DSC 255 - MACHINE LEARNING FUNDAMENTALS

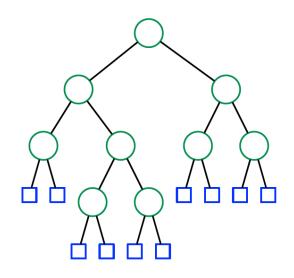
# **RANDOM FORESTS**

SANJOY DASGUPTA, PROFESSOR

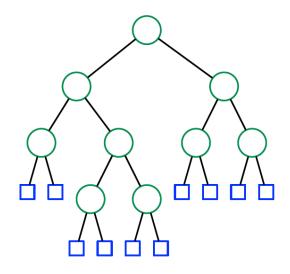


UC San Diego

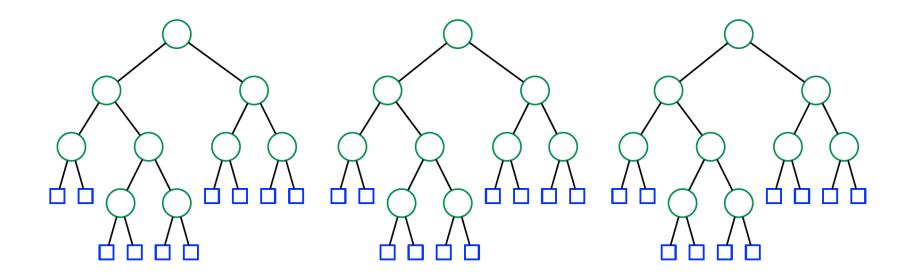
COMPUTER SCIENCE & ENGINEERING
HALICIOĞLU DATA SCIENCE INSTITUTE



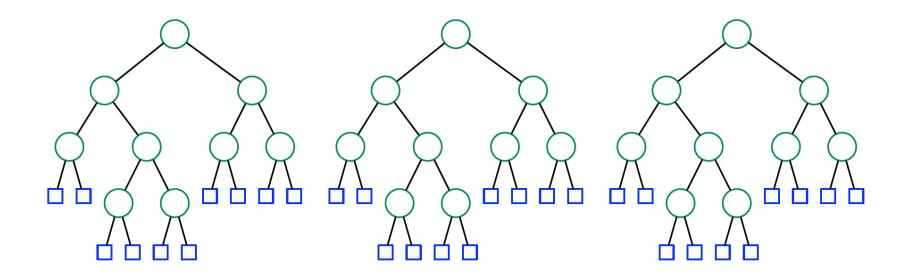
Decision tree.



**Decision tree.** Starts overfitting beyond a point.



- Decision tree. Starts overfitting beyond a point.
- Boosted decision trees.



- Decision tree. Starts overfitting beyond a point.
- Boosted decision trees. Learning is sequential, slow.

#### **Random Forests**

Given a data set S of n labeled points:

- For t = 1 to T:
  - $\triangleright$  Choose n' points randomly, with replacement, from S.
  - $\triangleright$  Fit a decision tree  $h_t$  to these points.
    - At each node restrict to one of k features chosen at random.

# Example settings:

- n'=n
- $k = \sqrt{d}$  for d-dimensional data

Final predictor: majority vote of  $h_1, ..., h_T$ .

## **Ecological Prediction Problem: "Covertype" Data**

# **Predict forest type:**

- Spruce-fir
- Lodgepole pine
- 5 other classes

# 54 cartographic/geological features:

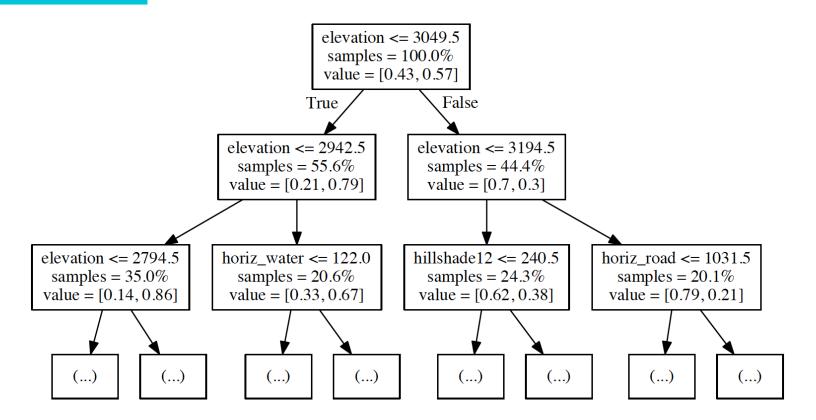
- Elevation, slope, amount of shade, ...
- Distance to water, road, ...
- Soil type

#### Data set details:

- 49,514 training points
- 445,627 test points



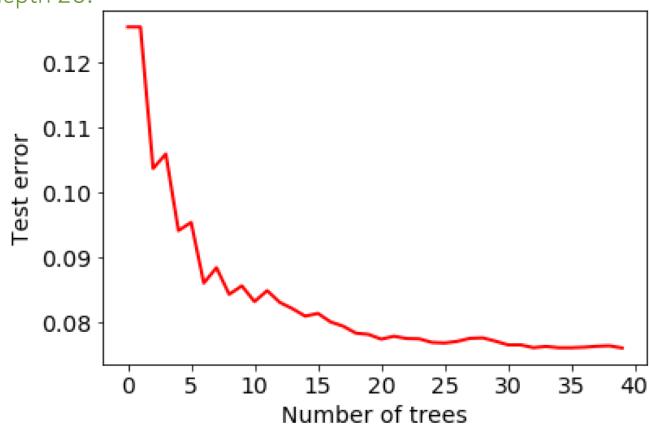
#### **Decision Tree**



Depth 20: training error 1%, test error 12.6%

## **Boosted Decision Trees**

Trees of depth 20.



#### **Random Forest**

#### Recall:

- Decision tree: depth 20, test error 12.6%
- Boosted decision trees, 10 trees, depth 20: test error 8.7%

Random forest setting: 10 trees, 50% features dropped, depth 40.

- Each individual tree has test error 15% to 17%
- Forest test error: 8.8%