# Predicting car accident severity

## Predicting car accident severity would have several applications

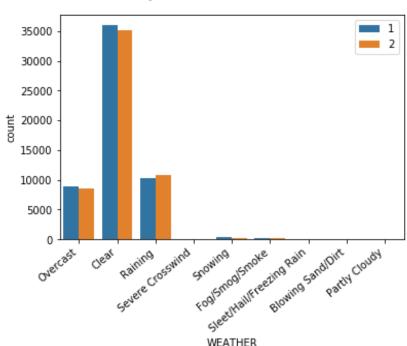
- Authorities could use this knowledge to take measures in order to reduce the number of severe accidents
- Navigation applications would improve their recommendations based on the probability that a severe accident might occur in any potential journey

### Data acquisition and cleaning

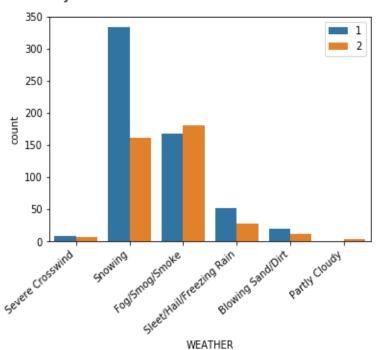
- Rows with empty values are deleted from the dataset
- ► The dataset is not balanced: severity 2 accidents (which caused injuries) are undersampled in the dataset. In order to solve this severity 1 rows are undersampled
- ► Features which are codes of identification or describe the accident are eliminated from the dataset

# Weather and road conditions have little impact on severity, when conditions are not extreme



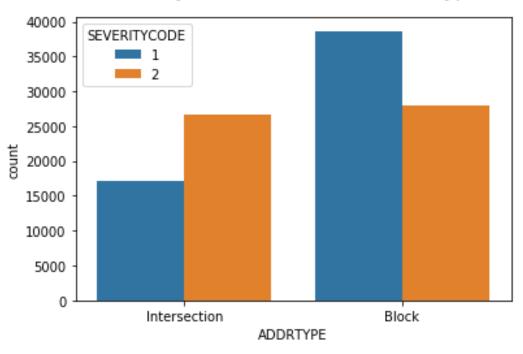


#### Severity code in terms of Extreme Weather Conditions

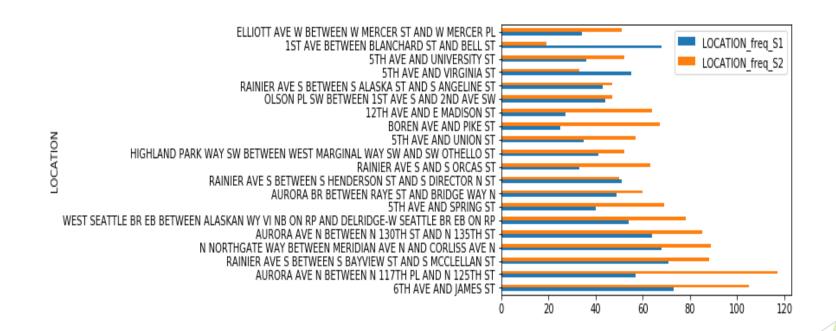


## Accidents in intersections tend to be more severe

#### Severity code in terms of Adress Type

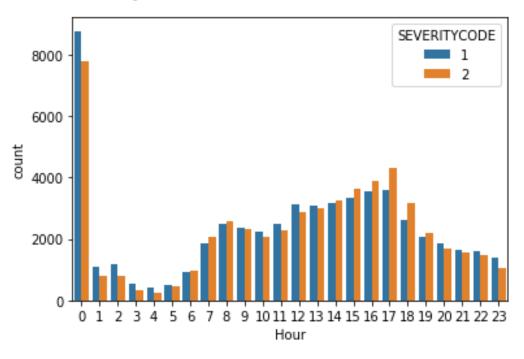


# There are black spots where accidents are more frequent



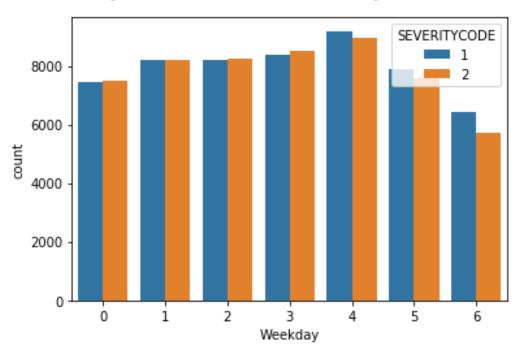
### Accidents at night are less frequent

#### Severity code in terms of Hour of the accident



# Accidents in the weekend are less frequent

Severity code in terms of Weekday of the accident



### Classification models

	KNN	Logistic Regression	Decision Tree
Accuracy	0.68	0.66	0.69
Confusion matrix	[[11203, 5572][ 5095, 11270]]	[[11671, 5104][ 6308, 10057]]	[[10477, 6298] [ 4451, 11914]]

- Performance of the three models is similar
- -Feature importance shows that *LOCATION\_FREQ\_S1* and *LOCATION\_FREQ\_S2* were the best predictors by far

### Conclusion and discussion

- Information about the location is the most valuable in order to predict severity
- Accidents in intersections tend to be more severe
- More features which describe the location of the accident would be useful
- ► For instance: speed limit in each location