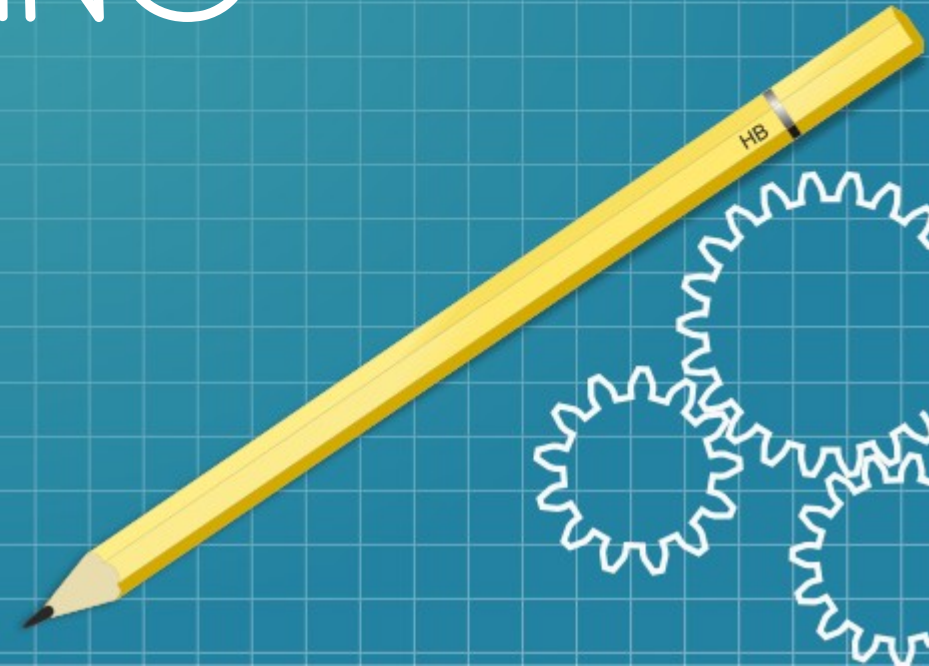


NETWORK ANALYSIS USING MACHINE LEARNING



A yellow pencil and a pink eraser are positioned in the top right corner of the white paper.

INTRODUCTION

REQUIREMENT

GRAPH EXPLORATORY ANALYSIS

MODEL DESCRIPTION & MODEL INTUTION

IMPLEMENTATION STEPS

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INTRODUCTION



Situation | Problem Definition : This project is all about predicting the types of a new protocol with the help of given input data set that contains all most 84 different features like packet length, flow duration, segment size, header length and many more.

Task: We were assigned with the task to come up with an efficient machine learning model which will result in maximum accuracy on a given data set.

Action: To do this work in an organised way, we enlisted all the relevant classification machine learning algorithm on a paper which fits the problem like SVM kernel, Decision Tree and Random forest . After that we started to analyse the performance of all those models meticulously.

Result: Eventually we got maximum accuracy of 98% on both training set and test set with help of random forest.

REQUIREMENT

Technology

- **Python**
- **Machine Learning**
- **Data Science**

Libraries

- **Numpy**
- **Pandas**
- **Sklern**
- **Matplotlib**
- **Seaborn**

Software Used

- **Notepad++**
- **Microsoft Office Word**
- **Google Colab**
- **Window**
- **Ubuntu**



GRAPH EXPLORATORY ANALYSIS



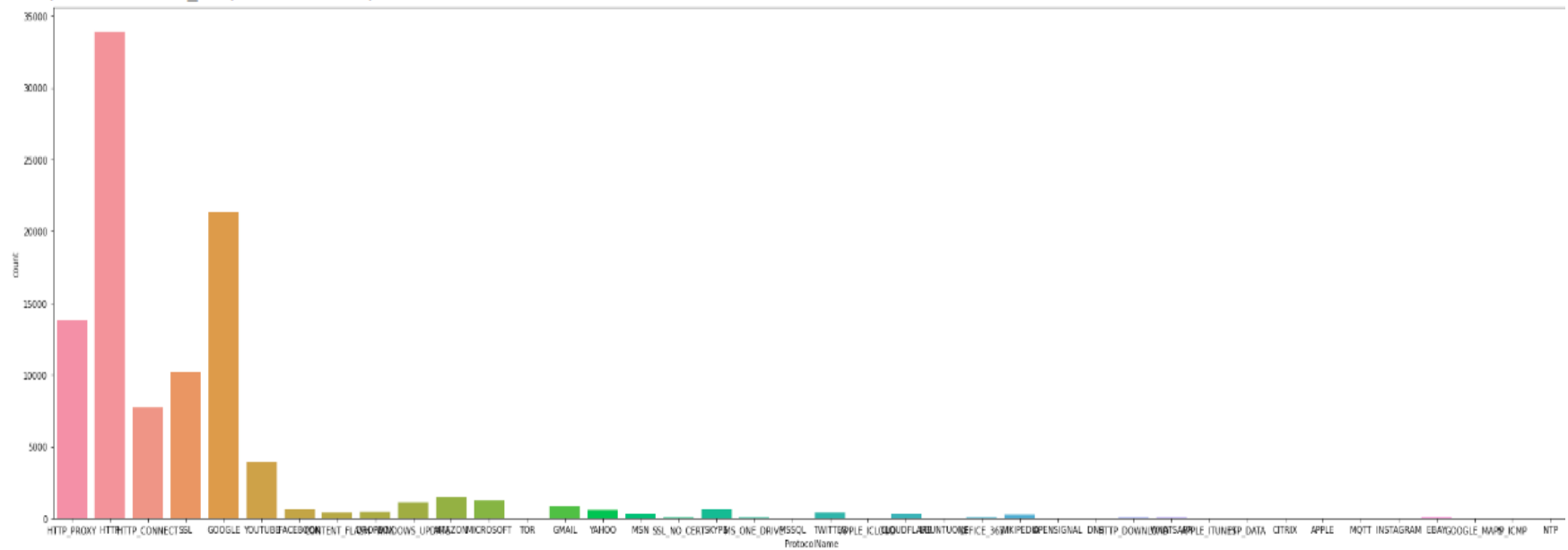
- It's also known as Exploratory Data analysis (EDA).
- It refers to the critical process of performing Initial investigations on data.
- It's used to discover patterns, to spot anomalies, to test hypothesis and to check assumptions with the help of summary statistics and graphical representations.
- Improves understanding of variables by extracting averages, mean, min. and max Values etc.
- Discover errors, outliers, and missing values in the data.
- Identify pattern by Visualizing data in graphs such as box plots, scatter plots, and histograms.



- Count plot of ProtocolName or Labels

```
plt.figure(figsize=(35, 10))  
sns.countplot(x="ProtocolName", data=df1)
```

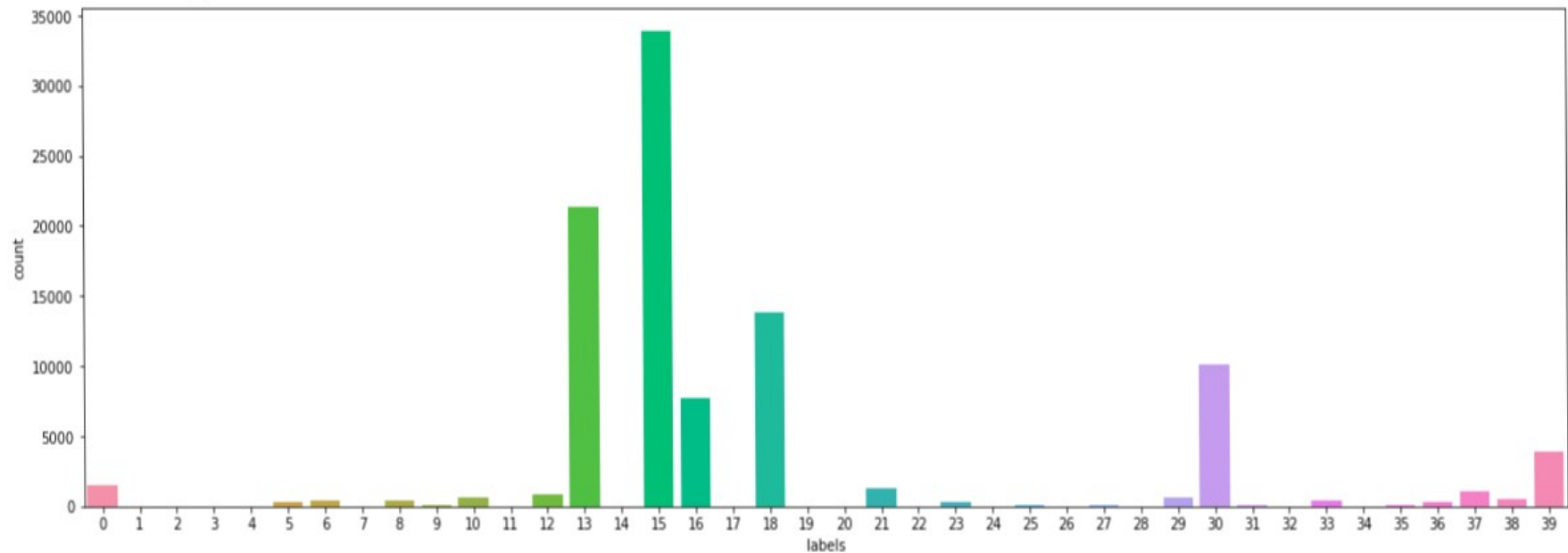
<matplotlib.axes._subplots.AxesSubplot at 0x7f522c131ad0>



- Label - Protocol Name

```
plt.figure(figsize=(20, 6))  
sns.countplot(x='labels', data=df1)
```

<matplotlib.axes._subplots.AxesSubplot at 0x7f522b2a1cd0>



- 15 – HTTP
- 13-Google
- 18-HTTP_Proxy
- 30-SSC
- 39-HTTP_Connect
- 0-Youtube

- Attribute)
- 
- A yellow pencil with a pink eraser is positioned diagonally across the frame. The pencil is sharpened and has a black eraser at the top. The eraser is a light pink color. The background is a blue grid pattern. The word 'Attribute)' is written in a black, sans-serif font on the left side of the image.

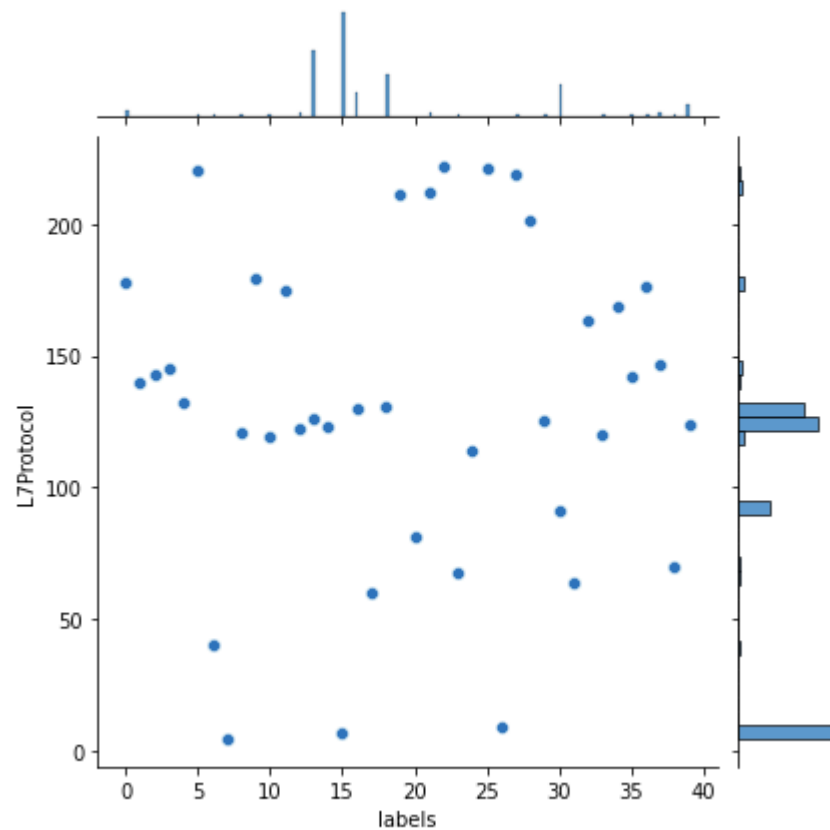
| | |
|-------------------------|----------|
| Init_Win_bytes_backward | 0.031210 |
|-------------------------|----------|



- Joint plot between labels and L7Protocol

```
sns.jointplot(x='labels', y="L7Protocol", data=df2)
```

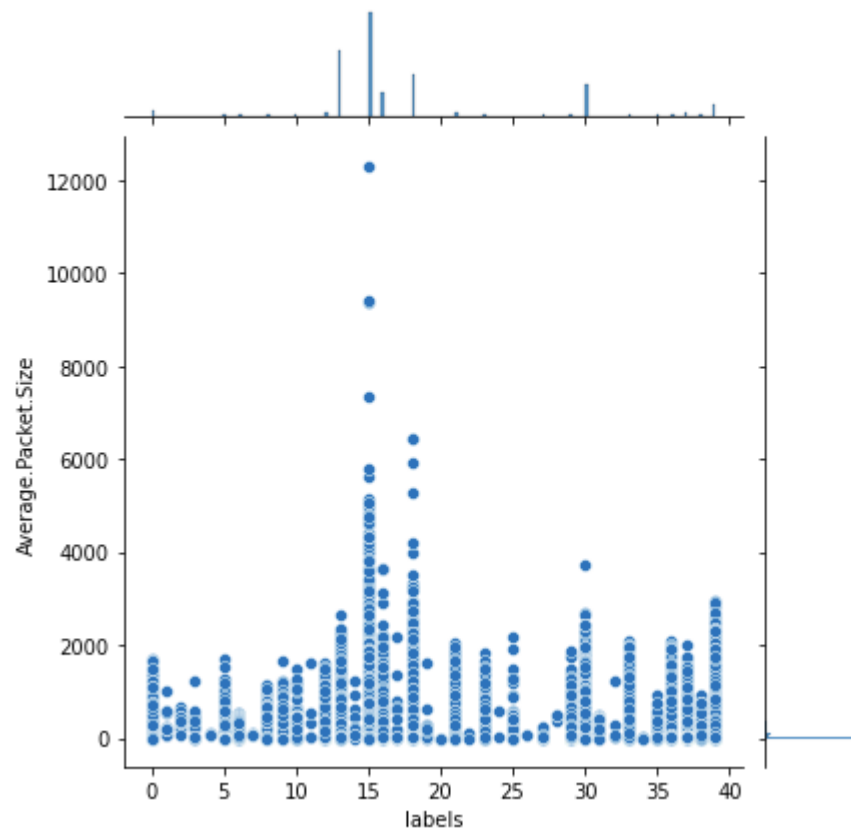
```
<seaborn.axisgrid.JointGrid at 0x7f522aa6f310>  
<Figure size 720x432 with 0 Axes>
```



- Joint plot between labels vs avg.packet.size

```
sns.jointplot(x="labels", y="Average.Packet.Size", data=df2)
```

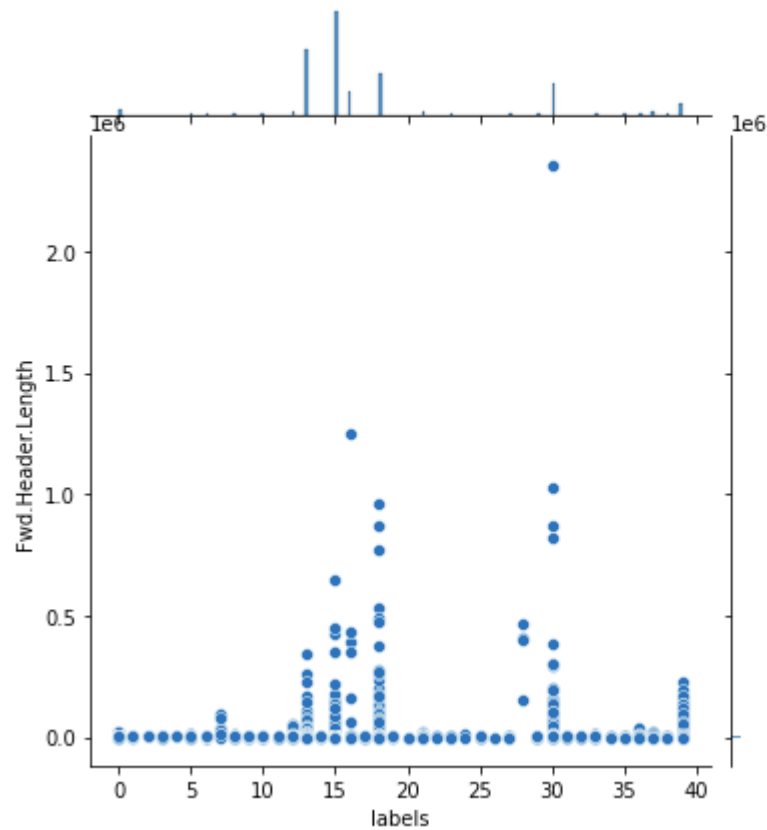
```
<seaborn.axisgrid.JointGrid at 0x7f522addf250>
```



- Joint plot between labels and forward.header.length

```
sns.jointplot(x='labels', y='Fwd.Header.Length', data=df2)
```

```
<seaborn.axisgrid.JointGrid at 0x7fc5c2e258d0>
```



MODEL DESCRIPTION (RANDOM FOREST)



- **What is random forest?**

Random forest is a supervised learning algorithm. The "forest" it builds, is an ensemble of decision trees, usually trained with the "bagging" method. The general idea of the bagging method is that a combination of learning models increases the overall result.

- **How Random Forest Works ?**

Random forest builds multiple decision trees and merges them together to get a more accurate and stable prediction

MODEL INTUITION (RANDOM FOREST)



STEP 1: Pick at random K data points from the Training set.



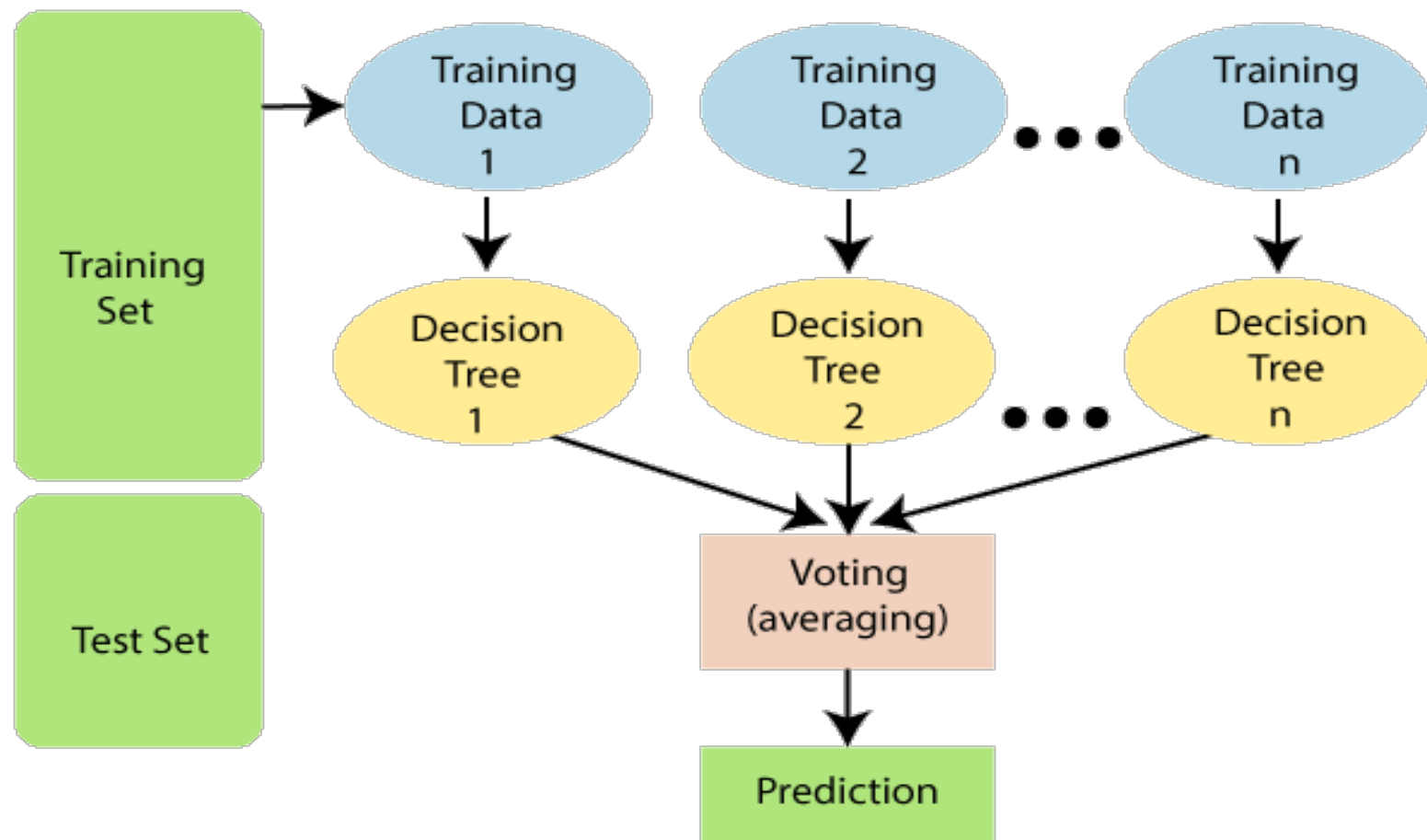
STEP 2: Build the Decision Tree associated to these K data points.



STEP 3: Choose the number N_{tree} of trees you want to build and repeat STEPS 1 & 2



STEP 4: For a new data point, make each one of your N_{tree} trees predict the category to which the data points belongs, and assign the new data point to the category that wins the majority vote.



IMPLEMENTATION STEPS



- **Importing The Libraries**
- **Importing The Dataset**
- **Splitting The Dataset Into Training Set and Test Set**
- **Removing non-integer column and Encoding The String Data Column Into Integer Column**
- **Feature Scaling**
- **Training The Random Forest Classification Model On Training Dataset**
- **Predicting A New Result**
- **Predicting The Test Set Result**
- **Building And Plotting The Confusion Matrix**
- **Data Visualisation**



CONCLUSION



- We have successfully implemented the random forest machine learning model on a given network data set. Which predict the types of protocol with an accuracy of 98 percentage.
- **PREVIEW** [CLICK ON PREVIEW TO SEE IMPLEMENTATION](#)





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