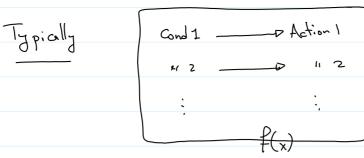
Lecture 1: Introductory remarks February 11, 2020 12:18 PM
Goals of this course
Machine learning
□ Descriptive ML
[Predictive ML
[] Generative ML
☐ Learning ☐ Notation
Goals
· Learn to use basic ML techniques
- IN I I I
■ Improve ML — New algorithms
New models
New metrics
New paradigms → Quantum ML



f(x): X is checked for { cond is and the proper action is implemented.

But how do we learn?

We see/observe

A) and try to fit with some known outcome

D Prediction

Instance 1
$$\triangle D$$
 Output 1 Output i \in Set of outputs

2 \longleftarrow Output 2 \longrightarrow XIt is smaller than

i instance \mathcal{D}_{3}^{2} .

We infer the relation.

Example

Apple — 0

Orrange — 0

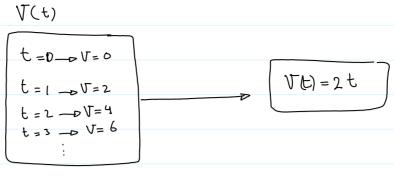
Snake — 1

Banana — 0

Elephant — 1

Input

This is known as "classification".



Regression

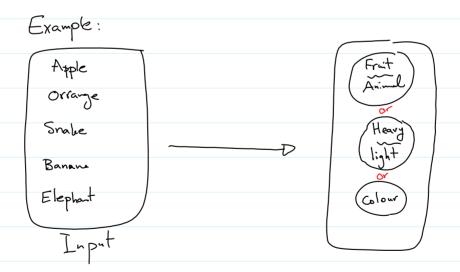
Both classification & regression come with

some target values and the goal is to estimate / predict

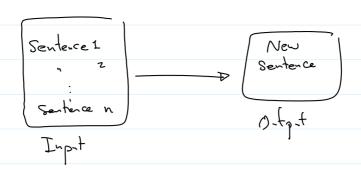
the outcome for some unseen instance.

B) try to recognize some pattern in the data.

Sometimes only the instances are given and the task is to draw some conclusion / describe the data.

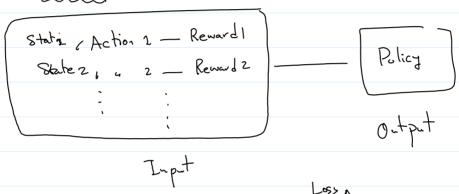


C try to mimic the data we get: learning to speak
draw
sing
play music



C try to learn a procedure: how to play a game, ...

* Learning ?!!



Example Minimite

Reward ~ - A loss

Charge in the



Policy go against the Jenivative

Notation -

$$X = \begin{bmatrix} x^{(1)} & \Rightarrow & Sample \\ x^{(2)} & \Rightarrow & Sample 2 \\ \vdots & \vdots & \vdots \\ x^{(n_s)} & \Rightarrow & Sample n_s \end{bmatrix}$$

$$X = \begin{bmatrix} x^{(1)} & \Rightarrow & Y^{(2)} \\ y^{(2)} & \vdots \\ y^{(n_s)} & \vdots \\ x^{(n_s)} & \Rightarrow & Target \end{bmatrix}$$

Measurement 1:
$$(P^{(1)}, V^{(1)}, n^{(1)}) \longrightarrow T^{(1)}$$

$$_{1}$$
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What's X & Y?

What's np & ns?