FIRST STEP TO PRACTICAL MACHINE LEARNING

KNOWLEDGE SHARING FOR CPE/SKE STUDENTS

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BEFORE WE START...

Make sure these are installed on your computer.

This page is a guide for installing on Windows

- Python 3.6: Download and install at https://www.python.org
- NumPy, Scipy, Matplotlib, Scikit-learn, MLxtend: Run pip install numpy scipy matplotlib sklearn mlxtend

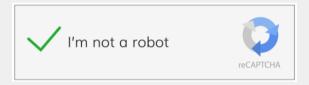
OUTLINE

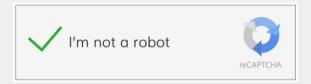
- 1 Introduction to Machine Learning
 - What is Machine Learning? Traditional programming approach Machine learning approach
- 2 Machine Learning Problems
 - Supervised learning
 - Unsupervised learning
 - Reinforcement learning
- 3 Model
- 4 Machine Learning Process
- 5 Problems for Machine Learning
 - Handwriting recognition

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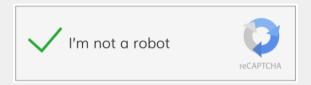


INTRODUCTION TO MACHINE LEARNING

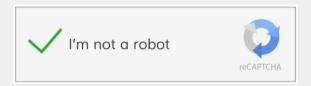




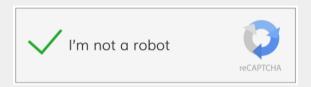
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 - ► Recaptcha helps stop millions of spam a day.

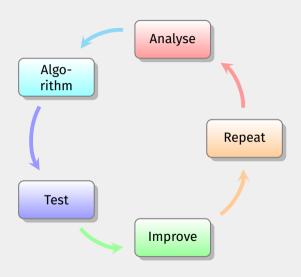


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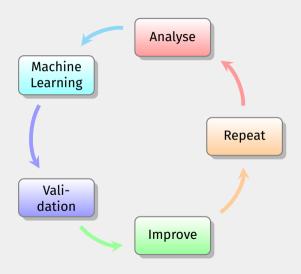


- This is Recaptcha.
 - ► Recaptcha helps stop millions of spam a day.
 - ▶ In some old days, we have to type Captcha texts to distinguish ourself from bots.
 - ► How is it possible that with a single click, an automated system can distinguish bots from humans?

TRADITIONAL PROGRAMMING APPROACH



MACHINE LEARNING APPROACH



IN OTHER WORDS...

Machine Learning

Machine Learning

= Data + Data analysis algorithm

Machine Learning

Data + Data analysis algorithmAdapt to change



MACHINE LEARNING PROBLEMS

Types of Machine Learning Problems

Types of Machine Learning problems

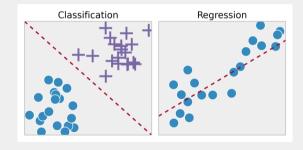
1. Supervised learning

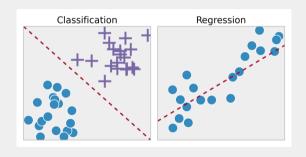
Types of Machine Learning Problems

- 1. Supervised learning
- 2. Unsupervised learning

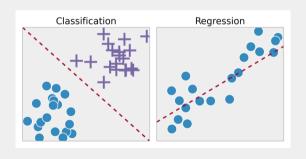
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- 1. Supervised learning
- 2. Unsupervised learning
- 3. Reinforcement learning

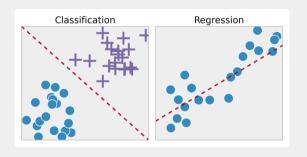




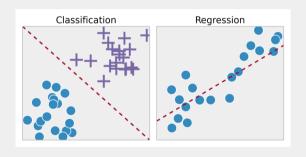
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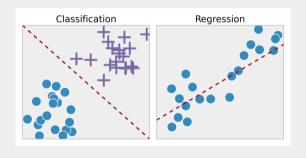
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- Two main supervised learning problems



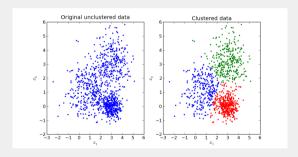
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- Two main supervised learning problems
 - ► Classification: On the discrete data



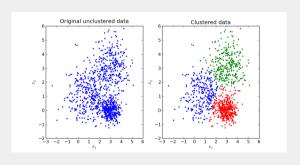
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 - ► Classification: On the discrete data
 - ► Regression: On the continuous data



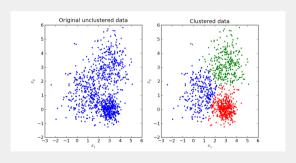
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 - ► Regression: On the continuous data
- Example problems: Spam E-mail detection, Facial recognition



Unsupervised Learning

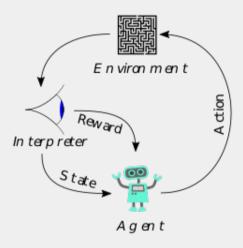


■ Discover hidden structure in non-labelled data.



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- Example: Clustering, Generative models

REINFORCEMENT LEARNING



MODEL

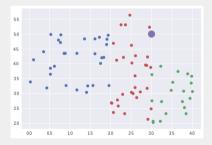
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MODEL

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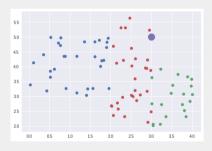
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Data

MODEL

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Determine which group should the purple dot be in (red/green/blue) by checking the colour of its nearest dot.

Data Method

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k-NN algorithm

To classify label of a data point, get *k* nearest data points to the data point, and select the major label among those data points.

Coding time!

MACHINE LEARNING PROCESS

MACHINE LEARNING PROCESS

- Train
- Test

(There'll be more of this, trust me.)

CHOOSING THE PARAMETER FOR k-NN ALGORITHM

What is the bad way to choose *k*?

- What if we choose k = # of all points?
 - ► What will happen if our dataset's got 3 labels of A, B, C with 10, 20, and 30 data points of each?
 - Answer: Our model will always answer the labels with the highest data point count.
- What if we choose k = 1?
 - ► Let's try!

Coding time!

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 - ► What will happen if we train on the testing set?
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 - Cheating! Like letting the model *remembers* the answer instead of **generalising** the data pattern.
 - ► In other words, don't test and train model on the same set of data.

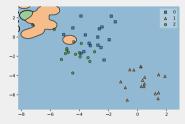
Choosing the best k

- **Train** with the training set, to let our model know how will the data looks like.
- **Test** with the testing set, to see on how our model performs.

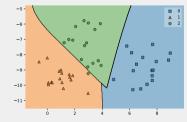
Warning! This is a simplified Machine Learning model training process, there are more to concerns!

OVERFITTING AND UNDERFITTING

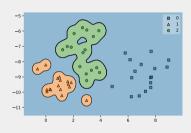
Which decision region is good?



Underfit: The model fails to recognise data pattern



Good fit: The model recognises data pattern **generally**



Overfit: The model **remembers** data pattern instead of generalising.

OVERFITTING AND UNDERFITTING

Good model must generalise