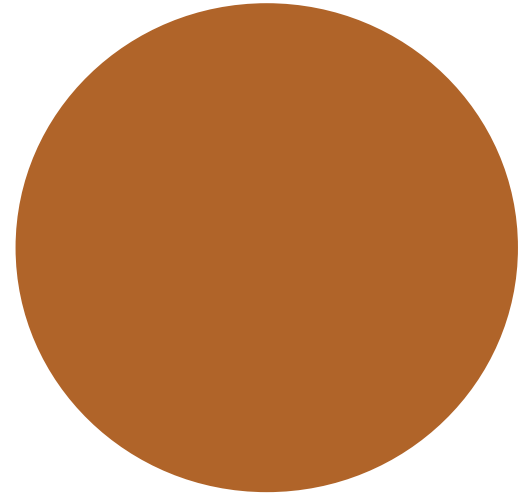


Combining Register and Survey Data to Understand Carbon-Intensive Time Use

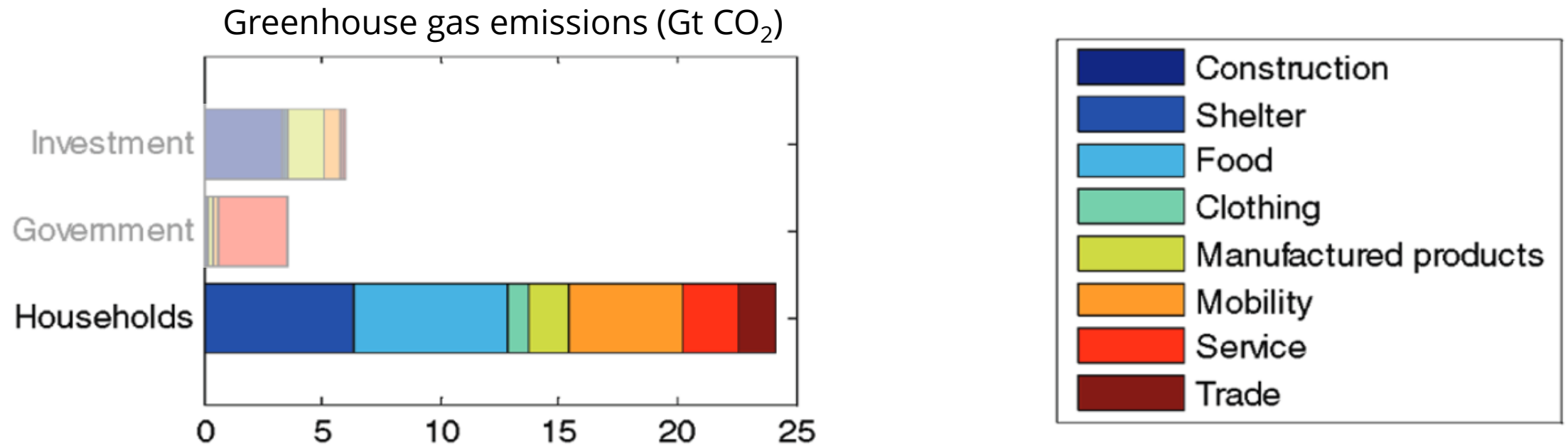
Maike Weiper

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

The Context



Consumption-Based Emissions



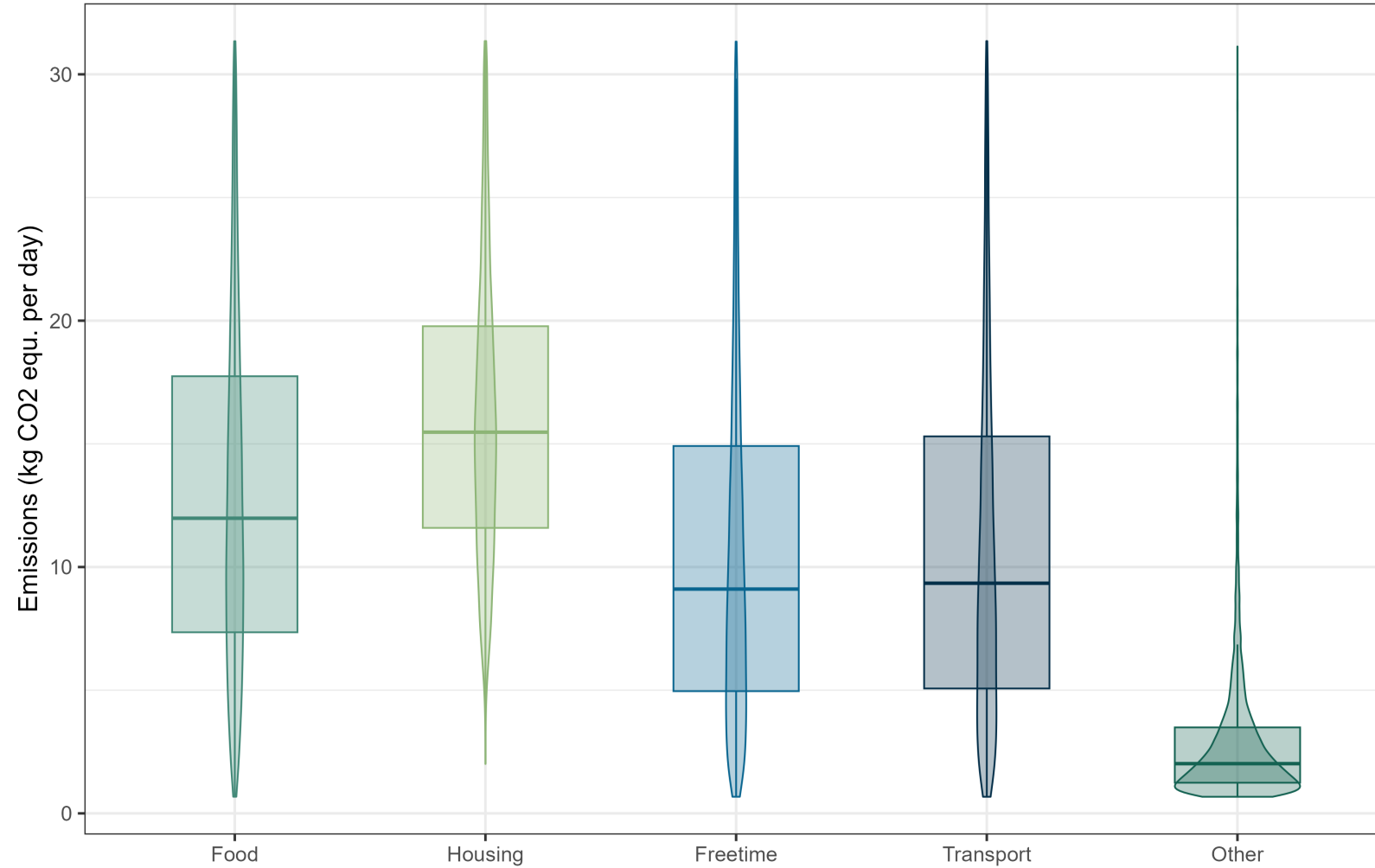
Research Questions

- 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

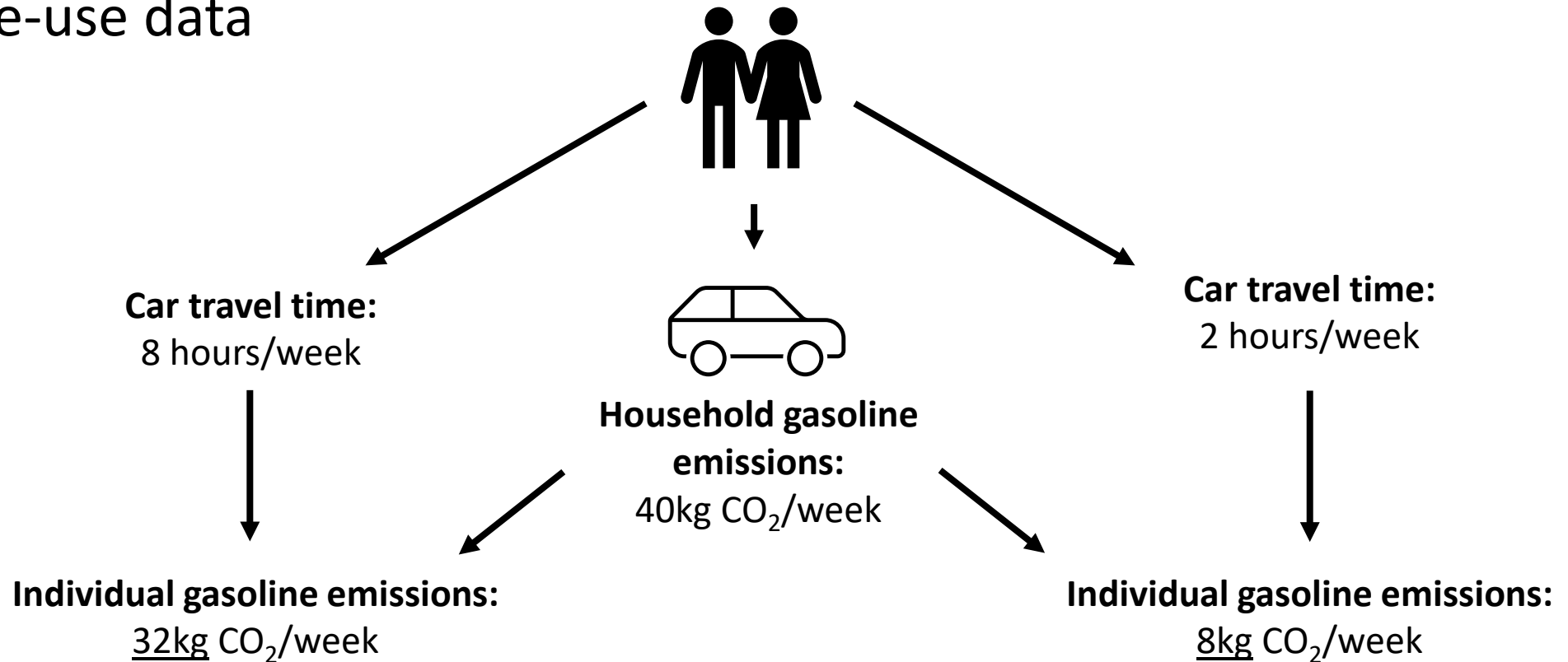


Research Questions

- 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

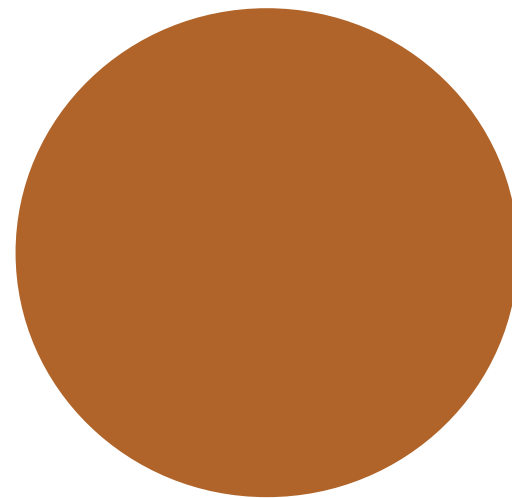
- Attribute household carbon emissions to individuals
- Time-use data

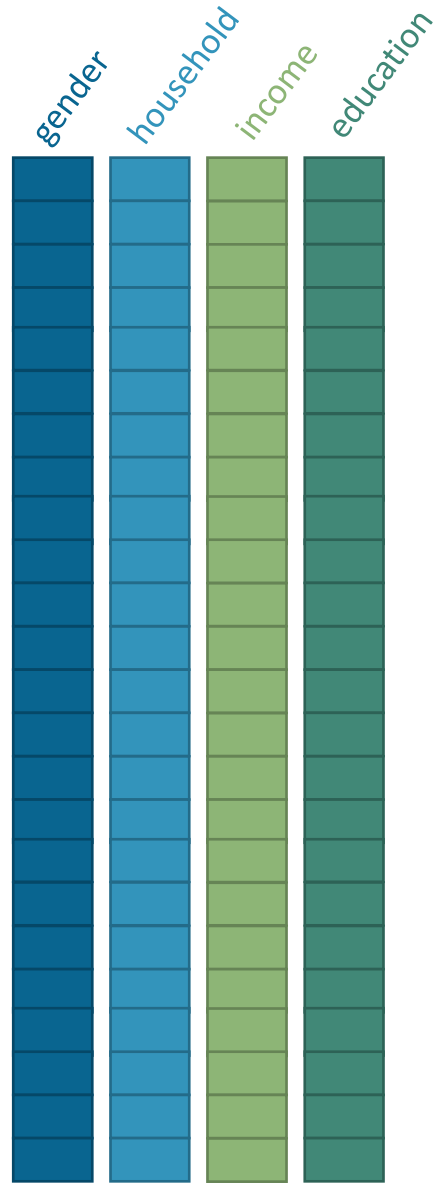


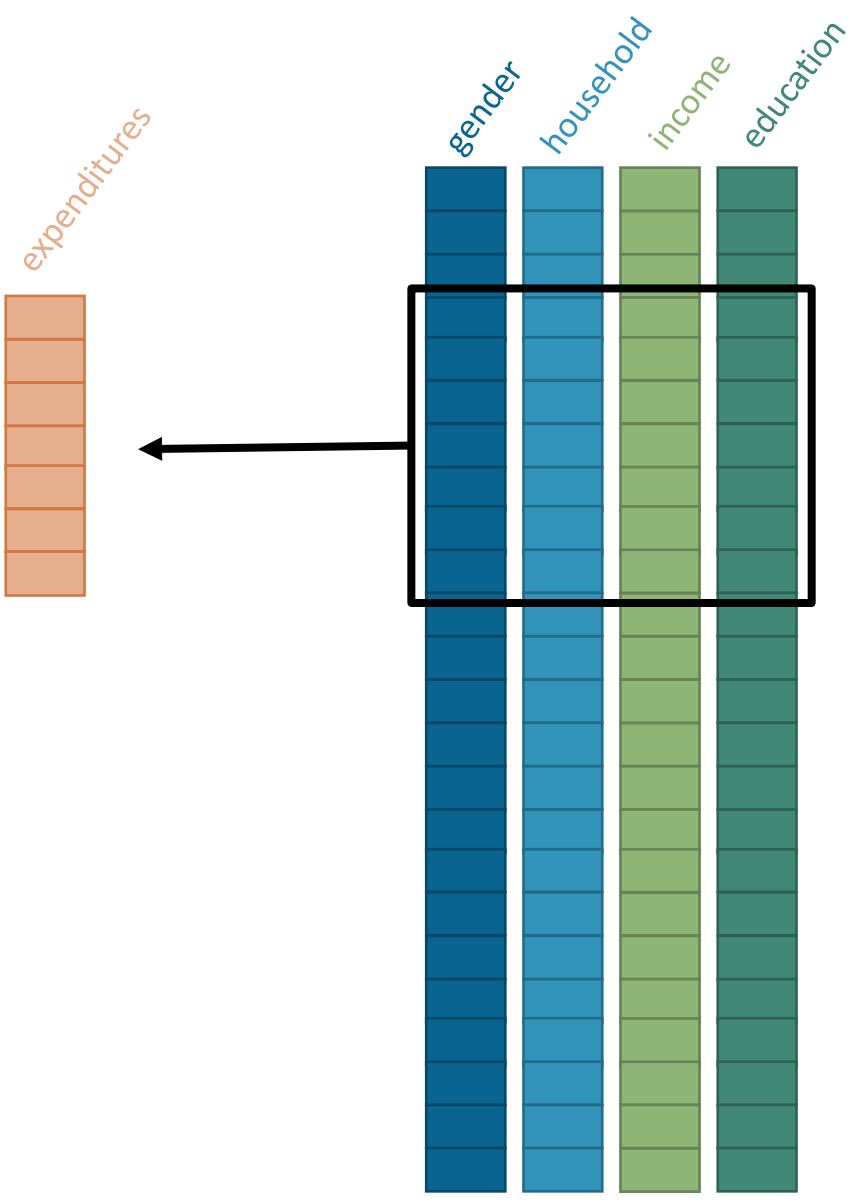
Individual Carbon Emissions

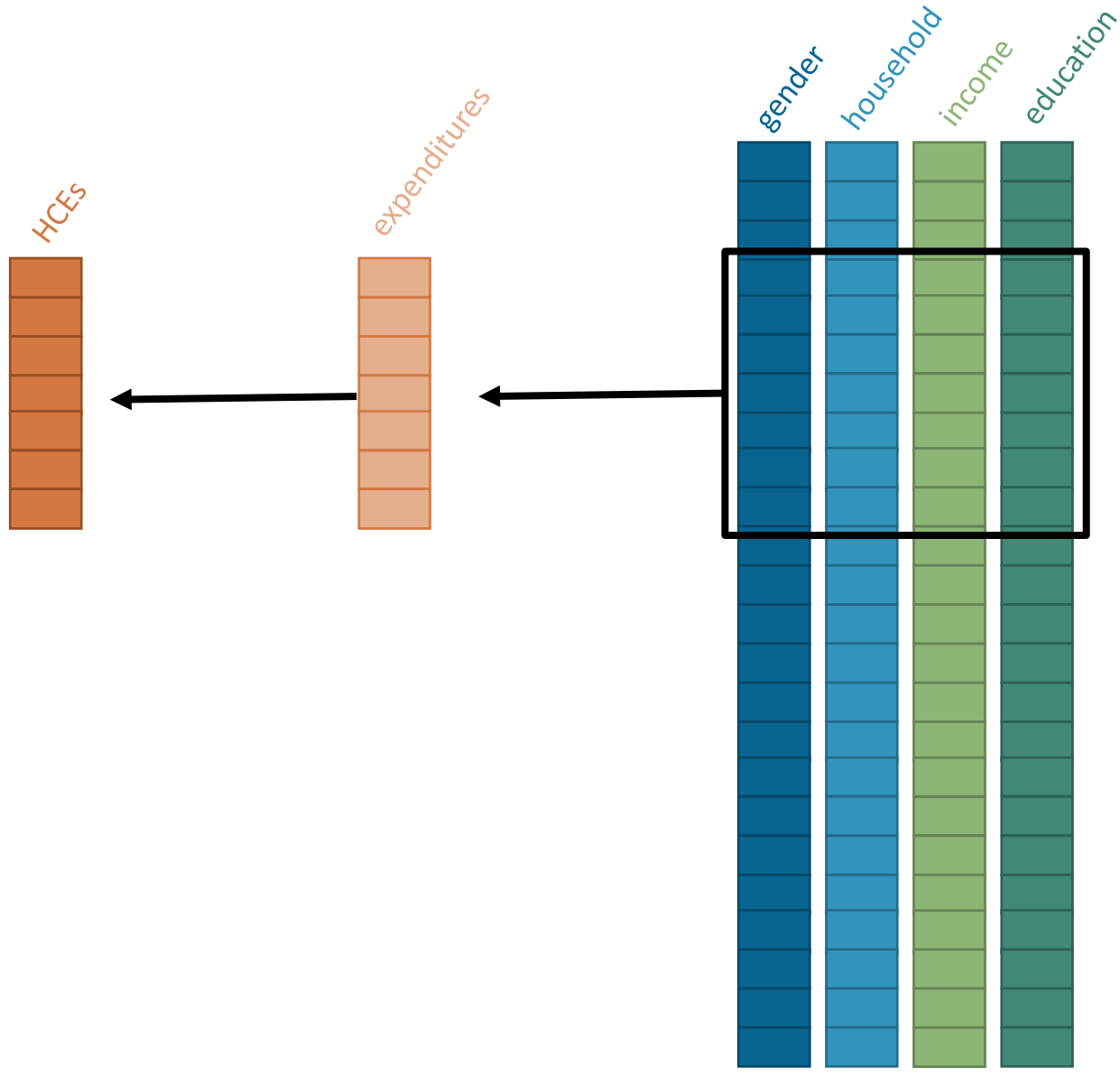
- Attribute household carbon emissions to individuals
- Time-use data
 - ~ 2300 individuals in 2016
 - 1 week of activities
 - 10 minute intervals

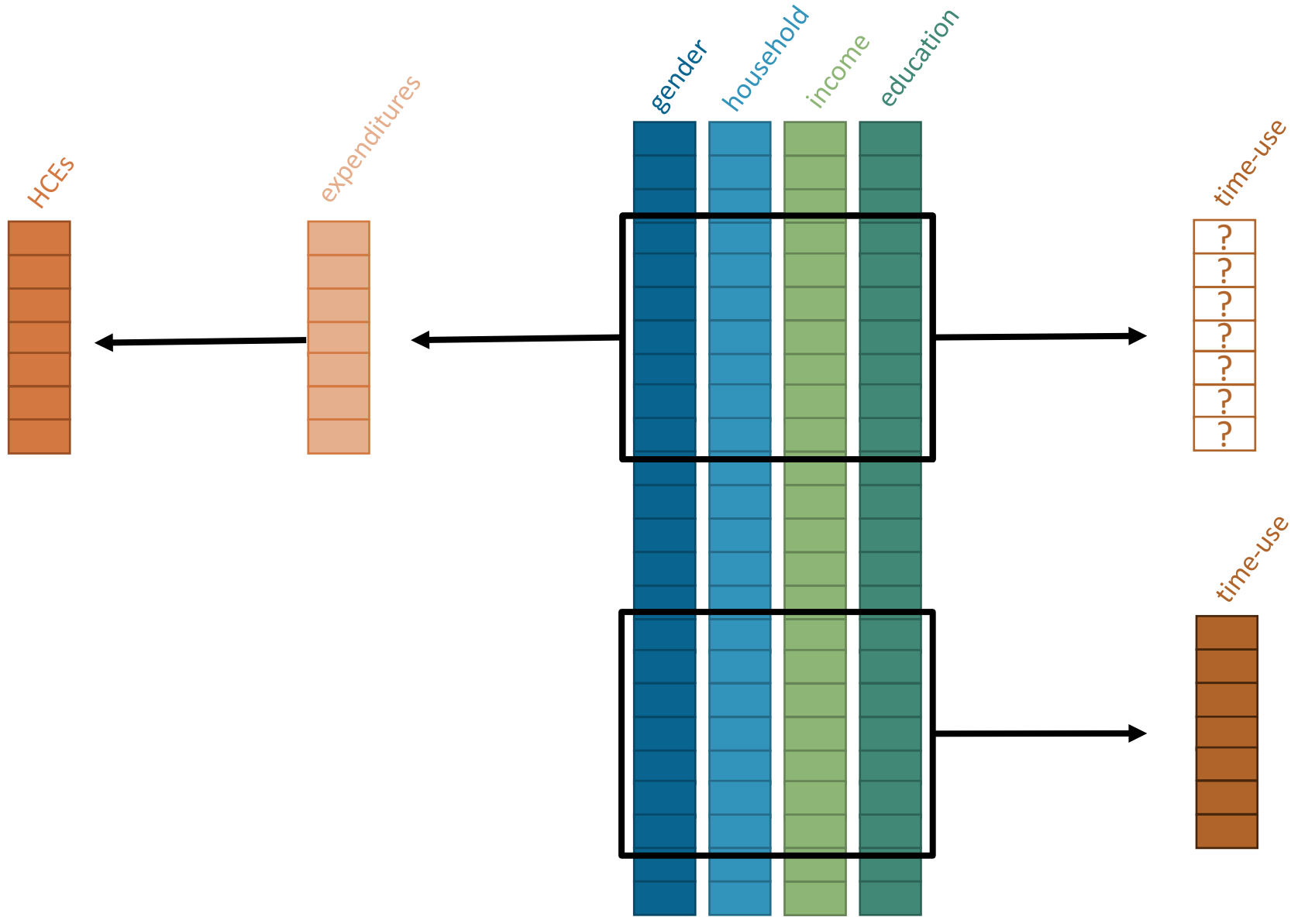
The Problem






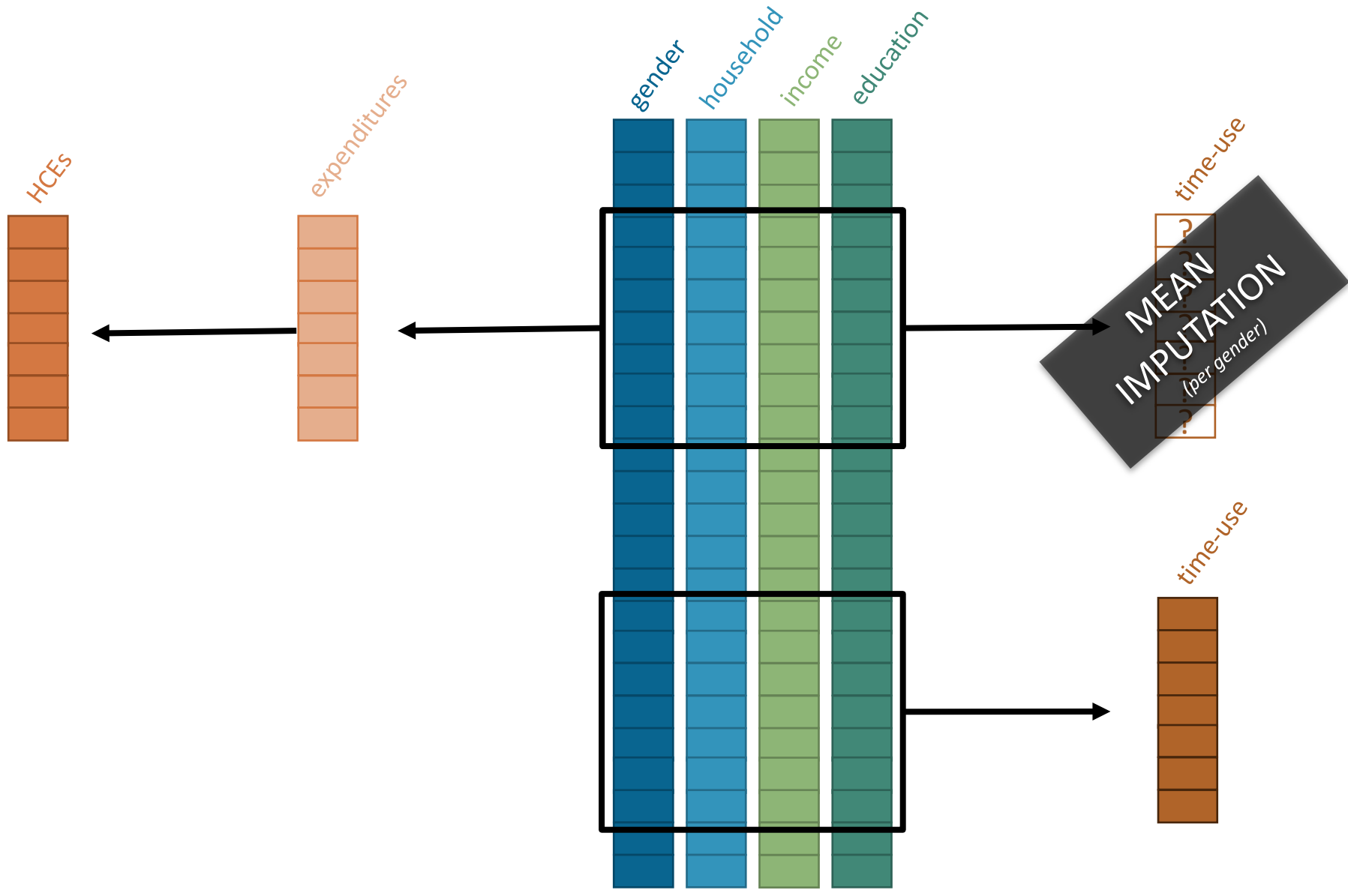


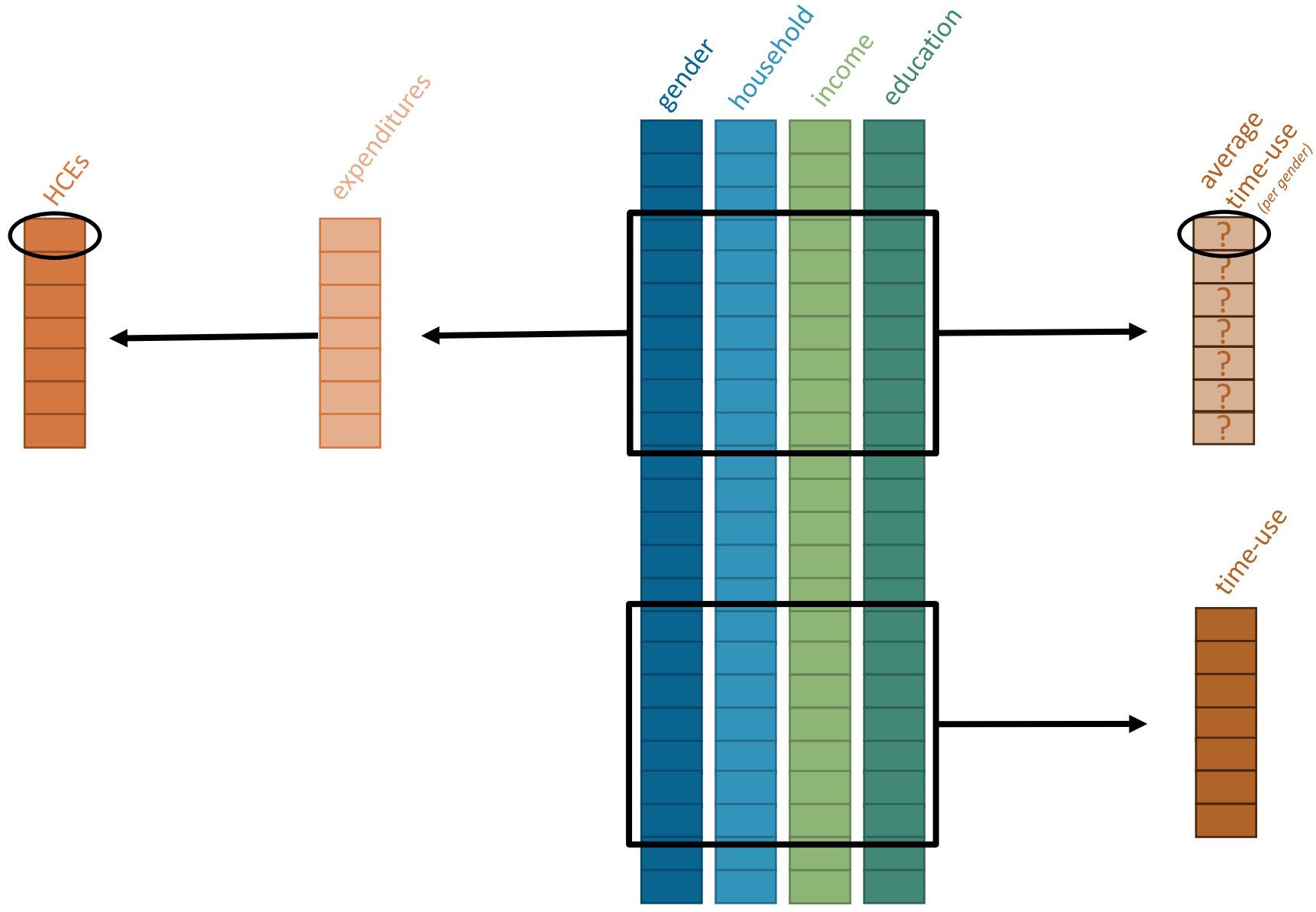





The background features several abstract geometric elements: a teal circle in the top left, an orange triangle in the top center, a large brown circle in the top right, a teal dashed line in the top right, an orange square outline in the middle right, and a teal dashed line in the bottom right.

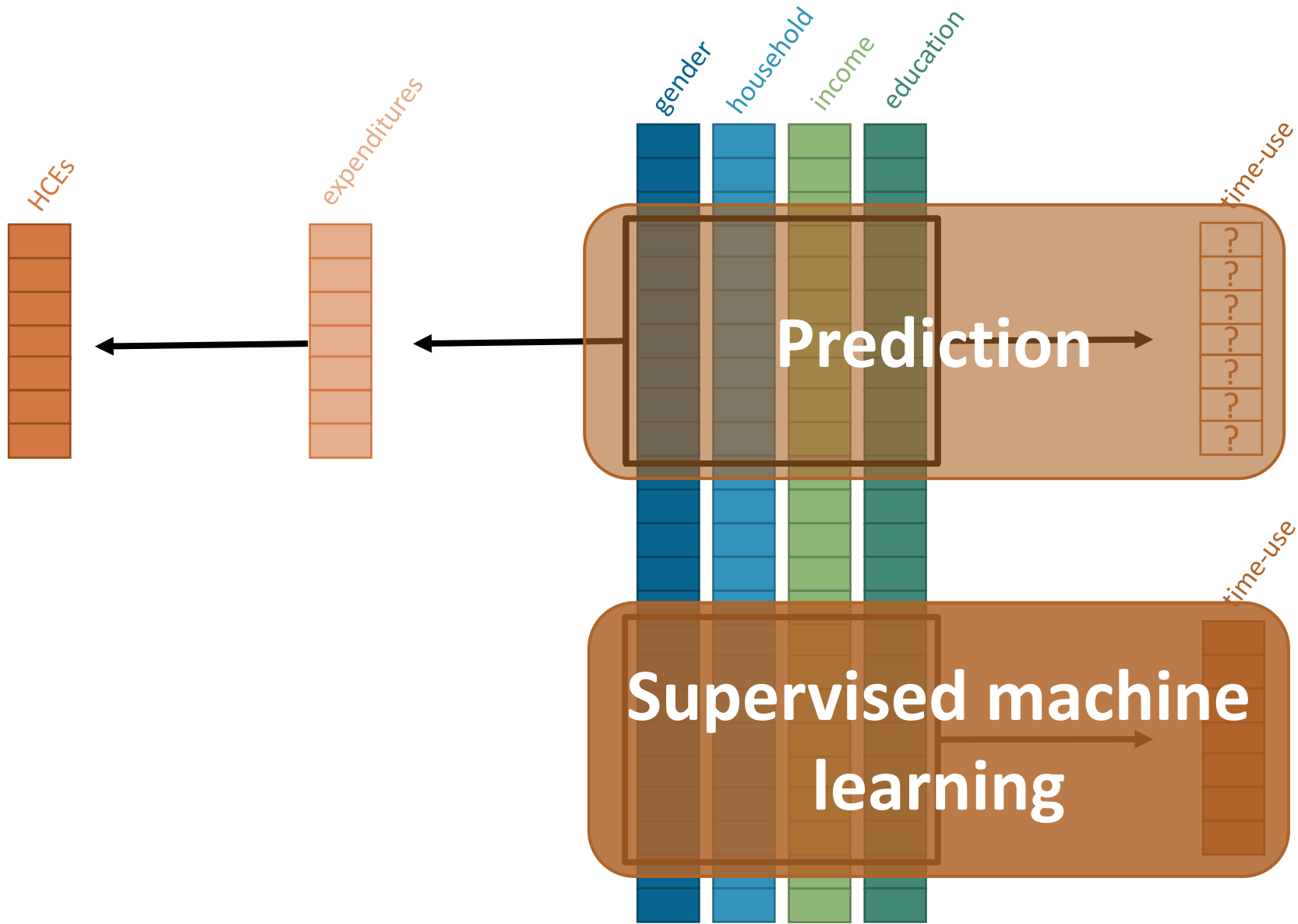
What others have
done...

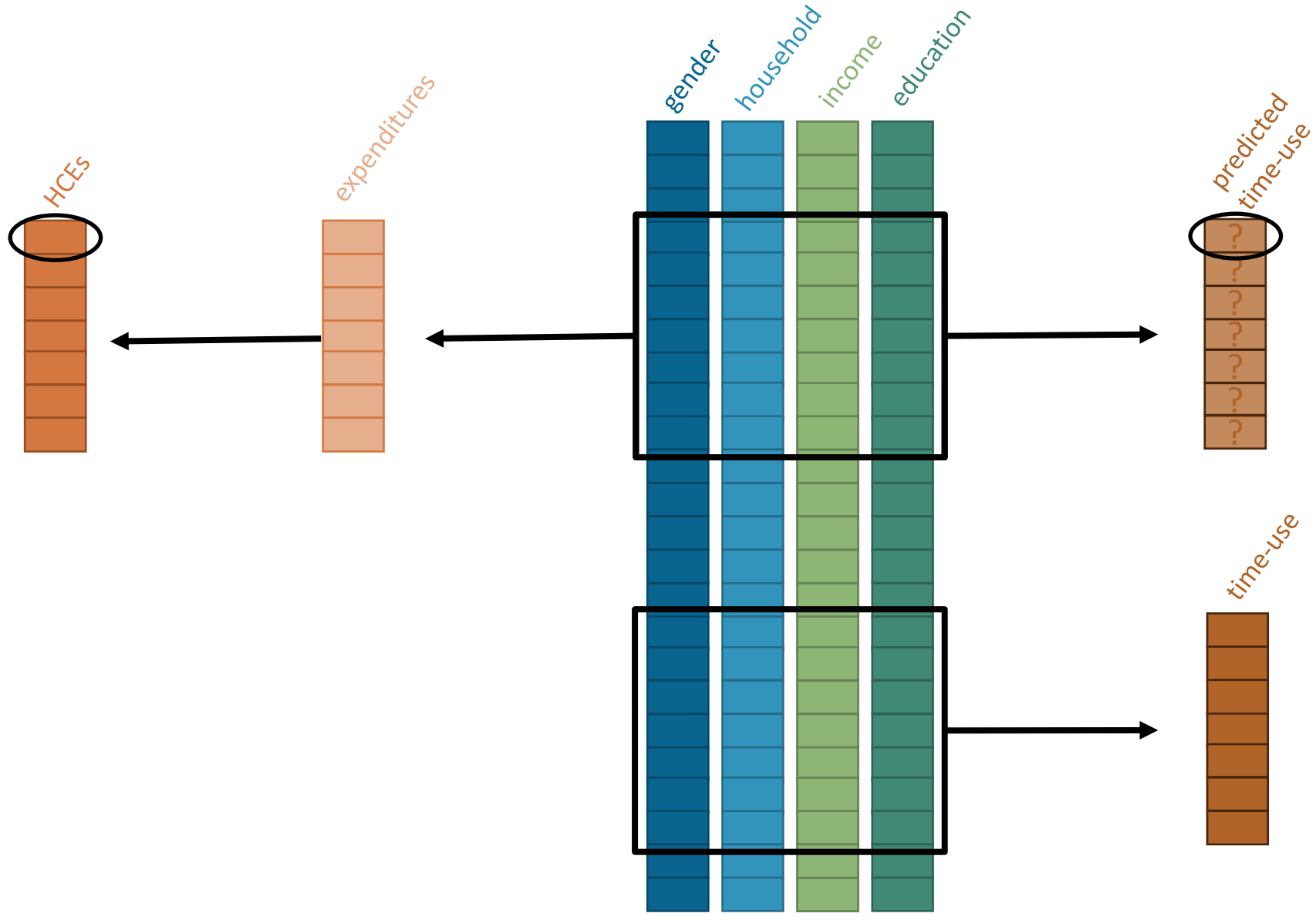




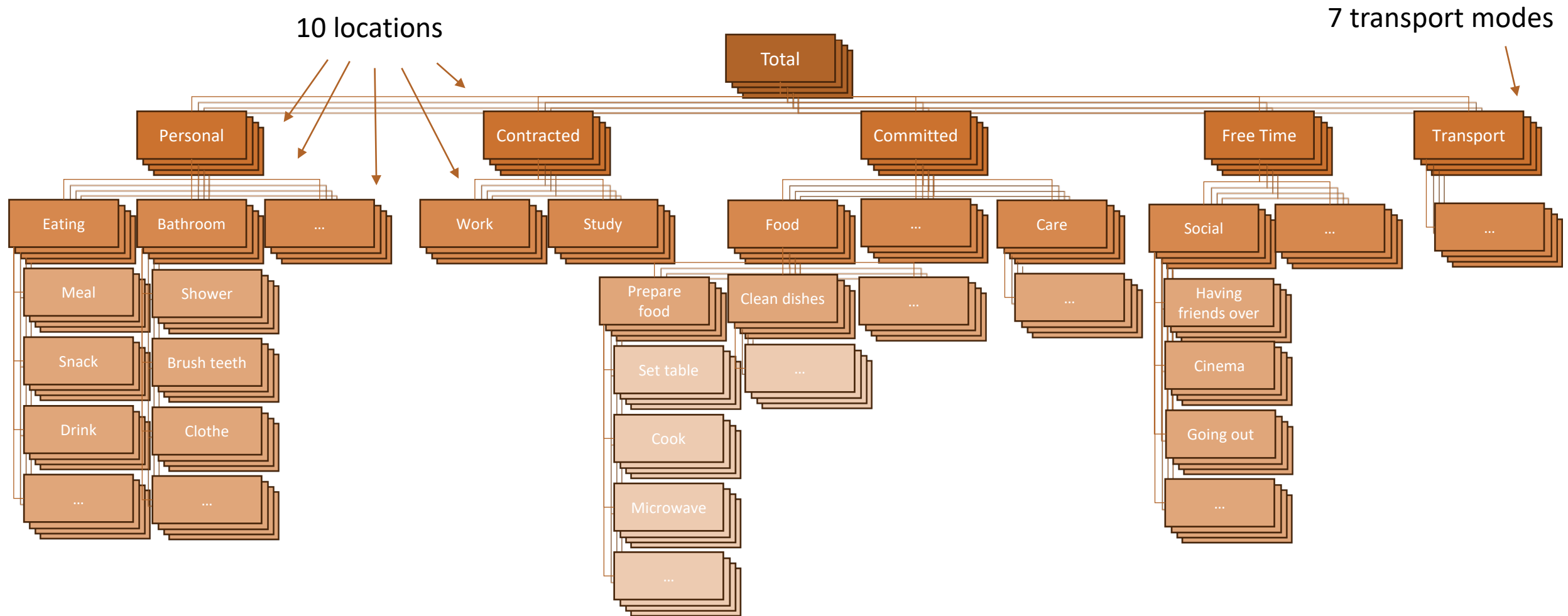


We think we can
do better!



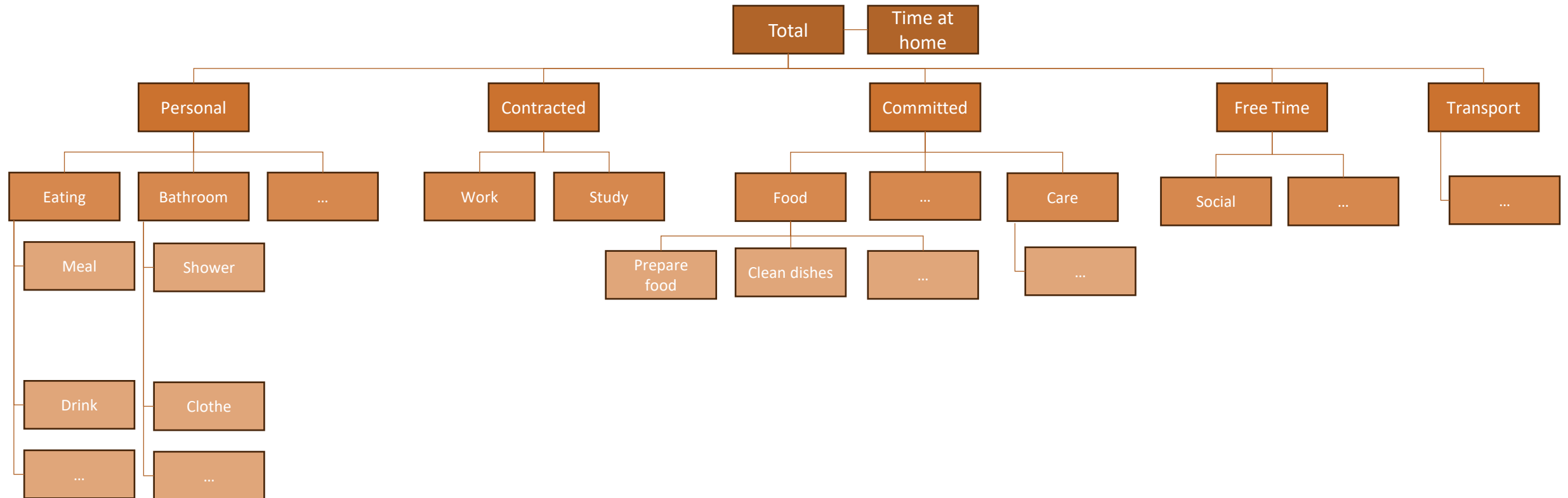


Predicting Time-Use



= 3323 unique activities

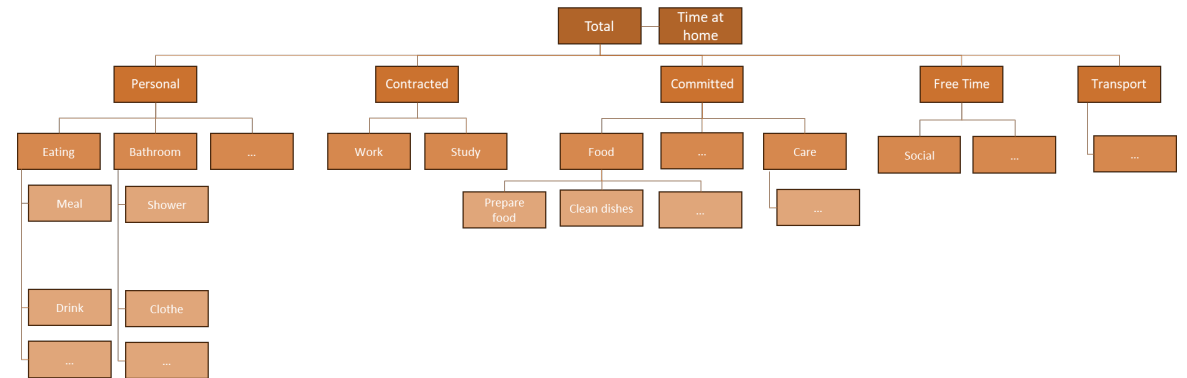
Predicting Time-Use

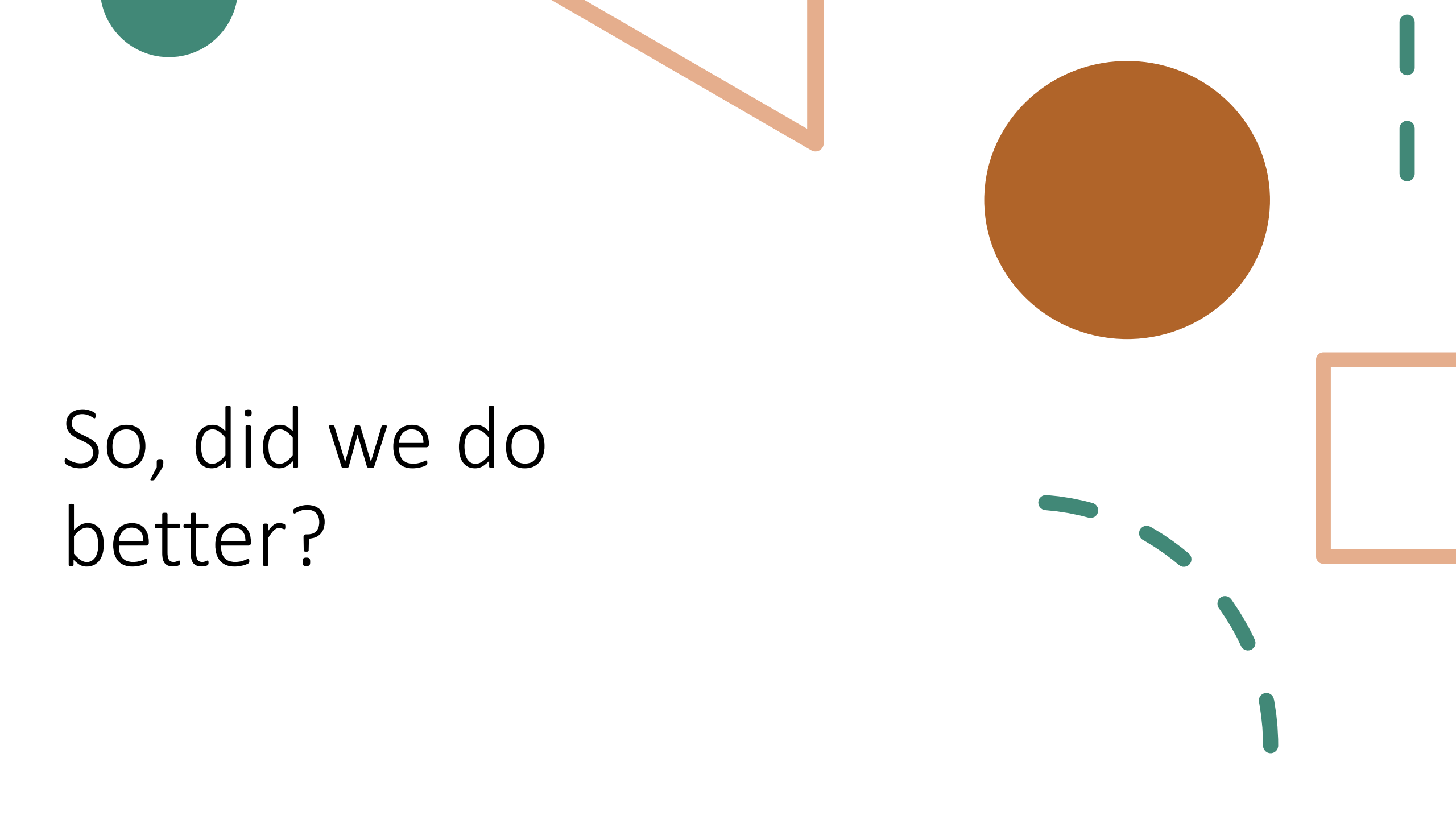


= 114 unique activities

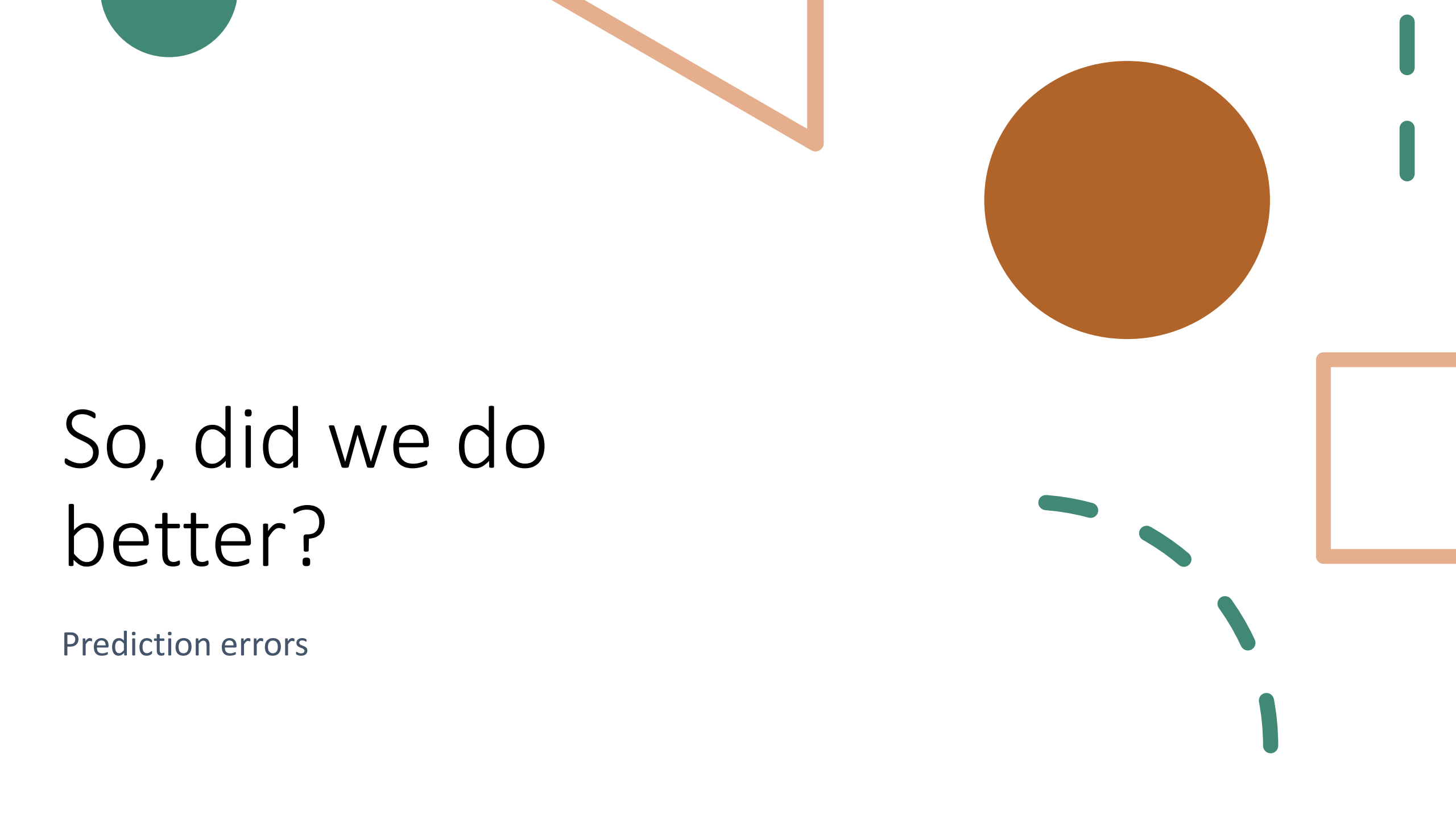
Predicting Time-Use

- 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





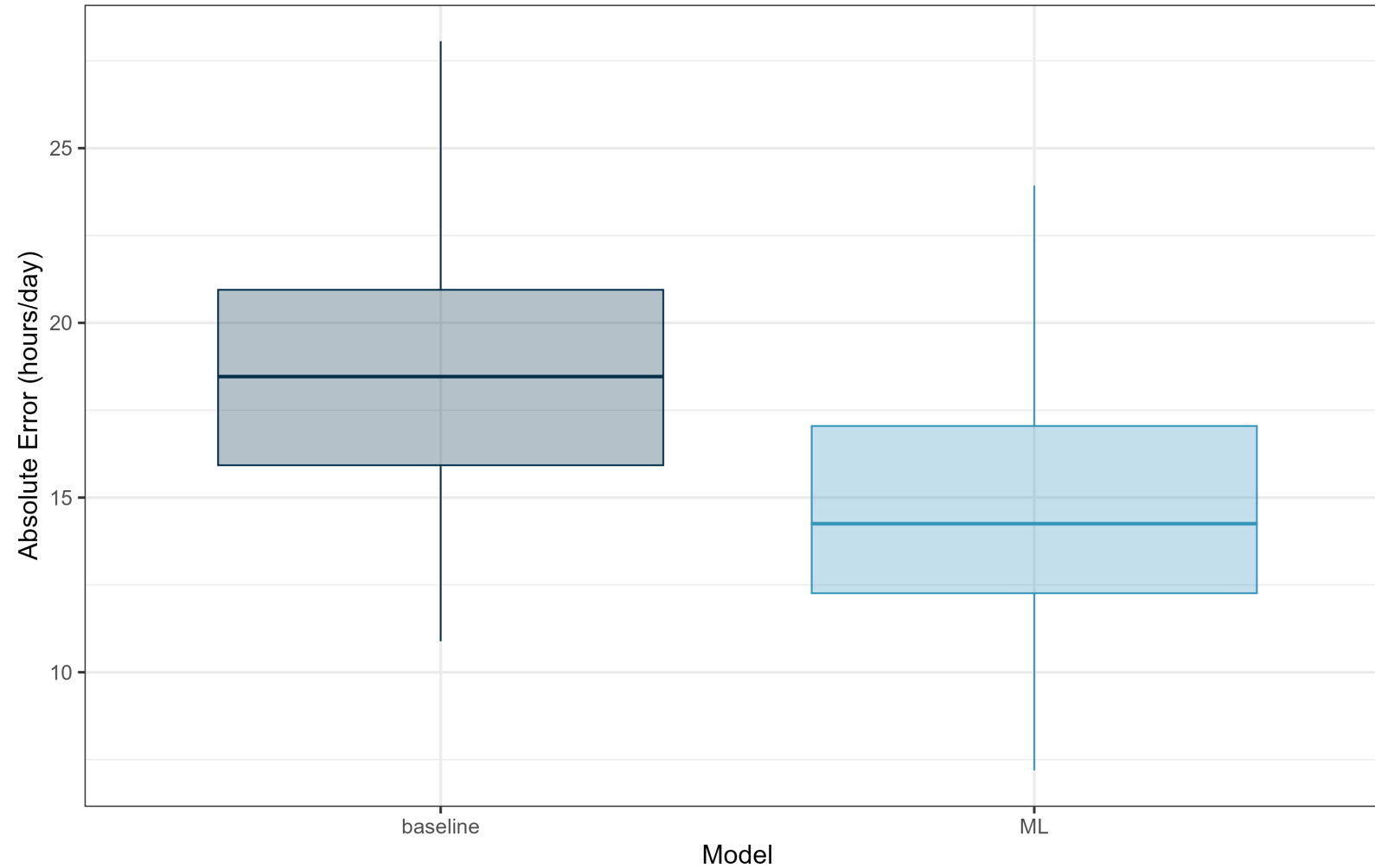
So, did we do
better?

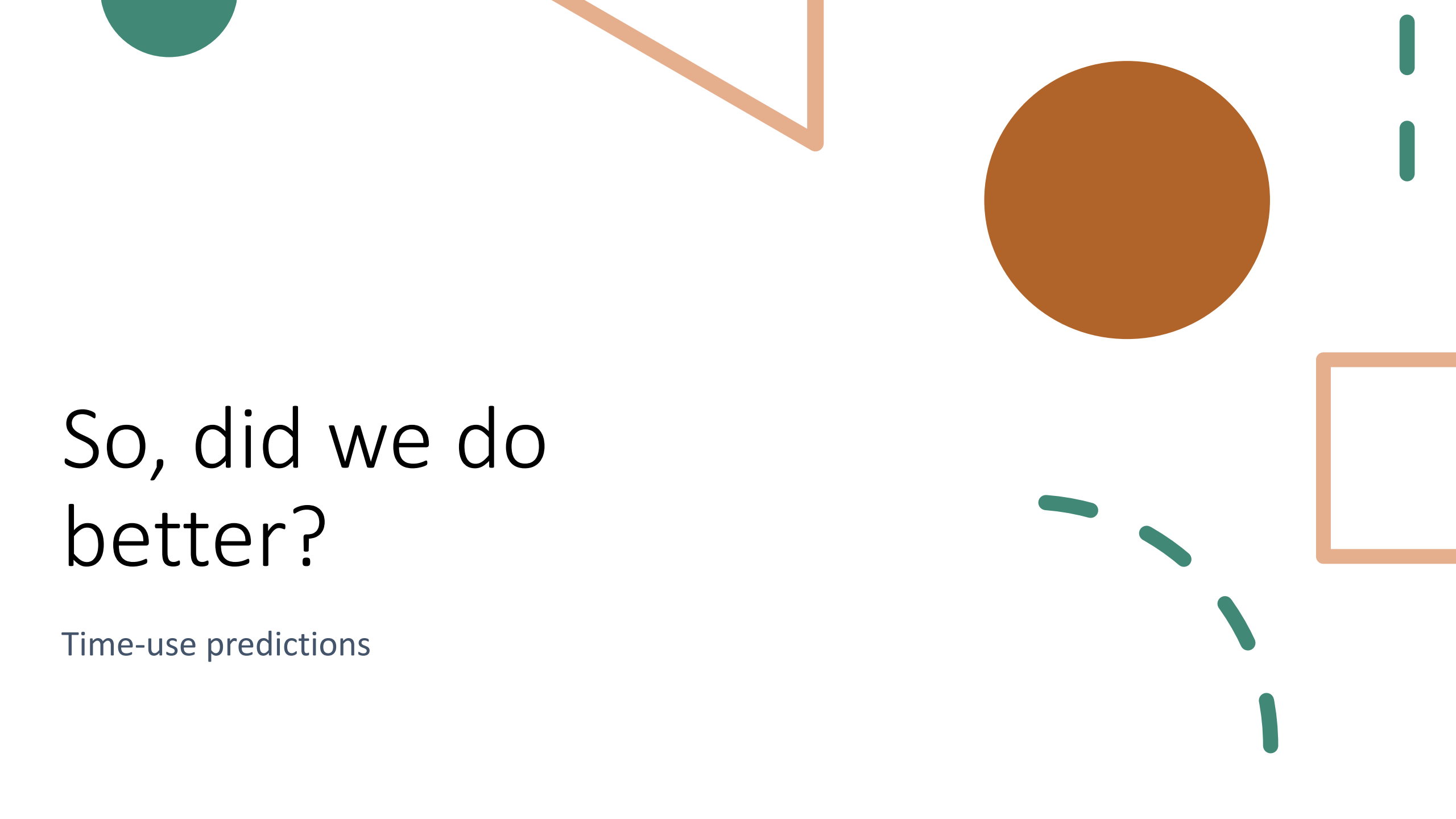


So, did we do
better?

Prediction errors

Predicted Time-Use (errors)

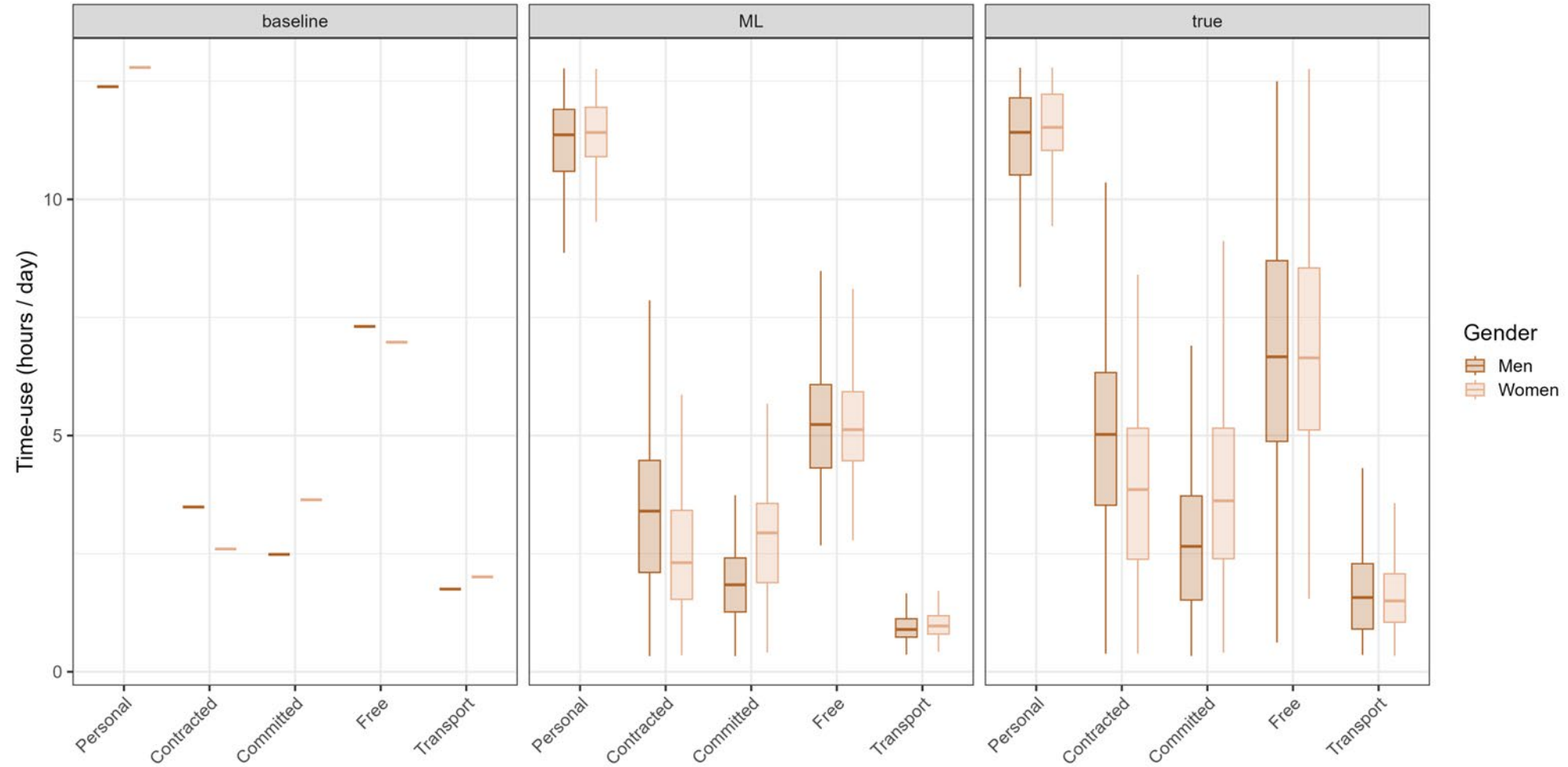




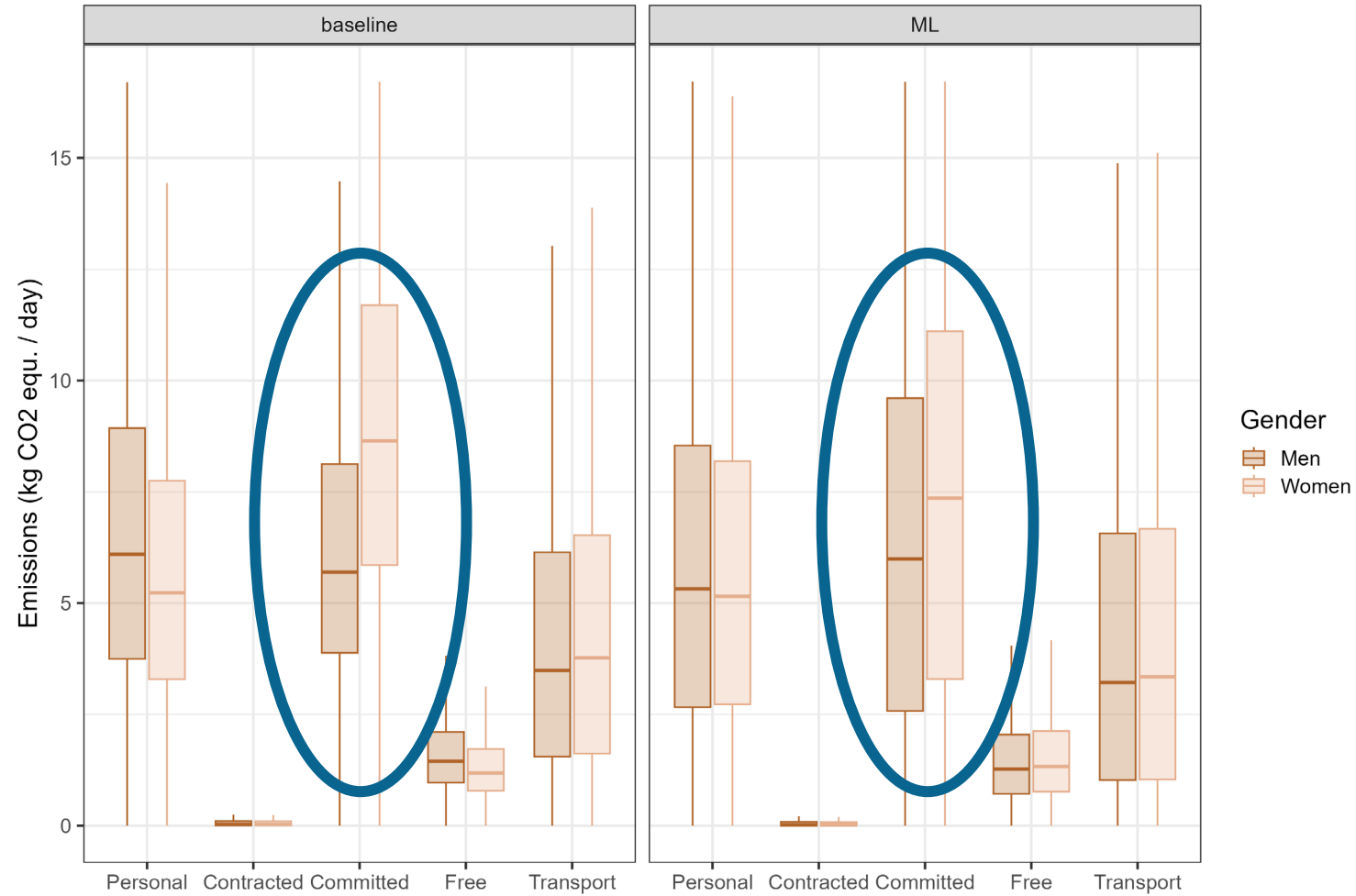
So, did we do
better?

Time-use predictions

Predicted Time-Use



Individual Carbon Emissions



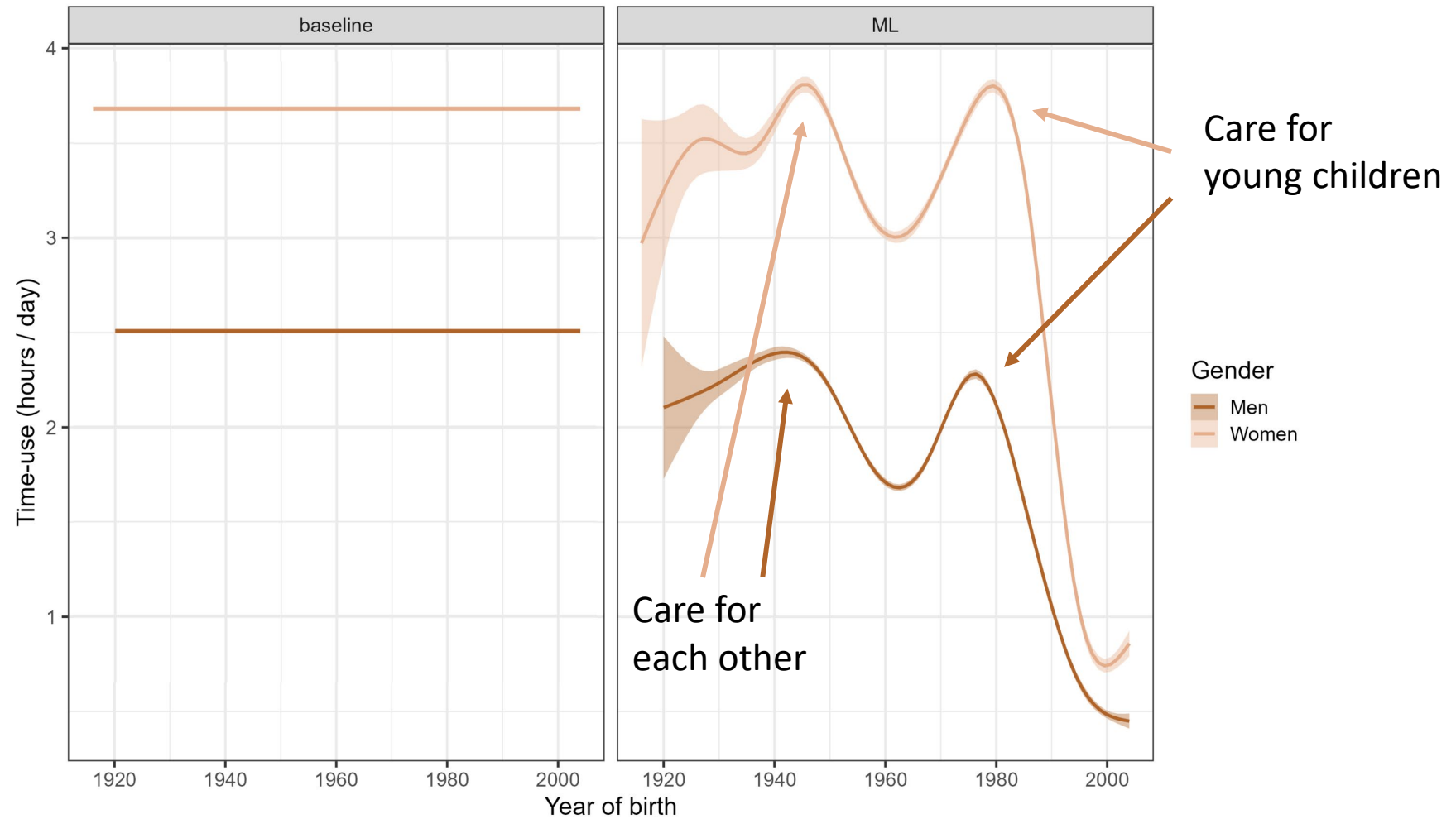


Implications

Committed time-use & emissions

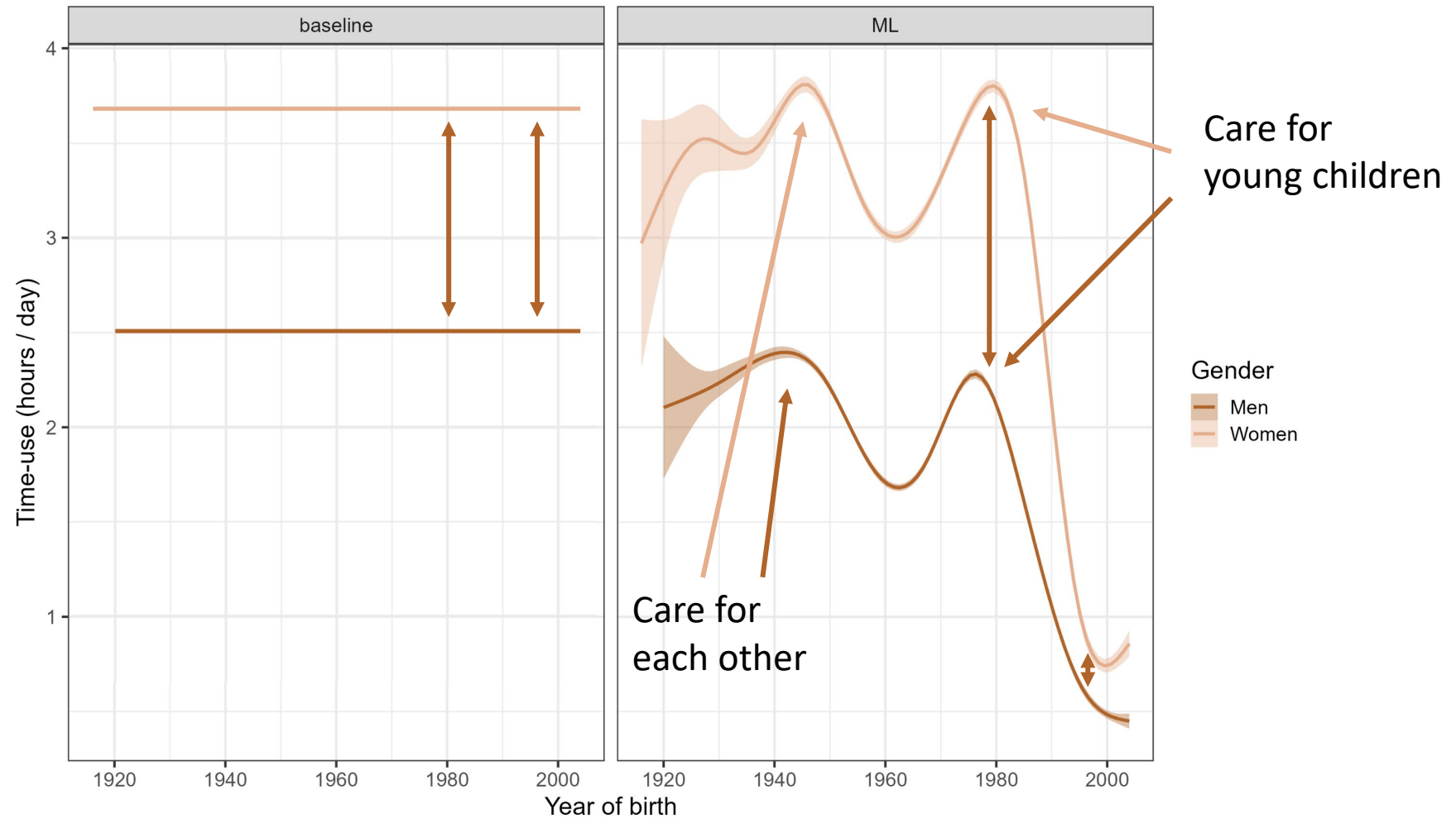
Committed Time-Use

Gender * Year of Birth



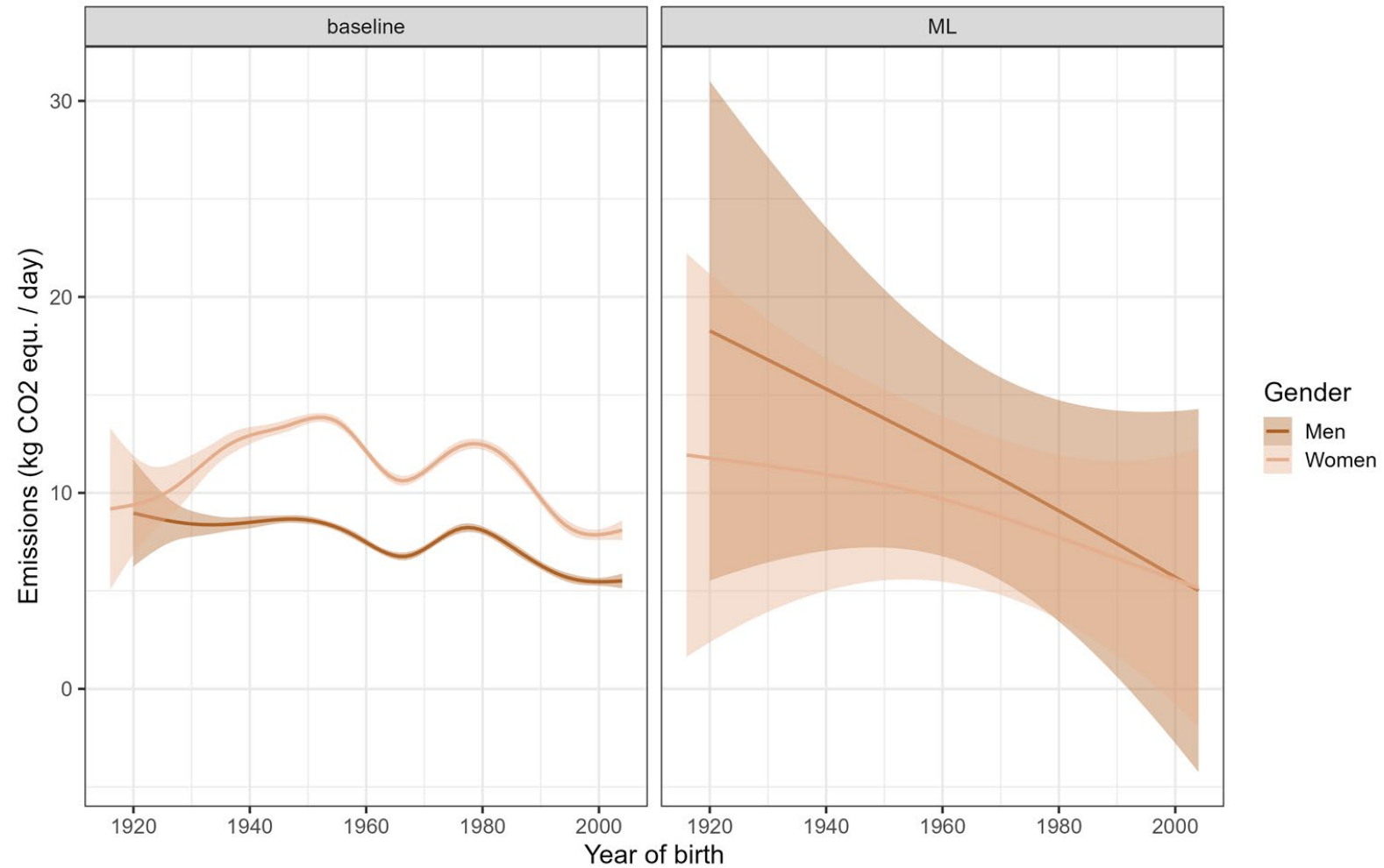
Committed Time-Use

Gender * Year of Birth



Committed Time-Use - Emissions

Gender * Year of Birth



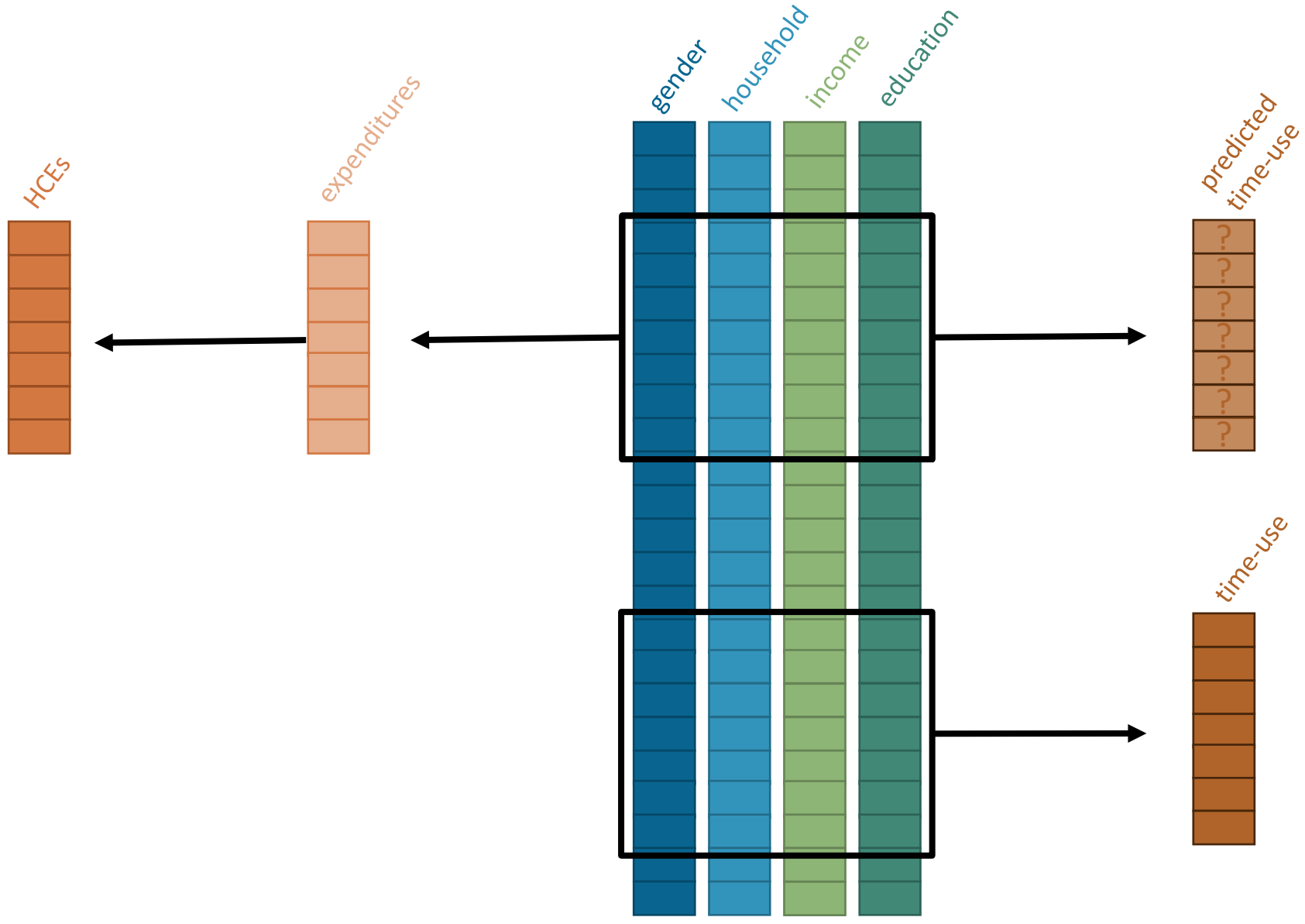


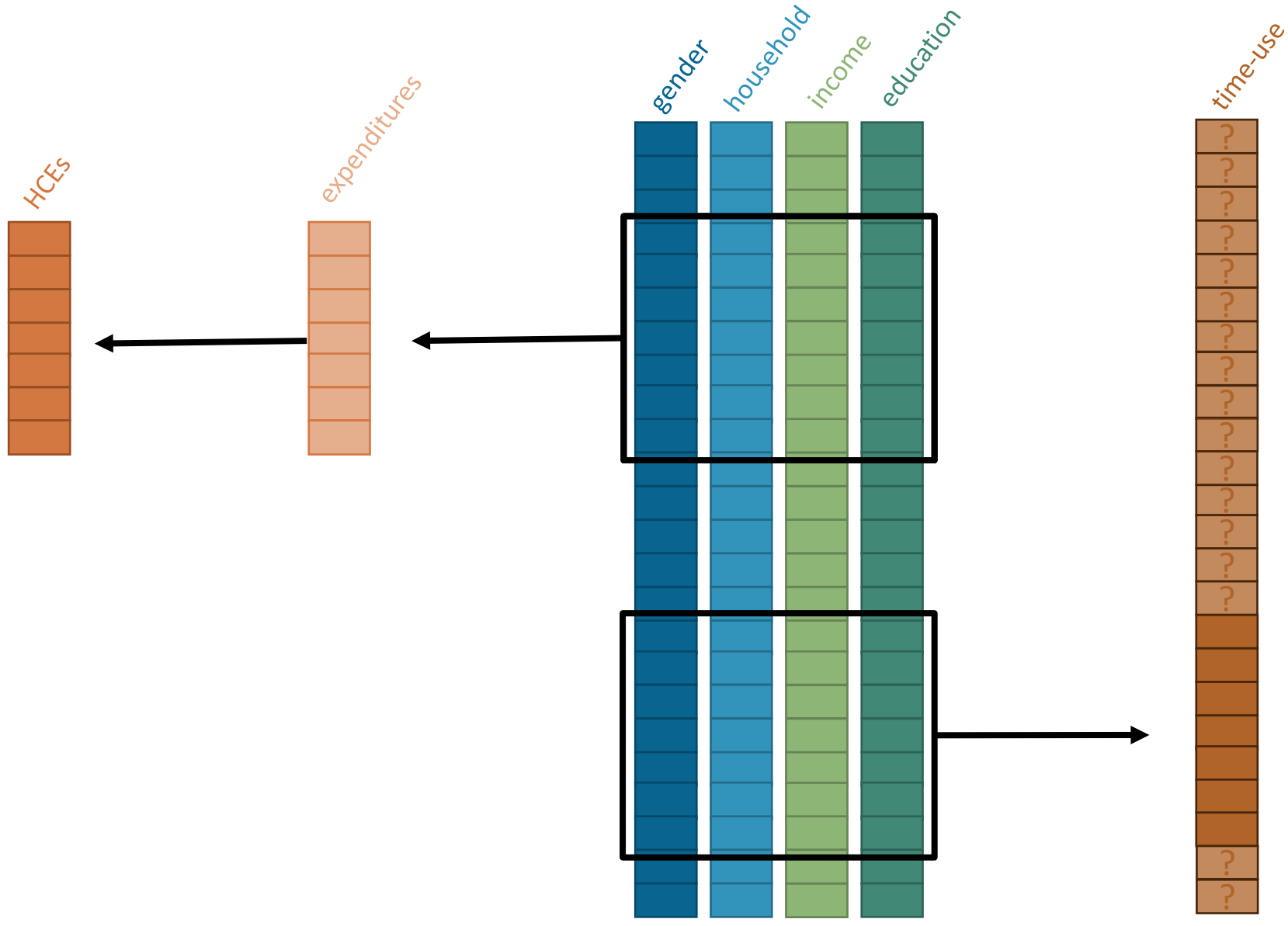
Next Steps

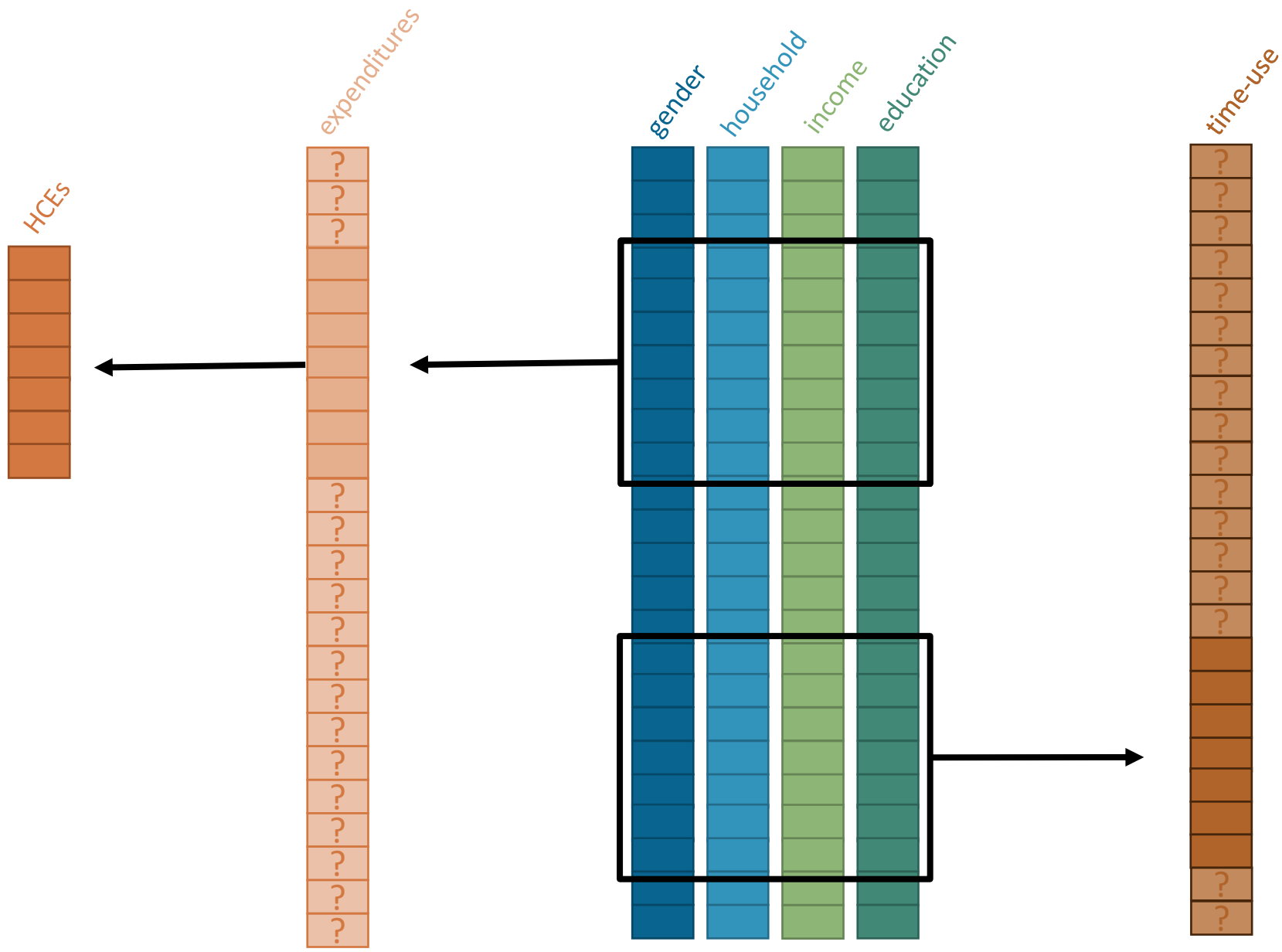


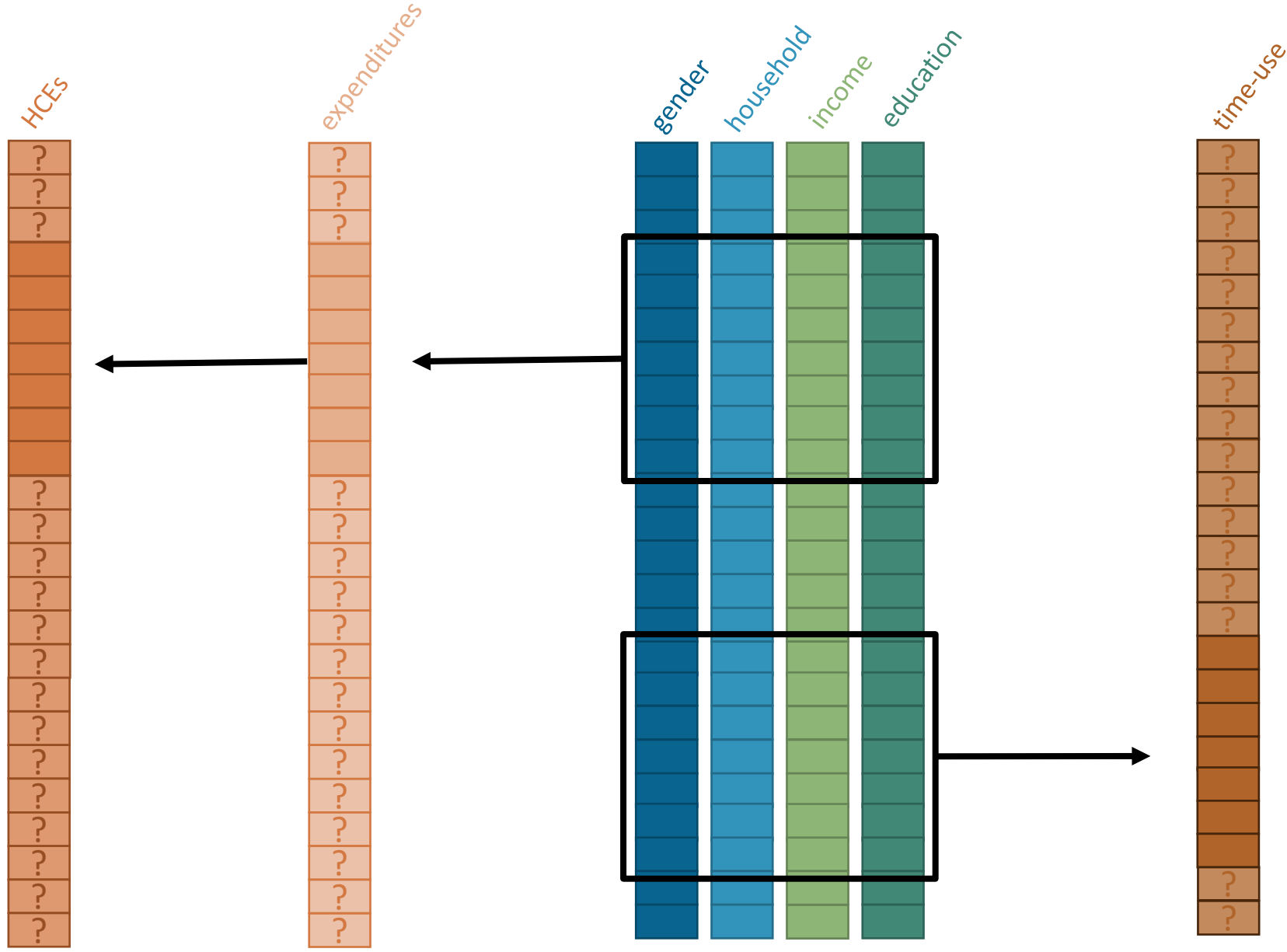
Next Steps

Calculate population-level individual carbon emissions









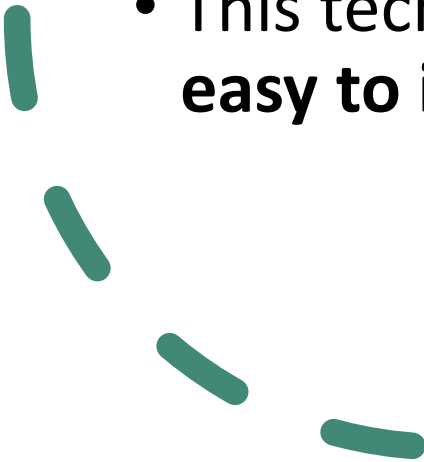


Next Steps

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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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