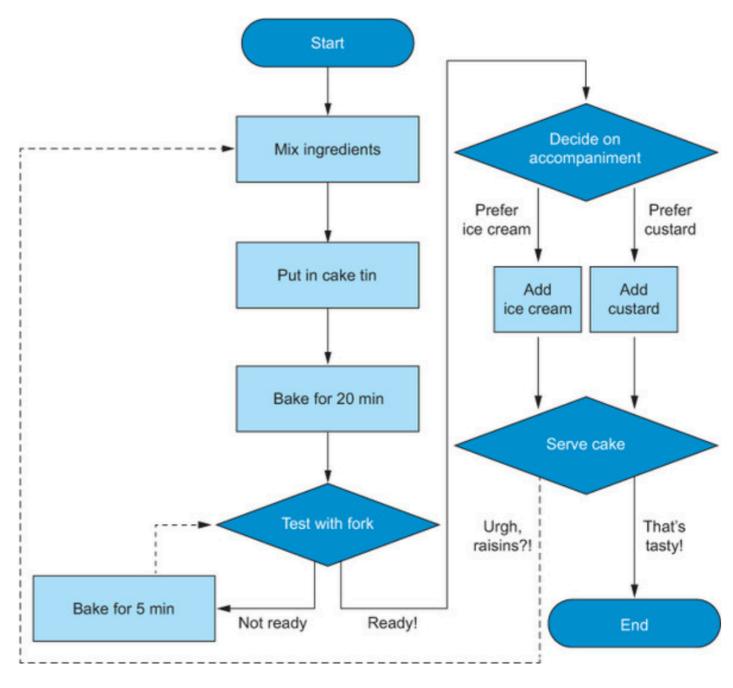
Machine learning: Regression

Alex Di Genova

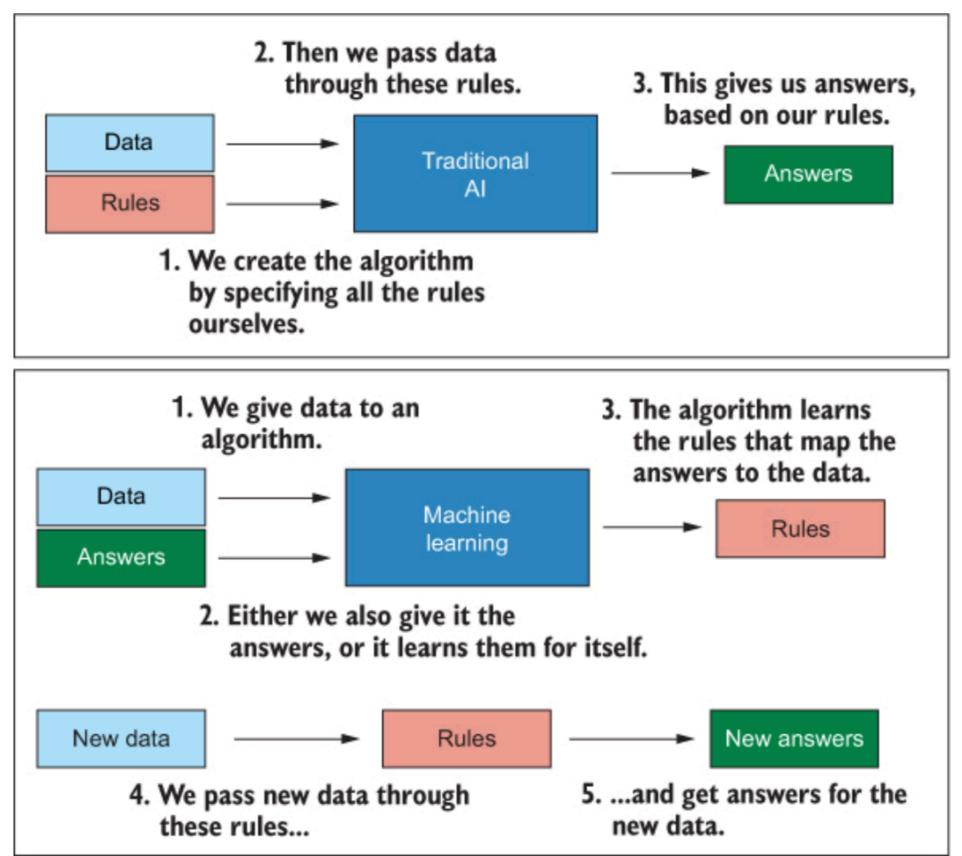
What is Machine learning?

 Machine learning, sometimes referred to as statistical learning, is a subfield of artificial intelligence (AI) whereby algorithms "learn" patterns in data to perform specific tasks.



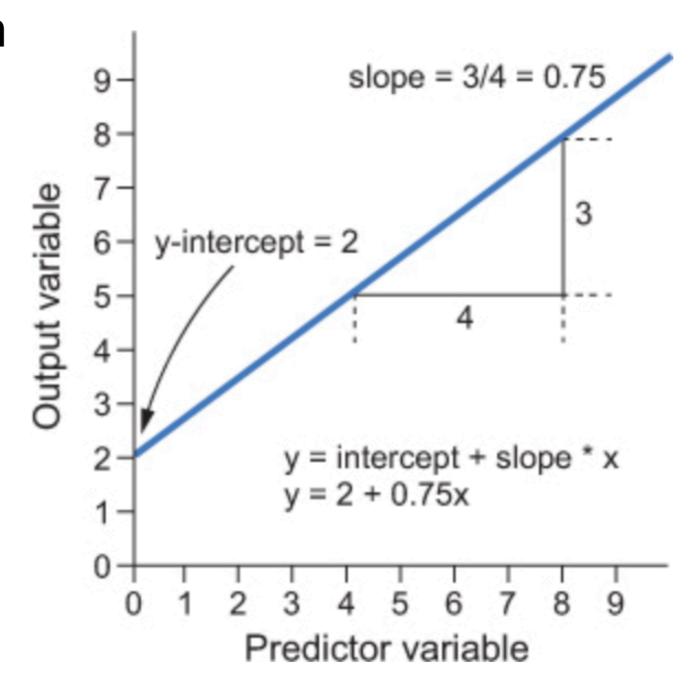
Recipe

What is Machine learning?



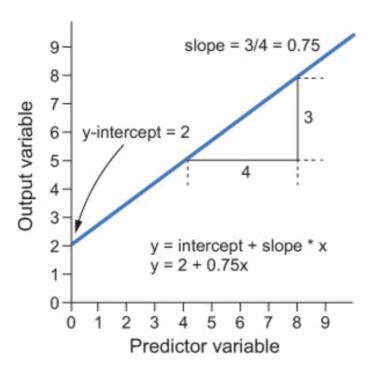
Model and algorithm

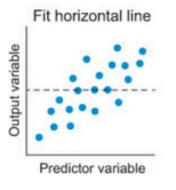
- A set of rules that a machine learning algorithm learns is a model.
- Once the model has been learned/trained, we can give it new observations, and it will output its predictions for the new data.
- Algorithm is the process by which a model learn.

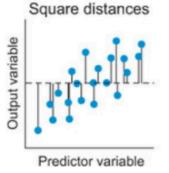


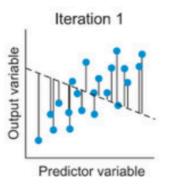
Machine learning Model and algorithm

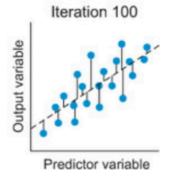
Y = intercept + slope X (parameters)

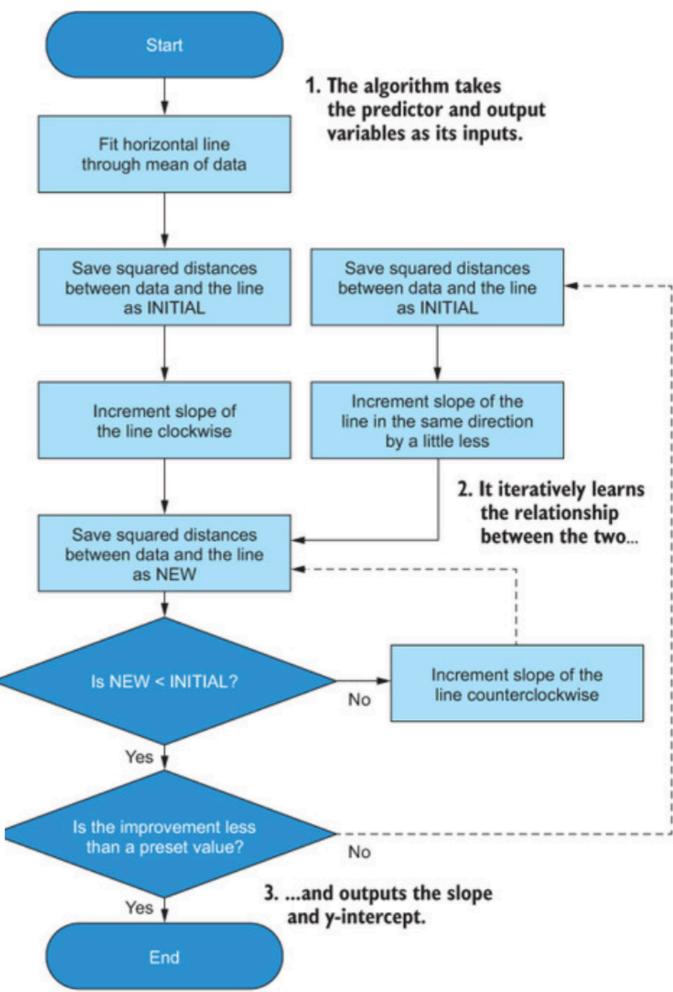








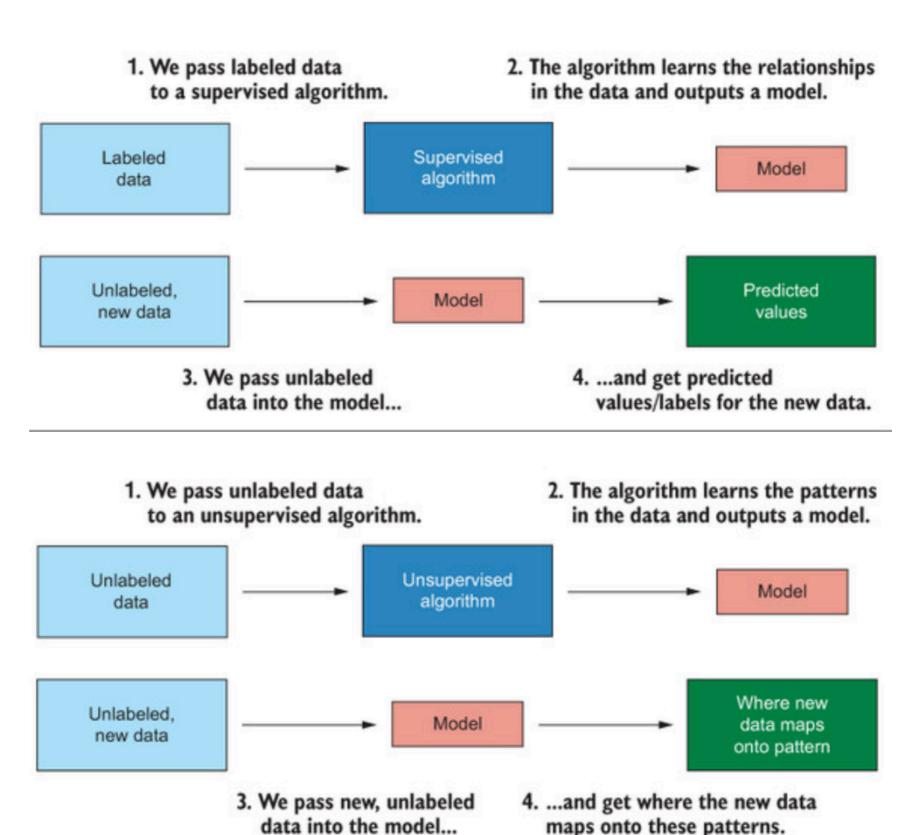




Machine learning algorithms

Classes

- Supervised
 - Classification
 - Regression
- Unsupervised
 - Dimension
 Reduction
 - Clustering
- Semi-supervised



Machine learning algorithms Classes

Supervised

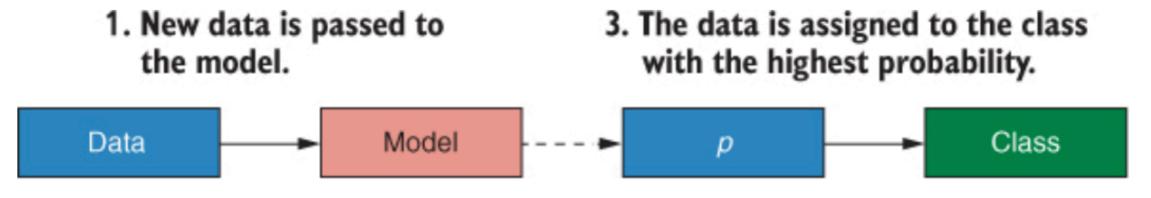
- Classification: take labeled data and learn patterns in the data that can be used to predict a categorical output variable.
- Regression: take labeled data and learn patterns in the data that can be used to predict a continuous output variable.

Unsupervised

- Dimension Reduction: take unlabeled and high-dimensional data (data with many variables) and learn a way of representing it in a lower number of dimensions (exploratory).
- Clustering: take unlabeled data and learn patterns of clustering in the data. A
 cluster is a collection of observations that are more similar to each other than
 to data points in other clusters.

Logistic Regression

 Logistic regression predicts the probability that a given input belongs to a particular class.



The model (indirectly) estimates the probability of the data belonging to each class.

Logistic regression

- Log-Odds
 - · Model the probability that a given input belongs to a particular class
 - Odds = p/1-p —- the ratio of the probability of the event occurring to the probability of it not occurring.
 - Log-Odds = log(p/1-p)
 - log-odds are modeled as a linear combination of the input variables:
 - $\log(1/1-p) = \beta 0 + \beta 1 \times 1 + \beta 2 \times 2 + ... + \beta n \times n$
- Sigmoid Function (Logistic function)
 - Maps the log-odds to a probability value between 0 and 1
 - $\sigma(z)= 1/(1+e^{-z})$ where $z=\beta 0+\beta 1 \times 1+\beta 2 \times 2+...+\beta n \times n$

Binary Classification

• If the probability is greater than a certain threshold (commonly 0.5), the input is classified as class 1; otherwise, it is classified as class 0.

Predict whether a student will pass (1) or fail (0) an exam

Data:

Hours Studied (x)	Pass/Fail (y)
1	0
2	0
3	0
4	1
5	1
6	1
7	1

1. Log-oods:

1.
$$log(1/1-p)=\beta 0 + \beta 1$$
 Hours Studied

2. Sigmoid Function:

1.p=
$$\sigma(\beta 0 + \beta 1 \cdot \text{Hours Studied}) = 1 / (1 + e^{-(\beta 0 + \beta 1 \cdot \text{Hours Studied})})$$

3. Binary Classification:

1. If p \geq 0.5 predict Pass (1). If p<0.5 predict Fail (0).

4.Let's asume model has been trained and the parameters are :

1.
$$\beta 0 = -6$$
 y $\beta 1 = 1$

Predict whether a student will pass (1) or fail (0) an exam

Data:

Hours Studied (x)	Pass/Fail (y)
1	0
2	0
3	0
4	1
5	1
6	1
7	1

1.Prediction Example

1. For a student who studied 3 Hours

2.Log-odds:

1.
$$z = log(1/1-p) = \beta 0 + \beta 1 3 = -6 + 1*3 = -3$$

3. Sigmoid Function:

$$1.p=\sigma(-3) = 1 / (1+e^{-(-3)}) = 1/(1+e^{-3}) = 1/21.08 = 0.04$$

4. Binary Classification:

1. If 0.04<0.5 the model predict Fail (0).

Book: Machine learning with R

Questions? Practice!!!