```
In [2]:
            import pandas as pd
            import numpy as np
            import csv
            import scipy.stats as scs
            import statsmodels.api as sm
            import statsmodels.formula.api as sms
            import scipy.stats as stats
            from math import sqrt
            from sklearn.preprocessing import OneHotEncoder
            from sklearn.tree import DecisionTreeClassifier
            from sklearn import tree
            from sklearn.feature_selection import SelectKBest, chi2
            from sklearn.metrics import accuracy_score, confusion_matrix, classification_
            import matplotlib.pyplot as plt
            import seaborn as sns
```

Question 1 What parts of Chicago have the most fatalities?

```
In [3]:
df = pd.read_csv(r'data\ChicagoCrashes.csv')
```

In [4]: df.describe()

| | Unnamed: 0 | CRASH_DATE_x | OCCUPANT_CNT | POSTED_SPEED_LIMIT | BEAT_OF |
|-------|--------------|--------------|--------------|--------------------|------------|
| count | 1.134909e+06 | 1.134909e+06 | 1.134909e+06 | 1.134909e+06 | 1.134909e- |
| mean | 9.897487e+05 | 2.018056e+03 | 1.415067e+00 | 2.888418e+01 | 1.233346e· |
| std | 5.947572e+05 | 1.283893e+00 | 1.418414e+00 | 5.913001e+00 | 6.996664e· |
| min | 0.000000e+00 | 2.015000e+03 | 0.000000e+00 | 0.000000e+00 | 1.110000e- |
| 25% | 4.751850e+05 | 2.017000e+03 | 1.000000e+00 | 3.000000e+01 | 7.250000e· |
| 50% | 9.654550e+05 | 2.018000e+03 | 1.000000e+00 | 3.000000e+01 | 1.212000e· |
| 75% | 1.493715e+06 | 2.019000e+03 | 2.000000e+00 | 3.000000e+01 | 1.821000e· |
| max | 2.115933e+06 | 2.020000e+03 | 6.000000e+01 | 9.900000e+01 | 2.535000e· |
| | | | | | |

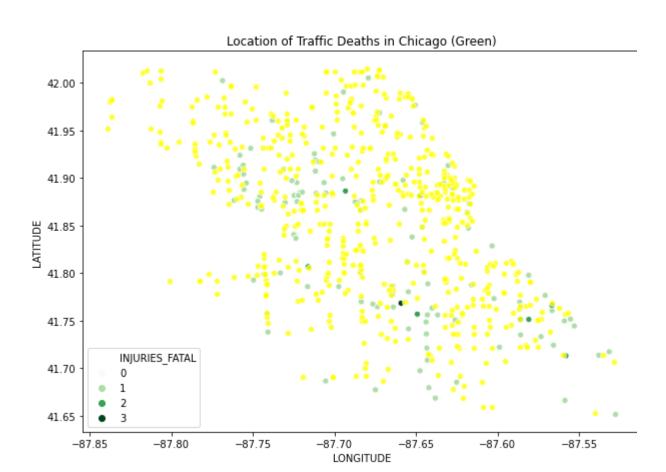
In [5]: df2 = df.sample(frac=0.0005)

```
In [6]: df.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1134909 entries, 0 to 1134908
Data columns (total 49 columns):
Column Non-Nu

| Data | COTUMNIS (COCAT 45 COTUMNIS). | | |
|------|-------------------------------|-----------------|-----------|
| # | Column | Non-Null Count | Dtype |
| | | | |
| 0 | Unnamed: 0 | 1134909 non-nul | l int64 |
| 1 | CRASH_DATE_x | 1134909 non-nul | l int64 |
| 2 | UNIT_TYPE | 1134909 non-nul | l object |
| 3 | MAKE | 1134909 non-nul | l object |
| 4 | MODEL | 1134909 non-nul | l object |
| 5 | VEHICLE_DEFECT | 1134909 non-nul | l object |
| 6 | VEHICLE_TYPE | 1134909 non-nul | l object |
| 7 | VEHICLE_USE | 1134909 non-nul | l object |
| 8 | MANEUVER | 1134909 non-nul | l object |
| 9 | OCCUPANT_CNT | 1134909 non-nul | l float64 |
| 10 | CRASH_DATE_y | 1134909 non-nul | l object |
| 11 | POSTED_SPEED_LIMIT | 1134909 non-nul | l int64 |
| 12 | TRAFFIC_CONTROL_DEVICE | 1134909 non-nul | l object |
| 13 | DEVICE_CONDITION | 1134909 non-nul | l object |
| 14 | WEATHER_CONDITION | 1134909 non-nul | l object |
| 15 | LIGHTING_CONDITION | 1134909 non-nul | l object |
| 16 | FIRST_CRASH_TYPE | 1134909 non-nul | l object |
| 17 | TRAFFICWAY_TYPE | 1134909 non-nul | l object |
| 18 | ALIGNMENT | 1134909 non-nul | l object |
| 19 | ROADWAY_SURFACE_COND | 1134909 non-nul | l object |
| 20 | ROAD_DEFECT | 1134909 non-nul | l object |
| 21 | REPORT_TYPE | 1134909 non-nul | l object |
| 22 | CRASH_TYPE | 1134909 non-nul | l object |
| 23 | DAMAGE | 1134909 non-nul | l object |
| 24 | PRIM_CONTRIBUTORY_CAUSE | 1134909 non-nul | l object |
| 25 | SEC_CONTRIBUTORY_CAUSE | 1134909 non-nul | l object |
| 26 | BEAT_OF_OCCURRENCE | 1134909 non-nul | l float64 |
| 27 | NUM_UNITS | 1134909 non-nul | l int64 |
| 28 | MOST_SEVERE_INJURY | 1134909 non-nul | l object |
| 29 | INJURIES_TOTAL | 1134909 non-nul | l float64 |
| 30 | INJURIES_FATAL | 1134909 non-nul | l float64 |
| 31 | INJURIES_INCAPACITATING | 1134909 non-nul | l float64 |
| 32 | INJURIES_NON_INCAPACITATING | 1134909 non-nul | l float64 |
| 33 | INJURIES_REPORTED_NOT_EVIDENT | 1134909 non-nul | l float64 |
| 34 | INJURIES_NO_INDICATION | 1134909 non-nul | l float64 |
| 35 | INJURIES_UNKNOWN | 1134909 non-nul | l float64 |
| 36 | CRASH_HOUR | 1134909 non-nul | l int64 |
| 37 | CRASH_DAY_OF_WEEK | 1134909 non-nul | l int64 |
| 38 | CRASH_MONTH | 1134909 non-nul | l int64 |
| 39 | LATITUDE | 1134909 non-nul | l float64 |
| 40 | LONGITUDE | 1134909 non-nul | l float64 |
| 41 | PERSON_ID | 1134909 non-nul | l object |
| 42 | PERSON_TYPE | 1134909 non-nul | l object |
| 43 | CRASH_DATE | 1134909 non-nul | l object |
| 44 | SEX | 1134909 non-nul | l object |
| 45 | SAFETY_EQUIPMENT | 1134909 non-nul | l object |
| 46 | AIRBAG_DEPLOYED | 1134909 non-nul | l object |
| 47 | EJECTION | 1134909 non-nul | l object |
| | | | |

```
48 INJURY_CLASSIFICATION
                                                1134909 non-null object
               dtypes: float64(11), int64(7), object(31)
               memory usage: 424.3+ MB
In [7]:
              df1 = df[df['INJURIES_FATAL'] > 0]
In [8]:
              plt.figure(figsize=(10,7))
              sns.scatterplot(x=df['LONGITUDE'],y=df['LATITUDE'],hue=df['INJURIES_FATAL'])
              plt.legend(loc='lower left')
              plt.title('Location of Traffic Deaths in Chicago')
              plt.show()
                                                Location of Traffic Deaths in Chicago
                  42.00
                  41.95
                  41.90
               LATITUDE
                  41.85
                  41.80
                  41.75
                             INJURIES_FATAL
                  41.70
                  41.65
                               -87.9
                                                 -87.8
                                                                    -87.7
                                                                                      -87.6
                                                            LONGITUDE
```



Question 1 Insights

We can see that there is no discernible pattern to location. In the Graph abc of Lake Michigan along the upper righthand side, with fatal accidents preser around the Downtown Chicago Area.

```
In [131]:

df.hist(figsize=(20,20),bins=50)

plt.show()

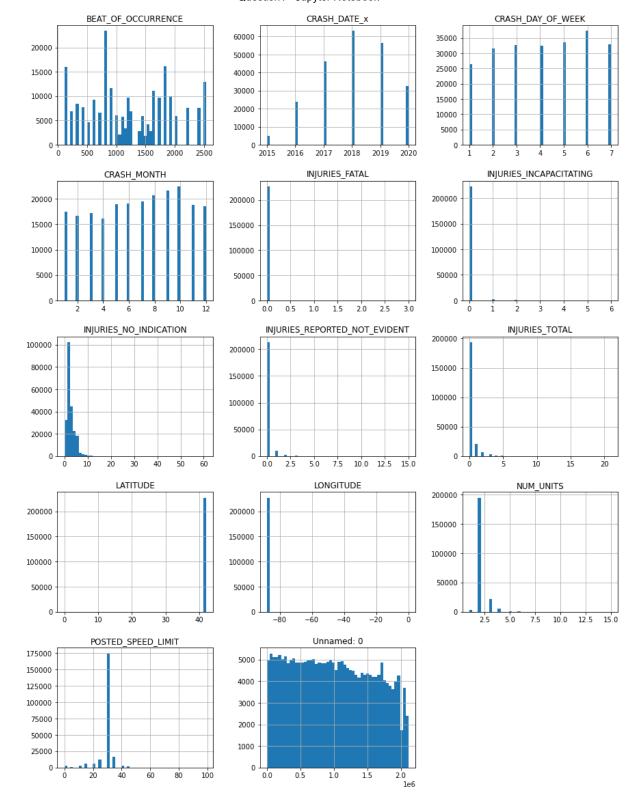
# quick observations - more likely to get in an accident on a Friday.

# after or around 3 PM to 5 PM (rush hour)

# October is most likely month in which to have an accident

# Speed limit in the Chicago city area is generally 35 MPH

# most accidents involve 1 person only.
```



Train Test Split and OneHotEncode

```
In [122]:
              # create a map
              # vehicle_defect_pairs = []
              # for ix, row in enumerate(df.select("VEHICLE_DEFECT").distinct().collect()).
                 pair = (ix, row.VEHICLE_DEFECT)
                  vehicle_defect_pairs.append(pair)
              # vehicle_defect_pairs
In [123]:
              # feature_list = []
              # for col in df.columns:
                  if col in (" c0", "CRASH RECORD ID", "RD NO x", "CRASH DATE x", "VEHICLE
                    continue
              #
              #
                  else:
                    feature list.append(col)
              # assembler = VectorAssembler(inputCols=feature list, outputCol="features")
In [100]:
              # # # Remove "object"-type features from df
              # cont features = [col for col in df.columns if df[col].dtype in [np.float64]
              # # # Remove "object"-type features from df
              # df_cont = df.loc[:, cont_features]
In [101]:
              # # Create df cat which contains only the categorical variables
              # features_cat = [col for col in df.columns if df[col].dtype in [np.object]]
              # other_ind = []
              # for col in features cat:
                    others = list(df[df[col].str.contains("OTHER")].index)
                    for oth in others:
                        if oth in other_ind:
                            continue
                        else: other_ind.append(oth)
              # df.drop(other_ind, inplace=True)
              # df_cat = df.loc[:, features_cat]
              # df_target = df.loc[:, ['INJURIES_FATAL']]
```

```
In [102]:
              # df_target['INJURIES_FATAL'] = df_target['INJURIES_FATAL'].astype('category
              # df = df.drop("INJURIES_FATAL", axis=1)
In [103]:
              # X = df[['MAKE', 'MODEL', 'VEHICLE_DEFECT', 'VEHICLE_TYPE', 'OCCUPANT_CNT',
              # X = df.drop(columns='INJURIES_FATAL')
              # target = df['INJURIES_FATAL']
In [124]:
              # create a map
              # vehicle_defect_pairs = []
              # for ix, row in enumerate(df.select("VEHICLE_DEFECT").distinct().collect()).
                pair = (ix, row.VEHICLE_DEFECT)
                  vehicle_defect_pairs.append(pair)
              # vehicle_defect_pairs
In [27]:
 In [ ]:
 In [ ]:
```