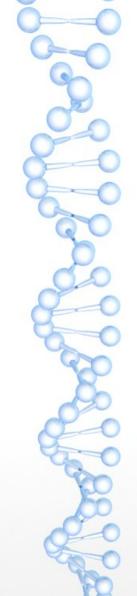


PREDICTION OF THE SEVERITY OF A TRAFFIC INCIDENT.



Problem origin

• Traffic accidents are tragedies that cost money, resources and even human lives.

- Have negative consequencies beyond the accident itself.
 - Traffic congestion.
 - Other accidents because of traffic diversion.
- Many factors involved, not all of them avoidable.

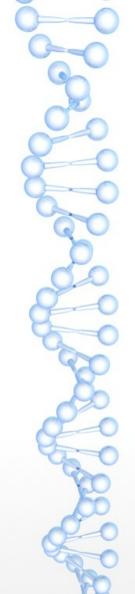


Possible target & stakeholder

Governments need to find a solution to this problem.

They must maintain the citizens safe.

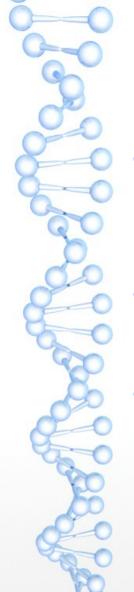
Saving resources as medical ones or money is essential.



Data source

 The original database is related to traffic collisions of the city of Seattle, provided by the state.

 The actual database is a simplified version of the original one.

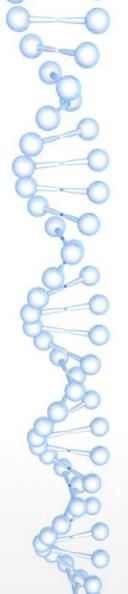


Data understanding

 The database consist of 194673 cases of different kind of accidents.

Each accident (or row) has 38 features.

 The main feature is the severity code, the aim for prediciton of this project.

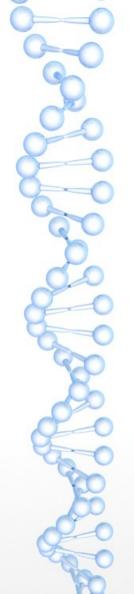


Choosing model

 In this problem the aim is to predict the severity index of an accident given certain information of it.

This is a classification kind of problem.

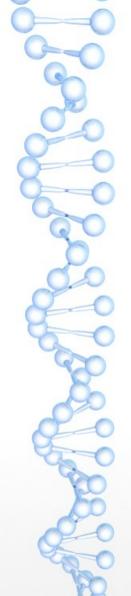
 Due to the non-binary target label (severity index) not all classification algorithms are valid



Choosing model

- Two algorithms are used.
 - K-Nearest Neighbors (KNN)
 - Support Vector Machine (SVM)

Both models are used for comparison.

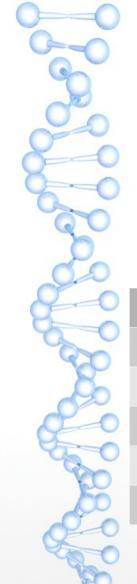


Select features

 The selected features for the model can be divided in two groups, focusing on the driver.

- Out of control: type of collision, the weather, road condition, light condition, etc.

In control: being under the influence of alcohol or drugs, speeding, distractions, etc.



Data preparation

 Many features are categorical ones, as they represent a situations or a status.

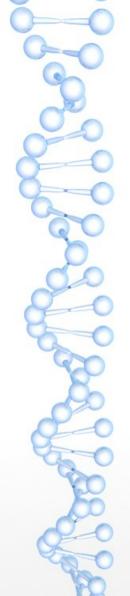
	SEVERITY CODE	ADDRTYP E	COLLISIO NTYPE	PERSONC OUNT	VEHCOUN T	PEDROW NOTGRNT	UNDERIN FL	INATTENT	WEATHER	ROADCO ND	LIGHTCO ND	SPEEDIN G
)	2	Intersection	Angles	2	2	NaN	N	NaN	Overcast	Wet	Daylight	NaN
	1	Block	Sideswipe	2	2	NaN	0	NaN	Raining	Wet	Dark-Street lights on	NaN
	1	Block	Parked car	4	3	NaN	0	NaN	Overcast	Dry	Daylight	NaN
)	1	Block	Other	3	3	NaN	N	NaN	Clear	Dry	Daylight	NaN
	2	Intersectio n	Angles	2	2	NaN	0	NaN	Raining	Wet	Daylight	NaN



Data preparation

 They must be turned into numerical features, labelling their values with numbers.

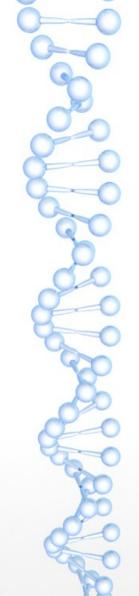
)	SEVERITY CODE	ADDRTYP E	COLLISIO NTYPE	PERSONC OUNT	VEHCOUN T	PEDROW NOTGRNT	UNDERIN FL	INATTENT	WEATHER	ROADCO ND	LIGHTCO ND	SPEEDIN G
)	2	0	0	2	2	0	1	0	0	0	0	0
	1	1	1	2	2	0	0	0	1	0	1	0
	1	1	2	4	3	0	0	0	0	1	0	0
)	1	1	3	3	3	0	1	0	2	1	0	0
	2	0	0	2	2	0	0	0	1	0	0	0



KNN - model

 After training the KNN model, with k=44, the results are tested.

- Mean accuracy of $\approx 0,753$.
- Jaccard index of $\approx 0,753$.
- F1-score of ≈ 0.843 .



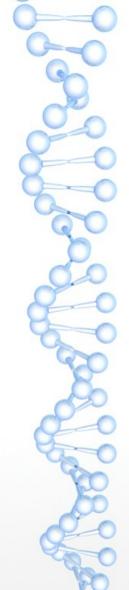
SVM - Model

After training the SVM model, the results are tested.

- Mean accuracy of $\approx 0,755$.

- Jaccard index of $\approx 0,755$.

F1-score of \approx 0,847.



Discussion & Conclusion

• Mean prediction success for both models of around the 75%.

 Acceptable but not ideal, need to test other techniques or features to be sure of the limits of this approach.