Predicting Vehicle Crash Type from Crash Report Incident

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Problem and Background

- Vehicle crashes are common, over **6 million** per year in the U.S.
- Resulted in **40,990 deaths** in the U.S. in 2023
- In order to prevent this, consider which factors influence the severity of crashes
 - Allows us to know how better to prevent injuries and fatalities
 - Identify which locations are more dangerous





Dataset Overview

- 24241 instances
- 42 attributes
- Vehicle Crashes occurring in the Town of Cary, North Carolina
- Classifying instances as "fatalities", "injury", or "crash"
 - Required preprocessing steps
- Examples of attributes:
 - o rdsurface
 - o rdcondition
 - lightcond
 - weather
 - trafcontrol

Preprocessing

- Removed columns
 - o tamainid
 - o records
 - contributing_factor
 - vehicle_type
 - o contrcirl_desc, contrcir2_desc, ...
 - o vehicle3, vehicle4, ...
- New/modified columns
 - o num_vehicles
 - o class
 - o ta_time
- Normalize data
- 24241 instances, 20 columns
- 0.18% fatalities, 13.34% injury, 86.47% crash

Dataset after Preprocessing

	В												М								
rdfeature	rdcharacter	rdolass	rdconfigur	rdsurface	rdcondition	lightcond	weather	trafeontri	lon2	lat2	zone	n	umpassengers	numpedestrians	vehicle1	vehicle2	workerea	ta_time month		ehicles of	lass
NO SPECIAL FEAT	L STRAIGHT, LEVEL	STATE SECONDAR	TWO-WAY, DIVIDED	D, COARSE ASPHALT	r DRY	DAYLIGHT	CLEAR	STOP SIGN	-0.06131	1 -0.01	4066 WAKE		0	-0.076324	PASSENGER CAR	SPORT UTILITY	NO	-0.333225	-1.605539	0.288377 c	rash
NO SPECIAL FEAT	L STRAIGHT, LEVEL'	PUBLIC VEHICULAR	ONE-WAY, NOT DIV	VI COARSE ASPHALT	DRY	DAYLIGHT	CLEAR	NO CONTROL PRES	0.01577	4 -0.01	4067	112	-0.635271	-0.076824	PASSENGER CAR	PASSENGER CAR	NO	-0.128985	-1.605539	0.288377 c	rash
NO SPECIAL FEAT	U STRAIGHT, LEVEL	PUBLIC VEHICULAR	TWO-WAY, NOT DI	VI SMOOTH ASPHAL	T DRY	DAYLIGHT	CLEAR	NO CONTROL PRES	0.01577	-0.01	4066	112	0	-0.076824	PICKUP	PASSENGER CAR	NO	-0.537455	-1.605539	0.288377 c	rash
NO SPECIAL FEAT	L STRAIGHT, GRADE	PUBLIC VEHICULAR	ONE-WAY, NOT DIV	VI SMOOTH ASPHAL	T DRY	DAYLIGHT	CLEAR	NO CONTROL PRES	0.01577	-0,01	4067	112	0	-0.076824	PASSENGER CAR	PASSENGER CAR	NO	-1.354428	-1.313731	0.288377 c	rash
NO SPECIAL FEAT	L STRAIGHT, LEVEL'	LOCAL STREET	TWO-WAY, NOT DI	VI SMOOTH ASPHAL	T DRY	DARK - ROADWAY N	CLEAR	NO CONTROL PRES	0.01577	-0.01	4066	122	-0.635271	-0.076824	FIRETRUCK	PASSENGER CAR	NO	1.300699	-1.313731	-1.476022 ci	rash
T-INTERSECTION	STRAIGHT, LEVEL'	LOCAL STREET	TWO-WAY,NOT DI	VI SMOOTH ASPHAL	T DRY	DAYLIGHT	CLEAR	STOP AND GO SIG	0.01577	4 -0.01	4066	120	0	-0.076824	SPORT UTILITY	PASSENGER CAR	NO	-1.354428	-1.313731	0.288377 c	rash
NO SPECIAL FEAT	U STRAIGHT, LEVEL	STATE SECONDAR	TWO-WAY, NOT DI	VI SMOOTH ASPHAL	T DRY	DAYLIGHT	CLEAR	NO CONTROL PRES	0.01577	-0.01	4066	118	0	-0.076824	PASSENGER CAR	PASSENGER CAR	NO	-0.945947	-1.313731	-1.476022 ci	resh
and the second s	F STRAIGHT, LEVEL			D, SMOOTH ASPHAL				STOP AND GO SIGI				112	0			PASSENGER CAR		-0.945947	-1.313731	0.288377 in	
	L STRAIGHT, LEVEL							STOP SIGN	0.01577			112	0	-0.076824		PASSENGER CAR		0.483737	-1.605539	0.288377 in	
	L STRAIGHT, LEVEL							NO CONTROL PRES				112	0.18559		PASSENGER CAR			0.483737	-1.605539	0.288377 c	
	U STRAIGHT, LEVEL'							NO CONTROL PRES			4066	112	0		PASSENGER CAR			0.892218	-1.313731	0.288377 c	
				VI SMOOTH ASPHAL				NO CONTROL PRES				112	-0.635271		TRUCK/TRAILER			0.279496	-1.313731	-1.476022 0	
	U STRAIGHT, LEVEL			D, SMOOTH ASPHAL				NO CONTROL PRES				116	0		PASSENGER CAR			-1.150188	-1.313731	-1.476022 ci	
	C STRAIGHT, LEVEL					DARK - LIGHTED RO		NO CONTROL PRES				112	0		PASSENGER CAR			1.300699	-1.313731	0.288377 0	
	L STRAIGHT, LEVEL'					DARK - LIGHTED RO		STOP SIGN	0.01577			112	0		SPORT UTILITY			1.709181	-1.313731	-1.476022 ci	
	L STRAIGHT, LEVEL							NO CONTROL PRES				112	0		PASSENGER CAR			0.075256	-1.313731	0.288377 c	
	L STRAIGHT, LEVEL							NO CONTROL PRES			4066 RALW		0.18559		PASSENGER CAR			0.483737	-1.313731	0.288377 c	
	U STRAIGHT, LEVEL							NO CONTROL PRES				120	0		SPORT UTILITY	PASSENGER CAR		0.483737	-1.313731	0.288377 0	
	L STRAIGHT, LEVEL							STOP SIGN	0.01577			112	1.006451		PASSENGER CAR		NO	-0.741706	-1.313731	0.288377 c	
	U STRAIGHT, LEVEL					DARK - LIGHTED RO		NO CONTROL PRES				117	0		SINGLE UNIT TRUC			-2.375631	-1.313731	0.288377 0	
	U STRAIGHT, LEVEL							NO CONTROL PRES				112	0		SPORT UTILITY			0.483737	-1.313731	0.288377 0	
	STRAIGHT, LEVEL					DARK - LIGHTED RO		STOP AND GO SIGI				115	0		PASSENGER CAR			0.892218	-1.313731	0.288377 0	
	U STRAIGHT, LEVEL							NO CONTROL PRES				112	0		PASSENGER CAR'		NO	0.075256	-1.313731	0.288377 c	
	U STRAIGHT, LEVEL					DARK - LIGHTED RO		NO CONTROL PRES				117	0		LIGHT TRUCK (MIN			1.300699	-1.313731	-1.476022 or	
	L STRAIGHT, LEVEL							NO CONTROL PRES				115	0		SPORT UTILITY		NO	-0.741705	-1.313731	0.288377 c	
	L STRAIGHT, LEVEL			D, COARSE ASPHALT				NO CONTROL PRES				116	0		PASSENGER CAR			0.687978	-1.313731	-1.476022 cr	
	L CURVE, GRADE					DARK - LIGHTED RO		NO CONTROL PRES				112	0		PASSENGER CAR			-2.579872	-1.313731	0.288377 0	
	U STRAIGHT, LEVEL					DARK - LIGHTED RO		NO CONTROL PRES				112	0	-0.076824		VAN	NO	1,300699	-1.313731	0.288377 0	
	C STRAIGHT, LEVEL					DARK - LIGHTED RO		NO CONTROL PRES				112	0			PASSENGER CAR		-2.579872	-1.313731	0.288377 0	
	STRAIGHT, LEVEL					DARK - LIGHTED RO		NO CONTROL PRES				112	0		SPORT UTILITY		11.0	1.096459	-1.313731	0.288377 in	
	U STRAIGHT, LEVEL							STOP SIGN	0.01577	-0.01	4066 WAKE		0		PASSENGER CAR			-1.354428	-1.313731	0.288377 c	
	L STRAIGHT, GRADE							NO CONTROL PRES		9	0	122	14.961093		PASSENGER CAR			0.279496	-1.313731	-1.476022 ci	
	T STRAIGHT, LEVEL							NO CONTROL PRES				112	0		SINGLE UNIT TRUC			-0.333225	-1.313731	-1.476022 ci	
	STRAIGHT, LEVEL			D, SMOOTH ASPHAL				STOP AND GO SIGI				116	0		PASSENGER CAR		110	0.687978	-1.313731	2.052776 in	
	SESTRAIGHT, HILLCRE					DARK - LIGHTED RO		STOP AND GO SIGI				112	3.469035		SPORT UTILITY			-0.741706	-1.313731	0.288377 in	
	L STRAIGHT, LEVEL					DARK - LIGHTED RO		NO CONTROL PRES			4066	111	0		SPORT UTILITY			-0.945947	-1.313731	-1.476022 ci	
	L STRAIGHT, LEVEL							NO CONTROL PRES			4066 RALW		0		PASSENGER CAR			0.279495	-1.313731	-1.476022 ci	
	F STRAIGHT, LEVEL					DARK - LIGHTED RO		STOP AND GO SIGI				112	0		PASSENGER CAR		1.12	0.687978	-1.313731	0.288377 c	
	U STRAIGHT, LEVEL'		TWO-WAY, DIVIDED		DRY			NO CONTROL PRES			4066	112	-0.635271		PASSENGER CAR			0.483737	-1.313731	3.817175 ci	
	L STRAIGHT, LEVEL			D, GROOVED CONCR		DARK - LIGHTED RO		NO CONTROL PRES				112	0		PASSENGER CAR			-1.96715	-1.313731	-1.476022 ci	
	L STRAIGHT, LEVEL							STOP AND GO SIGI				118	0	-0.076824		PASSENGER CAR		-1.354428	-1.313731	0.288377 ci	
				VI SMOOTH ASPHAL		DARK - LIGHTED RO		NO CONTROL PRES			4066	112	0		SPORT UTILITY			0.892218	-1.313731	0.288377 c	
	U STRAIGHT, LEVEL					DARK - LIGHTED RO		NO CONTROL PRES		1000	"Q +	113	0			PASSENGER CAR		1.50494	-1.313731	0.288377 c	
NO SPECIAL FEAT	L STRAIGHT, LEVEL'	STATE SECONDAR	TWO-WAY, DIVIDED	D, COARSE ASPHALT	r' DRY	DAYLIGHT	CLEAR	NO CONTROL PRES	0.01577	400	4068	112	0	-0.076824	SPORT UTILITY	PASSENGER CAR	NO	-0.128985	-1.313731	-1.476022 in	njury

Attribute Selection

- CorrelationAttributeEval
- 2) InfoGainAttributeEval
- 3) ReliefF
- 4) CfsSubsetEval
- 5) Self-Selected

CorrelationAttributeEval

Cutoff value of 0.02

1 rdfeature
2 rdcharacter
3 rdclass
4 _ rdconfigur
5 rdcondition
6 Iightcond
7 trafcontrl
8 zone
9 numpedestrians
10 vehicles
11 Class

$$r = rac{\sum \left(x_i - ar{x}
ight)\left(y_i - ar{y}
ight)}{\sqrt{\sum \left(x_i - ar{x}
ight)^2 \sum \left(y_i - ar{y}
ight)^2}}$$

```
=== Attribute Selection on all input data ===
Search Method:
       Attribute ranking.
Attribute Evaluator (supervised, Class (nominal): 21 class):
       Correlation Ranking Filter
Ranked attributes:
 0.128784
            9 trafcontrl
 0.11801
            3 rdclass
 0.109268
            1 rdfeature
 0.107156 20 vehicles
 0.106201
           4 rdconfigur
 0.069827
           14 numpedestrians
 0.035178
           12 zone
 0.029774
            2 rdcharacter
 0.025973
            7 lightcond
 0.023784
            6 rdcondition
 0.014696
           16 vehicle2
 0.013492
           15 vehicle1
 0.012947
            8 weather
 0.012734
           13 numpassengers
 0.006762
           10 lon2
 0.00541
           11 lat2
 0.002679
           19 month
 0.00234
            5 rdsurface
 0.002267 17 workarea
 0.000728 18 ta time
Selected attributes: 9,3,1,20,4,14,12,2,7,6,16,15,8,13,10,11,19,5,17,18 : 20
```

InfoGainAttributeEval

Cutoff value of 0.01

```
1 rdfeature
2 rdclass
3 rdconfigur
4 trafcontrl
5 numpassengers
6 vehicles
7 class
```

$IG(A) = \text{Entropy}(D) - \text{Entropy}_A(D)$

```
Ranked attributes:
0.037212
            3 rdclass
 0.023859
            1 rdfeature
 0.016529
            9 trafcontrl
0.015225
           20 vehicles
0.014701
            4 rdconfigur
0.011723 13 numpassengers
0.004443 12 zone
0.00321
           14 numpedestrians
0.001609
           15 vehicle1
0.001276
           18 ta time
            2 rdcharacter
 0.001193
           16 vehicle2
 0.001129
 0.001114
            5 rdsurface
0.000905
            6 rdcondition
 0.00082
           10 lon2
0.000737
           11 lat2
0.000721
            7 lightcond
 0.000557
            8 weather
 0.000152
           17 workarea
           19 month
Selected attributes: 3,1,9,20,4,13,12,14,15,18,2,16,5,6,10,11,7,8,17,19 : 20
```

ReliefF

Cutoff value of 0.02

```
rdfeature
    rdcharacter
   rdclass
4 rdconfigur
5 rdsurface
6 lightcond
7 weather
8 trafcontrl
9 | lon2
10 zone
11 vehicle1
12 vehicle2
13 ta time
14 vehicles
15 class
```

```
W(A) = W(A) - \frac{1}{m} \sum_{i=1}^{m} \left[ diff(A, \text{nearest\_hit}) - diff(A, \text{nearest\_miss}) \right]
```

```
Ranked attributes:
 0.0919194
            12 zone
             3 rdclass
 0.0831862
 0.0587109
           15 vehicle1
 0.0515362
           16 vehicle2
 0.0429486
            1 rdfeature
 0.0418798
             4 rdconfigur
 0.0382455
            19 month
 0.0334227
            5 rdsurface
 0.0305994
           7 lightcond
 0.0287035
             8 weather
 0.0221575
            18 ta time
 0.0204764
             2 rdcharacter
            20 vehicles
 0.0203199
 0.020221
             9 trafcontrl
 0.0088835
             6 rdcondition
 0.0062019
            14 numpedestrians
 0.0015324
            17 workarea
 0.0014038
            13 numpassengers
 0.0001509
            11 lat2
 0.0000133
            10 lon2
Selected attributes: 12,3,15,16,1,4,19,5,7,8,18,2,20,9,6,14,17,13,11,10 : 20
```

CfsSubsetEval

1 rdfeature
2 _ rdclass
3 _ rdconfigur
4 Trafcontrl
5 numpassengers
6 numpedestrians
7 vehicles

```
=== Attribute Selection on all input data ===
Search Method:
       Greedy Stepwise (forwards).
       Start set: no attributes
       Merit of best subset found:
                                      0.039
Attribute Subset Evaluator (supervised, Class (nominal): 21 class):
       CFS Subset Evaluator
       Including locally predictive attributes
Selected attributes: 1,3,4,9,13,14,20 : 7
                     rdfeature
                    rdclass
                     rdconfigur
                    trafcontrl
                    numpassengers
                    numpedestrians
                    vehicles
```

Self-Selected

- 1) weather
- 2) trafcontrl
- 3) lightcond
- 4) rdcondition (Road Condition)
- 5) rdfeature
- 6) vehicle1 and vehicle2
- 7) numpassengers
- 8) numpedestrians

Classifier Models

K-fold cross-validation

- Naive Bayes
 - A probabilistic classifier based on Bayes' theorem, assuming feature independence
- J48
- Builds decision tree by repeatedly splitting data based on attribute with the highest information gain
- RandomForest
 - Constructs multiple decision trees using random subsets of data and features, then aggregates votes
- DecisionTable
 - Summarizes data patterns by building and organizing conditions and corresponding actions in a table
- OneR
- Selects the single best-performing feature and builds a simple rule based on it

Results (Accuracy)

	Naive Bayes	J48	Random Forest	Decision Table	OneR
Self-Selected	85.12%	86.47%	85.18%	86.44%	86.44%
CorrelationAttri buteEval	82.76%	86.47%	84.89%	86.46%	86.43%
InfoGain	85.05%	86.47%	85.74%	86.45%	86.44%
ReliefF	84.44%	86.46%	84.32%	86.43%	86.43%
CfsSubsetEval	85.04%	86.47%	85.76%	86.45%	86.43%

Conclusion and Future Work

- Best model was CfsSubsetEval with Naive Bayes
 - 85.0519% Accuracy
 - 0.85 TPR
 - 0.785 FPR
 - 0.799 Precision
 - 0.850 Recall
 - 0.814 F-Measure
 - 0.114 MCC
- Injury and fatality classes have TP rate
- Tradeoff between majority and minority classes
- Address class imbalance

```
=== Confusion Matrix ===

a b c <-- classified as
20304 629 29 | a = crash
2920 310 5 | b = injury
41 2 1 | c = fatalities
```

Questions?