Triggers of Rogue Waves in Deep Water Envelope Equations

Will Cousins

Massachusetts Institute of Technology

Joint work with:

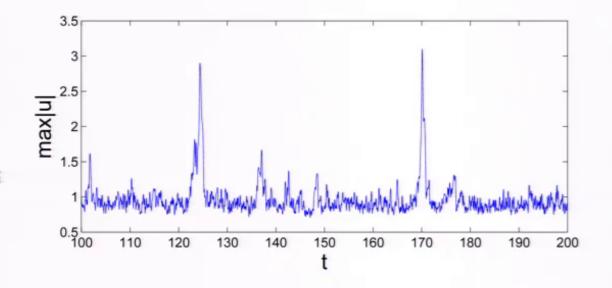
Themistoklis Sapsis, Mustafa Mohamad (MIT)
Amin Chabchoub (Swinburne University of Technology)





Extreme Events

- Dramatically atypical system responses
- Rare—live in the tails of the probability distribution
- Examples:
 - Freak/Rogue Ocean Waves
 - Financial market crashes
 - Extreme weather events (hurricanes, droughts, earthquakes)



High Dimensional, Nonlinear, Noisy Systems

Why should you care?

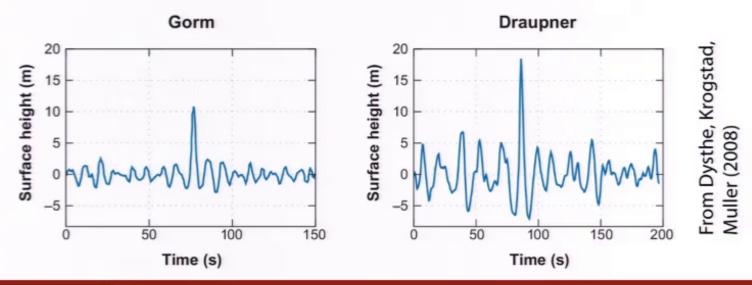
- Many systems demand such models, particularly when you want to represent extreme events
- Nonlinearity can generate non-Gaussian, heavy-tailed statistics (extreme events thus more likely)
- Combination of high dimensionality and nonlinearity make analysis challenging

Extreme Event Triggers

- We consider systems with intermittent, nonlinear, large magnitude events
- In some cases, it is possible to describe the triggers of these extreme events
- We will show you how you can use this information to perform
 - 1. Prediction
 - 2. Density Estimation (of the state and functionals thereof)

Rogue/Freak Ocean Waves

- Waves whose height is extremely large for a given wave field
- Common definition: > 4 times the standard deviation
- Such waves have caused catastrophic damage to people, ships, and coastal structures





Oceanic Rogue Waves

Rogue Waves – large elevation relative to a given sea state





Modified Nonlinear Schrodinger Equation

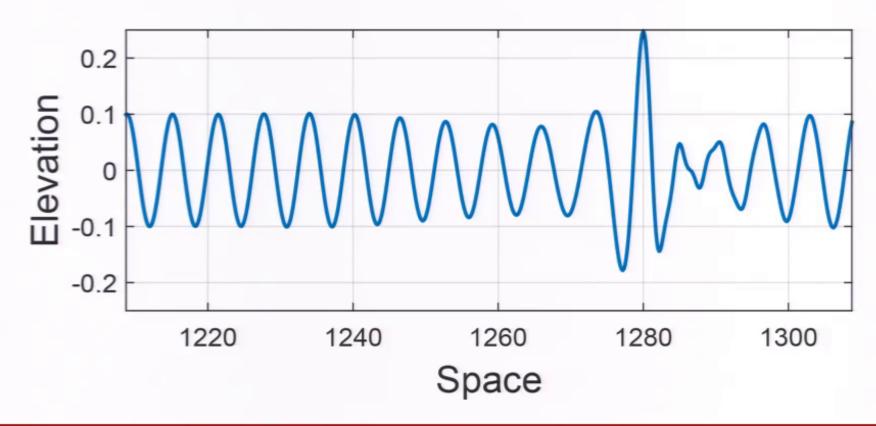
Dysthe derived closed equation for wave envelope u via perturbation method [Dysthe 1979]

Assumptions: Small steepness and slow varying u

$$0 = \frac{\partial u}{\partial t} + L(\partial_x, \partial_y) + \frac{i}{2}|u|^2 u + \frac{3}{2}|u|^2 \frac{\partial u}{\partial x} + \frac{1}{4}u^2 \frac{\partial u^*}{\partial x} + iu \frac{\partial \phi}{\partial x}\Big|_{z=0}$$

Note: Extreme waves are steep and vary quickly in *late* stages of their evolution, but less so in early stages. Envelope equations reproduce wave-tank extreme waves well [Chabchoub 2011]

Instabilities and Heavy Tails



WILL COUSINS



Instabilities and Heavy Tails

- Real wave fields aren't infinitesimally perturbed plane waves—energy distributed over range of frequencies
- Ex: Gaussian Spectrum:

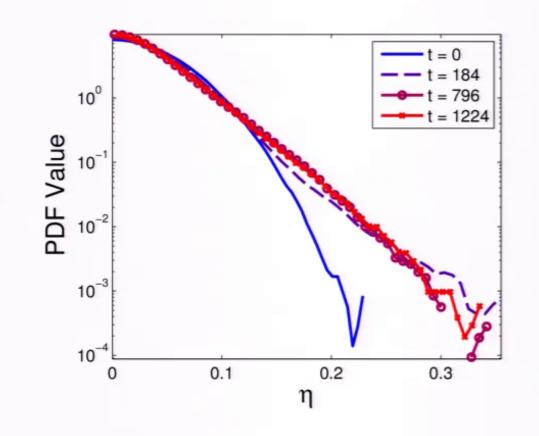
$$u(x,0) = \sum_{-N/2+1}^{N/2} \sqrt{2\Delta_k F(k\Delta_k)} e^{i(\omega_k x + \xi_k)}, \qquad F(k) = \frac{\epsilon^2}{\sigma\sqrt{2\pi}} e^{\frac{-k^2}{2\sigma^2}}$$

- ξ_k are random phases, meaning u(x,0) is normally distributed
- ε gives the magnitude of the envelope (wave steepness), σ determines the width of the power spectrum

Instabilities and Heavy Tails

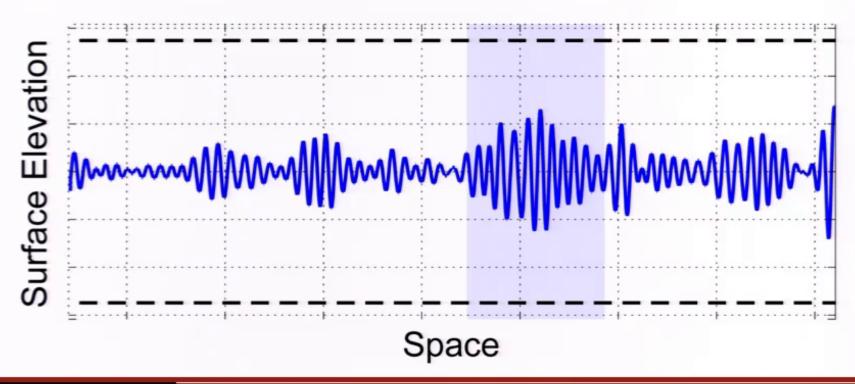
- Nonlinear interactions generate:
 - Energy transfer between modes
 - 2. Non-stationary statistics—heavy tails

[Alber 1978, Janssen 2003]



Focusing Wave Groups Trigger Rogue Waves

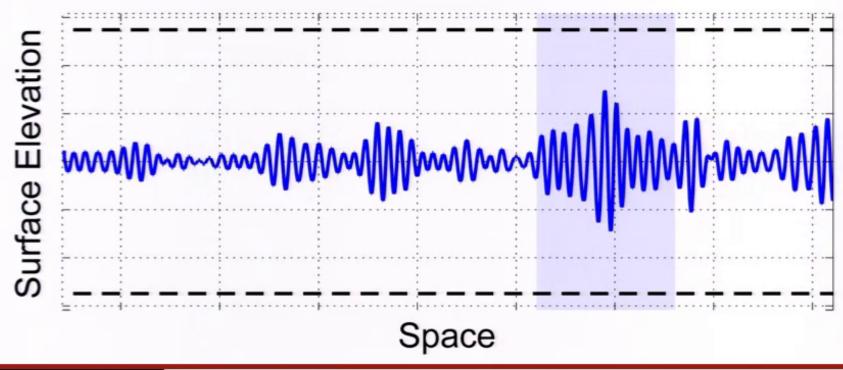
Rogue waves are triggered by focusing of benign-looking wave groups





Focusing Wave Groups Trigger Rogue Waves

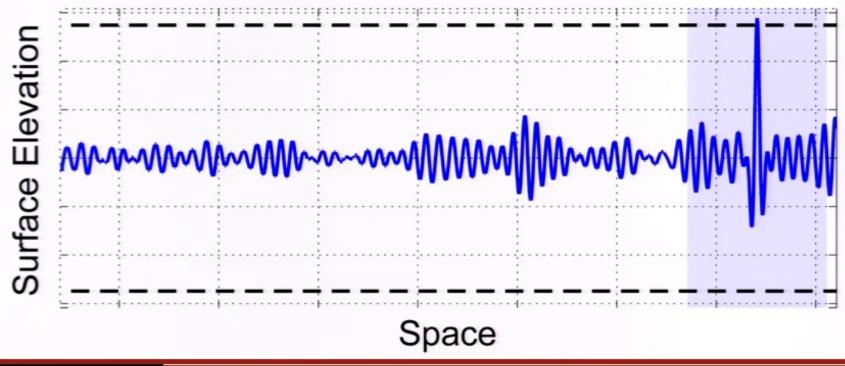
Rogue waves are triggered by focusing of benign-looking wave groups





Focusing Wave Groups Trigger Rogue Waves

Rogue waves are triggered by focusing of benign-looking wave groups

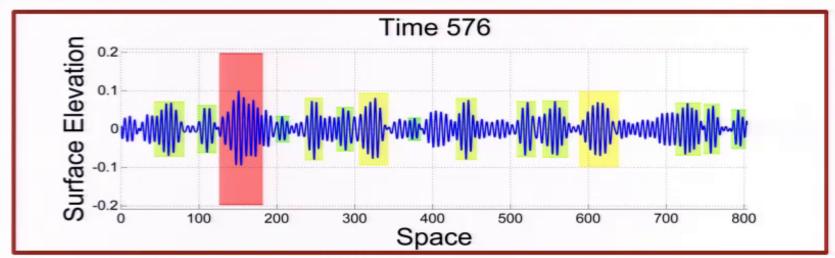




Our Simple Predictive Scheme

Our computationally cheap, group-based predictive scheme consists of two steps

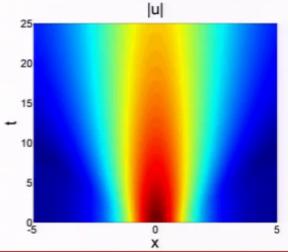
- Step 1 (offline): Quantify how a given wave group evolves as a function of its amplitude and length scale
- Step 2 (online): In an irregular wave field, pick out the groups. Use step 1 to predict future
 amplitude

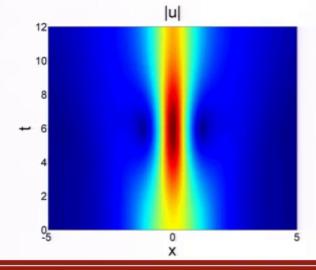


Localized Group Evolution

$$u(x,0) = A_0 \operatorname{sech}(x/L_0)$$

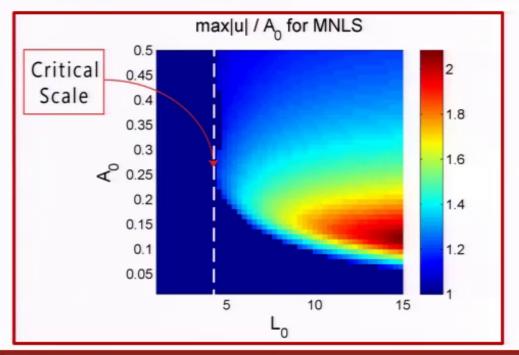
- For a given A₀ and L₀, how does this initial data evolve?
- Does group focus and increase in amplitude?
- If so, by how much? [Cousins, Sapsis 2015]





Localized Group Amplification Factors

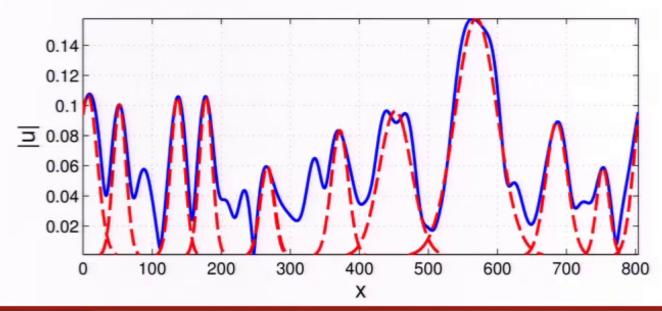
Construct a map u_{max}(A₀,L₀) using combination of offline numerical simulation and reduced order modeling





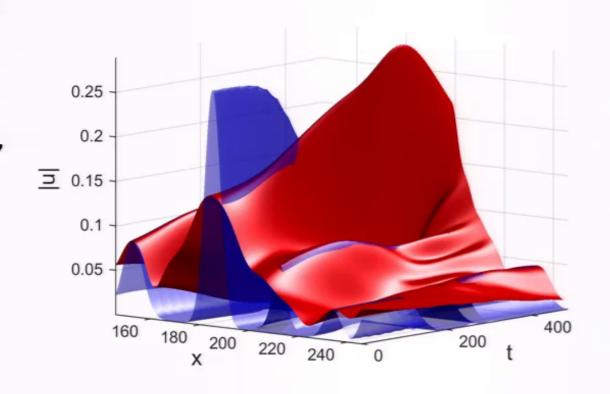
Group Detection

- Apple scale selection technique [Lindeberg 1998] to determine dominant groups in the wave field
- Gives locations, amplitude, and length scales of each group

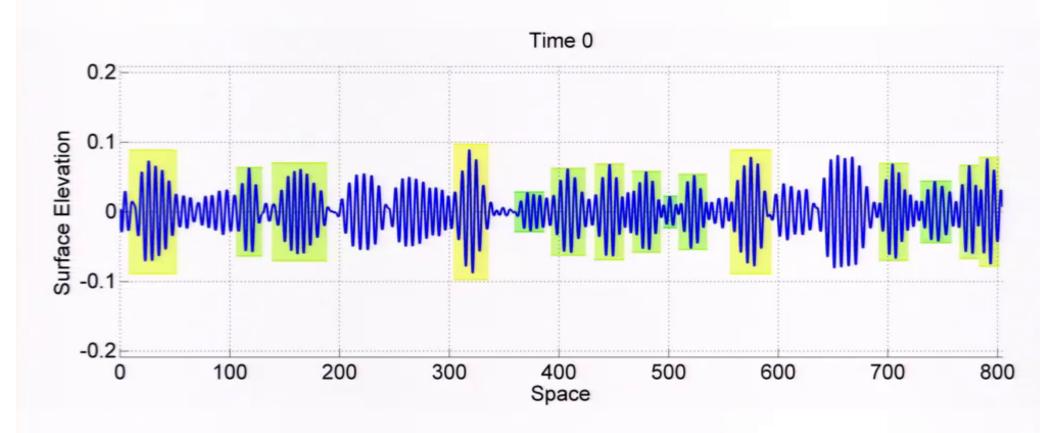


Prediction: Simulated Data

- Tested our scheme on 100 simulations of MNLS with Gaussian spectra
- 336 rogue waves occurred, predicted all of them in advance
- Average warning time: 24 wave periods
- 91 false positives (21.3%)



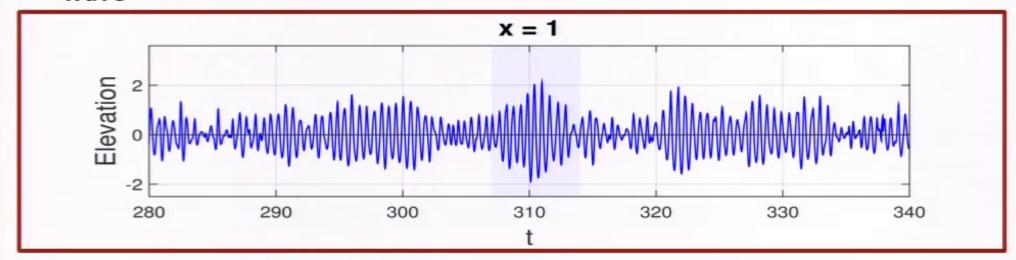
Prediction: Simulated Data





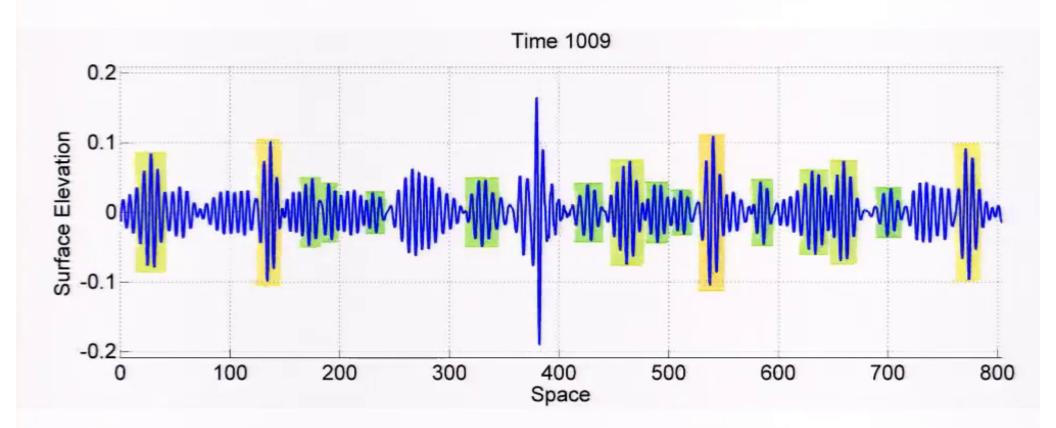
Prediction: Wave Tank Data

- We analyze data from an wave tank experiment
- Tank is 10 meters long, probes spaced 1 meter apart measure elevation time series
- Use measurements from one probe to cheaply predict downstream rogue wave



WILL COUSINS

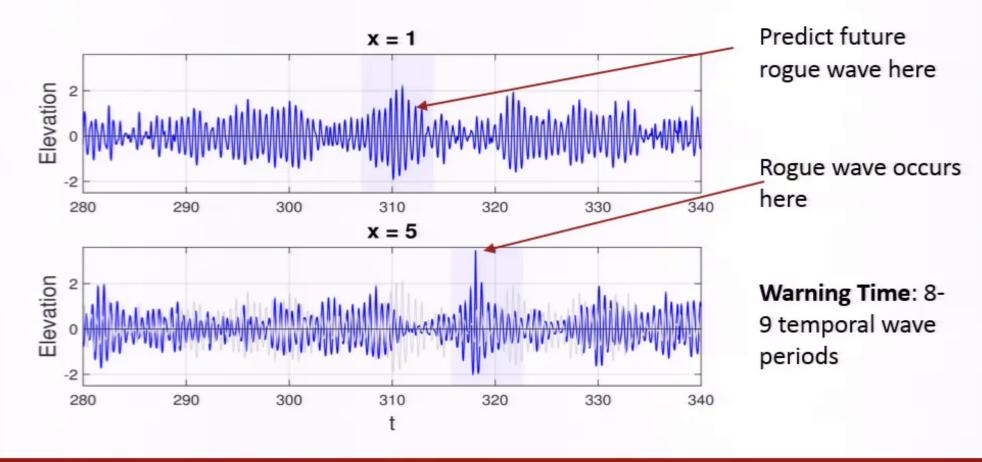
Prediction: Simulated Data



WILL COUSINS

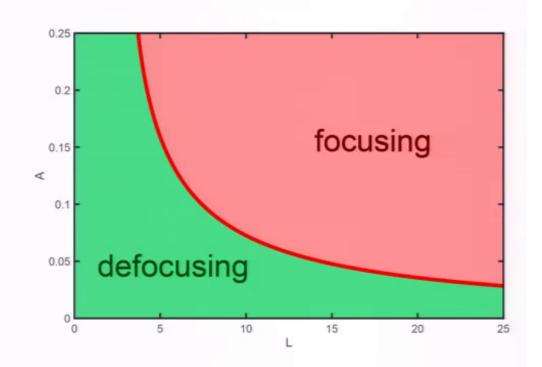


Prediction: Wave Tank Data



Estimation of Heavy Tailed Statistics

- Understanding the trigger of rogue waves allows us to predict them in advance
- We can also use this characterization of the trigger to efficiently compute statistics
- Use idea from Mohamad, Sapsis (2015): partition phase space into a stable and unstable regime



Estimation of Heavy Tailed Statistics

- u = envelope of wave elevation, let f(u) be functional of the wave field (i.e. response of ocean structure)
- We decompose the probability measure of f as follows

$$\rho(f(u)) = \rho(f(u)|\text{Foc. Group})\rho(\text{Foc. Group}) + \rho(f(u)|\text{¬Foc. Group})\rho(\text{¬Foc. Group})$$
Nonlinear/Unstable Part Linear/Stable Part



Evaluation of the Unstable Part

 $\rho(f(u)\big|\text{Foc. Group})\rho(\text{Foc. Group}) = \iint_{A,L\in\mathcal{F}} \rho(f(u)\big|A,L)\rho(A,L)dAdL$

