

# Machine Learning Assignment: GridSearchCV

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```
class sklearn.model_selection.GridSearchCV (estimator, param_grid, *, scoring=None,
n_jobs=None, refit=True, cv=None, verbose=0, pre_dispatch='2*n_jobs', error_score=nan,
return_train_score=False)
```

## Parameters:

- estimator- to implement the scikit-learn estimator interface; either the estimator needs to provide a score function, or a form of scoring must be passed;
- param\_grid- a dictionary with parameters names (str) as keys and lists of parameter settings to try as values;
- scoring- strategy to evaluate the performance of the cross-validated model on the test set;
- n\_jobs- number of cores to use in parallel for the grid search;
- refit- refit an estimator using the best-found parameters on the whole dataset;
- cv- acronym, cross validation strategy;
- pre\_dispatch- controls the number of jobs that get dispatched during parallel execution;
- return\_train\_score- Whether to return the training scores or not.

## Attributes:

- cv\_results\_- a dictionary with keys as column headers and values as columns, that can be imported into a pandas dataframe;
- best\_estimator\_- an estimator that was chosen by the search, i.e., estimator which gave highest score (or smallest loss if specified) on the remaining data. Not available if refit=False;
- best\_score\_- mean cross-validated score of the best\_estimator;
- best\_params\_- Parameter setting that gave the best results on the hold out data;
- scorer\_- a scorer function used on the “held out” data to choose the best parameters for the model;
- refit\_time\_- records seconds used for refitting the best model on the whole dataset;
- multimetric\_- whether or not the scorers compute several metrics.

## Methods:

- |                                |   |
|--------------------------------|---|
| • <b>decision_function(X)</b>  | Call decision_function on the estimator with the best-found parameters. |
| • <b>fit(X[, y, groups])</b>   | Run fit with all sets of parameters.                                    |
| • <b>get_params([deep])</b>    | Get parameters for this estimator.                                      |
| • <b>inverse_transform(Xt)</b> | Call inverse_transform on the estimator with the best found params.     |
| • <b>predict(X)</b>            | Call predict on the estimator with the best-found parameters.           |

- **predict\_log\_proba(X)** Call predict\_log\_proba on the estimator with the best-found parameters.
- **predict\_proba(X)** Call predict\_proba on the estimator with the best-found parameters.
- **score(X[, y])** Returns the score on the given data, if the estimator has been refit.
- **score\_samples(X)** Call score\_samples on the estimator with the best-found parameters.
- **set\_params(\*\*params)** Set the parameters of this estimator.
- **transform(X)** Call transform on the estimator with the best-found parameters.

### How does GridSearchCV work?

- GridSearchCV is the process of performing hyperparameter tuning in order to determine the optimal values for a given model that comes in Scikit-learn's(or SK-learn) model\_selection package.
- This function loops through predefined hyperparameters and fits your estimator (model) to your training set so one can select the best parameters from the listed hyperparameters.
- Predefined values for hyperparameters are passed to the GridSearchCV function.
- This is done by defining a dictionary in which a particular hyperparameter along with the values it can take exists.
- GridSearchCV tries all the combinations of the values passed in the dictionary and evaluates the model for each combination using the method of Cross-Validation.
- Therefore after using this function, we get accuracy/loss for every combination of hyperparameters and we can choose the one with the best performance.