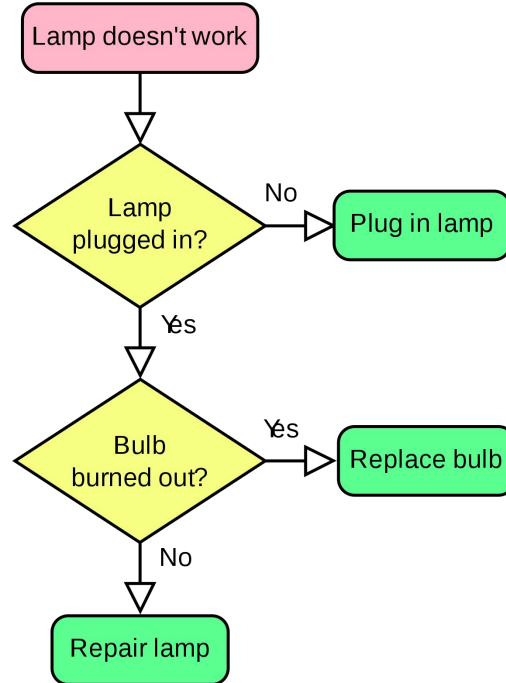


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or via fb/twitter (same  
username, photo from above)

Main fields of work: Computer  
Vision/Medical imaging.



# Machine learning - what is it exactly?



# Machine learning - what is it exactly?

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Is this a dog?



# Machine learning - what is it exactly?

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Is this a cat?

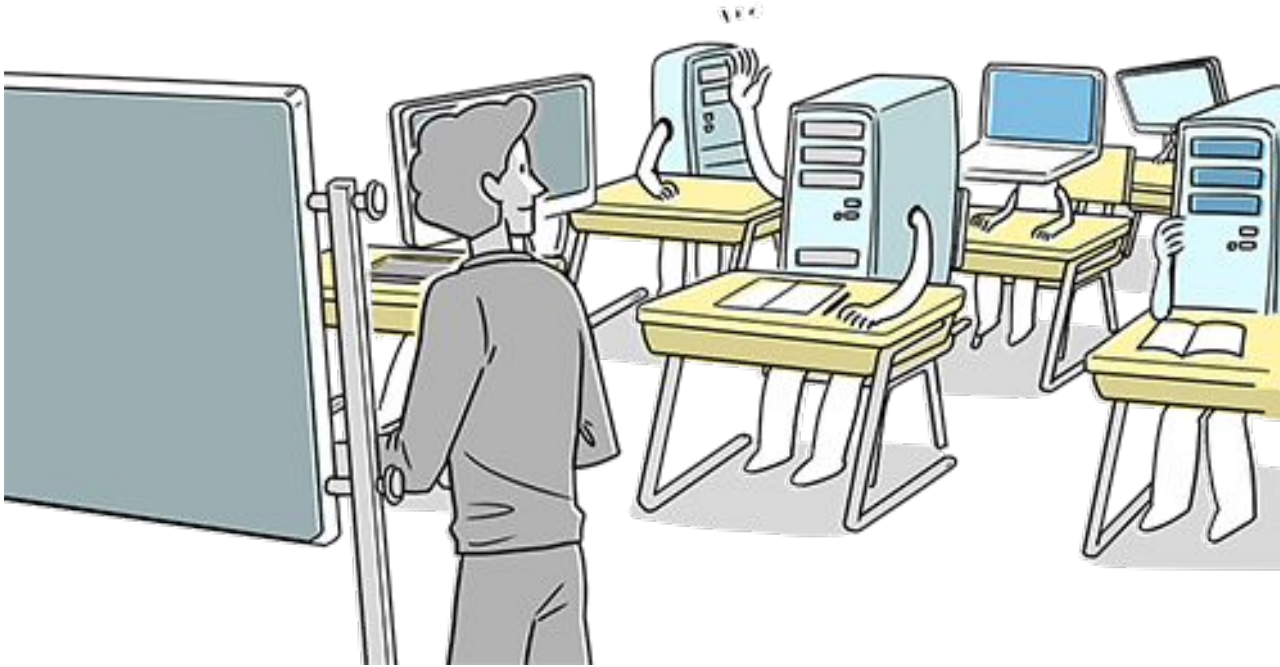


# Machine learning - what is it exactly?

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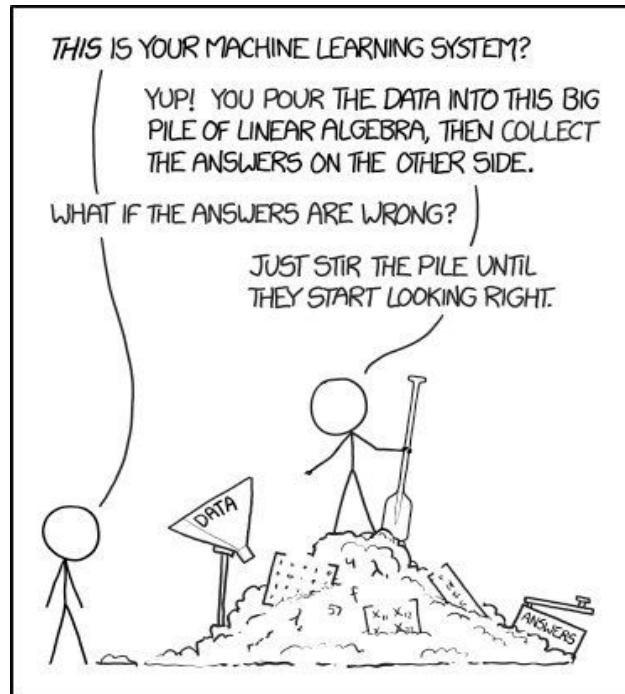
What about the computer?

# Machine learning - what is it exactly?



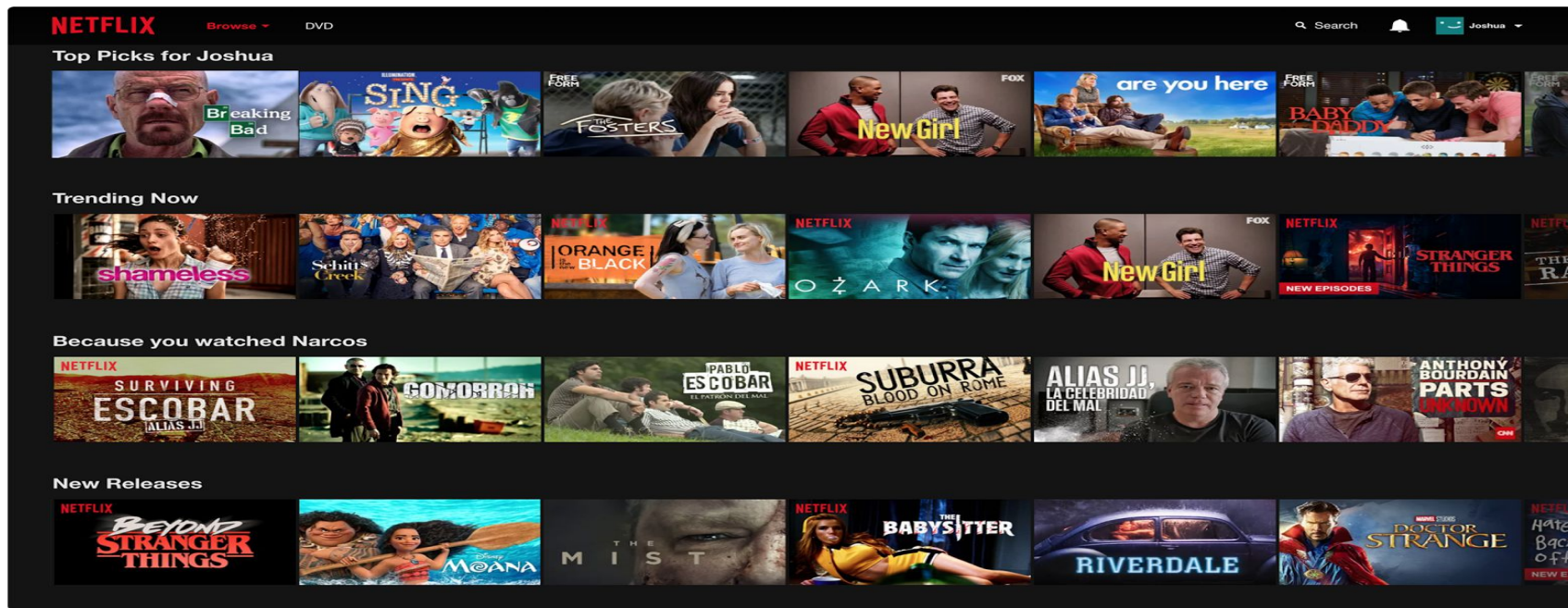
# Machine learning - what is it exactly?

Machine learning is the science of programming computers so they can learn from data how to perform some specified task - without being explicitly programmed.



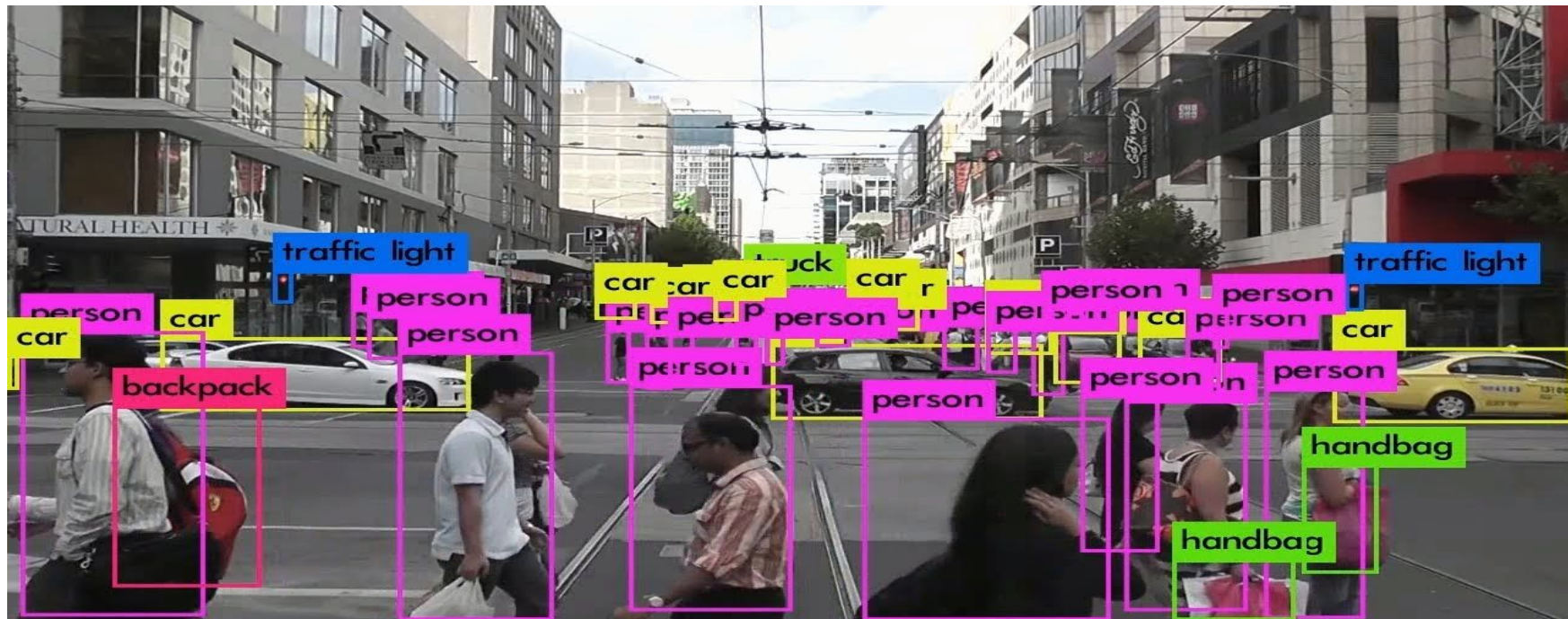


# ML - how it can be useful?



## Introduction to ML

# ML - how it can be useful?



Introduction to ML

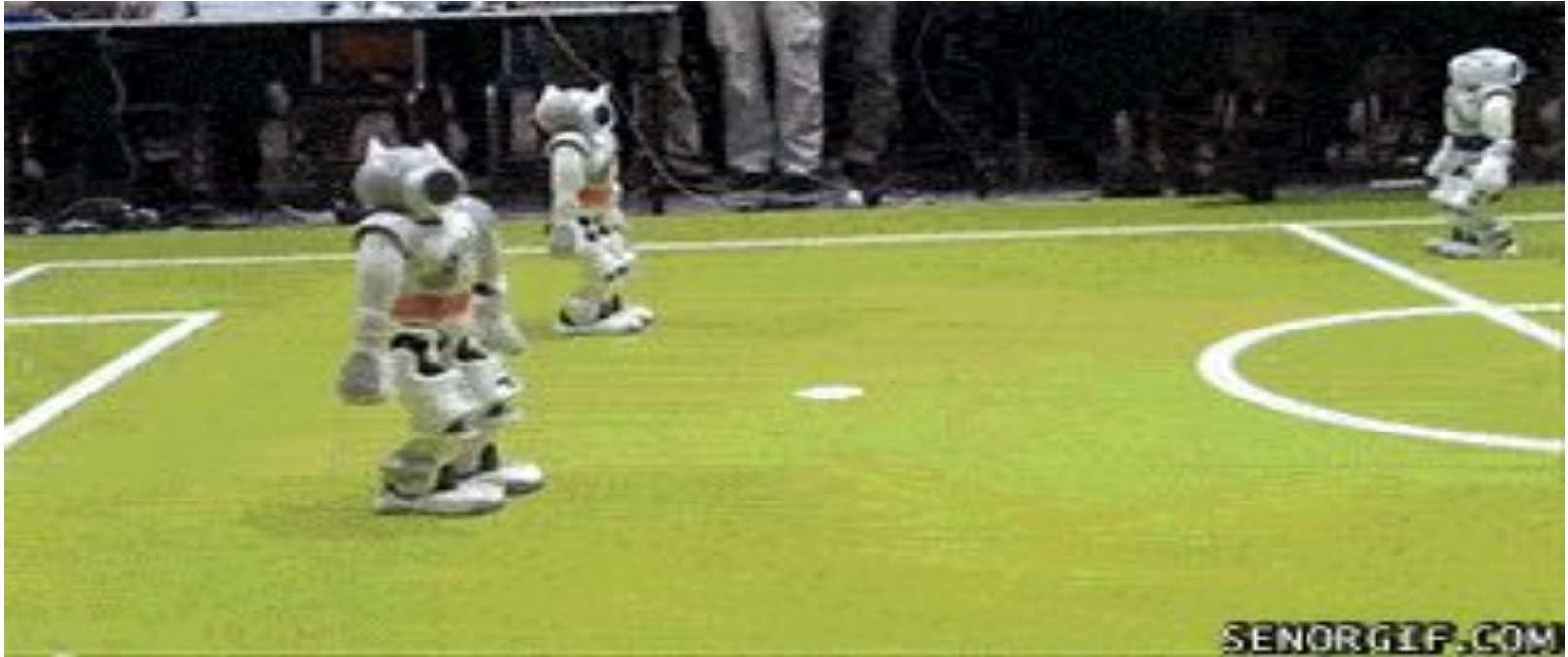
# ML - how it can be useful?



Introduction to ML



# ML - how it can be useful?



Introduction to ML

# Tasks ML is great for

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- Problems which existing solutions require a lot of manual tuning and lists of rules
- Problems with fluctuating environments, ML systems can update quite easily/automatically to new data.
- Complex problems, with no good solution at all with traditional approaches, e.g. image or audio recognition
- Reversing decision processes/finding associations/getting insights from data - so called data mining.

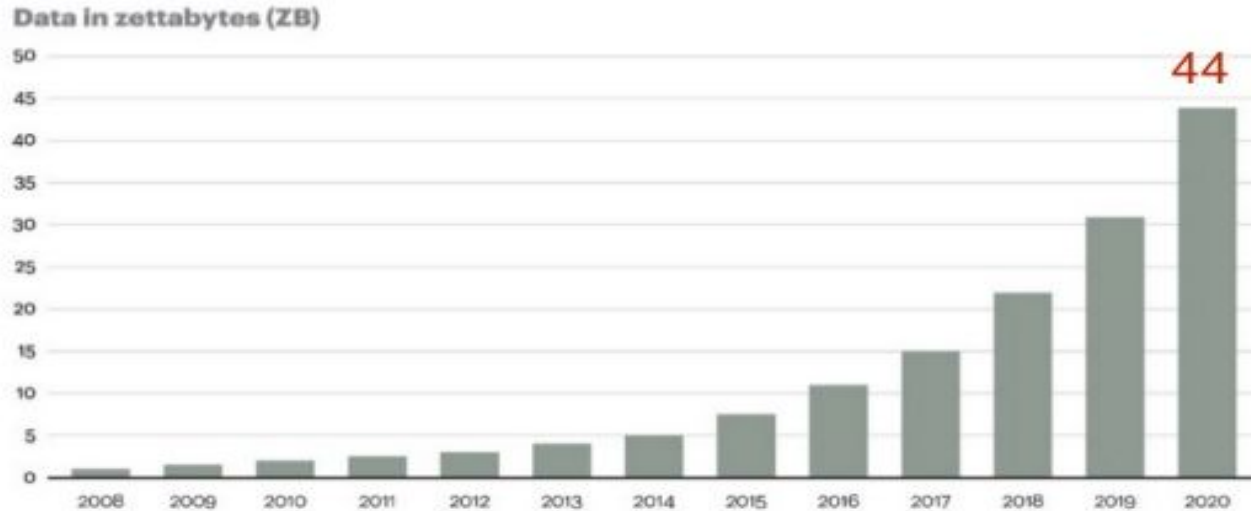
# Sudden popularity rise

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And *why* has ML become much more common and popular in recent years?

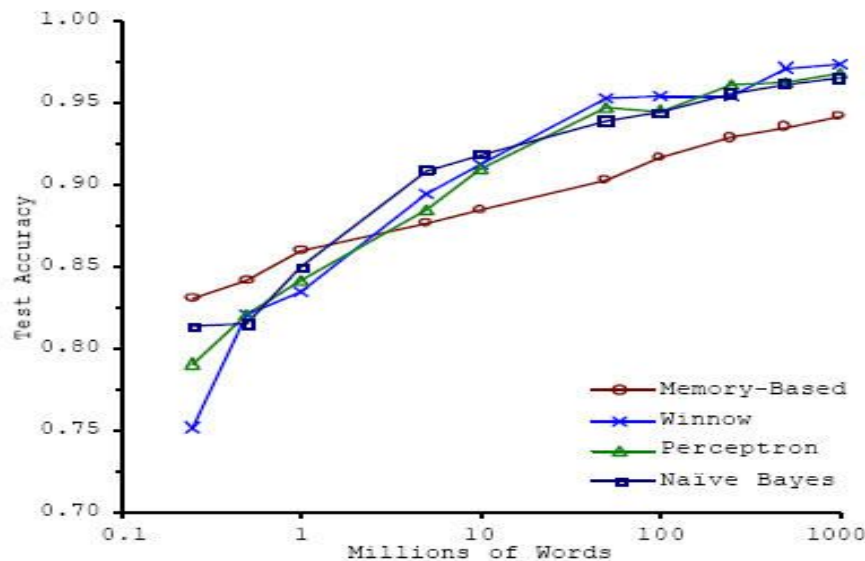
(besides the fact that we can do pretty cool things with it)

# Exponential data growth



44ZB = Google x 3000

# The unreasonable effectiveness of data



“The unreasonable effectiveness of data”, Banko & Brill, 2001

- Machine learning algorithms are often called “data hungry”.
- Their performance rise with the amount of training data, often outweighing selected algorithm and feature engineering.



## GOOGLE DATACENTER



1,000 CPU Servers  
2,000 CPUs • 16,000 cores

**600 kWatts**  
**\$5,000,000**

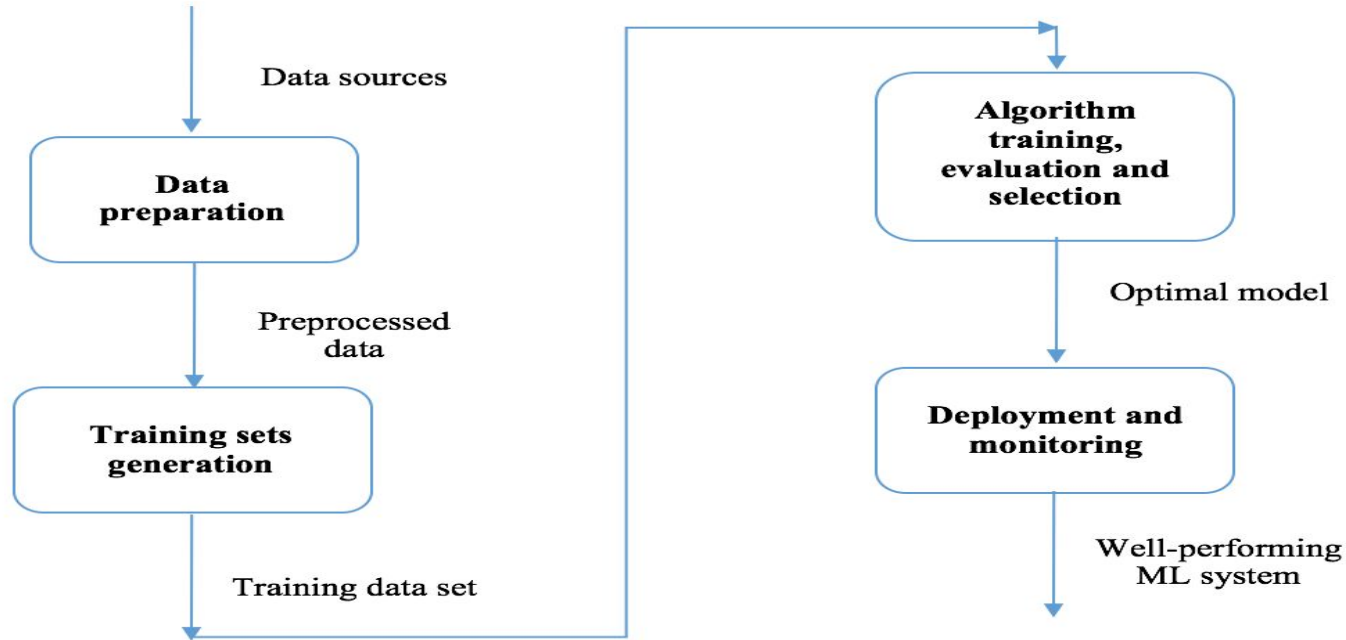
## STANFORD AI LAB



3 GPU-Accelerated Servers  
12 GPUs • 18,432 cores

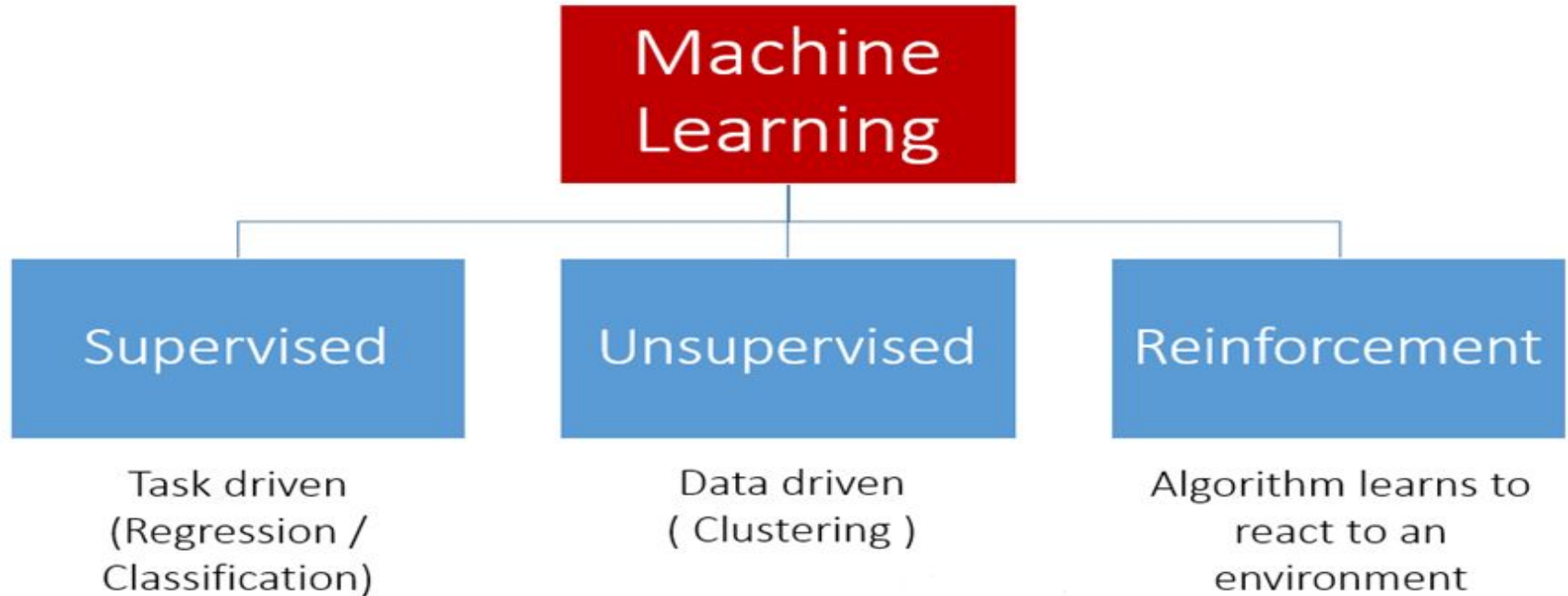
**4 kWatts**  
**\$33,000**

# Machine learning workflow schema



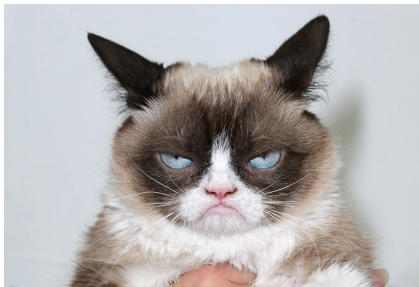
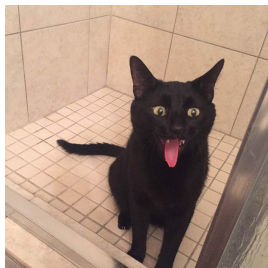
# Types of machine learning

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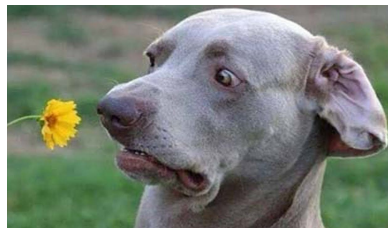


# Supervised learning

## Cats



## Dogs





# Unsupervised learning

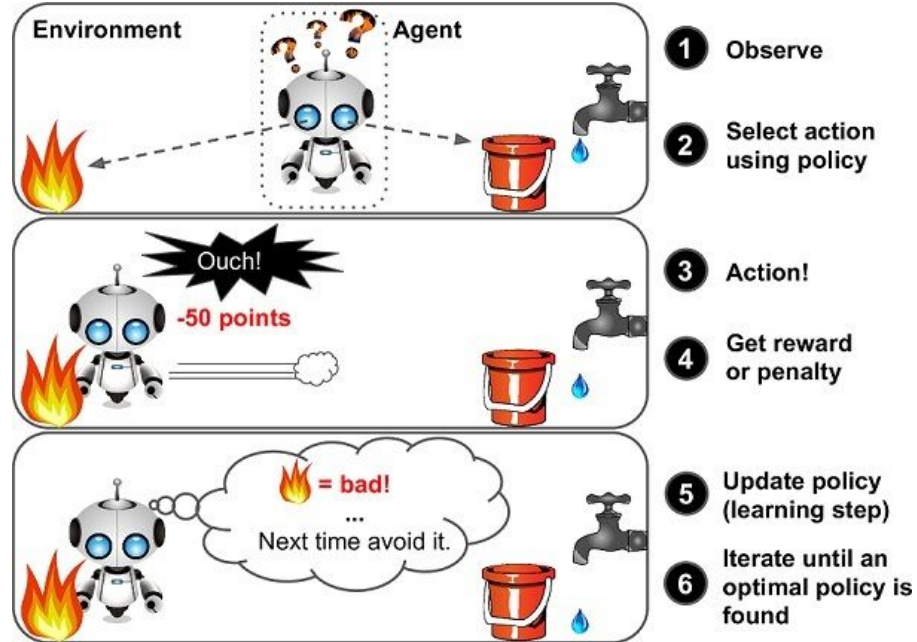


Introduction to ML

# Unsupervised learning



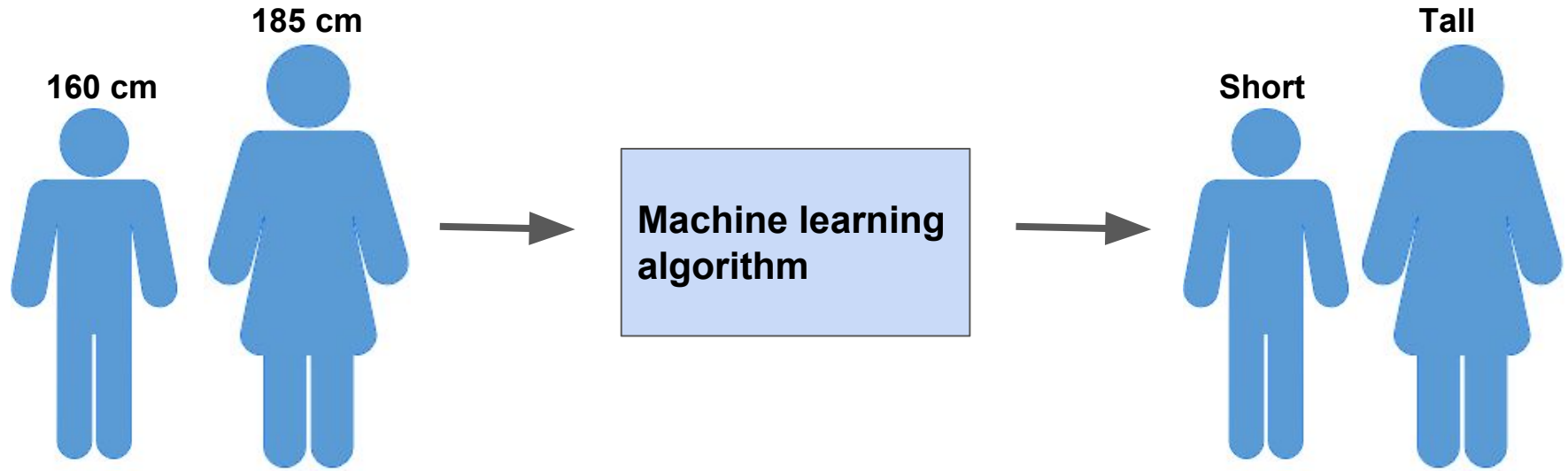
# Reinforcement learning





# Classification problem

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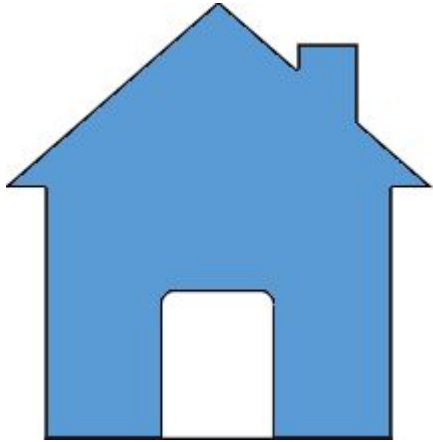


# Regression problem

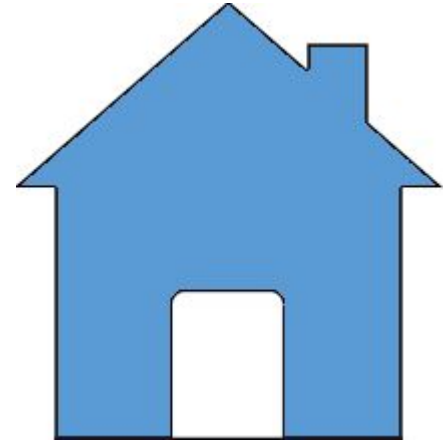
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Square meters: 50  
Condition: good  
Location: NY

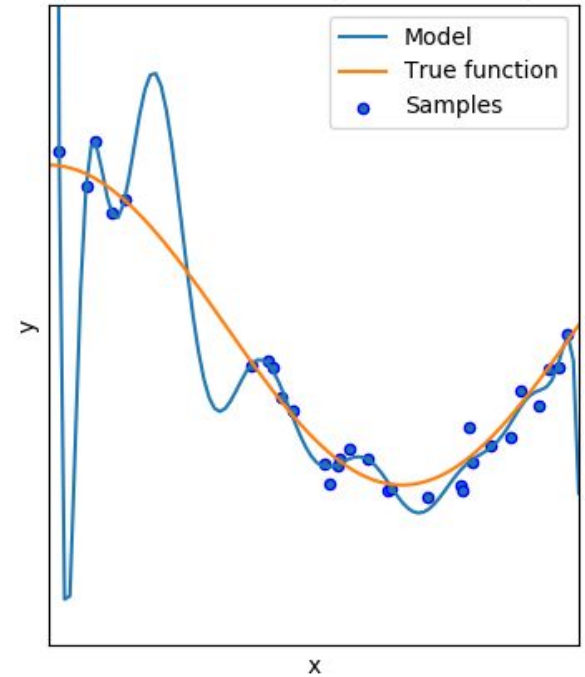
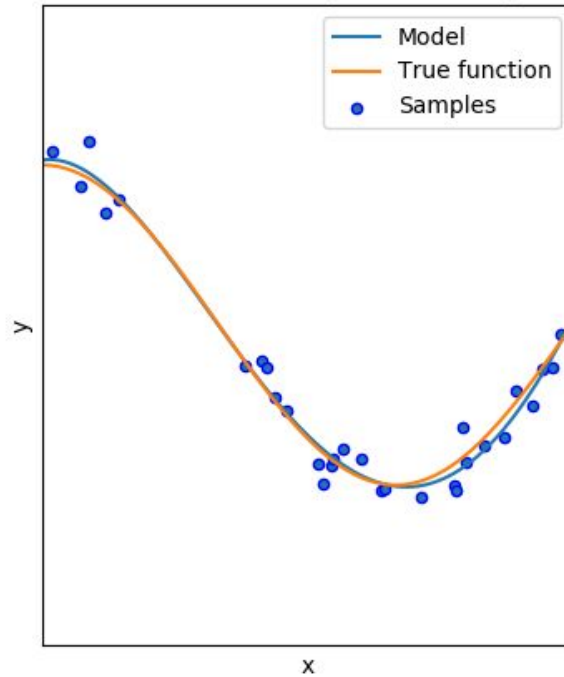
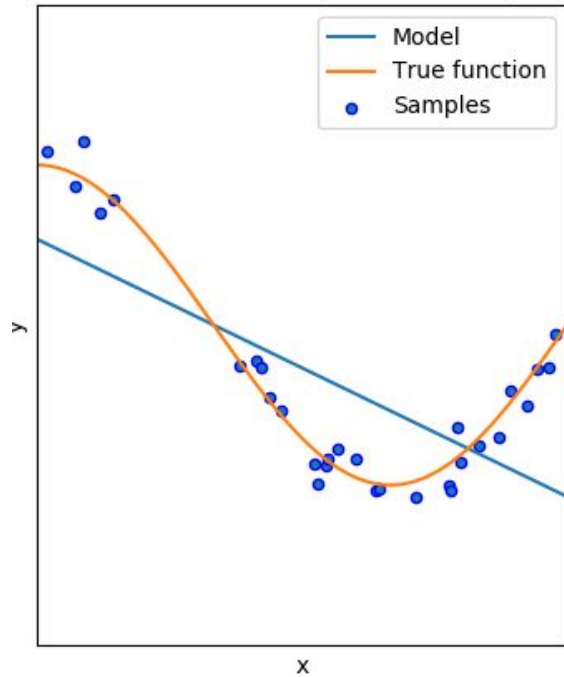
Price: 100 000\$  
(Continuous value)



**Machine learning  
algorithm**

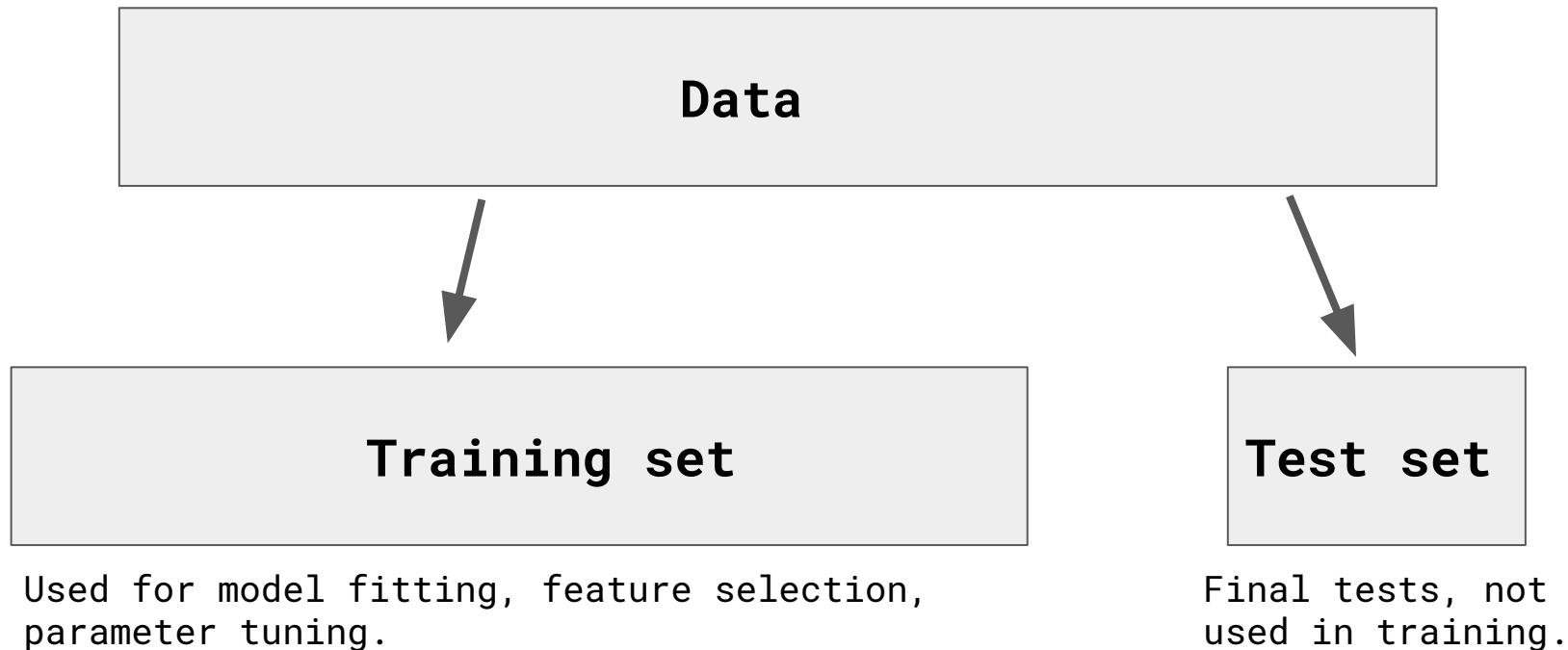


# Underfitting and overfitting



# Train/test splitting

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# Train/test splitting

Performance on training data vs test data



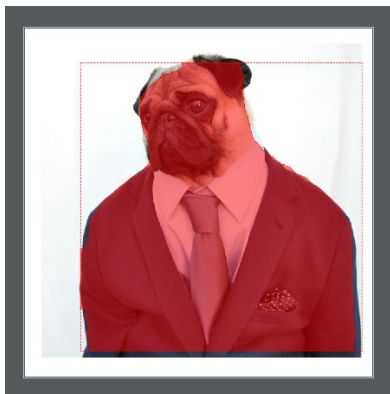
# But what if my machine is not enough...

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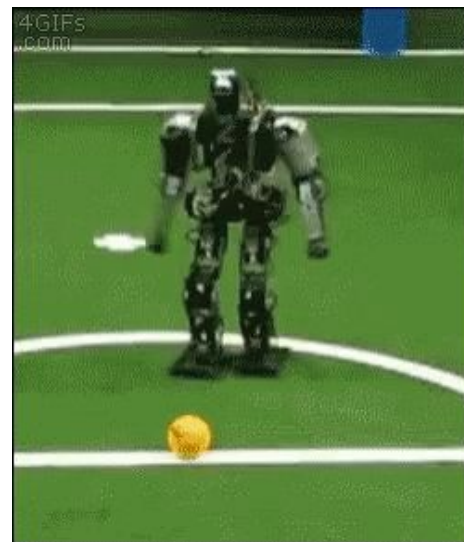
- Despite what it seems like, even “casual user” machines are often good enough at the beginning of the ML adventure.
- There are some computing resources available for free, e.g. Google Colab.
- Cloud resources (Paperspace, Azure, and so on) are quite cheap (and often have some free starter/student packages).
- And at last - get used to it (as after some threshold this problem affects pretty much everyone).



# Just remember - ML isn't a silver bullet...



“gibbon”  
99.3 % confidence



# Tools and stuff



And many, many others...



A 3D animated image of Shrek from the movie "Shrek". He is a green ogre with a large, friendly smile, showing his teeth. He has his arms outstretched to the sides, palms facing forward. He is wearing a white tunic with a brown, textured vest over it. The background shows a forest with trees having autumn-colored leaves (yellows and oranges) under a cloudy sky.

**ANY QUESTIONS?**