```
!pip install pyspark
!pip install tensorflowonspark
     Requirement already satisfied: pyspark in /usr/local/lib/python3.10/dist-packages (3.5.1
     Requirement already satisfied: py4j==0.10.9.7 in /usr/local/lib/python3.10/dist-packages
     Requirement already satisfied: tensorflowonspark in /usr/local/lib/python3.10/dist-packa
# Import necessary libraries
from pyspark.sql import SparkSession
from pyspark.context import SparkContext
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
import numpy as np
import tensorflowonspark as tfos
from tensorflowonspark import TFCluster
from pyspark.ml.feature import VectorAssembler
from pyspark.ml.feature import VectorSlicer
from pyspark.ml.classification import RandomForestClassifier
from pyspark.ml.evaluation import MulticlassClassificationEvaluator
from pyspark.ml.evaluation import BinaryClassificationEvaluator
from pyspark.context import SparkContext
from pyspark.sql.functions import count, when, isnull
import tensorflow as tf
import tensorflowonspark as tfos
import pyspark.sql.functions as F
from pyspark.sql.functions import col
import matplotlib.pyplot as plt
import seaborn as sns
import pandas as pd
# Create SparkSession
spark = SparkSession.builder \
    .appName("MalwareDetection") \
    .getOrCreate()
# Access SparkContext from SparkSession
sc = spark.sparkContext
# Load the dataset into a Spark DataFrame
dataset path = "Android Malware Detection.csv"
df = spark.read.csv(dataset path, header=True, inferSchema=True)
# Display the schema of the DataFrame
df.printSchema()
```

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++	++	+
0 0.0	0.0	0.0
1 0.0	0.0	0.0
2 0.6	0.0	0.0
3 0.6	0.0	0.0
4 0.0	0.0	0.0
5 0.6	0.0	0.0
6 0.6	0.0	0.0
7 0.6	0.0	0.0
8 0.6	0.0	0.0
9 0.6	0.0	0.0
10 0.6	0.0	0.0
11 0.6	0.0	0.0
12 0.6	0.0	0.0
13 0.0	0.0	0.0
14 0.6	0.0	0.0
15 0.6	0.0	0.0
16 0.6	0.0	0.0
17 0.6	0.0	0.0
18 0.6	0.0	0.0
19 0.6	0.0	0.0
++	++	+

only showing top 20 rows

```
# Count the number of rows in the dataset
row_count = df.count()
print("Number of rows in the dataset:", row_count)
```

Number of rows in the dataset: 4863

```
# Summary statistics
print("Summary Statistics:")
df.describe().show()
```

Summary Statistics:

+	summary	_c0	ACCESS_ALL_DOWNLOADS	ACCESS_CACHE_FILESYSTEM	ACCESS_CHECKIN_
i	count		'	4862	
	mean		8.227067050596463E-4	0.001234060057589	0.00452488
	stddev	1403.9715096824436	0.028674012029843828	0.03511111945893242	0.0671218
	min	0	0.0	0.0	
	max	4862	1.0	1.0	

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17	0.0	0.0	0.0
18	0.0	0.0	0.0
19	0.0	0.0	0.0
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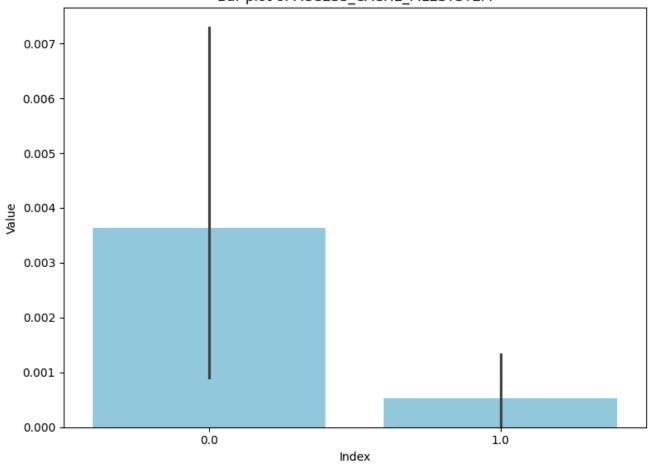
only showing top 20 rows

drop unnecessary columns

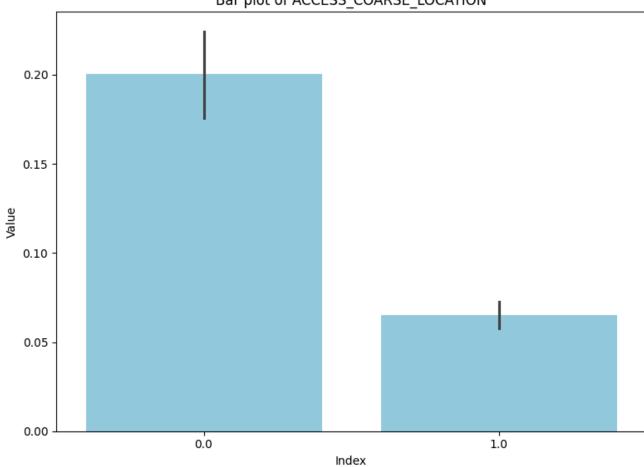
```
df = df_cleaned.drop('_c0')
# display new df
df.show()
     ACCESS_ALL_DOWNLOADS ACCESS_CACHE_FILESYSTEM ACCESS_CHECKIN_PROPERTIES ACCESS_COARSE_LC
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```

```
# Plot bar plots for selected features
selected_features = ["ACCESS_ALL_DOWNLOADS", "ACCESS_CACHE_FILESYSTEM", "ACCESS_COARSE_LOCAT
for feature in selected_features:
    plt.figure(figsize=(8, 6))
    sns.barplot(x=df.select('Label').rdd.flatMap(lambda x: x).collect(), y=df.select(feature
    plt.title(f'Bar plot of {feature}')
    plt.xlabel('Index')
    plt.ylabel('Value')
    plt.tight_layout()
    plt.show()
```



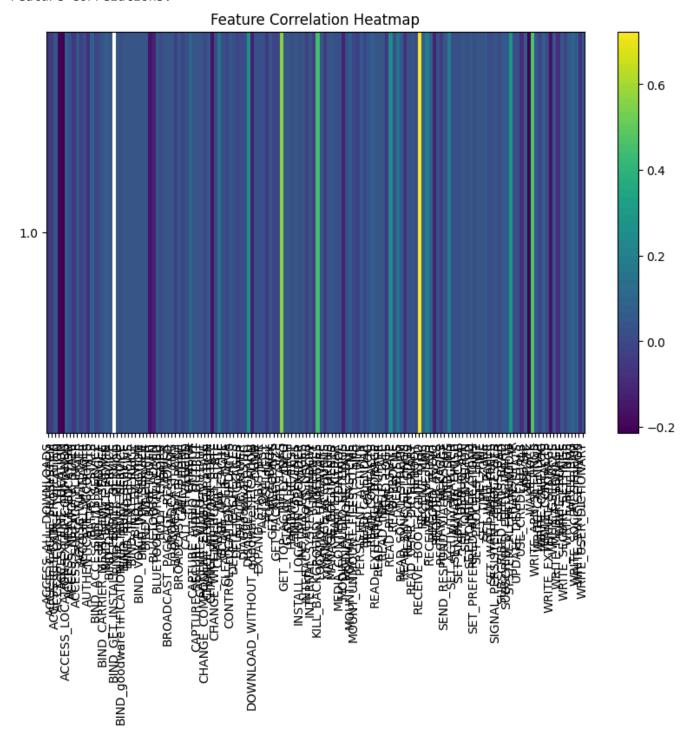






```
# Feature correlations
print("Feature Correlations:")
correlation_matrix = df.select([F.corr(col, 'Label').alias(col) for col in df.columns]).toPa
plt.figure(figsize=(10, 6))
plt.imshow(correlation_matrix, cmap='viridis', interpolation='nearest', aspect='auto')
plt.colorbar()
plt.xticks(range(len(correlation_matrix.columns)), correlation_matrix.columns, rotation='ver
plt.yticks(range(len(correlation_matrix.index)), correlation_matrix.index)
plt.title('Feature Correlation Heatmap')
plt.show()
```

Feature Correlations:

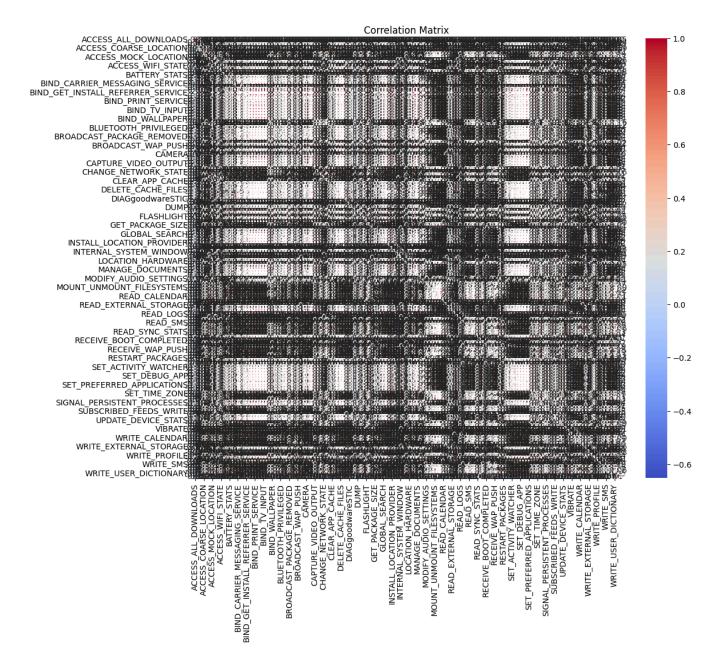


```
+----+
|Label|count|
+----+
| 0.0| 1098|
| 1.0| 3764|
+----+
```

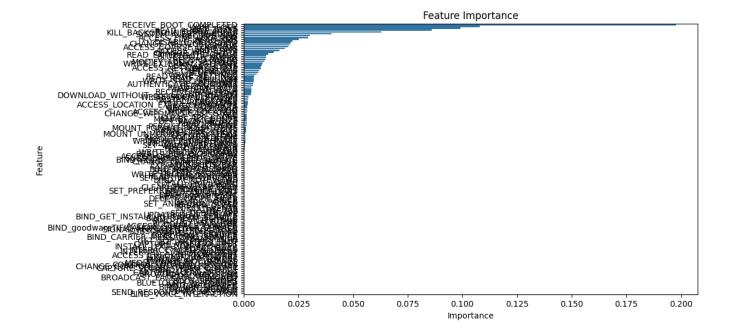
```
# Visualize the class distribution
class_distribution_pd = class_distribution.toPandas()
plt.figure(figsize=(6, 4))
sns.barplot(x="Label", y="count", data=class_distribution_pd)
plt.title("Class Distribution")
plt.xlabel("Label")
plt.ylabel("Count")
plt.show()
```

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```
# Correlation Analysis
correlation_matrix = df.drop("_c0").toPandas().corr()
plt.figure(figsize=(12, 10))
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt=".2f")
plt.title("Correlation Matrix")
plt.show()
```



```
# Assemble features
feature_cols = df.columns[1:-1] # Exclude _c0 and Label columns
assembler = VectorAssembler(inputCols=feature_cols, outputCol="features")
data_assembled = assembler.transform(df)
# Train a RandomForestClassifier
rf = RandomForestClassifier(labelCol="Label", featuresCol="features", numTrees=100)
model = rf.fit(data_assembled)
# Get feature importances
feature_importances = model.featureImportances.toArray()
# Visualize feature importances
feature_importance_pd = pd.DataFrame({"Feature": feature_cols, "Importance": feature_importa
feature_importance_pd_sorted = feature_importance_pd.sort_values(by="Importance", ascending=
plt.figure(figsize=(10, 6))
sns.barplot(x="Importance", y="Feature", data=feature_importance_pd_sorted)
plt.title("Feature Importance")
plt.xlabel("Importance")
plt.ylabel("Feature")
plt.show()
```



```
# Get the indices of the top 20 features based on their importances
top_20_indices = feature_importances.argsort()[-20:][::-1]

# Select only the top 20 features from the feature vector
slicer = VectorSlicer(inputCol="features", outputCol="selected_features", indices=top_20_inc
data_selected = slicer.transform(data_assembled)

# Train the RandomForestClassifier model using only the top 20 features
rf_selected = RandomForestClassifier(labelCol="Label", featuresCol="selected_features", num1
model_selected = rf_selected.fit(data_selected)
```

```
# Split the data into training and testing sets
train_data_selected, test_data_selected = data_selected.randomSplit([0.8, 0.2], seed=42)
```