Airplane Crash Analysis

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Introduction

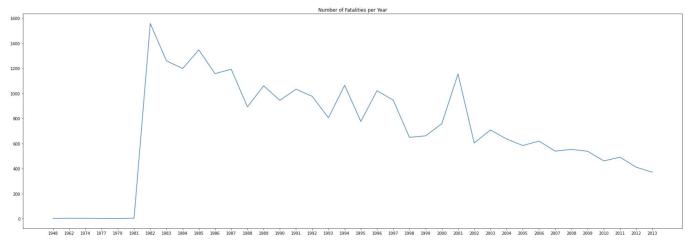
- National Transportation and Safety Board, consists of over 74 thousand accidents and incidents from 1948 to 2013 (https://data.ntsb.gov/avdata).
 - 23% (13089 rows) of the data does have fatalities (1 or more deaths), while the remaining 77% (57668 rows) is non-fatal airplane crashes.
 - Columns like: Date, Latitude, Longitude, Weather type, Amateur Built, Make, Model, Number of Engines, etc

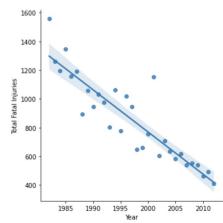


Methods

Answering 2 questions:

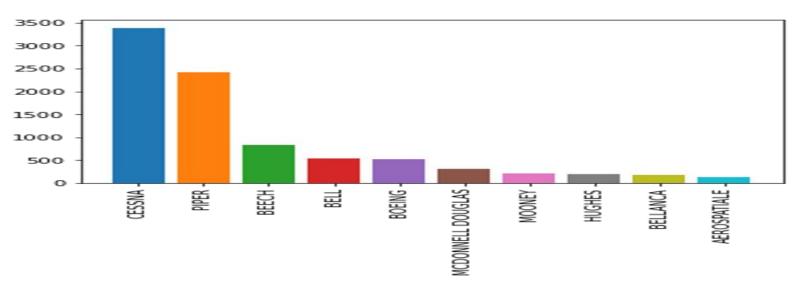
- 1) Are flights becoming safer over time?
 - Used Basic Data Analysis techniques to answer this question
 - Used graphs and linear models to get slope and then determine the trends based on the slope value.





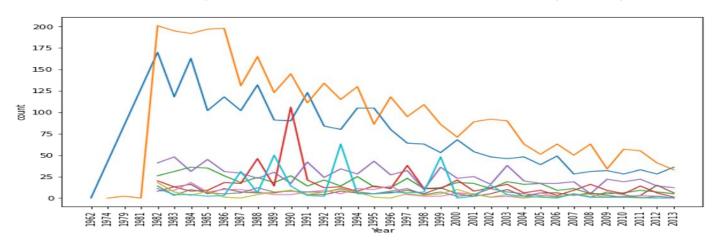
Further Research

Top 10 manufacturers with most fatality rate.



Things to know

- -Small flights are more prone to fatalities than the large commercial ones
- Improved design and safety measures over the years following FAA regulations.



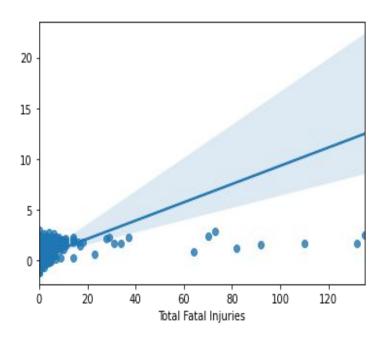


Methods (2)

- 2) Can we predict the number of fatalities or if a flight will contain any amount of fatalities?
 - Predicting number of fatalities:
 - Multiple Linear Regression
 - Predicting whether or not there will be fatalities on a flight (Binary Classify)
 - Multiple Logistic Regression
 - Decision Tree(s) and Random Forest(s)



Linear Regression



- Tried predicting no of fatalities with parameter like flight weather condition, no.of engines, location, Built, Phanse of flight etc.
- RMSE is found to be 3.4
- Most of the fatalities fall under 0-4 band which the model seems to have predicted correctly.

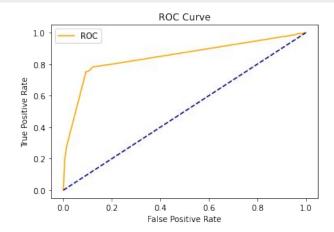
Logistic Regression

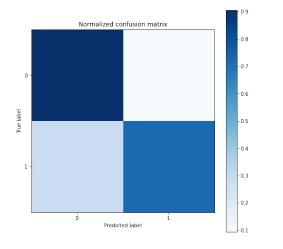
- Model designed to predict whether an airplane crash is fatal or not
- Predictions are based on
 - Aircraft Damage
 - Number of Engines
 - Weather Condition



Logistic Regression

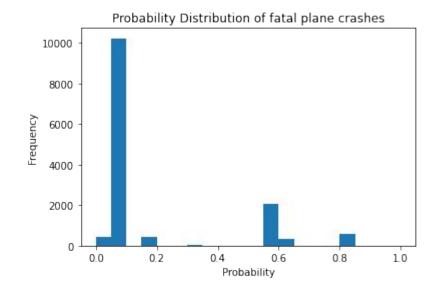
- Model resulted in 87% prediction accuracy with an AUC value of 83%
- Accuracy 10% better than baseline assuming all crashes are fatal
- Hyperparameter tuning using gridsearch marginally improved prediction accuracy
- Tuned model showed marginal improvement in accuracy





Logistic Regression

- Probability distribution is unbalanced as
 77% of crashes had no fatalities
- Utilized specific flight characteristics to predict likelihood of fatal plane crash
- Certain characteristics such as 'high aircraft damage' or 'bad weather' will lead to model predicting around 0.6 likelihood of fatal crash



Decision Tree

- Used Decision Tree to Binary Classify whether a plane will have any fatalities.
- Did compute DT with default parameters and with Hyperparameter tuning through Grid Search.
- Ultimately used the accuracy scores of these models to compare against Random Forest(s) since we know RF would outperform DTs.

```
{'criterion': 'entropy',
    'max_depth': 10,
    'max_leaf_nodes': None,
    'min_samples_leaf': 2,
    'splitter': 'best'}
The train accuracy is: 0.9020227895062274

The train accuracy is: 0.8853165630299604
```

Random Forest

Created 2 RFs:

- 1) RF with Default Parameters (used as baseline)
- 2) RF with Grid Search

```
{'criterion': 'gini',
  'max_depth': 20,
  'max_features': 'auto',
  'max_leaf_nodes': None,
  'min_samples_leaf': 5}
The train accuracy is: 0.9276212348732444

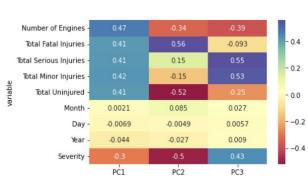
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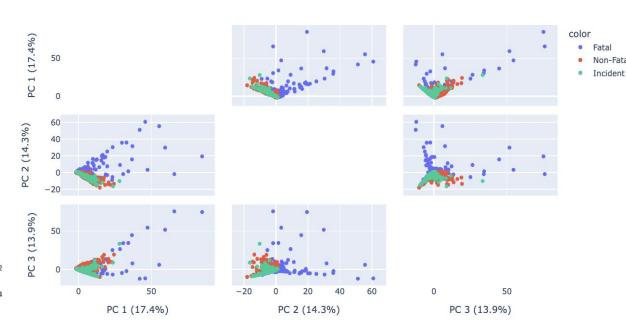
The train accuracy is: 0.9276212348732444
```

Stats	Random Forest w/ Grid Search Results
Accuracy	89.2312%
Precision	92.164%
Recall	94.822%
Specificity	64.8484%
False Negative Rate	5.177%
False Positive Rate	35.15%

Random Forest gave us the best results for Binary Classification!

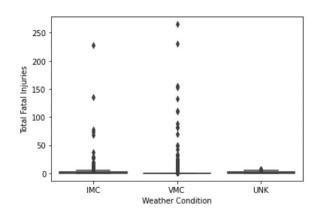
Principal Component **Analysis**

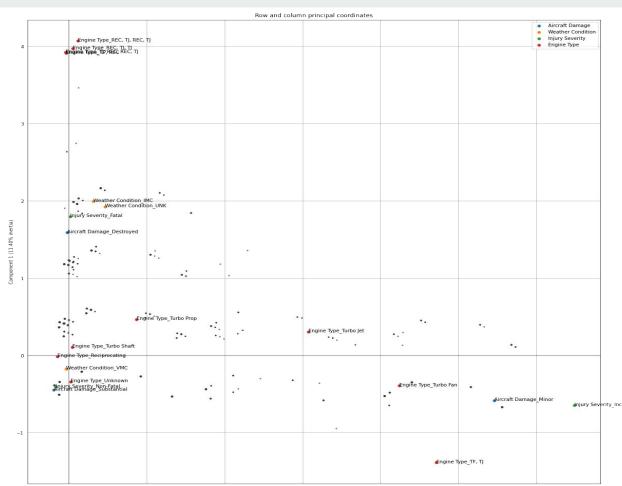




Non-Fatal

Multiple Correspondence Analysis





Component 0 (13 20% inertia)

Comparisons

When comparing the different binary classification methods such as Logistic regression, Random Forest and Decision Tree, it is completely reasonable that Random Forest outperformed both Decision Tree and Logistic Regression.

The performance of Random Forest is largely due to its emphasis on feature selection, ability to ignore linear relationships within predictors and utilization of ensemble learning.

Model	Accuracy
Logistic Regression	84%
Random Forest	89%
Decision Tree	85%

Conclusion

- Analysis focused on flight safety over time
- When predicting whether a flight is fatal, Random Forest drastically outperformed decision trees and logistic regression
- Aviation clearly becoming safer overtime
- When crashes do occur, it is feasible to predict if a crash will be fatal based on flight characteristics



