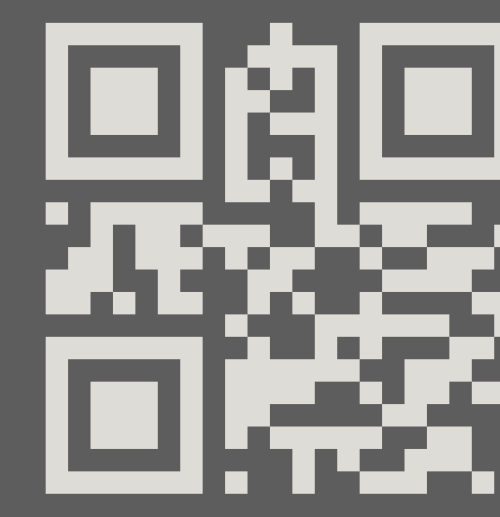


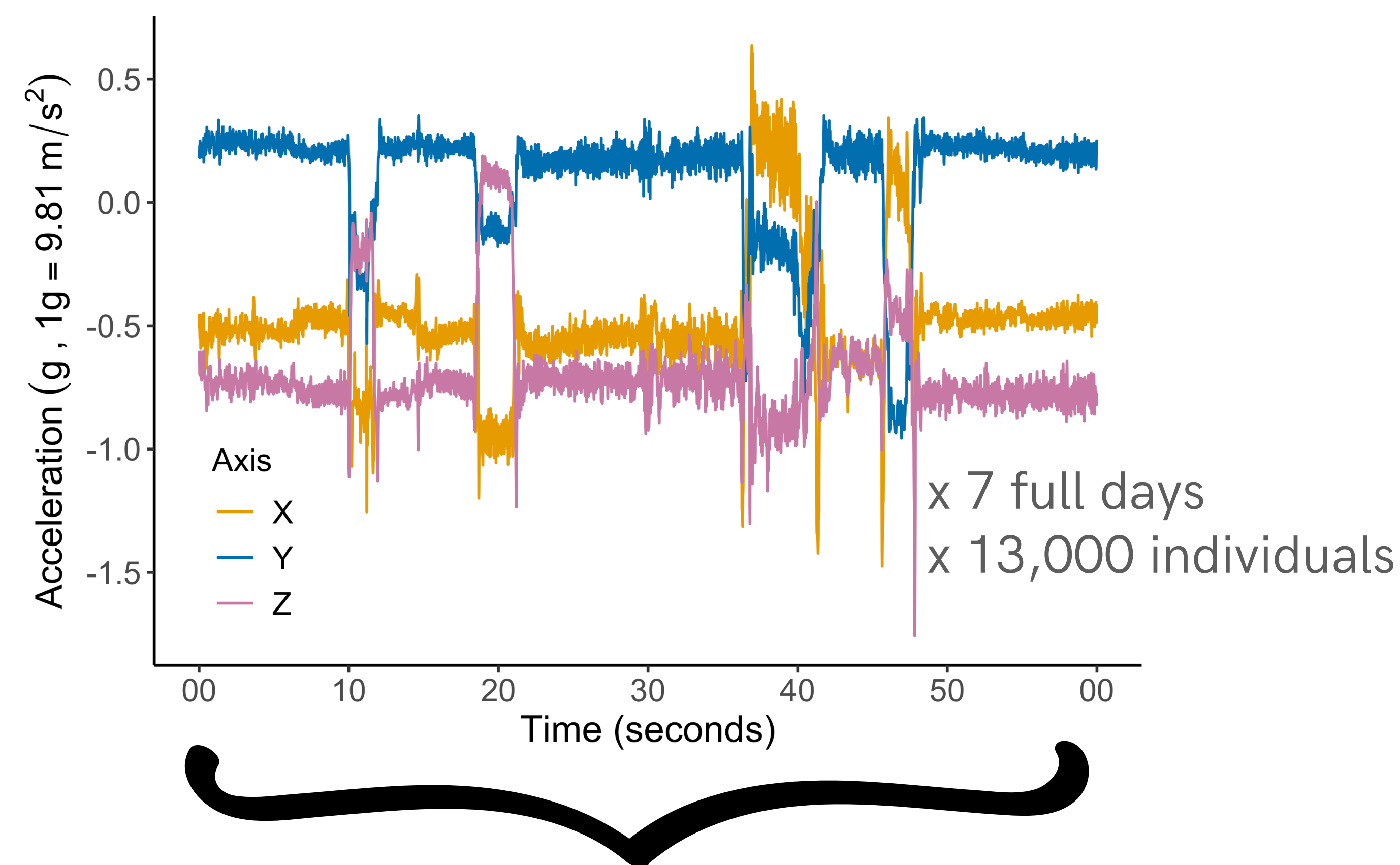
# FINGERPRINTING WALKING in a large epidemiological study



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## DATA

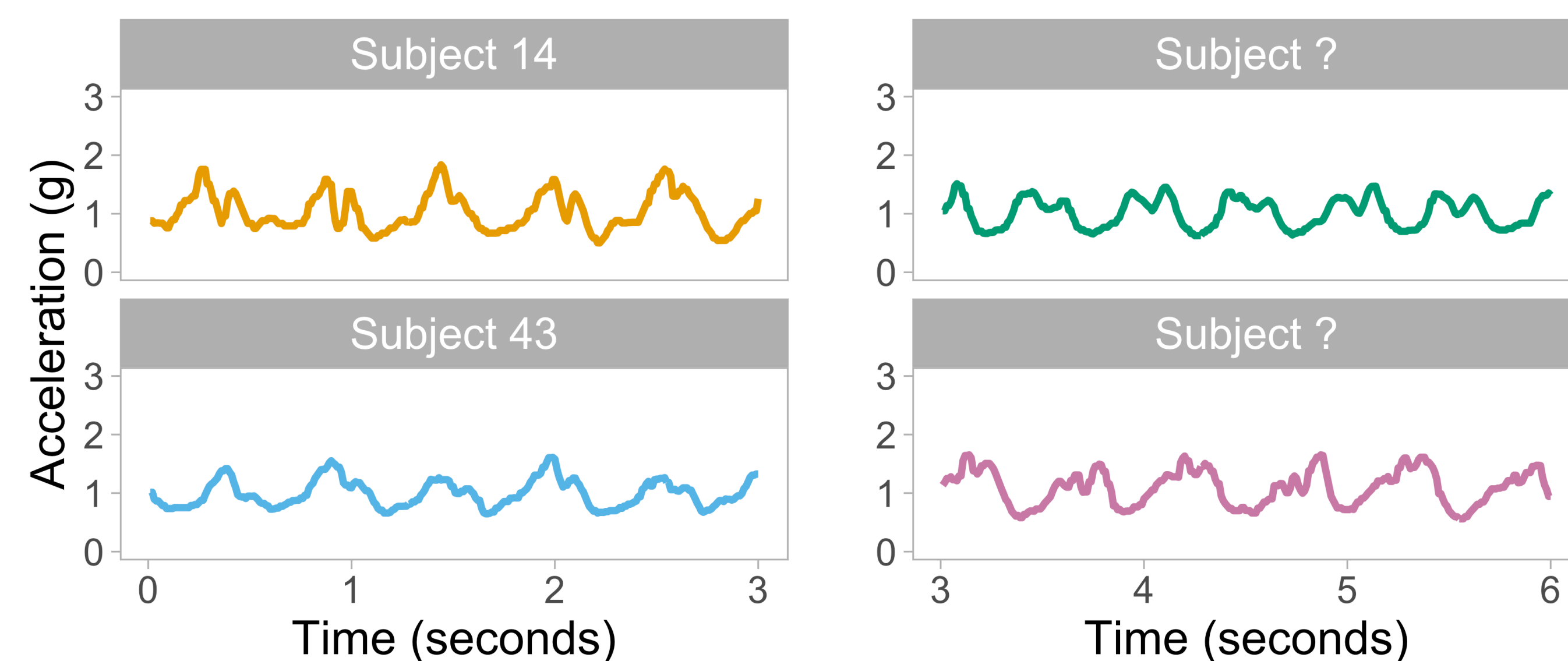
- Ntl Health & Nutrition Examination Survey (NHANES)
- Questionnaires and laboratory tests on large, **nationally representative** sample of Americans
- 2011-2014: survey included accelerometry sub-study



1 minute of data  
80 observations per second per dimension

## OBJECTIVE

Leverage accelerometry-derived walking patterns for biometric identification

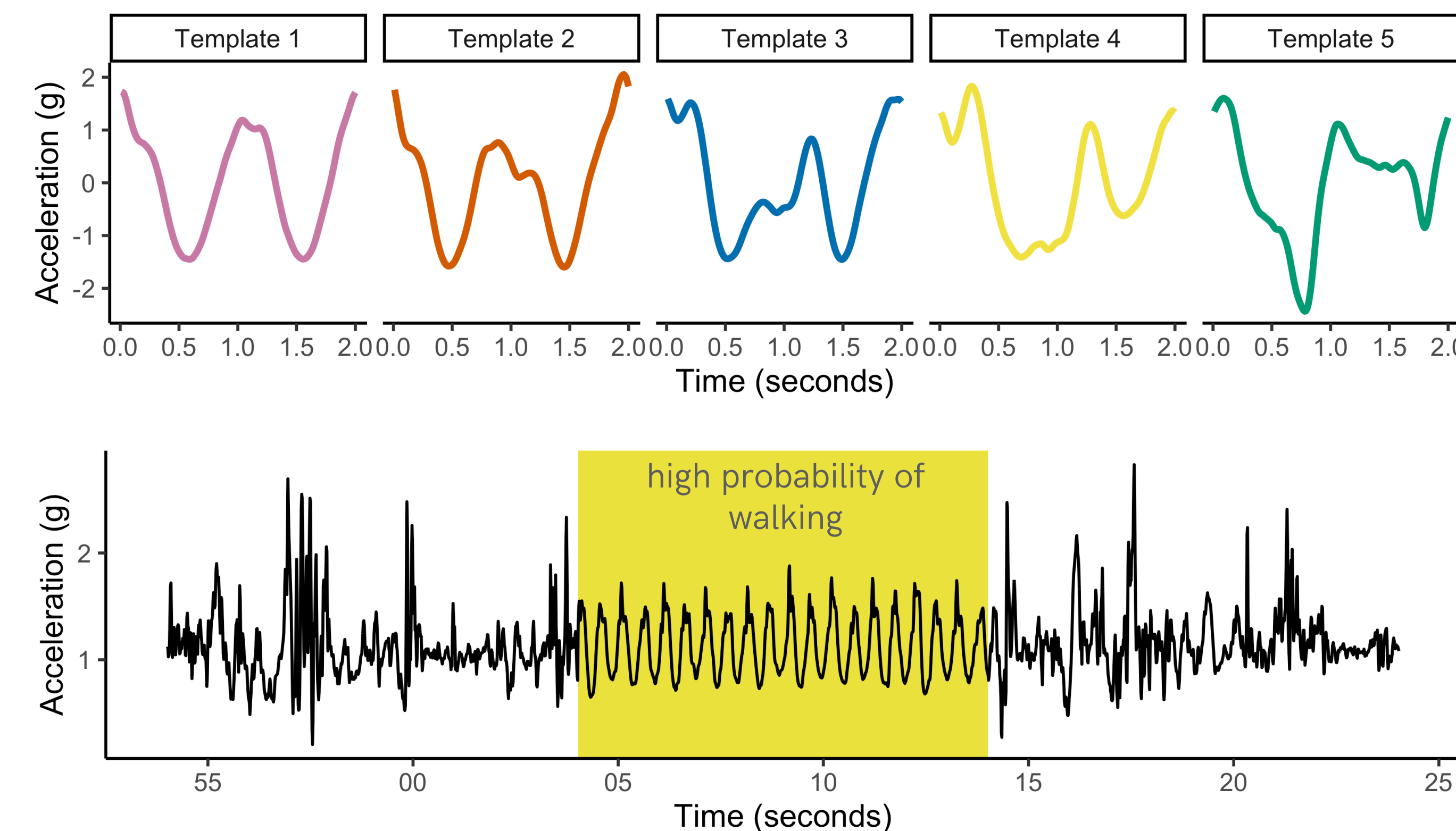


## METHODS

- Identify time periods with high probability of being walking
- Derive **grid-cell predictors** for identification models
- Implement various identification models under different train/test scenarios

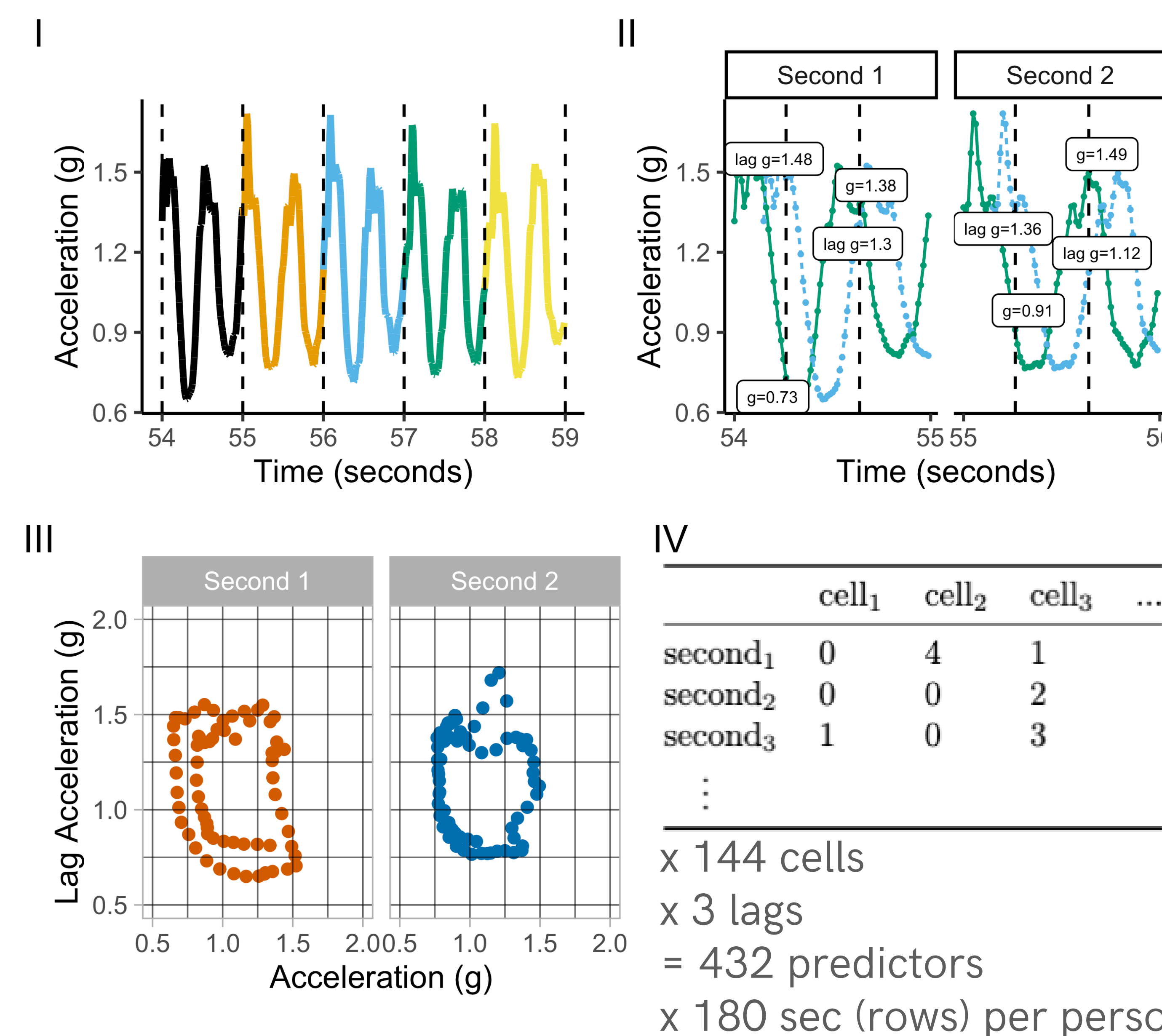
## 1. IDENTIFY WALKING

Use Adaptive Empirical Pattern Matching (ADEPT)<sup>1</sup> to identify walking periods from raw accelerometry data



## 2. DERIVE PREDICTORS

- Break walking periods into 1-second segments
- Compute empirical joint distribution of acceleration and lag acceleration at each 1/80th of a second for three lags
- Tabulate number of acceleration, lag acceleration pairs in each 0.25x0.25g square on the [0,3]gx[0,3]g grid
- Create predictor matrix

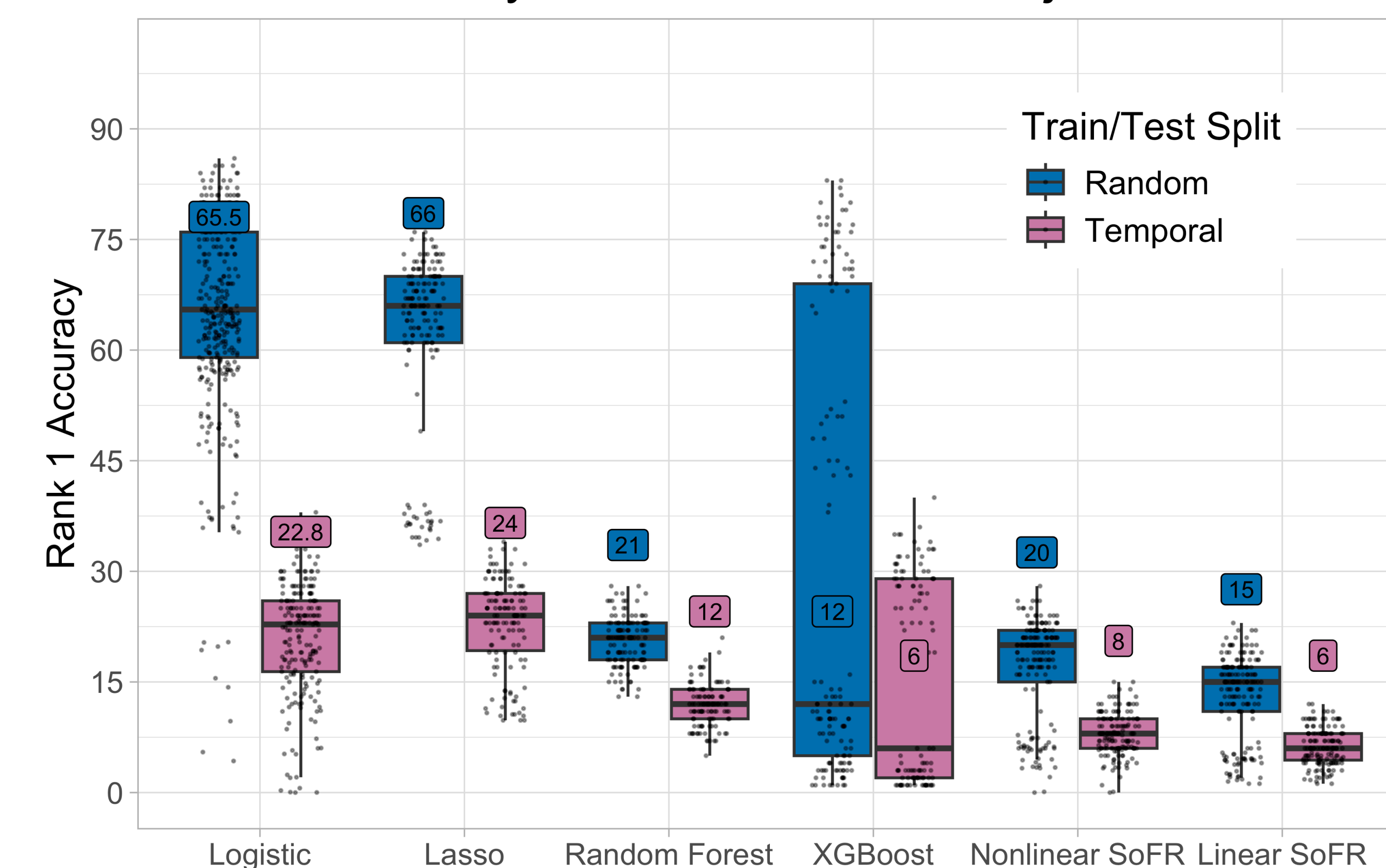


## 3. FIT MODELS

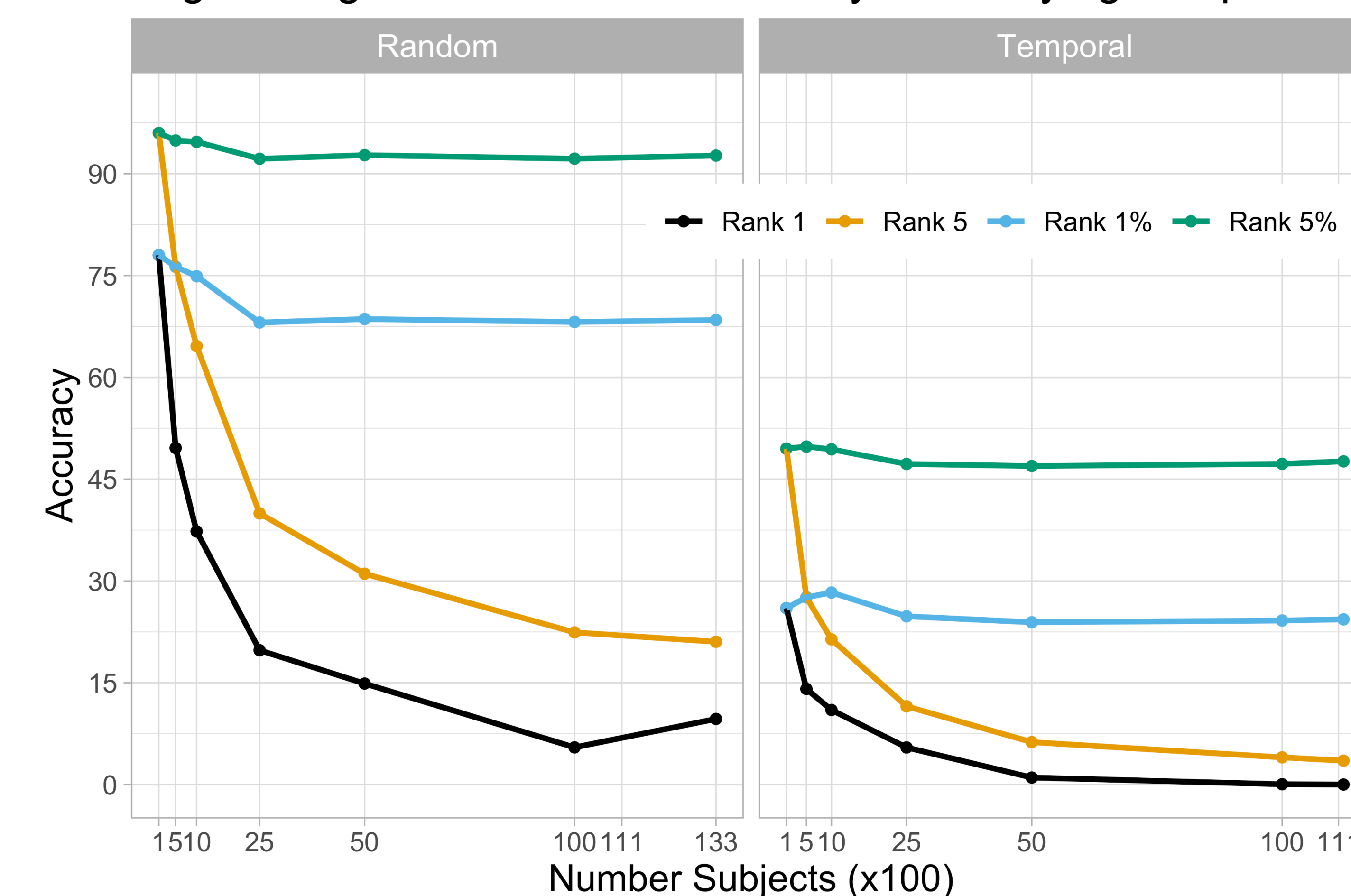
- One vs. rest classification paradigm with varying size subsets of the entire sample
- Logistic regression, machine learning (random forest, XGBoost), scalar on function regression
- Training and testing **randomly sampled**
- Training from days 1-3, testing from days 4-7 (**temporal**)

## RESULTS

Rank 1 accuracy in subsets of 100 subjects



Logistic regression models: accuracy with varying sample size



## TAKEHOME

- Accelerometry data can identify individuals from walking
- Implications for epidemiology, health, privacy/security
- Scalable to large datasets

