# CS156 (Introduction to AI), Spring 2021

# **Assignment\_4**

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## References and sources

https://www.geeksforgeeks.org/python-get-unique-values-list/

https://stackoverflow.com/questions/42521772/importerror-no-module-named-plotly-plotly-in-linuxmint17-3

https://towardsdatascience.com/pca-using-python-scikit-learn-e653f8989e60

https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#deprecate-loc-reindex-listlike

https://scikit-learn.org/stable/modules/generated/sklearn.model\_selection.KFold.html

https://scikit-

learn.org/stable/modules/generated/sklearn.metrics.plot\_confusion\_matrix.html#sklearn.metrics.plot\_co

## **Solution**

#### Code the solution

In [1]: import pandas as pd
import numpy as np

### importing the data

In [2]: dt=pd.read\_csv(r'C:\Users\gursi\Desktop\CS 156\CS 156\Assignment\Assignment 4\homework4
 dt.head()
Out[2]: id ASS1 SPX C6orf141 SP5 SP6 ITGA8 ATP2A1 ATP2A3 ITGA2 ...

	id	ASS1	SPX	C6orf141	SP5	SP6	ITGA8	ATP2A1	ATP2A3	ITGA2	•••
0	TCGA- AB- 2828- 03	3.935027	0.523329	0.0	0.000000	0.102277	2.686908	2.837357	7.444575	2.246715	
1	TCGA- AB- 2846- 03	3.372801	0.000000	0.0	0.000000	0.116270	2.083429	5.567935	8.361999	3.626850	
2	TCGA- AB- 2870- 03	4.198301	0.000000	0.0	0.000000	0.249176	1.546059	6.605116	9.138359	2.858430	
3	TCGA- AB- 2872- 03	4.115014	0.000000	0.0	0.222018	0.632254	2.158989	6.858708	9.201254	2.295229	
4	TCGA- AB- 2881- 03	3.662169	0.467823	0.0	0.000000	0.000000	1.183388	5.567935	9.040883	1.815837	

5 rows × 3002 columns

```
#to see all the different type of cancer
In [3]:
         from collections import Counter
         Counter(dt['Class'].tolist())
         #in this there are 12 different type of cancer
Out[3]: Counter({'Leukemia': 102,
                  'Bladder': 327,
                  'Breast': 875,
                  'Colon': 370,
                  'Glioblastoma': 117,
                  'Head&Neck': 415,
                  'Kidney': 420,
                  'LungAdeno': 424,
                  'LungSquamous': 390,
                  'Ovarian': 337,
                  'Rectal': 128,
                  'Uterine': 431})
         dz= dt.drop(['id','Class'],axis=1)
In [4]:
          dz.head()
In [5]:
                                           SP5
                                                    SP6
Out[5]:
              ASS1
                        SPX C6orf141
                                                           ITGA8
                                                                  ATP2A1
                                                                           ATP2A3
                                                                                     ITGA2
                                                                                              ITGA3
         0 3.935027 0.523329
                                   0.0 0.000000
                                               0.102277 2.686908
                                                                 2.837357 7.444575 2.246715 0.202624
         1 3.372801 0.000000
                                   0.0 0.000000 0.116270 2.083429 5.567935 8.361999 3.626850 0.449673
```

	ASS1	SPX	C6orf141	SP5	SP6	ITGA8	ATP2A1	ATP2A3	ITGA2	ITGA3	•
2	4.198301	0.000000	0.0	0.000000	0.249176	1.546059	6.605116	9.138359	2.858430	1.624238	
3	4.115014	0.000000	0.0	0.222018	0.632254	2.158989	6.858708	9.201254	2.295229	0.326149	
4	3.662169	0.467823	0.0	0.000000	0.000000	1.183388	5.567935	9.040883	1.815837	1.658433	

5 rows × 3000 columns

#### Pca

```
In [6]: #visulaization of the data with original dimensions - we use scatter plot matrix
   import plotly.express as px # install plotly first using pip install plotly
   import numpy as np
   import matplotlib.pyplot as plt
   from sklearn.decomposition import PCA

   pca=PCA(n_components=2)
   pca1_new=pca.fit_transform(dz)

#np.shape(pca1_new)[0]

#here we are creating a new dataframe of pca values
   principalDf = pd.DataFrame(data = pca1_new, columns = ['PCA1', 'PCA2'])
   principalDf.head()

Dz = pd.concat([principalDf, dt[['Class']]], axis = 1)

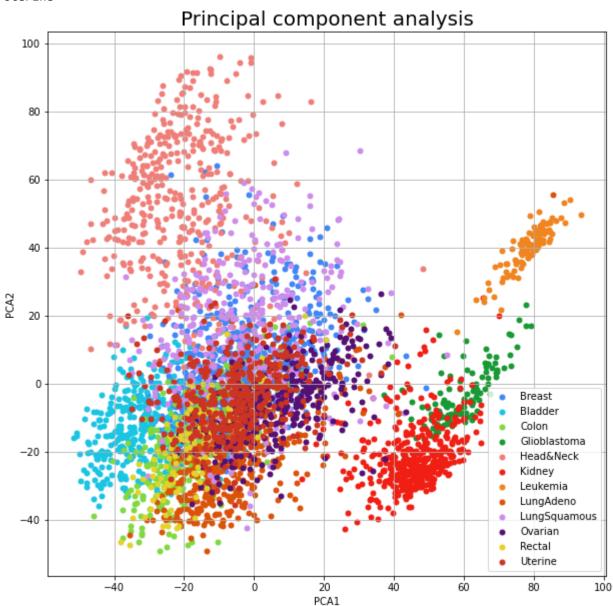
Dz.head()
```

# Out[6]: PCA1 PCA2 Class 0 82.045989 46.713051 Leukemia 1 76.722516 37.919092 Leukemia 2 76.643207 39.867655 Leukemia 3 74.817224 36.351105 Leukemia 4 79.694765 43.781020 Leukemia

#### Plotting the PCA

```
print(class1)
  indicesToKeep = Dz['Class'] == class1 # here only those indices will come as true
  #print(indicesToKeep)
  ax.scatter(Dz.loc[indicesToKeep, 'PCA1'], Dz.loc[indicesToKeep, 'PCA2'] , c = color
ax.legend(Class)
ax.grid()
```

Breast
Bladder
Colon
Glioblastoma
Head&Neck
Kidney
Leukemia
LungAdeno
LungSquamous
Ovarian
Rectal
Uterine



## Normalizing the data

In [9]: from sklearn.model\_selection import train\_test\_split

```
Assignment_4_Gursimran_Singh
           from sklearn.preprocessing import StandardScaler
           scaler=StandardScaler()
          X rescaled=scaler.fit transform(dz)
          print(len(X rescaled))
          4336
          Xtrain, Xtest, Ytrain, Ytest=train test split(X rescaled, dt['Class'], test size =0.2 , rand
In [28]:
          Xtrain.shape, Ytrain.shape
          Xtrain.shape, Xtest.shape, Ytrain.shape, Ytest.shape
          y1=Ytrain.to_numpy() # for cross validation we need Ytrain to be in numpy array
          print(Xtrain.shape)
          (3468, 3000)
```

#### **Cross Validation**

```
from sklearn.svm import LinearSVC
In [29]:
          model = LinearSVC(multi class='ovr', class weight='balanced')
          from sklearn.model selection import KFold
In [30]:
          # KFold(n_splits=10).get_n_splits(X_rescaled)
          kfolds = KFold(n splits =5)
          for train indices, validation indices in kfolds.split(Xtrain):
              #print("Training indices: "+str(train_indices))
              #print("Validation indices: "+str(validation_indices))
              X train set, X validation set = Xtrain[train indices], Xtrain[validation indices]
             # Ytrain[validation indices]
              Y train set, Y validation set = y1[train indices], y1[validation indices] #here pan
              model.fit( X train set, Y train set)
              print(model.score(X_validation_set,Y_validation_set))
          # here KFold method divide the kinto kfolds and return 2 parameter : 1) indices of trai
         0.9639769452449568
         0.9654178674351584
         0.9610951008645533
         0.9624819624819625
         0.9682539682539683
In [31]:
          y pred=model.predict(Xtest)
          print(model.score(Xtest,Ytest))
         0.9700460829493087
          from sklearn.metrics import plot confusion matrix
In [32]:
          #print(plot_confusion_matrix(Ytest, y_pred))
In [33]:
          plt.show()
```

titles options = [("Confusion matrix, without normalization", None), ("Normalized confu

In [34]:

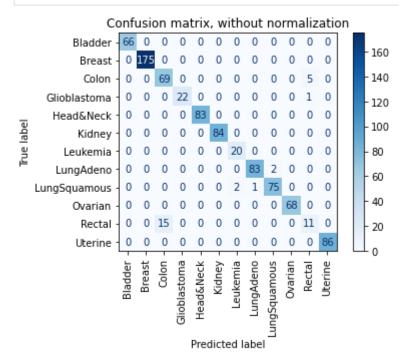
np.set printoptions(precision=2)

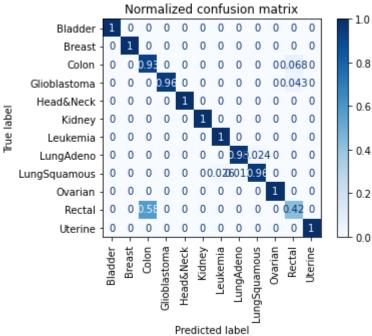
for title, normalize in titles options:

```
disp = plot_confusion_matrix(model, Xtest, Ytest, cmap=plt.cm.Blues, normalize=norm
disp.ax_.set_title(title)

#print(title)
#print(disp.confusion_matrix)

plt.show()
```





```
In [ ]:
In [ ]:
```