```
In [2]: import pandas as pd import matplotlib import matplotlib.pyplot as plt import numpy as np import random import seaborn as sns import warnings import time from collections import Counter from sklearn.preprocessing import StandardScaler, MinMaxScaler from sklearn.decomposition import PCA from scipy import stats
```

```
In [3]: seed = 101
```

```
In [4]: featureNames = ["word_freq_make", "word_freq_address", "word_freq_all", "word_freq_3d", "word_freq_our",
    "word_freq_over", "word_freq_remove", "word_freq_internet", "word_freq_order", "word_freq_mail",
    "word_freq_receive", "word_freq_will", "word_freq_people", "word_freq_report", "word_freq_addresses",
    "word_freq_free", "word_freq_business", "word_freq_email", "word_freq_you", "word_freq_credit",
    "word_freq_your", "word_freq_font", "word_freq_email", "word_freq_hp",
    "word_freq_bpl", "word_freq_george", "word_freq_650", "word_freq_lab", "word_freq_labs",
    "word_freq_telnet", "word_freq_857", "word_freq_data", "word_freq_415", "word_freq_85",
    "word_freq_technology", "word_freq_data", "word_freq_parts", "word_freq_pm", "word_freq_direct",
    "word_freq_cs", "word_freq_meeting", "word_freq_original", "word_freq_project", "word_freq_re",
    "word_freq_edu", "word_freq_table", "word_freq_conference", "char_freq_;", "char_freq_[",
    "char_freq_!", "char_freq_$", "char_freq_#", "capital_run_length_average",
    "capital_run_length_longest", "capital_run_length_total", "class"]
    dt = pd.read_csv('dataset/spambase.csv', header=None, names=featureNames)
```

In [5]: dt.describe()

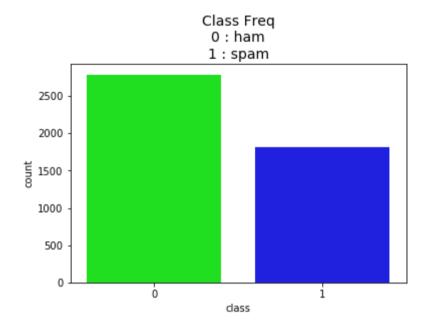
Out[5]:

| | word_freq_make | word_freq_address | word_freq_all | word_freq_3d | word_freq_our | word_freq_over | word_freq_remove | word_freq_internet \(\text{\cdot} \) |
|-------|----------------|-------------------|---------------|--------------|---------------|----------------|------------------|---------------------------------------|
| count | 4601.000000 | 4601.000000 | 4601.000000 | 4601.000000 | 4601.000000 | 4601.000000 | 4601.000000 | 4601.000000 |
| mean | 0.104553 | 0.213015 | 0.280656 | 0.065425 | 0.312223 | 0.095901 | 0.114208 | 0.105295 |
| std | 0.305358 | 1.290575 | 0.504143 | 1.395151 | 0.672513 | 0.273824 | 0.391441 | 0.401071 |
| min | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 |
| 25% | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 |
| 50% | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 |
| 75% | 0.000000 | 0.000000 | 0.420000 | 0.000000 | 0.380000 | 0.000000 | 0.000000 | 0.000000 |
| max | 4.540000 | 14.280000 | 5.100000 | 42.810000 | 10.000000 | 5.880000 | 7.270000 | 11.110000 |

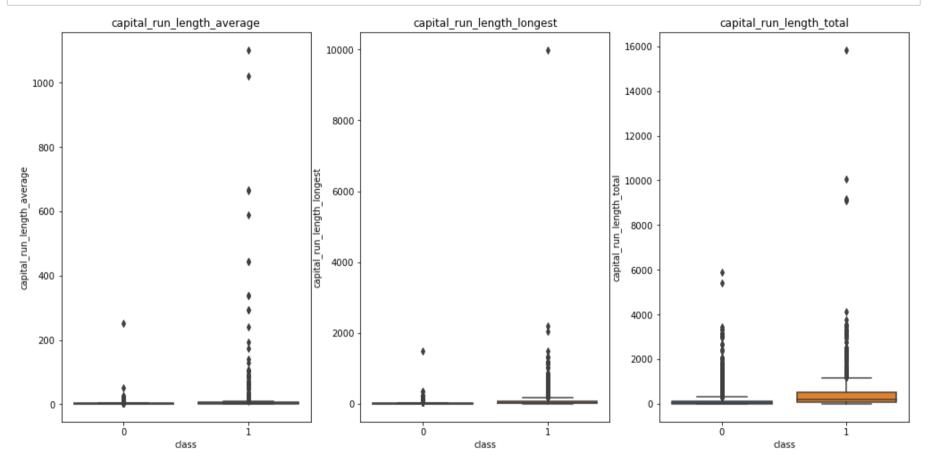
8 rows × 58 columns

In [6]: colors = ["#00FF00","#0000FF"]
sns.countplot('class',data = dt, palette = colors)
plt.title('Class Freq\n0 : ham\n1 : spam',fontsize=14)

Out[6]: Text(0.5, 1.0, 'Class Freq\n0 : ham\n1 : spam')



In [7]: f, axes = plt.subplots(1, 3, figsize=(17,8)) sns.boxplot(x="class",y="capital_run_length_average",data = dt, ax=axes[0]).set_title('capital_run_length_average') sns.boxplot(x="class",y="capital_run_length_longest",data = dt, ax=axes[1]).set_title('capital_run_length_longest') sns.boxplot(x="class",y="capital_run_length_total",data = dt, ax=axes[2]).set_title('capital_run_length_total') plt.show()



```
In [8]:

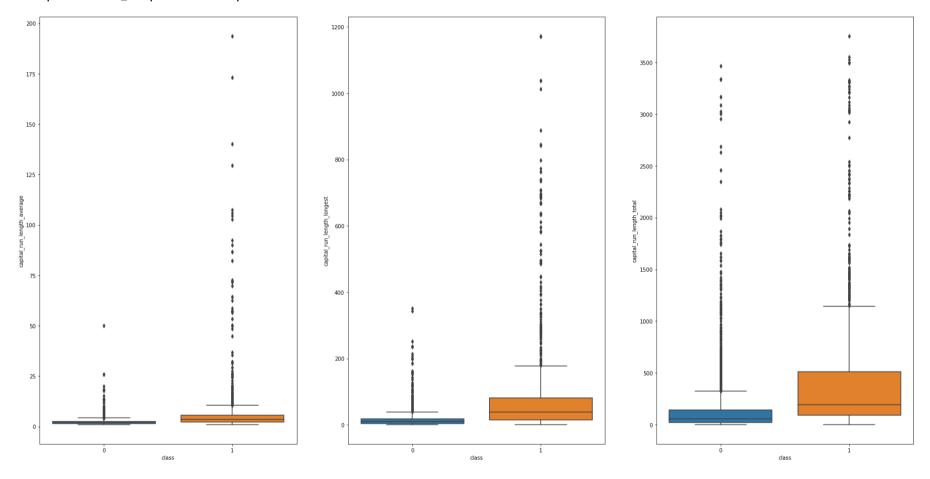
def removeOutlier(df, col_name, threshold, upper=True):
    if(upper==True):
        df = df.drop(df[(df[col_name] >= threshold)].index)
    else:
        df = df.drop(df[(df[col_name] < threshold)].index)
    return df

dt = removeOutlier(dt, 'capital_run_length_average',200)
    dt = removeOutlier(dt, 'capital_run_length_longest', 1400)
    dt = removeOutlier(dt, 'capital_run_length_total', 4000)</pre>
```

After Removing Outlier

```
In [9]: f, axes = plt.subplots(1, 3, figsize=(30,15))
sns.boxplot(x='class', y='capital_run_length_average',data=dt,ax = axes[0])
sns.boxplot(x='class', y='capital_run_length_longest',data=dt,ax = axes[1])
sns.boxplot(x='class', y='capital_run_length_total',data=dt,ax = axes[2])
```

Out[9]: <matplotlib.axes._subplots.AxesSubplot at 0x7f67abd5db20>



In [10]: mscaler = MinMaxScaler((0,100))

dt['capital_run_length_average'] = mscaler.fit_transform(np.array(dt['capital_run_length_average']).reshape(-1,1))

dt['capital_run_length_longest'] = mscaler.fit_transform(np.array(dt['capital_run_length_longest']).reshape(-1,1))

dt['capital_run_length_total'] = mscaler.fit_transform(np.array(dt['capital_run_length_total']).reshape(-1,1))

Visualization after Normalization

In [11]: dt.describe()

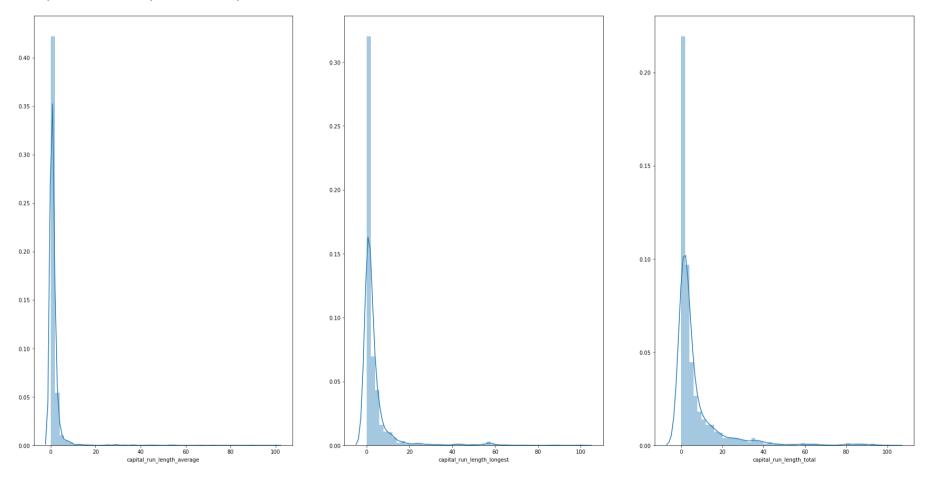
Out[11]:

| | word_freq_make | word_freq_address | word_freq_all | word_freq_3d | word_freq_our | word_freq_over | word_freq_remove | word_freq_internet | ١ |
|-------|----------------|-------------------|---------------|--------------|---------------|----------------|------------------|--------------------|---|
| count | 4580.000000 | 4580.000000 | 4580.000000 | 4580.000000 | 4580.000000 | 4580.000000 | 4580.000000 | 4580.000000 | - |
| mean | 0.104044 | 0.213707 | 0.277900 | 0.065675 | 0.311205 | 0.096162 | 0.114585 | 0.105603 | |
| std | 0.305758 | 1.293433 | 0.500914 | 1.398339 | 0.673004 | 0.274375 | 0.392245 | 0.401910 | |
| min | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 25% | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 50% | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 75% | 0.000000 | 0.000000 | 0.420000 | 0.000000 | 0.380000 | 0.000000 | 0.000000 | 0.000000 | |
| max | 4.540000 | 14.280000 | 5.100000 | 42.810000 | 10.000000 | 5.880000 | 7.270000 | 11.110000 | |
| | | | | | | | | | |

8 rows × 58 columns

In [12]: f, axes = plt.subplots(1,3,figsize=(30,15))
sns.distplot(dt['capital_run_length_average'],ax=axes[0])
sns.distplot(dt['capital_run_length_longest'],ax=axes[1])
sns.distplot(dt['capital_run_length_total'],ax=axes[2])

Out[12]: <matplotlib.axes._subplots.AxesSubplot at 0x7f67ae2fd8e0>

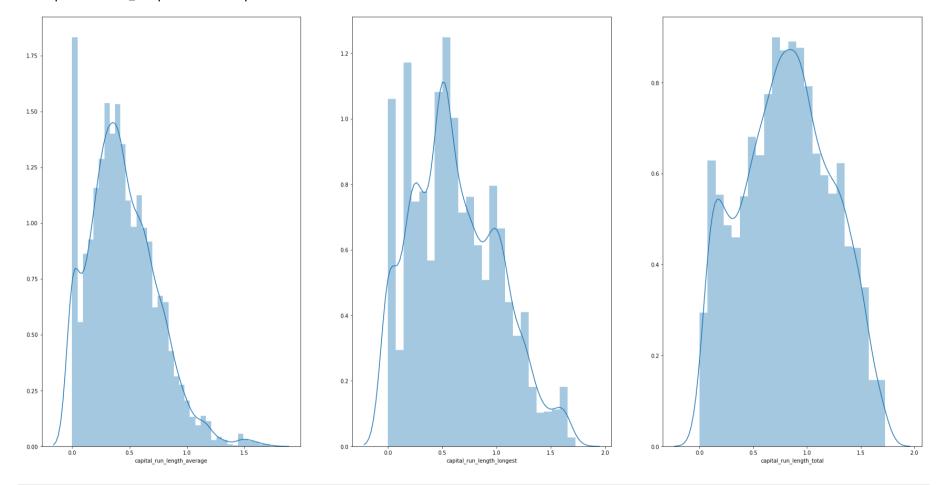


```
In [13]: dt['capital_run_length_average'] = np.log(np.array(dt['capital_run_length_average']+1).reshape(-1,1)) dt['capital_run_length_longest'] = np.log(np.array(dt['capital_run_length_longest']+1).reshape(-1,1)) dt['capital_run_length_total'] = np.log(np.array(dt['capital_run_length_total']+1).reshape(-1,1)) dt['capital_run_length_longest'] = np.log(np.array(dt['capital_run_length_longest']+1).reshape(-1,1)) dt['capital_run_length_total'] = np.log(np.array(dt['capital_run_length_total']+1).reshape(-1,1))
```

Transformation

In [14]: f, axes = plt.subplots(1,3,figsize=(30,15))
sns.distplot(dt['capital_run_length_average'],ax=axes[0])
sns.distplot(dt['capital_run_length_longest'],ax=axes[1])
sns.distplot(dt['capital_run_length_total'],ax=axes[2])

Out[14]: <matplotlib.axes._subplots.AxesSubplot at 0x7f67abde4610>



In [15]: dt = dt.reset_index() X = dt X = X.drop(columns='index') y = X.pop('class')

Choose the best parameter

```
In [17]: from sklearn.svm import SVC
from sklearn.model_selection import GridSearchCV
from sklearn.model_selection import StratifiedKFold

skf = StratifiedKFold(5)
svm_par = {'C': [0.01, 0.1, 0.5, 1, 10, 100], 'kernel': ['rbf', 'poly', 'sigmoid', 'linear']}
clf = GridSearchCV(SVC(random_state=seed), svm_par).fit(X, y).best_estimator_

In [18]: from sklearn.metrics import accuracy_score, precision_recall_fscore_support
def GetMetrics(estimator, X, y):
    y_pred = estimator.predict(X)
    acc = accuracy_score(y, y_pred)
    pre, rec, fs, sup = precision_recall_fscore_support(y, y_pred)
    return acc, pre, rec, fs
```

Training and Testing

print the scores

```
In [20]: names = ['Accuracy', 'Precision', 'Recall', 'f-score']

for i in range(len(scores)):
    print(names[i], scores[i], sep=' = ',)

Accuracy = 0.8427947598253275
    Precision = [0.90097087 0.7680798 ]
    Recall = [0.83303411 0.85793872]
    f-score = [0.86567164 0.81052632]

In []:
```