





The background of the slide is a dense, close-up photograph of succulent plants, likely Sedum or similar, with small, rounded, green leaves. A large, semi-transparent white rectangular box is centered over the image, serving as a backdrop for the title text. The title "Decision Trees" is written in a large, bold, dark green sans-serif font. The word "Decision" is on the top line, and "Trees" is on the bottom line. The text is centered within the white box.

# Decision Trees

Alex, Becca, Matt, Val





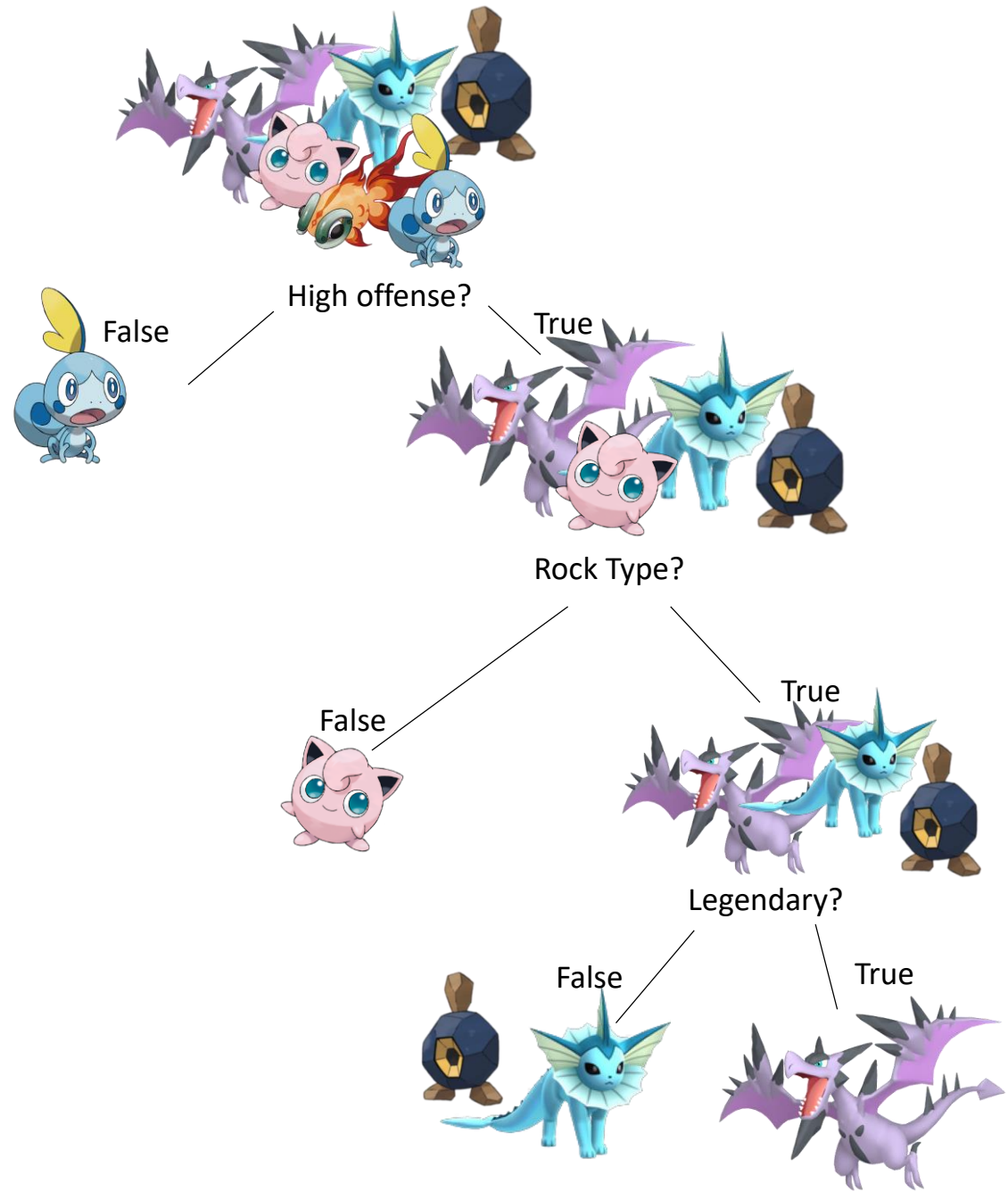
# Overview

## Decision Trees

- Nonparametric supervised learning algorithm
- Best for decision making and prediction
- Bases decisions on how a previous set of questions were answered

# Example

- Ash wants to choose one of his Pokechu to battle a Charizard:
  - High offense
  - Rock
  - Legendary





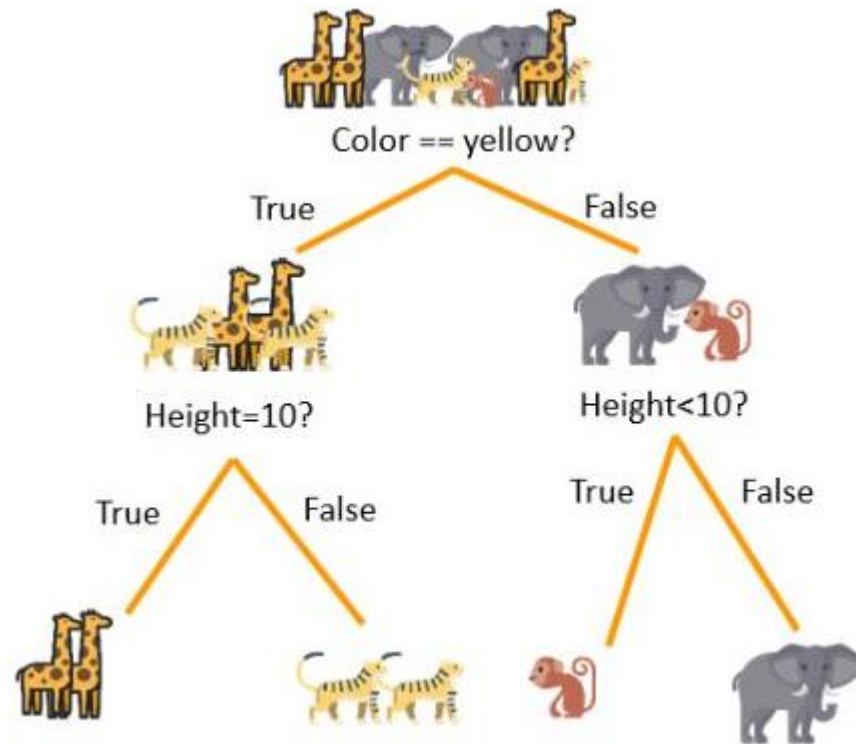
# Terms to Know

- Tree
  - Hierarchical structure mapping the possible outcomes of choices
- Nodes
  - Root node: the starting point of the tree
  - Decision node: Point where a decision must be made
  - End node: final outcomes of a decision path
- Entropy
  - Measure of disorder/randomness in a data set
- Information gain
  - Measure of how much information a feature provides about a class
    - Used to decide whether a feature should be used to split a node or not
- Greedy algorithm
  - Finds optimal solution at local level, but without “big picture” view

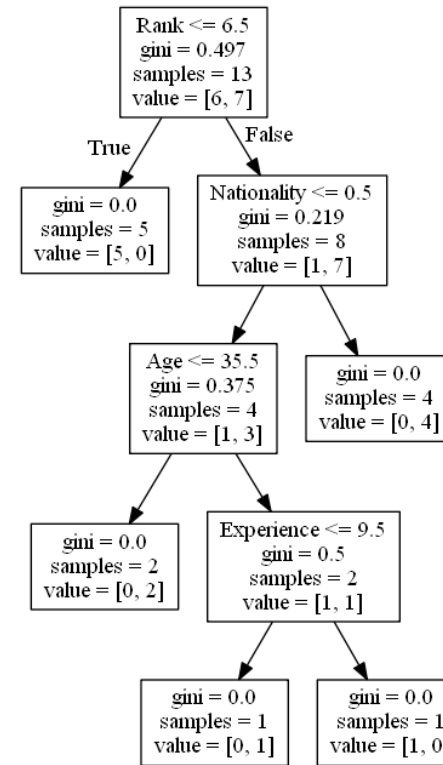


# Types

## Categorical



## Regression



# Test your Knowledge

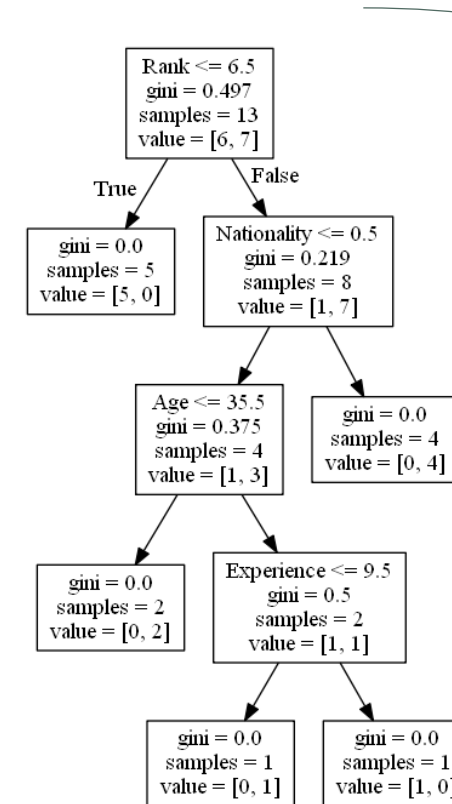
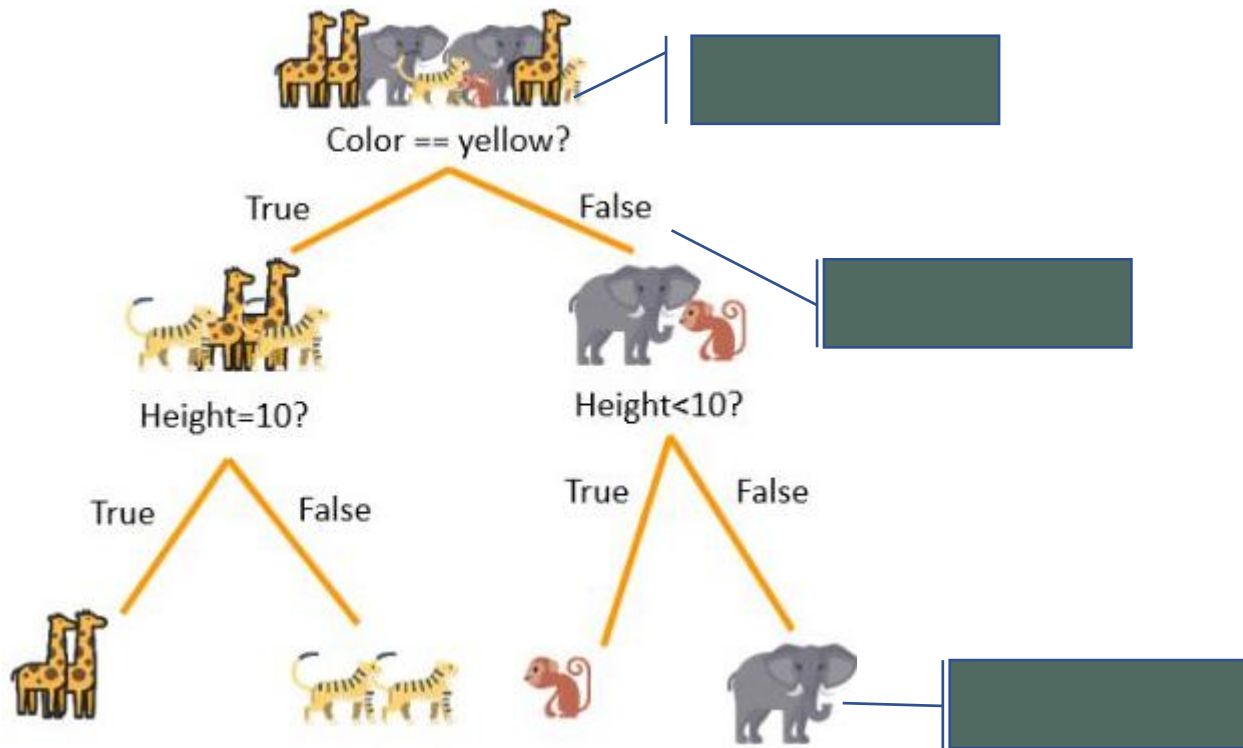
Tree

Root node

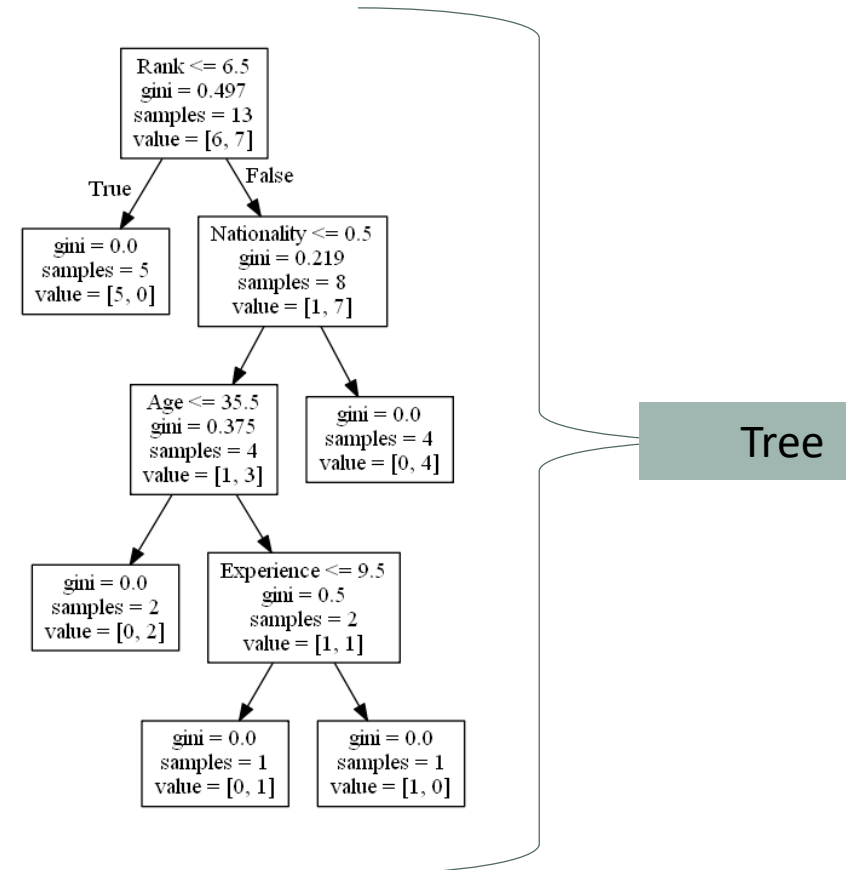
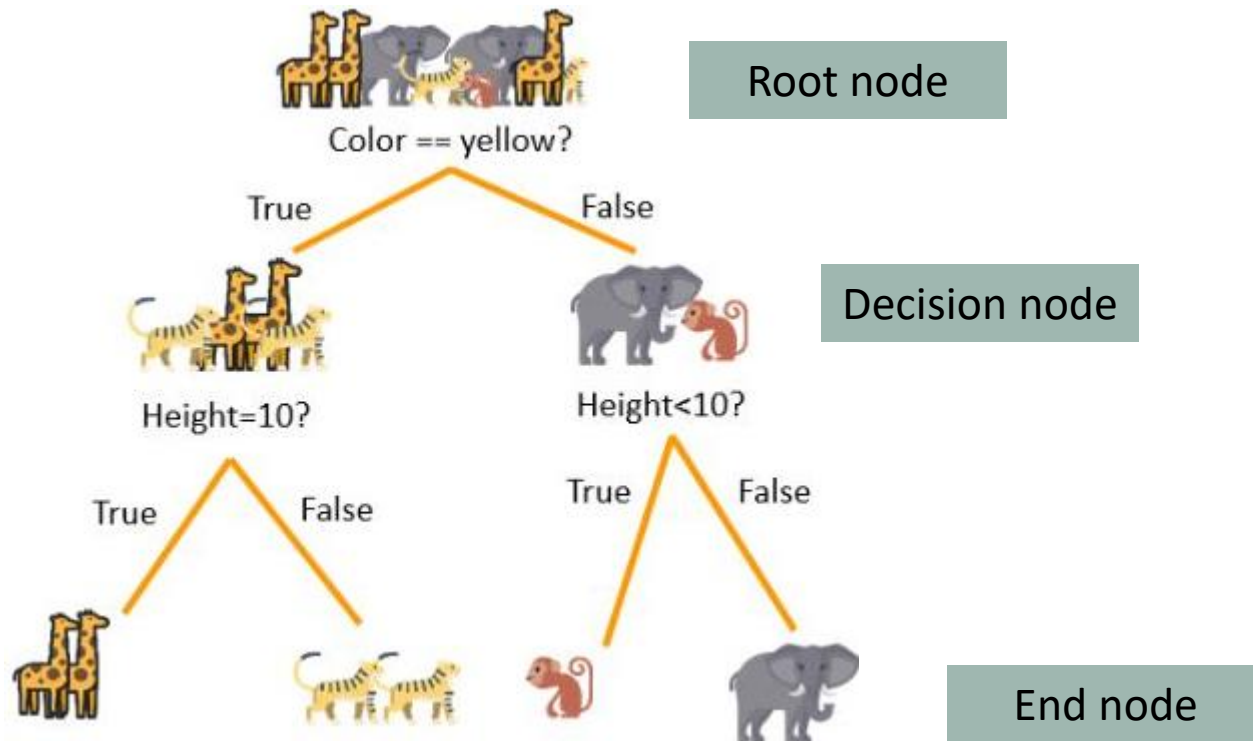
Decision node

End node

Entropy



# Answers





# Advantages and Disadvantages

- Advantages

- High accuracy
- Less up-front effort
  - Normalization not necessary
- Easy to depict and explain
- Flexible model, no prerequisites

- Disadvantages

- Does not support missing values (NaN)
- Can have high variance
- Takes more processing power/time
- Inadequate for continuous variables



# Hyperparameters

- Maximum depth: determines the maximum number of levels in the decision tree
- Minimum samples split: sets the minimum number of samples required to split an internal node
- Minimum samples leaf: sets the minimum number of samples required to be at a leaf node
- Maximum features: sets the maximum number of features considered for splitting a node
- Criterion: determines the metric used for evaluating the quality of a split (e.g. gini impurity, entropy)
- Splitter: determines the strategy used for splitting at each node (e.g. best, random)
- Class weight: determines the weights of classes in case of imbalanced data
- Random state: sets the random seed for reproducibility of results.



# Resources

- <https://www.geeksforgeeks.org/decision-tree/>
- [Video Tutorial](#)
- [Documentation \(Scikit Learn\)](#)

# Further Reading

- [Fisher Yates Algorithm \(article\)](#)
- [Simpson Index \(scientific paper\)](#)





**Questions?**