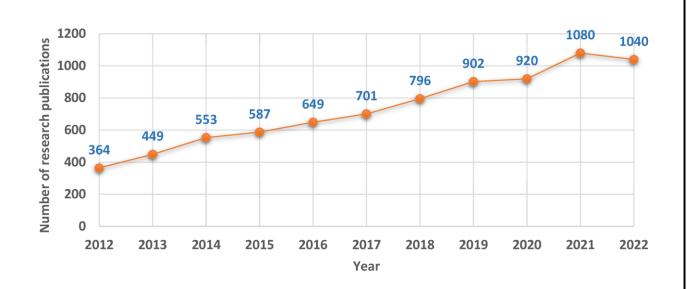


# DATA SCIENCE CHALLENGE | ACTIVITY RECOGNITION

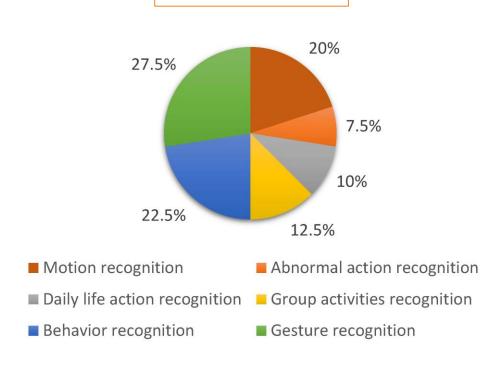
Pedro Matias

#### **MOTIVATION**

### Number of publications over time



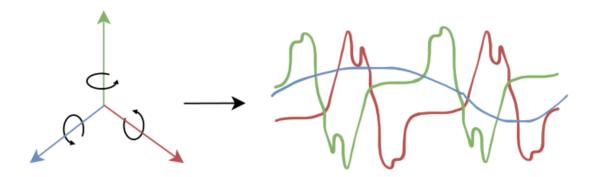
### Downstream Task



2

**PROBLEM** 

Accelerometer Data



Target Daily Activities

Sitting



Cycling



Walking



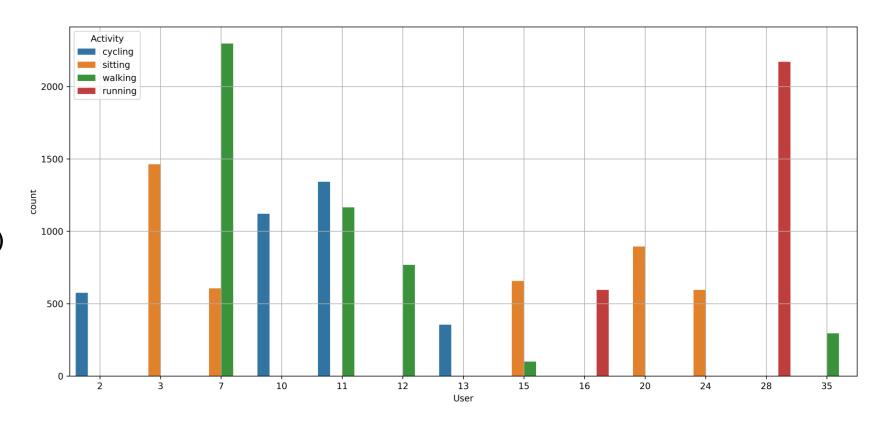
Running



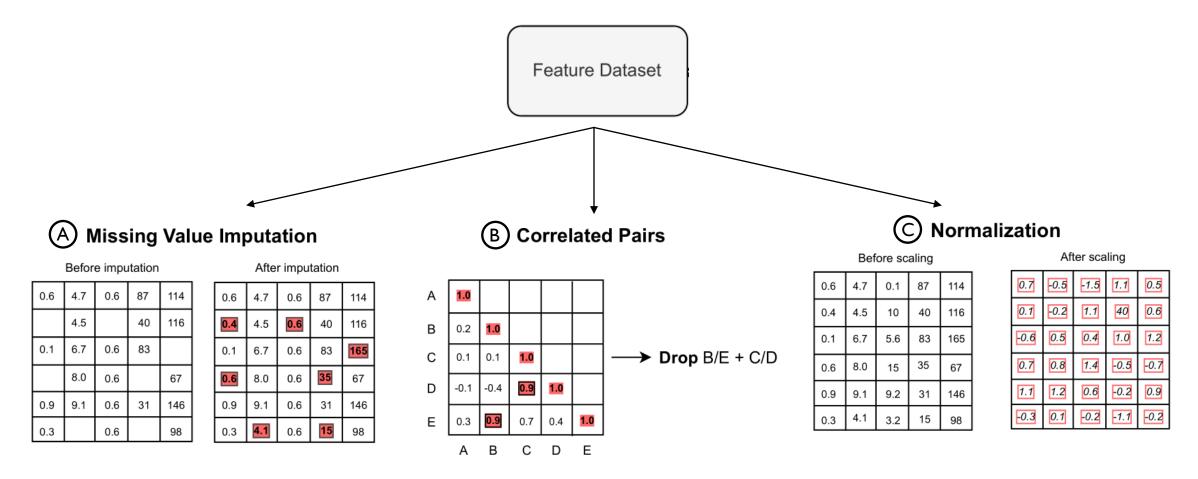
#### **DATASET**

### Characteristics

- A. 13 users
- **B.** 24 sessions
- C. 4 human activities
- **D.** 10.7 ± 4.4 min (duration)
- **E.** 15 000 samples
- **F.** 40 features



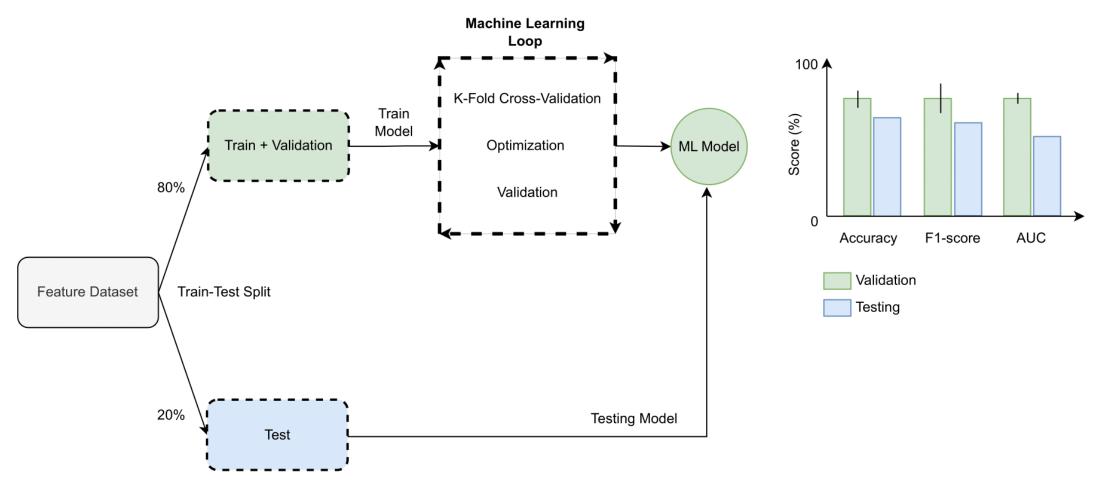
#### PREPROCESSING PIPELINE



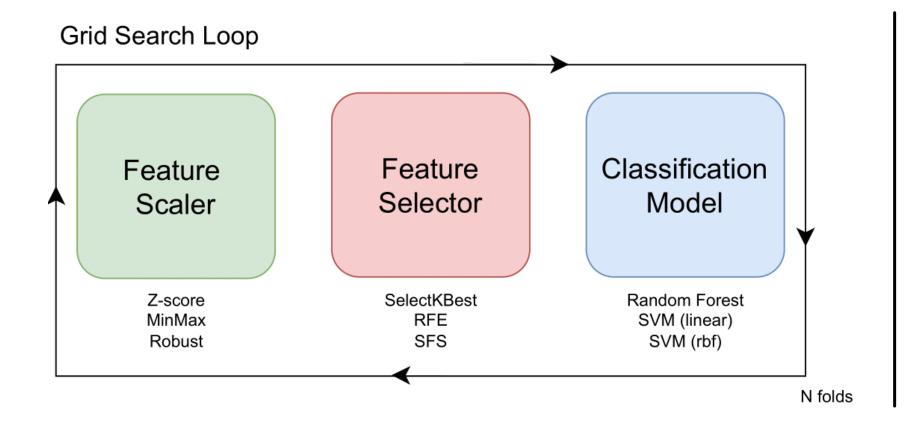
PEDRO MATIAS | DATA SCIENCE CHALLENGE

5

### ML PIPELINE



**GRID-SEARCH PIPELINE** 



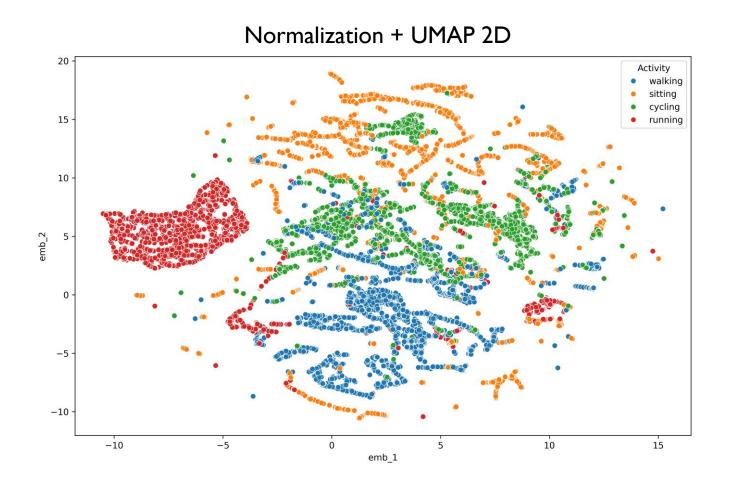
### **Notes**

5 folds (80/20%)

Split by Session

Optimization by FI-score (macro)

#### **EXPLORATORY DATA ANALYSIS**



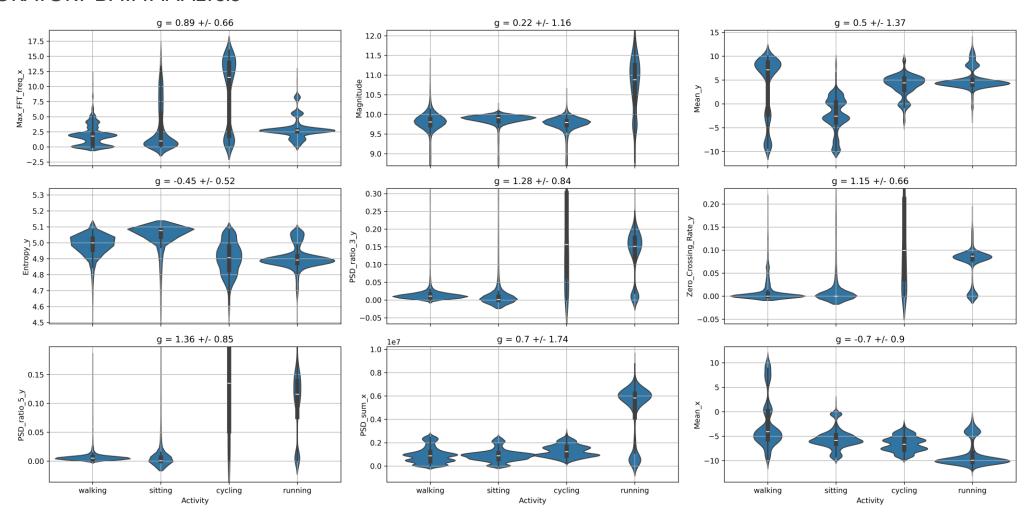
### **Notes**

Running potentially easier to identify;

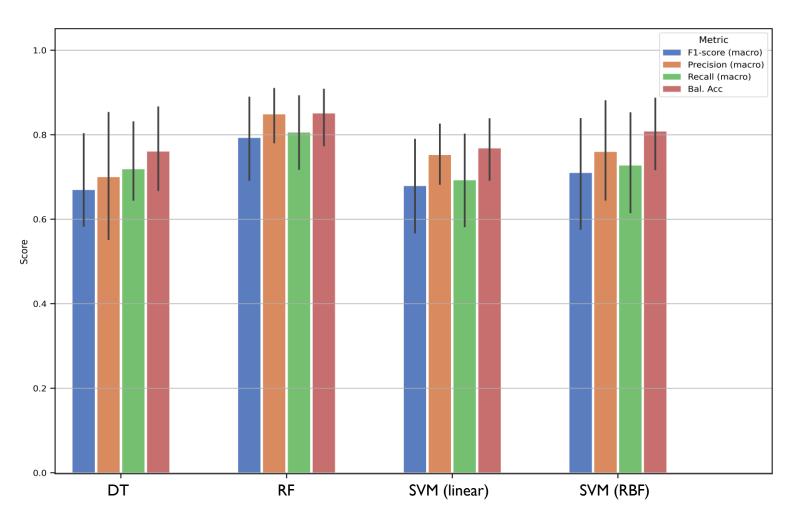
Walking and Sitting seem to be distinguishable as well;

**Cycling** and **Sitting** seem to be more overlapped. Physiological component can play and important role here.

#### **EXPLORATORY DATA ANALYSIS**



#### ML CLASSIFICATION RESULTS

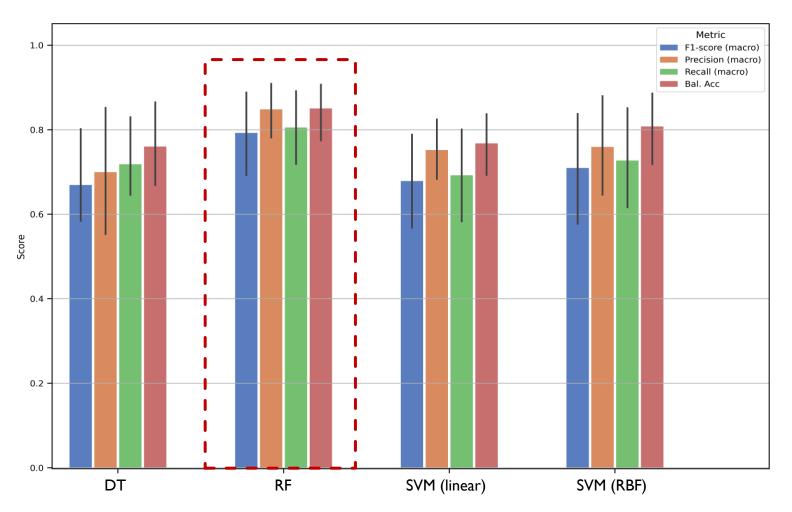


### **Remarks**

**Random Forest** performs better (79.2% ± 12.5% FI-score)

**To deploy**: model should be trained with all data, and a single external testing set should be used to report.

#### ML CLASSIFICATION RESULTS



### **Remarks**

**Random Forest** performs better (79.2% ± 12.5% FI-score)

**To deploy**: model should be trained with all data, and a single external testing set should be used to report.

#### PERFORMANCE IN REAL-WORLD SCENARIOS

- Performance of ML models in free-living scenarios often decreases when compared with validation scores;
- Some **features** are not **position-invariant.** This can affect real-world performances, if new incoming samples are collected with different device orientations;
- Reporting model performances transparently is crucial when deploying model into the real-world. Some examples: female vs male, young vs adult vs elderly, diseased vs healthy, device position, walking style;
- **To improve performance:** opening the context window (post-processing) may help reduce model failures (e.g., classification consistency over N consecutive windows; steps detected in walking windows; intra-user variability may help figure out most likely activities in specific times of day);
- Other approaches: position-invariant features (e.g., same metrics over signal magnitude), more activities (e.g., stand-to-sit/sit-to-stand; jumping; laying; indoors vs outdoors), ensemble methods or layered learning (model A to detect activity levels, model B to identify model ensembles), DL models (not in real-time but for offline processing).

# **THANK YOU**

## Contacts



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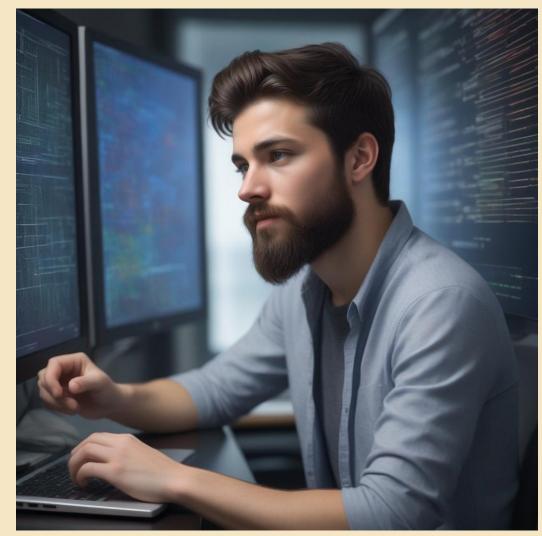
Pedro Matias



github/matiaspedro97



pedromatias



dataautogpt/OpenDalleV1.1