


```
import pandas as pd
df = pd.read_csv('cybersecurity_intrusion_data.csv')
```

```
df.head()
```



| | network_packet_size | login_attempts | session_duration | ip_reputation_score | failed_logins | unusual_time_access | attack_detected | prot |
|---|---------------------|----------------|------------------|---------------------|---------------|---------------------|-----------------|------|
| 0 | 599 | 4 | 492.983263 | 0.606818 | 1 | 0 | 1 | |
| 1 | 472 | 3 | 1557.996461 | 0.301569 | 0 | 0 | 0 | |
| 2 | 629 | 3 | 75.044262 | 0.739164 | 2 | 0 | 1 | |
| 3 | 804 | 4 | 601.248835 | 0.123267 | 0 | 0 | 1 | |
| 4 | 453 | 5 | 532.540888 | 0.054874 | 1 | 0 | 0 | |


Next steps:

[Generate code with df](#)

[View recommended plots](#)


[New interactive sheet](#)

```
df.info()
```




```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9537 entries, 0 to 9536
Data columns (total 14 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   network_packet_size                   9537 non-null   int64
1   login_attempts                       9537 non-null   int64
2   session_duration                     9537 non-null   float64
3   ip_reputation_score                  9537 non-null   float64
4   failed_logins                        9537 non-null   int64
5   unusual_time_access                  9537 non-null   int64
6   attack_detected                      9537 non-null   int64
7   protocol_type_TCP                    9537 non-null   bool
8   protocol_type_UDP                    9537 non-null   bool
9   encryption_used_DES                  9537 non-null   bool
10  browser_type_Edge                     9537 non-null   bool
11  browser_type_Firefox                  9537 non-null   bool
12  browser_type_Safari                   9537 non-null   bool
13  browser_type_Unknown                  9537 non-null   bool
dtypes: bool(7), float64(2), int64(5)
memory usage: 586.9 KB
```

```
df.describe()
```



| | network_packet_size | login_attempts | session_duration | ip_reputation_score | failed_logins | unusual_time_access | attack_detected |
|-------|---------------------|----------------|------------------|---------------------|---------------|---------------------|-----------------|
| count | 9537.000000 | 9537.000000 | 9537.000000 | 9537.000000 | 9537.000000 | 9537.000000 | 9537.000000 |
| mean | 500.430639 | 4.032086 | 792.745312 | 0.331338 | 1.517773 | 0.149942 | 0.447101 |
| std | 198.379364 | 1.963012 | 786.560144 | 0.177175 | 1.033988 | 0.357034 | 0.497220 |
| min | 64.000000 | 1.000000 | 0.500000 | 0.002497 | 0.000000 | 0.000000 | 0.000000 |
| 25% | 365.000000 | 3.000000 | 231.953006 | 0.191946 | 1.000000 | 0.000000 | 0.000000 |
| 50% | 499.000000 | 4.000000 | 556.277457 | 0.314778 | 1.000000 | 0.000000 | 0.000000 |
| 75% | 635.000000 | 5.000000 | 1105.380602 | 0.453388 | 2.000000 | 0.000000 | 1.000000 |
| max | 1285.000000 | 13.000000 | 7190.392213 | 0.924299 | 5.000000 | 1.000000 | 1.000000 |

```
df.shape
```



```
(9537, 14)
```

```
df = pd.get_dummies(df, drop_first=True)
```

```
# Check for missing values
print(df.isnull().sum())
# If any:
df.fillna(df.mean(), inplace=True)
```

```

network_packet_size    0
login_attempts         0
session_duration       0
ip_reputation_score    0
failed_logins          0
unusual_time_access    0
attack_detected        0
protocol_type_TCP      0
protocol_type_UDP      0
encryption_used_DES    0
browser_type_Edge      0
browser_type_Firefox   0
browser_type_Safari    0
browser_type_Unknown   0
dtype: int64

from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
X = df.drop('attack_detected', axis=1)
y = df['attack_detected']
scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)
X_train, X_test, y_train, y_test = train_test_split(X_scaled, y, test_size=0.2, random_state=42)

# Logistic Regression
from sklearn.linear_model import LogisticRegression
lr = LogisticRegression(max_iter=1000)
lr.fit(X_train, y_train)
lr_pred = lr.predict(X_test)

# Random Forest
from sklearn.ensemble import RandomForestClassifier
rf = RandomForestClassifier()
rf.fit(X_train, y_train)
rf_pred = rf.predict(X_test)

# SVM
from sklearn.svm import SVC
svm = SVC()
svm.fit(X_train, y_train)
svm_pred = svm.predict(X_test)

# KNN
from sklearn.neighbors import KNeighborsClassifier
knn = KNeighborsClassifier()
knn.fit(X_train, y_train)
knn_pred = knn.predict(X_test)

# Decision Tree
from sklearn.tree import DecisionTreeClassifier
dt = DecisionTreeClassifier()
dt.fit(X_train, y_train)
dt_pred = dt.predict(X_test)

from sklearn.metrics import accuracy_score, precision_score
print('Logistic Regression:', accuracy_score(y_test, lr_pred), precision_score(y_test, lr_pred))
print('Random Forest:', accuracy_score(y_test, rf_pred), precision_score(y_test, rf_pred))
print('SVM:', accuracy_score(y_test, svm_pred), precision_score(y_test, svm_pred))
print('KNN:', accuracy_score(y_test, knn_pred), precision_score(y_test, knn_pred))
print('Decision Tree:', accuracy_score(y_test, dt_pred), precision_score(y_test, dt_pred))

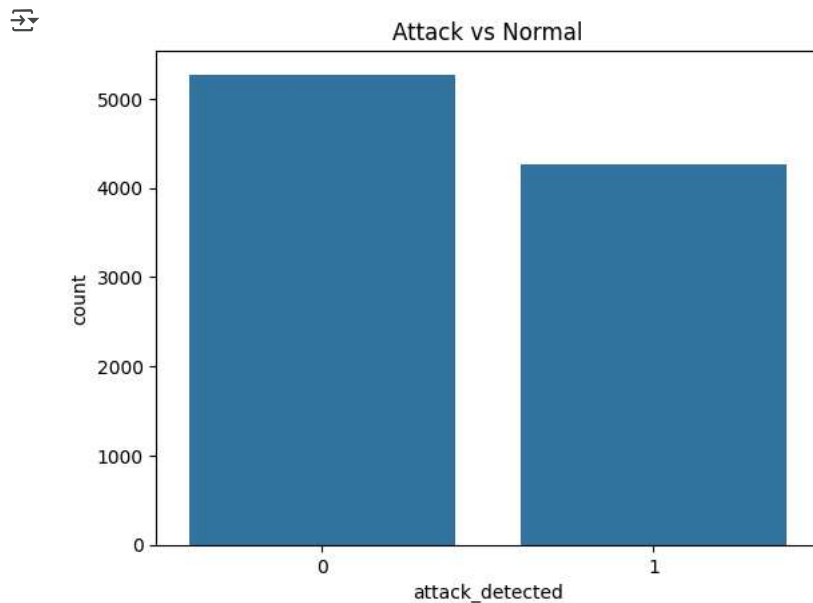
Logistic Regression: 0.7468553459119497 0.7477360931435963
Random Forest: 0.8930817610062893 0.9896449704142012
SVM: 0.8721174004192872 0.9380281690140845
KNN: 0.7971698113207547 0.8537666174298375
Decision Tree: 0.8265199161425576 0.8050171037628279

sample = X_test[5].reshape(1, -1)
print('Prediction:', rf.predict(sample))

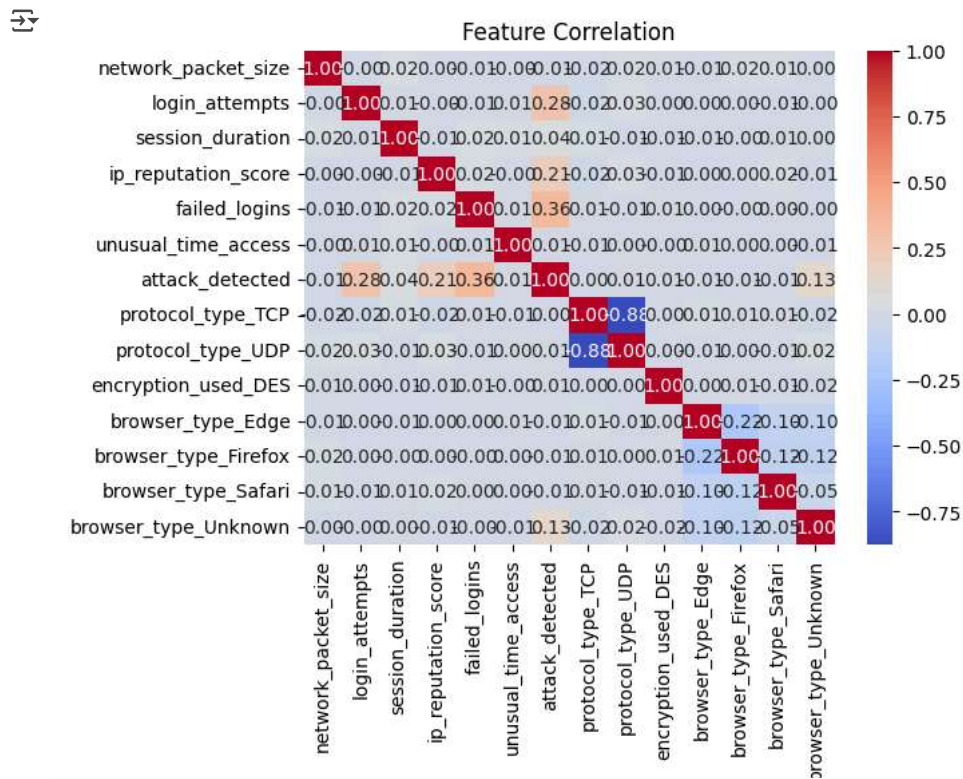
Prediction: [0]

```

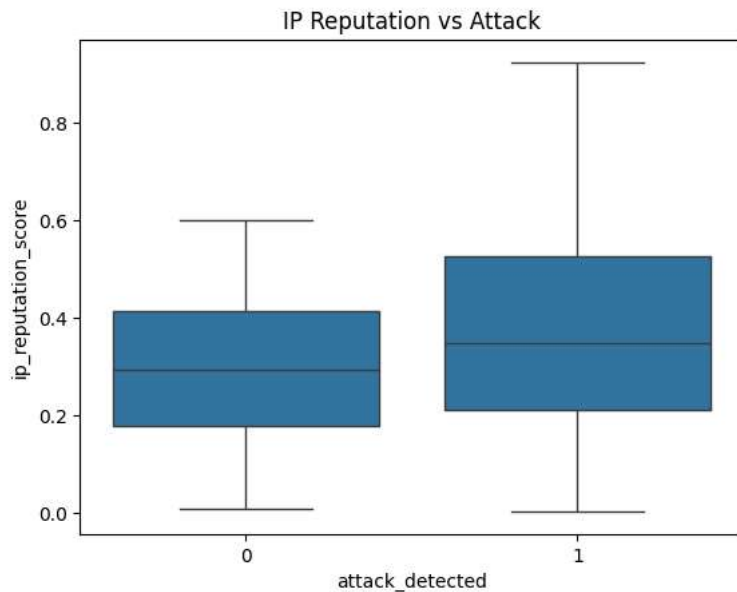
```
import seaborn as sns
import matplotlib.pyplot as plt
sns.countplot(x='attack_detected', data=df)
plt.title('Attack vs Normal')
plt.show()
```



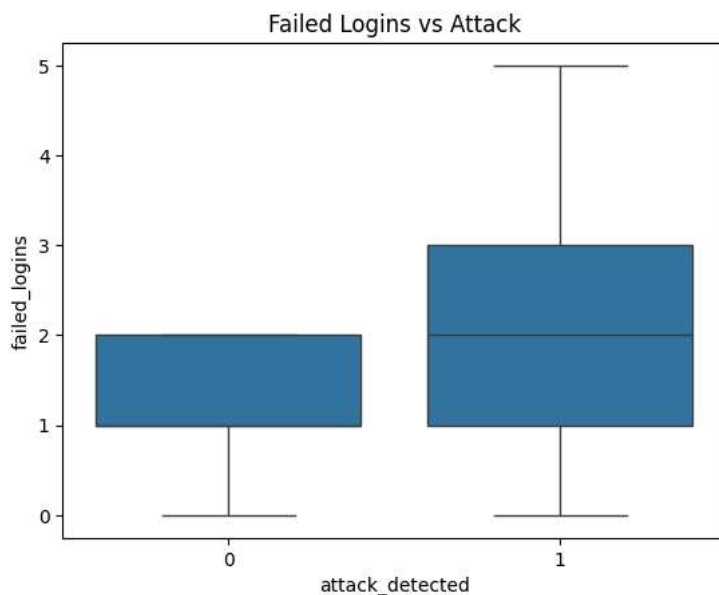
```
sns.heatmap(df.corr(), annot=True, fmt='.2f', cmap='coolwarm')
plt.title('Feature Correlation')
plt.show()
```



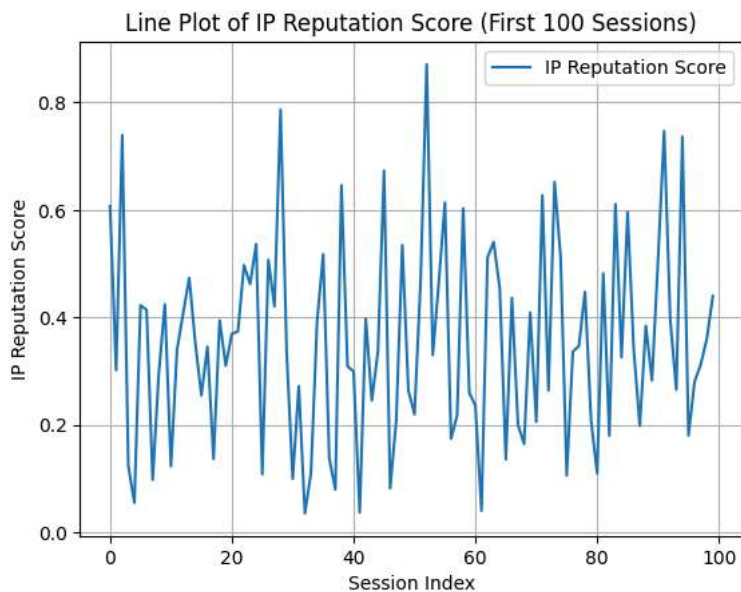
```
sns.boxplot(x='attack_detected', y='ip_reputation_score', data=df)
plt.title('IP Reputation vs Attack')
plt.show()
```



```
sns.boxplot(x='attack_detected', y='failed_logins', data=df)
plt.title('Failed Logins vs Attack')
plt.show()
```

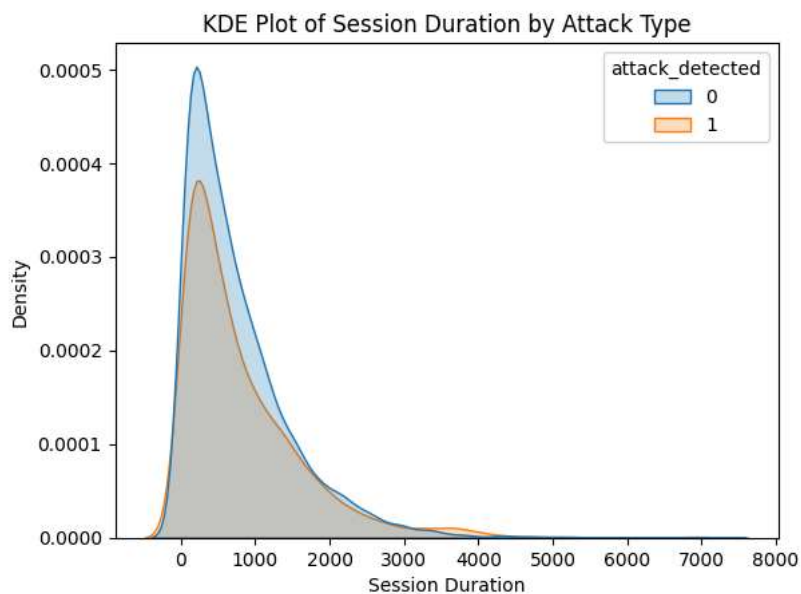


```
# Line plot for IP reputation scores of first 100 sessions
plt.plot(df['ip_reputation_score'][:100], label='IP Reputation Score')
plt.title('Line Plot of IP Reputation Score (First 100 Sessions)')
plt.xlabel('Session Index')
plt.ylabel('IP Reputation Score')
plt.legend()
plt.grid(True)
plt.show()
```



```
import seaborn as sns
import matplotlib.pyplot as plt
```

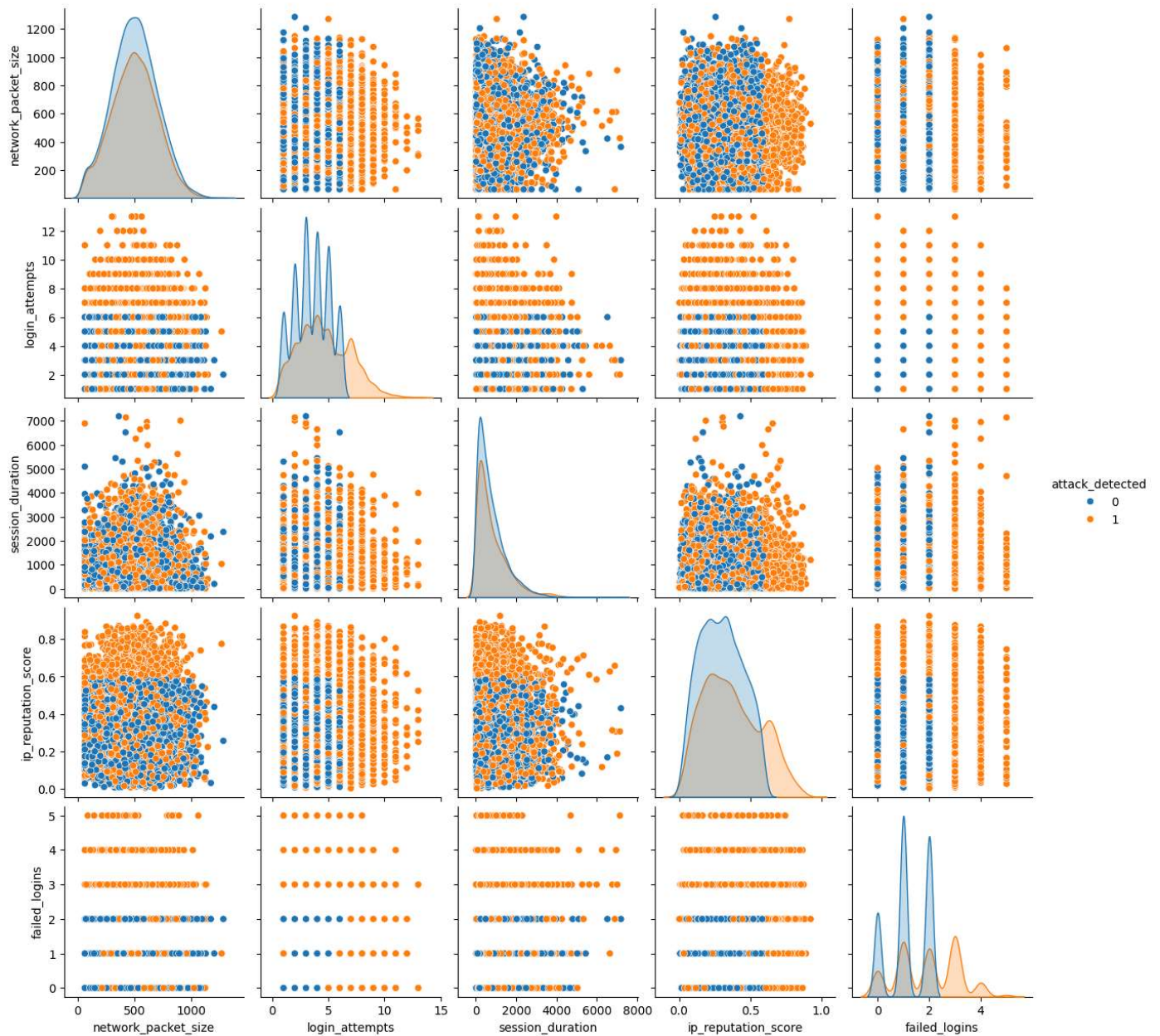
```
# KDE plot for session duration by attack class
sns.kdeplot(data=df, x='session_duration', hue='attack_detected', fill=True)
plt.title('KDE Plot of Session Duration by Attack Type')
plt.xlabel('Session Duration')
plt.ylabel('Density')
plt.show()
```



```
sns.pairplot(df[['network_packet_size', 'login_attempts', 'session_duration',
                 'ip_reputation_score', 'failed_logins', 'attack_detected']],
             hue='attack_detected')
plt.suptitle('Pairplot of Selected Features', y=1.02)
plt.show()
```



Pairplot of Selected Features



```
import matplotlib.pyplot as plt

plt.hist(df['session_duration'], bins=30, color='skyblue', edgecolor='black')
plt.title('Histogram of Session Duration')
plt.xlabel('Session Duration')
plt.ylabel('Frequency')
plt.grid(True)
plt.show()
```