



Università degli Studi di Cagliari, Italia

Department of Electrical and Electronic Engineering

Pattern Recognition & Application Lab

#### Electroencephalography signal processing based on textural features for monitoring the driver's state by a Brain-Computer Interface

Giulia Orrù, Marco Micheletto, Fabio Terranova, Gian Luca Marcialis





Milan, 10 | 15 January 2021



# The importance of vigilance

• Vigilance plays an essential role in most human activities, especially in driving.

- A considerable percentage of road accidents has been caused by sleep and fatigue.
- Drowsiness detection could give a massive benefit in the prevention of both non-fatal and fatal crashes.











Vehicle characteristics/ position

Driver eye/face monitoring

Physiological signals



#### Physiological signals: Electroencephalography (EEG)

• Most promising drowsiness indicator.

- Many approaches based on analysis in frequency domain.
- Delta, theta, and alpha activity characterizes the EEG signal during driver fatigue





#### SEED-VIG data set





Experimental setup: simulated driving scenario



Neuroscan system: channels configuration

Wei-Long Zheng and Bao-Liang Lu, A multimodal approach to estimating vigilance using EEG and forehead EOG. Journal of Neural Engineering, 14(2): 026017, 2017.





EEG-based Brain Computer Interface







EEG

Signal Processing

Preprocessing

Feature Extraction

Classification

Audio signal

Preprocessing

Butterworth bandpass filter

Feature Extraction

1D - Local Binary Pattern (Proposed method)

EEG-based Brain Computer Interface



(2)

#### Feature Extraction: 1D-Local Binary Pattern

$$LBP^{1D}(x[i]) = \sum_{r=0}^{\frac{P}{2}-1} \{sgn\left(x\left[i+r-\frac{P}{2}\right]-x[i]\right)2^r + sgn(x[i+r+1]-x[i])2^{r+\frac{P}{2}}\}$$
(1)

$$sgn = \begin{cases} 0, & x < 0\\ 1, & x \ge 0 \end{cases}$$

**P** represents the number of neighbouring samples thresholded against the centre one.















# Settings optimization: accuracy





# Generic user vs User-specific

Train dataset



Test dataset



Generic user



# Settings optimization: accuracy

#### best results highlighted







# New metrics for temporal response

Two distinct class transition considered:

Awake  $\rightarrow$  Tired (AT) Tired  $\rightarrow$  Drowsy (TD)

• *Hit rate* =  $\frac{H_{AT}}{n_{AT}} + \frac{H_{TD}}{n_{TD}}$ 

• 0 - delay hit rate

• Mean hit delay = 
$$\frac{\sum_{i=0}^{n_{AT}} \Delta_i^{AT} + \sum_{i=0}^{n_{TD}} \Delta_i^{TD}}{n_{AT} + n_{TD}}$$

Percentage of hits that happen on the first sample of the state change.

Mean of seconds of delay with which the response of the classifier correctly identifies a state change.

 $\Delta^{xy} = t(hit)_{xy} - t(change)_{xy}$ 

Rate of misclassified "awake" samples.

• False hit rate = 
$$\frac{FN_{(awake)}}{FN_{(awake)} + TP_{(awake)}}$$





#### Results: hit rate, 0-delay hit rate and mean hit delay





#### Results: False hit rate





# Conclusions



- 1D-LBP first application to driver's state monitoring.
- Introduction of novel performance parameters for the transition's detection and the related time delay.
- Strong effectiveness of the proposed method in detecting the awake-totired transitions (only 6 s of delay and the best hit rate).
- Overall performance is not yet good enough to develop a BCI for assessing the driver's vigilance in real environments.





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