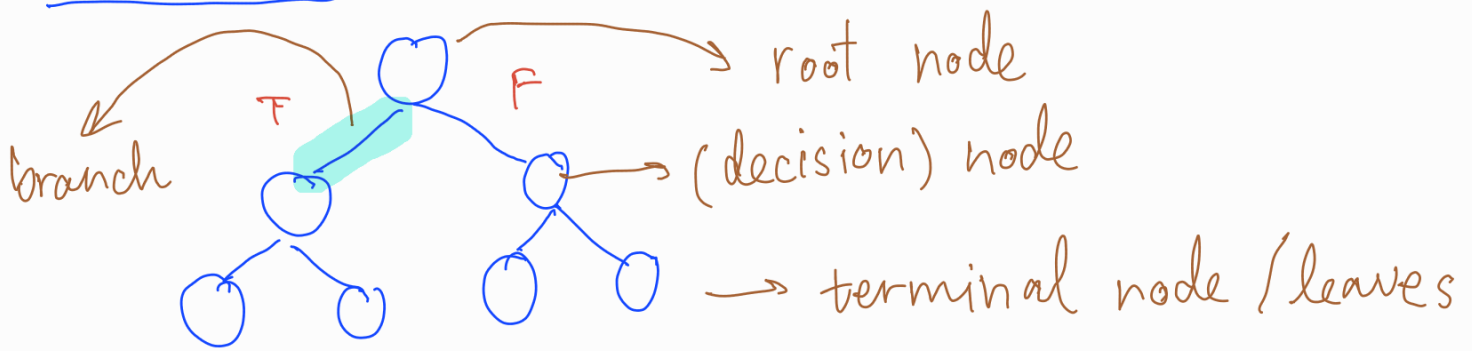
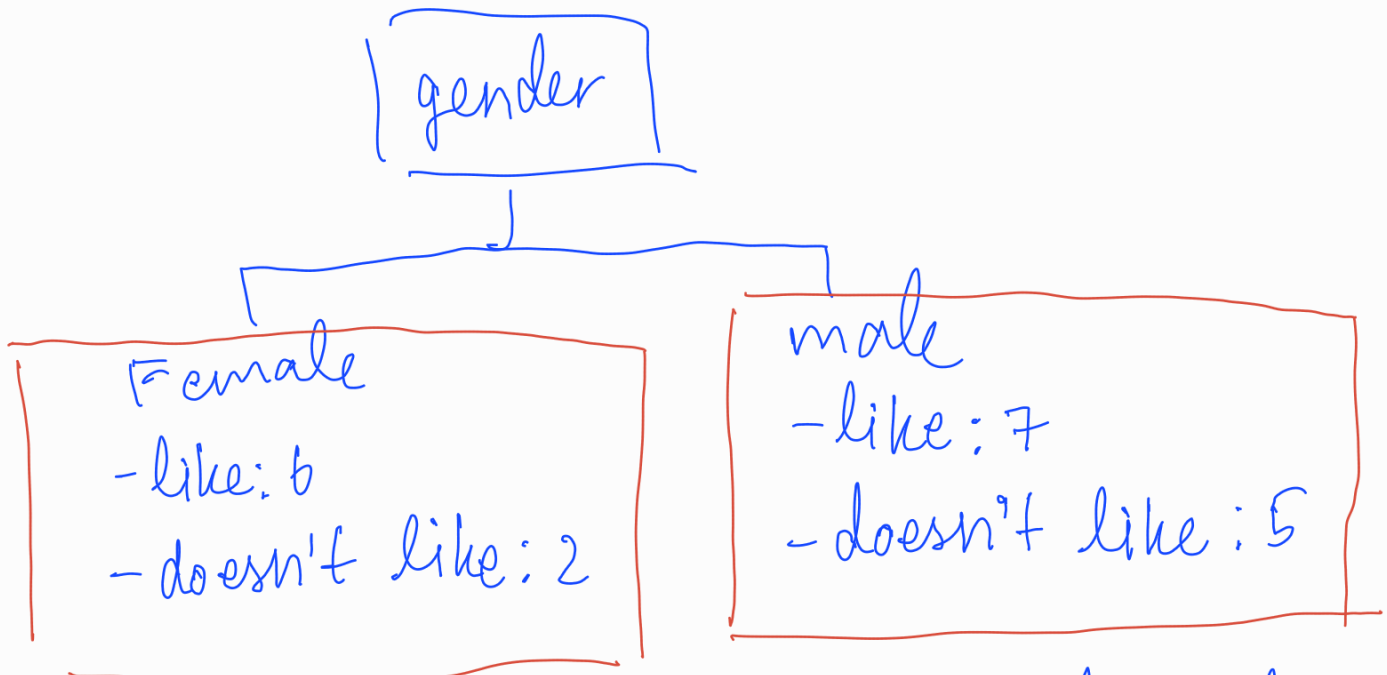


Decision tree . (binary)



create decision tree

- Predict who like Taylor Swift
 - attribute $\rightarrow$ gender, age, income



* We can see that Female and male node is n't impure

└ node can't split data purely (In female, it still have "like" & "doesn't like" group)

To calculate impurity of node, we used

"Gini Coefficient"

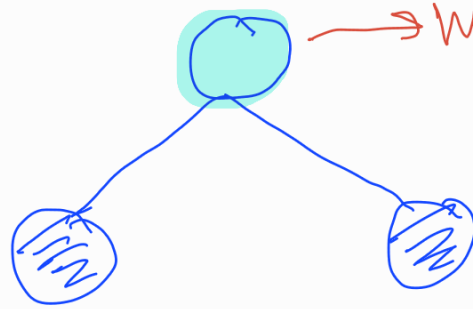
↳ low value, more pure

(0 = pure)

* weighted gini coefficient *

↳ compute gini coefficient for
parent node

⊗ → leave



→ weighted gini coefficient
(compute from gini
coeff of leaves)

* Decision tree can lead to "Overfitting"

* Decision tree can be used with
discrete & continuous value

To solve overfitting problem, they introduce

"CART algorithm"

↳ "pruning" help prevent overfit
(start from leave back to
parent node)

reduce the
tree size

Disadvantage of decision tree.

- for categorical data, it can be biased.
If we have more "like" data than "dislike", it leads to bias since DT is based on **majority vote**.
-

Random Forest

- ensemble of many decision trees
Aggregate the decision/result from decision trees and use majority vote/averaging to come up w/ one answer
- RF provide insight on importance of each attribute
- RF related to **bootstrapping & bagging**
(bootstrap aggregation)