Protocol Weka

TODO SCHRITTE VERVOLLSTÄNDIGEN!

# Feature Selection

## Feature Selection: 2-State

1. CFSEval verwenden (Auf ganzem Training Set, da sonst der Merit Score nicht angezeigt wird):

Führt zu 0 Merit. Interpretation: CfsEvalSubset erkennt keine „Korrelation“ (bzw. „Relatedness“) zwischen den Features und der Zielvariable.

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1. Ausweichen auf Wrapper:

**Annahme**: 10s mit NaiveBayes funktionieren am besten.

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### Selected Features:

1. delta
2. theta/gamma
3. alpha/gamma
4. num\_of\_peaks
5. mean\_temp

## Feature Selection: 5-State

1. CFSEval verwenden (Auf ganzem Training Set, da sonst der Merit Score nicht angezeigt wird):

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Führt zu 0.173 Merit (Sehr gering) und wählt nur ein Feature aus „theta/gamma“. Interpretation: CfsEvalSubset erkennt kaum „Korrelation“ (bzw. „Relatedness“) zwischen den Features und der Zielvariable.

1. Ausweichen auf Wrapper:

**Annahme**: 10s mit NaiveBayes funktionieren am besten.

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### Feature Selection:

1. alpha/gamma
2. max\_peak\_amplitude
3. pnn50

# Building Classifier

## 2-State

**Performance Majority Classifier (ZeroR):** 60.9756 %

Performance Naive Bayes: 74,3902% (10-k CV / Seed = Matrikel Nummer)

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## 5-State

**Performance Majority Classifier (ZeroR):** 35.3659 %

Performance Naive Bayes: 46.3415 % (10-k CV / Seed = Matrikel Nummer)

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# Manually Creating Leave One Subject Out Folds

pID: Needs to be converted from Numeric to Nominal first! (ID:39 in the Screenshots points at Column with pID. (Note: The column ID changes after feature selection.)

Train Set: All participants except one. Test Set: Left out participant.

*Repeat for 10 participants.*

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Figure 1 Filtering for Test Set Figure 2 Filtering For Train Set

Remove pID Column.

Saving 10 Training Sets and 10 Test Sets.

Repeat for 5 State Data Set.

# Validation

## Comparing Classification Results

For validation of the classifier performance, it is unclear what the base performance is. From the original paper it seems as if they took the performance of a majority classifier being used once as the baseline.

Therefore, I assume they did a one-sided T-test when testing for a significant difference in performance between the baseline and the accuracy samples of the cross-validations.

Another possible approach, would have been to build a majority classifier for each cross validation fold as a baseline model and do a (corrected) two-sided T-Test instead.

## Repeated Cross Validation

10 times 10-Fold CV

Validating the Naïve Bayes Classifier against a Majority Vote Classifier (ZeroR)

Source Code:

* PerInstanceValidation2State.java
* PerInstanceValidation5State.java

### 10 times 10-Fold CV Results: 2-State

### 10 times 10-Fold CV Results: 5-State

## Per Participant Cross Validation

(Aka Leave One Subject Out)

Source Code:

* PerParticipantValidation2State.java
* PerParticipantValidation5State.java

### Leave One Subject Out CV Results: 2-State

### Leave One Subject Out CV Results: 5-State

## Summary of Cross Validation Results

(\*One Sample T-Test P-Value < 0.05)

|  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- |
| **States** | **Per Instance CV** | | | **Per Participant CV** | | | **Majority Classifier** |
| *Accuracy* | *Cohen’s Kappa* | *Stdev* | *Accuracy* | *Cohen’s Kappa* | *Stdev* |
| Two | 76%\* | 0.48 | 13.1 | 71% | 0.11 | 16.4 | 60.98% |
| Five | 46\* | 0.26 | 14.8 | 48% | 0.22 | 21.6 | 35.37% |

**Results of the Original Study:**

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Comparing the results of the original study with my own, although there is a comparable improvement on baseline accuracy, in my validation the accuracy varied far stronger in the repeated cross validation (Stdev[2-State]: 13,5 vs 0,7; Stdev[5-State]: 16,3 vs 2,9) This means that the final performance of our model may be only a bit better (or worse) than a simple majority classifier.

# Comparing Time Windows

Again, I will need to select the time windows based on wrapper-based attribute selectors, since CFSEval does not work on my Data Set.

I assume they validated all time windows using 10 times 10-fold Cross Validation since the points for the 10s time window on their comparison graphic match the values in the table from the repeated cross validation.

## Wrapper Attribute Selection Protocol

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Save filtered Data Set as:

"[Project Path]\Normalized Data\[Time Window]\with\_pID\normalized\_interruption\_data\_ALL\_tw\_[Time Window]s\_pID\_3hz\_[2state/5state]\_filtered\_attributes.arff"

## Attribute Selection (TWs)

### Attribute Selection 10s 2-State:

CFSEval (Train Set): (Merit: 0)

Wrapper:

1. delta
2. theta/gamma
3. alpha/gamma
4. num\_of\_peaks
5. mean\_temp

### Attribute Selection 20s 2-State:

CFSEval (Train Set): (Merit: 0,272)

1. alpha/gamma
2. max\_peak\_amplitude
3. sum\_peak\_amplitude

Wrapper:

1. theta
2. delta/theta
3. delta/gamma
4. theta/alpha
5. theta/gamma
6. num\_of\_peaks
7. mean\_temp

### Attribute Selection 30s 2-State:

CFSEval (Train Set): (Merit: 0)

Wrapper:

1. alpha
2. theta/gamma
3. mean\_hr\_in\_bpm
4. num\_of\_peaks
5. mean\_temp

### Attribute Selection 45s 2-State:

CFSEval (Train Set): (Merit: 0)

Wrapper:

1. theta
2. alpha
3. delta/gamma
4. mean\_hr\_in\_bpm
5. num\_of\_peaks
6. mean\_temp

### Attribute Selection 60s 2-State:

CFSEval (Train Set): (Merit: 0)

Wrapper:

1. beta
2. delta/alpha
3. delta/gamma
4. mean\_hr\_in\_bpm
5. sdnn
6. pnn50
7. max\_temp
8. num\_of\_peaks
9. mean\_temp

### Attribute Selection 120s 2-State:

CFSEval (Train Set): (Merit: 0)

Wrapper:

1. delta/alpha
2. theta/beta
3. gamma/alpha
4. num\_of\_blinks
5. hr\_variance\_in\_bpm
6. max\_temp

### Attribute Selection 10s 5-State:

CFSEval (Train Set): (Merit: 0.173)

1. theta/gamma

Wrapper:

1. alpha/gamma
2. max\_peak\_amplitude
3. pnn50

### Attribute Selection 20s 5-State:

CFSEval (Train Set): (Merit: 0)

Wrapper:

1. delta/alpha
2. theta/delta
3. alpha/gamma
4. hr\_variance\_in\_bpm
5. max\_peak\_amplitude
6. sum\_peak\_amplitude
7. pnn50

### Attribute Selection 30s 5-State:

CFSEval (Train Set): (Merit: 0)

Wrapper:

1. theta/delta
2. hr\_variance\_in\_bpm

### Attribute Selection 45s 5-State:

CFSEval (Train Set): (Merit: 0)

Wrapper:

1. delta/theta
2. gamma/beta
3. num\_of\_peaks

### Attribute Selection 60s 5-State:

CFSEval (Train Set): (Merit: 0)

Wrapper:

1. beta/gamma
2. gamma/beta
3. num\_of\_peaks
4. sum\_peak\_amplitude
5. pnn20
6. mean\_temp

### Attribute Selection 120s 5-State:

CFSEval (Train Set): (Merit: 0)

Wrapper:

1. beta/gamma
2. gamma/beta
3. pnn50

### Attribute Selection 180s 5-State:

CFSEval (Train Set): (Merit: 0)

Wrapper:

1. theta/alpha
2. alpha/delta
3. gamma/beta
4. hr\_variance\_in\_bpm
5. max\_temp

## Results: Comparing Time Windows

As observed in the original study shorter time windows (<60s) perform better than longer ones. I would also have the same intuition as the original author, that the shorter time windows probably benefit from having less artifacts. As in the original study this trend is much weaker for the 5-State predictions.

|  |  |  |
| --- | --- | --- |
|  | **2-State** | **5-State** |
| **10s** | 76% | 46% |
| **20s** | 73% | 45% |
| **30s** | 76% | 44% |
| **45s** | 75% | 42% |
| **60s** | 75% | 41% |
| **120s** | 67% | 38% |
| **180s** | 66% | 42% |

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Automatisch generierte Beschreibung

Figure Graph from Original Paper