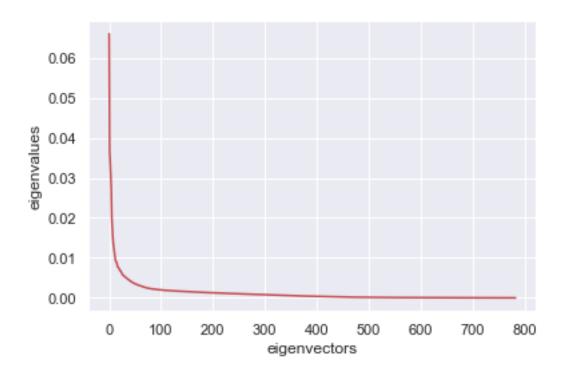
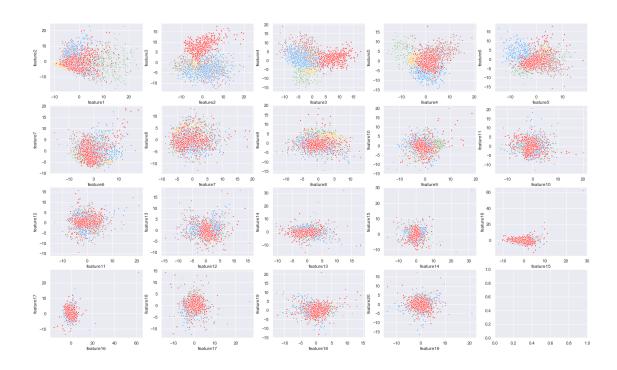
question2

February 28, 2020

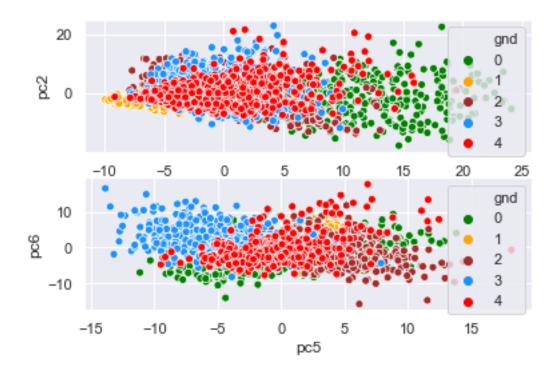
```
[1]: import numpy as np
     import pandas as pd
     import random
     import seaborn as sns; sns.set()
     from sklearn import neighbors
     from sklearn.model_selection import train_test_split
     from sklearn.metrics import accuracy_score
     import sklearn.preprocessing
     import matplotlib.pyplot as plt
     from sklearn.preprocessing import StandardScaler
[2]: handWritten = pd.read_csv("DataB.csv", sep=',')
     handWritten.drop(columns=[handWritten.columns[0]],axis=1,inplace=True)
     V = ['fea.'+str(i+1) for i in range(784)]
     R = 'gnd'
     VR = V + [R]
     SS = StandardScaler()
     handWritten[V] = SS.fit_transform(handWritten[V])
[3]: from sklearn.decomposition import PCA
     pca=PCA().fit(handWritten[V])
     y=pca.explained_variance_ratio_
     x=pca.components_
     plt.xticks = [range(0,20,1)]
     plt.xlabel('eigenvectors')
     plt.ylabel('eigenvalues')
     plt.plot(y,'r')
```

[3]: [<matplotlib.lines.Line2D at 0x129b4af90>]





[5]: <matplotlib.axes._subplots.AxesSubplot at 0x1247bc390>



```
import numpy.linalg as linalg
     def my_svd(X):
         n, m = X.shape
         eigValueU , eigVectorU = linalg.eigh(np.dot(X, X.T))
         eigValueV , eigVectorV = linalg.eigh(np.dot(X.T, X))
         indexU = eigValueU.argsort()[::-1]
         indexV = eigValueV.argsort()[::-1]
         eigValueU = eigValueU[indexU]
         eigVectorU = eigVectorU[:,indexU ]
         eigValueV = eigValueV[indexV]
         eigVectorV = eigVectorV[:,indexV]
         if n>m:
             sigma=np.sqrt(eigValueU)
         else:
             sigma=np.sqrt(eigValueV)
         return eigVectorU, sigma ,eigVectorV.T
[7]: import time
     from statistics import mean
[8]: data = handWritten[V].values
     #рса
     time_start = time.perf_counter()
```

[6]: #svd implementation

```
eigValueU , eigVectorU = linalg.eigh(np.dot(data.T, data))
     indexU = eigValueU.argsort()[::-1]
     eigValueU = eigValueU[indexU]
     eigVectorU = eigVectorU[:,indexU ]
     X_pca = eigVectorU.T.dot(data.T).T
     time_over = time.perf_counter()
     print("running time: "+str(time_over-time_start))
     print(X_pca)
    running time: 0.3689922300000035
    [[ 9.97069222e+00 6.18172201e+00 -4.99286326e+00 ... -2.84664097e-02
       6.97997657e-02 7.44883255e-02]
     [ 1.14159998e+01 6.94158705e+00 -5.06302886e+00 ... -2.28147473e-01
       4.55949595e-02 6.14775845e-02]
     [ 3.69011918e+00 4.69309729e+00 -2.90865640e+00 ... 1.15182625e-01
      -5.60241248e-03 3.14112641e-02]
     [-3.49421529 \text{e}-01 \quad 9.33681056 \text{e}-01 \quad 8.10744188 \text{e}+00 \ \dots \ -6.93521885 \text{e}-02
      -8.49495279e-02 -1.65756817e-02]
     [-3.11526327e+00 2.09047425e+00 6.27251911e+00 ... 1.26995790e-01
       1.46191708e-02 2.07556239e-03]
     [-5.64409375e+00 -2.46166632e-01 4.14018317e+00 ... 1.57632818e-02
      -1.77113186e-02 -1.11643749e-02]]
    dual pca via SVD
[9]: #dual_pca
     time_start = time.perf_counter()
     U,s,VT = my_svd(data.T)
     time_over = time.perf_counter()
     print("svd running time: "+str(time_over-time_start))
     time_start = time.perf_counter()
     X_{dual_PCA} = np.diag(s).dot(VT)[:784].T
     time_over = time.perf_counter()
     print("dual pca running time: "+str(time_over-time_start))
     print(X_dual_PCA)
    /usr/local/lib/python3.7/site-packages/ipykernel_launcher.py:16: RuntimeWarning:
    invalid value encountered in sqrt
      app.launch_new_instance()
    svd running time: 3.444919376999998
    dual pca running time: 0.5355880999999982
    [[ 9.97069222e+00 -6.18172201e+00 4.99286326e+00 ... 2.84664097e-02
      -6.97997657e-02 7.44883255e-02]
     [ 1.14159998e+01 -6.94158705e+00 5.06302886e+00 ... 2.28147473e-01
      -4.55949595e-02 6.14775845e-02]
     [ 3.69011918e+00 -4.69309729e+00 2.90865640e+00 ... -1.15182625e-01
       5.60241248e-03 3.14112641e-02]
```

```
[-3.49421529e-01 -9.33681056e-01 -8.10744188e+00 ... 6.93521885e-02
        8.49495279e-02 -1.65756817e-02]
      [-3.11526327e+00 -2.09047425e+00 -6.27251911e+00 ... -1.26995790e-01]
       -1.46191708e-02 2.07556239e-03]
      [-5.64409375e+00 2.46166632e-01 -4.14018317e+00 ... -1.57632818e-02
        1.77113186e-02 -1.11643749e-02]]
[10]: from sklearn.discriminant_analysis import LinearDiscriminantAnalysis
      transformer=LinearDiscriminantAnalysis()
      Data = pd.DataFrame(transformer.fit_transform(handWritten[V],handWritten[R]))
      Data.columns = ['feature'+str(i+1) for i in range(4)]
      Data['gnd'] = handWritten['gnd']
      fig, axes =plt.subplots(2,2,figsize=(25,15))
      for j in range(3):
          sns.scatterplot(x="feature"+str(j+1), y="feature"+str(j+2),
                          ax=axes[int(j/2)][j\%2],legend = 'full',hue="gnd",_\_
       →data=Data,palette=['green','orange','brown','dodgerblue','red'])
```