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Fundamentals of Machine Learning

Introduction

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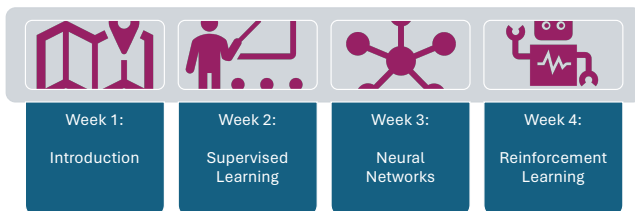


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Sessions



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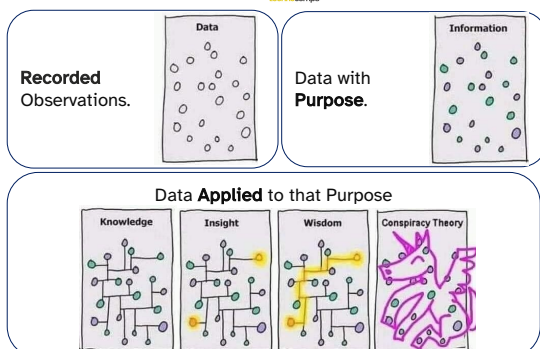
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Data





Machine Learning as a Tool

- It can be good to think of machine learning as a tool, using parts of other fields to solve real-world issues based on data.
- We may want to find patterns:
 - Detect anomalies or outliers
 - Cluster groups of similar things
 - Identify relationships between things
 - Apply a label to an observation
- These identified patterns are only useful if they allow us to then do something, we want actionable insights.

Machine Learning as a Tool

- A common example is to predict whether email is spam or not.
- Let's say that I've seen numerous emails over the last few years which tell me I can "make money quick" in some form or another. Those emails have turned out to be spam.
- My friends and colleagues send me genuine, non-spam emails, and they don't contain any mention of making money quick.

Machine Learning as a Tool

- We could create a **rule** that says:
IF subject contains “make money quick”
then the email is spam, otherwise it is not.
- Given a new email, we **apply our rule** and **predict** whether it should be labelled as spam or not.
- We **classify** it's **label** from the set of possible labels
{spam, non-spam}
- Based on something we observed (the subject content).

Machine Learning as a Tool

- However, it is quite weak rule, easy to bypass and potentially could get it wrong (misclassify).
- If I can spot that trend as a human, then what is the point of throwing machine learning at it?
- Instead we often want to find patterns/insights which humans could not find easily:
 - Complex patterns
 - Massive amounts of data

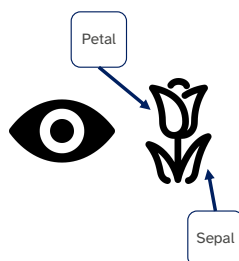
Data

- Dataum
 - “A single piece of information”
 - “A item of data”
 - “A single recorded observation”
- A dataum is often an observation or measurement of something, recording information about that thing
- A collection of multiple observations can form from data

Data

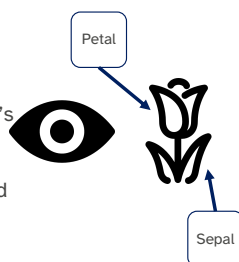
- Suppose we observe a flower
- What could we measure/record about that flower?

- Colour/ Number of petals
- Length/ Width of petals
- Length/ Width of sepals



Data

- Why might we want to?
- Well perhaps we have a hypothesis that we might be able to tell what the flower's **species** is by these measurements.
- Similar to our example example, we might want to predict the species based on our data, so that if we see a new flower, we can **classify it**



Data

- Our one observation becomes a single data point:
 - Colour: Black
 - Number of petals: 3
 - Length of petal: 1.5cm
 - Width of petal: 2cm
 - Length of sepal: 0.7cm
 - Width of sepal: 0.3cm
 - Species: Black rose





Data

- Our one observation becomes a single data point:

- Colour: Yellow
- Number of petals: 12
- Length of petal: 13cm
- Width of petal: 2cm
- Length of sepal: 3.5cm
- Width of sepal: 2cm
- Species: Sunflower



Data

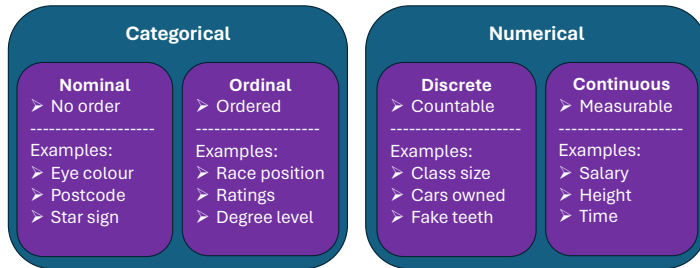
- If we then look at a lot of flowers, lots of observations perhaps we can create a large dataset on which we can apply our machine learning methods
- Maybe there are some trends in that data we collected which can help us solve our task



Data

- Data comes in a variety of shapes, sizes and types.
- We may collect numerical data, text data, images...
- We may collect pieces of information about a single observation. E.g. Name, age, address ...
- We may collect tens, thousands, or even millions of data points

Data Types



Dataset

Features

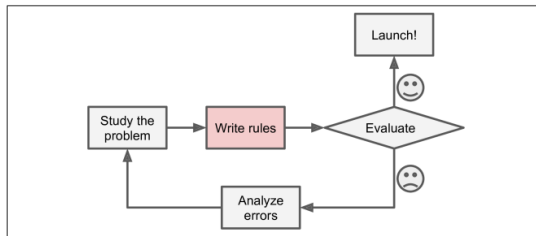
Sepal Length	Sepal Width	Petal Length	Petal Width	Species
5.1	5.5	1.4	0.2	setosa
4.9	3.0	1.4	0.2	setosa
7.0	3.2	4.7	1.4	versicolor
6.4	3.2	4.7	1.4	versicolor
6.3	3.3	6.0	2.5	virginica
5.8	2.7	5.1	1.9	virginica

• Samples
 • Instances
 • Data Points

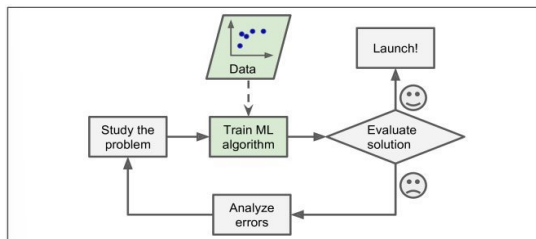
Machine Learning: Motivation



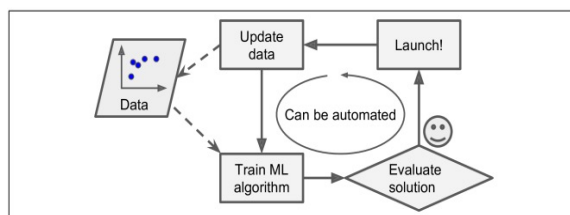
Traditional Approach



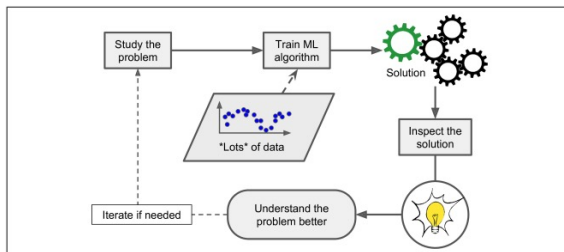
Machine Learning Approach



Automated Learning



Learning from Learning



Machine Learning: Types



Distinguishing Criteria

Three criteria to
separate machine
learning algorithms:

Human involvement during training
The machine learning paradigm

When data is used to learn.
Batch vs Online learning

How data is used to learn
Model-based vs Instance-based learning.

Learning Paradigms



Supervised Learning

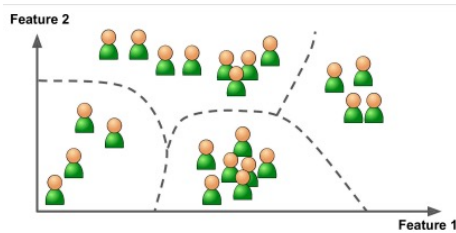


Unsupervised Learning

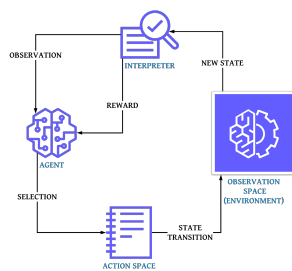


Reinforcement Learning

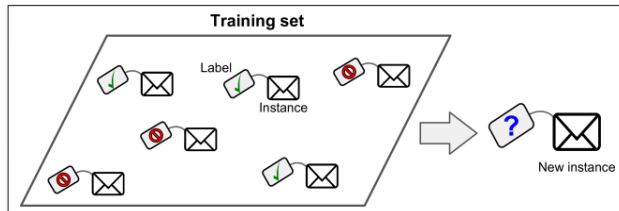
Unsupervised Learning



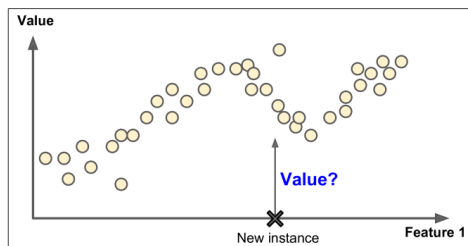
Reinforcement Learning



Supervised Learning: Classification



Supervised Learning: Regression



Supervised Learning: Training and Evaluation

