# Pros and cons of using temporal derivatives for brain functional correlations

J.K. Ochab

Marian Smoluchowski Institute of Physics Jagiellonian University, Cracow



Based on: JKO, W Tarnowski, MA Nowak, DR Chialvo

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#### Estimation of dynamic functional connectivity using Multiplication of Temporal Derivatives



James M. Shine <sup>a,b,\*</sup>, Oluwasanmi Koyejo <sup>b</sup>, Peter T. Bell <sup>a</sup>, Krzysztof J. Gorgolewski <sup>b</sup>, Moran Gilat <sup>a</sup>, Russell A. Poldrack <sup>b</sup>

<sup>&</sup>lt;sup>a</sup> Parkinson's Disease Research Clinic, Brain and Mind Research Institute, The University of Sydney, NSW, Australia

b Department of Psychology, Stanford University, Stanford, CA, USA

#### Dynamic functional connectivity

- functional magnetic resonance imaging (fMRI)
- blood-oxygen level-dependent (BOLD) signal
- spatial topologies of interacting regions (functional "networks")
- strength of ...... between pairs or sets of regions ("functional connectivity")
- absence of any task or stimuli ("resting-state")
- Dynamic = changing (time dependent)

Hutchison et al. (2013). *Dynamic functional connectivity: Promise, issues, and interpretations,* Neurolmage **80**, 360-378.

Chen, Rubinov, Chang (2017). *Methods and Considerations for Dynamic Analysis of Functional MR Imaging Data*, Neuroimaging Clinics of North America **27**, 547-560

#### SMA of MTD vs Pearson correlation

$$ds_{it} = s_{it+1} - s_{it}$$
  $\bar{ds}_i = \frac{1}{2w+1} \sum_{t'=t-w}^{t+\bar{w}} ds_{it'} = \frac{s_{it-w} - s_{it+w}}{2w+1} \to 0$  (1)

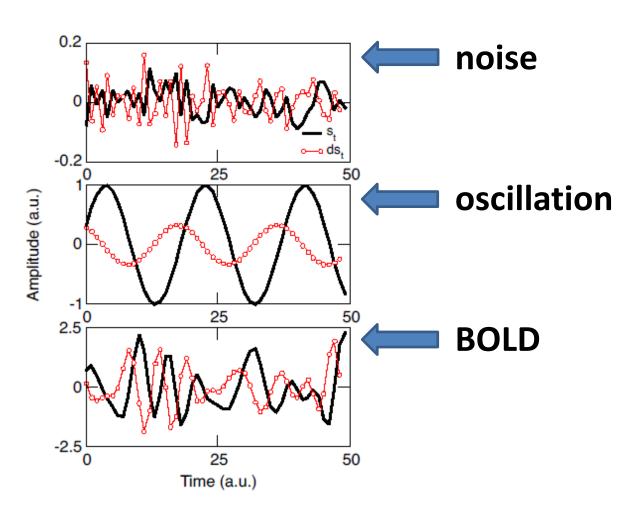
$$MTD_{ijt} = \frac{ds_{it}ds_{jt}}{\bar{\sigma}_i\bar{\sigma}_j} \tag{2}$$

$$SMA_{ijt} = \frac{1}{2w+1} \sum_{t'=t-w}^{t+w} MTD_{ijt'} = \frac{1}{2w+1} \sum_{t'=t-w}^{t+w} \frac{ds_{it'}}{\bar{\sigma}_i} \frac{ds_{jt'}}{\bar{\sigma}_j}, \tag{3}$$

$$r_{ij} = \frac{\sum_{t=1}^{T} (s_{it} - \bar{s}_i)(s_{jt} - \bar{s}_j)}{\sqrt{\sum_{t=1}^{T} (s_{it} - \bar{s}_i)^2} \sqrt{\sum_{t=1}^{T} (s_{jt} - \bar{s}_j)^2}} = \frac{1}{T - 1} \sum_{t=1}^{T} \left( \frac{s_{it} - \bar{s}_i}{\sigma_i} \right) \left( \frac{s_{jt} - \bar{s}_j}{\sigma_j} \right)$$

$$r_{ijt} = \frac{\sum_{t'=t-w}^{t+w} (ds_{it'} - \bar{d}s_i)(ds_{jt'} - \bar{d}s_j)}{\sqrt{\sum_{t'=t-w}^{t+w} (ds_{it'} - \bar{d}s_i)^2}} \sqrt{\sum_{t'=t-w}^{t+w} (ds_{jt'} - \bar{d}s_j)^2} = \frac{1}{2w+1} \sum_{t'=t-w}^{t+w} \frac{ds_{it'} - \bar{d}s_i}{\sigma_i} \frac{ds_{jt'} - \bar{d}s_j}{\sigma_j}$$

# 1. What's the difference? Simplest models



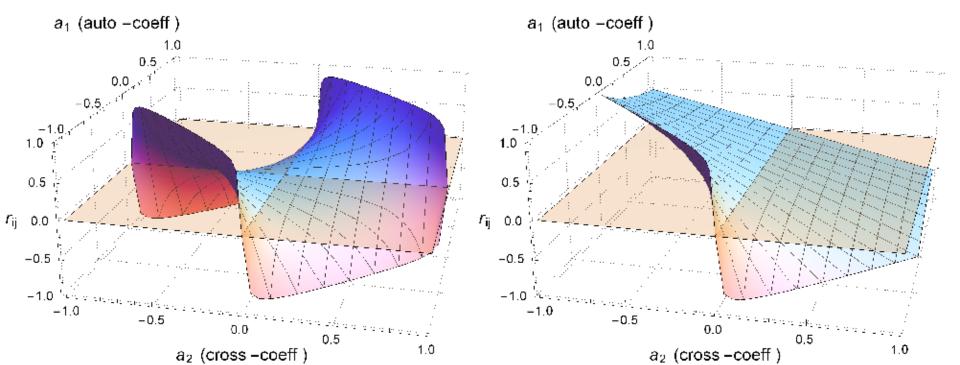
$$x_{it} = \sum_{k=1}^{q} \sum_{j=1}^{N} A_{ij}^{(k)} x_{j,t-k} + \xi_{it}$$
 **VAR(q)**

$$x_{1t} = a_1 x_{1,t-1} + a_2 x_{2t-1} + \xi_{1t}$$
  
 $x_{2t} = a_1 x_{2,t-1} + a_2 x_{1t-1} + \xi_{2t}$  q=1, N=2

We can analytically calculate correlations (for differences too)!

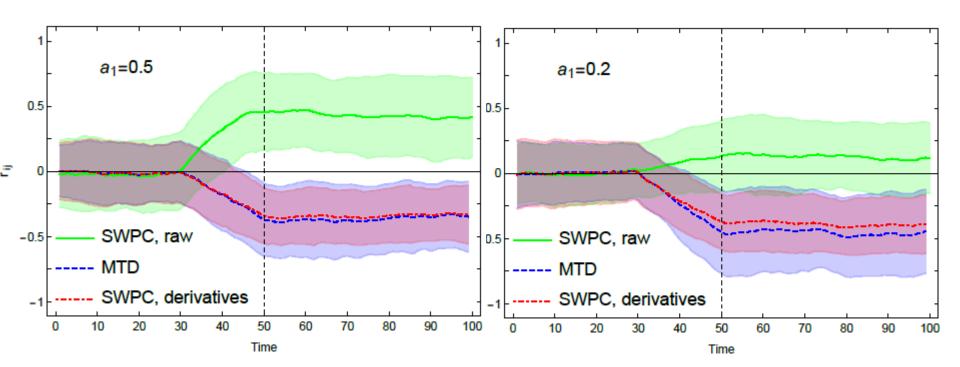
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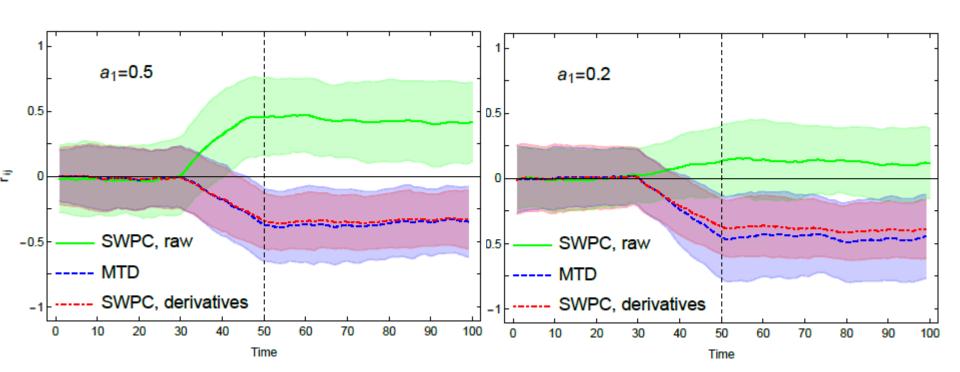
$$x_{it} = \sum_{k=1}^{q} \sum_{j=1}^{N} A_{ij}^{(k)} x_{j,t-k} + \xi_{it}$$
 VAR(q)

$$x_{1t} = a_1 x_{1,t-1} + a_2(t) x_{2t-1} + \xi_{1t}$$
  
 $x_{2t} = a_1 x_{2,t-1} + a_2(t) x_{1t-1} + \xi_{2t}$  q=1, N=2

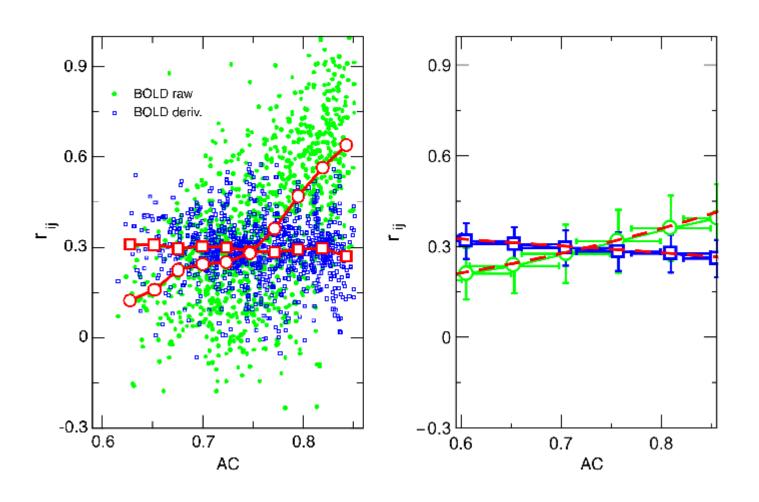


$$SMA_{ijt} = \frac{1}{2w+1} \sum_{t'=t-w}^{t+w} MTD_{ijt'} = \frac{1}{2w+1} \sum_{t'=t-w}^{t+w} \frac{ds_{it'}}{\bar{\sigma}_i} \frac{ds_{jt'}}{\bar{\sigma}_j}$$

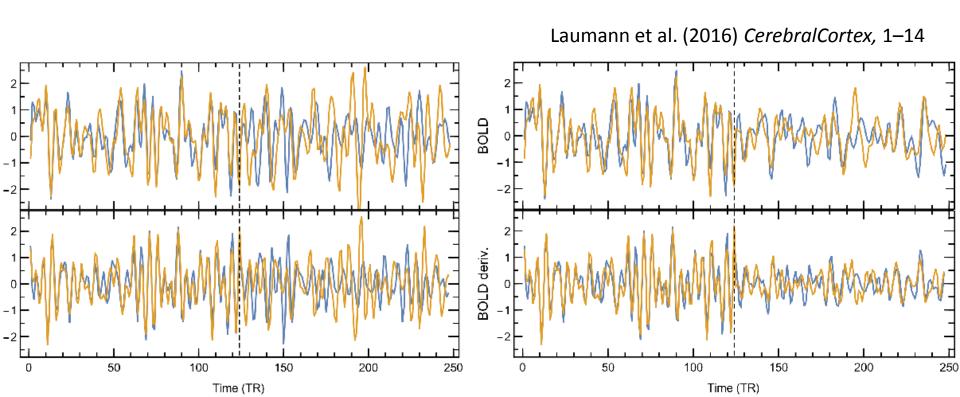
$$\frac{1}{2w+1} \sum_{t'=t-w}^{t+w} \frac{d\bar{s}_{it'}}{\bar{\sigma}_i} \frac{ds_{jt'}}{\bar{\sigma}_j}$$
Remember?



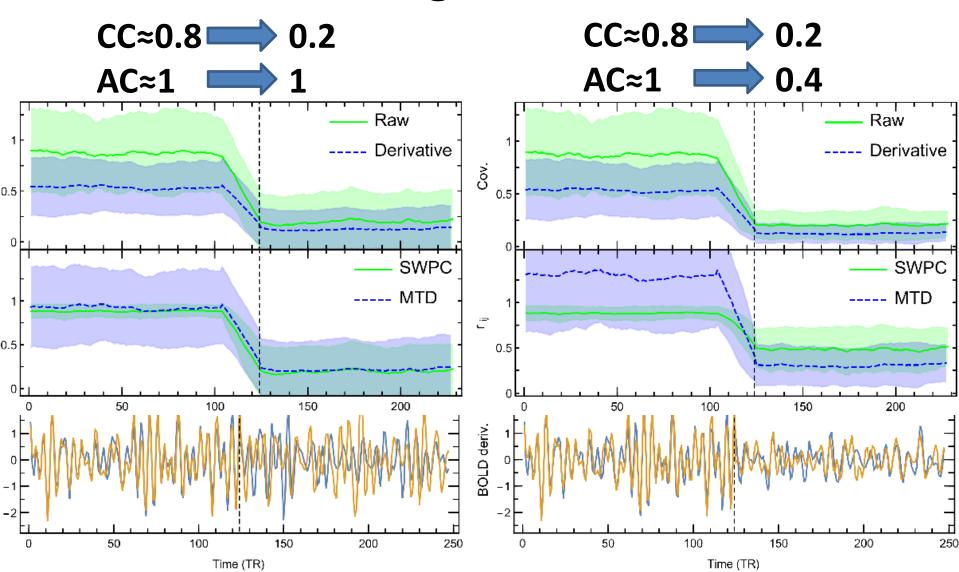
# 3. What's the difference? BOLD vs ARMA(1,1)x2



## 4. What's the difference? Surrogate BOLD



# 4. What's the difference? Surrogate BOLD



#### Conclusions

What's the difference? It depends.

- centering and windowed standardization decrease uncertainty of correlations
- differences: decrease signal-to-noise ratio
- differences: enhance stationarity, not affected by low frequency drifts
- differences: have lower sensitivity to autocorrelations (but worse than raw series for high autocorr.)

#### Based on:

JK Ochab, W Tarnowski, MA Nowak, DR Chialvo, On the pros and cons of using temporal derivatives to assess brain functional connectivity, submitted (2018); arXiv:1803.05048 [q-bio.NC]

### Thank you



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