

# Parameter Identification in Discrete-choice Models

## TheoryGuru applications

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### Background

Here we consider a discrete choice that depends on observed factors  $x$  and  $y$  as well as unobserved factors summarized by  $\sigma \epsilon$ , where  $\epsilon$  is normalized to have unit variance and distribution function  $F$ . The discrete choice is made if and only if:

$$\sigma \epsilon < \beta x + \gamma y$$

The log likelihood function for observing  $\{x_1, y_1, d_1, x_2, y_2, d_2, \dots, x_n, y_n, d_n\}$  is:

$$L\left[\left\{\frac{\beta}{\sigma}, \frac{\gamma}{\sigma}\right\}\right] = \sum_{i|d_i=0} \text{Log}\left[1 - F\left[\frac{\beta}{\sigma}x_i + \frac{\gamma}{\sigma}y_i\right]\right] + \sum_{i|d_i=1} \text{Log}\left[F\left[\frac{\beta}{\sigma}x_i + \frac{\gamma}{\sigma}y_i\right]\right]$$

### Setup

```
Get["http://economicreasoning.com"]
```

#### Proof & Logic Tools 6.2

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$$\begin{aligned} \text{LikelihoodisStrictlyConcave} &= \left( \text{Not@SameVector} \left[ \left\{ \left\{ \frac{\beta_1}{\sigma_1}, \frac{\gamma_1}{\sigma_1} \right\}, \left\{ \frac{\beta_2}{\sigma_2}, \frac{\gamma_2}{\sigma_2} \right\} \right\} \wedge 0 < \lambda < 1 \right] \Rightarrow \right. \\ &\quad \left. (1 - \lambda) L \left[ \left\{ \frac{\beta_1}{\sigma_1}, \frac{\gamma_1}{\sigma_1} \right\} \right] + \lambda L \left[ \left\{ \frac{\beta_2}{\sigma_2}, \frac{\gamma_2}{\sigma_2} \right\} \right] < L \left[ (1 - \lambda) \left\{ \frac{\beta_1}{\sigma_1}, \frac{\gamma_1}{\sigma_1} \right\} + \lambda \left\{ \frac{\beta_2}{\sigma_2}, \frac{\gamma_2}{\sigma_2} \right\} \right]; \right. \\ &\quad \left. (* \text{ for any } \left\{ \frac{\beta_1}{\sigma_1}, \frac{\gamma_1}{\sigma_1} \right\} \text{ and } \left\{ \frac{\beta_2}{\sigma_2}, \frac{\gamma_2}{\sigma_2} \right\} \text{ and } \lambda *) \right. \\ \text{Vector1MaxesLikelihood} &= L \left[ \left\{ \frac{\beta_1}{\sigma_1}, \frac{\gamma_1}{\sigma_1} \right\} \right] \geq L \left[ (1 - \lambda) \left\{ \frac{\beta_1}{\sigma_1}, \frac{\gamma_1}{\sigma_1} \right\} + \lambda \left\{ \frac{\beta_2}{\sigma_2}, \frac{\gamma_2}{\sigma_2} \right\} \right]; \\ \text{Vector2MaxesLikelihood} &= L \left[ \left\{ \frac{\beta_2}{\sigma_2}, \frac{\gamma_2}{\sigma_2} \right\} \right] \geq L \left[ (1 - \lambda) \left\{ \frac{\beta_1}{\sigma_1}, \frac{\gamma_1}{\sigma_1} \right\} + \lambda \left\{ \frac{\beta_2}{\sigma_2}, \frac{\gamma_2}{\sigma_2} \right\} \right]; \end{aligned}$$

## Results

### Coefficients not identified

$$\begin{aligned} \text{TheoryGuru} &\left[ \{0 < \lambda < 1, \right. \\ &\quad \text{LikelihoodisStrictlyConcave, Vector1MaxesLikelihood, Vector2MaxesLikelihood,} \\ &\quad \left. \text{SameVector} \left[ \left\{ \left\{ \frac{\beta_1}{\sigma_1}, \frac{\gamma_1}{\sigma_1} \right\}, \left\{ \frac{\beta_2}{\sigma_2}, \frac{\gamma_2}{\sigma_2} \right\} \right\} \Rightarrow \left( \frac{\beta_1}{\sigma_1} = \frac{\beta_2}{\sigma_2} \wedge \frac{\gamma_1}{\sigma_1} = \frac{\gamma_2}{\sigma_2} \right) \right], \right. \end{aligned}$$

$$\beta_1 = \beta_2 \vee \gamma_1 = \gamma_2]$$

True for some, False for others

$$\begin{aligned} \text{TheoryInstance} &\left[ \{0 < \lambda < 1, \right. \\ &\quad \text{LikelihoodisStrictlyConcave, Vector1MaxesLikelihood, Vector2MaxesLikelihood,} \\ &\quad \left. \text{SameVector} \left[ \left\{ \left\{ \frac{\beta_1}{\sigma_1}, \frac{\gamma_1}{\sigma_1} \right\}, \left\{ \frac{\beta_2}{\sigma_2}, \frac{\gamma_2}{\sigma_2} \right\} \right\} \Rightarrow \left( \frac{\beta_1}{\sigma_1} = \frac{\beta_2}{\sigma_2} \wedge \frac{\gamma_1}{\sigma_1} = \frac{\gamma_2}{\sigma_2} \right), \right. \right. \\ &\quad \left. \left. \sigma_1 > 0, \sigma_2 > 0 \right\}, \right. \end{aligned}$$

$$\text{Not}[\beta_1 = \beta_2 \vee \gamma_1 = \gamma_2]$$

SameVector[ $\left\{ \frac{\beta_1}{\sigma_1}, \frac{\gamma_1}{\sigma_1} \right\},$ $\left\{ \frac{\beta_2}{\sigma_2}, \frac{\gamma_2}{\sigma_2} \right\}$ ]	$\lambda$	$L \left[ \left\{ \frac{\beta_1}{\sigma_1}, \frac{\gamma_1}{\sigma_1} \right\} \right]$	$L \left[ \left\{ \left( (1 - \lambda) \frac{\beta_1}{\sigma_1} + \frac{\lambda \beta_2}{\sigma_2}, \right. \right. \right.$ $\left. \left. \left( (1 - \lambda) \frac{\gamma_1}{\sigma_1} + \frac{\lambda \gamma_2}{\sigma_2} \right) \right\} \right]$	$L \left[ \left\{ \frac{\beta_2}{\sigma_2}, \frac{\gamma_2}{\sigma_2} \right\} \right]$	$\beta_1$	$\beta_2$	$\gamma_1$	$\gamma_2$	$\sigma_1$	$\sigma_2$
True	$\frac{1}{2}$	0	-1	0	-1	$-\frac{1}{2}$	-2	-1	1	$\frac{1}{2}$

## Coefficients identified up to scale

```
TheoryGuru[{0 < λ < 1,
  LikelihoodisStrictlyConcave, Vector1MaxesLikelihood, Vector2MaxesLikelihood},
```

```
  SameVector[{ $\frac{\beta_1}{\sigma_1}, \frac{\gamma_1}{\sigma_1}$ }, { $\frac{\beta_2}{\sigma_2}, \frac{\gamma_2}{\sigma_2}$ }]]]
```

```
True
```

```
TheoryGuru[{0 < λ < 1,
  LikelihoodisStrictlyConcave, Vector1MaxesLikelihood, Vector2MaxesLikelihood,
  SameVector[{ $\frac{\beta_1}{\sigma_1}, \frac{\gamma_1}{\sigma_1}$ }, { $\frac{\beta_2}{\sigma_2}, \frac{\gamma_2}{\sigma_2}$ }]} ⇒ ( $\frac{\beta_1}{\sigma_1} = \frac{\beta_2}{\sigma_2} \wedge \frac{\gamma_1}{\sigma_1} = \frac{\gamma_2}{\sigma_2}$ )},
```

```
 $\frac{\beta_1}{\gamma_1} = \frac{\beta_2}{\gamma_2}$ ]
```

```
True
```

## Variable interpretations