BAX 452 Machine Learning

An Analysis for Car Crash and Implication for Municipal Management

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Executive Summary

This report aims to analyze the Car Crash Dataset provided by the City of Cary in North Carolina, to identify factors that contribute to car crashes and to gain municipal management insights. The car crash dataset includes information such as the severity of the car crash, time, road condition, and traffic control condition. We analyzed the dataset by EDA and a series of ML methods. It is worth mentioning that the focus of our analysis is not to predict the severity of future car crashes, but to identify and interpret how and why different factors contribute to car crashes and provide insights for municipal management.

Here are some key findings for our analysis:

- Interaction road feature, signal and stop traffic control, and curve road add to the Crash Score.
- Daytime has the highest Crash_Score, followed by late-early and overnight.
- The relative impact for interaction is lower for signal stop, and the relative impact for the work area is higher for late early.

Here are the suggestions based on our findings:

- The intersection has complex traffic conditions as well as a significantly higher Crash_Score.
 Thus, municipal management should consider signal and stop, two-way protected median, and lighting as traffic control and road management practice.
- Time of day largely affects the Crash_Score. Thus, municipal management could try different practices for different times, especially in the work area during rush hour.
- Municipal management should provide better lighting during night and more road management for other roads.
- It is of equal importance to improving the road condition as well as to raising awareness for safety. Municipal management could consider adding more signs and slogans as reminders.

Introduction

The traffic accident is a top ten cause of death. Statistics show that every year, there are approximately 1.37 billion people die due to traffic accidents, which accounts for about 2.2% of all the death. It has always been a focus of municipal management. However, there is no systematic and scientific way to analyze traffic accidents. In order to address this issue, the City of Cary in North Carolina provided a Car Crash Dataset, which records all of the car crashes in 2014 – 2019 and includes the severity of accidents and a series of factors could contribute to accidents, for example, time, road condition, and traffic control condition. We will use a series of techniques to develop predictive ML models and to investigate how different factors affect car crashes. Traditionally, municipal management tends to apply qualitative rather than quantitative analysis, as a result, our analysis will be the first and one of a kind. Hopefully, data analytics could identify and interpret factors that are overlooked or underestimated by traditional methods.

Data Characteristics

The dataset is provided by the Department of Transportation in the City of Cary in North Carolina, and it includes information for all of the car crashes in 2014 – 2019 with 14 dimensions and 23137 observations. The dependent variable Crash_Score is a numeric variable, which measures the extent of the car crashes, based on a series of factors, such as the number of vehicles involved, and the number of injuries or fatalities. The distributions of Crash_Score and the log of Crash_Score, it is shown that the distribution of the Crash_Score is right-skewed, with a median of 5.660, and a max of 53.070, which indicate that most car crashes are slight car crashes and a small proportion are severe car crashes. The log of Crash_Score is left-skewed (Figure 1).

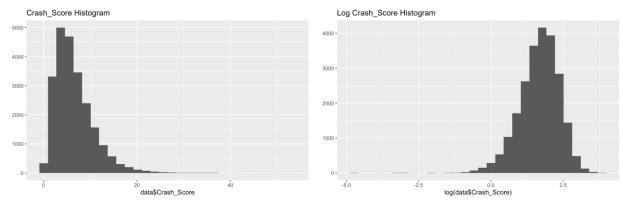


Figure 1 Crash_Score Histogram and the log of Crash_Score Histogram

Most of the independent variables are categorical variables, including year, month, time, road conditions, traffic control conditions, and weather. Here are some findings according to the EDA of independent variables (Appendix 1, 2, 3):

- There is no significant change over the years (from 2014-2019), nor significant change over the year (from Jan. to Dec.), in other words, no seasonality for car crashes.
- The crash score is affected by the time of the day, the median score is higher during the daytime (8 am to 8 pm), the median score is significantly lower for time 1 (midnight to 4 am), and relatively lower for time 2 (4 am to 8 am) and time 6 (8 pm to midnight).
- For the road feature, none traffic control has most car crashes, and the intersection has a significantly higher median score.
- For the road character, the straight level has a significantly higher median score.
- For the road class, the state highway has the most car crashes as well as the highest score.
- Surprisingly, for light, daylight has the highest median score, followed by dusk and down, and then dark light and dark not light.
- For the weather, rain and snow have a relatively higher median score, other has the lowest median score.

- Surprisingly, for the traffic control, signal and stop sign have the highest median score none has the lowest median score.
- None work area has more car crashes, while the work area has a higher median score.

We fitted a model with the unprocessed dataset, the in-sample R2 is relatively high, but the out-of-sample R2 is low. It seems that the model tends to overfit. Based on our observations, we decided to do a feature transformation and to combine some levels of the factors. Feature transformation could reduce the complexity of the model and prevent overfitting. Here is the process (Table 1):

Feature	Levels with Transformation
Time_of_Day	DAYTIME, OVERNIGHT, LATE-EARLY
Rd_Feature	INTERSECTION, OTHER
Rd_Character	STRAIGHT, CURVE
Rd_Surface	ASPHALT, OTHER
Weather	CLEAR-CLOUDY, RAIN-SNOW, OTHER
Traffic_Control	SIGNAL-STOP, OTHER

Table 1: Feature Transformation

Model Selection, Evaluation, and Interpretation

We used a series of techniques to select and evaluate generalized linear models (GLMs), and then interpret the model by the coefficients of variables. Although the predictive performance is not the main focus of our model, we used RMSE and out-of-sample R2 as the main performance metrics for us to select and evaluate models. The RMSE indicates how well the model can predict dependent variables in new datasets, and the out-of-sample R2 indicates how well the independent variables can interpret the variation for dependent variables in new datasets. We chose these metrics because we want the

model to generalize insights, rather than overfit our dataset. Additionally, we applied 10-fold cross-validation, to utilize the dataset better and to provide performance metrics more accurately.

Model 0: GLM with the unprocessed dataset

 $glm(Crash_Score \sim ., gaussian(), data = data)$

As we mentioned, we fitted a model with the unprocessed dataset, and the in-sample R2 is significantly lower than the out-of-sample R2, indicating overfit. As a result, we did feature transformation and reduced the level of factors (Appendix 5).

Model 1: GLM with the processed dataset

 $glm(Crash_Score \sim ., gaussian(), data = data2)$ $glm(Crash_Score \sim ., Gamma(log = "link"), data = data2)$

We fitted a model with the dataset after feature transformation, and both the RMSE and out-of-sample R2 are improved. Then, we benchmarked the performance for different distribution functions and link functions. As we mentioned, the distribution of Crash_Score is skewed, and Gamma distribution could be a better choice. After we investigated the visualizations (Figure 2, 3), it seems that with Gamma distribution and log link, the residuals have zero means and constant variables, and there are normal distributions for most residuals but not for extremes. As a result, we chose the model with Gamma distribution and log link, but we will also use Gaussian distribution for interpretation and intuition.

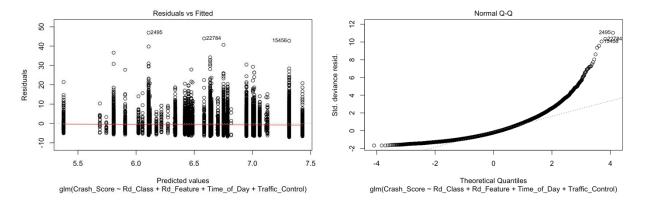


Figure 2: Visualization for GLM with Gaussian Distribution

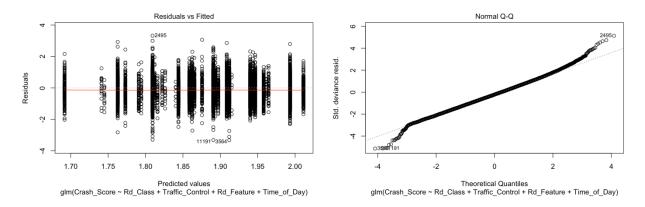


Figure 3: Visualization for GLM with Gamma Distribution and Log Link

Model 2: GLM with feature selection

$$glm(Crash_Score \sim Rd_Class + Traffic_Control + Rd_Feature$$

$$+ Time_of_Day, gaussian(), data = data2)$$
 $glm(Crash_Score \sim Rd_Class + Traffic_Control + Rd_Feature + Time_of_Day, Gamma(link$
 $= "log"), data = data2)$

After feature transformation and model selection, the in-sample R2 is still significantly higher than the out-of-sample R2. This is when we consider a feature selection to reduce the complexity of the model and prevent overfitting. We chose BIC forward selection and LASSO regression. The former starts with

no variables in the model and adds the variable that gives the most statistically significant improvement of the fit. The latter starts with all the variables in the model and shrinks the parameters with the penalty. Additionally, let us emphasize the focus of the analysis, to identify and interpret the key factors that contribute to car crashes. With this goal, we chose BIC over AIC for and the penalty is greater for each additional parameter; we also chose LASSO over ridge, for it could shrink the parameters of the useless variables to zero.

Although BIC forward selection and LASSO regression not significantly improve the model performance, it helps us identify key factors, including Rd_Class, Traffic_Control. Rd_Feature, and Time_of_Day. Here are some findings (Appendix 7):

- Interaction road feature adds to Crash_Score by approximately 0.36.
- Signal and stop traffic control add to Crash_Score by approximately 0.31.
- Daytime has the highest Crash_Score, followed by late-early, 0.30 lower, and overnight, 0.73
 lower.

Model 3: GLM with interaction

```
x_{train} = model. matrix(Crash\_Score \sim .+ (.)^2, trainData2) y_{train} = trainData2\$Crash\_Score glmnet(x\_train, y\_train, "gaussian", lambda = lasso.lam.min, alpha = 1)
```

The GLM without interaction provides similar insights as EDA. We want to further improve the model and extract more insights from the dataset, so we decided to take interaction into consideration. Car crashes are complicated and different factors affect each other. In the following interaction plots, it is shown that there are interactions between factors (Figure 4).

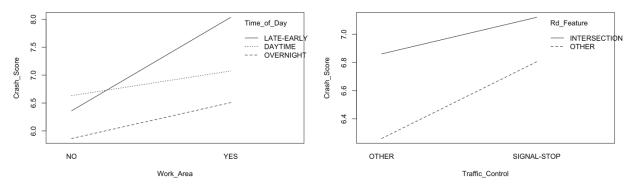


Figure 4: Interaction for different factors

To achieve this, we included all the variables and interactions and applied high dimension LASSO regression. Additionally, we used 10-fold cross-validation, to determine the penalty parameter, lambda, with minimum MSE (Figure 5).

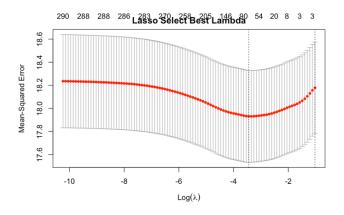


Figure 5: Selection for Lambda

With the high dimension LASSO regression, the in-sample R2 is significantly improved, and the out-of-sample R2 is further improved (Appendix 8). It helps us investigate how a single factor contributes to car crashes as well as how multiple factors interact, which provides insights that could be overlooked or underestimated by traditional methods. Here are some findings (Appendix 8):

 Although interaction, signal and stop both add to Crash_Score, the relative impact for interaction is lower for signal stop.

- The relative impact for the work area is higher for late early.
- The relative impact for light is higher during overnight.
- Road class and road condition interacts with each other. The ice, snow, slush road condition has
 relative high impact on other road than US and state highway.

Recommendation

Based on findings in Model 2 (Appendix 7) and Model 3 (Appendix 8), the following suggestions could be provided to the municipal management:

- The intersection has complex traffic conditions as well as a significantly higher Crash_Score. As a result, municipal management needs to pay additional attention to the intersection. Here are some recommended practices, signal and stop reduce the relative impact of the intersection, besides, two-way protected median and lighting could be considered.
- Time of day largely affects the Crash_Score, the highest is daytime, followed by late early and
 overnight. Municipal management could try different practices for different times. Additionally,
 time of day interacts with many factors. For example, more attention is needed in the work area
 during the rush hour.
- Bad lighting could largely contribute to Crash_Score during the night, and bad weather could
 largely contribute to Crash_Score for other roads than US and state highway. For example, there
 are more snowplows on highways than other roads during extreme weather. Thus, municipal
 management should provide better lighting during night and more road management for other
 roads.
- It seems surprising that curve road has a lower Crash_Score than straight road, as people are more cautious during curve road. Thus, it is of equal importance to improving the road condition

as well as to raising awareness for safety. Municipal management could consider adding more signs and slogans as reminders.

Furthermore, although it is a relatively comprehensive dataset with a large number of dimensions and observations, there are two major limitations. Firstly, Crash_Score is abstract. It is stated that Crash_Score is related to the extent of the car crash. However, it is not intuitive when we illustrate the effect of a variable as "Interaction road feature added to Crash_Score by approximately 0.36". What is an increase of 0.36 supposed to mean? What are the differences between Crash_Scores of 1, 5, and 10? Answers to these questions will help us better analyze the car crash issue. Secondly, as we mentioned, car crashes are complicated, there are various factors, even incidents, could contribute to traffic accidents. As a result, more dimensions for the dataset could be helpful for our analysis. For example, work and non-work areas could be an overgeneralized concept. If geospatial data is included in our dataset, we could analyze which area needs additional attention and provide customized traffic control practices for different areas.

Conclusion

By analyzing the Car Crash Dataset provided by the City of Cary in North Carolina, our report successfully identified and interpreted how different factors and interactions affect the severity of traffic accidents. Firstly, we applied EDA and gained initial insights from the dataset. Secondly, we used methods such as feature transformation, feature selection, and model selection to improve model performance and prevent overfitting. Then, we applied high dimension LASSO regression to investigate the interaction and extract further insights from the dataset. Finally, we provided several suggestions for municipal management. Furthermore, if we have a better understanding of the Crash_Score and more dimensions in the dataset, we will be able to generate more insights.

References

Road Safety Facts, https://www.asirt.org/safe-travel/road-safety-facts/

Appendix

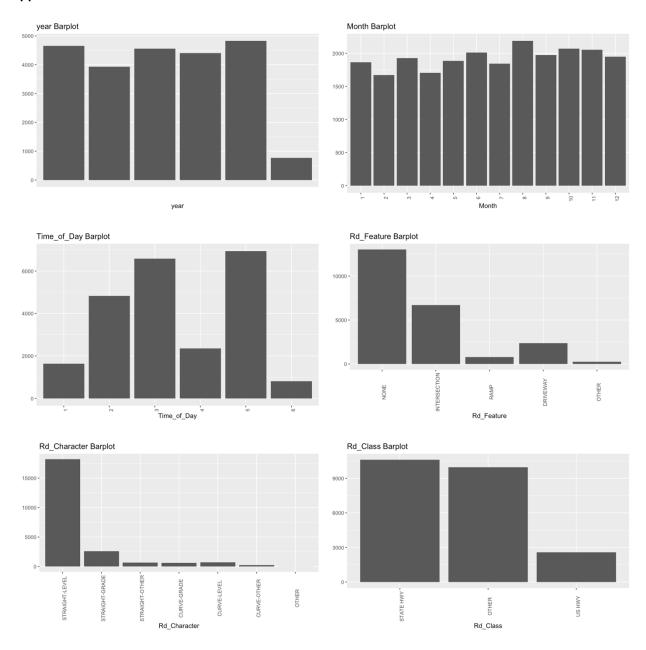
Appendix 1: Summary of Dataset

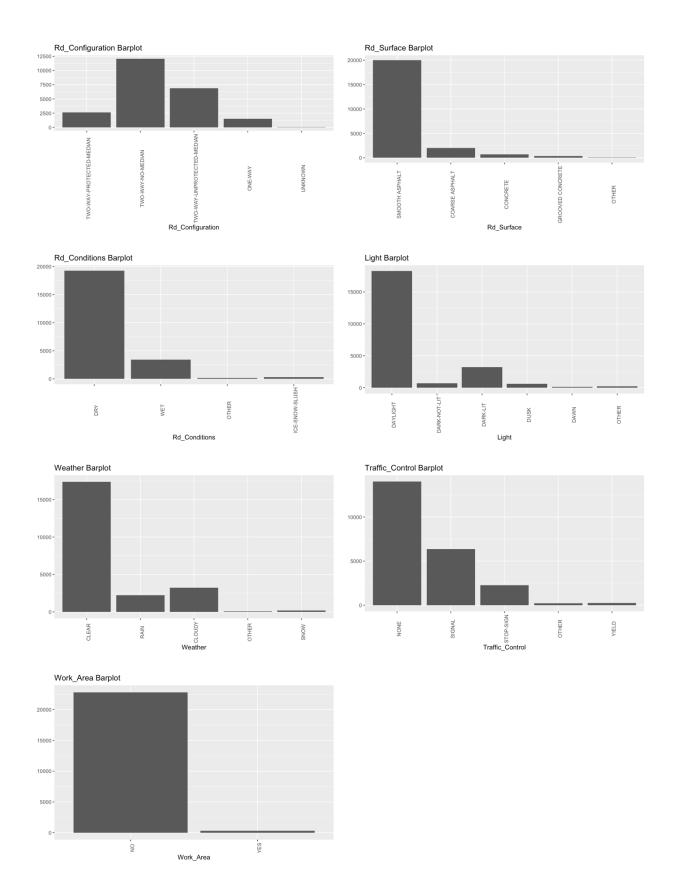
```
Crash Score
               year
                       Month
                                Time of Day
                                               Rd Feature
Rd Character
Min.: 0.010 Min.: 2014 Min.: 1.00 Min.: 1.000 NONE
                                                     :13025 STRAIGHT-
LEVEL:18215
1st Qu.: 3.540 1st Qu.:2015 1st Qu.: 3.00 1st Qu.:3.000 DRIVEWAY : 2373
CURVE-GRADE: 643
Median: 5.660 Median: 2016 Median: 7.00 Median: 4.000 INTERSECTION: 6702
CURVE-LEVEL: 725
Mean: 6.567 Mean: 2016 Mean: 6.56 Mean: 4.034 OTHER: 259 CURVE-
OTHER : 239
3rd Qu.: 8.600 3rd Qu.:2017 3rd Qu.:10.00 3rd Qu.:5.000 RAMP
                                                          : 778
OTHER
       : 13
Max. :53.070 Max. :2019 Max. :12.00 Max. :6.000
                                                        STRAIGHT-
GRADE: 2622
                                      STRAIGHT-OTHER: 680
  Rd Class
                    Rd Configuration
                                       Rd Surface
                                                     Rd Conditions
STATE HWY:10603 TWO-WAY-NO-MEDIAN
                                       :12076 SMOOTH ASPHALT :20007
DRY
        :19262
OTHER: 9960 ONE-WAY: 1496 COARSE ASPHALT: 1997 ICE-SNOW-
SLUSH: 322
US HWY: 2574 TWO-WAY-PROTECTED-MEDIAN: 2627 CONCRETE
                                                            : 692
OTHER
        TWO-WAY-UNPROTECTED-MEDIAN: 6882 GROOVED CONCRETE: 371
WET
        : 3419
        UNKNOWN
                          : 56 OTHER
                                         : 70
    Light
            Weather
                      Traffic Control Work Area
DAYLIGHT :18262 CLEAR :17393 NONE :14028 NO :22823
DARK-LIT : 3219 CLOUDY: 3234 OTHER : 228 YES: 314
DARK-NOT-LIT: 708 OTHER: 85 SIGNAL: 6352
```

DAWN : 140 RAIN : 2230 STOP-SIGN: 2269 DUSK : 602 SNOW : 195 YIELD : 260

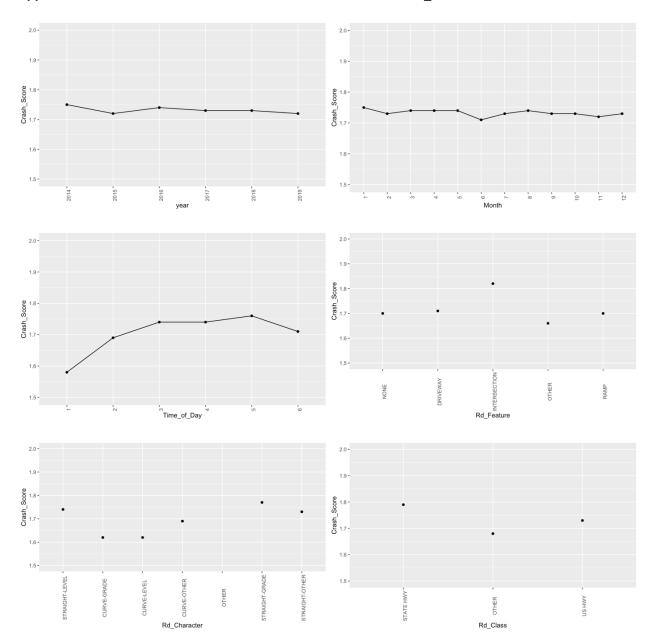
OTHER: 206

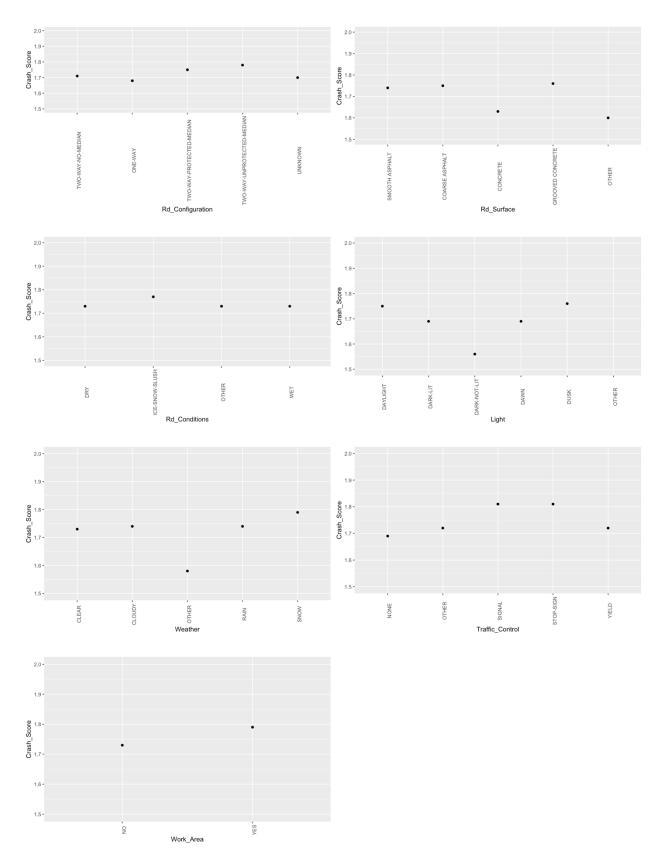
Appendix2: EDA Bar Plot for the Count of Observations





Appendix 3: EDA Line Plot and Scatter Plot for the Median of Crash_Score





Appendix 4: Summary of Dataset with Feature Transformation

Month Crash Score year Time_of_Day Rd Feature Rd Character Min.: 0.010 Min.: 2014 Min.: 1.00 DAYTIME: 18345 OTHER :16435 STRAIGHT:21517 1st Qu.: 3.540 1st Qu.:2015 1st Qu.: 3.00 OVERNIGHT: 808 INTERSECTION: 6702 CURVE : 1620 Median: 5.660 Median: 2016 Median: 7.00 LATE-EARLY: 3984 Mean: 6.567 Mean: 2016 Mean: 6.56 3rd Qu.: 8.600 3rd Qu.:2017 3rd Qu.:10.00 Max. :53.070 Max. :2019 Max. :12.00 Rd Configuration Rd Surface Rd Conditions Rd Class Light STATE HWY:10603 TWO-WAY-NO-MEDIAN :12076 ASPHALT:22004 :19262 DAYLIGHT :18262 OTHER : 9960 ONE-WAY : 1496 OTHER : 1133 ICE-SNOW-SLUSH: 322 DARK-LIT: 3219 US HWY: 2574 TWO-WAY-PROTECTED-MEDIAN: 2627 OTHER : 134 DARK-NOT-LIT: 708 TWO-WAY-UNPROTECTED-MEDIAN: 6882 WET : 3419 DAWN : 140 UNKNOWN DUSK : 56 : 602 OTHER : 206 Traffic Control Work Area Weather CLEAR-CLOUDY:20627 OTHER :14516 NO:22823 OTHER: 85 SIGNAL-STOP: 8621 YES: 314 RAIN-SNOW: 2425

Appendix 5: Performance for Model 0

GLM with Gaussian Distribution

```
[1] "AIC"
[1] 119370.1
[1] "RMSE"
[1] 4.25114
[1] "R2"
[1] 0.0166
[1] "OOS_R2"
[1] 0.01191
```

Appendix 6: Performance for Model 1

GLM with Gaussian Distribution

```
[1] "AIC"
[1] 119369.4
[1] "RMSE"
[1] 4.25061
[1] "R2"
[1] 0.01521
[1] "OOS_R2"
[1] 0.01215
```

GLM with Gamma Distribution and Log Link

```
[1] "AIC"
[1] 111848.9
[1] "RMSE"
[1] 4.25092
[1] "R2"
[1] 0.01565
[1] "OOS_R2"
[1] 0.01201
```

Appendix 7: Performance for Model 2

GLM with BIC forward selection

```
[1] "AIC"
[1] 119385.6
[1] "RMSE"
[1] 4.25201
[1] "R2"
[1] 0.01265
[1] "OOS_R2"
[1] 0.01191
```

```
Call:
glm(formula = Crash_Score ~ Rd_Class + Rd_Feature + Time_of_Day +
    Traffic_Control, family = gaussian(), data = data2)

Deviance Residuals:
    Min 1Q Median 3Q Max
```

```
-7.116 -2.990 -0.888 1.993 46.962
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)
                 Rd ClassOTHER -0.52929 0.06449 -8.207 2.38e-16 ***
Rd_ClassUS HWY 0.11414 0.09647 1.183 0.237
Rd FeatureINTERSECTION 0.36731 0.08050 4.563 5.07e-06 ***
Time of DayOVERNIGHT -0.73084 0.15299 -4.777 1.79e-06 ***
Time of DayLATE-EARLY -0.30067 0.07449 -4.036 5.45e-05 ***
Traffic ControlSIGNAL-STOP 0.31143 0.07598 4.099 4.17e-05 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for gaussian family taken to be 18.08283)
 Null deviance: 423606 on 23136 degrees of freedom
Residual deviance: 418256 on 23130 degrees of freedom
AIC: 132650
Number of Fisher Scoring iterations: 2
```

```
[1] "AIC"
[1] 111874
[1] "RMSE"
[1] 4.25218
[1] "R2"
[1] 0.01285
[1] "OOS_R2"
[1] 0.01183
```

```
Call:
glm(formula = Crash_Score ~ Rd_Class + Traffic_Control + Rd_Feature +
    Time_of_Day, family = Gamma(link = "log"), data = data2)

Deviance Residuals:
    Min 1Q Median 3Q Max
-3.3222 -0.5540 -0.1431 0.2772 3.3251
```

```
Coefficients:
             Estimate Std. Error t value Pr(>|t|)
                 1.890659 0.008845 213.758 < 2e-16 ***
(Intercept)
                    Rd ClassOTHER
Rd ClassUS HWY 0.017964 0.014688 1.223 0.221
Traffic ControlSIGNAL-STOP 0.049331 0.011569 4.264 2.02e-05 ***
Rd_FeatureINTERSECTION 0.053241 0.012257 4.344 1.41e-05 ***
Time of DayOVERNIGHT -0.117649 0.023294 -5.051 4.44e-07 ***
Time of DayLATE-EARLY -0.046650 0.011342 -4.113 3.92e-05 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for Gamma family taken to be 0.4191914)
 Null deviance: 9740.7 on 23136 degrees of freedom
Residual deviance: 9615.7 on 23130 degrees of freedom
AIC: 124303
Number of Fisher Scoring iterations: 5
```

GLM with LASSO Regression

```
[1] "RMSE"
[1] 4.38792
[1] "R2"
[1] 0.0152273
[1] "OOS R2"
[1] 0.01172
[1] "Coefficient"
27 x 1 sparse Matrix of class "dgCMatrix"
(Intercept)
                         30.587004828
(Intercept)
                       -0.011851525
year
Month
                         -0.001069361
Time_of_DayOVERNIGHT
                                 -0.376405738
Time of DayLATE-EARLY
                                -0.125524940
Rd FeatureINTERSECTION
                                 0.380489862
Rd CharacterCURVE
                              -0.245784850
Rd ClassOTHER
                           -0.505853226
```

Rd ClassUS HWY 0.133399307 Rd ConfigurationONE-WAY -0.191381117 Rd ConfigurationTWO-WAY-PROTECTED-MEDIAN 0.086848286 Rd ConfigurationTWO-WAY-UNPROTECTED-MEDIAN -0.020658867 Rd ConfigurationUNKNOWN 0.154234645 Rd SurfaceOTHER -0.293156710 Rd ConditionsICE-SNOW-SLUSH Rd ConditionsOTHER 0.432660392 Rd ConditionsWET -0.073925175 LightDARK-LIT -0.329201985 LightDARK-NOT-LIT -0.492735150 LightDAWN -0.110753748 LightDUSK LightOTHER -0.663746014 WeatherOTHER -0.558324341 WeatherRAIN-SNOW 0.146435986 Traffic_ControlSIGNAL-STOP 0.273934004 Work_AreaYES 0.462645783

Appendix 8: Performance for Model 3

High Dimension LASSO Regression

[1] "RMSE"	
[1] 4.384924	
[1] "R2"	
[1] 0.02173758	
[1] "OOS R2"	
[1] 0.01306927	
[1] "Coefficient"	
305 x 1 sparse Matrix of class "dgCMatrix"	
	1
(Intercept)	6.656291e+00
(Intercept)	
year .	
Month	:
Time_of_DayOVERNIGHT	
Time_of_DayLATE-EARLY	-1.726661e-04
Rd_FeatureINTERSECTION	2.455170e-01
Rd CharacterCURVE	

Dd ClassOTUED	4 6024700 01
Rd_ClassOTHER Rd_ClassUS HWY	-4.692470e-01
Rd_ConfigurationONE-WAY	
Rd ConfigurationTWO-WAY-PROTECTED-N	MEDIAN
Rd_ConfigurationTWO-WAY-UNPROTECTE	
Rd_ConfigurationUNKNOWN	
Rd_SurfaceOTHER	
Rd_ConditionsICE-SNOW-SLUSH	
Rd_ConditionsOTHER	
Rd_ConditionsWET	
_ LightDARK-LIT	-3.332283e-02
LightDARK-NOT-LIT	-2.983988e-01
LightDAWN	
LightDUSK	
LightOTHER	-2.987457e-01
WeatherOTHER	
WeatherRAIN-SNOW	
Traffic_ControlSIGNAL-STOP	2.308190e-01
Work_AreaYES	
year:Month	
year:Time_of_DayOVERNIGHT	
year:Time_of_DayLATE-EARLY	-8.481234e-06
year:Rd_FeatureINTERSECTION	4.680128e-06
year:Rd_CharacterCURVE	-3.727260e-06
year:Rd_ClassOTHER	-6.221184e-06
year:Rd_ClassUS HWY	
year:Rd_ConfigurationONE-WAY	
year:Rd_ConfigurationTWO-WAY-PROTEC	
year:Rd_ConfigurationTWO-WAY-UNPROT	ECTED-MEDIAN .
year:Rd_ConfigurationUNKNOWN	
year:Rd_SurfaceOTHER	
year:Rd_ConditionsICE-SNOW-SLUSH	
year:Rd_ConditionsOTHER	
year:Rd_ConditionsWET	
year:LightDARK-LIT	
year:LightDARK-NOT-LIT	
year:LightDAWN	
year:LightDUSK	
year:LightOTHER	-2.655948e-06
year:WeatherOTHER	

```
year:WeatherRAIN-SNOW
year:Traffic ControlSIGNAL-STOP
year:Work AreaYES
Month:Time of DayOVERNIGHT
Month: Time of DayLATE-EARLY
Month:Rd FeatureINTERSECTION
Month:Rd CharacterCURVE
Month:Rd ClassOTHER
                                          -2.266645e-03
Month:Rd ClassUS HWY
Month:Rd ConfigurationONE-WAY
Month:Rd ConfigurationTWO-WAY-PROTECTED-MEDIAN
                                                         2.327427e-02
Month:Rd ConfigurationTWO-WAY-UNPROTECTED-MEDIAN
Month:Rd ConfigurationUNKNOWN
Month:Rd SurfaceOTHER
Month:Rd ConditionsICE-SNOW-SLUSH
Month:Rd ConditionsOTHER
Month:Rd ConditionsWET
Month:LightDARK-LIT
                                       -1.280446e-02
Month:LightDARK-NOT-LIT
Month:LightDAWN
Month:LightDUSK
Month:LightOTHER
Month:WeatherOTHER
Month:WeatherRAIN-SNOW
Month:Traffic ControlSIGNAL-STOP
                                    8.479576e-03
Month: Work AreaYES
Time of DayOVERNIGHT:Rd FeatureINTERSECTION
Time_of_DayLATE-EARLY:Rd_FeatureINTERSECTION
Time of DayOVERNIGHT:Rd CharacterCURVE
Time of DayLATE-EARLY:Rd CharacterCURVE
                                                  -1.498426e-01
Time of DayOVERNIGHT:Rd ClassOTHER
Time of DayLATE-EARLY:Rd ClassOTHER
Time of DayOVERNIGHT:Rd ClassUS HWY
                                                   -4.460607e-01
Time of DayLATE-EARLY:Rd ClassUS HWY
                                                   2.689500e-02
Time of DayOVERNIGHT:Rd ConfigurationONE-WAY
Time of DayLATE-EARLY:Rd ConfigurationONE-WAY
Time of DayOVERNIGHT:Rd ConfigurationTWO-WAY-PROTECTED-MEDIAN
9.214905e-01
Time_of_DayLATE-EARLY:Rd_ConfigurationTWO-WAY-PROTECTED-MEDIAN
Time of DayOVERNIGHT:Rd ConfigurationTWO-WAY-UNPROTECTED-MEDIAN
```

Time_of_DayLATE-EARLY:Rd_ConfigurationTWO-WAY-UNPROTECTED-MEDIAN -	
7.651017e-02	
Time_of_DayOVERNIGHT:Rd_ConfigurationUNKNOWN .	
Time_of_DayLATE-EARLY:Rd_ConfigurationUNKNOWN .	
Time_of_DayOVERNIGHT:Rd_SurfaceOTHER .	
Time_of_DayLATE-EARLY:Rd_SurfaceOTHER 3.886565e-03	
Time_of_DayOVERNIGHT:Rd_ConditionsICE-SNOW-SLUSH 3.126225e-01	
Time_of_DayLATE-EARLY:Rd_ConditionsICE-SNOW-SLUSH .	
Time_of_DayOVERNIGHT:Rd_ConditionsOTHER 1.702176e+00	
Time_of_DayLATE-EARLY:Rd_ConditionsOTHER .	
Time_of_DayOVERNIGHT:Rd_ConditionsWET .	
Time_of_DayLATE-EARLY:Rd_ConditionsWET -3.696563e-01	
Time_of_DayOVERNIGHT:LightDARK-LIT -6.237493e-01	
Time_of_DayLATE-EARLY:LightDARK-LIT -1.648334e-01	
Time_of_DayOVERNIGHT:LightDARK-NOT-LIT -4.494685e-01	
Time_of_DayLATE-EARLY:LightDARK-NOT-LIT .	
Time of DayOVERNIGHT:LightDAWN .	
Time of DayLATE-EARLY:LightDAWN .	
Time_of_DayOVERNIGHT:LightDUSK .	
Time_of_DayLATE-EARLY:LightDUSK .	
Time_of_DayOVERNIGHT:LightOTHER .	
Time_of_DayLATE-EARLY:LightOTHER .	
Time_of_DayOVERNIGHT:WeatherOTHER -1.873626e+00	
Time of DayLATE-EARLY:WeatherOTHER .	
Time_of_DayOVERNIGHT:WeatherRAIN-SNOW .	
Time_of_DayLATE-EARLY:WeatherRAIN-SNOW .	
Time_of_DayOVERNIGHT:Traffic_ControlSIGNAL-STOP 2.039222e-01	
Time of DayLATE-EARLY:Traffic ControlSIGNAL-STOP 1.024350e-01	
Time_of_DayOVERNIGHT:Work_AreaYES .	
Time of DayLATE-EARLY:Work AreaYES 5.117929e-03	
Rd_FeatureINTERSECTION:Rd_CharacterCURVE .	
Rd FeatureINTERSECTION:Rd ClassOTHER .	
Rd_FeatureINTERSECTION:Rd_ClassUS HWY 1.211543e-01	
Rd_FeatureINTERSECTION:Rd_ConfigurationONE-WAY .	
Rd_FeatureINTERSECTION:Rd_ConfigurationTWO-WAY-PROTECTED-MEDIAN .	
Rd_FeatureINTERSECTION:Rd_ConfigurationTWO-WAY-UNPROTECTED-MEDIAN	
1.228891e-01	
Rd_FeatureINTERSECTION:Rd_ConfigurationUNKNOWN -5.485124e-02	
Rd_FeatureINTERSECTION:Rd_SurfaceOTHER .	
Rd_FeatureINTERSECTION:Rd_ConditionsICE-SNOW-SLUSH .	

Pd FooturaINTERSECTION:Pd ConditionsOTHER	-6.401051e-01
Rd_FeatureINTERSECTION:Rd_ConditionsOTHER Rd_FeatureINTERSECTION:Rd_ConditionsWET	1.479159e-02
Rd_FeatureINTERSECTION:LightDARK-LIT	1.4791396-02
Rd_FeatureINTERSECTION:LightDARK-NOT-LIT	•
	·
Rd_FeatureINTERSECTION:LightDAWN	•
Rd_FeatureINTERSECTION:LightDUSK	-6.385495e-01
Rd_FeatureINTERSECTION: LightOTHER	-0.3854956-01
Rd_FeatureINTERSECTION:WeatherOTHER	. 4 363507 - 04
Rd_FeatureINTERSECTION:WeatherRAIN-SNOW	1.262507e-01
Rd_FeatureINTERSECTION:Traffic_ControlSIGNAL-STOP	•
Rd_FeatureINTERSECTION:Work_AreaYES	•
Rd_CharacterCURVE:Rd_ClassOTHER	•
Rd_CharacterCURVE:Rd_ClassUS HWY	-1.126252e-01
Rd_CharacterCURVE:Rd_ConfigurationONE-WAY	
Rd_CharacterCURVE:Rd_ConfigurationTWO-WAY-PROTE	
Rd_CharacterCURVE:Rd_ConfigurationTWO-WAY-UNPRO	OTECTED-MEDIAN -
6.207546e-02	
Rd_CharacterCURVE:Rd_ConfigurationUNKNOWN	
Rd_CharacterCURVE:Rd_SurfaceOTHER	
Rd_CharacterCURVE:Rd_ConditionsICE-SNOW-SLUSH	
Rd_CharacterCURVE:Rd_ConditionsOTHER	
Rd_CharacterCURVE:Rd_ConditionsWET	
Rd_CharacterCURVE:LightDARK-LIT	-2.585268e-02
Rd_CharacterCURVE:LightDARK-NOT-LIT	
Rd_CharacterCURVE:LightDAWN	
Rd_CharacterCURVE:LightDUSK	
Rd_CharacterCURVE:LightOTHER	
Rd_CharacterCURVE:WeatherOTHER	
Rd_CharacterCURVE:WeatherRAIN-SNOW	-4.750677e-01
Rd_CharacterCURVE:Traffic_ControlSIGNAL-STOP	-3.084268e-02
Rd_CharacterCURVE:Work_AreaYES	7.838758e-01
Rd ClassOTHER:Rd ConfigurationONE-WAY	
Rd ClassUS HWY:Rd ConfigurationONE-WAY	-2.499031e-01
Rd_ClassOTHER:Rd_ConfigurationTWO-WAY-PROTECTED	D-MEDIAN 1.810607e-
01	
Rd_ClassUS HWY:Rd_ConfigurationTWO-WAY-PROTECTE	D-MEDIAN .
Rd_ClassOTHER:Rd_ConfigurationTWO-WAY-UNPROTEC	
Rd_ClassUS HWY:Rd_ConfigurationTWO-WAY-UNPROTE	
4.558398e-01	
Rd_ClassOTHER:Rd_ConfigurationUNKNOWN	

Rd_ClassUS HWY:Rd_ConfigurationUNKNOWN	
Rd_ClassOTHER:Rd_SurfaceOTHER	-5.307316e-01
Rd_ClassUS HWY:Rd_SurfaceOTHER	-3.30/3106-01
Rd_ClassOTHER:Rd_ConditionsICE-SNOW-SLUSH	2.197920e-01
Rd_ClassUS HWY:Rd_ConditionsICE-SNOW-SLUSH	2.1373200 01
Rd_ClassOTHER:Rd_ConditionsOTHER	·
Rd_ClassUS HWY:Rd_ConditionsOTHER	·
Rd_ClassOTHER:Rd_ConditionsWET	·
Rd_ClassUS HWY:Rd_ConditionsWET	•
Rd_ClassOTHER:LightDARK-LIT	·
Rd_ClassUS HWY:LightDARK-LIT	•
Rd_ClassOTHER:LightDARK-NOT-LIT	·
Rd_ClassUS HWY:LightDARK-NOT-LIT	-7.629586e-02
Rd_ClassOTHER:LightDAWN	7.0233000 02
Rd_ClassUS HWY:LightDAWN	-5.748051e-01
Rd_ClassOTHER:LightDUSK	3.7400310 01
Rd_ClassUS HWY:LightDUSK	·
Rd_ClassOTHER:LightOTHER	·
Rd_ClassUS HWY:LightOTHER	·
Rd_ClassOTHER:WeatherOTHER	•
Rd_ClassUS HWY:WeatherOTHER	·
Rd_ClassOTHER:WeatherRAIN-SNOW	·
Rd_ClassUS HWY:WeatherRAIN-SNOW	·
Rd_ClassOTHER:Traffic_ControlSIGNAL-STOP	
Rd_ClassUS HWY:Traffic_ControlSIGNAL-STOP	
Rd_ClassOTHER:Work_AreaYES	
Rd_ClassUS HWY:Work_AreaYES	2.459279e-01
Rd_ConfigurationONE-WAY:Rd_SurfaceOTHER	
Rd_ConfigurationTWO-WAY-PROTECTED-MEDIAN:R	d SurfaceOTHER
Rd ConfigurationTWO-WAY-UNPROTECTED-MEDIA	_
Rd_ConfigurationUNKNOWN:Rd_SurfaceOTHER	
Rd_ConfigurationONE-WAY:Rd_ConditionsICE-SNOV	W-SLUSH .
Rd_ConfigurationTWO-WAY-PROTECTED-MEDIAN:R	
Rd_ConfigurationTWO-WAY-UNPROTECTED-MEDIA	_
Rd_ConfigurationUNKNOWN:Rd_ConditionsICE-SNC	_
Rd_ConfigurationONE-WAY:Rd_ConditionsOTHER	
Rd_ConfigurationTWO-WAY-PROTECTED-MEDIAN:R	d ConditionsOTHER -
1.420038e+00	
Rd_ConfigurationTWO-WAY-UNPROTECTED-MEDIA	N:Rd ConditionsOTHER .
Rd_ConfigurationUNKNOWN:Rd_ConditionsOTHER	

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Rd ConfigurationONE-WAY:Rd ConditionsWET
Rd ConfigurationTWO-WAY-PROTECTED-MEDIAN:Rd ConditionsWET
Rd ConfigurationTWO-WAY-UNPROTECTED-MEDIAN:Rd ConditionsWET
Rd ConfigurationUNKNOWN:Rd ConditionsWET
Rd ConfigurationONE-WAY:LightDARK-LIT
                                                  -1.399179e-02
Rd ConfigurationTWO-WAY-PROTECTED-MEDIAN:LightDARK-LIT -2.227619e-
01
Rd ConfigurationTWO-WAY-UNPROTECTED-MEDIAN:LightDARK-LIT
                                                              -3.531091e-
01
Rd_ConfigurationUNKNOWN:LightDARK-LIT
Rd_ConfigurationONE-WAY:LightDARK-NOT-LIT
Rd ConfigurationTWO-WAY-PROTECTED-MEDIAN:LightDARK-NOT-LIT
Rd ConfigurationTWO-WAY-UNPROTECTED-MEDIAN:LightDARK-NOT-LIT
Rd ConfigurationUNKNOWN:LightDARK-NOT-LIT
Rd ConfigurationONE-WAY:LightDAWN
Rd ConfigurationTWO-WAY-PROTECTED-MEDIAN:LightDAWN
Rd ConfigurationTWO-WAY-UNPROTECTED-MEDIAN:LightDAWN
Rd ConfigurationUNKNOWN:LightDAWN
Rd ConfigurationONE-WAY:LightDUSK
                                                 5.997676e-01
Rd ConfigurationTWO-WAY-PROTECTED-MEDIAN:LightDUSK
Rd_ConfigurationTWO-WAY-UNPROTECTED-MEDIAN:LightDUSK
Rd ConfigurationUNKNOWN:LightDUSK
Rd ConfigurationONE-WAY:LightOTHER
Rd ConfigurationTWO-WAY-PROTECTED-MEDIAN:LightOTHER
                                                          -1.602921e+00
Rd_ConfigurationTWO-WAY-UNPROTECTED-MEDIAN:LightOTHER
Rd ConfigurationUNKNOWN:LightOTHER
Rd_ConfigurationONE-WAY:WeatherOTHER
                                                   -9.420533e-01
Rd ConfigurationTWO-WAY-PROTECTED-MEDIAN:WeatherOTHER
Rd ConfigurationTWO-WAY-UNPROTECTED-MEDIAN:WeatherOTHER
Rd ConfigurationUNKNOWN:WeatherOTHER
Rd ConfigurationONE-WAY:WeatherRAIN-SNOW
Rd ConfigurationTWO-WAY-PROTECTED-MEDIAN:WeatherRAIN-SNOW
Rd ConfigurationTWO-WAY-UNPROTECTED-MEDIAN:WeatherRAIN-SNOW
Rd ConfigurationUNKNOWN:WeatherRAIN-SNOW
Rd ConfigurationONE-WAY:Traffic ControlSIGNAL-STOP
                                                       -2.499343e-01
Rd ConfigurationTWO-WAY-PROTECTED-MEDIAN:Traffic ControlSIGNAL-STOP
Rd_ConfigurationTWO-WAY-UNPROTECTED-MEDIAN:Traffic_ControlSIGNAL-STOP .
Rd_ConfigurationUNKNOWN:Traffic_ControlSIGNAL-STOP
Rd_ConfigurationONE-WAY:Work AreaYES
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Rd_ConfigurationTWO-WAY-PROTECTED-MEDIAN:Wor	rk_AreaYES 1.917471e-
01	
Rd_ConfigurationTWO-WAY-UNPROTECTED-MEDIAN:V	Work_AreaYES .
Rd_ConfigurationUNKNOWN:Work_AreaYES	
Rd_SurfaceOTHER:Rd_ConditionsICE-SNOW-SLUSH	
Rd_SurfaceOTHER:Rd_ConditionsOTHER	
Rd_SurfaceOTHER:Rd_ConditionsWET	-8.372690e-02
Rd_SurfaceOTHER:LightDARK-LIT	
Rd_SurfaceOTHER:LightDARK-NOT-LIT	-9.233593e-02
Rd_SurfaceOTHER:LightDAWN	
Rd_SurfaceOTHER:LightDUSK	-1.428736e-01
Rd_SurfaceOTHER:LightOTHER	
Rd_SurfaceOTHER:WeatherOTHER	
Rd_SurfaceOTHER:WeatherRAIN-SNOW	
Rd_SurfaceOTHER:Traffic_ControlSIGNAL-STOP	
Rd SurfaceOTHER:Work AreaYES	
Rd_ConditionsICE-SNOW-SLUSH:LightDARK-LIT	-3.768304e-01
Rd_ConditionsOTHER:LightDARK-LIT	
Rd_ConditionsWET:LightDARK-LIT	
Rd_ConditionsICE-SNOW-SLUSH:LightDARK-NOT-LIT	1.077994e+00
Rd_ConditionsOTHER:LightDARK-NOT-LIT	
Rd_ConditionsWET:LightDARK-NOT-LIT	
Rd_ConditionsICE-SNOW-SLUSH:LightDAWN	
Rd_ConditionsOTHER:LightDAWN	
Rd_ConditionsWET:LightDAWN	
Rd ConditionsICE-SNOW-SLUSH:LightDUSK	
Rd_ConditionsOTHER:LightDUSK	2.736671e-01
Rd_ConditionsWET:LightDUSK	
Rd_ConditionsICE-SNOW-SLUSH:LightOTHER	
Rd_ConditionsOTHER:LightOTHER	
Rd_ConditionsWET:LightOTHER	
Rd_ConditionsICE-SNOW-SLUSH:WeatherOTHER	
Rd_ConditionsOTHER:WeatherOTHER	
Rd_ConditionsWET:WeatherOTHER	-2.477240e-01
Rd_ConditionsICE-SNOW-SLUSH:WeatherRAIN-SNOW	
Rd_ConditionsOTHER:WeatherRAIN-SNOW	
Rd_ConditionsWET:WeatherRAIN-SNOW	6.471902e-03
Rd_ConditionsICE-SNOW-SLUSH:Traffic_ControlSIGNAL	STOP .
Rd_ConditionsOTHER:Traffic_ControlSIGNAL-STOP	
Rd_ConditionsWET:Traffic_ControlSIGNAL-STOP	

Rd_ConditionsICE-SNOW-SLUSH:Work_AreaYES	
Rd_ConditionsOTHER:Work_AreaYES	6.774153e+00
Rd_ConditionsWET:Work_AreaYES	
LightDARK-LIT:WeatherOTHER	
LightDARK-NOT-LIT:WeatherOTHER	
LightDAWN:WeatherOTHER	
LightDUSK:WeatherOTHER	
LightOTHER:WeatherOTHER	
LightDARK-LIT:WeatherRAIN-SNOW	1.776019e-01
LightDARK-NOT-LIT:WeatherRAIN-SNOW	
LightDAWN:WeatherRAIN-SNOW	
LightDUSK:WeatherRAIN-SNOW	
LightOTHER:WeatherRAIN-SNOW	
LightDARK-LIT:Traffic_ControlSIGNAL-STOP	2.205936e-01
LightDARK-NOT-LIT:Traffic_ControlSIGNAL-STOP	
LightDAWN:Traffic_ControlSIGNAL-STOP	
LightDUSK:Traffic_ControlSIGNAL-STOP	
LightOTHER:Traffic_ControlSIGNAL-STOP	-2.812067e-01
LightDARK-LIT:Work_AreaYES	5.713643e-01
LightDARK-NOT-LIT:Work_AreaYES	
LightDAWN:Work_AreaYES	
LightDUSK:Work_AreaYES	
LightOTHER:Work_AreaYES	
WeatherOTHER:Traffic_ControlSIGNAL-STOP	-4.918214e-02
WeatherRAIN-SNOW :Traffic_ControlSIGNAL-STOP	5.792116e-02
WeatherOTHER:Work_AreaYES	
WeatherRAIN-SNOW :Work_AreaYES	
Traffic_ControlSIGNAL-STOP:Work_AreaYES	