

Wavelet p-leader non Gaussian multiscale expansions for Heart rate variability analysis in congestive Heart failure patients

H. WENDT¹, P. ABRY², K. KIYONO³, J. HAYANO⁴,
E. WATANABE⁵, Y. YAMAMOTO⁶

¹ CNRS, University of Toulouse, France

² CNRS, Ecole Normale Supérieure de Lyon, France

³ Osaka University, Graduate School of Engineering Science, Japan

⁴ Nagoya City University, Graduate School of Medical Sciences, Japan

⁵ Fujita Health University, School of Medicine, Toyoake, Japan

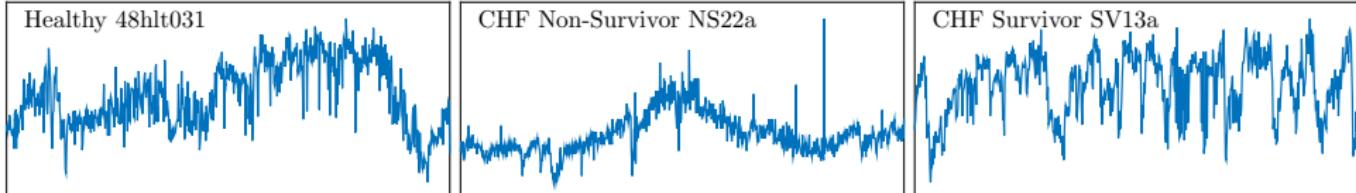
⁶ University of Tokyo, Graduate School of Education, Japan

Tokyo, 11/2018



Motivation: Heart rate variability (HRV)

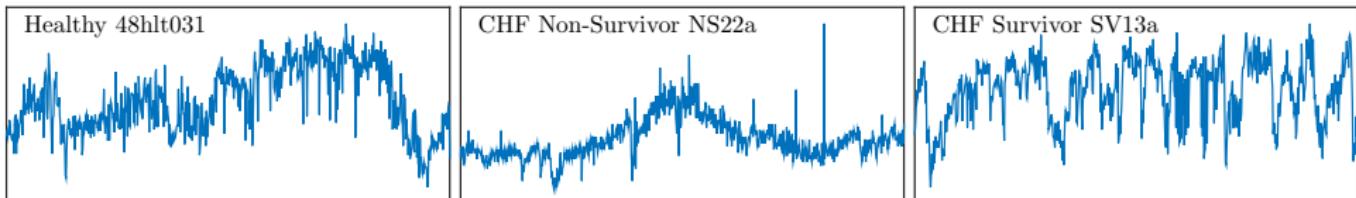
- ▶ Statistical characterization of temporal dynamics of HRV
 - ▶ power spectrum (bands, slope)
 - ▶ entropy (rate)
 - ▶ long-memory, scale-free and multifractal
 - ▶ non Gaussianity
 - ▶ ... [[Akselrod81](#),[Costa05](#),[Ivanov99](#),[Yamamoto94](#),...]



Data courtesy Y. Yamamoto (U Tokyo), K. Kiyono (U Osaka)

Motivation: Heart rate variability (HRV)

- ▶ Statistical characterization of temporal dynamics of HRV
 - ▶ power spectrum (bands, slope)
 - ▶ entropy (rate)
 - ▶ long-memory, scale-free and multifractal
 - ▶ non Gaussianity
 - ▶ ... [Akselrod81, Costa05, Ivanov99, Yamamoto94, ...] ...
- ▶ Non Gaussian multiscale representations
 - ▶ large continuum of time scales
 - ▶ no a priori scale-free dynamics
 - ▶ combining range of statistical orders



Data courtesy Y. Yamamoto (U Tokyo), K. Kiyono (U Osaka)

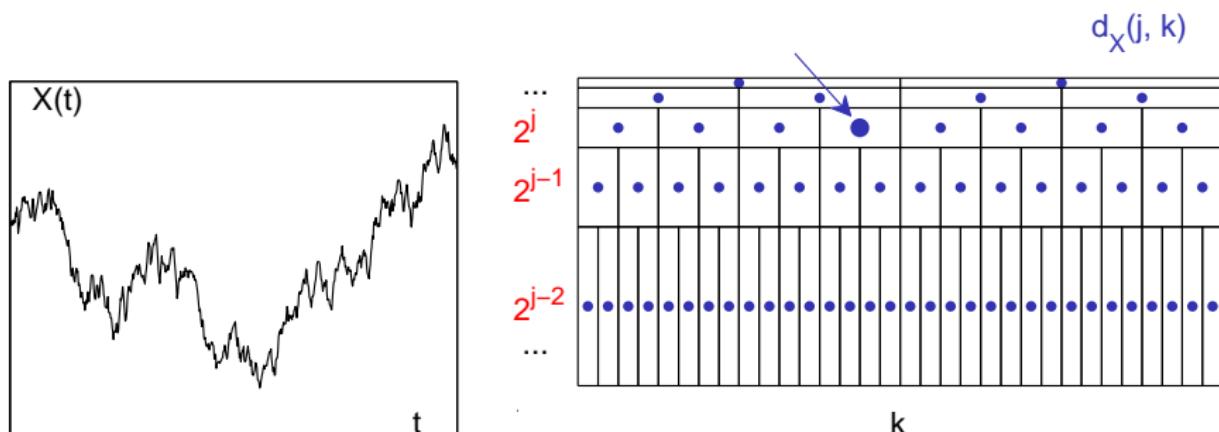
Scale-free dynamics and multifractal analysis

Wavelet coefficients and wavelet p-leaders

- Discrete wavelet transform:

- mother wavelet $\psi(t)$ with vanishing moments N_ψ :
 $\int_{\mathcal{R}} t^k \psi(t) dt \equiv 0, \quad \forall k \in \mathbb{N}, k < N_\psi; \quad \int_{\mathcal{R}} t^{N_\psi} \psi(t) dt \neq 0$
- wavelet basis $\{\psi_{j,k}(t) = 2^{-j/2} \psi(2^{-j}t - k)\}_{(j,k) \in \mathbb{N}^2}$
- wavelet coefficients (L^1 -normalized)

$$d(j, k) = \langle 2^{-j/2} \psi_{j,k} | X \rangle$$



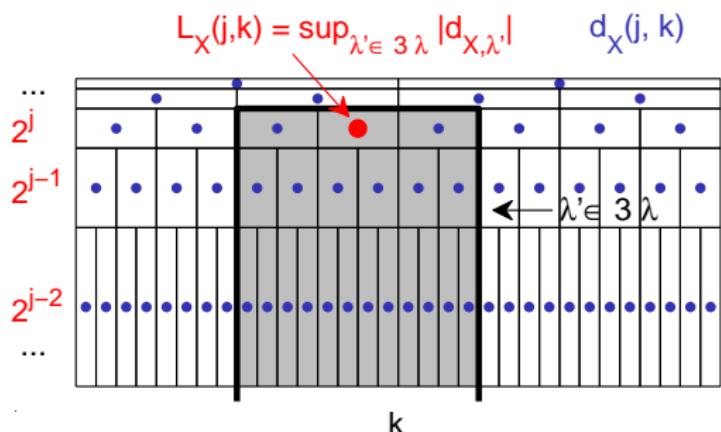
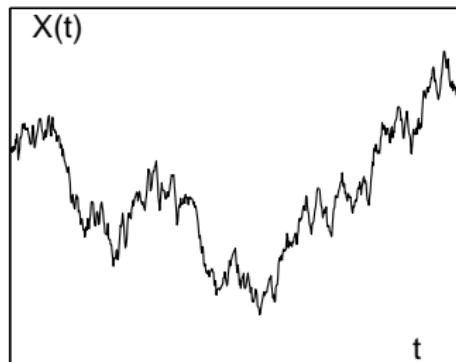
Scale-free dynamics and multifractal analysis

Wavelet coefficients and wavelet p-leaders

- Discrete wavelet transform: DWT coefficients $d(j, k)$
- wavelet p-leaders $\{\ell(j \cdot, \cdot)\}$ [Jaffard16]

$$\ell^{(p)}(j, k) \triangleq \left(\sum_{\lambda_{j',k'} \subset 3\lambda_{j,k}} |d(j', k')|^p 2^{-(j-j')} \right)^{1/p}$$

shown: wavelet leaders $p = +\infty$



Scale-free dynamics and multifractal analysis

Wavelet coefficients and wavelet p-leaders

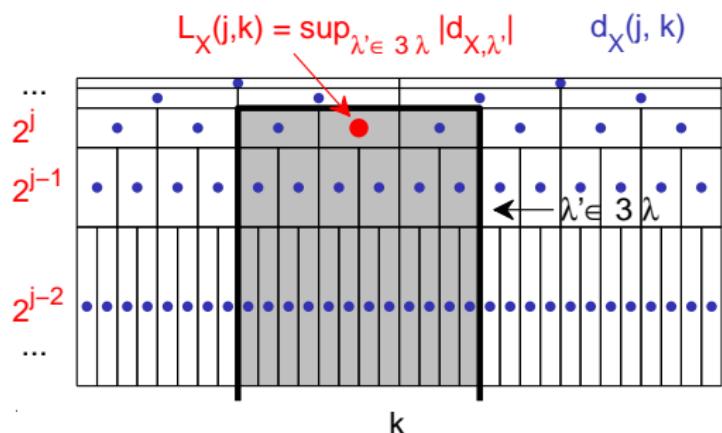
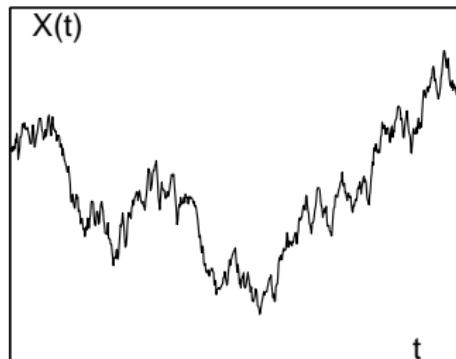
- Discrete wavelet transform: DWT coefficients $d(j, k)$
- wavelet p-leaders $\{\ell(j \cdot, \cdot)\}$ [Jaffard16]

$$\ell^{(p)}(j, k) \triangleq \left(\sum_{\lambda_{j',k'} \subset 3\lambda_{j,k}} |d(j', k')|^p 2^{-(j-j')} \right)^{1/p}$$

→ link with scale invariance

→ link with multifractals

shown: wavelet leaders $p = +\infty$



Scale-free dynamics and multifractal analysis

Moments, cumulants and scale invariance

- ▶ $\ell_s(k)$: p -leader coefficient $\ell^{(p)}(j, k)$ at scale $s = 2^j$
 - ▶ $L_q(s)$: log of q -th moments $\mathbb{E}\ell_s^q$, assumed finite ($\mathbb{E}\ell_s^q < \infty$)

$$L_q(s) \triangleq \log(\mathbb{E}\ell_s^q)$$

- ▶ $C_m(s)$: m -th cumulant of $\log(\ell_s)$

$$C_m(s) \triangleq \text{Cum}_m \log(\ell_s)$$

- ▶ Standard generating function expansion argument links them:

$$L_q(s) = \log(\mathbb{E}\ell_s^q) = \log \mathbb{E} e^{q \log(\ell_s)} = \sum_{m=1}^{\infty} C_m(s) \frac{q^m}{m!} \quad (1)$$

- ▶ Scale-free temporal dynamics \implies specific scale-dependence

$$L_q(s) = \kappa_q + \zeta(q) \log(s) \quad \text{scaling exponent } \zeta(q) \quad (2)$$

$$C_m(s) = c_m^0 + c_m \log(s) \quad \text{log-cumulant } c_m \quad (3)$$

Scale-free dynamics and multifractal analysis

Moments, cumulants and scale invariance

- ▶ $\ell_s(k)$: p -leader coefficient $\ell^{(p)}(j, k)$ at scale $s = 2^j$
 - ▶ $L_q(s)$: log of q -th moments $\mathbb{E}\ell_s^q$, assumed finite ($\mathbb{E}\ell_s^q < \infty$)

$$L_q(s) \triangleq \log(\mathbb{E}\ell_s^q)$$

- ▶ $C_m(s)$: m -th cumulant of $\log(\ell_s)$

$$C_m(s) \triangleq \text{Cum}_m \log(\ell_s)$$

- ▶ Standard generating function expansion argument links them:

$$L_q(s) = \log(\mathbb{E}\ell_s^q) = \log \mathbb{E} e^{q \log(\ell_s)} = \sum_{m=1}^{\infty} C_m(s) \frac{q^m}{m!} \quad (1)$$

- ▶ Scale-free temporal dynamics \implies specific scale-dependence

$$L_q(s) = \kappa_q + \zeta(q) \log(s) \quad \text{scaling exponent } \zeta(q) \quad (2)$$

$$C_m(s) = c_m^0 + c_m \log(s) \quad \text{log-cumulant } c_m \quad (3)$$

Scale-free dynamics and multifractal analysis

Moments, cumulants and scale invariance

- ▶ $\ell_s(k)$: p -leader coefficient $\ell^{(p)}(j, k)$ at scale $s = 2^j$
 - ▶ $L_q(s)$: log of q -th moments $\mathbb{E}\ell_s^q$, assumed finite ($\mathbb{E}\ell_s^q < \infty$)

$$L_q(s) \triangleq \log(\mathbb{E}\ell_s^q)$$

- ▶ $C_m(s)$: m -th cumulant of $\log(\ell_s)$

$$C_m(s) \triangleq \text{Cum}_m \log(\ell_s)$$

- ▶ Standard generating function expansion argument links them:

$$L_q(s) = \log(\mathbb{E}\ell_s^q) = \log \mathbb{E} e^{q \log(\ell_s)} = \sum_{m=1}^{\infty} C_m(s) \frac{q^m}{m!} \quad (1)$$

- ▶ Scale-free temporal dynamics \implies specific scale-dependence

$$L_q(s) = \kappa_q + \zeta(q) \log(s) \quad \text{scaling exponent } \zeta(q) \quad (2)$$

$$C_m(s) = c_m^0 + c_m \log(s) \quad \text{log-cumulant } c_m \quad (3)$$

Scale-free dynamics and multifractal analysis

Multifractal spectrum

e.g., [Jaffard16,Wendt07]

- Local regularity of $X(t)$ at t_0

p-exponent

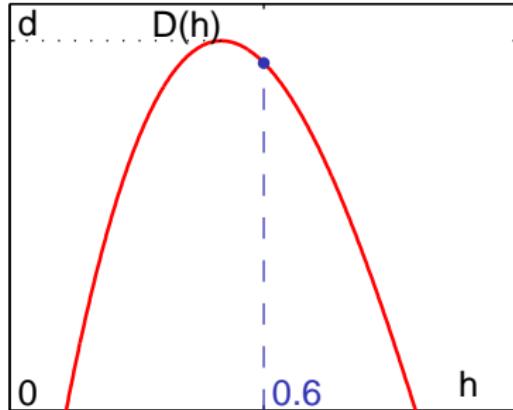
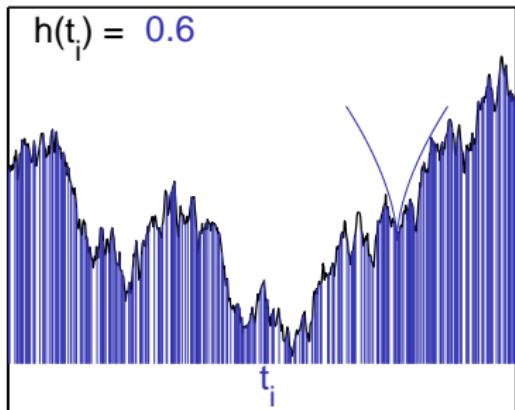
(Hölder exponent: $p = +\infty$)

$$h_p(t_0) \triangleq \sup_{\alpha} \left\{ \alpha : \left(\frac{1}{s} \int_{B(0,s)} |X(t) - X(t_0)|^p dt \right)^{\frac{1}{p}} < Cs^{\alpha} \right\}, \quad s > 0$$

- Multifractal Spectrum $\mathcal{D}(h_p)$: Fluctuations of regularity $h_p(t)$

- Set of points that share same regularity $\{t_i | h_p(t_i) = h\}$
- Fractal (or Hausdorff) Dimension of each set:

$$\mathcal{D}(h) \triangleq \dim_H \{t : h_p(t) = h\}$$



Scale-free dynamics and multifractal analysis

Multifractal spectrum and scale invariance

Key property of p-leaders ℓ_s : (under uniform regularity conditions)

$$h_p(t_0) = \liminf_{j \rightarrow -\infty} \frac{\log(\ell_{s=2^j}(k(t_0)))}{\log(2^j)}$$

Moments $\mathbb{E} \ell_s^q \rightarrow$ empirical moments $S_n(s, q) \triangleq \frac{n}{s} \sum_k \ell_s^q(k)$

$$\begin{aligned} \Rightarrow S_n(s, q) &\simeq \sum_h s^{1-\mathcal{D}(h)} s^{hq}, \\ &\simeq \sum_h s^{1-\mathcal{D}(h)+hq}, \\ &\sim_{s \rightarrow 0} \kappa_q s^{\zeta(q)}, \quad \text{scaling exponents } \zeta(q) \end{aligned}$$

Saddle-point argument: \Rightarrow Legendre transform

$$\zeta(q) = \inf_q (1 + hq - \mathcal{D}(h))$$

Multifractal formalism:

$$\mathcal{D}(h) \leq \inf_q (d + hq - \zeta(q)) \tag{4}$$

Scale-free dynamics and multifractal analysis

Multifractal spectrum and scale invariance

Key property of p-leaders ℓ_s : (under uniform regularity conditions)

$$h_p(t_0) = \liminf_{j \rightarrow -\infty} \frac{\log(\ell_{s=2^j}(k(t_0)))}{\log(2^j)}$$

Moments $\mathbb{E}\ell_s^q \rightarrow$ empirical moments $S_n(s, q) \triangleq \frac{n}{s} \sum_k \ell_s^q(k)$

$$\begin{aligned} \implies S_n(s, q) &\simeq \sum_h s^{1-\mathcal{D}(h)} s^{hq}, \\ &\simeq \sum_h s^{1-\mathcal{D}(h)+hq}, \\ &\sim_{s \rightarrow 0} \kappa_q s^{\zeta(q)}, \quad \text{scaling exponents } \zeta(q) \end{aligned}$$

Saddle-point argument: \Rightarrow Legendre transform

$$\zeta(q) = \inf_q (1 + hq - \mathcal{D}(h))$$

Multifractal formalism:

$$\mathcal{D}(h) \leq \inf_q (d + hq - \zeta(q)) \tag{4}$$

Scale-free dynamics and multifractal analysis

Multifractal spectrum and scale invariance

Key property of p-leaders ℓ_s : (under uniform regularity conditions)

$$h_p(t_0) = \liminf_{j \rightarrow -\infty} \frac{\log(\ell_{s=2^j}(k(t_0)))}{\log(2^j)}$$

Moments $\mathbb{E}\ell_s^q \rightarrow$ empirical moments $S_n(s, q) \triangleq \frac{n}{s} \sum_k \ell_s^q(k)$

$$\begin{aligned} \implies S_n(s, q) &\simeq \sum_h s^{1-\mathcal{D}(h)} s^{hq}, \\ &\simeq \sum_h s^{1-\mathcal{D}(h)+hq}, \\ &\sim_{s \rightarrow 0} \kappa_q s^{\zeta(q)}, \quad \text{scaling exponents } \zeta(q) \end{aligned}$$

Saddle-point argument: \Rightarrow Legendre transform

$$\zeta(q) = \inf_q (1 + hq - \mathcal{D}(h))$$

Multifractal formalism:

$$\mathcal{D}(h) \leq \inf_q (d + hq - \zeta(q)) \tag{4}$$

Departures from Gaussian: Cumulants $C_m(s)$?

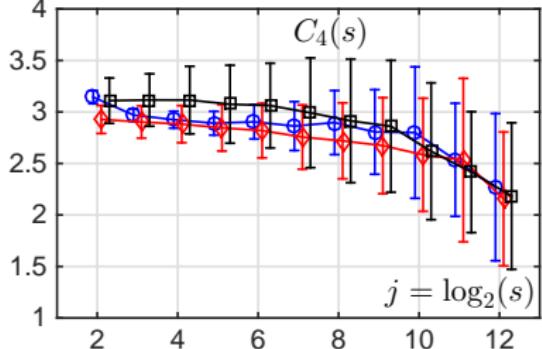
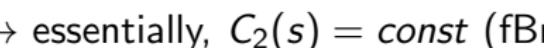
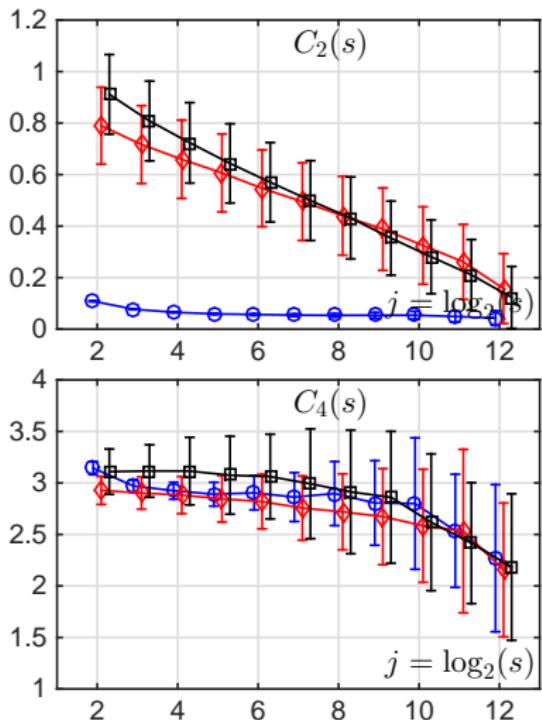
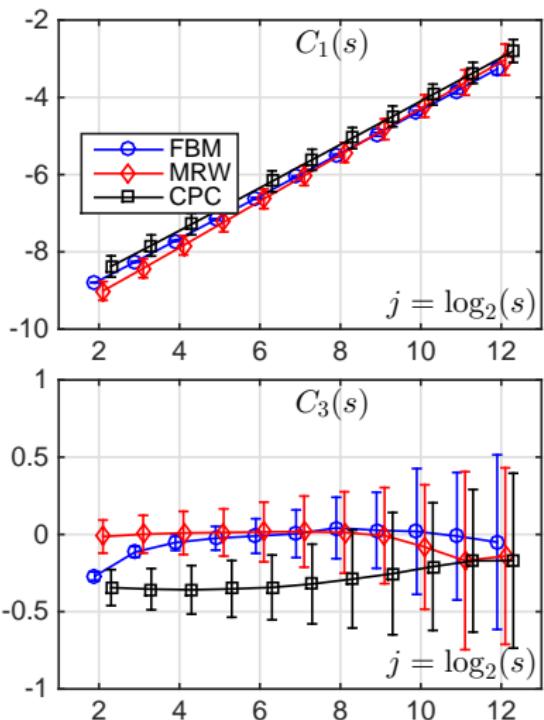
- ▶ Cumulants $C_m(s)$ provide description of **statistics of $\log \ell_s$**
 - ▶ $C_1(s)$ - mean of $\log \ell_s$
→ linear, Gaussian (mean & variance/covariance) properties
 - ▶ $C_2(s)$ - variance
 - ▶ $C_3(s)$ - skewness
 - ▶ $C_4(s)$ - kurtosis, etc ...
- ▶ Illustration for Gaussian / non Gaussian scale-free models
 - ▶ Fractional Brownian motion (fBm):
Gaussian, self-similar
 $C_1(s) = f(s)$, $C_m(s) = \text{const}$ for $m \geq 2$
 - ▶ Multifractal random walk (MRW):
non Gaussian multifractal of a special log-normal type
 $C_1(s), C_2(s) = f(s)$, $C_m(s) = \text{const}$ for $m \geq 3$;
 - ▶ Compound Poisson cascade (CPC):
non Gaussian multifractal of more general form
 $C_m(s) = f(s) \neq \text{const} \forall m \geq 1$

Departures from Gaussian: Cumulants $C_m(s)$?

- ▶ Cumulants $C_m(s)$ provide description of **statistics of $\log \ell_s$**
 - ▶ $C_1(s)$ - mean of $\log \ell_s$
→ linear, Gaussian (mean & variance/covariance) properties
 - ▶ $C_2(s)$ - variance
 - ▶ $C_3(s)$ - skewness
 - ▶ $C_4(s)$ - kurtosis, etc ...
- ▶ Illustration for Gaussian / non Gaussian scale-free models
 - ▶ Fractional Brownian motion (fBm):
Gaussian, self-similar
 $C_1(s) = f(s)$, $C_m(s) = \text{const}$ for $m \geq 2$
 - ▶ Multifractal random walk (MRW):
non Gaussian multifractal of a special log-normal type
 $C_1(s), C_2(s) = f(s)$, $C_m(s) = \text{const}$ for $m \geq 3$;
 - ▶ Compound Poisson cascade (CPC):
non Gaussian multifractal of more general form
 $C_m(s) = f(s) \neq \text{const} \forall m \geq 1$

Scale-free dynamics and multifractal analysis

Departures from Gaussian: Cumulants $C_m(s)$?



→ essentially, $C_2(s) = \text{const}$ (fBm) vs. $C_2(s) = f(s)$ (MRW, CPC)

Non Gaussian multiscale expansion: Definition

Definition. Let $P \in \mathbb{N}^+$, $\mathbf{q} = (q_1, \dots, q_{2P})$ with $q_i \neq 0$ and $q_i \neq q_j$:

$$\mathsf{L}_{\mathbf{q}}^{(2P)}(s) \triangleq \log \left(\prod_{i=1}^P \frac{(\mathbb{E} \ell_s^{q_{2i-1}})^{\frac{1}{q_{2i-1}}}}{(\mathbb{E} \ell_s^{q_{2i}})^{\frac{1}{q_{2i}}}} \right) \quad (5)$$

[Wendt18]

- Making use of (1) $(L_q(s) = \log(\mathbb{E} \ell_s^q) = \sum_{m=1}^{\infty} C_m(s) \frac{q^m}{m!})$

$$\begin{aligned} \mathsf{L}_{\mathbf{q}}^{(2P)}(s) &= \sum_{i=1}^P \frac{L_s(q_{2i-1})}{q_{2i-1}} - \frac{L_s(q_{2i})}{q_{2i}} \\ &= \sum_{m=2}^{\infty} C_m(s) \frac{\sum_{i=1}^P q_{2i-1}^{m-1} - q_{2i}^{m-1}}{m!} \end{aligned} \quad (6)$$

⇒ infinite **weighted** sum of the cumulants $C_m(s)$

Non Gaussian multiscale expansion: Definition

Definition. Let $P \in \mathbb{N}^+$, $\mathbf{q} = (q_1, \dots, q_{2P})$ with $q_i \neq 0$ and $q_i \neq q_j$:

$$\mathsf{L}_{\mathbf{q}}^{(2P)}(s) \triangleq \log \left(\prod_{i=1}^P \frac{(\mathbb{E} \ell_s^{q_{2i-1}})^{\frac{1}{q_{2i-1}}}}{(\mathbb{E} \ell_s^{q_{2i}})^{\frac{1}{q_{2i}}}} \right) \quad (5)$$

[Wendt18]

- Making use of (1) $(L_q(s) = \log(\mathbb{E} \ell_s^q) = \sum_{m=1}^{\infty} C_m(s) \frac{q^m}{m!})$

$$\begin{aligned} \mathsf{L}_{\mathbf{q}}^{(2P)}(s) &= \sum_{i=1}^P \frac{L_s(q_{2i-1})}{q_{2i-1}} - \frac{L_s(q_{2i})}{q_{2i}} \\ &= \sum_{m=2}^{\infty} C_m(s) \frac{\sum_{i=1}^P q_{2i-1}^{m-1} - q_{2i}^{m-1}}{m!} \end{aligned} \quad (6)$$

⇒ infinite **weighted** sum of the **cumulants** $C_m(s)$

Non Gaussian multiscale representations

Non Gaussian multiscale expansion: Properties [Wendt18]

Property 0: Evolution across scales s of $\mathbf{L}_q^{(2P)}(s)$ indicator for non Gaussian properties.

⇒ Gaussian data: $\mathbf{L}_q^{(2P)}(s)$ equals a constant.

Property 1: $\mathbf{L}_q^{(2P)}(s)$ independent of $C_1(s)$

⇒ only quantifies nonlinear properties, departures from Gaussian

$$(q_{2i}^{m-1} - q_{2i+1}^{m-1} = 0 \text{ for } m = 1)$$

Property 2: Relative weighting of $C_m(s)$ in $\mathbf{L}_q^{(2P)}(s)$ can be tuned

⇒ quantify the strength or nature of deviation from Gaussian

Property 3: $\mathbf{L}_q^{(2P)}(s)$ permits to probe cumulants of high order

⇒ using only moments $L_q(s) = \log(\mathbb{E}\ell_s^q)$ of low orders q

Non Gaussian multiscale representations

Non Gaussian multiscale expansion: Properties [Wendt18]

Property 0: Evolution across scales s of $\mathbf{L}_q^{(2P)}(s)$ indicator for non Gaussian properties.

⇒ Gaussian data: $\mathbf{L}_q^{(2P)}(s)$ equals a constant.

Property 1: $\mathbf{L}_q^{(2P)}(s)$ independent of $C_1(s)$

⇒ only quantifies nonlinear properties, departures from Gaussian
$$(q_{2i}^{m-1} - q_{2i+1}^{m-1} = 0 \text{ for } m = 1)$$

Property 2: Relative weighting of $C_m(s)$ in $\mathbf{L}_q^{(2P)}(s)$ can be tuned

⇒ quantify the strength or nature of deviation from Gaussian

Property 3: $\mathbf{L}_q^{(2P)}(s)$ permits to probe cumulants of high order

⇒ using only moments $L_q(s) = \log(\mathbb{E}\ell_s^q)$ of low orders q

Non Gaussian multiscale expansion: Properties [Wendt18]

Property 0: Evolution across scales s of $\mathbf{L}_q^{(2P)}(s)$ indicator for non Gaussian properties.

⇒ Gaussian data: $\mathbf{L}_q^{(2P)}(s)$ equals a constant.

Property 1: $\mathbf{L}_q^{(2P)}(s)$ independent of $C_1(s)$

⇒ only quantifies nonlinear properties, departures from Gaussian
$$(q_{2i}^{m-1} - q_{2i+1}^{m-1} = 0 \text{ for } m = 1)$$

Property 2: Relative weighting of $C_m(s)$ in $\mathbf{L}_q^{(2P)}(s)$ can be tuned

⇒ quantify the strength or nature of deviation from Gaussian

Property 3: $\mathbf{L}_q^{(2P)}(s)$ permits to probe cumulants of high order

⇒ using only moments $L_q(s) = \log(\mathbb{E}\ell_s^q)$ of low orders q

Non Gaussian multiscale expansion: Properties [Wendt18]

Property 0: Evolution across scales s of $\mathbf{L}_q^{(2P)}(s)$ indicator for non Gaussian properties.

⇒ Gaussian data: $\mathbf{L}_q^{(2P)}(s)$ equals a constant.

Property 1: $\mathbf{L}_q^{(2P)}(s)$ independent of $C_1(s)$

⇒ only quantifies nonlinear properties, departures from Gaussian
 $(q_{2i}^{m-1} - q_{2i+1}^{m-1} = 0 \text{ for } m = 1)$

Property 2: Relative weighting of $C_m(s)$ in $\mathbf{L}_q^{(2P)}(s)$ can be tuned
⇒ quantify the strength or nature of deviation from Gaussian

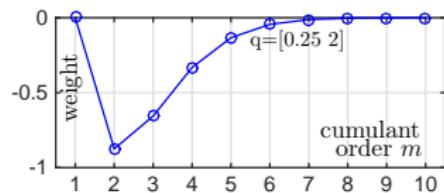
Property 3: $\mathbf{L}_q^{(2P)}(s)$ permits to probe cumulants of high order

⇒ using only moments $L_q(s) = \log(\mathbb{E}\ell_s^q)$ of low orders q

Non Gaussian multiscale representations

Non Gaussian multiscale expansion: Examples

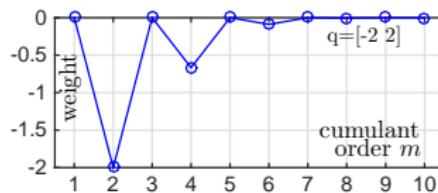
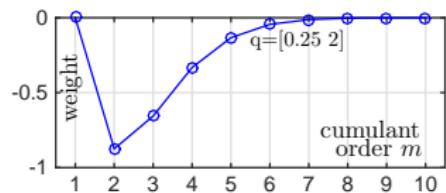
notation	moments \mathbf{q}	cumulants C_m in (6) and interpretation
$\mathbf{L}_{\mathbf{q}}^{(2)}(s)$	(0.25, 2)	$m \geq 2$ <i>any deviation from Gaussian</i>



Non Gaussian multiscale representations

Non Gaussian multiscale expansion: Examples

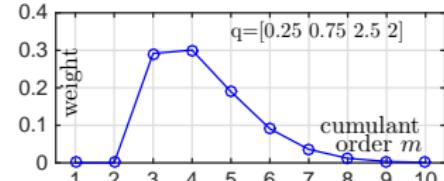
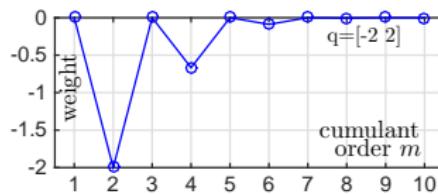
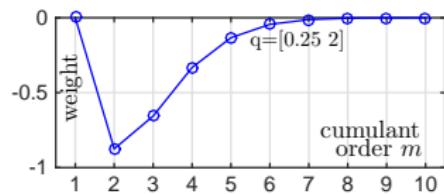
notation	moments \mathbf{q}	cumulants C_m in (6) and interpretation
$\mathbf{L}_{\mathbf{q}}^{(2)}(s)$	(0.25, 2)	$m \geq 2$ <i>any deviation from Gaussian</i>
$\mathbf{L}_{\mathbf{q}}^{(2)*}(s)$	(-2, 2)	even order $m = 2, 4, \dots$ <i>mainly symmetric properties of non Gaussian</i>



Non Gaussian multiscale representations

Non Gaussian multiscale expansion: Examples

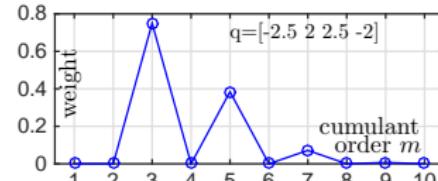
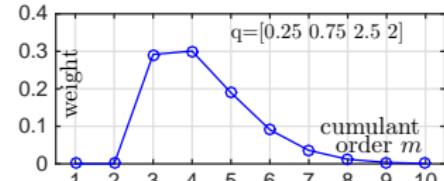
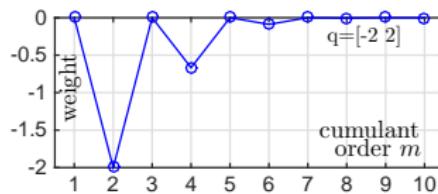
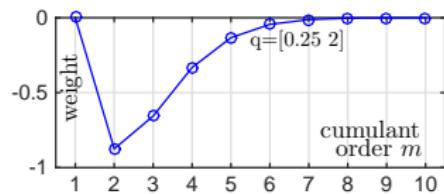
notation	moments \mathbf{q}	cumulants C_m in (6) and interpretation
$\mathbf{L}_{\mathbf{q}}^{(2)}(s)$	(0.25, 2)	$m \geq 2$ <i>any deviation from Gaussian</i>
$\mathbf{L}_{\mathbf{q}}^{(2)*}(s)$	(-2, 2)	even order $m = 2, 4, \dots$ <i>mainly symmetric properties of non Gaussian</i>
$\mathbf{L}_{\mathbf{q}}^{(4)}(s)$	(0.25, 0.75, 2.5, 2)	$m \geq 3$ <i>non log-normal nature of non Gaussian</i>



Non Gaussian multiscale representations

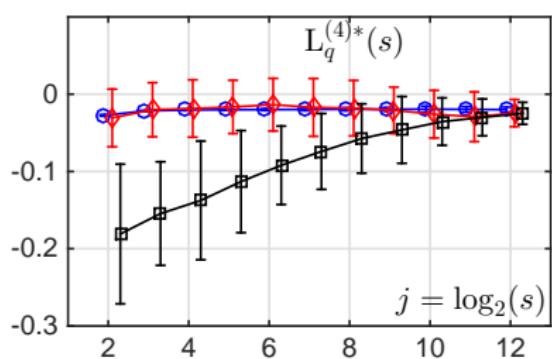
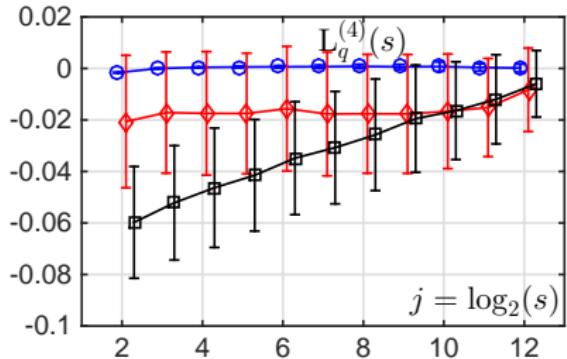
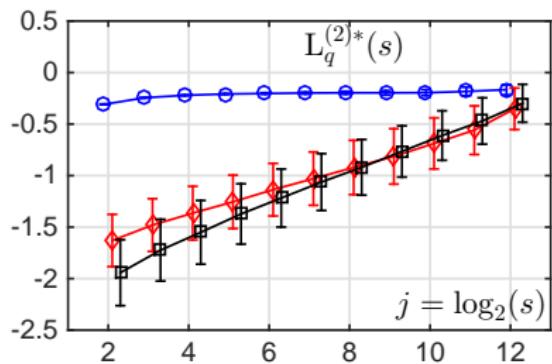
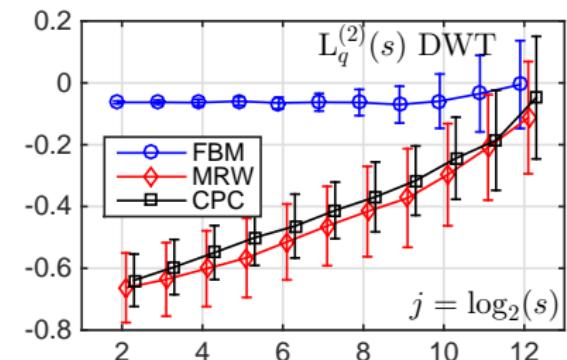
Non Gaussian multiscale expansion: Examples

notation	moments \mathbf{q}	cumulants C_m in (6) and interpretation
$\mathbf{L}_{\mathbf{q}}^{(2)}(s)$	(0.25, 2)	$m \geq 2$ <i>any deviation from Gaussian</i>
$\mathbf{L}_{\mathbf{q}}^{(2)*}(s)$	(-2, 2)	even order $m = 2, 4, \dots$ <i>mainly symmetric properties of non Gaussian</i>
$\mathbf{L}_{\mathbf{q}}^{(4)}(s)$	(0.25, 0.75, 2.5, 2)	$m \geq 3$ <i>non log-normal nature of non Gaussian</i>
$\mathbf{L}_{\mathbf{q}}^{(4)*}(s)$	(-2.5, 2, 2.5, -2)	$m \geq 3$ of odd order: $m = 3, 5, \dots$ <i>asymmetry of non Gaussian</i>



Non Gaussian multiscale representations

Non Gaussian multiscale expansion: Illustration



⇒ discriminate: Gaussian (fBm) - non G LN (MRW) - non G (CPC)

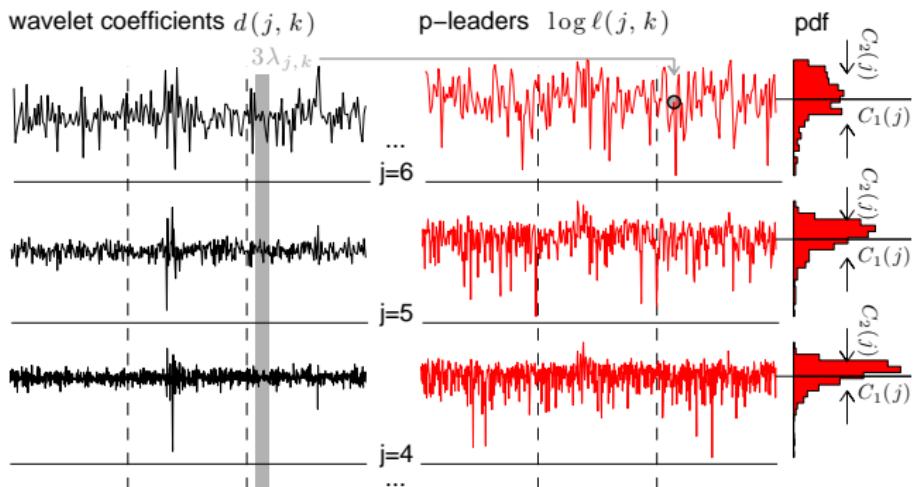
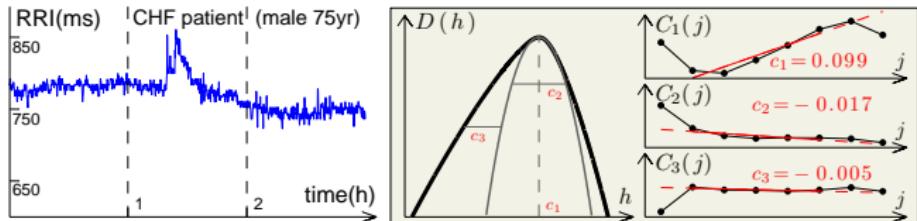
Application HRV analysis in CHF

Congestive Heart failure (CHF) patient HRV data

- ▶ cohort of 108 CHF patients
 - ▶ Fujita Health University Hospital, 2000-2001
 - ▶ 61 male and 47 female
 - ▶ age 21-92 (66.1 ± 14.8) years
 - ▶ similar health status before hospital discharge
 - ▶ 39 (36.1%) dead in follow-up of 33 ± 17 (range 1 – 59) months
clinical details: [Kiyono 2008]
- ▶ age-matched healthy (HM) control group of 90 subjects
- ▶ 24-hour Holter ECG recording
 - ▶ recorded prior to Hospital discharge
 - ▶ outlier / detection error corrected, no sustained tachyarrhythmias
- ▶ analyzed time series:
 - ▶ primitive $y_n = \sum_{i=1}^n x_n$ of RR inter-arrival time values
 $X \equiv \{x_n, n = 1, \dots, N\}$
interpolated at 4Hz (cubic spline)

Application HRV analysis in CHF

Estimation



$$N_\psi = 3, \text{ octaves } j_1 = 4 \leq j \leq j_2 = 9 \text{ (5.3s to 170.7s)}$$

Application HRV analysis in CHF

Spectral, Entropy and multifractal features

Estimates	LF	HF	LF/HF	α_{PSD}	sampEN	apEN
HM	0.057	0.028	3.001	2.314	0.267	0.324
CHF	0.064	0.160	0.713	1.362	0.431	0.600
NS	0.064	0.132	0.692	1.476	0.378	0.538
SV	0.064	0.176	0.725	1.300	0.463	0.637
WRS p-value	LF	HF	LF/HF	α_{PSD}	sampEN	apEN
CHF vs. HM	0.435	< 1e-8	< 1e-8	< 1e-8	< 1e-8	< 1e-8
NS vs. SV	0.406	0.173	0.927	0.327	0.062	0.127

Estimates	c_1	c_2	c_3	c_4
HM	0.255	-0.032	0.005	-0.005
CHF	0.079	-0.058	-0.008	-0.022
NS	0.053	-0.060	0.009	-0.070
SV	0.095	-0.057	-0.017	0.006
WRS p-value	c_1	c_2	c_3	c_4
CHF vs. HM	< 1e-8	0.276	3.0e-4	0.111
NS vs. SV	0.156	0.720	0.932	0.068

Application HRV analysis in CHF

Spectral, Entropy and multifractal features

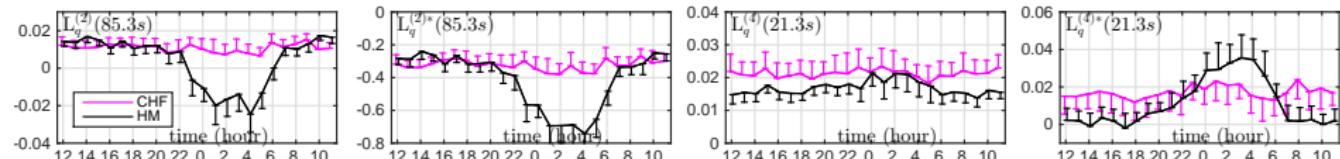
Estimates	LF	HF	LF/HF	α_{PSD}	sampEN	apEN
HM	0.057	0.028	3.001	2.314	0.267	0.324
CHF	0.064	0.160	0.713	1.362	0.431	0.600
NS	0.064	0.132	0.692	1.476	0.378	0.538
SV	0.064	0.176	0.725	1.300	0.463	0.637
WRS p-value	LF	HF	LF/HF	α_{PSD}	sampEN	apEN
CHF vs. HM	0.435	< 1e-8	< 1e-8	< 1e-8	< 1e-8	< 1e-8
NS vs. SV	0.406	0.173	0.927	0.327	0.062	0.127

Estimates	c_1	c_2	c_3	c_4
HM	0.255	-0.032	0.005	-0.005
CHF	0.079	-0.058	-0.008	-0.022
NS	0.053	-0.060	0.009	-0.070
SV	0.095	-0.057	-0.017	0.006
WRS p-value	c_1	c_2	c_3	c_4
CHF vs. HM	< 1e-8	0.276	3.0e-4	0.111
NS vs. SV	0.156	0.720	0.932	0.068

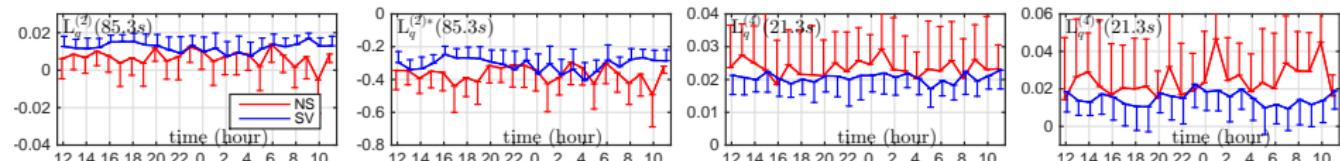
Application HRV analysis in CHF

Non Gaussian multiscale expansion: Circadian evolution

CHF vs. HM



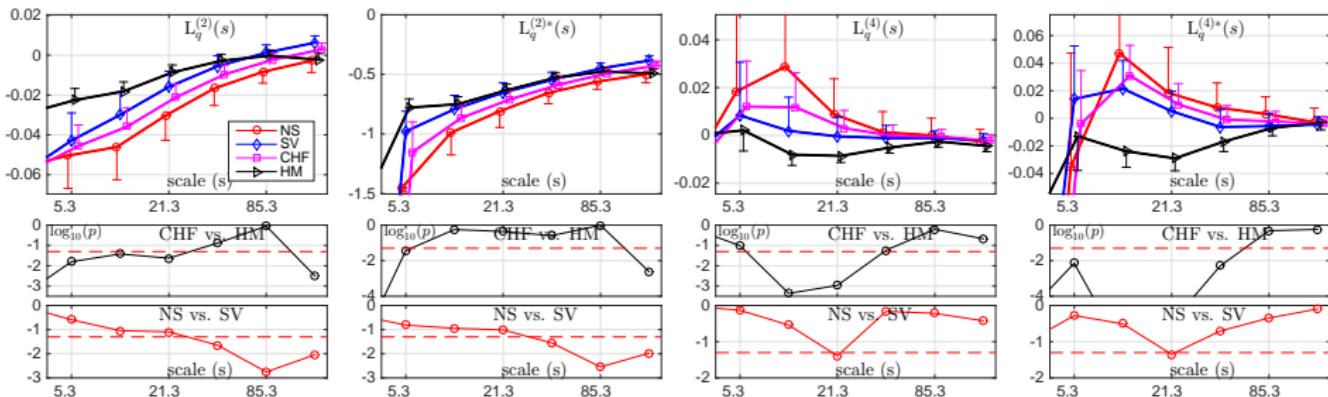
NS vs. SV



- ▶ HM: clear day-night evolution for all non Gaussian indices
- ▶ HM nighttime increase in non Gaussianity driven by:
 - ▶ purely asymmetric properties ($L_q^{(4)*}$)
 - ▶ even order properties ($L_q^{(4)}$)
 - ▶ stronger deviation from log-Normal
- ▶ No circadian modulation for CHF patients

Application HRV analysis in CHF

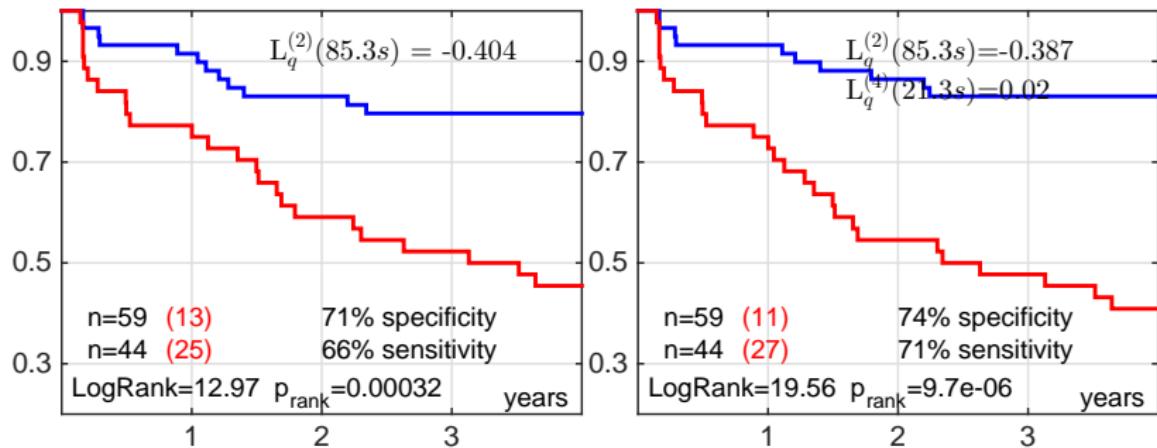
Non Gaussian multiscale expansion: 6h daytime



- Degree of non Gaussianity ($L_q^{(2)}$)
 - stronger non G for CHF than for HM, and NS than SV
 - **significant** for NS vs. SV (scale 85.3s)
- Nature of non Gaussianity ($L_q^{(4)}$)
 - non log-Normal type
 - stronger non LN for CHF than for HM, and NS than SV
 - **significant** for NS vs. SV (scale 21.3s)

Application HRV analysis in CHF

Non Gaussian multiscale expansion: discrimination



- ▶ Best results for spectral and entropy based analyses:
 - ▶ specificity $\leq 40\%$, $p > 0.01$ (sampEN, apEN)
- ▶ Degree of non Gaussianity ($L_q^{(2)}$) consistent with [Kiyono2008]
 - ▶ sensitivity (specificity) of 66% (71%), $p = 0.0003$
- ▶ Combined use of non Gaussian expansion indices ($L_q^{(2)}$ & $L_q^{(4)}$)
 - ▶ sensitivity (specificity) of 71% (74%), $p < 10^{-5}$

Conclusions

- ▶ Non Gaussian multiscale representations
 - ▶ use of continuum of scales without a priori scale-free dynamic assumptions
 - ▶ joint analysis of customizable ranges of (higher) statistical orders
 - ▶ e.g., degree and nature of departure from Gaussian
- ▶ Nonlinear dynamics of HRV (90 healthy subjects, 108 CHF patients)
 - ▶ circadian evolution of non Gaussian indices for healthy subjects only
 - ▶ specific non Gaussian indices possess high discriminative abilities between NS and SV CHF patients
 - ▶ for specific time scales (21.3s and 85.3s)
 - ▶ substantially improving upon spectral, entropy and multifractal based indices

Conclusions

- ▶ Non Gaussian multiscale representations
 - ▶ use of continuum of scales without a priori scale-free dynamic assumptions
 - ▶ joint analysis of customizable ranges of (higher) statistical orders
 - ▶ e.g., degree and nature of departure from Gaussian
- ▶ Nonlinear dynamics of HRV (90 healthy subjects, 108 CHF patients)
 - ▶ circadian evolution of non Gaussian indices for healthy subjects only
 - ▶ specific non Gaussian indices possess high discriminative abilities between NS and SV CHF patients
 - ▶ for specific time scales (21.3s and 85.3s)
 - ▶ substantially improving upon spectral, entropy and multifractal based indices

Bibliography

- [Kiyono08] K. Kiyono, J. Hayano, E. Watanabe, Z. R. Struzik, and Y. Yamamoto, "Non-Gaussian heart rate as an independent predictor of mortality in patients with chronic heart failure," *Heart Rhythm*, 2008.
- [Jaffard16] S. Jaffard, C. Melot, R. Leonarduzzi, H. Wendt, P. Abry, S. Roux, M. Torres, "p-exponent and p-leaders, part i: Negative pointwise regularity." *Physica A*, 2016.
- [Wendt07] H. Wendt, P. Abry, and S. Jaffard, "Bootstrap for empirical multifractal analysis," *IEEE Signal Proc. Mag.*, 2007.
- [Wendt18] H. Wendt, P. Abry, K. Kiyono, J. Hayano, E. Watanabe, Y. Yamamoto, "Wavelet p-Leader Non Gaussian Multiscale Expansions for Heart Rate Variability Analysis in Congestive Heart Failure Patients," *IEEE T. Biomedical Eng.*, 2018.
- [Akselrod81] S. Akselrod, D. Gordon, F. A. Ubel, D. C. Shannon, A. C. Berger, and R. J. Cohen, "Power spectrum analysis of heart rate fluctuation: a quantitative probe of beat-to-beat cardiovascular control," *Science*, 1981.
- [Costa05] M. D. Costa, A. L. Goldberger, and C.-K. Peng, "Multiscale entropy analysis of biological signals," *Physical review E*, 2005.
- [Ivanov99] P. C. Ivanov et al., "Multifractality in human heart rate dynamics," *Nature*, 1999.
- [Yamamoto94] Y. Yamamoto and R. L. Hughson, "On the fractal nature of heart rate variability in humans: effects of data length and β -adrenergic blockade," *Am. J. Physiol.*, 1994.

herwig.wendt@irit.fr

www.irit.fr/~Herwig.Wendt/