



ML in Aviation

How to improve safety

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Summary



- Background in Aviation (aeronautics)
- ML in Aviation business cases
 - Forecasting (Ticket sales, profit, etc.)
- What about safety?

Outline



1. Business Problem
2. Data
3. Results
4. Conclusions

Business Problem

Using machine learning (ML) specifically supervised learning.

Can we use classification algorithms to classify past aviation accidents as fatal or non-fatal based on features of that accident?

Data

< AviationData.csv (19.57 MB)  

Detail Compact Column 31 of 31 columns ▾

About this file

This file contains information from 1962 and later about civil aviation accidents and selected incidents within the United States, its territories and possessions, and in international waters.

▲ Injury.Severity ▾	▲ Aircraft.damage ▾	▲ Aircraft.Category ▾	▲ Registration.Num... ▾	▲ Make ▾
Injury Severity	Aircraft level of Damage	the category of airplane, airplane, helicopters...	Registration Number or ID	the manufacturer name
Non-Fatal 75%	Substantial 72%	[null] 66%	[null] 5%	Cessna 31%
Fatal(1) 10%	Destroyed 21%	Airplane 29%	NONE 0%	Piper 17%
Other (13156) 15%	Other (5687) 7%	Other (4146) 5%	Other (81390) 95%	Other (45084) 52%
Fatal	Substantial	Airplane	N13VT	Velocity
Fatal	Destroyed	Helicopter	N13AT	Bell
Minor	Substantial	Airplane	N56517	Maule
Non-Fatal	Substantial	Airplane	N3477E	Cessna
Minor	Substantial	Airplane	N8183C	Piper
Minor	Substantial	Airplane	N7569A	Abbott Gerry
Fatal	Substantial	Airplane	N8088G	Cessna

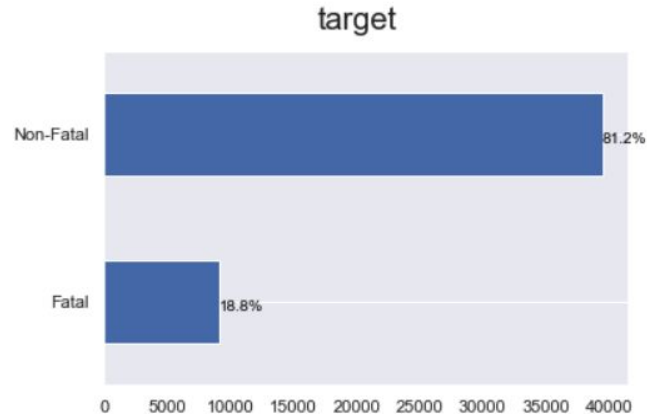


- The NTSB has a database from accidents since 1962
- Pulled from Kaggle

Final shape (4867,17)

```
data_2.columns
```

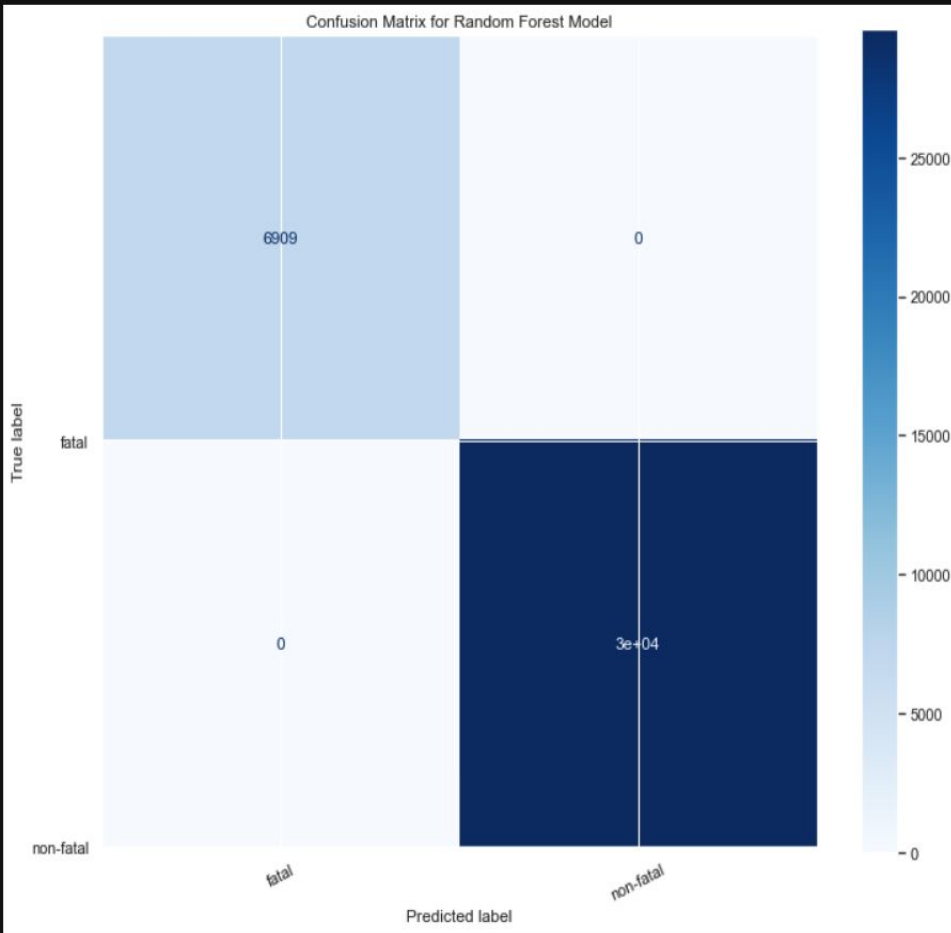
```
Index(['Unnamed: 0', 'target', 'location', 'aircraft_damage', 'make', 'model',  
      'number_of_engines', 'engine_type', 'weather_conditions',  
      'phase_of_flight', 'Year', 'Month', 'Day', 'injuries', 'pax_onboard',  
      'fatality_percentage', 'survived', 'amateur_built'],  
      dtype='object')
```



Initial Results

results

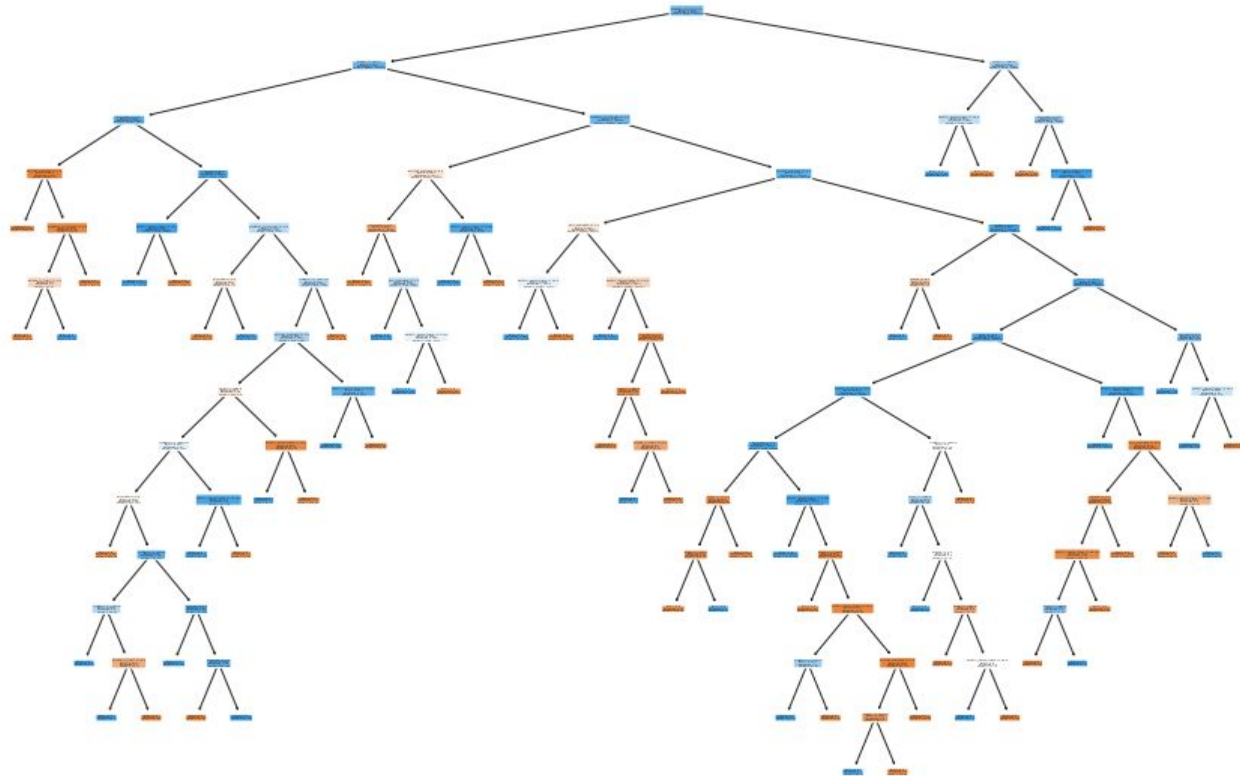
	model	train_precision	train_recall	train_accuracy	train_f1	train_time	test_precision	test_recall	test_accuracy	test_f1	test_time
0	knn	1.000000	1.000000	1.000000	1.000000	0.023560	0.923077	0.957389	0.900148	0.939920	4.634440
1	logistic_regression	0.998683	0.999662	0.998658	0.999173	2.625757	0.998591	0.999597	0.998521	0.999094	0.001482
2	decision_tree	1.000000	1.000000	1.000000	1.000000	0.058102	0.999597	0.999799	0.999507	0.999698	0.002410
3	random_forest	1.000000	1.000000	1.000000	1.000000	0.937998	1.000000	0.999799	0.999836	0.999899	0.043773
4	naive_bayes	0.987394	0.659053	0.716763	0.790484	0.023357	0.988198	0.657903	0.714497	0.789913	0.004797
5	adaboost	1.000000	1.000000	1.000000	1.000000	0.900672	0.999799	0.999799	0.999671	0.999799	0.041151
6	gradient_boosting	1.000000	0.999899	0.999918	0.999949	3.212887	0.999899	0.999799	0.999753	0.999849	0.011706



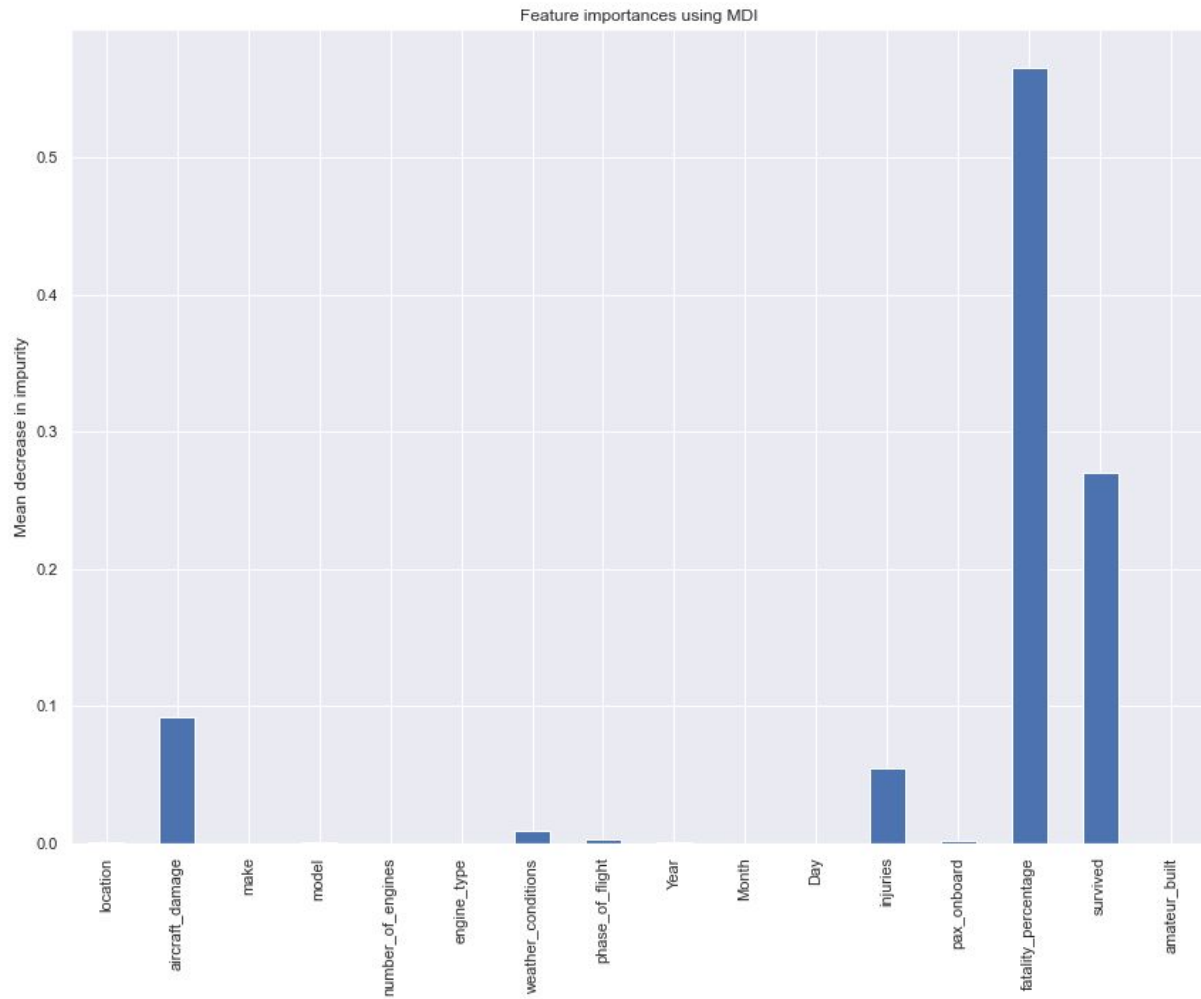
Analysis Random forest

```
{'model': 'random_forest',  
 'train_precision': 1.0,  
 'train_recall': 1.0,  
 'train_accuracy': 1.0,  
 'train_f1': 1.0,  
 'train_time': 0.9320237636566162,  
 'test_precision': 1.0,  
 'test_recall': 0.9997985292636244,  
 'test_accuracy': 0.9998356344510191,  
 'test_f1': 0.9998992544831755,  
 'test_time': 0.04516482353210449}
```

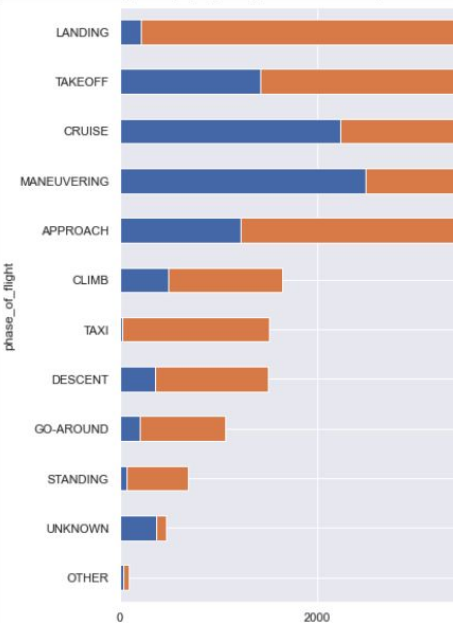

Visualization of my RF model



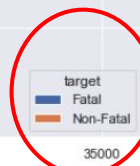
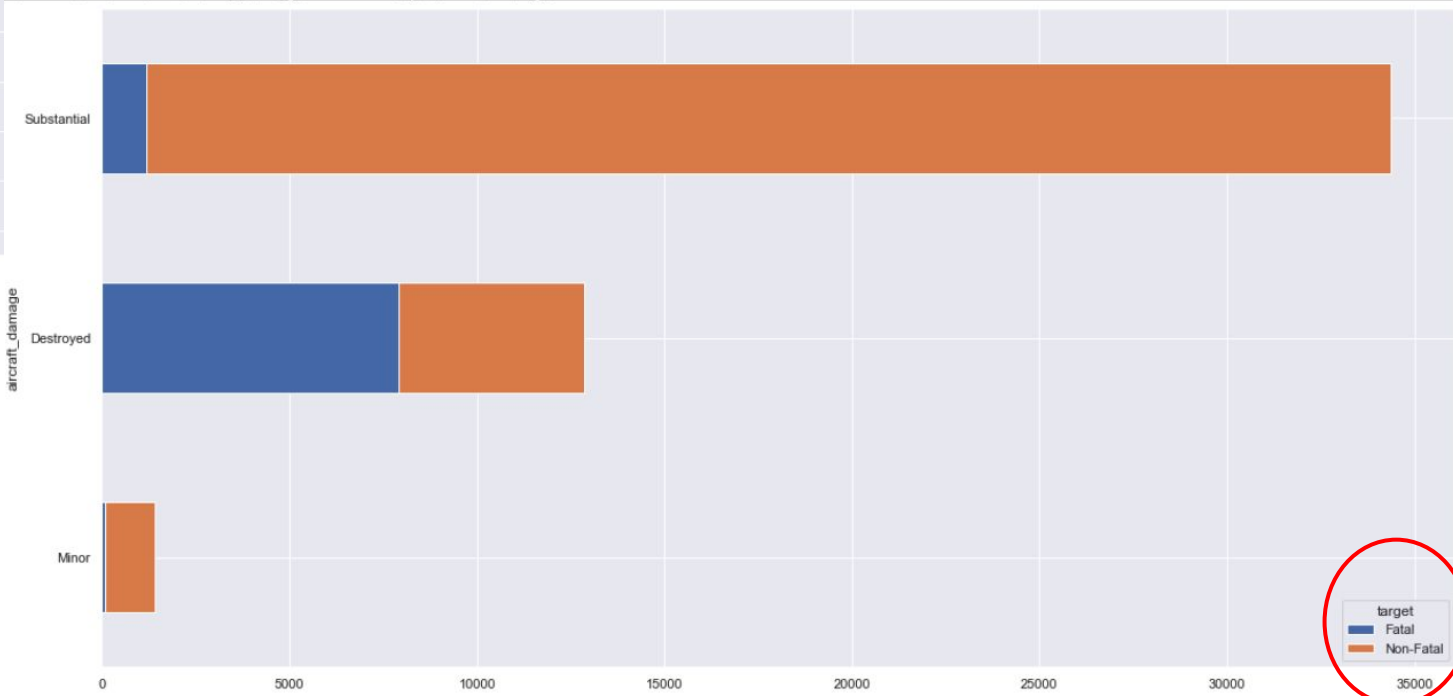
Model evaluation & improving



Bar Plot of phase_of_flight with respect to target



Bar Plot of aircraft_damage with respect to target



Conclusions

1. Successfully ran 7 vanilla ML models to learn how we can improve aviation safety.
 2. The above models returned acceptable model performance.
 3. Through initial classification modeling, we learned which features are important in classifying a fatal or non-fatal aviation accident.
 4. Although we can classify a accident with good model performance. Further investigation and feature engineering is required on the fatality_percentage feature.
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Next Steps

1. Find or create more data specifically in the aircraft_damage category
2. Use imputation to replace any unknown data
3. Productionize model with prediction function
4. Look into multiclass classification

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Thank You!
