

# Predictive Analytics

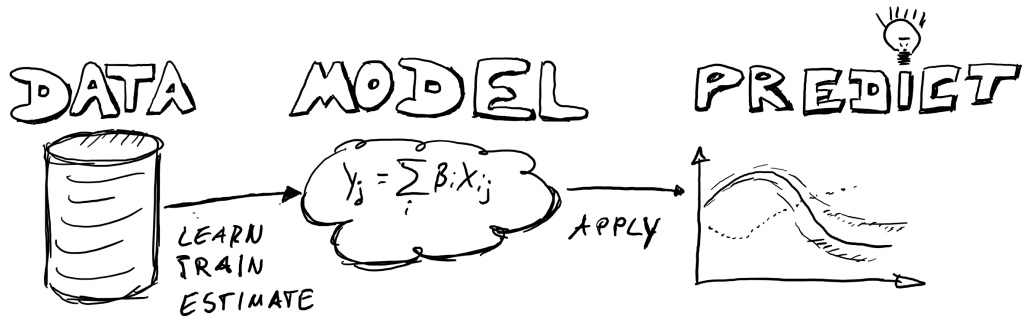
## Exercise 3: Binary Choice

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# Our Overall Objective



## To achieve this objective, we ...

- 1 imply a (theory-driven) **model** structure: select and specify variables  $X$ , i.e., deterministic utility  $V$ . For instance
  - $V_{ni} = \beta_1 TT_i$  or,
  - $V_{ni} = \beta_1 TT_i + \beta_2 TC_i$
- 2 **estimate** model coefficients  $\hat{\beta}$ :
  - maximum likelihood procedure: select values of  $\hat{\beta}$  such that the assumed model structure fits (replicates) the choices  $y_{ni}$  in the **data** as good as possible
  - roughly sayin': highest choice probability  $P_{ni}$  for the actual chosen alternative ( $y_{ni} = 1$ )
- 3 compute deterministic utility values  $\hat{\beta}X_{ni}$  and choice probabilities:

$$P_{ni} = \frac{e^{\hat{\beta}X_{ni}}}{\sum_j e^{\hat{\beta}X_{nj}}}$$

- 4 consider different scenarios (values) of  $X$  and **predict** corresponding  $P_{ni}$

# Mode Choice Example

- ▶ Estimate a series of models along the following slides
- ▶ Try to implement the different specifications
  
- ▶ Get data and R script from moodle

# Specifications

## ► Baseline specification

$$\begin{aligned}U_{n,\text{rail}} &= \beta_0 + \beta_1 \text{TC}_{n,\text{rail}} + \epsilon_{n,\text{rail}} \\U_{n,\text{car}} &= \beta_1 \text{TC}_{n,\text{car}} + \epsilon_{n,\text{car}}\end{aligned}$$

## ► Specification 1

$$\begin{aligned}U_{n,\text{rail}} &= \beta_0 + \beta_1 \text{TC}_{n,\text{rail}} + \beta_2 \text{TT}_{n,\text{rail}} + \epsilon_{n,\text{rail}} \\U_{n,\text{car}} &= \beta_1 \text{TC}_{n,\text{car}} + \beta_2 \text{TT}_{n,\text{car}} + \epsilon_{n,\text{car}}\end{aligned}$$

## ► Specification 2

$$\begin{aligned}U_{n,\text{rail}} &= \beta_0 + \beta_1 \text{TC}_{n,\text{rail}} + \beta_2 \text{TT}_{n,\text{rail}} + \epsilon_{n,\text{rail}} \\U_{n,\text{car}} &= \beta_0 + \beta_1 \text{TC}_{n,\text{car}} + \beta_2 \text{TT}_{n,\text{car}} + \epsilon_{n,\text{car}}\end{aligned}$$

# Specifications

## ► Specification 3

$$\begin{aligned}U_{n,\text{rail}} &= \beta_0 + \beta_1 \text{TC}_{n,\text{rail}} + \beta_2 \text{TT}_{n,\text{rail}} + \epsilon_{n,\text{rail}} \\U_{n,\text{car}} &= \beta_1 \text{TC}_{n,\text{car}} + \beta_3 \text{TT}_{n,\text{car}} + \epsilon_{n,\text{car}}\end{aligned}$$

## ► Specification 4

$$\begin{aligned}U_{n,\text{rail}} &= \beta_0 + \beta_4 \text{gender}_n + \beta_5 \text{business}_n + \beta_1 \text{TC}_{n,\text{rail}} + \beta_2 \text{TT}_{n,\text{rail}} + \epsilon_{n,\text{rail}} \\U_{n,\text{car}} &= \beta_1 \text{TC}_{n,\text{car}} + \beta_3 \text{TT}_{n,\text{car}} + \epsilon_{n,\text{car}}\end{aligned}$$

# Challenge Questions

- 1 Try to implement the following specification 5

$$\begin{aligned}U_{n,\text{rail}} &= \beta_0 + \beta_4 \text{gender}_n + \beta_5 \text{business}_n + \beta_6 \text{income}_n + \beta_1 \text{TC}_{n,\text{rail}} + \beta_2 \text{TT}_{n,\text{rail}} + \epsilon_{n,\text{rail}} \\U_{n,\text{car}} &= \beta_7 \text{income}_n + \beta_1 \text{TC}_{n,\text{car}} + \beta_3 \text{TT}_{n,\text{car}} + \epsilon_{n,\text{car}}\end{aligned}$$

- 2 Specify and estimate a model based on specification 5 that also considers whether wifi is available (see  $\text{service}_{\text{rail}}$  in apollo manual)
- 3 How does your model improve? Interpret the meaning of the coefficient!