
Machine Learning and AI

- Methods and Algorithms -

Personnal Notes
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■ Add "you cannot backup death" explanations	6
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Intro

This document will use the following classification for the machine learning algorithms. However their might be some changes. For exemple, some of them will be part of the commons algorithms and not from their real class.

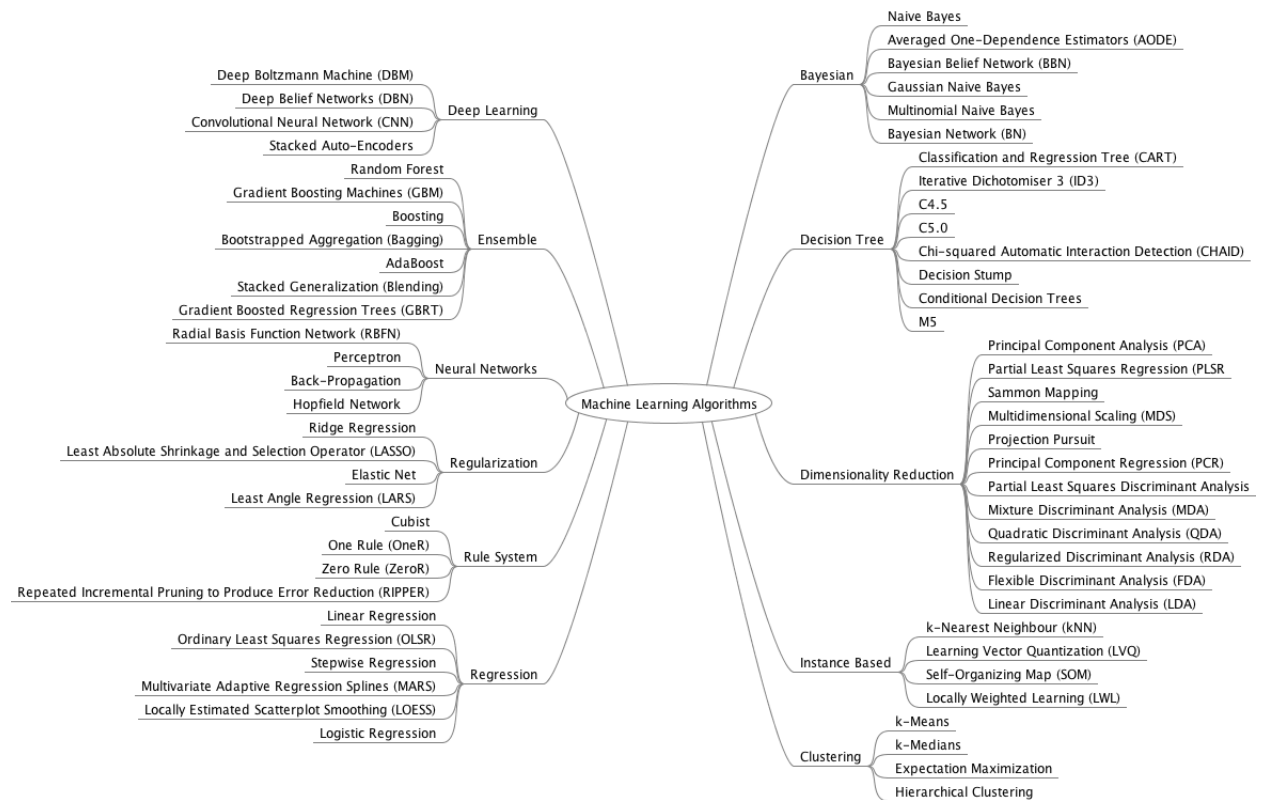


Figure 1 – Simple graph for algorithms classification in ML

Chapter 1

Common Machine Learning algorithms

This chapter is dedicated to the most common ML algorithms, a major part of the notes come from the mml-books.com

Find better paragraph layout

Add bibtex reference

1.1 Linear Regression

1.1.1 Maximum Likelihood Estimation (MLE)

Closed-Form Solution

Maximum A Posteriori Estimation (MAP)

1.2 Gradient Descent

1.2.1 Simple Gradient Descent

1.2.2 Gradient Descent with Momentum

1.2.3 Stochastic Gradient Descent

1.3 Model Selection and Validation

1.3.1 Cross-Validation

1.3.2 Marginal Likelihood

1.4 Bayesian Linear Regression

1.4.1 Mean and Variance

1.4.2 Sample function

Chapter 2

Reinforcement Learning

2.1 Markov Reward and Decision Process

2.1.1 State Value Function Closed-form

For a Markov Reward Process $(\mathcal{S}, \mathcal{P}, \mathcal{R}, \gamma)$, defining the Return R_t and the State Value Function $v(s) = \mathbb{E}[R_t | S_t = s]$

Then we have, in a vector form :

$$\mathbf{v} = (\mathbb{1} - \gamma \mathcal{P})^{-1} \mathcal{R}$$

2.1.2 Iterative Policy Evaluation Algorithm

2.2 Dynamic Programming in RL

2.2.1 Policy Iteration Algorithm

2.2.2 Value Iteration Algorithm

2.2.3 Asynchronous Backup in RL

Prioritised Sweeping

Real-time Dynamic Programming

2.2.4 Properties and drawbacks of Dynamic Programming

2.3 Model-Free Learning

2.3.1 Monte-Carlo Algorithms

(First Visit) Monte-Carlo Policy Evaluation

Add "you cannot backup death" explanations

Every Visit Monte-Carlo Policy Evaluation

Batch vs Online Monte-Carlo

Incremental Monte-Carlo Update

Runing Mean for Non-Stationnary World

2.3.2 Monte-Carlo Control Algorithms

Monte-Carlo Policy Improvement

Greedy Policy Improvement over State Value Function

Greed Policy Improvement over State-Action Value Function

Don't forget Starting to explore

Exploring Starts Problem

On-Policy ϵ -greedy first-visit Monte-Carlo control Algorithm

Monte-Carlo Batch Learning to Control

Monte-Carlo Iterative Learning to Control

2.3.3 Temporal Difference Learning

Add Comparison between MC and TD learning

Temporal Difference Value Function Estimation Algorithm

2.3.4 Temporal Difference Learning Control Algorithm

SARSA - On Policy learning Temporal Difference Control

SARSA-Lambda

Hindsight Experience Replay

Q-Learning: Off-Policy Temporal Difference Learning

2.4 Reinforcement Learning with Function Approximation

2.4.1 Exemple of features

Coarse Coding

Tile Coding

Radial-Basis Function

Deep Learning

2.4.2 Monte-Carlo with Value Function Approximation

2.4.3 Temporal Difference Learning with Value Function Approximation

2.4.4 Q-Learning with FA

2.4.5 subsection name

2.5 Deep Learning Reinforcement Learning

2.5.1 Experience Replay

2.5.2 Target Network

2.5.3 Clipping of Rewards

2.5.4 Skipping of Frames