cs-nslkdd-random-forest

May 20, 2023

```
[4]: # Load the dataset
     df = pd.read_csv("preprocessed_dataset_NSLKDD.csv")
[5]: import pandas as pd
     from sklearn.ensemble import RandomForestClassifier
     from sklearn.metrics import classification_report, confusion_matrix
     from sklearn.preprocessing import OneHotEncoder
     import matplotlib.pyplot as plt
     import seaborn as sns
     # Create feature matrix X and target vector y
     X = df.drop('subclass', axis=1)
     y = df['subclass']
     \# Perform one-hot encoding on the categorical columns in X
     categorical_cols = X.select_dtypes(include='object').columns
     encoder = OneHotEncoder(drop='first', sparse=False)
     X_encoded = pd.DataFrame(encoder.fit_transform(X[categorical_cols]),__
      Golumns=encoder.get_feature_names(categorical_cols))
     X = pd.concat([X.drop(categorical cols, axis=1), X encoded], axis=1)
     # Create and train the Random Forest model
     model = RandomForestClassifier()
     model.fit(X, y)
     # Predict the test set
     y_pred = model.predict(X)
     # Evaluate the model
     accuracy = model.score(X, y)
     classification = classification_report(y, y_pred)
     confusion = confusion_matrix(y, y_pred)
     # Print the evaluation metrics
     print("Accuracy:", accuracy)
     print("\nClassification Report:")
     print(classification)
```

```
# Plot the confusion matrix
plt.figure(figsize=(8, 6))
sns.heatmap(confusion, annot=True, fmt='d', cmap='Blues')
plt.xlabel('Predicted')
plt.ylabel('True')
plt.title('Confusion Matrix')
plt.show()
# Plot the 2x2 confusion matrix
confusion_2x2 = confusion[:2, :2] # Extract the 2x2 portion of the confusion_
 \rightarrow matrix
tn, fp, fn, tp = confusion 2x2.ravel() # Unpack the values of TN, FP, FN, TP
plt.figure(figsize=(4, 4))
sns.heatmap(confusion_2x2, annot=True, fmt='d', cmap='Blues')
plt.xlabel('Predicted')
plt.ylabel('True')
plt.title('2x2 Confusion Matrix')
plt.text(0, 0, f'TN: {tn}', horizontalalignment='center',
 ⇔verticalalignment='center')
plt.text(0, 1, f'FN: {fn}', horizontalalignment='center', u
 ⇔verticalalignment='center')
plt.text(1, 0, f'FP: {fp}', horizontalalignment='center',
 ⇔verticalalignment='center')
plt.text(1, 1, f'TP: {tp}', horizontalalignment='center', u
 ⇔verticalalignment='center')
plt.show()
# Filter out the "normal" class from the target column
attacks = df[df['subclass'] != 'normal']['subclass']
# Count the occurrences of each attack class
attack_counts = attacks.value_counts()
# Plot the distribution of attacks
plt.figure(figsize=(10, 6))
sns.barplot(x=attack_counts.index, y=attack_counts.values)
plt.xlabel('Attack Class')
plt.ylabel('Count')
plt.title('Distribution of Attacks (Excluding "normal" Class)')
plt.xticks(rotation=90)
plt.show()
```

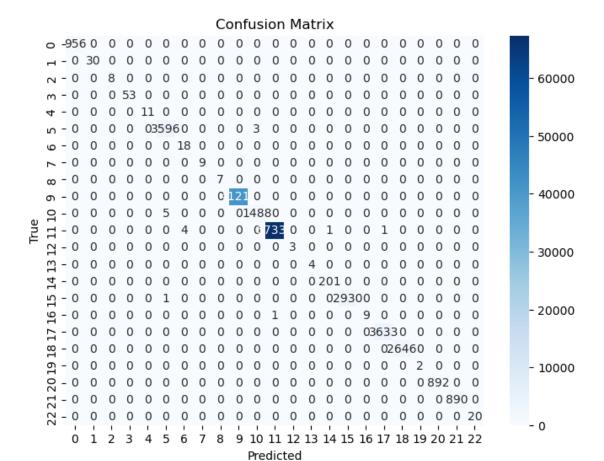
C:\Users\pappu\anaconda3\lib\site-packages\sklearn\utils\deprecation.py:87: FutureWarning: Function get_feature_names is deprecated; get_feature_names is deprecated in 1.0 and will be removed in 1.2. Please use get_feature_names_out

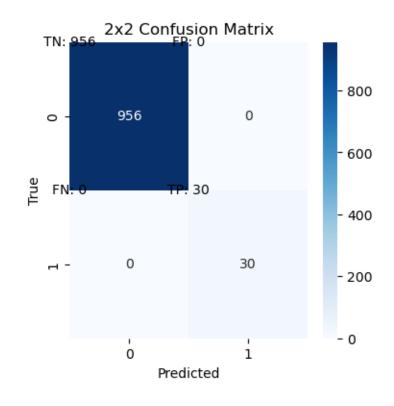
instead.
warnings.warn(msg, category=FutureWarning)

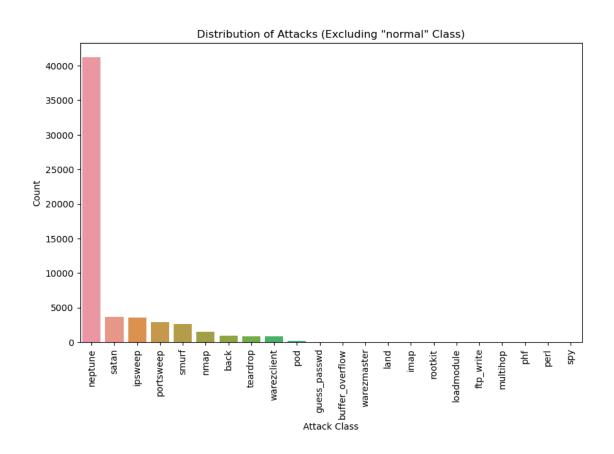
Accuracy: 0.9998729886562994

Classification Report:

	precision	recall	f1-score	support
back	1.00	1.00	1.00	956
buffer_overflow	1.00	1.00	1.00	30
${ t ftp_write}$	1.00	1.00	1.00	8
${ t guess_passwd}$	1.00	1.00	1.00	53
imap	1.00	1.00	1.00	11
ipsweep	1.00	1.00	1.00	3599
land	0.82	1.00	0.90	18
loadmodule	1.00	1.00	1.00	9
multihop	1.00	1.00	1.00	7
neptune	1.00	1.00	1.00	41214
nmap	1.00	1.00	1.00	1493
normal	1.00	1.00	1.00	67343
perl	1.00	1.00	1.00	3
phf	1.00	1.00	1.00	4
pod	1.00	1.00	1.00	201
portsweep	1.00	1.00	1.00	2931
rootkit	1.00	0.90	0.95	10
satan	1.00	1.00	1.00	3633
smurf	1.00	1.00	1.00	2646
spy	1.00	1.00	1.00	2
teardrop	1.00	1.00	1.00	892
warezclient	1.00	1.00	1.00	890
warezmaster	1.00	1.00	1.00	20
accuracy			1.00	125973
macro avg	0.99	1.00	0.99	125973
weighted avg	1.00	1.00	1.00	125973







[]: