

# 7<sup>th</sup> Discrete Choice Modelling Workshop, EPFL 2011

## Modelling regret effects in route-choice with real-time and feedback information

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# Outline

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# Background

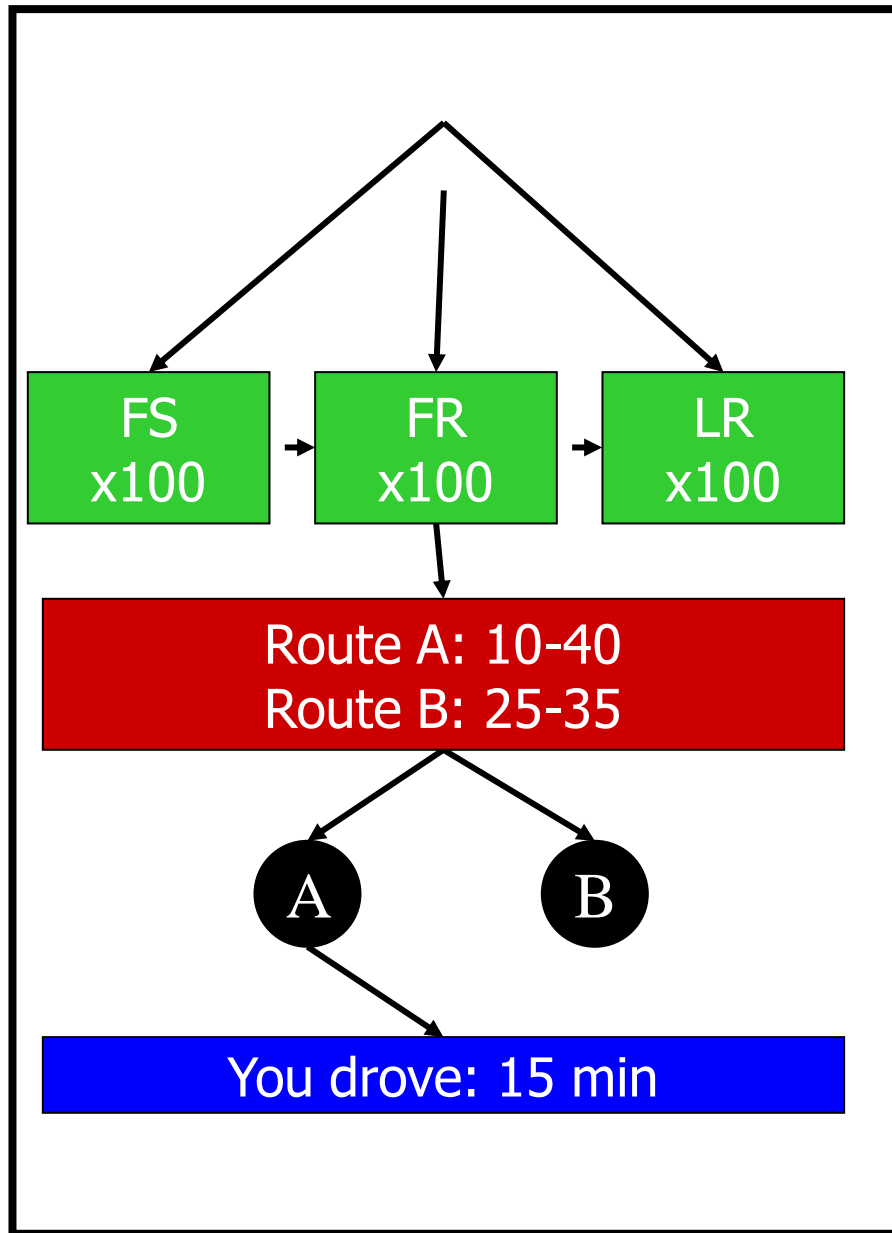
- Like EUT and Prospect Theory, Regret Theory (RT) is a model of human decision making under uncertainty.
- The three theories explain situations where choices are based on information providing a description of the alternatives.
- RT postulates choices are influenced not only by the attractiveness of a considered alternative as EUT, but also by the regret associated with not choosing a foregone one [i.e. Regret Aversion].
- But..in order to compare 'what is' with 'what would have been..', the DM needs to learn from experience what the foregone alternative implies.
- The 'trigger' for regret is not that obvious.

# Experiment setup

- ❑ Choice between a faster and a slower route (5 min. mean dif.)
- ❑ 24 participants
- ❑ Panel: 100 repeated choice-trials in 3 scenarios.

Scen.	Description	Range FAST [mean $\pm$ min]	Range SLOW [mean $\pm$ min]
<i>Fast and Safe</i>	Low var. on FAST	25 $\pm$ 5	30 $\pm$ 15
<i>Fast and Risky</i>	High var. on FAST	25 $\pm$ 15	30 $\pm$ 5
<i>Low Risk</i>	Low var. on both	25 $\pm$ 5	30 $\pm$ 5

- ❑ **Information:** in each trial 2 sources are always available:
  1. Descriptive: Travel time range expected on each route
  2. Experiential: Feedback on actual travel times of chosen route



Data was not designed with the objective of testing RT.

If regret is a significant effect, this is a strong indication to the relevance of regret in similar experienced-based route-choice decisions

# Modelling approach

1. EU - expected utility
2. EMU - expected modified utility

utility ( $U$ ) of alternative  $i$  for person  $m$  in response  $t$  is:

$$U_{imt} = \alpha_{im} + \beta_{im} X_{imt}$$

$\beta$  - fixed coefficients for alternatives' attributes -  $X$ ;

$\alpha$  - random coefficients  $\alpha \sim N(0, \sigma_\alpha^2)$

$$EU_{imt} = \left( \sum_{j=1, j \in S}^J p_{jt} U_{imjt} \right) + \varepsilon_{imt}$$

$p_j [0,1]$  is the probability that state-of-the world  $j$  will occur at response  $t$  out of the set of  $J$  possible states of the world -  $S$

# Modelling cont.

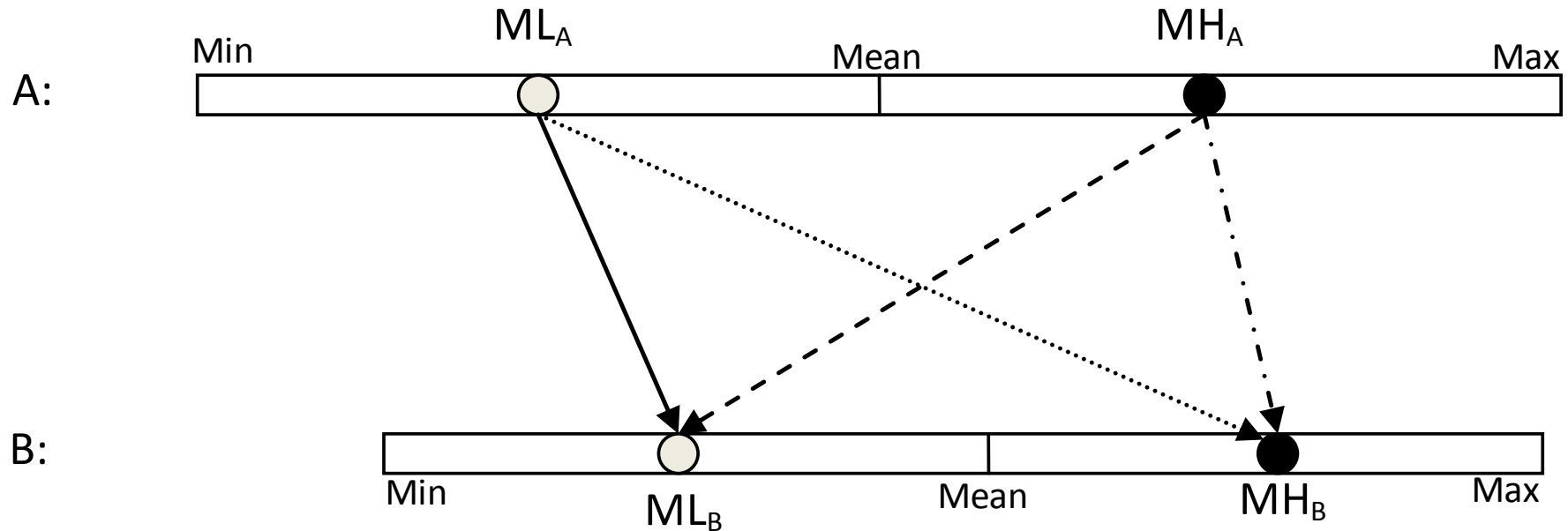
*Modified Utility (MU)* depends on both the considered ( $i$ ) and foregone ( $k$ ) alternatives. Following Chorus (2010), the modified utility ( $MU$ ) is:

$$MU_{ikmt} = \alpha_{im} + \beta_{im} X_{imt} + \{1 - e^{[-\rho(\beta_{im} X_{imt} - \beta_{im} X_{kmt})]}\}$$

$\rho \in [0, +\infty]$  is a regret aversion parameter. Higher values imply that person  $m$  will become more and more sensitive to regret

$$EMU_{ikmt} = \left( \sum_{j=1, j \in S}^J p_{jt} MU_{ikmt} \right) + \varepsilon_{imt}$$

# Specification



Assume that the DM regards two points on the TT range as being identified with the possible states of the world - one below (i.e. the first quarter) and the other above (i.e. the 3<sup>rd</sup> quarter) the mean value



# Specification cont.

- Four models are specified.

## 1. Simple EU model (control)

$$EU_A = \alpha_A + 0.5(\beta MH_A + \beta ML_A) + \varepsilon$$

## 2. Description-based RT model

$$EMU_A = \alpha_A + 0.25(\beta MH_A + 1 - e^{[-\rho(\beta MH_A - \beta MH_B)]}) + 0.25(\beta MH_A + 1 - e^{[-\rho(\beta MH_A - \beta ML_B)]}) \\ + 0.25(\beta ML_A + 1 - e^{[-\rho(\beta ML_A - \beta MH_B)]}) + 0.25(\beta ML_A + 1 - e^{[-\rho(\beta ML_A - \beta ML_B)]}) + \varepsilon$$

## 3. Description and experienced-based RT model

$$EMU_A = \alpha_A + 0.25(\beta MH_A + 1 - e^{\{-\rho[w(\beta MH_A - \beta MH_B) + (1-w)(\beta F_A - \beta F_B)]\}}) \\ + 0.25(\beta MH_A + 1 - e^{\{-\rho[w(\beta MH_A - \beta ML_B) + (1-w)(\beta F_A - \beta F_B)]\}}) \\ + 0.25(\beta ML_A + 1 - e^{\{-\rho[w(\beta ML_A - \beta MH_B) + (1-w)(\beta F_A - \beta F_B)]\}}) \\ + 0.25(\beta ML_A + 1 - e^{\{-\rho[w(\beta ML_A - \beta ML_B) + (1-w)(\beta F_A - \beta F_B)]\}}) + \varepsilon$$

$0 < w < 1$  is a weight attributed to the descriptive information ( $MH_i$ ,  $ML_i$ );  $(1-w)$  is the weight for feedbacks

$F_i$  is the feedback received for Route  $i$  the last time  $i$  is chosen

# Specification cont.

$w=1$  means only descriptive information affects regret (the same as Model II).

$w=0$ , means only feedbacks affect regret. The information provided ex-ante is not responsible for generating regret.

$w$  is estimated exogenously (trial and error)

- 4. Effect of risk on regret :** Regret coefficients specified for each scenario ( $s$ ).

$$\begin{aligned}
 EMU_A = \alpha_A + \sum_{s=1}^3 & \left[ 0.25(\beta MH_{As} + 1 - e^{\{-\rho_s[w(\beta MH_{As} - \beta MH_{Bs}) + (1-w)(\beta F_{As} - \beta F_{Bs})]\}}) \right. \\
 & + 0.25(\beta MH_{As} + 1 - e^{\{-\rho_s[w(\beta MH_{As} - \beta ML_{Bs}) + (1-w)(\beta F_{As} - \beta F_{Bs})]\}}) \\
 & + 0.25(\beta ML_{As} + 1 - e^{\{-\rho_s[w(\beta ML_{As} - \beta MH_{Bs}) + (1-w)(\beta F_{As} - \beta F_{Bs})]\}}) \\
 & \left. + 0.25(\beta ML_{As} + 1 - e^{\{-\rho_s[w(\beta ML_{As} - \beta ML_{Bs}) + (1-w)(\beta F_{As} - \beta F_{Bs})]\}}) \right] + \varepsilon
 \end{aligned}$$

# Estimation

- Biogeme 2.0
- Mixed logit model with non linear utilities
- Log likelihood maximization:

$$LL(\beta, \rho, \sigma) = \sum_{m=1}^M \log(P_{mi})$$

$$= \sum_{m=1}^M \log \left\{ \int_{\alpha} \left[ \prod_{t=1}^T \left( \frac{e^{EMU_{it}}}{\sum_{k=1, i \in K}^K e^{EMU_{kt}}} \right) \right] d\alpha \right\}$$

$M=24$  participants,  $T=300$  trials,  $K=2$  alternative routes

- Simulated LL using 1000 Halton draws.

# Results

No	Coef.	Est.	Std err*	t-test	p-value
1	$\beta$	-0.545	0.055	-9.83	< 0.001
	$\sigma_\alpha$	1.28	0.18	7.14	< 0.001
	$LL_0$	-4940.8			
	$LL_\beta$	-2086.3			
2	$\beta$	-1.24	0.148	-8.39	< 0.001
	$\sigma_\alpha$	1.32	0.181	7.26	< 0.001
	$\rho$	-0.134	0.0073	-18.46	< 0.001
	$LL_0$	-4940.8			
	$LL_\beta$	-1969.2			
3	$\beta$	-0.471	0.055	-8.58	< 0.001
	$\sigma_\alpha$	1.23	0.174	7.05	< 0.001
	$\rho$	0.0777	0.019	4.2	< 0.001
	w	0			
	$LL_0$	-4940.8			
	$LL_\beta$	-1985.6			

No	Coef.	Est.	Std err*	t-test	p-value
4	$\beta$	-0.45	0.0575	-7.82	< 0.001
	$\sigma_\alpha$	1.27	0.173	7.31	< 0.001
	$\rho_1$	0.0359	0.0254	1.41	0.160
	$\rho_2$	0.0913	0.0208	4.39	< 0.001
	$\rho_3$	0.313	0.064	4.89	< 0.001
	w	0			
	$LL_0$	-4940.8			
	$LL_\beta$	-1869.4			

# Conclusions

- Effect of regret do occur in the observed data.
- Regret is associated more with experiential feedback than with the descriptive information regarding the expected travel time ranges.
- Accounting for effects of risk Regret is more apparent in situations involving less risk, whereas riskier situation seem to inhibit regret.
- More research in understanding the relations between Regret, learning and risk attitudes.

Thanks,  
Merci.



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