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Explainability AI TD4 Presentation

Our Dataset

平 :: > US_Accidents_March23.csv (3.06 GB) Column Compact 10 of 46 columns Detail # Start_Lng ☐ Start_Time Time End_Time # Start_Lat Severity Shows start time of the Shows end time of the Shows latitude in GPS Shows longitude in GPS bws the severity of the coordinate of the start accident in local time accident in local time coordinate of the start ident, a number zone. End time here refers ween 1 and 4, where 1 point. point. icates the least impact to when the impact of accident on traffic flow traffic (i.e., short delay

2016-02-08 2023-04-01

- Accidents US
- 46 Columns
- very large width

Specificity

Size: 3 GB

Data Explorer

3.06 GB

US_Accidents_March23.csv

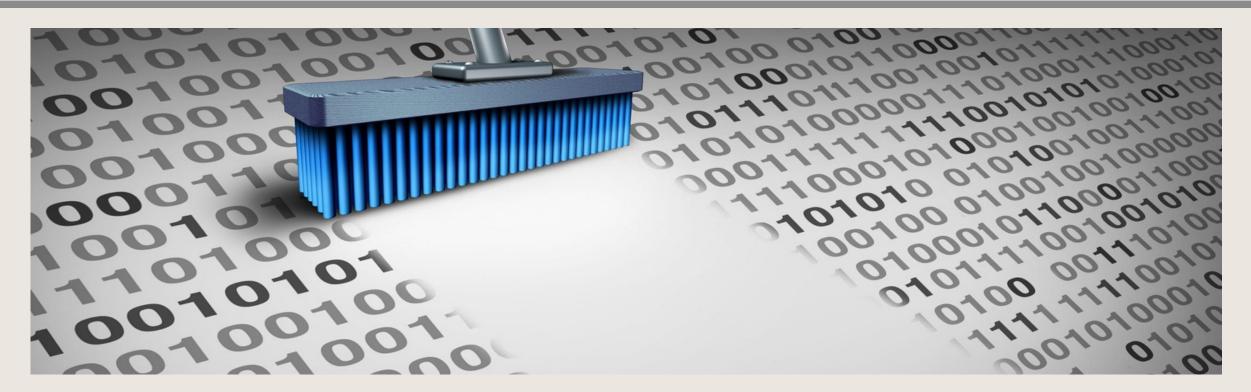
7 Million lines

df.shape

(7728394, 46)

first: drop columns

```
# Delete the columns we don't need
df = df.drop(['Street', 'City', 'Description','County', 'Zipcode', 'Country', 'Bump', '
```



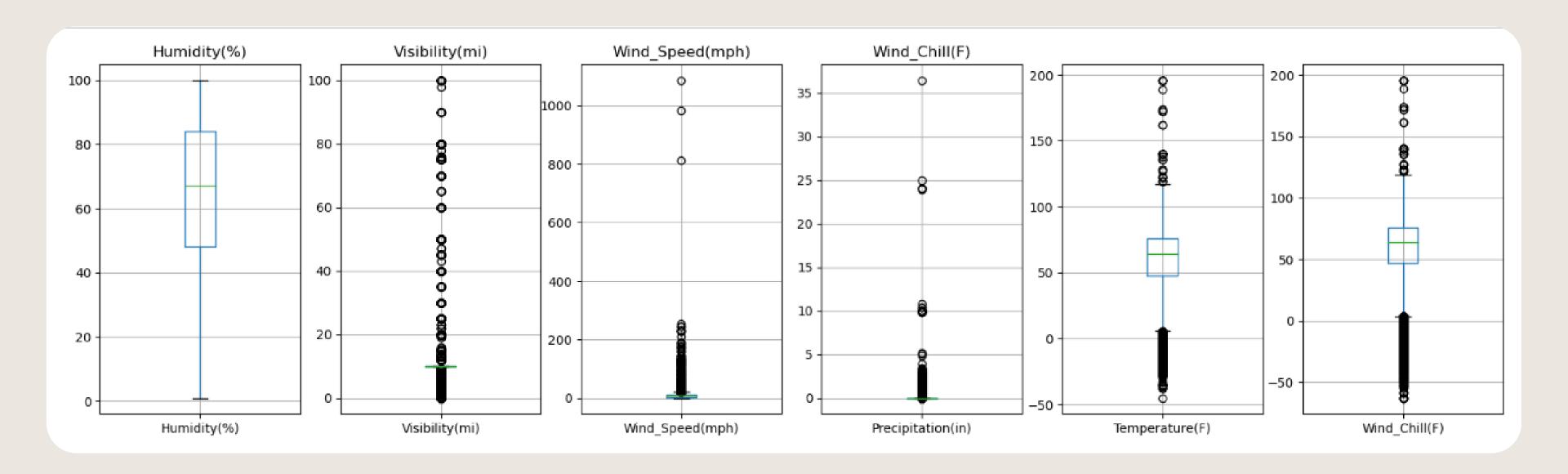
Missing values

```
print(df.isnull().sum())
Entrée [11]:
             ID
                                              0
             Source
             Severity
             Start_Time
             End Time
             Start Lat
             Start_Lng
             End Lat
                                       3402762
             End_Lng
                                       3402762
             Distance(mi)
                                              0
             State
                                              0
             Timezone
                                           7808
             Weather_Timestamp
                                        120228
             Temperature(F)
                                        163853
             Wind Chill(F)
                                       1999019
```

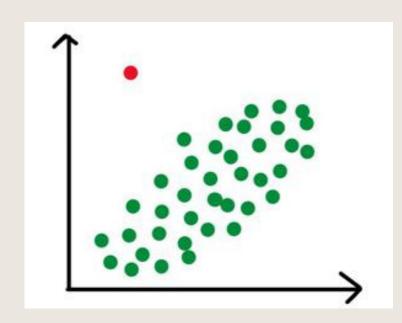
Missing values

```
.fillna(newdf['Start_Lat'])
    fillna('unknown')
fillna(newdf['Temperature(F)'].mean())
    fillna('###')
dropna(subset=['Precipitation(in)'])
dropna(subset=['Temperature(F)'])
```

OUTLIERS



OUTLIERS



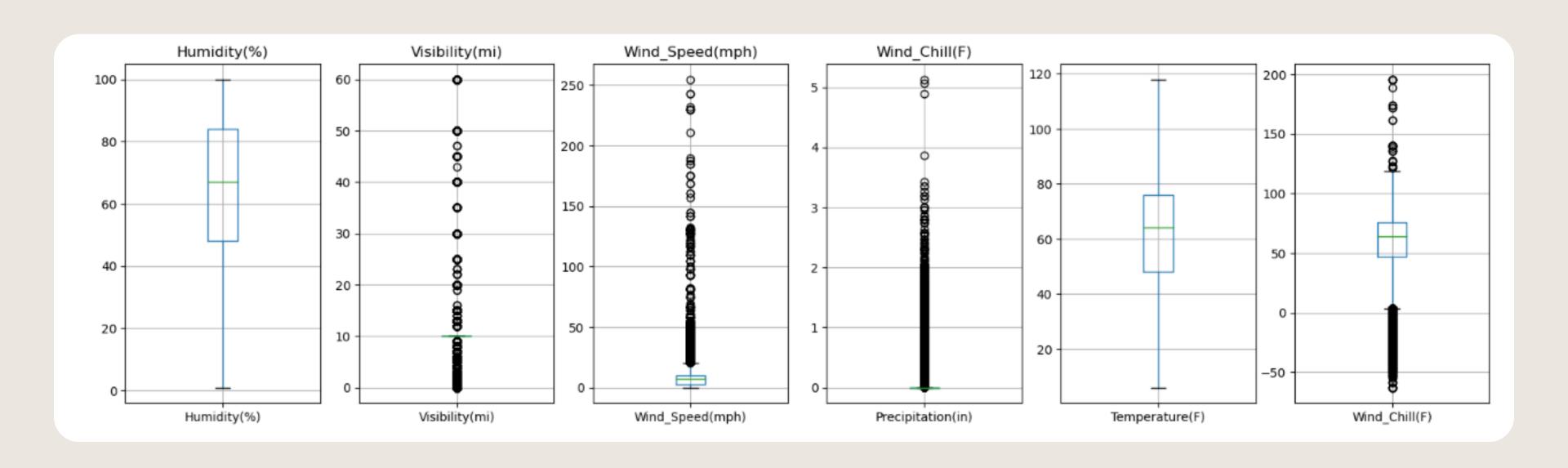
Outlier cleaning sample:

```
Q1= newdf['Wind_Speed(mph)'].quantile(0.25)
Q3= newdf['Wind_Speed(mph)'].quantile(0.75)
IQR = Q3 - Q1
moyenne = newdf['Wind_Speed(mph)'].mean()

threshold = 1.5

newdf.loc[(df['Wind_Speed(mph)'] < Q1 - threshold * IQR), 'Wind_Speed(mph)'] = moyenne
newdf.loc[(df['Wind_Speed(mph)'] > Q3 + threshold * IQR), 'Wind_Speed(mph)'] = moyenne
```

OUTLIERS - After data cleaning



Specificity



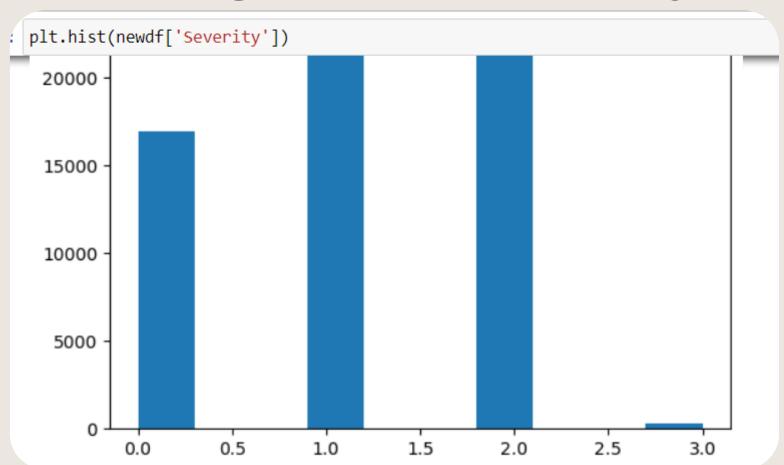
very huge amount of data(7m)

divide the dataset

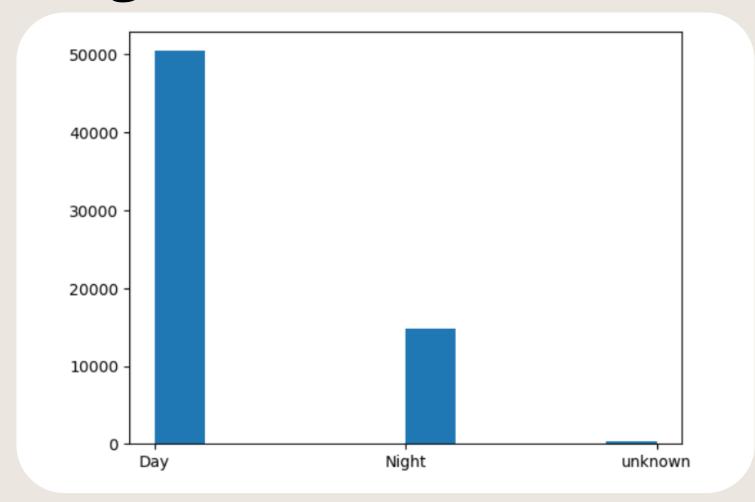
Distortion of result (plots)

Univariate Analysis

Histogram of Severity

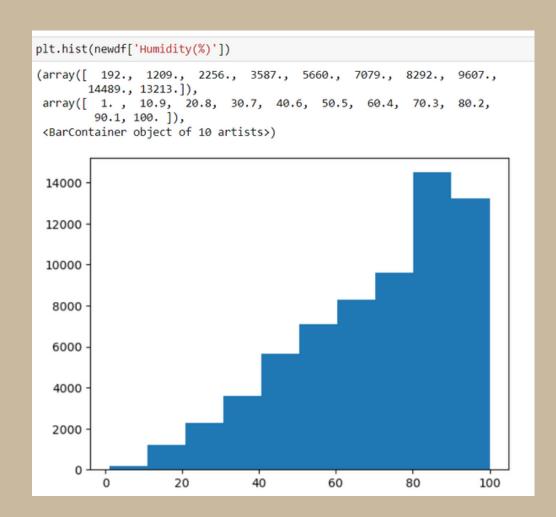


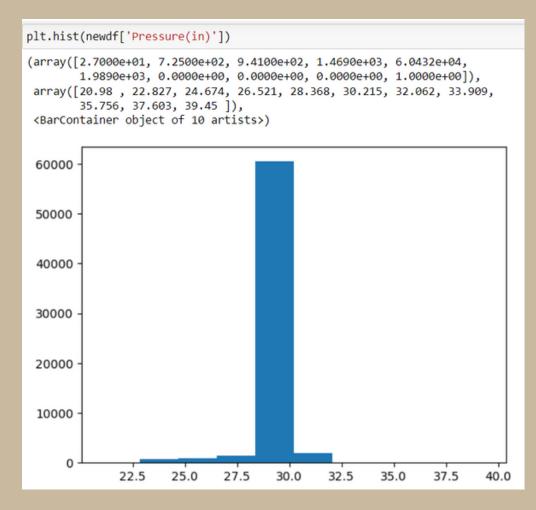
Histogram of Time of the day



Bivariate Analysis







LINEAR REGRESSION

Variable choice

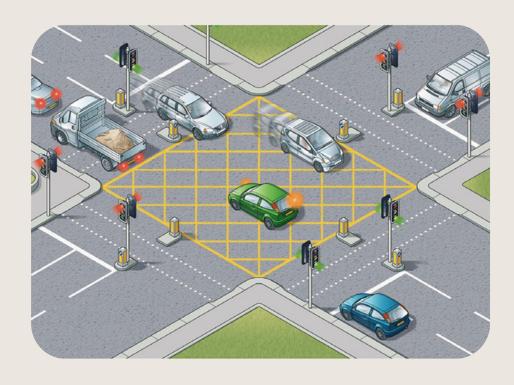
Weather-related

- Visibility (mi)
- Weather condition*
- Precipitation (in)



Environment-related

- Crossing (bool)
- Junction (bool)
- State*
- Stop (bool)
- Traffic Signal (bool)
- Station (bool)



Target: Severity

LINEAR REGRESSION

Our results

Training R^2 score: 0.24430378517137308

Testing R^2 score: 0.23833267491677146

Training score: Measures how well the model fits the training data.

Testing score: Prediction of new, unseen data

LinearRegression
LinearRegression()

Low score for both Training and Testing --> maybe there is a non-linear relationship on Severity

LINEAR REGRESSION

Coefficients

Intercept: 0.7858510659522875

Value of Severity if all independent variables as set to 0

Our coefficients with the highest values:

```
Coefficient for State_WV: 0.9754979967003913
Coefficient for State_WY: 1.8590767528192031
Coefficient for Weather_Condition_Light Thunderstorms and Rain: 0.7000497
Coefficient for Weather Condition Light Thunderstorms and Snow: 1.2477980
```

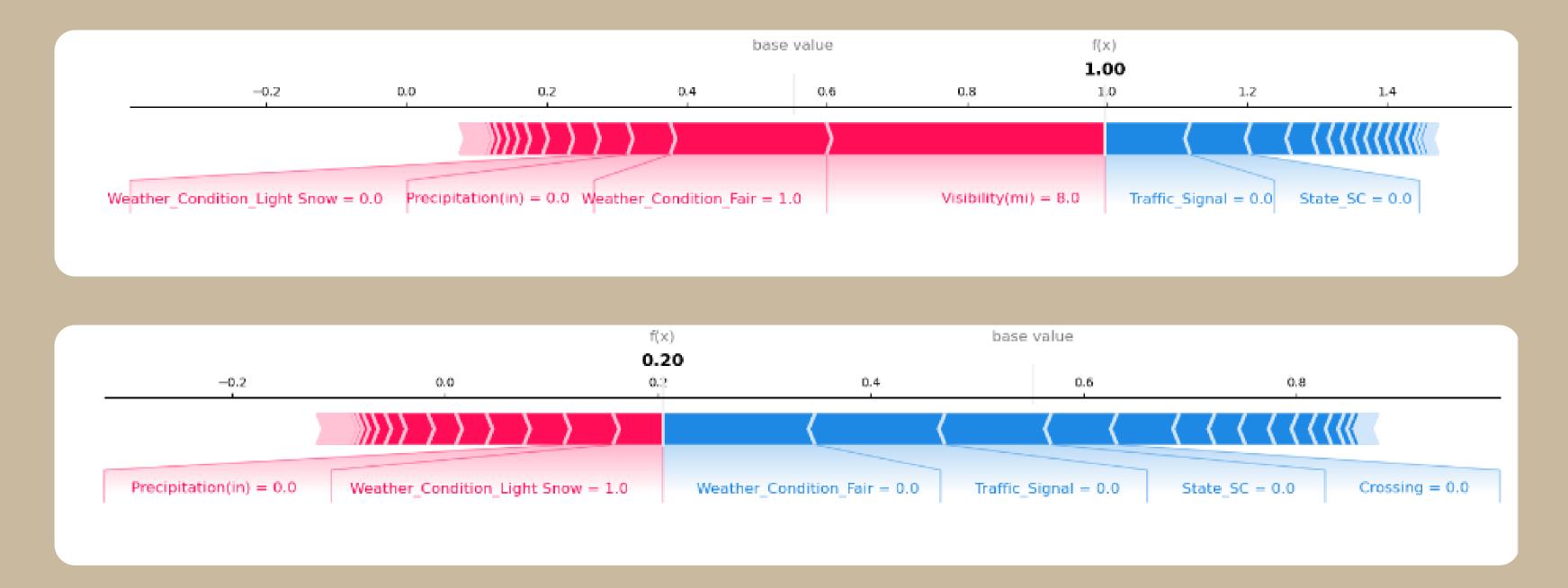
Complex Model - XgBoost

Why XgBoost?

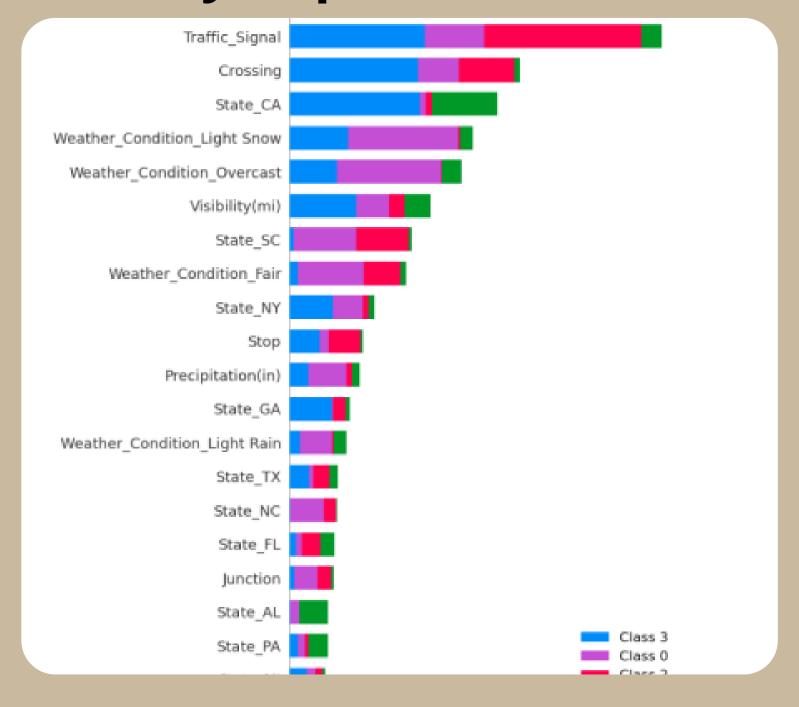
--> Series of decision trees, non-linear, high performance

Accuracy: 0.5707860028970039

Examples of forceplot



Identify important variables



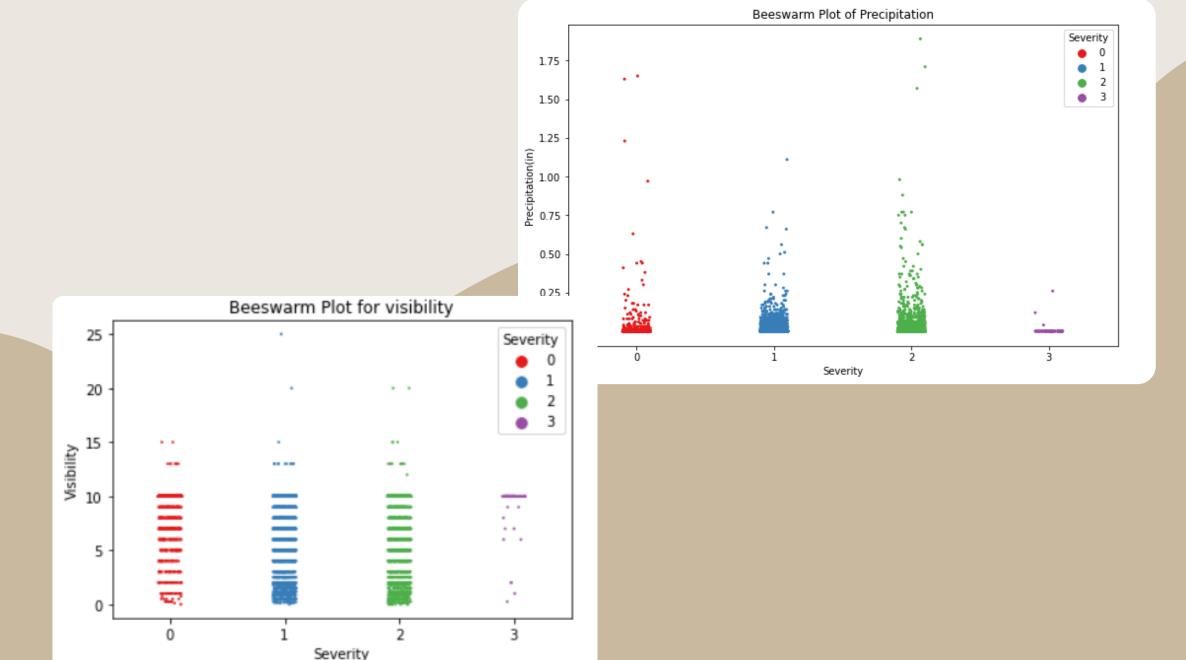
Comparing to linear regression:

- Weather conditions and States are still important
- Traffic_Signals and Crossings have an impact, we couldn't see it in the linear regression
- Visibility and Precipitation also appears to be impactful

Beeswarm plots

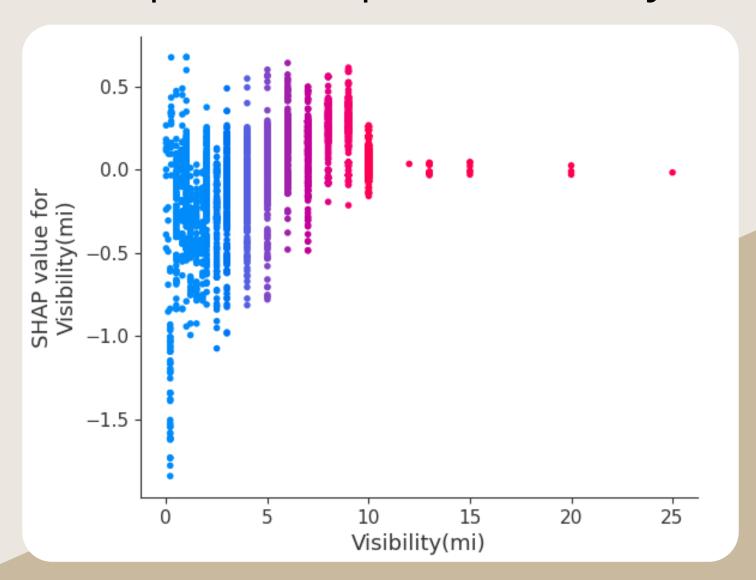


Binary values -> impossible to represent in a stripplot

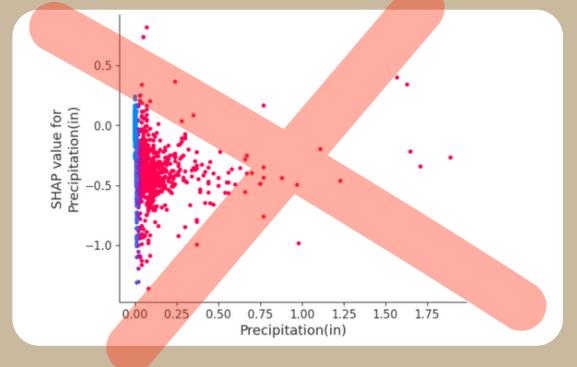


Dependance plots

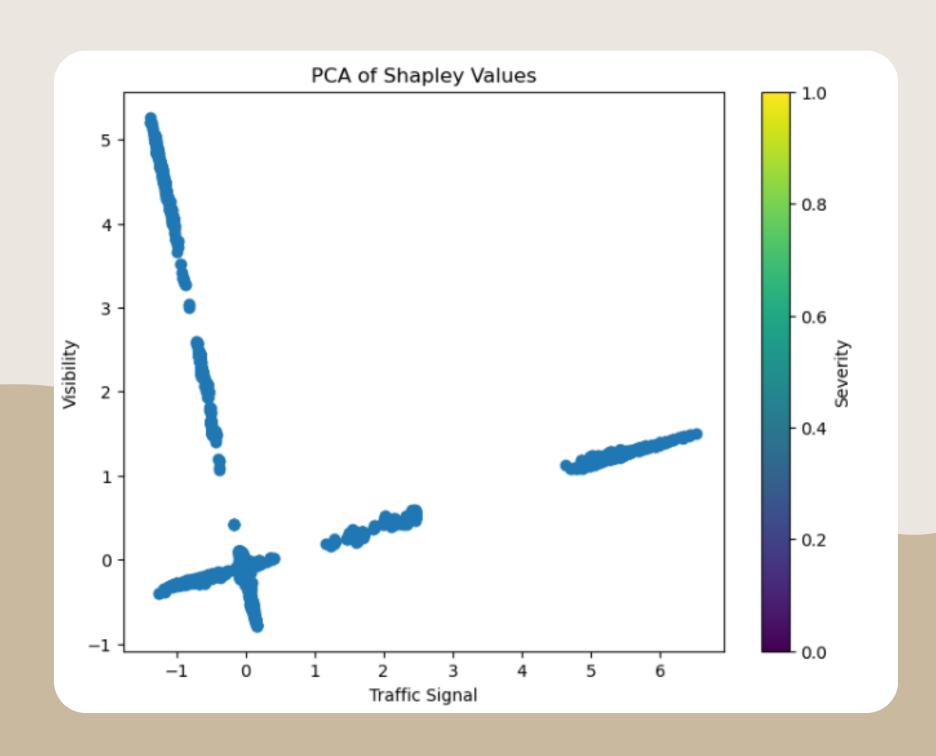
Dependance plot of Visibility

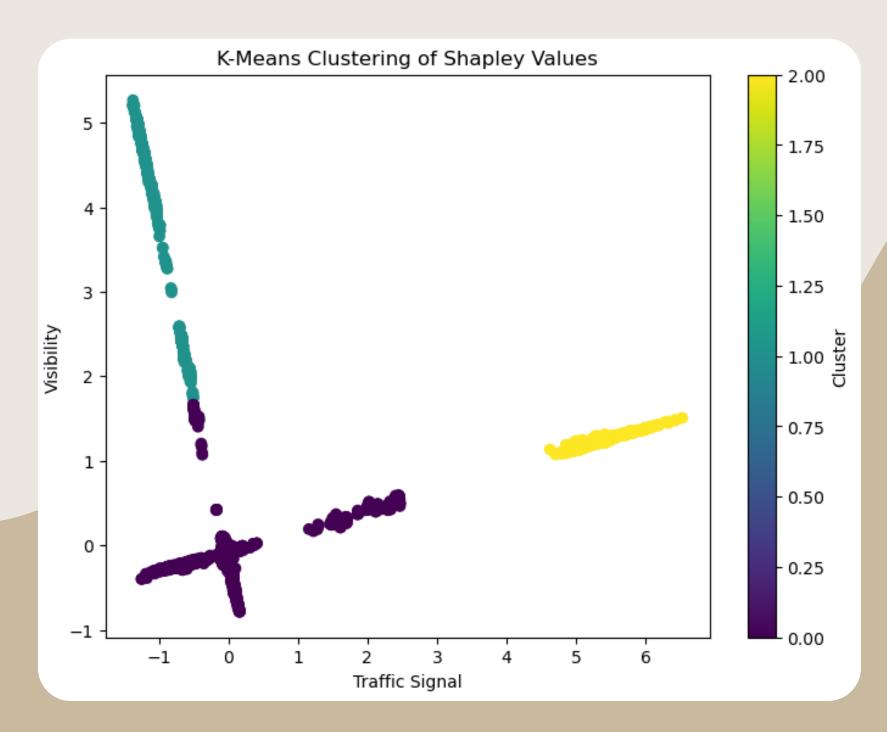


Dependance plot of Precipitation



Clustering of Shapley values





Thank you for Listening