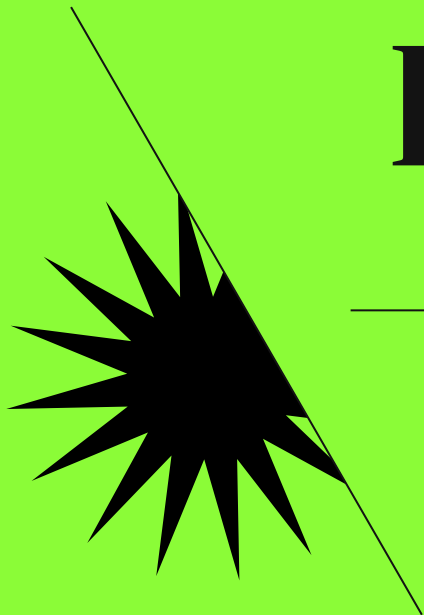


# Recipe Site Traffic



Zean Ni Zhehan



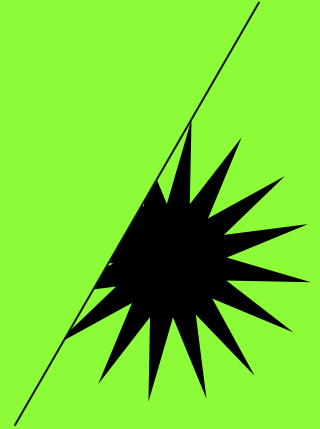
# Contents Table

1. Background
2. Goals
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5. Conclusion & Recommendation

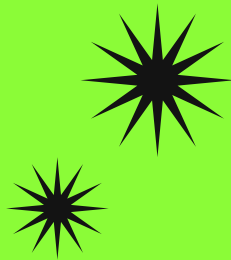




# Background



## Background



01

### Online recipes

Tasty Bytes, an online recipe startup features new recipes on their homepage website every day

02

### Traffic boost

On days that they feature a popular recipe, traffic increases by as much as 40%

03

### Challenge

However, it is difficult to predict ahead of time which recipes have high traffic

04

### Traffic

A typical binary classification problem

Predict whether a recipe will receive a high traffic based on the data collected from previously published recipes.

The criterion for success was the correct classification of **75%** of the recipes.



## **Project Goals**



# Validation & EDA

---

```
RangeIndex: 947 entries, 0 to 946
```

```
Data columns (total 8 columns):
```

#	Column	Non-Null Count	Dtype
0	recipe	947 non-null	int64
1	calories	895 non-null	float64
2	carbohydrate	895 non-null	float64
3	sugar	895 non-null	float64
4	protein	895 non-null	float64
5	category	947 non-null	object
6	servings	947 non-null	object
7	high_traffic	947 non-null	int64

```
dtypes: float64(4), int64(2), object(2)
```

```
memory usage: 59.3+ KB
```

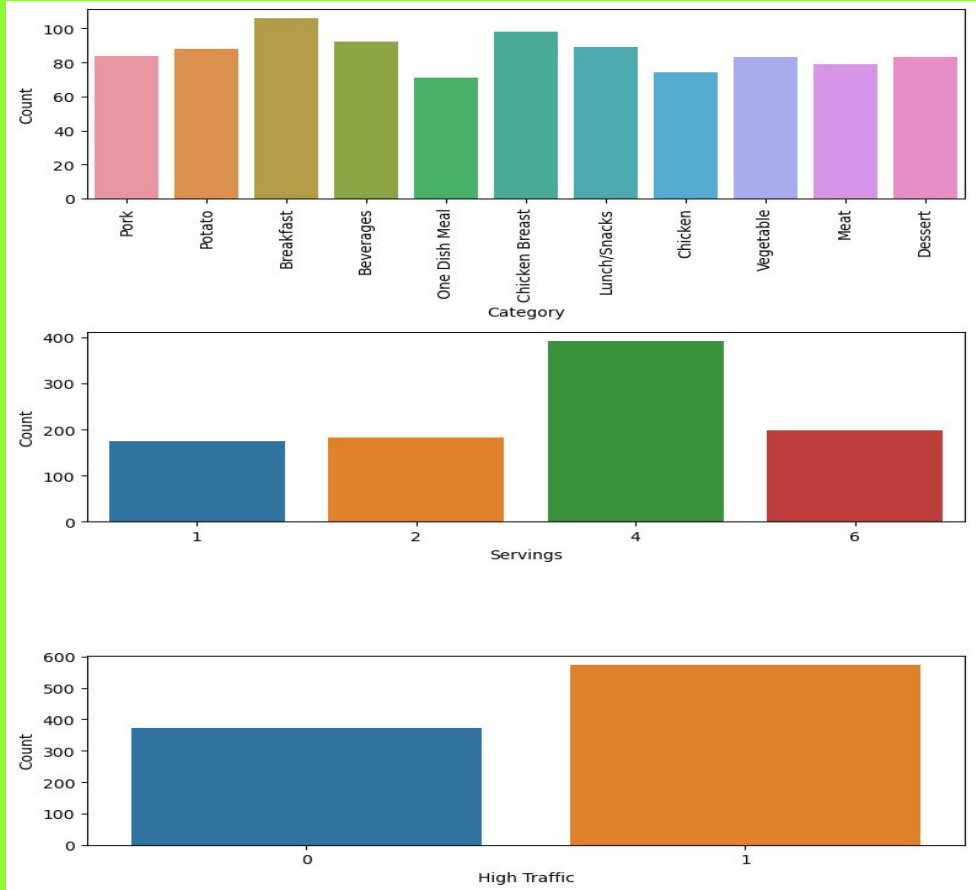
```
No of missing values in calories is: 52
```

```
No of missing values in carbohydrate is: 52
```

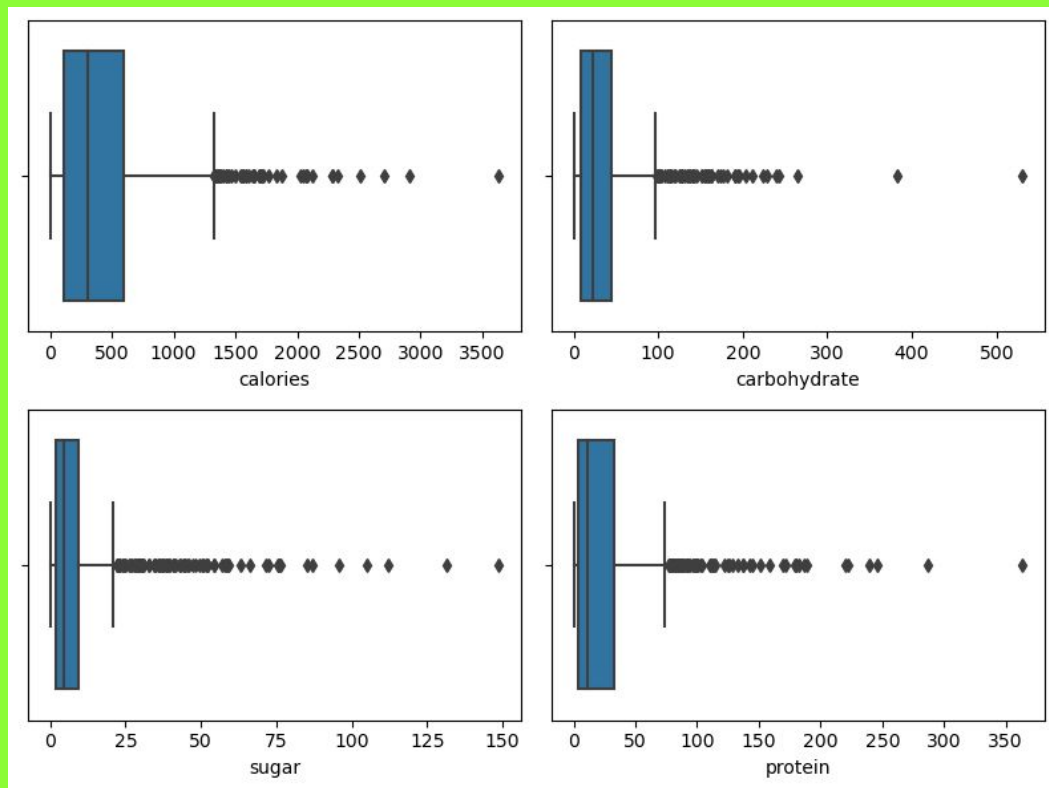
```
No of missing values in sugar is: 52
```

```
No of missing values in protein is: 52
```

# Validation & EDA

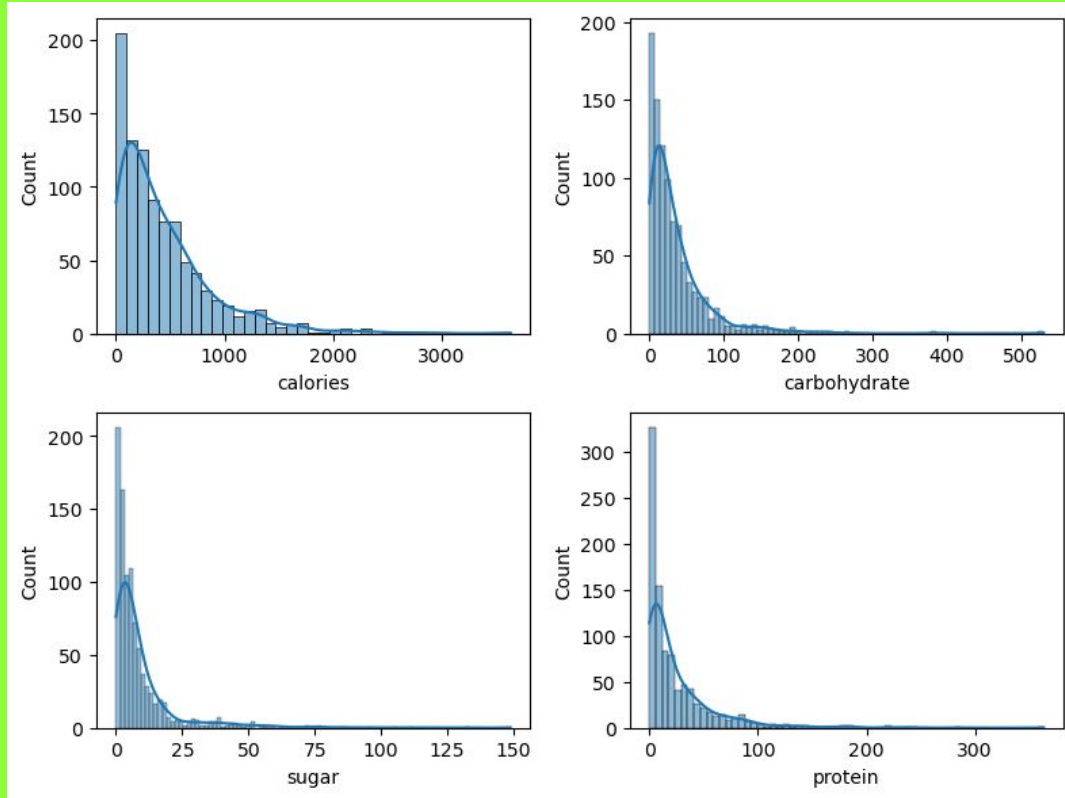


# Validation & EDA



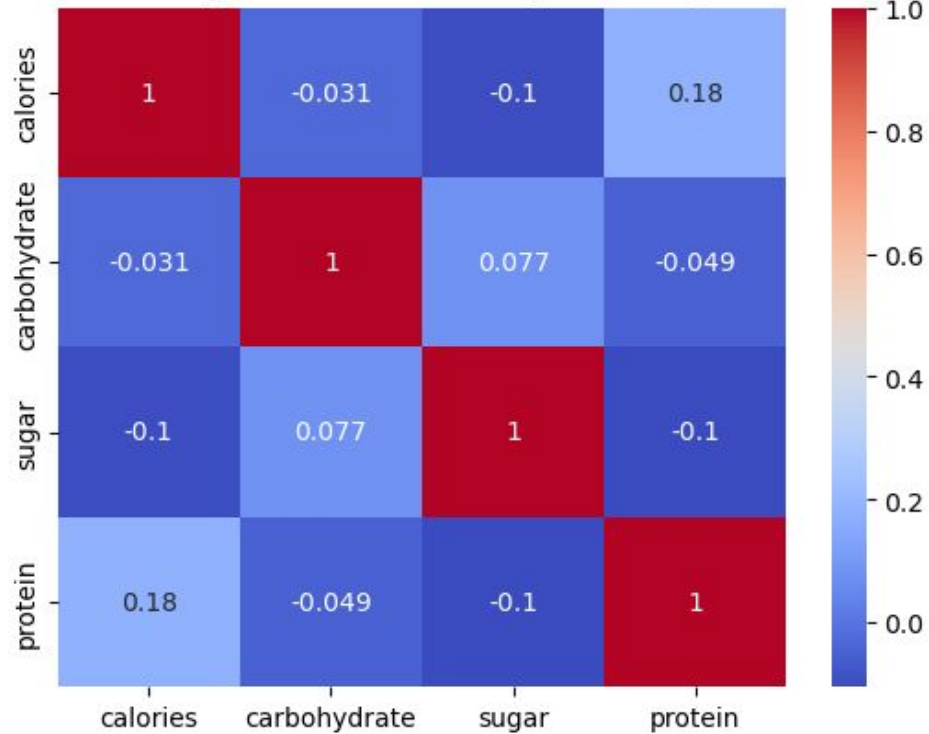


# Validation & EDA

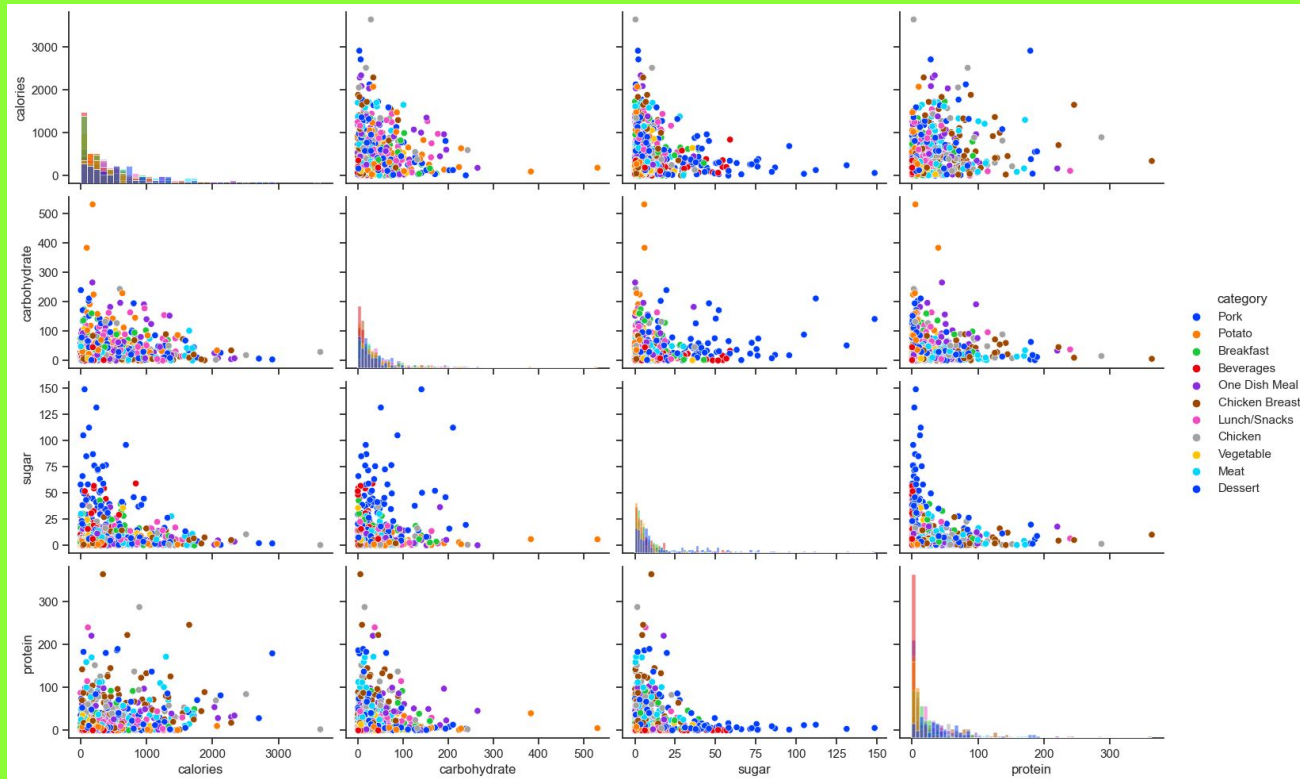


# Validation & EDA

Correlation Heatmap of Calories, Carbohydrate, Sugar, and Protein



# Validation & EDA



# Statistical Test

calories:  $t = -2.29$ ,  $p = 0.0225$

carbohydrate:  $t = -2.42$ ,  $p = 0.0156$

sugar:  $t = 2.22$ ,  $p = 0.0269$

protein:  $t = -1.35$ ,  $p = 0.1761$

# Statistical Test

```
contingency_table = pd.crosstab(df['category'], df['high_traffic'])  
chi2, p, dof, expected = chi2_contingency(contingency_table)  
chi2, p
```

```
(320.22296286253834, 8.182067546493786e-63)
```

```
contingency_table = pd.crosstab(df['servings'], df['high_traffic'])  
chi2, p, dof, expected = chi2_contingency(contingency_table)  
chi2, p
```

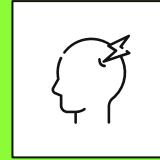
```
(2.7369889309788054, 0.4339779666711946)
```



# Feature Engineering



**health?**



**Complex?**



# Modeling



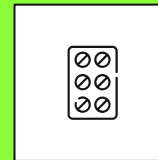
## Logistic Reg

```
confussion matrix: [[ 45  32]
 [ 11 102]]
accuracy:  0.7736842105263158
precision:  0.7611940298507462
recall:  0.9026548672566371
f1:  0.8259109311740891
```



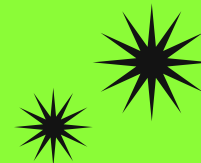
## Random Forest

```
confussion matrix: [[ 45  32]
 [ 11 102]]
accuracy:  0.7736842105263158
precision:  0.7611940298507462
recall:  0.9026548672566371
f1:  0.8259109311740891
```



## DNN

```
confussion matrix: [[42 35]
 [16 97]]
accuracy:  0.7315789473684211
precision:  0.7348484848484849
recall:  0.8584070796460177
f1:  0.7918367346938775
```





# Thanks!

