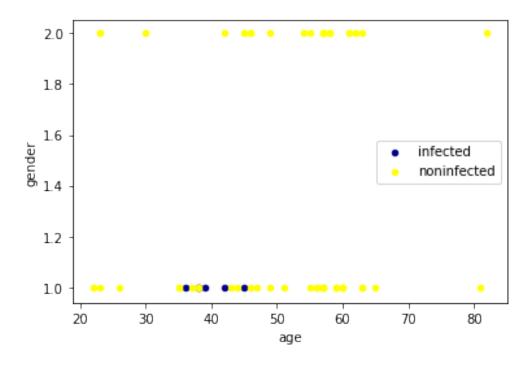
## Phase\_1

May 14, 2020

```
[9]: import pandas as pd
      import pylab as pl
      import numpy as np
      import scipy.optimize as opt
      from sklearn import preprocessing
      from sklearn.model_selection import train_test_split
      %matplotlib inline
      import matplotlib.pyplot as plt
 [6]: #!pip install ibm_watson wget
      !wget -0 italy_normal2.csv https://drive.google.com/open?
       \rightarrowid=1CX6txmOBM4f1ZITxNCa_YVDdb76nFrGu
     --2020-05-05 18:12:19--
     https://drive.google.com/open?id=1CX6txmOBM4f1ZITxNCa_YVDdb76nFrGu
     Resolving drive.google.com (drive.google.com)... 172.217.1.174,
     2607:f8b0:400b:809::200e
     Connecting to drive.google.com (drive.google.com)|172.217.1.174|:443...
     connected.
     HTTP request sent, awaiting response... 307 Temporary Redirect
     Location: https://drive.google.com/file/d/1CX6txmOBM4f1ZITxNCa_YVDdb76nFrGu/view
     ?usp=drive_open [following]
     --2020-05-05 18:12:20-- https://drive.google.com/file/d/1CX6txmOBM4flZITxNCa_YV
     Ddb76nFrGu/view?usp=drive_open
     Reusing existing connection to drive.google.com:443.
     HTTP request sent, awaiting response... 200 OK
     Length: unspecified [text/html]
     Saving to: 'italy_normal2.csv'
                              <=>
                                                   ] 68.79K --.-KB/s
                                                                           in 0.02s
     italy_normal2.csv
     2020-05-05 18:12:21 (4.20 MB/s) - 'italy_normal2.csv' saved [70445]
[10]: covid19 = pd.read_csv("italy_normal2.csv")
      covid19.head()
```

```
[10]:
        age
            gender
                    employmentstatus tested Construction Deliveringtohomes
     0
         57
                 2
                                        4.0
         63
                 2
                                        4.0
                                                       0
                                                                         0
     1
                                  1
     2
         57
                 2
                                        4.0
                                                                         0
                                  1
                                                       0
     3
         23
                 2
                                        4.0
                                                       0
                                                                         0
                                  1
     4
         60
                 1
                                  1
                                        4.0
                                                       0
                                                                         0
                   Healthcare Logisticsothertransportation
                                                          Manufacturing
        Foodretail
     0
                            0
                                                                      0
                0
                            0
     1
                                                        0
                                                                      0
     2
                0
                            0
                                                        0
                                                                      0
     3
                0
                            0
                                                        0
                                                                      0
     4
                0
                            0
                                                        0
                                                                      0
        Chronicobstructivepulmonary disease(COPD) Diabetes Epilepsy
     0
                                            0.0
                                                     0.0
                                                               0.0
                                            0.0
                                                     0.0
                                                               0.0
     1
     2
                                            0.0
                                                     0.0
                                                               0.0
     3
                                            0.0
                                                     0.0
                                                               0.0
     4
                                            0.0
                                                     0.0
                                                               0.0
        Heartdisease Highbloodpressure Highcholesterol HIVAids \
                0.0
                                                           0.0
     0
                                  0.0
                                                  0.0
     1
                0.0
                                  0.0
                                                  0.0
                                                           0.0
                0.0
     2
                                  0.0
                                                  0.0
                                                           0.0
     3
                0.0
                                  0.0
                                                  0.0
                                                           0.0
     4
                0.0
                                  0.0
                                                  1.0
                                                           0.0
        Mentalhealthcondition MultipleSclerosis another
                         0.0
                                           0.0
     0
                                                   1.0
                         0.0
                                           0.0
                                                   0.0
     1
     2
                         0.0
                                           0.0
                                                   1.0
                         0.0
                                           0.0
     3
                                                   1.0
                                           0.0
                                                   0.0
     4
                         0.0
     [5 rows x 29 columns]
[11]: ax = covid19[covid19['tested'] == 1][0:50].plot(kind='scatter', x='age',__
     covid19[covid19['tested'] == 4][0:50].plot(kind='scatter', x='age', y='gender',_
```

plt.show()



```
[13]: #to fix the data from any nan or infinte value (very important pre process lone)
    covid = covid19.replace(r'^\s*$', np.NaN, regex=True)
    covid = covid.fillna(0)
    np.any(np.isnan(covid))
    np.all(np.isfinite(covid))
    covid.dtypes
```

[13]:	age	int64
	gender	int64
	employmentstatus	int64
	tested	float64
	Construction	int64
	Deliveringtohomes	int64
	Foodretail	int64
	Healthcare	int64
	Logisticsothertransportation	int64
	Manufacturing	int64
	Policingorprisons	int64
	Publictransport	int64
	School	int64
	Socialcare	int64
	notworking	int64
	Arthritis	float64
	Asthma	float64
	Cancer	float64

```
Chronicobstructivepulmonary disease(COPD)
                                          float64
     Diabetes
                                          float64
     Epilepsy
                                          float64
                                          float64
     Heartdisease
     Highbloodpressure
                                          float64
     Highcholesterol
                                          float64
     HIVAids
                                          float64
     Mentalhealthcondition
                                          float64
     MultipleSclerosis
                                          float64
                                          float64
     another
     dtype: object
[127]: feature_df = covid[['age', 'gender', 'employmentstatus', 'tested', __
      'Deliveringtohomes', 'Foodretail', 'Healthcare',
           'Logisticsothertransportation', 'Manufacturing', 'Policingorprisons',
           'Publictransport', 'School', 'Socialcare', 'notworking', 'Arthritis',
           'Asthma', 'Cancer', 'Cysticfibrosis',
           'Chronicobstructivepulmonary disease(COPD)', 'Diabetes', 'Epilepsy',
           'Heartdisease', 'Highbloodpressure', 'Highcholesterol', 'HIVAids',
           'Mentalhealthcondition', 'MultipleSclerosis', 'another']]
     X = np.asarray(feature_df)
     X[0:5]
[127]: array([[57., 2., 1., 4., 0., 0., 0., 0., 0., 0., 0.,
            0., 1., 0., 0., 0., 0., 0., 0.,
                                             0., 0.,
                                                     0.,
            0., 0.,
                    1.],
                     1., 4., 0., 0., 0., 0., 0., 0., 0.,
           [63., 2.,
                                                              0.,
                         0., 0., 0., 0., 0.,
                                             0.,
                                                 0.,
            0., 0.,
                     1.,
                                                     0.,
                                                              0.,
                     0.],
                0.,
                     1., 4., 0., 0., 0., 0., 0., 0., 0.,
           [57., 2.,
                     0.,
            0., 0.,
                         0.,
                             0., 0., 0., 0.,
                                             0.,
                                                 0.,
                                                     0.,
                                                          0.,
            0., 0.,
                     1.],
                     1., 4., 0., 0., 0., 0., 0., 0., 0.,
           [23., 2.,
            0., 1.,
                     0.,
                         0.,
                             0., 0., 0., 0.,
                                             0.,
                                                 0.,
                                                     0.,
            0., 0.,
                    1.],
           [60., 1., 1., 4., 0., 0., 0., 0., 0., 0., 0., 0.]
            0., 1., 0., 0., 0., 0., 0., 0., 0., 0., 1.,
            0., 0., 0.]])
[128]: y = np.asarray(covid['tested'])
     y [0:100]
```

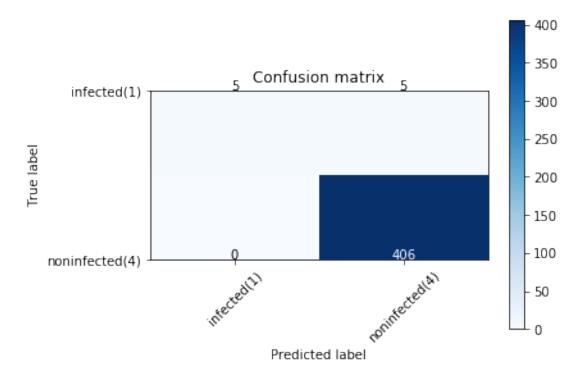
float64

Cysticfibrosis

```
4, 4, 4, 4, 2, 4, 4, 4, 4, 4, 4, 4])
[129]: X_train, X_test, y_train, y_test = train_test_split( X, y, test_size=0.2,__
       →random_state=4)
       print ('Train set:', X_train.shape, y_train.shape)
       print ('Test set:', X_test.shape, y_test.shape)
      Train set: (1672, 29) (1672,)
      Test set: (419, 29) (419,)
[130]: from sklearn import svm
       clf = svm.SVC(kernel='rbf')
       clf.fit(X_train, y_train)
      /home/jupyterlab/conda/envs/python/lib/python3.6/site-
      packages/sklearn/svm/base.py:196: FutureWarning: The default value of gamma will
      change from 'auto' to 'scale' in version 0.22 to account better for unscaled
      features. Set gamma explicitly to 'auto' or 'scale' to avoid this warning.
        "avoid this warning.", FutureWarning)
[130]: SVC(C=1.0, cache_size=200, class_weight=None, coef0=0.0,
         decision_function_shape='ovr', degree=3, gamma='auto_deprecated',
         kernel='rbf', max_iter=-1, probability=False, random_state=None,
         shrinking=True, tol=0.001, verbose=False)
[131]: | yhat = clf.predict(X_test)
       yhat [0:5]
[131]: array([4, 4, 4, 4, 4])
[132]: from sklearn.metrics import classification_report, confusion_matrix
       import itertools
[133]: def plot_confusion_matrix(cm, classes,
                                 normalize=False,
                                 title='Confusion matrix',
                                 cmap=plt.cm.Blues):
           11 11 11
           This function prints and plots the confusion matrix.
           Normalization can be applied by setting `normalize=True`.
           if normalize:
               cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
               print("Normalized confusion matrix")
           else:
               print('Confusion matrix, without normalization')
```

4, 4, 4, 4, 4, 4, 4, 4, 4, 4, 4, 4, 3, 4, 4, 4, 4, 4, 4, 4, 4, 4,

```
print(cm)
           plt.imshow(cm, interpolation='nearest', cmap=cmap)
           plt.title(title)
           plt.colorbar()
           tick_marks = np.arange(len(classes))
           plt.xticks(tick_marks, classes, rotation=45)
           plt.yticks(tick_marks, classes)
           fmt = '.2f' if normalize else 'd'
           thresh = cm.max() / 2.
           for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
               plt.text(j, i, format(cm[i, j], fmt),
                         horizontalalignment="center",
                         color="white" if cm[i, j] > thresh else "black")
           plt.tight_layout()
           plt.ylabel('True label')
           plt.xlabel('Predicted label')
[135]: # Compute confusion matrix
       cnf_matrix = confusion_matrix(y_test, yhat, labels=[2,4])
       np.set_printoptions(precision=2)
       print (classification_report(y_test, yhat))
       # Plot non-normalized confusion matrix
       plt.figure()
       plot_confusion_matrix(cnf_matrix,__
         \neg classes = ['infected(1)', 'noninfected(4)'], normalize = False, \quad title = 'Confusion_{\sqcup} 
        →matrix')
                     precision
                                  recall f1-score
                                                      support
                          0.00
                                    0.00
                                               0.00
                                                             1
                  1
                  2
                          0.83
                                    0.50
                                               0.62
                                                            10
                  3
                          0.00
                                    0.00
                                               0.00
                                                             2
                  4
                          0.98
                                     1.00
                                               0.99
                                                           406
                          0.98
                                               0.98
                                                           419
                                    0.98
         micro avg
         macro avg
                          0.45
                                    0.38
                                               0.40
                                                           419
                                    0.98
                                               0.98
                                                          419
      weighted avg
                          0.97
      Confusion matrix, without normalization
      [[ 5
              5]
       [ 0 406]]
```



```
[136]: #f1_score from sklearn library
from sklearn.metrics import f1_score
f1_score(y_test, yhat, average='weighted')
```

/home/jupyterlab/conda/envs/python/lib/python3.6/sitepackages/sklearn/metrics/classification.py:1143: UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted samples. 'precision', 'predicted', average, warn\_for)

[136]: 0.9756083878995574

```
[137]: #jaccard index for accuracy
from sklearn.metrics import jaccard_similarity_score
jaccard_similarity_score(y_test, yhat)
```

[137]: 0.9809069212410502

[]: