

Statistical Signal Processing  
(2009-2010)  
Handout #8: Hilbert-Huang Transform

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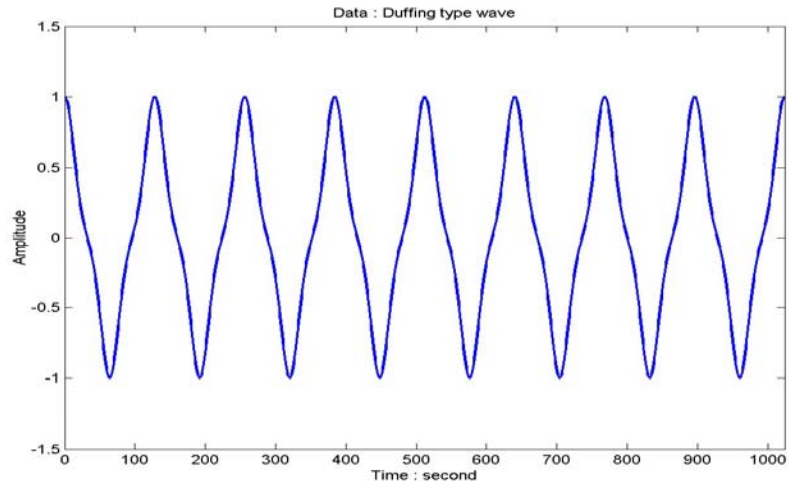
## Outline

- A motivational example: Why Hilbert-Huang transform?
- Hilbert-Huang transform
  - The Empirical Mode Decomposition (EMD) method
  - Hilbert Spectral Analysis
- Comparison with Fourier transform and wavelet transform
- Further information



## A motivational example

- $x(t) = \cos(\omega t + 0.3 \sin(2\omega t))$



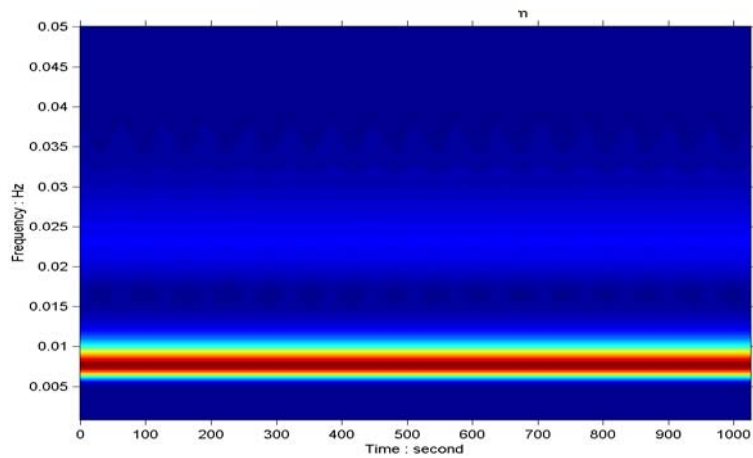
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See SSP\_Handout8\_HHT.m

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## A motivational example

- Wavelet transform cannot exactly reveal how the frequency is varying with time.

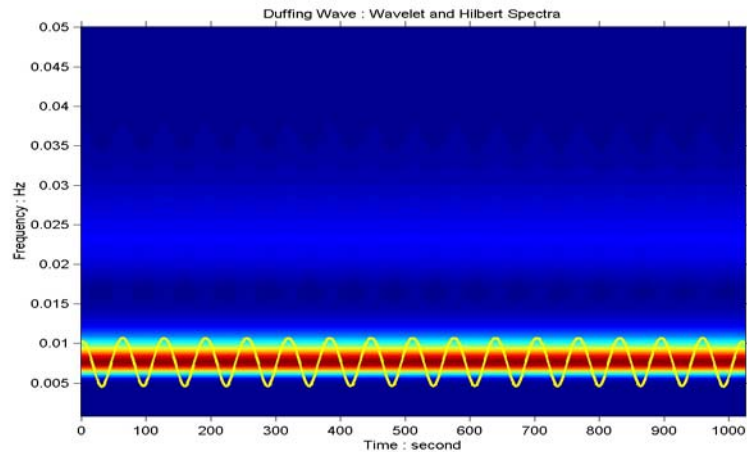


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## A motivational example

- Hilbert-Huang transform deals with this problem successfully.



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## Hilbert-Huang Transform

- The HHT provides a new method of analyzing nonstationary and nonlinear time series data.
- The HHT consists of the Empirical Mode Decomposition (EMD) method (from Norden E. Huang) and Hilbert Spectral Analysis (HSA) (from David Hilbert (1862-1943)).
- In particular, the HHT uses the EMD method to decompose a signal into so-called intrinsic mode functions (IMFs), and uses the HSA method to obtain instantaneous frequency data.

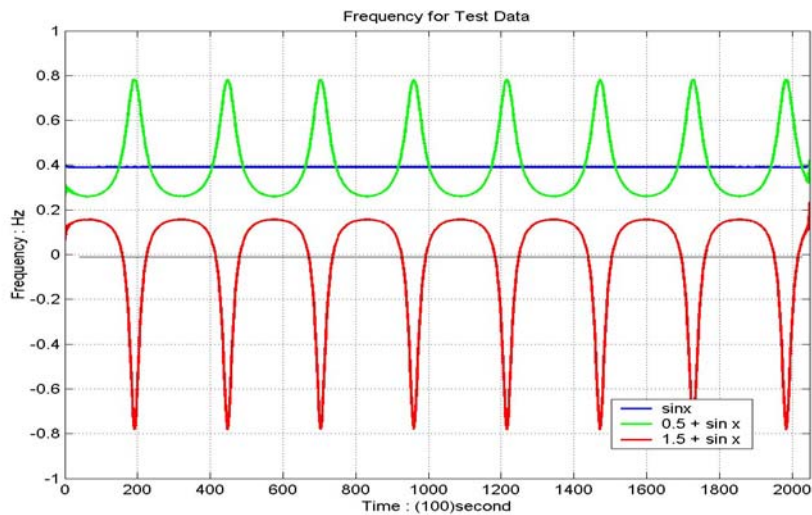
Why not use the HSA directly?



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## Hilbert-Huang Transform

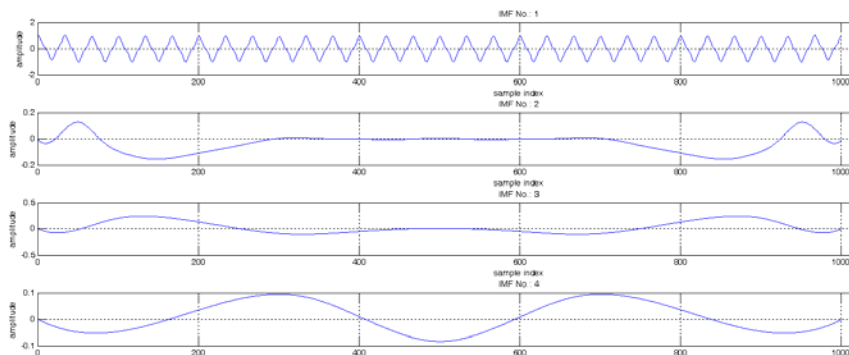


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## Hilbert-Huang Transform: The EMD method

- The EMD method is based on a simple assumption that any data consists of different simple intrinsic modes of oscillations. Each intrinsic mode, linear or nonlinear, represents a simple oscillation, which will have the same number of extrema and zero crossings.

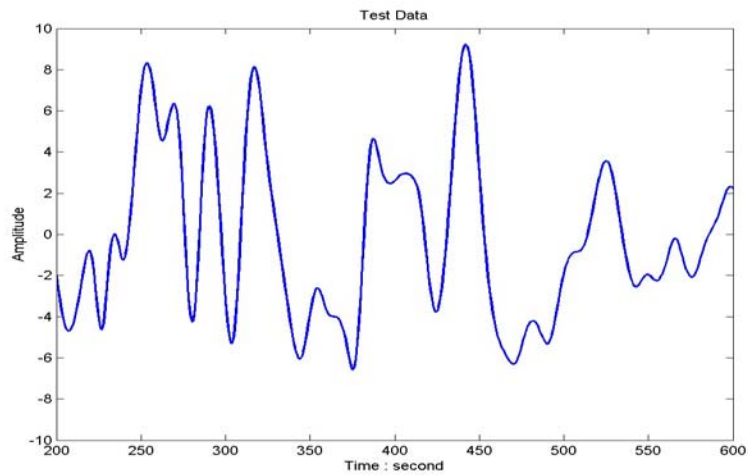


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## Hilbert-Huang Transform: The EMD method

- Original data

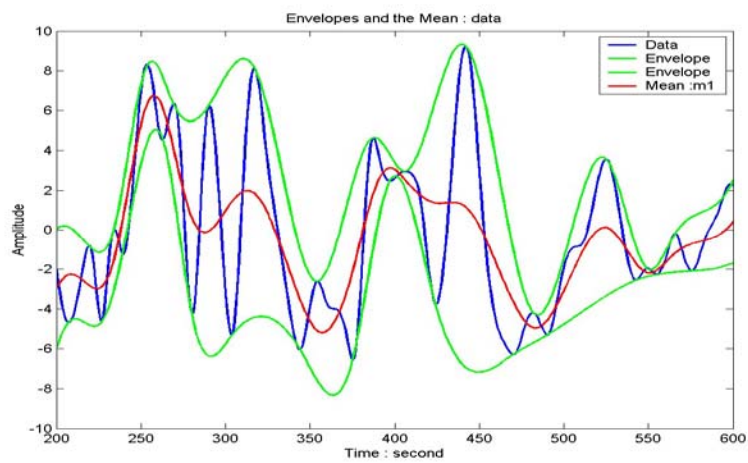


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## Hilbert-Huang Transform: The EMD method

- Identify all the local extrema, and connect all the local maxim (minima) by a cubic spline as shown the upper (lower) envelope.

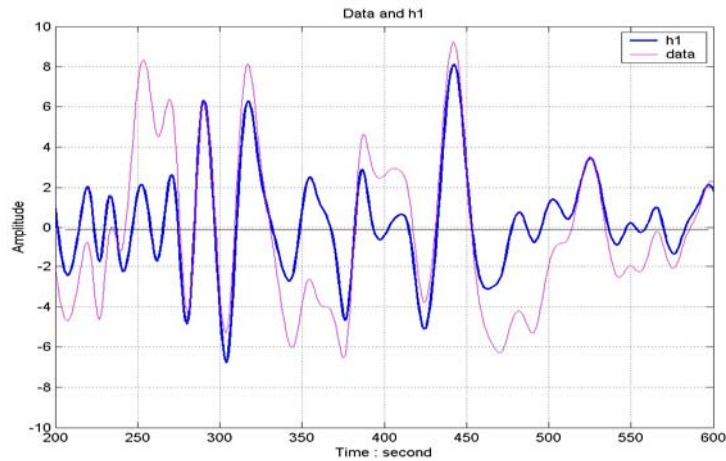


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## Hilbert-Huang Transform: The EMD method

- The mean of the upper and lower envelope is designated as  $m_1$ , and the difference between the data  $x(t)$  and  $m_1$  is the first component  $h_1 = x(t) - m_1$ .



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## Hilbert-Huang Transform: The EMD method

- Repeat the same process till the resulted component  $h_k$  satisfies the requirements for the intrinsic mode function (IMF): the number of extrema and the number of zero-crossings much either equal or differ at most by one.

$$x(t) - m_1 = h_1,$$

$$h_1 - m_2 = h_2,$$

.....

.....

$$h_{k-1} - m_k = h_k.$$

$$\Rightarrow h_k = c_1.$$

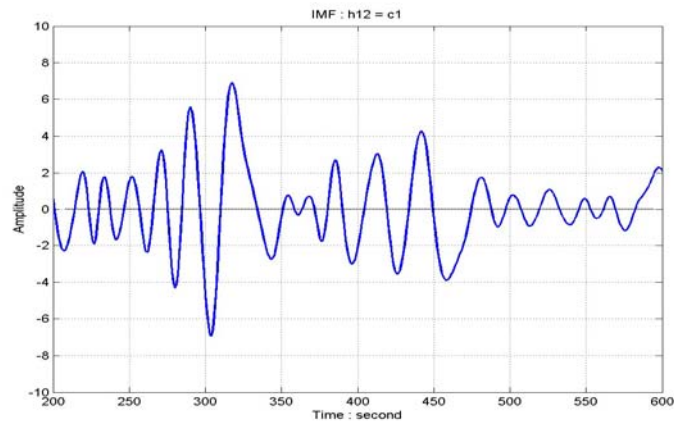


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## Hilbert-Huang Transform: The EMD method

- Repeat the same process till the resulted component  $h_k$  satisfies the requirements for the intrinsic mode function (IMF): the number of extrema and the number of zero-crossings much either equal or differ at most by one.



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## Hilbert-Huang Transform: The EMD method

- Repeat the above steps to obtain all the IMF components till no more IMFs can be extracted.

	IMF1	IMF2	IMF3	IMF n
	$X(t)$	$X(t) - c_1 = r_1$	$r_1 - c_2 = r_2$	$r_{n-2} - c_{n-1} = r_{n-1}$
0	$X(t) - m_1 = h_1$	$r_1 - m_2 = h_2$	$r_2 - m_3 = h_3$	$r_{n-1} - m_n = h_n$
1	$h_1 - m_{11} = h_{11}$	$h_2 - m_{21} = h_{21}$	$h_3 - m_{31} = h_{31}$	$h_n - m_{n1} = h_{n1}$
2	$h_{11} - m_{12} = h_{12}$	$h_{21} - m_{22} = h_{22}$	$h_{31} - m_{32} = h_{32}$	$h_{n1} - m_{n2} = h_{n2}$
3	$h_{12} - m_{13} = h_{13}$	$h_{22} - m_{23} = h_{23}$	$h_{32} - m_{33} = h_{33}$	$h_{n2} - m_{n3} = h_{n3}$
..	.....			
k	$h_{1(k-1)} - m_{1k} = h_{1k}$	$h_{2(k-1)} - m_{2k} = h_{2k}$	$h_{3(k-1)} - m_{3k} = h_{3k}$	$h_{n(k-1)} - m_{nk} = h_{nk}$
IMF	$h_{1k} = c_1$	$h_{2k} = c_2$	$h_{3k} = c_3$	$h_{nk} = c_n$



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## Hilbert-Huang Transform: The EMD method

- Repeat the above steps to obtain all the IMF components till no more IMFs can be extracted.

$$x(t) - c_1 = r_1 ,$$

$$r_1 - c_2 = r_2 ,$$

...

$$r_{n-1} - c_n = r_n .$$

$$\Rightarrow x(t) - \sum_{j=1}^n c_j = r_n .$$



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## Hilbert-Huang Transform: The EMD method

- To be able to analyze data from the nonstationary and nonlinear processes and reveal their physical meaning, the method has to be **Adaptive**.
- Adaptive requires *a posteriori* (not *a priori*) basis. But the present established mathematical paradigm is based on *a priori* basis.
- Only *a posteriori* basis could fit the varieties of nonlinear and nonstationary data without resorting to the mathematically necessary (but physically nonsensical) harmonics.



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## Hilbert-Huang Transform: The HSA method

- The Hilbert transform of any real-valued signal is

$$H[x(t)] = \frac{1}{\pi} PV \int_{-\infty}^{\infty} \frac{x(\tau)}{t-\tau} d\tau$$

Here PV indicates the principle value of the singular integral.

Let  $[\alpha, \beta]$  be a real interval and let  $f$  be a complex-valued function defined on  $[\alpha, \beta]$ . If  $f$  is unbounded near an interior point  $\zeta$  of  $[\alpha, \beta]$ , the integral of  $f$  over  $[\alpha, \beta]$  does not always exist. However, it may happen that the symmetric limit

$$\lim_{\epsilon \rightarrow 0^+} \left( \int_{\alpha}^{\zeta-\epsilon} f(x) dx + \int_{\zeta+\epsilon}^{\beta} f(x) dx \right)$$

exists. If it does, it is called the principle value of  $f$  from  $\alpha$  to  $\beta$ , and is denoted as

$$PV \int_{\alpha}^{\beta} f(x) dx$$



## Hilbert-Huang Transform: The HSA method

- With the Hilbert transform,

$$x(t) = A(t) \cos \Phi(t)$$

where

$$A(t) = \sqrt{(x(t))^2 + (H[x(t)])^2}$$

$$\Phi(t) = \arctan \left( \frac{H[x(t)]}{x(t)} \right)$$

- The instantaneous frequency is the time derivative of the phase.

$$\omega = \frac{d\Phi(t)}{dt}$$

Hence, we have

$$x(t) = \operatorname{Re} \left\{ A(t) e^{i \int \omega(t) dt} \right\}$$



## Hilbert-Huang Transform: The HSA method

- Given the period of a wave as  $T$ ; the frequency is defined as

$$\omega = \frac{1}{T}.$$

- The definition of instantaneous frequency is equivalent to defining velocity as

$$\text{Velocity} = \frac{\text{distance}}{\text{time}}; \text{ mean velocity } \quad \text{Newton} \Rightarrow v = \frac{dx}{dt}$$

$$\text{Frequency} = \frac{1}{\text{period}}; \text{ mean frequency}$$

$$\text{HHT defines the phase function} \Rightarrow \omega = \frac{d\theta}{dt}$$



## Hilbert-Huang Transform: The HSA method

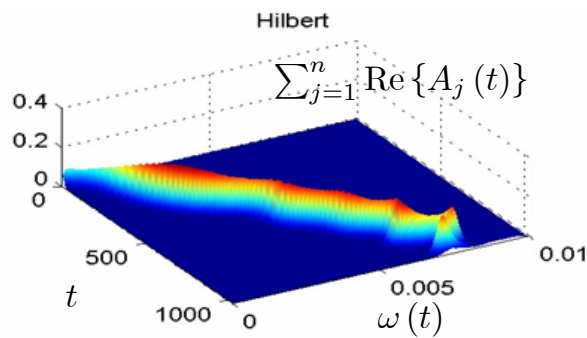
- To be able to analyze data from the nonstationary and nonlinear processes and reveal their physical meaning, the method has to be **local**.
- Locality requires **differential operation** to define properties of a function.
- Take frequency for example. The present established mathematical paradigm is based on Integral transform. But integral transform suffers the limitation of the uncertainty principle.
- Instantaneous Frequency** offers a total different view for nonlinear data: instantaneous frequency with no need for harmonics and unlimited by uncertainty.



## Hilbert-Huang Transform: The HSA method

- The HSA is the time-frequency representation of the original signal,

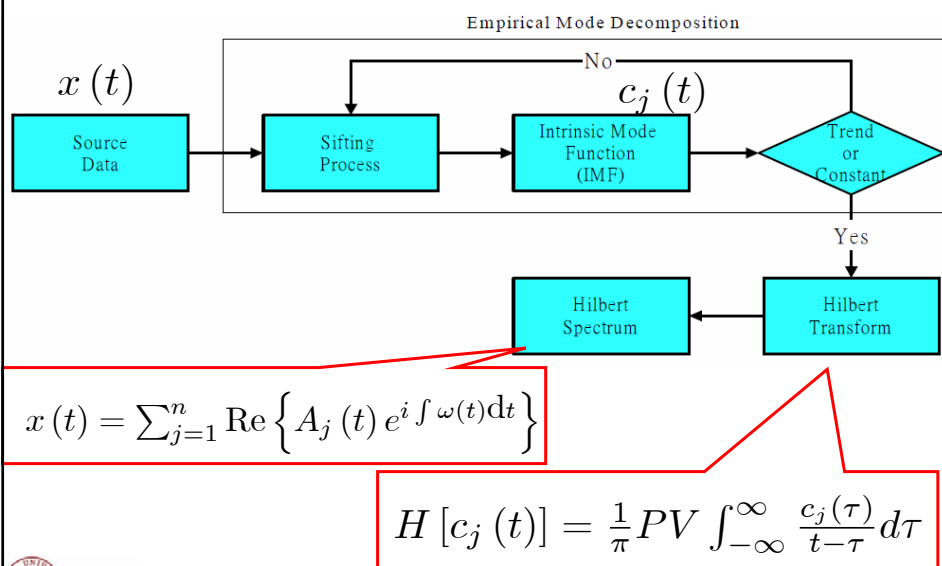
$$x(t) = \sum_{j=1}^n c_j = \sum_{j=1}^n \text{Re} \left\{ A_j(t) e^{i \int \omega(t) dt} \right\}$$



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## Hilbert-Huang Transform: The computation flowchart



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## Comparison

	Fourier	Wavelet	Hilbert
<b>Basis</b>	a priori	a priori	Adaptive
<b>Frequency</b>	Convolution: Global	Convolution: Regional	Differentiation: Local
<b>Presentation</b>	Energy-frequency	Energy-time- frequency	Energy-time- frequency
<b>Nonlinear</b>	no	no	yes
<b>Non-stationary</b>	no	yes	yes
<b>Uncertainty</b>	yes	yes	no
<b>Harmonics</b>	yes	yes	no

See SSP\_Handout8\_HHT.m for another example.



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## Further information

- N.E. Huang, A Plea for Adaptive Data Analysis: An Introduction to HHT (*presented by Dr. Huang at PKU in the summer of 2008*)
- C.M. Hsieh, Hilbert Huang Transform, 2007 (in Chinese).
- N.E. Huang, Introduction to the Hilbert-Huang transform and its related mathematical problems, Chapter 1 in Hilbert-Huang Transform and Its Applications, N.E. Huang and S. P. Shen (Eds), World Scientific Publishing Company 2005. (*A short but well-written introductory-level article on HHT*).
- A comprehensive website including Matlab program, papers, course materials at <http://rcada.ncu.edu.tw/research1.htm>



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