



Overview of Probabilistic Coastal Flooding Hazard Assessment Methods

- Norberto C. Nadal-Caraballo, PhD
- Victor M. Gonzalez, PE
- Madison O. Campbell
- Efrain Ramos-Santiago

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US Army Corps
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DISCOVER | DEVELOP | DELIVER

Outline

- Introduction
- Atmospheric forcing
- Probabilistic Coastal Flooding Hazard Assessment Methods
 - Extreme Value Analysis
 - Joint Probability Method / Probabilistic Coastal Hazard Analysis
 - ▶ TC parameterization
 - ▶ Data sources
 - ▶ Probability distributions
 - ▶ Synthetic TCs
 - ▶ Hazard Curves
- Summary
 - Logic-tree approach



MANTOLOKING, NJ



NOVEMBER 2, 2012



JULY 12, 2018



Background: Motivation

- Between 1980 and 2019, 44 tropical-cyclone billion-dollar disasters (1.1 event/year) have have affected the U.S.
- TCs are the 2nd most frequent disaster event in the U.S., after severe storms (113).
- Have caused more damage (\$945.9 billion, CPI-adjusted) than all other weather and climate disasters combined (\$808.7 billion, CPI-adjusted).
- Per event losses is also dominated by tropical cyclones (\$21.5 billion per event).

Rank	Event	Year	CPI-Adjusted Cost (Billions)	Deaths
1	Hurricane Katrina	2005	\$168.8	1,833
2	Hurricane Harvey	2017	\$130.0	89
3	Hurricane Maria	2017	\$93.6	2,981
4	Hurricane Sandy	2012	\$73.5	159
5	Hurricane Irma	2017	\$52.0	97

Introduction: Quantification of Coastal Hazards

- Coastal hazards due to tropical cyclones, extratropical cyclones, and other extreme storms: storm surge, surge + tide, wave action, currents, wind, rainfall, and flooding.
- Federal agencies and stakeholders need up-to-date, accurate, easily accessible coastal hazards information on a national scale, including uncertainty.
- Probabilistic coastal flooding hazard assessment methods:
 - Coastal storm risk management
 - Reduce vulnerability to storm damage
 - Develop strategies to enhance resiliency
- Extreme coastal storms are rare in space and time, and are not well represented in historical observation records.



Probabilistic Coastal Flooding Hazard Assessment

- Region-specific coastal hazards

Type of Atmospheric Forcing	
South Atlantic Coast Gulf of Mexico, Pacific Ocean Tropical Cyclones Extratropical Cyclone	Great Lakes Extratropical Cyclones
North Atlantic Coast Tropical and Extratropical Cyclones	Pacific Coast Extratropical Cyclones Tropical Cyclones

Probabilistic Coastal Flooding Hazard Assessment

- Standard of Practice

Probabilistic Methods

Tropical Cyclones	Extratropical Cyclones	Convective Storms; Seiches
<ul style="list-style-type: none"> Synthetic events Numerical simulation <ul style="list-style-type: none"> e.g., ADCIRC Probabilistic method <ul style="list-style-type: none"> Multivariate e.g., JPM, PCHA 	<ul style="list-style-type: none"> Observations “Pseudo-synthetics” Numerical simulation <ul style="list-style-type: none"> e.g., ADCIRC Probabilistic method <ul style="list-style-type: none"> Univariate e.g., EVA (GPD) 	<ul style="list-style-type: none"> Observations Probabilistic method <ul style="list-style-type: none"> Univariate e.g., EVA (GPD)

Probabilistic Coastal Flooding Hazard Assessment

- Hydrodynamic Models

High Certainty

ADCIRC

High Spatial
Correlation

Metamodel

Low Spatial
Correlation

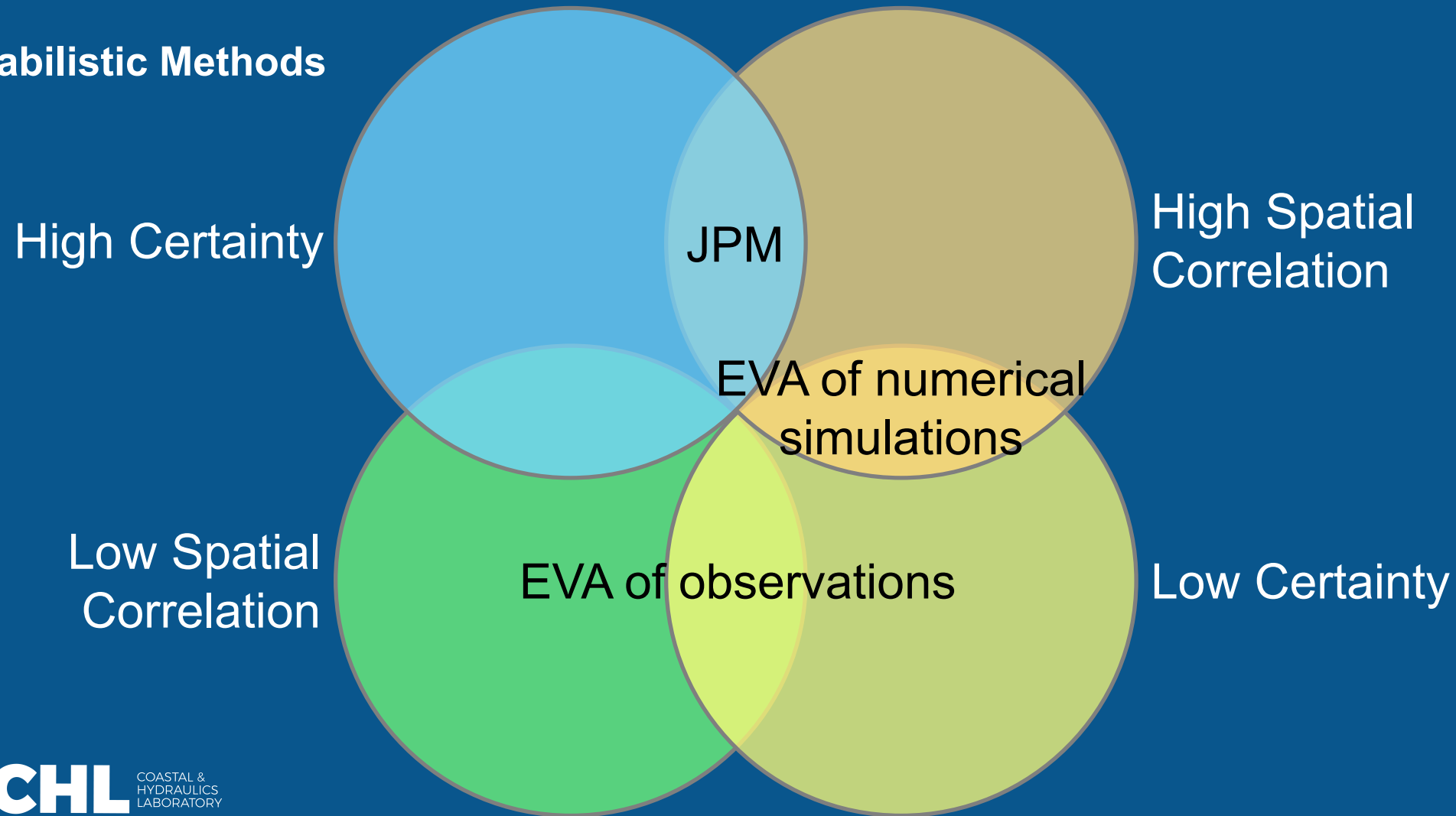
SLOSH

Low Certainty



Probabilistic Coastal Flooding Hazard Assessment

- Probabilistic Methods



Probabilistic Coastal Flooding Hazard Assessment

- **Standard of Practice: Probabilistic Methods**

Tier 1 – Low Fidelity

- **Univariate EVA of historical observations**
 - Benchmarking, high-frequency events, convective storms, seiches

Tier 2 – Medium Fidelity

- **JPM with low-fidelity numerical simulations**
 - Benchmarking
- **EVA of high-fidelity numerical simulations**
 - Extratropical cyclones

Tier 3 – High Fidelity

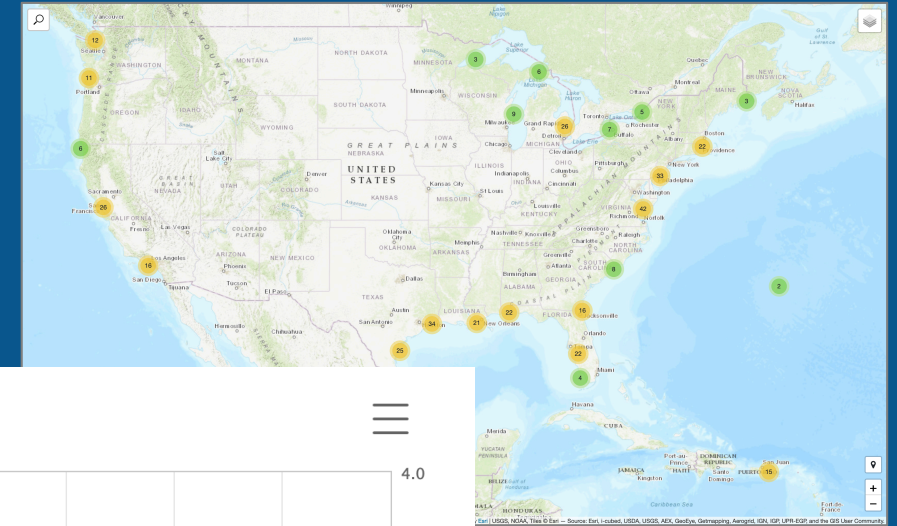
- **JPM or PCHA with high-fidelity numerical simulations (including Focus Studies)**
 - Tropical cyclones



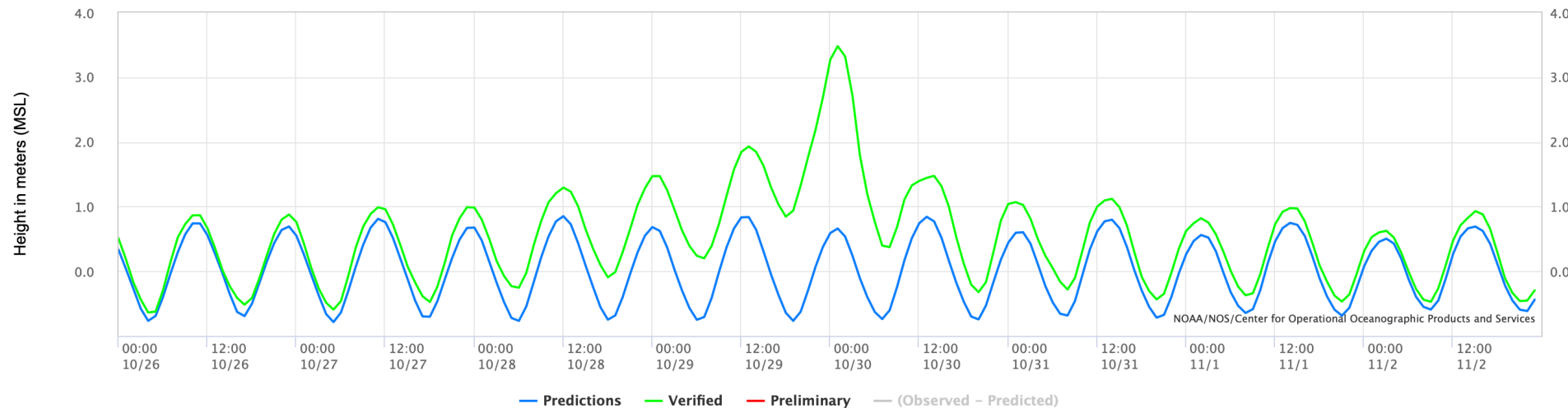
Extreme Value Analysis (EVA)

Data Sources: Water levels

- NOAA CO-OPS Tides & Currents
 - Predicted astronomical tides
 - Water level observations



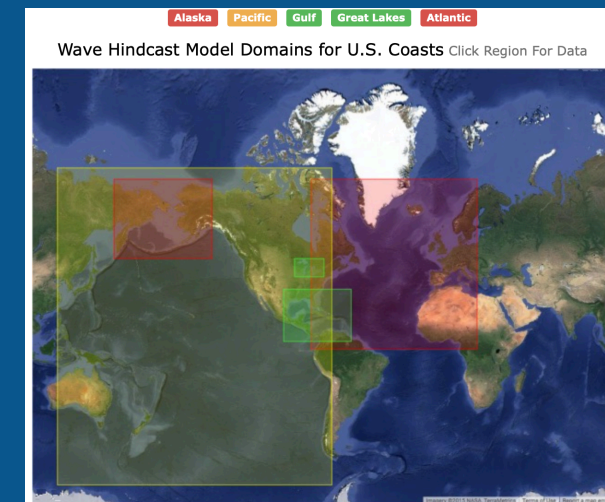
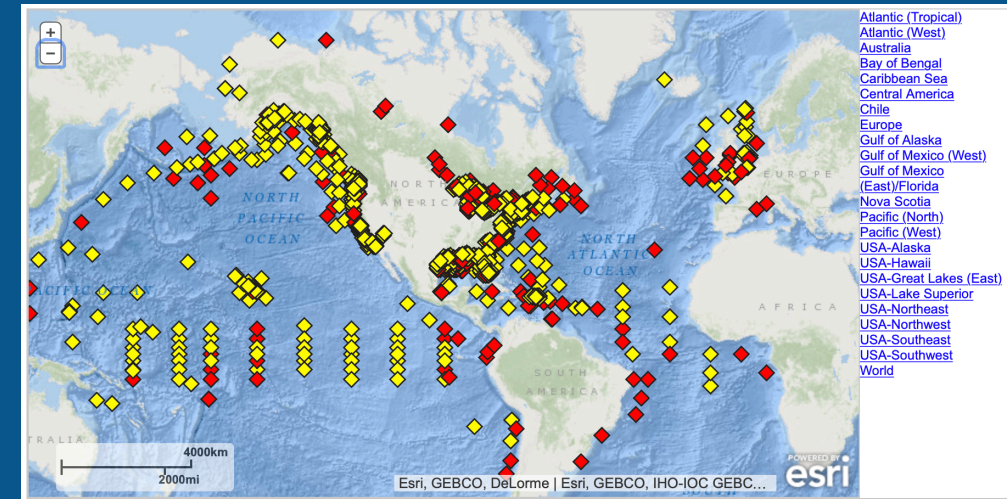
NOAA/NOS/CO-OPS
Verified Hourly Heights at 8518750, The Battery NY
From 2012/10/26 00:00 GMT to 2012/11/02 23:59 GMT



Extreme Value Analysis

Data Sources: Waves

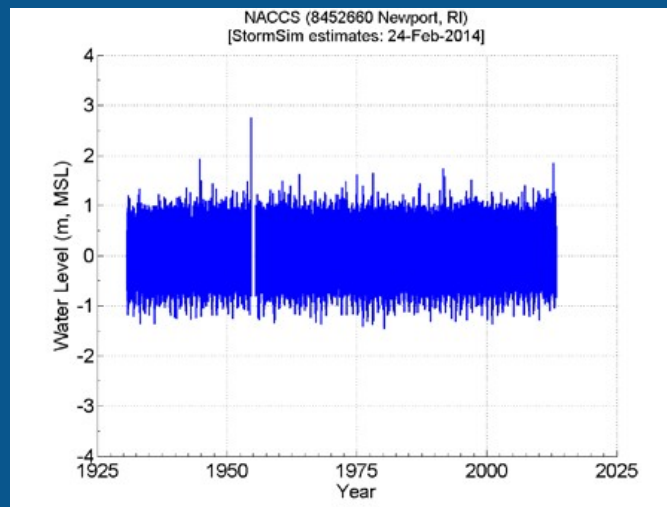
- **NOAA National Data Buoy Center**
 - Wave height, peak period, direction
 - Wave spectra
- **USACE-ERDC Wave Information Studies (WIS)**
 - Long-term hindcast (20-30 years)
 - Continuous, hourly wave climatology
 - Resolution: every few miles along the coast



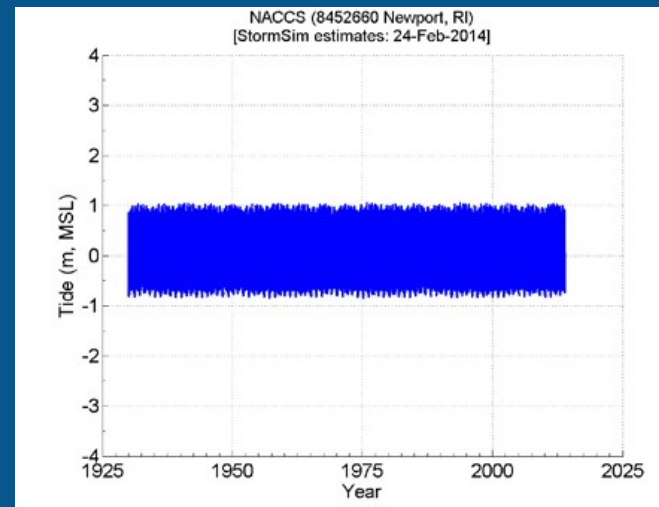
Extreme Value Analysis

Observed Water Level vs. Non-tidal Residuals (NTR)

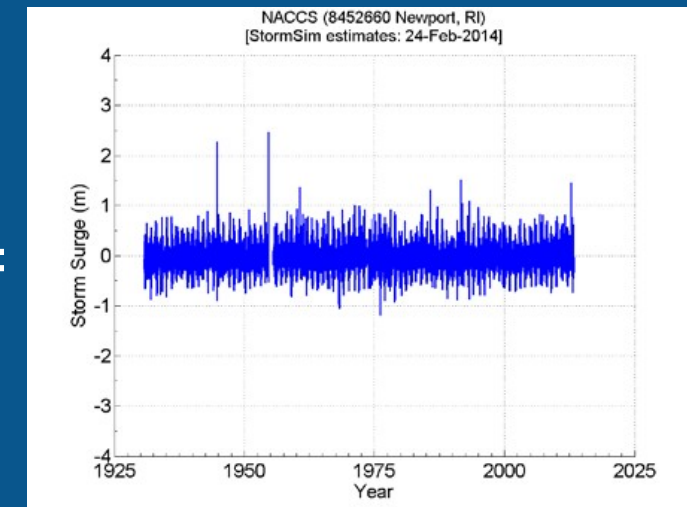
- Non-Tidal Residuals = Observed WL¹ – Predicted WL²
 1. Detrended Time Series (zero-crossing at year 1992.5)
 2. Astronomical Tide Time Series (NTDE of 1983-2001)



Observed



Predicted

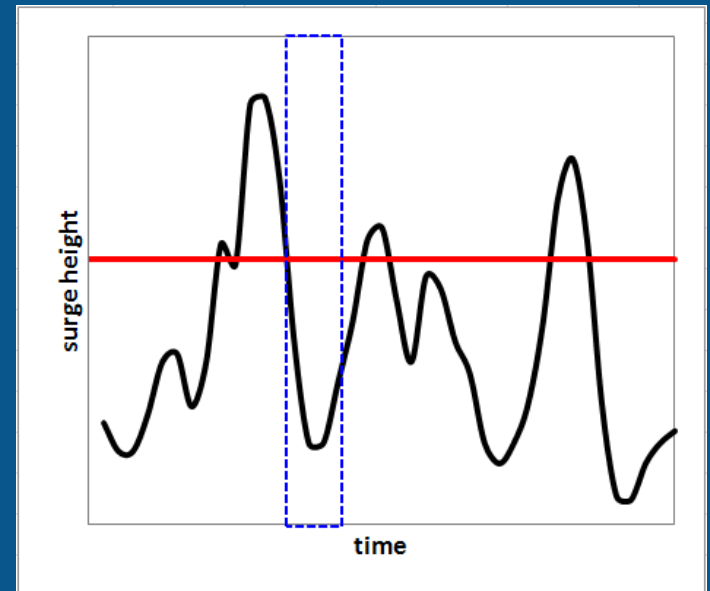


Residuals

Extreme Value Analysis

Sampling Methods

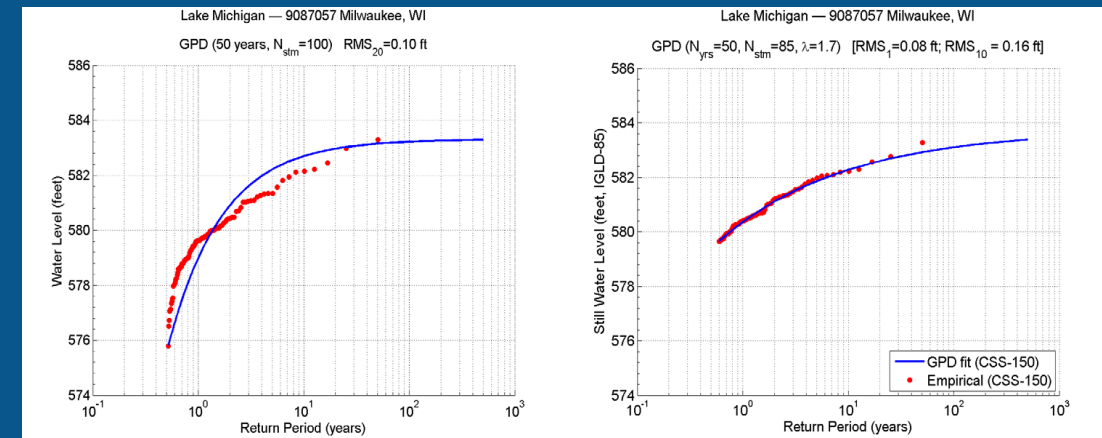
- **Block Maxima Series (BMS)**
 - Annual Maxima
 - ▶ Sample each year's peak event, over the duration of the data
 - Monthly Maxima
 - ▶ Sample each month's peak event
 - Generalized Extreme Value (GEV) Distribution
- **Partial Duration Series (PDS)**
 - Peaks-Over-Threshold (POT)
 - ▶ Sample independent & identically-distributed (iid) peak events
 - ▶ Automated threshold selection methods (e.g., MRL)
 - Generalized Pareto Distribution (GPD)



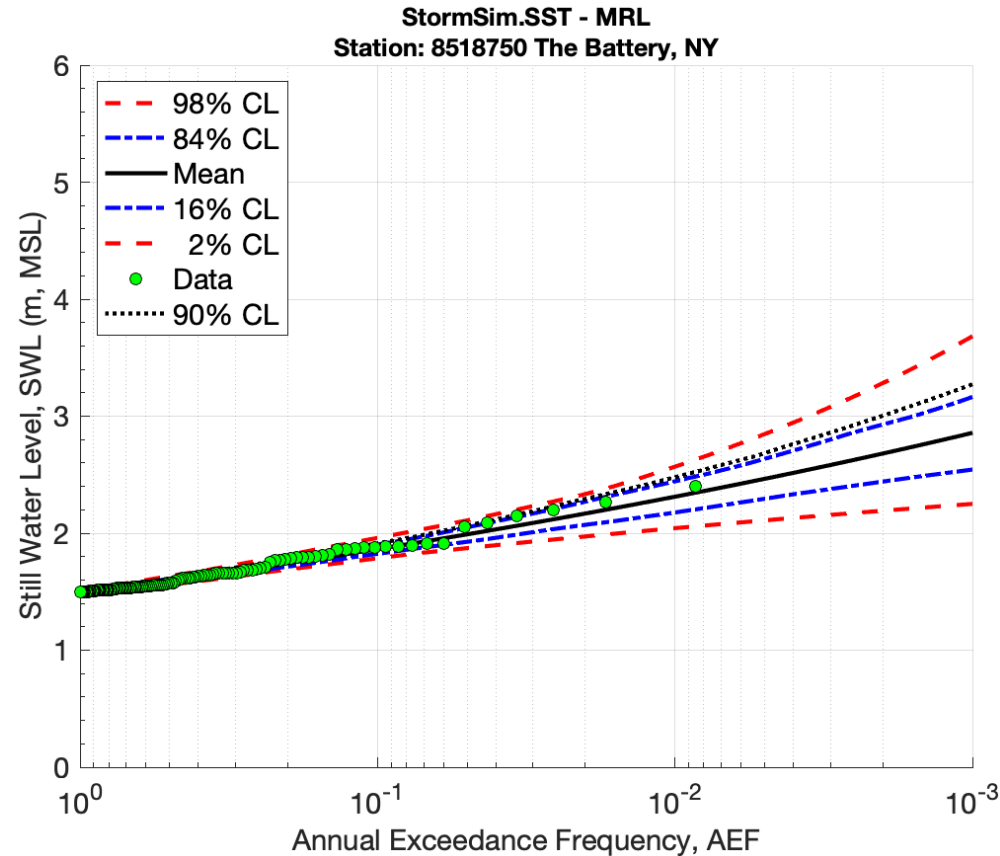
Extreme Value Analysis

Sampling Methods

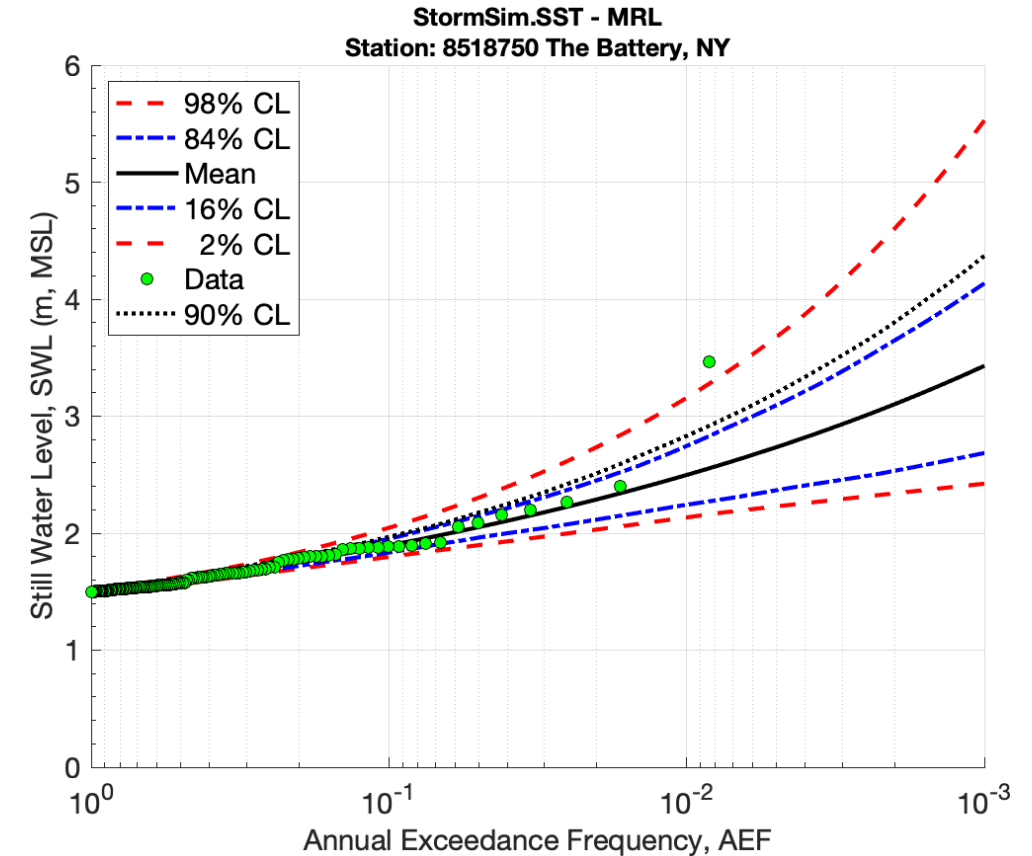
- **Block Maxima Series (BMS)**
 - Significant extreme events can potentially be discarded
 - Can result in mixed populations of extreme and non-extreme events
 - Often results in over-conservative estimates
 - Sensitive to short record lengths
- **Partial Duration Series (PDS)**
 - Maximizes usage of available data
 - Can incorporate BMS
 - Flexible / complex – user defined threshold
 - Less sensitive to short record lengths
 - Less uncertainty in water level predictions



Extreme Value Analysis: Historical Water Levels



1% AEP @ 90% CL = 2.45 m
0.1% AEP @ 90% CL = 3.16 m



1% AEP @ 90% CL = 2.82 m
0.1% AEP @ 90% CL = 4.45 m

Joint Probability Method (JPM)

Background

- **Characterization of individual TC parameters in the 1950's – Myers (1954)**
 - Hydrometeorological report submitted to USACE
 - Track, location, pressure deficit, radius of maximum winds, etc.
- **JPM - pioneered in the late 1960's – Russell (1968)**
 - Dissertation on the probability distributions of hurricane effects
 - Full Monte Carlo Simulation
 - Wind, surge, and wave loads on offshore structures
- **JPM – adopted for coastal storm surge studies in the 1970's (NOAA)**



Joint Probability Method (JPM)

Background

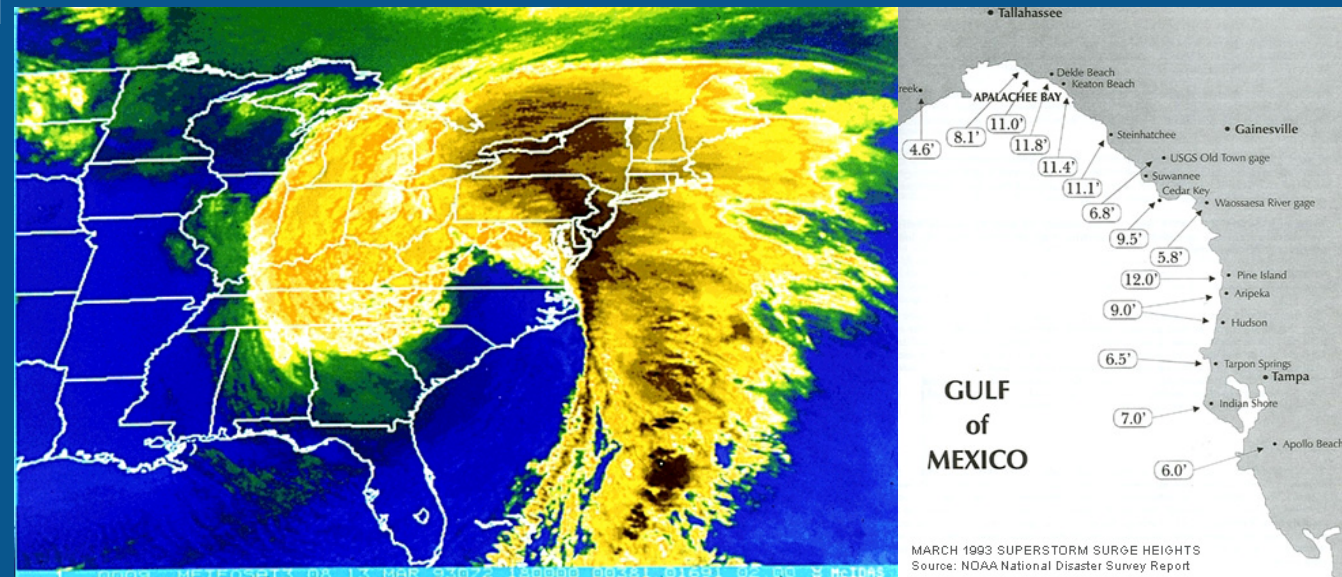
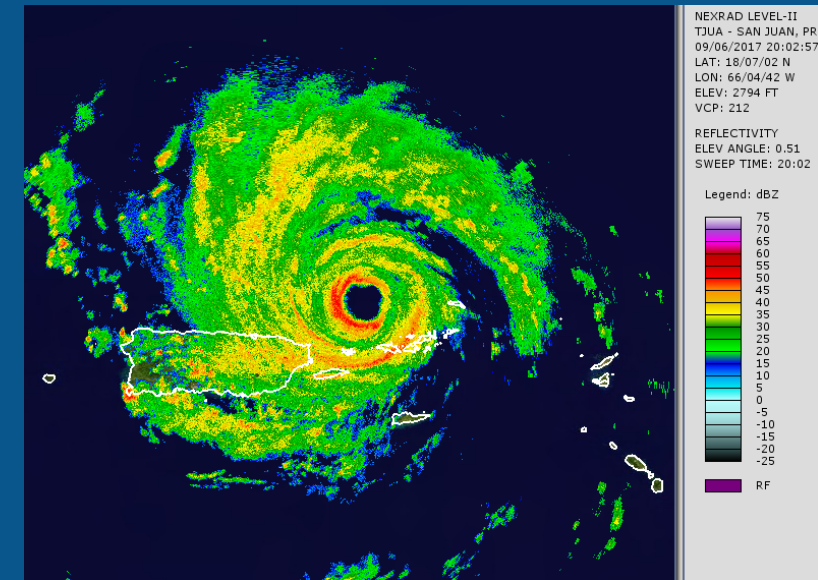
- **JPM with Optimal Sampling (JPM-OS) – aftermath of Hurricane Katrina (IPET 2009)**
 - Response Surface (RS) – USACE/Resio (2007)
 - Bayesian Quadrature (BQ) – FEMA/Toro (2008)
 - Stochastic Track Model (STM) – Vickery (2008)
- **Probabilistic Coastal Hazard Analysis (PCHA) Framework (USACE) – builds upon JPM-OS**
 - Meta-Gaussian Copula (MGC)
 - Gaussian Process Metamodel (GPM)



JPM – Atmospheric Forcing

Storm Climatology

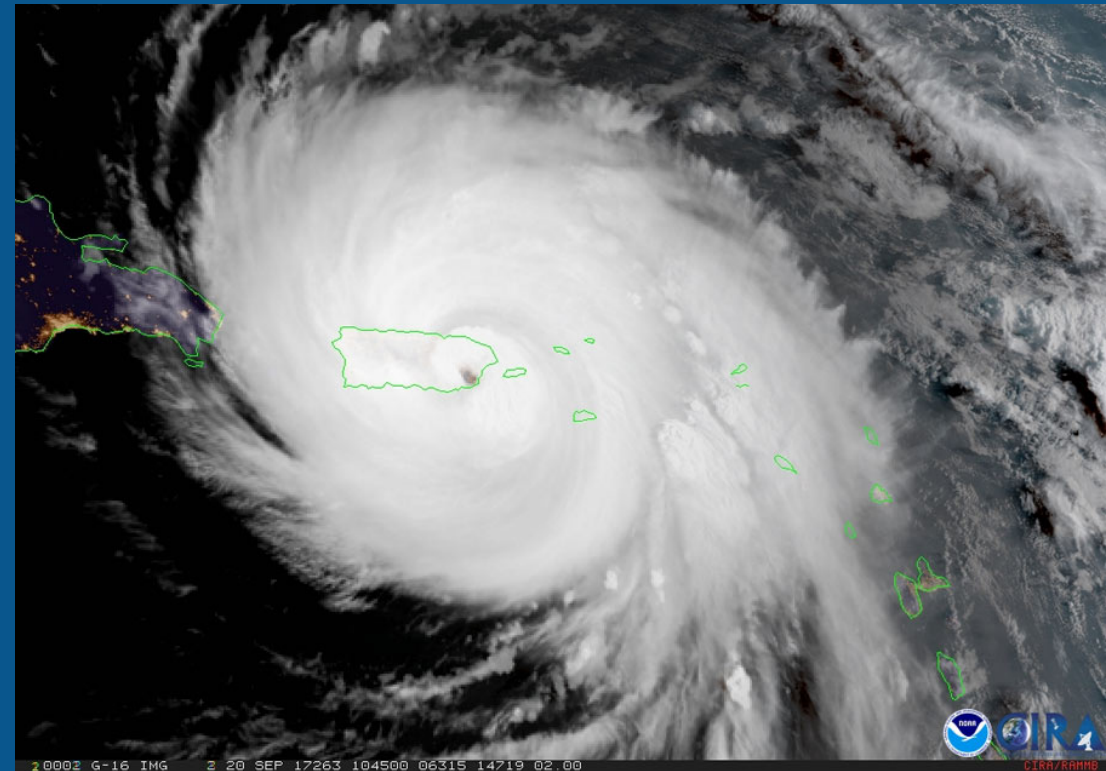
- Tropical Cyclones (TC)
 - “Genesis” as tropical storms
 - ▶ Tropical storms / Hurricanes
 - ▶ Transitioning tropical cyclones
 - ▶ Post-tropical cyclones
 - Extratropical transition has occurred
- Extratropical Cyclones (XC)
 - “Pure” extratropical storms
 - ▶ e.g., Nor’easters
 - No tropical cyclogenesis



JPM – Representation of Tropical Cyclone Forcing

TC Parameterization: $\text{Response} = f(x_o, \Delta p, R_{max}, V_t, \theta)$

1. Reference location (x_o)
2. Central pressure deficit (Δp)
3. Radius of maximum winds (R_{max})
4. Translational speed (V_t)
5. Heading direction (θ)



Joint Probability Method

JPM Integral

$$\lambda_{r(\hat{x}) > R} = \lambda \int P[r(\hat{x}) + \varepsilon > r | \hat{x}, \varepsilon] f_{\hat{x}}(\hat{x}) f_{\varepsilon}(\varepsilon) d\hat{x} d\varepsilon$$

$$\approx \sum_i^n \lambda_i P[r(\hat{x}) + \varepsilon > r | \hat{x}, \varepsilon]$$

where:

$\lambda_{r(\hat{x}) > R}$ = AEP of TC response r

$\hat{x} = f(x_o, \Delta p, R_{max}, V_t, \theta)$

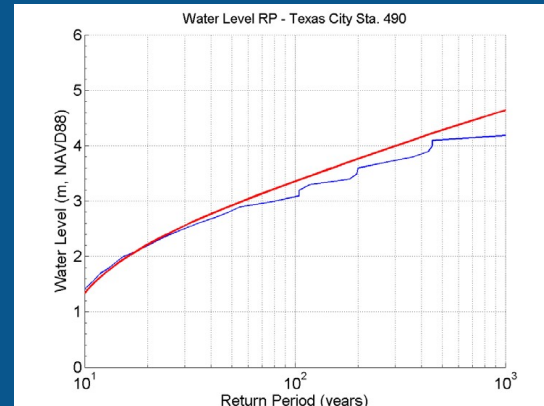
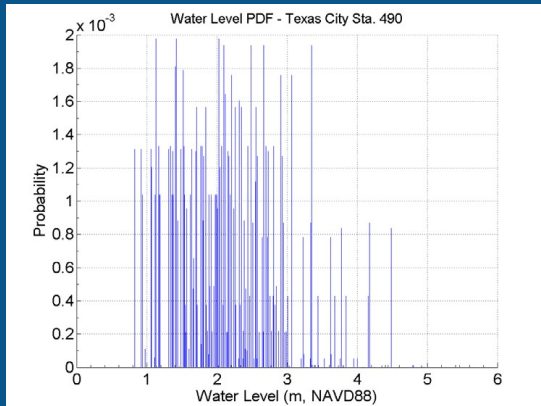
λ = SRR (storms/yr/km)

$\hat{\lambda}_i$ = probability mass (storms/yr) or λp_i ,

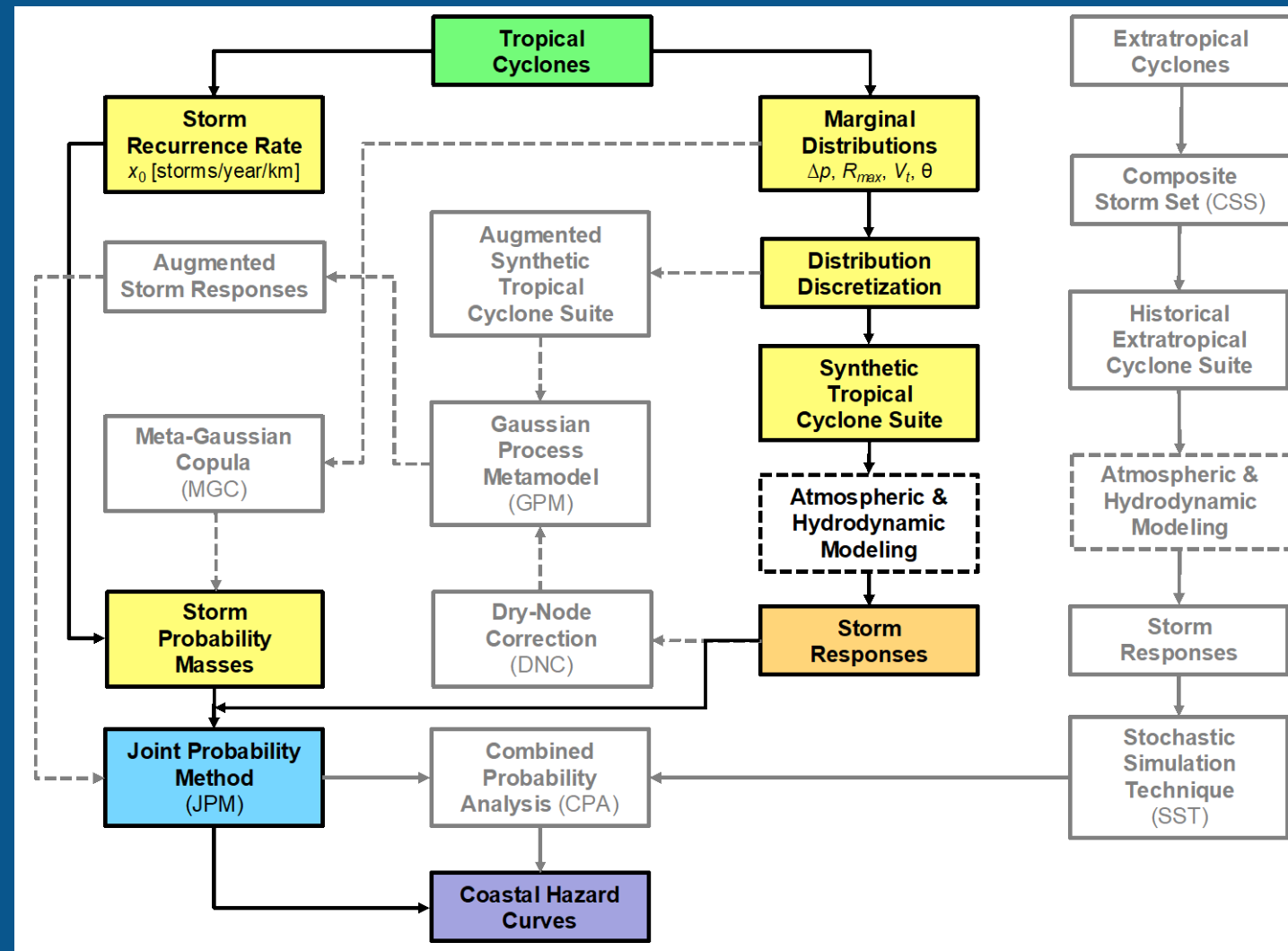
with p_i = product of discrete probability and TC track spacing (km)

$P[r(\hat{x}) + \varepsilon > r | \hat{x}, \varepsilon]$ conditional probability that storm i with parameters \hat{x}_i generates a response larger than r

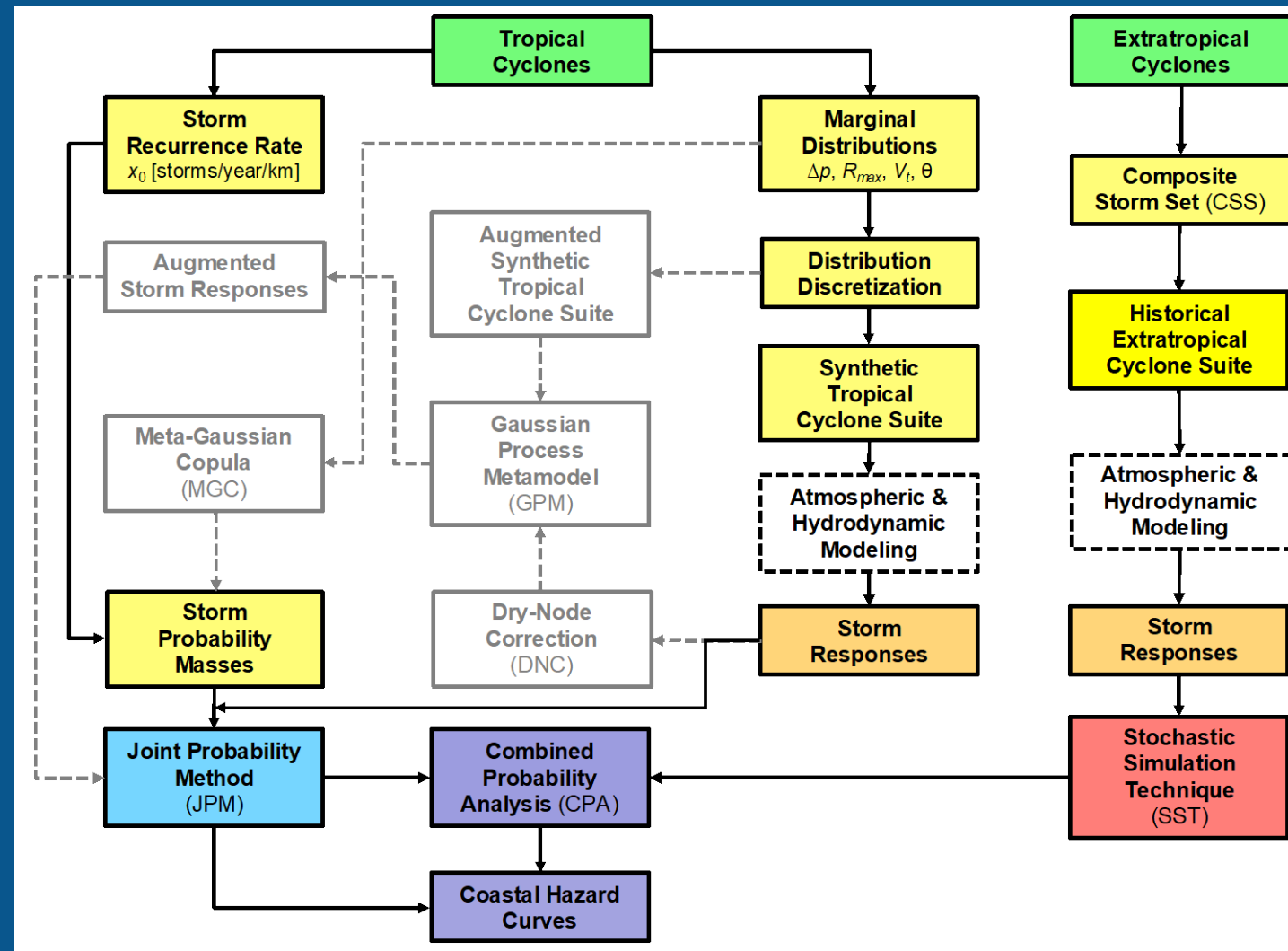
ε = unbiased error or aleatory uncertainty of r



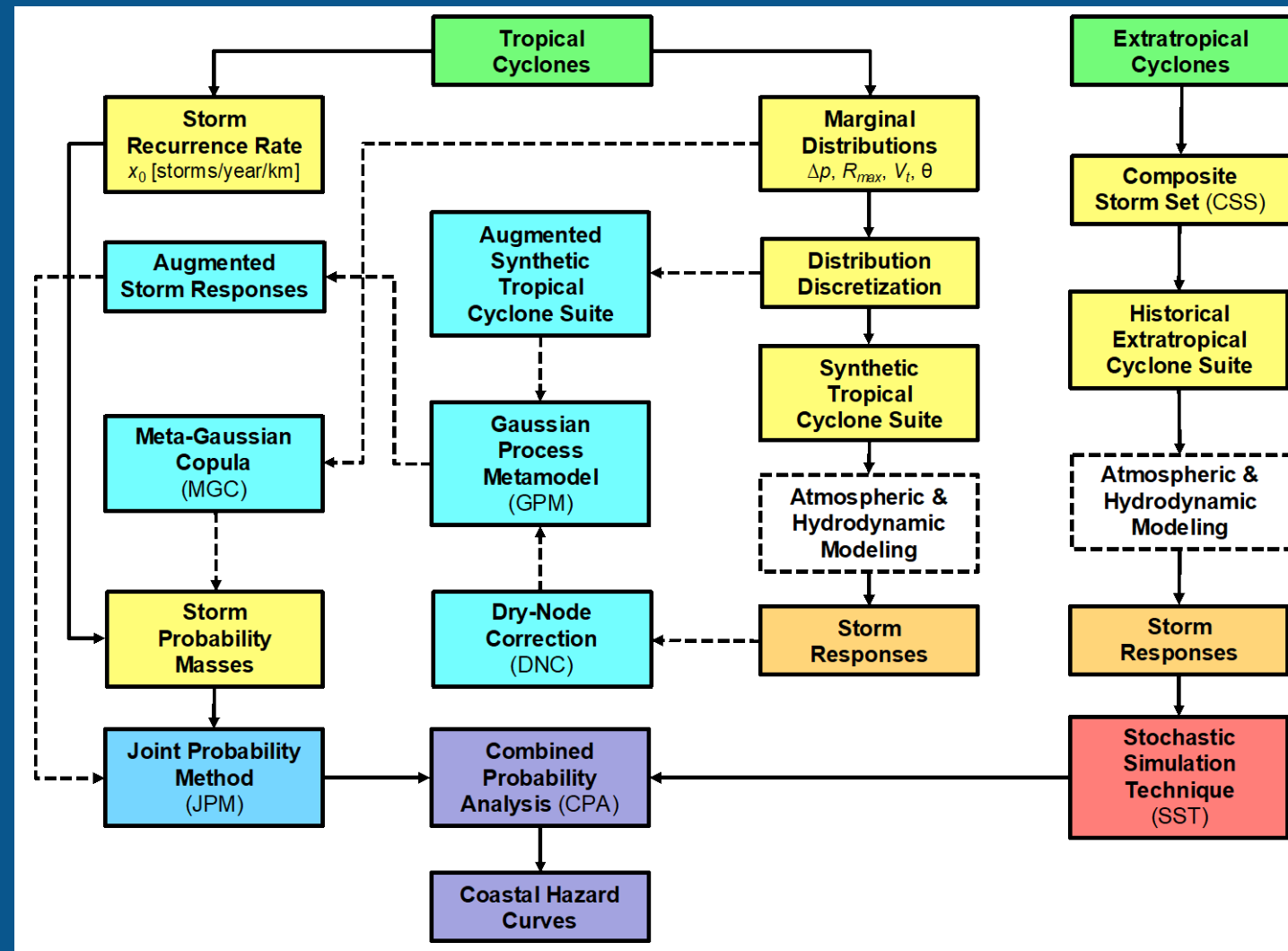
JPM – Joint Probability Analysis of TCs



JPM – Joint Probability Analysis of TCs and XC



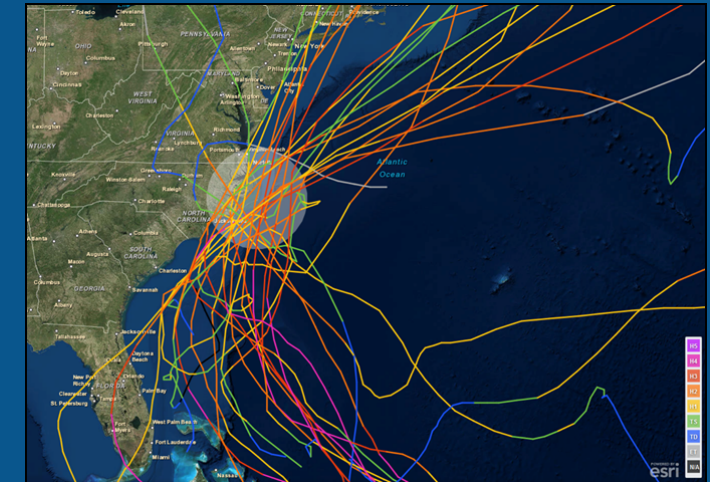
Probabilistic Coastal Hazard Analysis (PCHA)



JPM – Data Sources

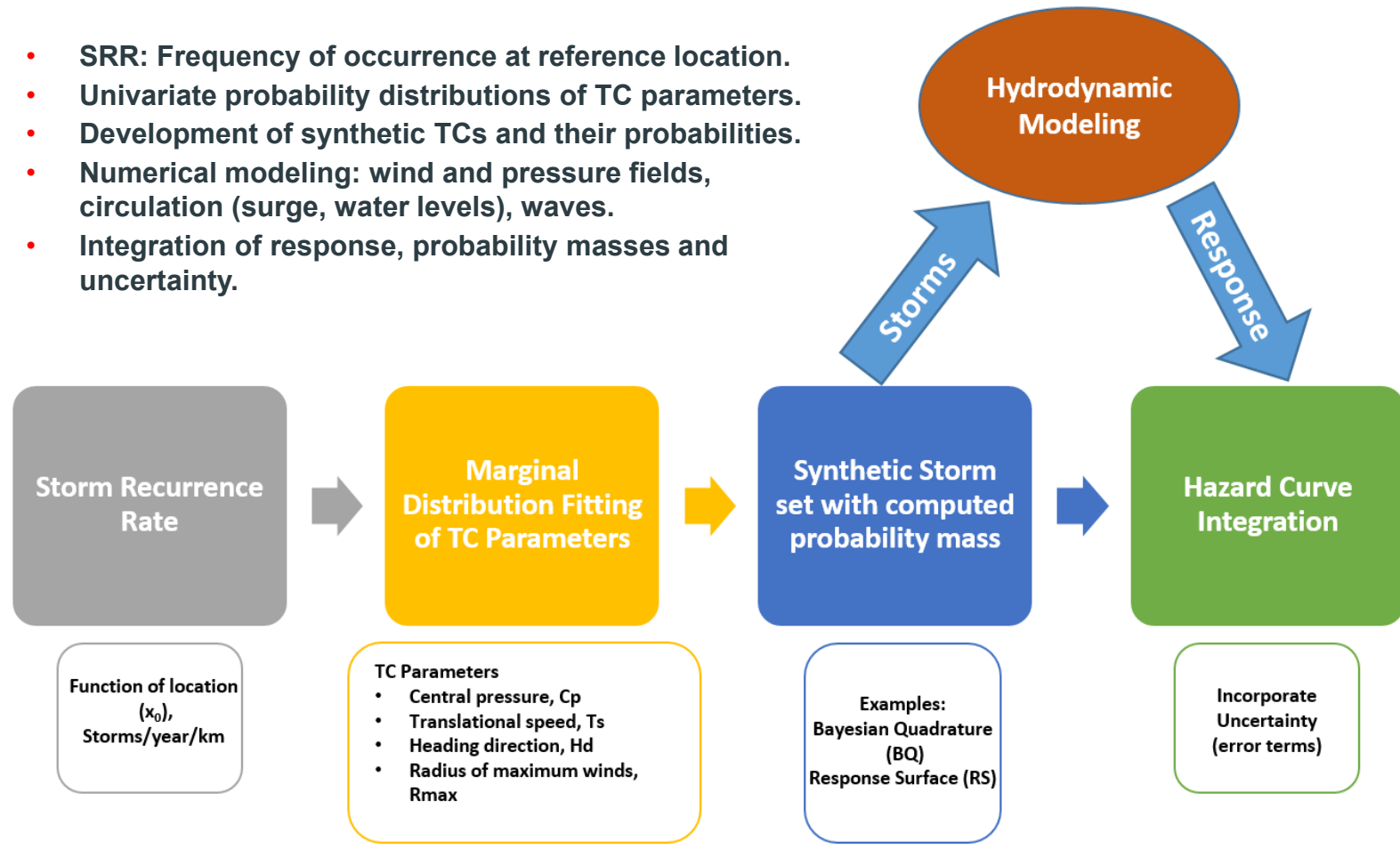
TC Data Sources

- NHC HURricane DATA 2nd generation (HURDAT2)
 - TC parameters: max wind speed, central pressure, lat, lon
- Automated Tropical Cyclone Forecast (ATCF)
 - Best track data: 2019
- Colorado State (CSU) Extended Best Track (EBTRK)
 - R_{max} (1988 – 2018)
- Coastal Hazards System (CHS) Gaussian Process Metamodel Fills in gaps in central pressure and estimates R_{max}
 - Period: 1851 – 2019



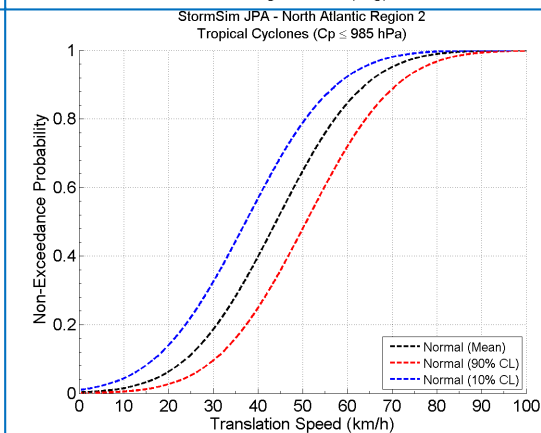
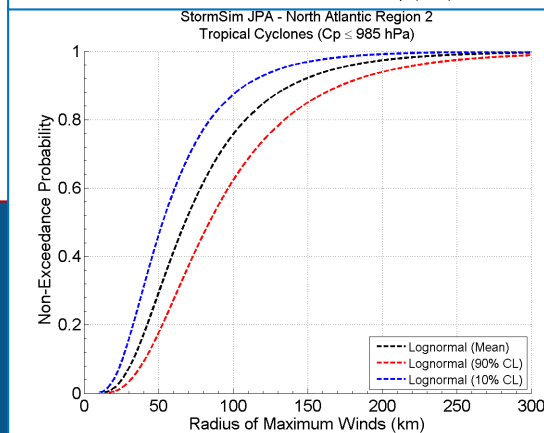
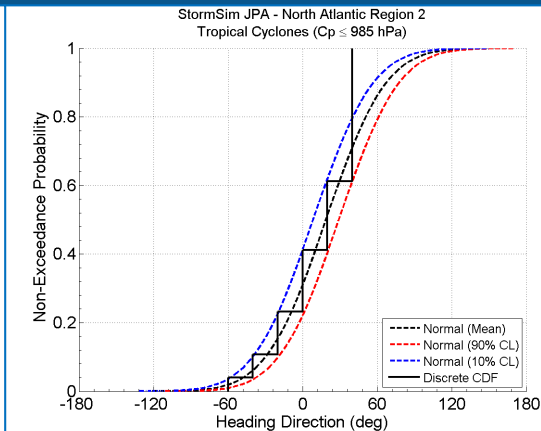
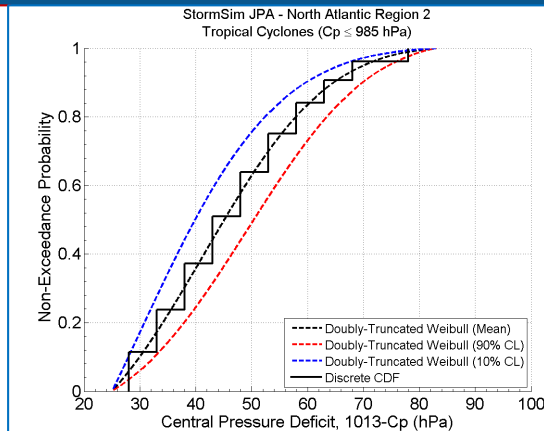
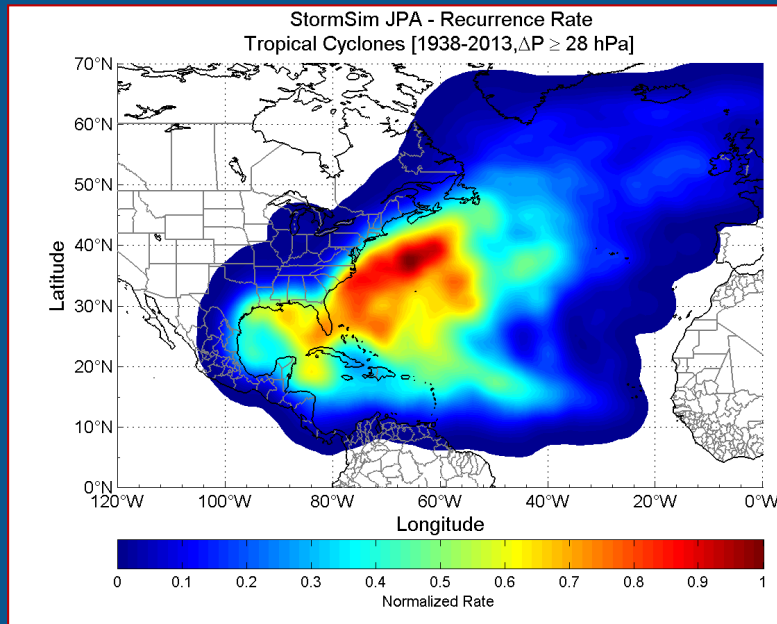
Joint Probability Method – Basic Components

- SRR: Frequency of occurrence at reference location.
- Univariate probability distributions of TC parameters.
- Development of synthetic TCs and their probabilities.
- Numerical modeling: wind and pressure fields, circulation (surge, water levels), waves.
- Integration of response, probability masses and uncertainty.



JPM – TC Parameterization

$$\text{Response} = f(x_o, \Delta p, R_{max}, V_t, \theta)$$



Marginal
Distribution Fitting
of TC Parameters

Joint Probability Method

Synthetic Storm
set with computed
probability mass

Storm Development / Discretization Approach

- **Response Surface (RS)**
 - Storm generation based on expert judgement
 - Storm surge response surface, interpolated in the $(\Delta p, R_{max})$ 2D space
- **Bayesian Quadrature (BQ)**
 - Coarse grid storm surge “benchmark”
 - Multiple storm sets generated by BQ; select the one closest to the “benchmark”
- **Stochastic Track Model (STM)**
 - Coarse grid storm surge “benchmark”
 - Discretization based on expert judgement, or generated by STM itself
- **Probabilistic Coastal Hazard Analysis (PCHA)**
 - Hybrid storm-generation approach: $\theta, \Delta p$ (regular spacing); R_{max}, V_t (BQ)
 - Discretization of the entire multi-parameter space $(\Delta p, R_{max}, V_t, \theta)$; Augmented TC Suite



Joint Probability Method

Synthetic Storm
set with computed
probability mass

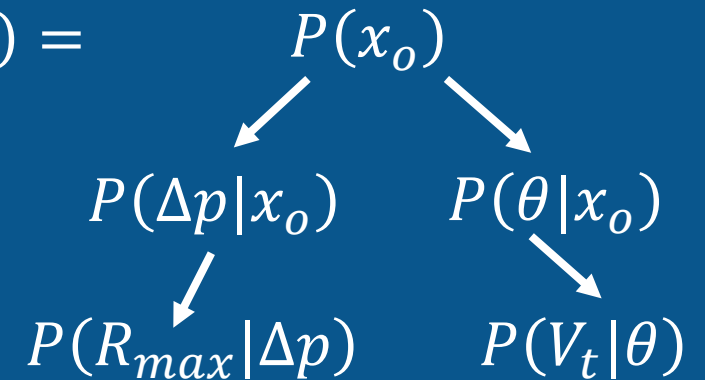
Storm Probabilities

- Assumed Independence

$$P(x_o, \Delta p, R_{max}, V_t, \theta) = P(x_o) \cdot P(\Delta p) \cdot P(R_{max}) \cdot P(V_t) \cdot P(\theta)$$

- Correlation Tree (1:1 Dependence)

$$P(x_o, \Delta p, R_{max}, V_t, \theta) =$$



- MGC (Multivariate Dependence)

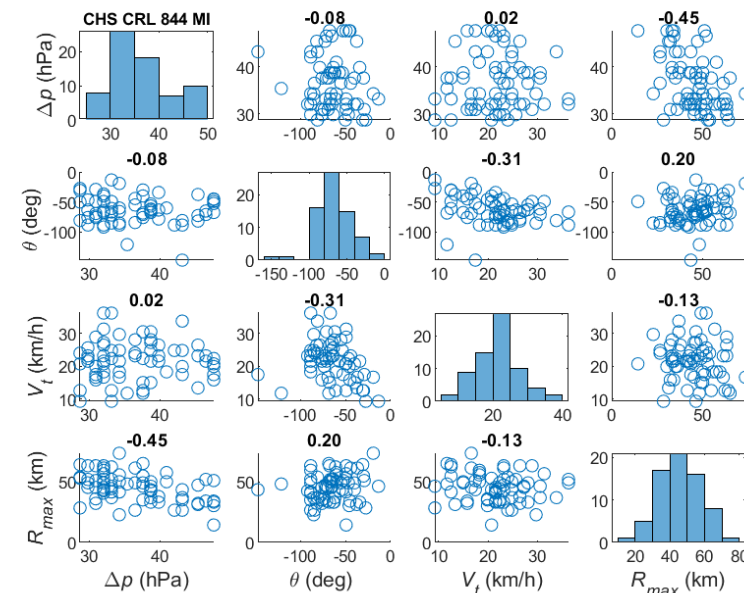
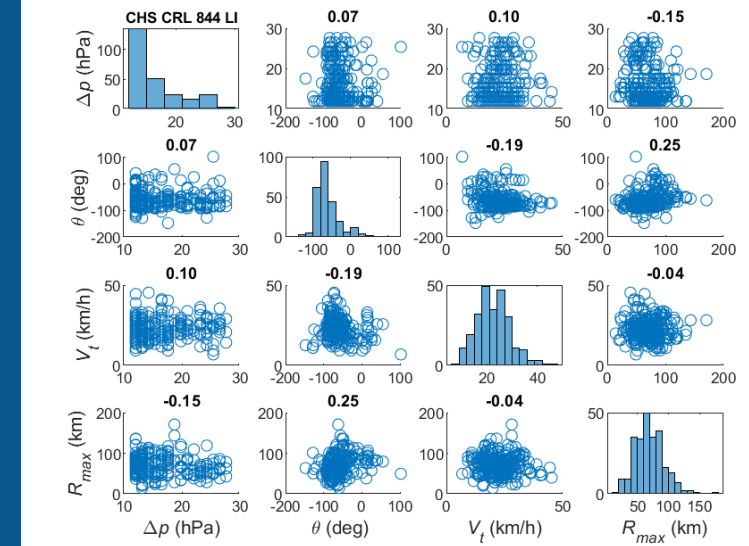
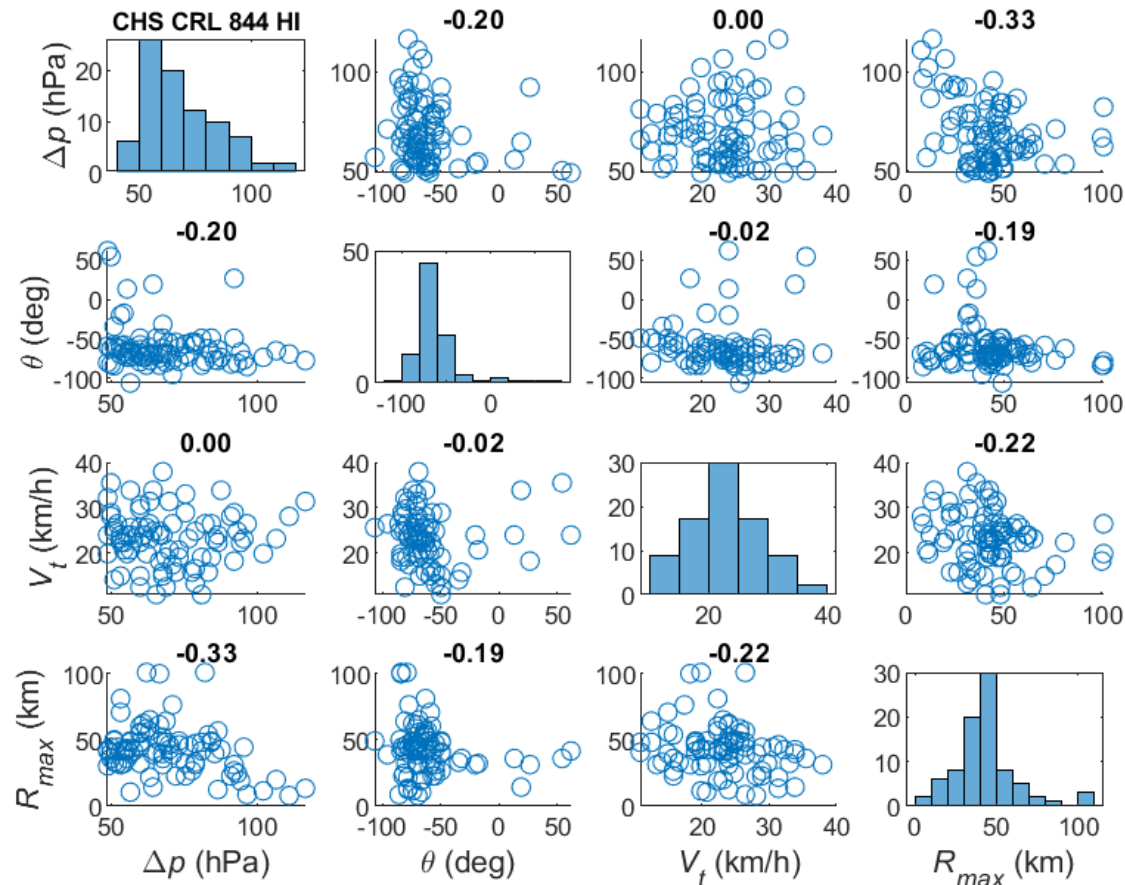
$$P(x_o) \cdot \Phi_R(\Phi^{-1}(u_1), \dots, \Phi^{-1}(u_n))$$

$$R = \begin{pmatrix} 1 & \rho_{1,2} & \dots & \rho_{1,n} \\ \rho_{2,1} & 1 & \dots & \rho_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ \rho_{n,1} & \rho_{n,2} & \dots & 1 \end{pmatrix}$$



PCHA Example: SACS PR-USVI

Meta-Gaussian Copula: Correlation Matrix

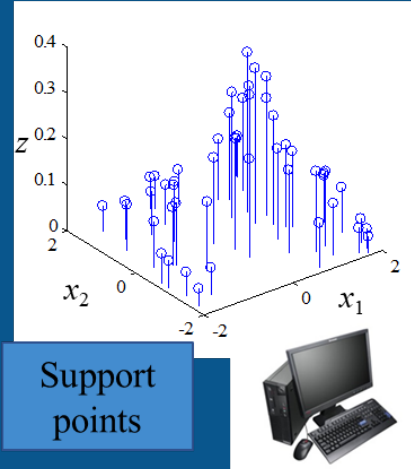
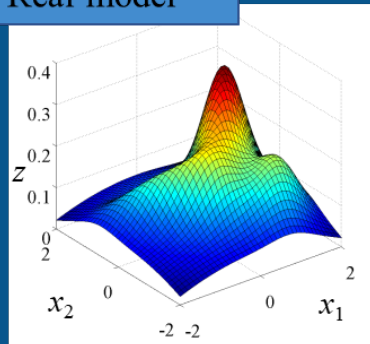


Synthetic Storm
set with computed
probability mass

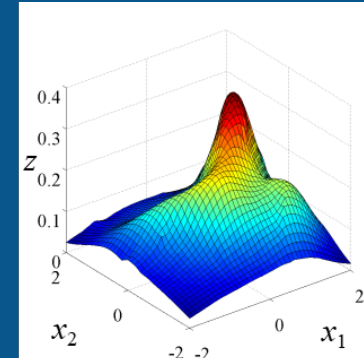
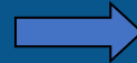
PCHA Example: SACS PR-USVI

Gaussian Process Metamodel (GPM)

Real model



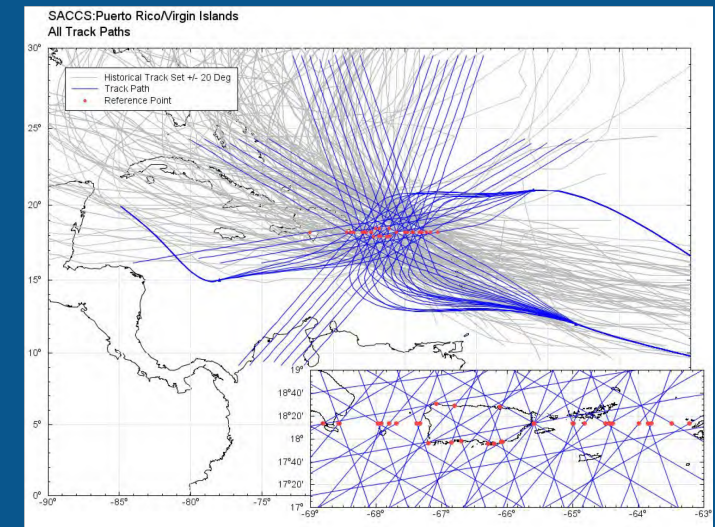
Support points



GPM approximation (global)
+ correction (local)



Synthetic Storm
set with computed
probability mass

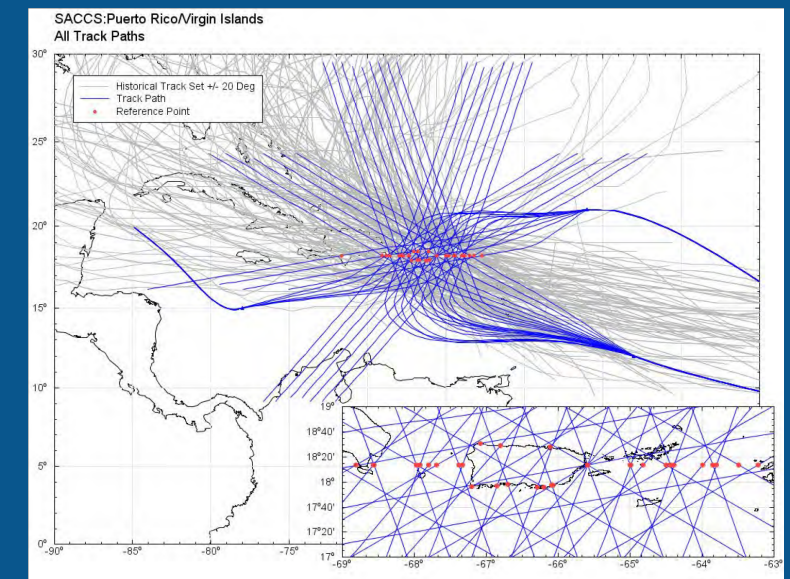
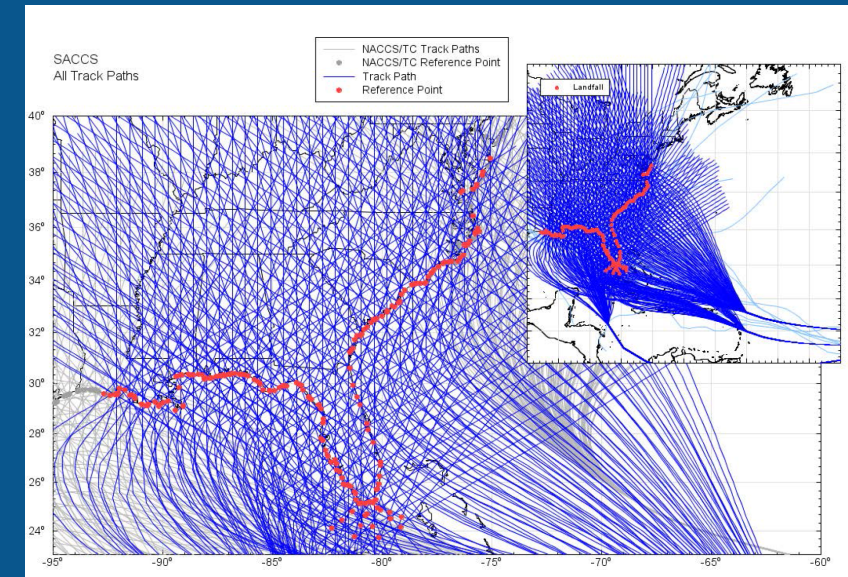
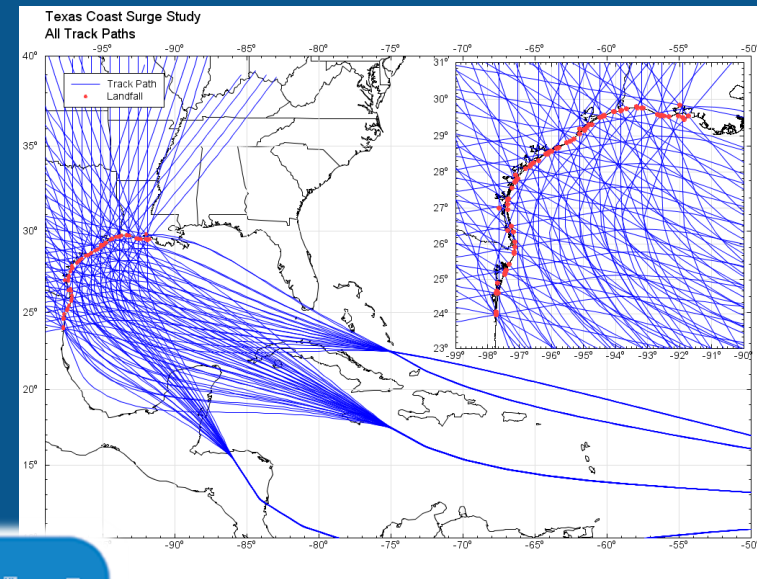
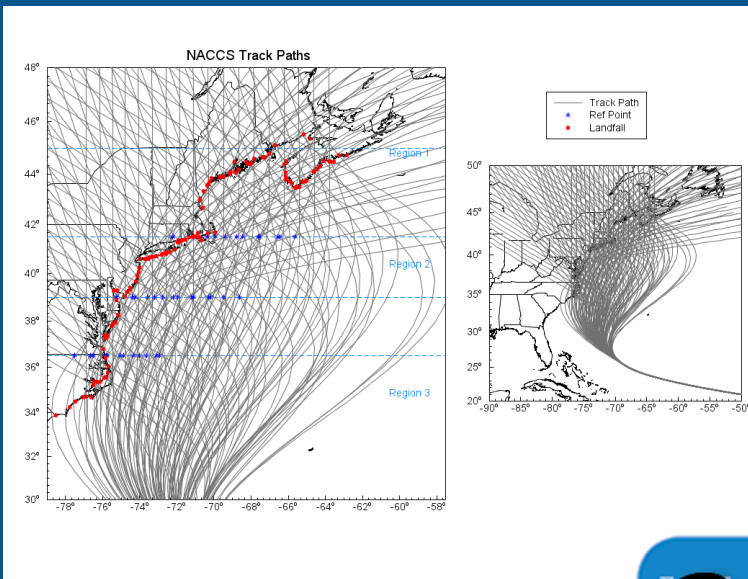


TC Parameter	Full Suite 300 TCs	Augmented Suite 348,000 TCs
θ (deg)	-60:20:60	-60:20:60
Δp (hPa)	8:10:148; 18:10:138	8:5:148
R_{max} (km)	8 to 143.6	10:5:155
V_f (km/h)	8 to 40	5:5:50

Joint Probability Method

Synthetic TC Suites: Master Tracks

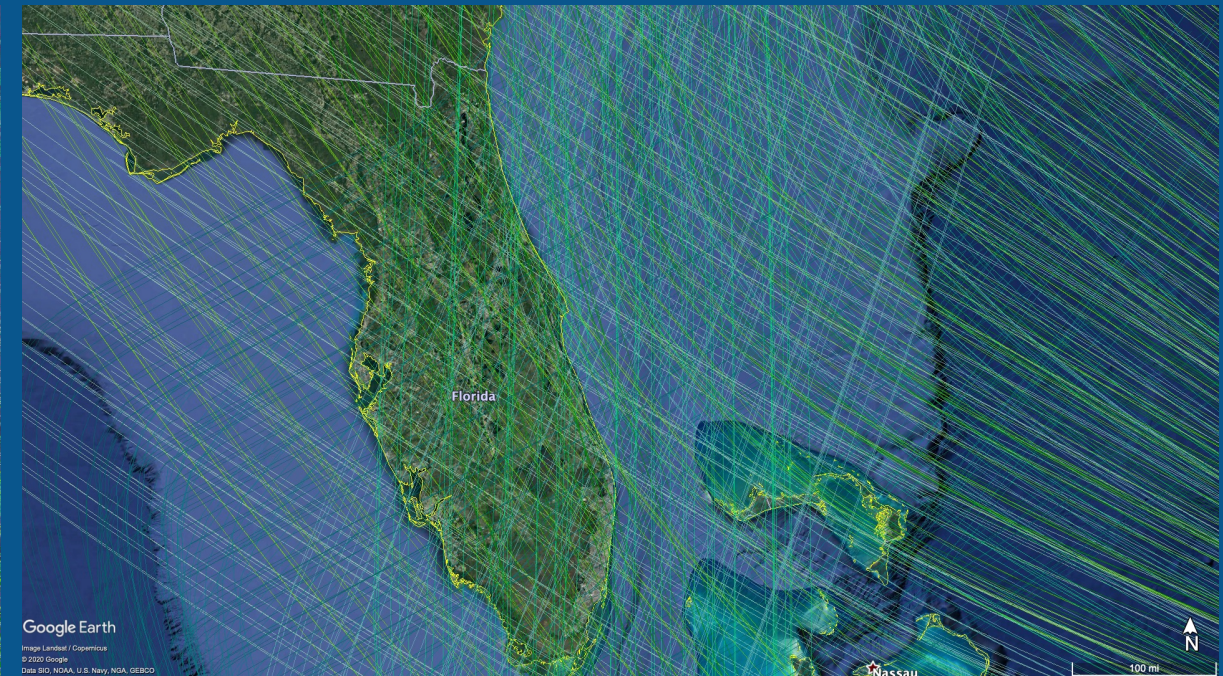
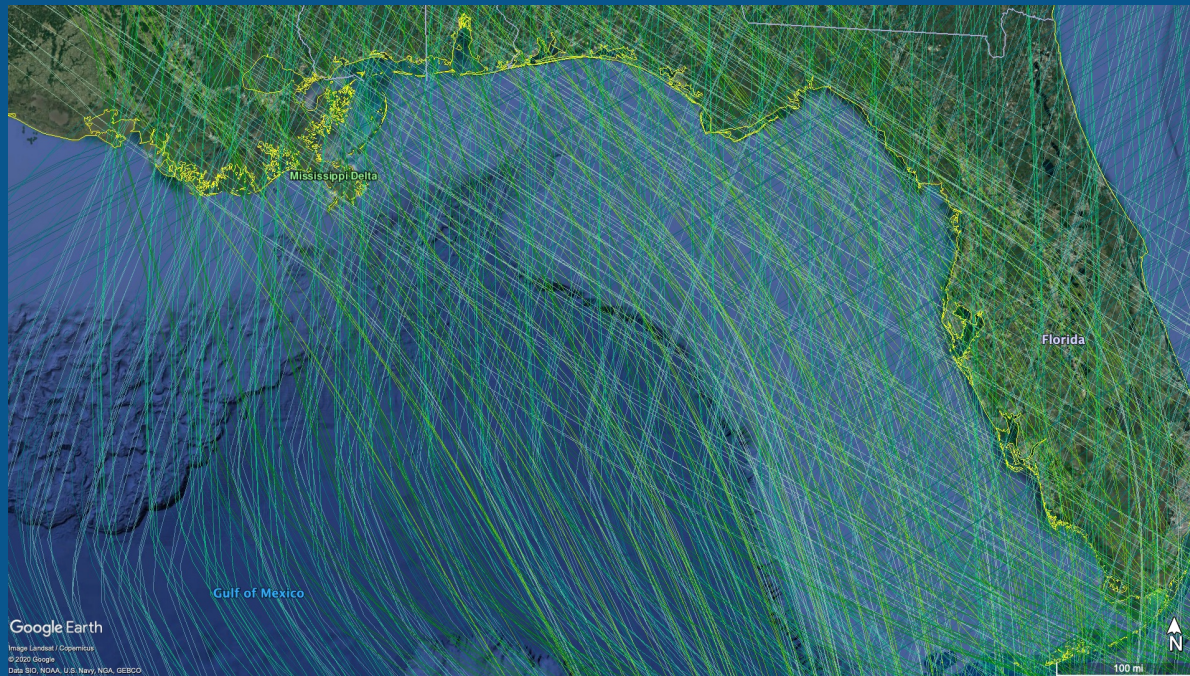
Synthetic Storm
set with computed
probability mass



Joint Probability Method

Synthetic TC Suites: R_{max} path

Synthetic Storm
set with computed
probability mass

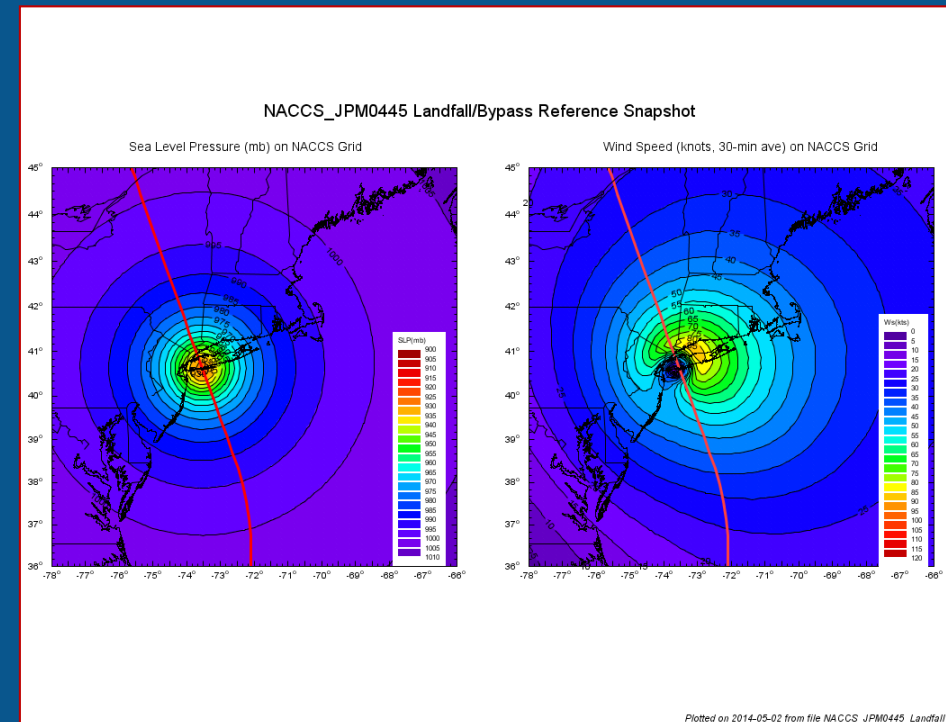
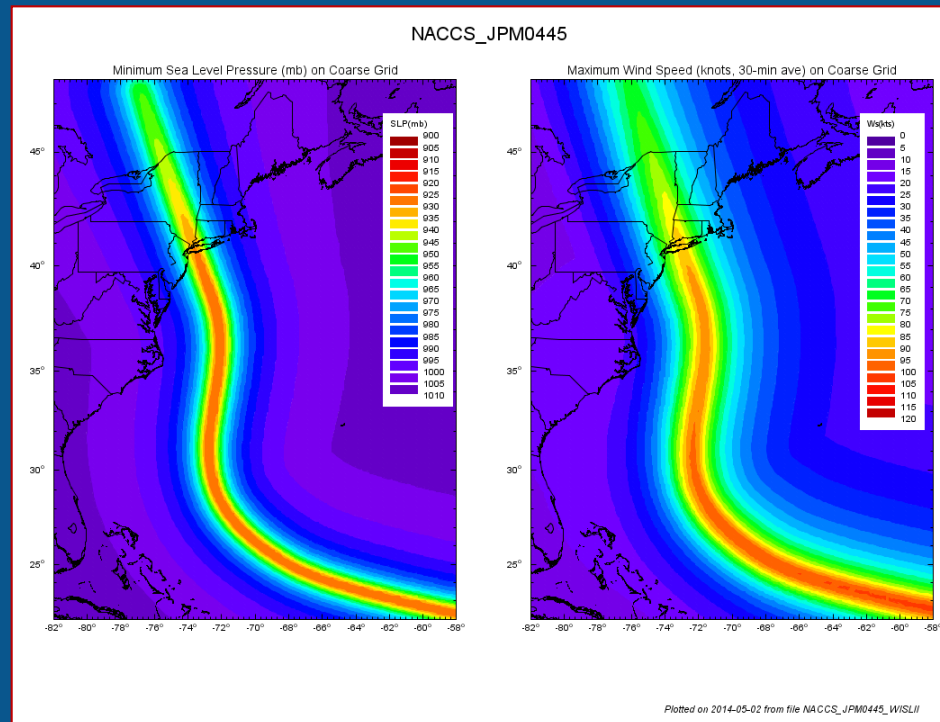


Joint Probability Method

Hydrodynamic
Modeling

Atmospheric simulation of synthetic TCs

- Wind & Pressure Fields

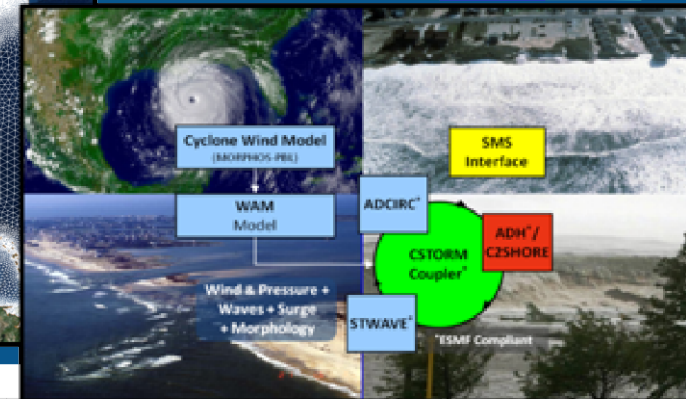
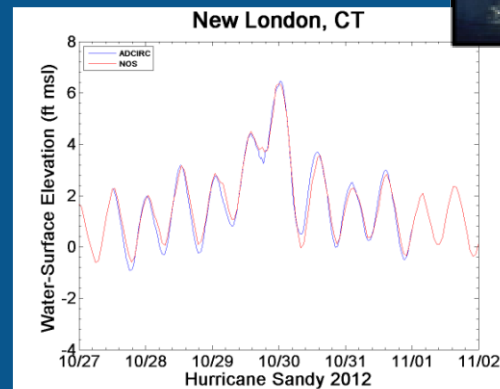
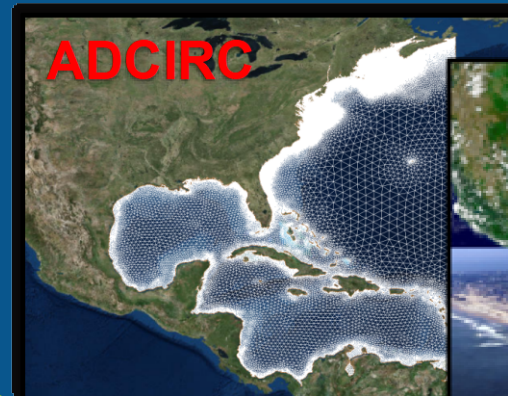
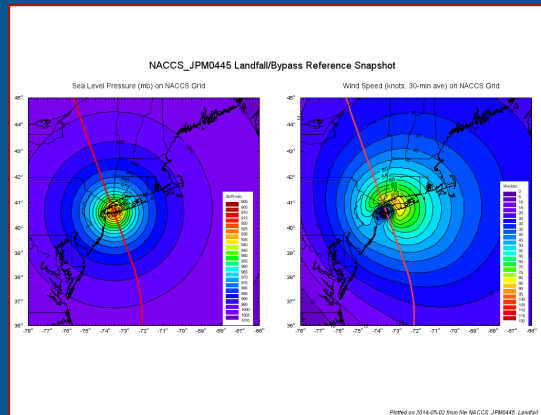


Joint Probability Method

Hydrodynamic simulation of synthetic TCs

- Storm Surge/Water Level & Waves

Hydrodynamic
Modeling



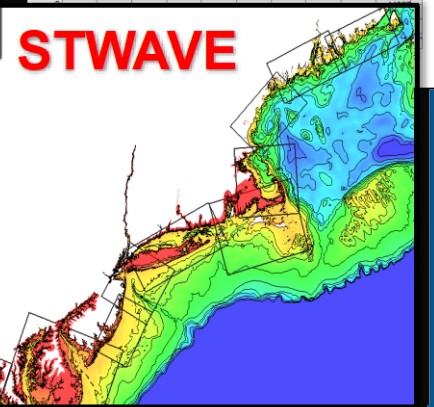
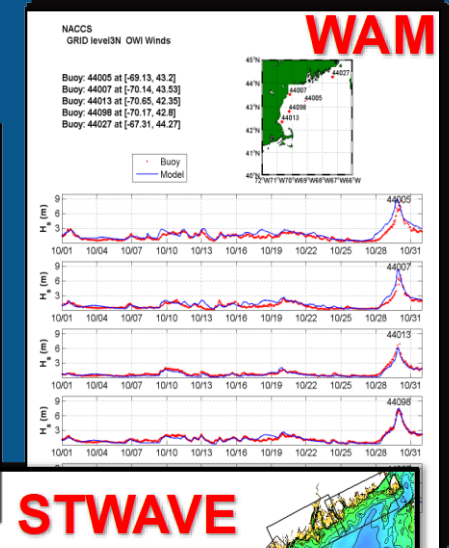
CSTORM-MS:
Coastal **STORM** Modeling System

WAM:
WAve Prediction **M**odel

STWAVE:
STeady-State Spectral **WA**VE model

ADCIRC:
ADvance **CIRC**ulation Model

3M nodes, min res ~20 m.



PCHA Example: SACS PR-USVI

ADCIRC



San Juan



Resolution Before: 70-100 m

Resolution After: 30-85 m

Notes:

- The largest city in Puerto Rico, contains significant amount of critical infrastructure

Hydrodynamic
Modeling

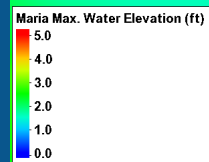


Base ADCIRC Mesh – Courtesy of Dr. Juan Gonzalez-Lopez

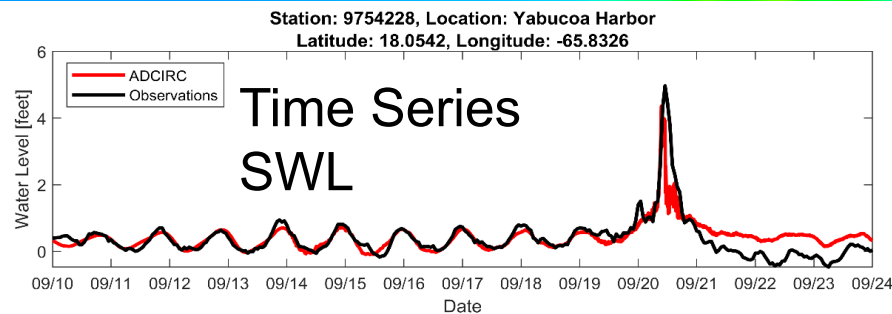
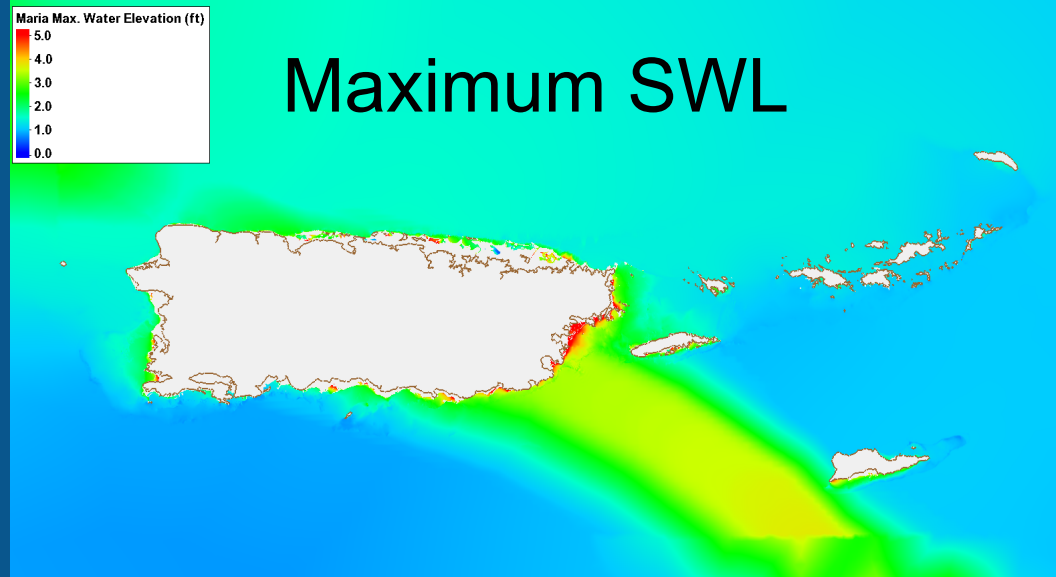
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PCHA Example: SACS PR-USVI

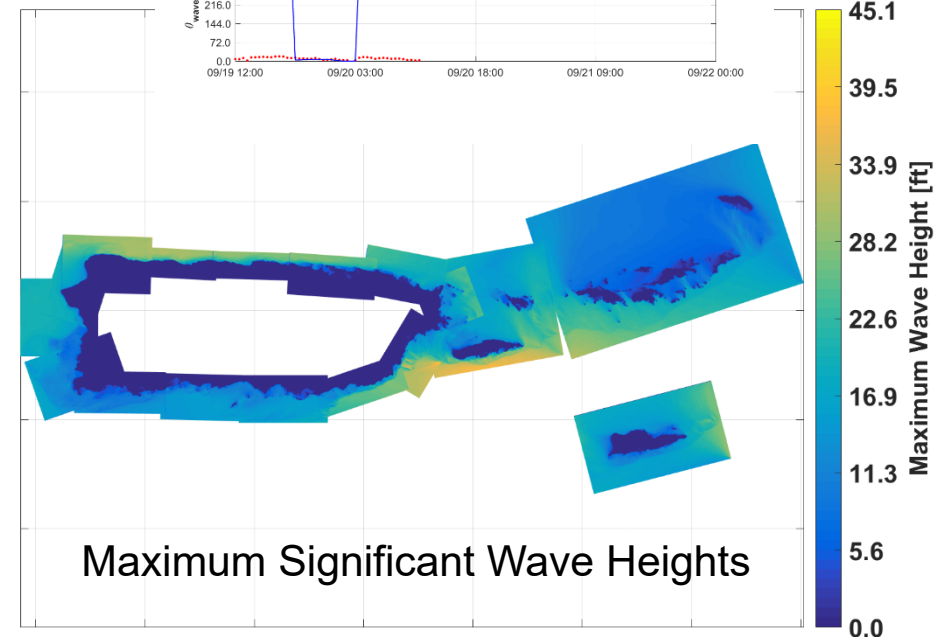
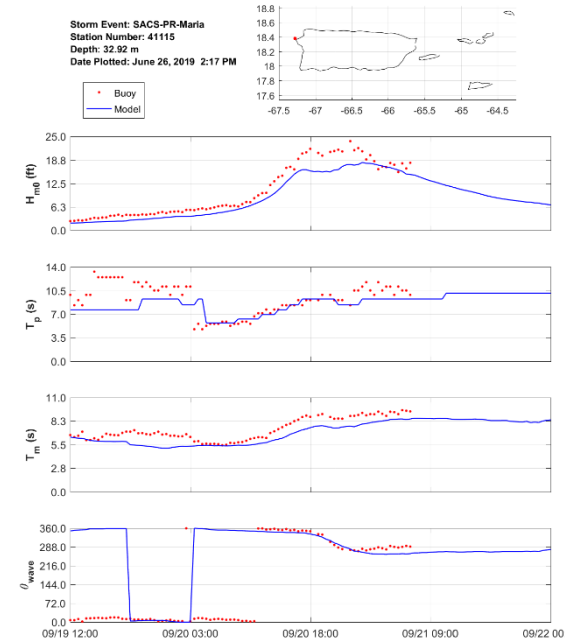
Hurricane Maria



Maximum SWL



Storm Event: SACS-PR-Maria
Station Number: 41115
Depth: 32.82 m
Date Plotted: June 26, 2019 2:17 PM



Maximum Significant Wave Heights

Hydrodynamic
Modeling

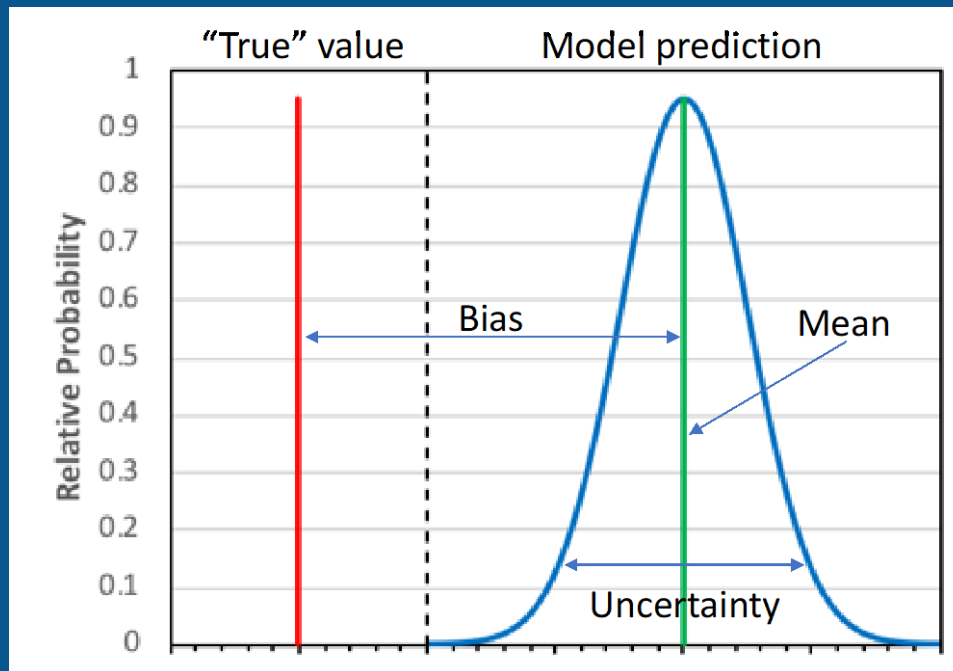
Time Series
Waves

Joint Probability Method

Hazard Curve
Integration

Quantification of Uncertainty

- Bias** – measure of model accuracy; defined as the average error or difference between model predictions and the “true” values.
- Uncertainty** – represents the likely extent or spread of the model error.



$$f(X) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{X-\mu}{\sigma}\right)^2}$$

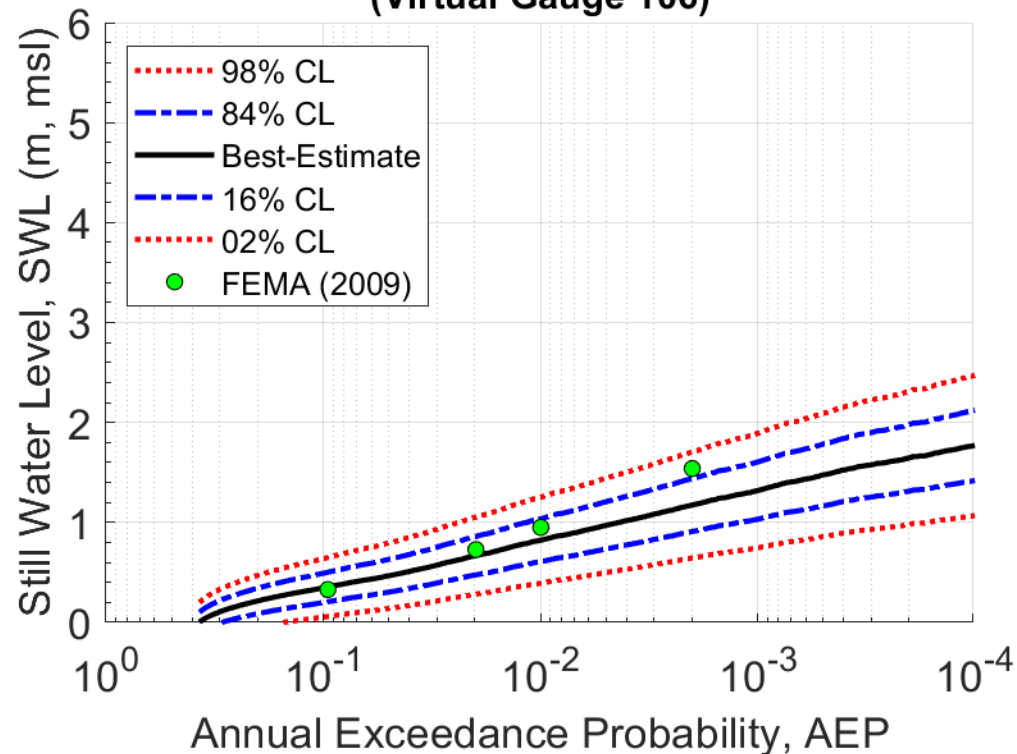
$$\sigma_{\varepsilon a} = \sqrt{\sigma_{\varepsilon 1}^2 + \sigma_{\varepsilon 2}^2}$$

Uncertainty	North Atlantic Coast Comp. Study (2015)	Sabine Pass to Galveston Bay Wave and Water Level Modeling Study (2015)	South Atlantic Coast Study: Puerto Rico and the U.S. Virgin Islands (ongoing)*	Flood Insurance Study: Coastal Counties, Texas (2011)	FEMA Region II Storm Surge Project (2014)	Mississippi Coastal Analysis Project (2008)
Hydrodynamic Modeling	0.48 m	0.91 m (combined with meteorological modeling)	0.20 m (constant) 0.30 (proportional)	0.56 to 0.76	0.39	0.23 m
Meteorological Modeling	0.38 m	-	0.14 (proportional) 0.09 (constant)	0.07 to 0.30	0.54	0.36 m
Storm Track Variation	0.25 m	0.09 m	N/A	0.20 x wave setup	N/A	N/A
Holland B	0.15 x storm surge elevation	0.17 x surge elevation	N/A	0.15 x surge elevation	N/A	0.15 x surge elevation
Astronomical Tide	variable	0.20 m	0.11 m	N/A	N/A	0.20 m

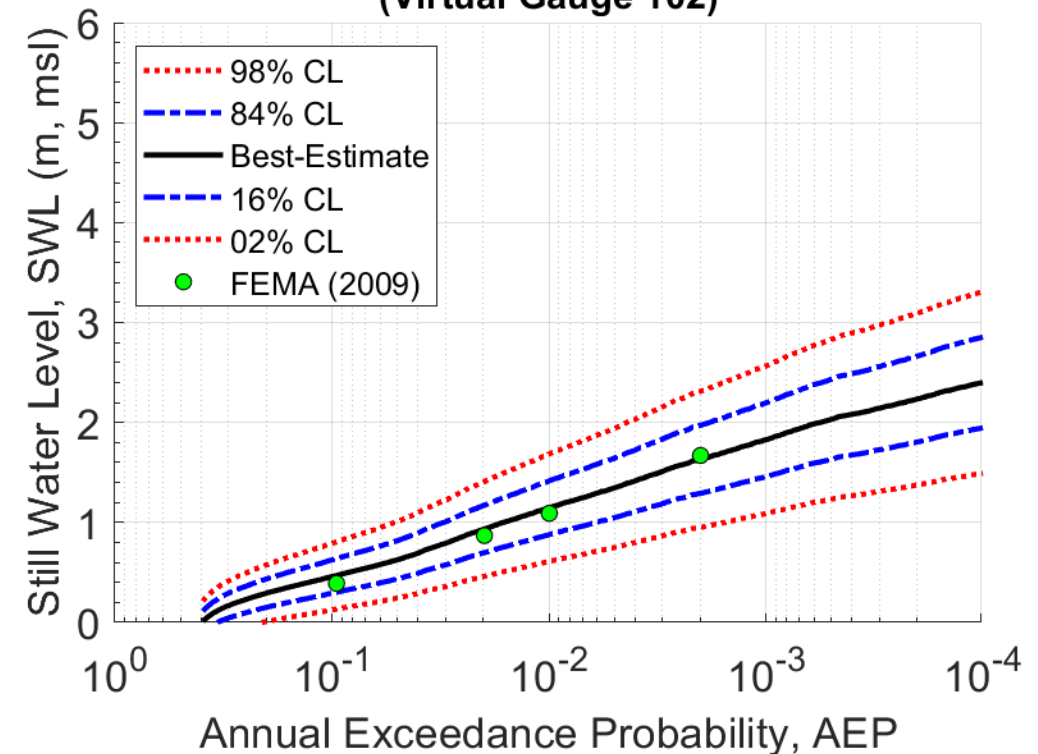
PCHA Example: SACS PR-USVI

Hazard Curve
Integration

CHS-SACS: Puerto Rico & U.S. Virgin Islands
(Virtual Gauge 106)



CHS-SACS: Puerto Rico & U.S. Virgin Islands
(Virtual Gauge 102)



USACE Coastal Flooding Hazard Studies



Comprehensive Studies

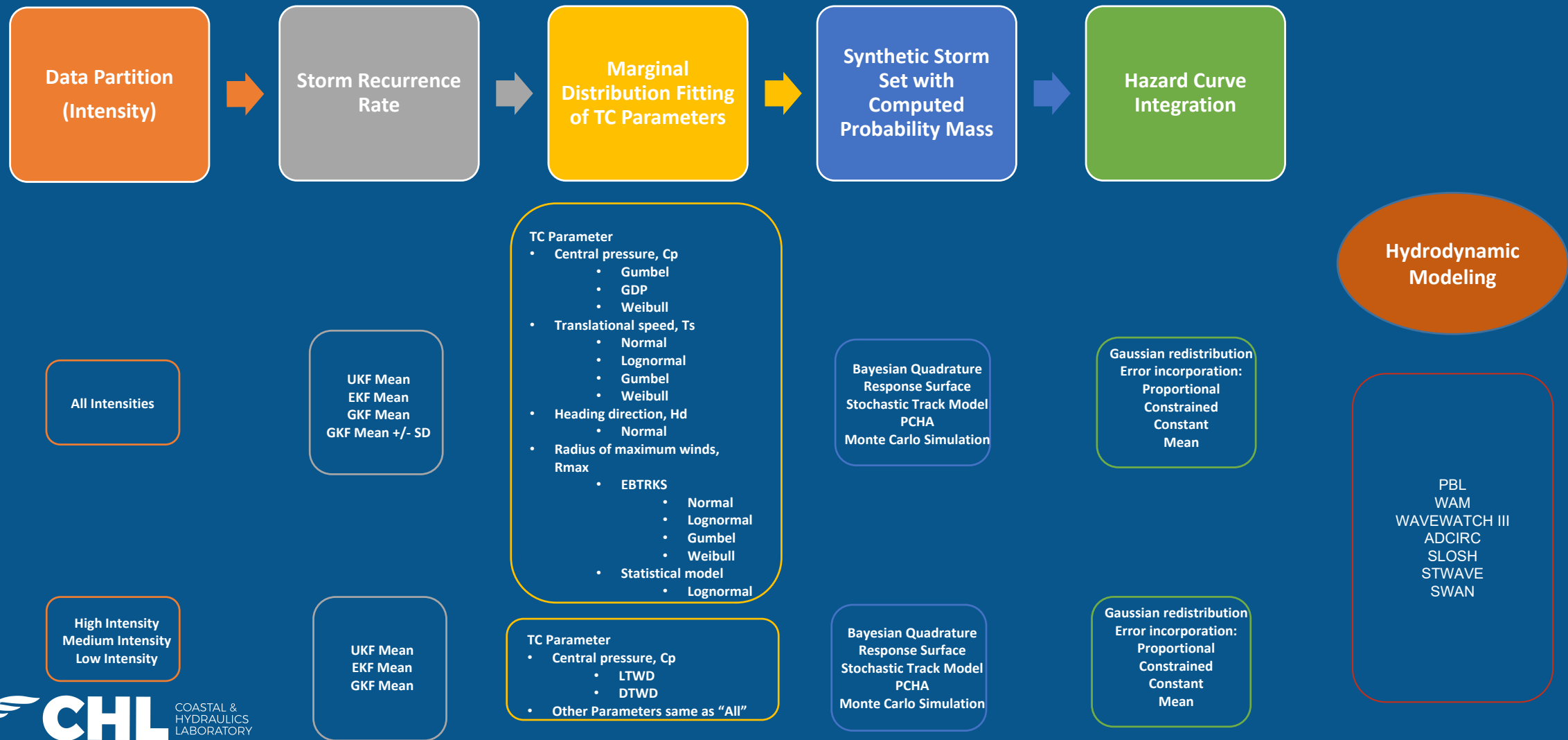
- North Atlantic Coast Comprehensive Study (NACCS)
- Coastal Texas Study (CTXS)
- South Atlantic Coastal Study (SACS)

Focus Studies

- Sabine to Galveston – Pre-construction, Engineering & Designs (PED)
- Coastal Texas – Galveston Bay Coastal Spine Feasibility Study
- New Jersey Back Bays Coastal Storm Risk Management (CSRM) Study
- Nassau County CSRM Study
- NY-NJ Harbor and Tributaries Feasibility Study



Data, Models, and Methods...





U.S. ARMY



Questions?

Dr. Norberto C. Nadal-Caraballo

Leader, Coastal Hazards Group

Norberto.C.Nadal-Caraballo@erdc.dren.mil



US Army Corps
of Engineers



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