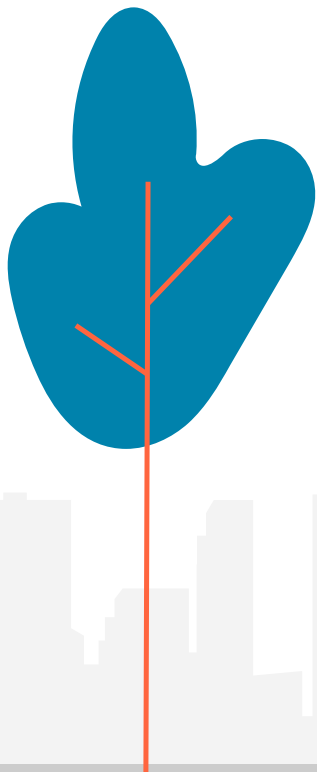
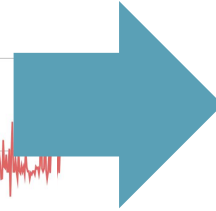
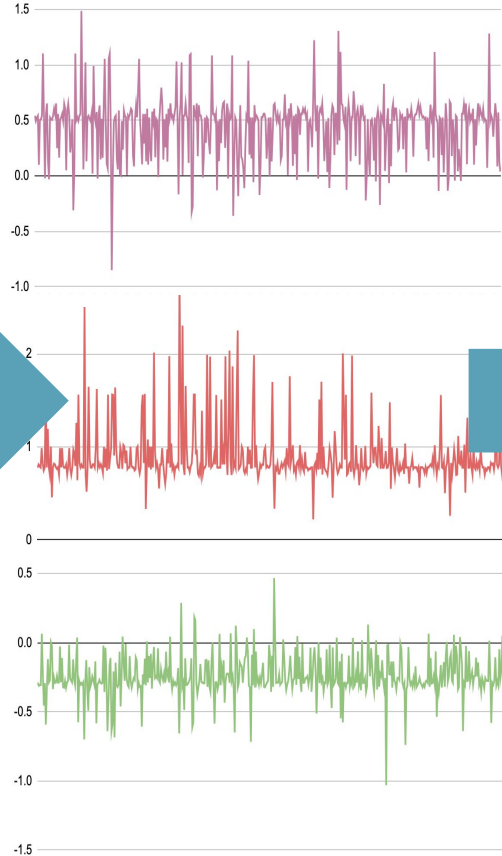
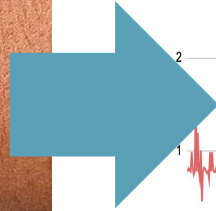


Stepping Up: Improving Step Count ML Algorithms

Brendan Callender, Jady Ellis,
Martin Hsu, Kirina Sirohi



Using Machine Learning to Count Steps

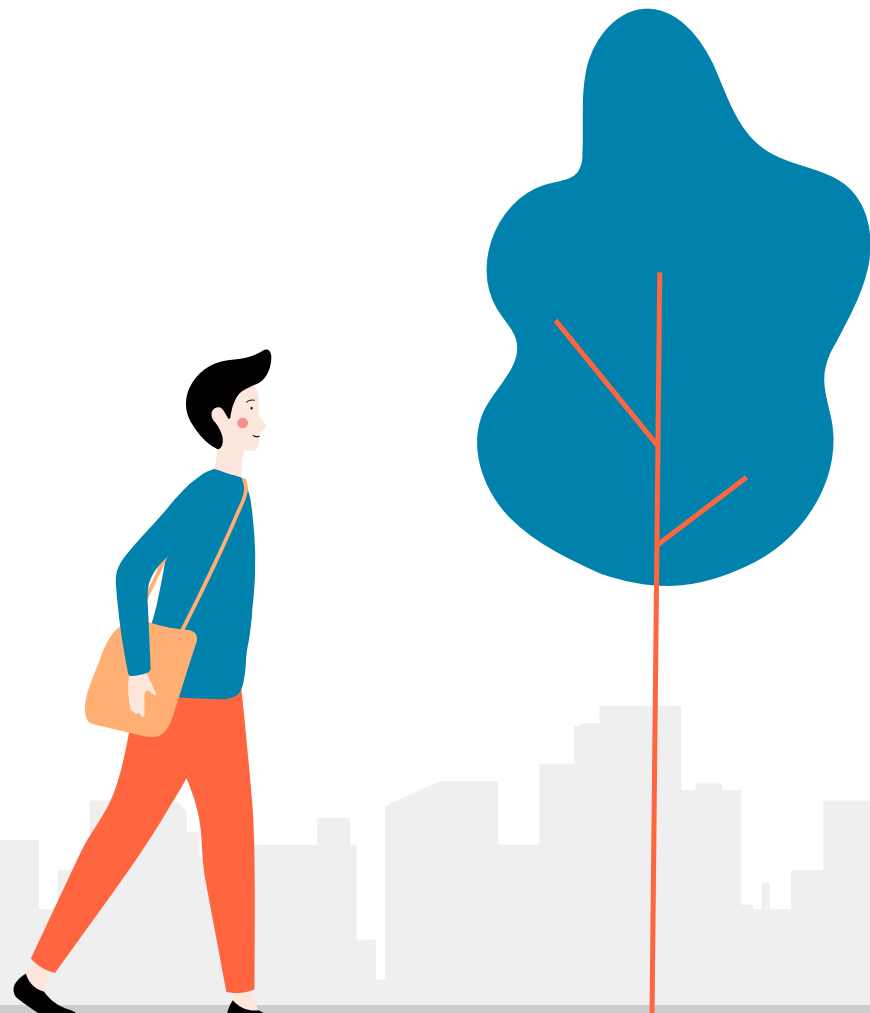


**8887
Steps**



01

Data



Accelerometer Data (ACT24)



Device Data

- Wrist-worn accelerometers
- 80Hz → 80 readings / second
- 3 axis → X, Y, Z



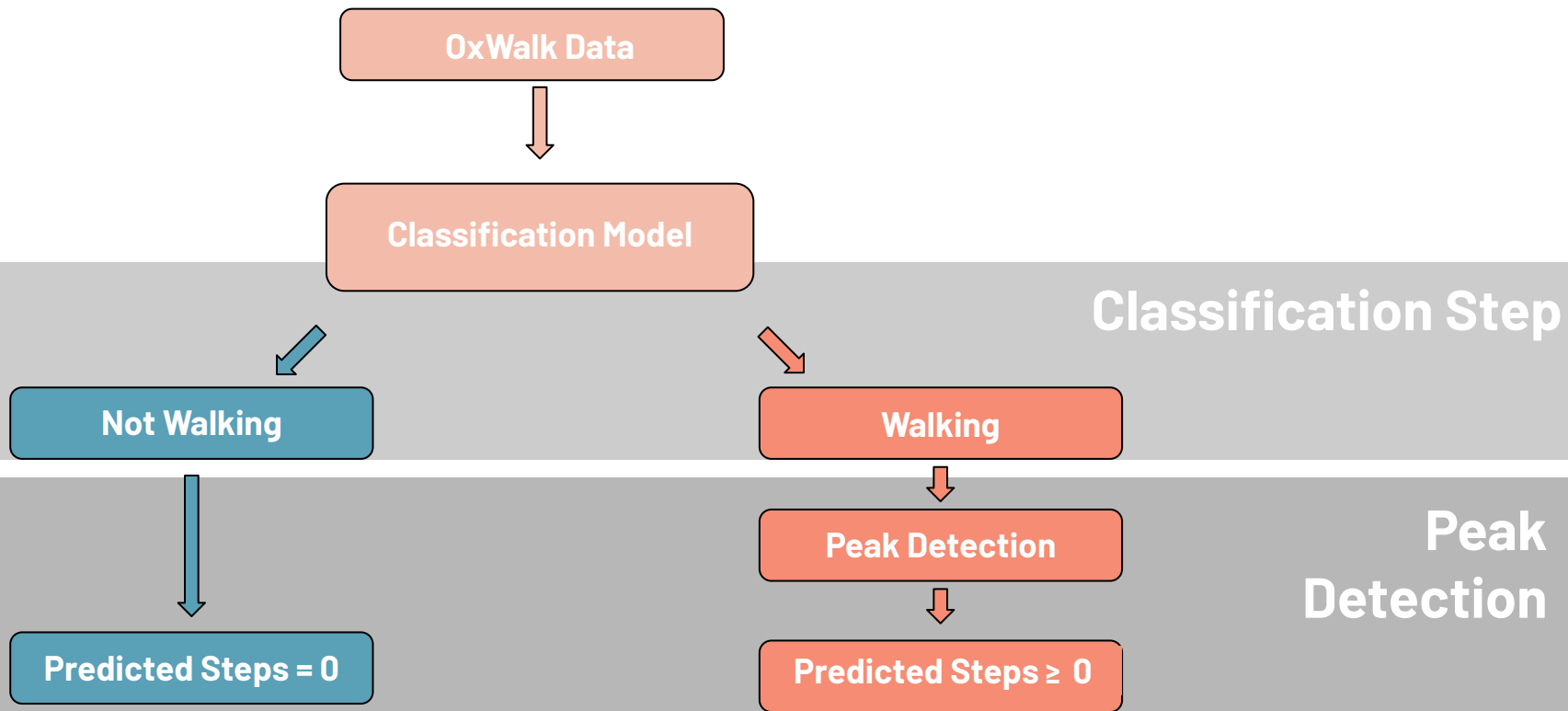
Ground Truth

- 24 people, 2x3 hour video recordings
- Labeled by Dr. Keadle's Team
- Steps
- Activity type, posture, intensity

02 Evaluation of UK Biobank Algorithm



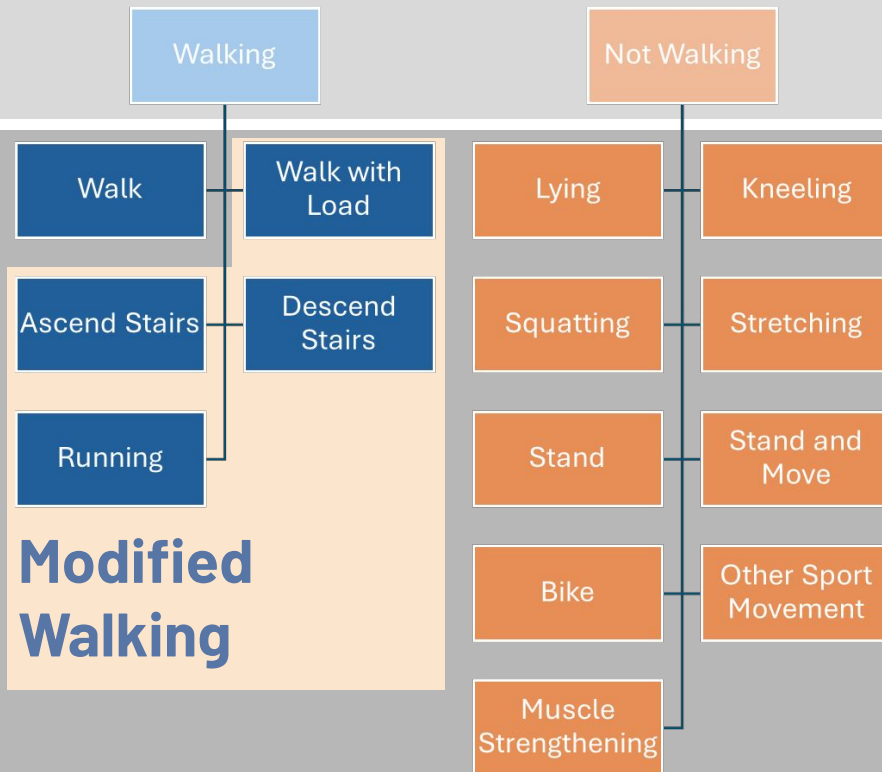
UK Biobank

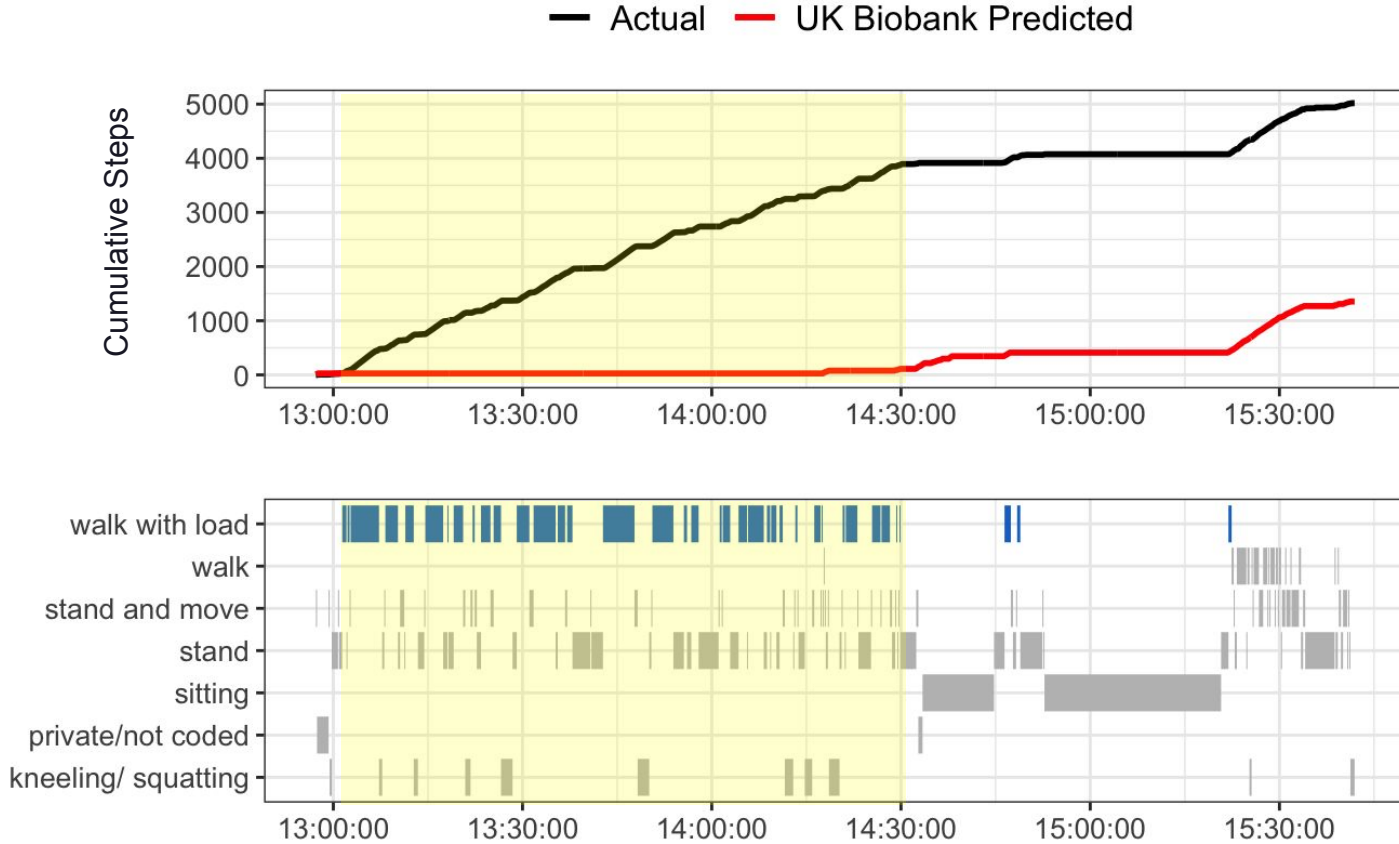


ACT24 Taxonomy is More Granular than Biobank

UK Biobank Postures

Dr. Keadle's Postures





**Biobank
algorithm
struggles with
modified
walking**

Modified walking has a large impact on error

When tested on ACT24, **UK Biobank algorithm**
is estimated to be **off by**



2.02

The infographic features two large circles, one blue and one orange, each containing a white numerical value. Below each circle is a text label. The background at the bottom shows a grey silhouette of a city skyline.

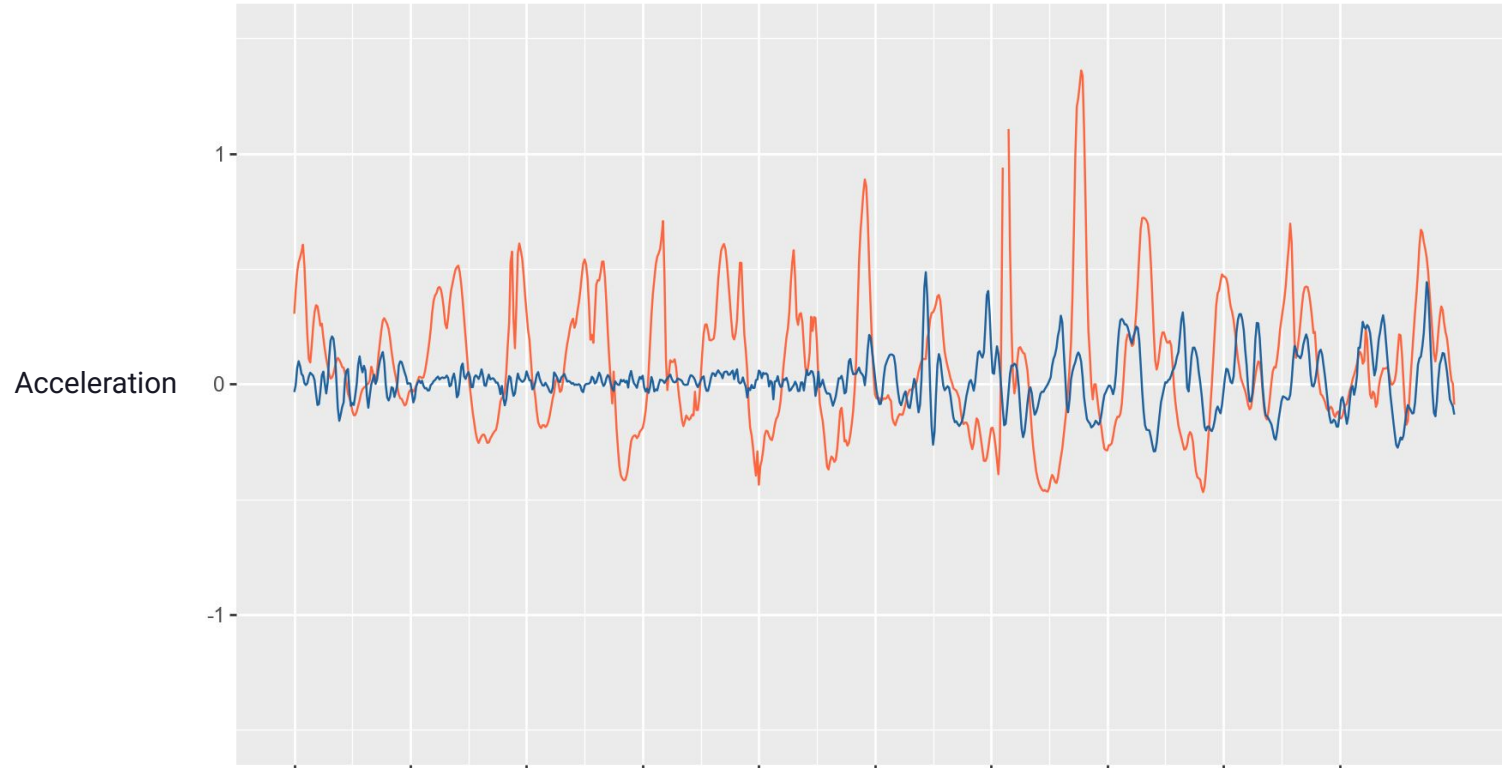
steps per minute
overall

22.90

steps per minute
in periods of **modified**
walking

Different Postures Convey Distinct Signals

Walking vs **Walking with Load**



New Research Answers/Questions


- **Q: How can we improve step counting?**
 - A: Focus on modified walking
 - A: Classify on more types of postures/activities
- **Q: How many categories should we classify on?**
 - Q: How many categories are too few? Too many?
 - Q: What should the categories be?

03

Developing Models



Process

1. **Define** different levels of classification granularity
 2. **Modify** algorithm to handle more categories and use ACT24 data instead of UK Biobank data
 3. **Cross-validate** on ACT24 data
 4. **Train** on ACT24, **test** on UK Biobank data
 5. **Calculate** error metrics, **compare** across granularity levels and to original algorithm
- 

Three Classification Levels



**More
Broad**

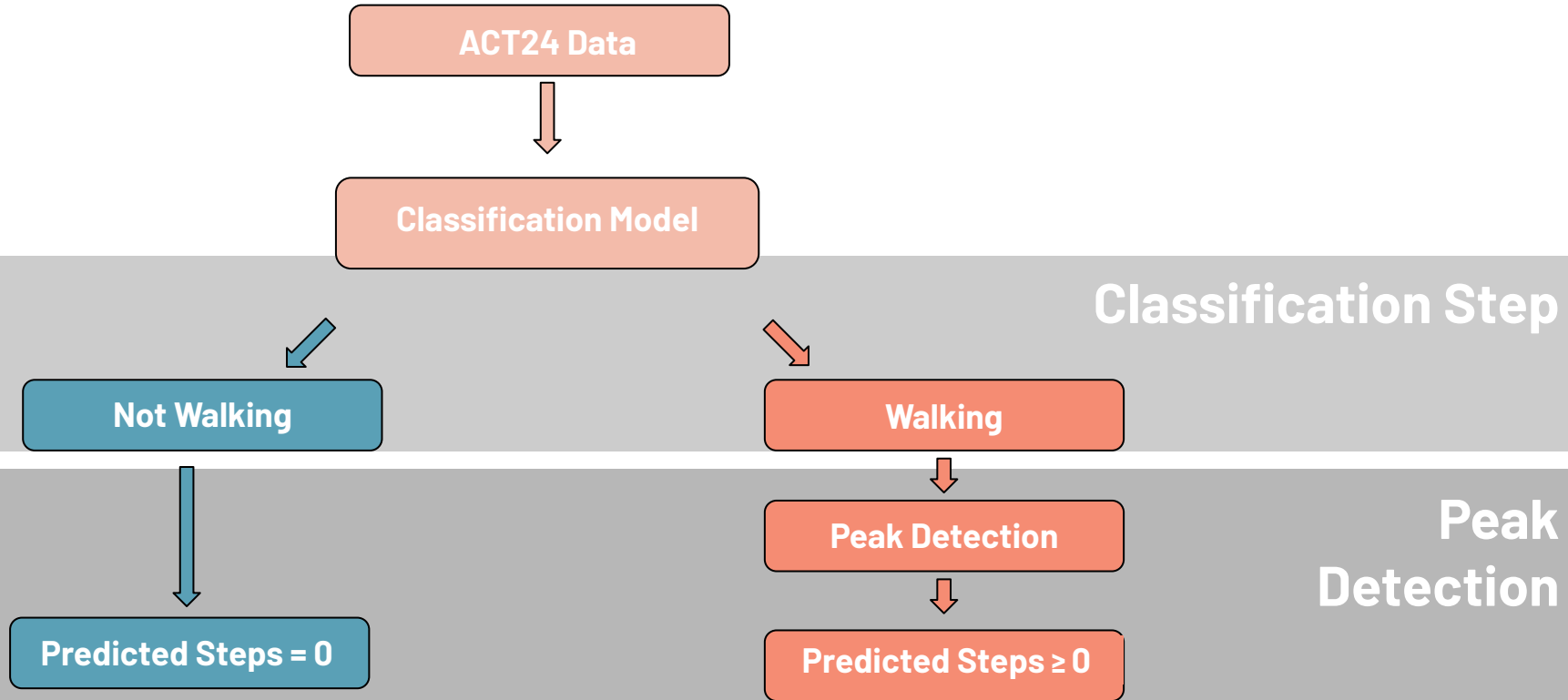
Walk/Not Walk

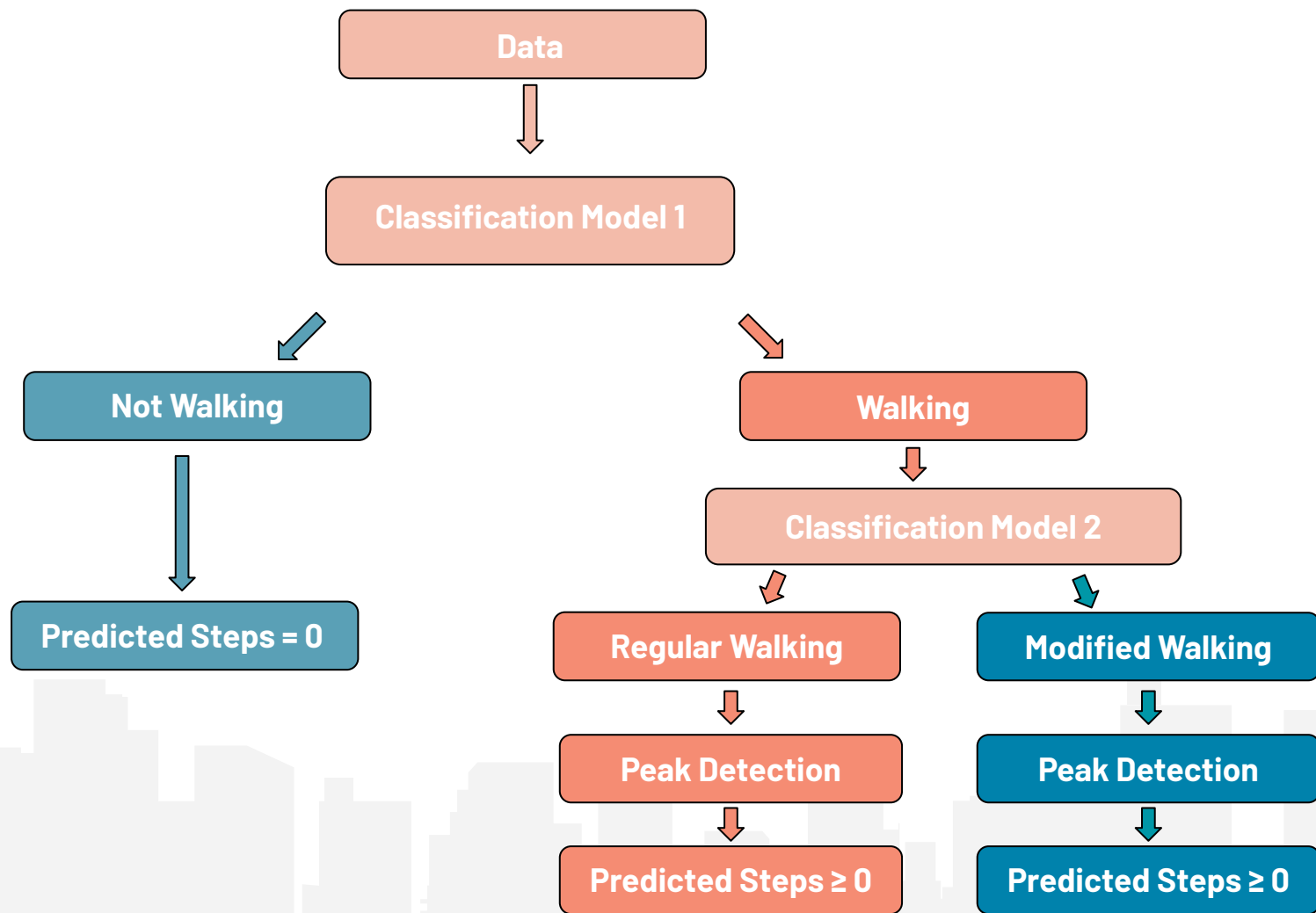
Condensed Postures

**More
Granular**

All Postures

ACT24 Walk/Not Walk Classification





ACT24 Walk/Not Walk Evaluation

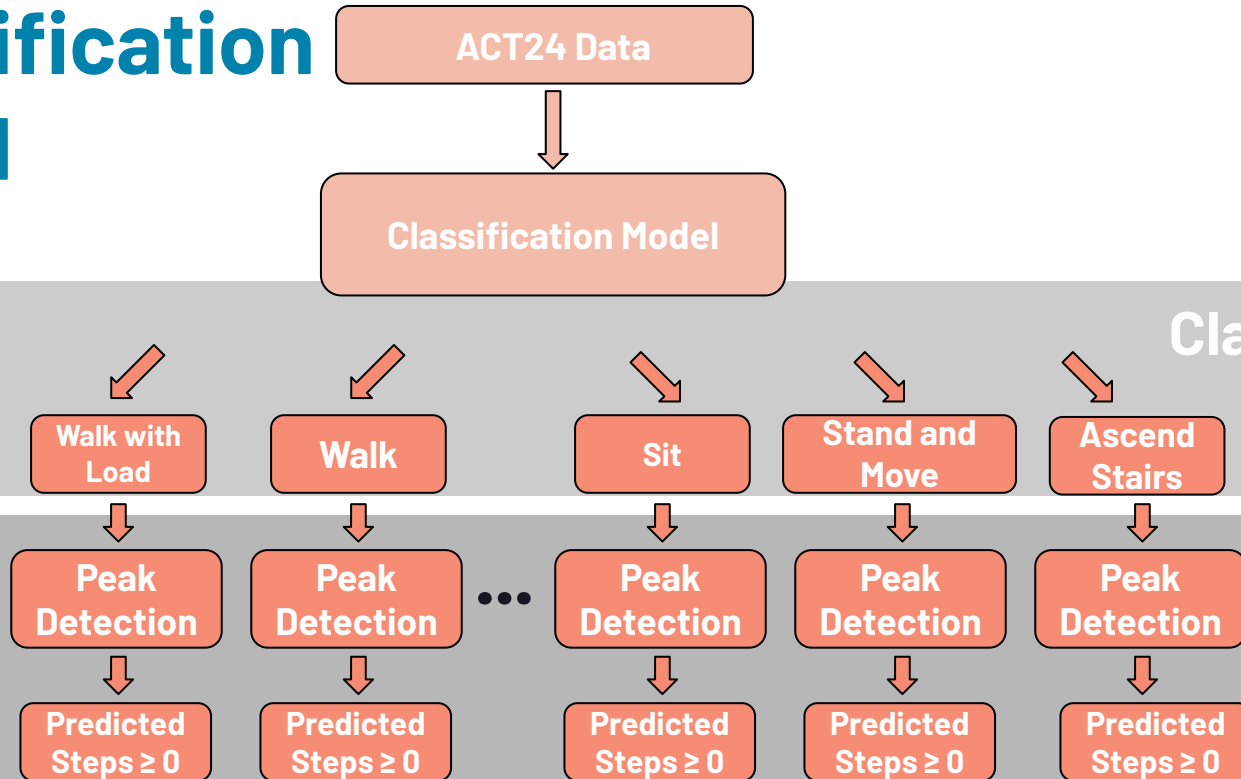
Pros

- Exposed to modified walking w/ ACT24 training data
- Binary classification problem (low classification error)
- Less sensitive to changes in dataset (low variance)

Cons

- Walking serves as a catchall category (high peak detection error)

All Posture Classification Model



Classification Step

Peak
Detection

All Posture Classification Evaluation

Pros

- Walking no longer a catch-all category
- Fined tuned peak detection for each category (lower peak detection error)

Cons

- Large number of classification categories
- Highly sensitive to changes in the dataset (high variance, overfitting, high classification error)
- More computationally expensive

Condensed Postures Mapping

Walk → **Walk**

Stand, Sit, Stretch, Kneel/Squat, Lying → **No Movement**

Stand and Move → **Stand and Move**

Ascend Stairs → **Ascend Stairs**

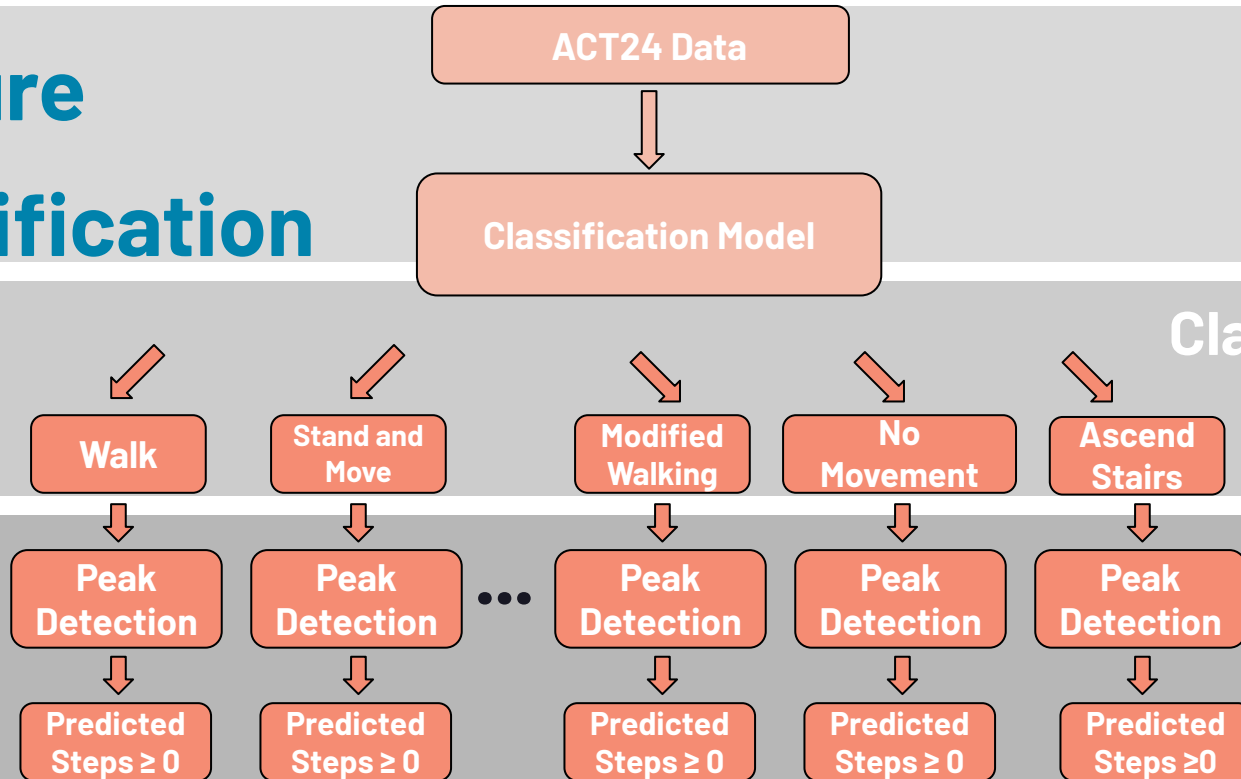
Descend Stairs, Walk with Load → **Modified Walking**

Bike → **Bike**

Muscle Strengthening → **Muscle Strengthening**

Other Sport Movement → **Other Sport Movement**

Condensed Posture Classification



Condensed Posture Classification Evaluation

Pros

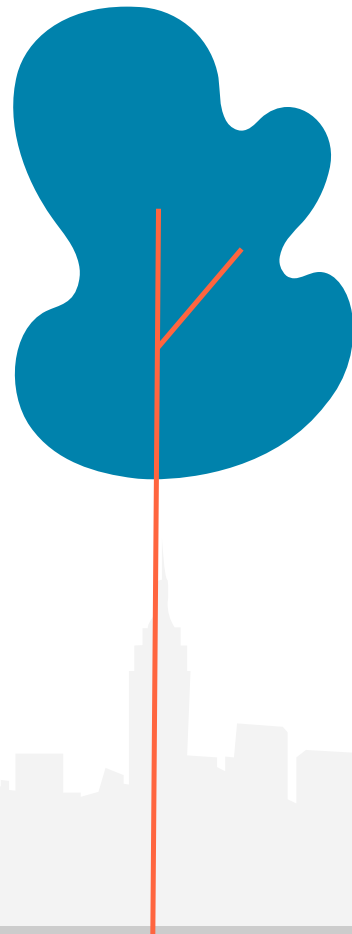
- Exposed to modified walking in training on ACT24
- Less classification categories
- Less computationally expensive than all postures

Cons

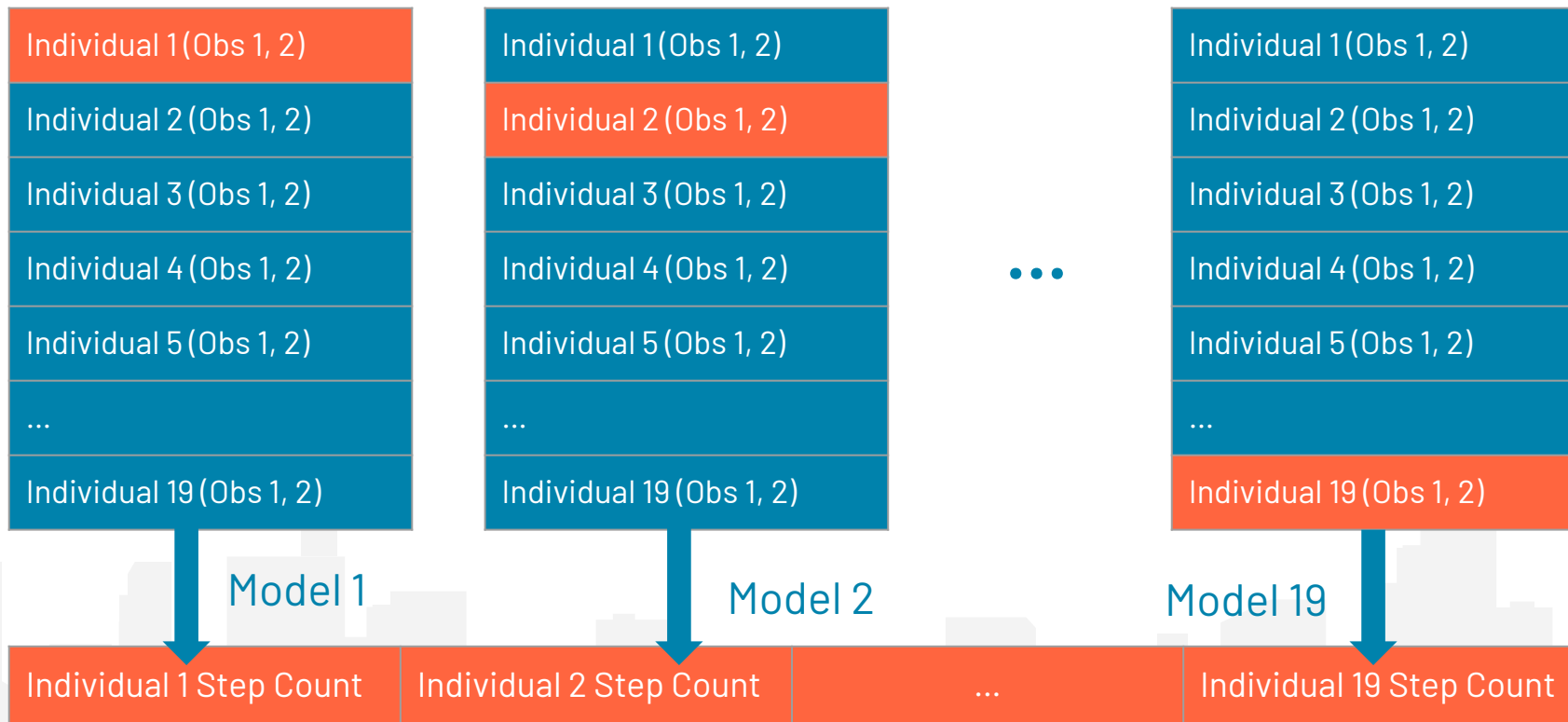
- Might still question overfitting to postures
- Could overlook certain unique posture signal tendencies

04

Model Testing



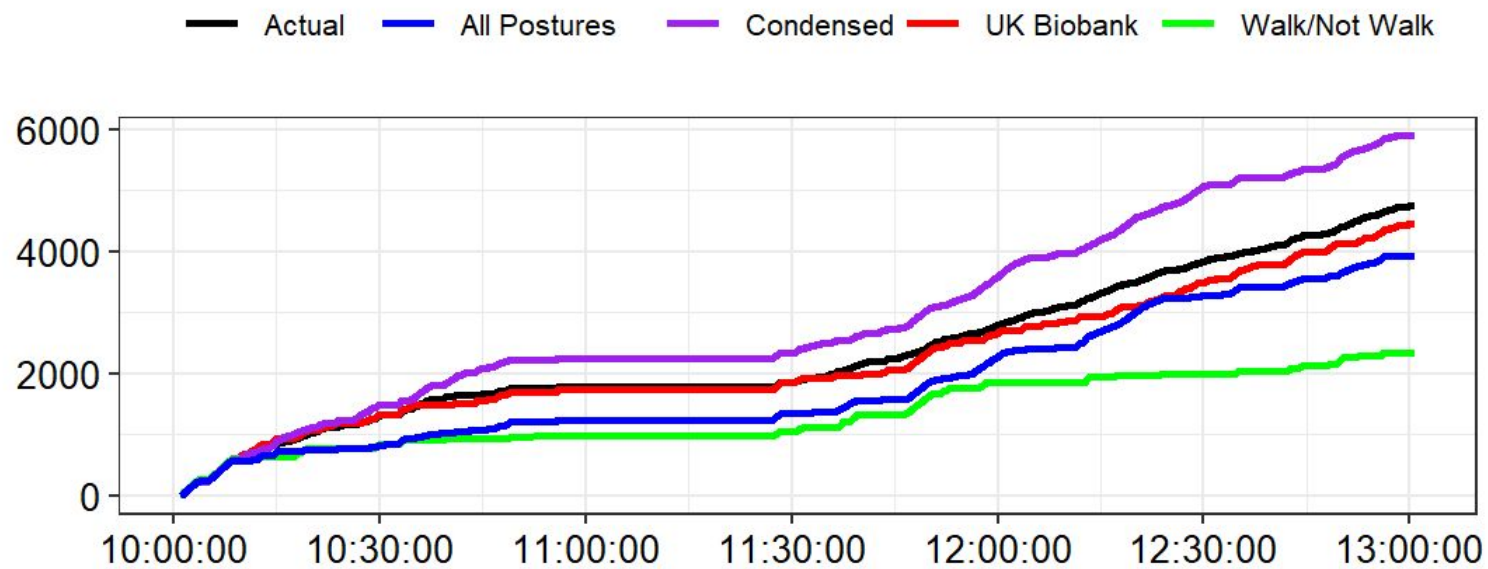
Leave-One-Out Cross Validation



"External" Predicted Step Counts for Entire ACT24 Dataset

Cross-Validation Pros/Cons

- **Test** on data with ground truth that **includes postures**
- **Computationally expensive** (Fit 57 Neural Networks)
 - 9 hour runtime with GPU acceleration!
- Postures **not equally represented** in each individual
 - “Running” only has one 10-second epoch in one individual
 - Model changes heavily depending on which individual excluded



On average, the models are estimated to be off by

Walk/Not Walk

3.901

All Postures

3.213

Condensed
Postures

2.578

steps per minute overall

On average, the models are estimated to be off by

Walk/Not Walk

25.458

All Postures

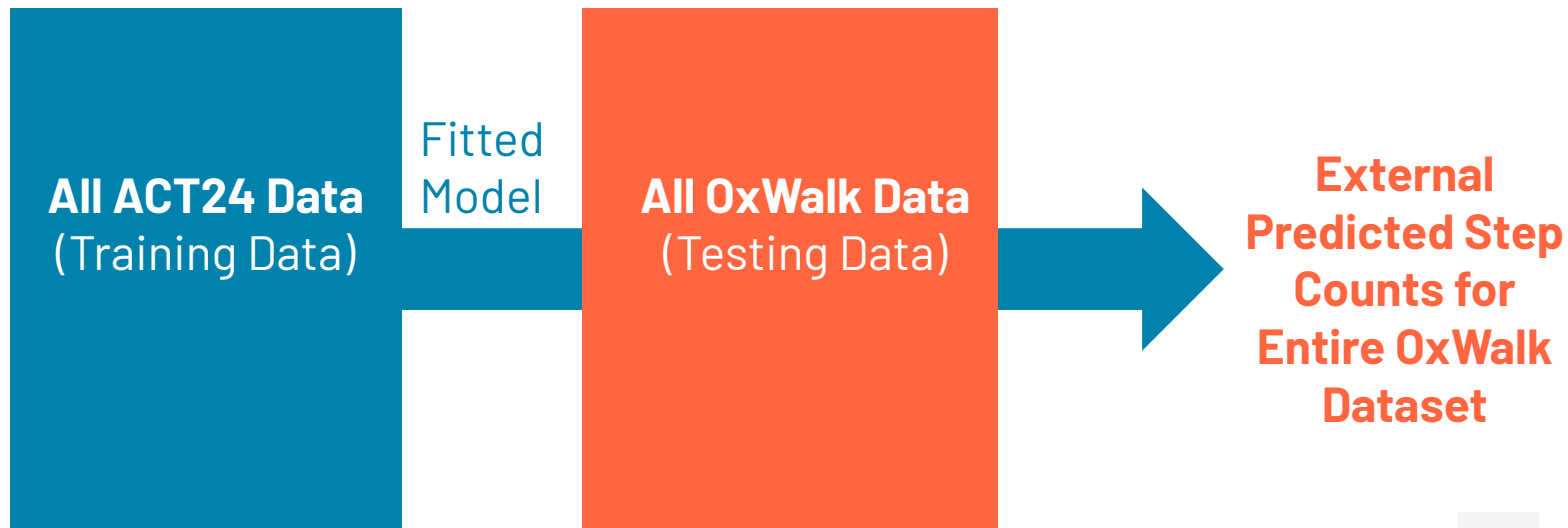
24.707

Condensed
Postures

25.852

steps per minute in periods of
modified walking

External Data Testing



- **Test** on **true external** UK Biobank accelerometer data (**OxWalk**)
- **No posture ground truth**

On average, the models are estimated to be off by...

Walk/Not Walk

3.810

All Postures

0.231

Condensed
Postures

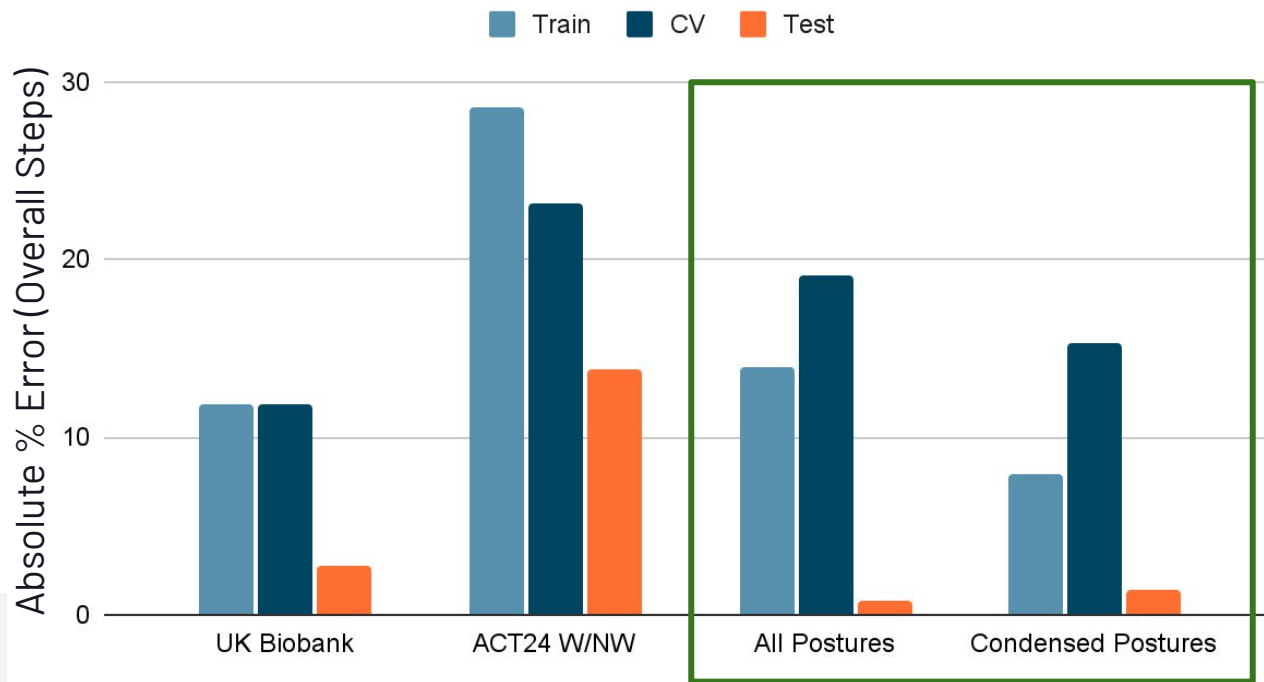
0.400

steps per minute overall

A light gray silhouette of a city skyline with various building shapes of different heights, spanning the width of the bottom of the slide.

05 Results





**In general,
classifying on
more postures
results in
noticeable
improvement**

Ranking Models (Test Data)

Model	1st	2nd	3rd
Condensed	18	12	5
All Postures	14	17	5
Walk / Not Walk	3	6	29

Results

- Classifying on **more** categories **improves** step counting outcomes
 - **Unclear** if there are specific improvements for **modified walking**
- We determined **preference** for **condensed** postures (not too few, not too many)
 - Metrics show **mixed** preference between all postures and condensed
 - Condensed is less computationally expensive
 - Less classification error



Final Thoughts

- ACT24 training data does not **represent** postures/activities equally
 - Having **more data** that represents more postures may **improve/change** model step counting algorithm outcomes
- **More combinations** of condensed postures can be explored
- **Other model specifications** have not been explored
 - Using **less** algorithm levels (e.g. one pass)
 - Using **more** algorithm levels (e.g. walk/not walk -> regular/modified walk -> peak detection)

Thanks!

