









Decision Tree cheat sheet

- The Decision Tree is the as like the way we make decisions day-to day life.
- It is one the of supervised ML algo.
- It looks like an if-else ladder code in python.
- tree-based algorithm.
- classification and regression algorithm
- e.g. Doctor treating cancer patient deciding which medicines to give on certain decisions
- Works better for classification.
- Non Parametric
- We prefer over logistic regression when there is a lot of categorial-mixed data.
- No impact of outliers.
- We try to achieve homogeneity in DT.
- To achieve homogeneity we use:

1 Entropy = -[\sum Pi log Pi]

2 Gini = 1- [\sum Pi²]

- By using entropy and Gini we calculate Information Gain(IG) of Decision Node.
- IG = H(s) wt. avg * (entropy of each feature)
- If IG is maximum then leaf node is tending to be pure.
- Leaf node carries the classifica.







- Decision node used to make decision.
 - If IG or Gini has minimum value i. e less impurity then Root or Decision node is best node.
 - Decision Tree tends to overfit a lot.
 - To overcome overfitting we use regularization

Tree Truncation: Hyperparameter Tuning: Min sample split, Min sample leaf, max depth

Tree Pruning: Cost Complexity Pruning i.e. CCP alpha. As CCP alpha values get increases the leaf node will get pruning.

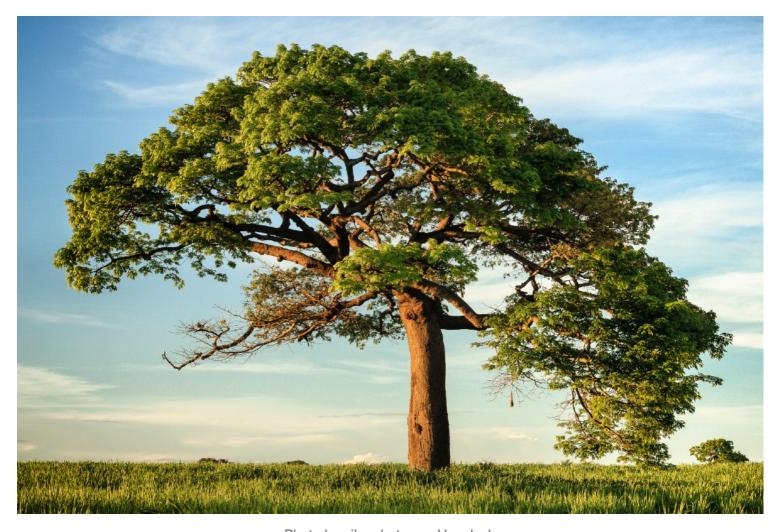


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Advantages and Disadvantages

Advantages:

- 1. No effect of outliers.
- 2.Classification and regression.
- 3.Non Parametric

- 4.Easy to implement and understand.5.We can visualize the tree.
- 6.Scaling is not required.

Disadvantages:

- 1.Tends to overfit a lot.
- 2. Low Bias and High Variance mostly.
- 3.Unstable a lot.

ID3 for multiclass. CART for binary.

Towards Data Science

Decision Tree

Data Science

Machine Learning

Artificial Intelligence