

Key concepts & study plan



Experimental design



Data collection & processing



Model specification & estimation



Interpretation & application

### Software for estimation

- Model estimation is the key step in any choice study
- □ Users rely on software for this
- Several options exist, with big differences in capabilities

### Choice modelling: history

- Used across different disciplines for over four decades
- □ Initially, small number of (advanced) users
- □ Similarly, small number of software packages
- □ Explosion in number of users and fields since 1990s
- $\hfill\square$  Ever broader group of users access to ever more advanced models

### Choice modelling: software split

- □ Fragmentation of community in terms of software, also along discipline lines
- Advanced users develop their own code
- Others split between commercial software and freeware tools
  - Commercial packages generally more powerful but with limitations in terms of available model structures or possibility for customisation
  - Freeware packages may have limitations in performance but benefit from more regular developments to accommodate new model structures

## ALogit

- Commercial software package
- By far the fastest package
- Can work with many 1,000s of alternatives and very large samples
- Widely used solution for large scale modelling
- Limited set of model structures, and restrictions on flexibility for model specification
- Limited ability for user to code models
- Classical estimation only
- Windows only



www.alogit.com

## Biogeme

- □ Free package, runs using Python
- □ Very powerful and flexible
- Large user base
- Runs on Windows and MacOS
- Classical estimation only
- Requires some Python knowledge



https://biogeme.epfl.ch/

## NLogit

- Commercial software package
- Powerful solution for applied modelling
- Includes graphical interface
- Limited ability for user to code models
- Classical estimation only
- Windows only



https://www.limdep.com/

### Other packages

- ALogit, Biogeme and NLogit are longstanding packages developed by leading choice modellers
- Also many smaller packages in Matlab, Python, R and Stata
- □ Generally focus on just a small number of features
- Big differences in flexibility, speed and accessibility
- □ Incomplete list of examples
  - Matlab: DCE, Kenneth Train's code
  - R: mixl, RSGHB
  - Python: Pylogit
  - Stata: bayesmlogit, clogit, gmnl, mixlogit



### Black boxes are a problem!

- Many packages are black box tools
- □ User has little or no knowledge of what goes on "under the hood"
- $\hfill\square$  Has made advanced models accessible to a broader group of users
- Disconnect between theory and software increases risk of misinterpretations and misspecifications
- Can also hide relevant nuances of modelling process and mistakenly give the impression that choice models are "easy tools" to use

#### Bayesians vs the rest

- Existing software almost exclusively allows use of only either classical estimation techniques or Bayesian techniques
- □ Fragmentation again runs largely in parallel with discipline boundaries
- Has only served to further contribute to lack of interaction/dialogue between classical and Bayesian communities



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### Existing limitations led to development of Apollo

- □ Culmination of 20 years of work
- Started with Ox code developed by Hess at Imperial College, translated into R at University of Leeds
- Combined with code developed by Palma at Pontificia Universidad Católica de Chile
- Major continuous further developments





- Couldn't base name on single model family as *Apollo* is more general
- Failure to come up with clever acronym led us to Greek mythology
- Apollo was the Greek god of prophecy

### Approach in Apollo

Free access: *Apollo* is a completely free package which does not rely on commercial statistical software as a host environment.

Big community: Apollo relies on R, a free software environment for statistical computing and graphics, which is very widely used across disciplines and works well across different operating systems.

Transparent, yet accessible: Apollo is neither a blackbox nor does it require expert econometric skills. The user can see as much or as little detail of the underlying methodology as desired, but the link between inputs and outputs remains.

Ease of use: Apollo combines easy to use R functions with new intuitive functions without unnecessary jargon or complexity.

### Approach in Apollo

Modular nature: Apollo uses the same code structure independently of whether the simplest multinomial logit model is to be estimated, or a complex structure using random coefficients and combining multiple model components.

Fully customisable: Apollo provides functions for many well known models but the user is able to add new structures and still make use of the overall code framework. This for example extends to coding expectation-maximisation routines.

Discrete and continuous: *Apollo* incorporates functions not just for commonly used discrete choice models but also for a family of models that looks jointly at discrete and continuous choices.

### Approach in Apollo

Novel structures: Apollo goes beyond standard choice models by incorporating the ability to estimate Decision Field Theory (DFT) models, a popular accumulator model from mathematical psychology.

Classical and Bayesian: Apollo does not restrict the user to either classical or Bayesian estimation but easily allows changing from one to the other.

Easy multi-threading: Apollo allows users to split the computational work across multiple processors without making changes to the model code.

Not limited to estimation: Apollo provides a number of pre and post-estimation tools, including diagnostics as well as prediction/forecasting capabilities and posterior analysis of model estimates.

## **Files required**

□ A data file (various formats can be read into R)

- not needed if data is read from a package
- A model file
  - contains all details on the model and tells R what routines to run
- □ Easier if data and model files are in the same directory

### Data format

- Apollo uses "wide" format
- All necessary information to calculate likelihood of a single observation should be contained in a single row of the data
- In more practical terms, for an MNL model, this means that attributes for all alternatives should be contained in each row
- Most common format in choice modelling, uses less space, and more general in allowing for a mixture of different dependent variables in the same data
- $\hfill\square$  Some other software uses "long" format, with one row per alternative
  - Apollo is not set up for such data, but reshaping the data is a straightforward task using apollo\_longToWide

#### Resources: www.apollochoicemodelling.com



### Documentation

- Website includes a detailed manual and shorter research paper
- In addition, syntax help for all functions is available directly in R
  - e.g. ?apollo\_mnl

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unctionality	Character, Betting instructing Apolio what processing to apply to the likelihood function. This is in general controlled by the functions that cell apollo_processilities, through the user can also cell apollo_processilities manually with a given functionality for isotropicotogoing. Proceeding was net:
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	<ul> <li>"estimate": For model estimation, produces likelihood of the full model, at the level of individual decision-makers, after averaging across draws.</li> </ul>
	<ul> <li>"gradient.": For model estimation, produces analytical gradients of the likelihood, where possible.</li> </ul>
	<ul> <li>"evalpath": Propares output for post-estimation reporting.</li> </ul>
	<ul> <li>"prediction": For model prediction, produces probabilities for individual alternatives and individual model components (f multiple components are present) at the level of an observation, after overaging across draws.</li> </ul>
	<ul> <li>*preprocessa': Prepares likelihood functions for use in estimation.</li> </ul>
	<ul> <li>* saw": For debugging, produces probabilities of all alternatives and individual model components at the level of an observation, at the level of individual chave.</li> </ul>
	<ul> <li>"separt": Prepares output summarising model and choiceset structure.</li> </ul>
	<ul> <li>"shares_LL": Produces overall model likelihood with constants only.</li> </ul>
	<ul> <li>"validates ": Validates model specification, produces likelihood of the full model, at the level of individual decision-makers, after averaging across draws.</li> </ul>
	<ul> <li>"sere_ttt": Produces overall model likelihood with all parameters at zero.</li> </ul>