

```
In [1]: import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score
from sklearn.neighbors import KNeighborsClassifier
from sklearn.preprocessing import StandardScaler
import seaborn as sns
```

```
In [2]: df = pd.read_csv("C:/Users/Adamin/OneDrive/Desktop/emails.csv")
```

```
In [3]: df.head()
```

```
Out[3]:
```

	Email No.	the	to	ect	and	for	of	a	you	hou	...	connevey	jay	valued	lay	i
0	Email 1	0	0	1	0	0	0	2	0	0	...	0	0	0	0	
1	Email 2	8	13	24	6	6	2	102	1	27	...	0	0	0	0	
2	Email 3	0	0	1	0	0	0	8	0	0	...	0	0	0	0	
3	Email 4	0	5	22	0	5	1	51	2	10	...	0	0	0	0	
4	Email 5	7	6	17	1	5	2	57	0	9	...	0	0	0	0	

5 rows × 3002 columns



```
In [4]: df.isnull().sum()
```

```
Out[4]: Email No.      0
the                0
to                0
ect                0
and                0
..
military          0
allowing          0
ff                0
dry               0
Prediction        0
Length: 3002, dtype: int64
```

```
In [5]: X = df.iloc[:,1:3001] # word frequency features
X
```

Out[5]:

	the	to	ect	and	for	of	a	you	hou	in	...	enhancements	connevey	jay
0	0	0	1	0	0	0	2	0	0	0	...	0	0	(
1	8	13	24	6	6	2	102	1	27	18	...	0	0	(
2	0	0	1	0	0	0	8	0	0	4	...	0	0	(
3	0	5	22	0	5	1	51	2	10	1	...	0	0	(
4	7	6	17	1	5	2	57	0	9	3	...	0	0	(
...
5167	2	2	2	3	0	0	32	0	0	5	...	0	0	(
5168	35	27	11	2	6	5	151	4	3	23	...	0	0	(
5169	0	0	1	1	0	0	11	0	0	1	...	0	0	(
5170	2	7	1	0	2	1	28	2	0	8	...	0	0	(
5171	22	24	5	1	6	5	148	8	2	23	...	0	0	(

5172 rows × 3000 columns



In [6]: `Y = df.iloc[:, -1].values # 1 = spam, 0 = not spam`
`Y`

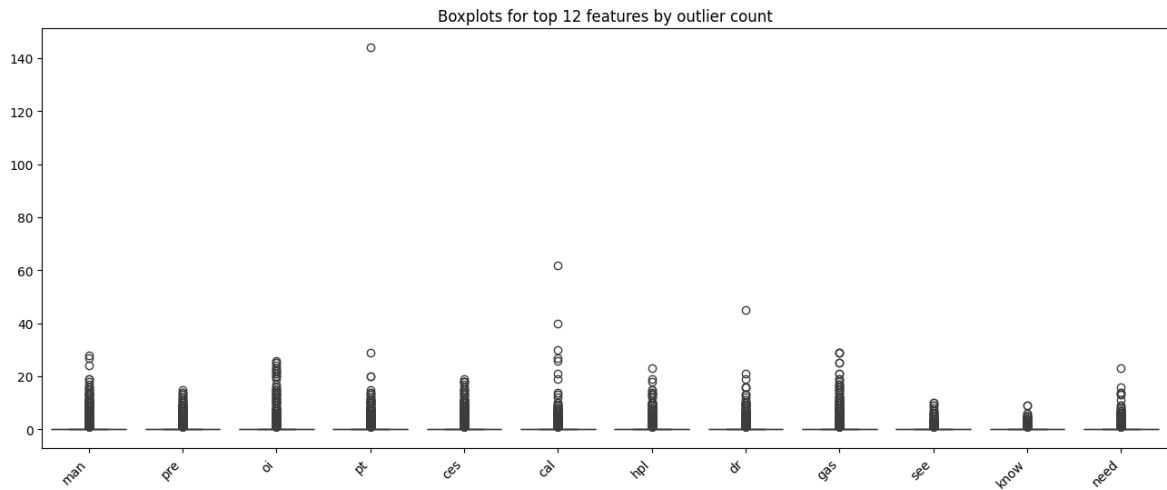
Out[6]: `array([0, 0, 0, ..., 1, 1, 0], shape=(5172,))`

In [8]: `df_numeric = df[numeric_cols]`
`numeric_cols = df.select_dtypes(include=['int64', 'float64']).columns.tolist()`
`# Visualize outliers`
`import matplotlib.pyplot as plt`
`import seaborn as sns`

`# compute IQR outlier counts (you already had this)`
`Q1 = df_numeric.quantile(0.25)`
`Q3 = df_numeric.quantile(0.75)`
`IQR = Q3 - Q1`
`lower = Q1 - 1.5 * IQR`
`upper = Q3 + 1.5 * IQR`
`outlier_mask = ((df_numeric < lower) | (df_numeric > upper))`
`outlier_counts = outlier_mask.sum().sort_values(ascending=False)`

`# pick top N features`
`topN = 12`
`top_features = outlier_counts.head(topN).index.tolist()`

`plt.figure(figsize=(16,6))`
`sns.boxplot(data=df_numeric[top_features])`
`plt.title(f"Boxplots for top {topN} features by outlier count")`
`plt.xticks(rotation=45, ha='right')`
`plt.show()`



```
In [9]: # Split data
X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size=0.25, random
```

```
In [10]: from sklearn.metrics import classification_report, confusion_matrix
```

```
# ----- Support Vector Machine -----
svc = SVC(C=1.0, kernel='rbf', gamma='auto')
svc.fit(X_train, y_train)
svc_pred = svc.predict(X_test)
```

```
In [11]: print("SVM Accuracy:", accuracy_score(y_test, svc_pred))
print("SVM Classification Report:\n", classification_report(y_test, svc_pred))
print("SVM Confusion Matrix:\n", confusion_matrix(y_test, svc_pred))
```

SVM Accuracy: 0.8932714617169374

SVM Classification Report:

	precision	recall	f1-score	support
0	0.90	0.96	0.93	913
1	0.87	0.74	0.80	380
accuracy			0.89	1293
macro avg	0.89	0.85	0.87	1293
weighted avg	0.89	0.89	0.89	1293

SVM Confusion Matrix:

```
[[872 41]
 [ 97 283]]
```

```
In [12]: # ----- K-Nearest Neighbors -----
knn = KNeighborsClassifier(n_neighbors=7)
knn.fit(X_train, y_train)
knn_pred = knn.predict(X_test)
```

```
In [13]: print("KNN Accuracy:", knn.score(X_test, y_test))
print("KNN Classification Report:\n", classification_report(y_test, knn_pred))
print("KNN Confusion Matrix:\n", confusion_matrix(y_test, knn_pred))
```

KNN Accuracy: 0.8662026295436969

KNN Classification Report:

	precision	recall	f1-score	support
0	0.93	0.87	0.90	913
1	0.74	0.85	0.79	380
accuracy			0.87	1293
macro avg	0.83	0.86	0.85	1293
weighted avg	0.87	0.87	0.87	1293

KNN Confusion Matrix:

```
[[798 115]
 [ 58 322]]
```

```
In [14]: scaler = StandardScaler()
X_train_s = scaler.fit_transform(X_train)
X_test_s = scaler.transform(X_test)

ks = [1, 3, 5]

results = {}
for k in ks:
    knn = KNeighborsClassifier(n_neighbors=k, n_jobs=-1)
    knn.fit(X_train_s, y_train)           # X_train_s must be scaled features
    y_pred = knn.predict(X_test_s)       # X_test_s must be scaled features

    acc = accuracy_score(y_test, y_pred)
    cm = confusion_matrix(y_test, y_pred)
    report = classification_report(y_test, y_pred, zero_division=0)

    results[k] = acc

print(f"\nK = {k}:")
print(f"  Accuracy = {acc:.4f}")
print("  Confusion Matrix:")
print(cm)
print("  Classification Report:")
print(report)
```

```

C:\Users\Adamin\AppData\Roaming\Python\Python313\site-packages\joblib\externals\loky\backend\context.py:136: UserWarning: Could not find the number of physical cores for the following reason:
[WinError 2] The system cannot find the file specified
Returning the number of logical cores instead. You can silence this warning by setting LOKY_MAX_CPU_COUNT to the number of cores you want to use.
  warnings.warn(
    File "C:\Users\Adamin\AppData\Roaming\Python\Python313\site-packages\joblib\externals\loky\backend\context.py", line 257, in _count_physical_cores
      cpu_info = subprocess.run(
        "wmic CPU Get NumberOfCores /Format:csv".split(),
        capture_output=True,
        text=True,
      )
    File "C:\Program Files\Python313\Lib\subprocess.py", line 554, in run
      with Popen(*popenargs, **kwargs) as process:
        ~~~~^~~~~~
    File "C:\Program Files\Python313\Lib\subprocess.py", line 1039, in __init__
      self._execute_child(args, executable, preexec_fn, close_fds,
      ~~~~~~^~~~~~
      pass_fds, cwd, env,
      ^~~~~~
    ...<5 lines>...
      gid, gids, uid, umask,
      ^~~~~~
      start_new_session, process_group)
      ^~~~~~
    File "C:\Program Files\Python313\Lib\subprocess.py", line 1554, in _execute_child
      hp, ht, pid, tid = _winapi.CreateProcess(executable, args,
      ~~~~~~^~~~~~
      # no special security
      ^~~~~~
    ...<4 lines>...
      cwd,
      ^~~~
      startupinfo)
      ^~~~~~

```

```

K = 1:
  Accuracy = 0.9026
  Confusion Matrix:
[[819  94]
 [ 32 348]]
  Classification Report:

```

	precision	recall	f1-score	support
0	0.96	0.90	0.93	913
1	0.79	0.92	0.85	380
accuracy			0.90	1293
macro avg	0.87	0.91	0.89	1293
weighted avg	0.91	0.90	0.90	1293

```

K = 3:
  Accuracy = 0.8639
  Confusion Matrix:
[[758 155]
 [ 21 359]]
  Classification Report:

```

	precision	recall	f1-score	support
0	0.97	0.83	0.90	913
1	0.70	0.94	0.80	380
accuracy			0.86	1293
macro avg	0.84	0.89	0.85	1293
weighted avg	0.89	0.86	0.87	1293

```

K = 5:
  Accuracy = 0.8345
  Confusion Matrix:
[[716 197]
 [ 17 363]]
  Classification Report:

```

	precision	recall	f1-score	support
0	0.98	0.78	0.87	913
1	0.65	0.96	0.77	380
accuracy			0.83	1293
macro avg	0.81	0.87	0.82	1293
weighted avg	0.88	0.83	0.84	1293

In []: