

Spectral Methods in Time Series Analysis

Using the SSA-MTM Toolkit for Time Series Analysis

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Singular Spectrum Analysis

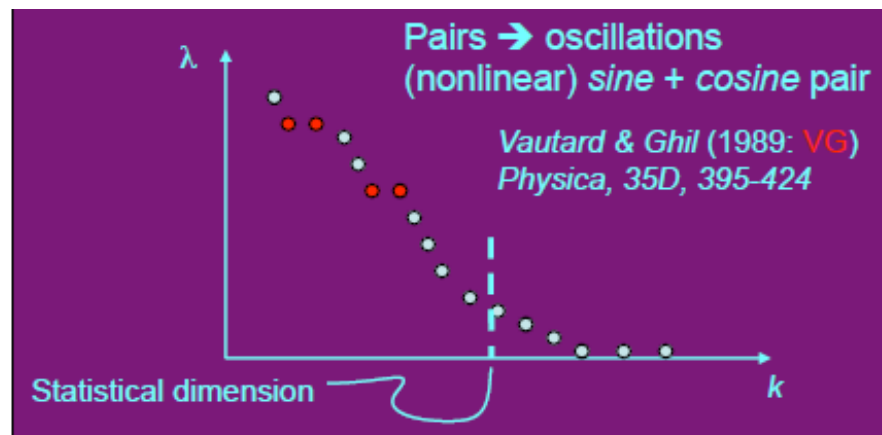
- Allows to get spectral information on the time series

time series $X(t)$: $t = 1, \dots, N$

- $X(t)$ embedded in a vector space of dimension M :
 - Diagonalization of the $M \times M$ lag-covariance matrix C_x
 - Calculation of the M eigenvectors E_k of the lag-covariance matrix C_x ; E_k are called temporal empirical orthogonal functions (EOFs).

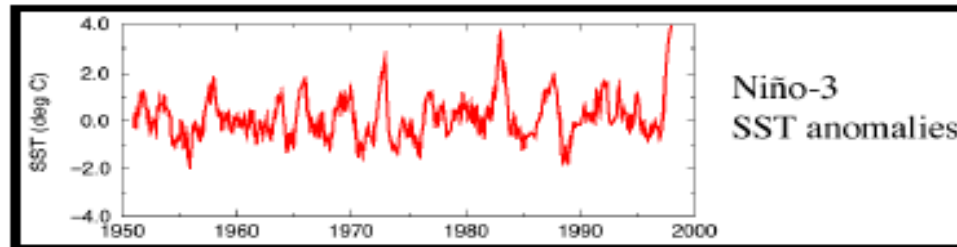
Decomposition and Reconstruction

- Projecting the time series onto each EOF yields the corresponding temporal principal components (PCs).
- The eigenvalues λ_k of C_x account for the partial variance in the direction E_k .
- A pair of nearly equal λ_k and associated EOFs in approximate phase quadrature characterize an oscillatory mode.

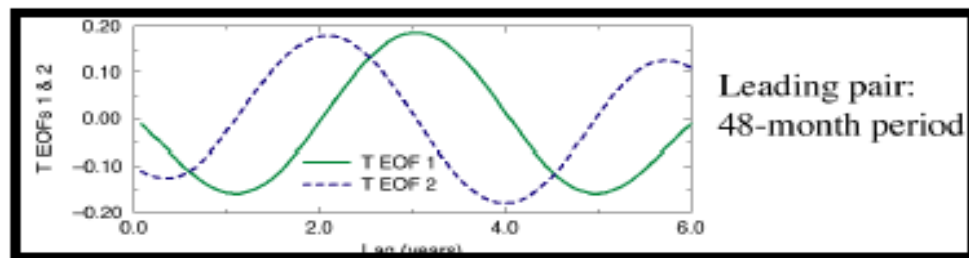


From M_Ghil-SSA-BESS
Presentation

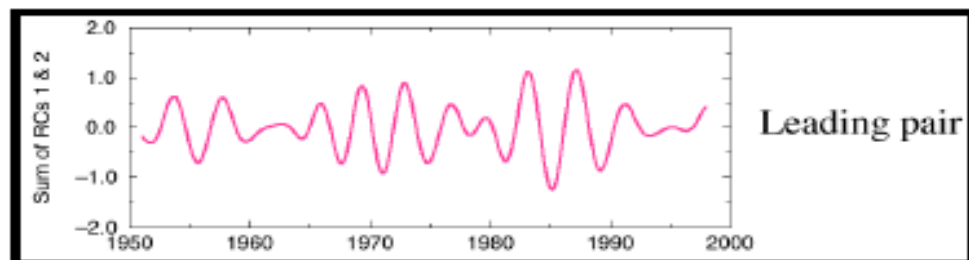
Time series



T-EOFs



RCs





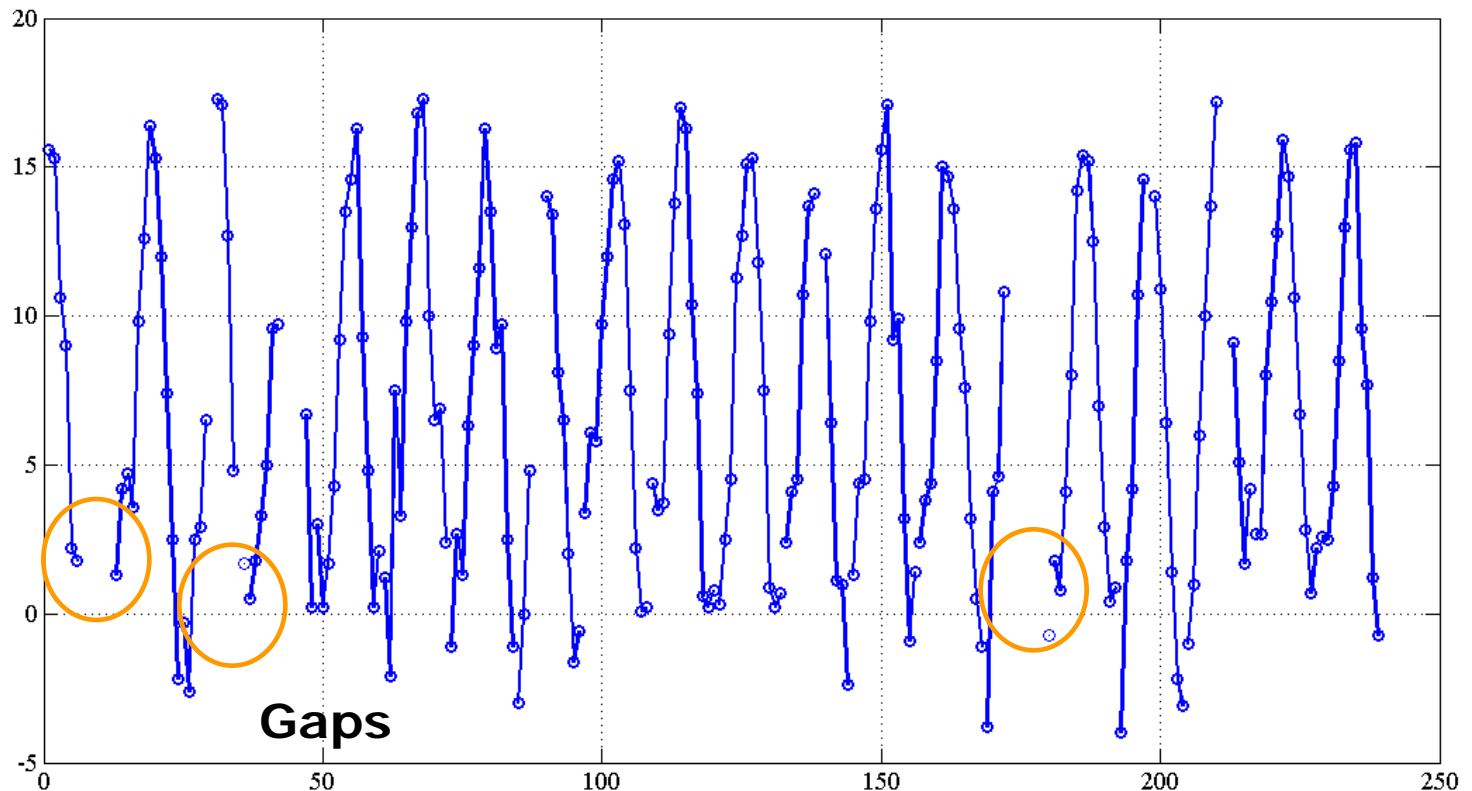
Window Length (M) and Principal Component Number (PC)

- Larger window size improves resolution but at the cost of increasing variance of the spectral estimates.
- The minimum number of PCs that properly reproduces the signal should be used.

SSA-MTM Toolkit for TS Analysis

<http://www.atmos.ucla.edu/tcd/ssa/>

- The Singular Spectrum Analysis - MultiTaper Method Toolkit is a software program to analyze short, noisy time series.



Monthly Mean Temperature in Acceglio, NW Italy (1989-2008)

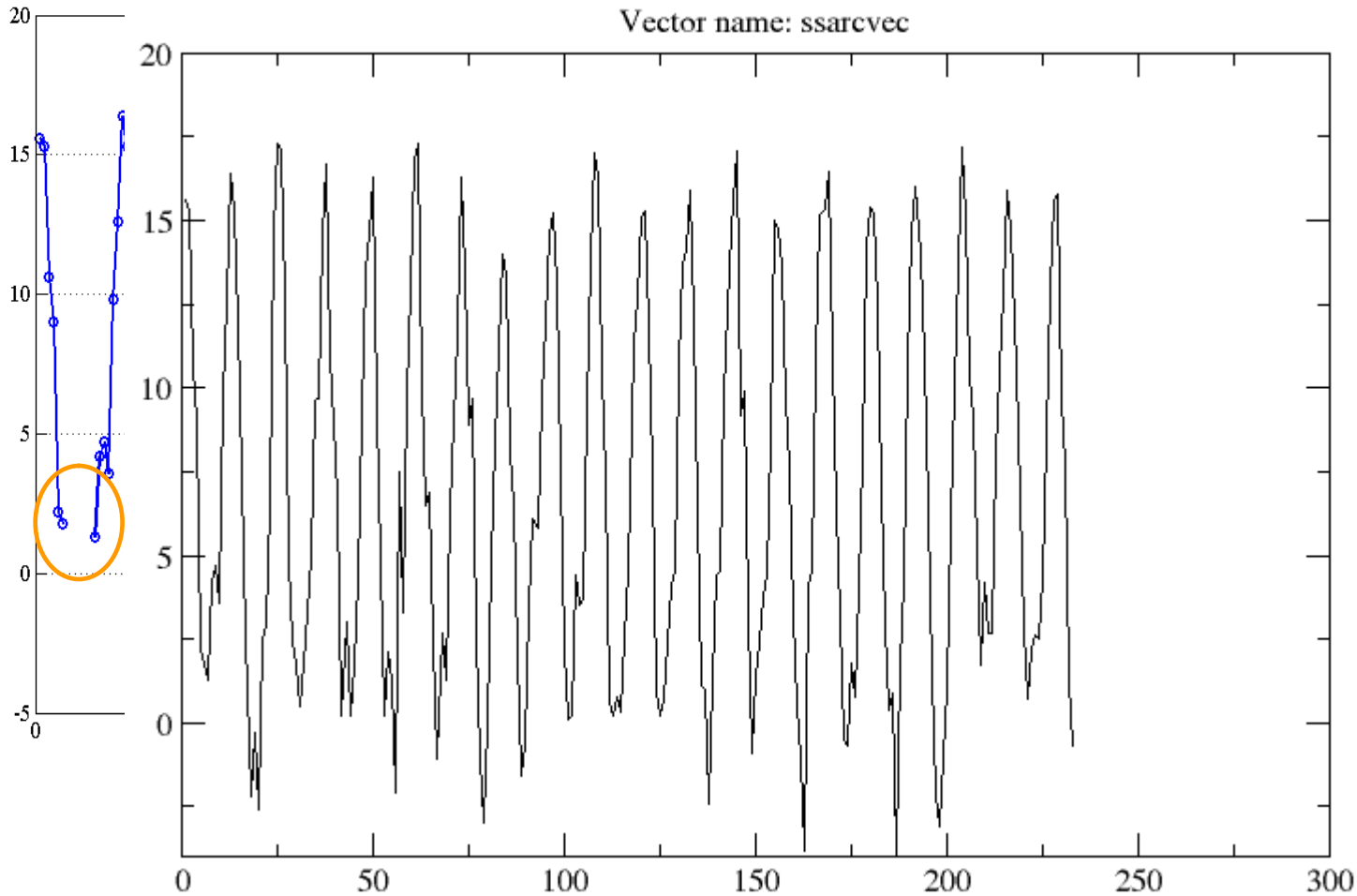
Gap filling

- Decomposition of the time series into PCs.
- Reconstruction of the signal using first significant components and so the missing values are estimated.

Temperature Time Series

Reconstruction and Gap-Filling (M:60, PC:12)

Vector (y values) vs. index (x values).

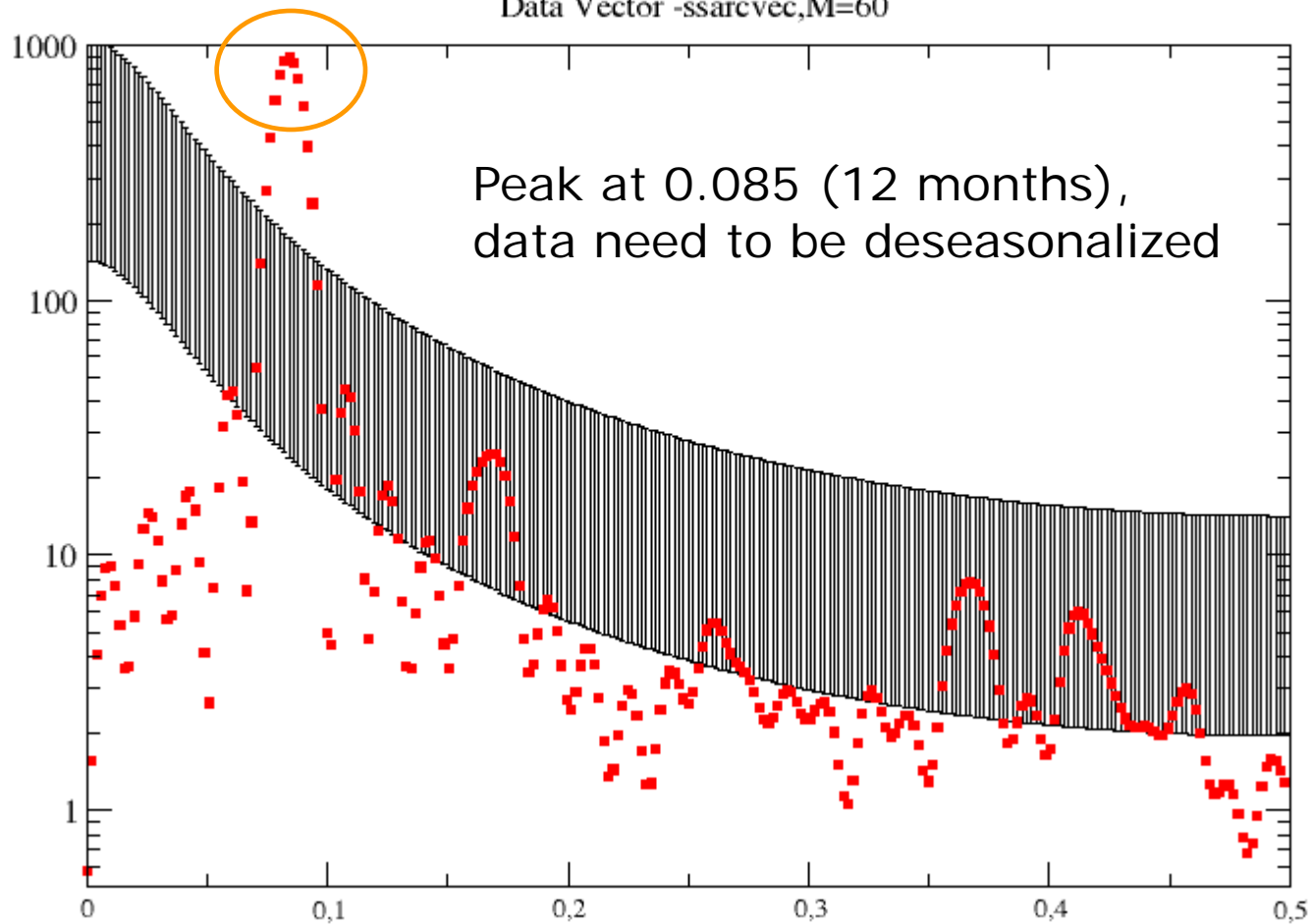


Temperature Time Series

B-T Correlogram – Reconstructed TS

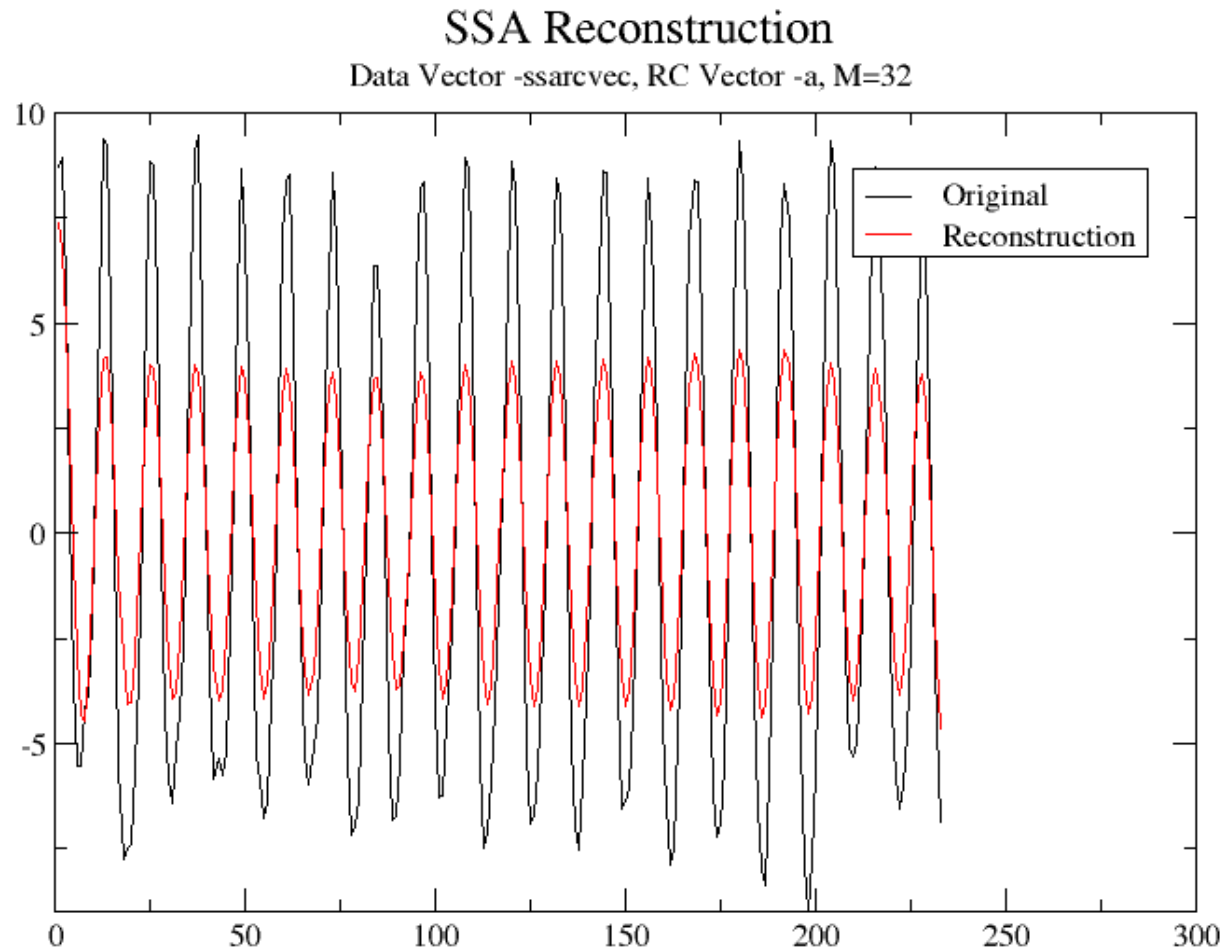
Blackman-Tukey Spectrum

Data Vector -ssarcvec,M=60



Temperature Time Series

Principal Component (PC) = 1

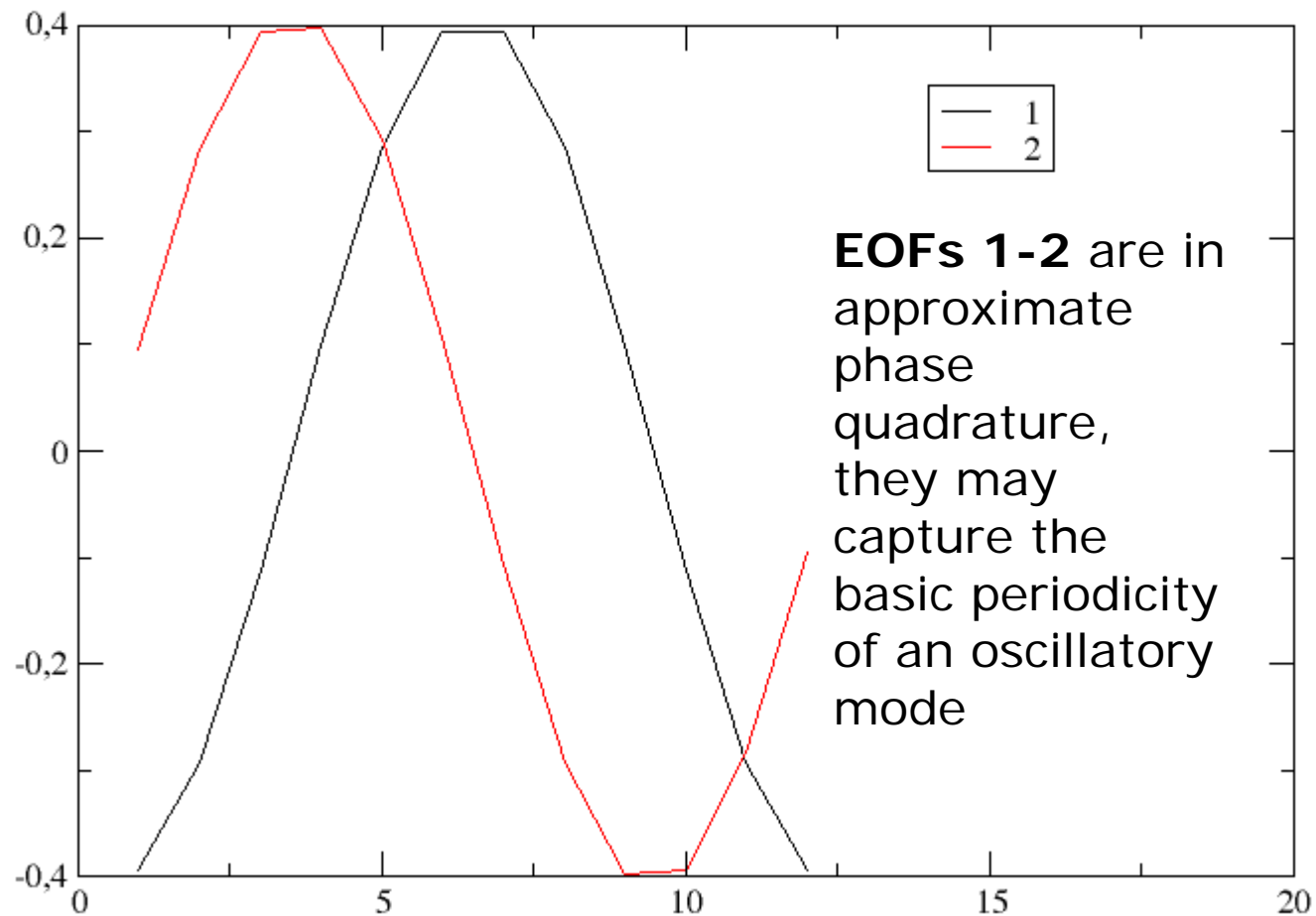


Analysis of **PC 1** which captures the seasonal cycle of the time series

Temperature Time Series

EOFs 1-2

SSA T-EOFs



Deseasonalization of Time Series

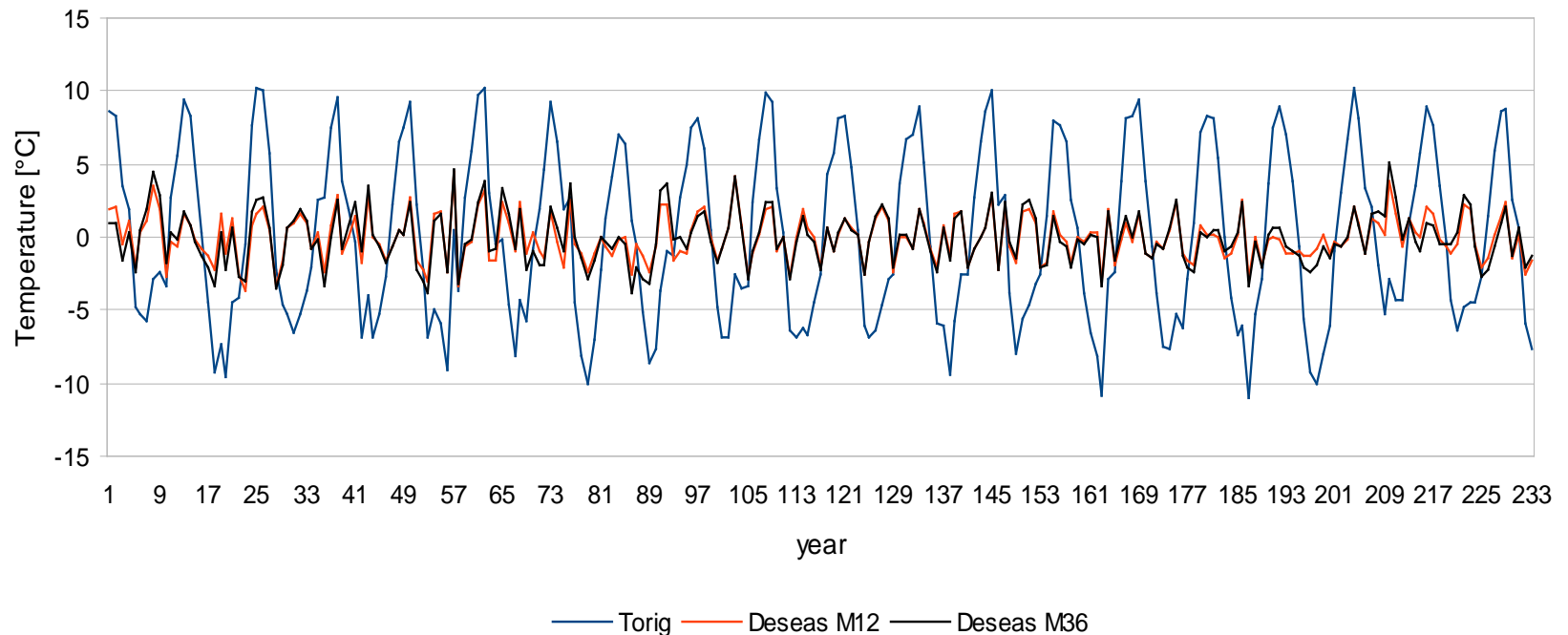
Seasonal components of the TS need to be removed, therefore PC 1-2 are taken away from the original TS.

- Deconstruction of the time series in PCs using a window length of 12 months ($M=12$).
- Reconstruction of the seasonal cycle using the PC1-2.
- Subtraction of PC1-2 from the original time series.

Temperature Time Series

Deseasonalization

Original VS deseasonalized time series



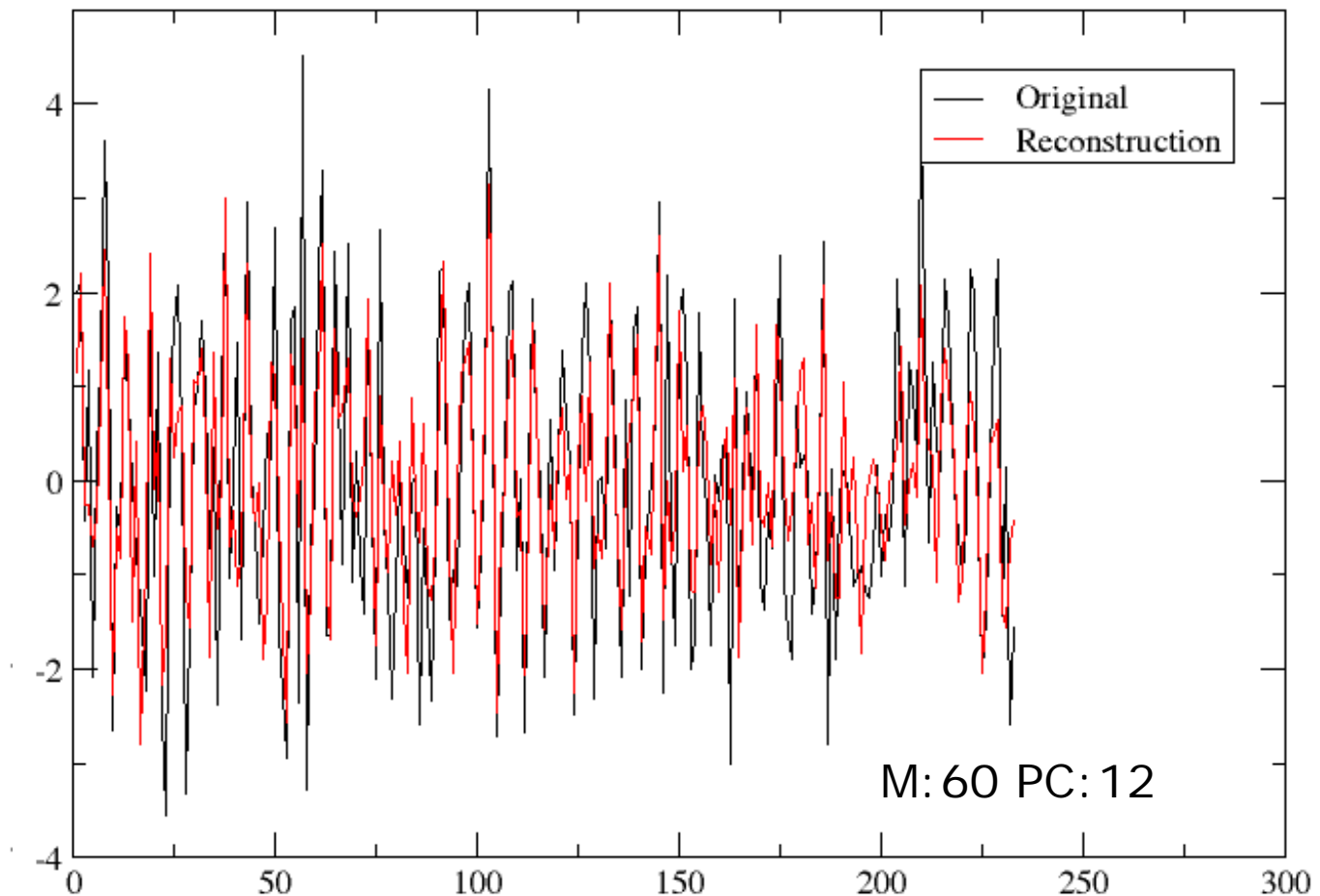
M=12 and M=36 produce similar deseasonalized time series

Deseasonalized Temperature TS

Reconstruction

SSA Reconstruction

Data Vector -tdes, RC Vector -ssarvec, M=60

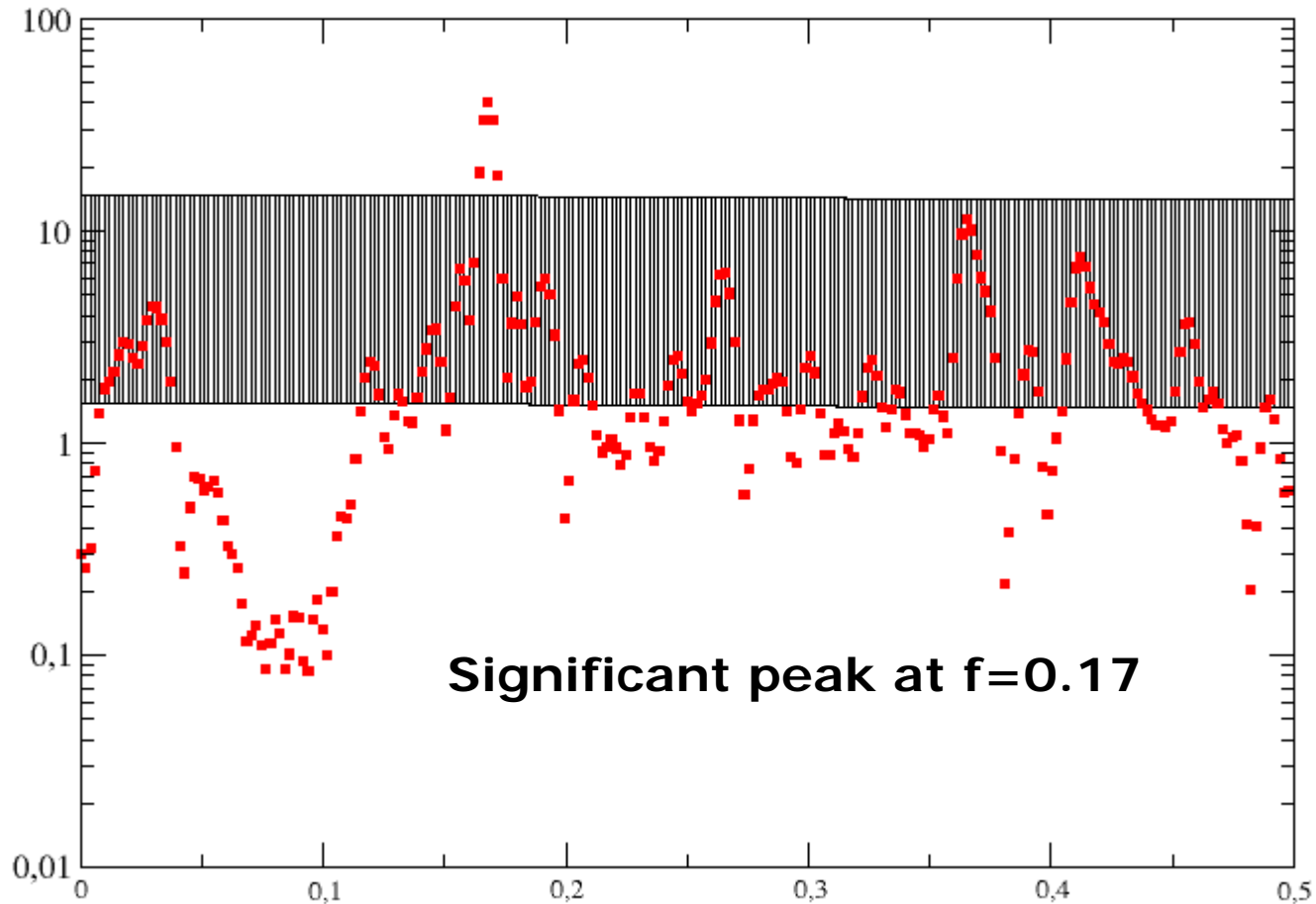


Temperature Time Series

B-T Correlogram – Deseasonalized TS

Blackman-Tukey Spectrum

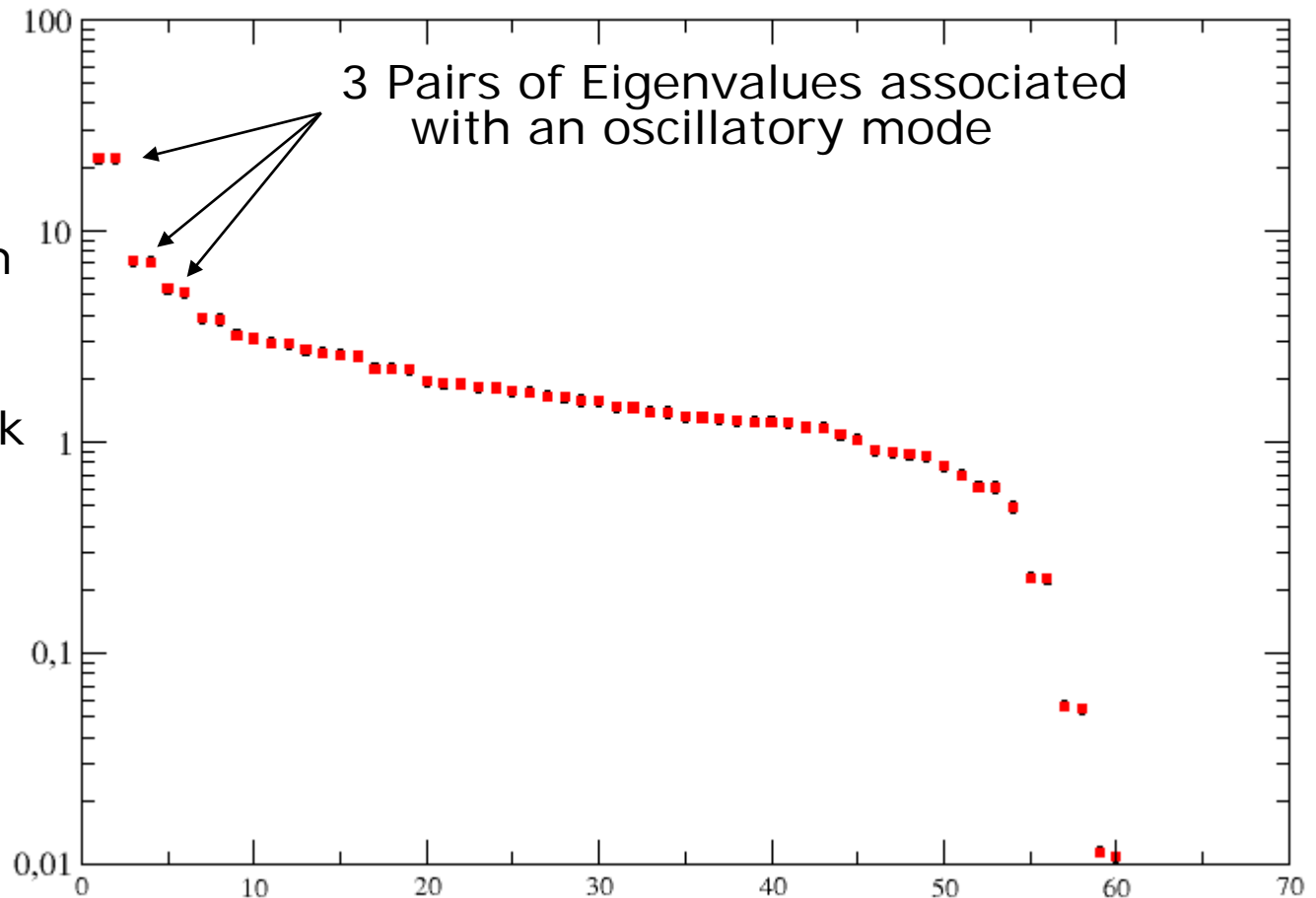
Data Vector -tdes,M=120



Deseasonalized Temperature TS Scree Diagram

SSA

Data Vector -tdes, M=60; AR(1) Error Bars test

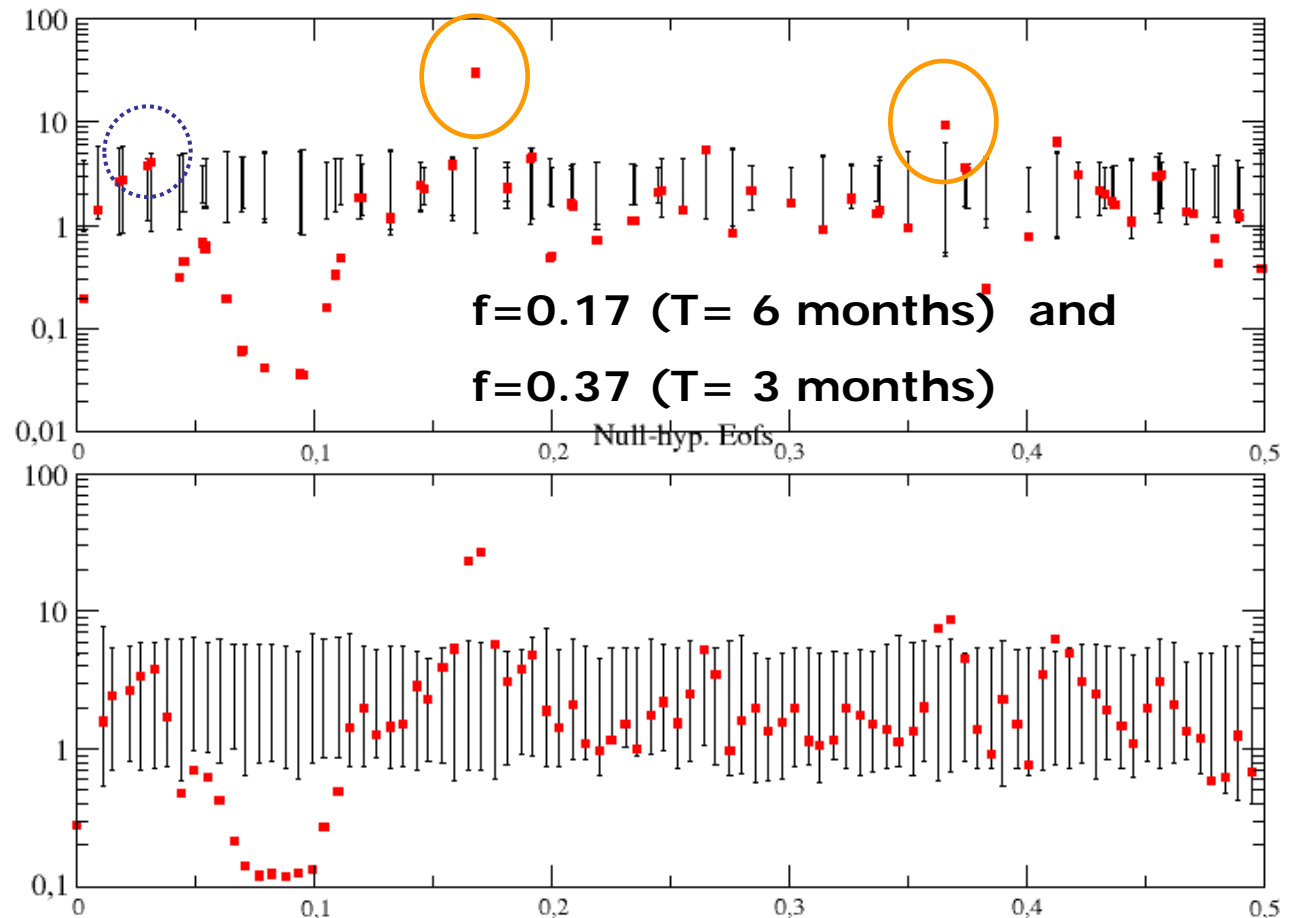


Eigenspectrum
displayed in
order of
eigenvalue rank

Deseasonalized Temperature TS

MCSSA Significance Test

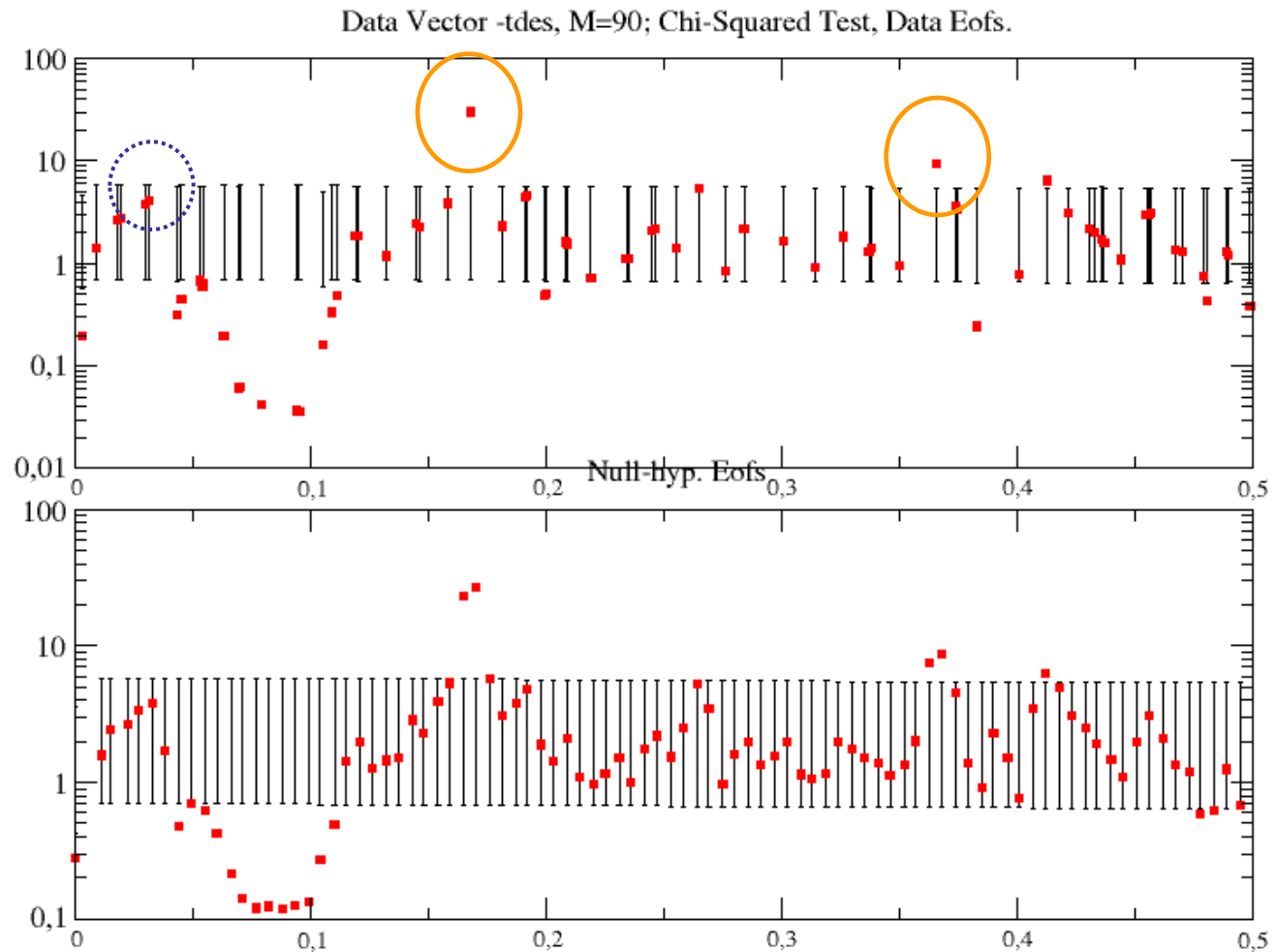
Data Vector -tdes, M=90; MCSSA Test, Data Eofs.



Tests significance against red noise null hypothesis

Deseasonalized Temperature TS

Chi-Squared Significance Test





Conclusions

- The Toolkit has provided a means to fill gaps and remove the seasonal cycle of the temperature TS.
- However, in order to obtain spectral peaks at lower frequencies this temperature TS should be much longer.

Thank you for your attention!

