

DATA 1030 Final Project: Anomaly Detection of Falling People

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<https://github.com/plumol/data1030-final-project>

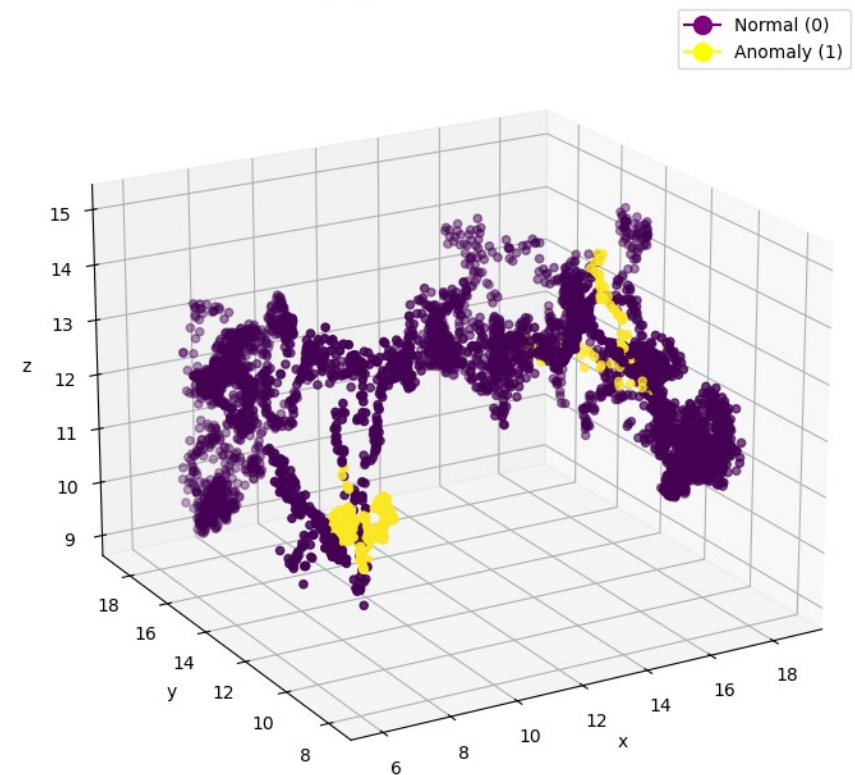
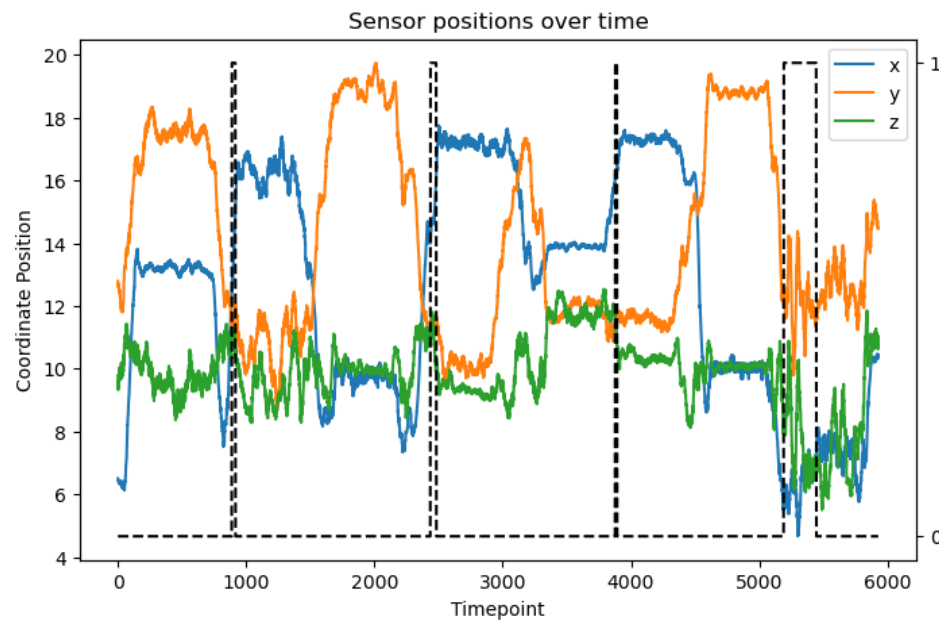
Dataset: Localization Data for Person Activity

- Anomaly detection: how can we predict for a very small positive class in a very unbalanced dataset?
- Background: This dataset^[1] was created to develop strategies for safer “smart” care environments for the elderly.
 - Contains positional data from 4 physical sensors (ankles, belt, and chest area) performing various physical tasks.
 - Labels for “falling” are considered positive class while all other activities (walking, standing, sitting, laying down) are negative
- Problem: Classification!

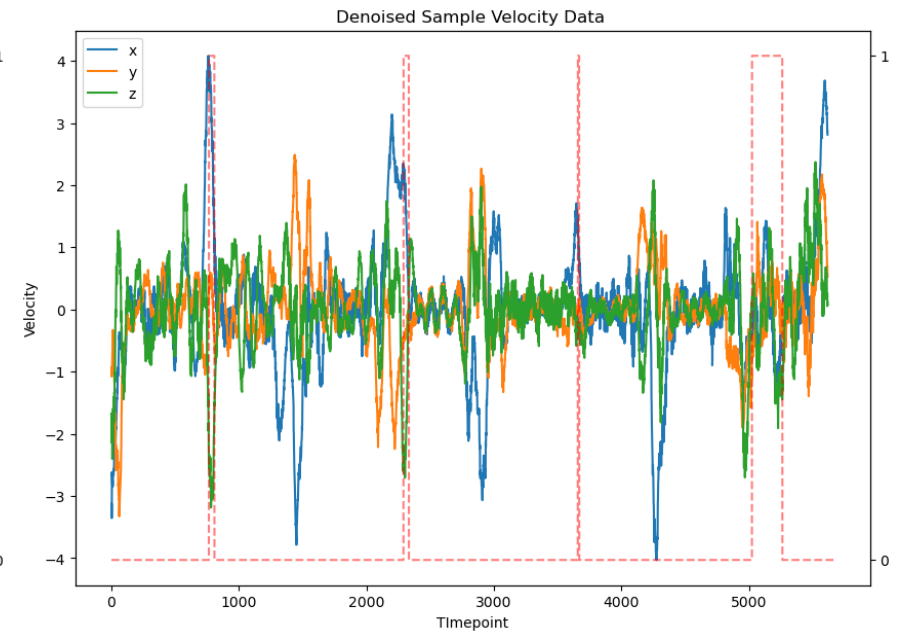
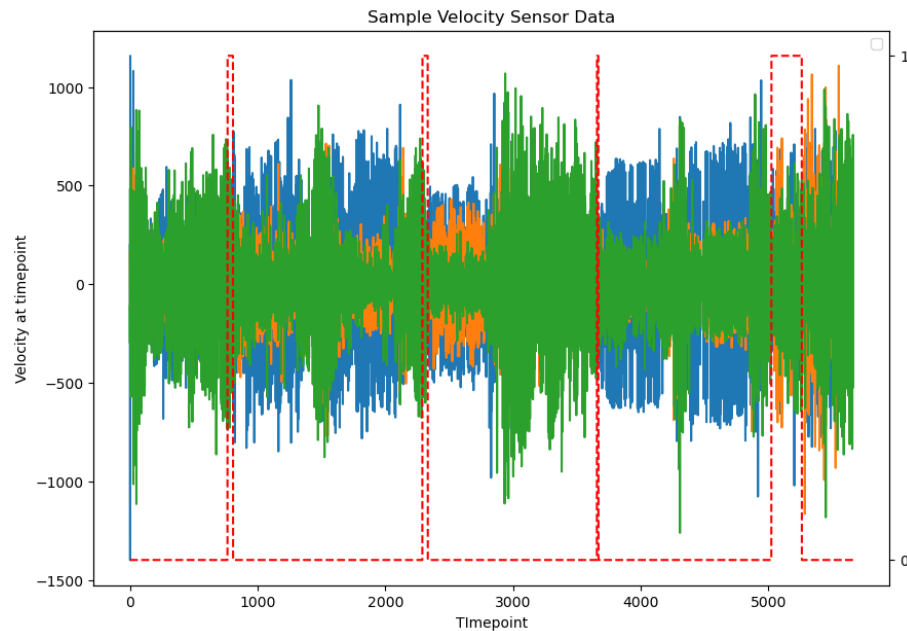
[1] Vidulin, V., Lustrek, M., Kaluza, B., Piltaver, R., & Krivec, J. (2010). Localization Data for Person Activity [Dataset]. UCI Machine Learning Repository. <https://doi.org/10.24432/C57G8X>.

Recap EDA- Sensor Positions (2D + 3D)

3D projection of sensor data



Recap EDA- Sensor Velocity and Dataset



$$V = \frac{x_2 - x_1}{dt} \quad dt = 0.027s$$

$$MA = \frac{1}{k} \sum_{\{i=n-k+1\}}^n p_i$$

k = window_size
p_i = value at i

Cross Validation and Model Training

- Dataset is split into 25 total users: 20 training/ 5 testing + shuffle for CV and 0.8 train-test-split

| Type | Users | Samples | Normal % | Anomaly % |
|----------|-------|---------|-------------------|----------------|
| Training | 20 | 134229 | 95.1% (127656) | 4.9% (6573) |
| Testing | 5 | 30030 | 94.6% (28420) | 5.4% (1610) |

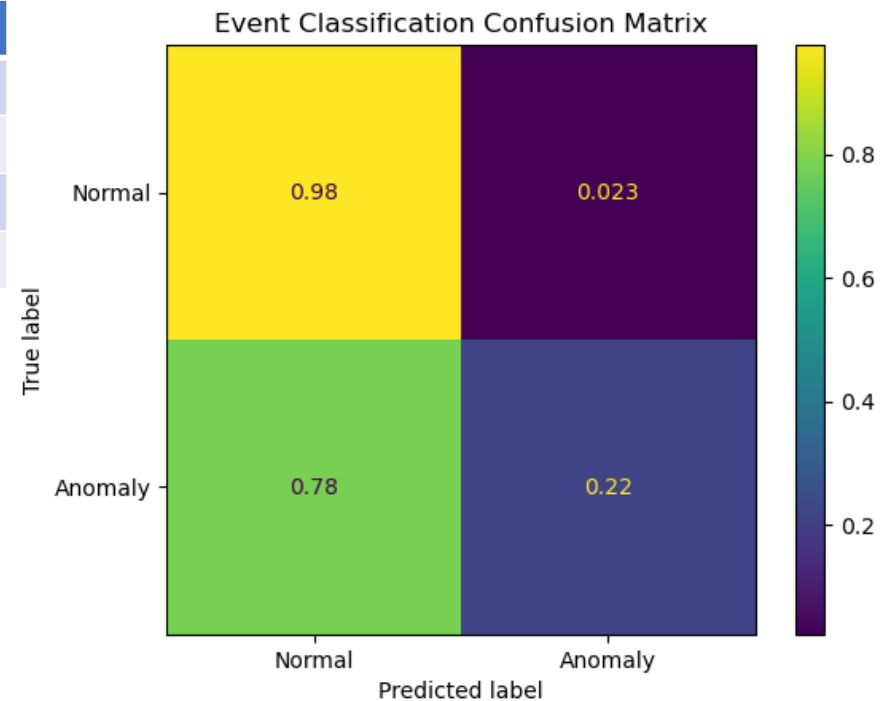
| Algorithm | Parameters |
|--------------------|---|
| LogisticRegression | C = [0.1, 1, 10] |
| RandomForest | max_features = [0.25, 0.5, 1.0] Max_depth = [5, 10, 30] |
| KNearestClassifier | n_neighbors = [1, 5, 10, 30] weights = ['uniform', 'distance'] |
| XGBoostClassifier | n_estimators = [100, 500, 1000, 2000] max_depth = [3, 4, 5] |

Results

| Model | F1 score mean | F1 score std |
|--------------------------|---------------|---------------|
| LogisticRegression | <0.0000 | <0.0000 |
| RandomForest | 0.2562 | 0.0154 |
| KNearestClassifier | 0.2011 | 0.0112 |
| XGBoostClassifier | 0.2857 | 0.0272 |

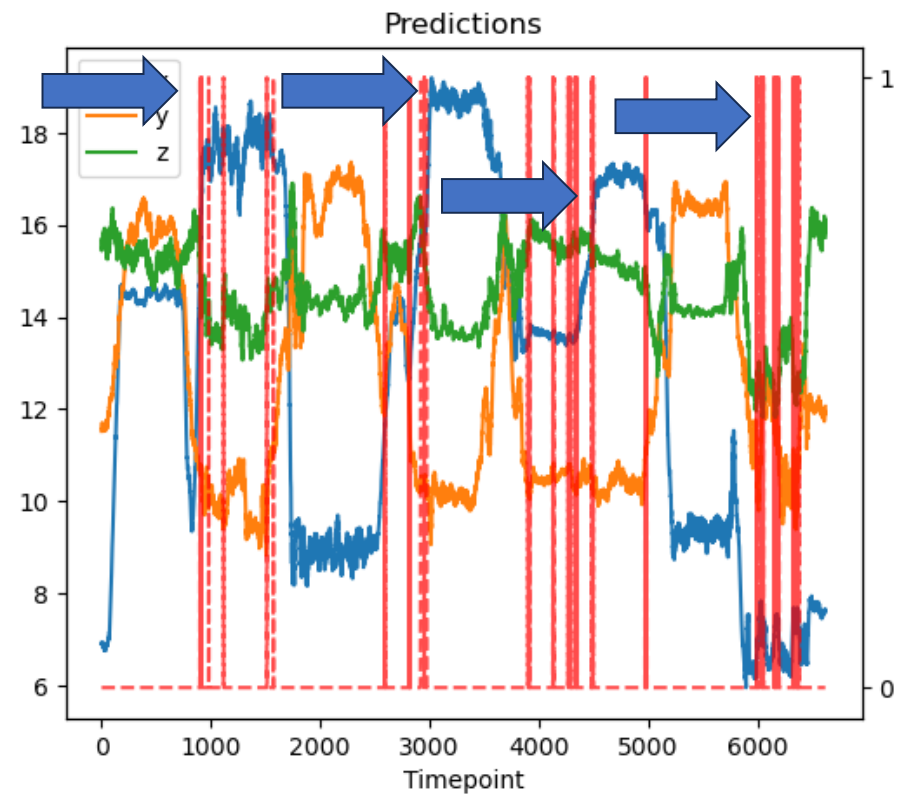
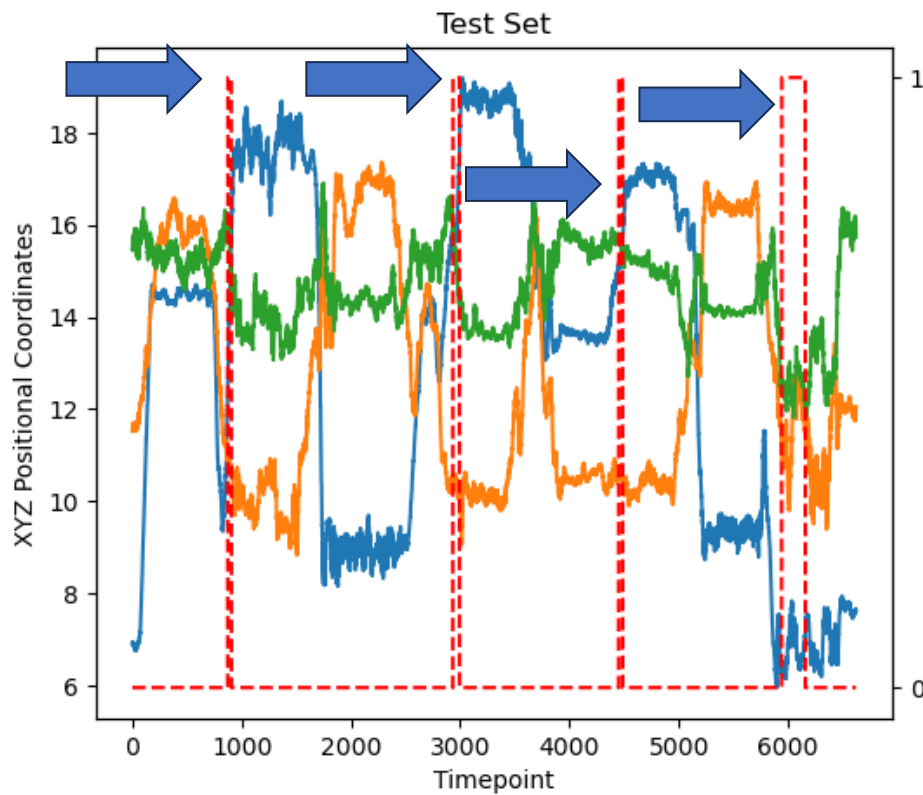
| Model | Best Parameters |
|--------------|---|
| RandomForest | max_depth = 30 max_features = 1.0 |
| KNN | n_neighbors = 5 weights = 'distance' |

| Model | Best Parameters |
|--------------------------|--------------------------------------|
| XGBoostClassifier | n_estimators = 2000 max_depth = 5 |

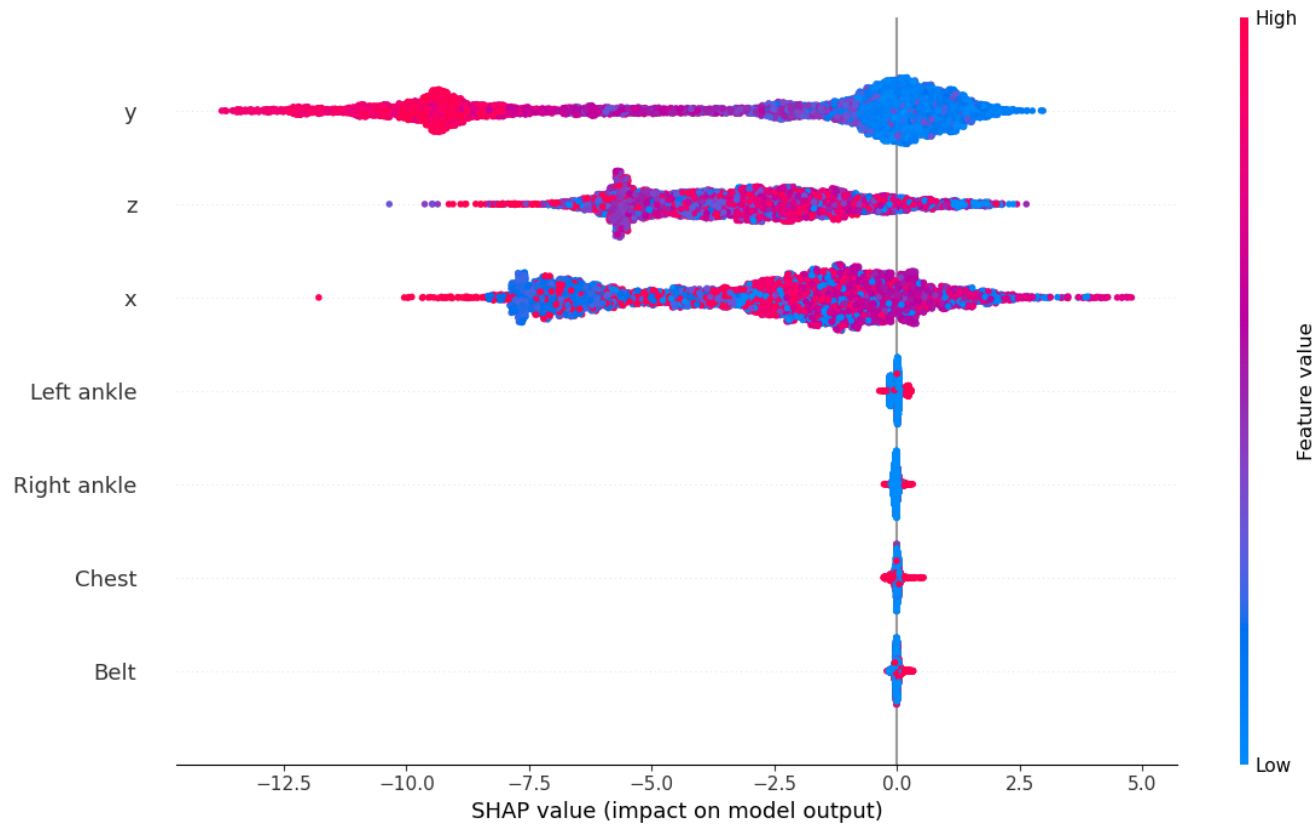


Predictions on a test dataset

Test Data vs Predictions



Interpretability: SHAP Values



Future Directions

- Alternative training: velocity data or acceleration data may be better predictors for 'falling' events
 - May produce more meaningful interpretable results on SHAP
- More complex models: deep neural networks and specialized architectures (ex. Deep Autoencoding Gaussian Mixture Model)