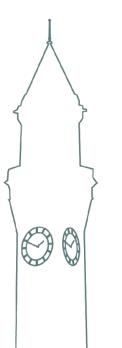
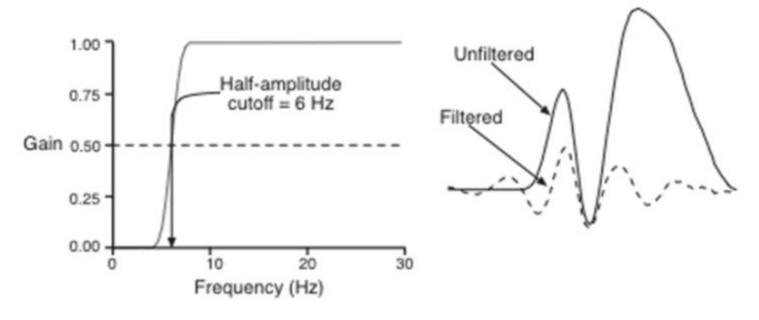
Can we investigate pre-stimulus activity just prior to t=0?

Steffen Buergers, Giulio Degano



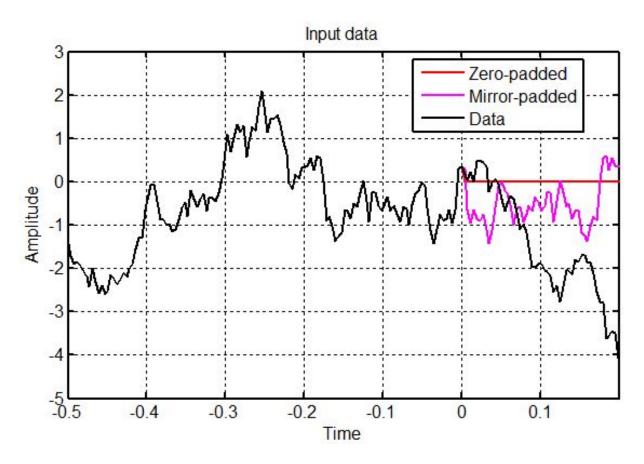
The problem of post-stim artifacts

Information bleeds into pre-stimulus period when analysing oscillatory activity!





Avoiding post-stimulus artifacts





Frequency sliding

Hilbert transform (or fft or wavelet)

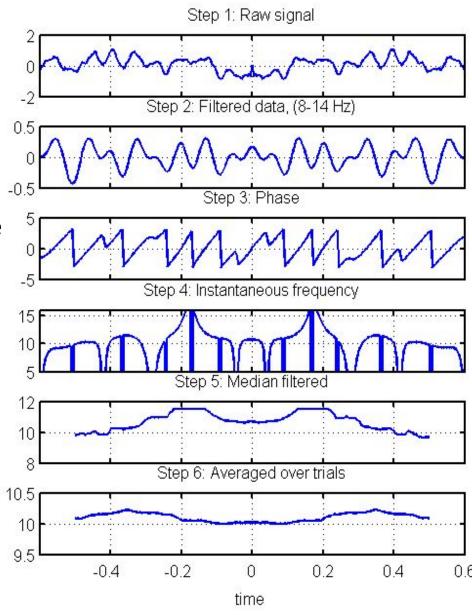
Take temporal **derivative of phase** angle time series

Apply 10 **median filters** with different orders

Take the **median** of the median filter values

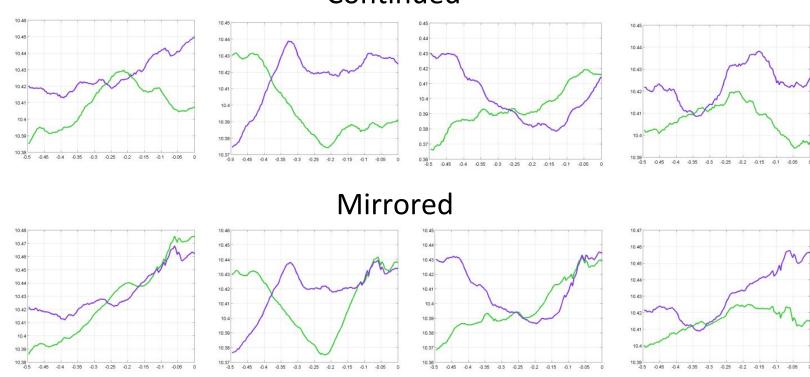
Average over trials





Frequency sliding (pink noise simulations, N=20, n=125)

Continued

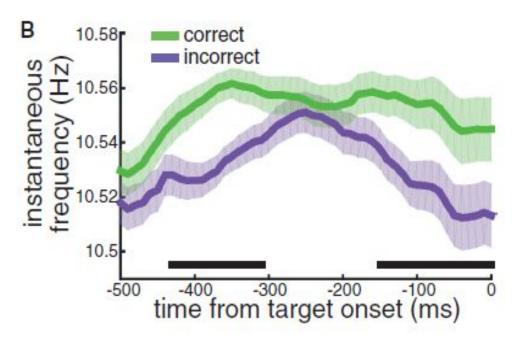


Time (-0.5 to 0s)



Frequency

Frequency sliding



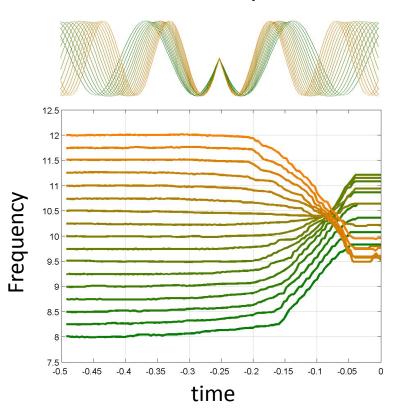
Significant difference at roughly 0.03 Hz

Samaha et al., 2015

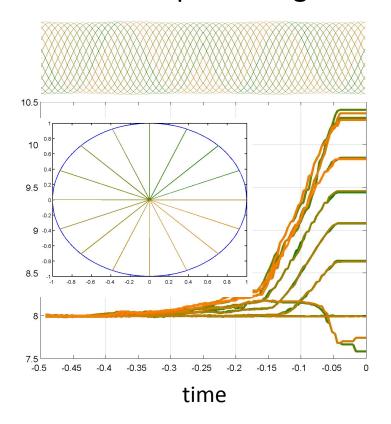


Frequency sliding: Mirror padding

Different frequencies



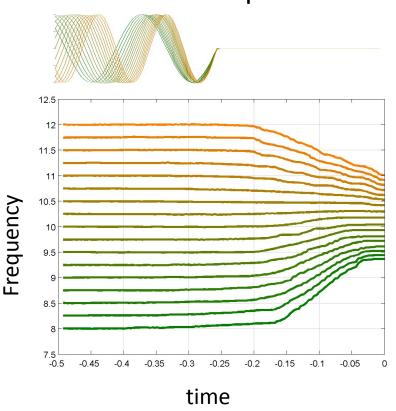
Different phase angles



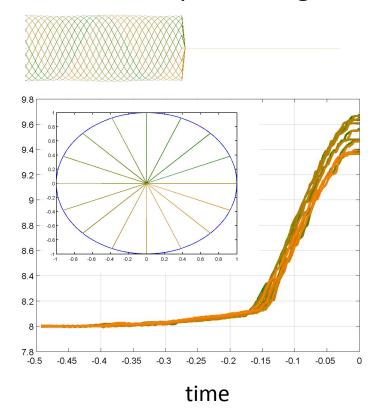


Frequency sliding: Zero padding

Different frequencies

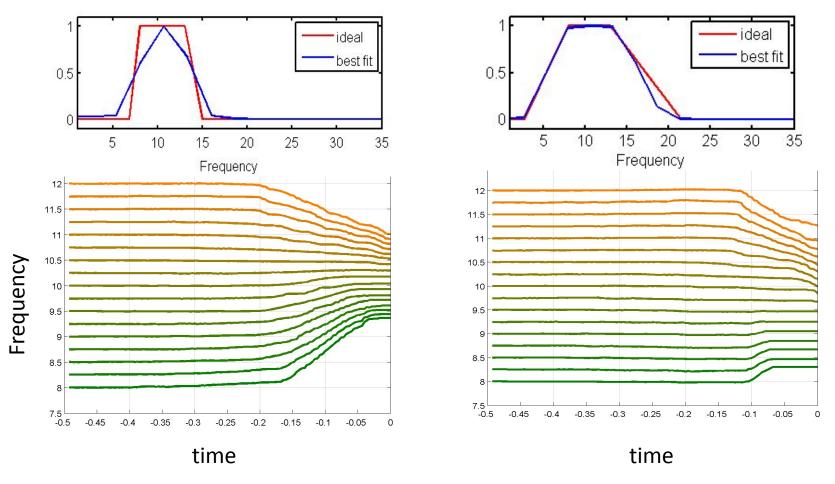


Different phase angles





Frequency sliding: Zero padding





Simulation summary:

Mirror-padding distorts data more than zero-padding

In both cases it is hard to interpret a difference between conditions, especially for small sample sizes

So should we ignore pre-stimulus activity?



Maybe we can forecast the signal

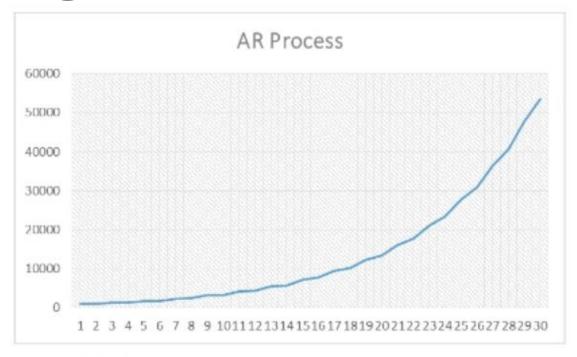
EEG is very noisy, and we want to predict data on single trials

Even a poor prediction should be better than zero-padding (if unbiased)





Autoregressive moving average modeling (ARMA)

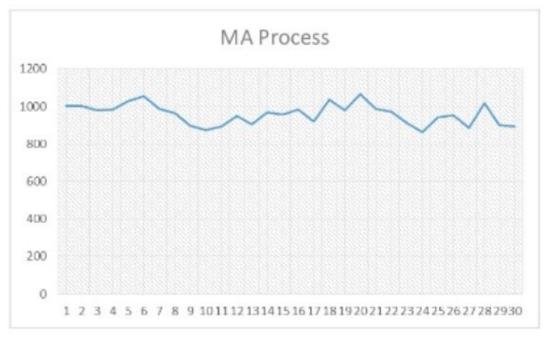


AR(1)
$$y_t = a1^* y_{t-1}$$

AR(2) $y_t = a1^* y_{t-1} + a2^* y_{t-2}$
AR(3) $y_t = a1^* y_{t-1} + a2^* y_{t-2} + a3^* y_{t-2}$



Autoregressive moving average modeling (ARMA)



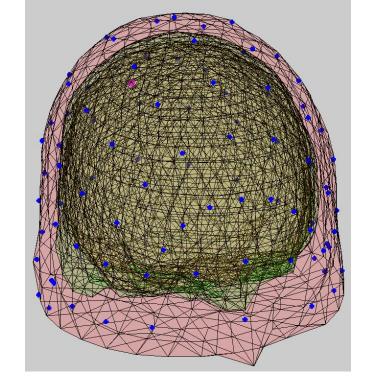
$$\begin{aligned} &\text{MA(1)}\ \epsilon_{t} = \text{b1*}\epsilon_{t\text{-}1} \\ &\text{MA(2)}\ \epsilon_{t} = \text{b1*}\epsilon_{t\text{-}1} + \text{b2*}\epsilon_{t\text{-}2} \\ &\text{MA(3)}\ \epsilon_{t} = \text{b1*}\epsilon_{t\text{-}1} + \text{b2*}\epsilon_{t\text{-}2} + \text{b3*}\epsilon_{t\text{-}3} \end{aligned}$$



Simulations with ft_dipolefitting

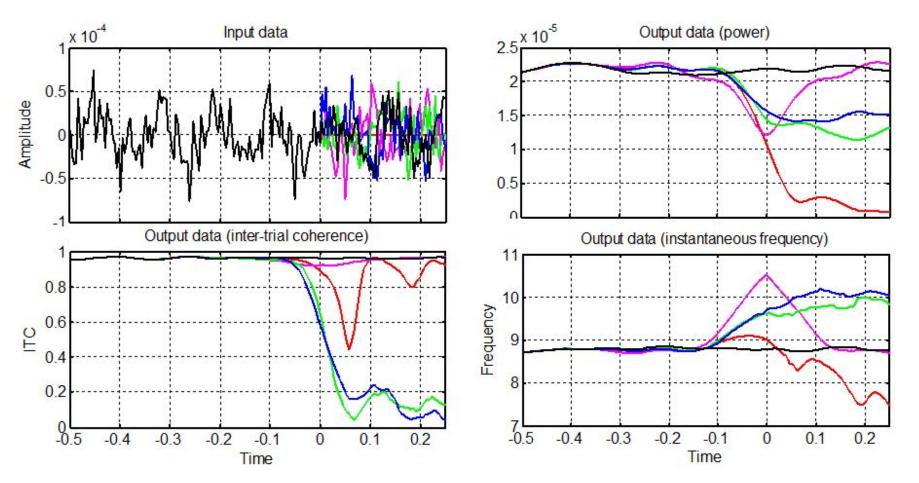
Simulate sine wave at frequency F with phase P and certain SNR using BEM volume conduction model.

AR order = 7
ARMA: na = 25, nc = 10
(modeling by Giulio Degano)



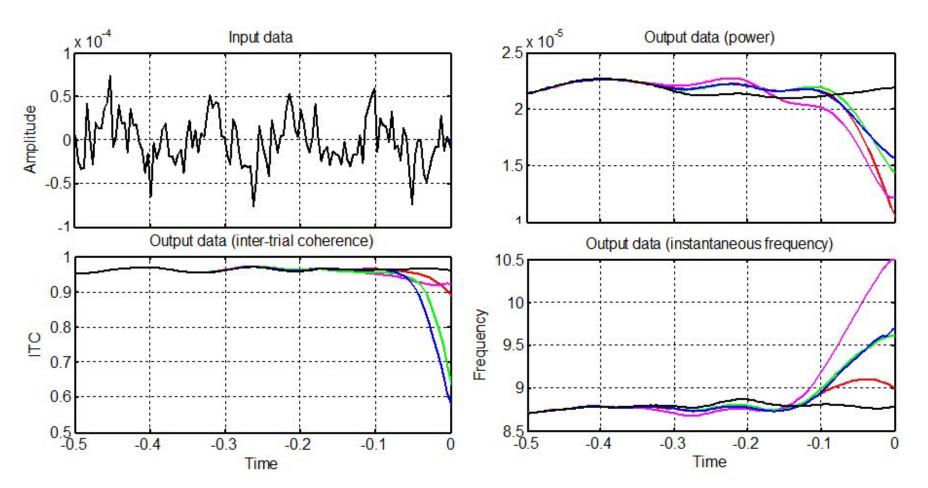






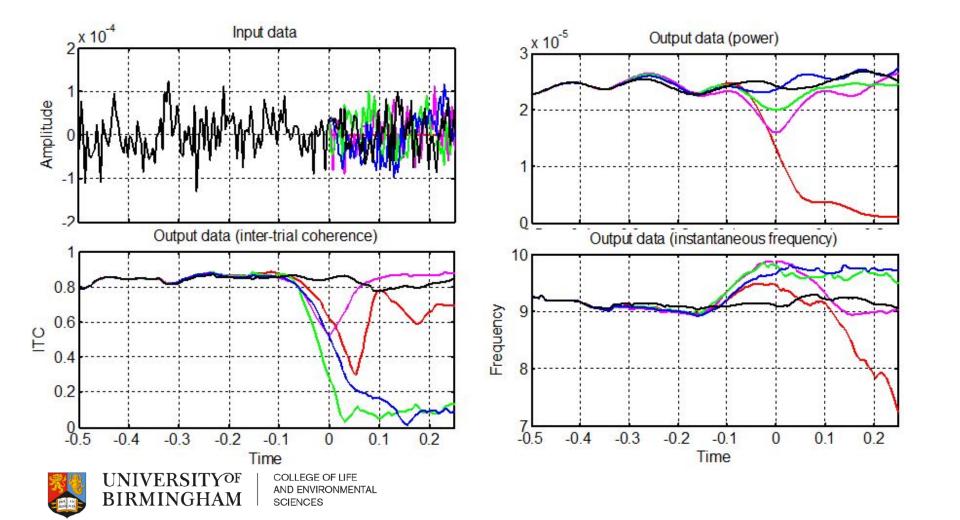




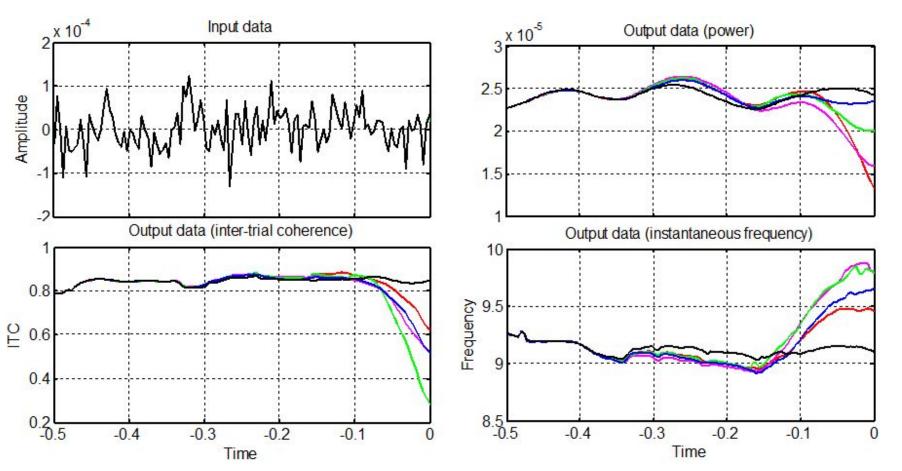






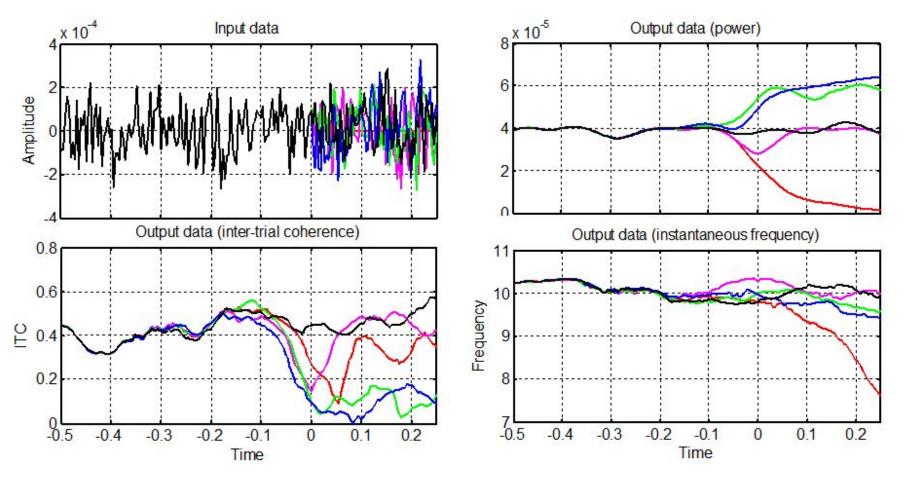








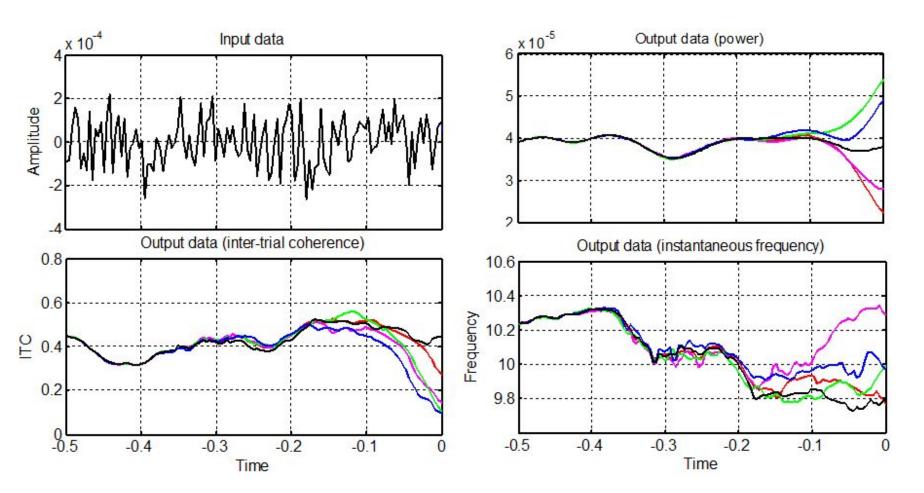






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Simulation summary:

For the given simulation and model specifications we observe significant distortions, similar to zero- or mirror-padding

ARMA and AR perform well for power, mediocre for frequency sliding and poorly for ITC

Real EEG data (no stimulus)

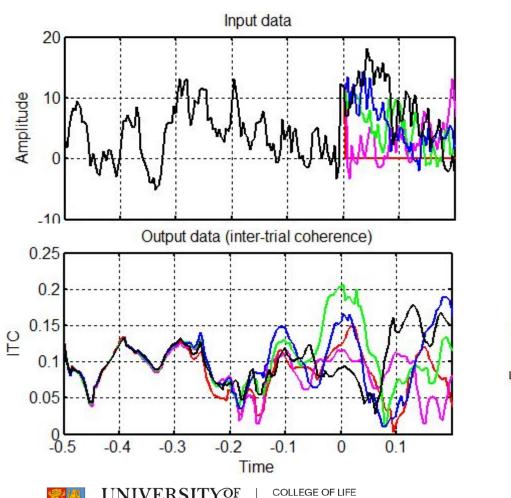
Sleep dataset of single participant

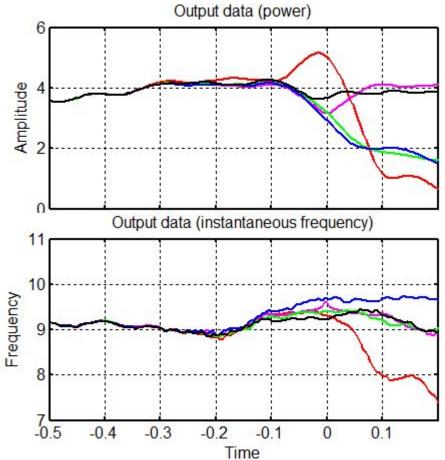
Number of trials: 79



Real EEG data (N = 1; n = 79)









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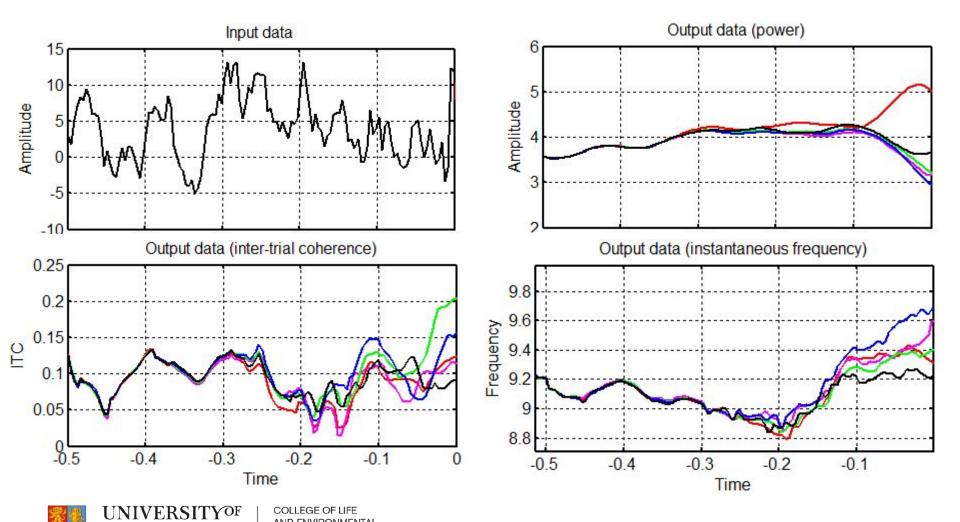
Real EEG data (N = 1; n = 79)

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Final summary:

All padding or forecasting methods introduce artifacts.

A quick test of AR and ARMA models gives complex results (no obvious advantage to zero-padding)

If this is a viable method, how do we choose parameters?



Forecasting methods - open discussion

Is it worth pursuing this idea further? What should we pay attention to?

Other possible forecasting methods: Support vector regression, neural networks, others?

Fin

Thank you!

References

Samaha, J., & Postle, B. R. (2015). The Speed of Alpha-Band Oscillations Predicts the Temporal Resolution of Visual Perception. *Current Biology*, *25*(22), 2985–2990. https://doi.org/10.1016/j.cub.2015.10.007