

# Extreme Wave Crests in Space and Time

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Different parts of this work with Prof. Rychlik and Dr. Hagberg

## Sea Models

Sea surface elevation is modelled by means of

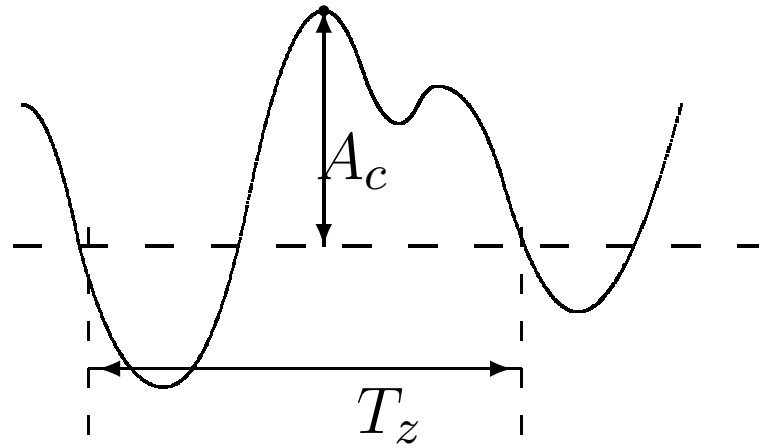
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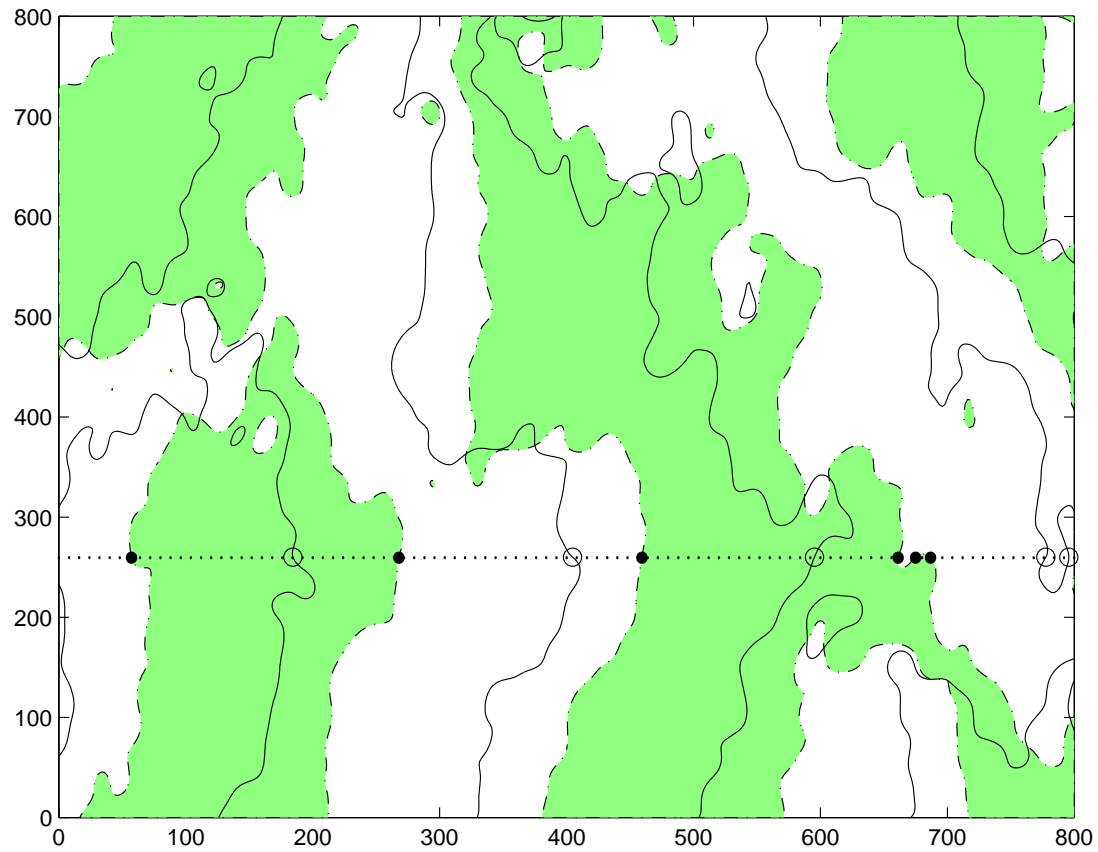
- Linear model,  $W(\boldsymbol{\tau}) = W(x, y, t)$  - Gaussian random field that is homogeneous (invariant under translation but not rotation)
- Second order model,  $W(\boldsymbol{\tau}) = W_l(\boldsymbol{\tau}) + W_q(\boldsymbol{\tau})$  - the linear field plus a quadratic correction term

## Wave characteristics in time



$T_z$ : wave period;  $A_c$ : crest height

# Wave characteristics in space



## Spatio-temporal wave characteristics

- $H_s = 4\sqrt{\lambda_{000}}$ , significant wave height
- $T_z = 2\pi\sqrt{\frac{\lambda_{000}}{\lambda_{002}}}$ , average zero down-crossing wave period
- $L_x = 2\pi\sqrt{\frac{\lambda_{000}}{\lambda_{200}}}$ , average zero down-crossing wave length along  $x$ -axis
- $L_y = 2\pi\sqrt{\frac{\lambda_{000}}{\lambda_{020}}}$ , average zero down-crossing wave length along  $y$ -axis
- $N = \frac{X}{L_X} \frac{Y}{L_Y} \frac{T}{L_T}$ ,  $N_T = \frac{X}{L_X} \frac{Y}{L_Y}$ ,  $N_X = \frac{Y}{L_Y} \frac{T}{L_T}$ ,  $N_Y = \frac{X}{L_X} \frac{T}{L_T}$

# Spatio-temporal wave characteristics

- $V_{drift} = (V_x, V_y) = (L_x/T_z, L_y/T_z) = \left( \sqrt{\frac{\lambda_{002}}{\lambda_{200}}}, \sqrt{\frac{\lambda_{002}}{\lambda_{020}}} \right)$

- $\mathbf{v}_{pr} = (v_x, v_y) = \left( -\frac{\lambda_{101}}{\lambda_{200}}, -\frac{\lambda_{011}}{\lambda_{020}} \right)$

- $\alpha_{xt} = -\frac{v_x}{V_x} = \frac{\lambda_{101}}{\sqrt{\lambda_{200}\lambda_{002}}}$  and  $\alpha_{yt} = -\frac{v_y}{V_y} = \frac{\lambda_{011}}{\sqrt{\lambda_{020}\lambda_{002}}}$

- $\alpha_{xt}, \alpha_{yt} \in [-1, 1]$

- $\alpha_{xt}^2 \approx 1$  for narrow-band sea

- $|\alpha_{xt}| = 1$  for a sea drifting along  $x$ -axis

- $\alpha_{xt} = 0$  and  $\alpha_{yt} = 0$  for confused sea (no organised movement of waves)

## Wave crest height and wave velocities

$$\begin{aligned} P\left(\max_{\boldsymbol{\tau} \in \mathbf{S}} W(\boldsymbol{\tau}) \geq u\right) &\leq \left(\frac{X}{L_x} + \frac{Y}{L_y} + \frac{T}{T_z}\right) e^{-\frac{8u^2}{H_s^2}} \\ &+ \sqrt{2\pi} \left(N_t + N_y \sqrt{1 - \alpha_{xt}^2} + N_x \sqrt{1 - \alpha_{yt}^2}\right) \frac{4u}{H_s} e^{-\frac{8u^2}{H_s^2}} \\ &+ 2\pi N \sqrt{1 - \alpha_{xt}^2 - \alpha_{yt}^2} \frac{16u^2}{H_s^2} e^{-\frac{8u^2}{H_s^2}} \end{aligned}$$

with  $\mathbf{S} = [0, X] \times [0, Y] \times [0, T]$



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with  $\mathbf{S} = [0, X] \times [0, Y] \times [0, T]$

• For  $X = Y = 0$ ,

$$P\left(\max_{t \in [0, T]} W(t) \geq u\right) \leq \frac{T}{T_z} e^{-\frac{8u^2}{H_s^2}}$$

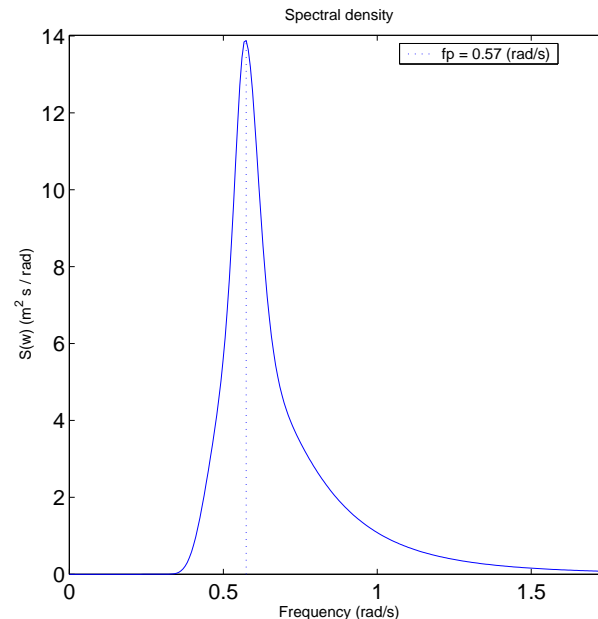
## Wave characteristics for spreading functions.

Area:  $100m \times 100m$  for 10 s

$D(\omega, \theta)$	$L_x$	$L_y$	$T_z$	$\alpha_{xt}$	$v_x$
box	124.87	154.45	8.7949	0.9228	-13.1023
cos2s	120.54	164.22	8.7988	0.89	-12.1936
sech2	117.62	172.45	8.7988	0.8751	-11.699
misses	117.25	173.63	8.7988	0.8658	-11.5381
poisson	110.89	199.68	8.7882	0.8202	-10.3492

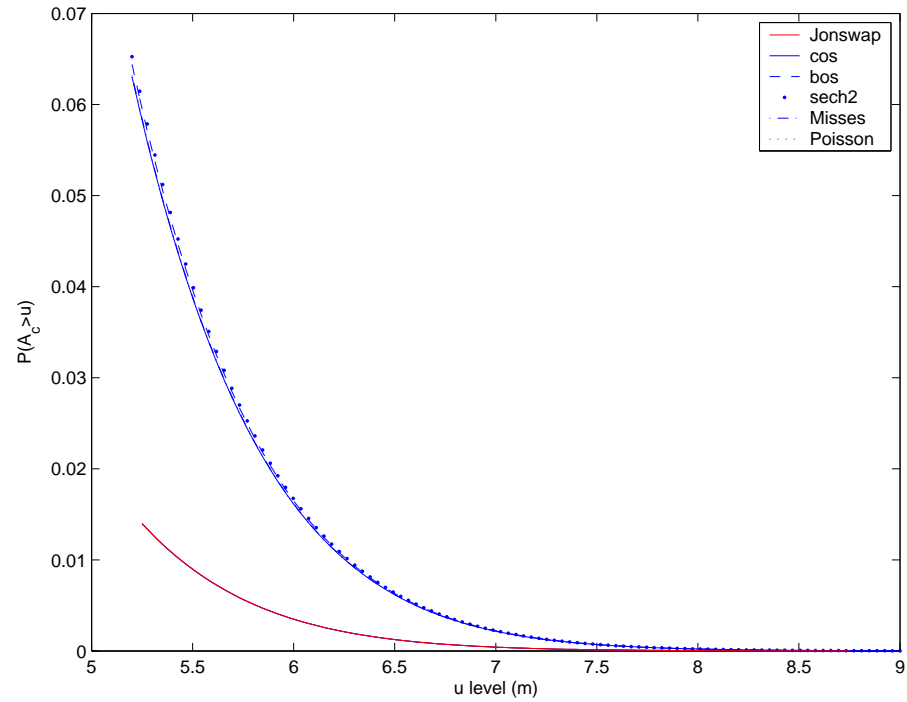
# Linear Sea model - Gaussian field

- $S(\omega, \theta) = S(\omega)D(\omega, \theta)$
- $S(\omega)$  *frequency spectrum* usually JONSWAP with  $H_s = 7m$   $T_p = 11s$  and  $\gamma = 2.3853$
- $D(\omega, \theta)$ , *spreading function*, like box,  $\cos^2$ ,  $\text{sech}^2$ , Misses, Poisson

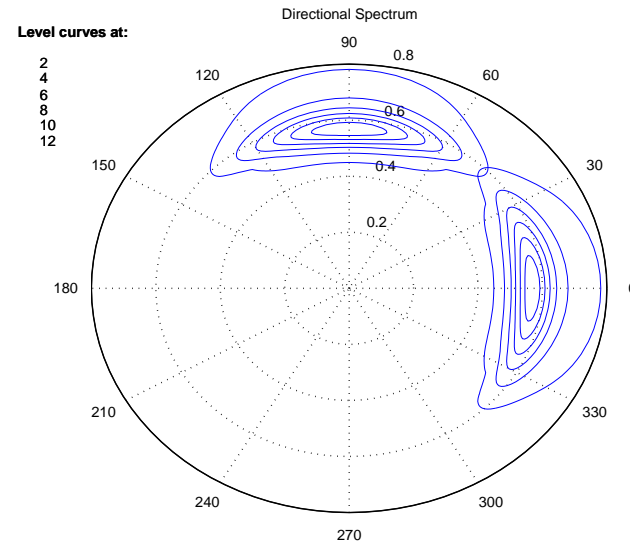
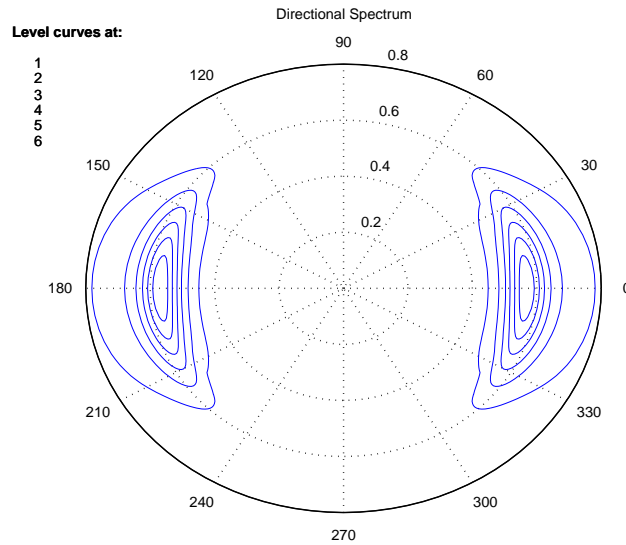


# Wave Crest in a Gaussian Sea

$100m \times 100m \times 10s$

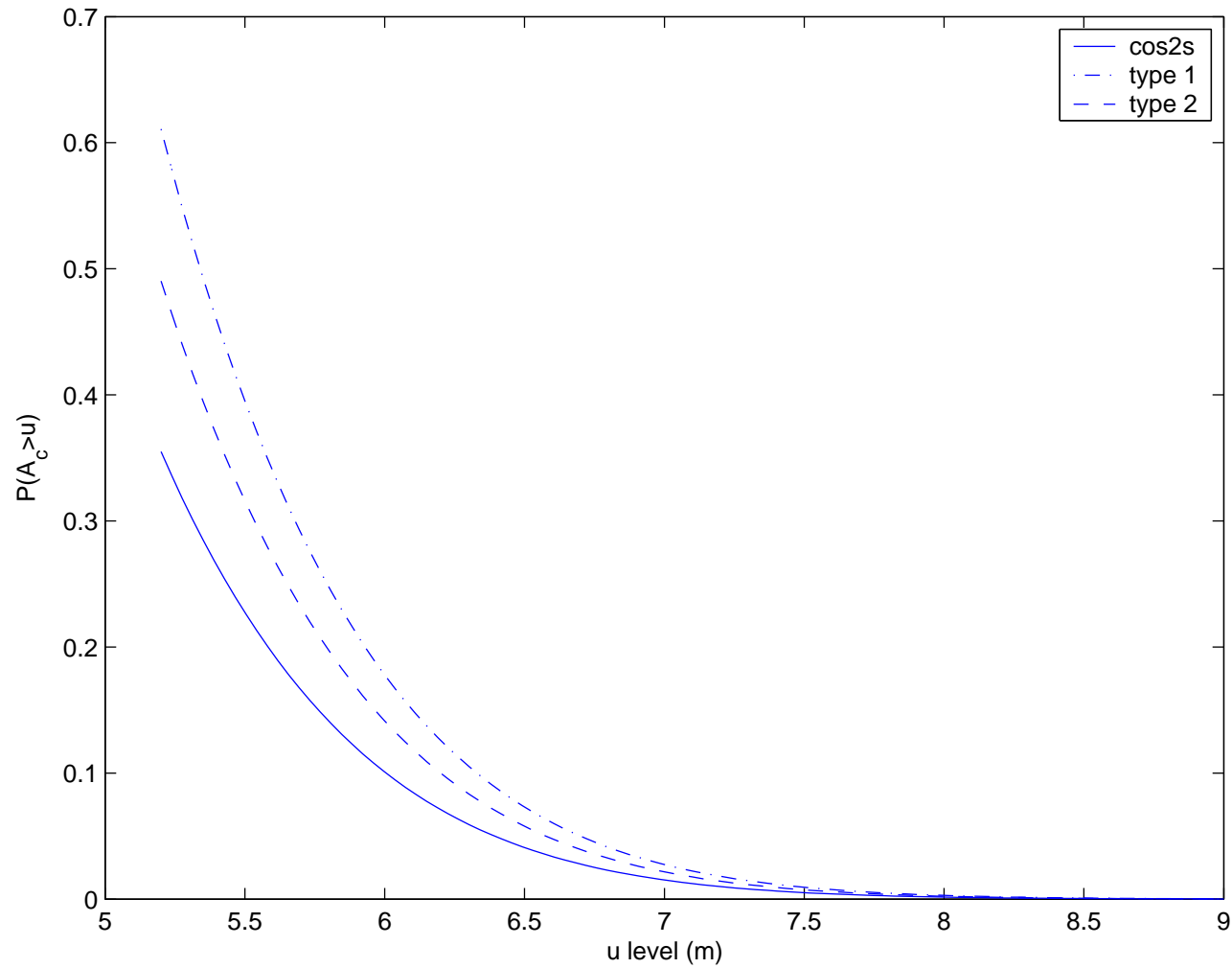


# Sea spectra



spectra	$L_x$	$L_y$	$\alpha_{xt}$	$\alpha_{yt}$	$v_x$	$v_y$
cos2s	120.54	164.22	0.89	0	-12.19	0
type 1	120.54	164.22	0	0	0	0
type 2	137.42	137.42	0.51	0.51	-7.92	-7.92

# Effect of directionality to distribution of high crests



# Freak Waves

Freak wave:  $A_c \geq 0.72 \times 2.1H_s$

- Buoy: For  $u = 1.5H_s \Rightarrow \frac{1}{\mu^+(u)} \approx 18.5$  years and if  $T_z \approx 10s \Rightarrow 60$  mil. waves
- Area  $5km \times 10km$ ,  $P_{xy}(u) = 0.00058 \Rightarrow \frac{1}{P_{xy}(u)} \approx 1723$  pictures or  $\approx 4.5$  mil. waves
- Area  $5km \times 10km$  for 37 min  $\Rightarrow \frac{1}{P_{xyt}(u)} \approx 1 \Rightarrow \approx 640000$  waves

## Conclusions

- Seas with different spreading functions resulted to the same global maximum distribution
- confused seas = highest waves (no organised movement)
- drifting seas (narrow-band) = waves moving along one direction with high speeds = lower extreme crest heights



## Problem

Find  $h^{crt}$  so that

$$P(\max_{0 \leq t \leq T} X(t) > h^{crt}) = p_0$$

- $X(t)$  - sea surface at time  $t$  as a second order sea
- $T = 1$  year
- $p_0 = 10^{-4}$ , then  $h^{crt}$  is called the 10 000 year wave crest

## Review of methods

When  $p_0$  is small,  $h^{crt}$  takes large values  $\Rightarrow$

- Statistical methods (i.e. Peaks Over Threshold (POT) method or yearly maxima) require crest height measurements over large periods of time

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- Statistical methods (i.e. Peaks Over Threshold (POT) method or yearly maxima) require crest height measurements over large periods of time
- Mathematical models: assume independence of wave crests, see Forristall (2000), Prevosto et al. (2000) and Krogstad and Barstow (2004)

## Rice method

Let  $N_T^+(h)$  be the number of upcrossings of the level  $h$  by the process  $X(t)$  during the time period  $[0, T]$ .

Then for any fixed time  $t_0 \in [0, T]$

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- $P(\max_{0 \leq t \leq T} X(t) > h) \leq E[N_T^+(h)]$
- Problem is equivalent to computing  $E[N_T^+(h)]$

## Computation of $E[N_T^+(h)]$

$$E[N_T^+(h)] = \int_0^T \mu_t^+(h) dt = \int_0^T \int_0^{+\infty} z f_{X(t), \dot{X}(t)}(h, z) dz dt$$

- The density  $f_{X(t), \dot{X}(t)}(h, z)$  includes two sources of variability
  - Variable sizes of sea waves during a sea state  
 $S_t := S_t(\omega, \theta)$  for fixed  $t$
  - Evolution of sea states with  $t$

## Computation of $E[N_T^+(h)]$

$S_t$ : random sequence of sea states. Then

$$\int_0^{+\infty} z f_{X(t), \dot{X}(t)}(h, z) dz = E\left[\int_0^{+\infty} z f_{X(t), \dot{X}(t)|S_t}(h, z) dz\right]$$

- Change rate of  $\{S_t\}$  much slower than of sea elevation, so we approximate  $f_{X(t), \dot{X}(t)|S_t}(h, z)$  by that of a second order sea with  $S_t$

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- $\int_0^{+\infty} z f_{X(t), \dot{X}(t)|S_t}(h, z) dz \approx \mu^+(h|S_t)$



## Estimation of $\mu^+(h|S)$

In the case of a stationary process and for fixed  $t$

$$\mu^+(h|S_t) = \int_0^{+\infty} z f_{X(t), \dot{X}(t)|S_t}(h, z) dz,$$

where  $X(t)$  is a second order sea

- saddle point method
- Breitung method:  $\mu^+(h|S) \approx c(\beta_h) \exp(-\beta_h^2/2)$  where  $\beta_h$  is the Hasofer-Linds safety index

## Evaluation of $E[N_T^+(h)]$

- $\{S_t\} \equiv (H_s(t), T_z(t)) \Rightarrow \mu^+(h|S_t) = \mu^+(h|H_s(t), T_z(t))$

Long term distribution means we look at a sea state chosen at random

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Hence

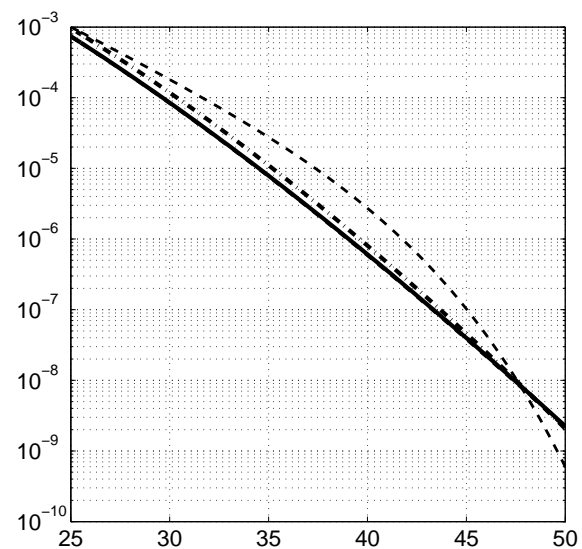
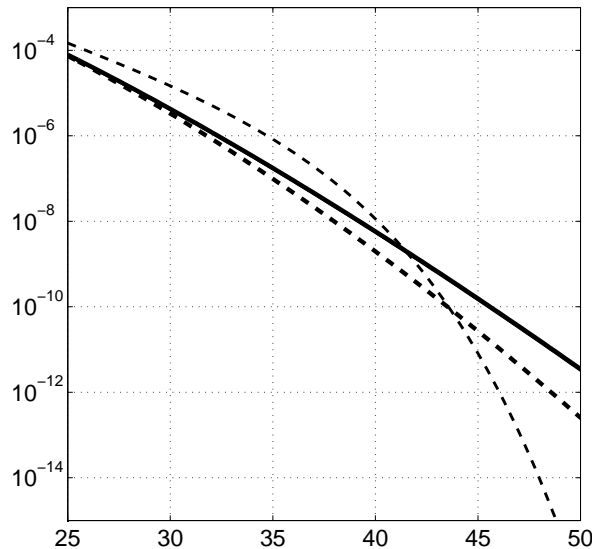
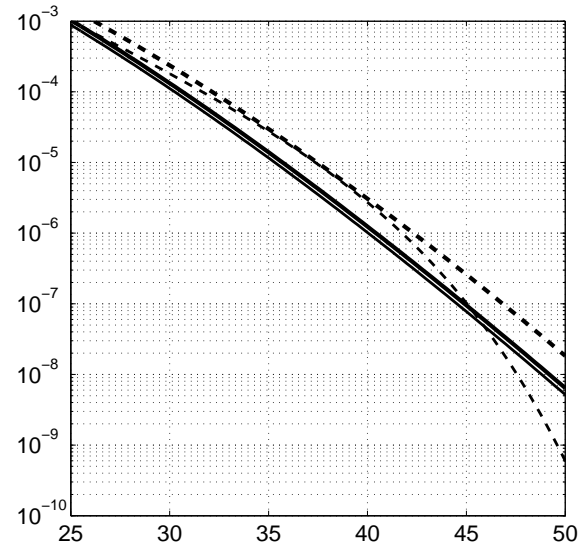
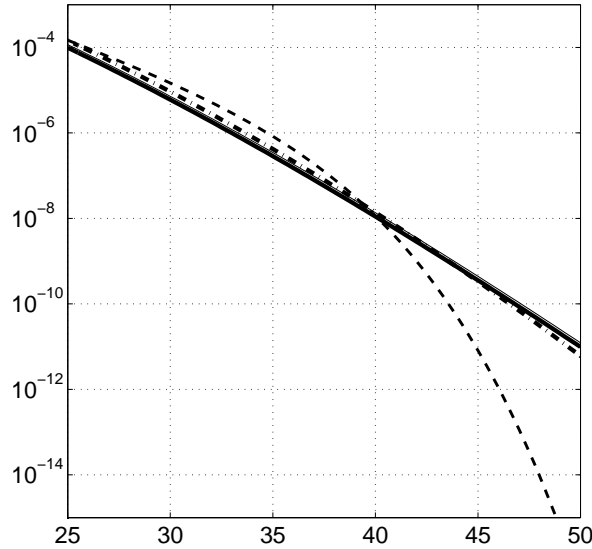
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Then
- $\mu^+(h) = \int \mu^+(h|h_s, t_z) f(h_s, t_z) dh_s dt_z$   
Hence
- $E[N_T^+(h)] \approx T\mu^+(h)$

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# Comparison of short term distribution



Solid: FORM and SORM, light dashed: Dawson, thick dashed: Farristall model

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## References

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- Baxevani, A., Hagberg, O. and Rychlik, I. (2005). Note on the distribution of extreme wave crests, *Proceedings of OMAE 2005*.
- Forristall, G. Z., (2000). Wave crest distributions: Observations and second order theory, *Journal of Physical Oceanography*, Vol. 30, pp. 329-360.
- Krogstad, H. E., and Barstow, S., (2004). Analysis and applications of second-order models for maximum crest height, *Journal of Offshore Meachanics and Arctic Engineering*, Vol. 126, pp. 66-71.



## References

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## Breitung method

Let  $g : \mathbb{R}^n \rightarrow \mathbb{R}$  be s.t.  $S = \{\mathbf{x} = (x_1, \dots, x_n); g(\mathbf{x}) = 0\}$  has a unique point  $\mathbf{x}_0$  of minimum distance from the origin.

Suppose  $\mathbf{Z}(t)$  is an  $n$ -dimensional, stationary, differentiable, Gaussian vector process, and  $\dot{\mathbf{Z}}(t)$  its derivative. Then for  $g(\mathbf{Z}(t)/\beta)$ ,  $\beta > 0$ , under some mild technical assumptions, we have:

$$\mu_{\beta}^{+}(0) = \frac{e^{-\beta^2/2}}{2\pi} (c + O(\beta^{-2})), \quad c = \sqrt{\frac{\mathbf{x}_0^T (\Sigma_{22} - \Sigma_{21} G_0 \Sigma_{12}) \mathbf{x}_0}{\det(I + P_0 G_0 P_0)}}$$

as  $\beta \rightarrow \infty$  where  $G_0 := \frac{1}{|\nabla g(\mathbf{x}_0)|} \left[ \frac{\partial^2 g}{\partial x_i \partial x_j}(\mathbf{x}_0) \right]_{i,j=1,2,\dots,n}$

$P_0 := I - \mathbf{x}_0 \mathbf{x}_0^T$  and  $\Sigma$  is the covariance matrix of  $(\mathbf{Z}(t), \dot{\mathbf{Z}}(t))$ .