Flight Delay Predictions



Structure





Our Stakeholders

Problem introduction

Prediction I: Flight delay: Yes or No?

The data: Flights & Delay

Prediction II: Flight delay: How much?



Conclusion

Stakeholder:





Tunisair is the flag carrier airline of Tunisia.

Operates scheduled international services to four continents.

As of April 2018 The fleet currently consists of 29 Airbus and Boeing aircrafts.

Problem introduction

Nobody likes flight delays as they bring many unfortunate things with them:

- Distressful waiting for passengers
- Missed transfer opportunities
- Additional costs for the airline
- Loss of customers
- Lower efficiency for airports (More organizational effort)





Tunisair Dataset

Records of national and international flights:

- Totally ~ 100k Flights
- 3 Years of Flight Data (Jan. 2016 Dec. 2018)
- 8 Various details (Features) about the flights
- Target variable Delay time of flights
- Dataset for geographic data about the airports



Data insights: Flights & Delay

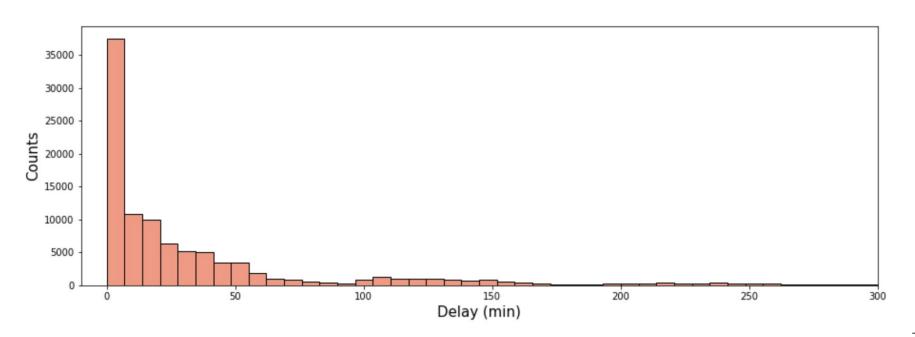
Why are the flights delayed?

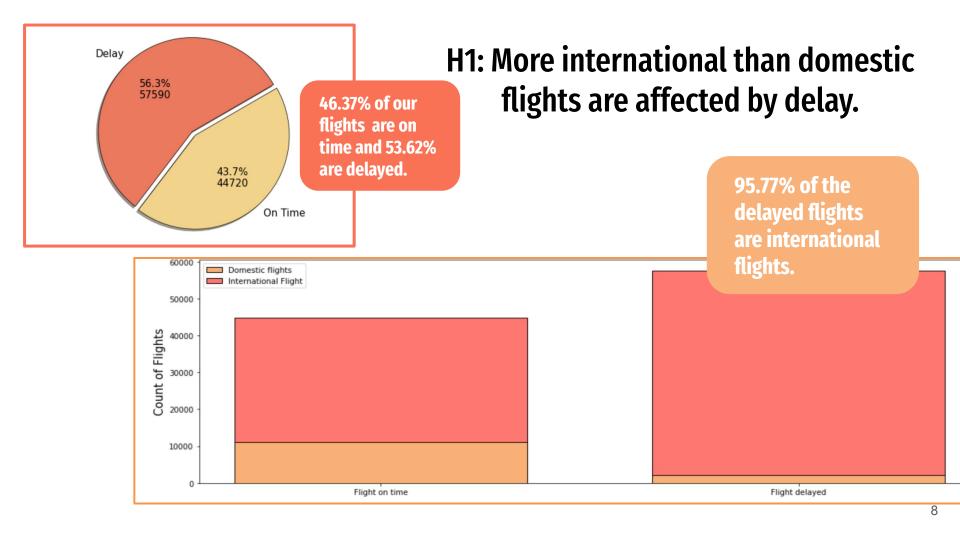
Hypotheses:

- I. More international flights than domestic flights are affected by delays.
- II. The season, day of the week and time of day play a role in the delay.
- III. The delay is due to an individual aircraft.

Data insights: Flights & Delay

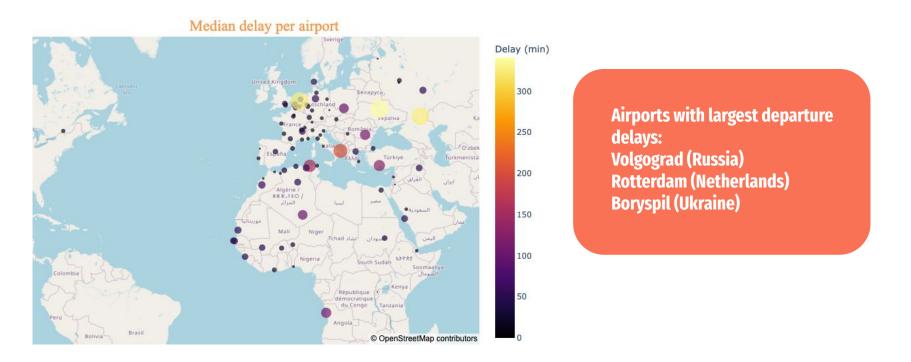
Overview of the time distribution of delays: Most flights are delayed less than one hour.



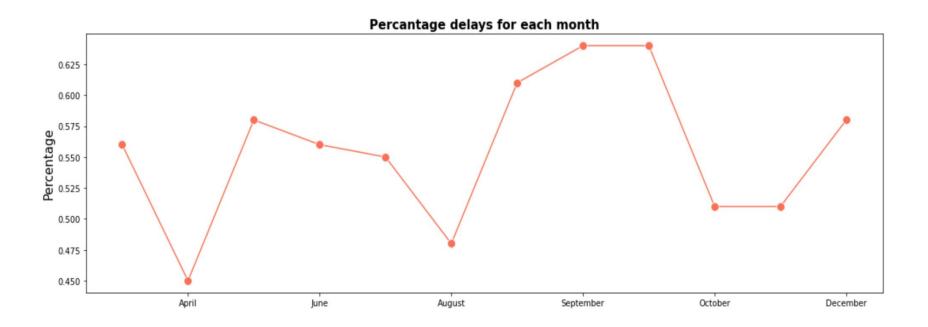


H1: More international than domestic flights are affected by delay.

Median delay per airport

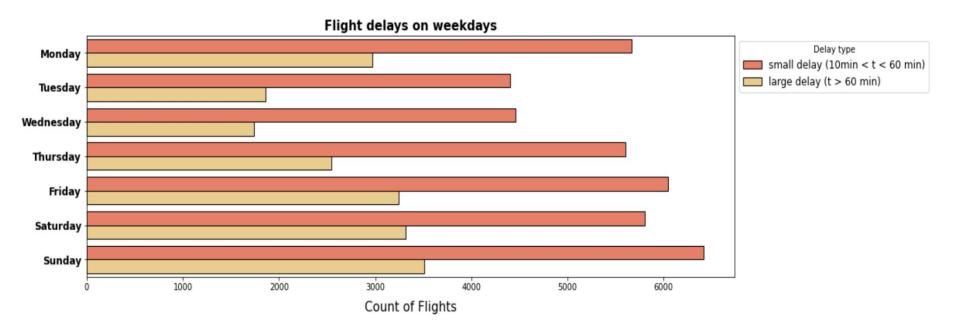


H2: The season, day of the week and time of day play a role in the delay.



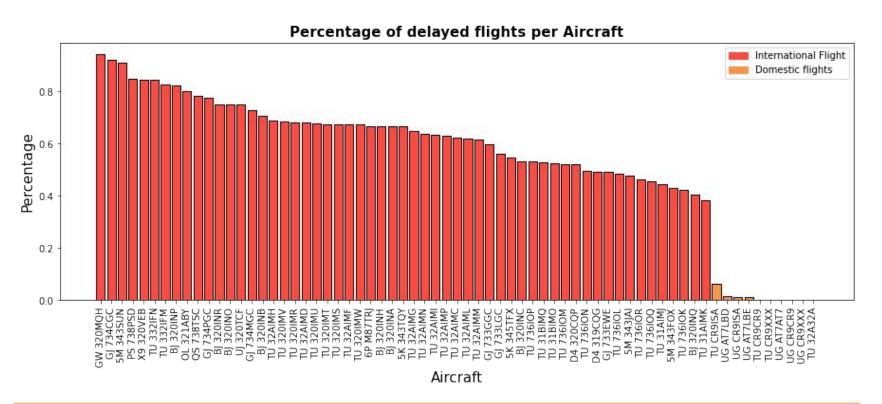
We can see a peak in percentage delays August/ September.

H2: The season, day of the week and time of day play a role in the delay.

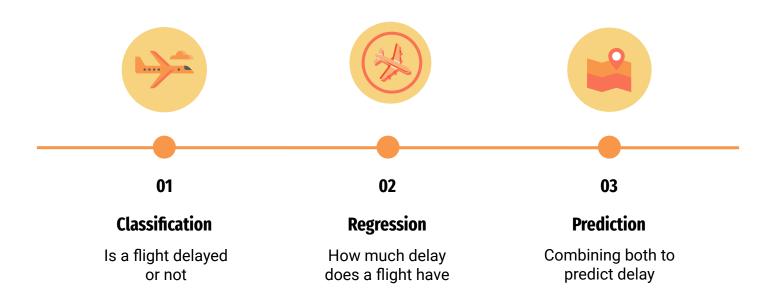


We can observe higher counts of flights that are delayed on the weekend and the beginning of the week.

H3: The delay is due to an individual aircraft.



Modeling



Baseline Model

Baseline Model: Classification
Our baseline classifies flights with an accuracy of 58%.

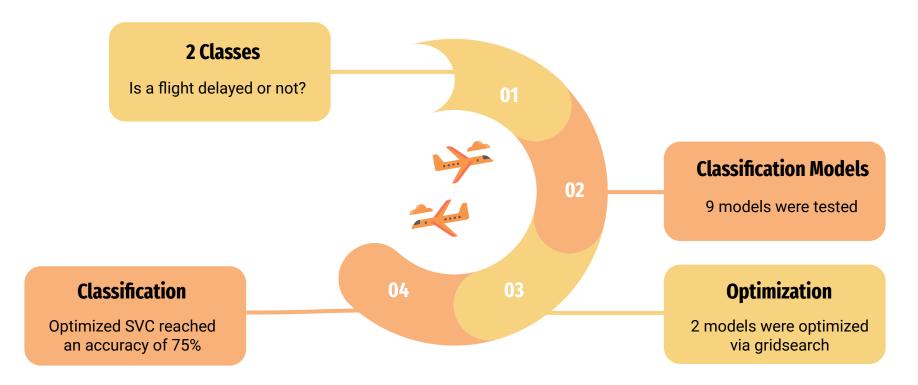
Baseline Model: Regression

Our baseline model predicts a delay with a root mean squared error of 140.09 minutes.



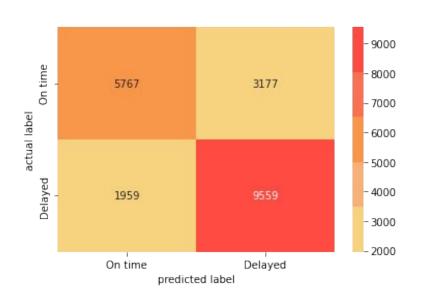
Those values we have to beat with our sophisticated models.

I - Classification



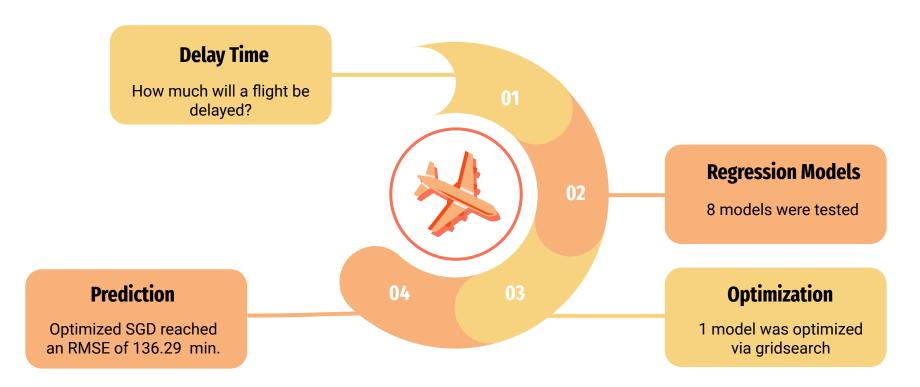
I - Classification

The best model classifies flights with an accuracy of 75%



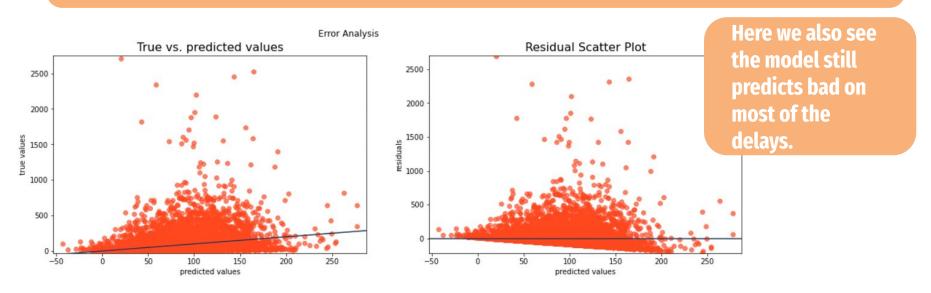
83% of delayed flights are detected

II - Regression



Prediction II: How much is the delay?

Our best model predicts the true value of the dealy with an root mean squared error of 136.29 minutes.



Conclusions



Predicting whether a flight is delayed or not works reasonably well from the existing data Predicting exact delay times is not possible.
Further data and analysis will be needed

Customer satisfaction and profits could be increased by deploying an automated delay predictor.

Thank you for listening!

