## spam\_classification\_binary

## April 24, 2018

## 1, 0. | spam, non-spam classes

word\_freq\_make: continuous. word\_freq\_address: continuous. word\_freq\_all: continuous. word\_freq\_3d: continuous. word\_freq\_our: continuous. word\_freq\_over: continuous. word\_freq\_remove: continuous. word\_freq\_internet: continuous. word\_freq\_order: continuous. ous. word\_freq\_mail: continuous. word\_freq\_receive: continuous. word\_freq\_will: continuous. word\_freq\_people: continuous. word\_freq\_report: continuous. word\_freq\_addresses: continuous. word\_freq\_free: continuous. word\_freq\_business: continuous. word\_freq\_email: continuous. word\_freq\_you: continuous. word\_freq\_credit: continuous. word\_freq\_your: continuous. word\_freq\_font: continuous. word\_freq\_000: continuous. word\_freq\_money: continuous. word\_freq\_hp: continuous. word\_freq\_hpl: continuous. word\_freq\_george: continuous. word\_freq\_650: continuous. word\_freq\_lab: continuous. word\_freq\_labs: continuous. ous. word\_freq\_telnet: continuous. word\_freq\_857: continuous. word\_freq\_data: continuous. word\_freq\_415: continuous. word\_freq\_85: continuous. word\_freq\_technology: continuous. word\_freq\_1999: continuous. word\_freq\_parts: continuous. word\_freq\_pm: continuous. word\_freq\_direct: continuous. word\_freq\_cs: continuous. word\_freq\_meeting: continuous. word\_freq\_original: continuous. word\_freq\_project: continuous. word\_freq\_re: continuous. word\_freq\_edu: continuous. word\_freq\_table: continuous. word\_freq\_conference: continuous. char\_freq\_;: continuous. char\_freq\_[: continuous. char\_freq\_!: continuous. char\_freq\_!: continuous. ous. char\_freq\_\$: continuous. char\_freq\_#: continuous. capital\_run\_length\_average: continuous. capital\_run\_length\_longest: continuous. capital\_run\_length\_total: continuous.

```
In [137]: df = pd.read_csv('https://archive.ics.uci.edu/ml/machine-learning-databases/spambase/s
         df.head()
Out[137]:
                 0.64 0.64.1 0.1 0.32
                                          0.2
                                                0.3
                                                     0.4
                                                           0.5
                                                                 0.6 ...
                                                                         0.40
           0.21 0.28
                         0.50 0.0 0.14 0.28
                                               0.21 0.07 0.00
                                                                0.94 ...
                                                                         0.00
         1 0.06 0.00
                         0.71 0.0 1.23 0.19 0.19
                                                    0.12 0.64
                                                                0.25 ...
                                                                         0.01
                         0.00 0.0 0.63 0.00 0.31 0.63
         2 0.00 0.00
                                                         0.31
                                                                0.63 ...
                                                                         0.00
            0.00
                 0.00
                         0.00 0.0 0.63 0.00
                                               0.31
                                                    0.63
                                                          0.31
                                                                0.63 ...
                                                                         0.00
                         0.00
                              0.0 1.85 0.00
                                               0.00 1.85 0.00
            0.00
                 0.00
                                                                         0.00
             0.41
                  0.42 0.778
                                0.43
                                      0.44
                                            3.756
                                                    61
                                                        278
           0.132
                   0.0
                        0.372 0.180
                                     0.048 5.114
                                                  101
                                                       1028
           0.143
                  0.0 0.276 0.184 0.010 9.821
                                                  485
                                                       2259
```

```
2 0.137
                    0.0 0.137 0.000 0.000 3.537
                                                    40
                                                         191 1
                    0.0 0.135 0.000 0.000 3.537
                                                         191 1
         3 0.135
                                                    40
                    0.0 0.000 0.000 0.000 3.000
                                                          54
         4 0.223
                                                    15
                                                             1
         [5 rows x 58 columns]
In [138]: x = df[list(df.columns)[:-1]]
         y = df['1']
         df['1'].value_counts()
Out[138]: 0
              2788
         1
              1812
         Name: 1, dtype: int64
In [139]: \# df['category'] = ('NSP', 'SP')[bool(df['1'].eq(0).all())]
         df.tail()
         # df.category.value_counts()
Out[139]:
                  0 0.64 0.64.1 0.1 0.32
                                            0.2 0.3 0.4 0.5 0.6 ...
                                                                           0.40 \
                      0.0
                            0.62 0.0 0.00 0.31 0.0 0.0
                                                            0.0 0.0 ...
         4595
               0.31
                                                                          0.000
         4596
                      0.0
                                                            0.0 0.0 ...
               0.00
                            0.00 0.0 0.00 0.00 0.0 0.0
                                                                          0.000
         4597
               0.30
                     0.0
                            0.30 0.0 0.00 0.00 0.0 0.0 0.0 0.0 ...
         4598
               0.96
                            0.00 0.0 0.32 0.00 0.0 0.0
                                                            0.0 0.0 ...
                      0.0
                                                                         0.000
         4599
               0.00
                      0.0
                            0.65 0.0 0.00 0.00 0.0 0.0
                                                            0.0 0.0 ...
                                                                         0.000
                0.41 0.42 0.778 0.43 0.44 3.756 61
                                                        278 1
         4595
               0.232
                      0.0 0.000
                                   0.0
                                         0.0 1.142
                                                     3
                                                         88 0
         4596
               0.000
                      0.0 0.353
                                   0.0
                                         0.0 1.555
                                                         14 0
         4597
               0.718
                       0.0 0.000
                                   0.0
                                         0.0 1.404
                                                     6 118 0
         4598
               0.057
                       0.0 0.000
                                   0.0
                                         0.0 1.147
                                                     5
                                                         78 0
         4599
               0.000
                                         0.0 1.250
                                                         40 0
                      0.0 0.125
                                   0.0
                                                     5
         [5 rows x 58 columns]
In [140]: x_train,x_test,y_train,y_test=train_test_split(x,y)
In [141]: # model = SGDClassifier()
         model = LogisticRegression()
         model.fit(x_train,y_train)
Out[141]: LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
                   intercept_scaling=1, max_iter=100, multi_class='ovr', n_jobs=1,
                   penalty='12', random_state=None, solver='liblinear', tol=0.0001,
                   verbose=0, warm_start=False)
In [142]: model.score(x_test,y_test)
Out[142]: 0.92
In [153]: 'Spam',len(df[df['1']==1]),'Not Spam',len(df[df['1']==0])
Out[153]: ('Spam', 1812, 'Not Spam', 2788)
```