ORCA 4500 - Foundations of Data Science (Homework 5)

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In [1]:
import pandas as pd
import numpy as np
from sklearn.pipeline import Pipeline
from sklearn.model selection import GridSearchCV, train test split
from sklearn.preprocessing import StandardScaler, PolynomialFeatures, OneHotEncoder
from sklearn.compose import ColumnTransformer
from sklearn.linear model import Ridge
from sklearn.ensemble import RandomForestRegressor
In [2]:
df = pd.read_csv('bb_agg.csv').drop(['Unnamed: 0'],axis=1)
In [3]:
df
Out[3]:
    yearID teamID H_bat 2B_bat 3B_bat HR_bat BB_bat SB_bat CS_bat GIDP_bat HR_ptch BB_ptch SO_ptch H_ptcl
      1997
                   1531
                           279
                                          161
                                                       126.0
                                                               72.0
                                                                       129.0
                                                                                 202
                                                                                         605
                                                                                                1050
                                                                                                       150
             ANA
                                   25
                                                 617
      1997
                   1490
                           268
                                          174
                                                 597
                                                       108.0
                                                               58.0
                                                                       143.0
                                                                                         450
  1
              ATL
                                   37
                                                                                 111
                                                                                                1196
                                                                                                       131
  2
      1997
              BAL
                    1498
                           264
                                   22
                                          196
                                                 586
                                                        63.0
                                                               26.0
                                                                       121.0
                                                                                 164
                                                                                         563
                                                                                                1139
                                                                                                       140
  3
      1997
             BOS
                   1684
                           373
                                   32
                                          185
                                                 514
                                                        68.0
                                                               48.0
                                                                       155.0
                                                                                 149
                                                                                         611
                                                                                                 987
                                                                                                       156
  4
      1997
             CHA
                   1498
                           260
                                   28
                                          158
                                                 569
                                                       106.0
                                                               52.0
                                                                       133.0
                                                                                 175
                                                                                         575
                                                                                                 961
                                                                                                       150
653
      2018
              SLN
                   1369
                           248
                                          205
                                                 525
                                                        63.0
                                                               32.0
                                                                        92.0
                                                                                 144
                                                                                         593
                                                                                                1337
                                                                                                       135
      2018
              TBA
                   1415
                           274
                                          150
                                                 540
                                                       128.0
                                                               51.0
                                                                       122.0
                                                                                 164
                                                                                         501
                                                                                                1421
                                                                                                       123
654
                                   43
                   1308
                                                       74.0
                                                               35.0
                                                                                 222
                                                                                         491
655
      2018
              TEX
                           266
                                   24
                                          194
                                                 555
                                                                       104.0
                                                                                                1121
                                                                                                       151
                   1336
      2018
             TOR
                           320
                                          217
                                                 499
                                                        47.0
                                                               30.0
                                                                       118.0
                                                                                 208
                                                                                         551
                                                                                                1298
                                                                                                       147
656
                                   16
      2018
             WAS
                   1402
                           284
                                   25
                                          191
                                                 631
                                                       119.0
                                                               33.0
                                                                       104.0
                                                                                 198
                                                                                         487
                                                                                                1417
                                                                                                       132
657
658 rows x 17 columns
In [4]:
y = df['W']
X = df.drop(['W', '3B bat', 'BB bat', 'G', 'teamID'], axis=1)
In [5]:
cat var = ['yearID','lgID']
cont var = X.columns.difference(cat_var)
# Is this the right place to split the data ?
X train, X test, y train, y test = train test split(X, y, test size=0.2, random state=42)
In [6]:
pipe cat 1 = Pipeline([('onehot',OneHotEncoder(handle unknown = 'ignore',sparse=False)),
                         ('poly', PolynomialFeatures (degree = 2, interaction only=True, include
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pipe cont 1 = Pipeline([('poly', PolynomialFeatures(degree=2, interaction only=True, includ

bias=False))])

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e bias=False)),
                       ('scaler', StandardScaler())])
cf 1 = ColumnTransformer([('cat',pipe_cat_1,cat_var),
                          ('cont',pipe_cont_1,cont_var)])
main pipe ridge 1 = Pipeline([('col transformer', cf 1),
                        ('ridge', Ridge(fit intercept=True))])
In [7]:
params ridge 1 = {'ridge alpha':np.logspace(-1,2,20)}
ridge cv 1 = GridSearchCV(main pipe ridge 1,params ridge 1,cv=5)
In [8]:
ridge_cv_1.fit(X_train,y_train)
Out[8]:
GridSearchCV(cv=5,
              estimator=Pipeline(steps=[('col transformer',
                                           ColumnTransformer(transformers=[('cat',
                                                                                Pipeline (steps
=[('onehot',
OneHotEncoder(handle unknown='ignore',
sparse=False)),
('poly',
PolynomialFeatures (include bias=False,
interaction only=True))]),
                                                                                ['yearID',
                                                                                'lgID']),
                                                                               ('cont',
                                                                               Pipeline (steps
=[('poly',
PolynomialFeatures(include bias=False,
interact...
                                                                               Index(['2B bat
', 'BB_ptch', 'CS_bat', 'GIDP_bat', 'HR_bat', 'HR_ptch', 'H_bat',
       'H_ptch', 'SB_bat', 'SO_ptch'],
      dtype='object'))])),
                                           ('ridge', Ridge())]),
              param grid={'ridge alpha': array([ 0.1
                                                                     0.14384499, 0.20691381
    0.29763514,
                       0.61584821, 0.88586679, 1.27427499,
         0.42813324,
        1.83298071, 2.6366509, 3.79269019, 5.45559478, 7.8475997, 11.28837892, 16.23776739, 23.35721469, 33.59818286, 48.32930239, 69.51927962, 100.
                                                                  ])})
In [9]:
ridge cv 1.best params
Out[9]:
{'ridge alpha': 33.59818286283781}
In [10]:
ridge cv 1.score(X test,y test)
Out[10]:
0.7968790227442946
```

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main pipe rf 1 = Pipeline([('col transformer', cf 1),
                       ('rf', RandomForestRegressor())])
In [12]:
params rf 1 = {'rf \max depth' : [1,2,3,4,5],
              'rf min samples split': [2,3,4,5,6,7,8]}
In [13]:
rf cv 1 = GridSearchCV(main pipe rf 1,params rf 1,cv=5)
In [14]:
rf cv 1.fit(X train, y train)
Out[14]:
GridSearchCV(cv=5,
             estimator=Pipeline(steps=[('col transformer',
                                         ColumnTransformer(transformers=[('cat',
                                                                           Pipeline (steps
=[('onehot',
OneHotEncoder(handle unknown='ignore',
sparse=False)),
('poly',
PolynomialFeatures(include bias=False,
interaction only=True))]),
                                                                           ['yearID',
                                                                             'lgID']),
                                                                           ('cont',
                                                                           Pipeline (steps
=[('poly',
PolynomialFeatures (include bias=False,
interaction only=True)),
('scaler',
StandardScaler())]),
                                                                           Index(['2B_bat
', 'BB_ptch', 'CS_bat', 'GIDP_bat', 'HR_bat', 'HR_ptch', 'H_bat',
       'H ptch', 'SB bat', 'SO ptch'],
      dtype='object'))])),
                                        ('rf', RandomForestRegressor())]),
             param grid={'rf max depth': [1, 2, 3, 4, 5],
                          'rf__min_samples_split': [2, 3, 4, 5, 6, 7, 8]})
In [15]:
rf cv 1.score(X test,y test)
Out[15]:
0.7209534277724245
```

Bonus Question

In [16]:

ın [II]:

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))])
pipe cont = Pipeline([('poly', PolynomialFeatures(interaction only=True, include bias=False
)),
                      ('scaler', StandardScaler())])
cf = ColumnTransformer([('cat', pipe cat, cat var),
                         ('cont', pipe cont, cont var)])
main pipe ridge = Pipeline([('col transformer',cf),
                       ('ridge', Ridge(fit intercept=True))])
In [17]:
params ridge = { 'ridge alpha':np.logspace(-1,2,20), 'col transformer cont poly degree
':[1,2,3,4,5]}
ridge cv = GridSearchCV(main pipe ridge,params ridge,cv=5)
In [18]:
ridge cv.fit(X train, y train)
Out[18]:
GridSearchCV(cv=5,
             estimator=Pipeline(steps=[('col_transformer',
                                          ColumnTransformer(transformers=[('cat',
                                                                             Pipeline(steps
=[('onehot',
OneHotEncoder(handle unknown='ignore',
sparse=False)),
('poly',
PolynomialFeatures (include bias=False,
interaction only=True))]),
                                                                             ['yearID',
                                                                              'lgID']),
                                                                            ('cont',
                                                                             Pipeline (steps
=[('poly',
PolynomialFeatures(include bias=False,
interact...
       'H ptch', 'SB bat', 'SO ptch'],
      dtype='object'))])),
                                         ('ridge', Ridge()))),
             param grid={'col transformer cont poly degree': [1, 2, 3, 4, 5],
                          'ridge alpha': array([ 0.1
                                                                  0.14384499, 0.20691381
    0.29763514,
         0.42813324,
                                     0.88586679,
                                                     1.27427499,
                       0.61584821,
         1.83298071, 2.6366509, 3.79269019, 5.45559478, 7.8475997, 11.28837892, 16.23776739, 23.35721469,
        33.59818286, 48.32930239, 69.51927962, 100.
                                                                ])})
In [19]:
ridge cv.best params
Out[19]:
{'col_transformer__cont__poly__degree': 2, 'ridge__alpha': 33.59818286283781}
In [20]:
ridge cv.score(X test,y test)
Out[20]:
0.7968790227442946
```

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In [21]:
main pipe rf = Pipeline([('col transformer',cf),
                       ('rf', RandomForestRegressor())])
In [22]:
params_rf = {'rf__max_depth' : [1,2,3,4,5],
           'rf min samples split': [2,3,4,5,6,7,8]}
In [23]:
rf cv = GridSearchCV(main pipe rf,params rf,cv=5)
In [24]:
rf cv.fit(X train, y train)
Out[24]:
GridSearchCV(cv=5,
             estimator=Pipeline(steps=[('col transformer',
                                         ColumnTransformer(transformers=[('cat',
                                                                           Pipeline(steps
=[('onehot',
OneHotEncoder(handle unknown='ignore',
sparse=False)),
('poly',
PolynomialFeatures(include bias=False,
interaction only=True))]),
                                                                           ['yearID',
                                                                            'lgID']),
                                                                          ('cont',
                                                                           Pipeline(steps
=[('poly',
PolynomialFeatures (include bias=False,
interaction only=True)),
('scaler',
StandardScaler())]),
                                                                           Index(['2B bat
', 'BB_ptch', 'CS_bat', 'GIDP_bat', 'HR_bat', 'HR_ptch', 'H_bat',
       'H ptch', 'SB bat', 'SO ptch'],
      dtype='object'))])),
                                        ('rf', RandomForestRegressor())]),
             param_grid={'rf__max_depth': [1, 2, 3, 4, 5],
                          'rf__min_samples_split': [2, 3, 4, 5, 6, 7, 8]})
In [25]:
rf cv.score(X test,y test)
Out[25]:
```

0.7159170715976091