

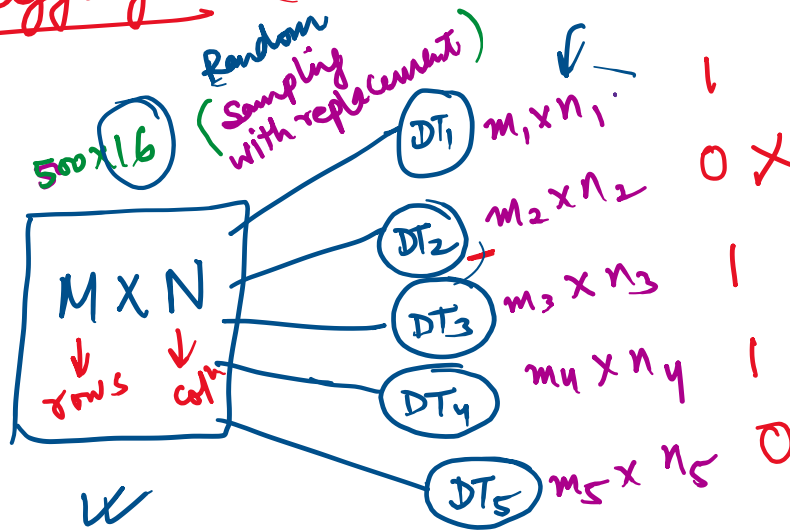
Ensemble Model.

DT. (Collection/Aggregation)

Bagging
(Bootstrap aggregation).
(Parallel)

Boosting
(Sequentially).

① Bagging. (Random Forest)



$$m_i \leq M$$
$$n_i \leq N$$

$$m_i \Rightarrow \left(\frac{2}{3}\right) M$$

$$n_i \Rightarrow \sqrt{N}$$

Random forest.

→ Forest is made of Trees & R.F. is made of D.T.

→ Ensemble model, Trained using Bagging technique.

→ Adv. - Combination of learning models.

DT: Searches for most imp. feature while splitting a node.

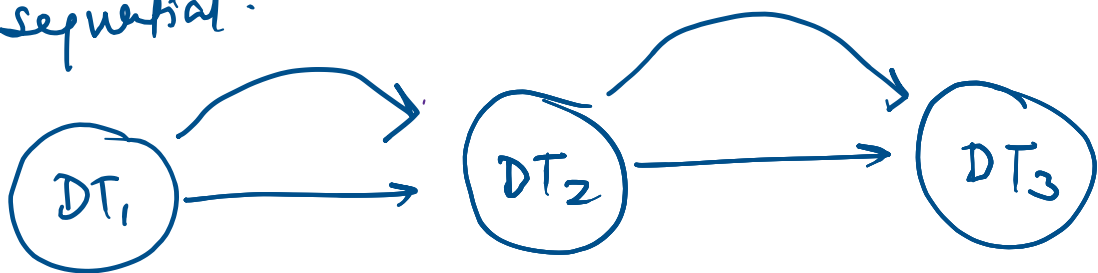
RF: Searches for most imp. feature among a random subset of feature.

⇓
(High diversity)

Most imp. features of DT is not included (by chance) in DT_3 , then also the split will happen. (based on best available feature).

② Boosting

→ sequential.



1000
→ 9
→ 7

Weak learners :

Strong learners :

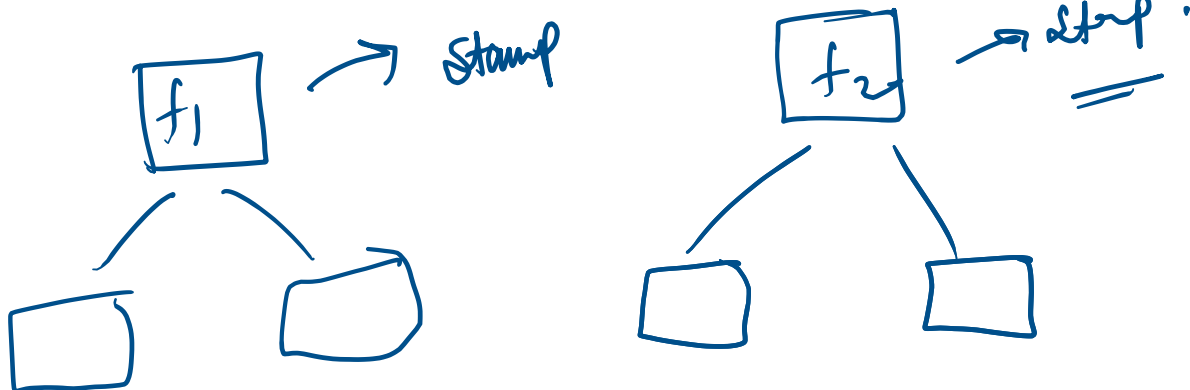
				y	y-hat
				0	0
				1	0
				1	1
				0	1
				0	0
				1	1

Step 1: Initialize equal weights.

	x_1	x_2	x_3	x_4	x_5	y	o/p	weights
1							yes	$\frac{1}{10}$
2							no	$\frac{1}{10}$
3							yes	$\frac{1}{10}$
4							no	$\frac{1}{10}$
5							yes	$\frac{1}{10}$
6							yes	$\frac{1}{10}$
7							no	$\frac{1}{10}$
8							yes	$\frac{1}{10}$
9							no	$\frac{1}{10}$
10							yes	$\frac{1}{10}$

$$\frac{1}{n} = \frac{1}{10}$$

Step 2: Create Base learners in sequential fashion.



Linear Entropy: High info gain.

$$e_t = \frac{9}{1} = \frac{1}{10}$$

= Sum of all the sample weights of that error row

Step 3:

Performance of Stamp.

$$= \frac{1}{2} \log_e \left(\frac{1 - e_t}{e_t} \right)$$

$$= \cancel{x} \leftarrow$$

Step 4:

↑ weights of incorrect classification
(weak learners)

↓ weight of correct classified
record (strong learners).

New sample weight (of incorrect classified _{now}) = old weight $\times e^{\text{Perf. dist.}}$

$$= \frac{1}{10} \times \cancel{e^x}$$

New sample wt. (strong learner) = old wt. $\times e^{-\text{Perf. str.}}$

$$= \frac{1}{10} \times e^{-x}$$

(Hyperparameter Tuning) in RF.

* Increasing the Predictive Power.

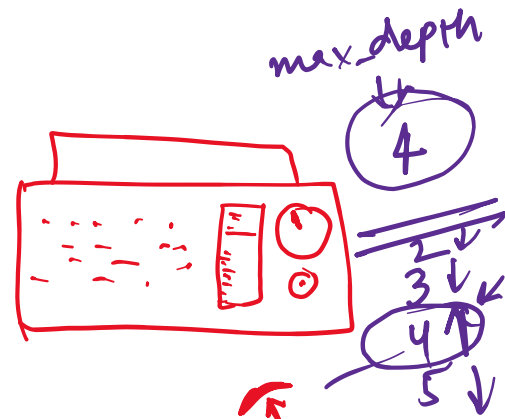
① n-estimators

→ More trees

✓ ↑ Perf.

✓ ↓ Computational Power

✓ ↓ Speed.



② max-features :

③ min-sample-split :

④ Max-depth.

② Model's Speed.

① n-jobs :

1

↓

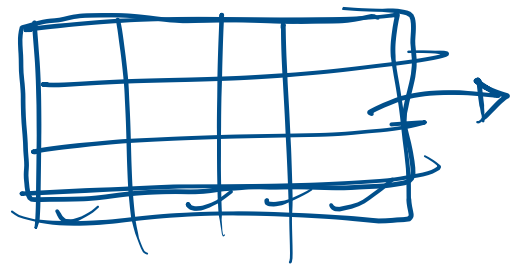
one Processor

-1

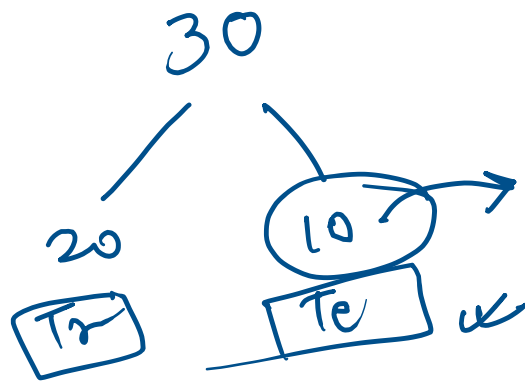
↓

no limit.

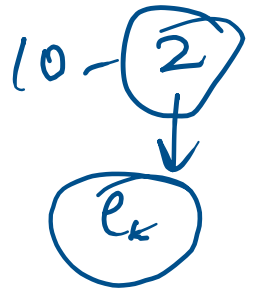
② oob_score
out of bag :-



60%
70%



oob-error-rate



Grid Search CV

- member of model-selection.
- Fits to the hyperparameters & fit the estimator on

↓

max_depth = (3, 4, 5, 8, 10)

[] = (10, 1, 15, 18)

[] = (3, 8, 9)

↓↓