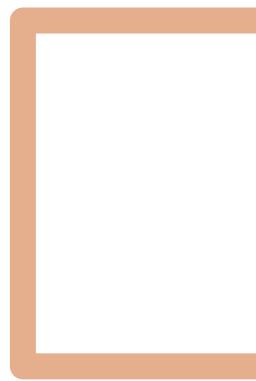
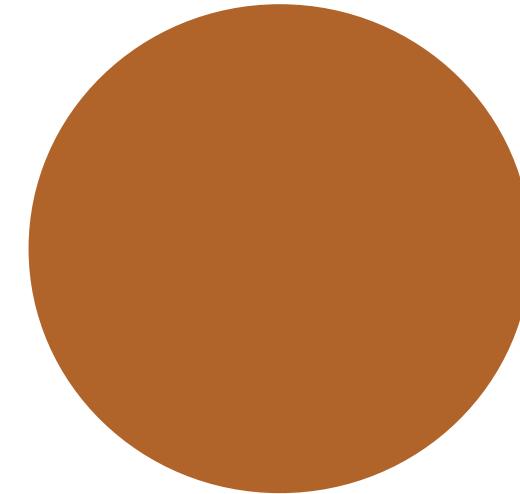
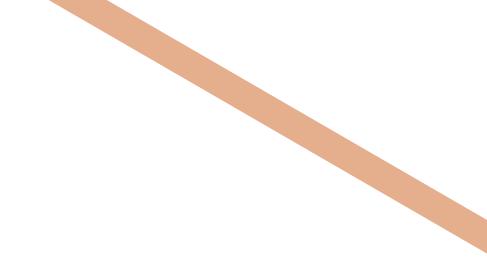


Combining Register and Survey Data to Understand Carbon- Intensive Time Use

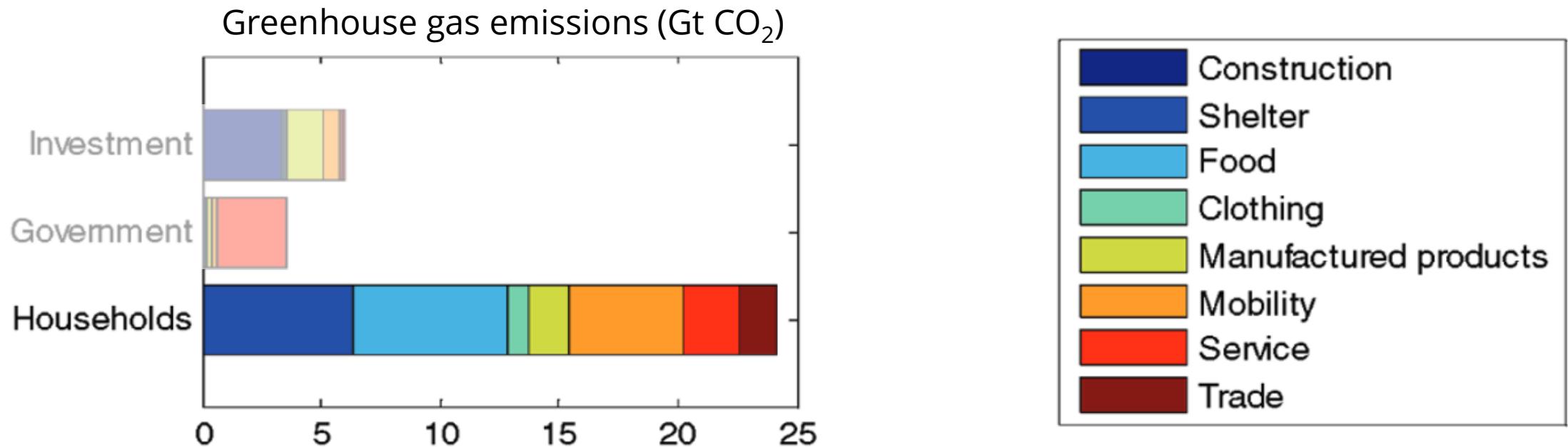
Maike Weiper

Erik-Jan van Kesteren
Javier Garcia Bernardo
Qixiang Fang
Jiamin Ou

The Context



Consumption-Based Emissions



Research Questions

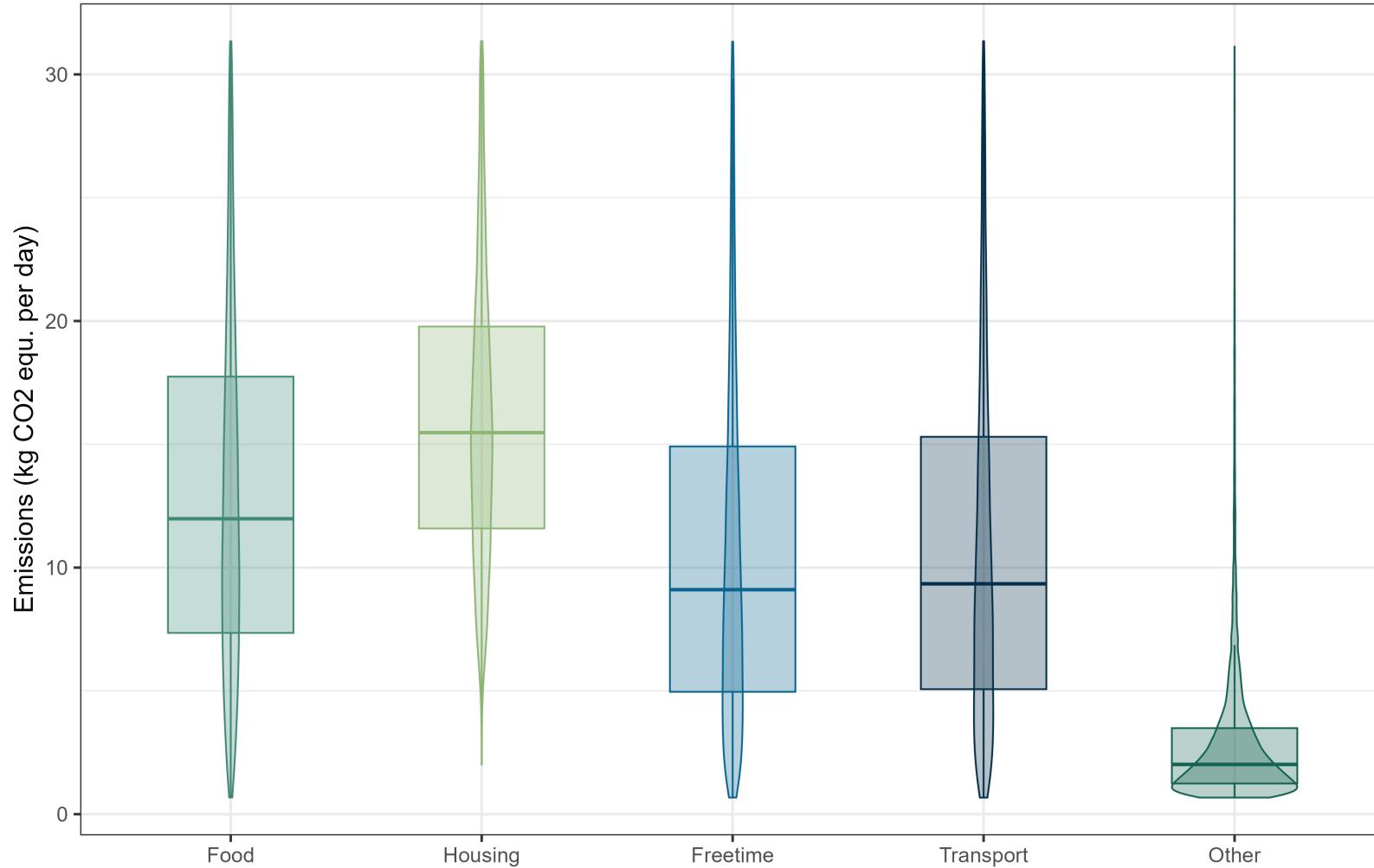
- What are the carbon emissions of Dutch households?
- How do household members contribute to their household carbon emissions?



Household Carbon Emissions

- Household expenditure survey
 - 15.000 households in 2015
 - 1 month expenditures
 - 155 products
- Environmentally-extended input-output analysis

Household Carbon Emissions

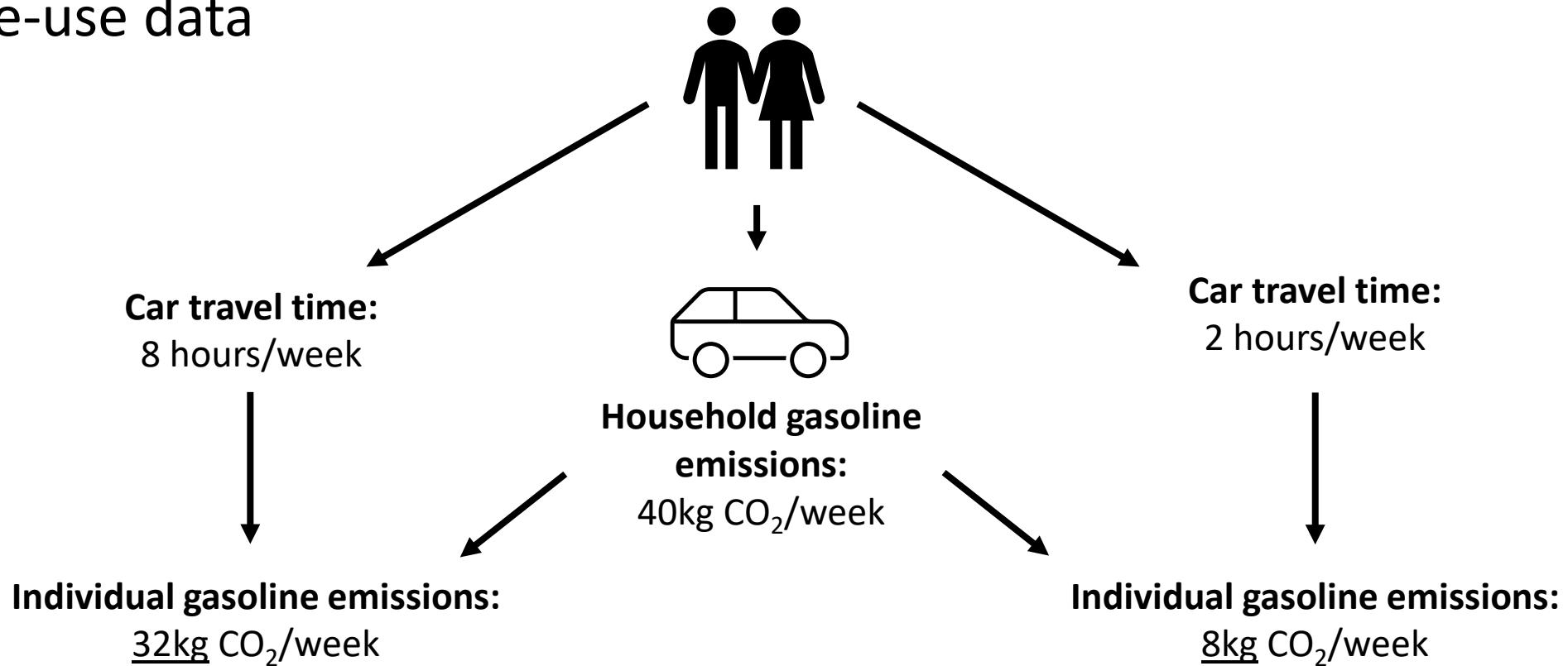


Research Questions

- What are the carbon emissions of Dutch households?
- How do household members contribute to their household carbon emissions?
 - What are the carbon emissions of individuals living in the Netherlands?

Individual Carbon Emissions

- Attribute household carbon emissions to individuals
- Time-use data

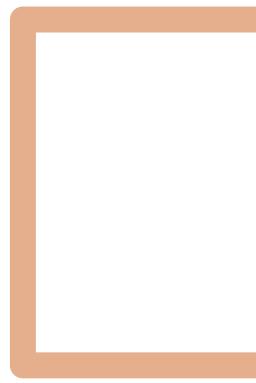
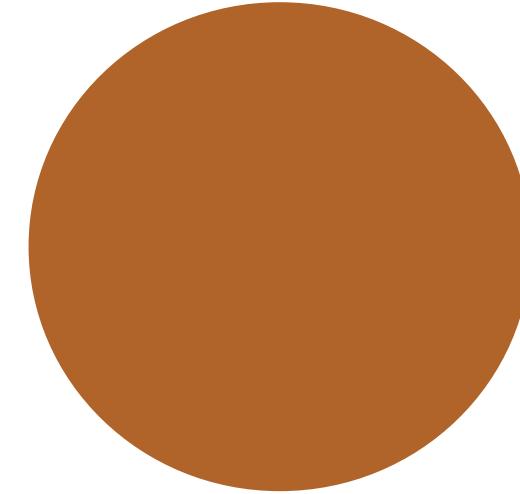
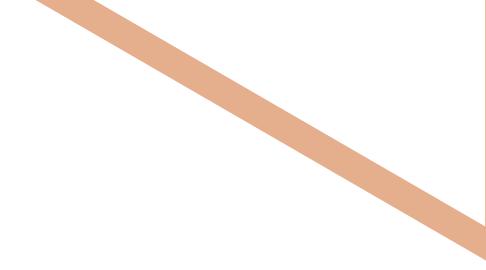


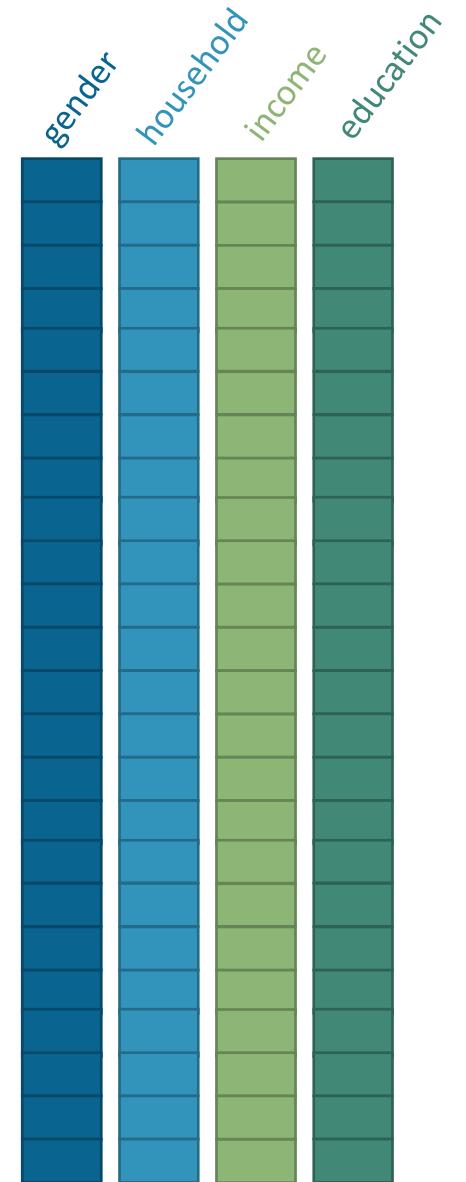


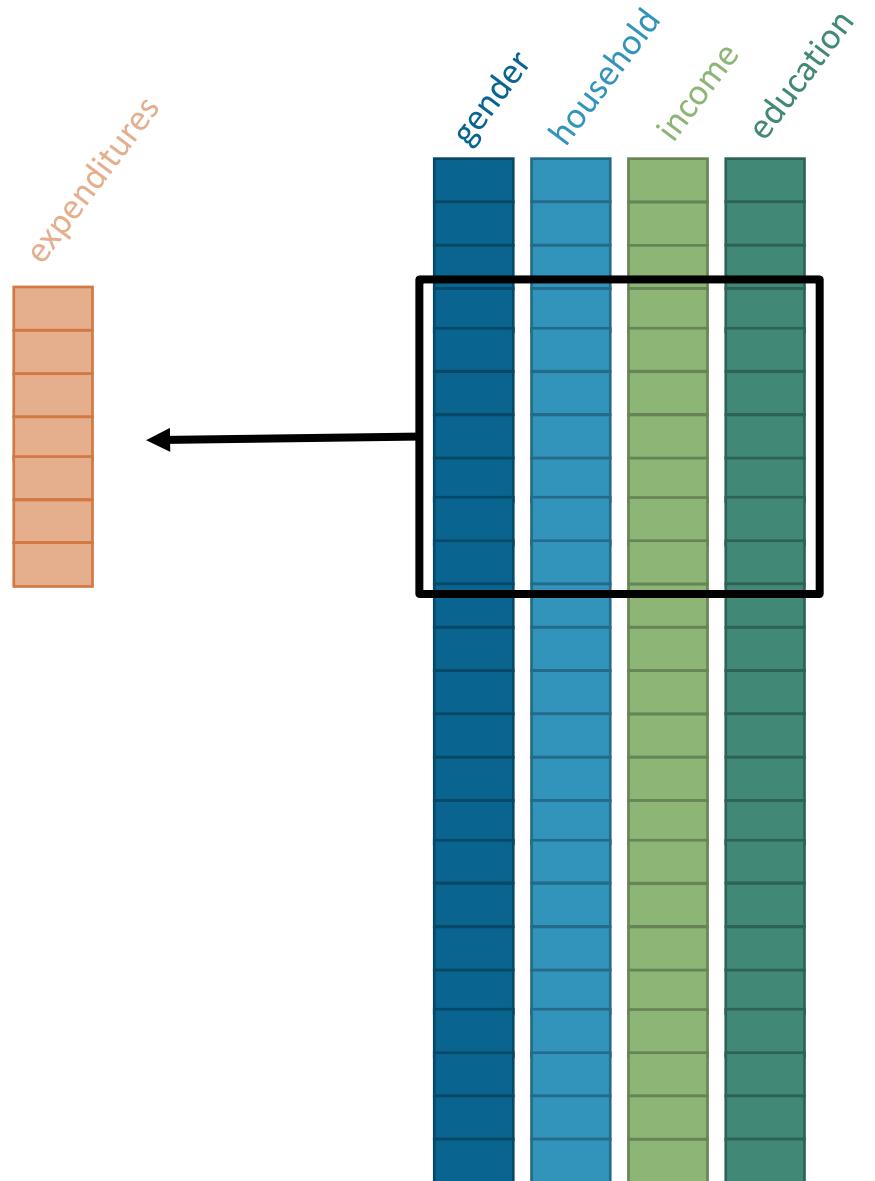
Individual Carbon Emissions

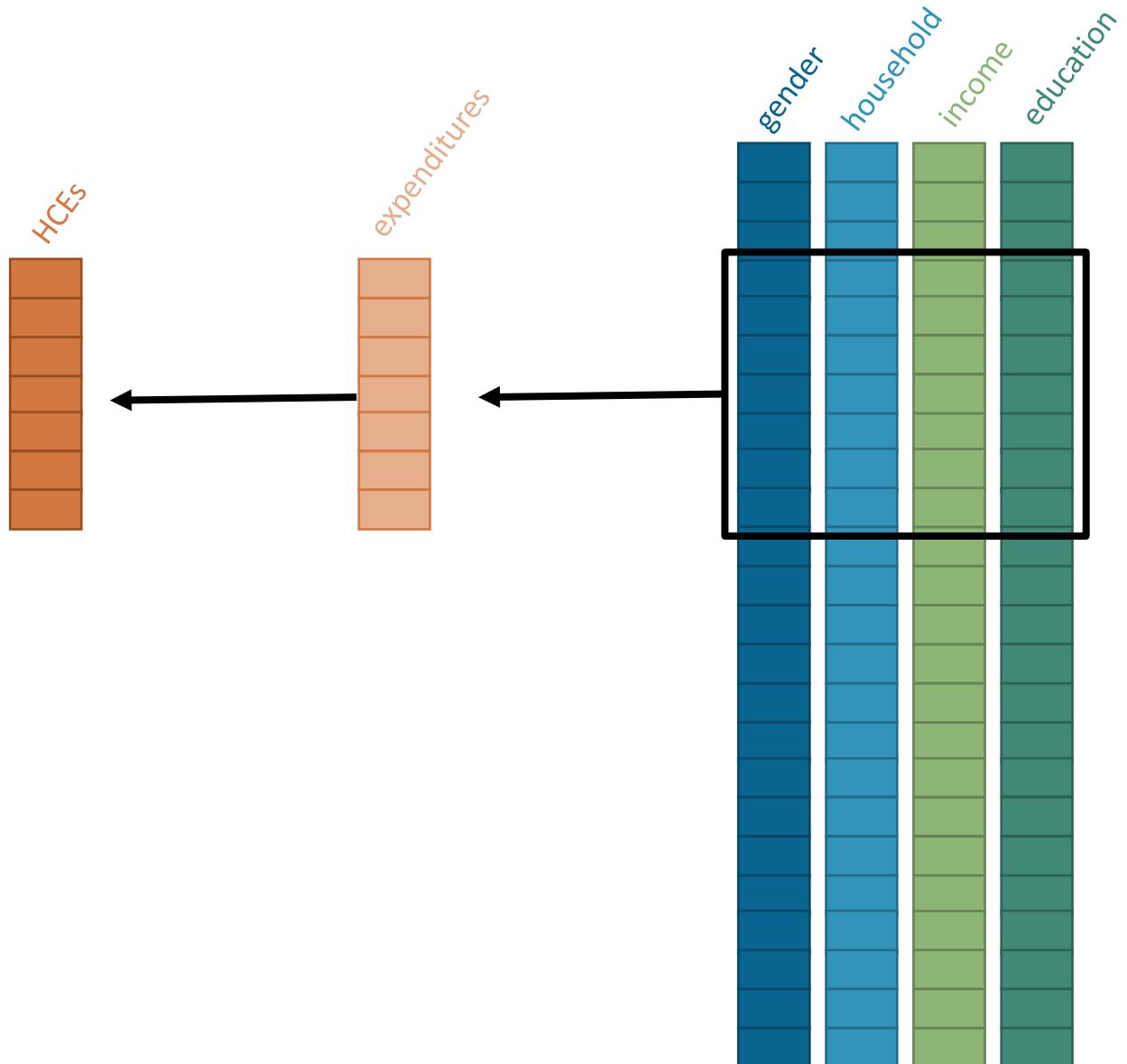
- Attribute household carbon emissions to individuals
- Time-use data
 - ~ 2300 individuals in 2016
 - 1 week of activities
 - 10 minute intervals

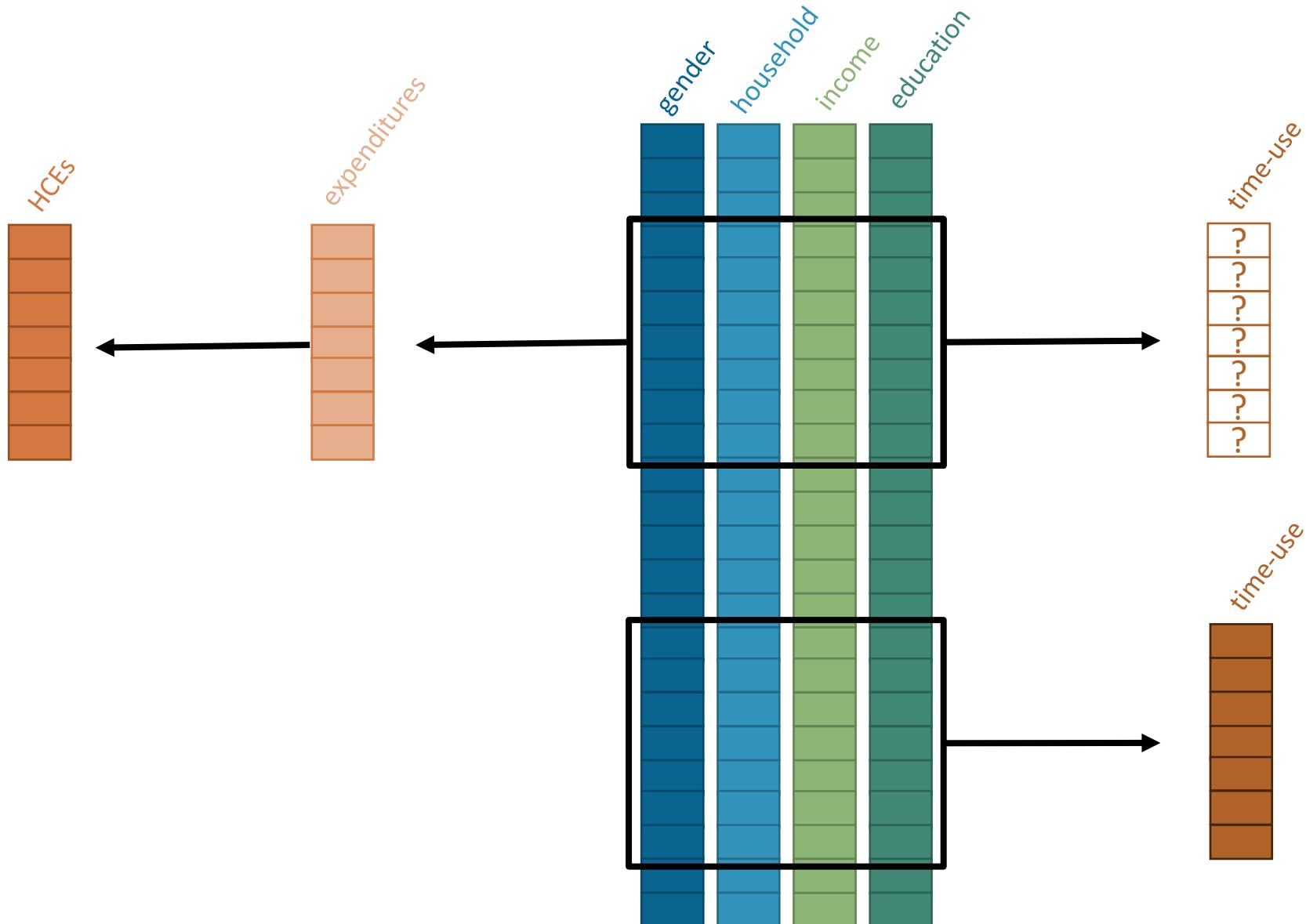
The Problem



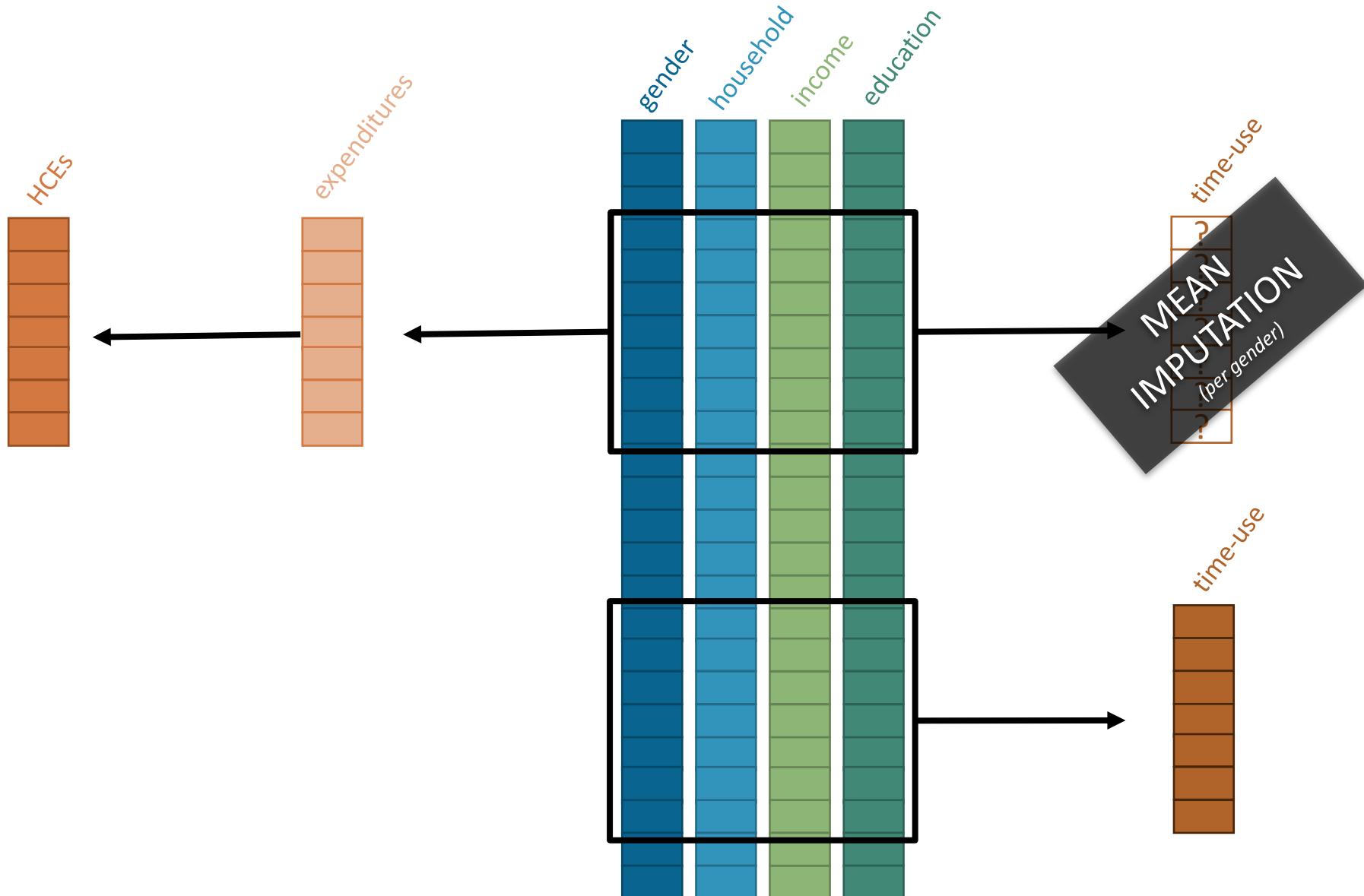


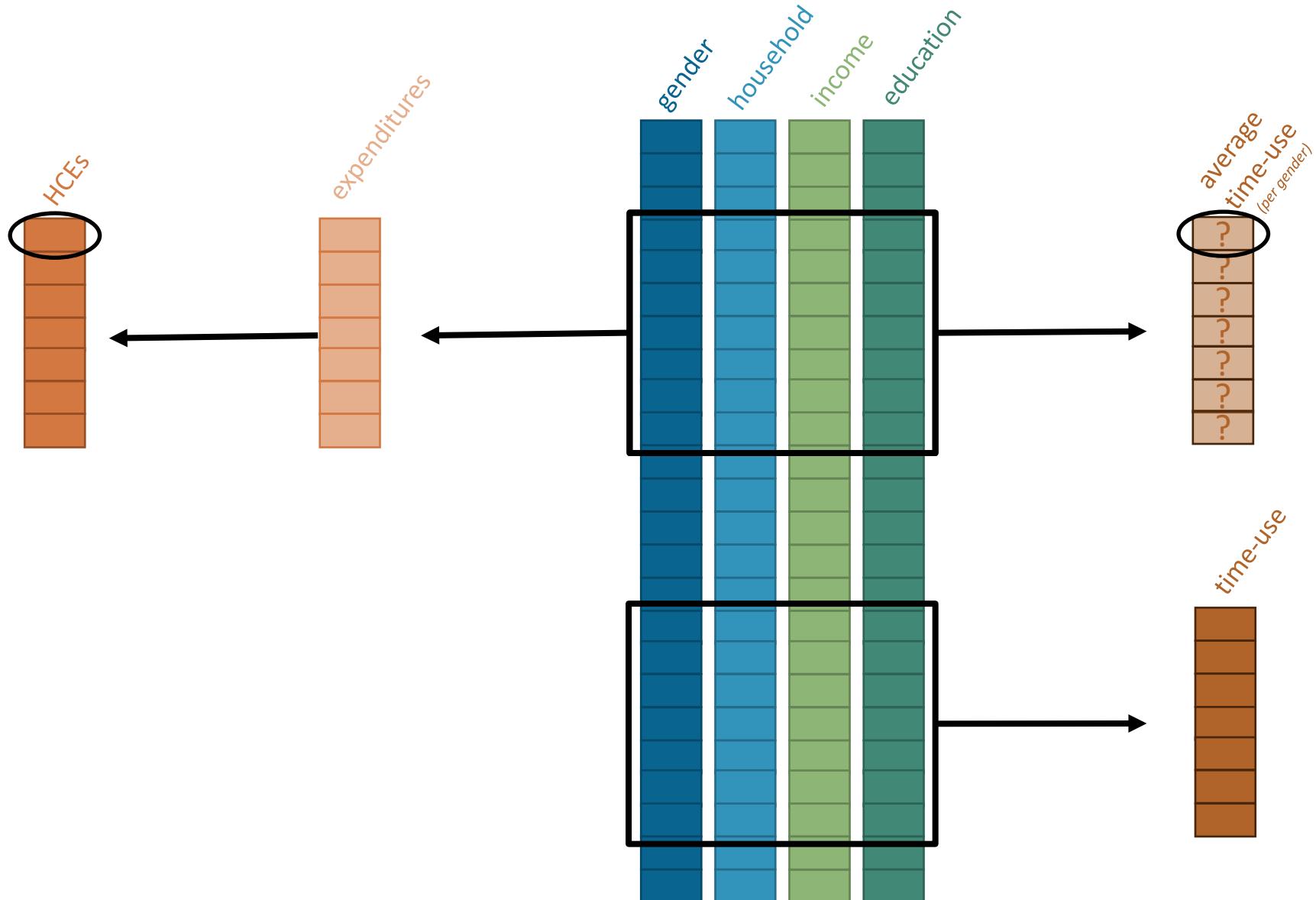




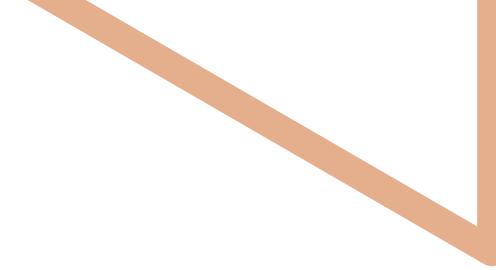
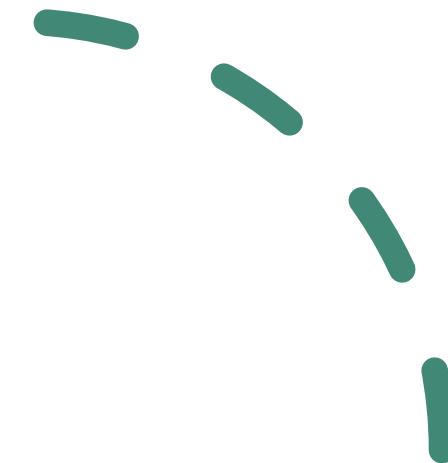
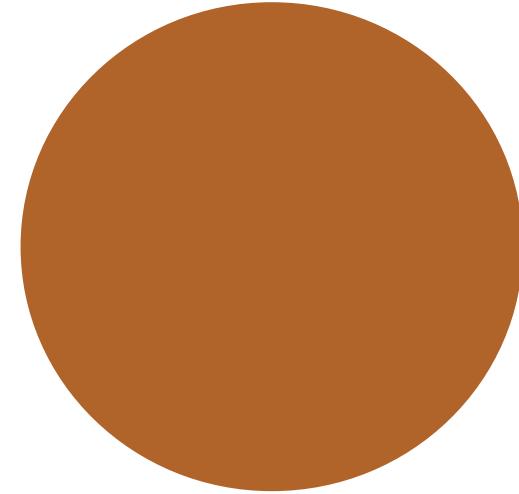


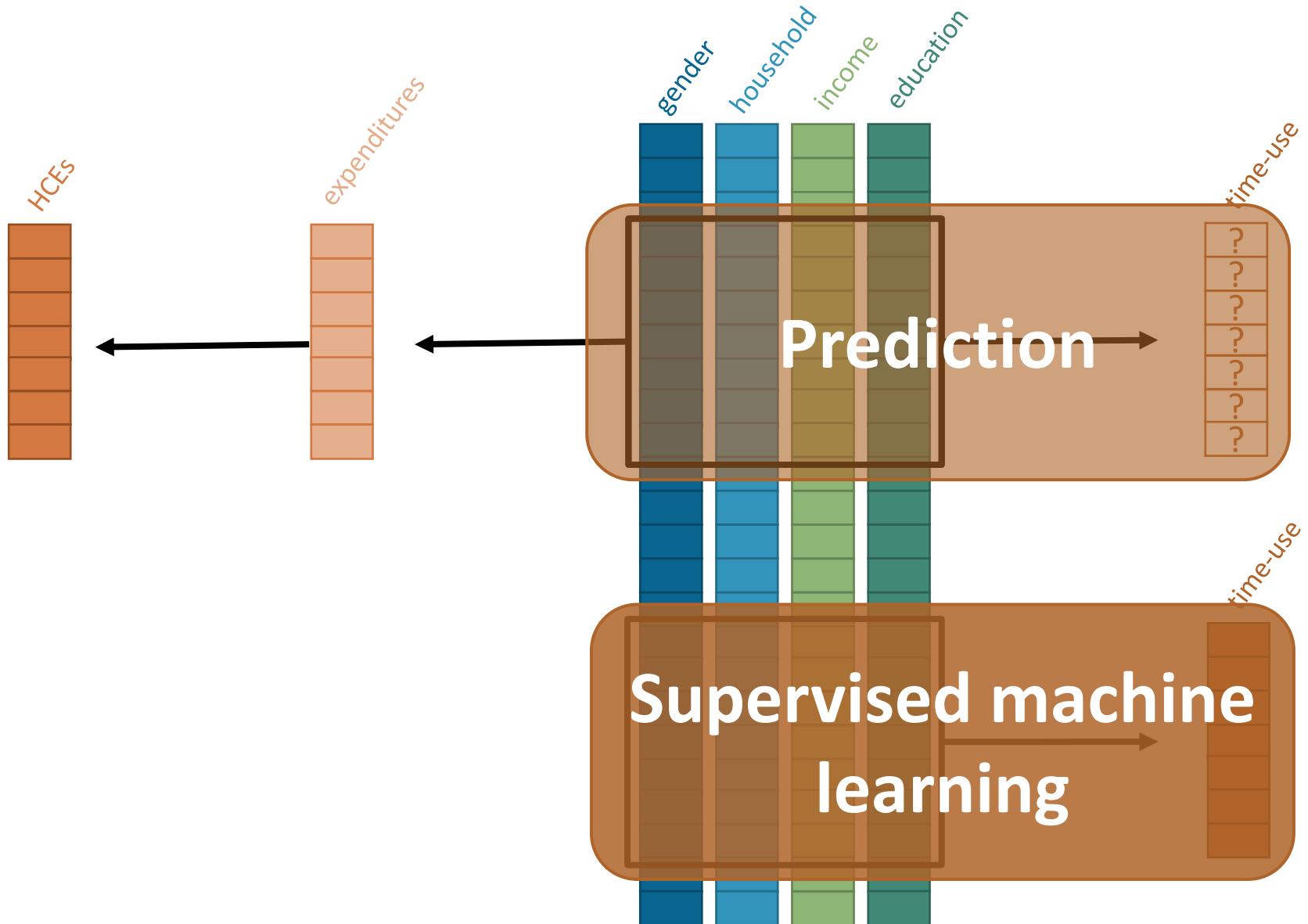
What others have
done...

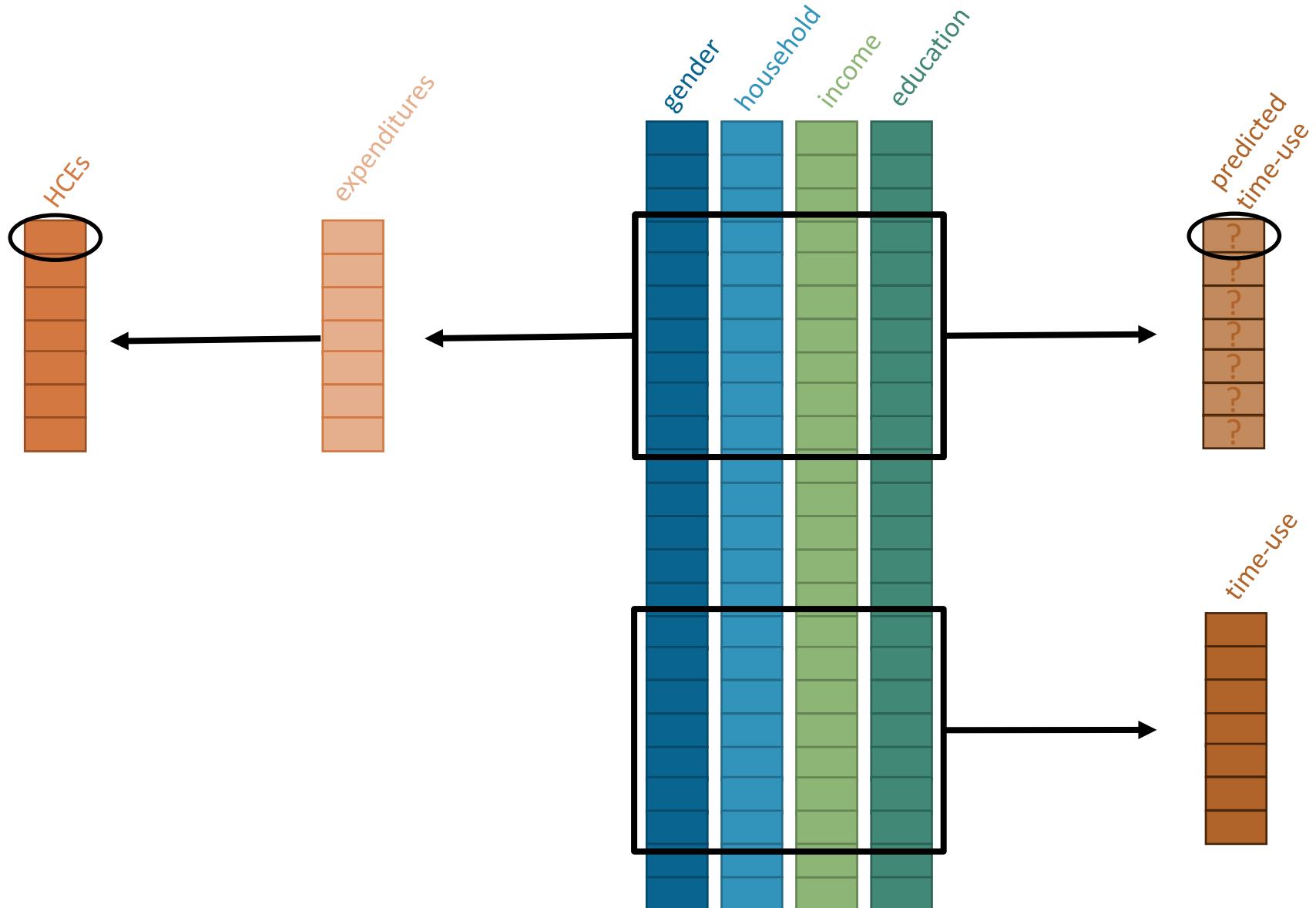




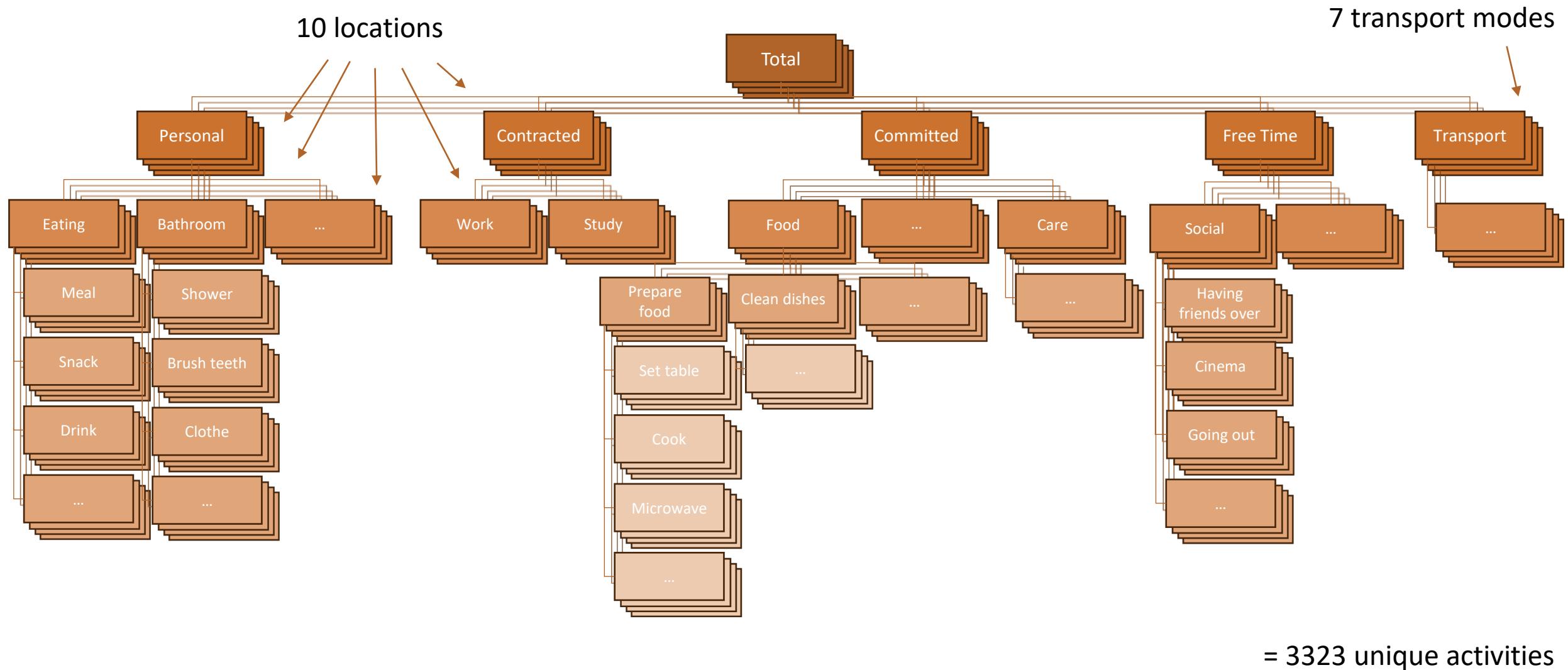
We think we can
do better!



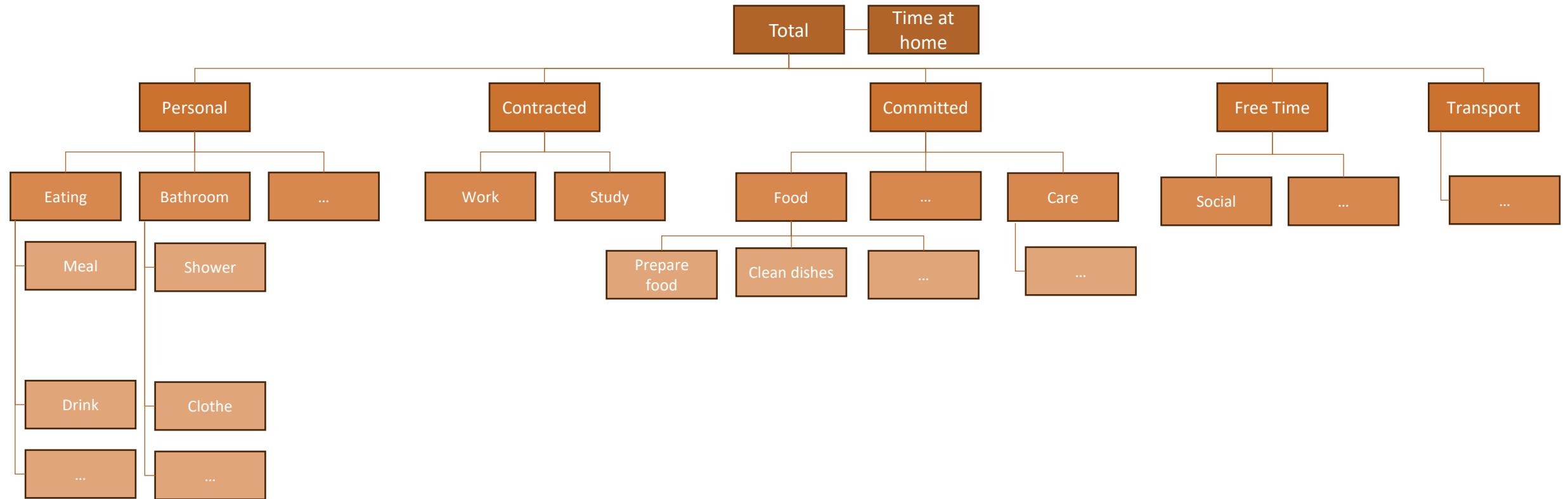




Predicting Time-Use



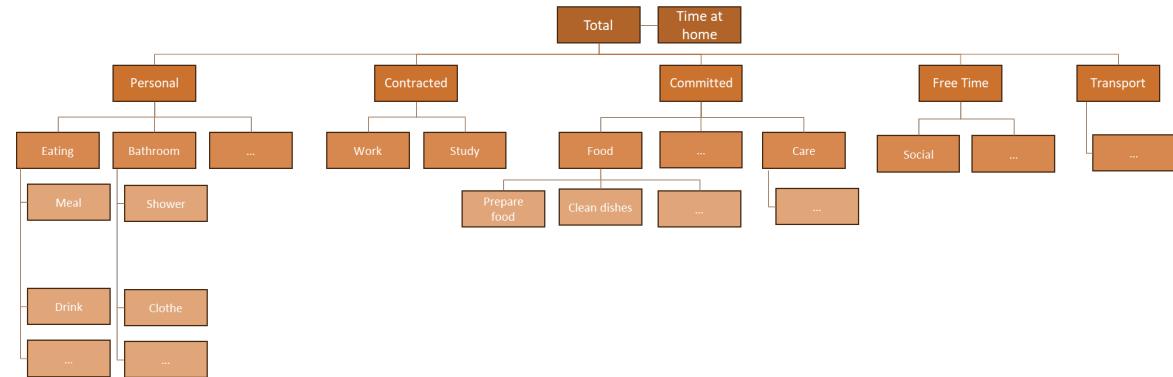
Predicting Time-Use



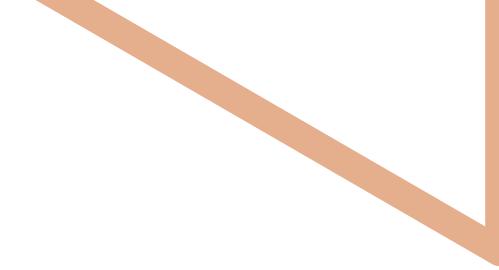
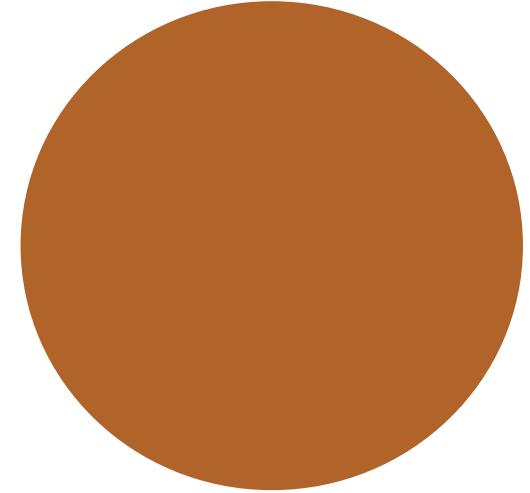
= 114 unique activities

Predicting Time-Use

- 114 activities to predict
 - Not normally distributed
- Features
 - 7 CBS datasets
 - 20 features (10 categorical)
 - Missing values
- Histogram-based gradient boosting regressor
 - Based on a number of sequential decision trees



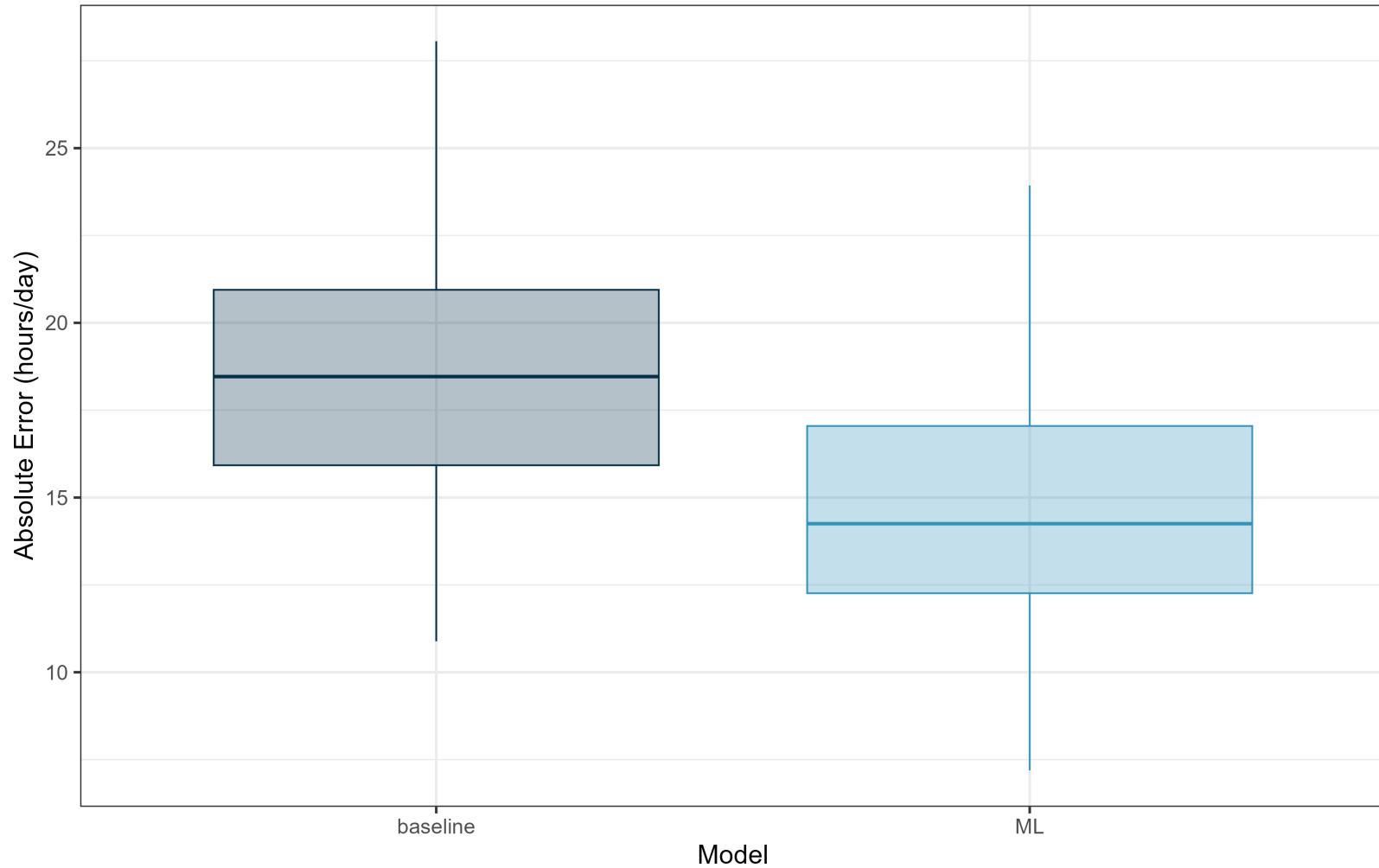
So, did we do
better?



So, did we do
better?

Prediction errors

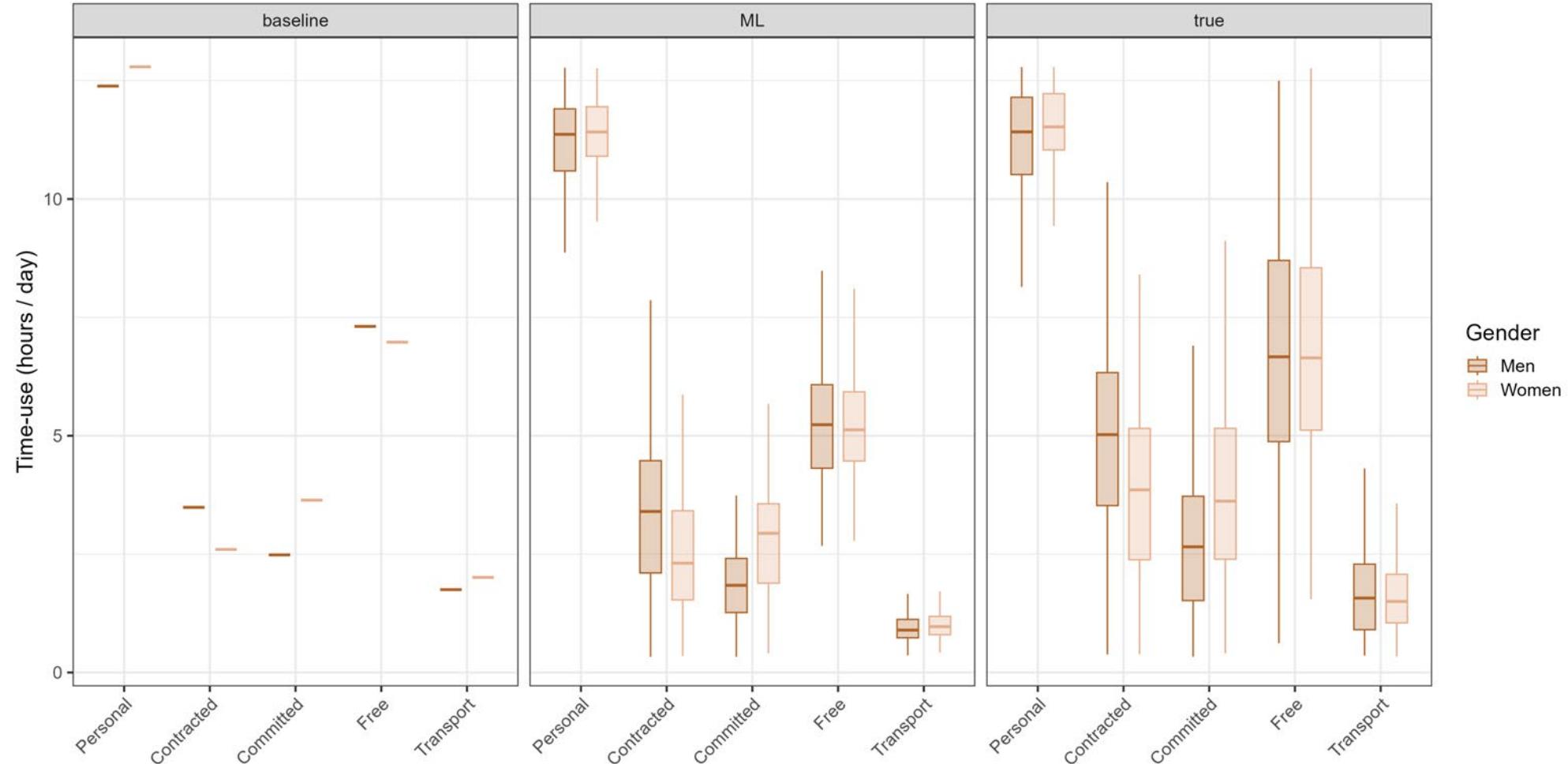
Predicted Time-Use (errors)



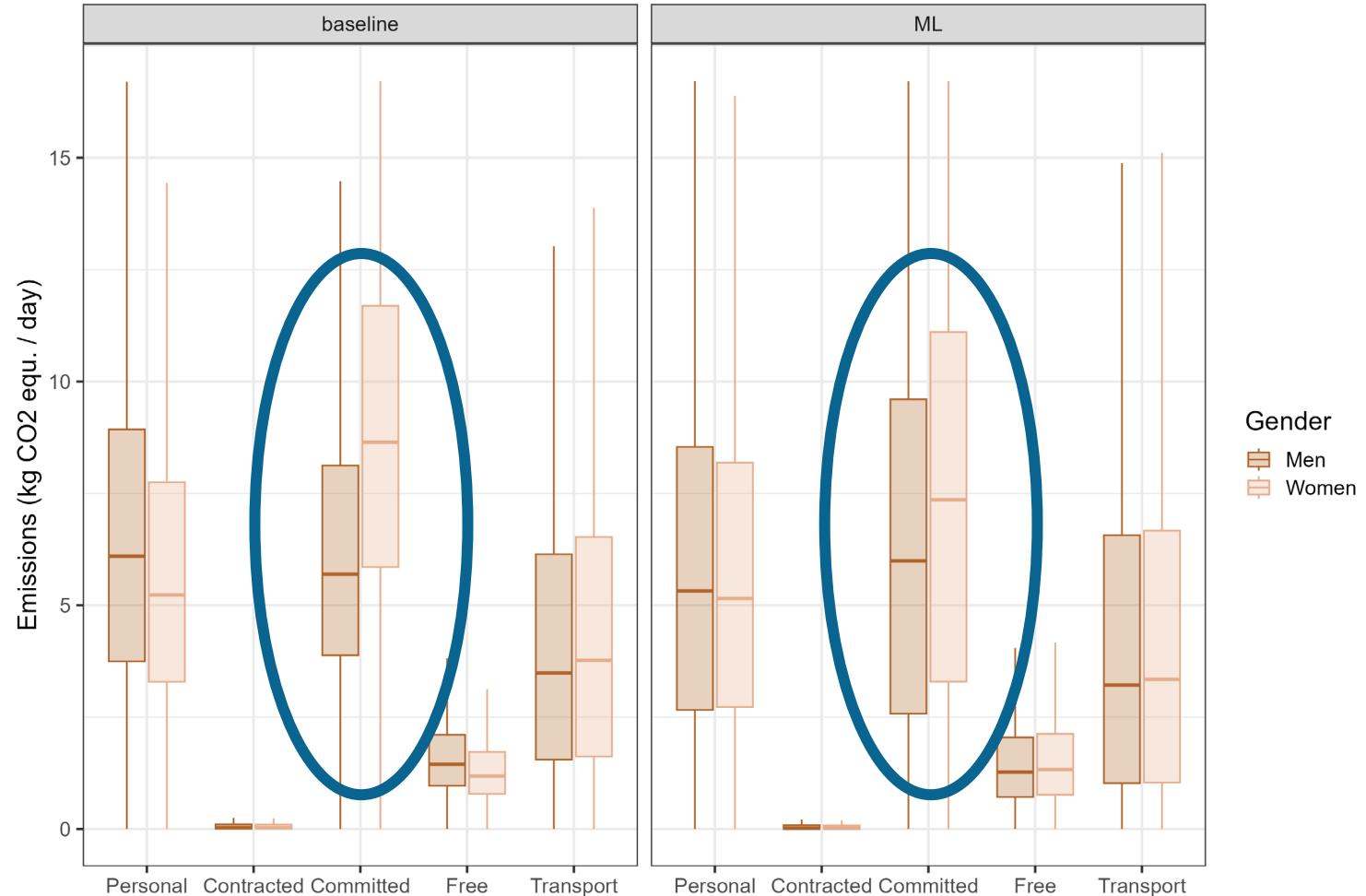
So, did we do
better?

Time-use predictions

Predicted Time-Use



Individual Carbon Emissions

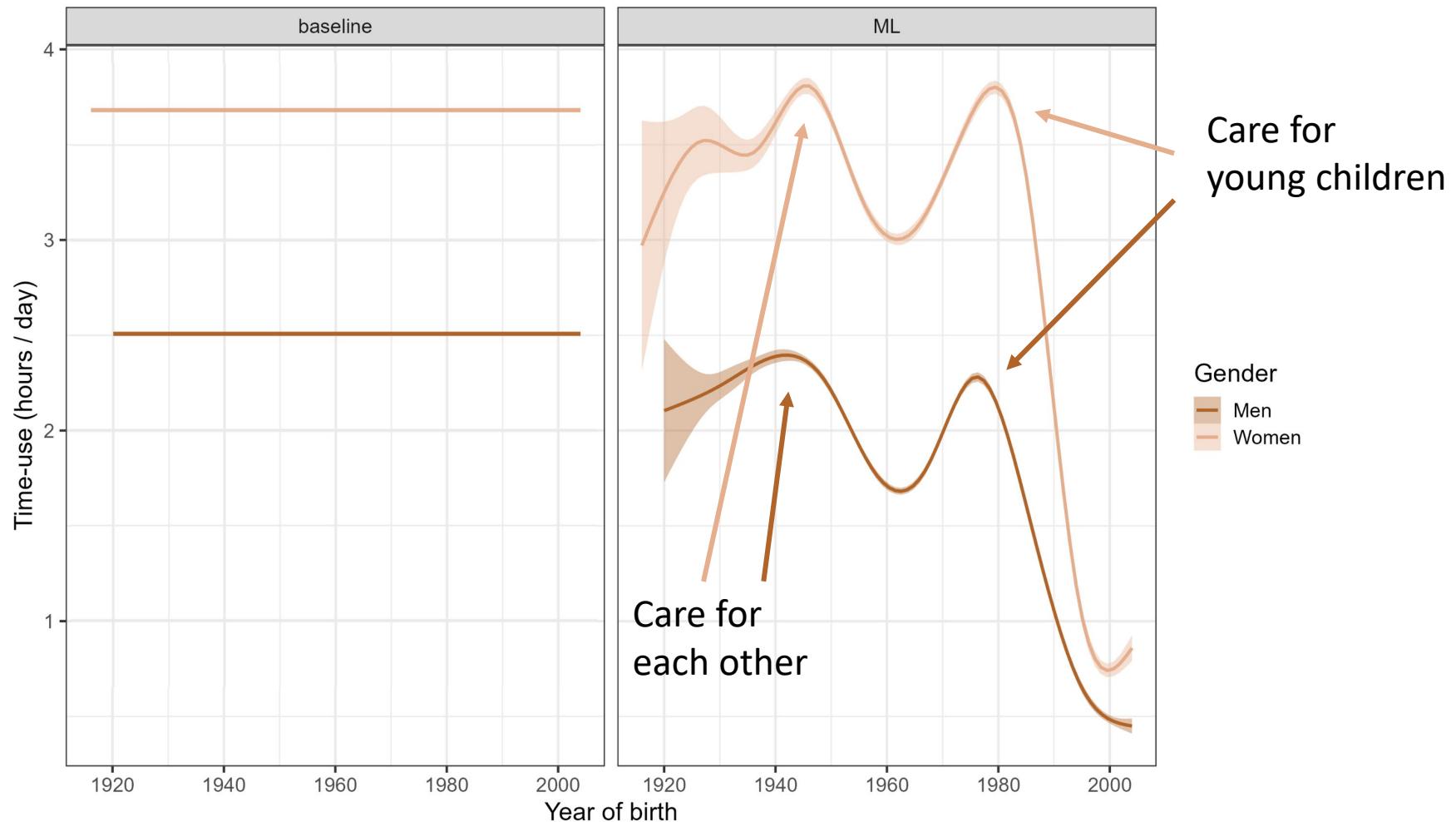


Implications

Committed time-use & emissions

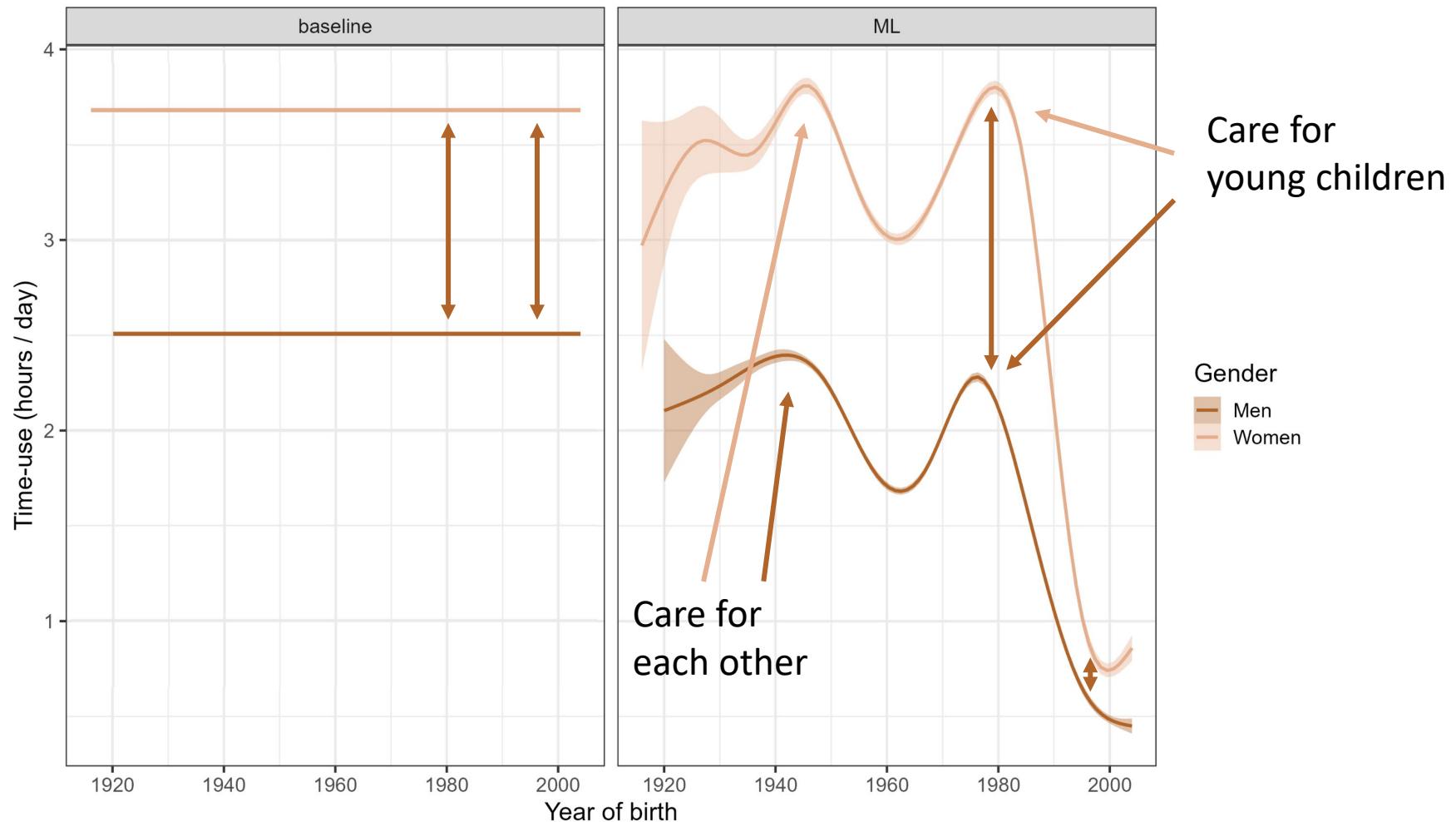
Committed Time-Use

Gender * Year of Birth



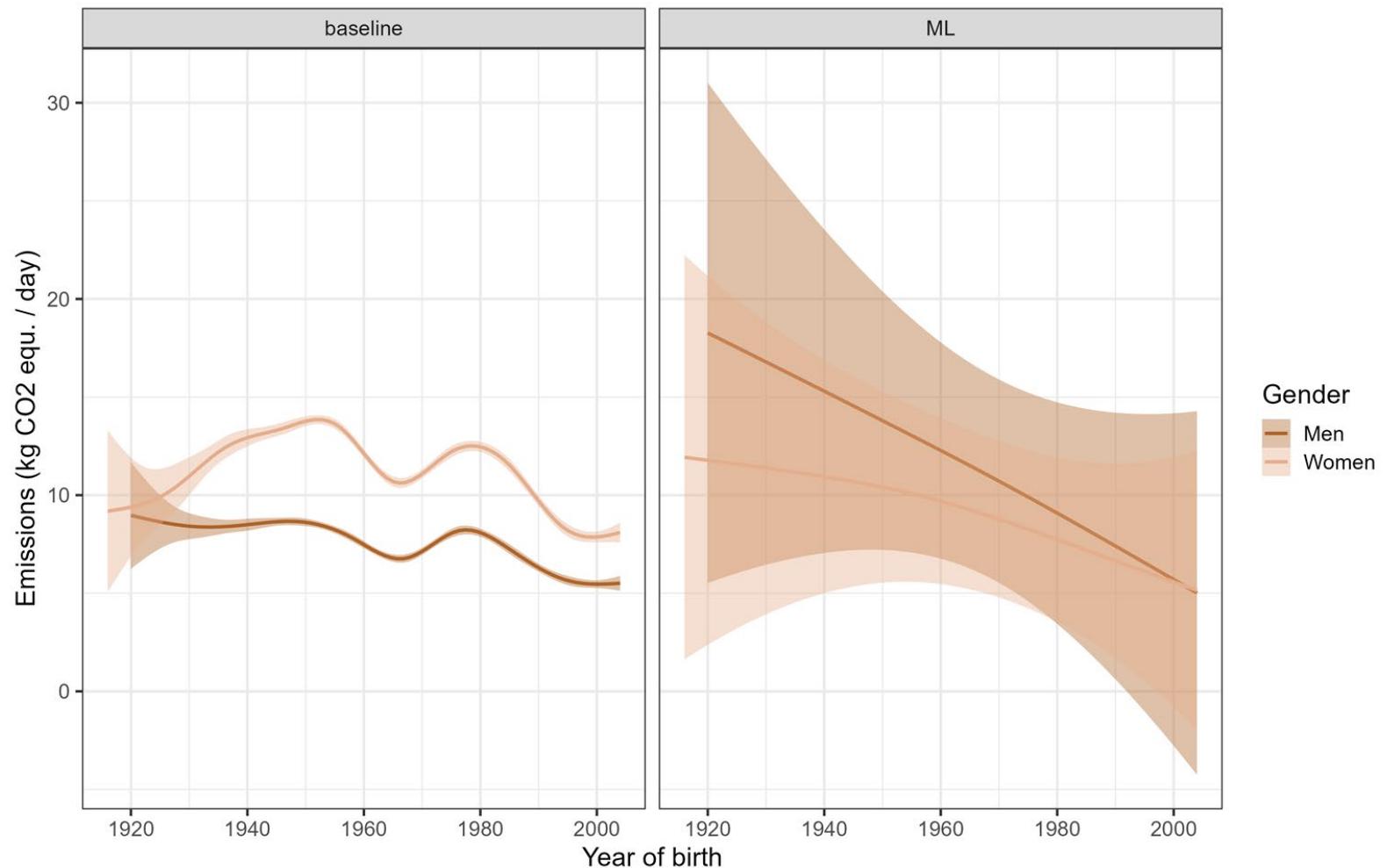
Committed Time-Use

Gender * Year of Birth

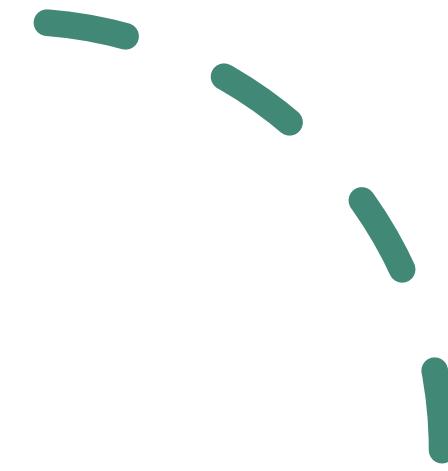
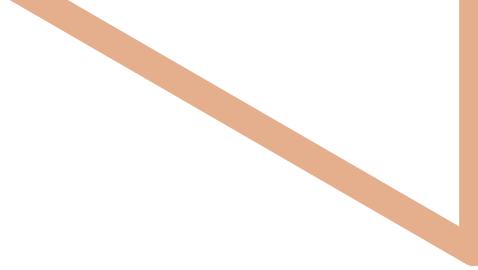
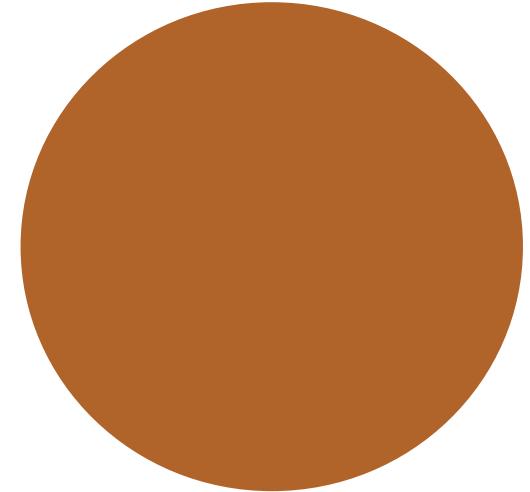


Committed Time-Use - Emissions

Gender * Year of Birth

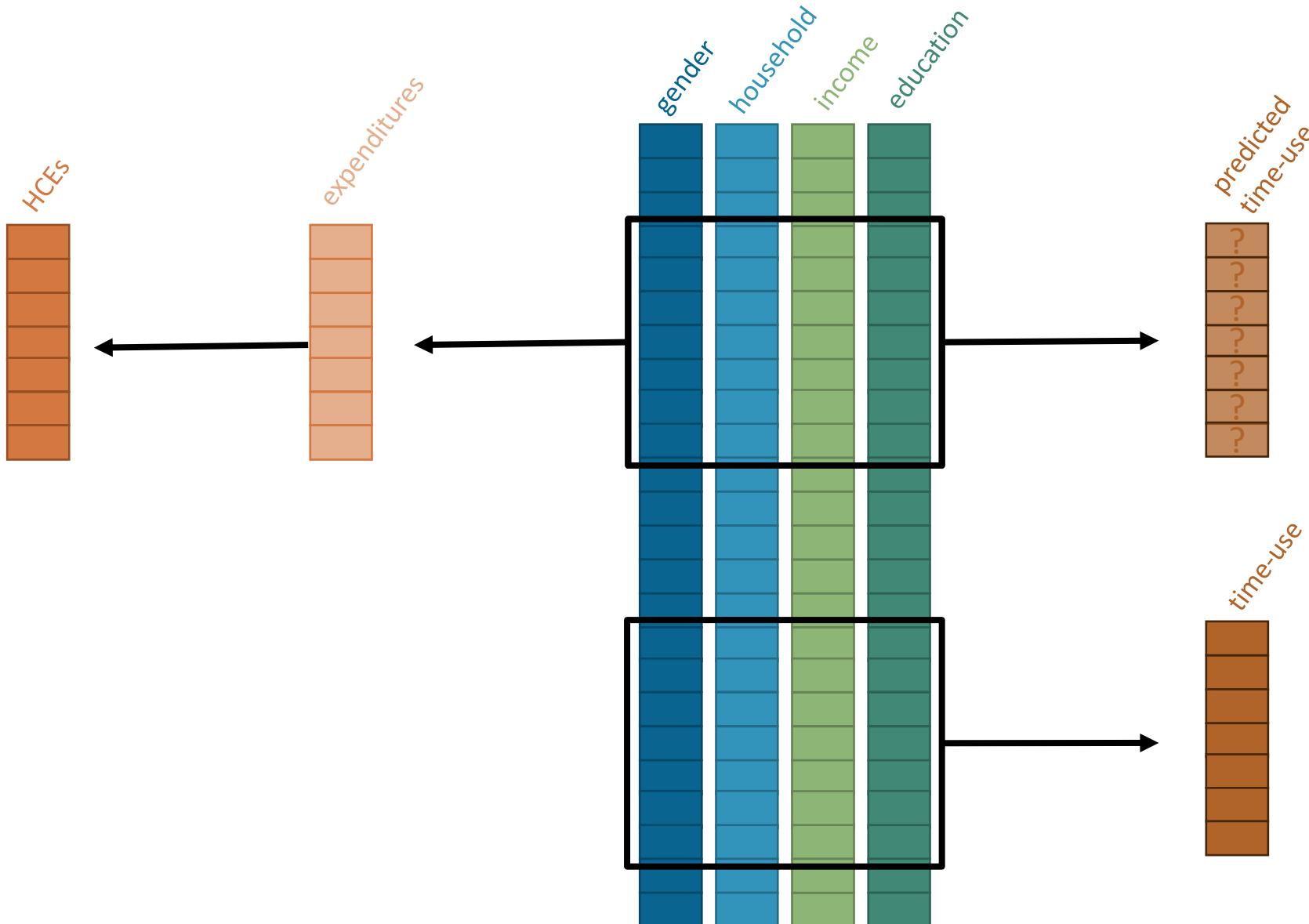


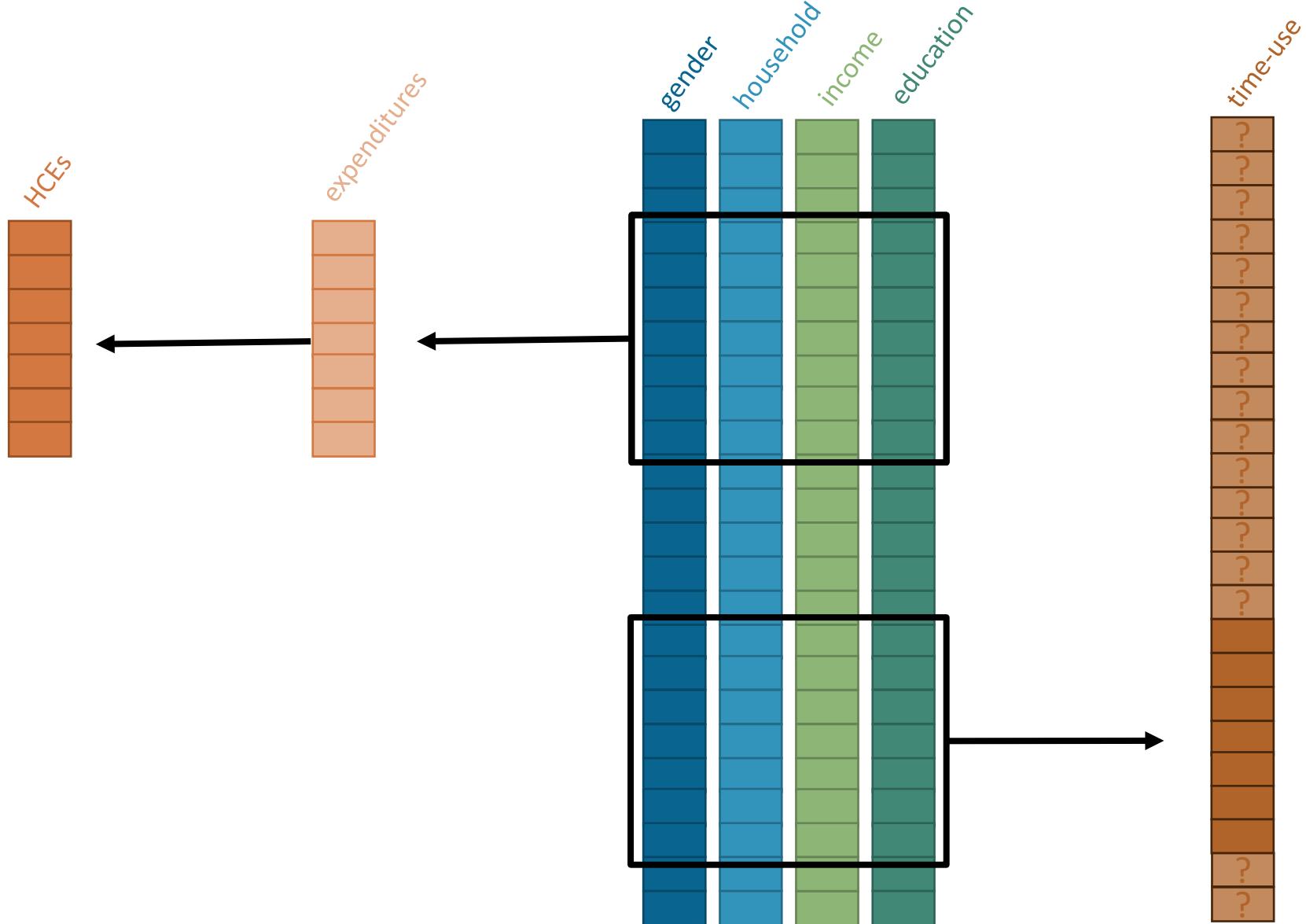
Next Steps

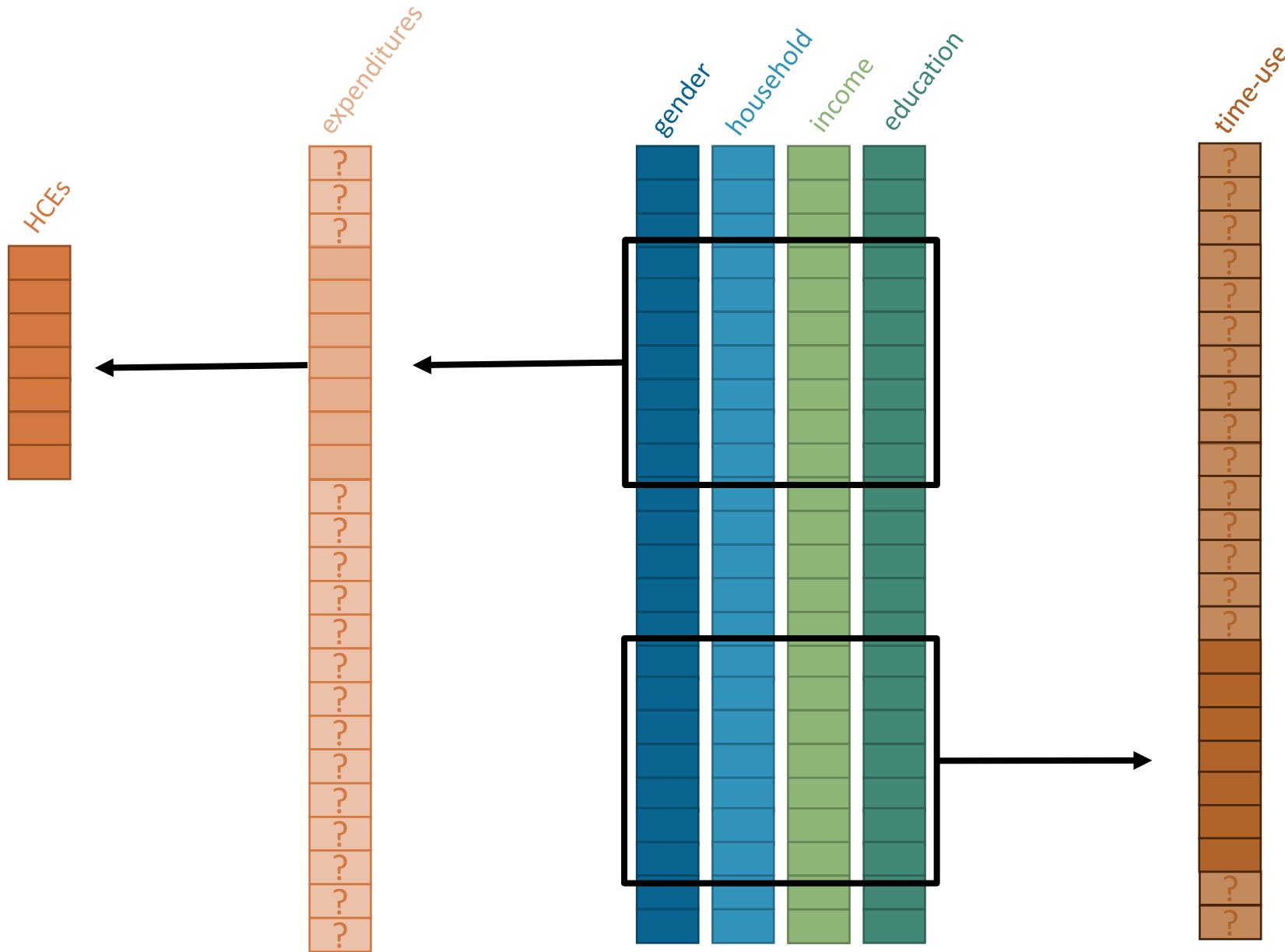


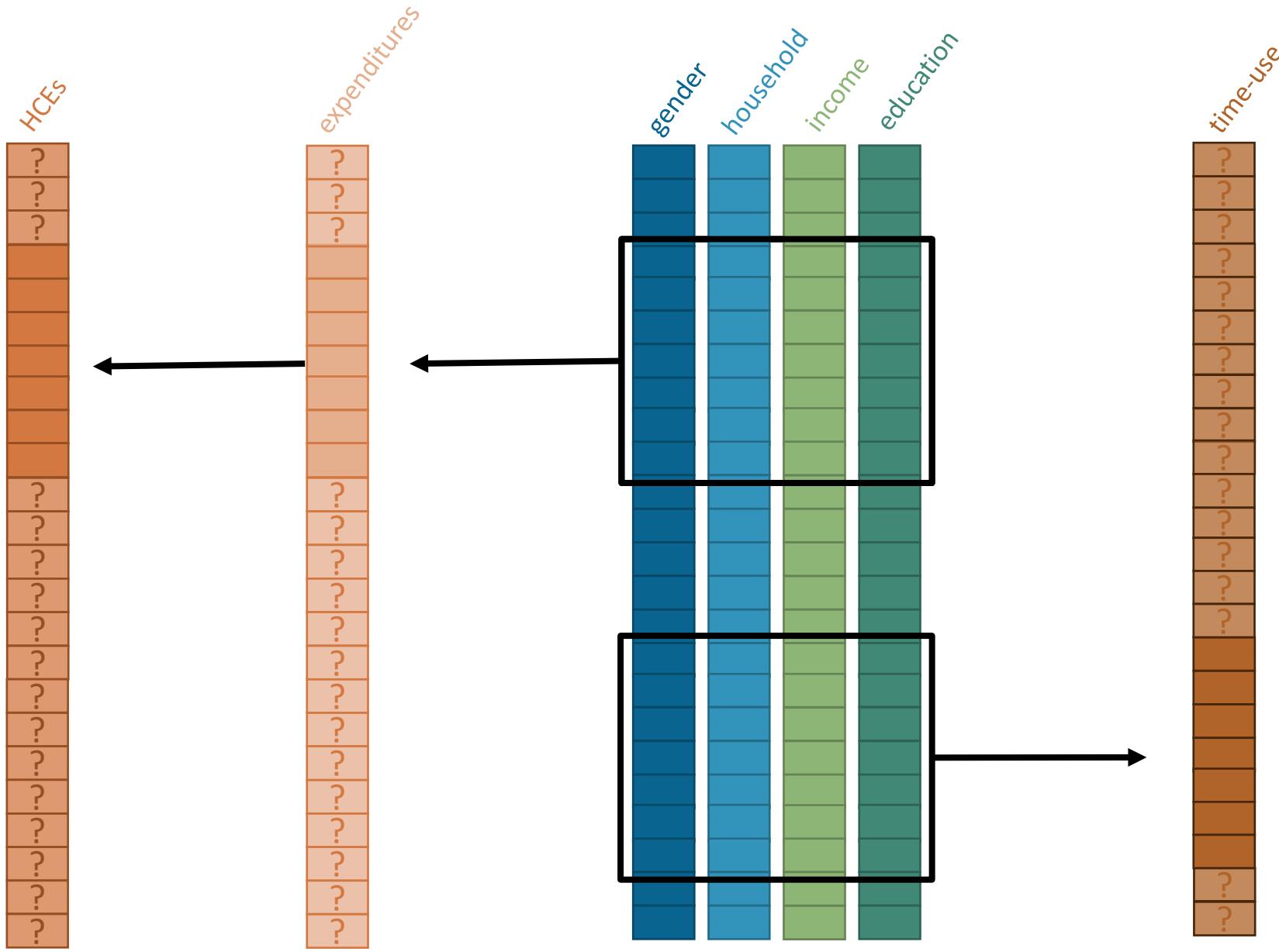
Next Steps

Calculate population-level individual carbon emissions









Next Steps

Estimate & account for uncertainty in predictions

Conclusions

- Imputing the average time-use of men & women **ignores** many **underlying household dynamics**
- Simple ML techniques can **improve the time-use predictions** by a lot
- This allows us to study time-use & individual carbon emissions on a much **larger (population) level**
- This technique can be applied to many **other contexts** & is relatively **easy to implement**

Combining Register and Survey Data to Understand Carbon-Intensive Time Use

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Erik-Jan van Kesteren, Javier Garcia Bernardo, Qixiang Fang, Jiamin Ou

m.l.v.weiper@uu.nl

https://github.com/sodascience/time_emissions

Thank you
for listening

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- Results based on calculations by Utrecht University using non-public microdata from Statistics Netherlands