

A plea for adequate data analysis methods

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A plea for adequate data analysis

- Whatever we want to do related to MCMD:
 - On forecast hazard occurrence
 - On increase Infrastructure resiliency
 - On quantify risk, and
 - On minimize hazard impact
- We need good data and, more importantly, **adequate data analysis methods.**

With the advance of IT and sensor technologies, there is an explosion of data.

We are now drowning in data yet still thirsty for information.

Scientific Activities

Collecting, analyzing, synthesizing, and theorizing are the core of scientific activities.

Data analysis is a key link in this continuous loop.

Data is the link to reality

Data analysis for engineering

- Data for design specifications
- Data for health monitoring
- Data for quantify risk
- Data to minimizing hazard impacts

Example: Health monitoring

- Most natural hazards are hard to predict; therefore, it is critical that we should increase the structures resiliency to withstand the impacts of the hazards. The easiest and most logic ways to maximize structural resiliency are these:
 - To design the structure intelligently, and
 - To keep the structure in healthy conditions and good repair.

Land Slide damages



**Photograph by R.L. Schuster,
U.S. Geological Survey, 1995.)**

Earthquake Damage

(www.cedim.de)



Hurricane Damage

Southern Grenada after Hurricane Ivan (Photo WW-2004).



Data Analysis

Data analysis is too important to be left to the mathematicians.

Why?!

Different Paradigms I

Mathematics vs. Science/Engineering

- **Mathematicians**

- Absolute proofs
- Logic consistency
- Mathematical rigor

- **Scientists/Engineers**

- Agreement with observations
- Physical meaning
- Working Approximations

Data Processing vs. Analysis

In pursue of mathematic rigor and certainty,
however, we lost sight of physics and are forced
to **idealize, but also deviate from, the reality.**

As a result, we are forced to live in a pseudo-real
world, in which all processes are

Linear and Stationary

On nonlinear oscillations

The key to structure health monitoring

How to define nonlinearity?

Based on Linear Algebra: nonlinearity is defined based on input vs. output.

But in reality, such an approach is not practical. The alternative is to define nonlinearity based on data characteristics.

Linear Systems

Linear systems satisfy the properties of **superposition** and **scaling**. Given two valid inputs

$$x_1(t) \text{ and } x_2(t)$$

as well as their respective outputs

$$y_1(t) = H\{x_1(t)\} \text{ and}$$

$$y_2(t) = H\{x_2(t)\}$$

then a linear system must satisfy

$$\alpha y_1(t) + \beta y_2(t) = H\{\alpha x_1(t) + \beta x_2(t)\}$$

for any scalar values **α** and **β** .

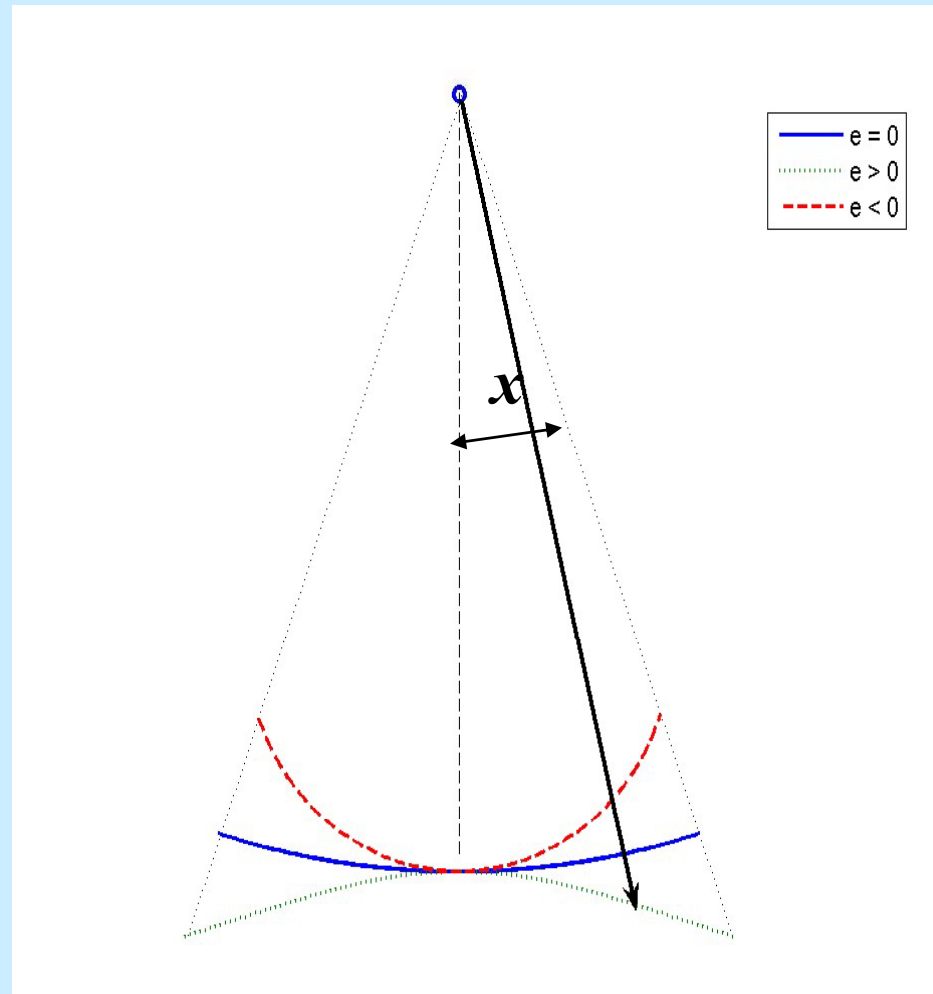
Characterizing Nonlinear Processes from Data

$$\frac{d^2 x}{dt^2} + x + \epsilon x^3 = \gamma \cos \omega t$$

$$\Rightarrow \frac{d^2 x}{dt^2} + x \left(1 + \epsilon x^2 \right) = \gamma \cos \omega t$$

\Rightarrow *Spring with position dependent constant,
intra-wave frequency modulation;
therefore, we need instantaneous frequency.*

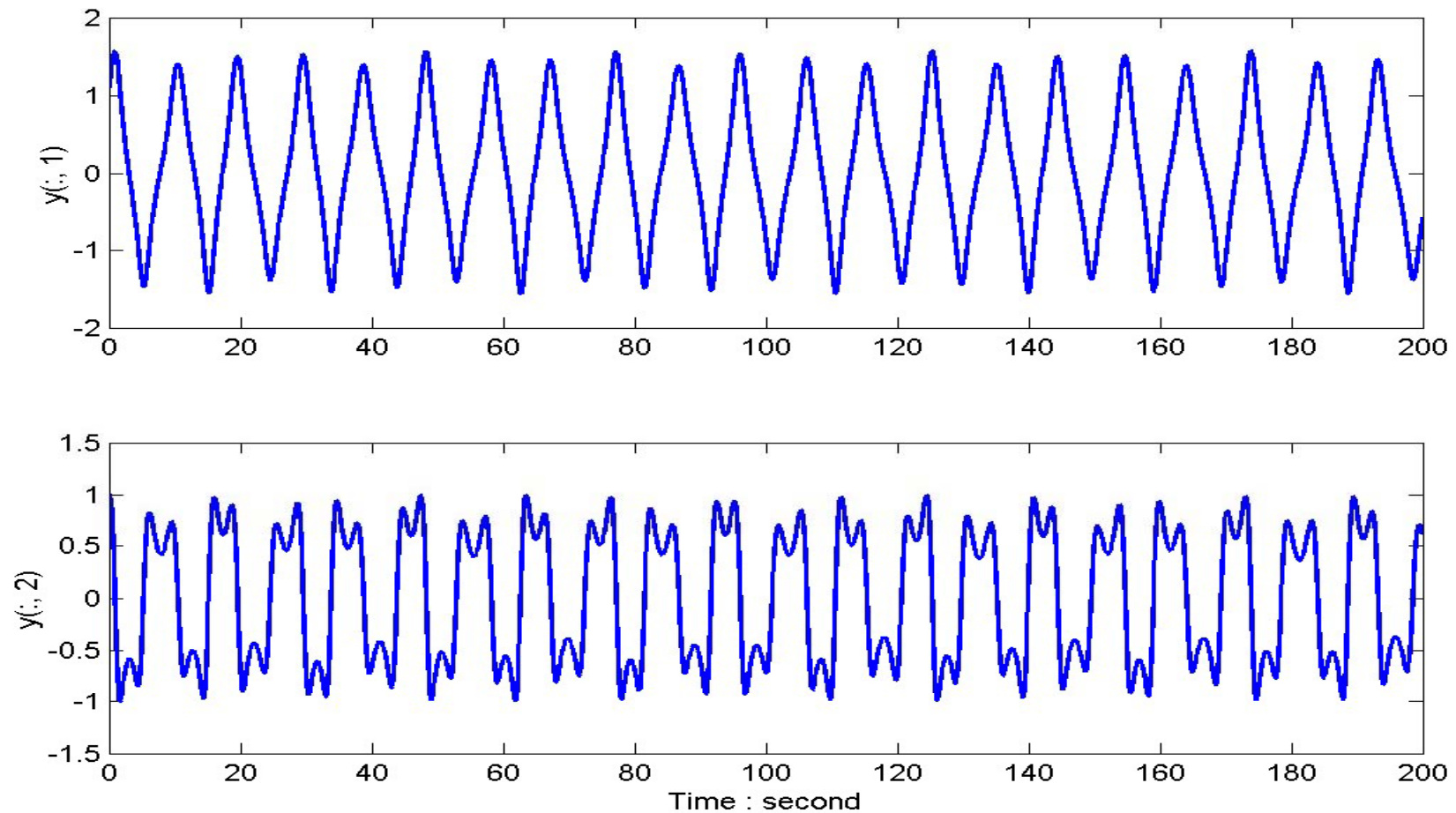
Duffing Pendulum: Intra-Wave Frequency Modulation



$$\frac{d^2 x}{dt^2} + x (1 + \epsilon x^2) = \gamma \cos \omega t .$$

Duffing Equation : Data

Duffing Equation : ODE23TB



Duffing Type Wave

Perturbation Expansion

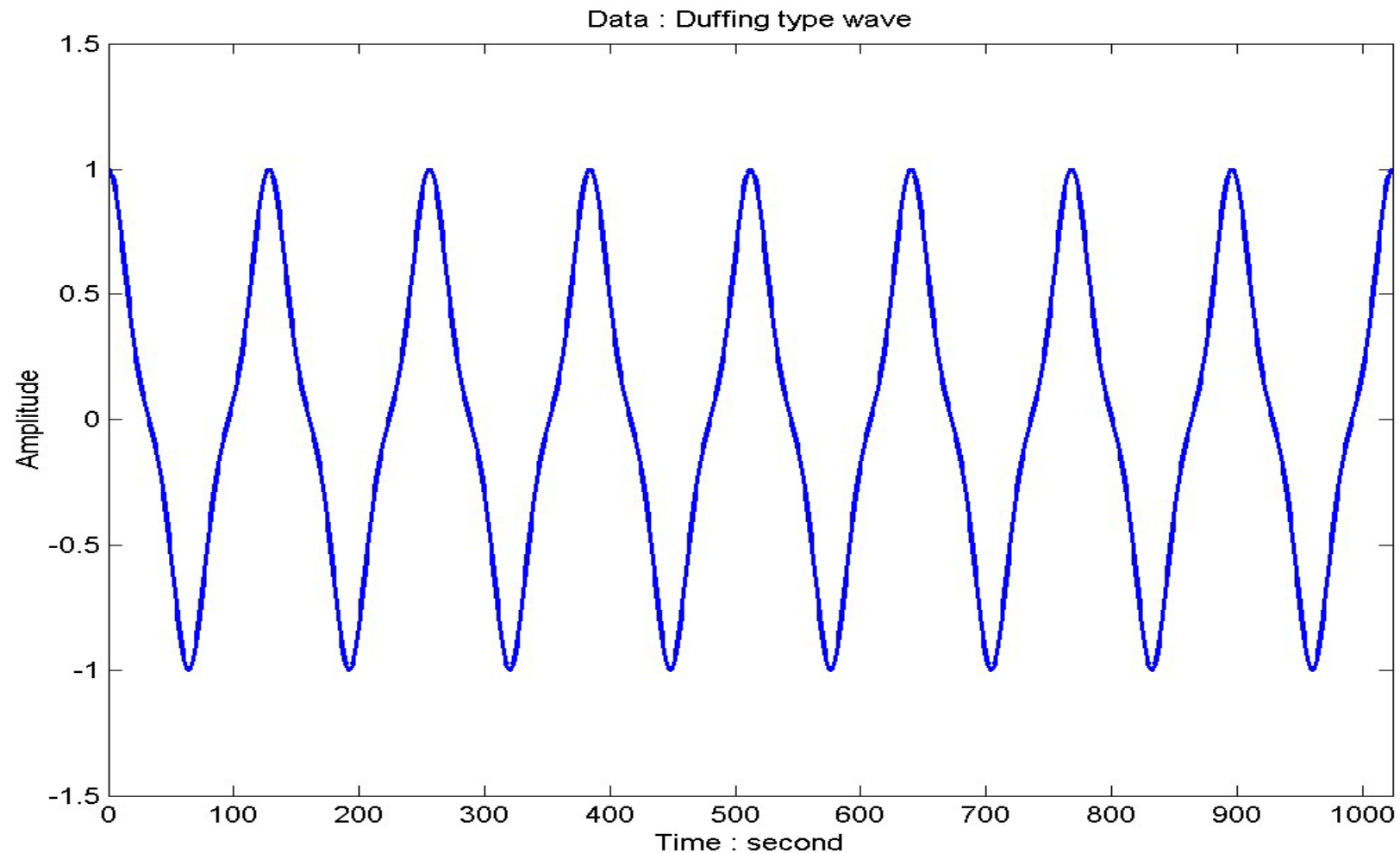
For $\varepsilon \ll 1$, we can have

$$\begin{aligned}x(t) &= \cos(\omega t + \varepsilon \sin 2\omega t) \\&= \cos \omega t \cos(\varepsilon \sin 2\omega t) - \sin \omega t \sin(\varepsilon \sin 2\omega t) \\&= \cos \omega t - \varepsilon \sin \omega t \sin 2\omega t + \dots \\&= \left(1 - \frac{\varepsilon}{2}\right) \cos \omega t + \frac{\varepsilon}{2} \cos 3\omega t + \dots\end{aligned}$$

This is very similar to the solution of Duffing equation .

Duffing Type Wave

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



Hilbert Transform : Definition

For any $x(t) \in L^p$,

$$y(t) = \frac{1}{\pi} \wp \int_{\tau} \frac{x(\tau)}{t - \tau} d\tau ,$$

then, $x(t)$ and $y(t)$ form the analytic pairs:

$$z(t) = x(t) + i y(t) = \textcolor{red}{a(t)} e^{i\theta(t)} ,$$

where

$$a(t) = \left(x^2 + y^2 \right)^{1/2} \text{ and } \theta(t) = \tan^{-1} \frac{y(t)}{x(t)} .$$

Instantaneous Frequency

$$\textit{Velocity} = \frac{\textit{distance}}{\textit{time}} ; \textit{mean velocity}$$

$$\textit{Newton} \Rightarrow v = \frac{dx}{dt}$$

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

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

So that both v and ω can appear in differential equations.

The combination of Hilbert Spectral Analysis and Empirical Mode Decomposition is designated as

HHT

(HHT vs. FFT)

Comparison between FFT and HHT

1. FFT :

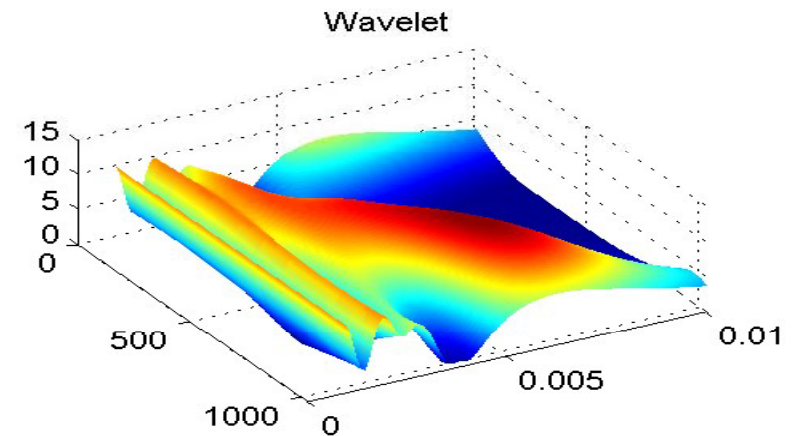
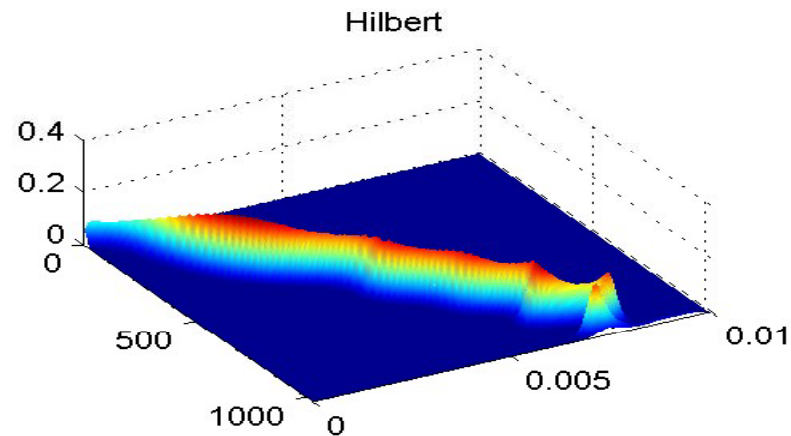
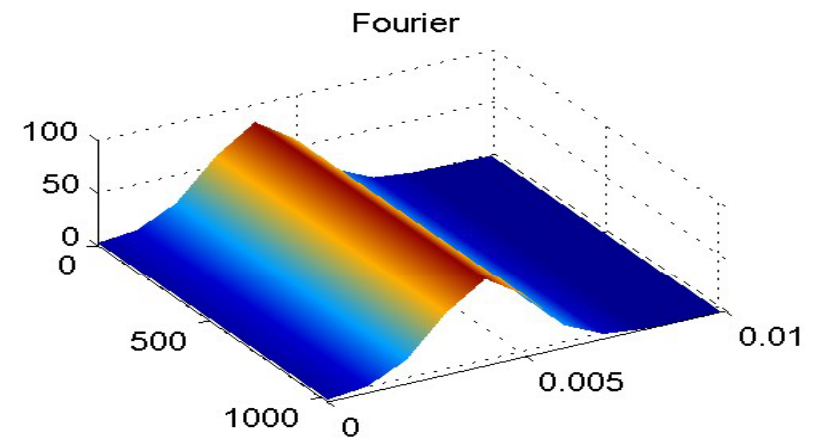
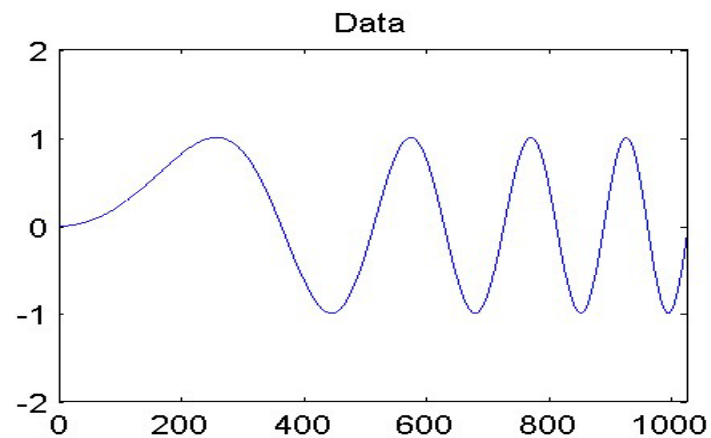
$$x(t) = \Re \sum_j a_j e^{i\omega_j t} .$$

2. HHT :

$$x(t) = \Re \sum_j a_j(t) e^{i \int_t \omega_j(\tau) d\tau} .$$

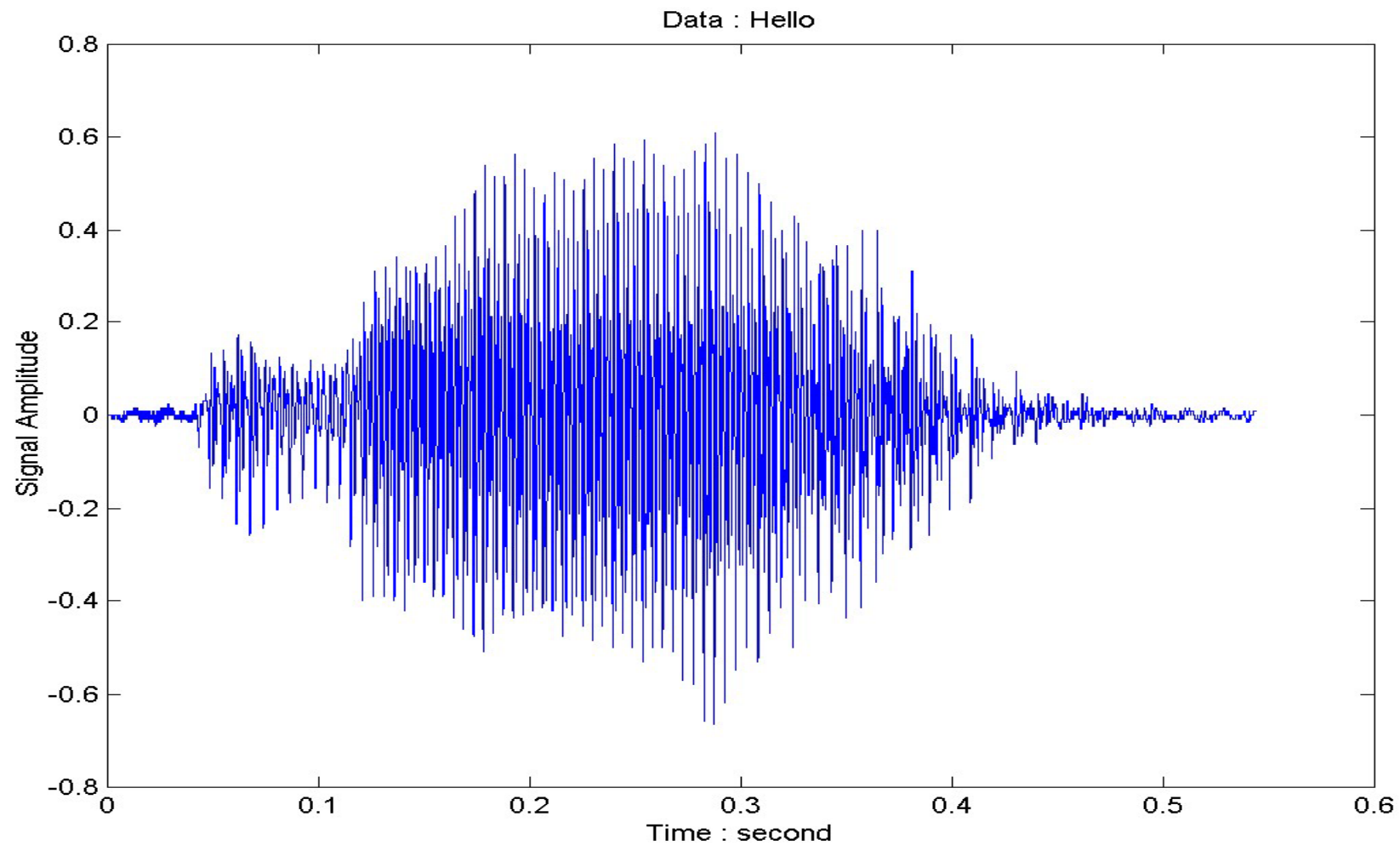
Comparisons: Fourier, Hilbert & Wavelet

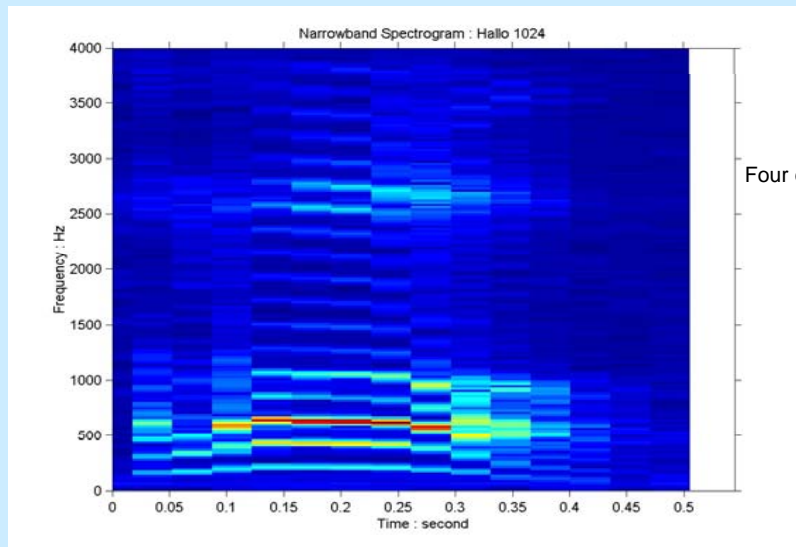
Comparison among Fourier, Hilbert, and Morlet Wavelet Spectra



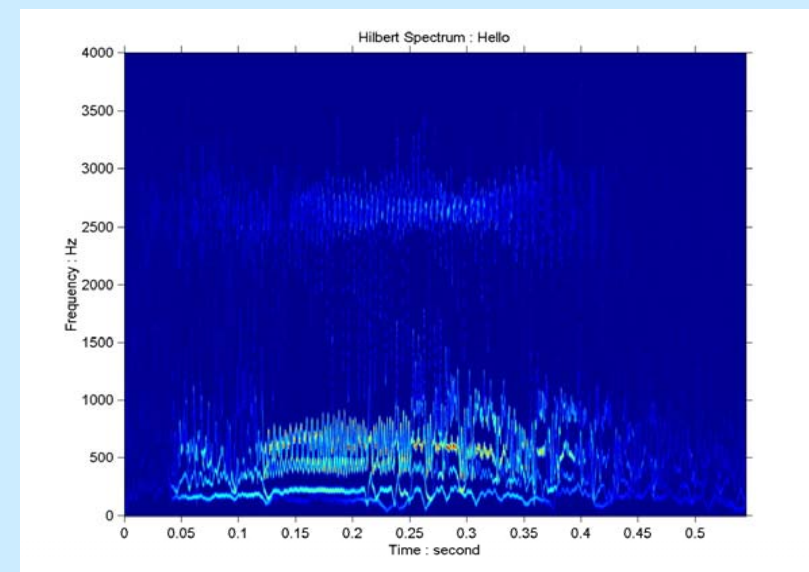
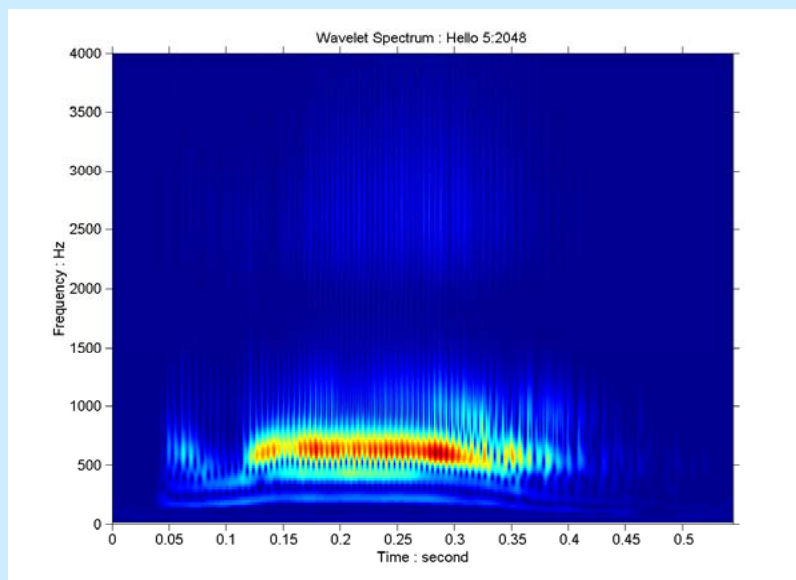
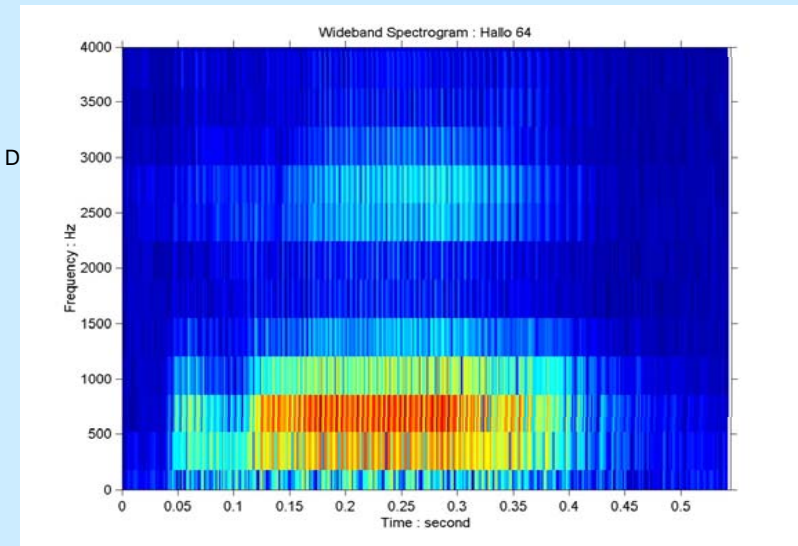
Speech Analysis

Hello : Data





Four comparisons D



Hilbert's View on Nonlinear Data

Duffing Type Wave

Perturbation Expansion

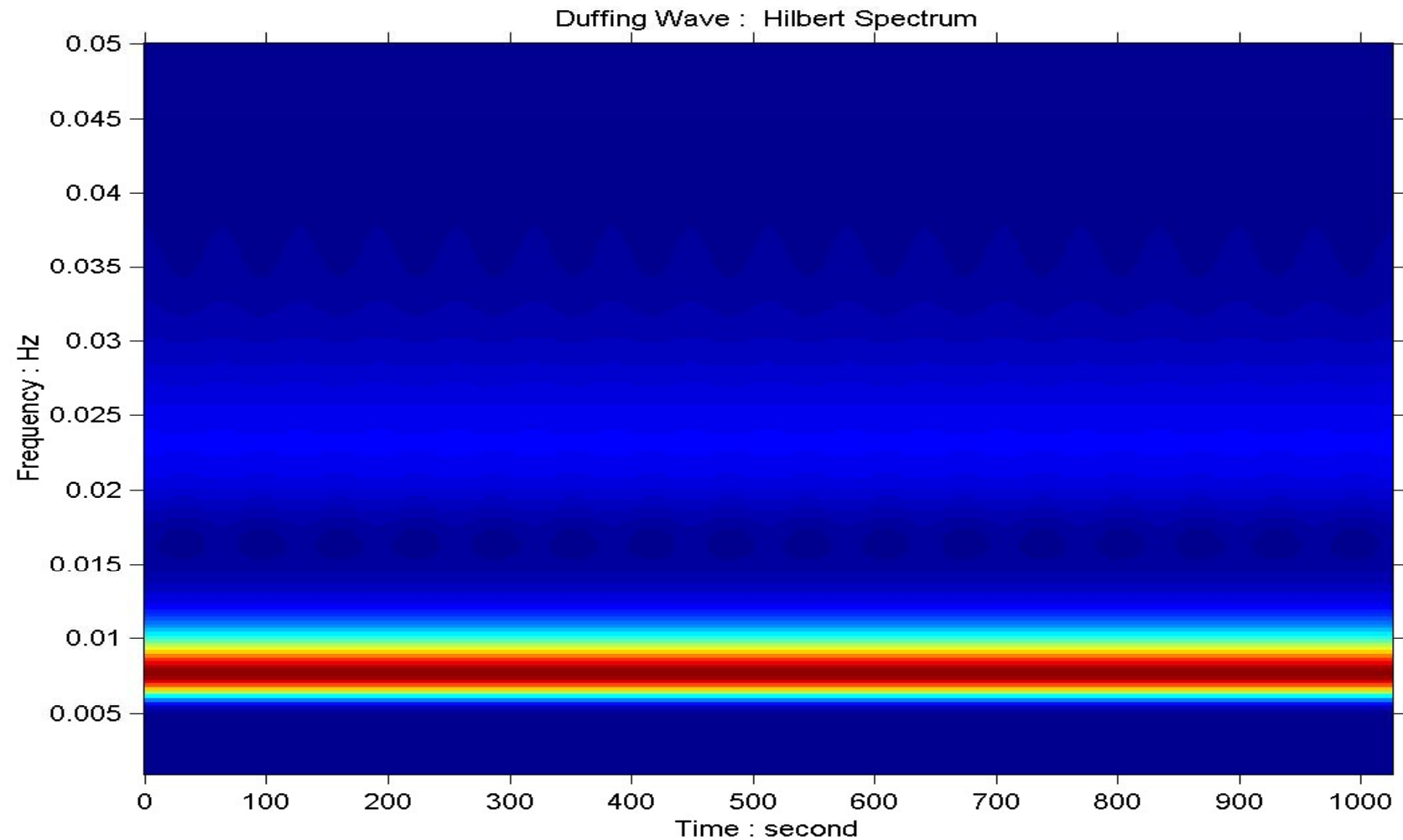
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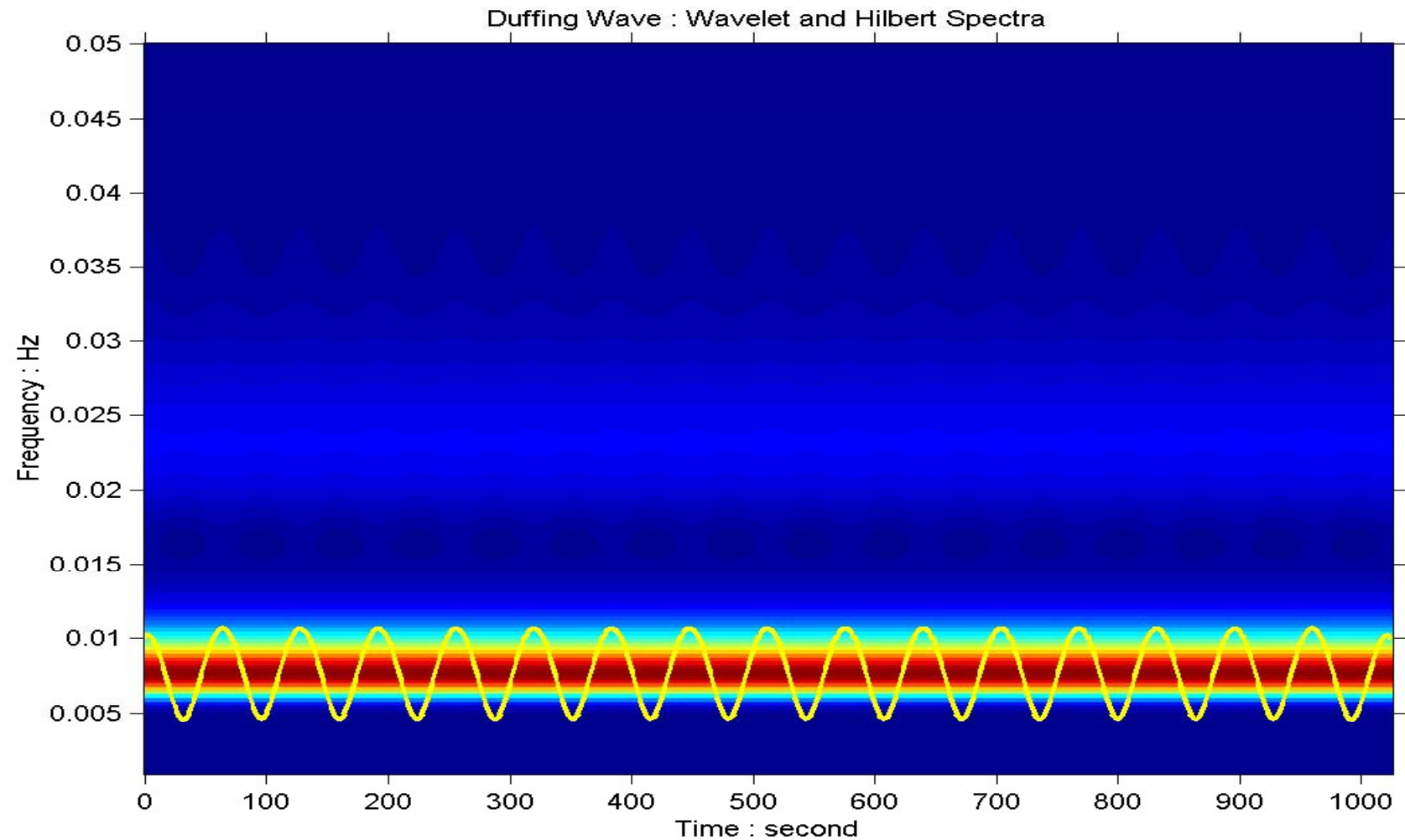
Duffing Type Wave

Wavelet Spectrum



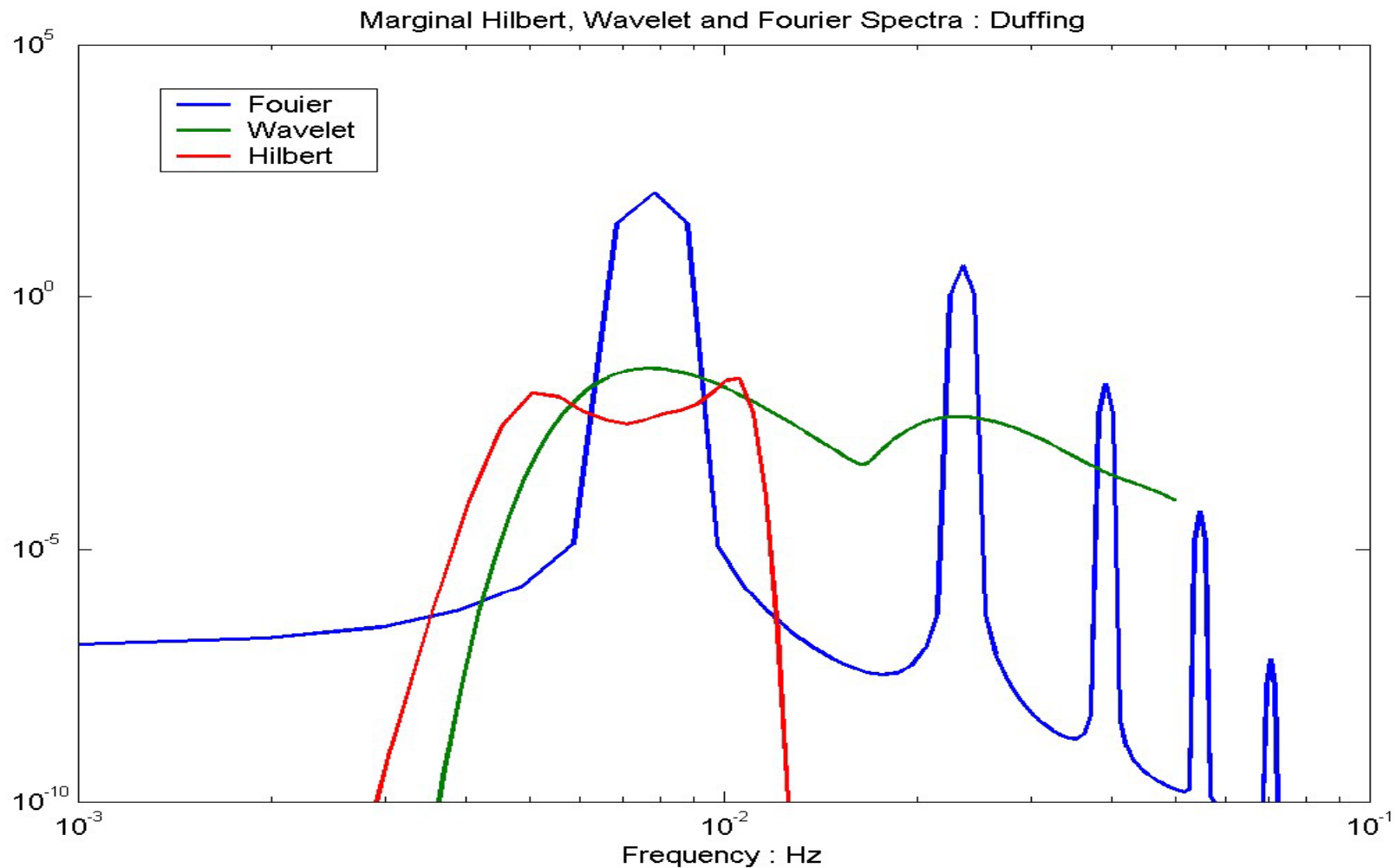
Duffing Type Wave

Hilbert Spectrum



Duffing Type Wave: Marginal Spectrum

Different harmonics from different Method: not physical



Degree of Nonlinearity

Quantify, rather than qualify,
nonlinearity

Degree of nonlinearity

Consider $x(t) = \cos(\omega t + \varepsilon \sin \eta \omega t)$

$$\Rightarrow IF = \frac{d\theta}{dt} = \omega(1 + \eta \varepsilon \cos \eta \omega t)$$

$$DN \text{ (Degree of Nonlinearity)} = \left\langle \left(\frac{IF - IF_z}{IF_z} \right)^2 \right\rangle^{1/2} = \frac{\eta \varepsilon}{\sqrt{2}}$$

Degree of Nonlinearity

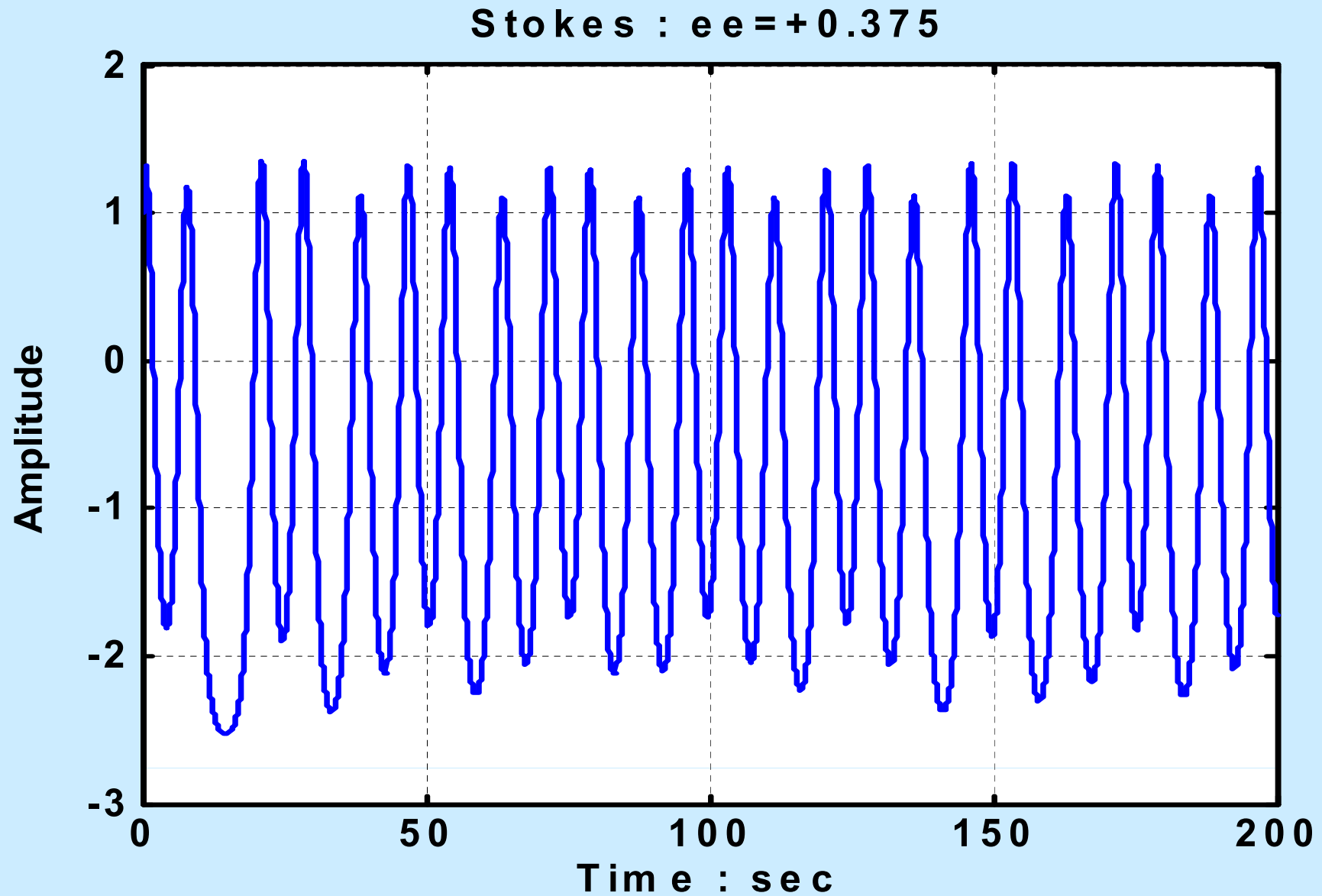
- We can determine **DN** precisely with Hilbert Spectral Analysis.
- We can also determine **ε** and **η** separately.
- **η** can be determined from the frequency modulation.
- **ε** can be determined from DN with **η** given.

Stokes Waves

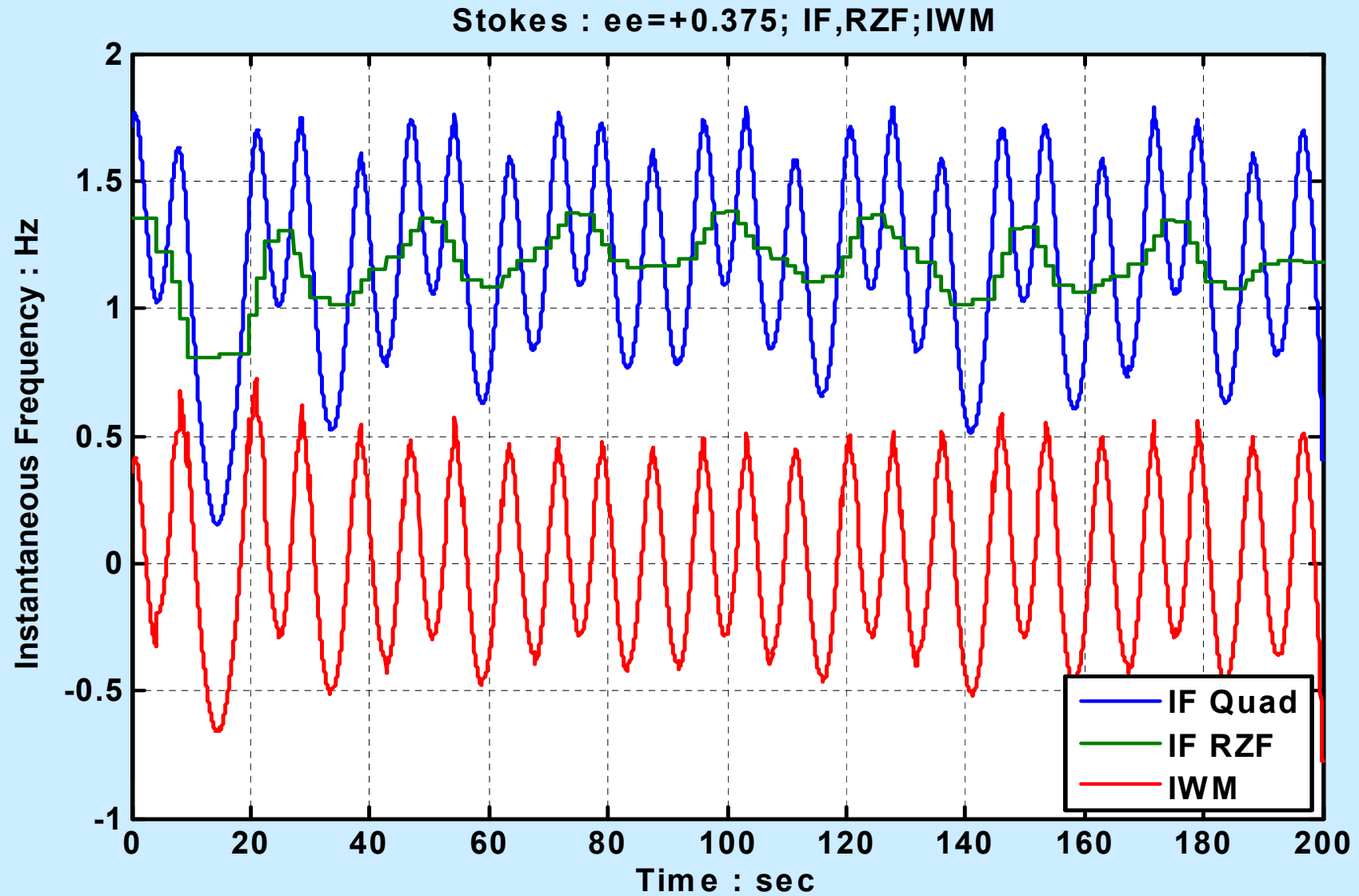
$$\frac{d^2 x}{dt^2} + x + \varepsilon x^2 = \gamma \cos \omega t$$

$$\frac{d^2 x}{dt^2} + x(1 + \varepsilon x) = \gamma \cos \omega t$$

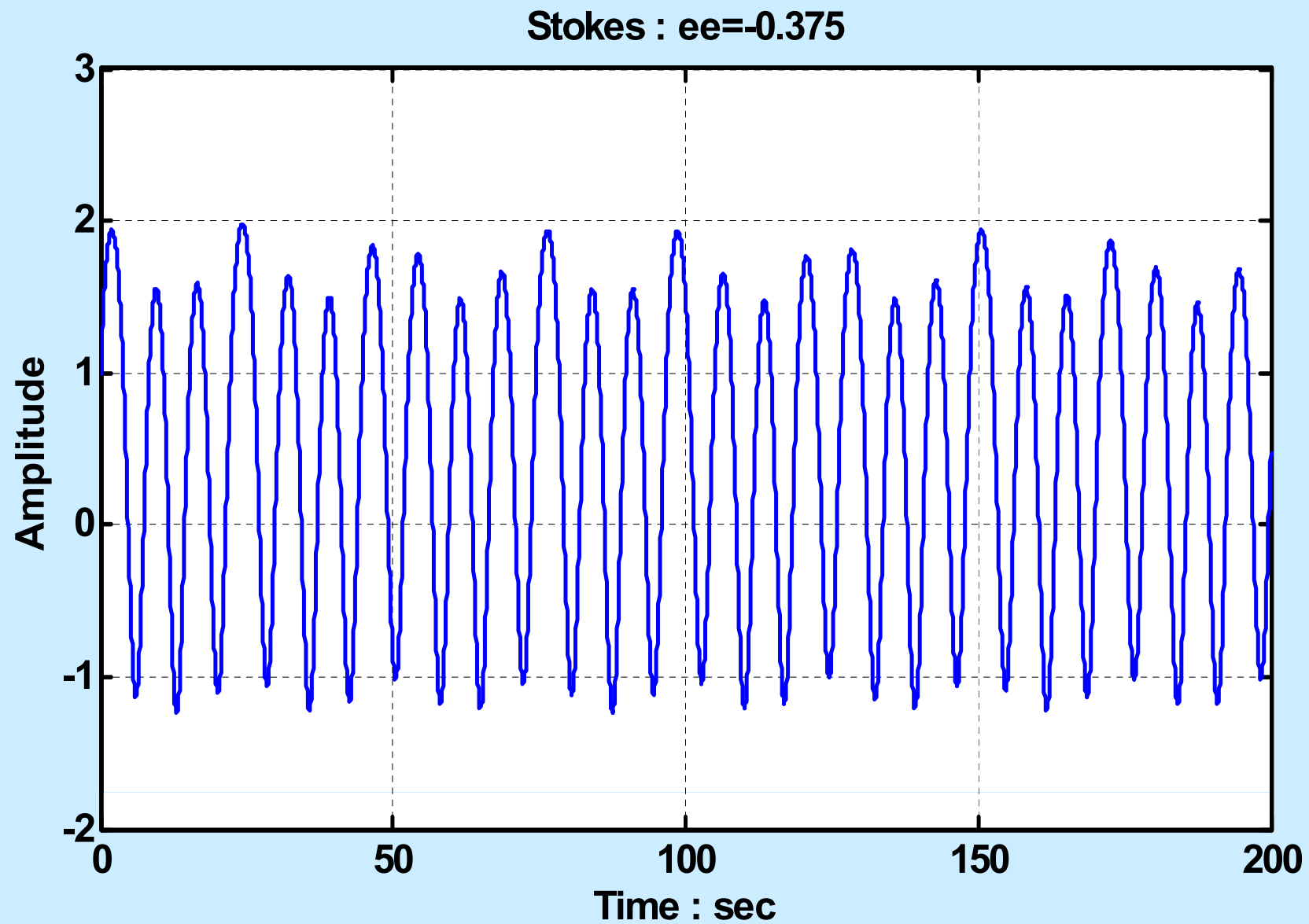
Stokes ee= +0.375; DN=0.0407



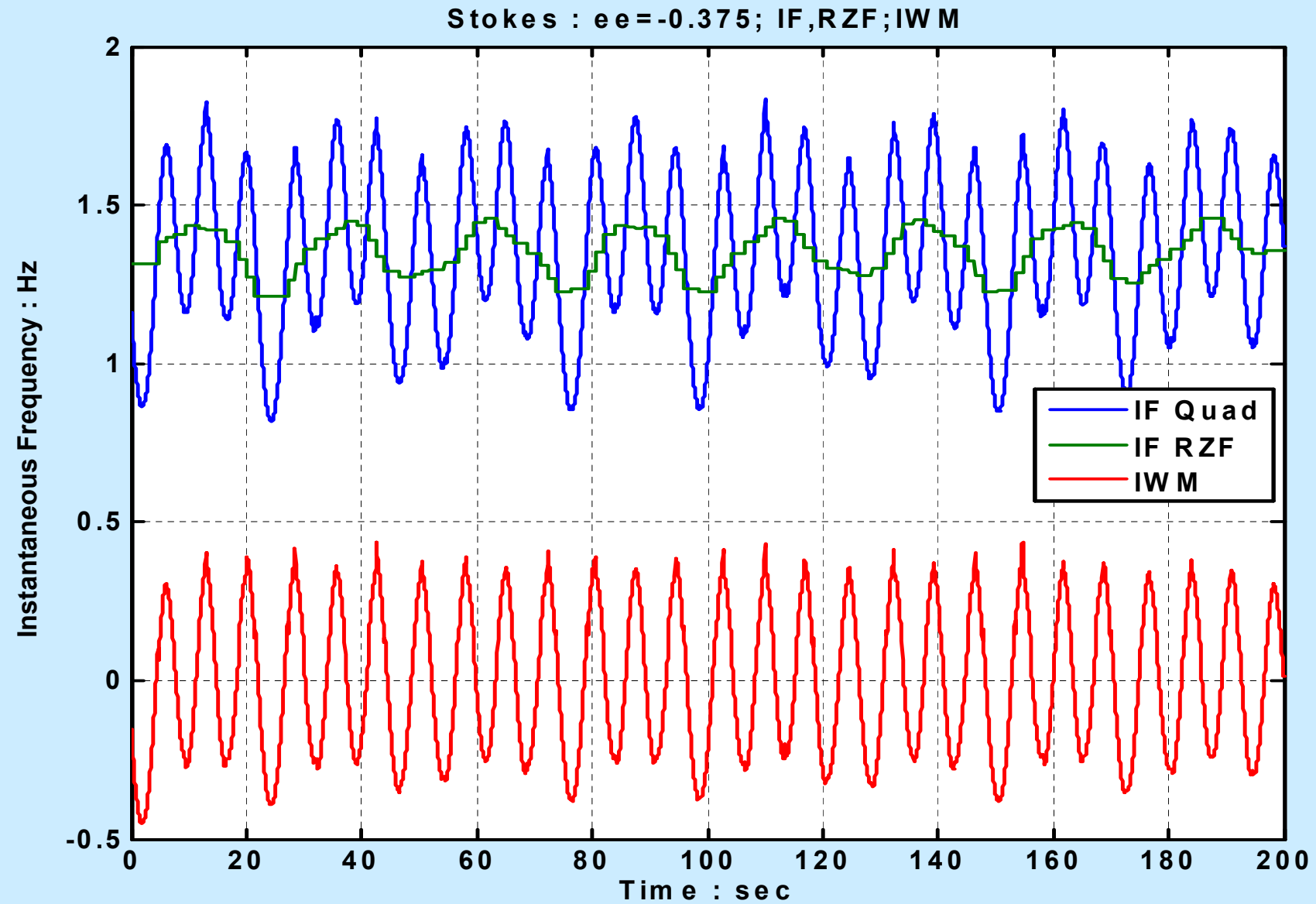
Stokes ee= +0.375; DN=0.0893



Stokes ee=-0.375; DN=0.0407



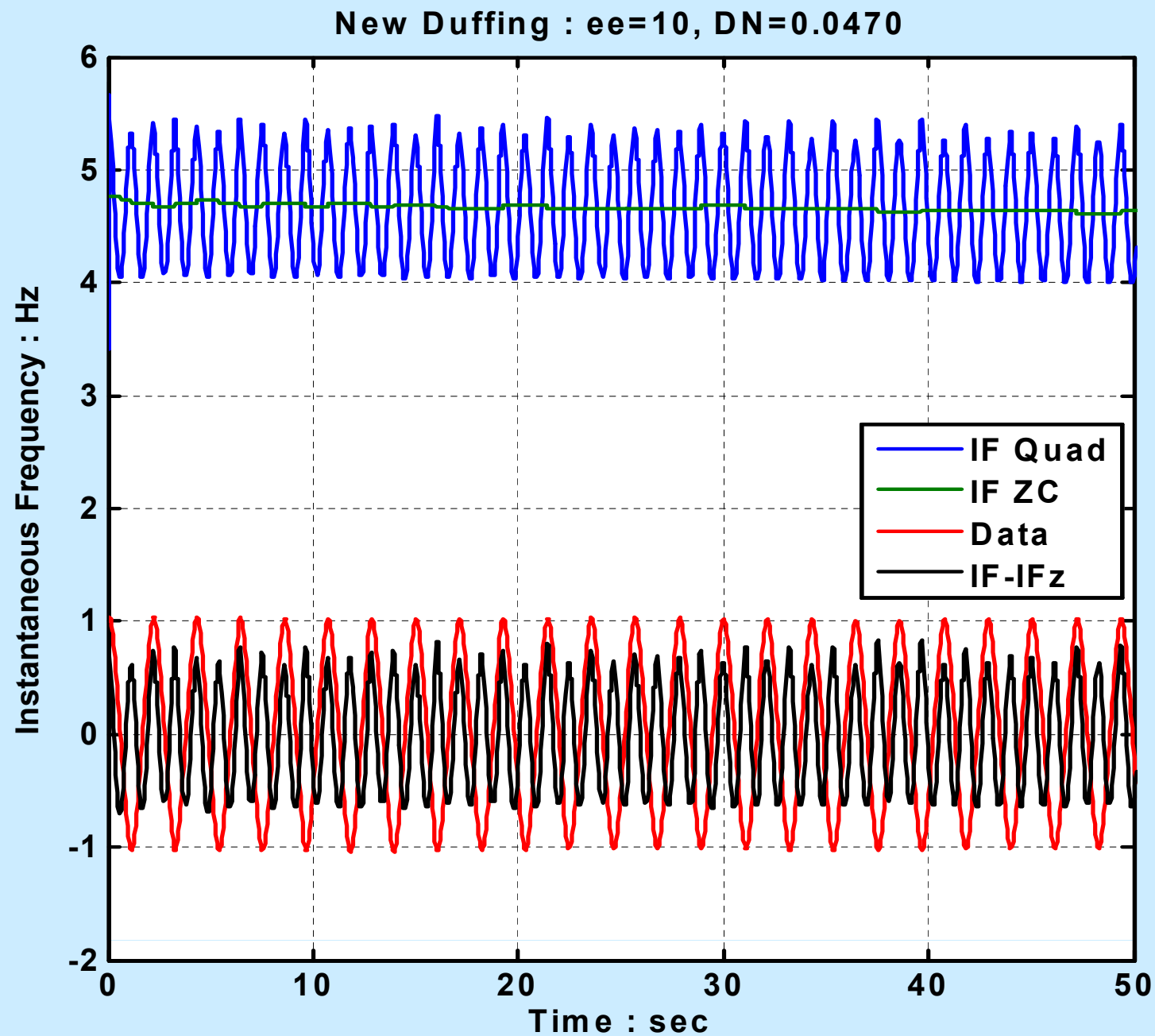
Stokes ee=-0.375; DN=0.0407



Duffing Waves

$$\frac{d^2 x}{dt^2} + x + \varepsilon x^3 = \gamma \cos \omega t$$

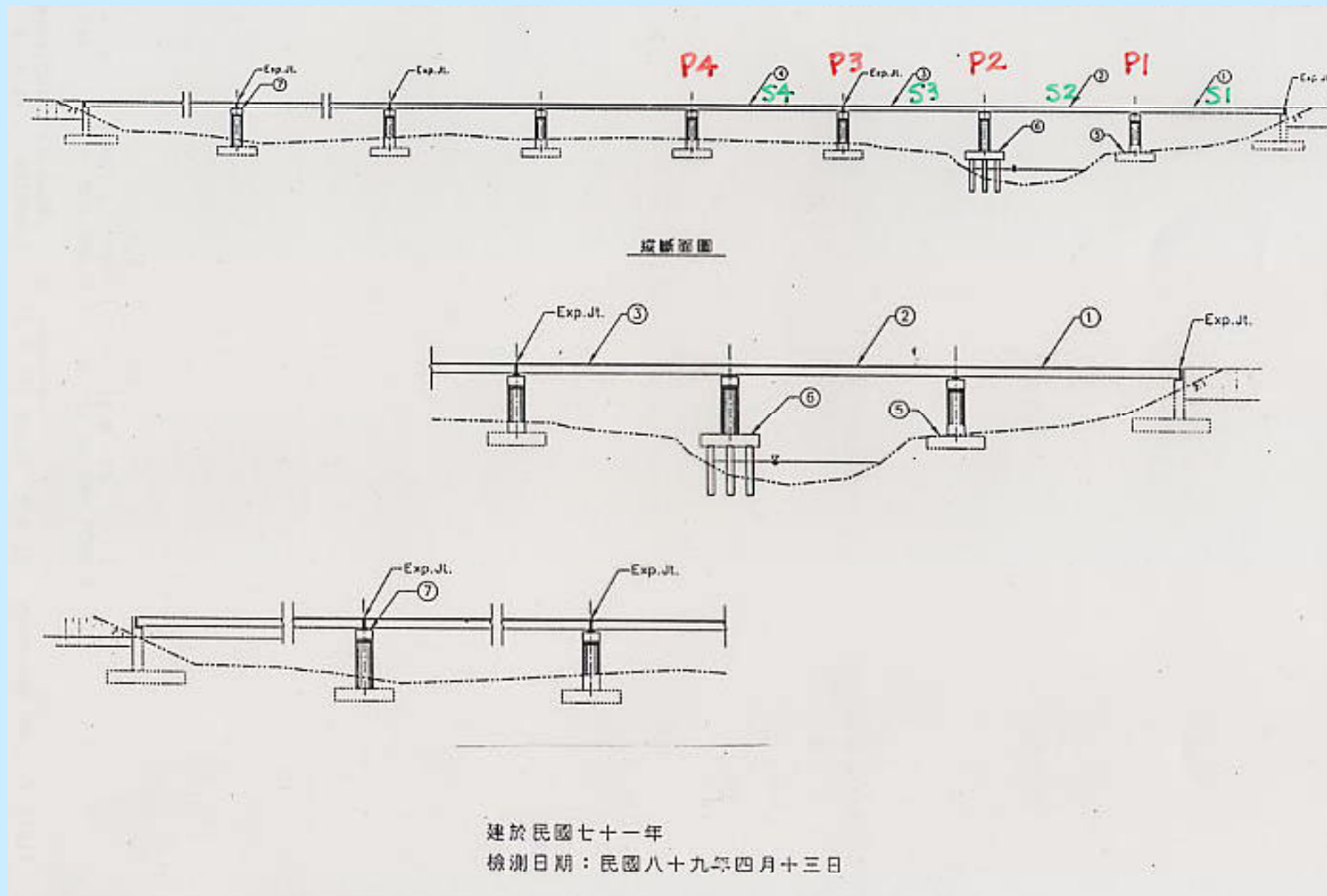
$$\frac{d^2 x}{dt^2} + x \left(1 + \varepsilon x^2 \right) = \gamma \cos \omega t$$



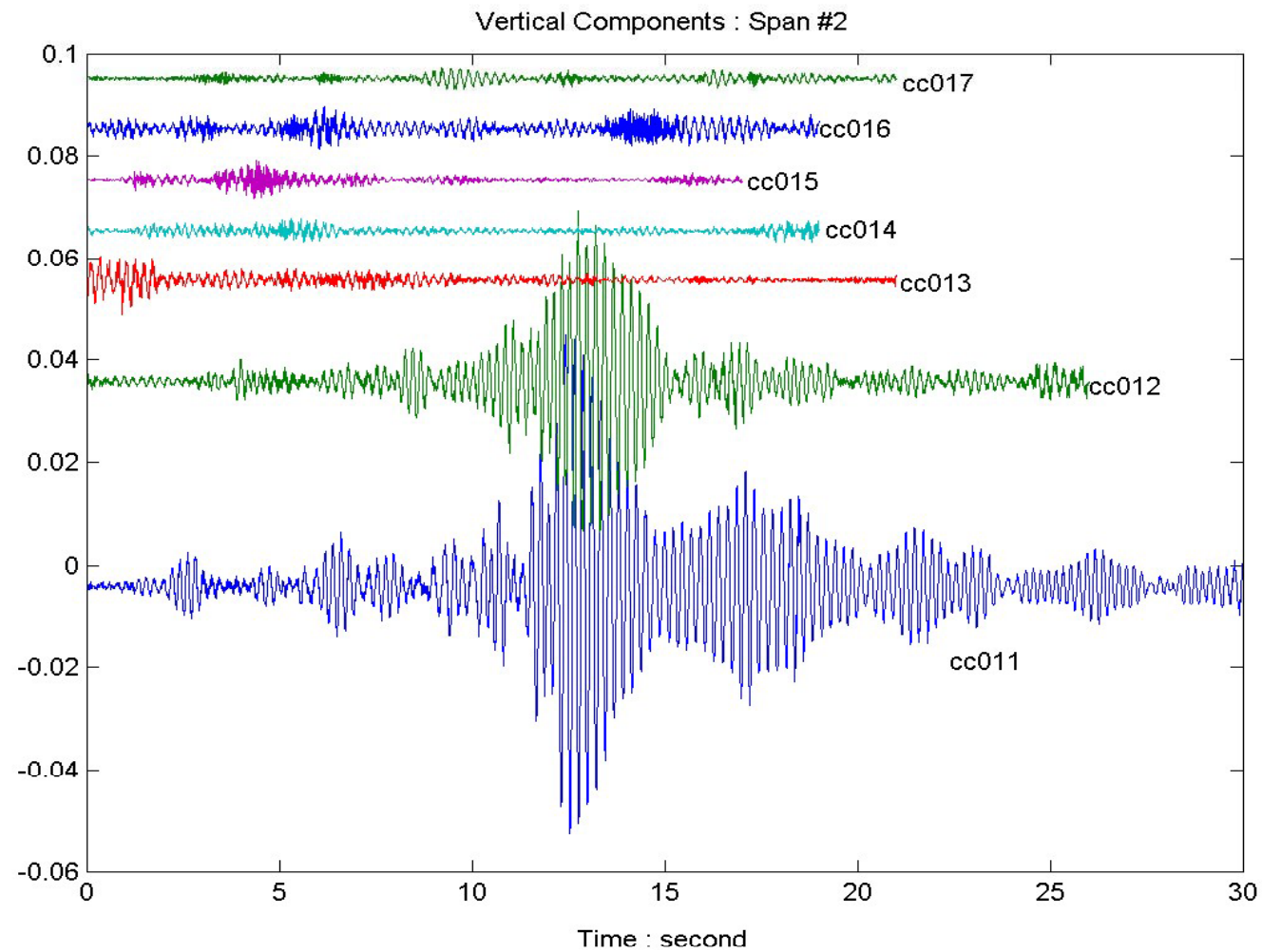
Case Study

A Bridge in Central Taiwan

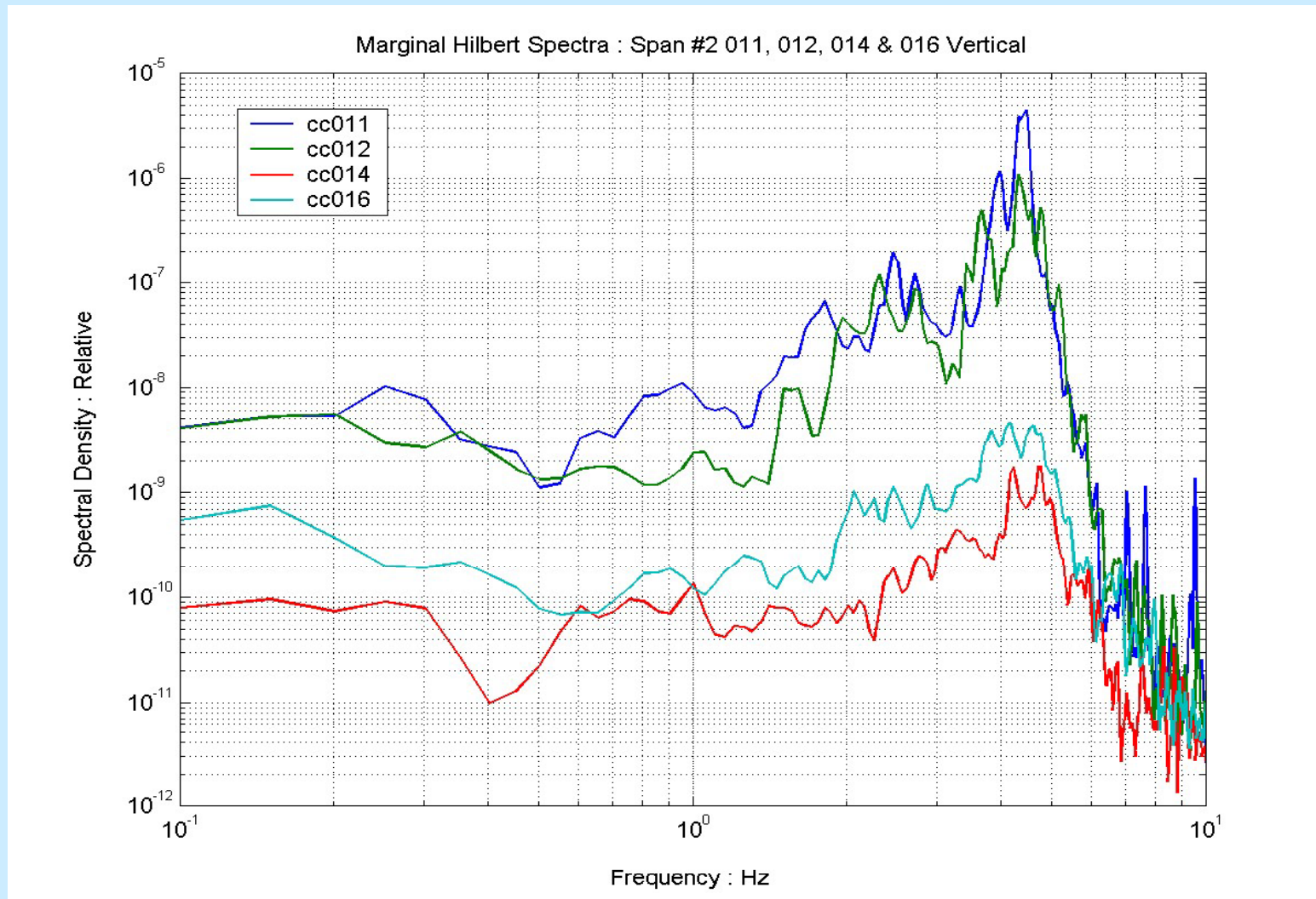
A Bridge in Southern Taiwan



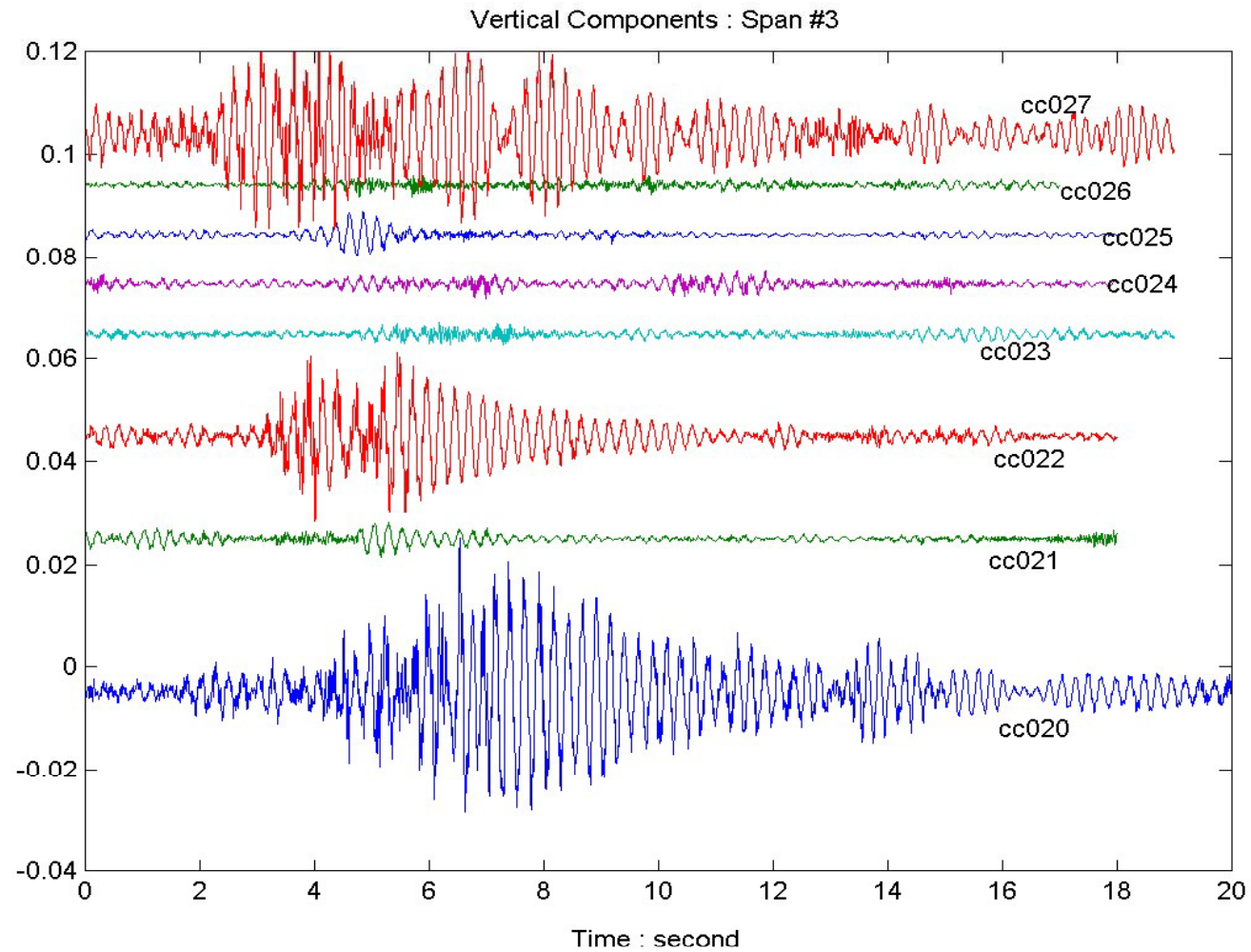
Data : Hsin Nan Bridge Span 2 Vertical



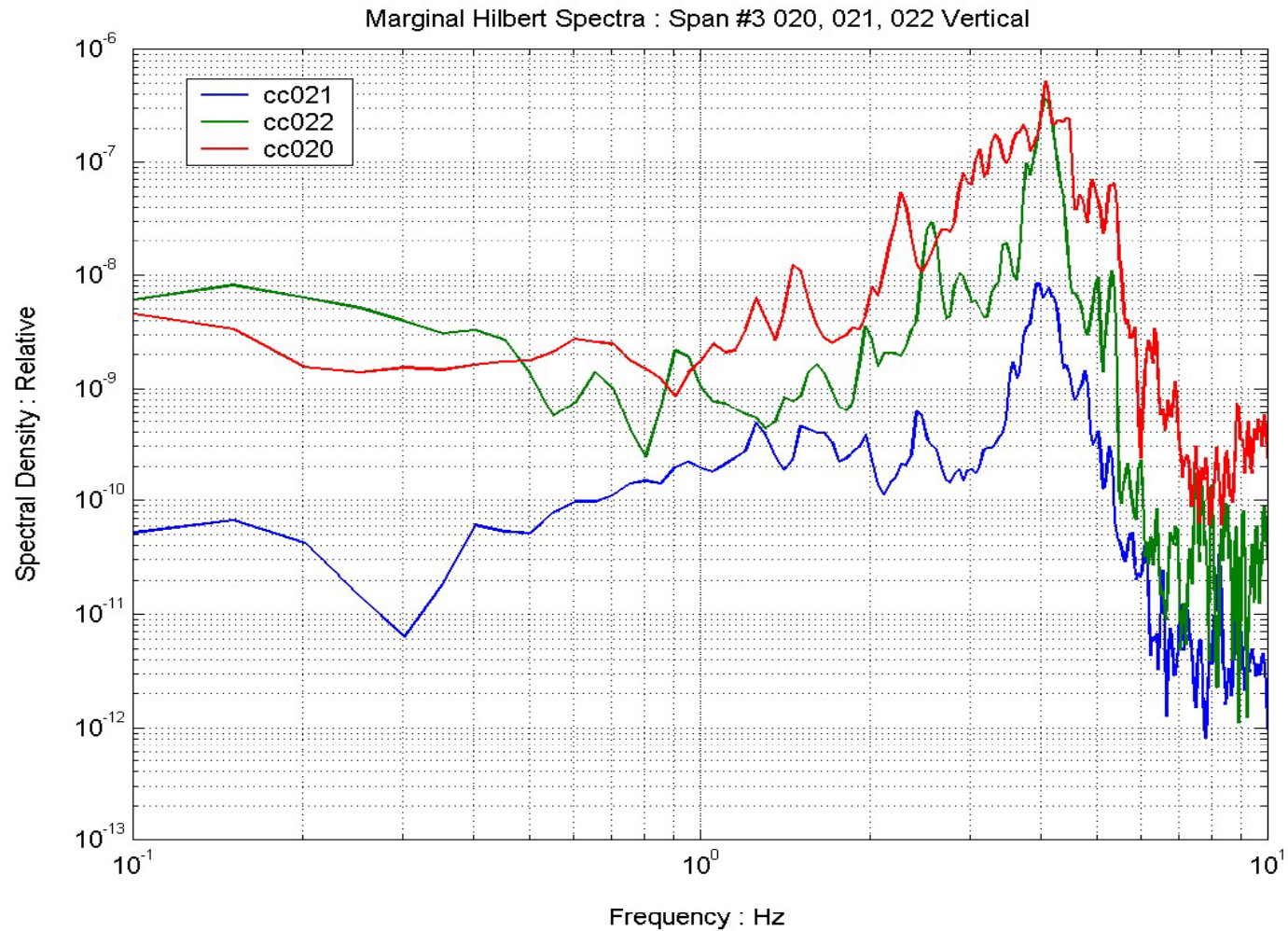
Marginal Hilbert Spectra : Hsin Nan Bridge Span 2 Vertical



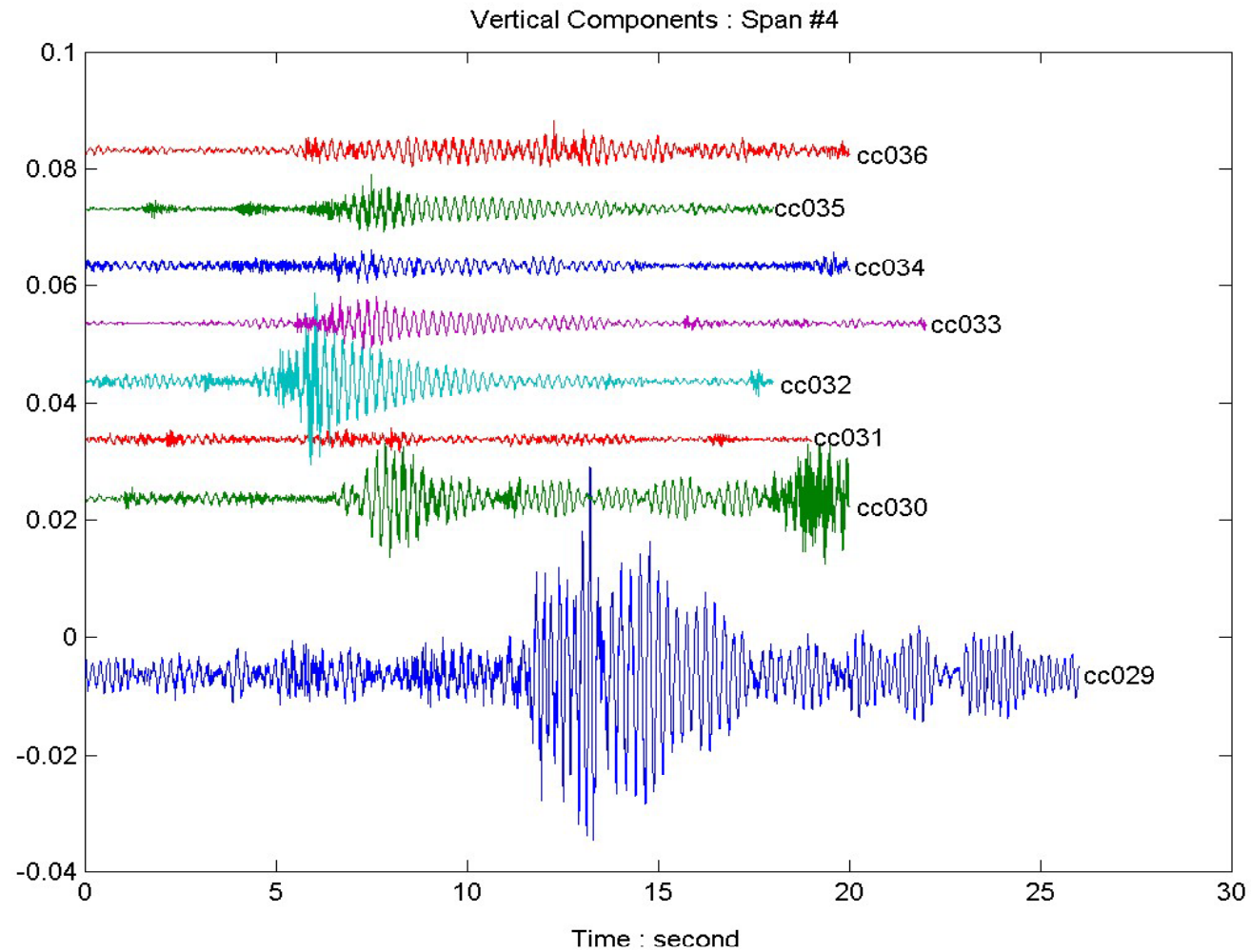
Data : Hsin Nan Bridge Span 3 Vertical



Marginal Hilbert Spectra : Hsin Nan Bridge Span 3 Vertical

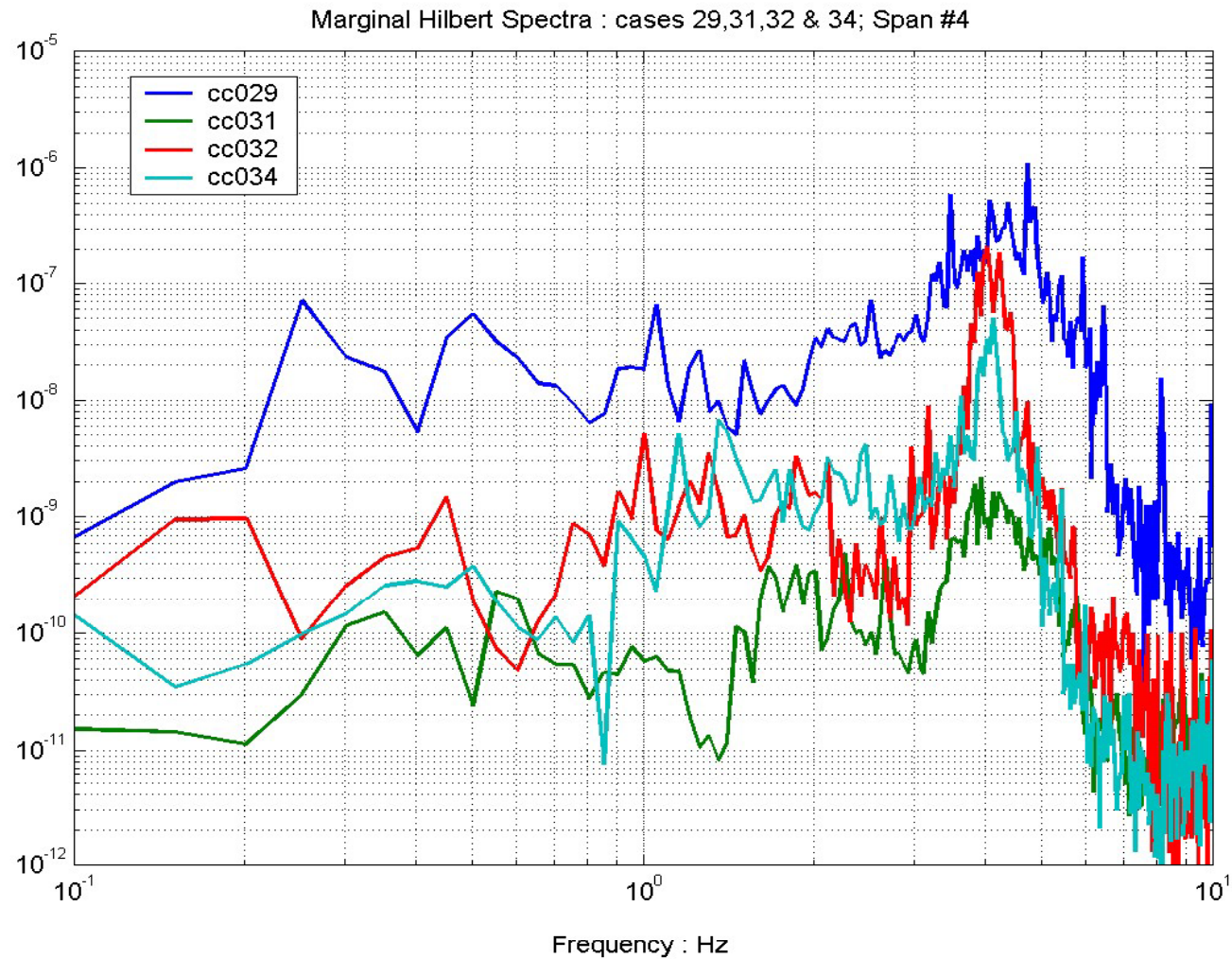


Data : Hsin Nan Bridge Span 4 Vertical



Marginal Hilbert Spectra : Hsin Nan Bridge Span 4

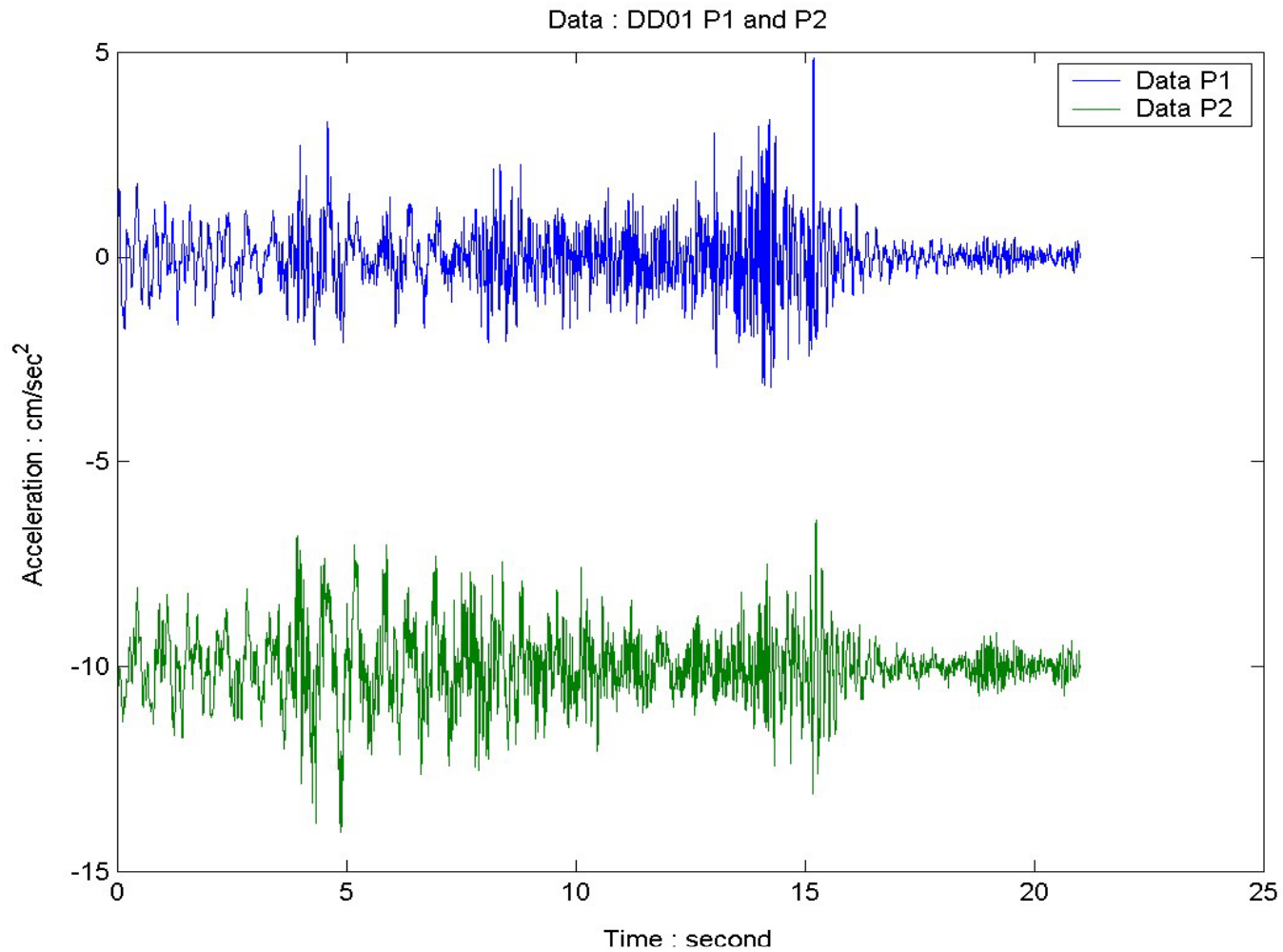
Vertical



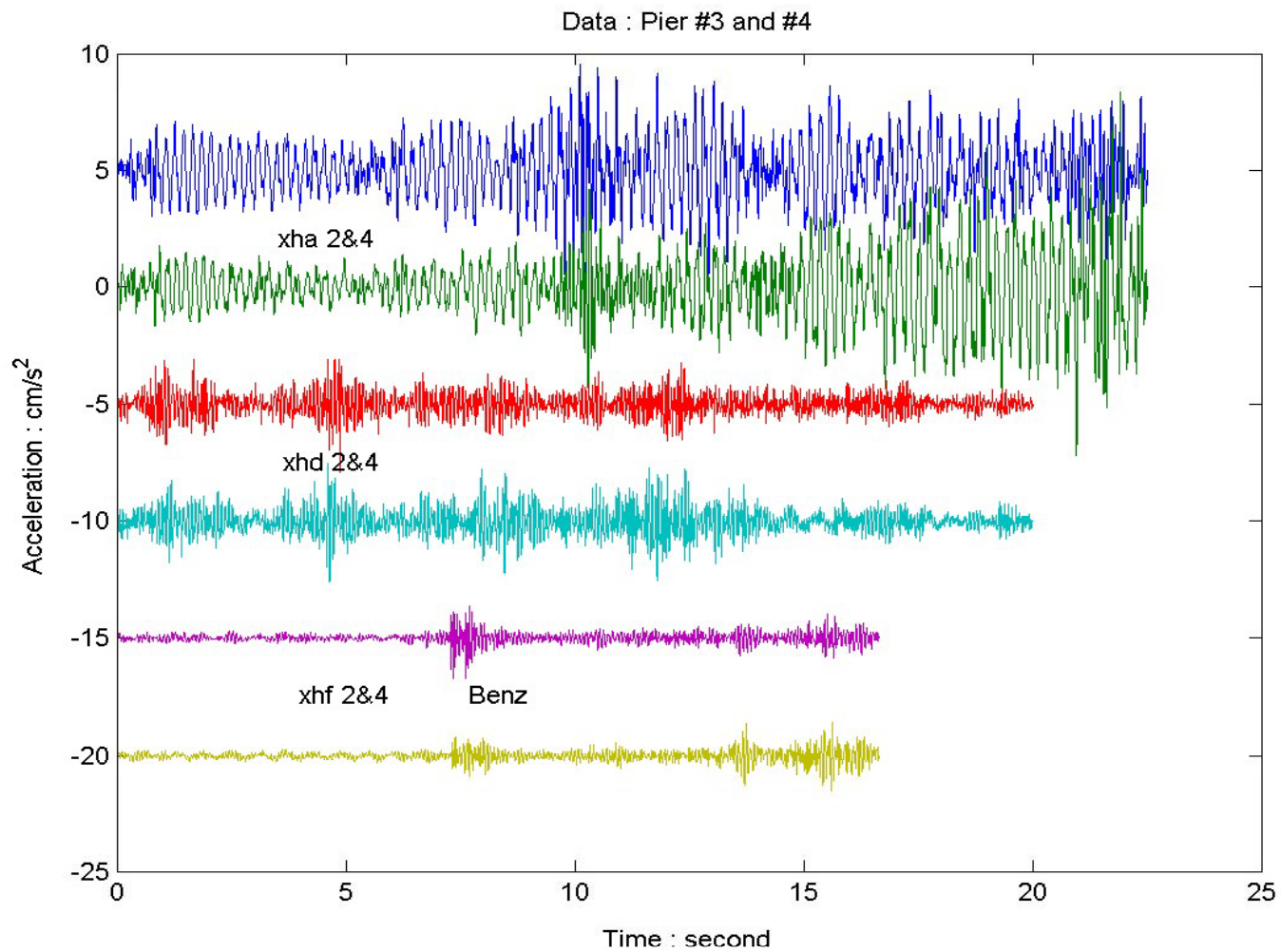
The foundations: the Piers

Check the piles

Data : Bridge Piers 1 & 2

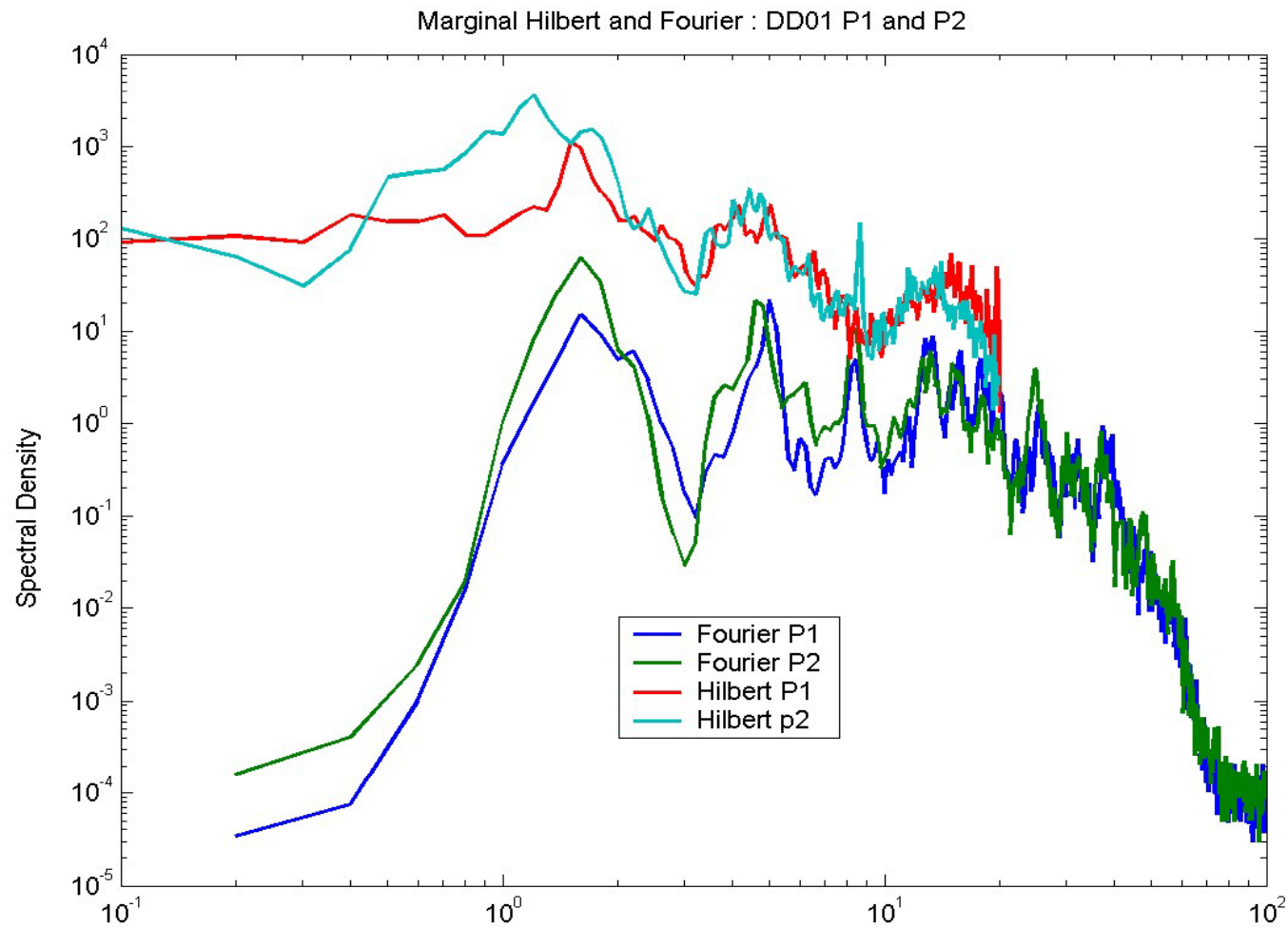


Data : Bridge Piers 3 & 4



Marginal Hilbert & Fourier Spectra :

Bridge Piers 1 & 2 Vertical



What This Means

- **Instantaneous Frequency** offers a total different view for nonlinear data: instantaneous frequency with no need for harmonics and unlimited by uncertainty.
- **Adaptive basis** is indispensable for nonstationary and nonlinear data analysis
- **HHT establishes a new paradigm of data analysis with which we can quantify nonlinearity**

Comparisons

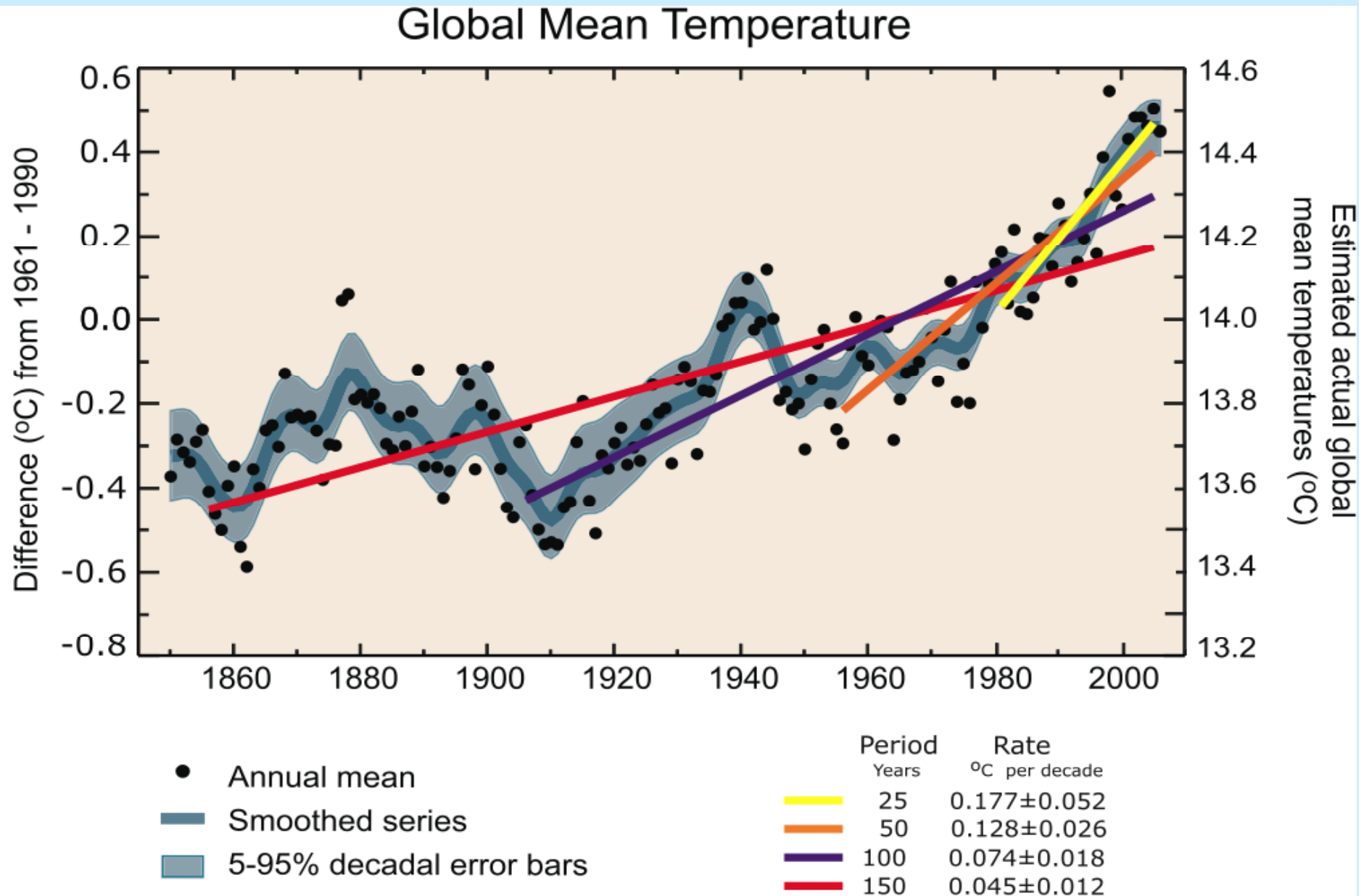
| | Fourier | Wavelet | Hilbert |
|----------------|-------------------------------|---------------------------------|---------------------------|
| Basis | a priori | a priori | Adaptive |
| Frequency | Integral transform: Global | Integral transform: Regional | Differentiation: Local |
| Presentation | Energy-frequency | Energy-time- frequency | Energy-time- frequency |
| Nonlinear | no | no | yes |
| Non-stationary | no | yes | yes |
| Uncertainty | yes | yes | no |
| Harmonics | yes | yes | no |

On Hazard Occurrence

Global Warming

HHT: Trend and prediction

IPCC Global Mean Temperature Trend



*“Note that for shorter recent periods,
the slope is greater, indicating
accelerated warming.”*

IPCC 4th Assessment Report 2007

The State-of-the arts: Trend

“One economist’s trend is another economist’s
cycle”

Engle, R. F. and Granger, C. W. J. 1991 *Long-run Economic Relationships*.
Cambridge University Press.

Regression method is arbitrary and ad hoc.

Philosophical Problem Anticipated

名不正則言不順

言不順則事不成

——孔夫子

On Definition

Without a proper definition,
logic discourse would be impossible.
Without logic discourse,
nothing can be accomplished.

Confucius

Definition of the Trend

Huang et al, *Proc. Roy. Soc. Lond.*, 1998

Wu et al. *PNAS* 2007

Within the given data span, the trend is an intrinsically fitted monotonic function, or a function in which there can be at most one extremum.

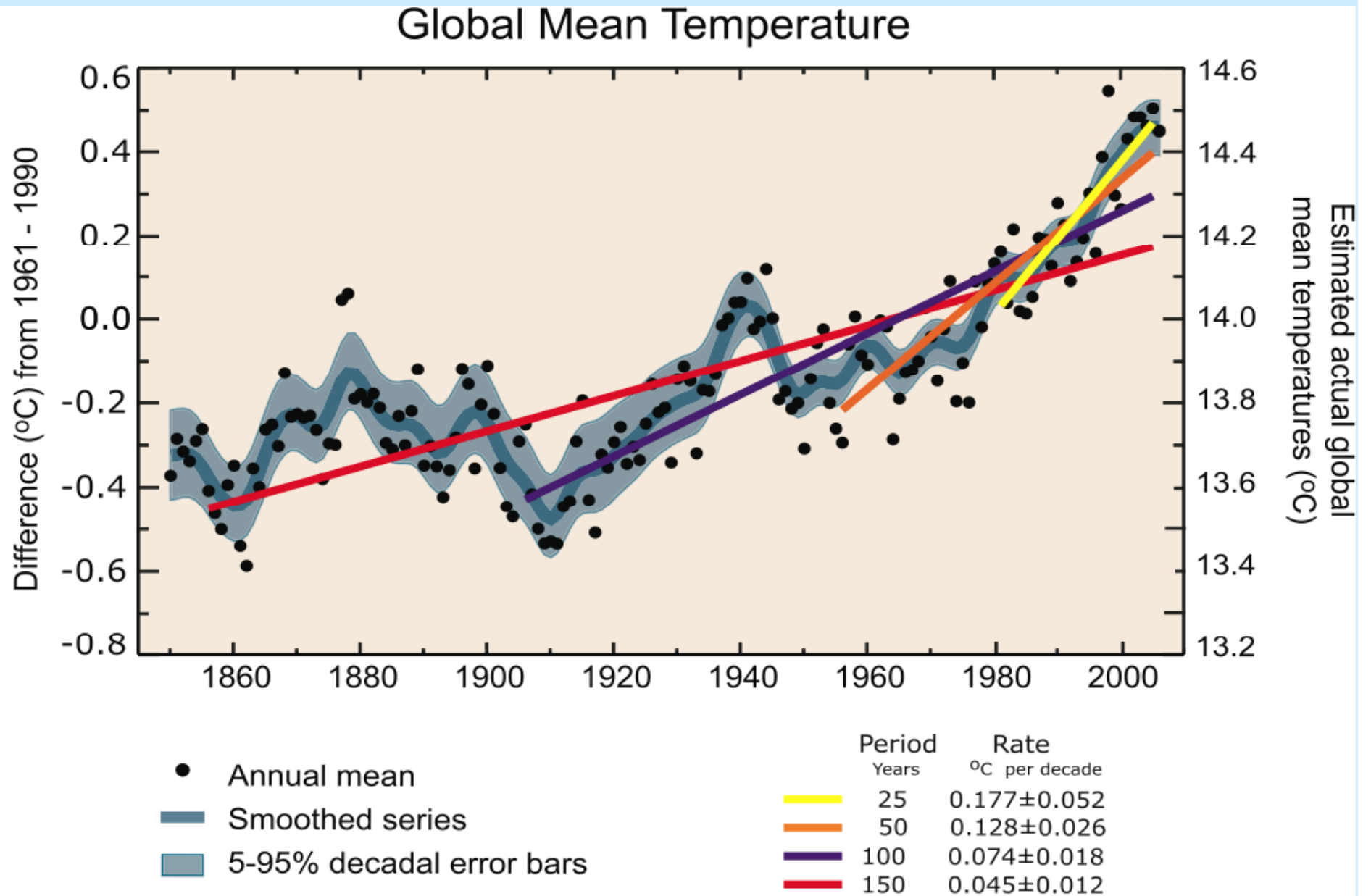
The trend should be an **intrinsic** and **local** property of the data; it is determined by the same mechanisms that generate the data.

Being local, it has to associate with a **local length scale**, and be valid only within that length span, and be part of a full wave length.

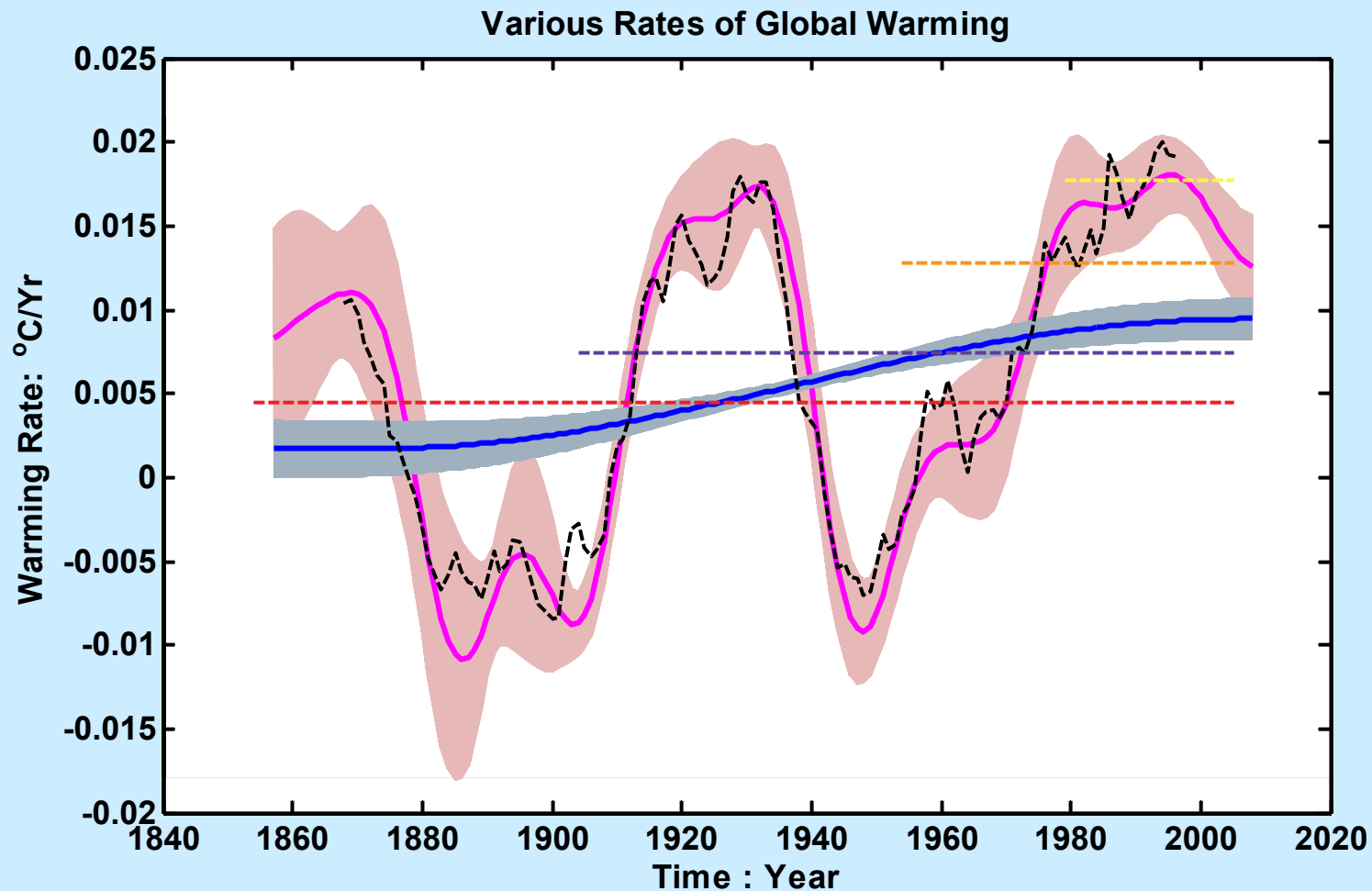
The method determining the trend should be **intrinsic**. Being intrinsic, the method for defining the trend has to be **adaptive**.

All traditional trend determination methods are **extrinsic**.

IPCC Global Mean Temperature Trend



Comparison between non-linear rate with multi-rate of IPCC

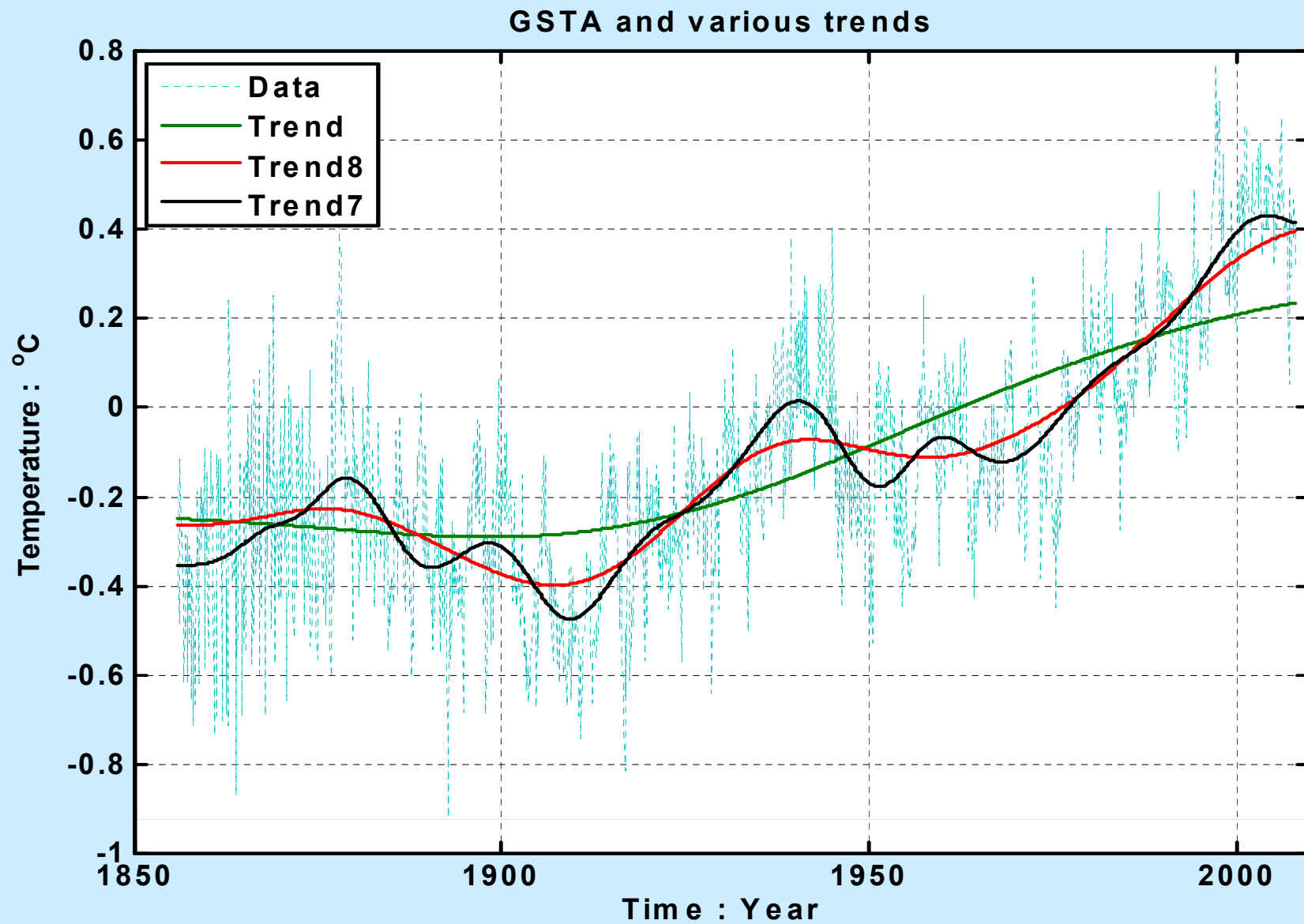


Blue shadow and blue line are the warming rate of non-linear trend.

Magenta shadow and magenta line are the rate of combination of non-linear trend and AMO-like components.

Dashed lines are IPCC rates.

GSAT Data and Various Trends



Annual Temperature Ranking : 2008

| GISS | NCDC | CRU | Rank |
|------|------|------|------|
| 2005 | 2005 | 1998 | 1 |
| 1998 | 1998 | 2005 | 2 |
| 2002 | 2002 | 2003 | 3 |
| 2007 | 2003 | 2002 | 4 |
| 2003 | 2006 | 2004 | 5 |
| 2006 | 2007 | 2006 | 6 |
| 2001 | 2004 | 2001 | 7 |
| 2004 | 2001 | 2007 | 8 |
| 2008 | 2008 | 1997 | 9 |
| 1997 | 1997 | 2008 | 10 |

Conclusion

Adaptive method is the only scientifically meaningful way to analyze data.

It is the only way to find out the underlying physical processes; therefore, it is indispensable in scientific research.

It is physical, direct, and simple; it is ideal for engineering applications.