* Hyperparameter Tuning *

Hyperparameter: - Params explicitly defined by user to control learning process.

not learned during training, but can be set before training begins.

La Used to improve model learning & values set before starting learning process of model.

La values cannot be changed during training process.

for ML models:

- 1) grid search cv
- 2 Randomized searchev

) Goidsearchev:-

to hyperparameters.

Lyperparam from grid of hyperparam values.

• Example: consider 2 parsams: count & vate.

count = [0.1, 0.2, 0.8, 0.4, 0.5]

rate = [0.1, 0.2, 0.8, 0.4]

Performance for all combinations in grid are:

	0.5	0.42	0.68	0.78	0.79
	0.4	0.40	0.67	0.825	0-81
	0.3	0.45	0.71	0.75	0.80
	0.2	0.35	0.72	0.62	0.61
count	0.1	0.42	0.59	0.61	0.78
		0.1	0.2	0.3	0.4
		rate			

Best performance = 0.825 (Highest).

Best combinal of param values:

count = 0.4rate = 0.3

• <u>Drawbacks</u> - go through all intermediate combinations of hyperparameter leads to computational expensive

2) Randomized Search CV:-

only fixed num. of phyperparameter setting.

find best set of hyperparams.

Reduce unnecessary computation.

· Implementation steps:

Ocreate hyperparameter gold-

param - dist = 3" max-depth": [3, None],

"max-features": randint (1,9),

"min-samples-leaf": randint (1,9),

"coiterion": ["gini", "entropy"]}

2 Instantiate Decision Tree Classifier -

sklearn. tree => Decision Tree Classifier tree = DTC()

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3 Instantiate Randomized Search CV object -

from sklear. model-selection import RSCV tree-cv = RSCV (tree, param_dist, cv = 5) tree-cv. fit (x, y)

4 Tuned Model -

best params = tree - cv. best - params -

best - model = tree_ev. best_estimator-

best - score = tree - cv. best - score-