# **Grid Search CV Api write up**

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### 1) Grid Search CV

Exhaustive search over specified parameter values for an estimator. The parameters of the estimator used to apply these methods are optimized by cross-validated grid-search over a parameter grid. Exhaustive search means it tries each and every possible combination and selects the best combination

### Code:-

sklearn.model\_selection.**GridSearchCV**(*estimator*, *param\_grid*, \*, *scoring=None*, n\_jo bs=None, refit=True, cv=None, verbose=0, pre\_dispatch='2\*n\_jobs', error\_score=nan, retur n\_train\_score=False)

# **Important Parameters are:-**

#### estimator

This is assumed to implement the scikit-learn estimator interface. Either estimator needs to provide a score function, or scoring must be passed. param\_grid Dictionary with parameters names ( str) as keys and lists of parameter settings to try as values, or a list of such dictionaries, in which case the grids spanned by each dictionary in the list are explored. This enables searching over any sequence of parameter settings.

#### scoring

Strategy to evaluate the performance of the cross-validated model on the test set.

CV

Determines the cross-validation splitting strategy.

# **Important Attributes are:-**

#### best\_estimator

Estimator that was chosen by the search, i.e. estimator which gave highest score (or smallest loss if specified) on the left out data. Not available if refit=False.

See refit parameter for more information on allowed values.

### best\_score

Mean cross-validated score of the best estimator

For multi-metric evaluation, this is present only if refit is specified.

This attribute is not available if refit is a function.

### best params

Parameter setting that gave the best results on the hold out data.

For multi-metric evaluation, this is present only if refit is specified.

## **Application:-**

# **Important Methods: -**

Fit(X, y)- fit the linear model.

Predict(X)-predict using linear model.

Score(X,y)-returns the coefficient of determination  $R^2$  of the prediction.