## **Machine Learning**

## Chap2

- The gradient of a function ::= a vector of partial derivatives

$$f([x^{(1)},x^{(2)}]) = ax^{(1)} + bx^{(2)} + c, \qquad \nabla f \text{ is given by the vector } \left[\frac{\partial f}{\partial x^{(1)}},\frac{\partial f}{\partial x^{(2)}}\right]$$

- Random variables
  - Usually expressed by italic capital letter
  - Expectation value

$$\mathbb{E}[X] \stackrel{\text{def}}{=} \sum_{i=1}^{k} [x_i \cdot \Pr(X = x_i)]$$
$$= x_1 \cdot \Pr(X = x_1) + x_2 \cdot \Pr(X = x_2) + \dots + x_k \cdot \Pr(X = x_k),$$

- Hyperparameters -> set by data analysts
- Parameters -> directly modified by learning algorithms
- Classification ::= assign labels to unlabeled examples
- Regression ::= predicting real valued labels (target)
- Model based & Instrumental Based Learning
  - Model-based:
    - Creates models
    - Develops parameter
    - Like w\* & b\* of SVM
  - Instrumental Based Learning
    - K-Nearest Neighbors (KNN)
      - Look at the close neighborhood of the input, output the most often ones.

- "Instead of estimating a model, store all training examples in memory and make predictions using a similarity measure".
- See demo <a href="https://github.com/zotroneneis/machine\_learning\_basics/blob/master/k\_nearest\_neighbour.ipynb">https://github.com/zotroneneis/machine\_learning\_basics/blob/master/k\_nearest\_neighbour.ipynb</a>