IPA 주관 인공지능센터 기본(fundamental) 과정

- GitHub link: here
- E-Mail: windkyle7@gmail.com

In [1]:

```
from sklearn.datasets import load iris
from sklearn.model selection import train test split
iris = load iris()
X train, X test, y train, y test =\
    train test split(iris.data, iris.target)
```

GridSearchCV 를 통해 LogisticRegression 모델의 최적의 Hyper parameter를 찾는다.

In [2]:

```
from sklearn.linear model import LogisticRegression
from sklearn.model selection import GridSearchCV
lr = LogisticRegression(solver='lbfgs', multi_class='auto', max_iter=1000)
param range = [0.001, 0.01, 0.1, 1.0, 10.0, 100.0, 1000.0]
param grid = {'C': param range}
params = {
    'estimator': lr,
    'param grid': param grid,
    'cv': 10,
    'iid': True,
    'return train score': True,
```

In [3]:

```
grid = GridSearchCV(**params)
grid.fit(X train, y train)
Out[3]:
GridSearchCV(cv=10, error score='raise-deprecating',
             estimator=LogisticRegression(C=1.0, class weight=None,
dual=False,
                                           fit intercept=True,
                                           intercept scaling=1, l1 ratio=None
                                           max_iter=1000, multi_class='auto',
                                           n jobs=None, penalty='12',
                                           random state=None, solver='lbfgs',
                                           tol=0.0001, verbose=0,
                                           warm start=False),
             iid=True, n jobs=None,
             param grid={'C': [0.001, 0.01, 0.1, 1.0, 10.0, 100.0,
1000.0]},
             pre dispatch='2*n jobs', refit=True, return train score=True,
```

1000 b

scoring=None, verbose=0)

최종 결과를 통해 최적의 Hyper parameter와 최적의 모델을 얻을 수 있다.

```
In [4]:
```

```
import pandas as pd
pd.DataFrame(grid.cv_results_)
```

Out[4]:

	mean_fit_time	std_fit_time	mean_score_time	std_score_time	param_C	params	split0_test_score	٤
0	0.003153	0.000772	0.000174	0.000052	0.001	{'C': 0.001}	0.750000	
1	0.005154	0.000359	0.000193	0.000072	0.01	{'C': 0.01}	0.916667	
2	0.009063	0.000708	0.000146	0.000018	0.1	{'C': 0.1}	1.000000	
3	0.013962	0.001782	0.000140	0.000003	1	{'C': 1.0}	1.000000	
4	0.017733	0.003538	0.000135	0.000007	10	{'C': 10.0}	1.000000	
5	0.024809	0.002786	0.000141	0.000016	100	{'C': 100.0}	1.000000	
6	0.037531	0.008949	0.000144	0.000026	1000	{'C': 1000.0}	1.000000	

7 rows × 31 columns

searchgrid

- Helps building parameter grids for scikit-learn grid search.
- https://searchgrid.readthedocs.io/en/latest/

In [5]:

```
from sklearn.base import BaseEstimator

class Dummy(BaseEstimator):
    def fit(self):
        pass

def score(self):
        pass
```

In [6]:

```
from sklearn.neighbors import KNeighborsClassifier
from sklearn.pipeline import Pipeline

knn = KNeighborsClassifier()
pipe = Pipeline([('clf', Dummy())])
```

```
param grid = [
    {
        'clf': [lr]
    },
    {
        'clf': [knn],
        'clf n neighbors': [3, 4, 5, 6, 7]
    },
]
params['estimator'] = pipe
params['param_grid'] = param_grid
In [7]:
grid = GridSearchCV(**params)
grid.fit(X_train, y_train)
Out[7]:
GridSearchCV(cv=10, error score='raise-deprecating',
             estimator=Pipeline(memory=None, steps=[('clf', Dummy())],
                                 verbose=False),
             iid=True, n_jobs=None,
             param grid=[{'clf': [LogisticRegression(C=1.0,
class weight=None,
                                                       dual=False,
                                                       fit intercept=True,
                                                       intercept scaling=1,
                                                       11 ratio=None,
                                                       max iter=1000,
                                                       multi class='auto',
                                                       n jobs=None,
penalty='12',
                                                       random state=None,
                                                       solver='lbfgs',
tol=0.0001,
                                                       verbose=0,
                                                       warm start=False)]},
                          {'clf': [KNeighborsClassifier(algorithm='auto',
                                                         leaf size=30,
                                                         metric='minkowski',
                                                         metric params=None,
                                                         n jobs=None,
                                                         n neighbors=3, p=2,
                                                        weights='uniform')],
                           'clf__n_neighbors': [3, 4, 5, 6, 7]}],
             pre dispatch='2*n jobs', refit=True, return train score=True,
             scoring=None, verbose=0)
In [8]:
pd.DataFrame(grid.cv results)
Out[8]:
```

					<u> </u>
0	0.013966	0.002585	0.000147	0.000025	LogisticRegression(C=1.0, class_weight=None, d
1	0.000286	0.000050	0.000573	0.000044	KNeighborsClassifier(algorithm='auto',

param_clf

mean_fit_time std_fit_time mean_score_time std_score_time

- me	ean_fit_time	std_fit_time	mean_score_time	std_score_time	leaf_si param_clf
2	0.000270	0.000003	0.000558	0.000011	KNeighborsClassifier(algorithm='auto', leaf_si
3	0.000274	0.000007	0.000562	0.000013	KNeighborsClassifier(algorithm='auto', leaf_si
4	0.000274	0.000012	0.000580	0.000026	KNeighborsClassifier(algorithm='auto', leaf_si
5	0.000272	0.000004	0.000571	0.000013	KNeighborsClassifier(algorithm='auto',