```
In [1]: pip install pandas scikit-learn
```

Defaulting to user installation because normal site-packages is not writeable Requirement already satisfied: pandas in c:\programdata\anaconda3\lib\site-package s (2.0.3)

Requirement already satisfied: scikit-learn in c:\programdata\anaconda3\lib\site-p ackages (1.3.0)

Requirement already satisfied: python-dateutil>=2.8.2 in c:\programdata\anaconda3 \lib\site-packages (from pandas) (2.8.2)

Requirement already satisfied: pytz>=2020.1 in c:\programdata\anaconda3\lib\site-p ackages (from pandas) (2023.3.post1)

Requirement already satisfied: tzdata>=2022.1 in c:\programdata\anaconda3\lib\site -packages (from pandas) (2023.3)

Requirement already satisfied: numpy>=1.21.0 in c:\programdata\anaconda3\lib\site-packages (from pandas) (1.24.3)

Requirement already satisfied: scipy>=1.5.0 in c:\programdata\anaconda3\lib\site-p ackages (from scikit-learn) (1.11.1)

Requirement already satisfied: joblib>=1.1.1 in c:\programdata\anaconda3\lib\site-packages (from scikit-learn) (1.2.0)

Requirement already satisfied: threadpoolctl>=2.0.0 in c:\programdata\anaconda3\lib\site-packages (from scikit-learn) (2.2.0)

Requirement already satisfied: six>=1.5 in c:\programdata\anaconda3\lib\site-packa ges (from python-dateutil>=2.8.2->pandas) (1.16.0)

Note: you may need to restart the kernel to use updated packages.

```
In [4]:
    import pandas as pd
    from sklearn.model_selection import train_test_split
    from sklearn.ensemble import RandomForestClassifier
    from sklearn.metrics import classification_report, accuracy_score
```

```
In [5]: data = pd.read_csv('data.csv')
```

```
In [23]: # Replace Labels
data['label'] = data['label'].replace({'good': 'benign', 'bad': 'malicious'})
```

In [33]: data

Out

33]:		url	label	features
	0	diaryofagameaddict.com	malicious	{'url_length': 22, 'num_digits': 0, 'num_speci
	1	espdesign.com.au	malicious	{'url_length': 16, 'num_digits': 0, 'num_speci
	2	iamagameaddict.com	malicious	{'url_length': 18, 'num_digits': 0, 'num_speci
	3	kalantzis.net	malicious	{'url_length': 13, 'num_digits': 0, 'num_speci
	4	slightlyoffcenter.net	malicious	{'url_length': 21, 'num_digits': 0, 'num_speci
	•••			
	420459	23.227.196.215/	malicious	{'url_length': 15, 'num_digits': 11, 'num_spec
	420460	apple-checker.org/	malicious	{'url_length': 18, 'num_digits': 0, 'num_speci
	420461	apple-iclods.org/	malicious	{'url_length': 17, 'num_digits': 0, 'num_speci
	420462	apple-uptoday.org/	malicious	{'url_length': 18, 'num_digits': 0, 'num_speci
	420463	apple-search.info	malicious	{'url_length': 17, 'num_digits': 0, 'num_speci

420464 rows × 3 columns

```
In [7]: def extract_url_features(url):
             return {
                  'url_length': len(url),
                  'num_digits': sum(c.isdigit() for c in url),
                  'num_special_chars': sum(c in '!@#$%^&*()-_=+[{]}|;:",<.>/?' for c in url),
                  'https_present': int('https' in url)
             }
In [8]: # Apply feature extraction
         data['features'] = data['url'].apply(extract_url_features)
         features_df = pd.DataFrame(data['features'].tolist())
In [10]: # Define the features (X) and target variable (y)
         X = features df
         y = data['label'] # Assuming 'Label' column contains 'malicious' or 'benign'
In [11]: # Split the data into training and test sets (80% training, 20% testing)
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_sta
In [12]: # Initialize the classifier
         clf = RandomForestClassifier(n_estimators=100, random_state=42)
         # Train the classifier
In [13]:
         clf.fit(X_train, y_train)
                   RandomForestClassifier
Out[13]:
         RandomForestClassifier(random_state=42)
In [14]: # Make predictions on the test set
         y_pred = clf.predict(X_test)
In [15]: # Evaluate the model
         print("Accuracy:", accuracy_score(y_test, y_pred))
         print(classification_report(y_test, y_pred))
         Accuracy: 0.862081267168492
                                  recall f1-score
                       precision
                                                       support
                  bad
                            0.76
                                      0.33
                                                0.46
                                                         14964
                            0.87
                                      0.98
                                                0.92
                                                         69129
                 good
                                                0.86
             accuracy
                                                         84093
                            0.81
                                      0.65
                                                0.69
                                                         84093
            macro avg
         weighted avg
                            0.85
                                      0.86
                                                0.84
                                                         84093
In [16]: # Load the new unlabeled dataset
         unlabeled data = pd.read csv('filtered file.csv')
In [17]:
         unlabeled_data
```

Out[17]: url https://www.google.com 1 https://www.youtube.com 2 https://www.facebook.com 3 https://www.baidu.com 4 https://www.wikipedia.org 450171 http://ecct-it.com/docmmnn/aptgd/index.php 450172 http://faboleena.com/js/infortis/jquery/plugin... 450173 http://faboleena.com/js/infortis/jquery/plugin... 450174 http://atualizapj.com/ 450175 http://writeassociate.com/test/Portal/inicio/I...

450176 rows × 1 columns

```
In [18]: # Extract features from the new URLs
    unlabeled_data['features'] = unlabeled_data['url'].apply(extract_url_features)
    unlabeled_features_df = pd.DataFrame(unlabeled_data['features'].tolist())

In [25]: # Predict Labels
    predictions = clf.predict(unlabeled_features_df)
    unlabeled_data['predicted_label'] = predictions

In [26]: # Replace predicted Labels
    unlabeled_data['predicted_label'] = unlabeled_data['predicted_label'].replace({'good In [27]: unlabeled_data]}.replace({'good In [27]: unlabeled_data]}.
```

Out[27]: features predicted label url {'url length': 22, 0 https://www.google.com malicious 'num digits': 0, 'num speci... {'url length': 23, 1 https://www.youtube.com malicious 'num_digits': 0, 'num_speci... {'url length': 24, 2 https://www.facebook.com malicious 'num_digits': 0, 'num_speci... {'url length': 21, 3 https://www.baidu.com malicious 'num_digits': 0, 'num_speci... {'url length': 25, https://www.wikipedia.org 4 benign 'num_digits': 0, 'num_speci... http://ecct-{'url length': 43, 450171 benign it.com/docmmnn/aptgd/index.php 'num_digits': 0, 'num_speci... {'url length': 159, 450172 http://faboleena.com/js/infortis/jquery/plugin... benign 'num_digits': 21, 'num_spe... {'url_length': 147, 450173 http://faboleena.com/js/infortis/jquery/plugin... benign 'num_digits': 20, 'num_spe... {'url_length': 22, 450174 http://atualizapj.com/ benign 'num_digits': 0, 'num_speci... {'url length': 143, 450175 http://writeassociate.com/test/Portal/inicio/I... benign 'num_digits': 9, 'num_spec... 450176 rows × 3 columns In [29]: # If you have true labels true_labels = pd.read_csv('urldata.csv') true labels['label'] = true labels['label'].replace({'good': 'benign', 'bad': 'mali unlabeled_data = pd.merge(unlabeled_data, true_labels, on='url') # Calculate accuracy In [31]: if 'label' in unlabeled data.columns: accuracy = accuracy_score(unlabeled_data['label'], unlabeled_data['predicted_la' print(f'Accuracy on unlabeled data: {accuracy:.2f}') else: print("True labels are not available for accuracy calculation.")

Accuracy on unlabeled data: 0.07