

# CAPSTONE PROJECT – CAR ACCIDENT SEVERITY

## 1 Data collection and understanding

### 1.1 Data source

We used shared data for Seattle city as basis to deal with the accidents data. At first glance at the CSV file, we could see what type of data we have with us. The target label for the data set is Severity, which describes fatality of an accident. The remaining columns have different types of attributes. Also noticed that the data had some unbalanced attributes which need to be attended in next steps.

We also used the collisions meta data available for the years to understand the nature of all attributes. Having about 1.95L data observations, we could plan for split of the observations that could be used to train and test the prospective model.

### 1.2 Data understanding

The dataset basics are provided as follows;

**Title:** Collisions—All Years

**Abstract:** All collisions provided by Traffic Records.

**Description:** This includes all types of collisions. Collisions are displayed at the intersection or mid-block of a segment in the Annexure.

**Timeframe:** 2004 to Present.

**Keyword(s):** SDOT, Seattle, Transportation, Accidents, Bicycle, Car, Collisions, Pedestrian, Traffic, Vehicle

**Types:** The data is a mix of numerical and categorical types.

The data set provides labelled data for severity of accident. It shows a dual class categorical type of variable. The attributes (36 columns) convey information mainly about;

- the incident: such as identification no., location coordinates, date, time etc.
- the collision: such as code, type, description, injuries, fatalities etc.
- the impact: such as count of pedestrians, cyclists, vehicles involved etc.
- preconditions: such as inattention, influence of drugs, road condition, weather, speeding etc.

Attributes are almost complete with the information such as name, data type, length and description as shown in next section. State Collision Code Dictionary comprising about 85 codes with descriptions is also provided in supplement.

With the given dataset, severity code is identified as the target variable (labelled or dependent) while rest of the fields are noted as independent variables or the attributes. The case objective along with given data does qualify it as a classification problem of the supervised machine learning. All columns that could influence the cause and impact of an accident need to be selected for training and testing the model.

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## 2 Data preparation

### 2.1 Basic insight of dataset

After reading data into Pandas data frame, it became a good start to explore the dataset. Following ways are followed to obtain essential insights of the data to help better understand the dataset;

#### Columns:

It provides list of columns that exist in the dataset.

#### Data types:

This step is to know the variety of types viz. object, float, int, bool and datetime64. In order to better learn about each attribute, it is necessary to know the data type of each column, which were identified using Python as in screen shot below;

```
[4]: df.dtypes

[4]: SEVERITYCODE      int64
      X              float64
      Y              float64
      OBJECTID        int64
      INCKEY          int64
      COLDETKEY        int64
      REPORTNO        object
      STATUS          object
      ADDRTYPE        object
      INTKEY          float64
      LOCATION        object
      EXCEPTSNCODE  object
      EXCEPTSNDESC  object
      SEVERITYCODE.1  int64
      SEVERITYDESC    object
      COLLISIONTYPE  object
      PERSONCOUNT   int64
      PEDCOUNT       int64
      PEDCYLCOUNT     int64
      VEHCOUNT       int64
      INCDATE         object
      INCDTM         object
      JUNCTIONTYPE    object
      SDOT_COLCODE    int64
      SDOT_COLDESC    object
      INATTENTIONIND  object
      UNDERINFL      object
      WEATHER         object
      ROADCOND        object
      LIGHTCOND       object
      PEDROWNOTGRNT   object
      SDOTCOLNUM      float64
      SPEEDING        object
      ST_COLCODE      object
      ST_COLDESC      object
      SEGLANEKEY      int64
      CROSSWALKKEY    int64
      HITPARKEDCAR    object
      dtype: object
```

#### Data description:

We could get statistical summary, such as count, unique value, column mean value, column standard deviation, etc of each column. It provides various summary statistics, excluding NaN (Not a Number) values.

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```
[6]: df.describe(include='all')
```

	SEVERITYCODE	X	Y	OBJECTID	INCKEY	COLDEKEY	REPORTNO	STATUS	ADDRTYPE	INTKEY	...	ROADCOND	LIGHTCOND	PEDROWNO
count	194673.000000	189339.000000	189339.000000	194673.000000	194673.000000	194673.000000	194673	194673	192747	65070.000000	...	189661	189503	
unique	NaN	NaN	NaN	NaN	NaN	NaN	194670	2	3	NaN	...	9	9	
top	NaN	NaN	NaN	NaN	NaN	NaN	1782439	Matched	Block	NaN	...	Dry	Daylight	
freq	NaN	NaN	NaN	NaN	NaN	NaN	2	189786	126926	NaN	...	124510	116137	
mean	1.298901	-122.330518	47.619543	108479.364930	141091.456350	141298.811381	NaN	NaN	NaN	37558.450576	...	NaN	NaN	
std	0.457778	0.029976	0.056157	62649.722558	86634.402737	86986.542110	NaN	NaN	NaN	51745.990273	...	NaN	NaN	
min	1.000000	-122.419091	47.495573	1.000000	1001.000000	1001.000000	NaN	NaN	NaN	23807.000000	...	NaN	NaN	
25%	1.000000	-122.348673	47.575956	54267.000000	70383.000000	70383.000000	NaN	NaN	NaN	28667.000000	...	NaN	NaN	
50%	1.000000	-122.330224	47.615369	106912.000000	123363.000000	123363.000000	NaN	NaN	NaN	29973.000000	...	NaN	NaN	
75%	2.000000	-122.311937	47.663664	162272.000000	203319.000000	203459.000000	NaN	NaN	NaN	33973.000000	...	NaN	NaN	
max	2.000000	-122.238949	47.734142	219547.000000	331454.000000	332954.000000	NaN	NaN	NaN	757580.000000	...	NaN	NaN	

11 rows x 38 columns

## Info:

Similar to above, but it provides a concise summary of the DataFrame as shown below;

```
RangeIndex: 194673 entries, 0 to 194672
Data columns (total 38 columns):
#   Column                Non-Null Count  Dtype
---  ---
0   SEVERITYCODE          194673 non-null int64
1   X                     189339 non-null float64
2   Y                     189339 non-null float64
3   OBJECTID              194673 non-null int64
4   INCKEY                194673 non-null int64
5   COLDEKEY              194673 non-null int64
6   REPORTNO              194673 non-null object
7   STATUS                194673 non-null object
8   ADDRTYPE              192747 non-null object
9   INTKEY                65070 non-null float64
10  LOCATION              191996 non-null object
11  EXCEPTRSNCODE       84811 non-null object
12  EXCEPTRSNDESC       5638 non-null object
13  SEVERITYCODE.1        194673 non-null int64
14  SEVERITYDESC           194673 non-null object
15  COLLISIONTYPE         189769 non-null object
16  PERSONCOUNT          194673 non-null int64
17  PEDCOUNT             194673 non-null int64
18  PEDCYLCOUNT           194673 non-null int64
19  VEHCOUNT             194673 non-null int64
20  INCDATE                194673 non-null object
21  INCDTTM               194673 non-null object
22  JUNCTIONTYPE          188344 non-null object
23  SDOT_COLCODE          194673 non-null int64
24  SDOT_COLDESC           194673 non-null object
25  INATTENTIONIND        29805 non-null object
26  UNDERINFL            189789 non-null object
27  WEATHER                189592 non-null object
28  ROADCOND              189661 non-null object
29  LIGHTCOND             189503 non-null object
30  PEDROWNOTGRNT         4667 non-null object
31  SDOTCOLNUM            114936 non-null float64
32  SPEEDING              9333 non-null object
33  ST_COLCODE            194655 non-null object
34  ST_COLDESC            189769 non-null object
35  SEGLANEKEY            194673 non-null int64
36  CROSSWALKKEY          194673 non-null int64
37  HITPARKEDCAR          194673 non-null object
dtypes: float64(4), int64(12), object(22)
memory usage: 56.4+ MB
```

## 2.2 Feature selection

In the first screening, it was noticed that some of the attributes were not significant to the cause of or to assess the impact of the severity. So these could be dropped for removing bias during design of the model. Rest of the columns were retained for further analysis.

## 2.3 Cleansing data

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In the second step, columns were analysed for missing values. Columns were treated for substituting missing values as shown in the table below;

Sr. No.	Attribute	Data type, length	Description	Wrangling Method	Rationale
1	OBJECTID	OBJECTID	ESRI unique identifier	Dropped	Insignificance
2	X	Longitude	ESRI geometry field	Dropped	Insignificance
3	Y	Latitude	ESRI geometry field	Dropped	Insignificance
4	ADDRTYPE	Text, 12	Collision address type: • Alley • Block • Intersection	Retained	1% missing data, replaced by max. frequency
5	INTKEY	Double	Key that corresponds to the intersection associated with a collision	Dropped	Insignificance
6	LOCATION	Text, 255	Description of the general location of the collision	Dropped	Insignificance
7	EXCEPTRSNCODE	Text, 10		Dropped	Insignificance
8	EXCEPTRSNDESC	Text, 300		Dropped	Insignificance
9	SEVERITYCODE	Text, 100	A code that corresponds to the severity of the collision: • 3—fatality • 2b—serious injury • 2—injury • 1—prop damage • 0—unknown	Retained	Target variable
10	SEVERITYDESC	Text	A detailed description of the severity of the collision	Retained	Target variable
11	COLLISIONTYPE	Text, 300	Collision type	Retained	2.5% missing data, replaced by max. frequency
12	PERSONCOUNT	Double	The total number of people involved in the collision	Retained	
13	PEDCOUNT	Double	The number of pedestrians involved in the collision. This is entered by the state.	Retained	
14	PEDCYLCOUNT	Double	The number of bicycles involved in the collision. This is entered by the state.	Retained	
15	VEHCOUNT	Double	The number of vehicles involved in the collision. This is entered by the state.	Retained	

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Sr. No.	Attribute	Data type, length	Description	Wrangling Method	Rationale
16	INJURIES	Double	The number of total injuries involved in the collision. This is entered by the state.	Retained	
17	SERIOUSINJURIES	Double	The number of serious injuries involved in the collision. This is entered by the state.	Retained	
18	FATALITIES	Double	The number of fatalities involved in the collision. This is entered by the state.	Retained	
19	INCDATE	Date	The date of the incident.	Dropped	Insignificance
20	INCDTTM	Text, 30	The date and time of the incident.	Dropped	Insignificance
21	JUNCTIONTYPE	Text, 300	Category of junction at which collision took place	Retained	3.3% missing data, replaced by max. frequency
22	SDOT_COLCODE	Text, 10	A code given to the collision by SDOT.	Dropped	Insignificance
23	SDOT_COLDESC	Text, 300	A description of the collision corresponding to the collision code.		
24	INATTENTIONIND	Text, 1	Whether or not collision was due to inattention (Y/N).	Dropped	85% data is missing
25	UNDERINFL	Text, 10	Whether or not a driver involved was under the influence of drugs or alcohol.	Dropped	only 3% observations are influencing
26	WEATHER	Text, 300	A description of the weather conditions during the time of the collision.	Retained	2.6% missing data, replaced by max. frequency
27	ROADCOND	Text, 300	The condition of the road during the collision.	Retained	2.6% missing data, replaced by max. frequency
28	LIGHTCOND	Text, 300	The light conditions during the collision.	Retained	2.7% missing data, replaced by max. frequency
29	PEDROWNOTGRNT	Text, 1	Whether or not the pedestrian right of way was not granted. (Y/N)	Dropped	97.6% data missing
30	SDOTCOLNUM	Text, 10	A number given to the collision by SDOT.	Dropped	Insignificance

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Sr. No.	Attribute	Data type, length	Description	Wrangling Method	Rationale
31	SPEEDING	Text, 1	Whether or not speeding was a factor in the collision. (Y/N)	Dropped	only 3% observations are influencing, rest data unavailable
32	ST_COLCODE	Text, 10	A code provided by the state that describes the collision. For more information about these codes, please see the State Collision Code Dictionary.	Dropped	Insignificance
33	ST_COLDESC	Text, 300	A description that corresponds to the state's coding designation.	Retained	2.5% missing data, replaced by max. frequency
34	SEGLANEKEY	Long	A key for the lane segment in which the collision occurred.	Dropped	Insignificance
35	CROSSWALKKEY	Long	A key for the crosswalk in which the collision occurred.	Dropped	Insignificance
36	HITPARKEDCAR	Text, 1	Whether or not the collision involved hitting a parked car. (Y/N)	Dropped	only 3.7% observations are influencing
37	STATUS	Text, 10	Matched, Unmatched	Dropped	Insignificance
38	REPORTNO	Long	Sr. No. of report for internal purposes	Dropped	Insignificance

### 2.4 Transforming data

The last step in data cleansing would be to check and make sure that all data is in the correct format (int, float, text or other). To use categorical variables for regression analysis, indicator variables (or dummy variable) were used for transforming categorical variables into binary values (0s and 1s). This would make the data ready for next tests of correlation and determining significance. The results are as shown in screen shots below;

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```
[15]: df_clean.info(max_cols=157)
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 194673 entries, 0 to 194672  
Data columns (total 156 columns):  
# Column
```

	Non-Null Count	Dtype
0 SEVERITYCODE	194673 non-null	int64
1 SEVERITYDESC	194673 non-null	object
2 PERSONCOUNT	194673 non-null	int64
3 PEDCOUNT	194673 non-null	int64
4 PEDCYLCOUNT	194673 non-null	int64
5 VEHCOUNT	194673 non-null	int64
6 ADDRTYPE_Alley	194673 non-null	uint8
7 ADDRTYPE_Block	194673 non-null	uint8
8 ADDRTYPE_Intersection	194673 non-null	uint8
9 COLLISIONTYPE_Angles	194673 non-null	uint8
10 COLLISIONTYPE_Cycles	194673 non-null	uint8
11 COLLISIONTYPE_Head On	194673 non-null	uint8
12 COLLISIONTYPE_Left Turn	194673 non-null	uint8
13 COLLISIONTYPE_Other	194673 non-null	uint8
14 COLLISIONTYPE_Parked Car	194673 non-null	uint8
15 COLLISIONTYPE_Pedestrian	194673 non-null	uint8
16 COLLISIONTYPE_Rear Ended	194673 non-null	uint8
17 COLLISIONTYPE_Right Turn	194673 non-null	uint8
18 COLLISIONTYPE_Sideswipe	194673 non-null	uint8
19 JUNCTIONTYPE_At Intersection (but not related to intersection)	194673 non-null	uint8
20 JUNCTIONTYPE_At Intersection (intersection related)	194673 non-null	uint8
21 JUNCTIONTYPE_Driveway Junction	194673 non-null	uint8
22 JUNCTIONTYPE_Mid-Block (but intersection related)	194673 non-null	uint8
23 JUNCTIONTYPE_Mid-Block (not related to intersection)	194673 non-null	uint8
24 JUNCTIONTYPE_Ramp Junction	194673 non-null	uint8
25 JUNCTIONTYPE_Unknown	194673 non-null	uint8
26 SDOT_COLDESC_DRIVERLESS VEHICLE RAN OFF ROAD - HIT FIXED OBJECT	194673 non-null	uint8
27 SDOT_COLDESC_DRIVERLESS VEHICLE RAN OFF ROAD - NO COLLISION	194673 non-null	uint8
28 SDOT_COLDESC_DRIVERLESS VEHICLE STRUCK MOTOR VEHICLE FRONT END AT ANGLE	194673 non-null	uint8
29 SDOT_COLDESC_DRIVERLESS VEHICLE STRUCK MOTOR VEHICLE LEFT SIDE AT ANGLE	194673 non-null	uint8
30 SDOT_COLDESC_DRIVERLESS VEHICLE STRUCK MOTOR VEHICLE LEFT SIDE SIDESWIPE	194673 non-null	uint8
31 SDOT_COLDESC_DRIVERLESS VEHICLE STRUCK MOTOR VEHICLE REAR END	194673 non-null	uint8
32 SDOT_COLDESC_DRIVERLESS VEHICLE STRUCK MOTOR VEHICLE RIGHT SIDE AT ANGLE	194673 non-null	uint8
33 SDOT_COLDESC_DRIVERLESS VEHICLE STRUCK MOTOR VEHICLE RIGHT SIDE SIDESWIPE	194673 non-null	uint8
34 SDOT_COLDESC_DRIVERLESS VEHICLE STRUCK OBJECT IN ROADWAY	194673 non-null	uint8
35 SDOT_COLDESC_DRIVERLESS VEHICLE STRUCK PEDESTRIAN	194673 non-null	uint8
36 SDOT_COLDESC_MOTOR VEHICLE STRUCK PEDESTRIAN	194673 non-null	uint8
37 SDOT_COLDESC_MOTOR VEHICLE OVERTURNED IN ROAD	194673 non-null	uint8
38 SDOT_COLDESC_MOTOR VEHICLE RAN OFF ROAD - HIT FIXED OBJECT	194673 non-null	uint8
39 SDOT_COLDESC_MOTOR VEHICLE RAN OFF ROAD - NO COLLISION	194673 non-null	uint8
40 SDOT_COLDESC_MOTOR VEHICLE STRUCK MOTOR VEHICLE, FRONT END AT ANGLE	194673 non-null	uint8
41 SDOT_COLDESC_MOTOR VEHICLE STRUCK MOTOR VEHICLE, LEFT SIDE AT ANGLE	194673 non-null	uint8
42 SDOT_COLDESC_MOTOR VEHICLE STRUCK MOTOR VEHICLE, LEFT SIDE SIDESWIPE	194673 non-null	uint8
43 SDOT_COLDESC_MOTOR VEHICLE STRUCK MOTOR VEHICLE, REAR END	194673 non-null	uint8
44 SDOT_COLDESC_MOTOR VEHICLE STRUCK MOTOR VEHICLE, RIGHT SIDE AT ANGLE	194673 non-null	uint8
45 SDOT_COLDESC_MOTOR VEHICLE STRUCK MOTOR VEHICLE, RIGHT SIDE SIDESWIPE	194673 non-null	uint8
46 SDOT_COLDESC_MOTOR VEHICLE STRUCK OBJECT IN ROAD	194673 non-null	uint8
47 SDOT_COLDESC_MOTOR VEHICLE STRUCK PEDALCYCLIST, FRONT END AT ANGLE	194673 non-null	uint8
48 SDOT_COLDESC_MOTOR VEHICLE STRUCK PEDALCYCLIST, LEFT SIDE SIDESWIPE	194673 non-null	uint8
49 SDOT_COLDESC_MOTOR VEHICLE STRUCK PEDALCYCLIST, REAR END	194673 non-null	uint8
50 SDOT_COLDESC_MOTOR VEHICLE STRUCK PEDALCYCLIST, RIGHT SIDE SIDESWIPE	194673 non-null	uint8
51 SDOT_COLDESC_MOTOR VEHICLE STRUCK TRAIN	194673 non-null	uint8
52 SDOT_COLDESC_NOT ENOUGH INFORMATION / NOT APPLICABLE	194673 non-null	uint8
53 SDOT_COLDESC_PEDALCYCLIST OVERTURNED IN ROAD	194673 non-null	uint8
54 SDOT_COLDESC_PEDALCYCLIST RAN OFF ROAD - HIT FIXED OBJECT	194673 non-null	uint8
55 SDOT_COLDESC_PEDALCYCLIST STRUCK MOTOR VEHICLE FRONT END AT ANGLE	194673 non-null	uint8
56 SDOT_COLDESC_PEDALCYCLIST STRUCK MOTOR VEHICLE LEFT SIDE AT ANGLE	194673 non-null	uint8
57 SDOT_COLDESC_PEDALCYCLIST STRUCK MOTOR VEHICLE LEFT SIDE SIDESWIPE	194673 non-null	uint8
58 SDOT_COLDESC_PEDALCYCLIST STRUCK MOTOR VEHICLE REAR END	194673 non-null	uint8
59 SDOT_COLDESC_PEDALCYCLIST STRUCK MOTOR VEHICLE RIGHT SIDE AT ANGLE	194673 non-null	uint8
60 SDOT_COLDESC_PEDALCYCLIST STRUCK MOTOR VEHICLE RIGHT SIDE SIDESWIPE	194673 non-null	uint8
61 SDOT_COLDESC_PEDALCYCLIST STRUCK OBJECT IN ROAD	194673 non-null	uint8
62 SDOT_COLDESC_PEDALCYCLIST STRUCK PEDALCYCLIST FRONT END AT ANGLE	194673 non-null	uint8
63 SDOT_COLDESC_PEDALCYCLIST STRUCK PEDALCYCLIST REAR END	194673 non-null	uint8
64 SDOT_COLDESC_PEDALCYCLIST STRUCK PEDESTRIAN	194673 non-null	uint8
65 WEATHER_Blowing Sand/Dirt	194673 non-null	uint8
66 WEATHER_Clear	194673 non-null	uint8
67 WEATHER_Fog/Smog/Smoke	194673 non-null	uint8
68 WEATHER_Other	194673 non-null	uint8
69 WEATHER_Overcast	194673 non-null	uint8
70 WEATHER_Partly Cloudy	194673 non-null	uint8
71 WEATHER_Raining	194673 non-null	uint8
72 WEATHER_Severe Crosswind	194673 non-null	uint8
73 WEATHER_Sleet/Hail/Freezing Rain	194673 non-null	uint8
74 WEATHER_Snowing	194673 non-null	uint8
75 WEATHER_Unknown	194673 non-null	uint8
76 ROADCOND_Dry	194673 non-null	uint8
77 ROADCOND_Ice	194673 non-null	uint8
78 ROADCOND_Oil	194673 non-null	uint8
79 ROADCOND_Other	194673 non-null	uint8



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79	ROADCOND_Other	194673	non-null	uint8
80	ROADCOND_Sand/Mud/Dirt	194673	non-null	uint8
81	ROADCOND_Snow/Slush	194673	non-null	uint8
82	ROADCOND_Standing Water	194673	non-null	uint8
83	ROADCOND_Unknown	194673	non-null	uint8
84	ROADCOND_Wet	194673	non-null	uint8
85	LIGHTCOND_Dark - No Street Lights	194673	non-null	uint8
86	LIGHTCOND_Dark - Street Lights Off	194673	non-null	uint8
87	LIGHTCOND_Dark - Street Lights On	194673	non-null	uint8
88	LIGHTCOND_Dark - Unknown Lighting	194673	non-null	uint8
89	LIGHTCOND_Dawn	194673	non-null	uint8
90	LIGHTCOND_Daylight	194673	non-null	uint8
91	LIGHTCOND_Dusk	194673	non-null	uint8
92	LIGHTCOND_Other	194673	non-null	uint8
93	LIGHTCOND_Unknown	194673	non-null	uint8
94	ST_COLDESC_All Other Multi Vehicle	194673	non-null	uint8
95	ST_COLDESC_All other non-collision	194673	non-null	uint8
96	ST_COLDESC_Breakage of any part of the vehicle resulting in injury or in further property damage	194673	non-null	uint8
97	ST_COLDESC_Domestic animal other (cat, dog, etc)	194673	non-null	uint8
98	ST_COLDESC_Entering at angle	194673	non-null	uint8
99	ST_COLDESC_Fire started in vehicle	194673	non-null	uint8
100	ST_COLDESC_Fixed object	194673	non-null	uint8
101	ST_COLDESC_From opposite direction - all others	194673	non-null	uint8
102	ST_COLDESC_From opposite direction - both going straight - one stopped - sideswipe	194673	non-null	uint8
103	ST_COLDESC_From opposite direction - both going straight - sideswipe	194673	non-null	uint8
104	ST_COLDESC_From opposite direction - both moving - head-on	194673	non-null	uint8
105	ST_COLDESC_From opposite direction - one left turn - one right turn	194673	non-null	uint8
106	ST_COLDESC_From opposite direction - one left turn - one straight	194673	non-null	uint8
107	ST_COLDESC_From opposite direction - one stopped - head-on	194673	non-null	uint8
108	ST_COLDESC_From same direction - all others	194673	non-null	uint8
109	ST_COLDESC_From same direction - both going straight - both moving - rear-end	194673	non-null	uint8
110	ST_COLDESC_From same direction - both going straight - both moving - sideswipe	194673	non-null	uint8
111	ST_COLDESC_From same direction - both going straight - one stopped - rear-end	194673	non-null	uint8
112	ST_COLDESC_From same direction - both going straight - one stopped - sideswipe	194673	non-null	uint8
113	ST_COLDESC_From same direction - one left turn - one straight	194673	non-null	uint8
114	ST_COLDESC_From same direction - one right turn - one straight	194673	non-null	uint8
115	ST_COLDESC_Non-domestic animal (deer, bear, elk, etc)	194673	non-null	uint8
116	ST_COLDESC_Not stated	194673	non-null	uint8
117	ST_COLDESC_One car entering driveway access	194673	non-null	uint8
118	ST_COLDESC_One car entering parked position	194673	non-null	uint8
119	ST_COLDESC_One car leaving driveway access	194673	non-null	uint8
120	ST_COLDESC_One car leaving parked position	194673	non-null	uint8
121	ST_COLDESC_One parked-one moving	194673	non-null	uint8
122	ST_COLDESC_Other object	194673	non-null	uint8
123	ST_COLDESC_Pedalcyclist All Other Involvements ONE UNIT - PEDALCYCLIST ONLY or PEDALCYCLIST STR	194673	non-null	uint8
124	ST_COLDESC_Pedalcyclist Strikes Moving Vehicle	194673	non-null	uint8
125	ST_COLDESC_Pedalcyclist Strikes Pedalcyclist or Pedestrian	194673	non-null	uint8
126	ST_COLDESC_Person fell, jumped or was pushed from vehicle	194673	non-null	uint8
127	ST_COLDESC_Railway Vehicle Strikes Pedalcyclist	194673	non-null	uint8
128	ST_COLDESC_Railway Vehicle Strikes Pedestrian	194673	non-null	uint8
129	ST_COLDESC_Railway Vehicle Strikes Vehicle	194673	non-null	uint8
130	ST_COLDESC_Same direction -- both turning left -- both moving -- rear end	194673	non-null	uint8
131	ST_COLDESC_Same direction -- both turning left -- both moving -- sideswipe	194673	non-null	uint8
132	ST_COLDESC_Same direction -- both turning left -- one stopped -- rear end	194673	non-null	uint8
133	ST_COLDESC_Same direction -- both turning left -- one stopped -- sideswipe	194673	non-null	uint8
134	ST_COLDESC_Same direction -- both turning right -- both moving -- rear end	194673	non-null	uint8
135	ST_COLDESC_Same direction -- both turning right -- both moving -- sideswipe	194673	non-null	uint8
136	ST_COLDESC_Same direction -- both turning right -- one stopped -- rear end	194673	non-null	uint8
137	ST_COLDESC_Same direction -- both turning right -- one stopped -- sideswipe	194673	non-null	uint8
138	ST_COLDESC_Strikes or Was Struck by Object from the Load of Another Vehicle	194673	non-null	uint8
139	ST_COLDESC_Strikes or Was Struck by a Part of Another Vehicle (Not from Load)	194673	non-null	uint8
140	ST_COLDESC_Vehicle - Pedalcyclist	194673	non-null	uint8
141	ST_COLDESC_Vehicle Hits City Road or Construction Machinery	194673	non-null	uint8
142	ST_COLDESC_Vehicle Hits Other Road or Construction Machinery	194673	non-null	uint8
143	ST_COLDESC_Vehicle Hits State Road or Construction Machinery	194673	non-null	uint8
144	ST_COLDESC_Vehicle Strikes All Other Non-Domestic Animal	194673	non-null	uint8
145	ST_COLDESC_Vehicle Strikes Deer	194673	non-null	uint8
146	ST_COLDESC_Vehicle Strikes Pedalcyclist	194673	non-null	uint8
147	ST_COLDESC_Vehicle Strikes Railway Vehicle	194673	non-null	uint8
148	ST_COLDESC_Vehicle Struck by City Road or Construction Machinery	194673	non-null	uint8
149	ST_COLDESC_Vehicle Struck by Other Road or Construction Machinery	194673	non-null	uint8
150	ST_COLDESC_Vehicle backing hits pedestrian	194673	non-null	uint8
151	ST_COLDESC_Vehicle going straight hits pedestrian	194673	non-null	uint8
152	ST_COLDESC_Vehicle hits Pedestrian - All Other Actions	194673	non-null	uint8
153	ST_COLDESC_Vehicle overturned	194673	non-null	uint8
154	ST_COLDESC_Vehicle turning left hits pedestrian	194673	non-null	uint8
155	ST_COLDESC_Vehicle turning right hits pedestrian	194673	non-null	uint8

dtypes: int64(5), object(1), uint8(150)  
memory usage: 36.8+ MB

## 2.5 Test of correlation and significance

To get a better measure of the important characteristics, we would look at the correlation of the variables vis-a-vis target variable i.e. Accident Severity. 2 measures are followed for the analysis.

### A. Pearson Correlation:



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The Pearson Correlation measures the linear dependence between two variables X and Y of 'int64' or 'float64' types.

The resulting coefficient is a value between -1 and 1 inclusive, where:

1: Total positive linear correlation.

0: No linear correlation, the two variables most likely do not affect each other.

-1: Total negative linear correlation.

The closeness to terminal values (-1 and 1) would decide strength of the correlation.

### B. P-value:

The P-value is the probability value that the correlation between these two variables is statistically significant. Normally, we choose a significance level of 0.05, which means that we are 95% confident that the correlation between the variables is significant. We would use "stats" module in the "Scipy" library to get the P-value.

By convention, when the

p-value is  $< 0.001$ : we say there is strong evidence that the correlation is significant.

the p-value is  $< 0.05$ : there is moderate evidence that the correlation is significant.

the p-value is  $< 0.1$ : there is weak evidence that the correlation is significant.

the p-value is  $> 0.1$ : there is no evidence that the correlation is significant.

```
[19]: from scipy import stats

dict={"Column Name":["Pearson Correlation Coefficient","P-value"]}
for column in df_corr.columns.values.tolist():
    pearson_coef, p_value = stats.pearsonr(df_corr[column], df_corr['SEVERITYCODE'])
    dict[column]=[pearson_coef,p_value]
df_dict = pd.DataFrame(dict)
df_dict_clean=df_dict.transpose(copy=True)
df_dict_clean.rename(columns={0: 'Pearson Correlation Coefficient',1:'P-value'}, inplace=True)
df_dict_clean.reset_index(inplace=True)
df_dict_clean.drop([0,1], axis=0, inplace=True)
df_dict_clean
```

```
[21]: df_dict_clean[df_dict_clean['Pearson Correlation Coefficient'] > 0.2]
```

```
[21]:
```

	index	Pearson Correlation Coefficient	P-value
3	PEDCOUNT	0.246338	0
4	PEDCYLCOUNT	0.214218	0
10	COLLISIONTYPE_Cycles	0.213271	0
15	COLLISIONTYPE_Pedestrian	0.245441	0
20	JUNCTIONTYPE_At Intersection (intersection rel...	0.201628	0
36	SDOT_COLDESC_MOTOR VEHICLE STRUCK PEDESTRIAN	0.243571	0

## 2.6 Conclusion: Important Variables

By now we have a better idea of what our data looks like and which variables are important to take into account when predicting the 'Severity' class.

## CAPSTONE PROJECT – CAR ACCIDENT SEVERITY

As we move into building machine learning modelling steps to automate our analysis, feeding the model with variables that meaningfully affect our target variable would help improve the model's prediction performance.