

Latent Space Unsupervised Semantic Segmentation - Appendix

1 HYPERPARAMETERS SEARCH AND SELECTION

Table S1 and Table S2 show the hyperparameters considered for the offline and online algorithms respectively.

Offline Models	Data Scalers	NW		TC		Step-Size		Autoencoder	Extraction Algorithm
	All Datasets	All datasets		EMG Datasets	UCI	All datasets		All Datasets	All Datasets
LS-USS	No Scaler	50	300	1000	800	-	-	Fully Connected Convolutional	REA LREA
	Standard Scaler	100	400	1500	1200				
	Robust Scaler	150	500	2000	1600				
	Min Max Scaler	200							
FLUSS	No Scaler	50	300	1000	800	-	-	-	REA LREA
	Standard Scaler	100	400	1500	1200				
	Robust Scaler	150	500	2000	1600				
	Min Max Scaler	200							
LFMD	No Scaler	50	300			25	200	Fully Connected Convolutional	REA LREA
	Standard Scaler	100	400	-	-	50	250		
	Robust Scaler	150	500			100	350		
	Min Max Scaler	200				150	500		

Table S1. Hyperparameters considered for the CPD algorithms in the offline setting. NW is the sub-sequence length. TC is the temporal constraint.

Online Models	Data Scalers	NW		TC		Step-size		Autoencoder	NW		Extraction Algorithm
	All Datasets	All datasets		EMG datasets	UCI	All datasets		All Datasets	All datasets		All Datasets
LS-USS (ϵ -real time)	No Scaler	50	300	1000	800	-	-	Fully Connected Convolutional	-0.5	-2.0	LTEA
	Standard Scaler	100	400	1500	1200				-1.0	-2.5	
	Robust Scaler	150	500	2000	1600				-1.5	-3.0	
	Min Max Scaler	200									
LS-USS - Online	No Scaler	50	300	1000	800	-	-	Fully Connected Convolutional	-0.5	-2.0	LTEA
	Standard Scaler	100	400	1500	1200				-1.0	-2.5	
	Robust Scaler	150	500	2000	1600				-1.5	-3.0	
	Min Max Scaler	200									
FLUSS (ϵ -real time)	No Scaler	50	300	1000	800	-	-	-	-0.5	-2.0	LTEA
	Standard Scaler	100	400	1500	1200				-1.0	-2.5	
	Robust Scaler	150	500	2000	1600				-1.5	-3.0	
	Min Max Scaler	200									
FLOSS	No Scaler	50	300	1000	800	-	-	-	-0.5	-2.0	LTEA
	Standard Scaler	100	400	1500	1200				-1.0	-2.5	
	Robust Scaler	150	500	2000	1600				-1.5	-3.0	
	Min Max Scaler	200									
LFMD	No Scaler	50	300			25	200	Fully Connected Convolutional	-0.5	-2.0	LTEA
	Standard Scaler	100	400	-	-	50	250		-1.0	-2.5	
	Robust Scaler	150	500			100	350		-1.5	-3.0	
	Min Max Scaler	200				150	500				

Table S2. Hyperparameters considered for the CPD algorithms in the online setting. NW is the sub-sequence length. TC is the temporal constraint. The threshold parameter is the threshold used in the LTEA extraction algorithm.

Tables S3 and S4 shows the hyperparameters selected for the offline and online algorithms after doing hyperparameter search. The selected parameters is the ones used for the comparisons between the different models in Section VI and VII.

UCI						
CPD Algorithm	Extraction Algorithm	Data Scaler	NW	TC	Step-size	Autoencoder
LFMD	LREA	Robust Scaler	50	-	250	Fully Connected
LS-USS	REA	Min Max Scaler	50	800	1	Fully Connected
FLUSS	LREA	Robust Scaler	50	1200	1	-
EMG Artificial						
CPD Algorithm	Extraction Algorithm	Data Scaler	NW	TC	Step-size	Autoencoder
LFMD	REA	Standard Scaler	200	-	500	Conv. Model
LS-USS	LREA	Standard Scaler	50	1500	1	Conv. Model
FLUSS	LREA	Standard Scaler	50	1500	1	-
EMG						
CPD Algorithm	Extraction Algorithm	Data Scaler	NW	TC	Step-size	Autoencoder
LFMD	LREA	Standard Scaler	50	-	100	Fully Connected
LS-USS	LREA	Robust Scaler	50	2000	1	Conv. Model
FLUSS	LREA	Standard Scaler	50	1500	1	-
Dance Artificial						
CPD Algorithm	Extraction Algorithm	Data Scaler	NW	TC	Step-size	Autoencoder
LFMD	LREA	Min Max Scaler	100	-	100	Fully Connected
LS-USS	LREA	No Scaler	50	800	1	Conv. Model
FLUSS	LREA	Standard Scaler	50	800	1	-
Dance						
CPD Algorithm	Extraction Algorithm	Data Scaler	NW	TC	Step-size	Autoencoder
LFMD	LREA	Min Max Scaler	400	-	100	Fully Connected
LS-USS	LREA	Min Max Scaler	400	1600	1	Fully Connected
FLUSS	REA	Robust Scaler	400	800	1	-

Table S3. Hyperparameters selected for the compared CPD algorithms in the offline setting. NW is the sub-sequence length. TC is the temporal constraint.

2 TIME SERIES SEGMENTATION EXAMPLES

UCI							
CPD Algorithm	Extraction Algorithm	Data Scaler	NW	TC	Step-size	Autoencoder	Threshold
LFMD	LTEA	Min Max Scaler	100	-	100	Fully Connected	-2
FLUSS	LTEA	Robust Scaler	50	1600	1	-	-1
FLOSS	LTEA	Robust Scaler	150	1600	1	-	-1
LS-USS	LTEA	Min Max Scaler	150	800	1	Fully Connected	-1.5
LS-USS Online	LTEA	Robust Scaler	50	1200	1	Fully Connected	-0.5
EMG Artificial							
CPD Algorithm	Extraction Algorithm	Data Scaler	NW	TC	Step-size	Autoencoder	Threshold
LFMD	LTEA	Standard Scaler	300	-	100	Fully Connected	-0.5
FLUSS	LTEA	Standard Scaler	50	1500	1	-	-1
FLOSS	LTEA	Standard Scaler	100	2000	1	-	-0.5
LS-USS	LTEA	Standard Scaler	50	2000	1	Conv. Model	-0.5
LS-USS Online	LTEA	Standard Scaler	50	1500	1	Conv. Model	-1
EMG							
CPD Algorithm	Extraction Algorithm	Data Scaler	NW	TC	Step-size	Autoencoder	Threshold
LFMD	LTEA	Standard Scaler	500	-	500	Fully Connected	-1.5
FLUSS	LTEA	Standard Scaler	50	1000	1	-	-2
FLOSS	LTEA	Standard Scaler	200	1000	1	-	-2
LS-USS	LTEA	Standard Scaler	300	1000	1	Fully Connected	-2
LS-USS Online	LTEA	Standard Scaler	200	1500	1	Fully Connected	-2
Dance Artificial							
CPD Algorithm	Extraction Algorithm	Data Scaler	NW	TC	Step-size	Autoencoder	Threshold
LFMD	LTEA	Min Max Scaler	50	-	50	Fully Connected	-0.5
FLUSS	LTEA	Min Max Scaler	100	800	1	-	-0.5
FLOSS	LTEA	No Scaler	50	800	1	-	-0.5
LS-USS	LTEA	Min Max Scaler	50	800	1	Conv. Model	-0.5
LS-USS Online	LTEA	Min Max Scaler	50	800	1	Conv. Model	-0.5
Dance							
CPD Algorithm	Extraction Algorithm	Data Scaler	NW	TC	Step-size	Autoencoder	Threshold
LFMD	LTEA	Min Max Scaler	125	-	200	Fully Connected	-1.5
FLUSS	LTEA	Robust Scaler	200	1200	1	-	-1.5
FLOSS	LTEA	Robust Scaler	125	1200	1	-	-2
LS-USS	LTEA	Robust Scaler	200	1200	1	Fully Connected	-2
LS-USS Online	LTEA	Min Max Scaler	200	1600	1	Fully Connected	-1.5

Table S4. Hyperparameters selected for the compared CPD algorithms in the online setting. NW is the sub-sequence length. TC is the temporal constraint. The threshold parameter is the threshold used in the LTEA extraction algorithm.

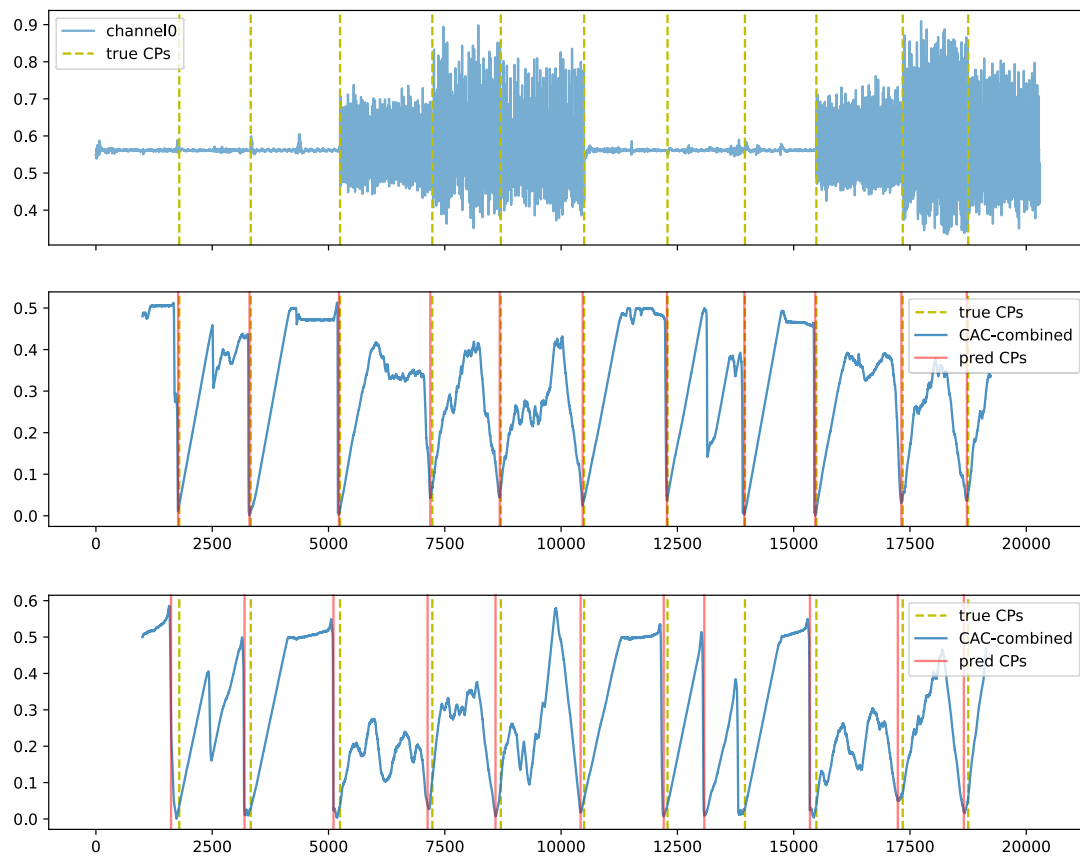


Figure S1. Top: Channel 0 of subject 4 in the UCI dataset. Middle: Segmentation using LS-USS – LREA- Bottom: Segmentation using LS-USS – LTEA

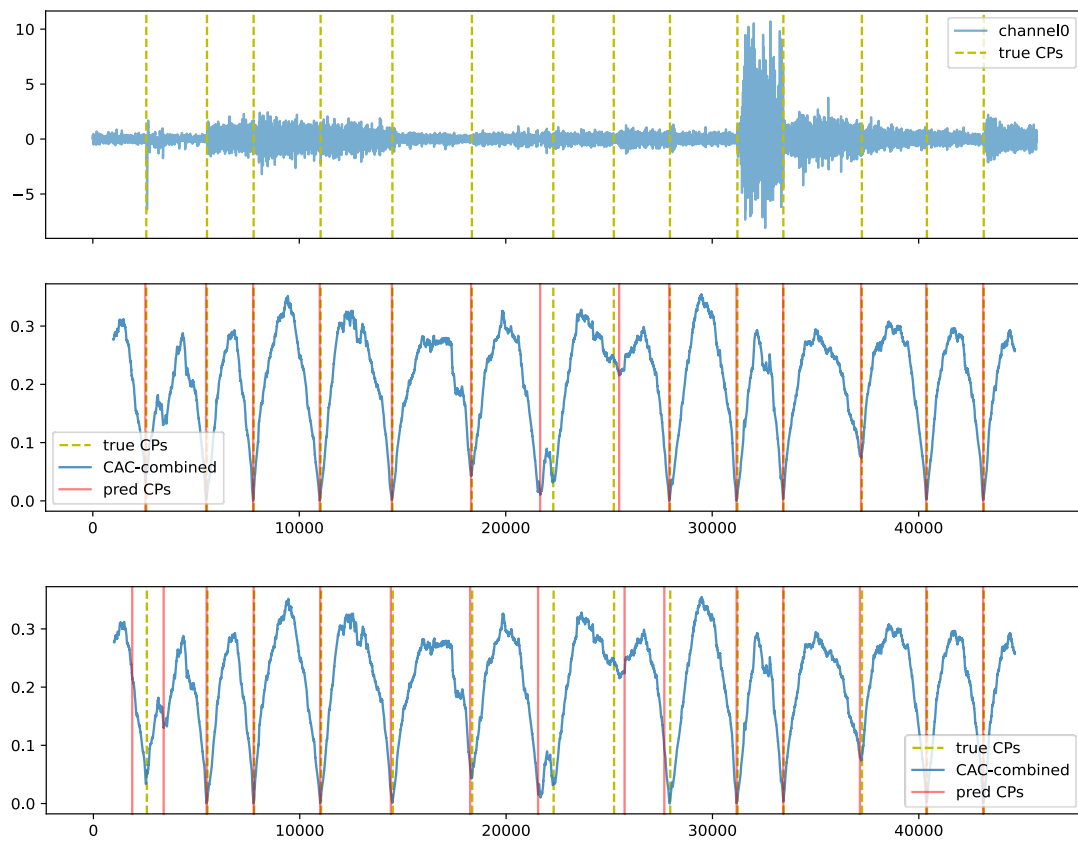


Figure S2. Top: Channel 0 of gesture19 in the EMG Artificial dataset. Middle: Segmentation using LS-USS – LREA- Bottom: Segmentation using LS-USS – LTEA.

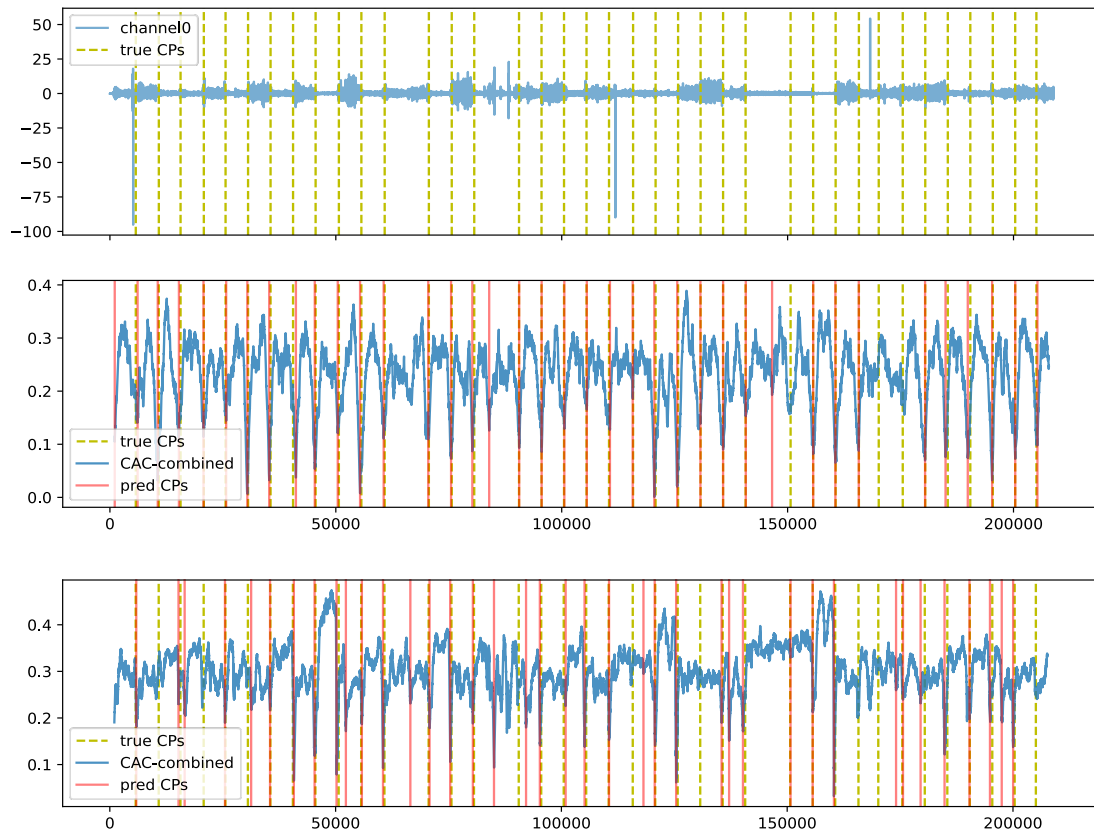


Figure S3. Top: Channel 0 of participant3_evaluation3 in the EMG dataset. Middle: Segmentation using LS-USS – LREA- Bottom: Segmentation using LS-USS – LTEA.

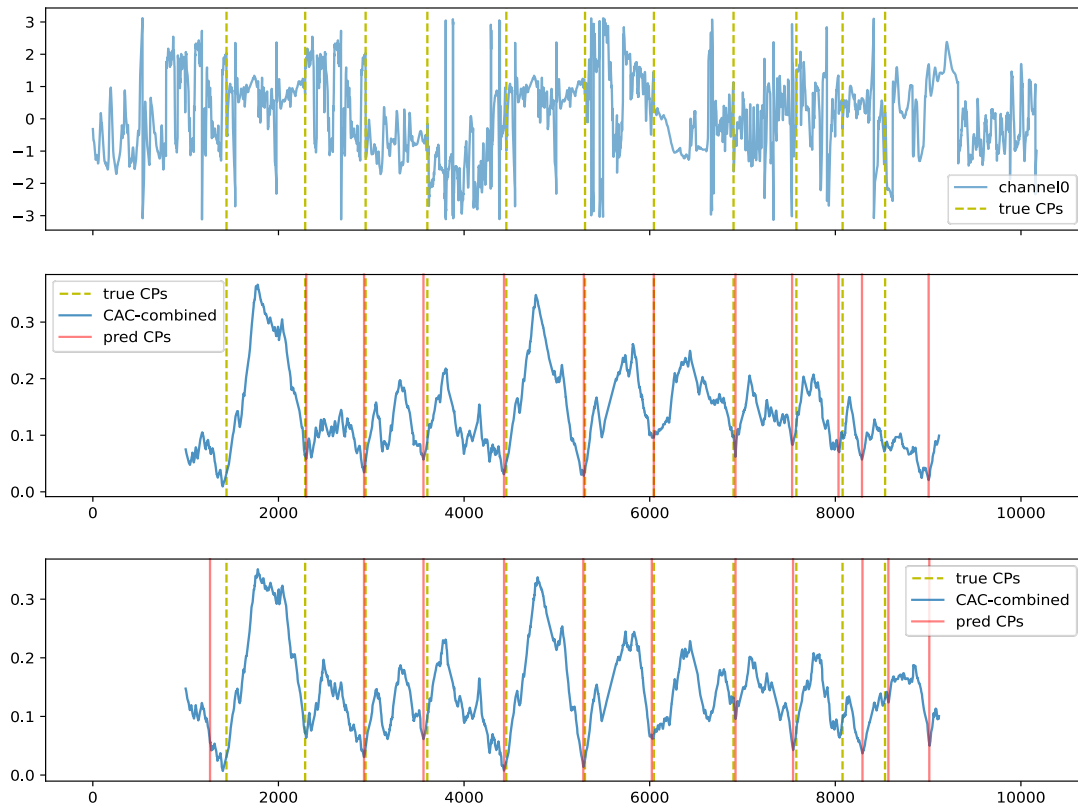


Figure S4. Channel 0 of d41 in the Dance Artificial dataset. Middle: Segmentation using LS-USS – LREA- Bottom: Segmentation using LS-USS – LTEA.

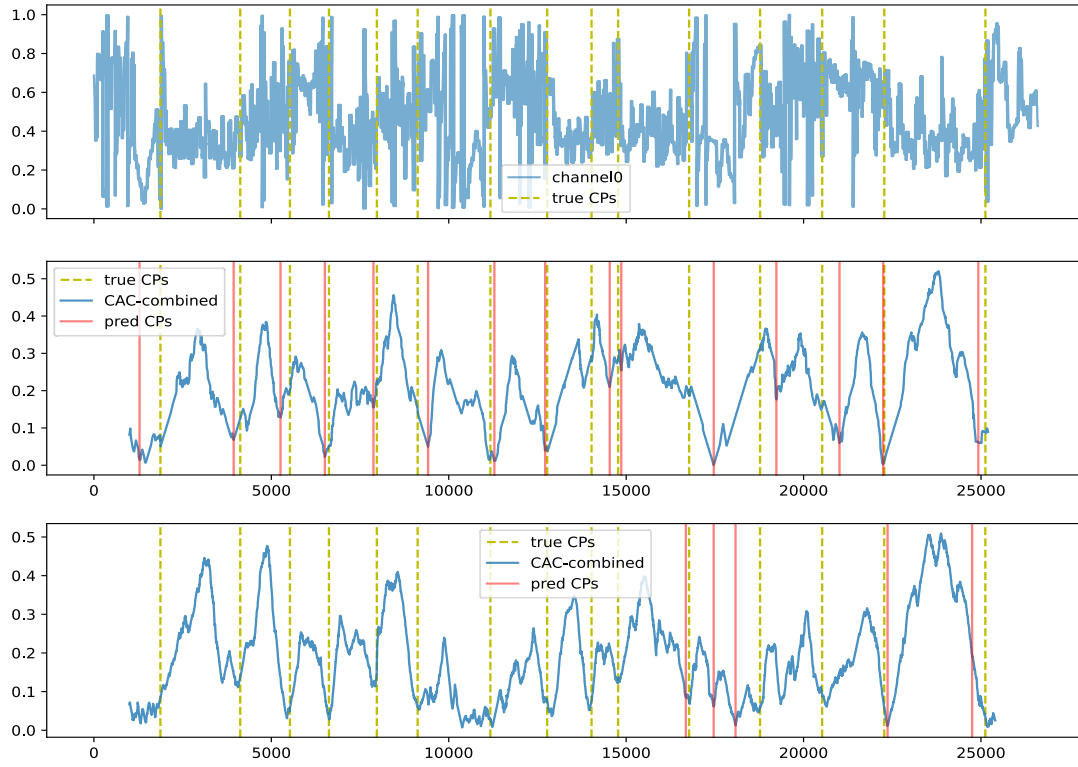


Figure S5. Top: Channel 0 of d24 in the Dance dataset. Middle: Segmentation using LS-USS – LREA- Bottom: Segmentation using LS-USS – LTEA (shows the effects of setting the threshold parameter to high).