```
In [1]:
import warnings
warnings.filterwarnings('ignore')
```

# Random Forest & Ansemble(Clone Coding Iris data)

```
In [2]:
```

```
import pandas as pd
from sklearn import datasets
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn import metrics
from IPython.display import Image
```

## In [3]:

```
아이리스 종류: ['setosa' 'versicolor' 'virginica']
target: [0:setosa, 1:versicolor, 2:virginica]
데어터 수: 150
데이터 열 이름: ['sepal length (cm)', 'sepal width (cm)', 'petal len
gth (cm)', 'petal width (cm)']
```

# Out[3]:

#### sepal length sepal width petal length petal width species 1 4.9 3.0 1.4 0.2 0 2 4.7 3.2 1.3 0.2 1.5 3 46 3 1 0.2 0 0

### In [4]:

```
x = data[['sepal length', 'sepal width', 'petal length', 'petal width']]
y = data['species']

# 테스트 데이터 30%
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.3)
print(len(x_train))
print(len(x_test))
print(len(y_train))
print(len(y_test))
```

```
105
45
105
45
In [5]:
forest = RandomForestClassifier(n estimators=100)
forest.fit(x_train, y_train)
Out[5]:
RandomForestClassifier(bootstrap=True, ccp alpha=0.0, class weight=
                      criterion='gini', max depth=None, max featu
res='auto',
                      max leaf nodes=None, max samples=None,
                      min impurity decrease=0.0, min impurity spl
it=None,
                      min samples leaf=1, min samples split=2,
                      min weight_fraction_leaf=0.0, n_estimators=
100,
                      n jobs=None, oob score=False, random state=
None,
                      verbose=0, warm start=False)
In [6]:
# 예측
y_pred = forest.predict(x test)
print(y pred)
print(list(y test))
1 1 0 0 1 2 1 0]
[2, 1, 2, 2, 1, 2, 2, 1, 0, 1, 2, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 1,
1, 2, 2, 2, 0, 2, 0, 0, 0, 1, 0, 2, 1, 2, 1, 1, 1, 0, 0, 1, 2, 1, 0
In [7]:
# 정확도 확인
print('정확도 :', metrics.accuracy_score(y_test, y_pred))
정확도 : 0.911111111111111
Random Forest & Ansemble(Regression)
<u>참고사항</u>
Train Data
In [8]:
train = pd.read_csv("train.csv")
```

In [9]:

```
train
Out[9]:
     battery_power blue clock_speed dual_sim fc four_g int_memory m_dep mobile_wt n_cores ... px_hei
0
           842
                      2.2
                             0 1
                                     0
                                                       188
                                                              2 ...
                                                 0.6
   1
          1021
                      0.5
                             1 0
                                     1
                                            53
                                                 0.7
                                                       136
                                                              3 ...
 2
                                                 0.9
           563
                      0.5
                              1 2
                                                       145
                                                              5 ...
           615
               1
                      2.5
                             0 0
                                            10
                                                 0.8
                                                       131
  4
          1821
                      1.2
                             0 13
                                                 0.6
                                                       141
 1995
           794
                      0.5
 1996
                                                       187
                                                              4 ...
          1965
                      2.6
                                            39
                                                 0.2
               1
                             1 0
 1997
          1911
                      0.9
                                                 0.7
                                                       108
                                                              5 ...
 1998
          1512
                      0.9
                             0 4
                                     1
                                                       145
                                            46
                                                 0.1
                                                 0.9
                                                              6 ...
                      2.0
                                                       168
2000 rows × 21 columns
                                                                    F
In [10]:
train.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2000 entries, 0 to 1999
Data columns (total 21 columns):
             Non-Null Count Dtype
 # Column
                    -----
    battery_power 2000 non-null int64
 0
 1 blue 2000 non-null int64
 2 clock_speed
                  2000 non-null float64
 3 dual_sim 2000 non-null int64
 4 fc
5 four_g
                    2000 non-null int64
                  2000 non-null int64
 6 int_memory
                   2000 non-null int64
 7 m dep
                   2000 non-null float64
 8 mobile wt
                   2000 non-null int64
 9 n cores
                    2000 non-null int64
 10 pc
                    2000 non-null int64
 11 px height
                   2000 non-null int64
                   2000 non-null int64
 12 px width
 13 ram
                   2000 non-null int64
 14 sc h
                   2000 non-null int64
 15 sc_w
                   2000 non-null
                                   int.64
 16 talk_time
                   2000 non-null
                                   int.64
 17 three_g
                                   int64
                   2000 non-null
 18 touch screen 2000 non-null
                                   int64
 19 wifi 2000 non-null 20 price_range 2000 non-null
                                    int64
                                   int64
dtypes: float64(2), int64(19)
memory usage: 328.2 KB
In [11]:
train.columns
Out[11]:
Index(['battery power', 'blue', 'clock speed', 'dual sim', 'fc', 'f
       'int memory', 'm dep', 'mobile wt', 'n cores', 'pc', 'px hei
```

aht',

```
'px width', 'ram', 'sc h', 'sc w', 'talk time', 'three g',
         'touch screen', 'wifi', 'price range'],
        dtype='object')
In [12]:
x train = train.drop("price range", axis = 1)
x_train
Out[12]:
      battery_power blue clock_speed dual_sim fc four_g int_memory m_dep mobile_wt n_cores pc px_he
                                                                            2 2
   0
             842
                   0
                           2.2
                                    0
                                       1
                                             0
                                                            0.6
                                                                    188
                           0.5
            1021
                                                      53
                                                            0.7
                                                                    136
                                                                            3
   2
             563
                           0.5
                                    1 2
                                                      41
                                                            0.9
                                                                    145
                                                                            5 6
   3
             615
                   1
                           2.5
                                    0 0
                                                      10
                                                            0.8
                                                                    131
                                                                            6 9
             1821
                           1.2
                                    0 13
                                                      44
                                                            0.6
                                                                    141
                                                                            2 14
 1995
             794
                           0.5
                                    1 0
                                                       2
                                                            0.8
                                                                    106
                                                                            6 14
 1996
            1965
                   1
                           2.6
                                    1 0
                                             0
                                                      39
                                                            0.2
                                                                    187
                                                                            4 3
 1997
             1911
                           0.9
                                                      36
                                                            0.7
                                                                    108
                                                                            8 3
 1998
                   0
                           0.9
                                    0 4
                                                                            5 5
            1512
                                             1
                                                      46
                                                            0.1
                                                                    145
             510
                           2.0
                                                            0.9
                                                                            6 16
                                                      45
                                                                    168
2000 \text{ rows} \times 20 \text{ columns}
4
In [13]:
y train = train["price_range"]
for i in range(len(y_train)):
    if y train[i] == 0:
         y_train[i] = "Low Cost"
     elif y_train[i] == 1:
        y_train[i] = "Medium Cost"
     else:
        y_train[i] = "Very High Cost"
y_train
Out[13]:
0
             Medium Cost
          Very High Cost
1
2
          Very High Cost
3
          Very High Cost
              Medium Cost
                . . .
1995
                 Low Cost
1996
          Very High Cost
1997
          Very High Cost
1998
                 Low Cost
1999
          Very High Cost
Name: price range, Length: 2000, dtype: object
Test Data
In [14]:
test = pd.read_csv("train.csv")
In [15]:
```

```
|x test = test.drop("price range", axis = 1)
x test
Out[15]:
      battery_power blue clock_speed dual_sim fc four_g int_memory m_dep mobile_wt n_cores pc px_he
   0
                           2.2
                                   0
                                      1
                                                                          2
             842
                                                          0.6
                                                                  188
            1021
                           0.5
                                   1
                                                     53
                                                          0.7
                                                                  136
                                                                          3
                                                                             6
   2
             563
                           0.5
                                   1
                                                     41
                                                          0.9
                                                                  145
                                                                          5 6
   3
             615
                           2.5
                                   0 0
                                            0
                                                     10
                                                          0.8
                                                                  131
                                                                          6 9
                   1
   4
            1821
                           1.2
                                   0 13
                                                     44
                                                          0.6
                                                                  141
                                                                          2 14
 1995
             794
                           0.5
                                                          0.8
                                                                  106
                                            0
 1996
            1965
                   1
                           2.6
                                   1 0
                                                     39
                                                                  187
                                                                          4 3
                                                          0.2
 1997
                                                                          8 3
            1911
                                                          0.7
                                                                  108
 1998
            1512
                   0
                           0.9
                                   0 4
                                            1
                                                                          5 5
                                                     46
                                                          0.1
                                                                  145
 1999
             510
                           2.0
                                                          0.9
                                                                  168
                                                                          6 16
2000 \text{ rows} \times 20 \text{ columns}
4
In [16]:
y_test = test["price_range"]
for i in range(len(y_test)):
    if y_test[i] == 0:
        y_test[i] = "Low Cost"
     elif y_test[i] == 1:
       y_test[i] = "Medium Cost"
     else:
        y_test[i] = "Very High Cost"
y_test
Out[16]:
             Medium Cost
0
1
          Very High Cost
          Very High Cost
3
          Very High Cost
             Medium Cost
1995
                Low Cost
          Very High Cost
1996
1997
          Very High Cost
1998
                 Low Cost
1999
       Very High Cost
Name: price range, Length: 2000, dtype: object
GridSearch CV를 통한 랜덤포레스트 하이퍼파라미터
튜닝
In [17]:
from sklearn.ensemble import RandomForestClassifier
from sklearn.model_selection import GridSearchCV
```

```
from sklearn.ensemble import RandomForestClassifier
from sklearn.model_selection import GridSearchCV
from sklearn.metrics import accuracy_score
In [18]:
```

```
rf clf = RandomForestClassifier(random state = 0, n jobs = -1)
grid cv = GridSearchCV(rf clf, param grid = params, cv = 3, n jobs = -1)
grid_cv.fit(x_train, y_train)
print('최적 하이퍼 파라미터: ', grid_cv.best_params_)
print('최고 예측 정확도: {:.4f}'.format(grid_cv.best_score_))
최적 하이퍼 파라미터: {'max_depth': 12, 'min_samples_leaf': 8, 'min_s
amples_split': 8, 'n_estimators': 100}
최고 예측 정확도: 0.9100
In [20]:
best param clf = RandomForestClassifier(n estimators = 100,
                               max_depth = 10,
                               min samples leaf = 8,
                               min samples split = 20,
                               n_{jobs} = -1)
In [21]:
best_param_clf.fit(x_train, y_train)
prediction = best_param_clf.predict(x_test)
print("예측 정확도: {:.4f}".format(accuracy score(y test, prediction)))
예측 정확도: 0.9660
Random Forest의 각 특성 중요도 시각화
In [22]:
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
In [23]:
ftr_importances_values = best_param_clf.feature_importances_
ftr_importances = pd.Series(ftr_importances_values, index = x_train.columns)
ftr top20 = ftr importances.sort values(ascending=False)[:20]
plt.figure(figsize=(8,6))
plt.title('Top 20 Feature Importances')
sns.barplot(x=ftr_top20, y=ftr_top20.index)
plt.show()
                          Top 20 Feature Importances
       ram
 battery_power
   px_height
    px_width
   mobile_wt
  int memory
  clock_speed
```

pc
talk\_time
sc\_w
m\_dep
sc\_h
fc
n\_cores
touch\_screen
blue
wifi

```
dual_sim
     four a
     three_g
                                   04
                                                06
In [24]:
from sklearn.tree import export graphviz
In [25]:
model = RandomForestClassifier()
model.fit(x train, y train)
estimator = model.estimators [3]
export_graphviz(estimator, out_file='tree.dot',
               feature names = x train.columns,
               class_names = y_train.unique(),
              max depth = 3, # 표현하고 싶은 최대 depth
              precision = 3, # 소수점 표기 자릿수
              filled = True, # class별 color 채우기
              rounded=True, # 박스의 모양을 둥글게
In [33]:
from subprocess import call
call(['dot', '-Tpng', 'tree.dot','-o', 'tree.png', '-Gdpi=600'])
FileNotFoundError
                                            Traceback (most recent c
all last)
<ipython-input-33-0a11d1de2db7> in <module>
      1 from subprocess import call
----> 2 call(['dot', '-Tpng', 'tree.dot','-o', 'tree.png', '-Gdpi=6
00'])
~\Anaconda3\lib\subprocess.py in call(timeout, *popenargs, **kwargs
    337
             retcode = call(["ls", "-1"])
    338
--> 339
             with Popen(*popenargs, **kwargs) as p:
    340
    341
                     return p.wait(timeout=timeout)
~\Anaconda3\lib\subprocess.py in init (self, args, bufsize, exec
utable, stdin, stdout, stderr, preexec fn, close fds, shell, cwd, e
nv, universal newlines, startupinfo, creationflags, restore signals
, start new session, pass fds, encoding, errors, text)
    798
                                          c2pread, c2pwrite,
    799
                                          errread, errwrite,
--> 800
                                          restore signals, start new
session)
    801
                 except:
    802
                     # Cleanup if the child failed starting.
~\Anaconda3\lib\subprocess.py in execute child(self, args, executa
ble, preexec_fn, close_fds, pass_fds, cwd, env, startupinfo, creati
onflags, shell, p2cread, p2cwrite, c2pread, c2pwrite, errread, errw
rite, unused restore signals, unused start new session)
   1205
   1206
                                                    os.fspath(cwd) i
f cwd is not None else None,
-> 1207
                                                    startupinfo)
   1208
                     finally:
   1200
                         # obild is lamaked offer the secretic as
```

```
1209
                        # Unita is launched. Close the parent's co
py of those pipe
FileNotFoundError: [WinError 2] 지정된 파일을 찾을 수 없습니다
In [34]:
# Display in jupyter notebook
from IPython.display import Image
In [35]:
Image(filename = 'tree.png')
_____
                                          Traceback (most recent c
FileNotFoundError
all last)
<ipython-input-35-04b3b31865e2> in <module>
---> 1 Image (filename = 'tree.png')
~\Anaconda3\lib\site-packages\IPython\core\display.py in init (s
elf, data, url, filename, format, embed, width, height, retina, unc
onfined, metadata)
   1202
               self.unconfined = unconfined
   1203
               super(Image, self). init (data=data, url=url, fil
ename=filename,
-> 1204
                        metadata=metadata)
   1205
   1206
                if self.width is None and self.metadata.get('width'
, {}):
~\Anaconda3\lib\site-packages\IPython\core\display.py in init (s
elf, data, url, filename, metadata)
    625
                   self.metadata = {}
    626
--> 627
               self.reload()
    628
               self. check data()
    629
~\Anaconda3\lib\site-packages\IPython\core\display.py in reload(sel
   1233
                """Reload the raw data from file or URL."""
   1234
                if self.embed:
-> 1235
                    super(Image, self) .reload()
   1236
                    if self.retina:
   1237
                        self. retina shape()
~\Anaconda3\lib\site-packages\IPython\core\display.py in reload(sel
                """Reload the raw data from file or URL."""
    650
    651
                if self.filename is not None:
--> 652
                   with open (self.filename, self. read flags) as
f:
    653
                        self.data = f.read()
                elif self.url is not None:
    654
FileNotFoundError: [Errno 2] No such file or directory: 'tree.png'
```