211112262 VAIBHAV PATEL

```
ML LAB1
[1]: import numpy as
     np import pandas
     as pd
     import matplotlib.pyplot as plt
     import math
[5]: train_dirs=[]
     test_dirs=[]
     headers=[0, 0, 0, 0, 0]
     for dir in os.listdir("./"):
         if(dir.find("5-fold")!=-1):
             train_dirs.append("./"+dir+"/train/")
test_dirs.append("./"+dir+"/test/")
    0.0.1 Linear Regression
[1]: def LinearRegression(train_file, test_file, header):
         train_df = pd.read_csv(train_file, header=header, delimiter=",")
         test_df = pd.read_csv(test_file, header=header, delimiter=",")
         X_train = train_df.iloc[:, :-1].values
         y_train = train_df.iloc[:, -1].values
         X_test = test_df.iloc[:, :-1].values
         y_test = test_df.iloc[:, -1].values
         from sklearn.linear model import LinearRegression
         regressor = LinearRegression()
         regressor.fit(X_train, y_train)
         y_pred =
         regressor.predict(X_test)
         from sklearn.metrics import mean_squared_error, mean_absolute_error,_
      r2 score
         mse = mean squared error(v test v nred)
```

```
r2score = r2_score(y_test, y_pred)
         return np.array([mse, mae, r2score])
[7]: for train_dir, test_dir, header in zip(train_dirs, test_dirs, headers):
         train_files=os.listdir(train_dir)
         test_files=os.listdir(test_dir)
         val=np.zeros(3)
         for train, test in zip(train_files, test_files):
             val+=LinearRegression(train_dir+train, test_dir+test, header)
         print(train_dir)
         val/=len(train_files)
         val[0]=math.sqrt(val[0])
         val=pd.DataFrame(val, index=["RMSE", "MSE", "R2"], columns=["Values"])
         print(val)
         print("
                                             \n")
    ./ CONCIECE-3-10IU/ CLAIII/
             Values
    RMSE 10.441372
    MSE
           8.272734
    R2
           0.603040
    ./diabetes-5-fold/train/
            Values
    RMSE
          0.63949
    8
    MSE 0.501970
    R2 -0.000552
    ./mortgage-5-fold/train/
            Values
    RMSE
          0.12182
    4
    MSE 0.083536
    R2
          0.998404
    ./plastic-5-fold/train/
            Values
    RMSE 1.531465
    MSE
         1.232442
    R2
          0.798437
    ./stock-5-fold/train/
            Values
```

mae = mean_absolute_error(y_test, y_pred)

```
RMSE 2.346750
MSE 1.832574
R2 0.870620
```

0.0.2 Polynomial Regression of degree 2 and 3

```
[8]: def PolynomialRegression(train_file, test_file, header, degree):
         train_df = pd.read_csv(train_file, header=header, delimiter=",")
         test_df = pd.read_csv(test_file, header=header, delimiter=".")
         X_train = train_df.iloc[:, :-1].values
         y_train = train_df.iloc[:, -1].values
         X_test = test_df.iloc[:, :-1].values
         y_test = test_df.iloc[:, -1].values
         from sklearn.linear model import LinearRegression
         from sklearn.preprocessing import PolynomialFeatures
         poly_reg=PolynomialFeatures(degree=degree)
         X_poly=poly_reg.fit_transform(X_train)
         regressor = LinearRegression()
         regressor.fit(X_poly, y_train)
         y_pred = regressor.predict(poly_reg.transform(X_test))
         from sklearn.metrics import mean_squared_error, mean_absolute_error,
      r2 score
         mse = mean_squared_error(y_test, y_pred)
         mae = mean_absolute_error(y_test, y_pred)
         r2score = r2_score(y_test, y_pred)
         return np.array([mse, mae, r2score])
[9]: Polynomial Regression of degree 2
     for train_dir, test_dir, header in zip(train_dirs, test_dirs, headers):
         train_files = os.listdir(train_dir)
         test_files = os.listdir(test_dir)
         val = np.zeros(3)
         for train, test in zip(train_files, test_files):
             val += PolynomialRegression(train_dir + train, test_dir + test, header,__
      2)
```

```
print(train_dir)
         val /= len(train_files)
         val[0] =
         math.sqrt(val[0])
         val = pd.DataFrame(val, index=["RMSE", "MSE", "R2"], columns=["Values"])
         print(val)
     ./concrete-5-fold/train/
            Values
     RMSE
          7.66514
     3
     MSE 5.892674
     R2
          0.786115
     ./diabetes-5-fold/train/
            Values
     RMSE
          0.56129
     7
     MSE 0.456880
     R2
          0.226230
     ./mortgage-5-fold/train/
            Values
     RMSE
          0.12061
     6
     MSE 0.055462
     R2
          0.998544
     ./plastic-5-fold/train/
            Values
     RMSE 1.528545
     MSE 1.226209
     R2
          0.799254
     ./stock-5-fold/train/
            Values
     RMSE 1.127098
     MSE 0.901557
     R2
          0.970154
     Polynomial Regression of degree 3
[10]: for train_dir, test_dir, header in zip(train_dirs, test_dirs, headers):
         train_files = os.listdir(train_dir)
```

test_files = os.listdir(test_dir)

```
val = np.zeros(3)
    for train, test in zip(train_files, test_files):
        val += PolynomialRegression(train_dir + train, test_dir + test, header,_
  3)
    print(train_dir)
    val /= len(train_files)
    val[0] =
    math.sqrt(val[0])
    val = pd.DataFrame(val, index=["RMSE", "MSE", "R2"], columns=["Values"])
    print(val)
    print("
./concrete-5-fold/train/
       Values
RMSE
     6.72537
9
MSE 4.674355
R2
     0.834883
./diabetes-5-fold/train/
       Values
RMSE
     0.83851
1
MSE 0.620331
R2 -0.519498
./mortgage-5-fold/train/
       Values
RMSE
     2.59757
MSE 0.512271
R2
     0.338858
./plastic-5-fold/train/
       Values
RMSE 1.473863
MSE 1.166224
R2
     0.813267
./stock-5-fold/train/
       Values
RMSE 0.880364
MSE 0.661540
R2
     0.981784
```

0.0.3 Regularization in Linear Regression

```
[24]: def Regularization(train_file, test_file, header):
         train_df = pd.read_csv(train_file, header=header, delimiter=",")
         test_df = pd.read_csv(test_file, header=header, delimiter=",")
         X_train = train_df.iloc[:, :-1].values
         y_train = train_df.iloc[:, -1].values
         X_test = test_df.iloc[:, :-1].values
         y_test = test_df.iloc[:, -1].values
         from sklearn.linear_model import Ridge
         alphas = np.array([2**i \text{ for } i \text{ in range}(-18, 51, 2)])
         best_mse, best_alpha=float('inf'), None
         from sklearn.metrics import mean_squared_error
         for alpha in alphas: regressor=Ridge(alpha=alpha)
             regressor fit(X_train, y_train)
             y_pred=regressor.predict(X_test)
             mse=mean_squared_error(y_test, y_pred)
             if mse<best_mse:</pre>
                  best_mse, best_alpha=mse, alpha
        return np.array([best_mse, best_alpha])
[25]: for train_dir, test_dir, header in zip(train_dirs, test_dirs, headers):
         train_files = os.listdir(train_dir)
         test_files = os.listdir(test_dir)
         val = np.zeros(2)
         for train, test in zip(train_files, test_files):
             val = Regularization(train_dir + train, test_dir + test, header)
         print(train_dir)
         val[0]=math.sqrt(val[0])
         val = pd.DataFrame(val, index=["Best RMSE", "Best Alpha"],_
       columns=["Values"])
         print(val)
         print("_____\n")
     Values
   Best RMSE
                 10.222187
   Best Alpha 1024.000000
```

```
./diabetes-5-fold/train/
              Values
Best RMSE
            0.63437
Best Alpha 0.000004
./mortgage-5-fold/train/
              Values
Best RMSE
            0.11068
Best Alpha 1.000000
./plastic-5-fold/train/
             Values
Best RMSE
            1.51057
Best Alpha 0.000004
./stock-5-fold/train/
             Values
Best RMSE
            2.31511
7
Best Alpha 0.000004
```

0.0.4 Ridge on Linear Reg for Best MAE

```
[26]: def Regularization(train_file, test_file, header):
    train_df = pd.read_csv(train_file, header=header, delimiter=",")
    test_df = pd.read_csv(test_file, header=header, delimiter=",")

X_train = train_df.iloc[:, :-1].values
    y_train = train_df.iloc[:, -1].values

X_test = test_df.iloc[:, :-1].values

y_test = test_df.iloc[:, -1].values

from sklearn.linear_model import Ridge
    alphas = np.array([2**i for i in range(-18, 51, 2)])

best_mse, best_alpha=float('inf'), None

from sklearn.metrics import mean_absolute_error
for alpha in alphas:
    regressor=Ridge(alpha=alpha)
    regressor-Fidge(alpha=alpha)
    regressor.predict(X_test)
```

```
mse=mean_absolute_error(y_test, y_pred)
             if mse<best_mse:</pre>
                  best_mse, best_alpha=mse, alpha
        return np.array([best_mse, best_alpha])
[27]: for train_dir, test_dir, header in zip(train_dirs, test_dirs, headers):
         train_files = os.listdir(train_dir)
         test_files = os.listdir(test_dir)
         val = np.zeros(2)
         for train, test in zip(train_files, test_files):
             val = Regularization(train_dir + train, test_dir + test, header)
         print(train_dir)
         val = pd.DataFrame(val, index=["Best MAE", "Best Alpha"],_
       columns=["Values"])
          print(val)
         print("_____\n")
     ./ CUTICIETE-3-1010/ LI all I/
                      Values
    Best MAE
                   8.084741
   Best Alpha 4096.000000
     ./diabetes-5-fold/train/
                  Values
    Best MAE
                0.47644
    Best Alpha 4.00000
     ./mortgage-5-fold/train/
                   Values
     Best MAE
                 0.07921
     Best Alpha 1.000000
     ./plastic-5-fold/train/
                    Values
    Best MAE
                 1.234517
   Best Alpha 64.000000
     ./stock-5-fold/train/
                     Values
    Best MAE
                  1.816376
   Best Alpha 256.000000
```

0.0.5 Ridge for Linear Reg, Best R2 score

```
[28]: def Regularization(train_file, test_file, header):
         train_df = pd.read_csv(train_file, header=header, delimiter=".")
         test_df = pd.read_csv(test_file, header=header, delimiter=",")
         X_train = train_df.iloc[:, :-1].values
         y_train = train_df.iloc[:, -1].values
         X_test = test_df.iloc[:, :-1].values
         y_test = test_df.iloc[:, -1].values
         from sklearn.linear model import Ridge
          alphas = np.array([2**i for i in range(-18, 51, 2)])
          best_mse, best_alpha=-float('inf'), None
         from sklearn.metrics import r2_score
         for alpha in alphas:
              regressor=Ridge(alpha=alpha)
             regressor fit(X_train, y_train)
             y_pred=regressor.predict(X_test)
             mse=r2_score(y_test, y_pred)
             if mse>best mse:
                  best_mse, best_alpha=mse, alpha
[29]: for train_dir, test_dir, header in zip(train_dirs, test_dirs, headers):
         train_files = os.listdir(train_dir)
         test_files = os.listdir(test_dir)
         val = np.zeros(2)
         for train, test in zip(train_files, test_files):
             val = Regularization(train_dir + train, test_dir + test, header)
         print(train_dir)
         val = pd.DataFrame(val, index=["Best R2 Score", "Best Alpha"],
       columns=["Values"])
         print(val)
          print("_____\n")
     ./ CUTICIECE-3-1010/ CLAILI/
                         Values
     Best R2 Score 0.674538
     Best
                         Alpha
     1024.000000
```

```
./diabetes-5-fold/train/
                    Values
             R2
     Best
                     Score
     0.319099 Best Alpha
     0.000004
     ./mortgage-5-fold/train/
                    Values
     Best
             R2
                     Score
     0.998849 Best Alpha
     1.000000
     ./plastic-5-fold/train/
                    Values
             R2
                     Score
     Best
     0.790441 Best Alpha
     0.000004
     ./stock-5-fold/train/
                    Values
     Best
             R2
                     Score
     0.866915 Best Alpha
     0.000004
[17]: import warnings
     warnings.filterwarnings('ignore')
     0.0.6 Ridge for All using MSE
[30]: def PolynomialRidge(train_file, test_file, header, degree):
         train_df = pd.read_csv(train_file, header=header, delimiter=",")
         test_df = pd.read_csv(test_file, header=header, delimiter=",")
         X_train = train_df.iloc[:, :-1].values
         y_train = train_df.iloc[:, -1].values
         X_test = test_df.iloc[:, :-1].values
         y_test = test_df.iloc[:, -1].values
         from sklearn.linear_model import Ridge
```

from sklearn.preprocessing import PolynomialFeatures

poly_reg=PolynomialFeatures(degree=degree)
X_poly=poly_reg.fit_transform(X_train)

```
alphas=np.array([2**i for i in range(-18, 30)])
                             best_alpha, best_mse=None, float('inf')
                             for alpha in alphas:
                                         regressor = Ridge(alpha=alpha)
                                         regressor.fit(X_poly, y_train)
                                        y_pred = regressor.predict(poly_reg.transform(X_test))
                                        from sklearn.metrics import mean_squared_error,
                                         mean_absolute_error,_
                      r2_score
                                         mse = mean_squared_error(y_test, y_pred)
                                         if(mse<best_mse):</pre>
                                                     best_mse, best_alpha=mse, alpha
                                                  .... ...... /[| .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | ... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | .... | ... | ... | .... | .... | .... | .... | .... | .... | .... | .... | ....
[31]: for train_dir, test_dir, header in zip(train_dirs, test_dirs, headers):
                             train_files = os.listdir(train_dir)
                             test_files = os.listdir(test_dir)
                             val = np.zeros(2)
                             for train, test in zip(train_files, test_files):
                                         val += PolynomialRidge(train_dir + train, test_dir + test, header, 2)
                             print(train_dir)
                             val /= len(train_files)
                             val[0] =
                             math.sqrt(val[0])
                             val = pd.DataFrame(val, index=["RMSE", "Alpha"], columns=["Values"])
                             print(val)
                print("-----
                                                     Values
                RMSE
                7.646834e+00
                Alpha 3.355445e+06
                ./diabetes-5-fold/train/
                                                     Values
                                       5.275634e-
                RMSE
                01
                                                       Alpha
                1.677722e+06
                ./mortgage-5-fold/train/
                                                Values
                RMSE
                                         0.075369
                Alpha 153.900001
```

```
./plastic-5-fold/train/
                 Values
     RMSE
               1.527185
     Alpha 26265.600002
     ./stock-5-fold/train/
             Values
     RMSE
            1.11370
     6
     Alpha 5.200002
[21]: for train_dir, test_dir, header in zip(train_dirs, test_dirs, headers):
         train_files = os.listdir(train_dir)
         test_files = os.listdir(test_dir)
         val = np.zeros(2)
         for train, test in zip(train_files, test_files):
             val += PolynomialRidge(train_dir + train, test_dir + test, header, 3)
         print(train_dir)
         val /= len(train_files)
         val[0] =
         math.sqrt(val[0])
         val = pd.DataFrame(val, index=["RMSE", "Alpha"], columns=["Values"])
     print("----\n")
                 Values
     RMSE
     6.089751e+00
     Alpha 1.074266e+08
     ./diabetes-5-fold/train/
                 Values
     RMSE
            6.234885e-
                  Alpha
     1.090521e+08
     ./mortgage-5-fold/train/
                 Values
     RMSE
            6.824711e-
     02
                  Alpha
     1.250165e+08
     ./plastic-5-fold/train/
```

0.0.7 Ridge for all using MAE

```
[22]: def PolynomialRidge(train_file, test_file, header, degree):
         train_df = pd.read_csv(train_file, header=header, delimiter=",")
         test_df = pd.read_csv(test_file, header=header, delimiter=",")
         X_train = train_df.iloc[:, :-1].values
         y_train = train_df.iloc[:, -1].values
         X_test = test_df.iloc[:, :-1].values
         y_test = test_df.iloc[:, -1].values
         from sklearn.linear_model import Ridge
         from sklearn.preprocessing import PolynomialFeatures
         poly_reg=PolynomialFeatures(degree=degree)
         X_poly=poly_reg.fit_transform(X_train)
          alphas=np.array([2^{**i} for i in range(-18, 30)])
         best_alpha, best_mse=None, float('inf')
         for alpha in alphas:
             regressor = Ridge(alpha=alpha)
             regressor fit(X_poly, y_train)
             y_pred = regressor.predict(poly_reg.transform(X_test))
             from sklearn.metrics import mean_squared_error, mean_absolute_error,
       r2_score
              mse = mean_absolute_error(y_test, y_pred)
              if(mse<best_mse):</pre>
                  best_mse, best_alpha=mse, alpha
         return np.array([best_mse, best_alpha])
```

```
[23]: for train_dir, test_dir, header in zip(train_dirs, test_dirs, headers):
         train_files = os.listdir(train_dir)
         test_files = os.listdir(test_dir)
         val = np.zeros(2)
         for train, test in zip(train_files, test_files):
             val += PolynomialRidge(train_dir + train, test_dir + test, header, 3)
         print(train_dir)
         val /= len(train_files)
         val = pd.DataFrame(val, index=["MAE", "Alpha"], columns=["Values"])
         print(val)
         print("____\n")
     ./ כטווכו פנפ-ט-וטוט/ נו מווו/
                  Values
     MAE
            4.500617e+0
     0 Alpha
            1.074791e+0
     ./diabetes-5-fold/train/
                  Values
     MAE
            4.660749e-01
     Alpha
            1.078460e+0
     ./mortgage-5-fold/train/
                  Values
            4.360656e-02
     MAE
     Alpha
            1.183318e+0
     ./plastic-5-fold/train/
              Values
     MAE
            1.164341
     Alpha 0.362501
     ./stock-5-fold/train/
                 Values
              0.647212
    MAE
   Alpha 7782.400196
[32]: for train_dir, test_dir, header in zip(train_dirs, test_dirs, headers):
         train_files = os.listdir(train_dir)
```

test_files = os.listdir(test_dir)

```
val = np.zeros(2)
         for train, test in zip(train_files, test_files):
              val += PolynomialRidge(train_dir + train, test_dir + test, header, 2)
         print(train_dir)
         val /= len(train_files)
         val = pd.DataFrame(val, index=["MAE", "Alpha"], columns=["Values"])
         print(val)
         print("_____\n")
     ./concrete-5-fold/train/
                  Values
     MAE
            5.847408e+0
     1 Alpha
            3.355445e+0
     6
     ./diabetes-5-fold/train/
                  Values
     MAE
            2.783232e-01
     Alpha
            1.677722e+0
     ./mortgage-5-fold/train/
                Values
    MAE
             0.005681
   Alpha 153.900001
     ./plastic-5-fold/train/
                  Values
    MAE
               2.332295
   Alpha 26265.600002
     ./stock-5-fold/train/
              Values
    MAE
           1.240341
    Alpha 5.200002
          Ridge for all using R2 score
[33]: def PolynomialRidge(train_file, test_file, header, degree):
         train_df = pd.read_csv(train_file, header=header, delimiter=",")
         test_df = pd.read_csv(test_file, header=header, delimiter=",")
```

```
X_train = train_df.iloc[:, :-1].values
         y_train = train_df.iloc[:, -1].values
         X_test = test_df.iloc[:, :-1].values
         y_test = test_df.iloc[:, -1].values
         from sklearn.linear_model import Ridge
         from sklearn.preprocessing import PolynomialFeatures
          poly_reg=PolynomialFeatures(degree=degree)
         X_poly=poly_reg.fit_transform(X_train)
          alphas=np.array([2**i \text{ for } i \text{ in } range(-18, 30)])
         best_alpha, best_mse=None, -float('inf')
         for alpha in alphas:
              regressor = Ridge(alpha=alpha)
              regressor.fit(X_poly, y_train)
             y_pred = regressor.predict(poly_reg.transform(X_test))
             from sklearn.metrics import mean_squared_error, mean_absolute_error,
       r2 score
              mse = r2\_score(y\_test, y\_pred)
              if(mse>best_mse):
                  best_mse, best_alpha=mse, alpha
[34]: for train_dir, test_dir, header in zip(train_dirs, test_dirs, headers):
         train_files = os.listdir(train_dir)
         test_files = os.listdir(test_dir)
         val = np.zeros(2)
         for train, test in zip(train_files, test_files):
              val += PolynomialRidge(train_dir + train, test_dir + test, header, 2)
         print(train_dir)
         val /= len(train_files)
         val = pd.DataFrame(val, index=["R2", "Alpha"], columns=["Values"])
         print(val)
         print("____\n")
     ./ เบเเตยเฮ-ม-เบเน/ เเลเเ/
                  Values
     R2
            7.871333e-01
     Alpha
            3.355445e+0
```

```
./diabetes-5-fold/train/
                  Values
     R2
            3.330702e-01
     Alpha
            1.677722e+0
     ./mortgage-5-fold/train/
                Values
    R2
             0.999395
   Alpha 153.900001
     ./plastic-5-fold/train/
                  Values
    R2
               0.799613
   Alpha 26265.600002
     ./stock-5-fold/train/
              Values
    R2
           0.970870
    Alpha 5.200002
[35]: for train_dir, test_dir, header in zip(train_dirs, test_dirs, headers):
          train_files = os.listdir(train_dir)
          test_files = os.listdir(test_dir)
          val = np.zeros(2)
          for train, test in zip(train_files, test_files):
              val += PolynomialRidge(train_dir + train, test_dir + test, header, 3)
          print(train_dir)
          val /= len(train_files)
          val = pd.DataFrame(val, index=["R2 score", "Alpha"], columns=["Values"])
          print(val)
          print("____\n")
     ./ COTICTE LE-3-1010/ LFaIFI/
                     Values
     R2 score 8.650418e-
                     Alpha
     1.074266e+08
     ./diabetes-5-fold/train/
                     Values
     R2 score 1.826710e-
     01
                      Alpha
     1.090521e+08
```

./mortgage-5-fold/train/ Values R2 score 9.995106e-01 Alpha 1.250165e+08 ./plastic-5-fold/train/ Values R2 score 0.81367 1 Alpha 25.625002 ./stock-5-fold/train/ Values R2 score 0.983591 Alpha 16896.000001