## Walking Analysis with Machine Learning

GA DSIF 9 Capstone Project

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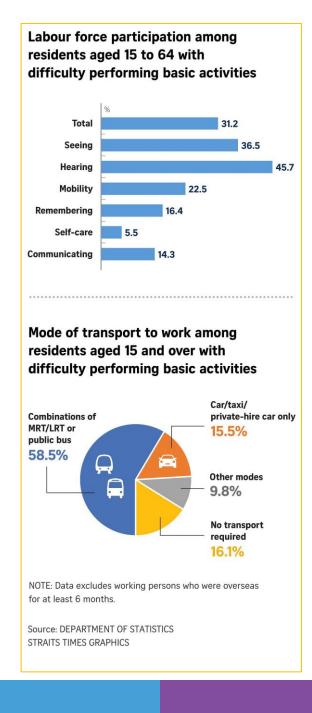
Prepared on: 23 May 2023



## Background and Problem Statement

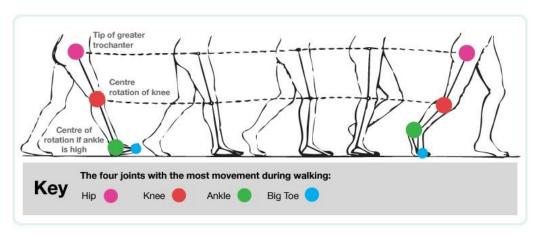
#### Problem Statement

- 1. In addition to Singapore's ageing population, there are many factors affecting mobility and quality of life, such as loss of muscle with age, osteoarthritis and injuries.
- 2. Among 150,000 household surveyed in 2020 population census, 62,500 people faced mobility challenges such as walking and climbing steps.



## Objectives

- 1. Using machine learning to identify key features that relates to different walking issues between Healthy Controls, Hip, Knee and Ankle respectively.
- 2. Explore possibility of affordable wearable sensors that target these key features for general public to monitor outcomes of treatments.



Source: https://americanbonehealth.org/exercise/the-mechanics-of-walking/

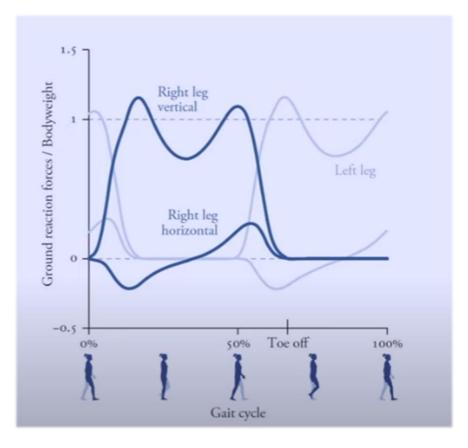
#### Measurements

There are 3 types **Ground reaction forces** (N):

- Vertical (Landing and push-off)
- Horizonal (Braking and propulsion)
- Rotation

There are 2 types of **Center of Pressures** (Cm):

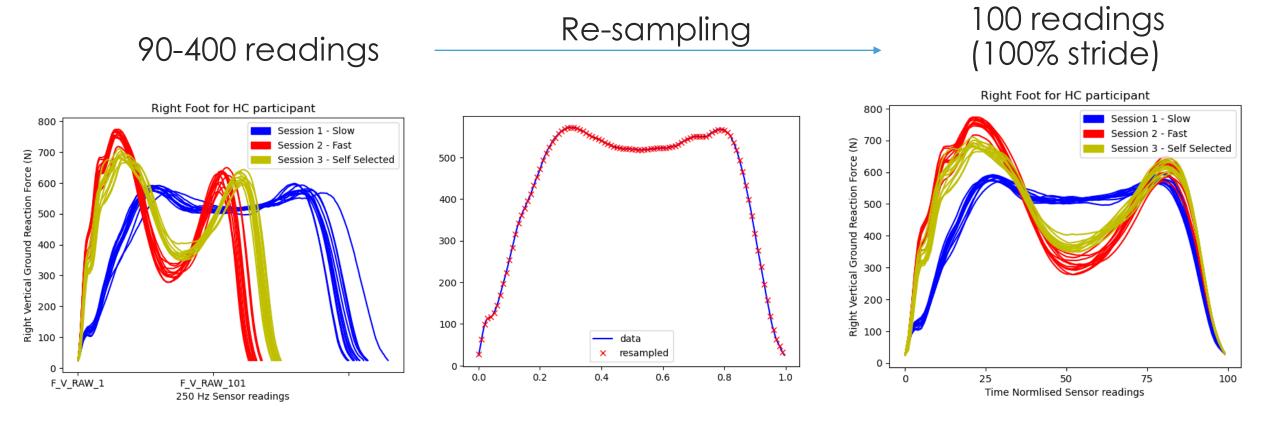
- Horizontal (Braking and propulsion)
- Rotation



Source: https://www.youtube.com/watch?v=7EdzFIRE8vY

# Data Processing and Exploration

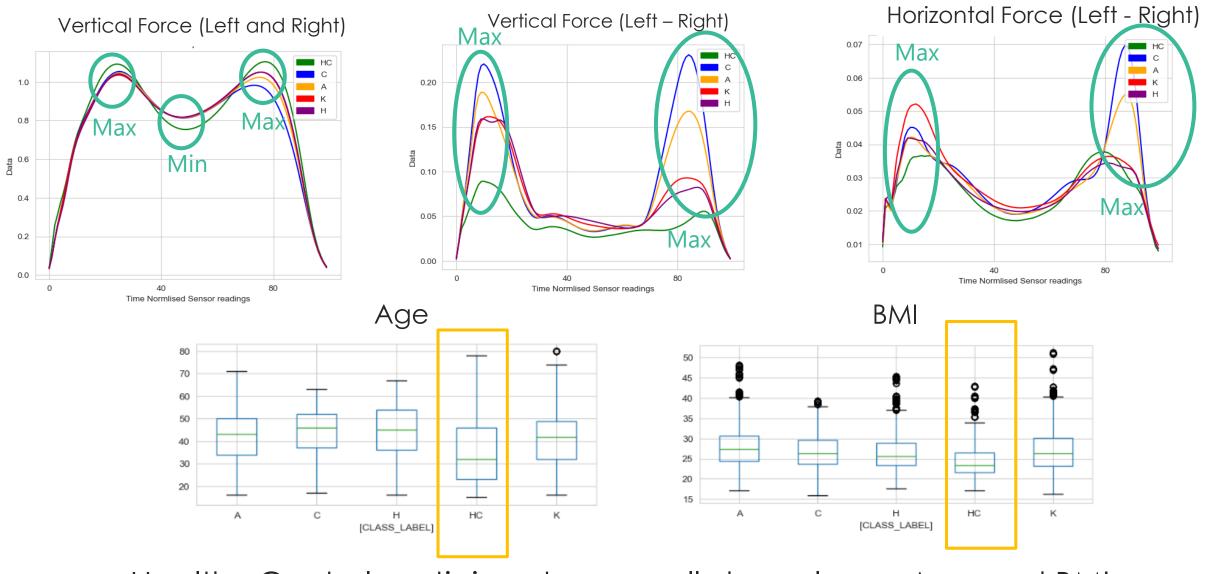
### Data Processing



The dataset is obtained from GaitRec, a large online database for gait analysis: Horsak, B., Slijepcevic, D., Raberger, AM. et al. GaitRec, a large-scale ground reaction force dataset of healthy and impaired gait. Sci Data 7, 143 (2020)

All GRF readings are subsequently **normalized** by weight of participant

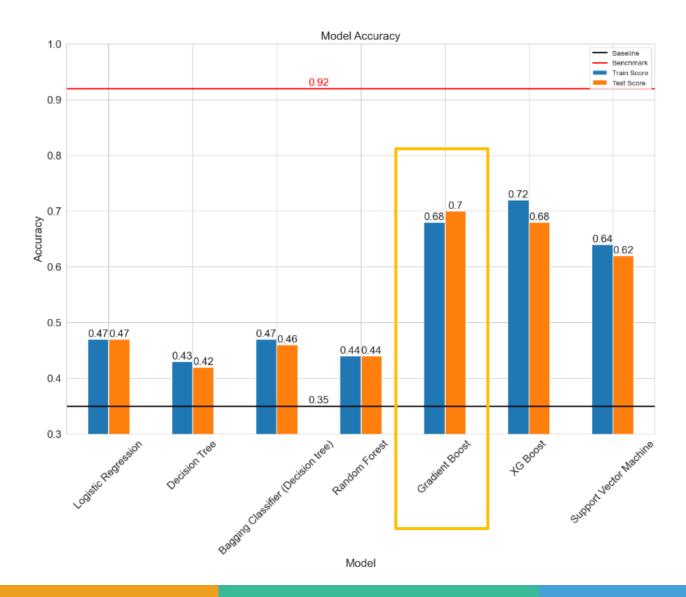
#### 12 Selected Features



Healthy Control participants generally have lower Age and BMI

## Model Evaluation and Conclusion

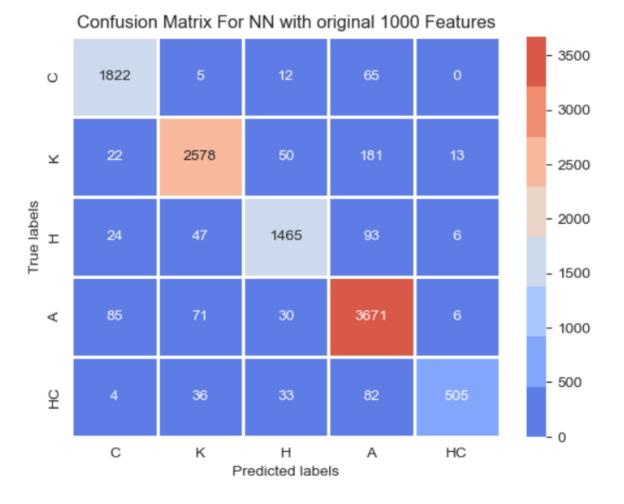
#### Model Evaluation



Gradient Boosted Classifier with accuracy of **70%** will be selected as the best alternative to the benchmark model of **92%** accuracy.

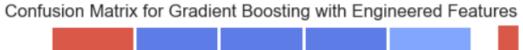
#### Benchmark: Neural Network with 1000 features

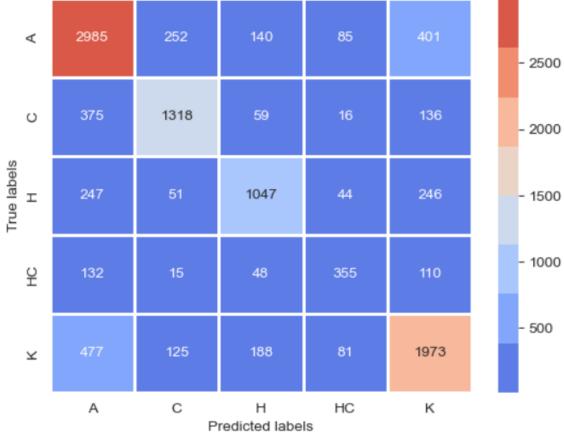
	precision	recall	f1-score	support
С	0.93	0.96	0.94	1904
K	0.94	0.91	0.92	2844
H	0.92	0.90	0.91	1635
Α	0.90	0.95	0.92	3863
HC	0.95	0.77	0.85	660
accuracy			0.92	10906
macro avg	0.93	0.89	0.91	10906
weighted avg	0.92	0.92	0.92	10906



#### Gradient Boosted Trees with 12 selected features

	precision	recall	f1-score	support
А	0.71	0.77	0.74	3863
C	0.75	0.69	0.72	1904
Н	0.71	0.64	0.67	1635
HC	0.61	0.54	0.57	660
K	0.69	0.69	0.69	2844
accuracy			0.70	10906
macro avg	0.69	0.67	0.68	10906
weighted avg	0.70	0.70	0.70	10906





#### Conclusion

- 1. Using key features like BMI, Age, maximum landing force, push force, together with differences in vertical and horizontal forces between left and right leg can approximately reach 70% accuracy in classification of gait.
- 2. There is feasibility in exploring more key features and affordable sensors that target these key features in solutions and research available currently.



#### Source for illustration:

Binelli, M.R., van Dommelen, R., Nagel, Y. *et al.* Digital manufacturing of personalised footwear with embedded sensors. *Sci Rep* **13**, 1962 (2023). https://doi.org/10.1038/s41598-023-29261-0

## Summary and Appendix

6. Compare with 7. Back to 1. Prepare and baseline and feature process data benchmark engineering 5. Train and 8. Train and 2. Feature Engineering evaluate models evaluate models 9. Model 3. Feature 4. Baseline Selection and

model

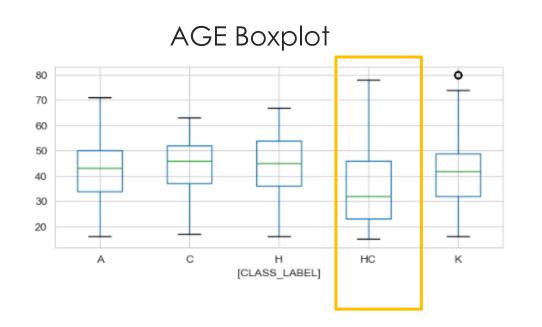
Conclusion

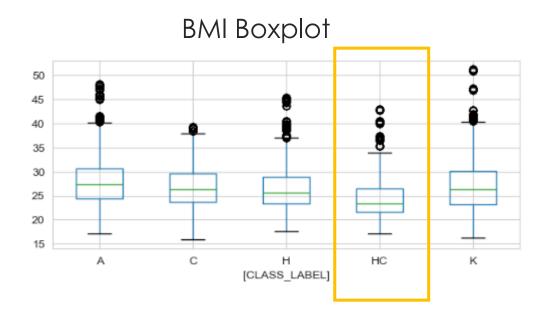
Selection

### Distribution of classes



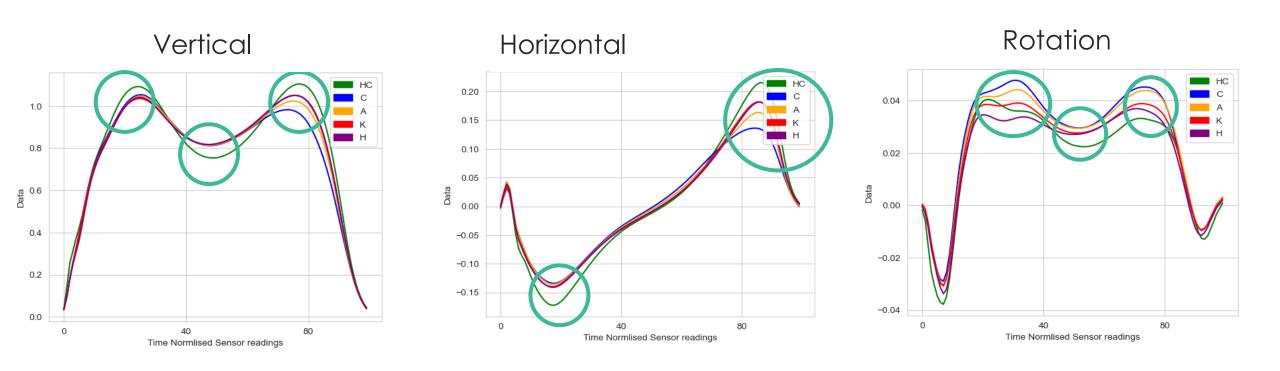
### Age and BMI





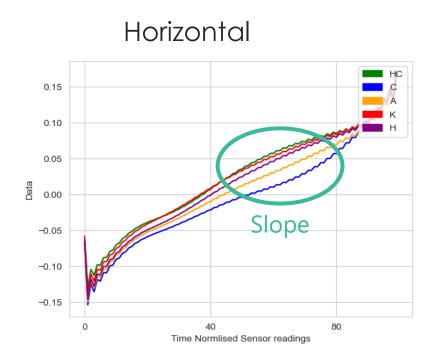
Healthy Control participants generally have lower Age and BMI

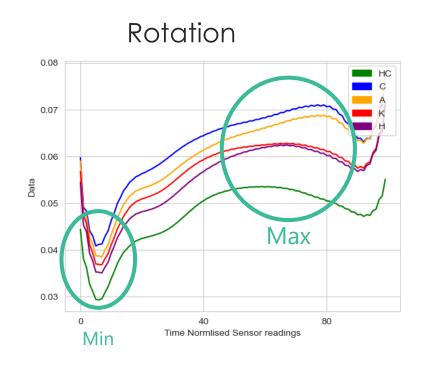
#### Ground Reaction Forces Measurements



**300** readings to **8** engineered features

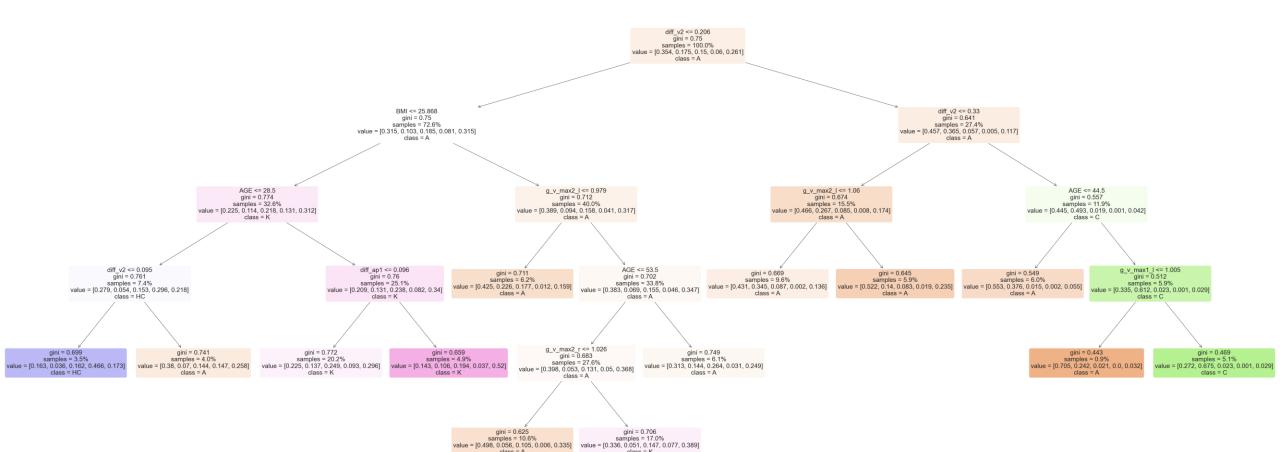
#### Center of Pressure Measurements





**200** readings to **3** engineered features

#### Decision Tree Classifier with 12 selected features



### Training History of Neural Network

