

# Mathematical Modeling of Behavior

Michel Bierlaire

Transport and Mobility Laboratory  
School of Architecture, Civil and Environmental Engineering  
Ecole Polytechnique Fédérale de Lausanne



# Outline

- 1 Motivation
  - In this course
  - Applications
  - Importance
- 2 Simple example
  - Choice problem
  - Data
  - Model specification
  - Probabilities
  - Model
  - Estimation
  - Testing
  - Maximum likelihood
  - Hypothesis testing
  - Application

# Motivation

## Human dimension in

- engineering
- business
- marketing
- planning
- policy making

## Need for

- behavioral *theories*
- quantitative *methods*
- operational mathematical *models*

# Motivation

## Concept of demand

- marketing
- transportation
- energy
- finance

## Concept of choice

- brand, product
- mode, destination
- type, usage
- buy/sell, product

# In this course...

## Focus

- Individual behavior (vs. aggregate behavior)
- Theory of behavior which is
  - **descriptive** (how people behave) and not normative (how they should behave)
  - **general**: not too specific
  - **operational**: can be used in practice for forecasting
- Type of behavior: **choice**

# Applications

## Mode choice in the Netherlands

- Context: car vs rail in Nijmegen
- Objective: sensitivity to travel time and cost, inertia.

## Mode choice in Switzerland

- Context: Car Postal
- Objective: demand forecasting

# Applications

## Swissmetro

- Context: new transportation technology
- Objective: demand pattern, pricing

## Residential telephone services

- Context: flat rate vs. measured
- Objective: offer the most appropriate service

## Airline itinerary choice

- Context: questionnaire about itineraries across the US
- Objective: help airlines and aircraft manufacturer to design a better offer

# Importance



## Daniel L. McFadden

- UC Berkeley 1963, MIT 1977, UC Berkeley 1991
- Laureate of *The Bank of Sweden Prize in Economic Sciences in Memory of Alfred Nobel 2000*
- Owns a farm and vineyard in Napa Valley
- “Farm work clears the mind, and the vineyard is a great place to prove theorems”



# Outline

- 1 Motivation
  - In this course
  - Applications
  - Importance
- 2 Simple example
  - Choice problem
  - Data
  - Model specification
  - Probabilities
  - Model
  - Estimation
  - Testing
  - Maximum likelihood
  - Hypothesis testing
  - Application

# Simple example

## Objectives

Introduce basic components of choice modeling:

- definition of the problem
- data
- model specification
- parameter estimation
- model application

## Application

Analysis of the market for smartphones

# Choice problem

## Choice

Consumer's choice to

- own a smartphone
- own another ("non-smart") mobile phone.

## Questions

- what is the current market penetration of smartphones relative to non-smart phones?
- how will the penetration change in the future?

# Data

## Population

- adults
- in the US
- owning a mobile phone

## Sample

- 2000 adults
- randomly selected

## Questions

Is your mobile phone a smartphone

- Yes,
- No.

What is your level of educational attainment?

- No high school diploma,
- High school graduate,
- College graduate.

# Data

## Contingency table

Smartphone	Education			
	Low ( $k = 1$ )	Medium ( $k = 2$ )	High ( $k = 3$ )	
Yes ( $i = 1$ )	75	500	510	1085
No ( $i = 2$ )	175	500	240	915
	250	1000	750	2000

## Market penetration in the sample

- $1085/2000 = 54.3\%$
- How do we predict? We need a model.

# Model specification

## Variables

### Dependent

- or endogenous
- what is explained
- here: choice to use a smartphone
- notation:  $i$
- nature: discrete
- 1 = “yes”; 2 = “no”

### Independent

- or exogenous
- explanatory
- here: level of education
- notation:  $k$
- nature: discrete
- 1 = “low”; 2 = “medium”; 3 = “high”

# Probabilities

## Marginal probability

- frequency of smartphone ownership in the population
- $P(i = 1)$
- Inference: use the sample to obtain an estimate
- $P(i = 1) \approx \hat{P}(i = 1) = 1085/2000 = 0.543$

## Joint probability

- frequency of smartphone ownership and medium level of education
- $P(i = 1, k = 2) \approx \hat{P}(i = 1, k = 2) = 500/2000 = 0.25$

## Conditional probability

- frequency of smartphone ownership in the population of people with medium level of education
- $P(i = 1|k = 2) \approx \hat{P}(i = 1|k = 2) = 500/1000 = 0.50$

# Model

$$\begin{aligned}P(i, k) &= P(i|k)P(k) \\ &= P(k|i)P(i)\end{aligned}$$

## Interpretation

- $P(i|k)$ : level of education explains smartphone ownership
- $P(k|i)$ : smartphone ownership explains level of education

## Model

- identify stable causal relationships between the variable
- here: we select  $P(i|k)$  as an acceptable behavioral model
- stability over time necessary to forecast



# Model

## Specification

$$P(i = 1 | k = 1) = \pi_1,$$

$$P(i = 1 | k = 2) = \pi_2,$$

$$P(i = 1 | k = 3) = \pi_3.$$

## Parameters

- $\pi_1, \pi_2, \pi_3$
- unknown
- must be estimated from data

# Model estimation

$$\pi_j = P(i = 1|k = j) \approx \hat{\pi}_j = \hat{P}(i = 1|k = j) = \frac{\hat{P}(i = 1, k = j)}{\hat{P}(k = j)}$$

Using the contingency table:

$$\begin{aligned}\hat{\pi}_1 &= 75/250 = 0.300, \\ \hat{\pi}_2 &= 500/1000 = 0.500, \\ \hat{\pi}_3 &= 510/750 = 0.680.\end{aligned}$$

# Quality of the estimates

## Informal checks

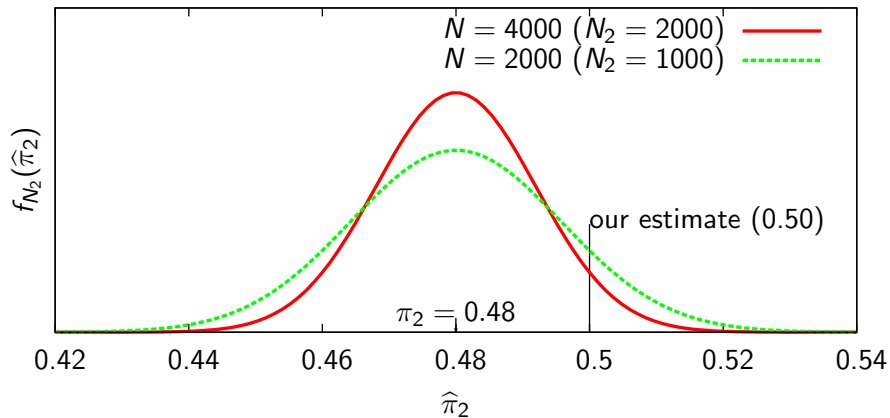
- Do these estimates make sense?
- Do they match our a priori expectations?
- Here: as years of education increases, there is a higher penetration of smartphones.

## Quality of the estimates

- How is  $\hat{\pi}_j$  different from  $\pi_j$ ?
- We have no access to  $\pi_j$
- For each sample, we would obtain a different value of  $\hat{\pi}_j$
- $\hat{\pi}_j$  is distributed.

# Quality of the estimates

## Distribution of $\pi_2$



# Quality of the estimates

## Distribution of $\pi_2$

- Smaller samples are associated with wider spread
- The larger the sample, the better the estimate
- In practice, impossible to repeat the sampling multiple times
- Distributions derived from theoretical results or simulation

## Properties

- Bernoulli (0/1) random variables
- Variance:  $\sigma_j^2 = \pi_j(1 - \pi_j)$
- Sample average: unbiased estimator
- Standard error of the estimator:  $\sqrt{\sigma^2/N}$
- Estimated standard error:

$$\hat{s}_{\pi_j} = \sqrt{\hat{\pi}_j(1 - \hat{\pi}_j)/N_j}$$

# Testing

## Estimates and standard errors

parameter	$\hat{\pi}_j$	$\hat{s}_{\pi_j}$
$\pi_1$	0.300	0.029
$\pi_2$	0.500	0.016
$\pi_3$	0.680	0.017

# Maximum likelihood estimation

## Likelihood function

$$\mathcal{L}^* = \prod_{n=1}^N P(i_n | k_n)$$

- Probability that our model reproduces exactly the observations
- For our example:

$$\mathcal{L}^* = (\pi_1)^{75} (1 - \pi_1)^{175} (\pi_2)^{500} (1 - \pi_2)^{500} (\pi_3)^{510} (1 - \pi_3)^{240}$$

# Maximum likelihood estimation

## Estimates

- Values of the parameters that maximize  $\mathcal{L}^*$ .
- In practice, the logarithm is maximized

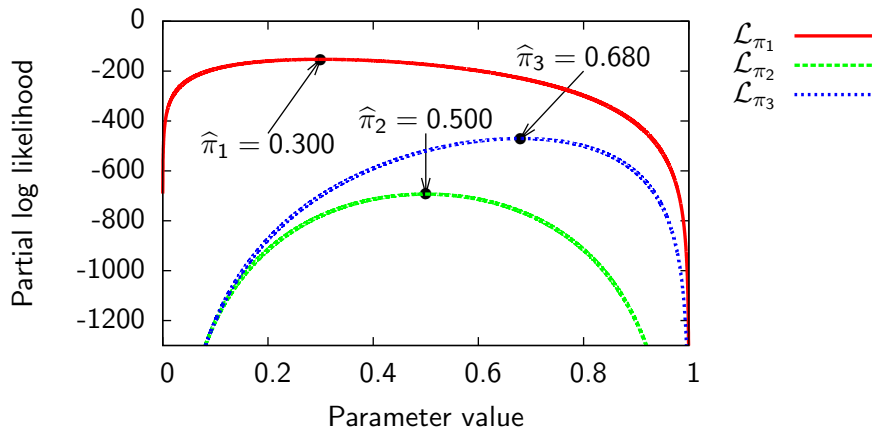
$$\mathcal{L} = \ln \mathcal{L}^* = \sum_{n=1}^N \ln P(i_n | k_n).$$

## Properties

- Consistency
- Asymptotic efficiency



# Maximum likelihood



# Hypothesis testing

## Null hypothesis

- Default hypothesis
- Is accepted except if the data tells otherwise
- Example: education has no effect on smartphone ownership
- Under the null hypothesis, we have a restricted model

$$\pi = \pi_1 = \pi_2 = \pi_3.$$

- We compare the unrestricted and the restricted model

# Hypothesis testing

## Unrestricted model

- Log likelihood function:

$$\mathcal{L} = 75 \ln(\pi_1) + 175 \ln(1 - \pi_1) + 500 \ln(\pi_2) + 500 \ln(1 - \pi_2) \\ + 510 \ln(\pi_3) + 240 \ln(1 - \pi_3)$$

- Estimates:  $\hat{\pi}_1 = 0.300$ ,  $\hat{\pi}_2 = 0.500$ ,  $\hat{\pi}_3 = 0.680$ .
- Maximum likelihood:  $-1316.0$

## Restricted model

- Log likelihood function:

$$\mathcal{L} = 1085 \ln(\pi) + 915 \ln(1 - \pi).$$

- Estimate:  $\hat{\pi} = 0.543$
- Maximum likelihood:  $-1379.1$

# Hypothesis testing

## Property

- If the null hypothesis is true
- the statistic

$$-2(\mathcal{L}^R - \mathcal{L}^U) = -2(-1379.0 + 1316.0) = 126.1$$

- is asymptotically distributed as  $\chi^2$  with degrees of freedom equal to the number of restrictions (2 here).

## Applying the test

- the critical value of the  $\chi^2$  distribution with 2 degrees of freedom at 99% significance is  $9.210 < 126.1$ .
- The null hypothesis is rejected with at least 99% confidence.
- Education *does* influence smartphone ownership.

# Model application

## Present scenario

- Level of education: low (12.5%), medium (50%), high (37.5%)
- Penetration rate:  
 $0.300 \times 12.5\% + 0.500 \times 50\% + 0.680 \times 37.5\% = 54.3\%$

## Future scenario

- Level of education will change in the future
- Level of education: low (10%), medium (40%), high (50%)
- Penetration rate:  $0.300 \times 10\% + 0.500 \times 40\% + 0.680 \times 50\% = 57\%$

## Note

- Causal relationship does not vary over time
- Values of the explanatory variables evolve over time

# Outline

- 1 Motivation
    - In this course
    - Applications
    - Importance
  - 2 Simple example
    - Choice problem
    - Data
- Model specification
  - Probabilities
  - Model
  - Estimation
  - Testing
  - Maximum likelihood
  - Hypothesis testing
  - Application