

# GROUP 15

## ML MODEL TO PREDICT TRAFFIC FLOW



# INTRODUCTION

## Machine Learning Goal :

To build a regression model that predicts traffic volume at a specific junction (Cross\_6) as our Target Variable using data from surrounding intersections (Cross\_1 to Cross\_5) and additional features such as weekday and time.

The model will help traffic authorities or smart city planners anticipate congestion and easily manage road usage.

**Dataset:** We used a Dataset from <https://zenodo.org/records/3653880> and manipulated it to add some missing values.

## Github

<https://github.com/Oweci/GROUP-10-REGRESSION-ML-MODEL/edit/main/README.md>

Our data has 11,519 rows and 8 columns

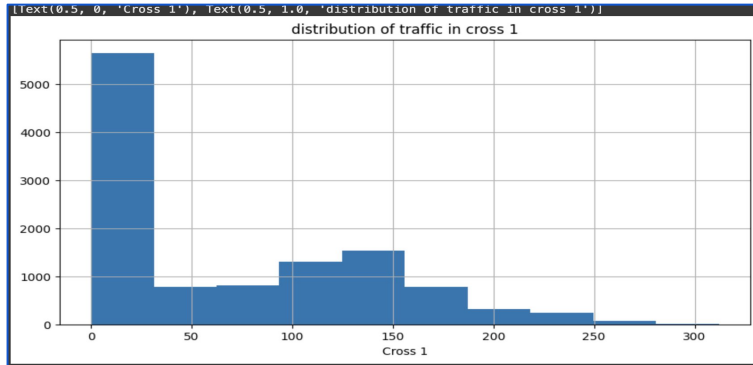
# DATASET DESCRIPTION

- Data represents traffic patterns across different junctions, time and days.
- Each row corresponds to traffic readings collected at one time interval.
- Traffic patterns vary depending on day of week and time.
- Tabular database (Numerical and Categorical)
- **Task** : Regression
- Our data has 11,519 rows and 8 columns

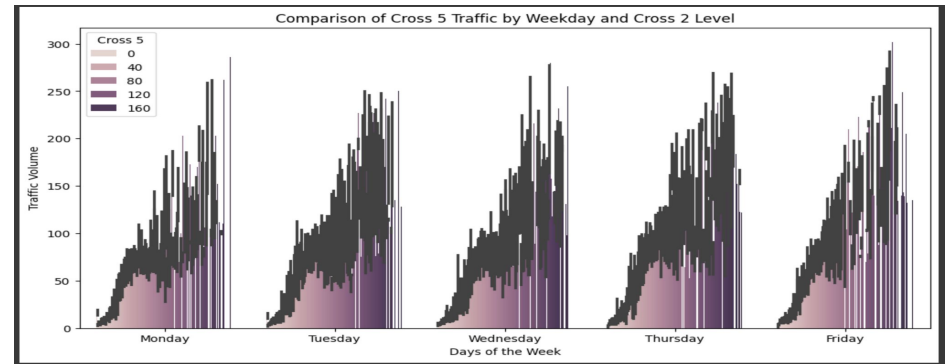
Columns	Description
Timestamp	Time durations
Weekday	Days if the week
Cross 1 to Cross 5	Cross roads (junctions)
Cross 6	Target Variable (Trying to predict the volume of cars )

# EDA and VISUALIZATION (Salha & Sisco)

1. **Univariate Visualization** : Histogram
2. **Multivariate Visualization** : Grouped Bar chart



The histogram of Cross 1 shows that the majority of traffic counts are higher on the end, The road at Cross 1 is mostly busy. High traffic volume is the normal condition, and low traffic volume is the rare condition.



The multivariate bar chart shows how traffic at Cross 2 varies across the days of the week, while also comparing it with cross 5. We then use the colors inside each bar to show how the traffic volume at Cross 2 is affected by the traffic levels at a different intersection, Cross 5.

# PREPROCESSING THE DATA (SISCO)

Used pipeline

```
#creating transformers
```

```
num_transformer = Pipeline(steps = [  
    ('imputer', numerical_meadian),  
    ('scaler', scaler)  
])
```

```
ordinal_transformer=Pipeline(  
    steps=[('ord_enc',ordinal_encoder)]  
)
```

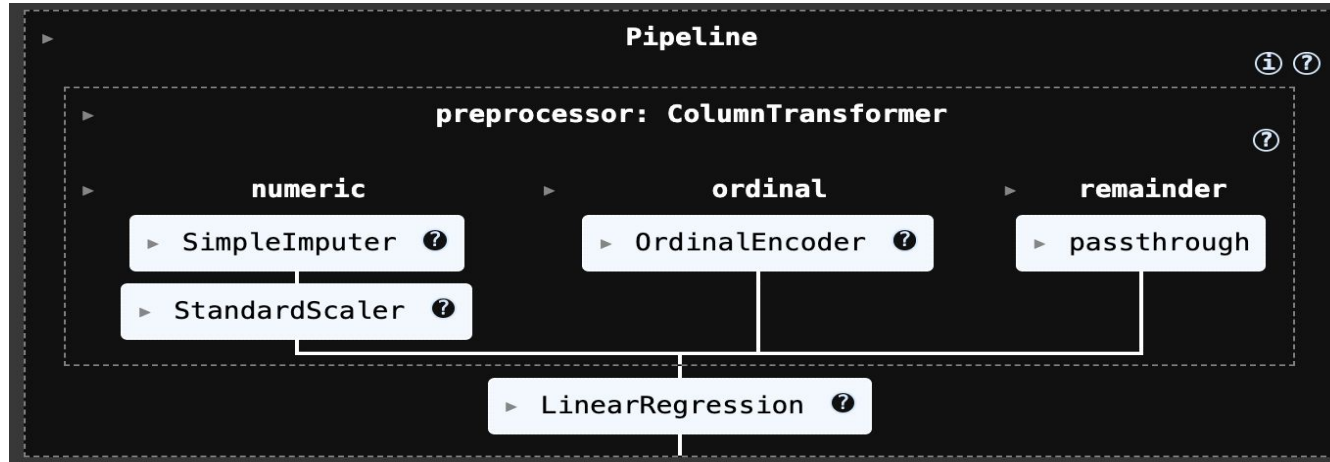
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```
#combining transformers
```

```
df_traffic_preprocessor = ColumnTransformer(  
    transformers = [  
        ('numeric', num_transformer, num_cols),  
        ('ordinal', ordinal_transformer, cat_cols)  
    ],  
    remainder='passthrough'
```

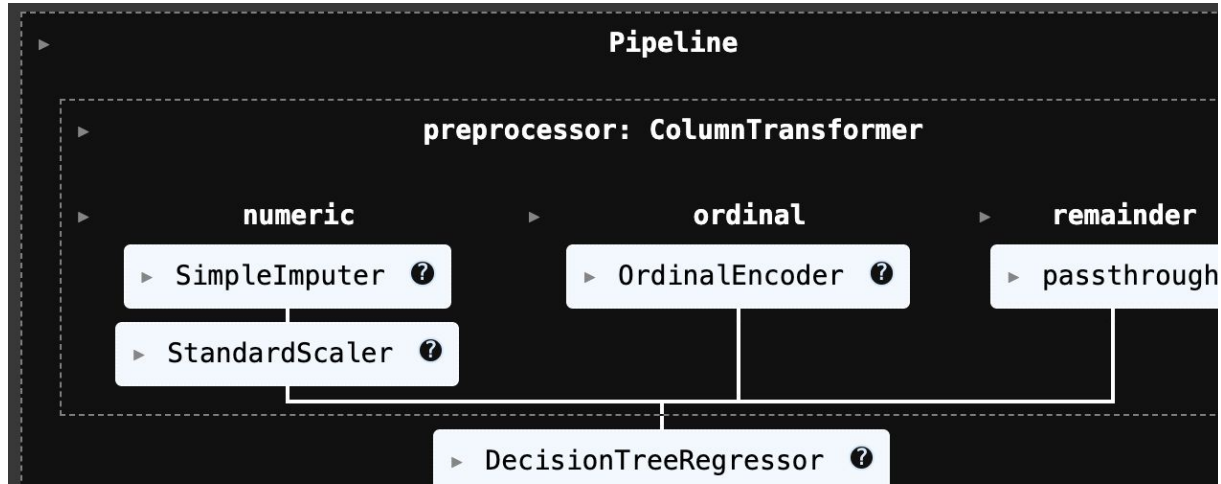
# Model Development - (SISCO)

First Algorithm (Linear Regression )



# Model Development cont.. (FLORENCE)

## Decision Tree Regressor



# Evaluation metrics (FLORENCE)

R squared score (Linear regression)

```
# Evaluate model performance on test data using R2 score (coefficient of determination)
r2 = r2_score(y_test, y_predict)
print("R2", r2)
```

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R<sup>2</sup>: 0.8540013437773113

R squared score (Decision tree regressor)

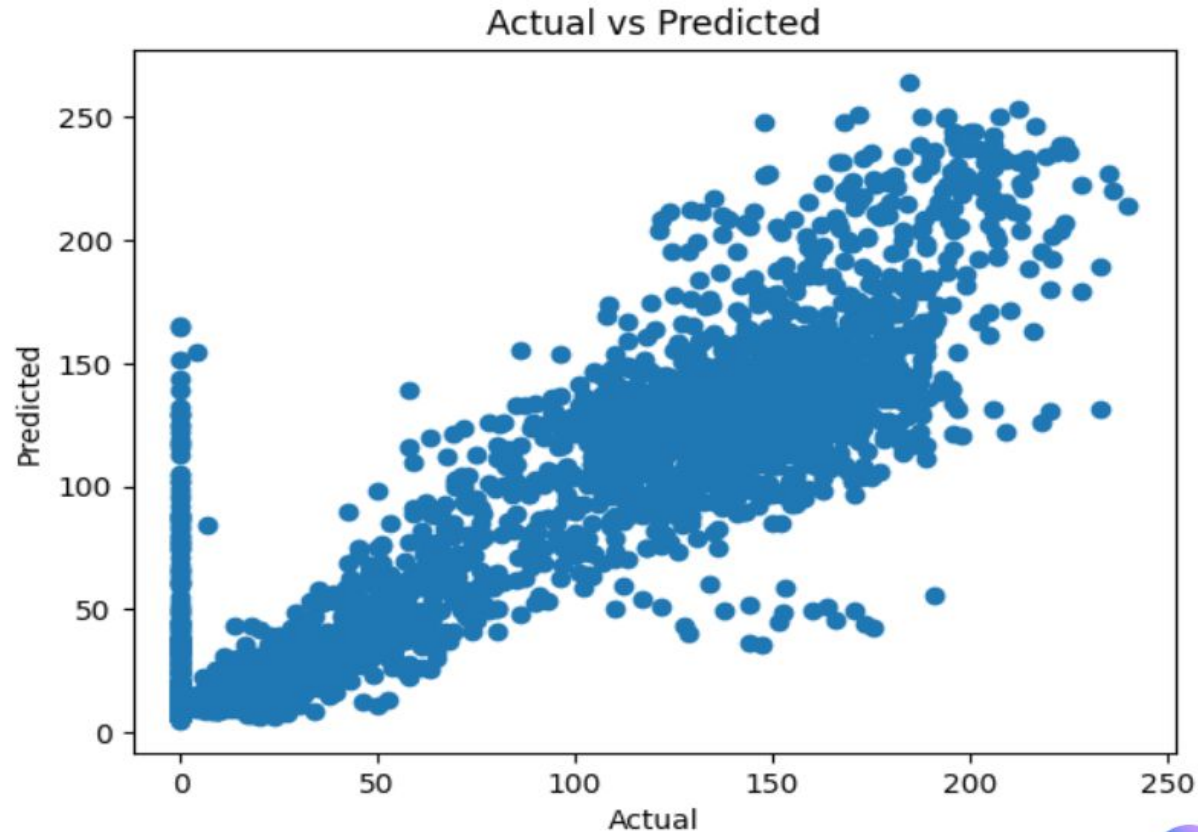
```
r2 = r2_score(y_test, y_predict_tree)
print("R2", r2)
```

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R<sup>2</sup>: 0.8476499006480733



# Scatter plot for actual and predicted values. (FLORENCE)



**THE END**