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# Grid Search CV Api write up

### **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**( $estimator, param\_grid, *, scoring=None, n\_j obs=None, refit=True, cv=None, verbose=0, pre_dispatch='2*n_jobs', error_score=nan, re turn_train_score=False)$ 

## **Important Parameters:**

#### 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.