

Statistics of extreme dynamic behaviour of marine structures

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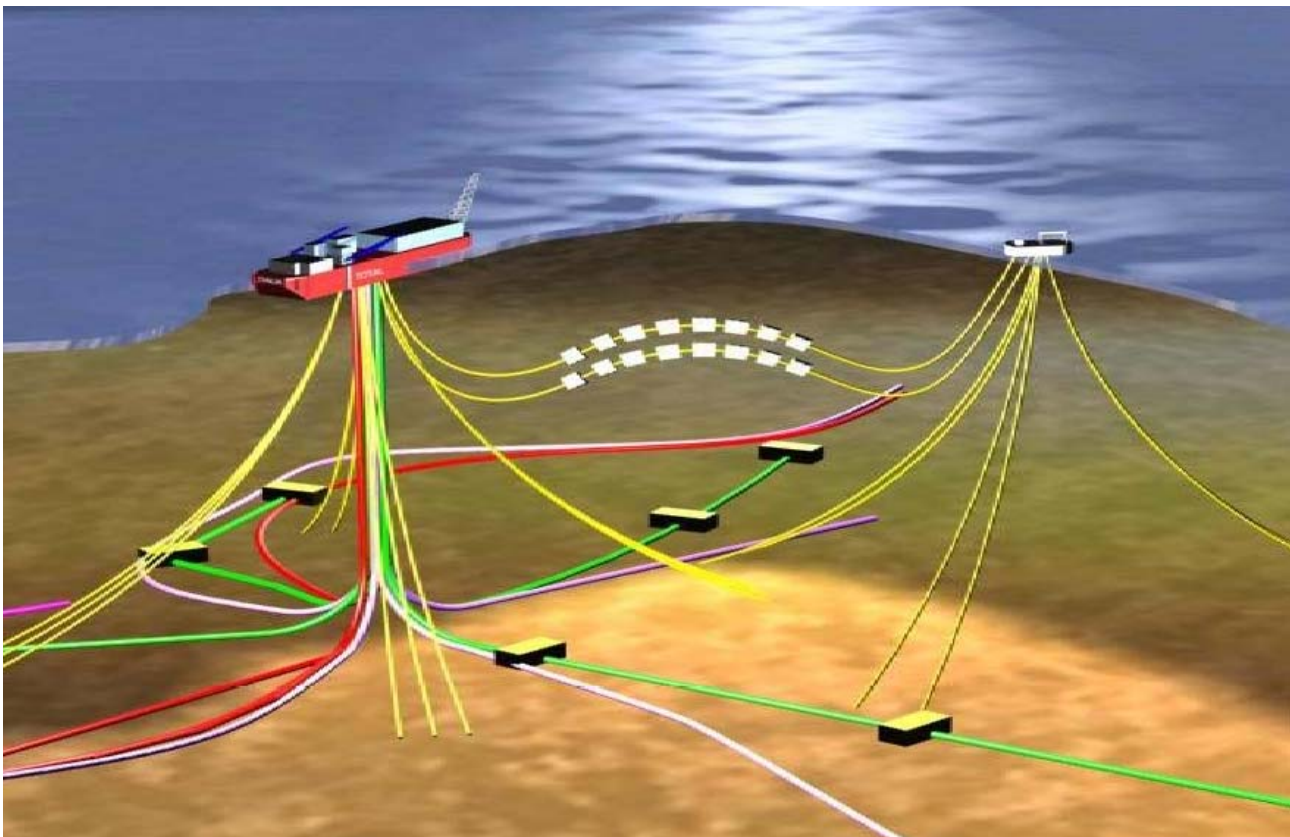
The estimation of the probability of failure of the mooring of a marine structure needs two steps. First, a probabilistic step, which corresponds to the calculation of the distributions of the maximum forces applied to the mooring lines during a given sea-state (short-term). A second step, which uses the statistics of the sea-state conditions (climatology) to estimate the probability of failure of the line, considering all the situations that the structure will encounter during its service life (long-term). To be feasible, this statistical long-term step needs a probabilistic short term calculations, not too much costly in computing time, but of course with a sufficient accuracy in all the sea-state situations.

The modelling of behaviour of the marine structures is taken more and more complex (dynamic, nonlinear), and so the calculation of the distributions of the maxima during a given sea-state needs adapted methodologies. The roll motion and the low frequency movement of a floating marine structure will be given as examples.

Comparisons between different methods and Gaussian hypotheses will be commented.

Statistics of extreme dynamic behaviour of marine structures

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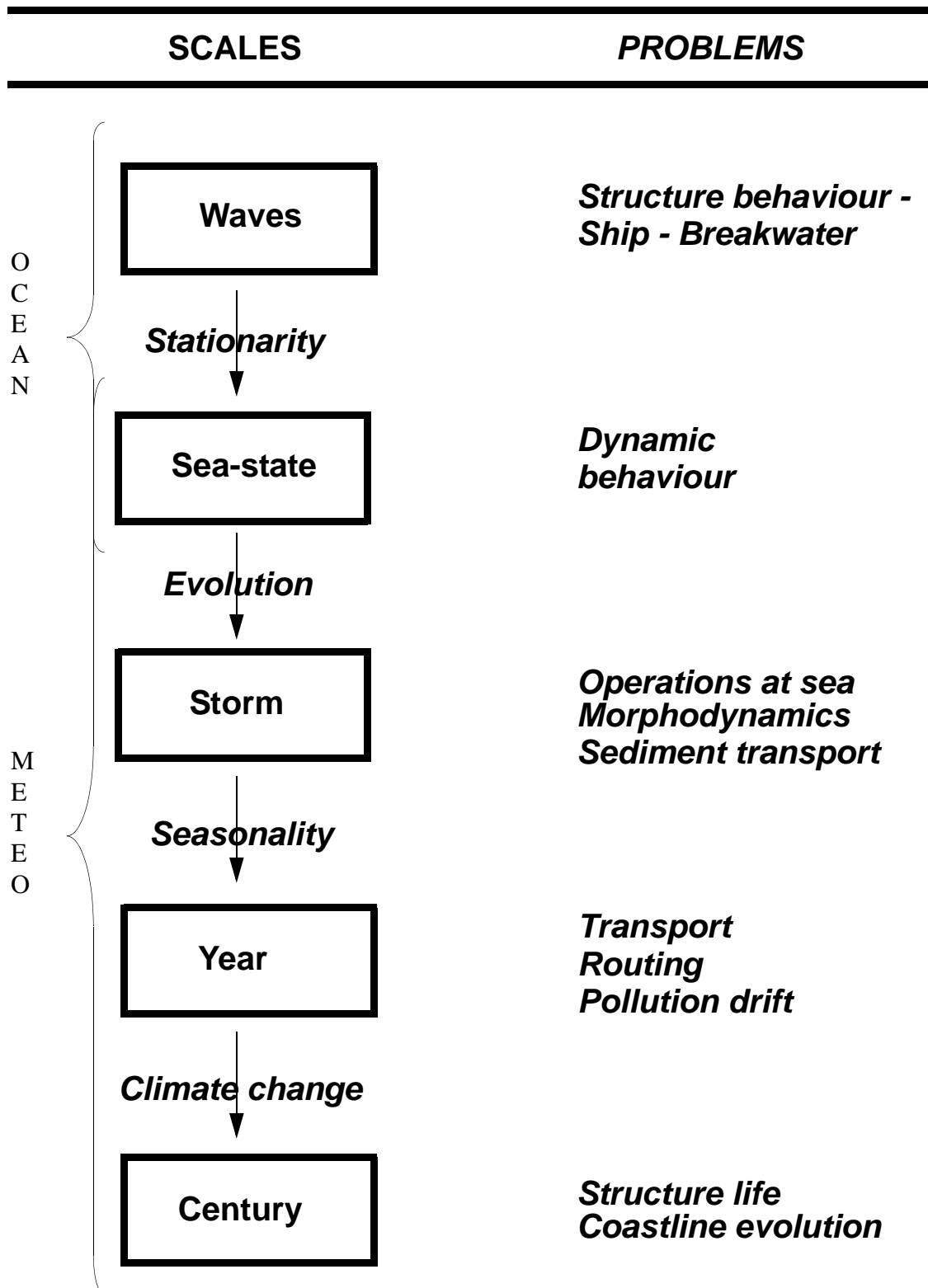


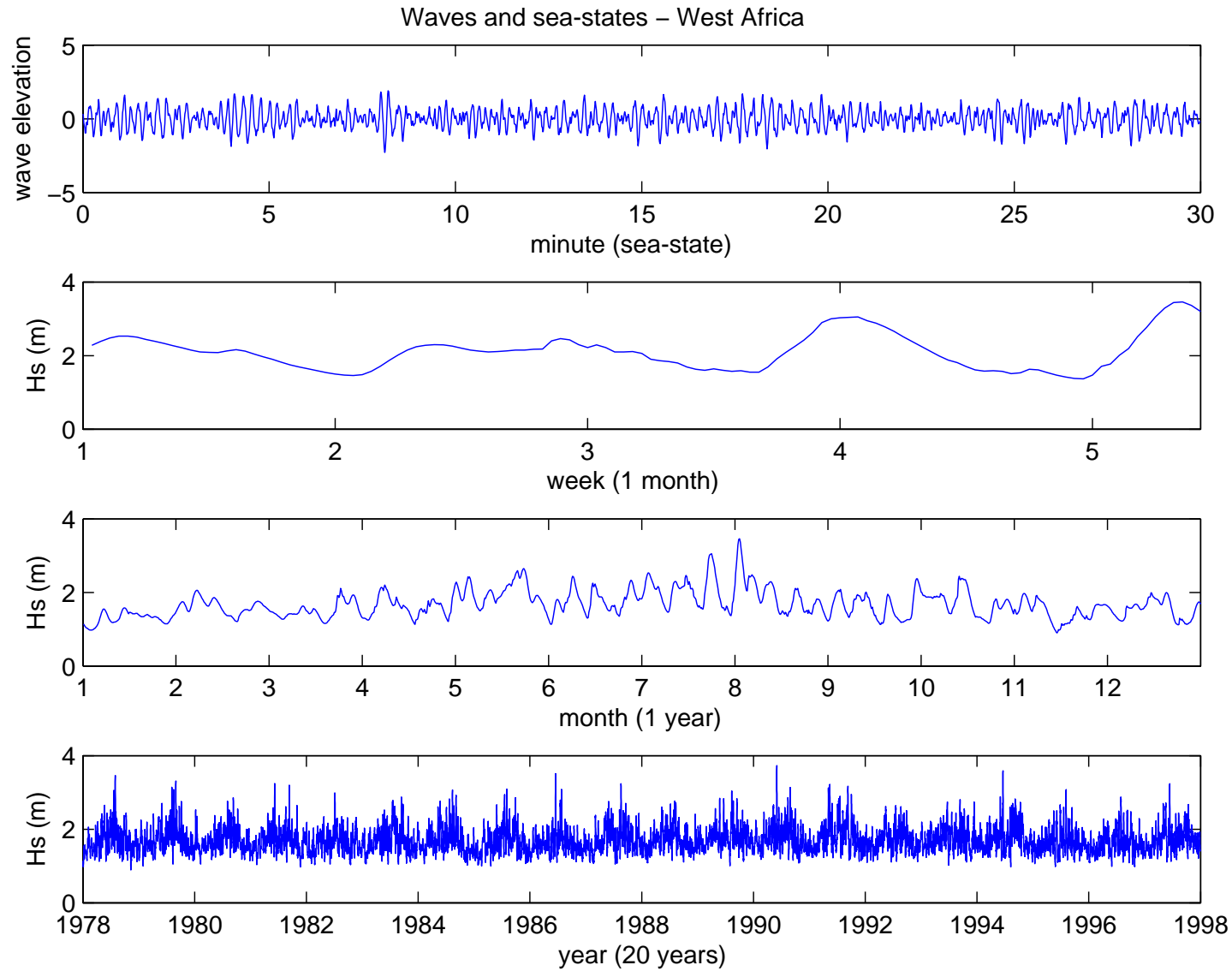
- **Combined effects of wind, waves, current**
- **Complex nonlinear dynamic behavior**
- **Design for extreme conditions**

Introduction of statistics and probabilities

- Random phenomena: wind, waves, current -> statistics/climatologies
- Time evolution (short/medium/long term-> process, dependence
- Simultaneity -> joint extremes
- Design -> probability of failure $P(X_{\max}^{\text{life}} > x_r)$

FIGURE 1. Scales for metocean conditions

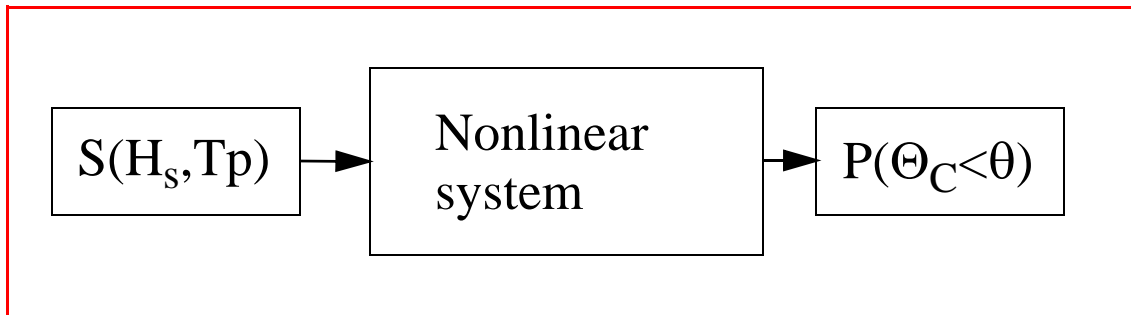




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Probability of failure $P(X_{\max}^{\text{life}} > x_r) < 10^{-?}$

Probabilistic step (short term):



Probabilistic step

Decomposition in stationary state (sea-states ~3hours)

Statistical step (long term):



$$P(S \leq s) = \int P(S \leq s | \Sigma = \sigma) p_{\Sigma}(\sigma) d\sigma \quad (1)$$

$$P(X_{\max}^{\text{sea-state}} \leq x) = \int P(X_{\max}^{\text{sea-state}} \leq x | H_S) p_{H_S}(h) dh \quad (2)$$

$$P(X_{\max}^{\text{life}} \leq x) = P(X_{\max}^{\text{sea-state}} \leq x)^{\# \text{sea-state/life}} \quad (3)$$

- H_s of the climatology $\ll H_s$ of failure \rightarrow extrapolation
- H_s dependence \rightarrow extremal index

Probabilistic step (short term):

- Rice series (factorial moments)

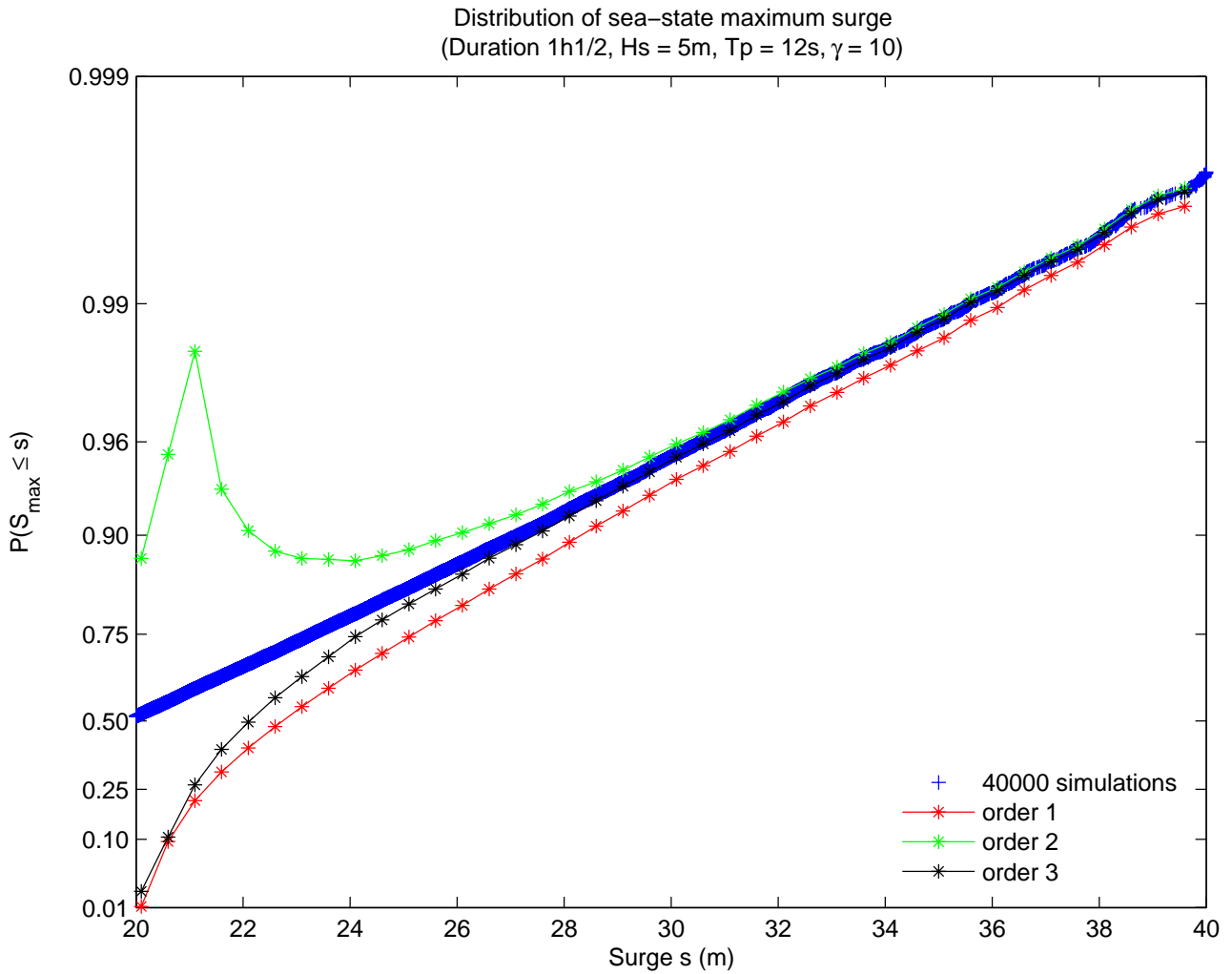
$$\begin{aligned} P(X_{\max}^{\text{sea-state}} > x) &= P(X_{t_0} > x) + P(N \geq 1 | X_{t_0} \leq x) \\ &= P(X_{t_0} > x) + P(N = 1) + P(N = 2) + P(N = 3) + \dots \\ &= P(X_{t_0} > x) + \sum_{n=1}^{\infty} \frac{(-1)^{n+1}}{n!} \mathbb{E}(N(N-1)\dots(N-n+1)) \end{aligned}$$

Rem. Azais, Wschebor -> Results for X Gaussian process

- Davies bound (x high, sea-state of short duration)

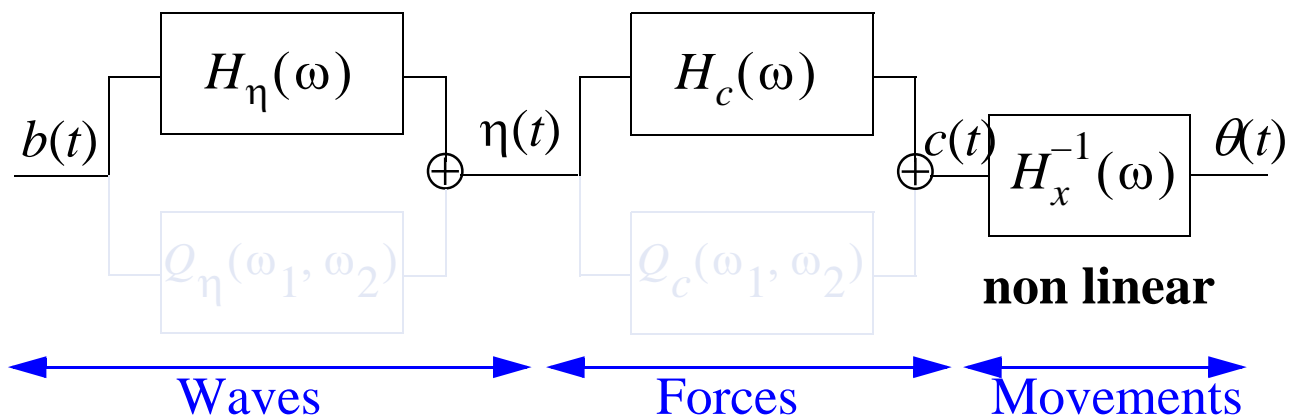
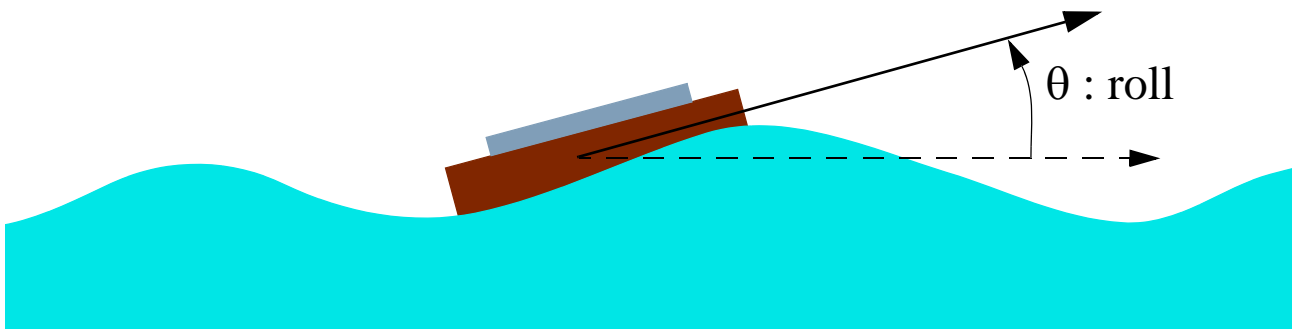
$$P(X_{\max}^{\text{sea-states}} > x) = \mathbb{E}(N) \quad (5)$$

to be verified for all sea-states participating in the long term step



Gumbel plot, 30m = 4 * standard-deviation

Extreme roll



Model:

$$I\ddot{\theta}(t) + T\dot{\theta}(t)|\dot{\theta}(t)| + K\theta(t) = C(t) \quad (1)$$

Excitation of roll, moment:

$$C(t) = h_c(t) \otimes \eta(t) \quad (2)$$

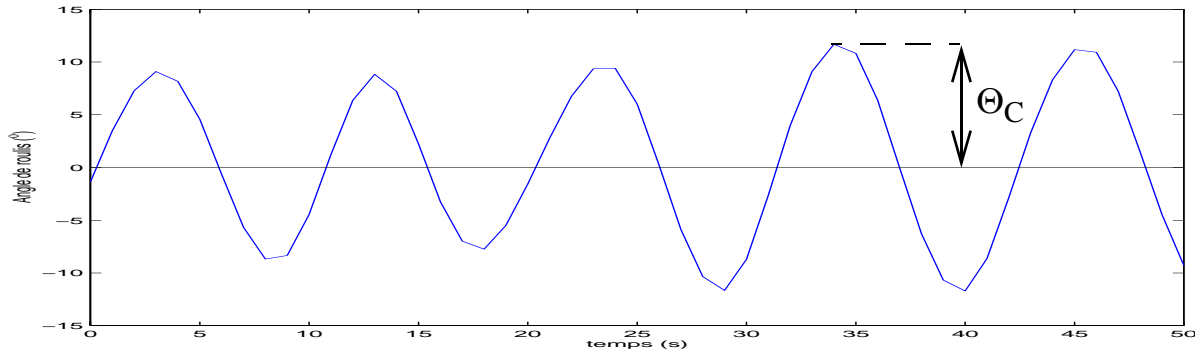
Wave elevation:

$$\eta(t) = h_\eta(t) \otimes b(t) \quad \text{with } b(t) \text{ Gaussian white noise} \quad (3)$$

and $|H(\omega)|^2 = S_\eta(\omega)$

Distribution of maxima :

$$P(\Theta_C < \theta) \quad (4)$$



Narrow band process:

$$P(\Theta_C > \theta) = \int_{\theta}^{\infty} f_{\Theta_C}(\theta) d\theta = \frac{\mathbb{E}(N_{\theta})}{\mathbb{E}(N_0)} = \frac{\mu^+(\theta)T}{\mu^+(0)T} \quad (5)$$

Rice, stationary case:

$$\mu^+(\theta) = \int_0^{\infty} \dot{\theta} f_{\Theta \dot{\Theta}}(\theta, \dot{\theta}) d\dot{\theta} \quad (6)$$

if joint distribution is difficult to obtain:

- **Independence hypothesis:**

$$\mu^+(\theta) = f_{\Theta}(\theta) \int_0^{\infty} \dot{\theta} f_{\dot{\Theta}}(\dot{\theta}) d\dot{\theta} \quad (7)$$

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- **Projection method (perturbation of independence)**

$$f_{\Theta\dot{\Theta}}(\theta, \dot{\theta}) = f_{\Theta}(\theta)f_{\dot{\Theta}}(\dot{\theta})s(\theta, \dot{\theta}) \quad (8)$$

- **Edgeworth, Gram-Charlier (Gaussian perturbation)**

Marginal and joint probability laws

- **Simulation (Monte Carlo)**
- **Equivalent Linearisation**
- **Volterra, Wiener kernel**
- **Fokker-Planck equation**
 - **Exact solution**
 - **Equivalent linear system**
 - **Stochastic averaging**
 - **Numeric (FE)**
- **Linearize&Match [Armand,Duthoit]**

Restricting hypotheses

- **White noise input**

$$I\ddot{\theta}(t) + T\dot{\theta}(t)|\dot{\theta}(t)| + K\theta(t) = C(t) \quad (9)$$

replaced by:

$$\begin{aligned} I\ddot{\theta}(t) + T\dot{\theta}(t)|\dot{\theta}(t)| + K\theta(t) - C(t) &= 0 \\ \ddot{C}(t) + 2\xi_{\eta}\omega_{\eta}\dot{C}(t) + \omega_{\eta}^2 C(t) &= B(t) \end{aligned} \quad (10)$$

- **Damping in $\dot{\theta}(t)|\dot{\theta}(t)|$**

$$I\ddot{\theta}(t) + T\dot{\theta}(t)|\dot{\theta}(t)| + K\theta(t) = C(t) \quad (11)$$

can be replaced by:

$$I\ddot{\theta}(t) + \alpha\dot{\theta}(t) + \Xi\dot{\theta}(t)^3 + K\theta(t) = C(t) \quad (12)$$

- **Weak damping**

Variations of $S_{\eta}(\omega)$ are very small in the frequency band of the response.

Input can be considered as a white noise.

Linearize&Match

- **Equivalent linear system**

$$I\ddot{\theta}(t) + T\dot{\theta}(t)|\dot{\theta}(t)| + K\theta(t) = C(t) \quad (13)$$

equivalent linear system:

$$I\ddot{\theta}(t) + \alpha_L\dot{\theta}(t) + K\theta(t) = C(t) \quad (14)$$

minimize for α_L

$$IE(\varepsilon^2), \text{ with } \varepsilon = T\dot{\theta}(t)|\dot{\theta}(t)| - \alpha_L\dot{\theta}(t) \quad (15)$$

one uses a Gaussian closure on $\dot{\theta}$:

$$\alpha_L = T \frac{4}{\sqrt{2\pi}} IE(\dot{\theta}^2)^{1/2} \quad (16)$$

- **Cubic equivalent system**

$$I\ddot{\theta}(t) + T\dot{\theta}(t)|\dot{\theta}(t)| + K\theta(t) = C(t) \quad (17)$$

replaced by:

$$I\ddot{\theta}(t) + \alpha\dot{\theta}(t) + \Xi\dot{\theta}(t)^3 + K\theta(t) = C(t) \quad (18)$$

$$\alpha = T \sqrt{\frac{2 IE(\dot{\theta}^2)}{\pi}}, \quad \Xi = T \sqrt{\frac{2}{9\pi IE(\dot{\theta}^2)}} \quad (19)$$

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- **Sequence of linear equivalent systems**

$$I\ddot{\theta}_i(t) + (\alpha + \beta_i \Xi \mathbb{IE}(\dot{\theta}_i^2(t)))\dot{\theta}_i(t) + K\theta_i(t) = C(t) \quad (20)$$

who verify:

$$\mathbb{IE}(\theta_i(t)^{2i}) = \mathbb{IE}(\theta(t)^{2i}) \quad (21)$$

in using a Gaussian closure.

- **β_i identification**

$$\theta = \theta_{(1)} + \theta_{(3)} + \theta_{(5)} + \dots \quad (22)$$

$$\mathbb{IE}(\theta^{2n}) = \mathbb{IE}(\theta_{(1)}^{2n}) + 2n \mathbb{IE}(\theta_{(1)}^{2n-1} \theta_{(3)}) + \mathcal{O}(\theta_{(1)}^{2(n+2)}) \quad (23)$$

$$\beta = 3 + 2(i-1) \frac{\int h(\tau) R_{\theta_{(1)} \dot{\theta}_{(1)}}^3(\tau) d\tau}{\mathbb{IE}(\dot{\theta}_{(1)}^2) \mathbb{IE}(\theta_{(1)}^2) \int h(\tau) R_{\theta_{(1)} \dot{\theta}_{(1)}}(\tau) d\tau} \quad (24)$$

- **Calculus of $\mathbb{IE}(\theta^{2i})$**

The $\mathbb{IE}(\dot{\theta}_i^2(t))$ are calculated with an iteratif scheme:

$$\mathbb{IE}(\dot{\theta}_i^2) = \int_{-\infty}^{\infty} \omega^2 \left| \frac{1}{-I\omega^2 + i\omega(\alpha + \beta_i \Xi \mathbb{IE}(\dot{\theta}_i^2)) + K} \right|^2 S_{CC}(\omega) d\omega$$

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then

$$\mathbb{E}(\theta_i^{2i}) = \int_{-\infty}^{\infty} \left| \frac{1}{-I\omega^2 + i\omega(\alpha + \beta_i \Xi \mathbb{E}(\dot{\theta}_i^2)) + K} \right|^2 S_{CC}(\omega) d\omega$$

which give the $\mathbb{E}(\theta_i^{2i})$ in using Gaussianity of the output of linear systems.

- **Roll maxima probability law**

- **Marginal law**

Maximum entropy

$$\max \left(- \int_{-\infty}^{\infty} f_{\Theta}(\theta) \log(f_{\Theta}(\theta)) d\theta \right) \quad (27)$$

with moment constraints

$$\int_{-\infty}^{\infty} f_{\Theta}(\theta) \theta^{2i} d\theta = \mathbb{E}(\theta_i^{2i}) \quad (28)$$

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- Maxima

$$P(\Theta_C > \theta) = \frac{\mu^+(\theta)}{\mu^+(0)} = \frac{\int_0^{\infty} \dot{\theta} f_{\Theta\dot{\Theta}}(\theta, \dot{\theta}) d\dot{\theta}}{\int_0^{\infty} \dot{\theta} f_{\Theta\dot{\Theta}}(0, \dot{\theta}) d\dot{\theta}} \quad (29)$$

in using

$$f_{\Theta\dot{\Theta}}(\theta, \dot{\theta}) = f_{\Theta}(\theta) f_{\dot{\Theta}}(\dot{\theta}) \quad (30)$$

or

$$f_{\Theta\dot{\Theta}}(\theta, \dot{\theta}) = f_{\Theta}(\theta) f_{\dot{\Theta}}(\dot{\theta}) s(\theta, \dot{\theta}) \quad (31)$$

where the $s(\theta, \dot{\theta})$ are estimated in introducing the joint moment $IE(\theta^2 \dot{\theta}^2)$ calculated from the fourth order linear equivalent system

$$IE(\theta^2 \dot{\theta}^2) = IE(\theta_{i=2}^2 \dot{\theta}_{i=2}^2) = IE(\theta_{i=2}^2) IE(\dot{\theta}_{i=2}^2) \quad (32)$$

• «Exact» closure

New equivalent linear systems by a closure with $f_{\dot{\Theta}}(\dot{\theta})$:

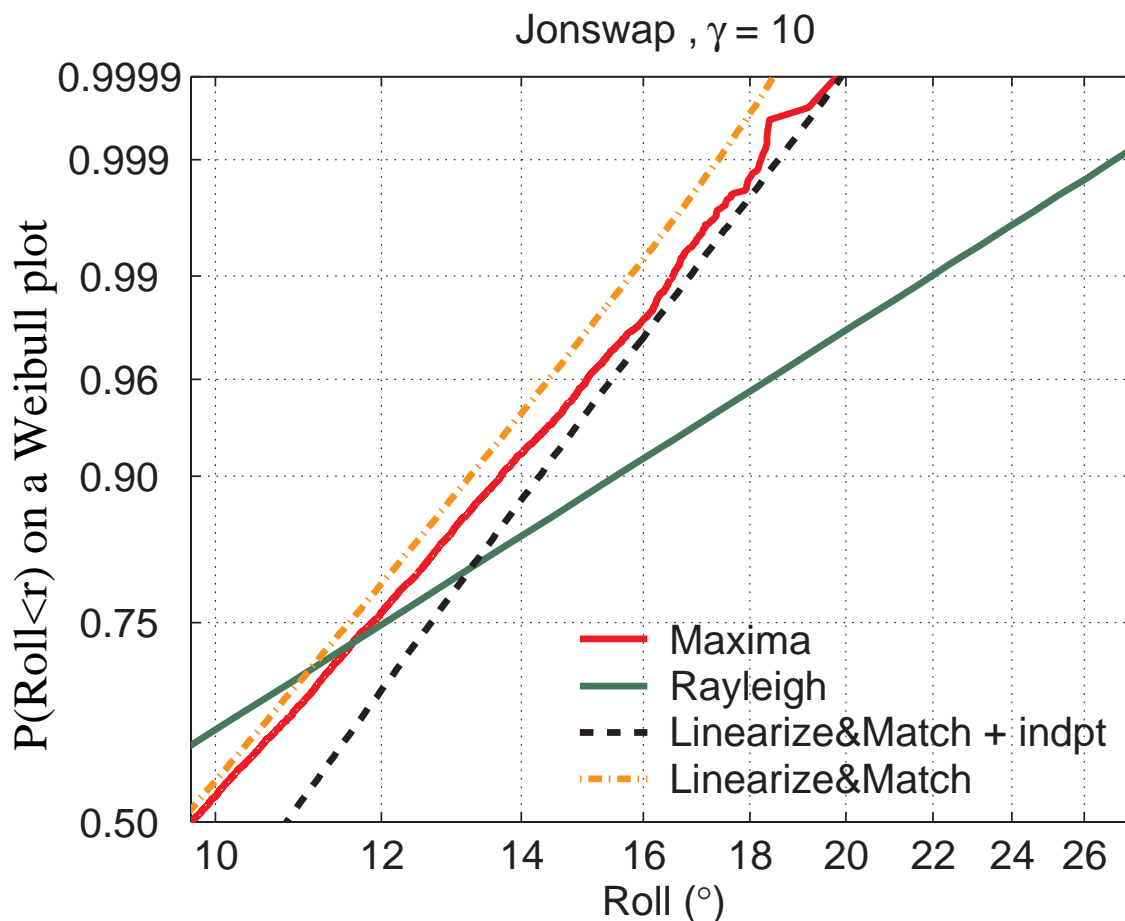
$$I\ddot{\theta}(t) + \alpha_L^{(2)} \dot{\theta}(t) + K\theta(t) = C(t) \quad (33)$$

which give new $IE(\theta_i^2)$ et $IE(\dot{\theta}_i^2)$, until convergence (fast).

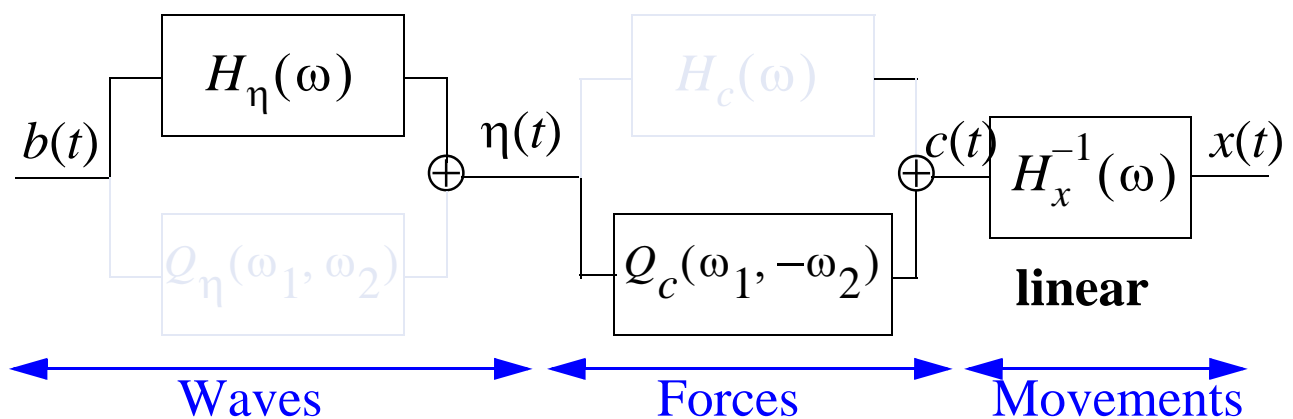
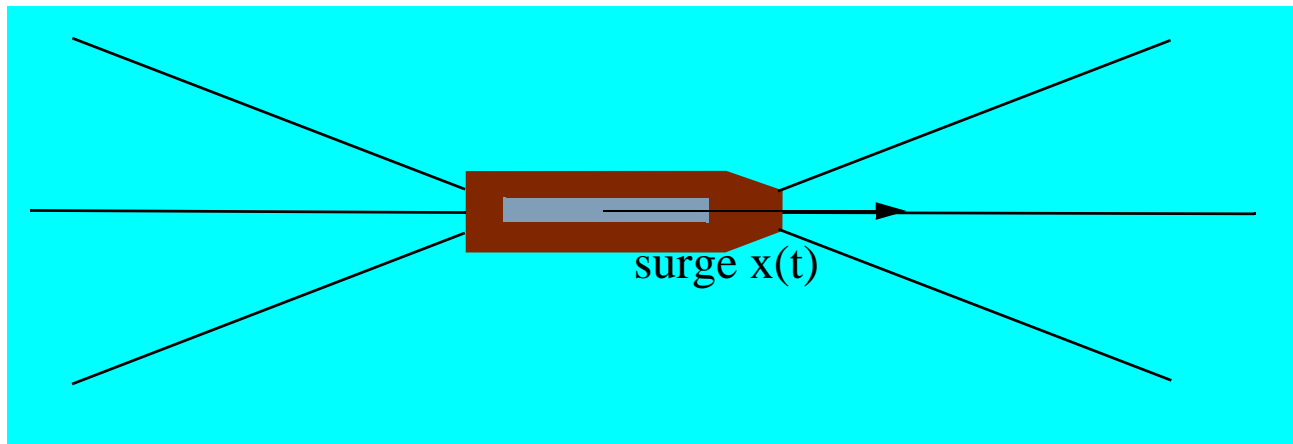
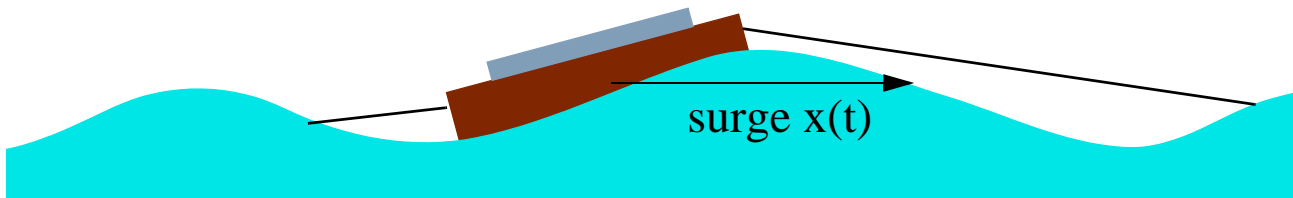
Moments Linearize&Match

“Exacts” moments (black numbers) $\gamma = 10, T_p = T_r$
 Linearize&Match (white numbers)
 Linearize&Match iterated (blue numbers).

$\mathbb{E}(\theta^k \dot{\theta}^l)$ normalised		$l(\dot{\theta})$							
		0		2		4		6	
$k(\theta)$	0			16.53	14.65 16.41	2.02	2.16	5.43	6.14
	2	52.37	45.96 51.80	0.69	0.72				
	4	2.06	2.14						
	6	5.66	6.18						



Extremes of low frequency movements



Global transfer between $b(t)$ and $x(t)$:

$$Q_T(\omega_1, -\omega_2) = H_\eta(\omega_1)H_\eta(-\omega_2)Q_c(\omega_1, -\omega_2)H_x^{-1}(\omega_1 - \omega_2)$$

i.e.:

$$x(t) = 2 \int_0^\infty \int_0^\infty Q_T(\omega_1, -\omega_2) B(\omega_1) e^{j\omega_1 t} \overline{B(\omega_2)} e^{-j\omega_2 t} d\omega_1 d\omega_2 \quad (35)$$

Eigen-decomposition of Q_T :

$$x(t) = \frac{1}{2} \sum_{i=1, n} \lambda_i (z_i(t)^2 + \tilde{z}_i(t)^2)$$

with $z_i(t)$ and $\tilde{z}_i(t)$ standard Gaussian processes (36)

$\tilde{z}_i(t)$ Hilbert transform of $z_i(t)$

$$\text{and } \mathbf{IE}(z_i z_j) = \mathbf{IE}(z_i \tilde{z}_i) = \mathbf{IE}(\tilde{z}_i \tilde{z}_j) = \mathbf{IE}(z_i \tilde{z}_j) = 0$$

• Calculus of $\mathbf{IE}(N_\beta)$

- Rice

$$\mu^+(\beta) = \int_0^\infty \dot{x} f_{x\dot{x}}(\beta, \dot{x}) d\dot{x} \quad (37)$$

Gaussian hypothesis, independence, projection method (with moments calculated from $x(t)$, EQ 36).

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- Lindgren (1980)

$$x(t) - \beta = \frac{1}{2} \sum_{i=1, n} \lambda_i (z_i(t)^2 + \tilde{z}_i(t)^2) - \beta = G(\beta) = 0 \quad (38)$$

$$\mu(\beta) = \frac{\beta^{n-1}}{(2\pi)^{n/2}} \int_{\partial G} \mathbb{E}(\|n(z)^T \dot{z}(t)\| \mid z(t) = \beta x) e^{-\frac{\beta}{2} x^T x} ds(x)$$

Breitung proposed une asymptotic approximation valid if the number of points of $G(\beta)$ at a minimal distance of 0 is finite.

- Hagberg (2004)

Particular case of quadratic forms

$$\sum_{i=1, n} \gamma_i w_i(t)^2 \quad (40)$$

$$\text{and } \gamma_1 \leq \dots \leq \gamma_k < \gamma_{k+1} = \dots = \gamma_n \quad (41)$$

$$\mu(\beta^2) = A(\beta) \int_{r \in S_{n-k-1}} I(r, \beta) ds(r) \quad (42)$$

$$I(r, \beta) = \int_{s^T \Gamma_k s < \beta^2} Q(s, r, \beta) e^{-\frac{1}{2} s^T (I_k - \Gamma_k) s} ds \quad (43)$$

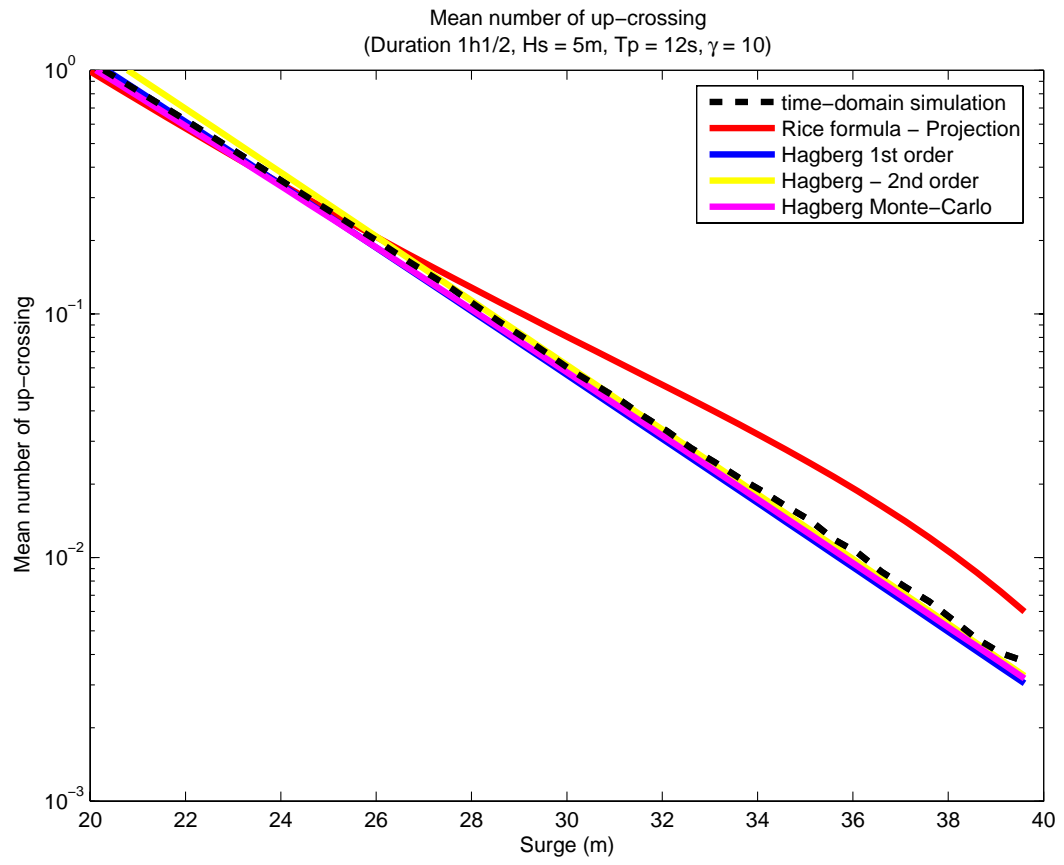
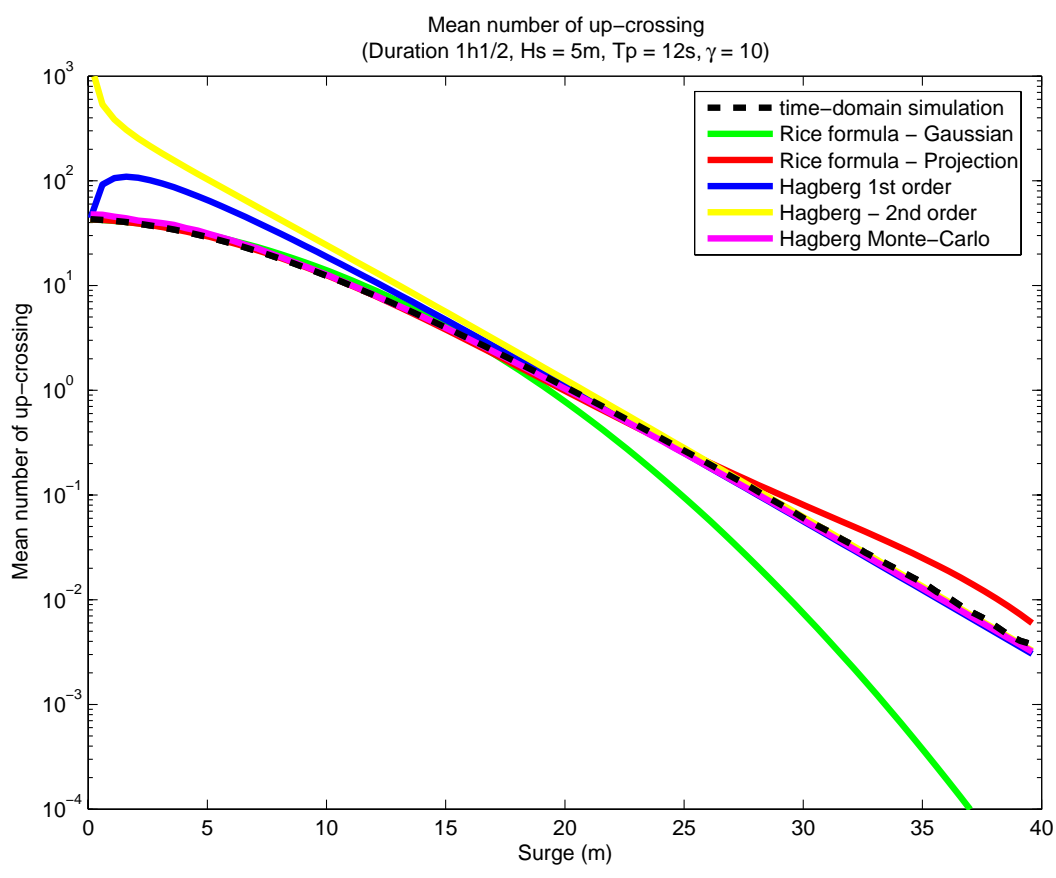
- asymptotic development

$$\mu(\beta^2) = A(\beta) \sum_j c_j \frac{1}{\beta^{2j}}, \quad c_0, c_1 \quad (44)$$

- Monte-Carlo integration



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Improvements

- Rice series, non Gaussian case
- Comparisons with higher damping (non Gaussian vicinity)
- Addition of a linear part:

$$x(t) = \frac{1}{2} \sum_{i=1, n} \lambda_i (z_i(t)^2 + \tilde{z}_i(t)^2) + \sum_{i=1, n} \gamma_i z_i(t)$$

- $\mathbf{IE}(N)$ for more complex systems

