Coursera - Applied Data Science Capstone

Ana Cristina Bittar de Oliveira

Capstone Project - Week 1

1. A description of the problem and a discussion of the background. (15 marks)

How to prevent or, at least, reduce the severity of road accidents is a very important question. A better understanding of the probable causes of collisions could be a good approach in order to suggest actions to be taken or improvements to be made in traffic engineering systems willing to prevent them.

The main potential consequences of a car collision are:

- Injuries
- Traffic jams
- Related costs (government, insurance companies, individuals, et al)

Seattle Traffic Management Division releases weekly updates about all collisions recorded by Traffic Records since 2004. The main goal is to analyze that data to identify patterns and make predictions about risk and severity of the accidents based on some common attributes. With this information an alert system could be developed to provide guidance on what to do in short, medium, or large term.

Some examples of actions to be taken could be:

- deviate traffic in pre-defined times or weather conditions (preventive)
- improve traffic signs
- reduce speed in specific roads in determined weather conditions
- reduce speed in specific roads (permanently)
- make structural changes on traffic
- educative campaigns and advertisement
- recommendations to improve the data collection
- other

The present work involves data analysis related to collisions in the city of Seattle. It would be out of scope to generalize this model for sites other than those included on the data to be analyzed, although intuition indicates that, for similar traffic systems, the results could be extended. The expected result is a model capable to predict an accident severity in pre-determined conditions.

2. A description of the data and how it will be used to solve the problem. (15 marks)

This project will use the dataset shared in the Capstone. The CSV file was obtained directly from Seattle Open Data Portal.

https://data-

seattlecitygis.opendata.arcgis.com/datasets/5b5c745e0f1f48e7a53acec63a0022ab 0/data

The dataset provides information about all types of collisions occurred in the city of Seattle since 2004 and is updated weekly. It was also provided metadata for better understanding of the dataset.

Attribute Information:

https://www.seattle.gov/Documents/Departments/SDOT/GIS/Collisions OD.pdf

The data will be used to make predictions about probability and severity of car accidents based mainly on:

- Weather conditions
- Road conditions
- Light conditions

Further analysis (maybe using additional datasets) could be made to check if it's possible to correlate some driver conditions (as speed, influence of drugs or alcohol, inattention) with other attributes like time of the day, day of the week, events like concerts or games (if the data is available) and define whether or not those attributes could be used for predictions. This is out of scope of the present report.

Some attributes are redundant, or the number of observations is irrelevant, and will be excluded.

Info about the original data file:

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 221389 entries, 0 to 221388
Data columns (total 40 columns):
    # Column Non-Null Count Dtype
```

0	X	213918 non-null 213918 non-null 221389 non-null	float64
	Y	213918 non-null	float64
2	OBJECTID INCKEY	221389 non-null	int64
3	INCKEY	221389 non-null	int64
4	COLDETKEY	221389 non-null	int.64
_	DEDODUMO	221200 11	a la -i a a +
6	STATUS ADDRTYPE INTKEY LOCATION EXCEPTRSNCODE	221389 non-null	object
7	ADDRTYPE	217677 non-null	object
8	INTKEY	71884 non-null	float64
9	LOCATION	216801 non-null	object
10	EXCEPTRSNCODE	100986 non-null	object
ΤТ	EXCEPTRSNDESC	II//9 non-null	object
	SEVERITYCODE		
13	SEVERITYDESC	221389 non-null	object
14	COLLISIONTYPE	195159 non-null	object
15	PERSONCOUNT	221389 non-null	int64
16	PEDCOUNT	221389 non-null	int64
17	PEDCYLCOUNT VEHCOUNT	221389 non-null	int64
18	VEHCOUNT	221389 non-null	int64
19	INJURIES	221389 non-null	int64
20	SERIOUSINJURIES	221389 non-null	int64
21	FATALITIES	221389 non-null	int64
22	FATALITIES INCDATE INCDTTM	221389 non-null	object
23	INCDTTM	221389 non-null	datetime64[ns]
24	JUNCTIONTYPE	209417 non-null	object
25	SDOT_COLCODE SDOT_COLDESC	221388 non-null	float64
26	SDOT_COLDESC	221388 non-null	object
27	TND $TMTMTMTMTMD$	30188 non-niill	ohiact
28	UNDERINFL WEATHER ROADCOND	195179 non-null	object
29	WEATHER	194969 non-null	object
30	ROADCOND	195050 non-null	object
31	LIGHTCOND	194880 non-null	object

```
32 PEDROWNOTGRNT 5192 non-null
                                    object
 33 SDOTCOLNUM
                   127205 non-null float64
 34 SPEEDING
                   9928 non-null
                                    object
 35 ST COLCODE
                    211976 non-null object
 36 ST_COLDESC
                    195159 non-null object
 37 SEGLANEKEY
                    221389 non-null int64
 38 CROSSWALKKEY
                   221389 non-null int64
 39 HITPARKEDCAR
                   221389 non-null object
dtypes: datetime64[ns](1), float64(5), int64(12), object(22)
memory usage: 67.6+ MB
```

The table below shows statistics about some numerical variables related to severity. In a supervised model those attributes could be used to define different weights for severity scale.

	PERSONCOUNT	PEDCOUNT	PEDCYLCOUNT	VEHCOUNT	INJURIES	SERIOUSINJURIES	FATALITIES
Count	221,389	221,389	221,389	221,389	221,389	221,389	221,389
Mean	2.227161	0.038136	0.027350	1.731057	0.373962	0.015209	0.001685
Std	1.470190	0.201815	0.164508	0.829259	0.732158	0.158072	0.044701
Min	0.0	0.0	0.0	0.0	0.0	0.0	0.0
25%	2.0	0.0	0.0	2.0	0.0	0.0	0.0
50%	2.0	0.0	0.0	2.0	0.0	0.0	0.0
75%	3.0	0.0	0.0	2.0	1.0	0.0	0.0
Max	93	6	2	15	78	41	5

Rows with missing values for the main attributes were removed.

	SEVERITYCODE	WEATHER	ROADCOND	LIGHTCOND
Count	194699	194699	194699	194699
Unique	5	11	9	9
Тор	1	Clear	Dry	Daylight
Freq.	133575	114556	128304	119384

All the observations with SEVERITYCODE = "0", described as "Unknown" were removed from dataset.

Some aspects that should be observed are:

- There's very detailed geographic information, but no demographic about the conductors
- The data went back 15 years, but traffic volume increases rapidly, but the first model won't differ the significance of most recent occurrences

First attempt will be an unsupervised model.