

WEEK 1

□ MACHINE LEARNING

→ As per Arthur Samuel (1959)

"Machine Learning is the field of study that gives computers the ability to learn without being explicitly programmed"

→ As per Tom Mitchell (1998)

"A computer program is said to learn from experience E with respect to some task T and some performance measure P , if its performance of T as measured by P , improves with experience E ."

★ EXAMPLE

For a game of checkers,

- The experience E would be the experience of having the program play tens of thousands of games itself
- The task T would be the task of playing checkers
- The performance measure P would be the probability that it wins the next game of checkers against some new opponent

→ Machine Learning can be broadly classified into 2 categories

SUPERVISED
LEARNING

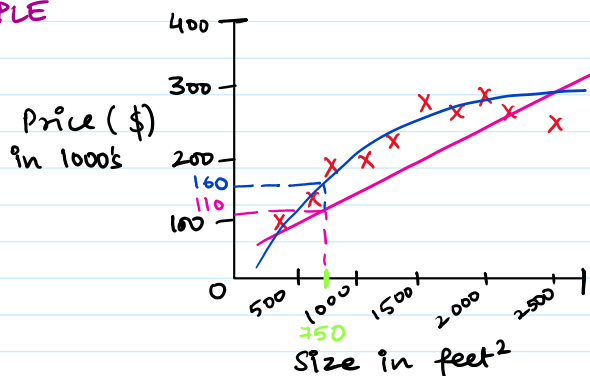
UNSUPERVISED
LEARNING

→ There are others as well but not as popular : • Reinforcement Learning
• Recommender Systems

□ SUPERVISED LEARNING

→ Supervised learning is probably the most common type of machine learning

★ EXAMPLE



House Price Prediction

- ↳ Data has been collected from city of Portland, Oregon
- ↳ Plot of the data looks like this

① Let's say you have a friend with a house of 750 sq feet and they are hoping to sell the house. They want to know how much they can get for the house.

→ How can a learning algorithm help here?

get ~ for the house.

→ How can a learning algorithm help here?

↳ One thing the learning algorithm might do is put a **straight line**

↳ Shows **110k** for the house

↳ What if instead of fitting a straight line, we plot a **polynomial function**

↳ Shows **160k** for the house

→ Each of the above fits would be a fine example of a learning algorithm

→ This is an example of "Supervised Learning Algorithm".

→ The term "Supervised" refers to the fact that we gave the algorithm a data set in which the 'right answers' were given.

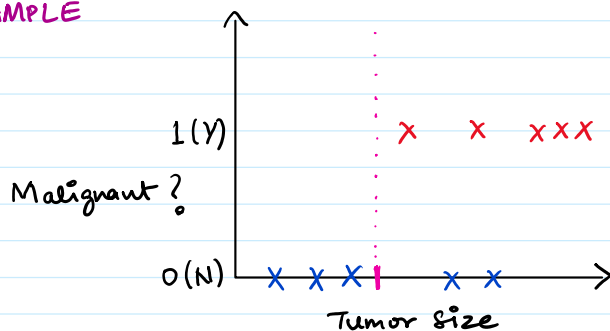
↳ We gave it a data set of houses in which for every example in this data set, we told the algorithm what is the right price

↳ And the task of the algorithm was just to produce more of such right answers

• To define a bit of terminology, this is called a **Regression** problem

predict continuous valued output
(example: price)

★ EXAMPLE



Breast Cancer (malignant, benign)

From medical records, predict whether breast cancer is malignant or benign

ⓐ For **this tumor size**, what is the chance that the tumor is malignant v/s benign?

• To define a bit of terminology, this is an example of a **Classification** problem

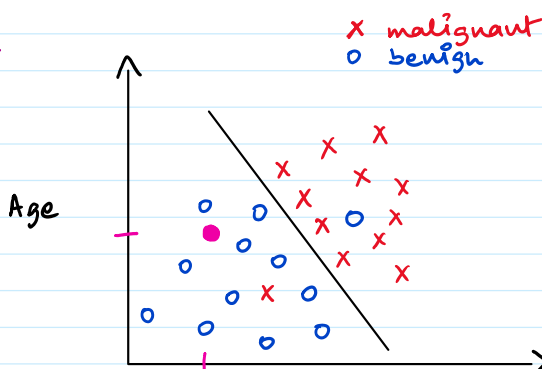
• In this example, we have only used **1 feature or attribute** to predict whether a tumor is malignant or benign

↳ We can have more such features

discrete valued output (0 or 1)
↑ classes/labels

<we can have more than 2 classes>

★ EXAMPLE



Breast Cancer (malignant, benign)

This time we have more features

ⓐ For **this particular point**, we have to tell what is the chance for the tumor to be malignant or benign.

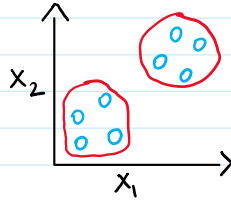
Tumor size

- In a dataset like this, what a learning algorithm might do is fit a straight line to the data to try to separate out the malignant tumors from the benign ones
↳ After doing so, we can say that the point in question has a higher chance of being benign

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□ UNSUPERVISED LEARNING

- In Unsupervised Machine Learning, we are given data which has no classes or labels



Given this dataset, the unsupervised learning algorithm might decide that the data lives in 2 different clusters

- The problem always is - there is some data, can you find some structure in the dataset?
→ Applications - Google News, Gene Array Categorization, Market Segmentation etc.
→ We derive structure from the data but we don't necessarily know the effects of the variables
→ With unsupervised learning, there is no feedback based on the prediction results