



An Econometric Analysis of US Airline Flight Delays with Time-of-Day Effects

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Outline

Introduction Model □ Factors Data Estimation Results Delay Change Decomposition Conclusions





Introduction:

Arrival Delays & Scheduled Arrivals







Goals

- Develop <u>statistical</u> model to forecast delay in the NAS
- Certain advantages over simulation models
 - Estimated over large number of days
 - Reduced data needs
 - Instantaneous calculation
 - Future year prediction based on application to a full year of days rather than a few sample days
- Compare results with simulation





Objectives

- Model daily average arrival delay
- Assess impacts of
 - Queuing
 - Volume
 - Weather (storm and terminal conditions)
 - Seasonal Effects
- Investigate time of day effects for queuing delay





Model

$AvDelay(t) = f(\overline{Queuing}(t), SFlights(t), GAFlights(t), MFlights(t), IFR(t), Windspeed(t), THX(t), \overline{THX}(t), \overline{Events}(t), \overline{Season}(t), \overline{Period}(t)) + v_t$





Queuing

Calculate deterministic queuing delay on a daily basis for each of 32 DOT airports

Based on

Quarter-hr demand profile (based on schedule and adjusted for cancelled flights)

Quarter-hr AAR





Queuing: Procedure

Airport level

- Construct Cumulative Arrival Demand Curve
- Construct Cumulative Arrival Count Curve
- Calculate Delay as Area Between Demand and Count Curves
- System level
 - Summed airport delays to get total delay
 - Divided by total arrivals to get average delay







En Route Weather (Storms)

- Based on Surface Summary of the Day from NOAA (National Oceanographic and Atmospheric Administration)
- Daily summary from ~1500 US weather stations
- Each station reports binary (yes/no) thunderstorm variable
- Used proportion reporting thunderstorms as storm metric
 - Construct area-specific thunderstorm metrics on lat-long quandrangles
 - □ Include each metric as separate explanatory variable









Terminal Weather

Proportion of the day under IFR Airport IFR proportion: proportion of time under IFR

- System IFR proportion: weight airport IFR proportion by # of operations
- Higher IFR proportion, expected higher delay due to lower AAR

Wind speed

- Airport wind speed: avg. wind speed of the day
- System wind speed: weight airport wind speed by # of operations
- □ Higher wind speed, expected higher delay





Other Effects

- Volume Scheduled arrivals **GA** operations Monthly fixed effects Based on month Effects of winds and other weather factors Yearly fixed effects Year-to-year trends not otherwise accounted for □ May reflect FAA performance among other factors
- UA strike effect
- 9/11 Effect





Estimation

Data

Daily data: Jan. 2000- June 2005

Excludes 12/20-25, 9/11-30/01, and standard/daylight savings transition days

Estimations

OLS model: Heteroskedasticity and autocorrelationGARCH model

Model form

$$y_{t} = \mathbf{X}_{t}^{'} \mathbf{\beta} + v_{t} \qquad h_{t} = ARCH0 + ARCH1 \cdot \varepsilon_{t-1}^{2}$$

$$v_{t} = \varepsilon_{t} - AR1 \cdot v_{t-1} \qquad + GARCH1 \cdot h_{t-1} + HET1 \cdot Q(t)$$

$$\varepsilon_{t} = \sqrt{h_{t}} \cdot e_{t} \qquad e_{t} \sim IN(0,1)$$





Estimation Results

	Variable	Baseline Standard		Time-of-Day Standard		Interaction Standard		Full Standard		
Category	(parameter for category									
	GARCH)	Estimate	Error	Estimate	Error	Estimate	Error	Estimate	Error	
	Intercept	-14.156	2.351	-15.405	2.268	-9.663	2.357	-9.532	2.265	
Queuing	Queuing	1.457	0.202			1.448	0.197			
	Queuing (00:0008:00)			0.895	0.377			0.861	0.375	
	Queuing (08:0012:00)			3.304	0.607			3.099	0.599	
	Queuing (12:0016:00)			2.236	0.446			2.286	0.443	
	Queuing (16:0024:00)			1.029	0.207			0.994	0.205	
	Queuing ²	-0.079	0.025	-0.074	0.024	-0.082	0.025	-0.072	0.024	
Volume	Scheduled arrivals (000)	0.964	0.157	1.009	0.153	0.771	0.163	0.675	0.159	
	GA operations (000)	0.098	0.030	0.131	0.030	0.073	0.030	0.118	0.029	
	IFR ratio	14.475	0.824	13.524	0.840					
Terminal	Wind speed	0.490	0.071	0.481	0.070					
Weather	Sch*IFR ratio					0.828	0.047	0.800	0.048	
	Sch*Wind speed					0.029	0.004	0.029	0.004	
En Route Weather	Storms in region 1	1.644	0.658	1.684	0.657					
	Storms in region 2	5.627	0.814	5.636	0.816					
	Storms in region 3	2.269	0.661	2.271	0.653					
	Storms in region 4	3.946	0.475	3.939	0.468					
	Storms in region 5	9.736	0.853	9.753	0.850					
	Storms in region 6	10.153	0.788	10.061	0.789					
	Military Operations (000)	-0.246	0.035	-0.314	0.035					
	Sch*Storms in region 1					0.120	0.037	0.106	0.037	
	Sch*Storms in region 2					0.358	0.046	0.345	0.046	
	Sch*Storms in region 3					0.137	0.037	0.125	0.037	
	Sch*Storms in region 4					0.240	0.027	0.229	0.026	
	Sch*Storms in region 5					0.529	0.047	0.520	0.045	
	Sch*Storms in region 6					0.605	0.044	0.583	0.044	
	Sch*Military Operations (000)					-0.015	0.002	-0.017	0.002	
Event	UA strike dummy	1.820	0.631	1.720	0.628	1.684	0.626	1.460	0.618	
	2nd half 2001 after 9/11dummy	0.042	0.883	0.281	0.875	-0.114	0.879	0.394	0.864	
R-Squared		0.7	0.720		0.724		0.727		0.729	

Note: "Scheduled arrivals" is abbreviated as "Sch."; Time dummies and GARCH variables are used but not shown here.





Key Findings

- □ Time of day queuing effects
 - □ Vary strongly by time-of-day
 - Greatest impact in morning (delay propagation)
 - Concave relationship
- Volume
 - Scheduled arrivals: 1000 additional scheduled flights increases average delay by 0.68-0.77 minutes on extremely good weather days
 - □GA operations: 1000 additional GA operations in increases average delay by 0.1 minutes
 - Military operations: strong negative relationship with delay (!)





Key Findings (cont)

- Terminal weather effects
 - Depend on volume
 - Have impact even when controlling for queuing
- En route weather effects
 - Depend on volume
 - □Wide geographic variation
- Other effects

2000 UA job action had pronounced effect
 Post 9/11 effect not significant (captured by other variables)









Conclusions

- Statistical models help us explain and predict delay changes
- Use of deterministic queuing increases fidelity of these models and allows time-of-day effects to be captured
- En route convective weather effects can be captured in fairly simple ways
- Substantial unexplained variation remains