```
import pandas as pd
df = pd.read_csv("//content/malware_classification_dataset_cleaned (2).csv")
df.head()
```



	file_name	family	obfuscation	api_calls_count	entropy	suspicious_api_calls	label
0	file_100177.exe	Worm	Metamorphism	336	4.87	RegSetValue	Malware
1	file_100425.doc	Ransomware	Packing	230	7.04	RegSetValue	Malware
2	file_100519.doc	Spyware	Metamorphism	148	5.72	ReadFile	Malware
3	file_100599.js	Ransomware	Polymorphism	19	4.00	WriteFile	Benign
4	file_100884.zip	Adware	Packing	462	4.27	WriteFile	Malware

import pandas as pd df = pd.read_csv("/content/malware_classification_dataset_cleaned (2).csv") df.info() df.describe()



<class 'pandas.core.frame.DataFrame'> RangeIndex: 4000 entries, 0 to 3999 Data columns (total 7 columns):

#	Column	Non-Null Count	Dtype			
0	file_name	4000 non-null	object			
1	family	4000 non-null	object			
2	obfuscation	3235 non-null	object			
3	api_calls_count	4000 non-null	int64			
4	entropy	4000 non-null	float64			
5	suspicious_api_calls	4000 non-null	object			
6	label	4000 non-null	object			
dtypes: fleet(4/1) int(4/1) object(5)						

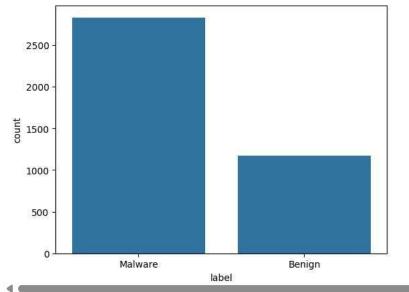
dtypes: float64(1), int64(1), object(5)

memory usage: 218.9+ KB

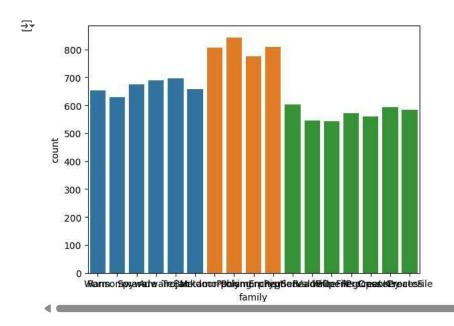
	api_calls_count	entropy
count	4000.000000	4000.000000
mean	253.657750	5.759140
std	142.572649	1.292099
min	10.000000	3.500000
25%	128.000000	4.630000
50%	250.000000	5.760000
75%	376.000000	6.880000
max	500.000000	8.000000

df.isnull().sum() df.duplicated().sum()

→ np.int64(0)

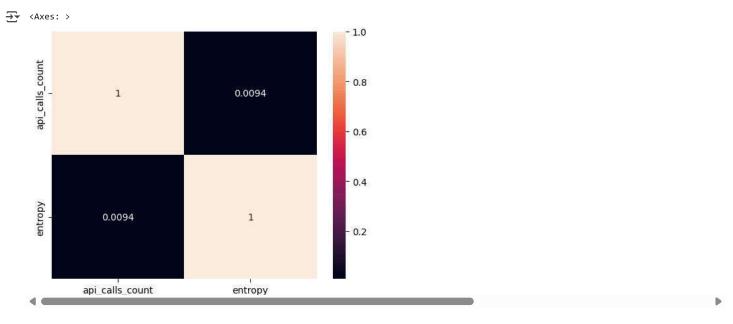
import seaborn as sns sns.countplot(x='label', data=df) 

for col in ['family', 'obfuscation', 'suspicious_api_calls']:
 sns.countplot(x=col, data=df)

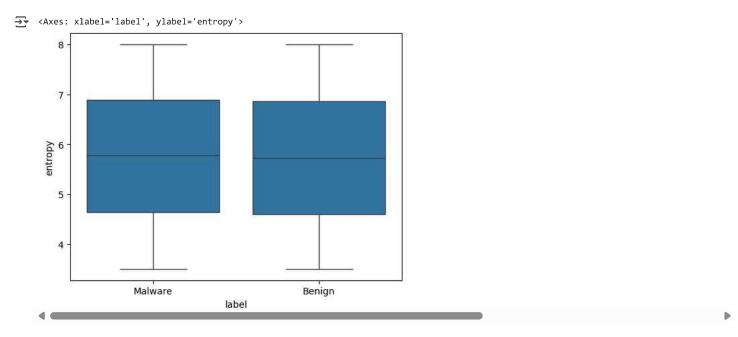


import seaborn as sns
import pandas as pd

numerical_features = df.select_dtypes(include=['number']).columns
sns.heatmap(df[numerical_features].corr(), annot=True)



sns.boxplot(x='label', y='entropy', data=df)



```
import pandas as pd
from sklearn.ensemble import RandomForestClassifier
from sklearn.model_selection import train_test_split
import seaborn as sns
from sklearn.preprocessing import LabelEncoder

X = df.drop('label', axis=1)
y = df['label']
encoder = LabelEncoder()

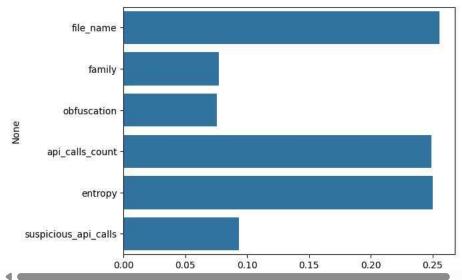
for col in X.select_dtypes(include=['object']).columns:
    X[col] = encoder.fit_transform(X[col])
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42) # 80% train, 20% test

model = RandomForestClassifier()
model.fit(X_train, y_train)
```

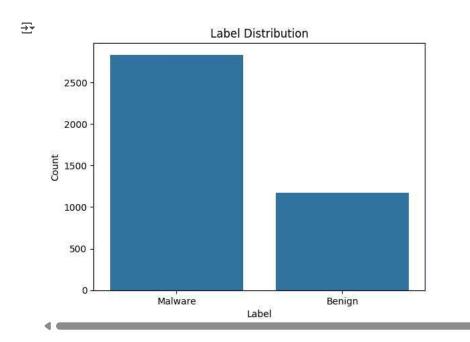
```
feature_importances = model.feature_importances_
feature_names = X_train.columns
```

sns.barplot(x=feature_importances, y=feature_names)

```
<Axes: ylabel='None'>
```



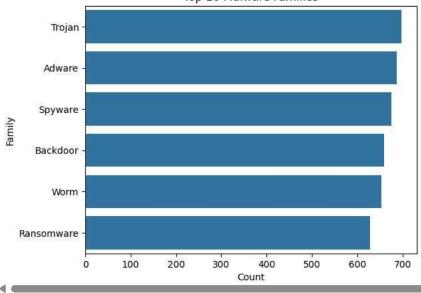
```
import seaborn as sns
import matplotlib.pyplot as plt
sns.countplot(x='label', data=df)
plt.title("Label Distribution")
plt.xlabel("Label")
plt.ylabel("Count")
plt.show()
```



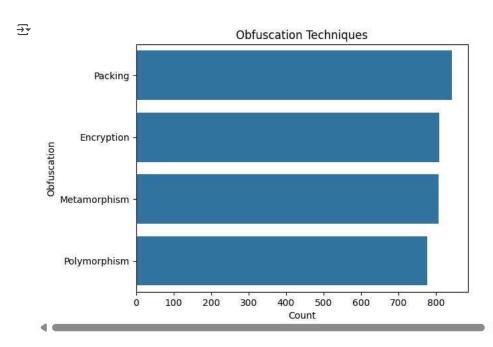
```
family_counts = df['family'].value_counts().head(10)
sns.barplot(x=family_counts.values, y=family_counts.index)
plt.title("Top 10 Malware Families")
plt.xlabel("Count")
plt.ylabel("Family")
plt.show()
```



Top 10 Malware Families



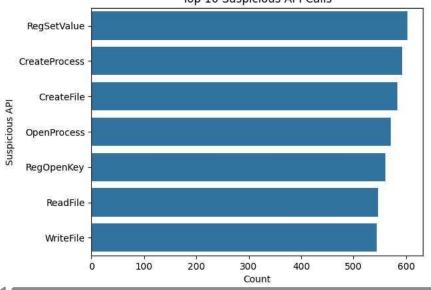
obfuscation_counts = df['obfuscation'].value_counts()
sns.countplot(y='obfuscation', data=df, order=obfuscation_counts.index)
plt.title("Obfuscation Techniques")
plt.xlabel("Count")
plt.ylabel("Obfuscation")
plt.show()



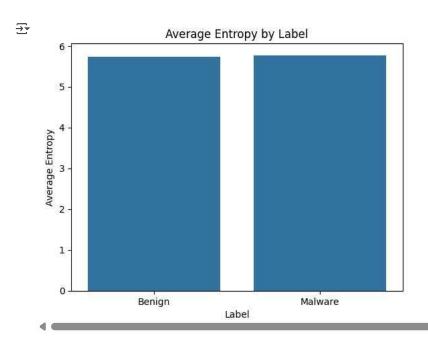
api_counts = df['suspicious_api_calls'].value_counts().head(10)
sns.barplot(x=api_counts.values, y=api_counts.index)
plt.title("Top 10 Suspicious API Calls")
plt.xlabel("Count")
plt.ylabel("Suspicious API")
plt.show()



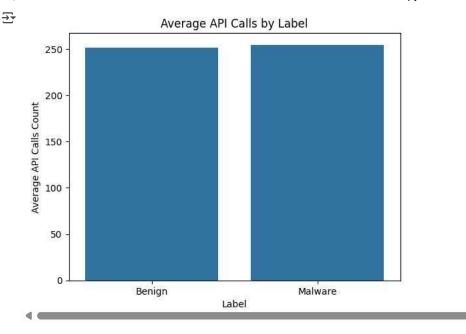
Top 10 Suspicious API Calls



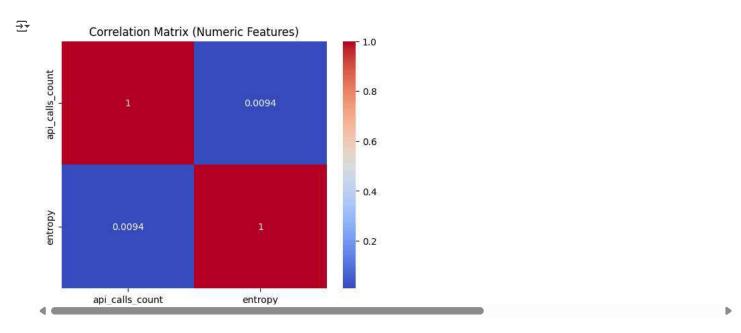
```
entropy_by_label = df.groupby('label')['entropy'].mean()
sns.barplot(x=entropy_by_label.index, y=entropy_by_label.values)
plt.title("Average Entropy by Label")
plt.xlabel("Label")
plt.ylabel("Average Entropy")
plt.show()
```



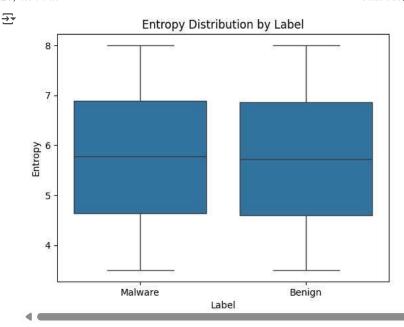
```
api_calls_by_label = df.groupby('label')['api_calls_count'].mean()
sns.barplot(x=api_calls_by_label.index, y=api_calls_by_label.values)
plt.title("Average API Calls by Label")
plt.xlabel("Label")
plt.ylabel("Average API Calls Count")
plt.show()
```



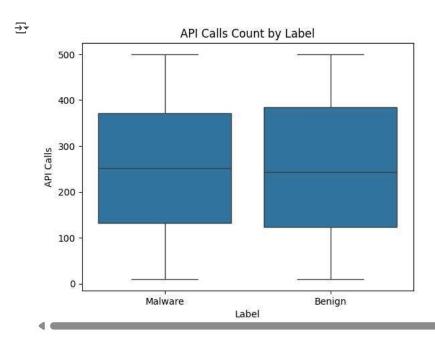
correlation_matrix = df[['api_calls_count', 'entropy']].corr()
sns.heatmap(correlation_matrix, annot=True, cmap="coolwarm")
plt.title("Correlation Matrix (Numeric Features)")
plt.show()



sns.boxplot(x='label', y='entropy', data=df)
plt.title("Entropy Distribution by Label")
plt.xlabel("Label")
plt.ylabel("Entropy")
plt.show()



```
sns.boxplot(x='label', y='api_calls_count', data=df)
plt.title("API Calls Count by Label")
plt.xlabel("Label")
plt.ylabel("API Calls")
plt.show()
```



```
from sklearn.ensemble import RandomForestClassifier
from sklearn.preprocessing import LabelEncoder

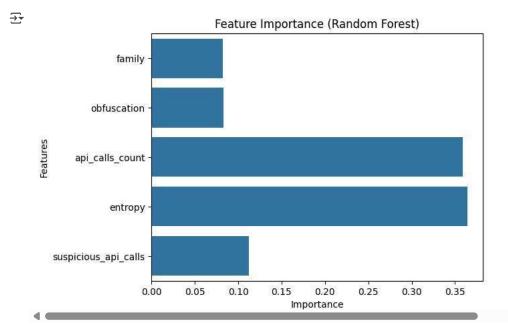
df_encoded = df.copy()
le = LabelEncoder()
for col in ['family', 'obfuscation', 'suspicious_api_calls', 'label']:
    df_encoded[col] = le.fit_transform(df_encoded[col])

X = df_encoded.drop(['file_name', 'label'], axis=1)
y = df_encoded['label']

model = RandomForestClassifier()
model.fit(X, y)

importances = model.feature_importances_
sns.barplot(x=importances, y=X.columns)
plt.title("Feature Importance (Random Forest)")
plt.xlabel("Importance")
```

```
plt.ylabel("Features")
plt.show()
```



import pandas as pd

Encryption

Metamorphism

246

221

563

586

```
class_grouped = df.groupby('label')[['entropy', 'api_calls_count']].mean()
print("1. Average Feature Values by Class:\n", class_grouped, "\n")
pivot_family_api = pd.pivot_table(df, values='api_calls_count',
                                  index='family',
                                  columns='label',
                                  aggfunc='mean')
print("2. Average API Calls per Family per Label:\n", pivot_family_api, "\n")
obf_label_freq = pd.pivot_table(df, values='file_name',
                                index='obfuscation',
                                columns='label',
                                aggfunc='count',
                                fill_value=0)
print("3. Frequency of Obfuscation Types by Label:\n", obf label freq, "\n")
pivot_entropy_api = pd.pivot_table(df, values='entropy',
                                   index='suspicious_api_calls',
                                   columns='label',
                                   aggfunc='mean')
print("4. Average Entropy by Suspicious API and Label:\n", pivot_entropy_api, "\n")
→ 1. Average Feature Values by Class:
                entropy api_calls_count
     label
     Benign
             5.728991
                             251.381523
     Malware 5.771590
                             254.597669
     2. Average API Calls per Family per Label:
      label
                      Benign
                                Malware
     family
     Adware
                 230.502793 252.734774
     Backdoor
                 248.135678 261.397826
     Ransomware
                250.528409 248.553097
                 261.882682 247.790323
     Spyware
                 258.513274 262.840764
     Trojan
                 256.342857 254.702032
     Worm
     3. Frequency of Obfuscation Types by Label:
      label
                    Benign Malware
     obfuscation
```

Packing

WriteFile

```
Polymorphism
                224
                         552
4. Average Entropy by Suspicious API and Label:
                        Benign Malware
label
suspicious_api_calls
CreateFile
                     5.738050 5.725412
CreateProcess
                     5.752626 5.650966
OpenProcess
                     5.882327 5.751845
ReadFile
                     5.555409 5.798372
RegOpenKey
                     5.589036 5.778883
RegSetValue
                     5.822623 5.842649
```

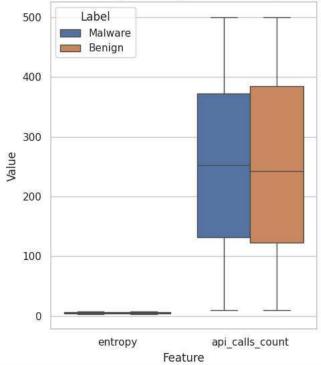
587

5.751220 5.862868

256



Statistical Analysis: Entropy and API Calls by File Type



```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score, roc_auc_score
import matplotlib.pyplot as plt
import seaborn as sns
```

```
df = pd.read_csv("malware_classification_dataset_cleaned (2).csv")
df_encoded = pd.get_dummies(df, columns=['obfuscation', 'family', 'suspicious_api_calls'])
X = df_encoded.drop(columns=['label', 'file_name'])
y = df_encoded['label'].map({'Benign': 0, 'Malware': 1})
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
model = RandomForestClassifier(n_estimators=100, random_state=42)
model.fit(X_train, y_train)
y_pred = model.predict(X test)
y_prob = model.predict_proba(X_test)[:, 1]
accuracy = accuracy_score(y_test, y_pred)
precision = precision_score(y_test, y_pred)
recall = recall_score(y_test, y_pred)
f1 = f1_score(y_test, y_pred)
roc_auc = roc_auc_score(y_test, y_prob)
print("Model Performance:")
print(f"Accuracy: {accuracy:.2f}")
print(f"Precision: {precision:.2f}")
print(f"Recall: {recall:.2f}")
print(f"F1 Score: {f1:.2f}")
print(f"ROC-AUC Score: {roc_auc:.2f}")
importances = model.feature_importances_
features = X.columns
importance_df = pd.DataFrame({'Feature': features, 'Importance': importances})
importance_df = importance_df.sort_values(by='Importance', ascending=False)
plt.figure(figsize=(6, 3))
sns.barplot(x='Importance', y='Feature', data=importance_df.head(10))
plt.title("Top 10 Feature Importances")
plt.xlabel("Importance")
plt.ylabel("Feature")
plt.tight_layout()
plt.show()

→ Model Performance:
     Accuracy: 0.64
     Precision: 0.69
     Recall: 0.87
     F1 Score: 0.77
     ROC-AUC Score: 0.48
                                               Top 10 Feature Importances
                                    entropy
                             api_calls_count
                 obfuscation Metamorphism
                        obfuscation Packing
                 obfuscation Polymorphism
                     obfuscation Encryption
               suspicious_api_calls_WriteFile
           suspicious api calls RegSetValue
         suspicious api calls CreateProcess
          suspicious api calls RegOpenKey
                                             0.0
                                                    0.1
                                                             0.2
                                                                     0.3
                                                        Importance
```

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

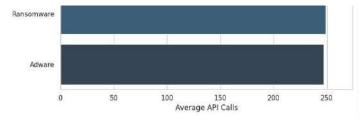
df = pd.read_csv("malware_classification_dataset_cleaned (2).csv")
```

```
sns.set(style="whitegrid")
plt.figure(figsize=(16, 24))
plt.subplot(3, 2, 1)
df_melted = df.melt(id_vars='label', value_vars=['entropy', 'api_calls_count'],
                    var_name='Feature', value_name='Value')
sns.boxplot(x='Feature', y='Value', hue='label', data=df_melted)
plt.title("Boxplot: Entropy & API Calls by Label")
plt.xlabel("Feature")
plt.ylabel("Value")
plt.subplot(3, 2, 2)
sns.histplot(data=df, x='entropy', hue='label', kde=True, bins=30, palette='Set2')
plt.title("Entropy Distribution")
plt.xlabel("Entropy")
plt.ylabel("Frequency")
plt.subplot(3, 2, 3)
sns.histplot(data=df, x='api_calls_count', hue='label', kde=True, bins=30, palette='Set1')
plt.title("API Call Count Distribution")
plt.xlabel("API Calls Count")
plt.ylabel("Frequency")
plt.subplot(3, 2, 4)
df corr = df.copy()
df_corr['label'] = df_corr['label'].map({'Benign': 0, 'Malware': 1})
corr_matrix = df_corr[['entropy', 'api_calls_count', 'label']].corr()
sns.heatmap(corr_matrix, annot=True, cmap='coolwarm', fmt=".2f")
plt.title("Correlation Heatmap")
plt.subplot(3, 2, 5)
label_counts = df['label'].value_counts()
plt.pie(label_counts, labels=label_counts.index, autopct='%1.1f%%', colors=['#66b3ff', '#ff9999'])
plt.title("Distribution of Malware vs Benign Files")
if 'family' in df.columns:
    plt.subplot(3, 2, 6)
    family_api = df.groupby('family')['api_calls_count'].mean().sort_values(ascending=False)
    \verb|sns.barplot(x=family_api.values, y=family_api.index, palette='Blues_d')|\\
    plt.title("Average API Call Count by Malware Family")
    plt.xlabel("Average API Calls")
    plt.ylabel("Family")
plt.tight_layout()
plt.show()
```

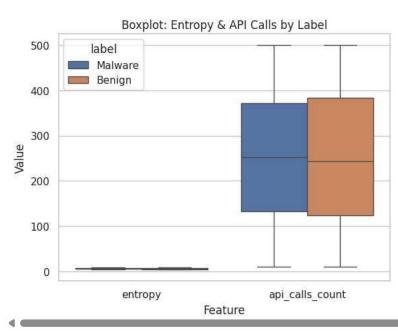
<ipython-input-40-3d741ec31307>:45: FutureWarning: Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `leg sns.barplot(x=family_api.values, y=family_api.index, palette='Blues_d') Boxplot: Entropy & API Calls by Label Entropy Distribution label label 500 Malware Malware Benign Benign 100 400 80 300 Frequency 60 Value 200 40 100 20 0 entropy api_calls_count Entropy Feature API Call Count Distribution Correlation Heatmap label 120 Malware ____ Benign 100 - 0.8 80 - 0.6 api calls count 60 0.4 40 0.2 label 20 100 200 300 400 label entropy api_calls_count API Calls Count Average API Call Count by Malware Family Distribution of Malware vs Benign Files Trojan Malware Backdoor 70.8% Worm Spyware 29.2%



Untitled3.ipynb - Colab







sns.histplot(data=df, x='entropy', hue='label', kde=True, bins=30, palette='Set2')
plt.title("Entropy Distribution")
plt.show()