

MS Thesis Progress Report for Sem 1, 2021-22

Data Analysis for Predicting Instabilities in Power Systems

Aryan Ritwajeet Jha

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Thesis supervisor: Dr. Nilanjan Senroy
Department of Electrical Engineering
IIT Delhi



Introduction

Bifurcations and Critical Slowing Down

Bifurcation: A qualitative change in the ‘motion’ of a dynamical System due to a quantitative change in one of its parameters. Serious bifurcations, called **Critical Bifurcations**, cause the system to become unstable from stable.



Bifurcations and Critical Slowing Down

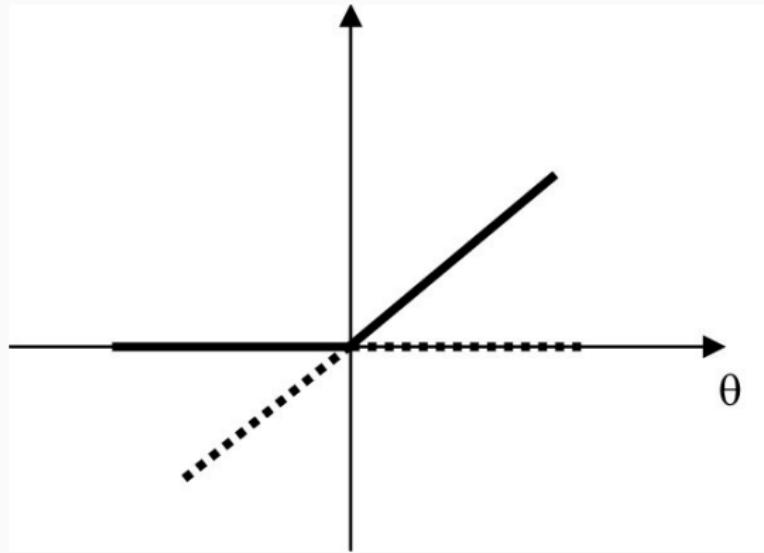


Figure 1: Bifurcation Diagram showing the Normal form of Transcritical Bifurcation



$$\frac{dx}{dt} = \theta x - x^2 \quad (1)$$

Bifurcations and Critical Slowing Down

Critical Slowing Down: Dynamical Systems exhibit early statistical warning signs before collapsing:

- Increased recovery times from perturbations.
- Increased signal variance from the mean trajectory.
- Increased flicker and asymmetry in the signal

The above three properties can be identified by increasing variance and autocorrelation in time-series measurements taken from the system.



Last Semester's Work

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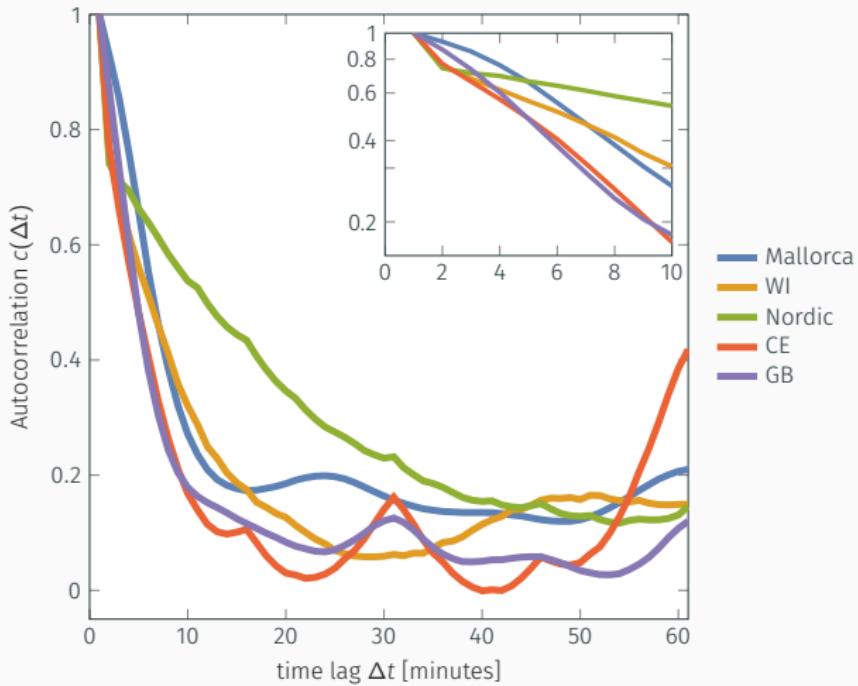


Figure 2: Autocorrelation decay of different synchronous regions.



Last Semester's Work

Table 1: Inverse-correlation values for different grids

Grid name	Inverse-correlation value $T^{-1} [min^{-1}]$
Mallorca	0.0654
Western Interconnection	0.0498
Nordic	0.0235
Continental Europe	0.0829
Great Britain	0.0879

Figure 3: Inverse correlation time is proportional to the damping constant of the grid.



Procedure

Procedure: Simulation

- Created and simulated the IEEE 9 Bus System in PSSE 34.
- Added stochastic disturbance to the loads (at Bus 5, 6 and 8) via white noise modeled as

$$(P_L)_i[k] = (P_L)_i[k - 1] + N(0, \sigma^2) \quad (2)$$

for load bus i at discrete sample k . $\sigma = 0.01$ pu.

- In order to drive the power grid towards bifurcation, steadily increased the three loads at different rates, between 20% to 30% per minute.
- Ran a time analysis simulation until critical bifurcation attained.
- Extracted the bus voltages.



Procedure: Detrending Bus Voltages

Passed the voltage signals through a Low Pass Filter in order to capture the slow changing trends not an effect of CSD. Eg. Change in bus voltages due to the gradual increasing of loads. Gaussian Kernel Smoothing Filter was used for the same.

$$h(n, \sigma_f) = \frac{1}{\sqrt{2\pi}\sigma_f} \exp^{\frac{-n^2}{2\sigma_f^2}} \quad (3)$$

where σ_f can be varied between 5 to 10 .

The detrended signal is obtained by subtracting the filtered signal from the original signal.

$$d[x] = x[k] - GKS(x[k]) \quad (4)$$



Procedure: Applying AR1 and Variance

Windows varying between $W = 1$ second to $W = 40$ seconds were used for computing autocorrelations and variances.

For every window, Autoregressive Model of Order 1 was fitted onto the detrended voltages using least squares of error approach.

$$d[k] = a_1 d[k - 1] + e[k] \quad (5)$$

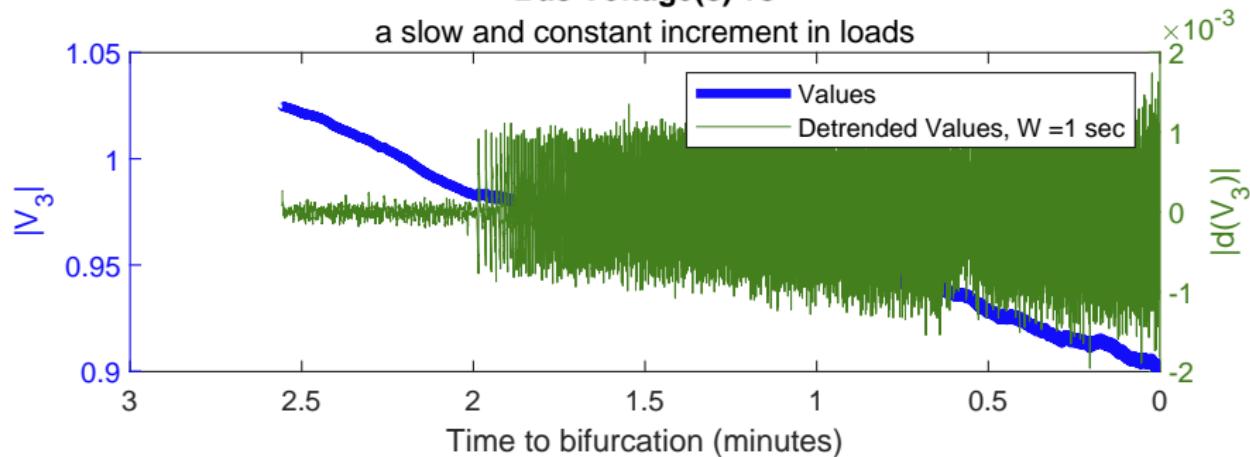
Variances were also computed for every window.

$$\sigma^2 = \frac{1}{n_k} \sum_{k=1}^{n_k} d[k]^2 \quad (6)$$

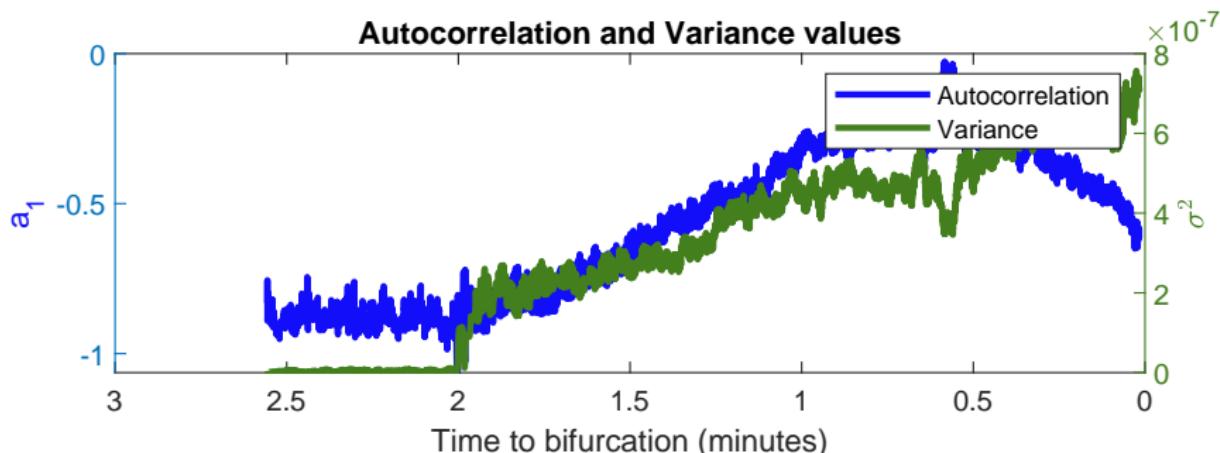


Results

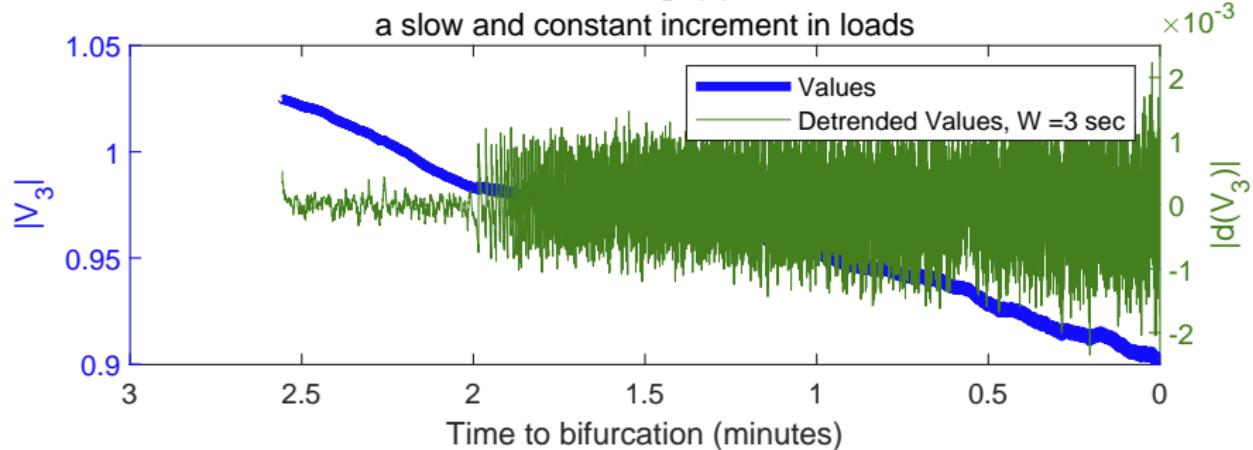
Bus Voltage(s) vs a slow and constant increment in loads



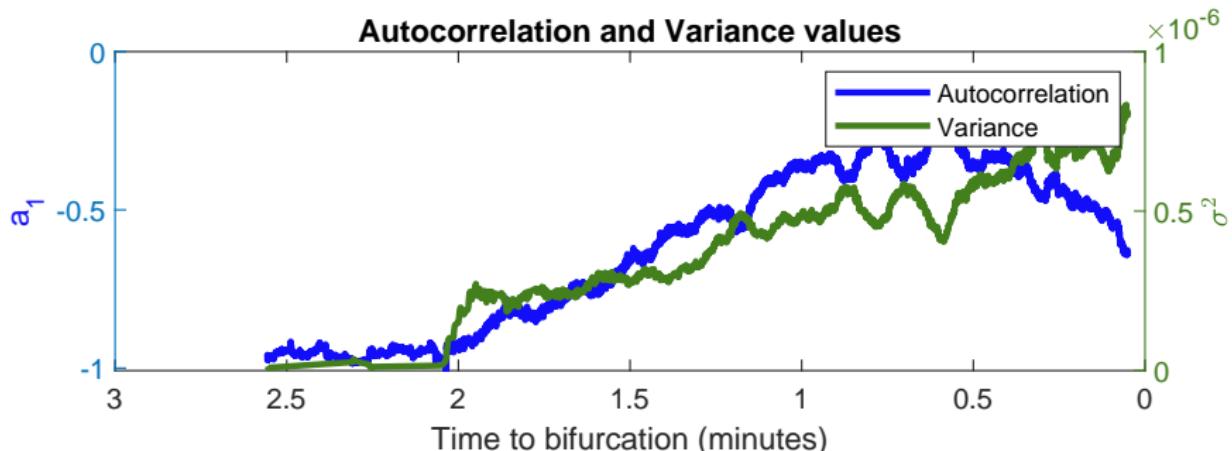
Autocorrelation and Variance values



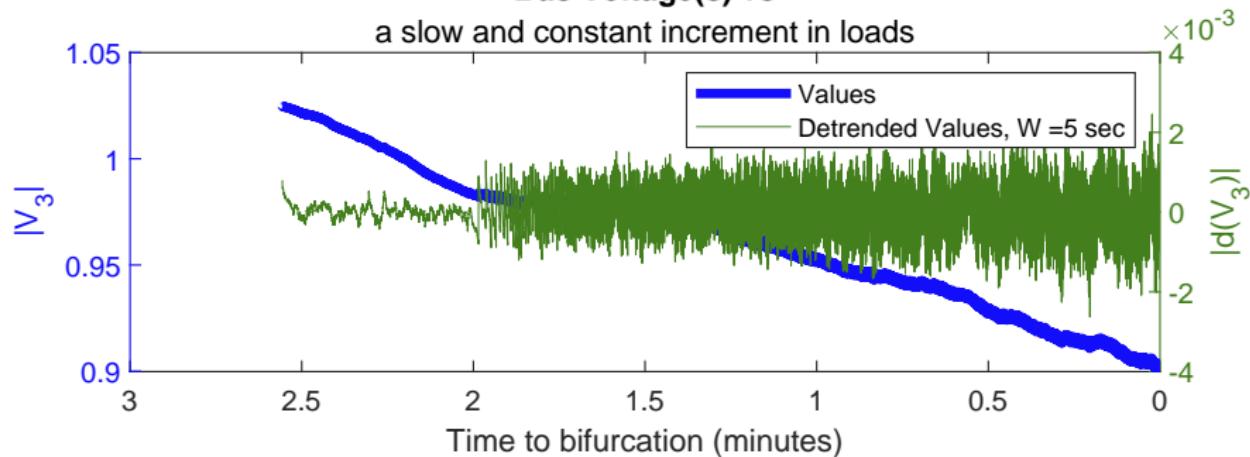
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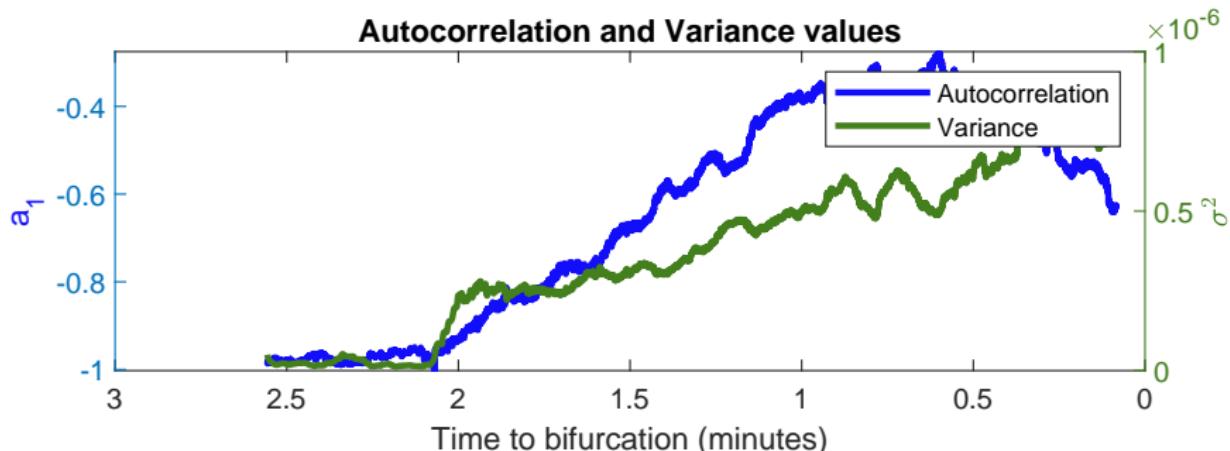
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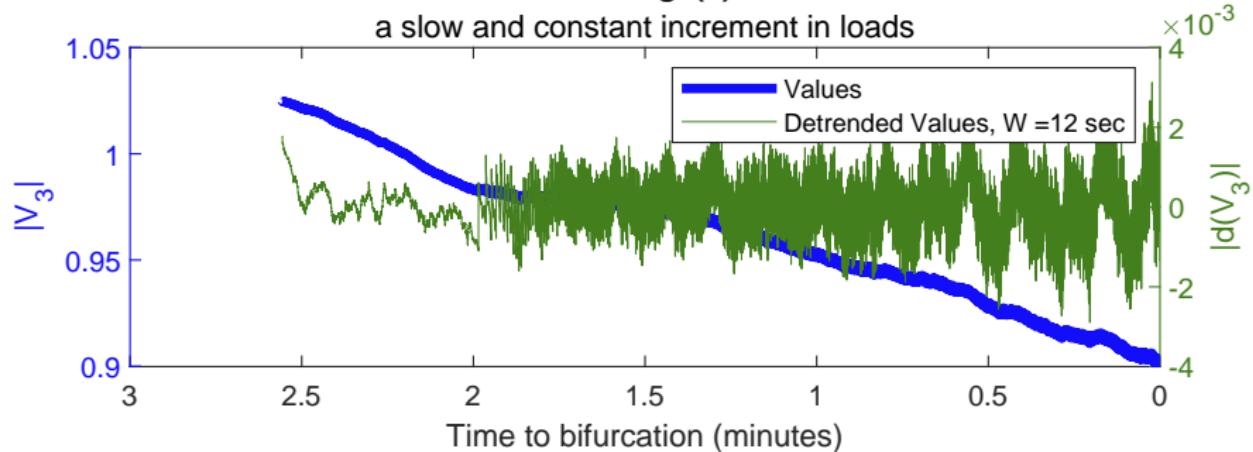
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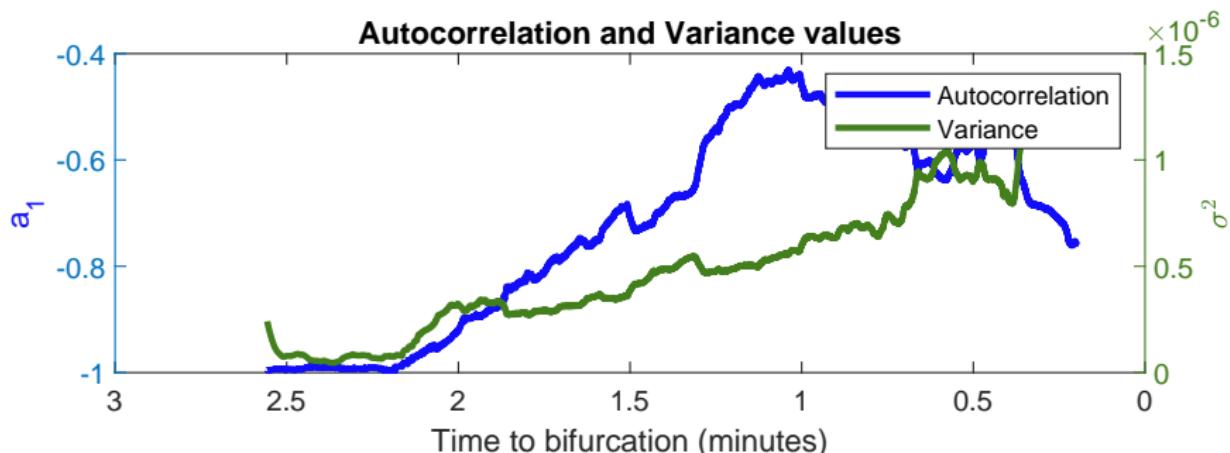
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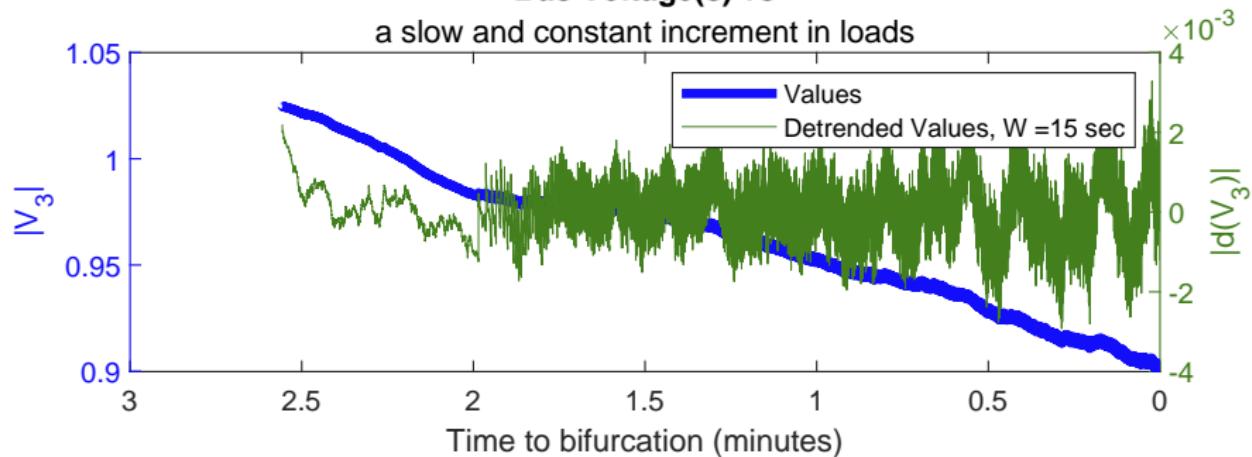
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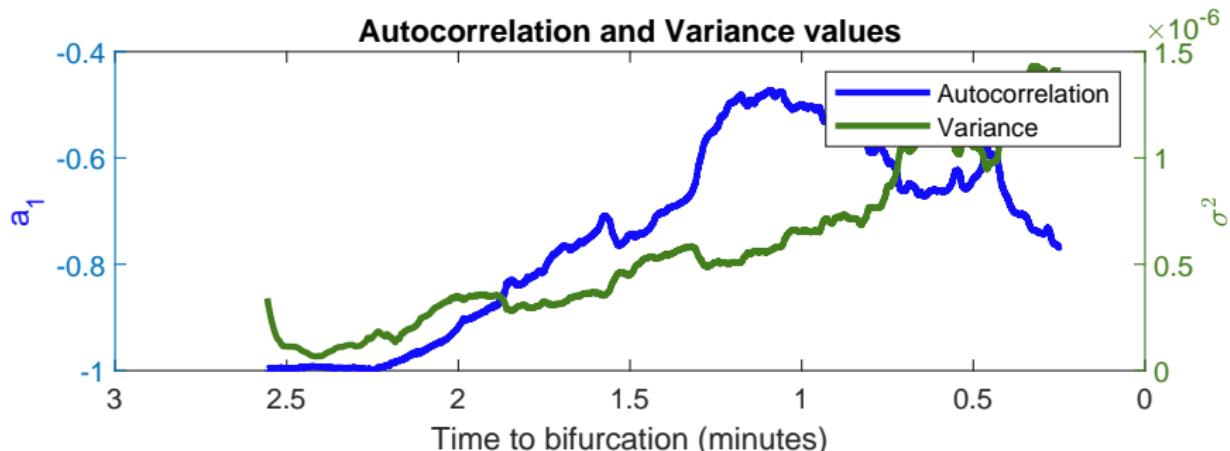
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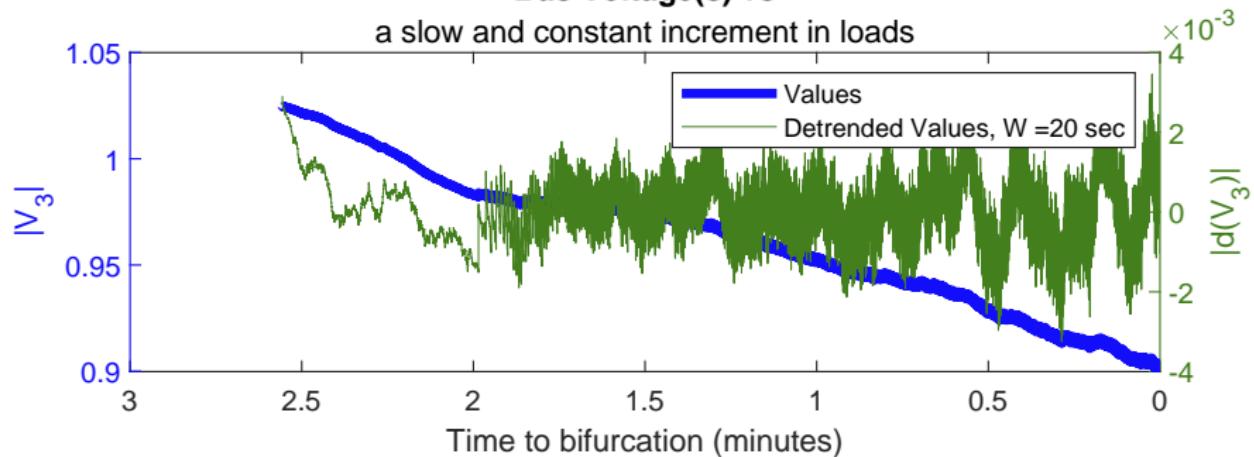
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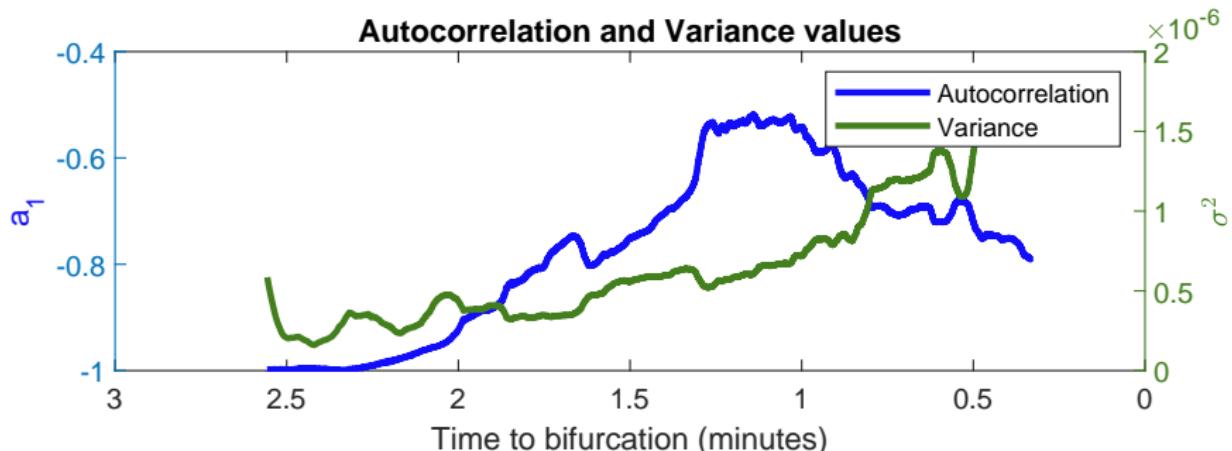
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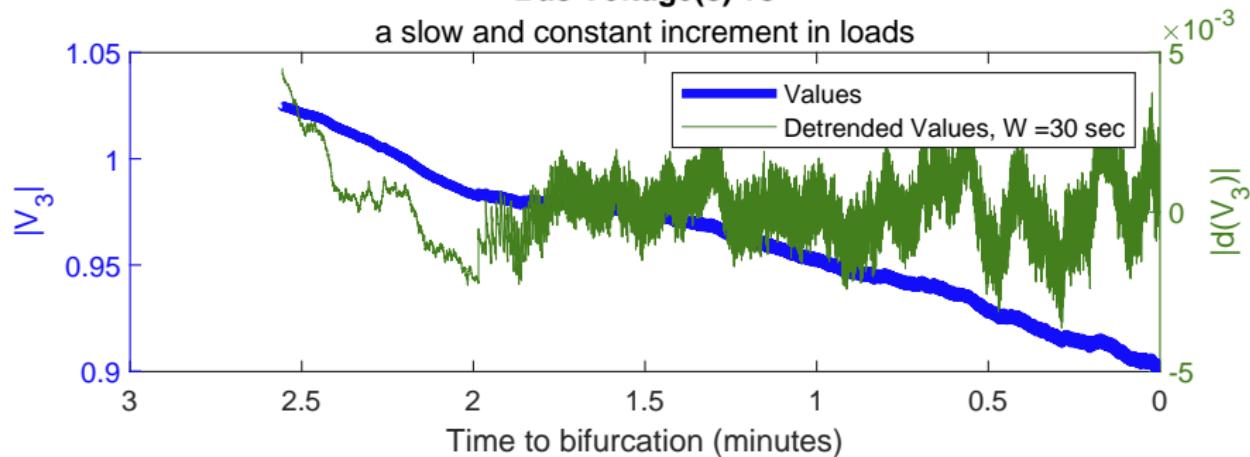
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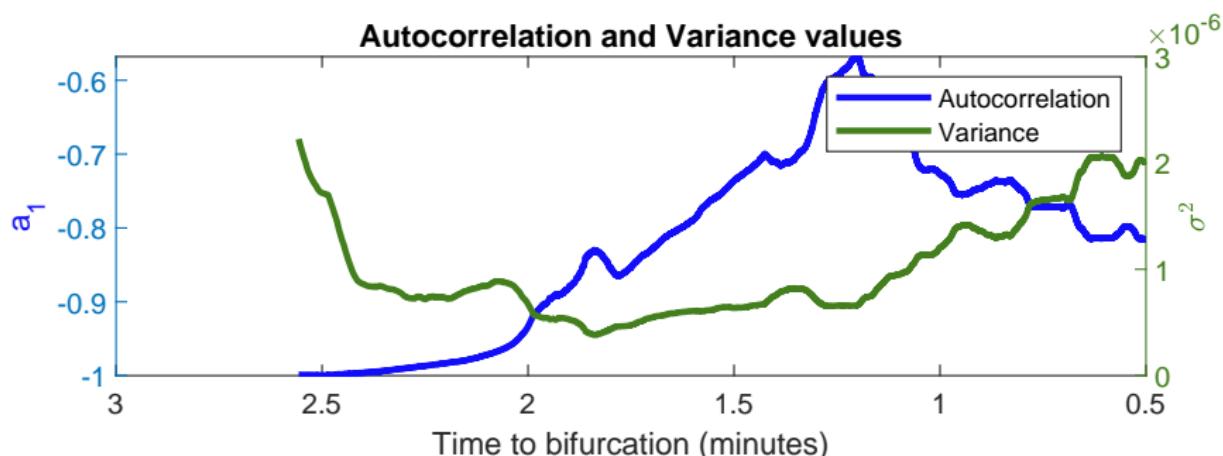
Autocorrelation and Variance values



Bus Voltage(s) vs a slow and constant increment in loads



Autocorrelation and Variance values



Conclusion and Future Work

Conclusion

Autocorrelation and Variance of the detrended bus voltage time series signal performed satisfactorily as early warning signs of Critical Slowing Down.

These statistical parameters may lose their effectiveness as EWS when the sampling window is outside the ‘suitable’ range. In this case this range may be said to be between 1 second to 15 seconds.



Future Work

- Examine CSD symptoms on line currents.
- Perform the analysis on bigger power grid models
- Attempt to construct an algorithm which can determine the ‘Time to Bifurcation’ for any particular power grid (supplied as a model) if given a stream of PMU data containing grid state variables.



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References

References i

- [1] Goodarz Ghanavati, Paul D. H. Hines, and Taras I. Lakoba. "Identifying Useful Statistical Indicators of Proximity to Instability in Stochastic Power Systems". In: *IEEE Transactions on Power Systems* 31.2 (2016), pp. 1360–1368. DOI: [10.1109/TPWRS.2015.2412115](https://doi.org/10.1109/TPWRS.2015.2412115).
- [2] Eduardo Cotilla-Sanchez, Paul D. H. Hines, and Christopher M. Danforth. "Predicting Critical Transitions From Time Series Synchrophasor Data". In: *IEEE Transactions on Smart Grid* 3.4 (2012), pp. 1832–1840. DOI: [10.1109/TSG.2012.2213848](https://doi.org/10.1109/TSG.2012.2213848).



References ii

- [3] Benjamin Schäfer et al. "Non-Gaussian power grid frequency fluctuations characterized by Lévy-stable laws and superstatistics". In: *Nature Energy* 3.2 (2018), pp. 119–126. ISSN: 2058-7546. DOI: [10.1038/s41560-017-0058-z](https://doi.org/10.1038/s41560-017-0058-z). URL: <http://dx.doi.org/10.1038/s41560-017-0058-z>.
- [4] Petr Vorobev et al. "Deadbands, Droop, and Inertia Impact on Power System Frequency Distribution". In: *IEEE Transactions on Power Systems* 34.4 (2019), pp. 3098–3108. DOI: [10.1109/TPWRS.2019.2895547](https://doi.org/10.1109/TPWRS.2019.2895547).
- [5] Francesca Madia Mele et al. "Impact of variability, uncertainty and frequency regulation on power system frequency distribution". In: *2016 Power Systems Computation Conference (PSCC)*. 2016, pp. 1–8. DOI: [10.1109/PSCC.2016.7540970](https://doi.org/10.1109/PSCC.2016.7540970).



References iii

- [6] Muhammad Adeen and Federico Milano. "On the Impact of Auto-Correlation of Stochastic Processes on the Transient Behavior of Power Systems". In: *IEEE Transactions on Power Systems* (2021), pp. 1–1. DOI: [10.1109/TPWRS.2021.3068038](https://doi.org/10.1109/TPWRS.2021.3068038).
- [7] Leonardo Rydin Gorjão et al. "Open database analysis of scaling and spatio-temporal properties of power grid frequencies". In: *Nature Communications* 11.1 (2020), p. 6362. ISSN: 2041-1723. DOI: [10.1038/s41467-020-19732-7](https://doi.org/10.1038/s41467-020-19732-7). URL: <https://doi.org/10.1038/s41467-020-19732-7>.

