

A Machine Learning approach to predict future falls for older care home residents across Japan: a collaborative study with a care company

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Introduction

- ❑ Aging and population decline -> major socio-economic problems in Japan ¹
- ❑ Increasing falls -> big challenge for elderly people ²
- ❑ Many elderly live in care homes in Japan ³
- ❑ Falls and related injuries are on the rise in care homes ⁴
- ❑ Goals -> collaborative project to predict future resident fallers (& non-fallers)
 - ❑ Possible ergonomic changes in care homes and updating caregivers' training programs

Conclusion

- ✓ Random Forests (RF) are better than Decision trees and most other algorithms
- ✓ 4 models are made – full model and models with reduced (topmost) predictors
- ✓ 37 predictors, 36 predictors, top 30 and top 20 predictors models
- ✓ Early & successful tool to do fall-related screening for incoming care home residents
- ✓ SHAPley (SHAP) plots ⁵ explain the model predictions in an intuitive way for caregivers

Methods

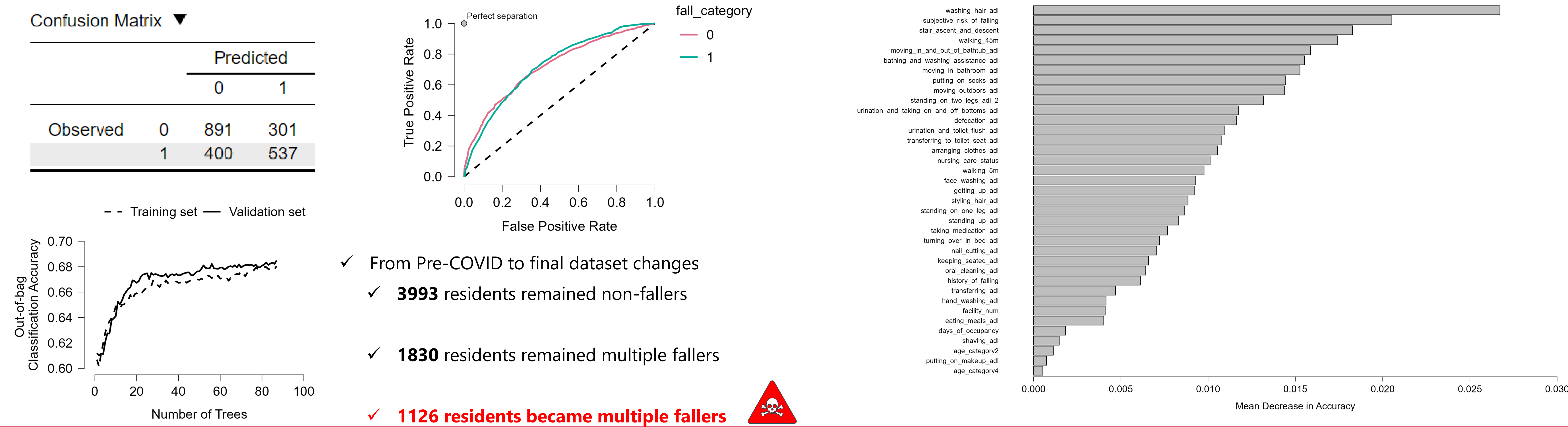
- ❑ N=10,648 residents in SOMPO Care database
- ❑ Average age -> 88 years old
- ❑ 2301 males and 8347 females (22%♂ and 78%♀)
- ❑ Dataset timeline ->
 - ❑ April 2019 - March 2020 (aka Pre-COVID zone)
 - ❑ August 2021 - July 2022 (last 1 full year of available data)
 - ❑ 6949 common residents
- ❑ Completely anonymized and delivered to AIST
- ❑ Non-fallers (0) & multiple fallers (1) -> binary classification problem

- ❑ Modeling algorithms considered -> Decision trees, Random forests, kNN, Naïve Bayes, Boosting, Light GBM
- ❑ 60% used for training, 20% for validation & 20% for pure testing
- ❑ Predictors used ->
 - ❑ mostly Activities of Daily Living (ADLs) of residents, especially if they can really do the activities or not
 - ❑ Nursing care status, subjective risk of falling, history of falling, facility type of residents, age-groups, among others
 - ❑ Predominantly categorical, also a few ordinal and numeric
 - ❑ One-hot encoding was done, as needed

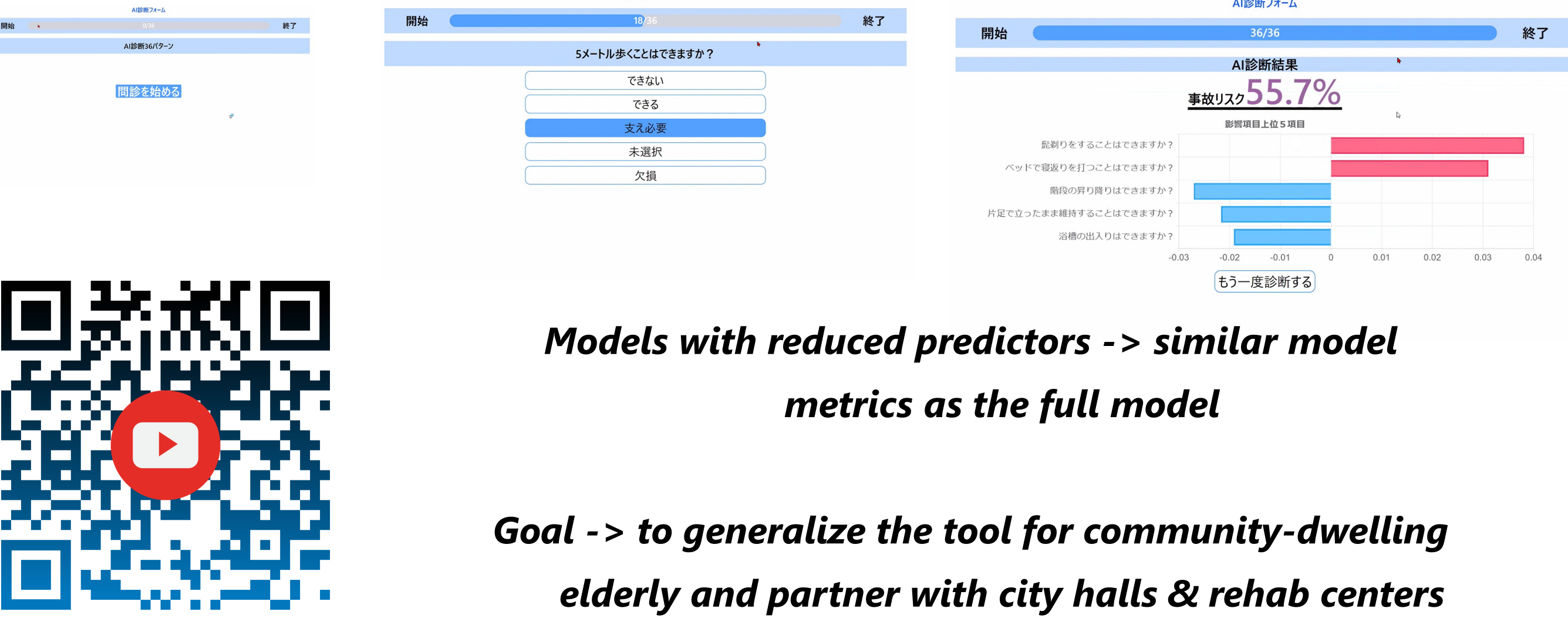
Predominantly from ADLs., we can identify future fallers (& non-fallers) with ~70% accuracy

Results

Better model metrics than ⁶ -> 66% specificity, 67% accuracy, precision, recall, F1 score, NPV and 73% AUC (averaged over both classes) -> test set, unseen during model training



Discussion and Future Work



References

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