

Emphasizing the Progression of Knowledge Over the Individual Work

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3rd Symposium on Activity-Based Modeling
TUM Science & Study Center Raitenhaslach
December 2024



IATBR *“Are we in crisis?”* workshop feedback

We asked:



Are we achieving our objectives?

Rating Poll 52 votes 52 participants

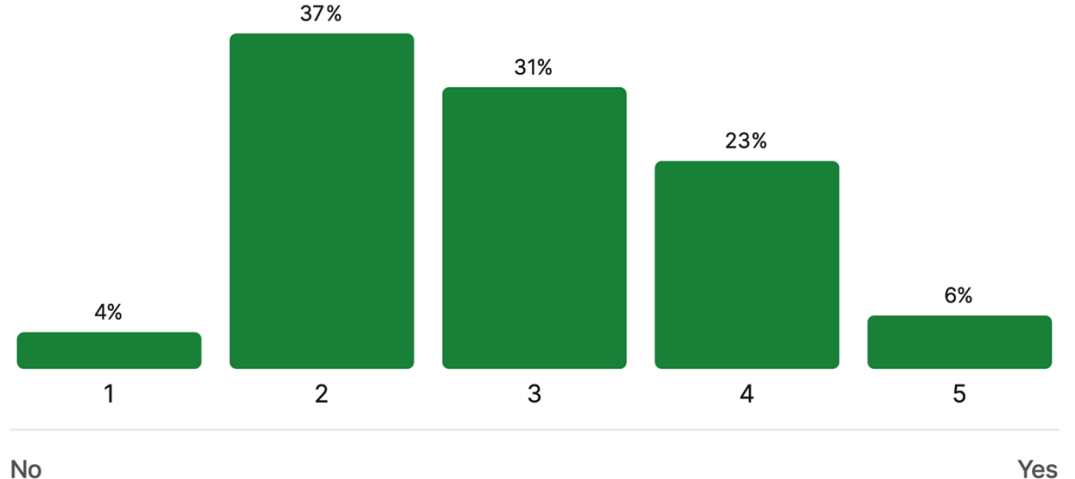
“What is the objective of our field?”

Score: 2.9

Majority responded:

“To inform policy and decision-making”

Are we getting there?



Lessons from other fields

Machine learning:

Standardised benchmarks

Open datasets


Economics:


Robustness checks


Psychology:

Replicability



 **BMC** Part of Springer Nature

 **BioData Mining**

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Research | [Open access](#) | Published: 11 December 2017

PMLB: a large benchmark suite for machine learning evaluation and comparison

Randal S. Olson , William La Cava, Patryk Orzechowski, Ryan J. Urbanowicz & Jason H. Moore

BioData Mining **10**, Article number: 36 (2017) | [Cite this article](#)

29k Accesses | **201** Citations | **46** Altmetric | [Metrics](#)

IATBR *"Are we in crisis?"* workshop feedback (II)

Top priorities identified at IATBR:

- Common datasets
- Open-source infrastructure

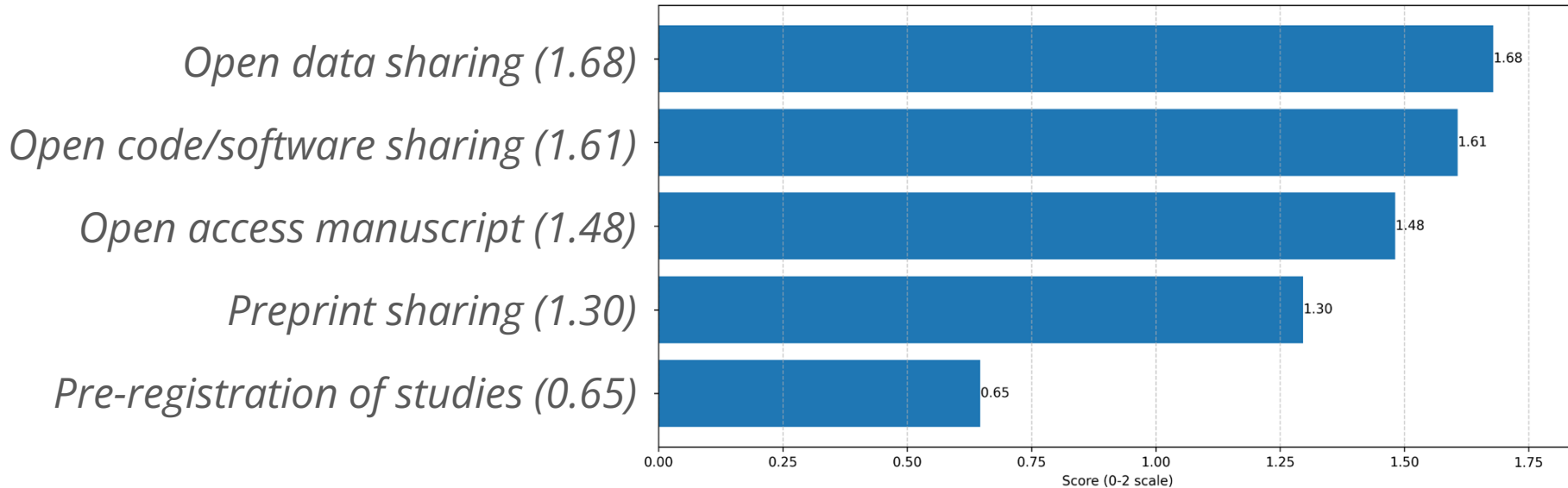
OUTLINE TODAY

- Effort #1: Open-science initiative
- Effort #2: Benchmarking infrastructure for mode choice

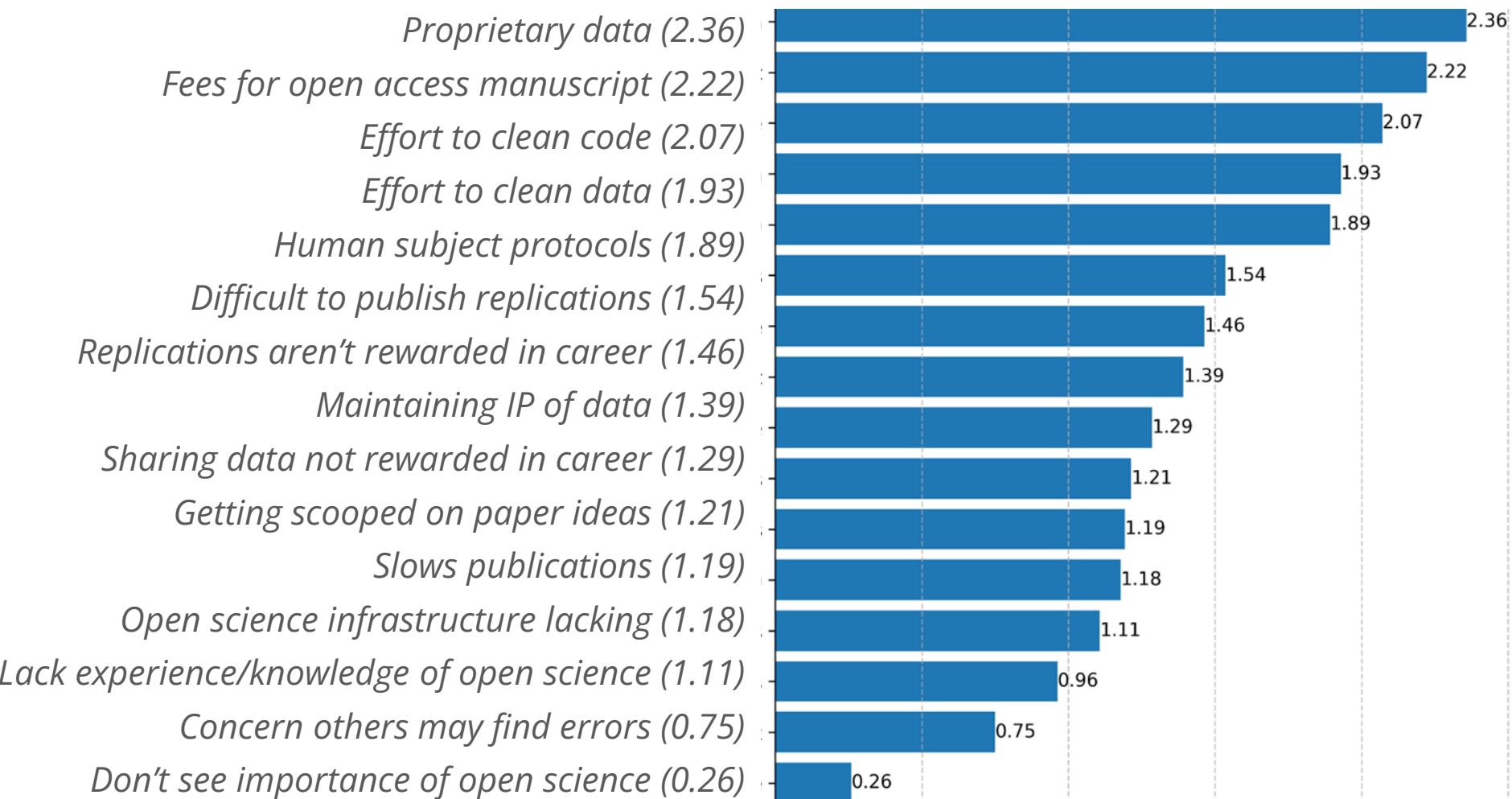
Open Science: collaborating differently

n=28 of you!

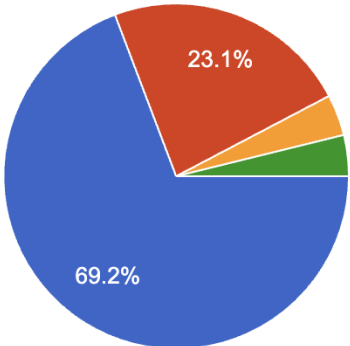
How important do you think this is for travel behavior research?
(composite score 0 = all say not important... 2 = all respond very important)



What limits your ability to practice open science? (0-3 point composite score)



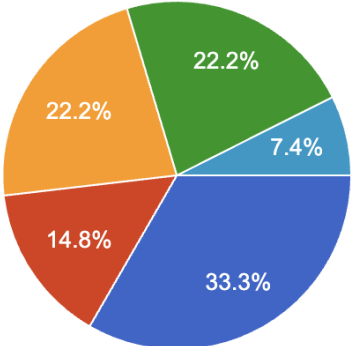
Great intentions to share...



Manuscript

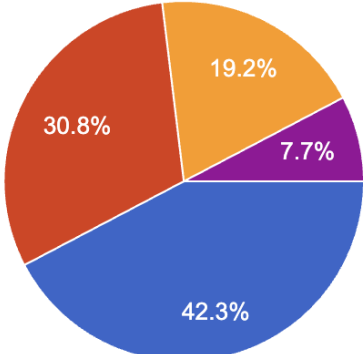
92% yes

n=26 of you



Data

48% yes

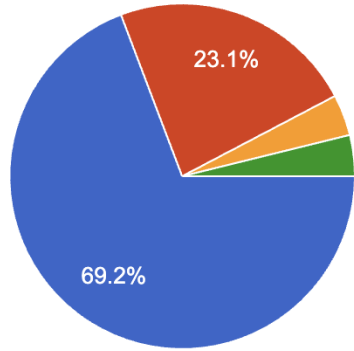


Code

73% yes

- Definitely yes
- Probably yes
- Probably no
- Definitely no

... not so great results



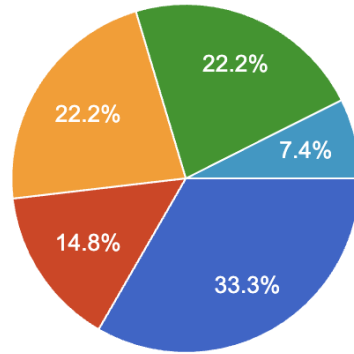
Manuscript

92% yes

n=26 of you

65% yes

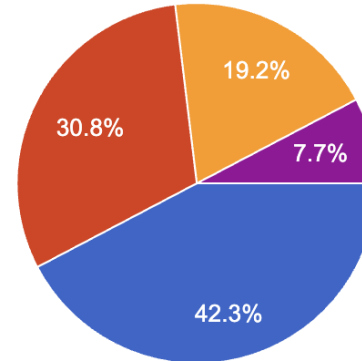
Last 20 papers in



Data

48% yes

15% yes



Code

73% yes

5% yes

- Definitely yes
- Probably yes
- Probably no
- Definitely no

Effort #1

IATBR Operation 100% OPEN SCIENCE

Open data and code is first, critical step.

Targeting 11 special issues from IATBR Vienna 2024.

Objectives

- Creating processes, definitions, and practices for open science in our field
- Increasing open science participation by nudging and working through concerns of researchers (e.g., intellectual property rights)
- Documenting limitations of open science in our field (e.g., proprietary software and data, protection of human subjects)

Effort #2: Rethinking our modeling practice

What would a benchmarking suite look like for our field?

Let's tackle mode choice!

Using PMLB Python interface

Install

Most users should install PMLB from the Python Package Index (PyPI) using `pip`:

```
$ pip install pmlb
```

For access to new features or datasets that are not yet merged into an official release, you can instead install from [the GitHub sources](#):

```
$ git clone https://github.com/EpistasisLab/pmlb
$ cd penn-ml-benchamrks/
$ pip install .
```

Usage

```
from pmlb import fetch_data

# Returns a pandas DataFrame
mushroom = fetch_data('mushroom')
mushroom.describe().transpose()
```

Source: Penn Machine Learning Benchmarks (PMLB)

Mode choice benchmarking sandbox (MCBS)

Python library built on Biogeme

COMPILES

- Datasets
- Methodologies

GENERATES DIRECT COMPARISONS

- Across datasets
(Swissmetro, London, Canada)
- Across model formulations
(Logit, Nested Logit, Mixed Logit)
- Across multiple metrics
(Goodness of fit, Value of Time, Forecast)

Mode Choice Benchmarking Sandbox (MCBS)

A Python package for benchmarking discrete choice models

[View on GitHub](#)

[Download .zip](#)

[Download .tar.gz](#)

Mode Choice Benchmarking Sandbox (MCBS)

MCBS is a specialized Python package designed to streamline the development and evaluation of transportation mode choice models. It provides researchers and practitioners with a standardized environment for implementing, testing, and comparing different discrete choice modeling approaches.

<https://carlosguirado.github.io/mode-choice-benchmarking-sandbox/>



Easy setup

Quick Start

Install MCBS using pip:

```
pip install mcbs
```

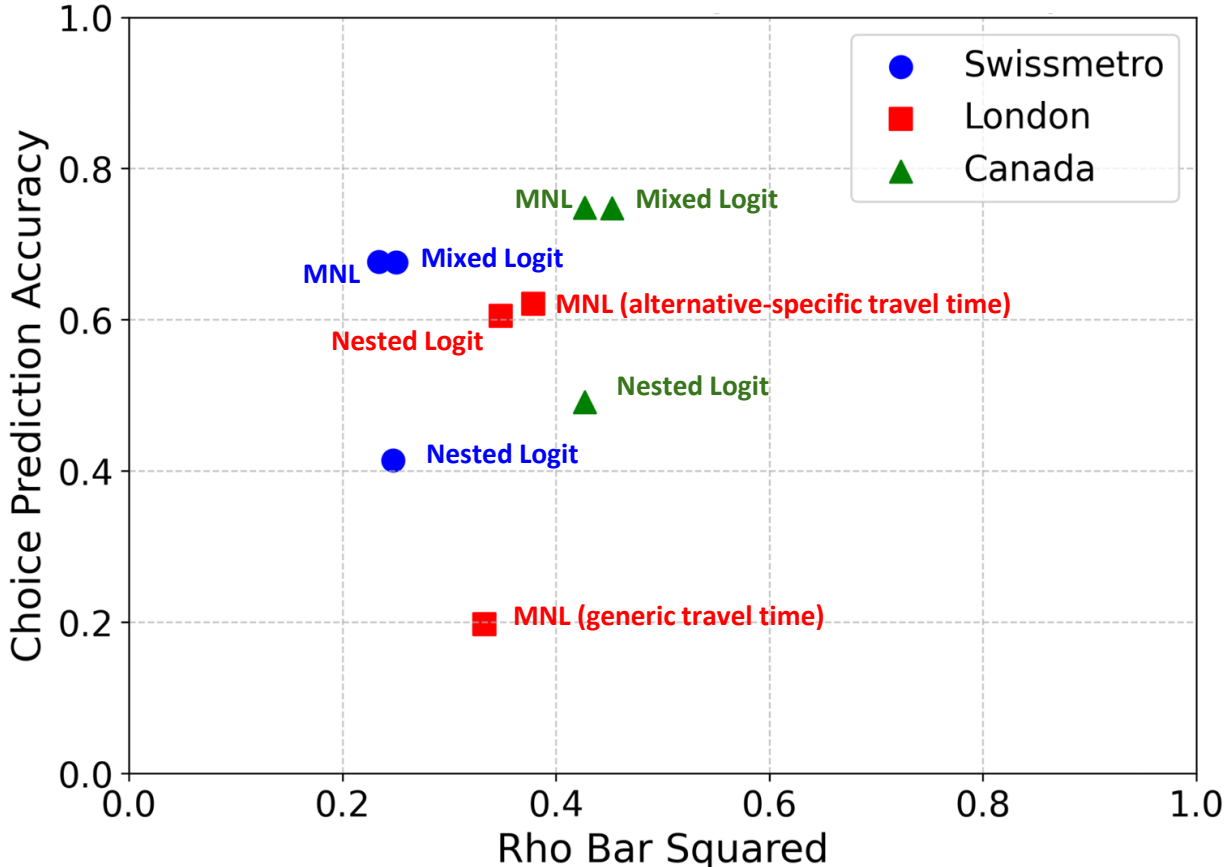
Basic usage example:

```
from mcbs import Benchmark

# Initialize benchmark with dataset
benchmark = Benchmark("swissmetro_dataset")

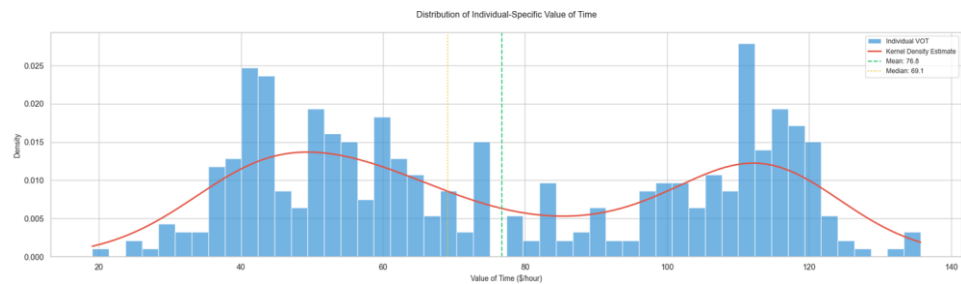
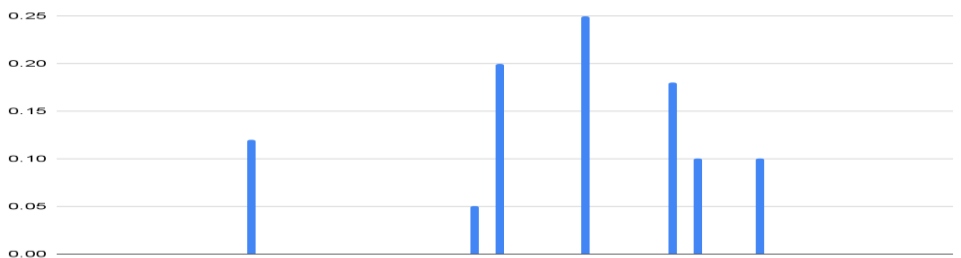
# Run your models
results = benchmark.run(models)
benchmark.compare_results(results)
```

METRIC #1: Goodness of Fit



METRIC #2: Value of Time

Sample density



Value of Time

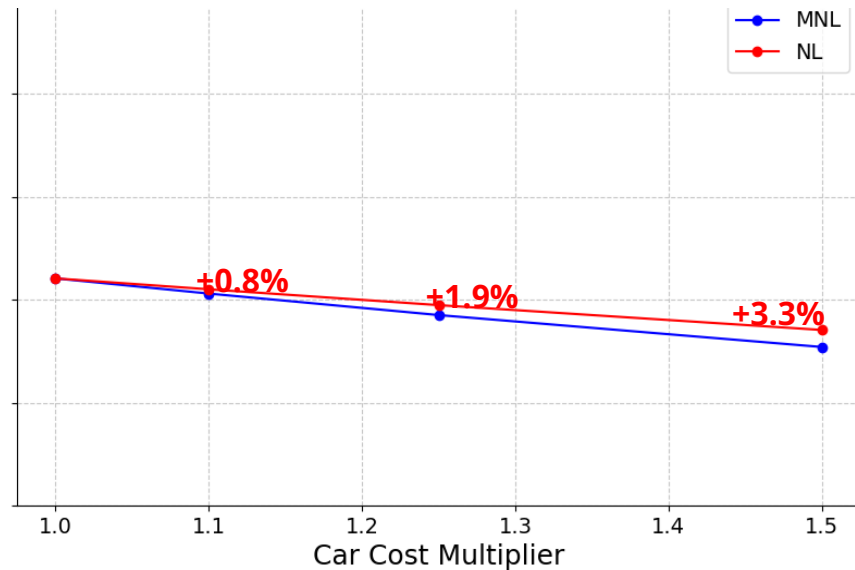
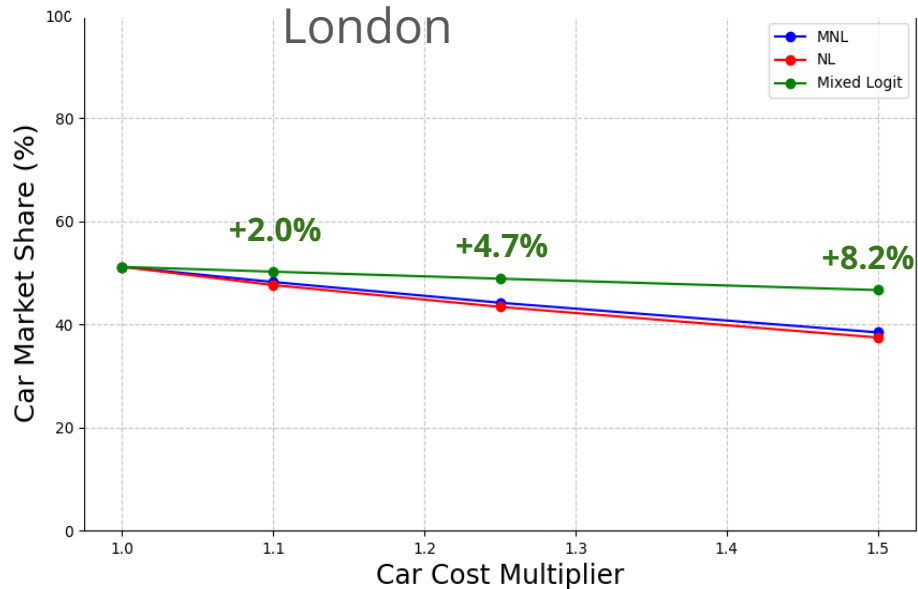
MODEL A:
Logit - Homogeneous

MODEL B:
Logit- Systematic heterogeneity

MODEL C:
Mixed logit
(posterior)

METRIC #3: Forecast (Congestion Pricing)

Canada
London



Reflection

Building such a sandbox isn't easy... but Rome wasn't built in a day.

ChatGPT estimates 10,000-20,000 scholarly papers on mode choice.

- What would our (mode choice) models look like today if we had such an infrastructure or explicit comparative culture for the last 50+ years?
- How would new modes such as Shared Mobility and Autonomous Vehicles have been integrated?

What would our impact be if we prioritized credibility?

Habit models from computational neuroscience

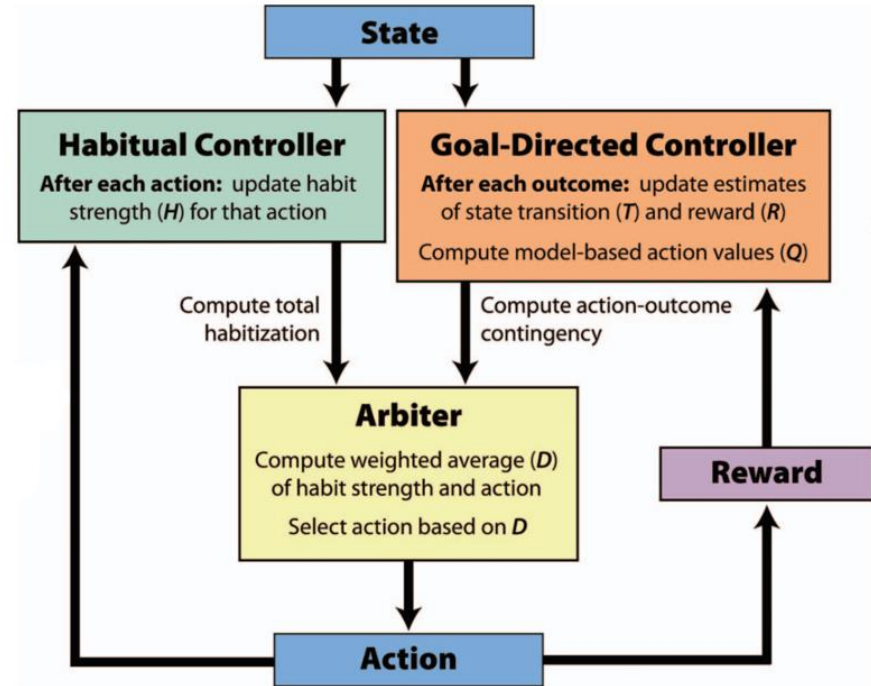
Qianhua Luo, Berkeley PhD student

Cool model!

- Dual system theory
- Models habit formation
- Application to Telecommute Decisions through the Pandemic

Open science reflection:

- Smartphone data, yet will publish estimation dataset/code – It's a pain!
- Obvious papers that should be compared on same dataset (Cherchi, Bansal) – Not done or valued!



Miller et al. (2019)

Discussion

How could the work in this symposium be coordinated to better achieve our objective of informing policy & planning?

What would a benchmarking sandbox for evolutionary behavior look like?

