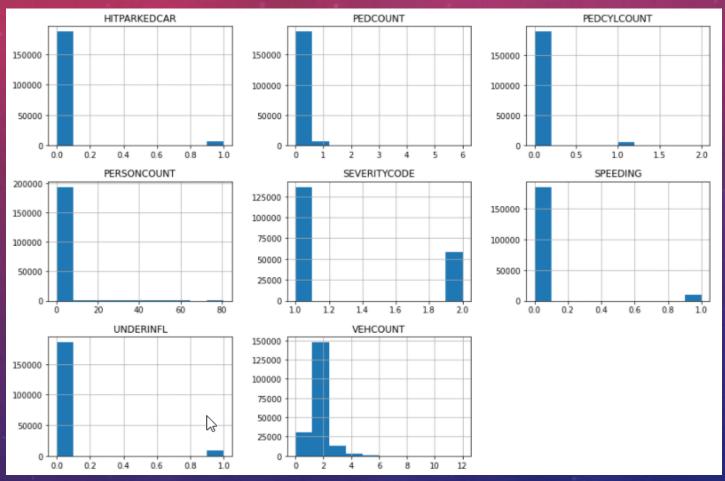


#### DATA EXPLORATION AND HYGIENE

- Example data came from the Coursera capstone project assignment
- Dataset contained 194,643 records with 38 fields.
  - Selected data for analysis contained all rows for 15 fields; 12 of these fields were analyzed for relevance.
  - Replaced all missing values with zeros and normalized true/false, yes/no, and mixed bool indicators to 1s or 0s
  - Replaced all missing values in text fields with 'Unknown'
  - Duplicate severity code was dropped.
  - One-hot encoding was done for weather and collision type.

```
OBJECTID
                    3 ... 219545 219546 219547]
SEVERITYCODE
[2 1]
SEVERITYDESC
['Injury Collision' 'Property Damage Only Collision']
['Intersection' 'Block' 'Alley' nan]
['Angles' 'Sideswipe' 'Parked Car' 'Other' 'Cycles' 'Rear Ended' 'Head On'
 'Unknown' 'Left Turn' 'Pedestrian' 'Right Turn'
 [ 2  4  3  0  1  5  6 16  8  7 11  9 12 17 26 22 10 37 13 36 28 14 53 19
[0 1 2 3 4 5 6]
PEDCYLCOUNT
[0 1 2]
VEHCOUNT
[2 3 1 0 4 7 5 6 8 11 9 10 12]
[0 1]
['Overcast' 'Raining' 'Clear' 'Unknown' 'Other' 'Snowing' 'Fog/Smog/Smoke'
 'Sleet/Hail/Freezing Rain' 'Blowing Sand/Dirt' 'Severe Crosswind
 'Partly Cloudy']
ROADCOND
['Wet' 'Dry' 'Unknown' 'Snow/Slush' 'Ice' 'Other' 'Sand/Mud/Dirt'
 'Standing Water' 'Oil']
LIGHTCOND
['Daylight' 'Dark - Street Lights On' 'Dark - No Street Lights' 'Unknown'
 'twilight' 'Dark - Street Lights Off' 'Other' 'Dark - Unknown Lighting'
SPEEDING
[0 1]
HITPARKEDCAR
[0 1]
```

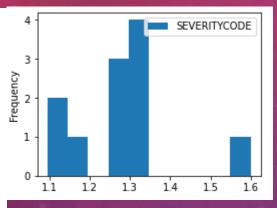
## HISTOGRAMS OF NUMERIC FIELDS



Numeric fields contained largely the same values and were therefore generally not useful.

## SEVERITY FOR WEATHER AND COLLISION TYPE

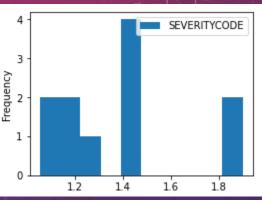
normalized severity	
	SEVERITYCODE
WEATHER	
Blowing Sand/Dirt	1.267857
Clear	1.322491
Fog/Smog/Smoke	1.328647
Other	1.139423
Overcast	1.315544
Partly Cloudy	1.600000
Raining	1.337185
Severe Crosswind	1.280000
Sleet/Hail/Freezing	Rain 1.247788
Snowing	1.188534
Unknown	1.094190



Normalized severity by we		
WEATHER	SEVERITYCODE	
Blowing Sand/Dirt	1	0.732143
	2	0.267857
Clear	1	0.677509
	2	0.322491
Fog/Smog/Smoke	1	0.671353
	2	0.328647
Other	1	0.860577
	2	0.139423
Overcast	1	0.684456
	2	0.315544
Partly Cloudy	2	0.600000
	1	0.400000
Raining	1	0.662815
	2	0.337185
Severe Crosswind	1	0.720000
	2	0.280000
Sleet/Hail/Freezing Rain	1	0.752212
	2	0.247788
Snowing	1	0.811466
v <sup>2</sup>	2	0.188534
Unknown	1	0.945928
	2	0.054072

- Note that the average of 1.6 indicates that there are more injuries than property damage (50/50 = 1.5)
- Injuries on partly cloudy days are 60%.
- Correlation is not causation as there are more often partially cloudy days in Seattle.

	SEVERITYCODE
COLLISIONTYPE	
Angles	1.392917
Cycles	1.876085
Head On	1.430830
Left Turn	1.394877
Other	1.257858
Parked Car	1.055473
Pedestrian	1.898305
Rear Ended	1.430361
Right Turn	1.206022
Sideswipe	1.134666
Unknown	1.212276
The second secon	



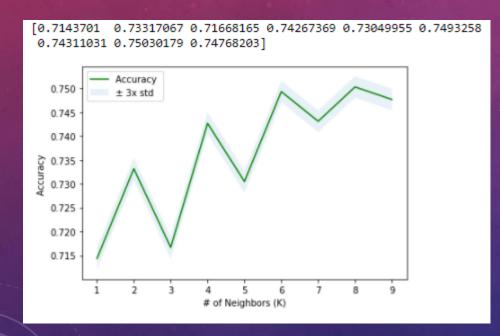
COLLITETONITYDE	CEVERTIVEORE	
COLLISIONTYPE		
Angles	1	0.607083
	2	0.392917
Cycles	2	0.876085
	1	0.123915
Head On	1	0.569170
	2	0.430830
Left Turn	1	0.605123
	2	0.394877
Other	1	0.742142
	2	0.257858
Parked Car	1	0.944527
	2	0.055473
Pedestrian	2	0.898305
	1	0.101695
Rear Ended	1	0.569639
	2	0.430361
Right Turn	1	0.793978
	2	0.206022
Sideswipe	1	0.865334
	2	0.134666

- Note that the averages of more than 1.5 indicate more injuries than property damage (50/50 = 1.5)
- Injuries are far more likely (almost 90%) when a pedestrian or cyclist is involved.

## KNN RESULTS

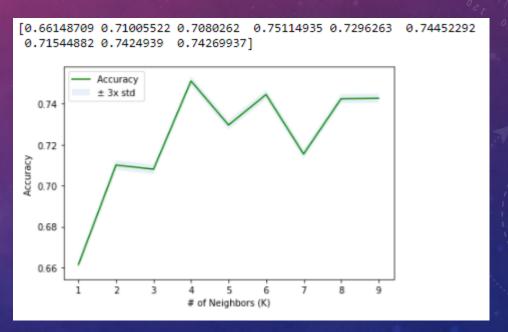
Models were created using KNN test/train data using dataframes that were one hot encoded for either weather or collision. They were also tested to determine the best k for testing

#### Weather



Train set acc: 0.7472999524843006 Test set acc: 0.7503017850263259

#### Collision Type



KNeighborsClassifier(n\_neighbors=4)
Train set acc: 0.7489758440457692
Test set acc: 0.7511493514832412

# CONCLUSION AND FUTURE DIRECTION

- The greatest indicator of whether or not there will be an injury is whether or not a pedestrian or cyclist
  is involved.
- Additional data including severity greater than 2 would be necessary to make a better analysis.
- More widespread (including other cities, states, or countries) would be beneficial to determine if something like address type or road conditions would have more or less of an effect.
- In the future if I were to redo this I would probably drop the pedestrian and cyclist data and attempt to
  determine which features had an effect solely on the remaining incidents.

