

# Optimal Pricing Rule for One-way Airline Tickets

Empirical evidences

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## Dynamic pricing in airline industry



Time based theory	Capacity based theory
<ul> <li>Gale and Holmes (1992, 1993)</li> <li>Gallego and van Ryzin (1994)</li> <li>Piga and Bachis (2007)</li> <li>Möller and Watanabe (2010)</li> </ul>	<ul> <li>Dana (1999)</li> <li>Escobari and Gan (2007)</li> <li>Deneckere and Peck (2012)</li> </ul>

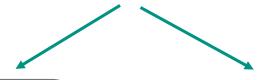
#### Combination of both theories

- Anjos, Cheng and Currie (2004)
- Alderighi, Nicolini and Piga (2014)
- Alderighi, Gaggero and Piga (2018)





Anjos, Cheng and Currie (2004, JORS)



**What** -> optimal pricing rule for airline tickets under oneway pricing

**How** -> analytical model + standard analytical method for constrained optimization

**Results** -> simple pricing rule; relationship between days before departure and distance from expected booking curve

#### **Dataset**



- → Sample routes / period / collection
  77,221 Ryanair flights, 81 out of 154 routes from UK over
  Jan 2004 Jun 2005; data collected with an electronic spider
- → Fares
  11 fares at different days to departure (weekly sampled)
- → Seats

  Available seats detected if less than 50





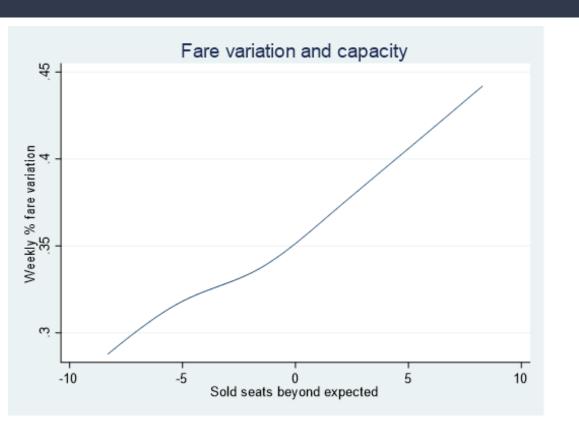
- > Fare hikes over time
- From day 42 to day 21 fares grow at a decreasing rate
- ➤ Inflection point at day 21
- ➤ Growth at increasing rates during the 2 weeks before departure





- ➤ Increasing number of seats sold when approaching the departure date
- Booking curve: mean calculated on available seats





- ➤ Anjos, Cheng and Currie (2004) conjecture satisfied
- > Fares follow a simple rule:
  - increase more when above the booking curve
  - o increase less when **below** the booking curve

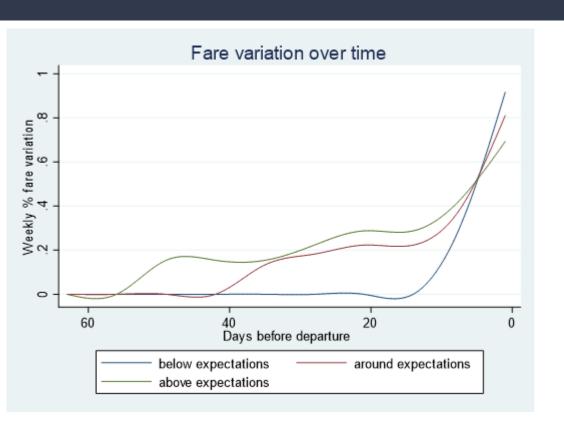






- Simple pricing rule continues to hold over different booking periods
- Relationship between sold seats beyond expected and fare variation evolves in time





- On average, fares are rised more and earlier when above the booking curve
- On average, no evidences for price drops (Biloktach, Gaggero & Piga, 2015, TM)

## Methodology



→ Sample selection

Aircraft capacity: 0 - 189

Available seats detected: if less than 50

Bias corrected with: Tobit regression (Alderighi et al., 2014)

→ Panel data approach

Log-lin random effect panel estimator with selection coefficient

 $\rightarrow$   $\Delta \ln (Fare) = sold seats beyond expected + D(days to departure) + selection coefficient$ 

#### Results

$\Delta \ln(\text{Fare})$	(1)	(2)	(3)
$\Delta E(Seats)$	0.020***	0.008***	
$\Delta E(Seats)^+$			0.015***
$\Delta E(Seats)^-$			-0.002
1 day to dep.		0.884***	0.860***
7 days to dep.		0.520***	0.496***
14 days to dep.		0.234***	0.214***
21 days to dep.		0.192***	0.177***
28 days to dep.		0.153***	0.144***
35 days to dep.		0.145***	0.139***
42 days to dep.		0.123***	0.119**
49 days to dep.		0.123**	0.123**
56 days to dep.		0.112**	0.114**
$\Lambda$	0.019***	0.007***	0.008***
Constant	0.864***	0.256***	0.217***
R-squared	.00475	.0364	.0372
Observations	107043	107043	107043



Standard Errors clustered by route. Significant at \*10%, \*\* 5%, and \*\*\*1%.



#### **Conclusions**

- Optimal simple pricing rule confirmed
- Both capacity and time theories prove to be relevant, especially if considered together
- > Fares hike over time, mainly during the 2 weeks before departure
- Symmetric fare variation when sold seats are above or under expectations

## Future steps



➤ Instrumental variable approach (endogeneity fare & available seats)

➤ Bayesian approach for estimating expected booking curve

Test the same model for other companies (e.g. full service)