

Choice Based Airline Revenue Management Modeling with Flexible Substitution Patterns

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Content



Introduction



Research Contribution

Development of a revenue maximization model

- ▶ that decides on the optimal seat allocation
- ▶ that sets the optimal prices for each offered seat
- ▶ based on integrating demand from discrete choice models

Assumptions

- ▶ Static model
- ▶ Single resource with capacity C
- ▶ Multiple products → fare classes
- ▶ Two compartments → business and economy class
- ▶ Demand is modeled as fare class choice → utility maximization
- ▶ Stochastically dependent demand

Approach

Motivation

- ▶ Simplifying assumptions in static single-resource RM models
 - Demand for different fare classes are stochastically independent
 - Demand does not depend on other available fare classes
- ▶ Constant substitution patterns between fare classes

Realization

1. **Simulation** based model for revenue optimization
2. Demand from a DCM^a with **not constant substitution patterns**

^aDiscrete Choice Model



Demand Model



Discrete Choice Models

- ▶ Allow to model multiproduct RM with *flexible*¹ capacity
- ▶ Stochastic demand models
- ▶ **Information asymmetrie** between seller and buyer

Seller

- ▶ Only partial information on choice decision
- ▶ Observes individual utility values from actual or stated choice decisions

Buyer

- ▶ Full information on choice decision
- ▶ Choice decision is influenced by
 - (i) Characteristics of the alternatives and the customer
 - (ii) Customers evaluation of alternatives
- ▶ Stochastically dependent demand structures

¹Ability to offer different products using the initial capacity C [BC03].

Properties

Multinomial Logit Model (MNL)

$$U_f = V_f + \varepsilon_f \quad \forall f$$

- ▶ Constant substitution
- ▶ ε_f is iid EV distributed
- ▶ Identical cross elasticities
- ▶ IIA property
- ▶ [TVR04]

Nested Logit Model (NL)

$$U_f = V_f + \boxed{\varepsilon_m + \varepsilon_{fm}} \quad \forall f \in C_m | m$$

- ▶ No constant substitution
- ▶ Correlation of pairs of f
- ▶ Generalization of MNL
- ▶ More realistic choice behaviour
- ▶ [WK01]



Application I: Modelling Demand - Fare Class Choice

Utility Function

Considered function of deterministic utility

$$U_f = \beta_{f,asc} + \beta_{price} \cdot x_{f,price} + \beta_{f,prp} \cdot x_{prp} + \beta_{f,gen} \cdot x_{gender} + \varepsilon_f \quad \forall f$$

- ▶ Generation of synthetic datasets based on U_f
- ▶ NL and MNL models differ according to ε_f
- ▶ Coefficient values are provided → "true" values

Results

- ▶ Two populations with 10.000 individuals each
- ▶ Choice behaviour according to NL and MNL models
- ▶ Datasets exhibit expected substitution patterns
- ▶ Assumption: NL population exhibits "true" behaviour

Alternatives, Variables & Nesting

- ▶ Choice set contains 4 fare class alternatives i :
 - ▶ Regular (1) and discount (2) fare in Business Class
 - ▶ Regular (3) and discount (4) fare in Economy Class
 - ▶ No choice/Outside Alternative (OA) (5)
- ▶ Provided coefficient values
 - ▶ $\beta_{price} = -0.0040$
 - ▶ $\beta_{1,prp} = 2.0, \beta_{2,prp} = 1.5, \beta_{3,prp} = 1.0, \beta_{4,prp} = 0.5, \beta_{5,prp} = 0$
 - ▶ $\beta_{1,gen} = 0.8, \beta_{2,gen} = 0.5, \beta_{3,gen} = 0.2, \beta_{4,gen} = -0.1, \beta_{5,gen} = 0$
- ▶ Assignment of alternatives to nests
 - ▶ Nest 1 = {1,2}
 - ▶ Nest 2 = {3,4}
 - ▶ Nest 3 = {5}



Tree Structure

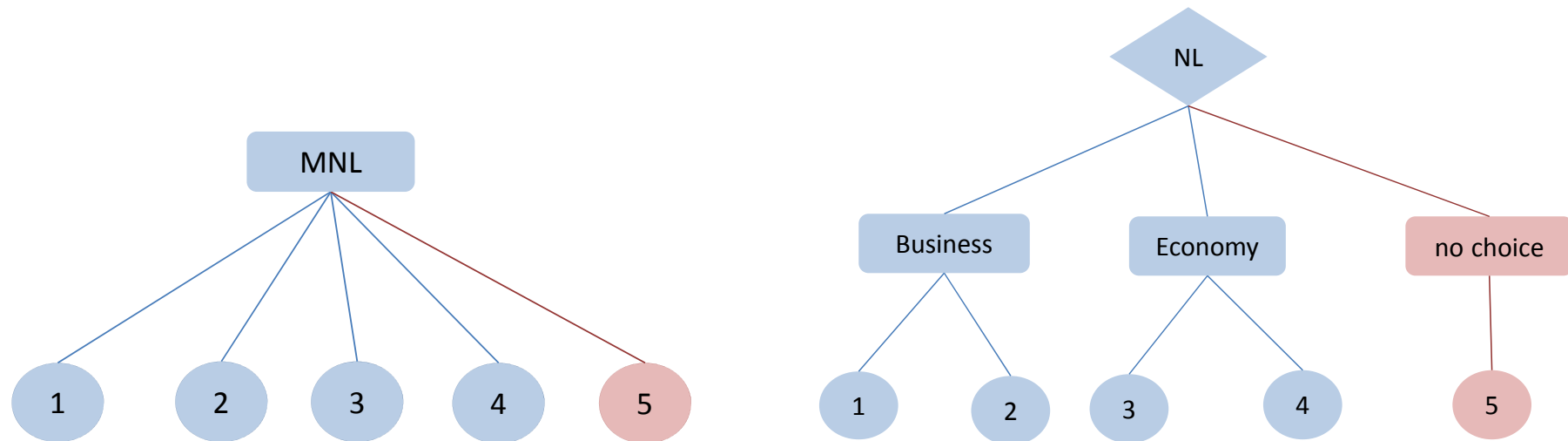


Figure: Tree structure for fare class choice in the MNL (left) and NL (right)

Demand Integration

1. Draw random sample of 2000 observations from **NL** population
2. Use sample as dataset for estimation
3. Estimate NL model from NL sample → NL coefficients
4. Estimate MNL model from NL sample → MNL coefficients
5. Put NL coefficients in optimization model with NL utility function
6. Put MNL coefficients in optimization model with MNL utility function
7. Compare results

Estimation Results

Parameter	Name	NL estimate	t-stat	MNL estimate	t-stat
1	$\beta_{asc,1}$	0.321	0.69	0.229	0.37
2	$\beta_{asc,2}$	1.83	4.40	2.51	8.23
3	$\beta_{asc,3}$	1.32	6.83	1.16	5.81
4	$\beta_{asc,4}$	1.98	14.31	2.15	17.44
5	β_{price}	-0.00459	-8.64	-0.00559	-15.75
6	$\beta_{prp,1}$	1.08	3.47	1.30	3.21
7	$\beta_{prp,2}$	0.361	1.95	0.347	1.80
8	$\beta_{prp,3}$	0.145	0.97	0.231	1.42
9	$\beta_{prp,4}$	-0.193	-1.60	-0.212	-1.74
10	$\beta_{gen,1}$	2.40	6.35	2.68	5.34
11	$\beta_{gen,2}$	1.47	7.53	1.45	7.16
12	$\beta_{gen,3}$	1.07	6.64	1.22	7.31
13	$\beta_{gen,4}$	0.516	4.14	0.477	3.80
14	NestA	1.73	1.77 ¹	—	—
15	NestB	1.36	1.71 ¹	—	—

¹t-test against 1

Application II: Simulation based RM Optimization Model

Preface

Generate utility values for two populations of individuals

- ▶ Choice behavior based on MNL and NL models
- ▶ Maximum utility determines choice decision
- ▶ Linear model formulation
- ▶ General formulation in terms of applied DCM
- ▶ Calculation of optimal seat allocation per fare class
 - Decision on fare classes offered to an individual
 - Fare classes are offered in a revenue maximizing manner
 - Choice sets vary across individuals
- ▶ Calculation of optimal price per fare class f
- ▶ Solved with Cplex in GAMS version 23.9

Definitions

Sets

S	Simulated sets of individuals
F	Set of fare classes indexed f , no choice/OA $f = 5$
N	Set of simulated individuals indexed $n \forall s \in S$

Parameters

$\bar{p}_f$	Upper bound for price variable in fare class f
L	Sufficiently large number
A	Sufficiently large number to guarantee $u_{snf} > 0$
C	Seat capacity of considered resource
d_{snf}	Deterministic utility plus A , without $\beta_{price} \cdot p_f$
β_{price}	Price coefficient



Definitions

Variables

u_{snf}	Utility individual n receives from choosing fare class f in s
$\tilde{u}_{sn}$	Maximum utility value for individual n in s
p_f	Price for a ticket in fare class f in s
π_{snf}	Price individual n pays for a ticket in fare class f in s
b_f	Maximum amount of bookings in fare class f in s
y_{snf}	$= 1$, if individual n chooses fare class f in s (0, else)
x_{snf}	$= 1$, if f is offered to n in s (0, else)
Z	Objective value

Objective Function

$$\max Z = \frac{1}{|S|} \cdot \sum_s \sum_n \sum_f \pi_{snf} \quad (1)$$

Price & Utility

$$u_{snf} - d_{snf} - \beta_{price} \cdot p_f \leq 0 \quad \forall s, n, f \quad (2)$$

$$u_{snf} - d_{snf} - \beta_{price} \cdot p_f + L \cdot (1 - x_{snf}) \geq 0 \quad \forall s, n, f \quad (3)$$

$$\pi_{snf} - \bar{p}_f \cdot y_{snf} \leq 0 \quad \forall s, n, f \quad (4)$$

$$\pi_{snf} - p_f \leq 0 \quad \forall s, n, f \quad (5)$$

$$\tilde{u}_{sn} - u_{snf} \geq 0 \quad \forall s, n, f \quad (6)$$

$$u_{snf} - L \cdot x_{snf} \leq 0 \quad \forall s, n, f \quad (7)$$

$$\tilde{u}_{sn} - u_{snf} - L \cdot (1 - y_{snf}) \leq 0 \quad \forall s, n, f \quad (8)$$

Remaining Constraints

$$\sum_n y_{snf} - b_f \leq 0 \quad \forall s, f < 5 \quad (9)$$

$$\sum_f y_{snf} = 1 \quad \forall s, n \quad (10)$$

$$y_{snf} - x_{snf} \leq 0 \quad \forall s, n, f \quad (11)$$

$$x_{snf} - x_{sn-1f} \leq 0 \quad \forall s, n > 1, f < 5 \quad (12)$$

$$b_f \geq 0 \quad \forall f \quad (13)$$

$$p_f \geq 0 \quad \forall f \quad (14)$$

$$\pi_{snf} \geq 0 \quad \forall s, n, f \quad (15)$$

$$x_{snf} \in \{0, 1\} \quad \forall s, n, f \quad (16)$$

$$y_{snf} \in \{0, 1\} \quad \forall s, n, f \quad (17)$$



Results



Overview

- ▶ Results for varying price coefficient β_p
- ▶ β_p changes according to start value (MNL/NL) by $+0.0005$
- ⇒ Change in individuals price sensitivity
 - ▶ Demand is constant with 10 generated individuals per simulation run
 - ▶ Available Capacity is $C = 5$
 - ▶ Simulated sets of individuals $S = 20$
 - ▶ Comparison for NL and MNL models



Open fare classes for number of individuals

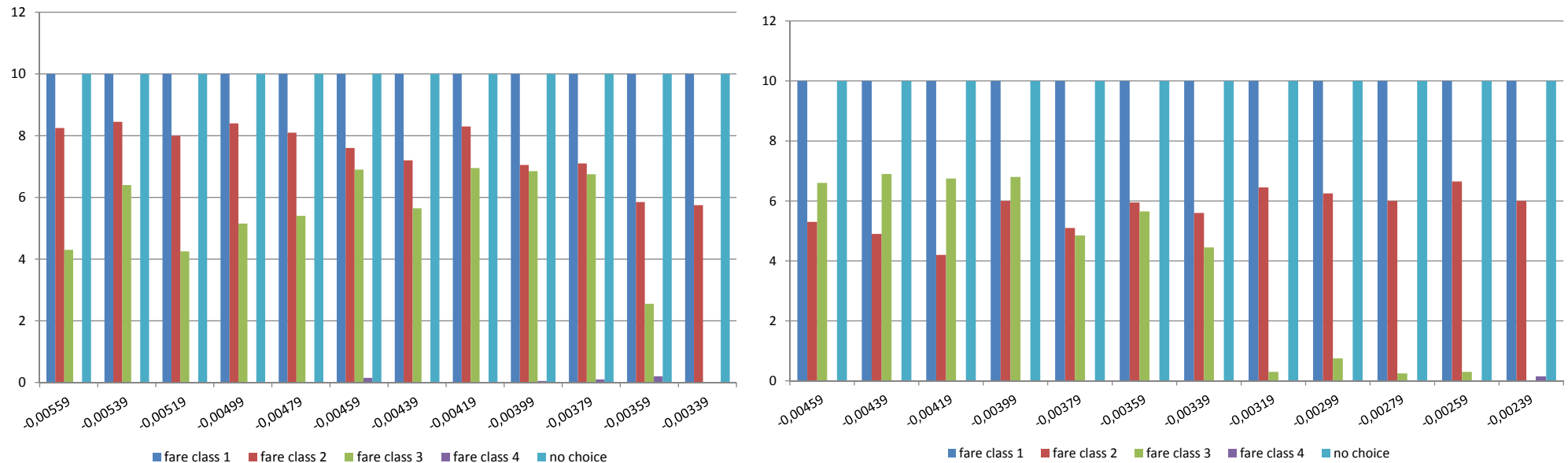


Figure: Open fare classes for MNL (left) and NL (right) utilities



Chosen fare classes on average

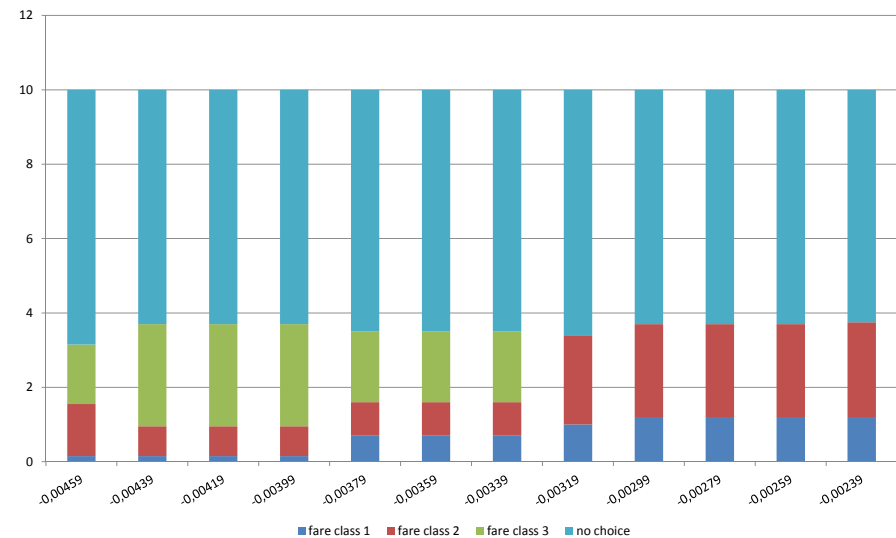
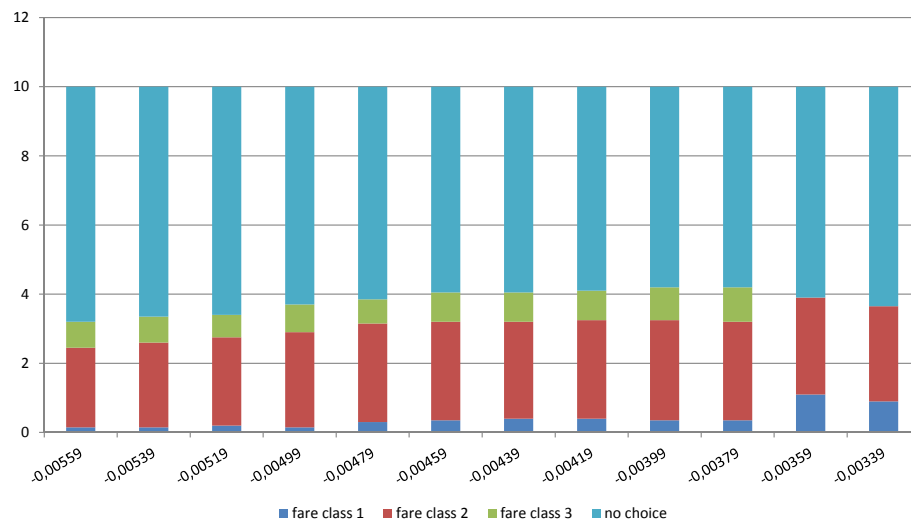


Figure: Number of bookings for MNL (left) and NL (right) utilities



Prices payed on average

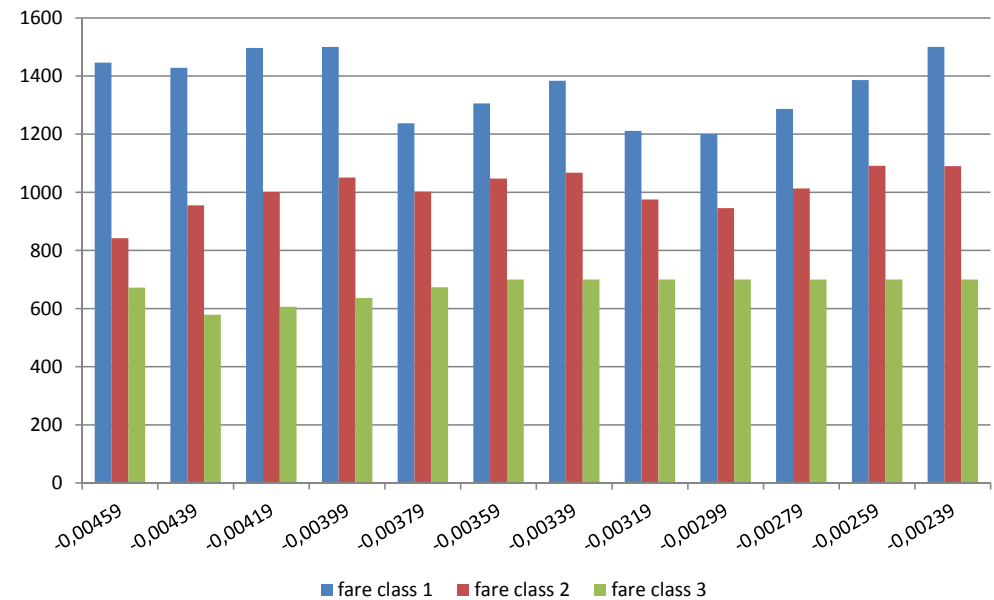
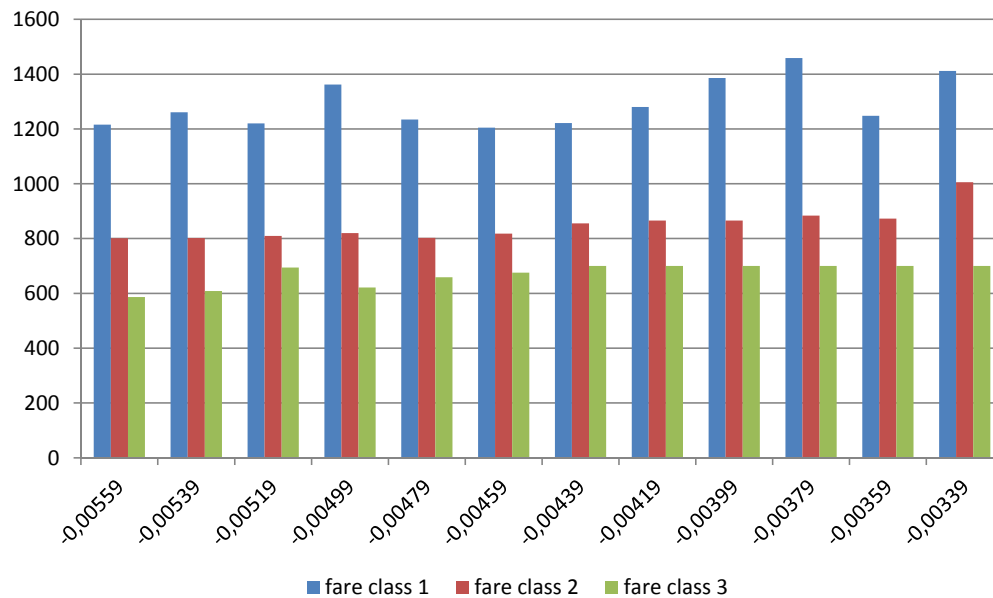
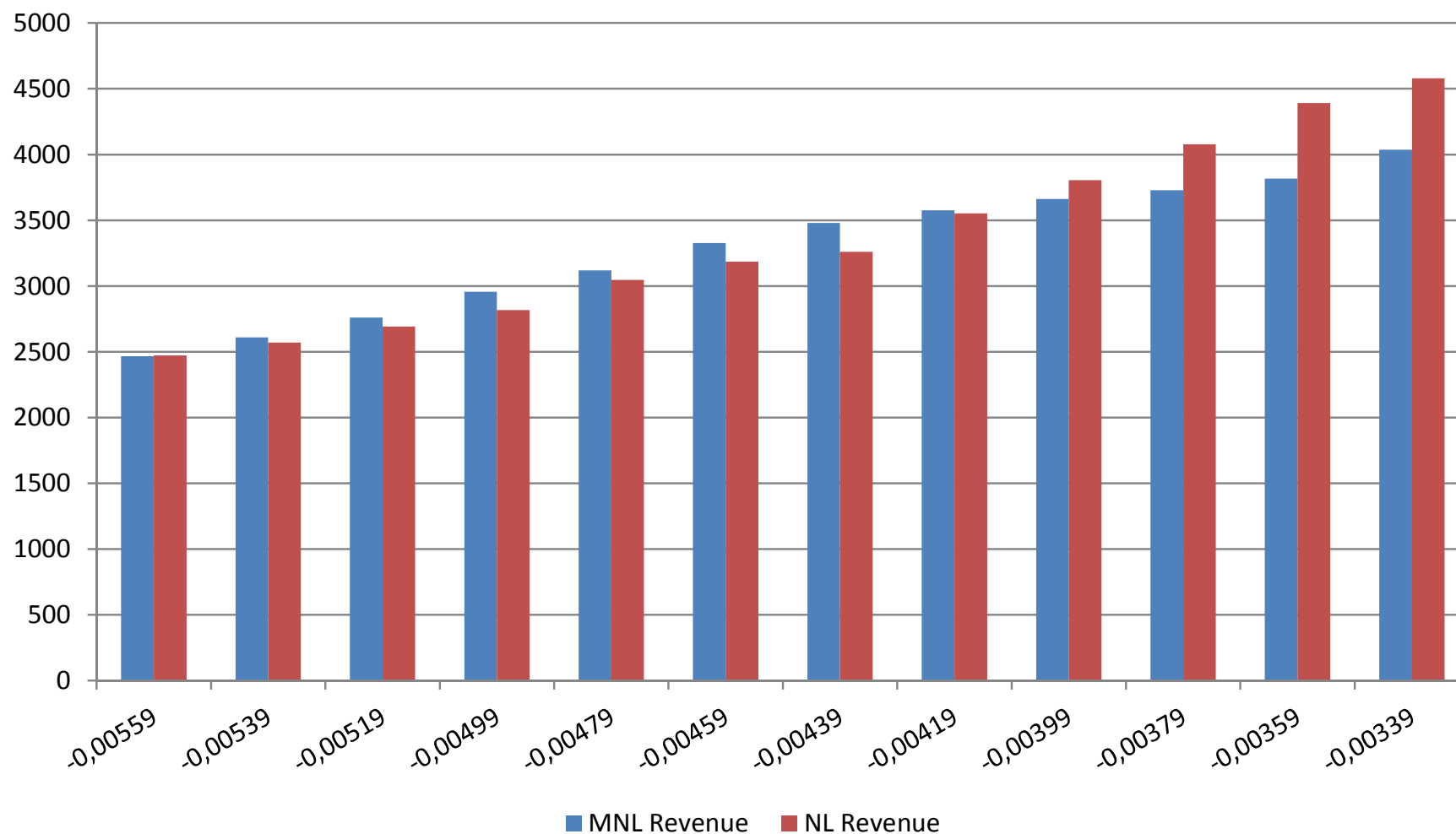


Figure: Prices payed on average for MNL (left) and NL (right) utilities



Average Revenues





Direct Elasticities

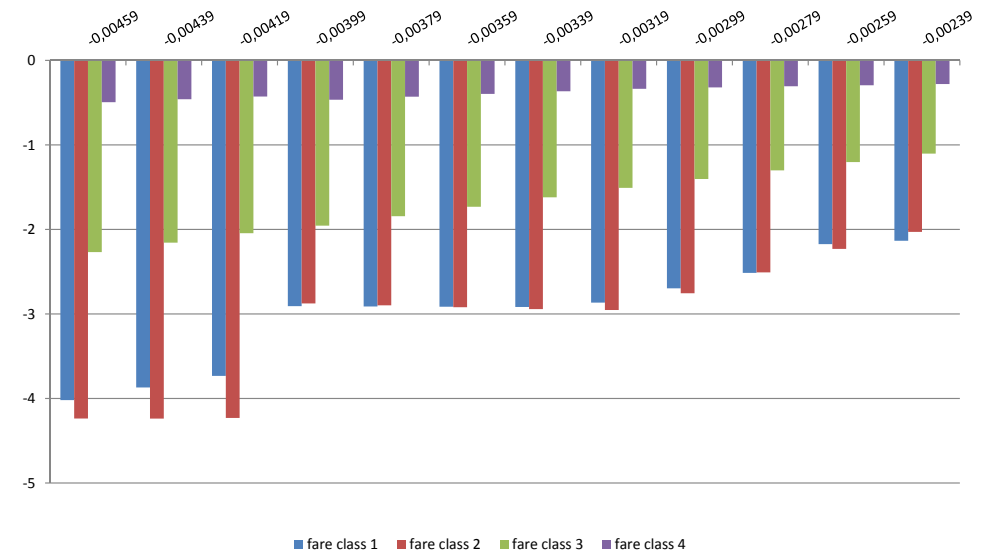
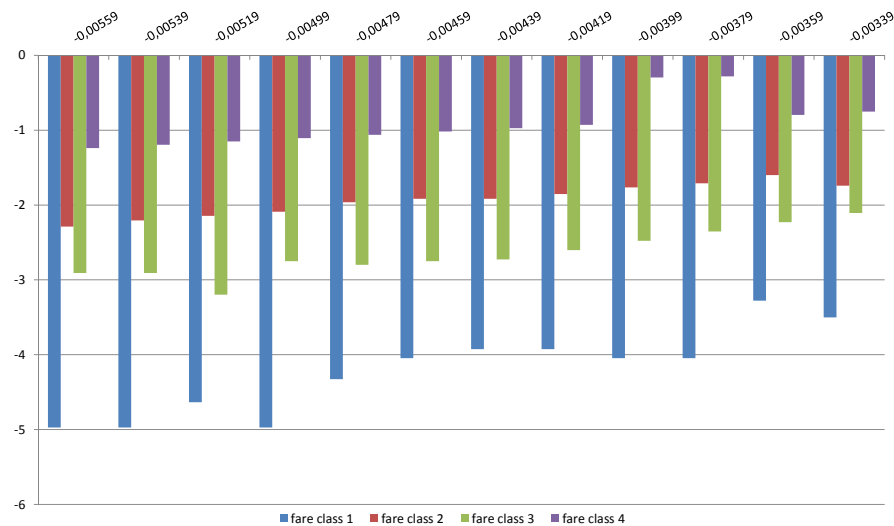


Figure: Direct elasticities for MNL (left) and NL (right) utilities



Cross Elasticities

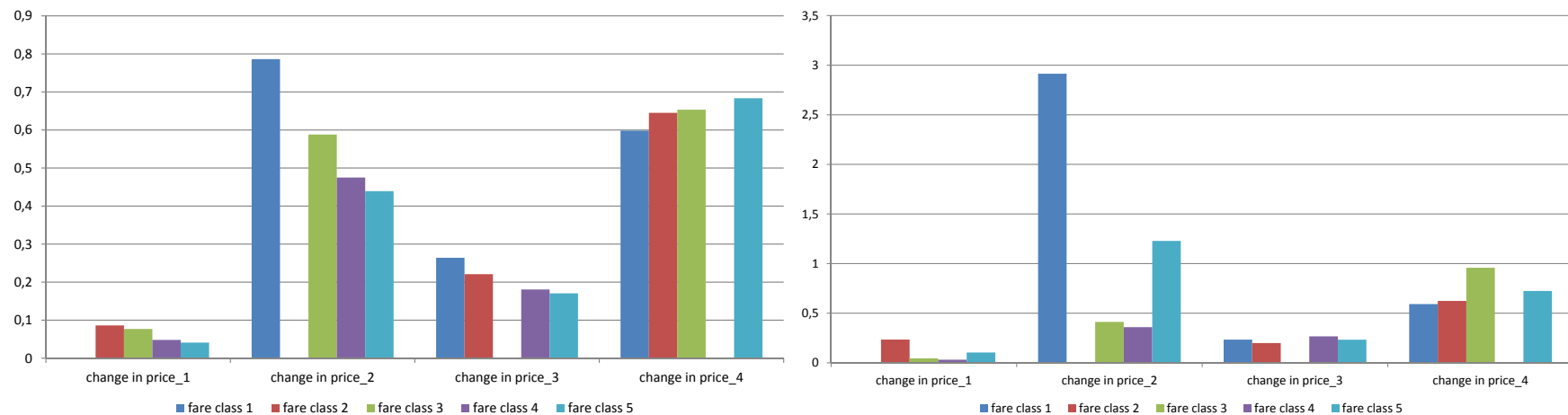


Figure: Cross elasticities for MNL (left) and NL (right) utilities



Thank you very much for your attention!



Coefficients

Coefficient	true value	MNL		NL	
		estimate	std	estimate	std
$\beta_{asc,1}$	0.50	0.869	0.193	0.388	0.194
$\beta_{asc,2}$	1.50	1.590	0.137	1.290	0.179
$\beta_{asc,3}$	1.50	1.510	0.0797	1.430	0.0671
$\beta_{asc,4}$	2.00	1.970	0.0555	1.890	0.0616
$\beta_{asc,5}$	0.00	.	.	.	.
β_{price}	-0.0040	-0.0042	0.000147	-0.0040	0.000226
$\beta_{purpose,1}$	2.00	1.960	0.150	1.970	0.142
$\beta_{purpose,2}$	1.50	1.640	0.0971	1.490	0.0853
$\beta_{purpose,3}$	1.00	1.130	0.0725	0.943	0.0652
$\beta_{purpose,4}$	0.50	0.600	0.0642	0.501	0.0548
$\beta_{purpose,5}$	0.00	.	.	.	.
$\beta_{gender,1}$	0.80	0.726	0.135	0.780	0.123
$\beta_{gender,2}$	0.50	0.494	0.0916	0.597	0.0822
$\beta_{gender,3}$	0.20	0.226	0.0701	0.193	0.0614
$\beta_{gender,4}$	-0.10	-0.0517	0.0614	-0.107	0.0530
$\beta_{gender,5}$	0.00	.	.	.	.
μ_1	1.9	.	.	1.760	0.169
μ_2	2.10	.	.	2.120	0.146
μ_3	1.00	.	.	1.00	.

Table: True coefficient values for the dataset generation of MNL and NL



Substitution Patterns I

MNL

Fare class	1	2	3	4	5
Market share	5.10	10.50	45.30	31.00	8.10

- ▶ Market shares when all fare classes are available

Fare class	1	2	3	4	5
Market share	6.96	14.97	65.56	0.00	12.51

- ▶ Alternative 4 is no longer available
 - ▶ Market shares of 1, 2, 3 and 5 rise proportionally
 - ▶ Ratio of substitution between any pair of alternatives is constant
- **Example:** ratio of alternatives $\frac{1}{3} = \frac{5.10}{45.30} = \frac{6.96}{65.56} = 0.11$

Substitution Patterns II

NL

Fare class	1	2	3	4	5
Market share	2.93	6.67	40.41	28.29	21.70

► Nest 1 = {1,2}, Nest 2 = {3,4}, Nest 3 = {5}

Fare class	1	2	3	4	5
Market share	3.53	8.26	60.03	0.00	28.18

- Market share of alternative in the same nest (=3) rises proportionally
- Flexible substitution *across* nests, constant substitution *within* nests
- **Example:** ratio of alternatives $\frac{1}{3} = \frac{2.93}{40.41} = 0.073 \neq \frac{3.53}{60.03} = 0.059$
- **Example:** ratio of alternatives $\frac{1}{2} = \frac{2.93}{6.67} = \frac{3.53}{8.26} = 0.43$



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