EC-Benchmark

August 10, 2020

Organizers:

Andreas F. Haselsteiner (University of Bremen; <u>a.haselsteiner@uni-bremen.de</u>) Ryan G. Coe (Sandia National Laboratories; ryan.coe@sandia.gov) Lance Manuel (University of Texas, Austin; lmanuel@mail.utexas.edu)













EC-Benchmark

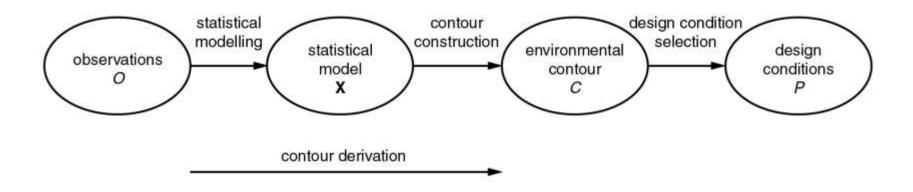
• When: Announced at OMAE2019, some participants submitted OMAE2020 papers

- Why: systematic comparison* of contour methods
 - "predictive power"
 - site-specific data
 - uncertainty in relation to limited datasets
 - defining multivariate extremes

Contribution	Authors ¹	Email of corresponding author	Exercise 1	Exercise 2
1	Wei Chai*, Bernt Leira	chaiwei@whut.edu.cn	X	
2	Clarindo Guilherme, Carlos Guedes Soares*	c.guedes.soares@centec.tecnico.ulisboa.pt	X	X
3	Ásta Hannesdóttir*	astah@dtu.dk	X	X
4	Andreas Haselsteiner*, Aljoscha Sander, Jan- Hendirk Ohlendorf, Klaus-Dieter Thoben	a.haselsteiner@uni-bremen.de	X	Х
5	Guillaume de Hauteclocque*	guillaume.de- hauteclocque@bureauveritas.com	х	
6	Ed Mackay*, Philip Jonathan	e.mackay@exeter.ac.uk	X	
7	Chi Qiao, Andrew Myers*	atm@neu.edu	X	
3	Anna Rode*, Arndt Hildebrandt, Boso Schmidt	anna.rode@stud.uni-hannover.de	X	
9	Erik Vanem*, Arne Bang Huseby	erik.vanem@dnvgl.com	X	X

^{*}not "picking a winner"

Environmental contours

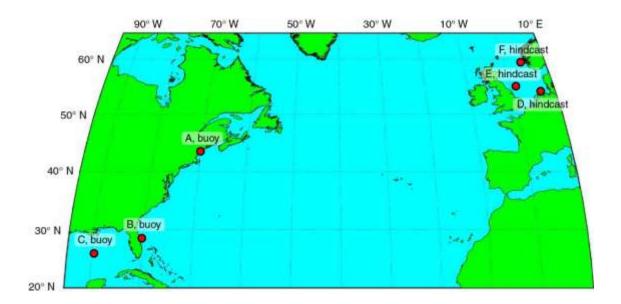


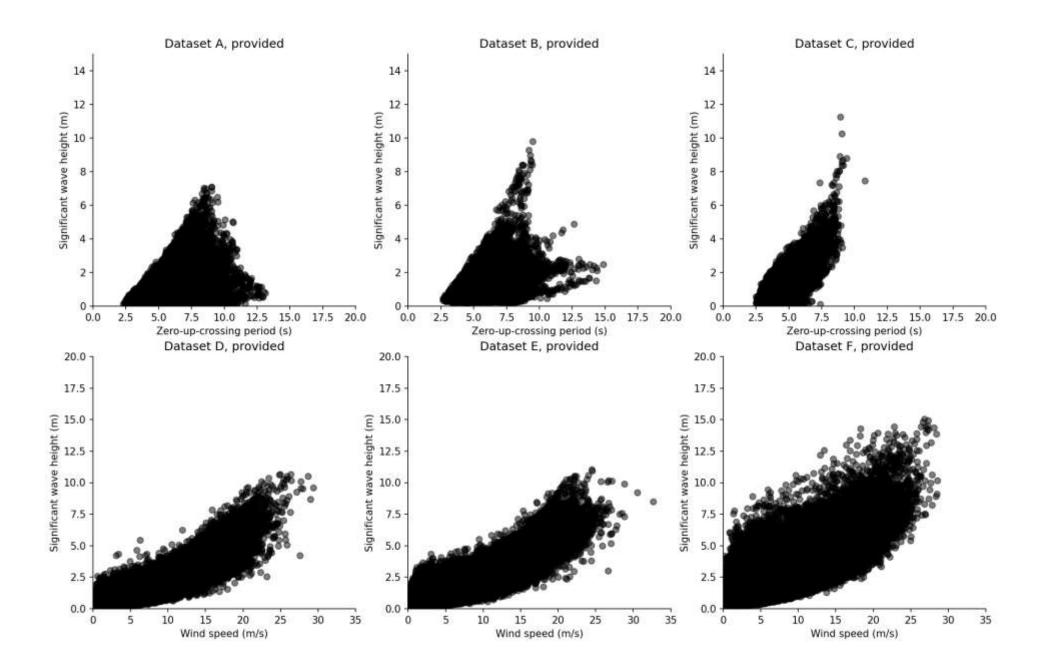
- Different ways to model the joint distribution
 - Different model structures
 - Different parameter estimation techniques
- Different ways to define a contour based on a joint distribution

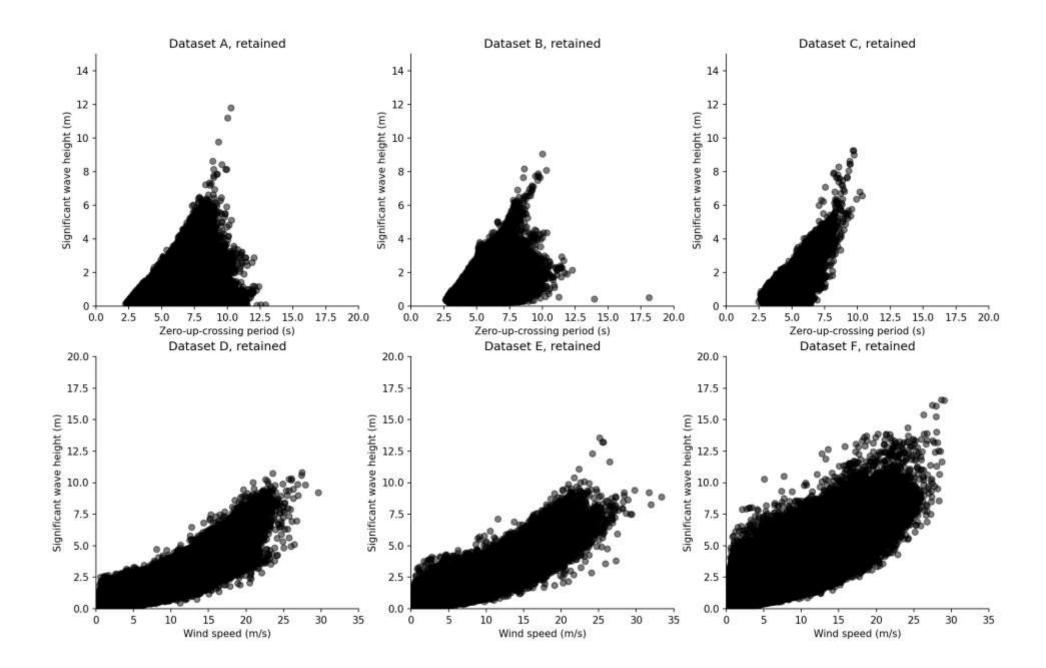
Exercises & Datasets

- Exercise 1: Calculate 12 contours
 - (A, B, C): Significant wave height & zerocrossing period for 1 & 20 yr return periods (10 years of data)
 - (D,E,F): Significant wave height & wind speed for 1 & 50 yr return periods (25 years of data)

- Exercise 2: Calculate uncertainty in contour for dataset D
 - 1. Draw random sample from data
 - 2. Calculate a 50-yr contour from sample
 - 3. (repeat 1000x)

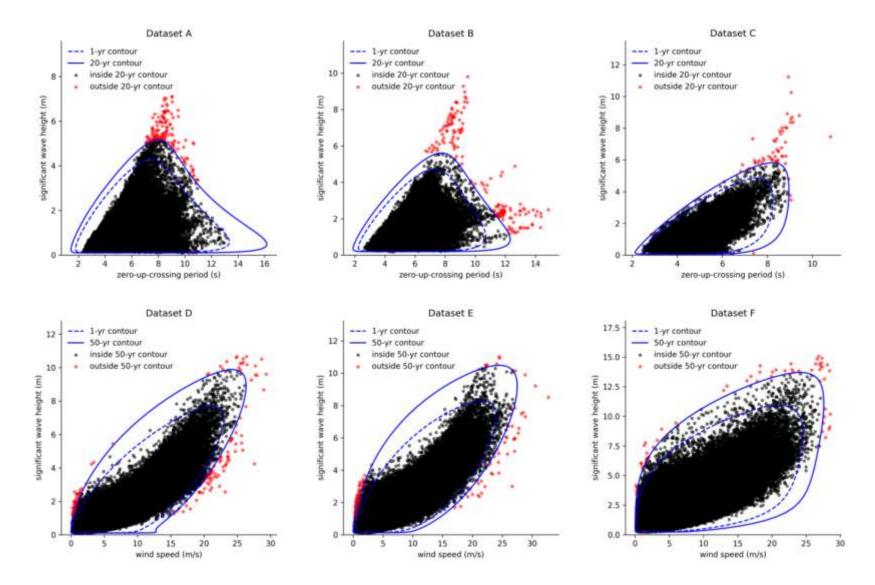




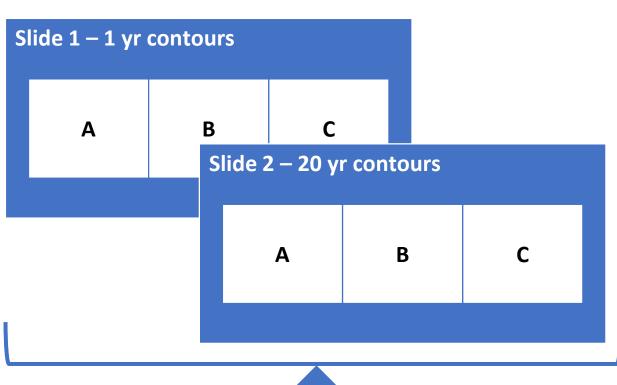


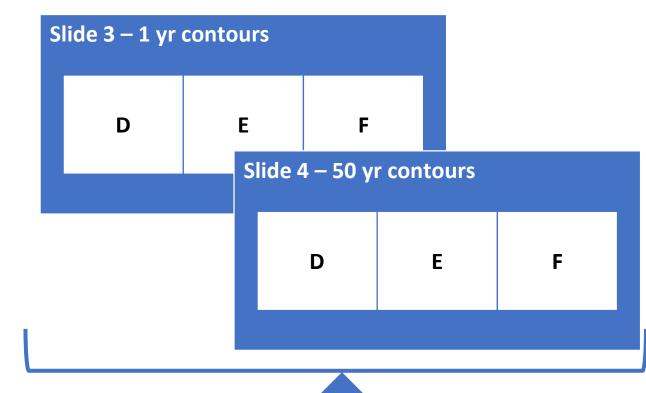
ii) Overview about the results

Exercise 1: Baseline results

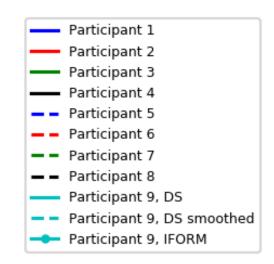


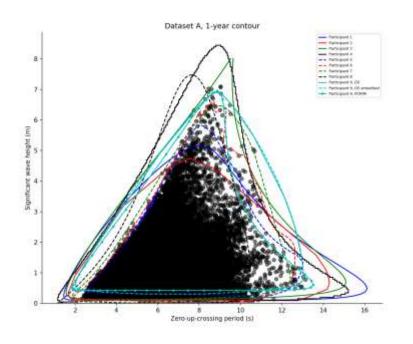
Overlay plots

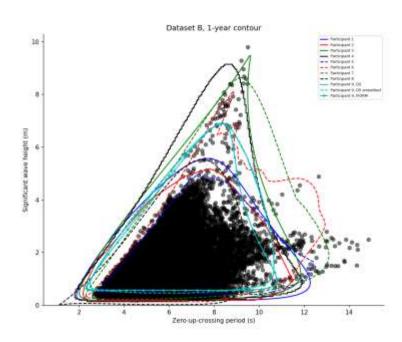


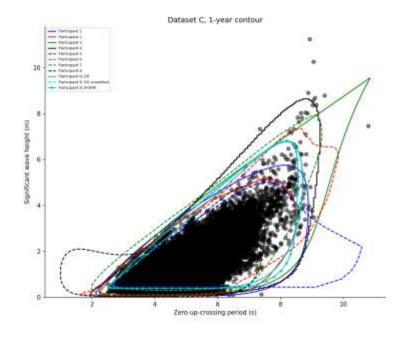


Overlay plots, datasets A-C, 1-yr contour

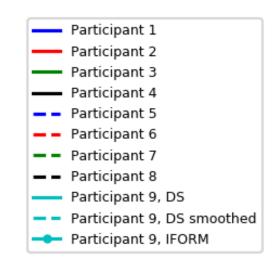


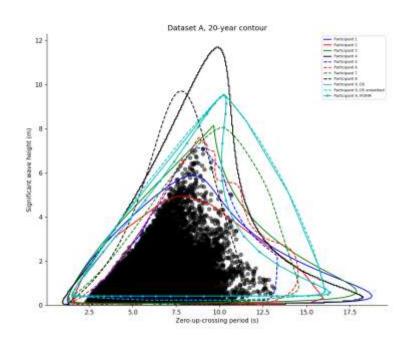


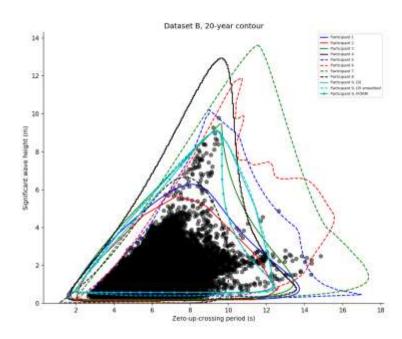


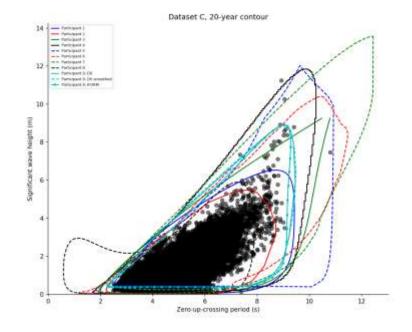


Overlay plots, datasets A-C, 20-yr contour

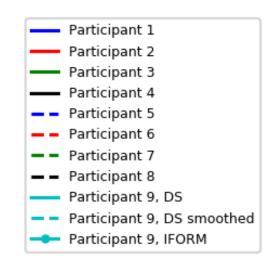


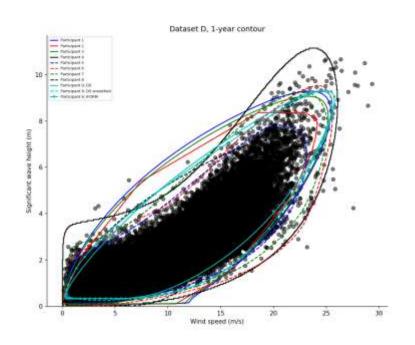


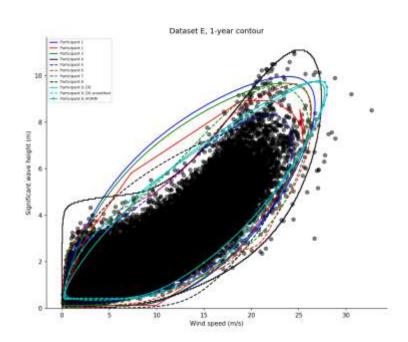


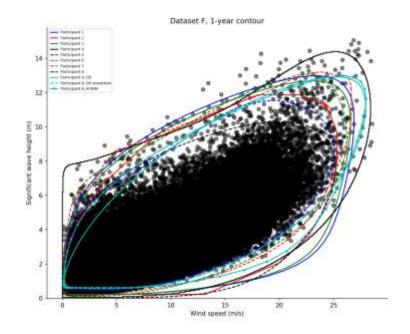


Overlay plots, datasets D-F, 1-yr contour

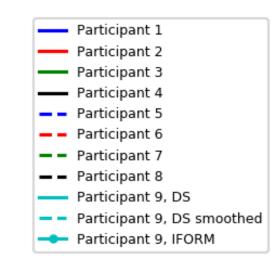


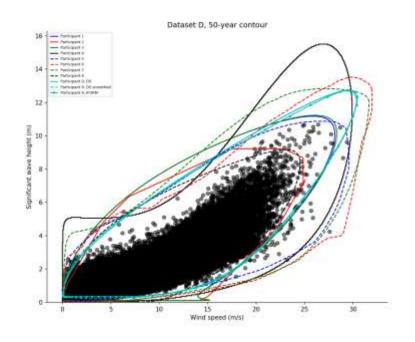


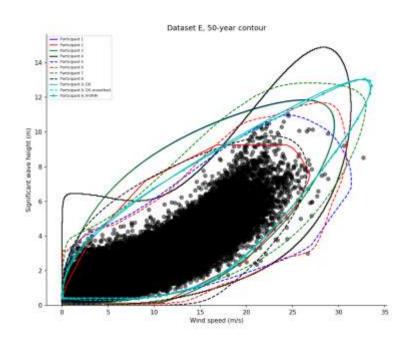


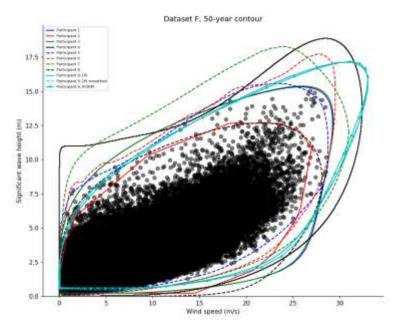


Overlay plots, datasets D-F, 50-yr contour









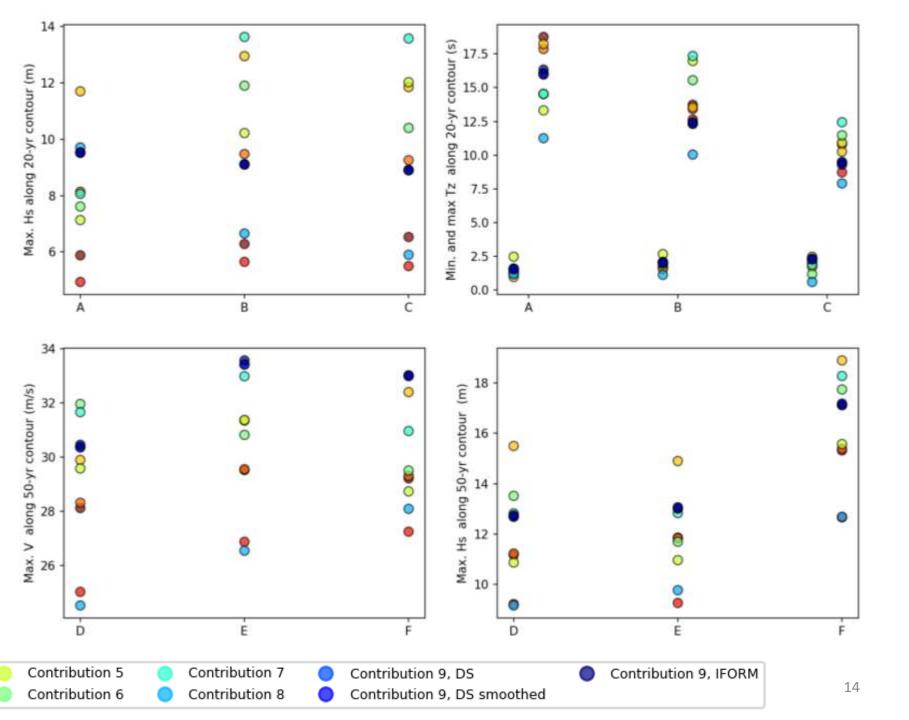
Maxima along the contours

Contribution 1

Contribution 2

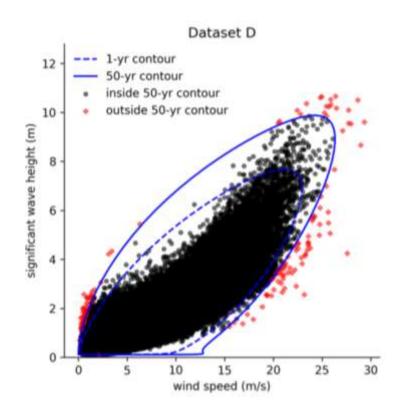
Contribution 3

Contribution 4



Points outside the contours

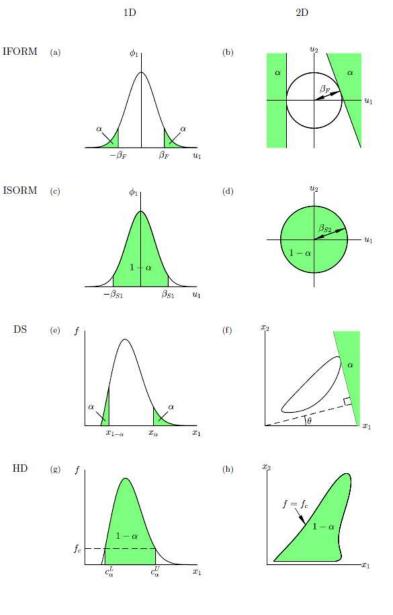
- Points exceeding the contour, n_E , at any region are counted
- Expected number of exceeding points can be calculated as $E[n_F] = N \cdot \alpha_T$
 - N = Number of data points a sample holds
 - α_T = Probability that an observations falls anywhere outside the contour
- Contour methods define the exceedance probability, α , differently
 - → Contours with same return period can have different expected number of points outside the contour, due to different contour methods



Points outside the contours

→ Contours with same return period can have different expected number of points outside the contour, due to different contour methods

Contour method	$\alpha_{ au}$ (total exceedance probability)
IFORM	$1 - \mathbf{\chi}_n^2(\beta_F^2) = 1 - \mathbf{\chi}_n^2([\Phi^{-1}(1-\alpha)]^2)$
ISORM	α
Direct Sampling	ca. similar to IFORM
Direct IFORM	ca. similar to IFORM
Highest density	α



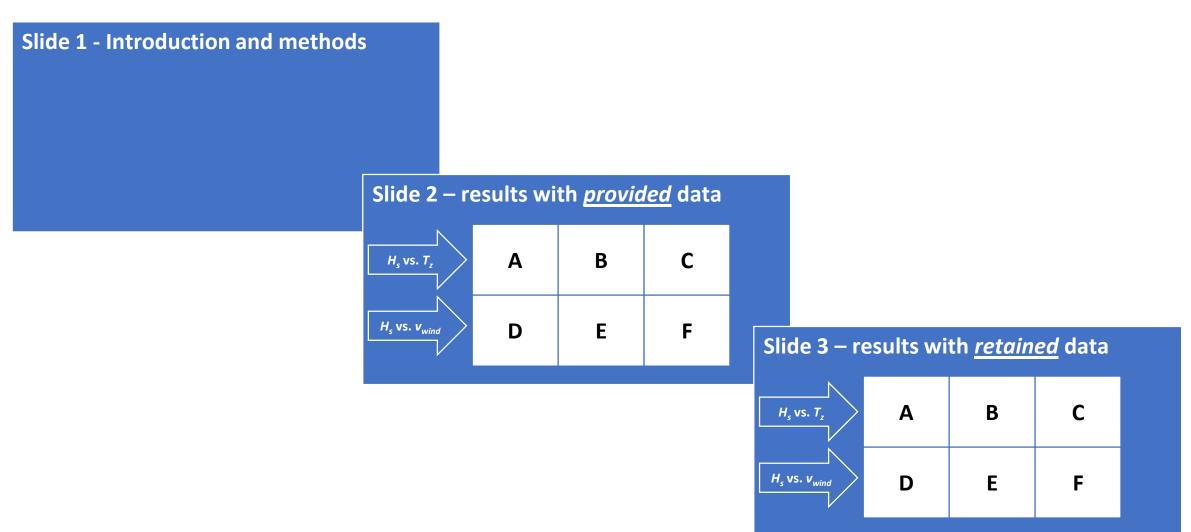
Points outside the H_s - T_z contours (provided & retained datasets)

Contribution	# points outside the 1-yr contour: mean (A, B, C)	Expected # points outside 1-yr contour	# points outside the 20-yr contour: mean (A, B, C)	Expected # points outside 20-yr contour
1	256.3 (388, 259, 122)	20 (ISORM)	114.0 (153, 127, 62)	1 (ISORM)
2	406.7 (596, 360, 264)	ca. 197 (DS)	286.7 (437, 236, 187)	ca. 11.5 (DS)
3	82.3 (97, 92, 58)	? (similar to IFORM?)	33.7 (40, 27, 34)	? (similar to IFORM?)
4	75.0 (25, 134, 66)	20 (HDC)	13.7 (1, 32, 8)	1 (HDC)
5	18295.3 (9381, 14710, 30795)	ca. 197 (Direct IFORM)	16764.0 (8503, 12573, 29216)	ca. 11.5 (Direct IFORM)
6	154.0 (170, 169, 123)	197 (IFORM)	16.7 (27, 14, 9)	11.5 (IFORM)
7	281.3 (370, 299, 175)	197 (IFORM)	35.3 (79, 4, 23)	11.5 (IFORM)
8	762.0 (599, 807, 880)	197 (IFORM)	322.0 (189, 368, 409)	11.5 (IFORM)
9 DS	22127.3 (22562, 23146, 20674)	ca. 197 (DS)	21966.3 (22504, 22933, 20462)	ca. 11.5 (DS)
9 DS smoothed	12031.7 (10459, 13838, 11798)	ca. 197 (DS)	6062.7 (5096, 7716, 5376)	ca. 11.5 (DS)
9 IFORM	22207.7 (22659, 23241, 20723)	197 (IFORM)	21994.7 (22514, 22990, 20480)	11.5 (IFORM)

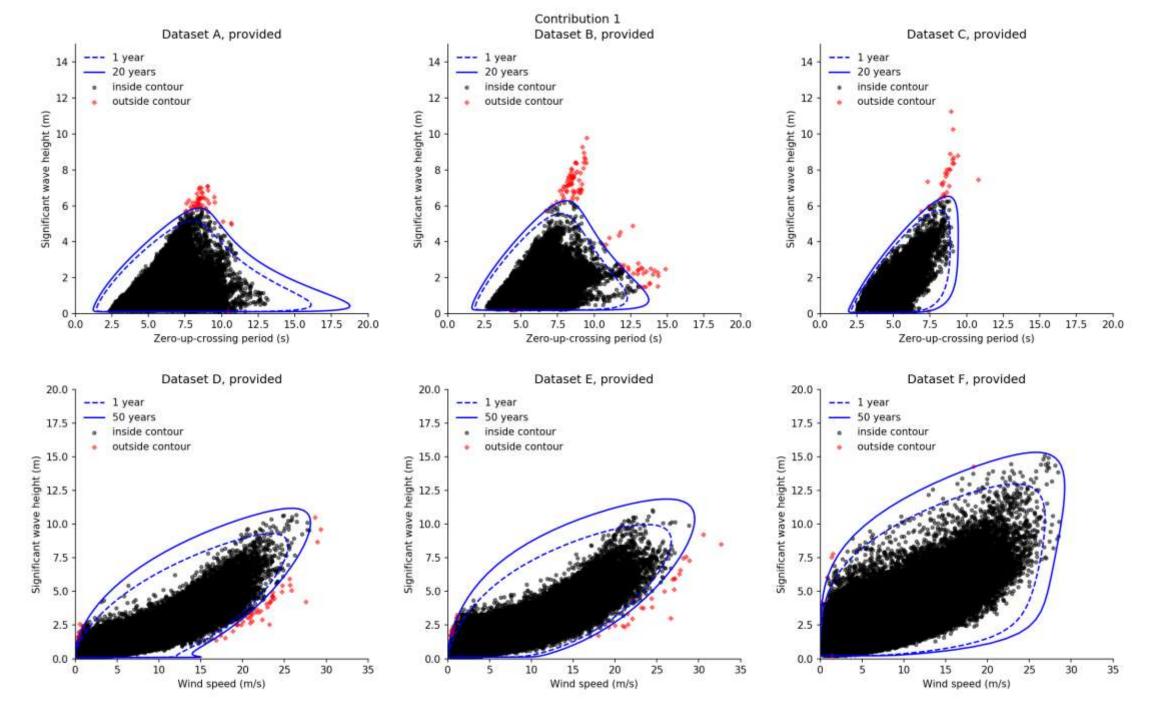
Points outside the V-H_s contours (provided & retained datasets)

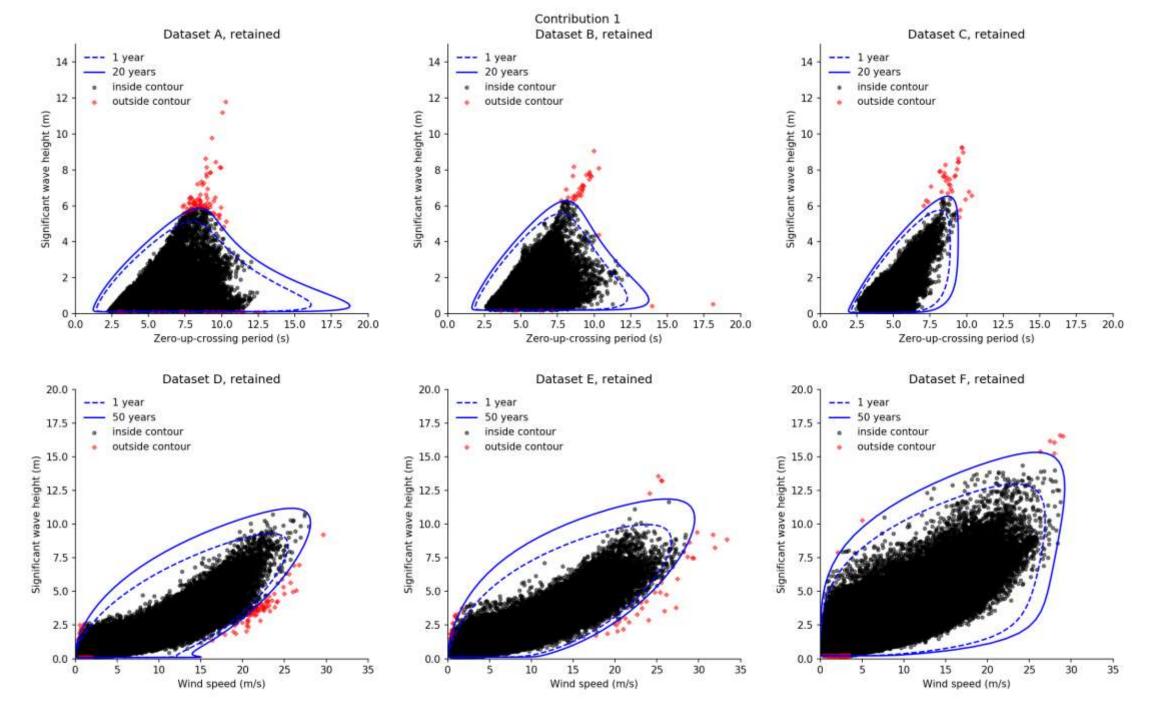
Contribution	# points outside the 1-yr contour: mean (D, E, F)	Expected # points outside 1-yr contour	# points outside the 50-yr contour: mean (D, E, F)	Expected # points outside 50-yr contour
1	424.3 (483, 519, 271)	50 (ISORM)	103.7 (154, 86, 71)	1 (ISORM)
2	1186.3 (1252, 1478, 829)	492 (IFORM)	529.7 (514, 720, 355)	12 (IFORM)
3	524.7 (592, 652, 330)	? (similar to IFORM?)	102.7 (154, 82, 72)	? (similar to IFORM?)
4	88.0 (64, 99, 101)	50 (HDC)	3.7 (3, 8, 0)	1 (HDC)
5	2235.0 (1875, 2036, 2794)	ca. 492 (Direct IFORM)	156.0 (174, 112, 182)	ca. 12 (Direct IFORM)
6	270.7 (265, 289, 258)	492 (IFORM)	8.3 (3, 17, 5)	12 (IFORM)
7	371.3 (294, 446, 374)	492 (IFORM)	8.7 (3, 16, 7)	12 (IFORM)
8	1729.7 (2977, 979, 1233)	492 (IFORM)	262.0 (509, 110, 167)	12 (IFORM)
9 DS	18191.7 (20337, 20481, 13757)	ca. 492 (DS)	15301.0 (17460, 17226, 11217)	ca. 12 (DS)
9 DS smoothed	12819.3 (12862, 14665, 10931)	ca. 492 (DS)	5771.3 (5092, 6782, 5440)	ca. 12 (DS)
9 IFORM	16267.3 (18427, 18404, 11971)	492 (IFORM)	15267.0 (17415, 17184, 11202)	12 (IFORM)

Participant results focus

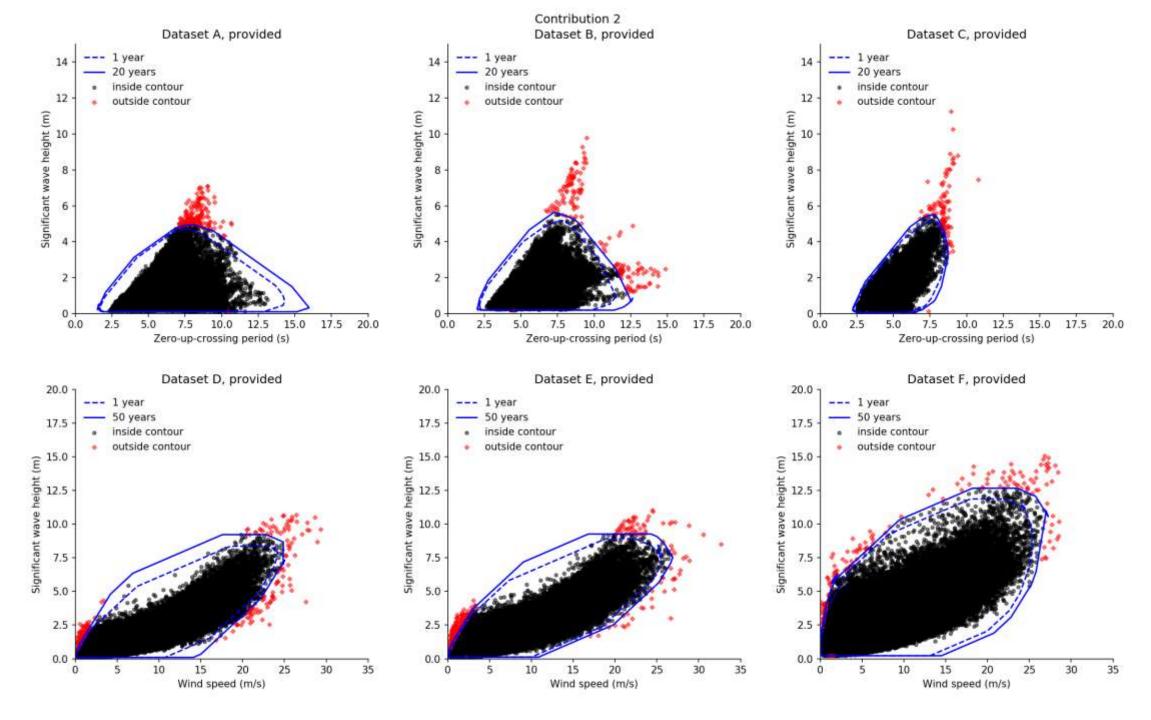


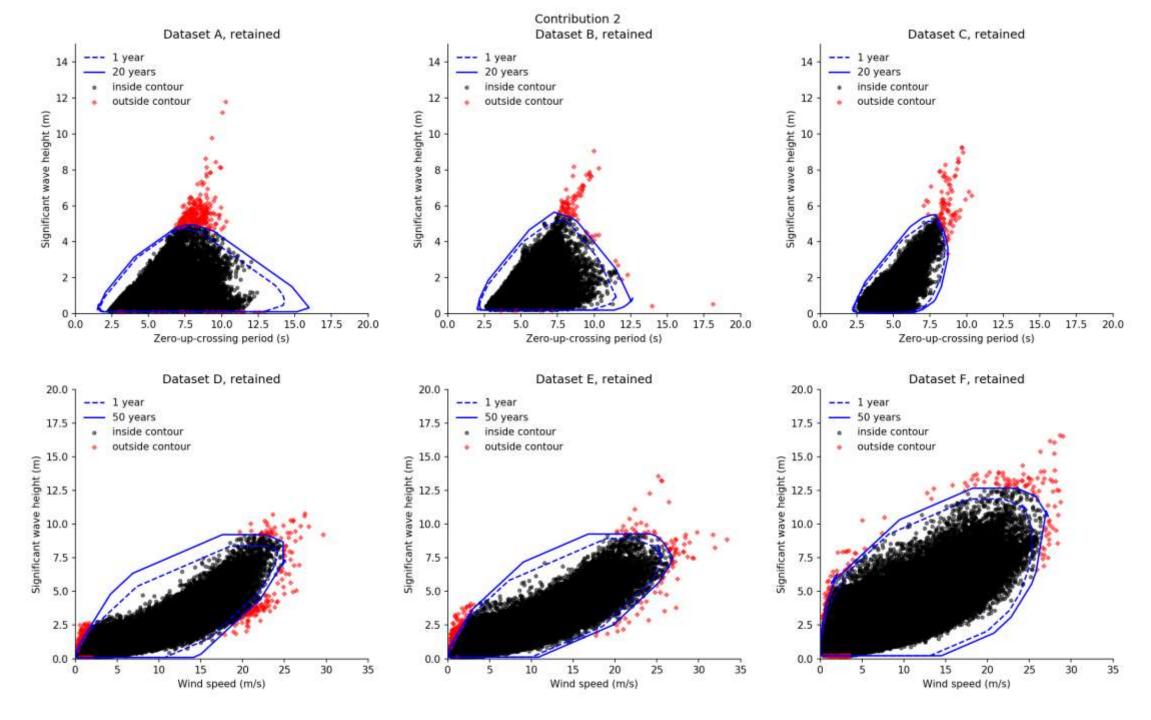
- Authors: Wei Chai & Bernt Leira
- Statistical model for datsets A-C: provided models in Table 2 of OMAE2019-96523
 - Previously published: OMAE2019-96523 (Andreas is the first author)
- Method to fit the model to datasets A-C: provided models in Table 2 of OMAE2019-96523
 - Previously published: OMAE2019-96523
- Statistical model for datasets D-F: provided models in Table 2 of OMAE2019-96523
 - Previously published: OMAE2019-96523
- Methods to fit the model to datasets D-F: OMAE2019-96523
 - Previously published: OMAE2019-96523
- Method used for contour construction: Inverse Second Order Reliability Method (ISORM)
 - Previously published: Chai, W, Leira B. J. Environmental contours based on inverse SORM, Marine Structures, 2018, 60: 34-51.
- Comment:



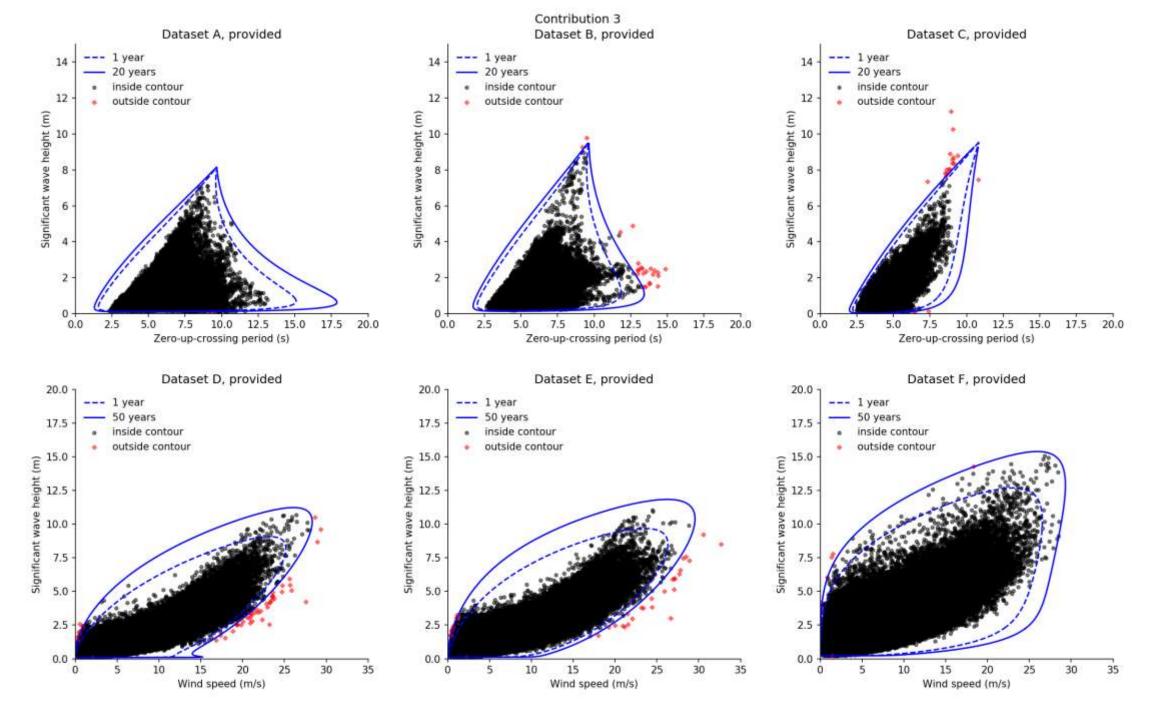


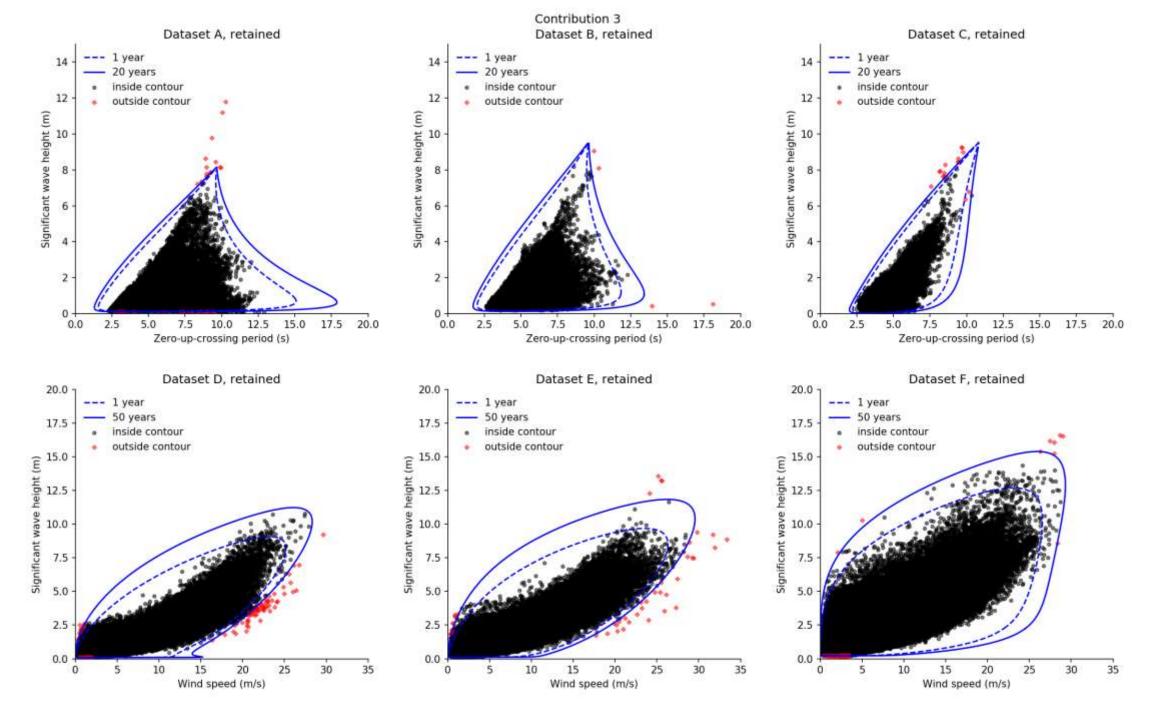
- Authors: Guilherme Clarindo & Carlos Guedes Soares
- Statistical model for datsets A-C: CMA joint distribution with 3P Weibull and conditioned lognormal distributions
 - Previously published: DNV-GL. 2017. Recommended practice DNVGL-RPC205: Environmental conditions and environmental loads. Tech report.
- Method to fit the model to datasets A-C: Maximum likelihood for Hs marginals, least squares for Tz dependence
 - Previously published:
- Statistical model for datsets D-F: CMA joint distribution with 3P Weibull and conditioned 2P Weibull distributions
 - Previously published: DNV-GL. 2017. Recommended practice DNVGL-RPC205: Environmental conditions and environmental loads. Tech report.
- Methods to fit the model to datasets D-F: Maximum likelihood for Hs marginals, least squares for U10 dependence
 - Previously published:
- Method used for contour construction: Direct Monte Carlo Simulations Approach
 - Previously published: Huseby, A. B., Vanem, E. & Natvig, B. 2013. A new approach to environmental contours for ocean engineering applications based on direct Monte Carlo simulations. Ocean Engineering. (60): 124–135.
- Comment:



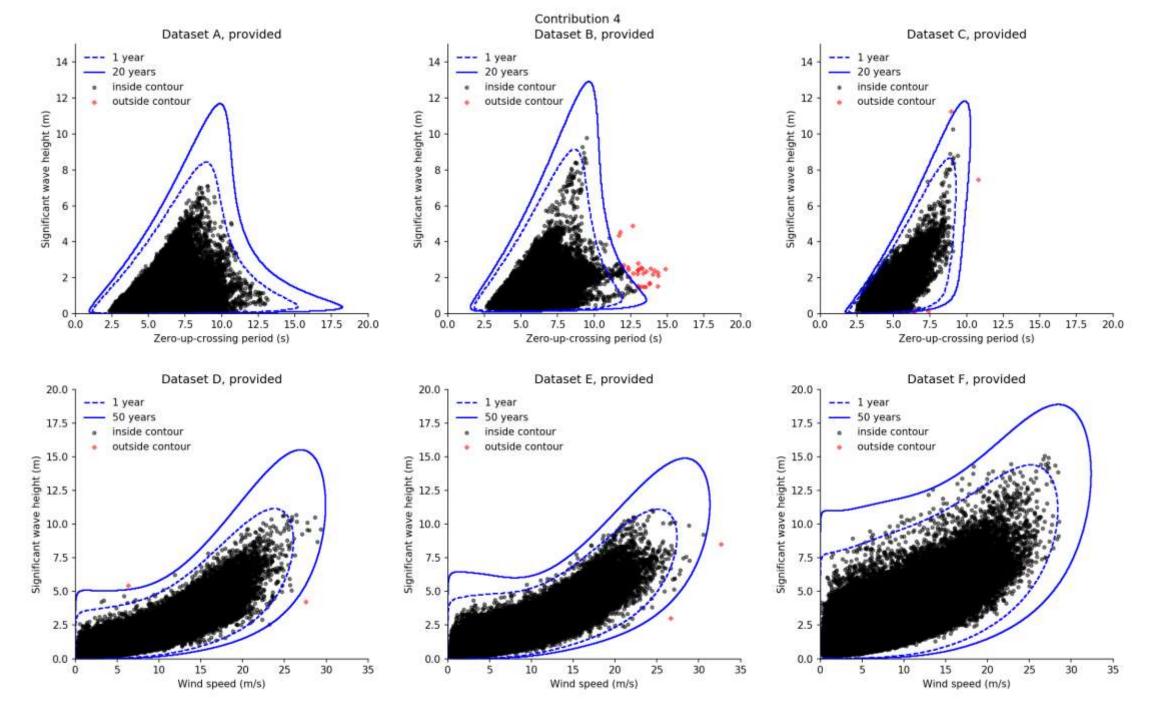


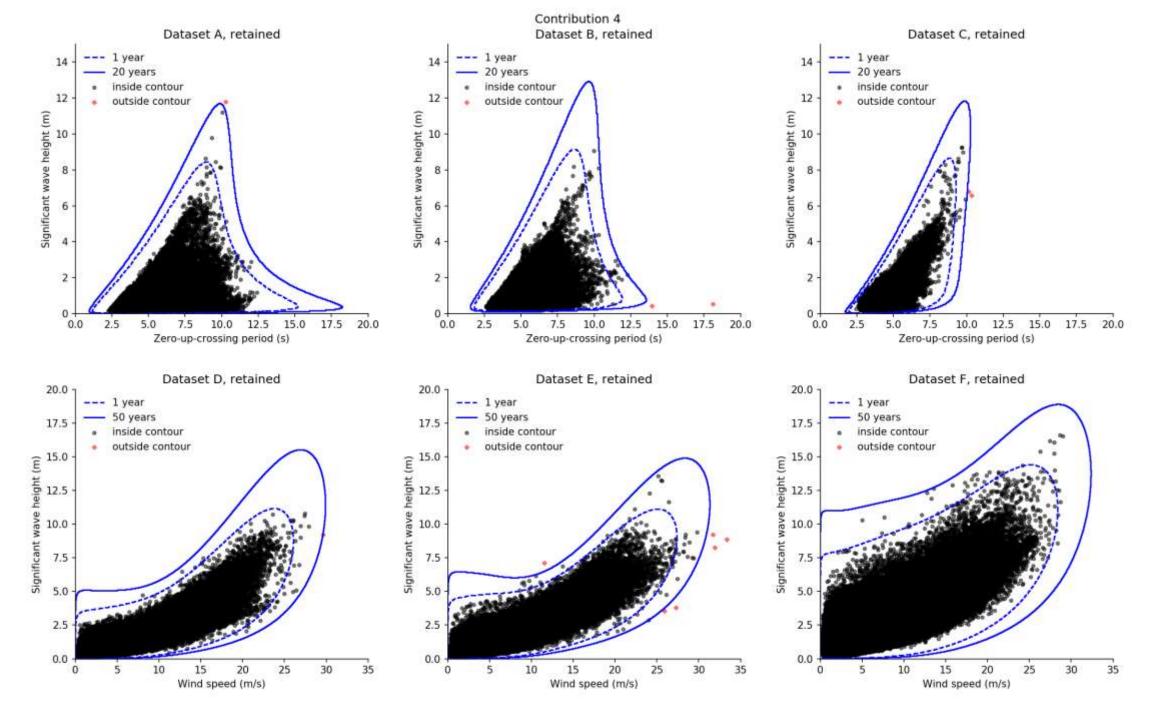
- Author: Ásta Hannesdóttir
- Statistical model for datasets A-C: CMA joint distribution with Weibull and lognormal distributions
 - Previously published: DNV-RP-C205, 2010
- Method to fit the model to datasets A-C: MLE for maginals, least squares for dependence functions
 - Previously published: Hannesdóttir, Á., Kelly, M., and Dimitrov, N.: Extreme wind fluctuations: joint statistics, extreme turbulence, and impact on wind turbine loads, Wind Energ. Sci., 4, 325–342, https://doi.org/10.5194/wes-4-325-2019, 2019
- Statistical model for datasets D-F: CMA joint distribution with Weibull 3-p. and Weibull 2-p.
 - Previously published: DNV-RP-C205, 2010
- Methods to fit the model to datasets D-F: MLE for maginals, least squares for dependence functions
 - Previously published:
- Method used for contour construction: Inverse directional simulation (IDS) contour method
 - Previously published: Dimitrov, N.: Inverse Directional Simulation: an environmental contour method providing an exact return period. Manuscript submitted for publication in Journal of Physics: Conference Series, 2020.
- Comment:





- Authors: Andreas Haselsteiner, Aljoscha Sander, Jan-Hendrik Ohlendorf & Klaus-Dieter Thoben
- Statistical model for datasets A-C: Global hierarchical model with an Exp. Wbl. distribution and an LN distribution
 - Previously published:
- Method to fit the model to datasets A-C: Weighted least squares for Hs, MLE for Tz, nonlinear least squares for dependence function
 - Previously published: Weighted least squares for Hs: Haselsteiner, A. F., & Thoben, K.-D. (2020). Predicting wave heights for marine design by prioritizing extreme events in a global model. Renewable Energy, 156, 1146–1157. https://doi.org/10.1016/j.renene.2020.04.112
- Statistical model for datasets D-F: Global hierarchical model with Exp Wbl. for V and Hs
 - Previously published:
- Methods to fit the model to datasets D-F: Weighted least squares for marignals, weighted least square for dependence
 - Previously published: Weighted least squares for Hs: Haselsteiner, A. F., & Thoben, K.-D. (2020). Predicting wave heights for marine design by prioritizing extreme events in a global model. Renewable Energy, 156, 1146–1157. https://doi.org/10.1016/j.renene.2020.04.112
- Method used for contour construction: Highest density contour method
 - Previously published: DOI: 10.1016/j.coastaleng.2017.03.002
- Comment:



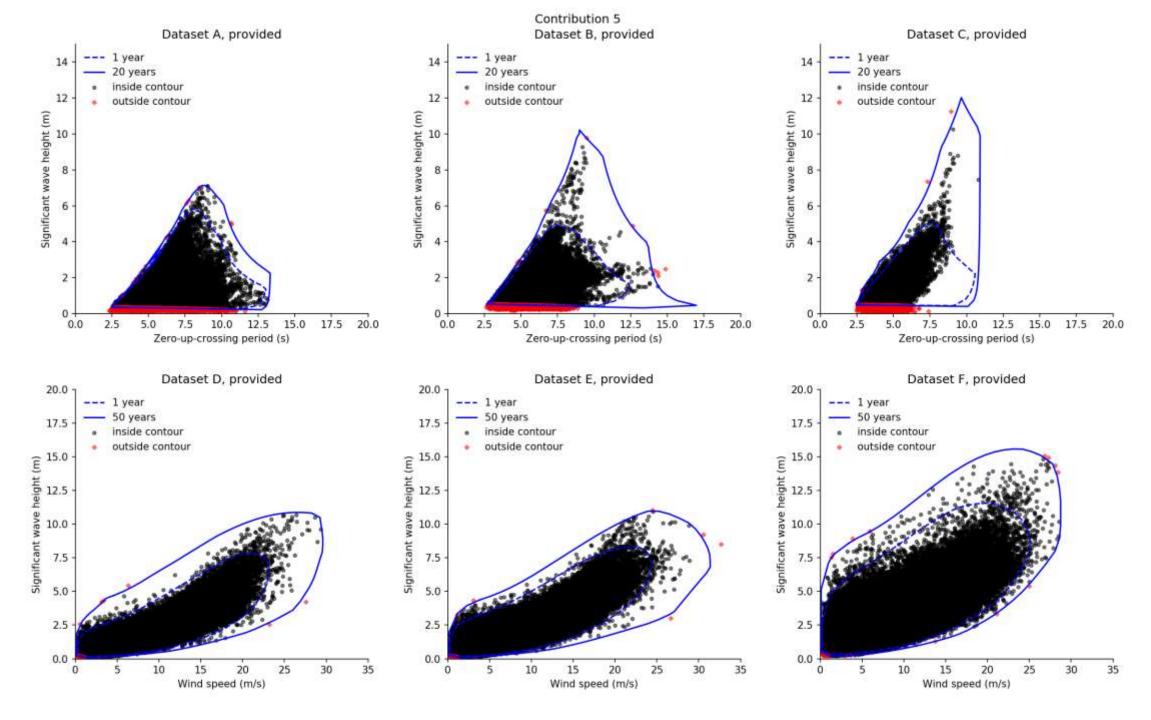


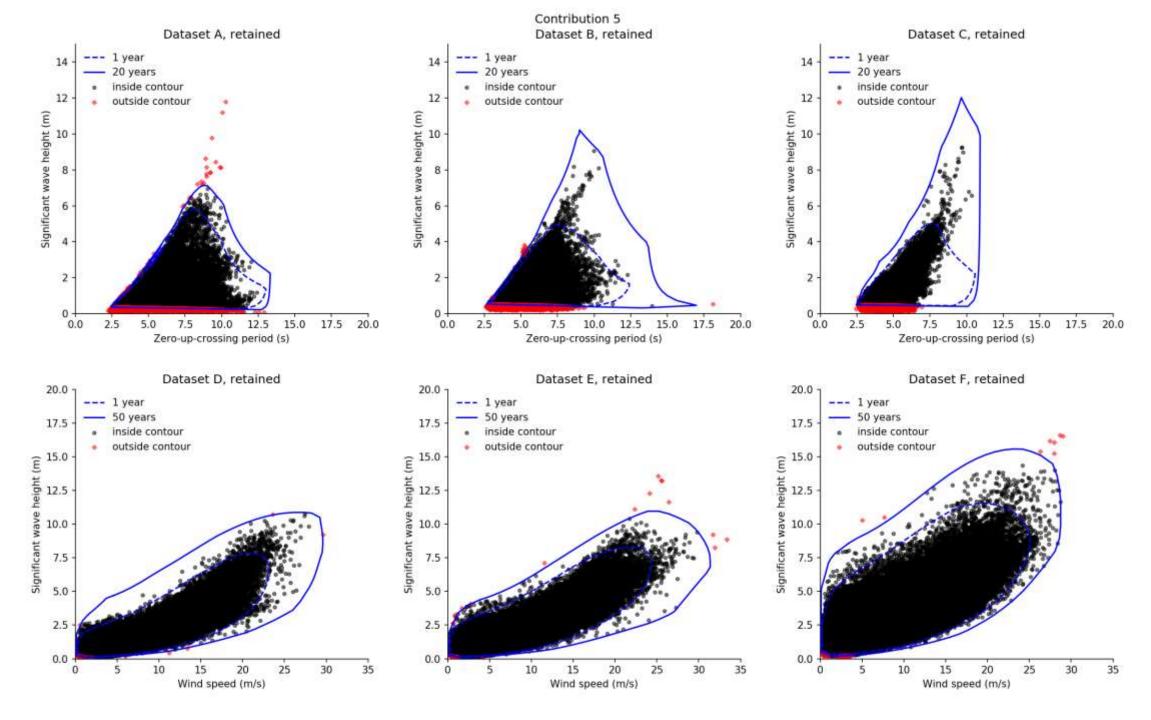
- Authors: Guillaume de Hauteclocque
- Statistical model for datasets A-C: Peak-Over-Threshold, Generalized Pareto Distribution
 - Previously published: Standard text book, used in OMAE2019-95993
- Method to fit the model to datasets A-C: MLE
 - Previously published: Standard text book, used in OMAE2019-95993
- Statistical model for datasets D-F: Peak-Over-Threshold, Generalized Pareto Distribution
 - Previously published: Standard text book, used in OMAE2019-95993
- Methods to fit the model to datasets D-F: MLE
 - Previously published: Standard text book, used in OMAE2019-95993
- Method used for contour construction: Direct IFORM with de-clustering
 - Previously published: OMAE2019-95993
- Comment:

Compared to OMAE2019-95993, different variable changes have been used to get dataset roughly convex:

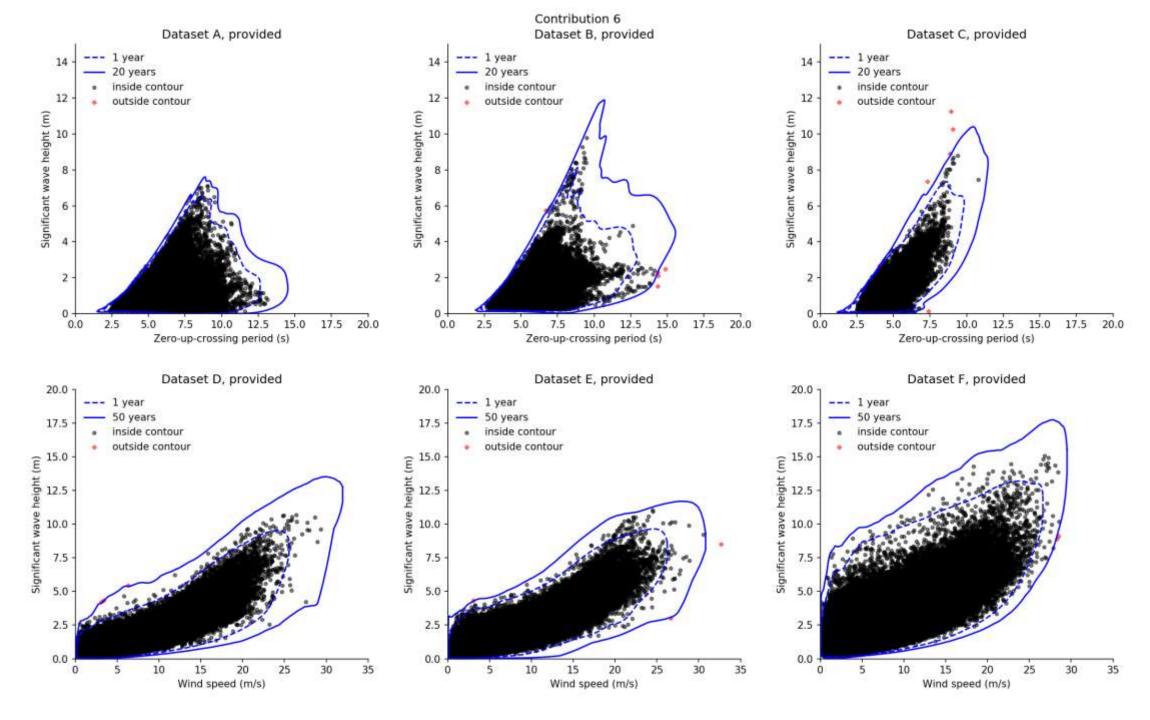
- (Hs, Tz) changed to (Hs, Hs*Tz)
- (Hs, Ws) changed to (Ws, Hs / f(Ws))

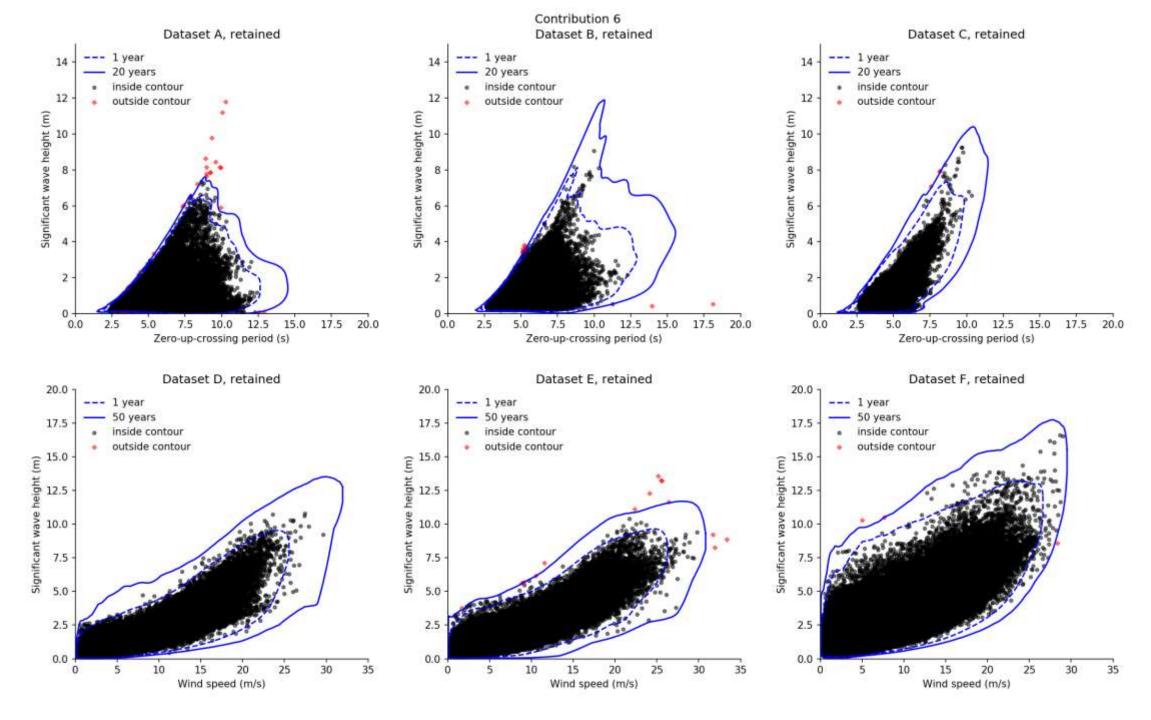
Attention has been paid only to the relevant part of the contour (the upper part)





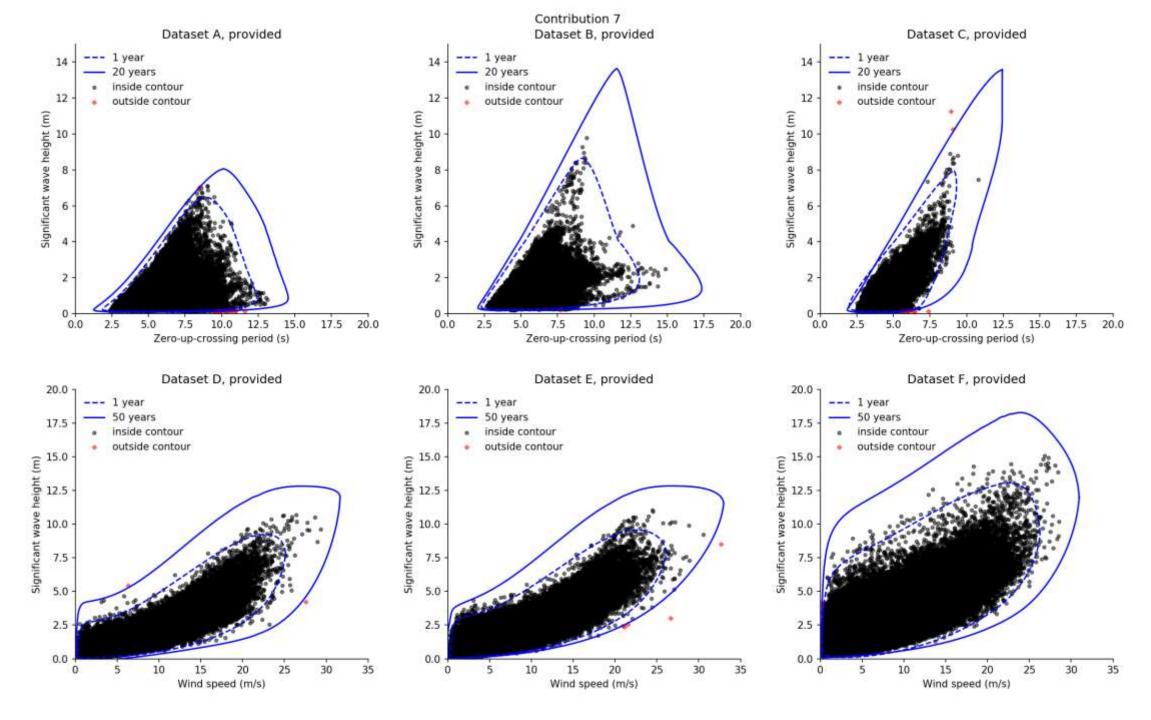
- Authors: Ed Mackay & Philip Jonathan
- Statistical model for datasets A-C: Storm resampling with non-stationary model for storms peaks
 - Previously published: Ref [1]
- Method to fit the model to datasets A-C: See reference
 - Previously published:
- Statistical model for datasets D-F: As above
 - Previously published:
- Methods to fit the model to datasets D-F:
 - Previously published:
- Method used for contour construction: IFORM
 - Previously published:
- Comment: [1] Mackay & Jonathan, 2020. Estimation Of Environmental Contours Using A Block Resampling Method. OMAE2020-18308.

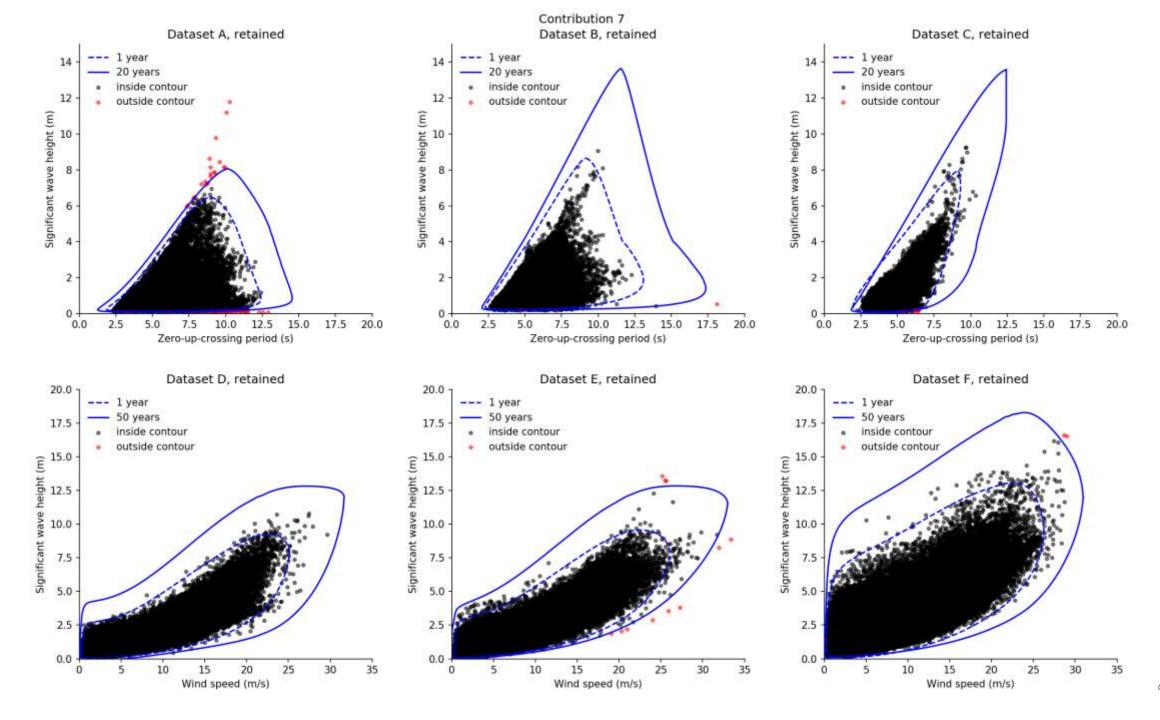




Exercise 1 – contribution 7

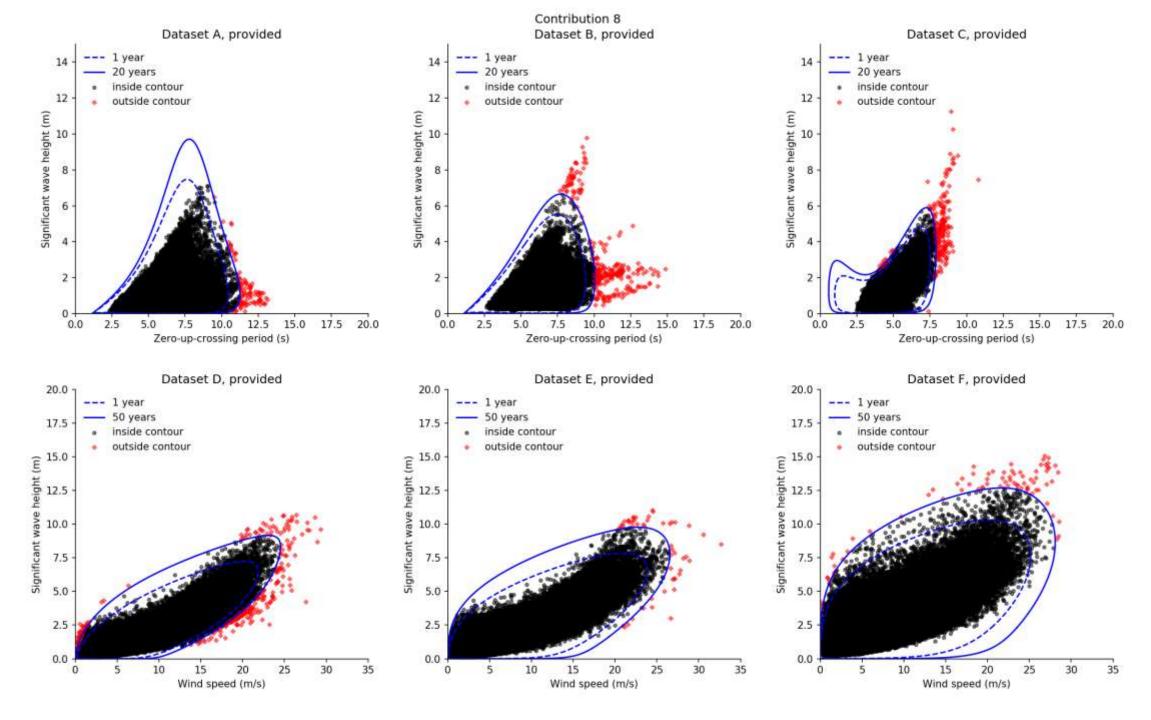
- Authors: Chi Qiao & Andrew Myers
- Statistical model for datasets A-C:
 - Previously published:
- Method to fit the model to datasets A-C:
 - Previously published:
- Statistical model for datasets D-F:
 - Previously published:
- Methods to fit the model to datasets D-F:
 - Previously published:
- Method used for contour construction: IFORM contour method
 - Previously published: Winterstein et al. (1993): Environmental parameters...
- Comment:
 - 1. A new framework based on I-FORM and Rosenblatt transformation is established for contour construction.
 - 2. Univariate distribution fitting is conducted using a method proposed by the authors in a manuscript that is currently under review.
 - 3. Lower half of the contour is constructed by MLE
 - 4. Upper half of the contour is constructed using a re-parameterization method proposed by the authors in Qiao, C., and A. T. Myers. "Re-parameterized Weibull distribution for modeling metocean extremes of multiple hazards with the Rosenblatt transformation."
 - 5. A Python package for interactive environmental fitting is being developed and will be published on PyPI

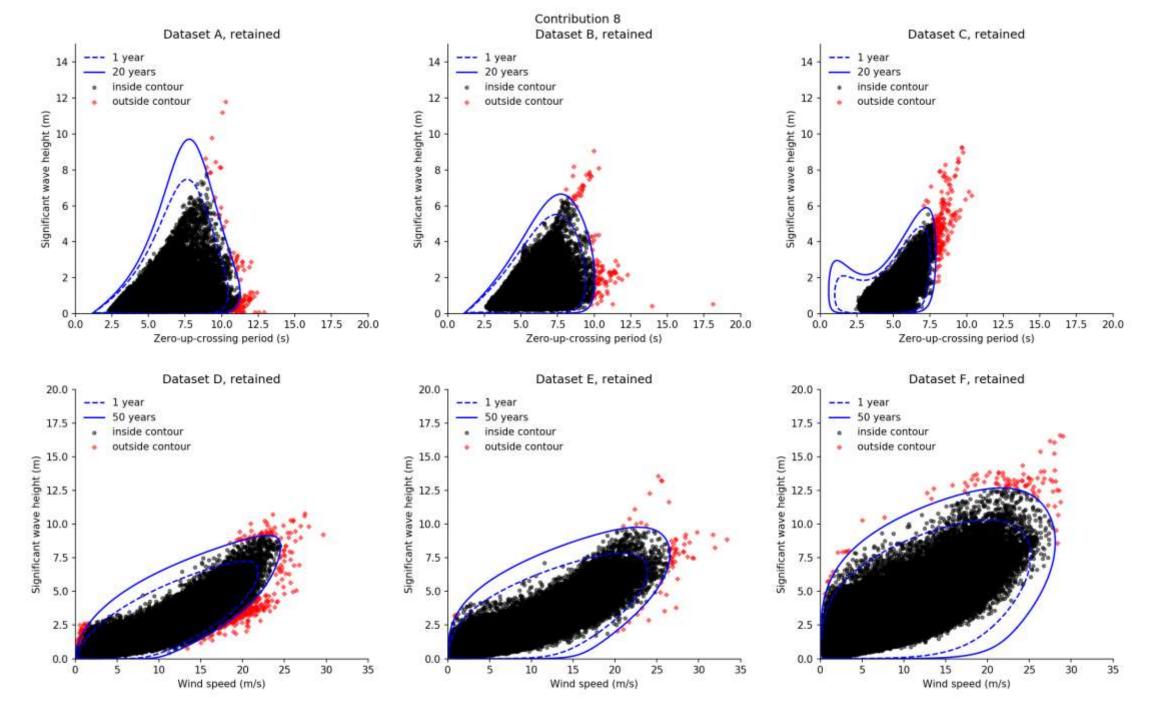




Exercise 1 – contribution 8

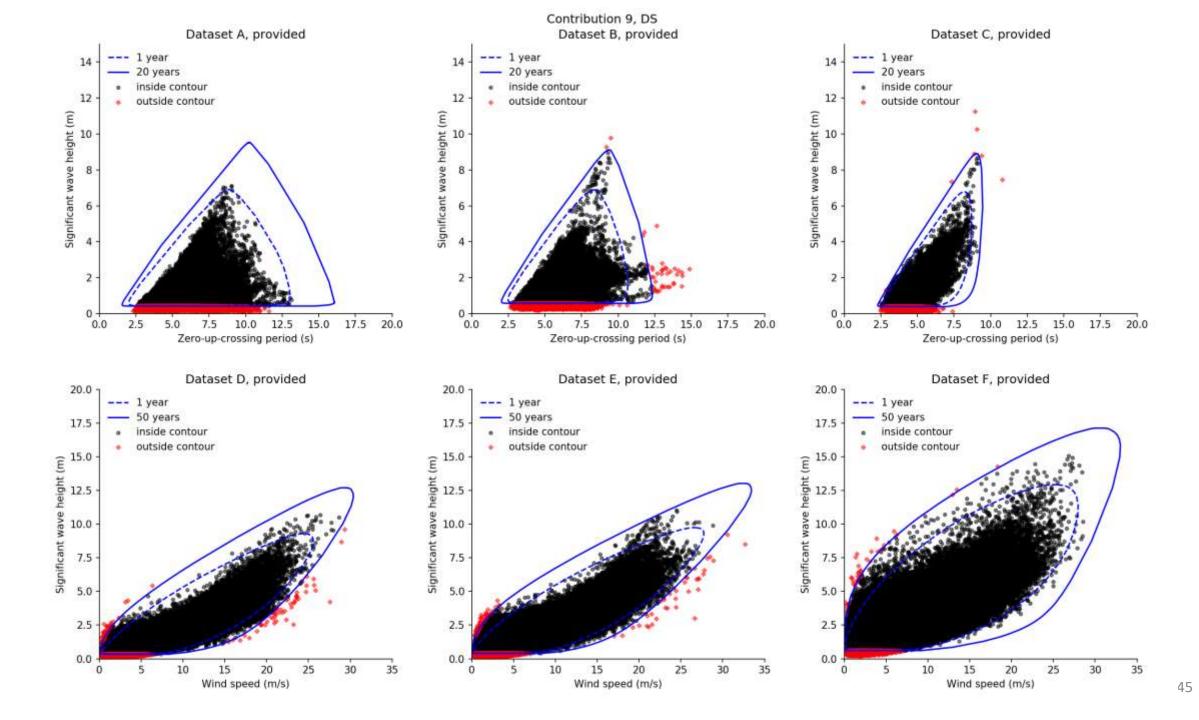
- Authors: Anna Rode, Arndt Hildebrandt & Boso Schmidt
- Statistical model for datasets A-C: 2-Parameter Weibull distribution
 - Previously published:
- Method to fit the model to datasets A-C: Maximum-Likelihood for marginals, OLS for dependence
 - Previously published:
- Statistical model for datasets D-F: 2-Parameter Weibull distribution
 - Previously published:
- Methods to fit the model to datasets D-F: Maximum-Likelihood for marginals, OLS for dependence
 - Previously published:
- Method used for contour construction: IFORM contour method
 - Previously published: Winterstein et. al, in "Environmental Parameters for extreme Response
- Comment:

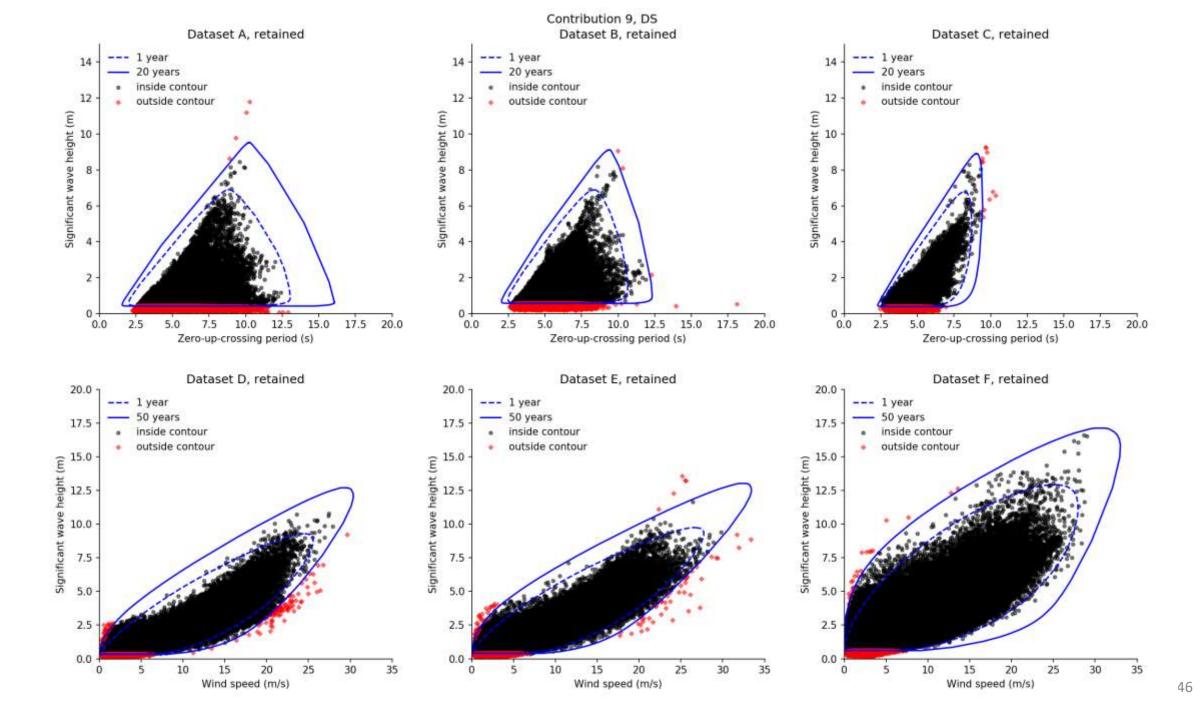


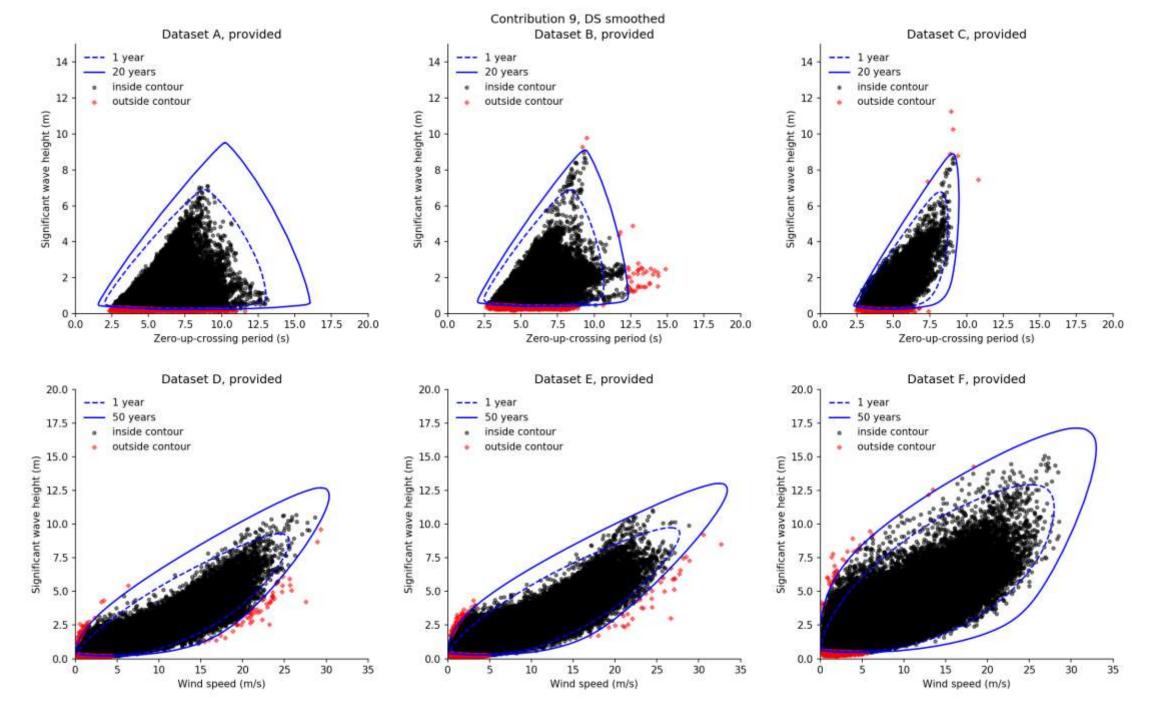


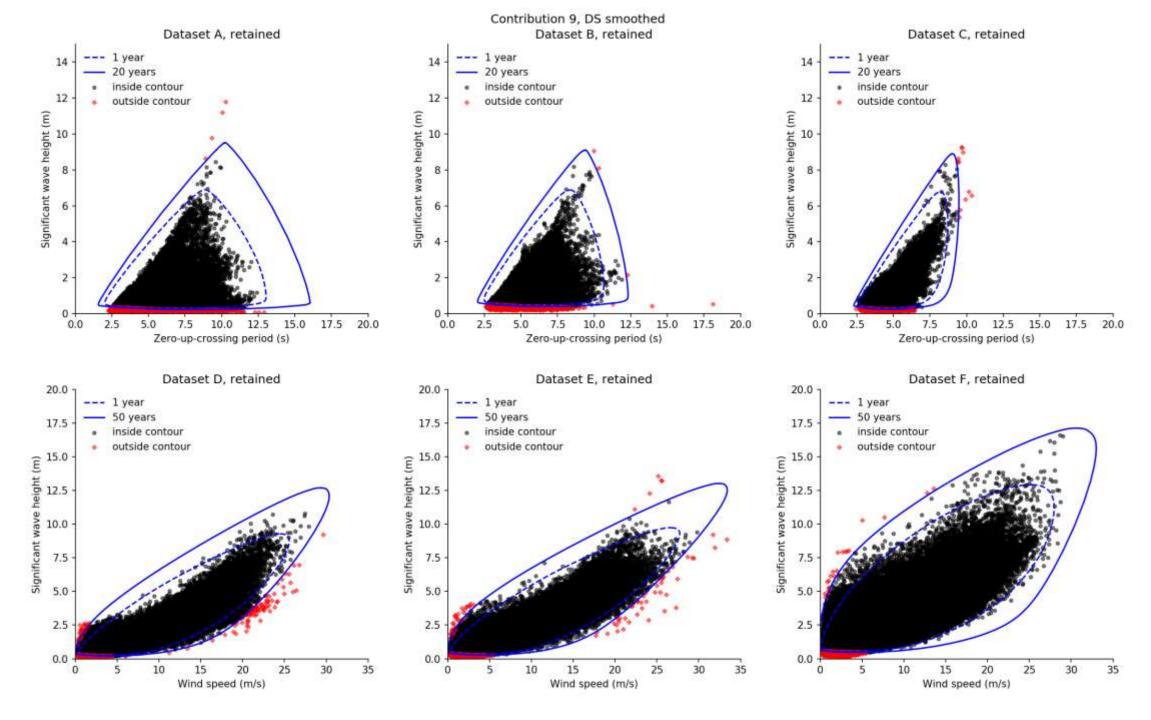
Exercise 1 – contribution 9

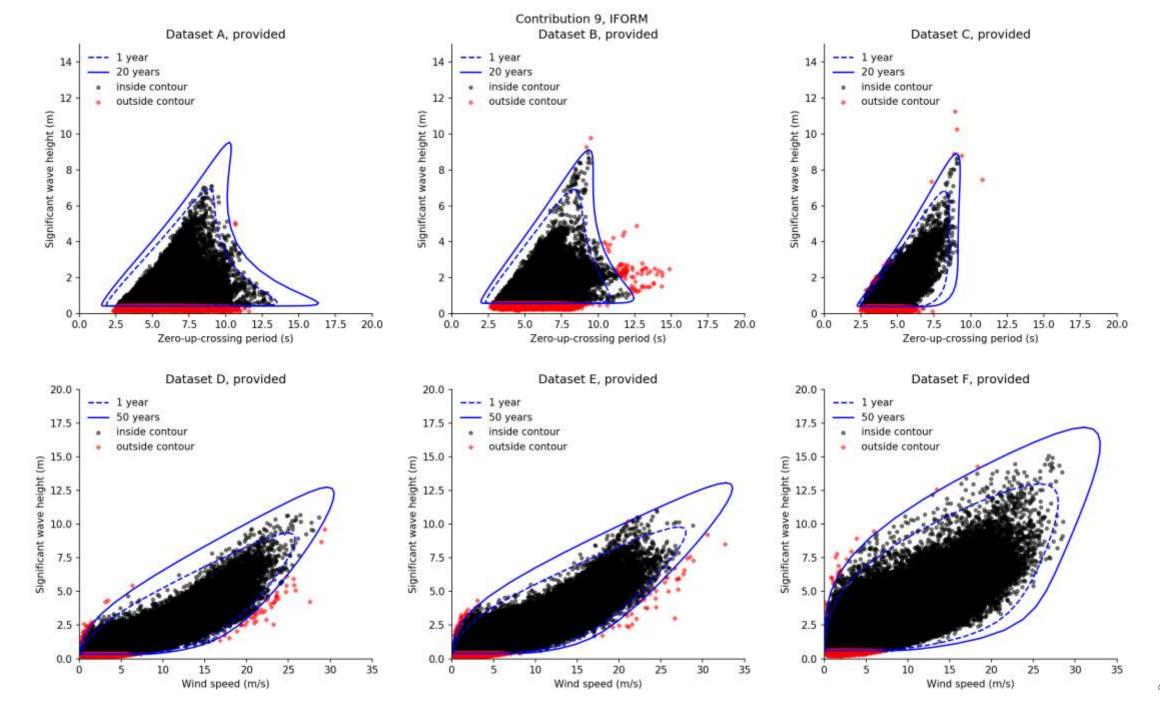
- Authors: Erik Vanem & Arne Bang Huseby
- Statistical model for datasets A-C: 3-parameter Weibul and conditional log-normal
 - Previously published: e.g. DNV GL RP C205, but many other places
- Method to fit the model to datasets A-C: GoF statistic and least square to binned data for conditional mod
 - Previously published: e.g. Vanem (2016); Marine Structures. vol. 49, pp. 180-205, 2016
- Statistical model for datasets D-F: 3-parameter Weibull and conditional 2-parameter Weibull
 - Previously published: e.g. DNV GL RP C205, but many other places
- Methods to fit the model to datasets D-F: GoF statistic and ML
 - Previously published: GoF: e.g. in Vanem (2015): . Journal of Ocean Engineering and Marine Energy, vol. 1(4), pp. 339-359, 2015. See also Vanem et al. (2019): Applied Ocean Research, 91, 101870, 2019
- Method used for contour construction: Direct sampling contours and IFORM contours
 - Previously published: E.g. Huseby et al. (2013; 2014): Ocean Engineering, vol 60, pp. 124-135, 2013; Structural Safety, vol. 54, pp. 32-45, 2015; Vanem and Bitner-Gregersen (2015): Journal of Offshore Mechanics and Arctic Engineering, vol. 137(5), 051601, pp. 1-8, 2015; Vanem (2017): Marine Structures. 56, pp. 137-162, 2017. See also earlier work by Haver and Winterstein.
- Comment:
 - For the direct sampling contours in Ex. 1, there is an initial and a smoothed version of the contours. Possibly, the smoothed versions do not need to enter the exercise.
 - For Ex. 2 results are only submitted for the direct sampling contours, and contours for mean, median, lower and upper are provided (where lower/upper corresponds to the 95% CI. Please note that the CIs are calculated from the C-function and not exactly as prescribed in the announcement.

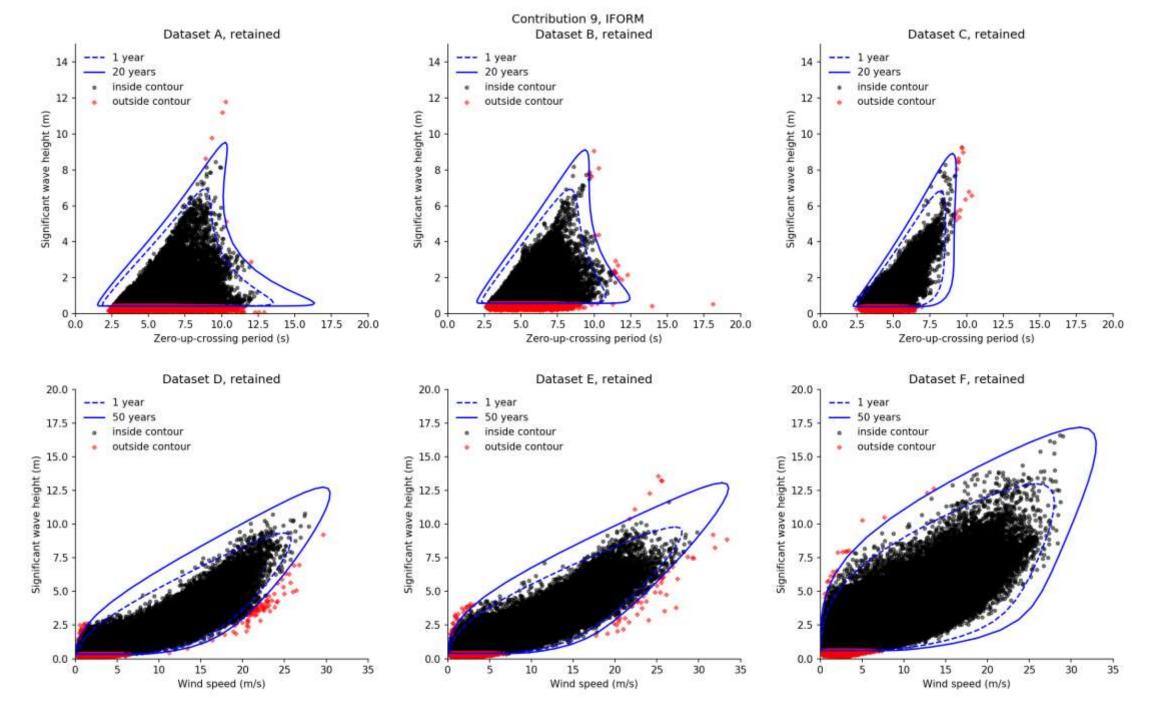




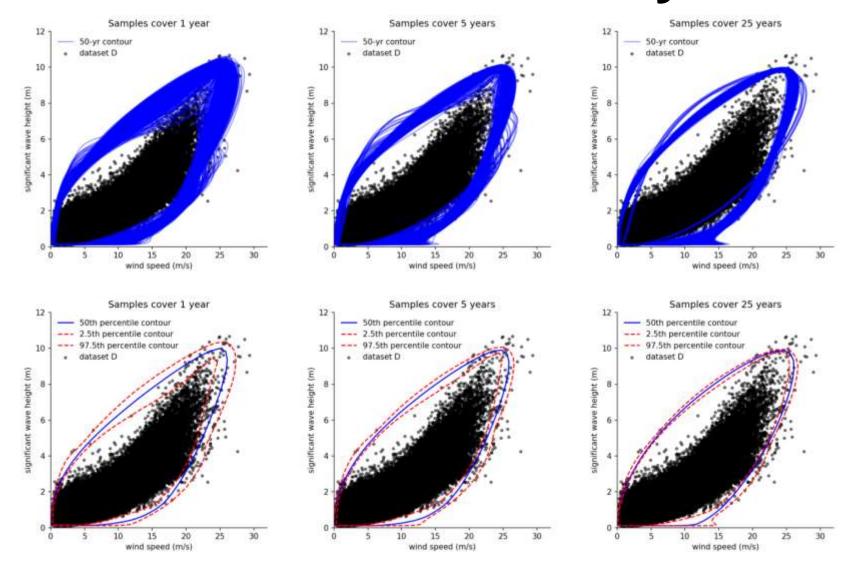




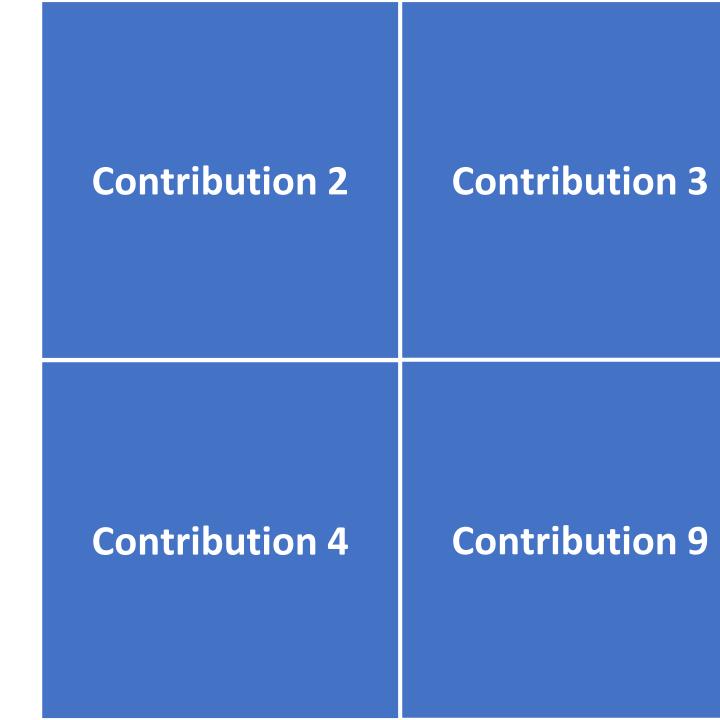




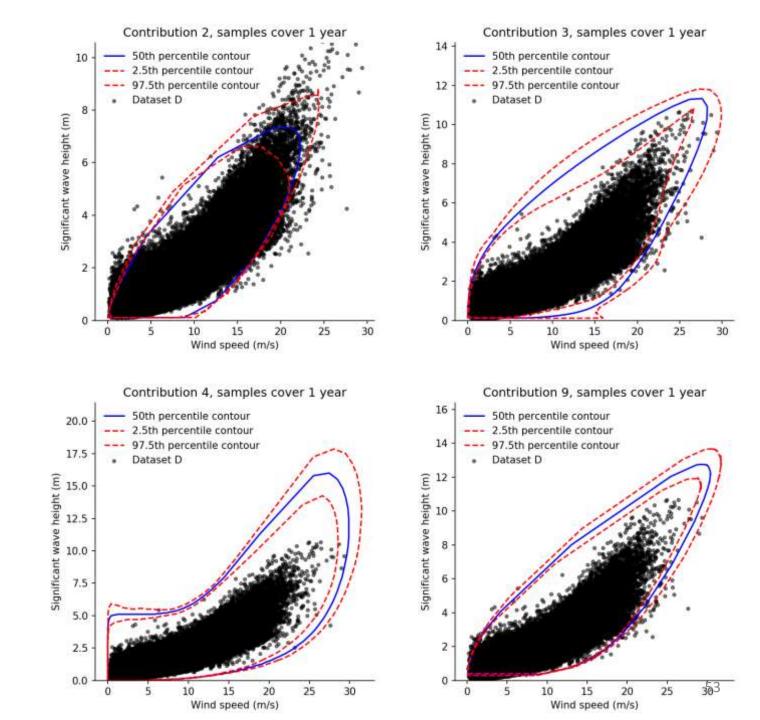
Exercise 2 - uncertainty



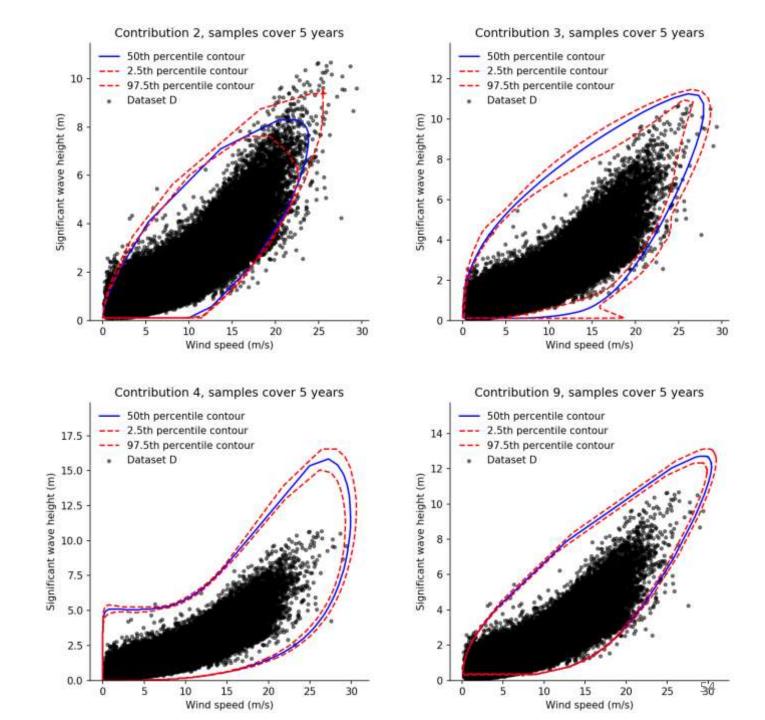
Exercise 2 uncertainty



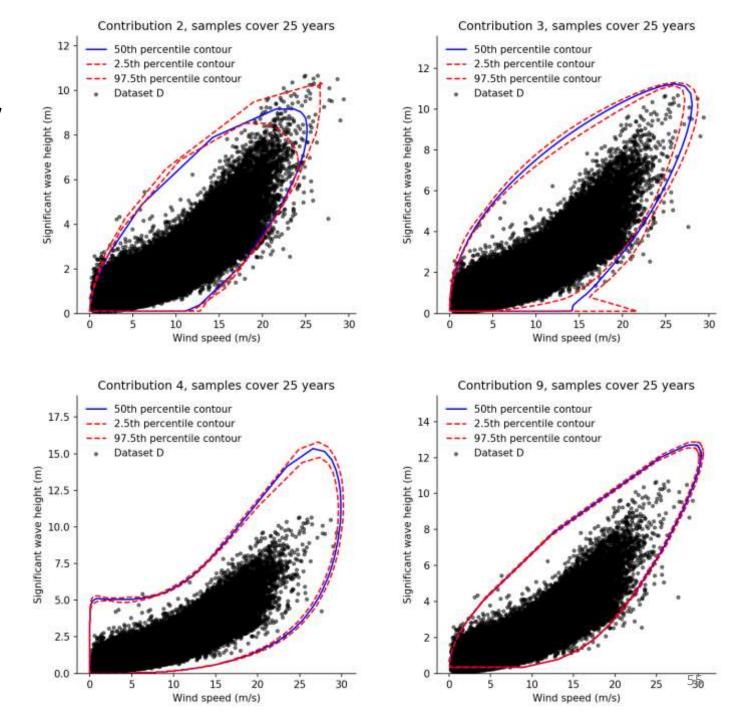
Exercise 2 uncertainty, 1 yr



Exercise 2 uncertainty, 5 yr



Exercise 2 uncertainty, 25 yr



iii) Overview about the proposed structure for the joint paper

Switch to https://github.com/ec-benchmark-organizers/ec-benchmark-organizers/ec-benchmark/blob/master/joint-paper-joint-paper-outline.md

Publishing the results

- We proposed a special issue entitled "Probabilistic Long-Term Extreme Conditions using Metocean Data" to Ocean Engineering
- If the journal is interested, we would propose to submit the joint paper
- Timeline
 - August 31st: Deadline for first complete draft
 - September 15th: Deadline for submitting comments on the first draft
 - September 30st: Deadline for revised draft
 - October 15th: Deadline for comments on revised draft
 - October 31st: Submission of the manuscript
- The special issue would also be a good place for standalone papers based on new models & methods that were proposed in individual contributions
- Proposed timeline: initial submissions between September 1 and November 16

iv) Decisions on the proposed structure, decisions on proposals for further analysis

Paper structure & further analysis

- Proposed main content [x]
- Proposal A: Assessment of statistical models in terms of marginal and projected variables []
 - Include the proposed analysis for participants who are interested in it

v) Decisions on the next steps and the timeline

Next steps and time line

- Sharing results
 - Share submitted files on GitHub repository [x]
 - Share whole/subset of this presentation on GitHub repository [x]
- Writing the paper
 - August 31st: Deadline for first complete draft [x]
 - October 31st: Submission of the manuscript [x]
- Move special issue deadline backwards

vi) Discussion and further points

Protocol for discussion

• Schedule another meeting for discussion in ca. two weeks