log_anomaly_detection_notebook

June 7, 2025

0.1 Step 1: Load Libraries and Dataset

[ ]:
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.preprocessing import LabelEncoder, StandardScaler
from sklearn.cluster import KMeans, DBSCAN
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import classification_report, confusion_matrix,␣
↪roc_auc_score, accuracy_score
from sklearn.model_selection import train_test_split

# Load the dataset
df = pd.read_csv("log_sample.csv")
df.head()
```

0.2 Step 2: Data Preprocessing

[ ]:
```
# Encode categorical features
df_encoded = df.copy()
le = LabelEncoder()
df_encoded['protocol'] = le.fit_transform(df_encoded['protocol'])
df_encoded['flags'] = le.fit_transform(df_encoded['flags'])
df_encoded['label'] = df_encoded['label'].map({'benign': 0, 'DoS': 1})

# Drop timestamp and IPs (not useful in clustering/classification for now)
df_encoded = df_encoded.drop(columns=["timestamp", "src_ip", "dst_ip"])
df_encoded.head()
```

0.3 Step 3: Clustering with K-Means

[ ]:
```
scaler = StandardScaler()
X_scaled = scaler.fit_transform(df_encoded.drop(columns=["label"]))

kmeans = KMeans(n_clusters=2, random_state=42)
df_encoded["cluster_kmeans"] = kmeans.fit_predict(X_scaled)
```

```
# Compare with actual labels
pd.crosstab(df_encoded['label'], df_encoded['cluster_kmeans'],␣
 ↪rownames=['Actual'], colnames=['Cluster'])
```

0.4 Step 4: Classification with Random Forest

[ ]:
```
X = df_encoded.drop(columns=["label", "cluster_kmeans"])
y = df_encoded["label"]

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,␣
 ↪random_state=42)

clf = RandomForestClassifier(n_estimators=100, random_state=42)
clf.fit(X_train, y_train)
y_pred = clf.predict(X_test)

print("Confusion Matrix:\n", confusion_matrix(y_test, y_pred))
print("\nClassification Report:\n", classification_report(y_test, y_pred))
print("ROC AUC Score:", roc_auc_score(y_test, clf.predict_proba(X_test)[:,1]))
```

0.5 Step 5: Evaluation ROC Curve, Confusion Matrix, Metrics Summary

[ ]:
```
#  Evaluation: ROC Curve, Confusion Matrix, Metrics Summary
from sklearn.metrics import roc_curve, auc, ConfusionMatrixDisplay

# ROC Curve
y_prob = clf.predict_proba(X_test)[:, 1]
fpr, tpr, _ = roc_curve(y_test, y_prob)
roc_auc = auc(fpr, tpr)

plt.figure(figsize=(8, 5))
plt.plot(fpr, tpr, color='darkorange', lw=2, label=f'ROC curve (AUC = {roc_auc:.
 ↪2f})')
plt.plot([0, 1], [0, 1], color='navy', lw=2, linestyle='--')
plt.xlabel("False Positive Rate")
plt.ylabel("True Positive Rate")
plt.title("ROC Curve")
plt.legend(loc="lower right")
plt.grid(True)
plt.show()

# Confusion Matrix
ConfusionMatrixDisplay.from_estimator(clf, X_test, y_test, cmap='Blues',␣
 ↪values_format='d')
plt.title("Confusion Matrix")
plt.show()
```

0.6 Pipeline Diagram: Visual Overview

[ ]:
```
#  Pipeline Diagram: Visual Overview
import matplotlib.pyplot as plt
import matplotlib.patches as mpatches

fig, ax = plt.subplots(figsize=(10, 6))
ax.axis('off')

# Define steps and coordinates
steps = [
    ("Raw Logs\n(CSV)", (0.1, 0.5)),
    ("Preprocessing\n+ Feature Selection", (0.3, 0.5)),
    ("Clustering\n(K-Means / DBSCAN)", (0.5, 0.7)),
    ("Classification\n(Random Forest)", (0.5, 0.3)),
    ("Evaluation\n(ROC, F1, Accuracy)", (0.7, 0.5)),
]

# Draw boxes
for text, (x, y) in steps:
    ax.add_patch(mpatches.FancyBboxPatch((x, y), 0.18, 0.15,
                                         boxstyle="round,pad=0.02",
                                         fc="skyblue", ec="black", lw=1.5))
    ax.text(x + 0.09, y + 0.075, text, ha='center', va='center', fontsize=10)

# Arrows between steps
arrows = [
    ((0.28, 0.575), (0.3, 0.575)),
    ((0.3, 0.575), (0.48, 0.725)),
    ((0.3, 0.575), (0.48, 0.325)),
    ((0.5, 0.725), (0.68, 0.575)),
    ((0.5, 0.325), (0.68, 0.575)),
]

arrow_style = dict(arrowstyle="->", lw=1.5, color="black")
for start, end in arrows:
    ax.annotate("", xy=end, xytext=start, arrowprops=arrow_style)

plt.title("Log-Based DoS Detection Pipeline", fontsize=14)
plt.show()
```