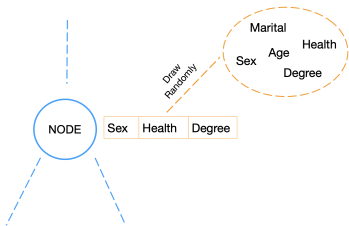


# Introduction to Machine Learning

## Random Forest: In a Nutshell

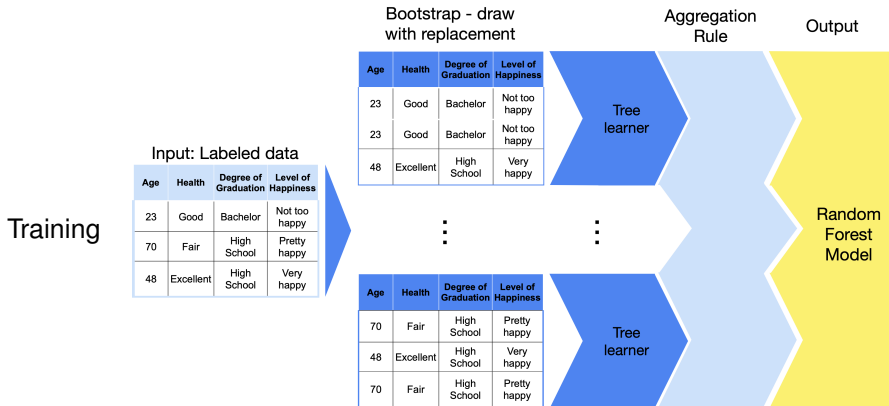


### Learning goals

- Understand basic concept of random forest
- Know basic aggregation rules
- Understand concept of feature importance

# LEARNING AND PREDICTION WITH RF

- Stabilizes tree learner by bagging (bootstrap aggregation)
- Randomizes tree learner and combines models into one meta model
- Can be adapted to learning task, i.e., classification or regression



# LEARNING AND PREDICTION WITH RF

Prediction

Input: Unlabeled data

Age	Health	Degree of Graduation	Level of Happiness
41	Fair	Bachelor	?
35	Good	Bachelor	?
22	Fair	High School	?

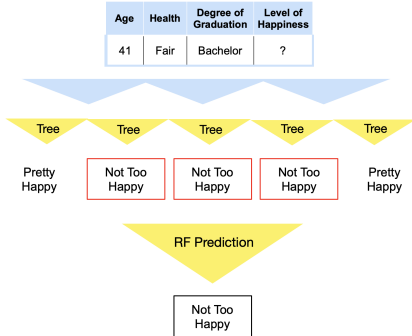
Random  
Forest  
Model

Prediction

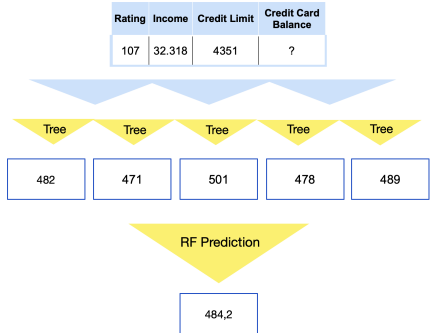
Level of Happiness
Not too happy
Pretty happy
Not too happy

# AGGREGATION RULES FOR DIFFERENT TASKS

Classification Task - Majority Vote

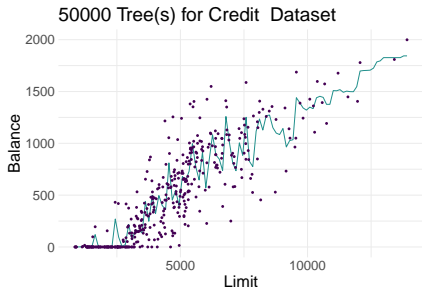
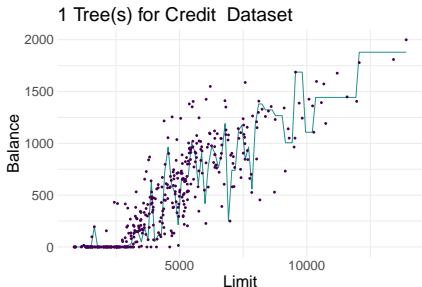


Regression Task - Averaging



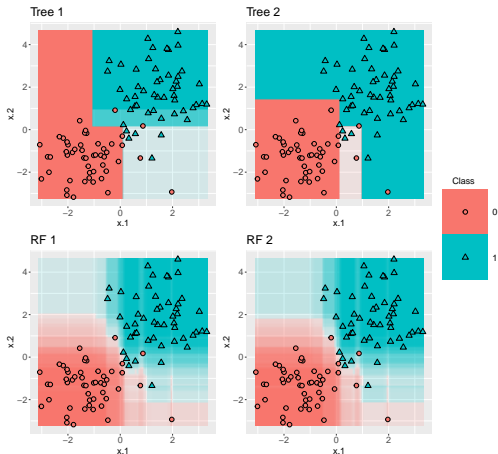
# PERFORMANCE OF RF

- In general: Increasing the ensemble size stabilizes the predictions
  - For regression tasks the stabilization is often not sufficient.



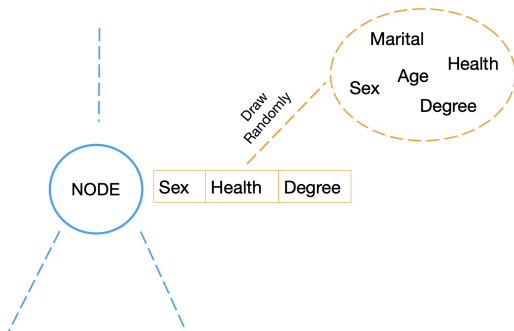
# PERFORMANCE OF RF

- RF performs well for classification tasks:
  - Two different trees → Quite different decision regions
  - Two different RFs → Similar decision regions



# PERFORMANCE OF RF

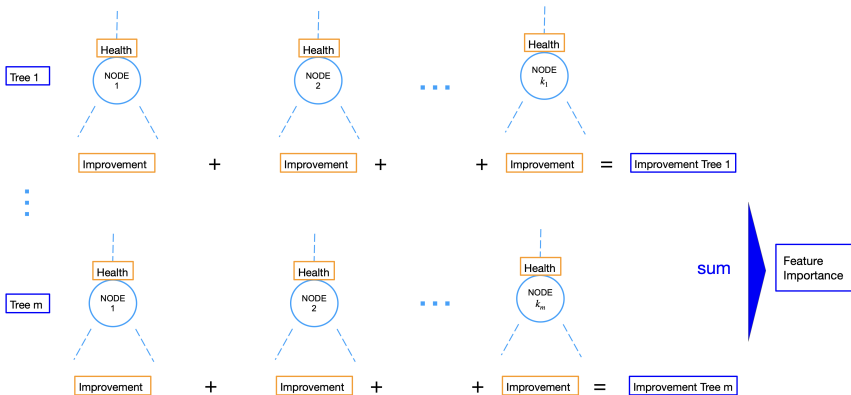
- Trees should be decorrelated, i.e., make mistakes in different directions
- Avoid correlation by
  - Bootstrap sampling
  - Randomized splits. In each node of each tree, consider different features for splitting:



# FEATURE IMPORTANCE

Several options, e.g., measure contribution of feature to model:

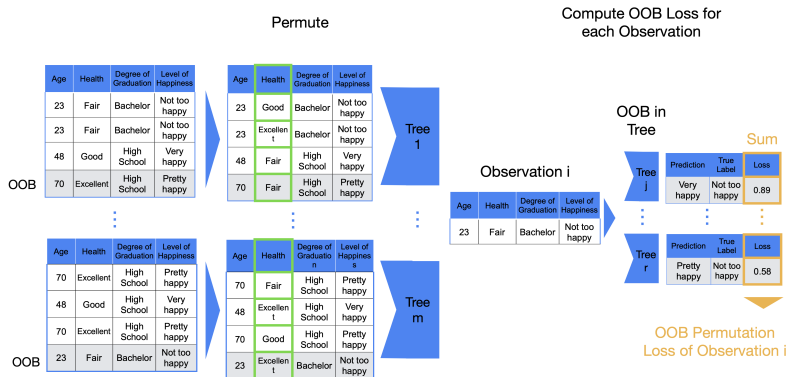
- Measure based on improvement in splitting criterion
- E.g. Feature importance of 'Health', search all nodes with 'Health' as splitting variable:



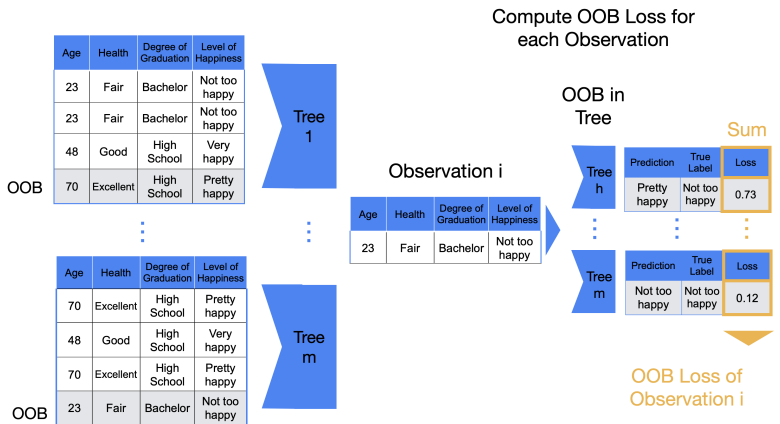


# FEATURE IMPORTANCE

- Measure based on OOB Loss



# FEATURE IMPORTANCE



Feature Importance  
'Health'

=

OOB Permutation Loss  
of all observations

-

OOB Loss of all  
observations