# Introduction to Machine Learning and Al

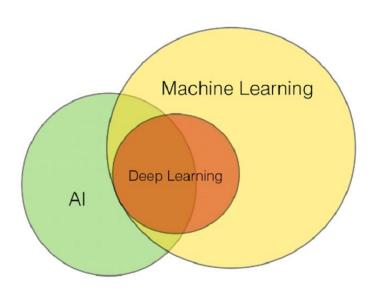


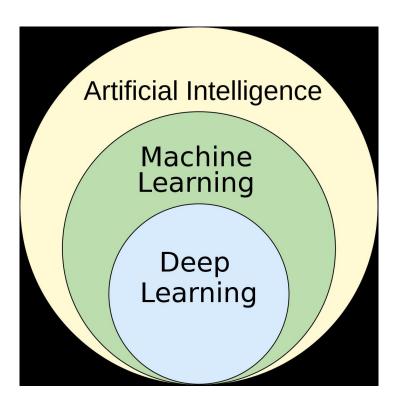
Silvan David Peter - Emmanouil Karystinaios

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## What is AI?

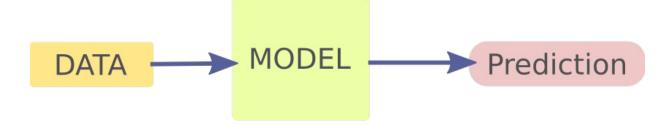






# **Learning from Data**

- What is Data?
- What are Predictions?
- What is Model?

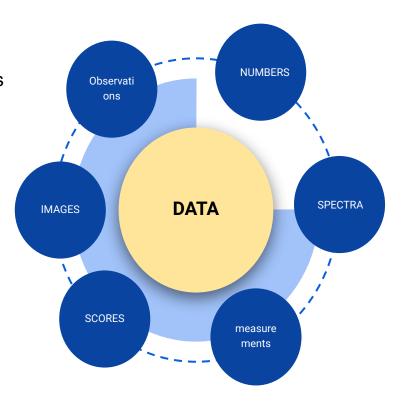




## What is Data?

Data is information.

What matters in Learning systems is that this information is represented numerically as a Number, Vector, Matrix, or tensor.





## **Predictions**

Prediction is information we can extract from data with the help of some model.

For example in Image recognition the data is an image of a dog, then the prediction would be the **label** "dog".

In systems these labels are represented by numbers, for example:

$$1 = dog$$
,  $2 = cat$ ,  $3 = car$ , etc..





## **Trainable Model**

A model simply refers to some real-valued function for example:

$$f(x) = x^2 + 1$$

A function is trainable if we can change the value of its parameters:

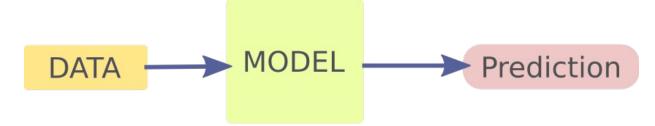
$$f(x) = ax^2 + b$$

In this case a, b



# **Learning from Data**

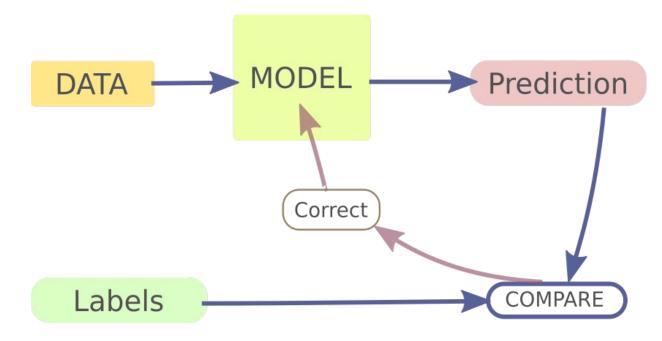
- What is Data?
  Data is Information
- What are Predictions? Information extracted from data with the help of a model.
- What is Model?
   Some kind of processing of data.





# **Supervised Learning**

- What are Labels?
- How do we compare?
- How do we correct?





# **Terminology Summary**

- Label = the variable we are predicting
- Features = the input variables describing our data
- Example/Sample = a particular instance of data, i.e. x
  - A labeled example has (features, label), i.e. (x, y)
  - A unlabeled examples has no label (used for making predictions on new data)
- Model = maps examples to predicted labels y'



### **Loss Function**

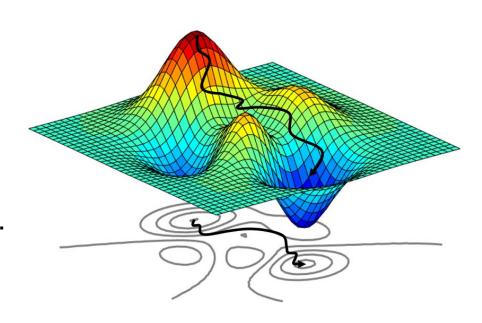
**Loss function** is a method of evaluating how well your algorithm models your dataset. If your predictions are totally off, your loss function will output a higher number. If they're pretty good, it'll output a lower number.



# **Optimization**

In the simplest case, an optimization problem consists of maximizing or minimizing a real function by systematically choosing input values from within an allowed set and computing the value of the function.

A standard algorithm for optimizing models in Machine Learning is called **Gradient Descent**. The most widely used is called **ADAM**.





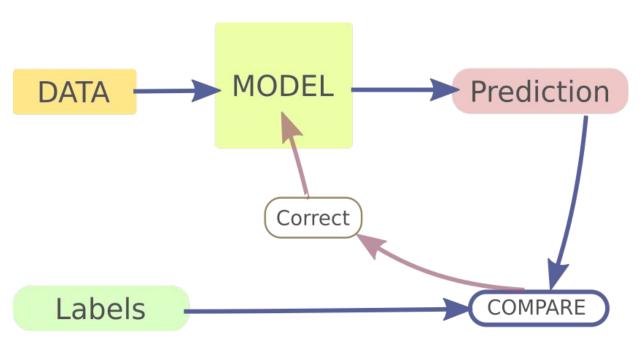
# **Supervised Learning**

- What are Labels?
- How do we compare?

**Loss Function** 

How do we correct?

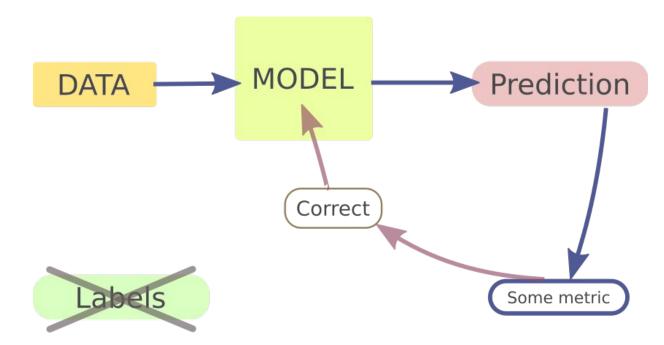
Optimization (ADAM)





# **Unsupervised Learning**

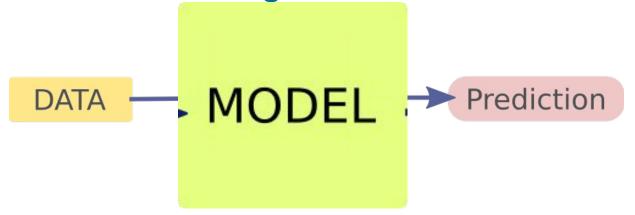
- How do we compare without Labels?
- Examples?





What is a Model in Machine Learning

- Scalar Multiplication
- Linear Transformation
- Non-LinearTransformation
- Neural Network





# **Regression vs Classification**

A **regression** model predicts continuous values. For example, regression models make predictions that answer questions like the following:

- What is the value of the Spotify stock?
- What is expected volume of the next track?

A **classification** model predicts discrete values. For example, classification models make predictions that answer questions like the following:

- Given a song will the user skip or not skip?
- Is this the genre of this song of a rock, pop or classical?



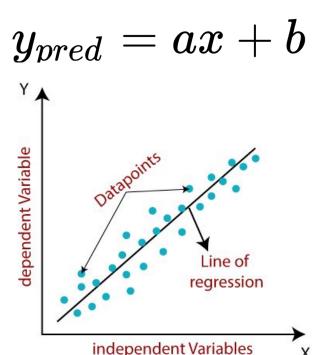
# **Linear Regression - One dimension**

Linear Regression is used to model the relationship between two variables and estimate the value of a response by using a line-of-best-fit.

The variables in question (x,y) here are assumed to have a linear relation.

The question we are asking in this scenario is:

Given f(x) = ax + b which values a, b are more suitable such as  $f(x) \approx y$ .





## **Linear Transformation**

When x and/or y are not just numbers, but vectors or matrices, then the previous example works exactly the same.

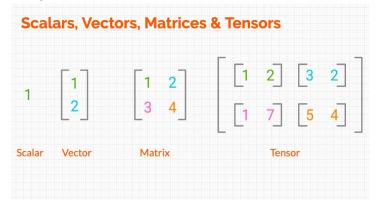
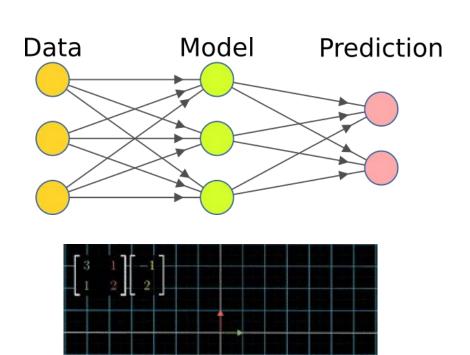
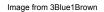


Image from Towardsdatascience.org



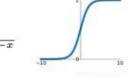




# **Non-Linear Operations**

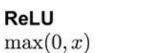
#### **Activation Functions**

#### Sigmoid



#### tanh

tanh(x)





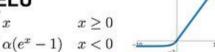
# Leaky ReLU max(0.1x, x)



#### Maxout

 $\max(w_1^T x + b_1, w_2^T x + b_2)$ 

#### ELU

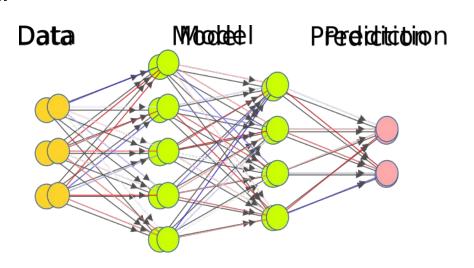


that take as input the result



# **Neural Network - Weights and Biases**

Weight is the parameter within a neural network that transforms input data within the network's hidden layers. A neural network is a series of nodes, or neurons. Within each node is a set of inputs, weight, and a bias value.





# **Stages of Learning**

#### **TRAINING**

During training we update the parameters of a model.

The training process is divided in two hierarchical sections:

- Epochs (Number of circles over the training data)
- Batches (Iterating over subsets of the training data)

#### **VALIDATION/TESTING**

During validation or testing we check if the model parameters after a session of training result to a good prediction. In other words we evaluate the model without changing its parameters.

#### **PREDICTION**

The prediction phase is the same as the validation phase but in this case we do not have a way of evaluating. After a model is trained, we feed it data for which we don't necessarily have labels and we get a prediction.





## **External Resources**

- https://www.youtube.com/watch?v=aircAruvnKk&list=PLZHQObOWTQDNU6R1\_67 000Dx ZCJB-3pi&ab channel=3Blue1Brown
  - Youtube Series for mathematical background of Neural Networks (3Blue1Brown)
- https://scikit-learn.org/
  - Introductory Python Package for Machine Learning.



