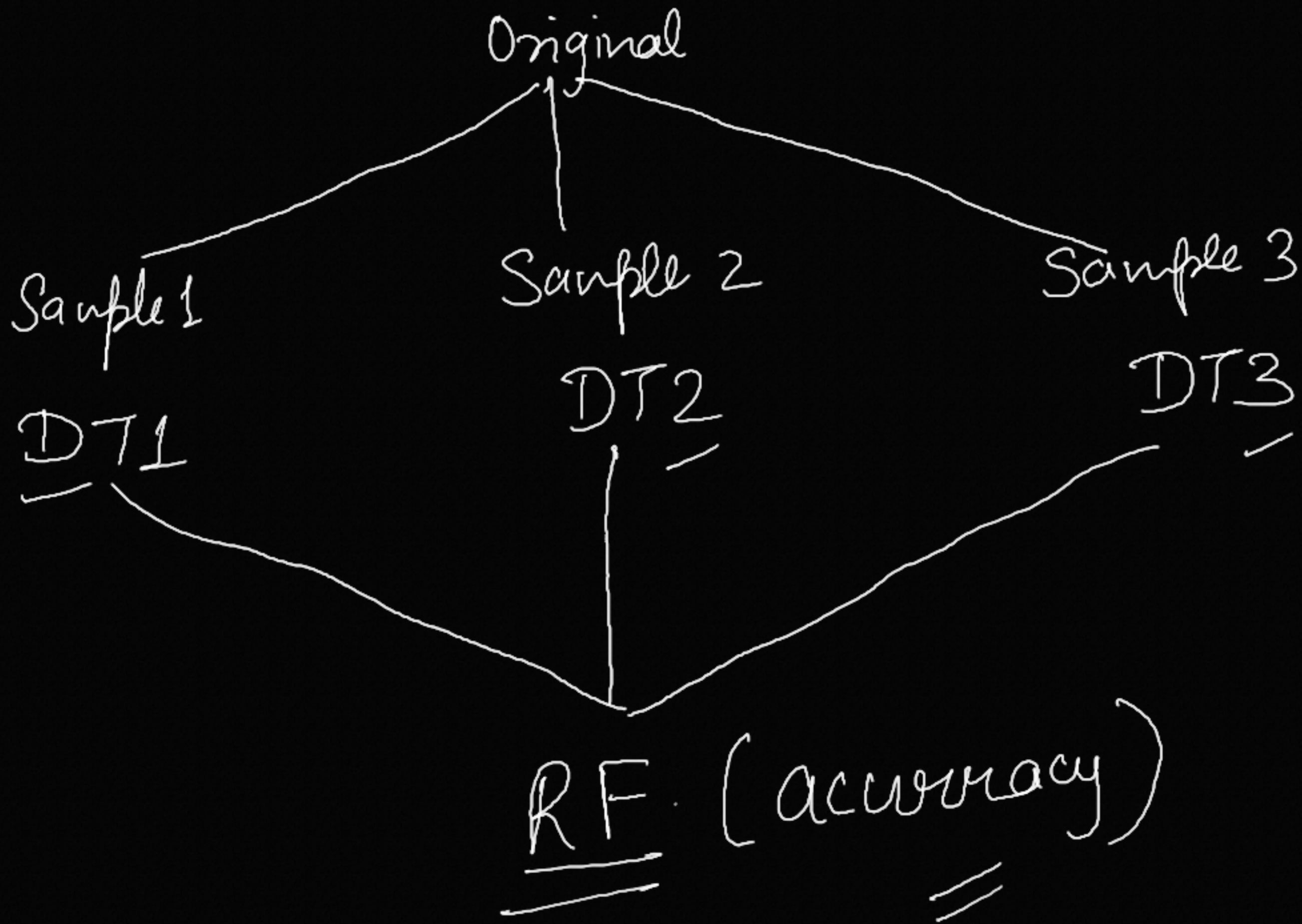
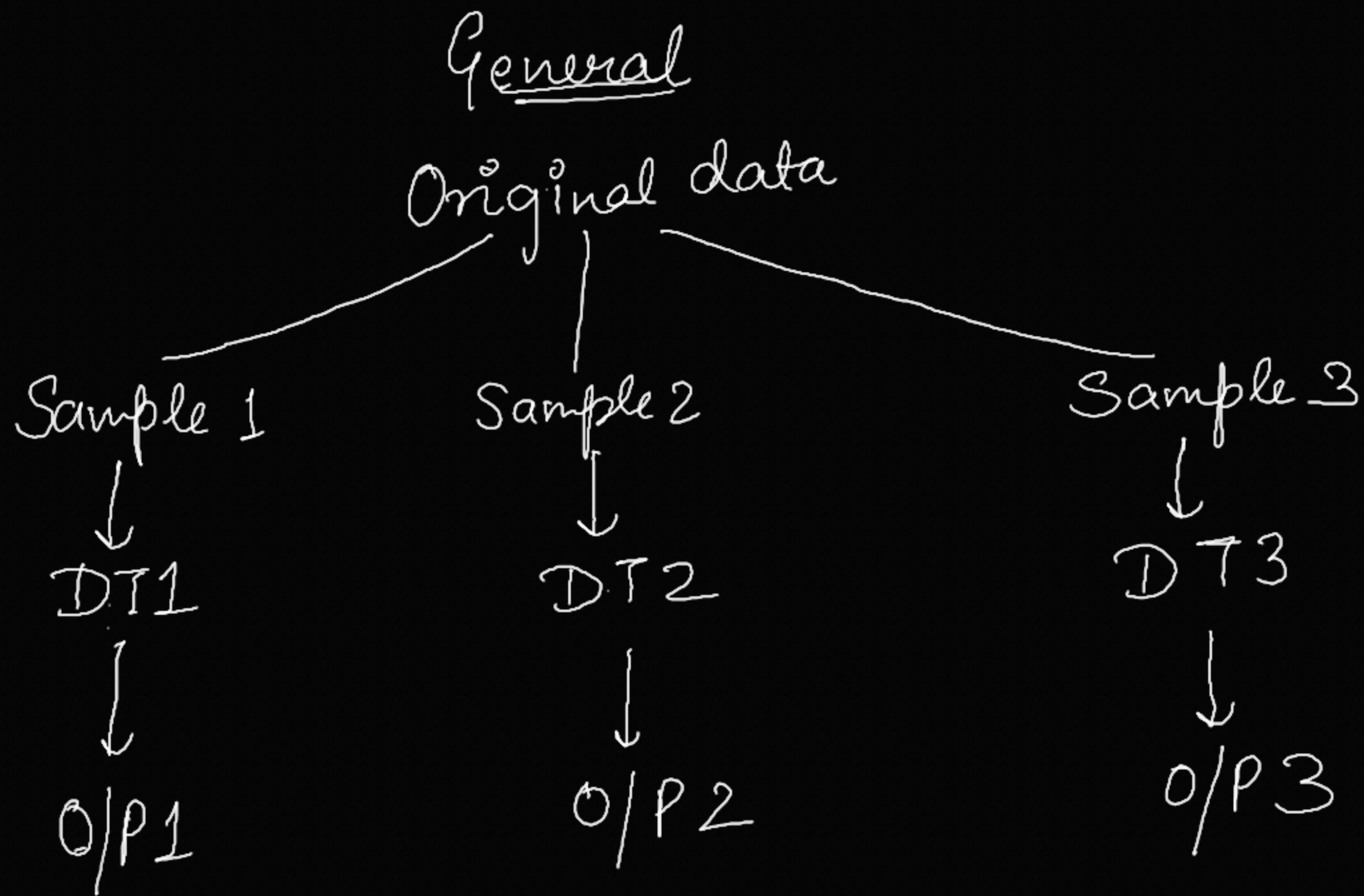


Random Forest

- * Regression and Classification
- * Introduced \rightarrow Leo Breiman & Adele Cutler
2001.
- * No assumptions
- * OOB error estimation \rightarrow Out of Bag
(Internal validation data)
($\frac{1}{3}$ rd of data)





Req → Mean , Classification → Voting

Mathematical

<u>feature 1</u>	<u>feature 2</u>	<u>Target</u>
1	2	0
2	3	10
3	4	1
4	5	1
5	6	1

① Create Samples/Subset

Subset 1

f1	f2	T
1	2	0
2	3	0
3	4	1
4	5	1

Subset 2

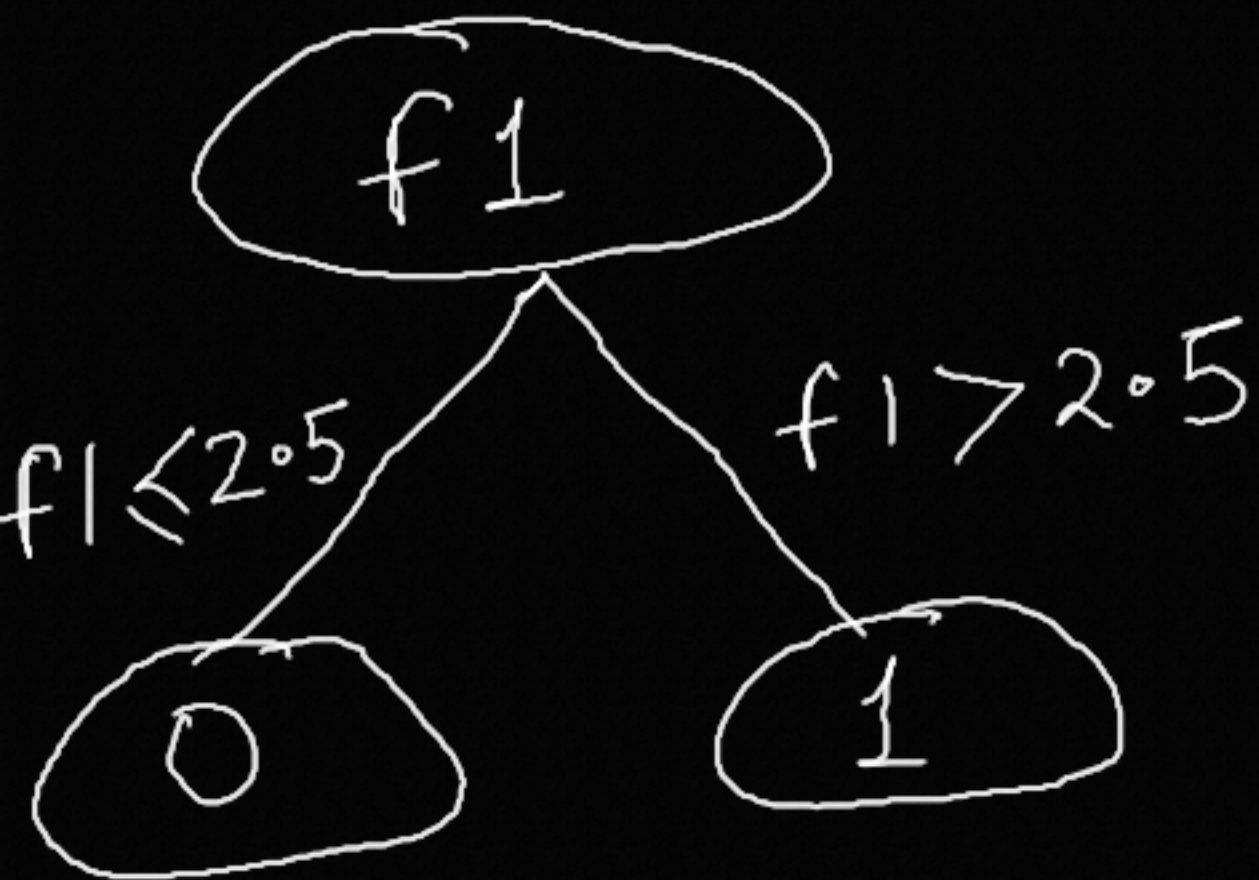
f1	f2	T
2	3	0
3	4	1
4	5	1
5	6	1

Subset 3

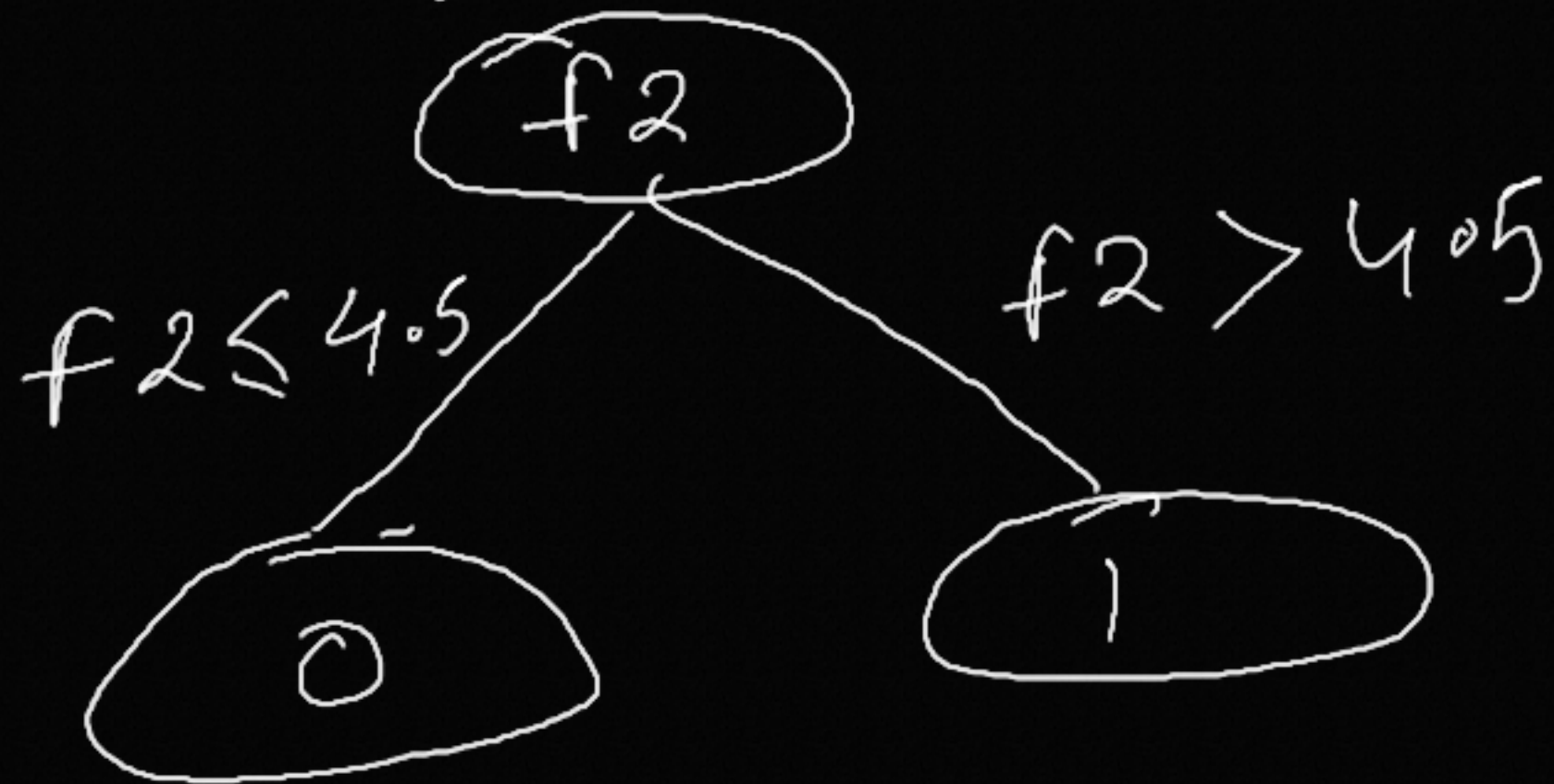
f1	f2	T
1	2	0
3	4	1
4	5	1
5	6	1

② Create DT for each sample

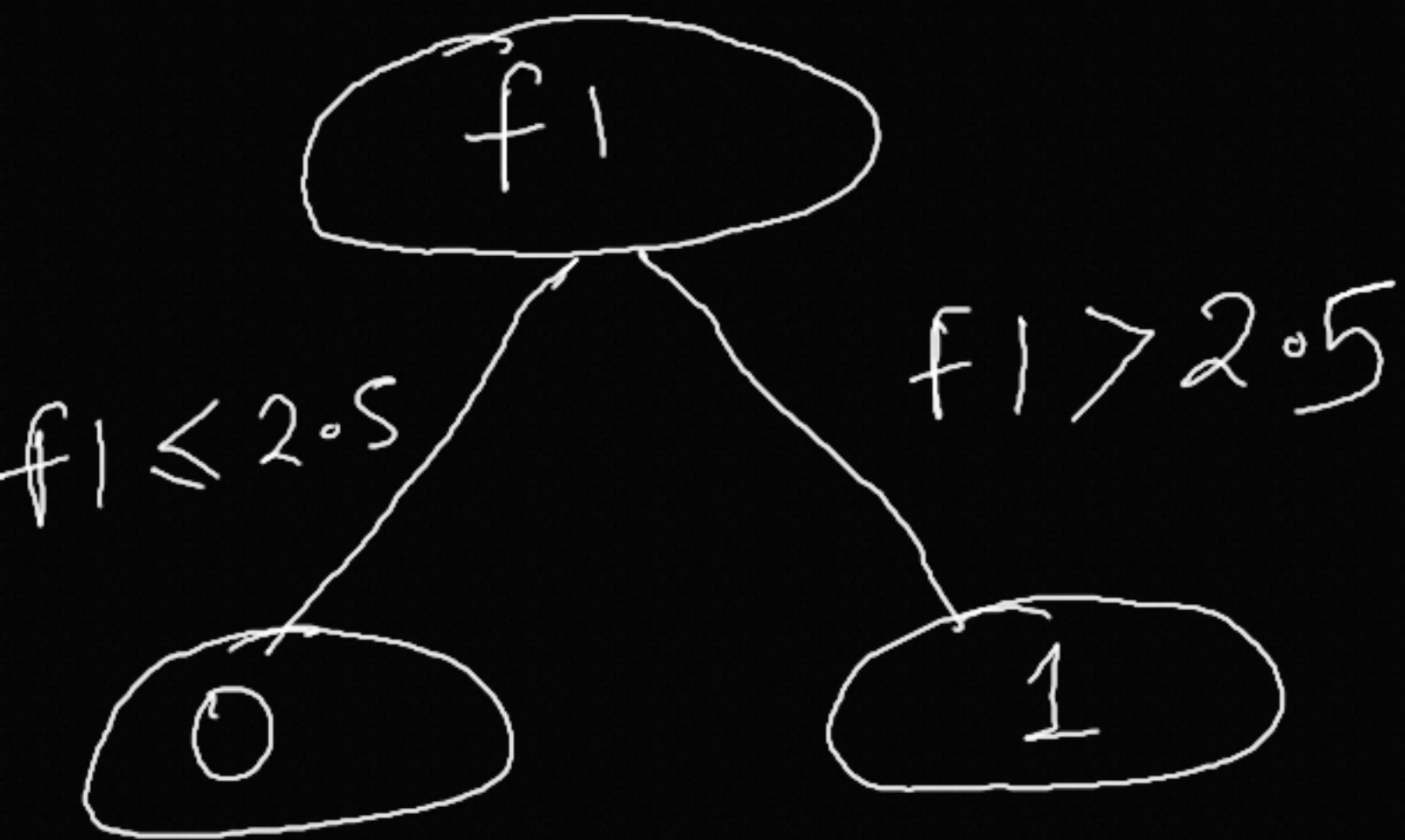
(a) DT1



(b) DT2



(c) DT3.



Model
Training
Completed

* Prediction

$$f_1 = 3, f_2 = 4$$

Apply on each DT

$$\text{DT 1} \rightarrow 1$$

$$\text{DT 2} \rightarrow 0$$

$$\text{DT 3} \rightarrow 1$$

Voting \rightarrow final = 1
o/p

When to Use

- * Complex & Non-linear data
- * Very large dataset
- * They can perform good even we have missing values
- * Prevent overfitting

When Not to Use

- * High-D data with less rows / data points
- * When majority features are categorical
- * Highly computational intensive
- * Large-scale problems

fact : Rf are also called "black-box model"