

# New cycles found in air temperature data and proxy series

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## The Continuous Wavelet transform

The wavelet analysis provides a two-dimensional unfolding of a one-dimensional signal by decomposing it into scale–time coefficients.

The continuous wavelet transform turns a signal  $s$  into a function  $W$

$$W[s](t, a) = \int s(x) \bar{\psi}\left(\frac{x-t}{a}\right) \frac{dx}{a},$$

where  $\bar{\psi}$  denotes the complex conjugate of the function  $\psi$ ,  $a$  the scale and  $t$  the time.

## Conditions on $\psi$

The function  $\psi$  must be integrable, square integrable and satisfy some admissibility condition. Such a function is called a wavelet. The Morlet wavelet is particularly well conditioned for frequency-based study. It satisfies the following equality

$$\hat{\psi}(\omega) = \exp\left(-\frac{(\omega - \Omega)^2}{2}\right) - \exp\left(-\frac{\omega^2 + \Omega^2}{2}\right),$$

where  $\Omega > 5$  is called the central frequency.

## Some Properties of the Wavelet Transform

- the wavelet transform is linear:

$$W[c(u + v)] = cW[s] + cW[v],$$

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- the wavelet transform allows to handle noisy data,
- the wavelet transform is blind to polynomial behaviors (up to a degree  $n$ , depending on  $\psi$ ):  $W[s + P] = W[s]$ , where  $P$  is a polynomial of degree  $\leq n$ .

Consequently, non-zero mean and linear tendencies do not affect the wavelet transform.

## The Scale Spectrum

For wavelets such as the Morlet wavelet, we have

$$W[\cos(\omega_0 t)](t, a) = \exp(i\omega_0 t) \hat{\psi}(a\omega_0),$$

so that the frequency  $\omega_0$  is given by the maximum of  $\hat{\psi}(a\omega_0)$ :  
 $a_\omega = \Omega/\omega_0$ . Consequently, the unknown frequency  $\omega_0$  can be obtained through the maximum of  $|W[\cos(\omega_0 t)]|$ .

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### Definition

The scale spectrum of a signal  $s$  is defined by

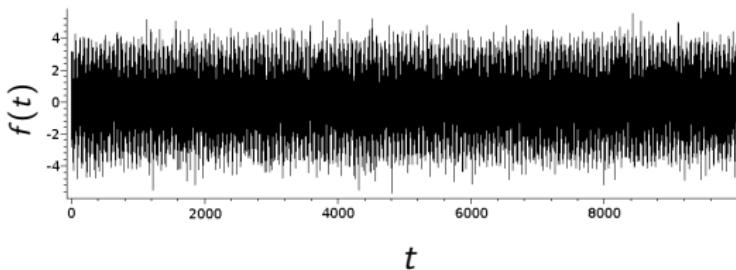
$$\Lambda(a) = E|W[s](t, a)|,$$

where  $E$  denotes the mean over the time  $t$ .

## For What Purpose?

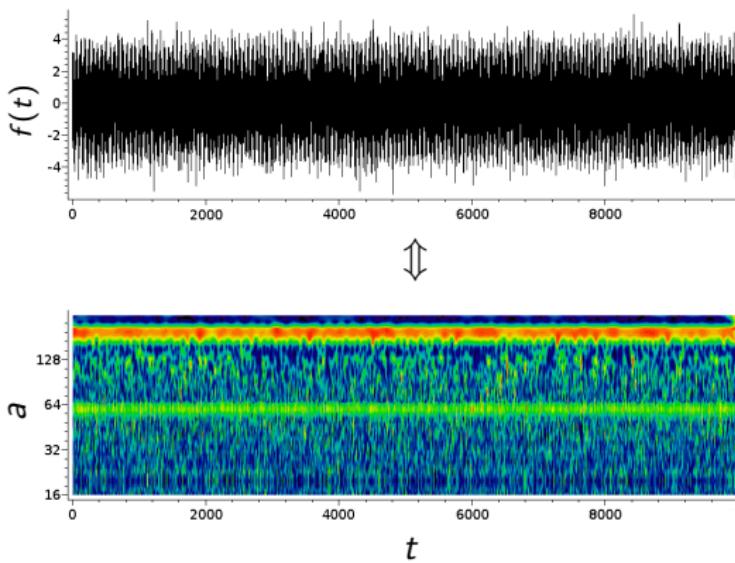
The scale spectrum should be useful for signals which are not stationary but whose characteristics do not evolve too quickly: the scale spectrum allows to recover frequencies even if they “kindly depend on the time”, i.e. if  $\omega_0 = \omega_0(t)$  with  $\frac{d}{dt}\omega_0 \ll 1$ , one should be able to recover  $E\omega_0$ .

# A Visual Example



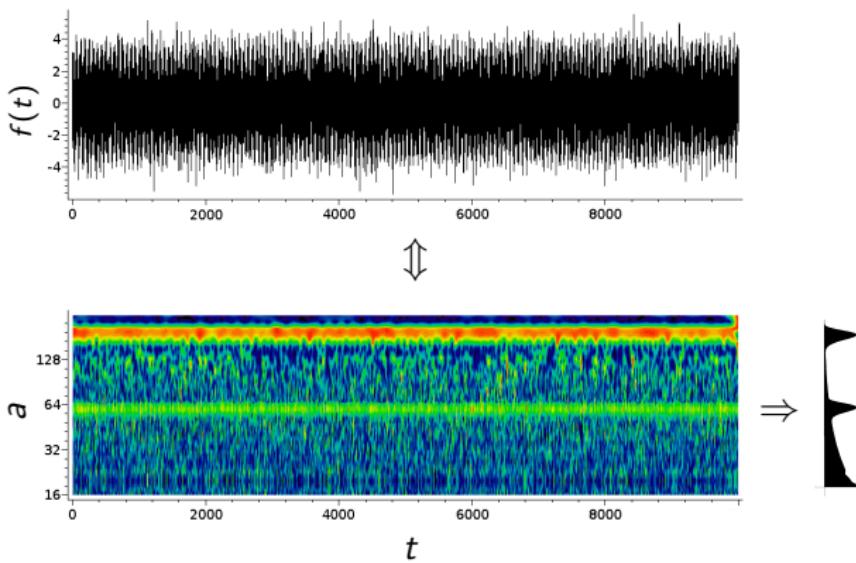
Example

## A Visual Example

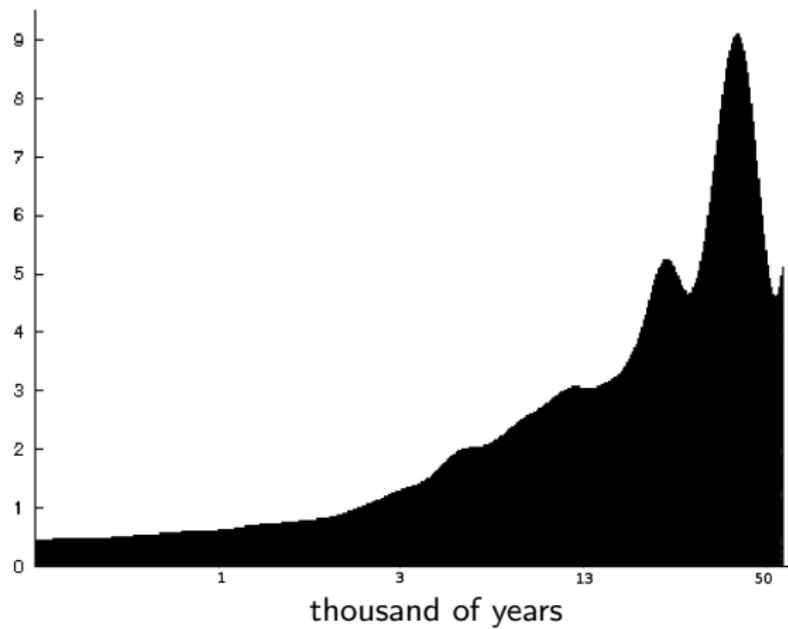


Example

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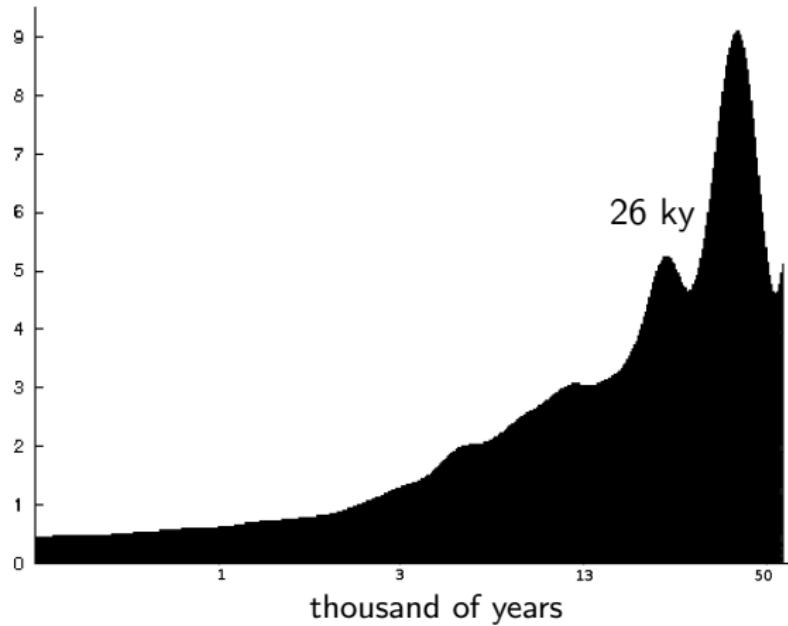


## Known long cycles are observed in the spectrum



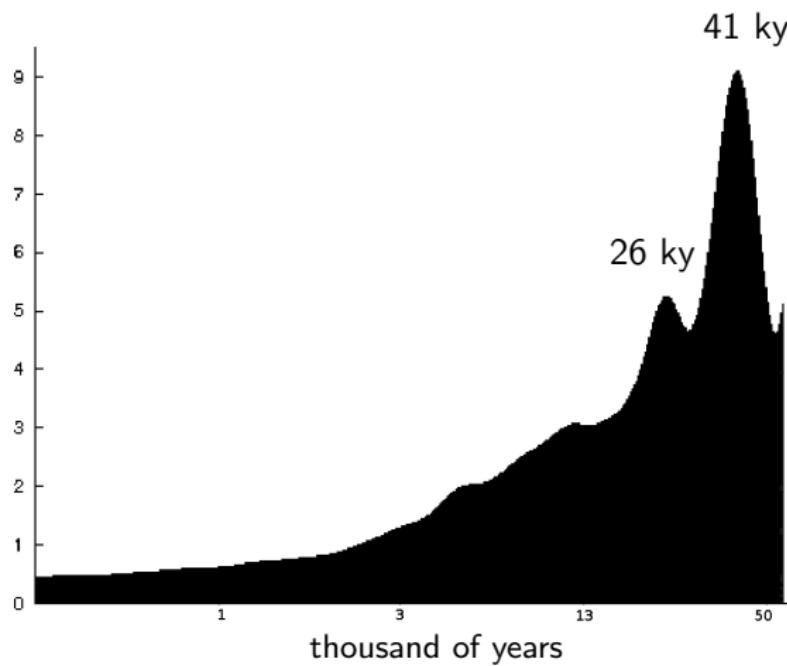
EPICA dome

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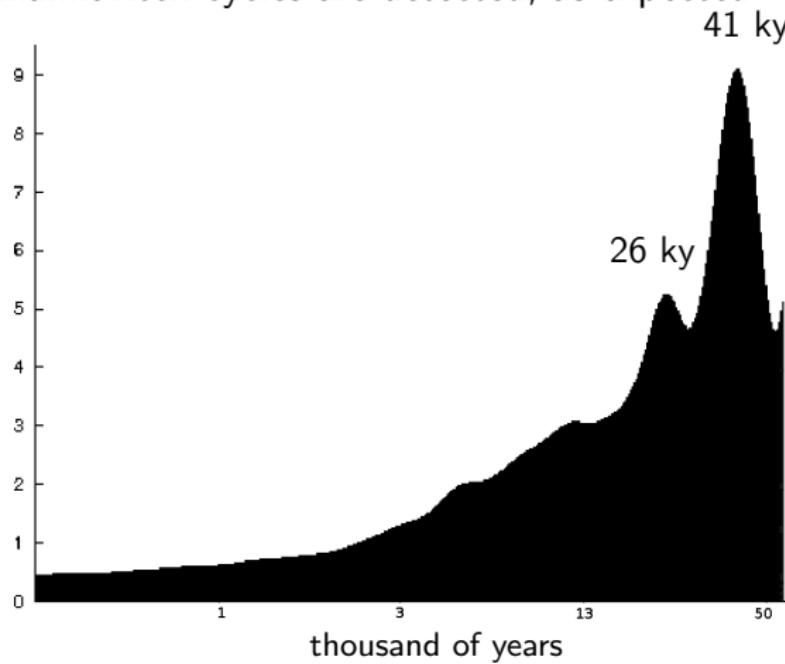
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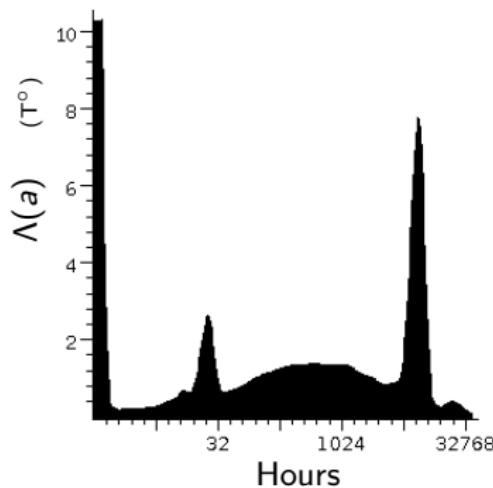
## Known long cycles are observed in the spectrum

Milankovitch cycles are detected, as expected



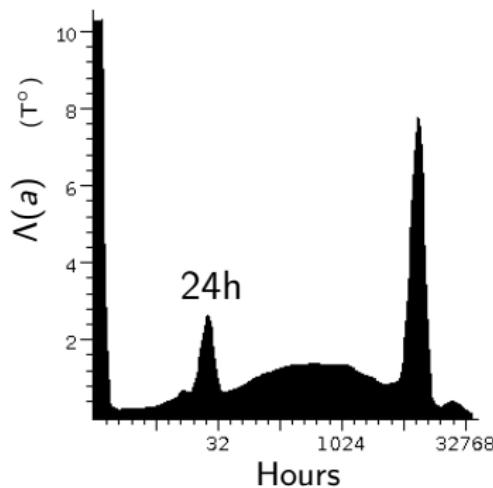
EPICA dome

## What About the Temperature Data?



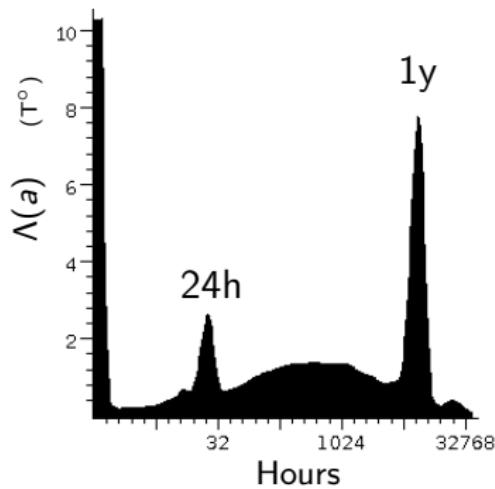
Bierset-aero, Belgium  
1950–2007 hourly-sampled data

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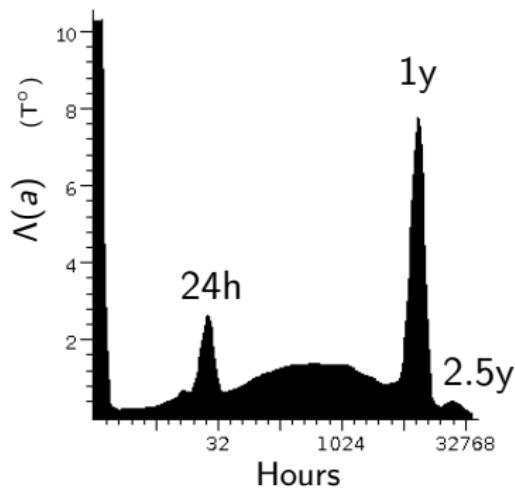
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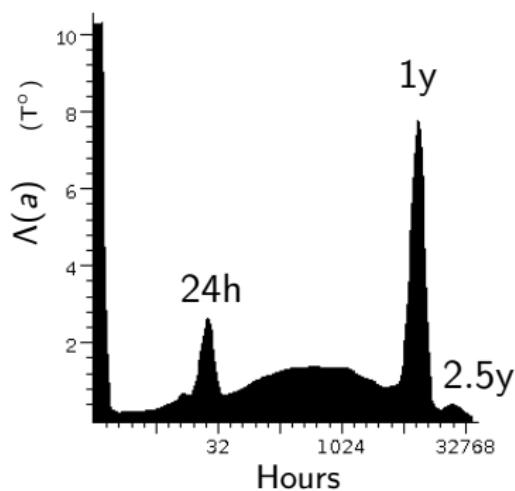
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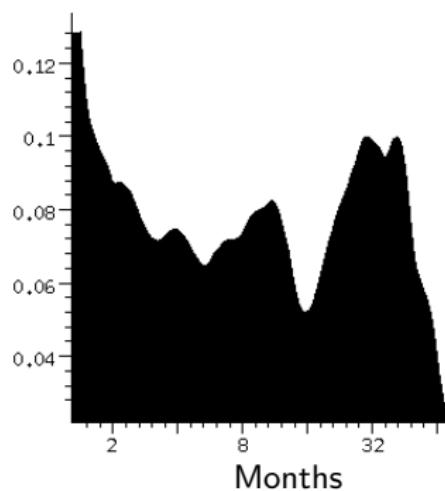


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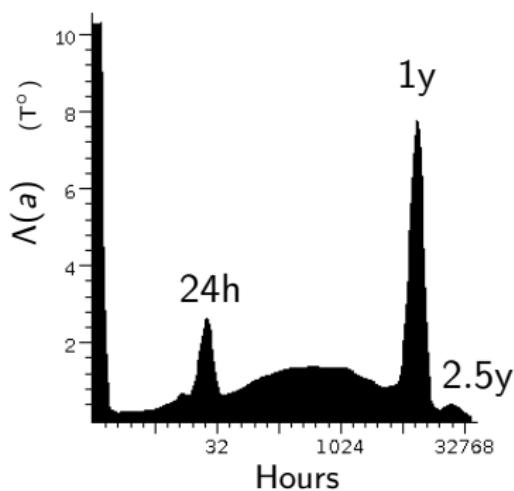


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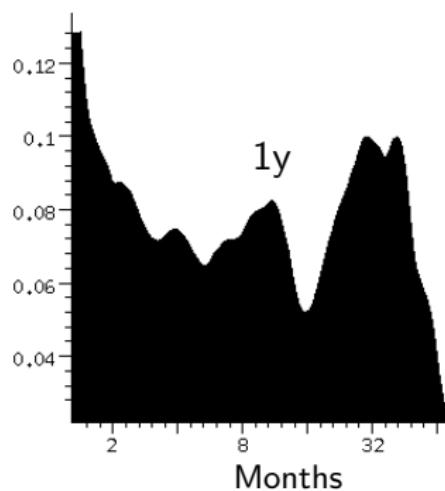


crutempg13  
1950–2007 monthly-sampled data

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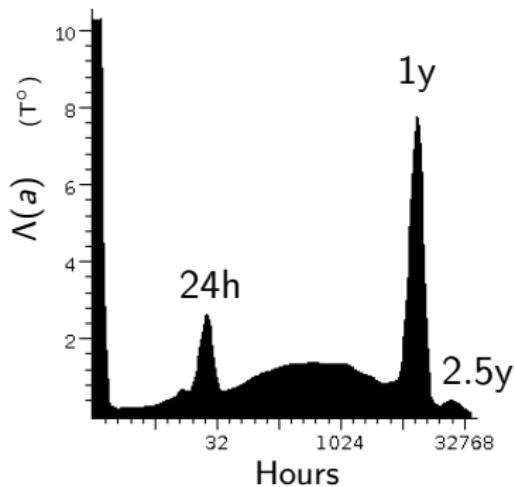


Bierset-aero, Belgium  
1950–2007 hourly-sampled data

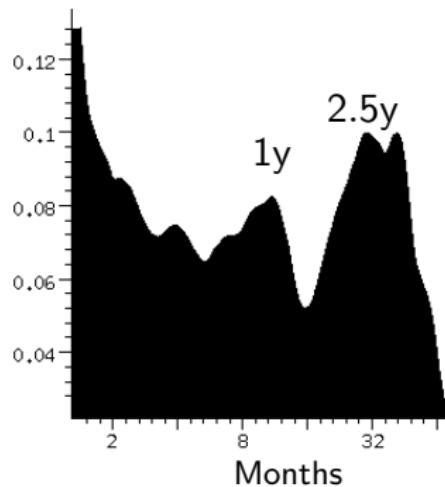


crutempgl3  
1950–2007 monthly-sampled data

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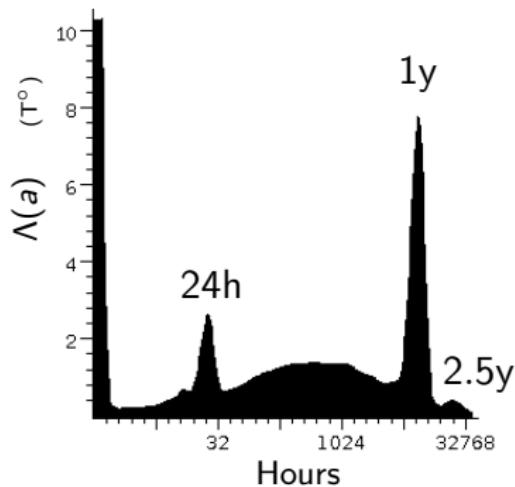


Bierset-aero, Belgium  
1950–2007 hourly-sampled data

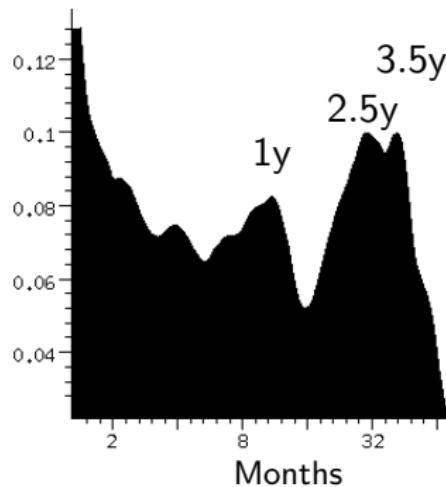


crutempgl3  
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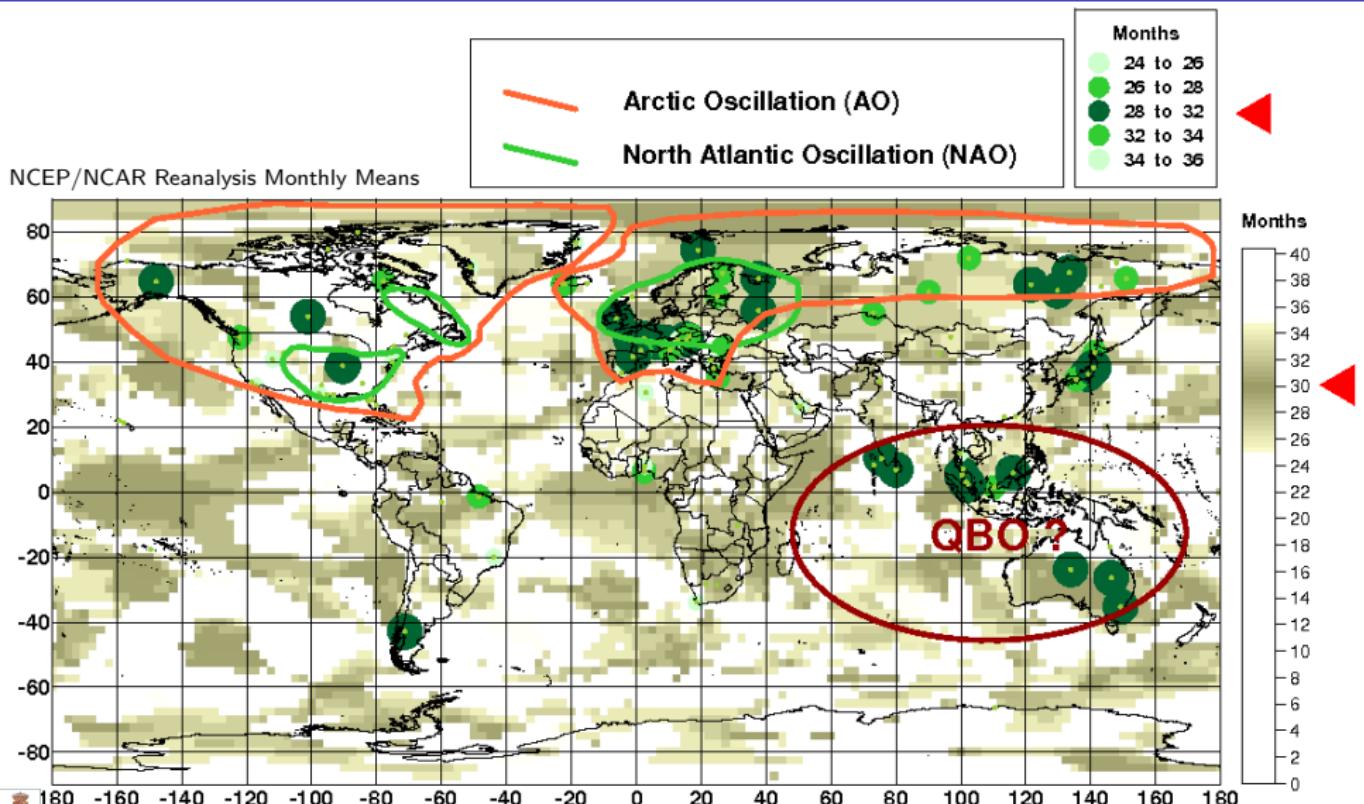


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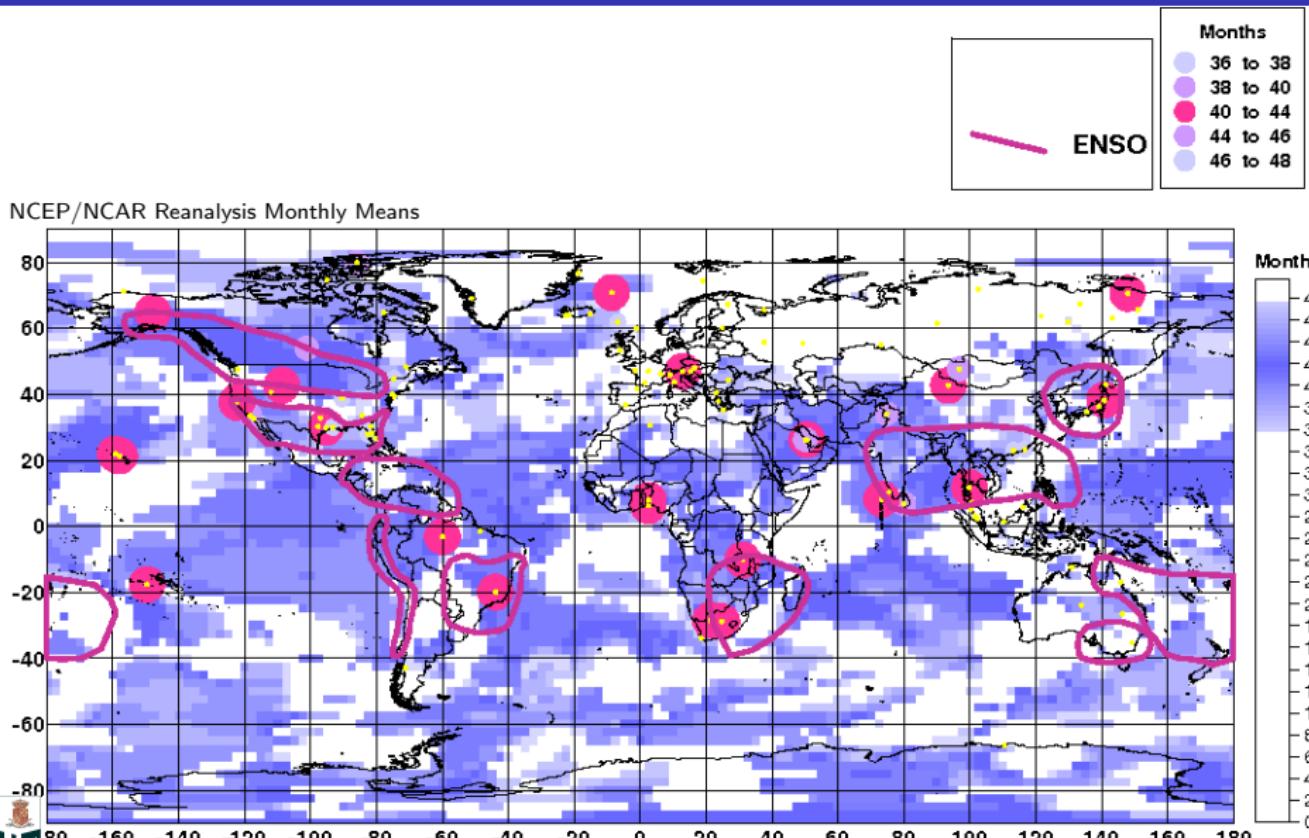


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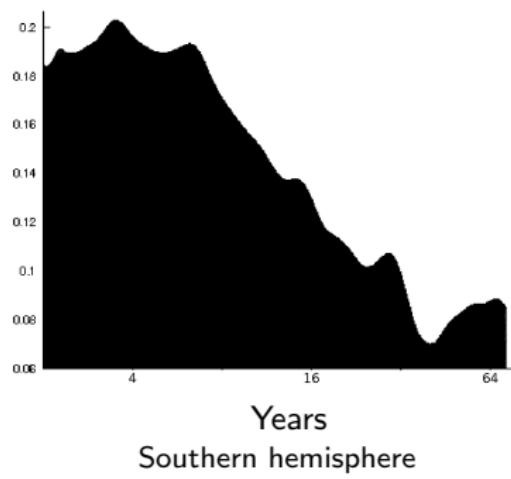
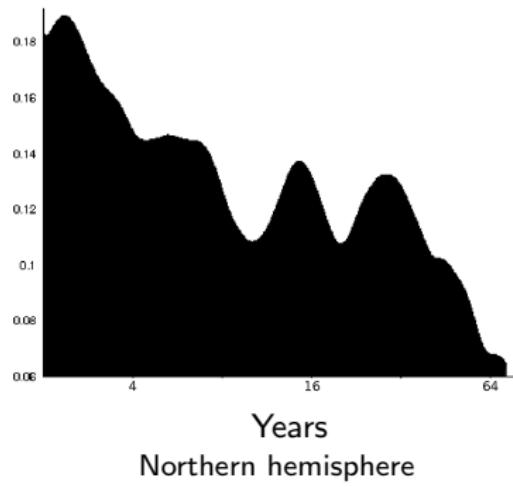
## 30 Months Cycle



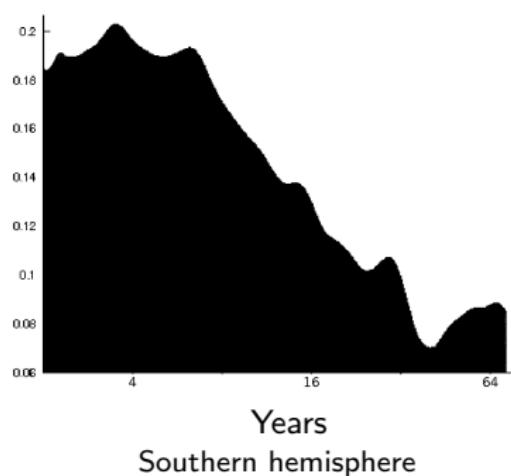
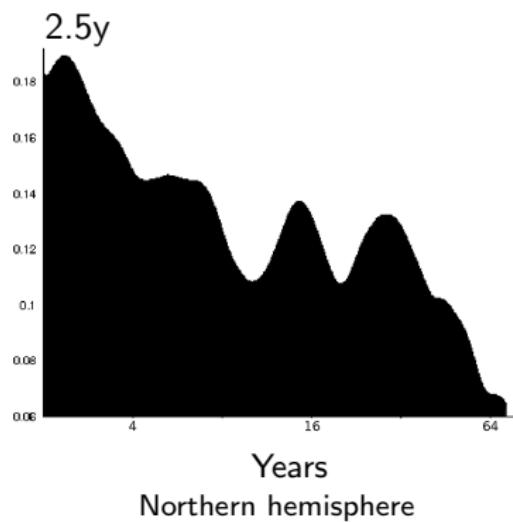
## 42 Months Cycle



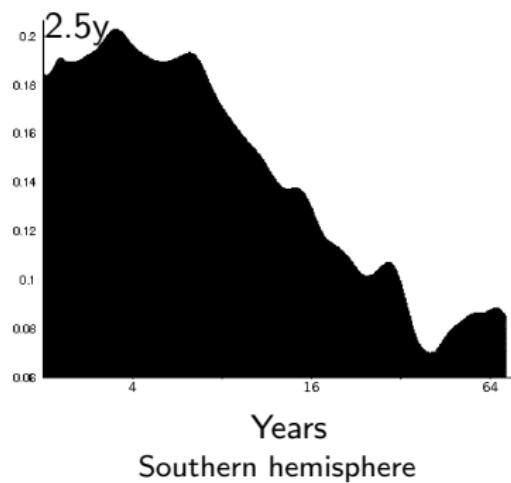
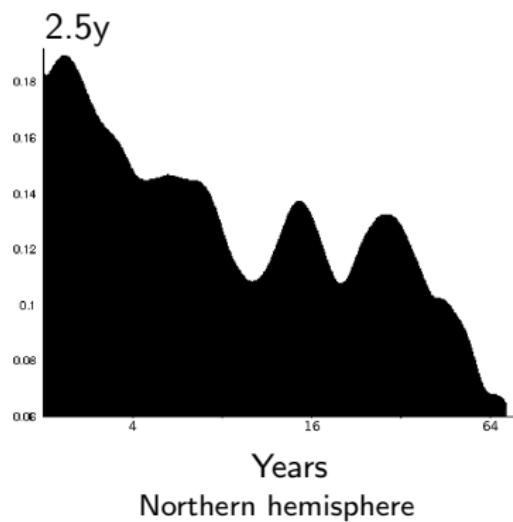
# Millennial Temperature Reconstructions of Jones et al.



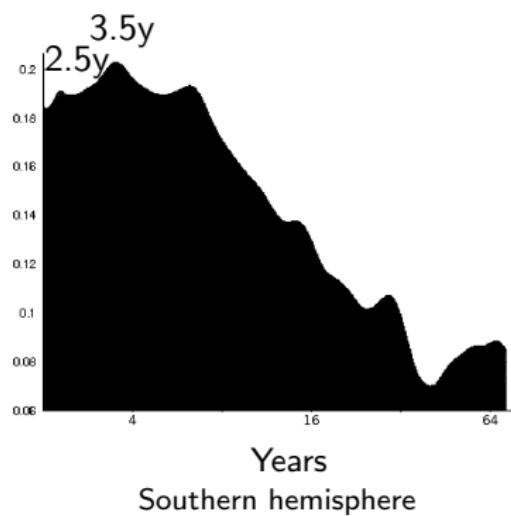
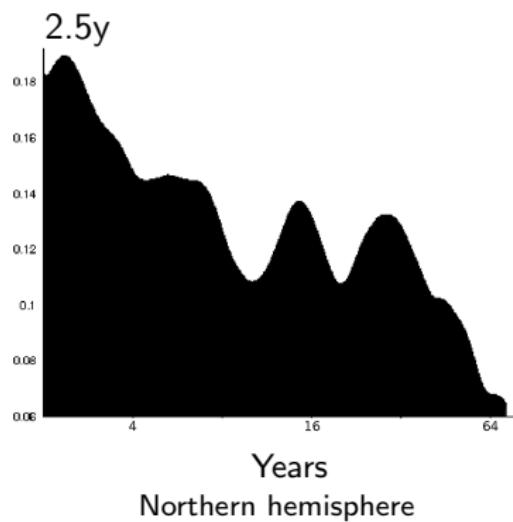
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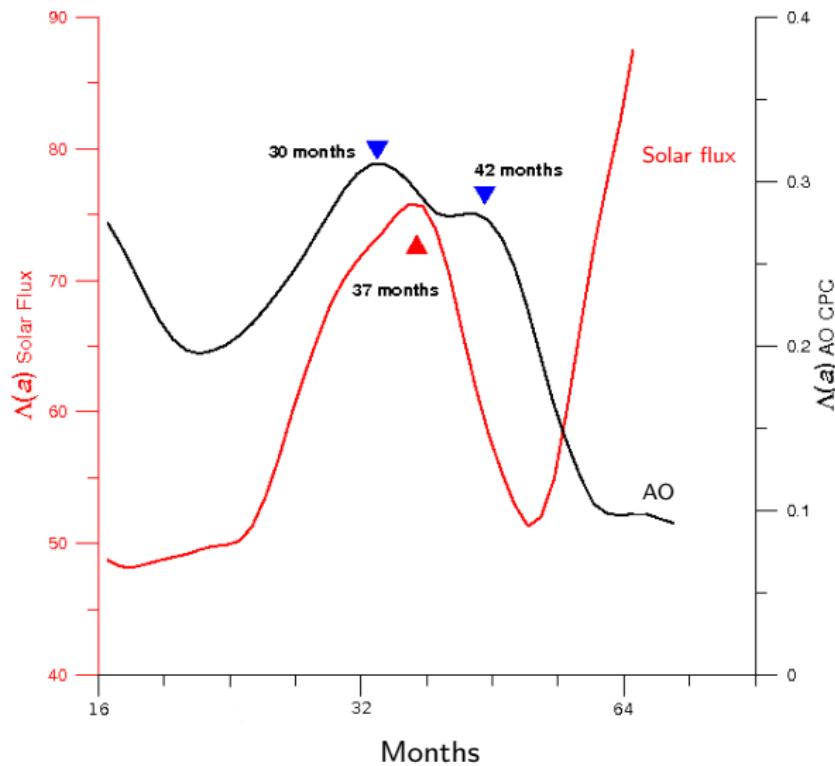
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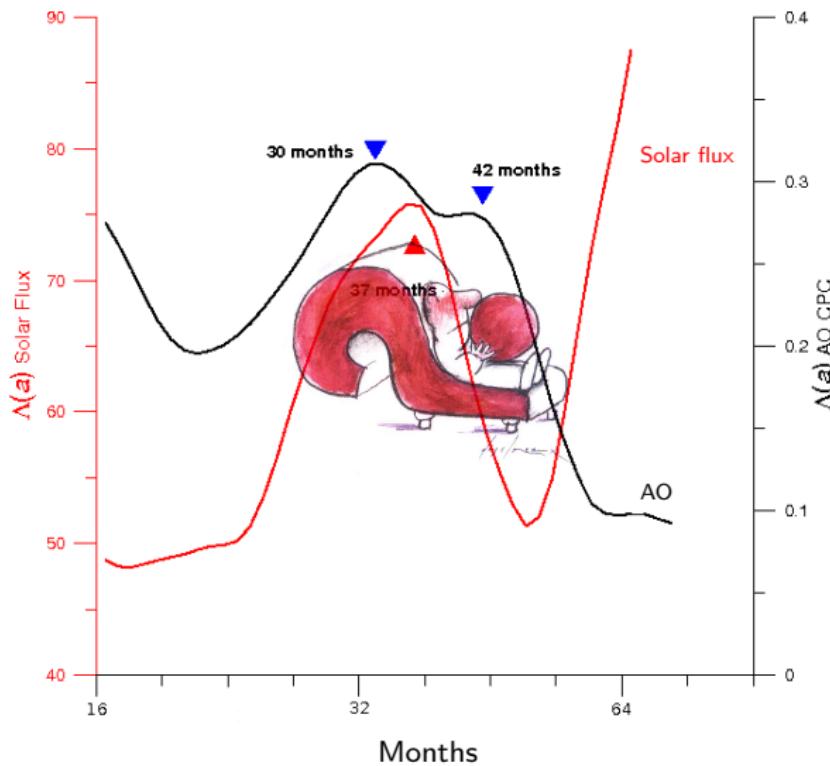
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## A Relation with the Sun?



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## For Further Reading



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Log-term variations in the correlation between NAO and solar activity: The importance of North-South solar activity asymmetry for atmospheric circulation, *ASR*, 40, 1152–66 (2007).



K. Labitzke,

On the solar-cycle–QBO Relationship: A summary, *J.A.S.-T.P., special issue*, 67, 45–54 (2005).



S. Nicolay, G. Mabille, X. Fettweis and M. Erpicum,

A 30 and a 43 months period cycles in air temperature time series using the Morlet wavelet method, *submitted*.



M. Paluš and D. Novotná,

Quasi-biennial oscillations extracted from the monthly NAO index and temperature records are phase-synchronized, *Nonlin. Processes Geophys.*, 13, 287–96 (2006).