- Introduction
 - Organization, tooling

- Introduction
 - Organization, tooling
 - Motivations and Use Cases

- Introduction
 - Organization, tooling
 - Motivations and Use Cases
 - Big Data movement

- Introduction
 - Organization, tooling
 - Motivations and Use Cases
 - Big Data movement
 - Ideas and context

- Introduction
 - Organization, tooling
 - Motivations and Use Cases
 - Big Data movement
 - Ideas and context

- Introduction
 - Organization, tooling
 - Motivations and Use Cases
 - Big Data movement
 - Ideas and context

- Introduction
 - Organization, tooling
 - Motivations and Use Cases
 - Big Data movement
 - Ideas and context

- Introduction
 - Organization, tooling
 - Motivations and Use Cases
 - Big Data movement
 - Ideas and context

- Introduction
 - Organization, tooling
 - Motivations and Use Cases
 - Big Data movement
 - Ideas and context

- Introduction
 - Organization, tooling
 - Motivations and Use Cases
 - Big Data movement
 - Ideas and context
- - Supervised Learning
 - Unsupervised Learning
 - The Goals of Unsupervised Learning
 - Reinforcement Learning
 - Summary



- 4 courses of 3 hours each, with practical examples
- 3 Practical works of 4 hours each, with R/RStudio and python/anaconda programs / libraries
- 3 short quizzes
- Slides from several MOOCs, advanced materials

- Introduction
 - Organization, tooling
 - Motivations and Use Cases
 - Big Data movement
 - Ideas and context
- Statistical Learning, Machine Learning frameworks
 - Supervised Learning
 - Unsupervised Learning
 - The Goals of Unsupervised Learning
 - Reinforcement Learning
 - Summary

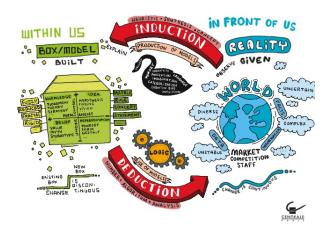


- How to understand data sets, correlation but not causation
- How to handle variety of data sets (see 3V description later)
- See materials for Supervised (*) and Unsupervised (*) use-cases
- (*) Definition to come later

- Introduction
 - Organization, tooling
 - Motivations and Use Cases
 - Big Data movement
 - Ideas and context
- Statistical Learning, Machine Learning frameworks
 - Supervised Learning
 - Unsupervised Learning
 - The Goals of Unsupervised Learning
 - Reinforcement Learning
 - Summary

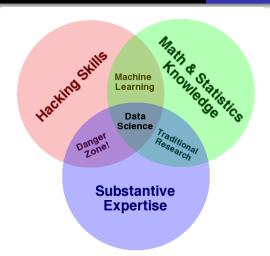
- 3V definition :
 - Volume, Velocity, Variety
 - + Veracity, ++
- Position of data mining vs ML vs Statistical Learning vs AI

- Introduction
 - Organization, tooling
 - Motivations and Use Cases
 - Big Data movement
 - Ideas and context
- Statistical Learning, Machine Learning frameworks
 - Supervised Learning
 - Unsupervised Learning
 - The Goals of Unsupervised Learning
 - Reinforcement Learning
 - Summary



• From a MOOC entitled What the manager can learn from Philosophy





Blog post http://drewconway.com/zia/2013/3/26/the-data-science-venn-diagram

Contents

Introduction

Contents

Introduction

Contents

Introduction

- Introduction
- 2 Statistical Learning, Machine Learning frameworks

- Introduction
- Statistical Learning, Machine Learning frameworks
 - Supervised Learning

- 1 Introduction
- 2 Statistical Learning, Machine Learning frameworks
 - Supervised Learning
 - Unsupervised Learning

- Introduction
- 2 Statistical Learning, Machine Learning frameworks
 - Supervised Learning
 - Unsupervised Learning
 - The Goals of Unsupervised Learning

- 1 Introduction
- 2 Statistical Learning, Machine Learning frameworks
 - Supervised Learning
 - Unsupervised Learning
 - The Goals of Unsupervised Learning
 - Reinforcement Learning

- Introduction
- 2 Statistical Learning, Machine Learning frameworks
 - Supervised Learning
 - Unsupervised Learning
 - The Goals of Unsupervised Learning
 - Reinforcement Learning
 - Summary

- Introduction
 - Organization, tooling
 - Motivations and Use Cases
 - Big Data movement
 - Ideas and context
- 2 Statistical Learning, Machine Learning frameworks
 - Supervised Learning
 - Unsupervised Learning
 - The Goals of Unsupervised Learning
 - Reinforcement Learning
 - Summary



Starting point :

- Outcome measurement Y (also called dependent variable, response, target);
- Vector of p predictor measurements X_i (also called inputs, regressors, covariates, features, independent variables). X is a matrix of dimension (N,p), where N is the number of measurements;
- In the regression problem, Y is quantitative (e.g price, blood pressure);
- In the classification problem, Y takes values or levels (categories) in a finite, unordered set (survived/died, digit 0-9, cancer class of tissue sample);
- We have training data $(x_1, y_1), ..., (x_N, y_N)$. These are observations (examples, instances) of these measurements.



- Introduction
 - Organization, tooling
 - Motivations and Use Cases
 - Big Data movement
 - Ideas and context
- 2 Statistical Learning, Machine Learning frameworks
 - Supervised Learning
 - Unsupervised Learning
 - The Goals of Unsupervised Learning
 - Reinforcement Learning
 - Summary



- In unsupervised learning, we observe only the features X1, X2, . . . , Xp.
- We are not interested in prediction, because we do not have an associated response variable Y.

- Introduction
 - Organization, tooling
 - Motivations and Use Cases
 - Big Data movement
 - Ideas and context
- 2 Statistical Learning, Machine Learning frameworks
 - Supervised Learning
 - Unsupervised Learning
 - The Goals of Unsupervised Learning
 - Reinforcement Learning
 - Summary



- The goal of Unsupervised Learning is to discover interesting things about the measurements: is there an informative way to visualize the data? Can we discover subgroups among the variables or among the observations?
- We discuss two methods :
 - principal components analysis (PCA), a tool used for data visualization or data pre-processing before supervised techniques are applied, and
 - **clustering**, a broad class of methods for discovering unknown subgroups in data.

- Introduction
 - Organization, tooling
 - Motivations and Use Cases
 - Big Data movement
 - Ideas and context
- Statistical Learning, Machine Learning frameworks
 - Supervised Learning
 - Unsupervised Learning
 - The Goals of Unsupervised Learning
 - Reinforcement Learning
 - Summary



- In reinforcement learning, observations are done one by one.
 (e.g. temporal series)
- After each prediction a reward is given
- Use Case : Chess or Go player
- toward common sense...

- Introduction
 - Organization, tooling
 - Motivations and Use Cases
 - Big Data movement
 - Ideas and context
- Statistical Learning, Machine Learning frameworks
 - Supervised Learning
 - Unsupervised Learning
 - The Goals of Unsupervised Learning
 - Reinforcement Learning
 - Summary



The 3 types of Machine Learning From the most difficult to the less

Unsupervised Learn.	Supervised Learn.	Reinforcement Learn.
Observations	+ targets (labels)	++ rewards
Understand	Predict	Decide
Clustering	Classification	Action Policy
	or Regression	Strategy

ML Cake

From the Yann Le Cun's lesson at College de France

Reinforcement Learning (cherry)

- The machine predicts a scalar reward given once in a while.
- A few bits for some samples

Supervised Learning (icing)

- The machine predicts a category or a few numbers for each input
- 10→10,000 bits per sample

Unsupervised Learning (cake)

- The machine predicts any part of its input for any observed part.
- Predicts future frames in videos
- Millions of bits per sample

