

Abstract

Accurately predicting the demand for aviation is a complex problem that is essential for the success of the private aviation industries. Factors such as seasonality and location affect the demand for private flights, but high-demand events and holidays introduce additional and often unexpected influences on these services. In European destinations, travel is heavily characterized by high-demand events and holidays. This research utilizes detailed characterization data centered around Europe containing over 1.1 million private flights between 2,016 locations from 2018 and 2019. Leveraging advanced data analysis techniques, this project constructs a spatio-temporal forecasting model to accurately predict the demand for private jet travel during high-demand events and holidays in European destinations. This research delivers valuable insights to providers of private aviation, enabling them to proactively respond to market fluctuations and optimize their operational strategies.

Research Scope

This research is interested in the interaction of demand factors for events that move across locations over time. Previous research has found factors that influence private aviation demand, but not during high demand events. Through an exploration of the private flight dataset and previous studies on luxury travel, this poster documents a normalization procedure. This methodology is then extended to forecasting future arrival counts.

Data Visualization

• 1,117,298 private jet flights from 2018 and 2019



Figure 1: Heat Maps of Total Arrivals per Day of the Week and Week of the Year



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Event-Driven Demand Modeling of European Private Aviation Travel

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Global Model

• A predictive model is built using the 2018 arrival data, with coefficients indicative of day of the week and month of the year.



Figure 2: Line Plot of 2019 Flight Data and Me

Arrivals(∂, μ) = $\beta_{\text{baseline}} * \partial_{\text{day of week}} * \mu_{\text{month}}$

Data Description and Methodology

- This method uses a base year to quantify events in context. For each 1,117,298 private jet flight arrival and departure records from 2018 and 2019
- The data set had several missing entry data that had to be discarded. This did not affect the subset used for this project.
- The arrival counts are given for each airport's ICAO identifier. Airports that receive majority private flights were chosen for analysis as more representative of dataset.
- After determining that time of year and day of week were relevant, a more sophisticated data normalization algorithm was created. This method uses a base year to quantify event demand in context. As a result, results from one event can be used to forecast similar, but distinct events. Such distinctions result from differences in location, time, economic conditions, and other idiosyncratic factors like event hype. For example, we used a soccer championship event in 2018 in Ukraine to forecast the demand spike for the 2019 championship in Madrid.

- 2250 - 2000 - 1750 - 1500 - 1250 - 1000 - 750 - 500 - 2250 - 2000 1750 - 1500

> - 1250 - 1000

- 750

	Baseline	Coefficient		
	β_{baseline}	830		
	Month	Coefficient		
ctual Data	January	0.983		
Data	February	1.138		
in mark 0.70	March	1.172		
lared: 0.76	April	1.201		
	May	1.439		
	June	1.731		
	July	1.814		
	August	1.529		
	September	1.554		
MANN	October	1.328		
TT INNAA	November	1.147		
	December	1.0		
	Day of the Week	Coefficient		
	Monday	1.140		
	Tuesday	1.214		
	Wednesday	1.303		
	Thursday	1.359		
350	Friday	1.468		
550	Saturday	1.0		
	Sunday	1.230		
odel Data	Table 1: Model Coefficients for Month and Day of the Week			

Algorithmic Normalization Method

This method uses a base year to quantify events in context. For each day of week, it evaluates forward and backward-looking ratios of arrival counts to see if they fall within a threshold. This outlines unique trends in seasonality at a given airport for events normalization.

Event Characterization



Locations	 2018: Ukraine and Russia: Boryspil International A Sikorsky International A Pulkovo Airport 2019: Spain: Adolfo Suárez Madrid–E 	
Date	May 26, 2018 and June 1, 2	
Factor of Change	Model: • 2018: 13.50 • 2019: 6.90	Algorithm 2018: 11

Future Work

- Sensitivity testing on algorithm's parameters
- Further break down and store the normalized event coefficients by days out from event, size of airport in yearly arrivals, and distance of airport from event center
- Create mapping algorithm for the correct application of coefficients in cases like an event being truncated or an event starting on different days of the week between years

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• Residual testing over more events with both methods to determine fit

