

## Data Collection and Preprocessing Phase

Date	15 July 2024
Team ID	SWTID1720090652
Project Title	Predictive Modelling for Fleet Fuel Management using Machine Learning
Maximum Marks	6 Marks

### Data Exploration and Preprocessing Template

Identifies data sources, assesses quality issues like missing values and duplicates, and implements resolution plans to ensure accurate and reliable analysis.

Section	Description																																				
Data Overview	<p>Basic statistics, dimensions, and structure of the data.</p> <p>The dataset contains 488 records with 12 column names (12 prospective features before cleaning). The columns are distance, consume, speed, temp_inside, temp_outside, specials, gas_type, AC, rain, sun, refill liters (contains majority null values), refill gas (contains majority null values)</p>																																				
Univariate Analysis	<p>The mean, median and mode of the corresponding columns are as follows:</p> <table><tr><th></th><th>Mean</th><th>median</th><th>mode</th></tr><tr><td>distance</td><td>19.652835</td><td>14.6</td><td>11.8</td></tr><tr><td>consume</td><td>4.912371</td><td>4.7</td><td>4.5</td></tr><tr><td>speed</td><td>41.927835</td><td>40.5</td><td>42.0</td></tr><tr><td>temp_inside</td><td>21.929521</td><td>22.0</td><td>21.5</td></tr><tr><td>temp_outside</td><td>11.358247</td><td>10.0</td><td>8.0</td></tr><tr><td>AC</td><td>0.077320</td><td>0.0</td><td>0.0</td></tr><tr><td>rain</td><td>0.123711</td><td>0.0</td><td>0.0</td></tr><tr><td>sun</td><td>0.082474</td><td>0.0</td><td>0.0</td></tr></table>		Mean	median	mode	distance	19.652835	14.6	11.8	consume	4.912371	4.7	4.5	speed	41.927835	40.5	42.0	temp_inside	21.929521	22.0	21.5	temp_outside	11.358247	10.0	8.0	AC	0.077320	0.0	0.0	rain	0.123711	0.0	0.0	sun	0.082474	0.0	0.0
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Bivariate Analysis	<pre> distance    distance    consume    speed    temp_inside    temp_outside \ distance      1.000000   -0.128967   0.562299     0.075305     0.088175 consume     -0.128967    1.000000  -0.227866    -0.161991    -0.320811 speed        0.562299  -0.227866    1.000000     0.059725     0.015411 temp_inside   0.075305  -0.161991   0.059725     1.000000     0.361308 temp_outside  0.088175  -0.320811   0.015411     0.361308     1.000000 AC           -0.025738   0.096591  -0.035408     0.297775     0.167562 rain         -0.019791   0.248118   0.009489    -0.037356    -0.186315 sun          0.081120  -0.170667   0.081618     0.246120     0.346903             AC      rain      sun distance  -0.025738 -0.019791  0.081120 consume   0.096591  0.248118 -0.170667 speed     -0.035408  0.009489  0.081618 temp_inside 0.297775 -0.037356  0.246120 temp_outside 0.167562 -0.186315  0.346903 AC         1.000000  0.242915  0.088598 rain       0.242915  1.000000 -0.112650 sun        0.088598 -0.112650  1.000000 </pre>
Multivariate Analysis	<p>On performing feature selection with SelectKBest and f_regression, we get the following result:</p> <pre> Selected features: Index(['speed', 'temp_inside', 'temp_outside', 'rain', 'sun'], dtype='object') </pre>
Outliers and Anomalies	<p>Identification is done by using the IQR method. We define the first and third quartile's boundaries and check for the values that lie outside and inside the boundaries.</p>
<b>Data Preprocessing Code Screenshots</b>	
Loading Data	<pre> df=pd.read_csv('/content/measurements.csv') df </pre>
Handling Missing Data	<pre> #fill missing values in 'temp_inside' column with mean temp_inside_mean = df['temp_inside'].mean() df['temp_inside'].fillna(temp_inside_mean, inplace=True) </pre>
Data Transformation	<p>Code for transforming variables (scaling, normalization).</p> <pre> # Standardize the features scaler = StandardScaler() X_train = scaler.fit_transform(X_train) X_test = scaler.transform(X_test)  # Initialize and train the SVR model </pre>

## Feature Engineering

```
# Drop unnecessary columns
df.drop(['refill gas', 'refill liters', 'specials'], axis=1, inplace=True)
```

```
# Replace commas with periods for numeric representations
for col in df.columns:
    if df[col].dtype == 'object': # Check if the column is of object type (likely string)
        df[col] = df[col].str.replace(',', '.', regex=True)
```

```
# Convert numeric columns to float after replacing commas
for col in df.select_dtypes(include=['object']).columns:
    try:
        df[col] = df[col].astype(float)
    except ValueError:
        pass # Handle columns that cannot be converted to float
```

```
# Convert categorical variable 'gas_type' into dummy variables
dum1 = pd.get_dummies(df['gas_type'])
df = pd.concat([df, dum1], axis=1)
df.drop('gas_type', axis=1, inplace=True)
```

## Save Processed Data

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

# Save the model and scaler
joblib.dump(model, 'svm_model.pkl')
joblib.dump(scaler, 'scaler.pkl')
```