

# Understanding the Incidents on Legacy Airlines with Machine Learning: Case Study

## Top 5 US Airlines



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## Background

- The air transportation system is part of most nations critical infrastructure
- Rigorous safety standards are effective; however minor accidents/incidents are somewhat frequent in comparison to major accidents even with standards in place
- There are known patterns pilots see anecdotally; but the data must be analyzed
- Incident have a substantial cost to the airlines, raise ticket prices, erode consumer confidence, and generate insurance claims
- Modeling minor accidents to identify causal and contributing data stands to save the industry substantial costs and increase margins

## Solution Model

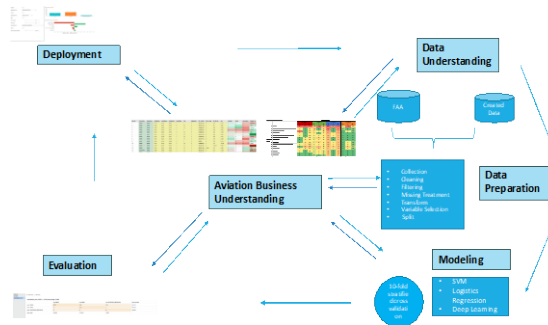
Criterion	Value	Standard Deviation
accuracy	0.994	0.003
classification_error	0.006	0.003
AUC	0.991	0.014
precision	1.0	0.0
recall	0.928	0.041
f_measure	0.962	0.022
sensitivity	0.928	0.041
specificity	1.0	0.0

## Literature

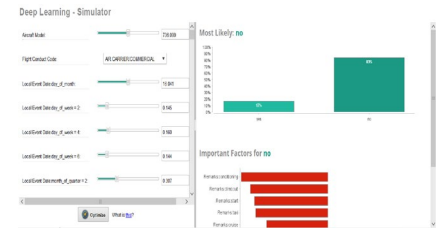
### Main Relevant Research Problems

- qualitative/quantitative approaches to detect anomalous aircraft behavior (Hwang et al, 2008)
- assess the safety, and quantify the risk associated
- causal models, collision risk models, human error models, and third-party risk models (Netjasov and Janic, 2008)
- automatically detect flight trajectory anomalies (Di Ciccio et al, 2016)
- A hybrid model blending SVM and DNN ensemble prediction for aviation incidents (Zhang et al. 2019)

Gap: Predicting Incident Damage Type and Impacting Variables is not addressed



## Results



## Results

	Method Accuracy		
	Deep Learning	Logistic Regression	SVM
Accuracy	99.4	97.1	100
Class Recall	92.7	64.4	100
AUC	0.99	0.87	1

## Results

Table 1: Text Mining Important Terms

Term Count	Topic	Topic ID
0.051	+flight attendant, +attendants, flight,	1
0.053	+engine, fire, evacuation, takeoff, NR2	2
0.056	aircraft, crew, landing, report,	3
0.056	emergency	4
0.056	head, +ch, +seat belt, +sqr, +sqr	5
0.058	slow, air turbulence, air, turbulence,	6
0.059	+encounter	7
0.059	+fall, overhead, bin, +passenger, +open	8
0.055	cart, +door, +service, service cart, galley	9
0.051	people, +injure, turbulence, fire,	10
0.051	+encounter	11

## Conclusions

- Minor damages category can be significantly predicted
- Weights by correlations define the essential variables in prediction for minor damages.
- The relations between variable is meaningful for SME