

## Comparing and cross-validating of results from LatentGOLD Choice

It is possible to estimate  $\mu$ RRM models in LatentGOLD (LG) Choice 5.1 . Its implementation is however somewhat different from the implementation described in Cranenburgh et al. 2015 (henceforth referred to as the standard implementation).

Equation 1 gives the implementation of the  $\mu$ RRM model in LG Choice 5.1 (see Vermunt and Magidson (2014)); equation 2 gives the standard implementation.

$R_i = \sum_{j \neq i} \sum_m \left( 1 + \exp \left( \beta_m^* [x_{jm} - x_{im}] \right) \right) \quad P_i = \frac{e^{-e^{\mu^*} R_i}}{\sum_J e^{-e^{\mu^*} R_j}}$	(1)
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$R_i = \sum_{j \neq i} \sum_m \left( 1 + \exp \left( \frac{\beta_m}{\mu} [x_{jm} - x_{im}] \right) \right) \quad P_i = \frac{e^{-\mu R_i}}{\sum_J e^{-\mu R_j}}$	(2)
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Hence, the implementation LG Choice 5.1 differs from the standard implementation in two important ways:

1. the taste parameter  $\beta_m^*$  is not divided by the scale parameter  $\mu$ .
2. the log of the scale is estimated, instead of the scale.

As a consequence, we cannot directly compare the results obtained using LG Choice 5.1 with the results obtained using the standard implementations. Below we show how to compare and cross-validate results obtained using LG Choice 5.1 with those obtained using the standard implementation.

### Point estimates

To compare and cross-validate the point estimates of LG Choice 5.1 we can simply multiply the taste parameter estimates with the  $e^{\mu^*}$  (the scale), see equation 3, where \* denotes the LG parameter estimate.

$\begin{bmatrix} \hat{\beta}_1 \\ \hat{\beta}_2 \\ \vdots \\ \hat{\beta}_M \\ \hat{\mu}_1 \end{bmatrix} = \begin{bmatrix} e^{\mu^*} \beta_1^* \\ e^{\mu^*} \beta_2^* \\ \vdots \\ e^{\mu^*} \beta_M^* \\ e^{\mu^*} \end{bmatrix}$	(3)
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Standard errors

To attain the standard error we apply the delta method (equation 4).

$\text{cov}(\Phi) = \Phi'^T \Omega \Phi'$	(4)
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where  $\Omega$  denotes the AVC matrix of the LG parameter,  $\Phi$  the vector of standard parameters, and  $\Phi'$  the matrix of first derivatives of the standard parameters towards the LG parameters. Given the implementation of the  $\mu$ RRM model in LG Choice we get:

$$\Phi = \left[ \exp(\mu^*) \beta_1^* \quad \exp(\mu^*) \beta_2^* \quad \cdots \quad \exp(\mu^*) \beta_M^* \quad \exp(\mu^*) \right] \Rightarrow$$

$$\Phi' = \begin{bmatrix} e^{\mu^*} & 0 & \cdots & 0 & 0 \\ 0 & e^{\mu^*} & \cdots & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \cdots & e^{\mu^*} & 0 \\ e^{\mu^*} \beta_1^* & e^{\mu^*} \beta_2^* & \cdots & e^{\mu^*} \beta_M^* & e^{\mu^*} \end{bmatrix} \quad \Phi'^T = \begin{bmatrix} e^{\mu^*} & 0 & \cdots & 0 & e^{\mu^*} \beta_1^* \\ 0 & e^{\mu^*} & \cdots & 0 & e^{\mu^*} \beta_2^* \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \cdots & e^{\mu^*} & e^{\mu^*} \beta_M^* \\ 0 & 0 & \cdots & 0 & e^{\mu^*} \end{bmatrix}$$

The standard errors are the square root of the diagonal element of the covariance matrix:

$$\begin{bmatrix} SE_{\beta_1} \\ SE_{\beta_2} \\ \vdots \\ SE_{\beta_M} \\ SE_{\mu} \end{bmatrix} = \begin{bmatrix} \sqrt{\text{COV}_{11}} \\ \sqrt{\text{COV}_{22}} \\ \vdots \\ \sqrt{\text{COV}_{MM}} \\ \sqrt{\text{COV}_{M+1,M+1}} \end{bmatrix}$$

## Application to the shopping choice data:

The 1-class model yields the following parameter estimates, and covariance matrix  $\Omega$ .

	<i>Est</i>	<i>SE</i>	
<i>FSG</i> *	0.9405	0.1996	$\Omega = \begin{bmatrix} 0.0399 & 0.0004 & -0.0028 & -0.0370 \\ 0.0004 & 0.0001 & -0.0001 & -0.0008 \\ -0.0028 & -0.0001 & 0.0004 & 0.0031 \\ -0.0370 & -0.0008 & 0.0031 & 0.0410 \end{bmatrix}$
<i>FSO</i> *	-0.0096	0.0087	
<i>TT</i> *	-0.0865	0.0195	
$\mu^*$	-1.9711	0.2024	

To attain the point estimates we apply equation 3. As expected, these point estimates correspond exactly with those obtained using the standard implementation, see Cranenburgh et al. 2015.

	<i>Est</i>
<i>FSG</i> *	0.1310
<i>FSO</i> *	0.0013
<i>TT</i> *	-0.0120
$\mu^*$	0.1393

To attain the standard errors we apply equation 4 to get the covariance matrix

$$\text{cov} = \begin{bmatrix} 0.13931 & 0 & 0 & 0.13102 \\ 0 & 0.13931 & 0 & 0.00134 \\ 0 & 0 & 0.13931 & -0.01205 \\ 0 & 0 & 0 & 0.13931 \end{bmatrix} \cdot \begin{bmatrix} 0.0399 & 0.0004 & -0.0028 & -0.0370 \\ 0.0004 & 0.0001 & -0.0001 & -0.0008 \\ -0.0028 & -0.0001 & 0.0004 & 0.0031 \\ -0.0370 & -0.0008 & 0.0031 & 0.0410 \end{bmatrix}$$

$$\text{cov} = \begin{bmatrix} 0.13931 & 0 & 0 & 0 \\ 0 & 0.13931 & 0 & 0 \\ 0 & 0 & 0.13931 & 0 \\ 0.13102 & 0.00134 & -0.01205 & 0.13931 \end{bmatrix}$$

$$\text{cov} = \begin{bmatrix} 0.00012749 & -0.00000655 & -0.00000038 & 0.00003028 \\ -0.00000655 & 0.00000172 & -0.00000068 & -0.00000787 \\ -0.00000038 & -0.00000068 & 0.00000331 & -0.00000866 \\ 0.00003028 & -0.00000787 & -0.00000866 & 0.00079570 \end{bmatrix}$$

Finally, the standard errors are obtained by taken the square root of the diagonal elements of the covariance matrix. As expected these results correspond exactly<sup>1</sup> with results in Cranenburgh et al. 2015.

	<i>SE</i>	<i>t - stat</i>
<i>FSG</i> *	0.01123	11.6647
<i>FSO</i> *	0.00111	1.2075
<i>TT</i> *	0.00174	-6.9384
$\mu^*$	0.02820	4.9397

<sup>1</sup> In Van Cranenburgh et al. (2015) there is a minor error in Table 5.1 pg 101. The reported t-statistic for the difference from one for the scale parameter equals 30.51, instead of 87.83.

- Van Cranenburgh, S., Guevara, C. A. & Chorus, C. G. (2015). New insights on random regret minimization models. *Transportation Research Part A: Policy and Practice*, 74(0), 91-109.
- Vermunt, J. K. & Magidson, J. (2014). *Upgrade manual for Latent GOLD Choice 5.0*. (Belmont, MA , USA).