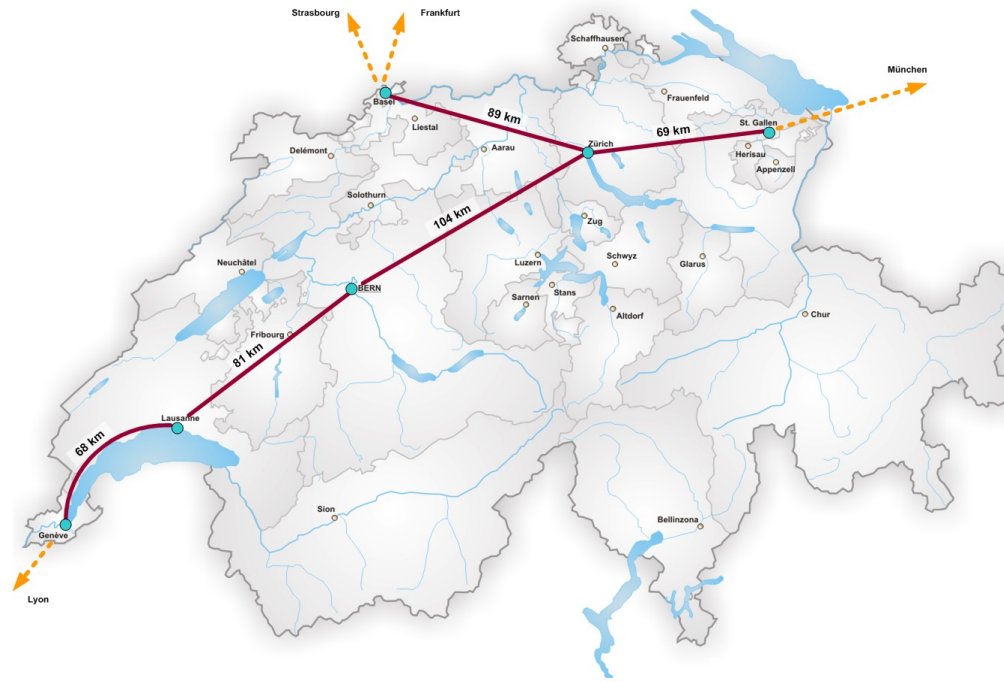


Swissmetro

- A MagLev underground system operating in partial vacuum to connect the major urban centers in Switzerland



MNL Model of Swissmetro Data

- SP data only – business and commuting trips
- Choice between **car**, **train**, and **Swissmetro (SM)**
- Considered variables:
 - Cost [sFr]
 - Travel time [min]
 - Headway [min] (Time between two vehicles at station for train and SM)
 - Annual Season Ticket (1: if individual has, 0: otherwise)
 - Senior (age > 65)

MNL Model of Swissmetro Data

- Estimation of Multinomial Logit Models (MNL)

- Start by considering **generic** attributes

Example code:

Ch05_Logit/Swissmetro/logit_SM_generic.ipynb

- Test **alternative-specific** attributes

Example code:

Ch05_Logit/Swissmetro/logit_SM_specific.ipynb

Note the change in B_COST changes

- Add and test **socio-economic characteristics**

Example code:

Ch05_Logit/Swissmetro/logit_SM_socioec.ipynb

Note the new coefficients

Model Specification

Variable	Description	Parameters to be Estimated		
		Car	Train	SM
	Constant	ASC_{Car}		ASC_{SM}
TT	Travel Time	β_{Time}	β_{Time}	β_{Time}
COST	Travel Cost	$\beta_{Cost-Car}$	$\beta_{Cost-Train}$	$\beta_{Cost-SM}$
HE	Headway		β_{HE}	β_{HE}
GA	Annual Season Ticket		β_{GA}	β_{GA}
SENIOR	Age > 65	β_{Age}		β_{Age}

Model Specification

$$V_{Car} = ASC_{Car} + \beta_{Time} TT_{Car} + \beta_{Cost-Car} COST_{Car} + \beta_{Age} SENIOR$$

$$V_{Train} = 0 + \beta_{Time} TT_{Train} + \beta_{Cost-Train} COST_{Train} + \beta_{HE} HE_{Train} + \beta_{GA} GA$$

$$V_{SM} = ASC_{SM} + \beta_{Time} TT_{SM} + \beta_{Cost-SM} COST_{SM} + \beta_{HE} HE_{SM} + \beta_{Age} SENIOR + \beta_{GA} GA$$

Model Specification

$$V_{Car} = ASC_{Car} + \beta_{Time} TT_{Car} + \beta_{Cost-Car} COST_{Car} + \beta_{Age} SENIOR$$

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$$V_{SM} = ASC_{SM} + \beta_{Time} TT_{SM} + \beta_{Cost-SM} COST_{SM} + \beta_{HE} HE_{SM} + \beta_{Age} SENIOR + \beta_{GA} GA$$

- Generic Time Coefficient

Model Specification

$$V_{Car} = ASC_{Car} + \beta_{Time} TT_{Car} + \beta_{Cost-Car} COST_{Car} + \beta_{Age} SENIOR$$

$$V_{Train} = 0 + \beta_{Time} TT_{Train} + \beta_{Cost-Train} COST_{Train} + \beta_{HE} HE_{Train} + \beta_{GA} GA$$

$$V_{SM} = ASC_{SM} + \beta_{Time} TT_{SM} + \beta_{Cost-SM} COST_{SM} + \beta_{HE} HE_{SM} + \beta_{Age} SENIOR + \beta_{GA} GA$$

- Generic Time Coefficient
- Alternative Specific Cost Coefficient

Estimation Results

- Initial log-likelihood:
-6958.42

- Final log-likelihood:
-4927.17

- $\bar{\rho}^2$: 0.291

- Number of
observations: 6759

Parameter	Estimate	Rob. t-stat
ASC_{Car}	-0.608	-4.24
ASC_{SM}	-0.135	-1.26
β_{Time}	-1.11	-9.19
$\beta_{Car-Cost}$	-0.936	-8.02
$\beta_{Train-Cost}$	-2.68	-15.2
$\beta_{SM-Cost}$	-1.04	-14.0
β_{HE}	-0.586	-5.55
β_{GA}	0.557	2.91
β_{Age}	-1.88	-17.3

Estimation Results

- Initial log-likelihood:
-6958.42

- Initial LL vs. $L(0)$

- Final log-likelihood:
-4927.17

- $\bar{\rho}^2$: 0.291

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Estimation Results

- What are the preferences over the three alternatives, all else being the same?
- What do the time and cost parameters show?
- How does frequency affect the train utility?
- Socio-economic characteristics?

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ASC_{SM}	-0.135	-1.26
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Estimation Results

- Train is preferred over Car and SM
- Time and cost coefficients are negative and significant
- Sensitivity to cost is the highest for train
- If rail-based modes are more frequent they have higher utility
- Having a season ticket increases the utility towards rail-based modes
- Seniors prefer train

Parameter	Estimate	Rob. t-stat
ASC_{Car}	-0.608	-4.24
ASC_{SM}	-0.135	-1.26
β_{Time}	-1.11	-9.19
$\beta_{Car-Cost}$	-0.936	-8.02
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Specification Testing: Generic vs. Alternative-Specific Test

- How can we justify the model specification with alternative specific cost coefficients as opposed to a generic specification?
- Likelihood ratio test
- **Restricted model:** Generic

$$V_{Car} = ASC_{Car} + \beta_{Time} TT_{Car} + \beta_{Cost} COST_{Car} + \beta_{Age} SENIOR$$

$$V_{Train} = 0 + \beta_{Time} TT_{Train} + \beta_{Cost} COST_{Train} + \beta_{HE} HE_{Train} + \beta_{GA} GA$$

$$V_{SM} = ASC_{SM} + \beta_{Time} TT_{SM} + \beta_{Cost} COST_{SM} + \beta_{HE} HE_{SM} + \beta_{Age} SENIOR + \beta_{GA} GA$$

- **Unrestricted model:** Alternative-Specific

$$V_{Car} = ASC_{Car} + \beta_{Time} TT_{Car} + \beta_{Cost-Car} COST_{Car} + \beta_{Age} SENIOR$$

$$V_{Train} = 0 + \beta_{Time} TT_{Train} + \beta_{Cost-Train} COST_{Train} + \beta_{HE} HE_{Train} + \beta_{GA} GA$$

$$V_{SM} = ASC_{SM} + \beta_{Time} TT_{SM} + \beta_{Cost-SM} COST_{SM} + \beta_{HE} HE_{SM} + \beta_{Age} SENIOR + \beta_{GA} GA$$

Specification Testing: Generic vs. Alternative-Specific Test

- $H_0: \beta_{Cost-Car} = \beta_{Cost-Train} = \beta_{Cost-SM}$
- Reject H_0 if $-2(L_R - L_U) > \chi^2_{((1-\alpha), df)}$

Restricted Model

```
Model: Multinomial Logit
Number of estimated parameters: 5
Number of observations: 6768
Number of individuals: 6768
Null log-likelihood: -6964.663
Init log-likelihood: -6964.663
Final log-likelihood: -5315.386
Likelihood ratio test: 3298.553
Rho-square: 0.237
Adjusted rho-square: 0.236
Final gradient norm: +2.161e-002
Diagnostic: Convergence reached...
Iterations: 69
Run time: 00:05
Variance-covariance: from analytical hessian
Sample file: swissmetro.dat
```

Unrestricted Model

```
Model: Multinomial Logit
Number of estimated parameters: 7
Number of observations: 6768
Number of individuals: 6768
Null log-likelihood: -6964.663
Init log-likelihood: -6964.663
Final log-likelihood: -5068.559
Likelihood ratio test: 3792.209
Rho-square: 0.272
Adjusted rho-square: 0.271
Final gradient norm: +1.262e-002
Diagnostic: Convergence reached...
Iterations: 81
Run time: 00:05
Variance-covariance: from analytical hessian
Sample file: swissmetro.dat
```

Specification Testing: Generic vs. Alternative-Specific Test

- $H_0: \beta_{Cost-Car} = \beta_{Cost-Train} = \beta_{Cost-SM}$
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Init log-likelihood: -6964.663
Final log-likelihood: -5068.559
Likelihood ratio test: 3792.209
Rho-square: 0.272
Adjusted rho-square: 0.271
Final gradient norm: +1.362e-002
Diagnostic: Convergence reached...
Iterations: 81
Run time: 00:05
Variance-covariance: from analytical hessian
Sample file: swissmetro.dat
```

- The degrees of freedom is $df = K_U - K_R$

Specification Testing: Generic vs. Alternative-Specific Test

- $H_0: \beta_{Cost-Car} = \beta_{Cost-Train} = \beta_{Cost-SM}$
- Reject H_0 if $-2(L_R - L_U) > \chi^2_{((1-\alpha), df)}$

Restricted Model

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Diagnostic: Convergence reached...
Iterations: 81
Run time: 00:05
Variance-covariance: from analytical hessian
Sample file: swissmetro.dat
```

- The degrees of freedom is $df = K_U - K_R$
- A typical threshold is $\alpha = 0.05$

Specification Testing: Generic vs. Alternative-Specific Test

- $H_0: \beta_{Cost-Car} = \beta_{Cost-Train} = \beta_{Cost-SM}$
- Reject H_0 if $-2(L_R - L_U) > \chi^2_{((1-\alpha), df)}$
 $-2(L_R - L_U) = -2(-5315.386 + 5068.559) = 493.654$
 $\chi^2_{((1-0.05), 2)} = 5.991$
- We reject H_0 . There exists sufficient statistical evidence that our model should include alternative specific cost coefficients

Specification Testing: Market Segmentation Test

- Is there variation across different market segments?
 - E.g. education levels, income categories, etc.,
- Estimate model on the full data set. Then estimate the same model on each group separately

Example code:

```
Ch06_SpecTesting/Swissmetro/  
    SpecTest_SM_segmentation.ipynb
```

Note how the database is duplicated and restricted into Male and Female observations

Specification Testing: Market Segmentation Test

- $H_0: \beta_{Male} = \beta_{Female}$
- Reject H_0 if $-2(L_R - L_U) > \chi^2_{((1-\alpha), df)}$

Model	Log likelihood	Number of coefficients
Male	-3680.002	9
Female	-1110.618	9
Restricted model	-4927.167	9

$$-2(L_R - L_U) = -2(-4927.167 + 3680.002 + 1110.618) \\ = 273.094$$

$$\chi^2_{((1-0.05), 9)} = 16.919$$

- The log likelihood of the unrestricted model is the sum of the log likelihoods of the models estimated on each of the 2 segments