## What is GridSearchCV?

GridSearchCV is a function in scikit-learn (sklearn) that performs hyperparameter tuning for machine learning models using cross-validation. It searches for the best combination of hyperparameters that results in the best performance of the model.

## Key arguments of GridSearchCV

Here are the key arguments of GridSearchCV:

- estimator: The machine learning model to tune. (e.g., KNeighborsRegressor())
- param\_grid: A dictionary with hyperparameter names as keys and lists of possible values as values. (e.g., {'n neighbors': [3, 5, 7], 'weights': ['uniform', 'distance']})
- cv: The number of folds for cross-validation. Can be an integer, a KFold object, or a StratifiedKFold object. (e.g., 5 for 5-fold cross-validation)
- **scoring**: The evaluation metric. Can be a string (e.g., 'neg\_mean\_squared\_error', 'r2') or a callable function.
- n\_jobs: The number of jobs to run in parallel. If -1, all CPUs are used.

## Example usage of GridSearchCV with KNeighborsRegressor

Here's an example:

```
from sklearn.model selection import GridSearchCV
from sklearn.neighbors import KNeighborsRegressor
from sklearn.datasets import load boston
from sklearn.model_selection import train_test_split
# Load Boston housing dataset
boston = load boston()
X, y = boston.data, boston.target
# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random state=42)
# Create a K-Nearest Neighbors Regressor model
knn model = KNeighborsRegressor()
# Define the hyperparameter grid
param grid = {
   'n neighbors': [3, 5, 7],
    'weights': ['uniform', 'distance']
# Perform grid search with 5-fold cross-validation
grid search = GridSearchCV(knn model, param grid, cv=5,
scoring='neg_mean_squared_error', n_jobs=-1)
grid search.fit(X train, y train)
# Print the best parameters and the corresponding score
print("Best parameters:", grid_search.best_params_)
print("Best score:", grid search.best score )
```

In this example, we:

- Load the Boston housing dataset.
- Split the data into training and testing sets using train\_test\_split.
- Create a KNeighborsRegressor model.
- Define a hyperparameter grid with two hyperparameters: n\_neighbors and weights.
- Perform a grid search with 5-fold cross-validation ( cv=5 ) using the negative mean squared error ( scoring='neg\_mean\_squared\_error' ) as the evaluation metric.
- Run the grid search in parallel using all available CPUs (n jobs=-1).

## **Output**

```
Best parameters: {'n_neighbors': 5, 'weights': 'uniform'}
Best score: -10.5311
```

In this output, we see the best combination of hyperparameters and the corresponding score.

That's it! You've now used <code>GridSearchCV</code> to perform hyperparameter tuning for a <code>KNeighborsRegressor</code> model using cross-validation.