

Techniques for EEG signal denoising

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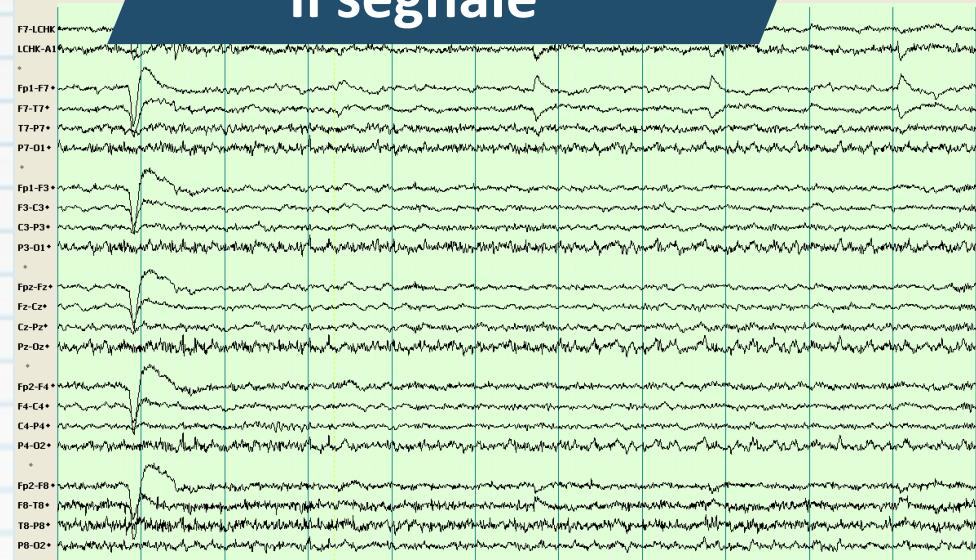
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Segnale EEG

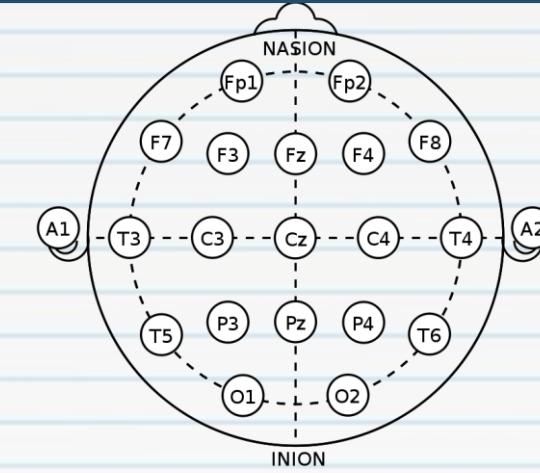
I sensori

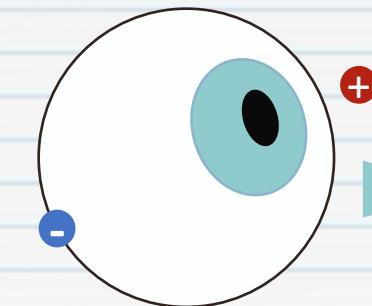


Il segnale

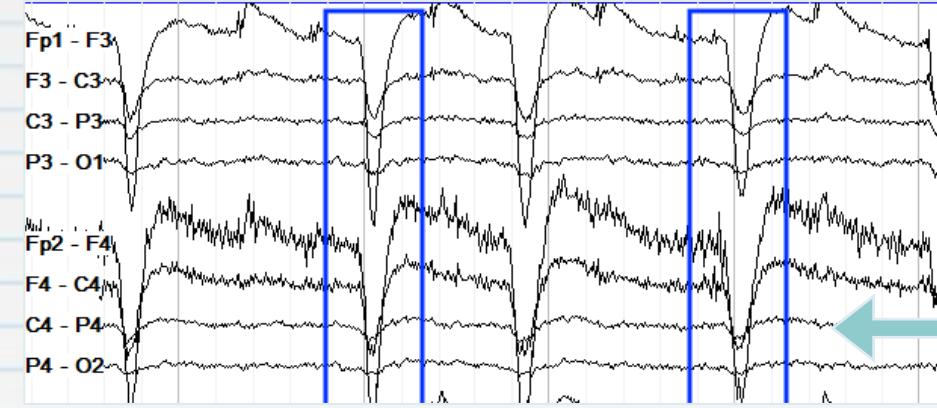


Lo standard





EOG – artefatti di origine oculare



Artefatti

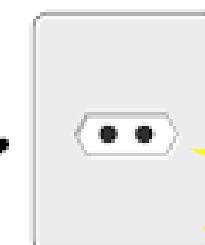


ECG o EKG – artefatti di origine cardiaca



EMG – artefatti di origine muscolare

artefatti dati da agenti esterni
(es. Malfunzionamento, corrente)



Artifact Avoidance

L' atto di guidare il soggetto a non eseguire movimenti che portano a contaminare il segnale (es. Non battere gli occhi, stare fermo, etc.).

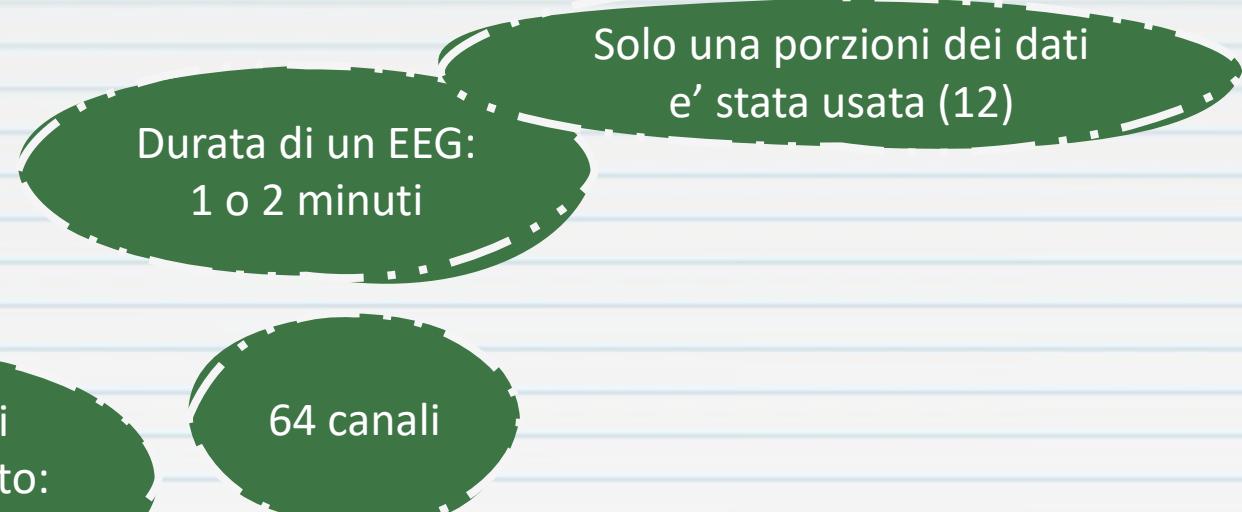
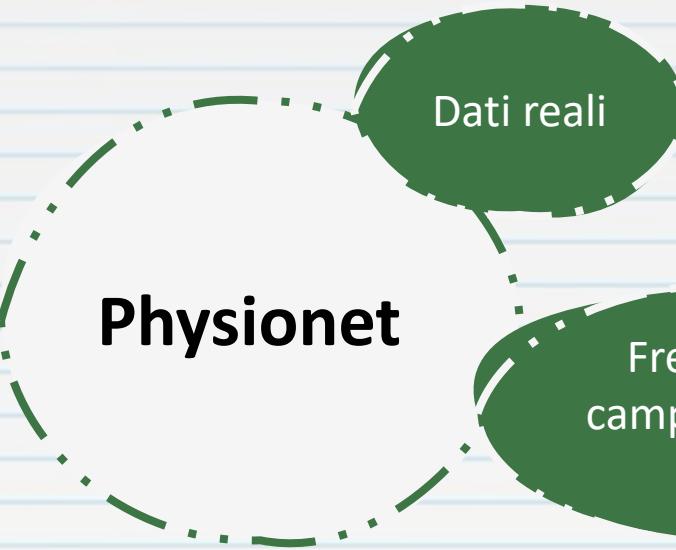
Artifact Rejection

L' atto di rifiutare le parti di segnale contaminate da artefatti mediante un controllo visuale da parte di un professionista o tramite procedure automatizzate.

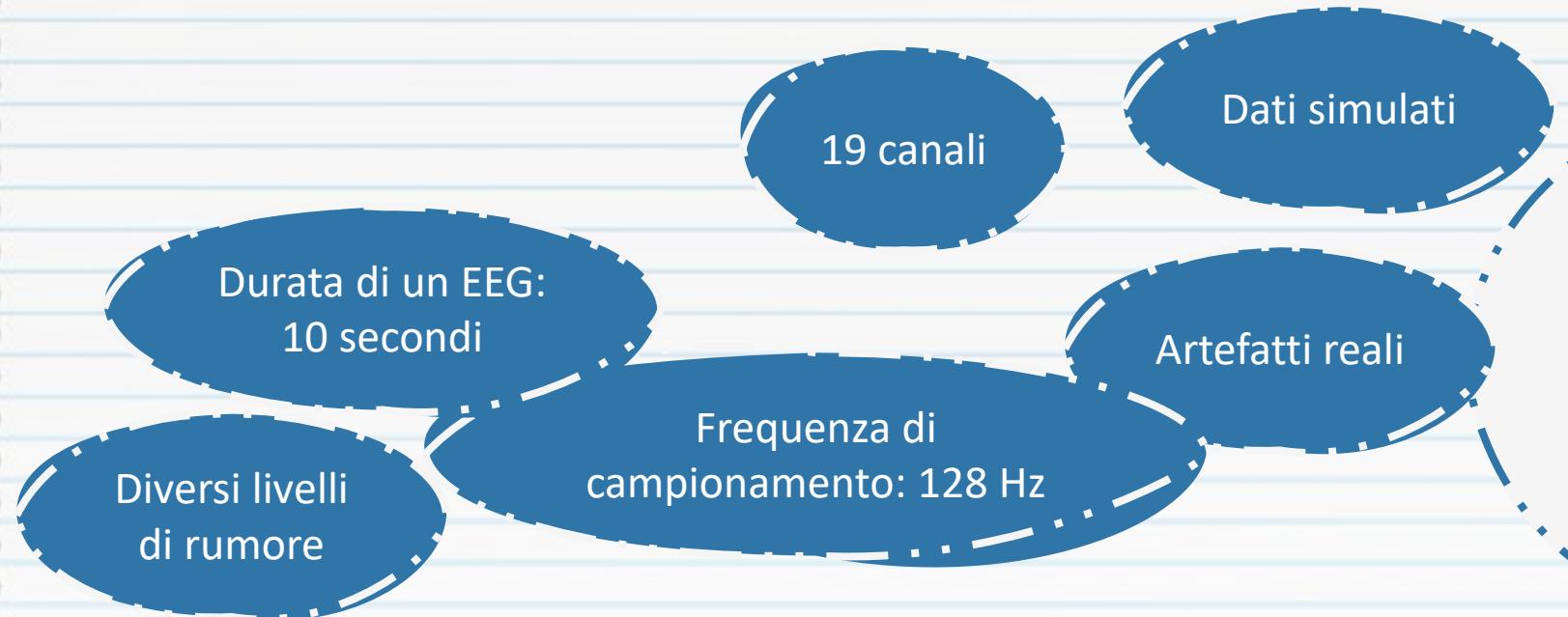
Rimozione degli artefatti

Tecniche per la rimozione di artefatti

- Filtraggio
- Regressione
- **Independent component analysis**
- **Empirical mode decomposition**
- **Wavelet transform**



I Data Set



**Data set
simulato^[1]**

Matlab

Indipendent Component Analysis (ICA)

Matlab extension EEGLAB

Classification ^[4]

EEGLAB extension ICLabel ^[4]

Empirical Mode Decomposition (EMD) - Threshold ^[3]

In Matlab: *emd()*

- Statistical Threshold
- Universal Threshold
- VK Threshold ^[3]

$$\lambda_k = P_k \text{std}(h_k)$$

$$P_k = e^{\frac{\lambda_k - S_k}{\lambda_k + S_k}}$$

$$S_k = \frac{\sum_k |h_k|}{N}$$

Wavelet Transform (WT)
– Threshold ^[2]

In Matlab: *wavedec()*, *waverec()*

- Statistical Threshold
- Universal Threshold

In Matlab: *wthcoef()*

Algoritmi

$$S = 1.5 * \text{std}(C_k)$$

Signal to noise ratio (dB)

$$SNR = 10 \log \left(\frac{\sum_{n=1}^N x(n)^2}{\sum_{n=1}^N y(n)^2} \right)$$

8.079
1.996

Coefficiente di correlazione

$$CC = \frac{\sum_{n=1}^N (x(n) - \text{mean}(x))(y(n) - \text{mean}(y))}{\sqrt{\sum_{n=1}^N (x(n) - \text{mean}(x))^2 \sum_{n=1}^N (y(n) - \text{mean}(y))^2}}$$

0.693
0.823

Root mean square error

$$RMSE = \sqrt{\frac{1}{N} \sum_{n=1}^N (x(n) - y(n))^2}$$

11.574
8.242

Mean Square Error

$$MSE = \frac{1}{N} \sum_{n=1}^N (y(n) - s(n))^2$$

30.445
134.733

EMD-ST
Vs
WT-ST

Metriche

Conclusioni

Indipendent components analysis (ICA):

- Presenta problemi con il data set simulato.
- Non sempre dalla scomposizione si ottengono le componenti artefattuali.

Wavelet Threshold:

- La rimozione degli artefatti oculari non e' ottimale.

Empirical mode decomposition (EMD) – Threshold:

- Presenta una buona rimozione dell'elemento artefattuale.

Sviluppi futuri

- Testare varianti delle tecniche proposte;
- Mirare alla rimozione di artefatti anche non oculari;
- Esplorare l'efficacia delle tecniche proposte anche in altri data set.



Grazie!

Bibliografia

- [1] Gallego-Jutglà, Esteve, et al. "Application of Multivariate Empirical Mode Decomposition for Cleaning Eye Blinks Artifacts from EEG Signals." *IJCCI (NCTA)*. 2011.
- [2] Khatun, Saleha, Ruhi Mahajan, and Bashir I. Morshed. "Comparative study of wavelet-based unsupervised ocular artifact removal techniques for single-channel EEG data." *IEEE journal of translational engineering in health and medicine* 4 (2016): 1-8.
- [3] Vijayasankar, A., and P. Rajesh Kumar. "Correction of blink artifacts from single channel EEG by EMD-IMF thresholding." *2018 Conference on Signal Processing And Communication Engineering Systems (SPACES)*. IEEE, 2018.
- [4] Pion-Tonachini, Luca, Ken Kreutz-Delgado, and Scott Makeig. "ICLabel: An automated electroencephalographic independent component classifier, dataset, and website." *NeuroImage* 198 (2019): 181-197.