A Unified Framework for Analyzing Non-Stationary Joint Extremes Using Transformed-Stationary EVA

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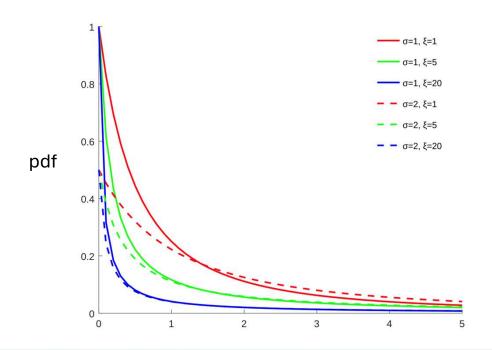






Probabilistic study of extremes

- Risk assessment of natural hazards: flooding, drought, etc.
- The theory: extreme value analysis (EVA)

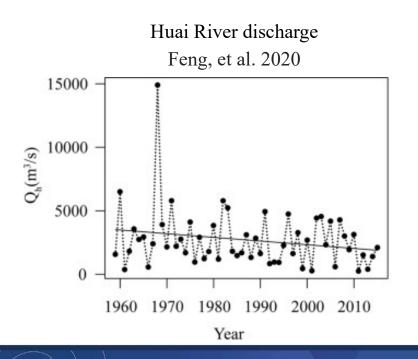


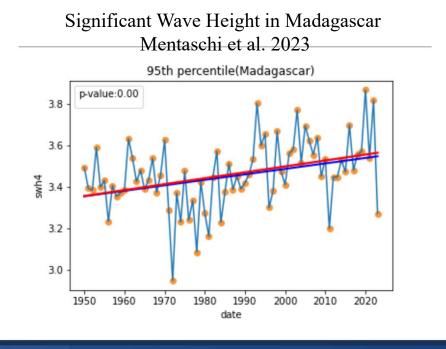




Stationarity requirement

- The classical EVA require stationarity of the underlying series
- Many datasets, especially those related to natural hazards, indicate non-stationarity







Transformed-stationary Extreme Value Analysis: tsEVA

A freely available toolbox in GitHub: https://github.com/menta78/tsEva

Geophysical Research Letters

RESEARCH LETTER

10.1002/2016GL072488

Key Points:

- Extreme waves will change along a large portion of the coasts generally increasing in the S. Hemisphere and decreasing in the N. Hemisphere
- The projected changes of extreme waves can be explained with a projected intensification of climatic patterns such as AAO, ENSO, and NAO

Supporting Information:

Supporting Information S1

Global changes of extreme coastal wave energy fluxes triggered by intensified teleconnection patterns

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Abstract In this study we conducted a comprehensive modeling analysis to identify global trends in extreme wave energy flux (WEF) along coastlines in the 21st century under a high emission pathway (Representative Concentration Pathways 8.5). For the end of the century, results show a significant increase up

ARTICLE

DOI: 10.1038/s41467-018-04692-w

nature climate change

COMMUNICATIONS

OPE

Global probabilistic projections of extreme sea levels show intensification of coastal flood hazard

Michalis I. Vousdoukas (1,2), Lorenzo Mentaschi (1,1), Evangelos Voukouvalas³, Martin Verlaan^{4,5},

Environmental Research Letters



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HIDHICUEN

LETTER

Extreme heat waves under 1.5 $^{\circ}\text{C}$ and 2 $^{\circ}\text{C}$ global warming

Alessandro Dosio^{1,4}, Lorenzo Mentaschi¹, Erich M Fischer² and Klaus Wyser³

- ¹ European Commission, Joint Research Centre, Ispra, Italy
- Institute for Atmospheric and Climate Science, ETH Zürich, Switzerland
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Analysis

https://doi.org/10.1038/s41558-022-01540-0

Cost-effective adaptation strategies to rising river flood risk in Europe

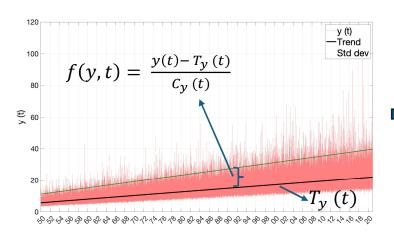
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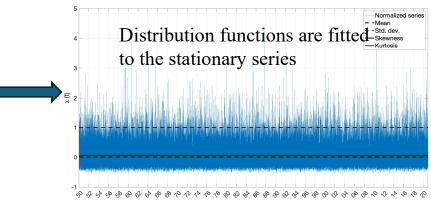
Check for updates

River flood risk in Europe could rise to unprecedented levels due to global warming and continued development in flood-prone areas. Here, we appraise the potential of four key adaptation strategies to reduce flood risk across



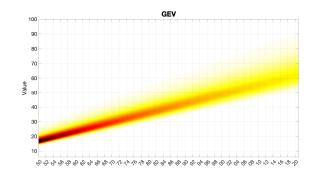
tsEVA methodology



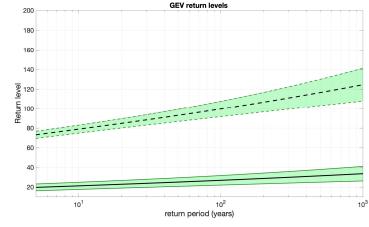


Transformation using time-varying normalization:

- A wide range of transformations
- Diagnosis of the type of nonstationarity
- No priori parametric assumption
- Different stage for non-stationary detection and fitting of EVD



Using a back-transformation, non-stationary distribution parameters are obtained

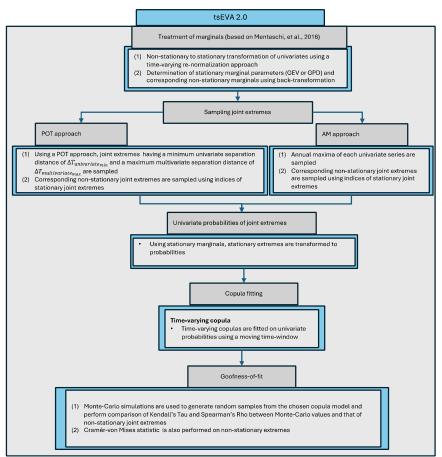


Return levels as a function of time

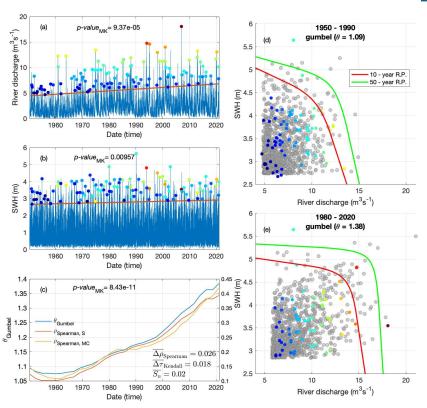


Extension of tsEVA for multivariate extremes: compound events

- (1) Each univariate is separately stationarized
- (2) Joint extremes are sampled using POT or AM approaches
- (3) Joint extremes are converted to probabilities using an appropriate CDF
- (4) A time-varying copula is then fitted on these set of probabilities based on pair-wise Kendal-taus and C-vine for multivariate Gumbel
- (5) Goodness-of-fit of the copula model is assessed using Cramer-von Mises metric and Monte-Carlo strategies

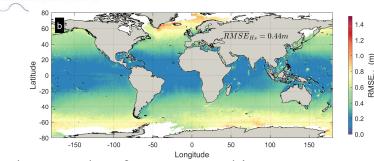








River data from Tilloy et al., 2025 (Hydrological Reanalysis)

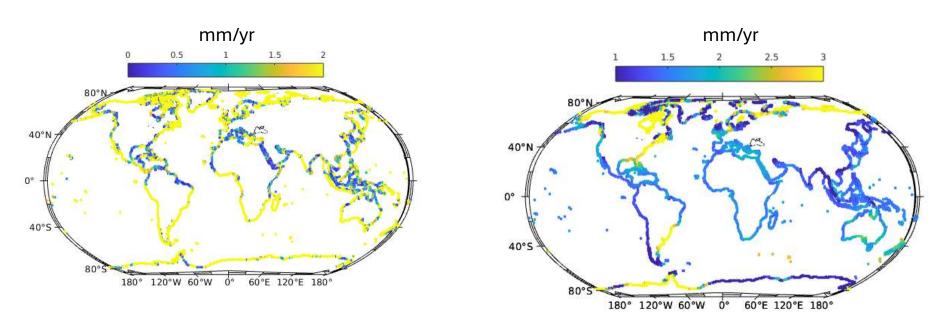


GPD for margin and a time-varying Gumbel copula (40-year time window) to model dependence

3-hourly wave data from Mentaschi et al., 2023



Bivariate application of tsEva



Trend of extremes of SWH

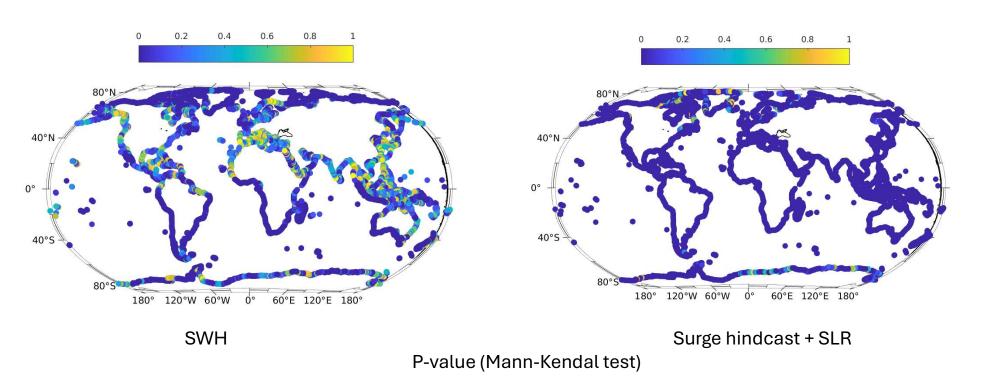
Trend of extremes of surge hindcast + SLR

Trend of the 99th percentile

Data source: Mentaschi, et al., 2023 forced by ERA-5 & bias-corrected by altimetry ~ 130 k coastal grid points



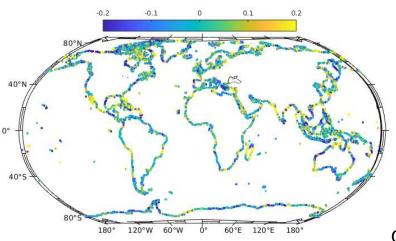
Bivariate application of tsEva

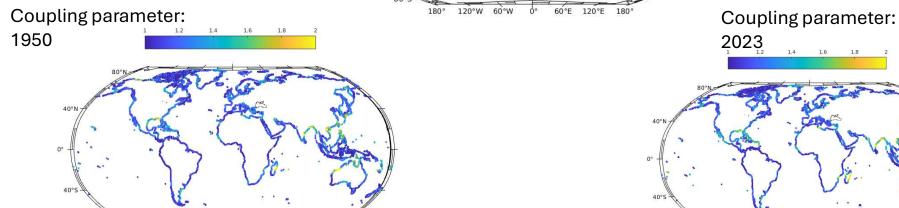


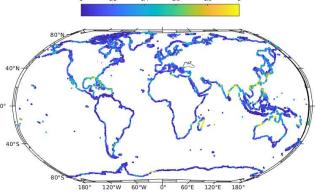
Data source: Mentaschi, et al., 2023 forced by ERA-5 & bias-corrected by altimetry \sim 130 k coastal grid points



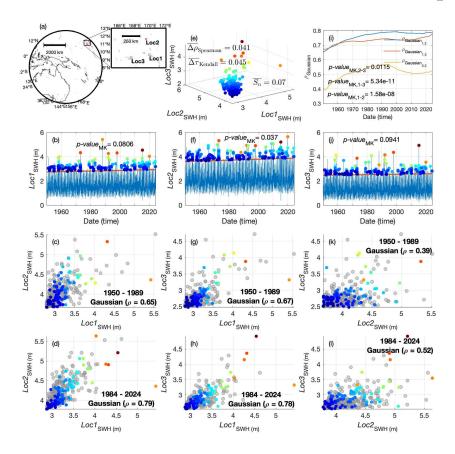
Bivariate application of tsEva

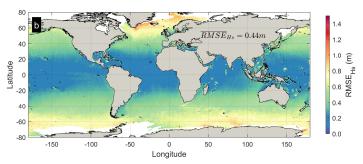












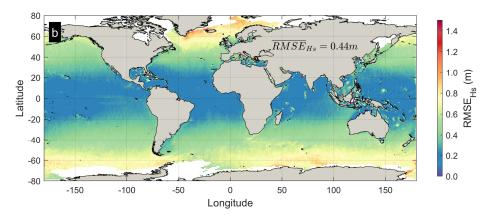
3-hourly SWH from Mentaschi et al., 2023 (1950 – 2024)

• Generalized Pareto Distribution (GPD) for univariate margins and a time-varying Gaussian copula (40-year time window) to model dependence



HERA hydrological reanalysis: Tilloy et al., 2025



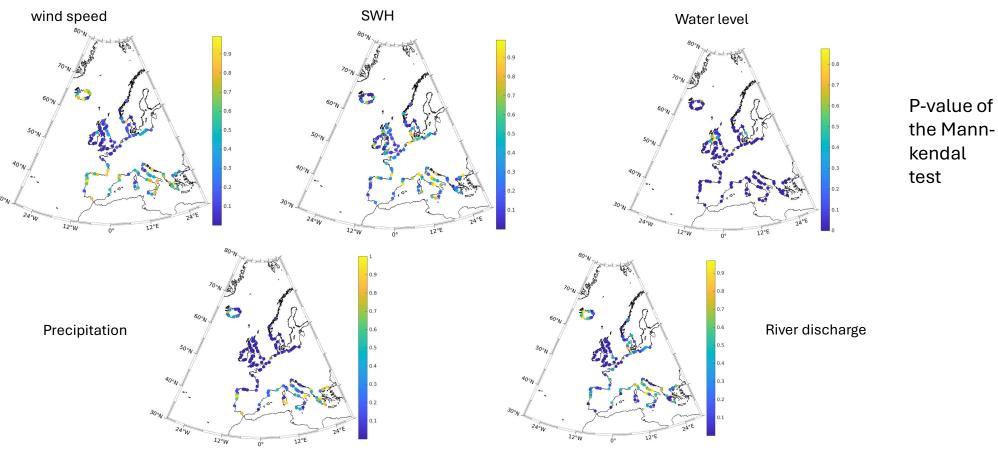


3-hourly SWH from Mentaschi et al., 2023 (1950 – 2024)

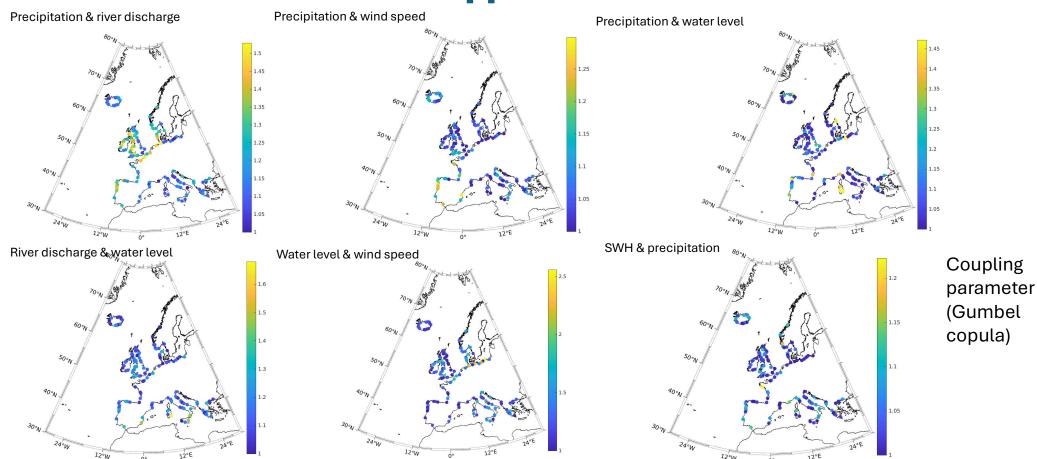
- All river mouthes (>10 m3/s) across Europe were identified
- Surge (+SLR)and SWH from nearest point were extracted
- Wave runup based on Stockdon, et al., 2006
- Instanteneous hourly wind speed and precipitation from ERA-5

160 grid points yielded full fiver-variate dataset

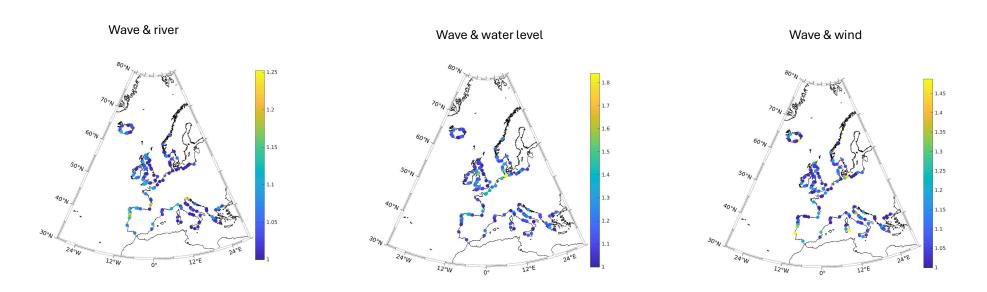






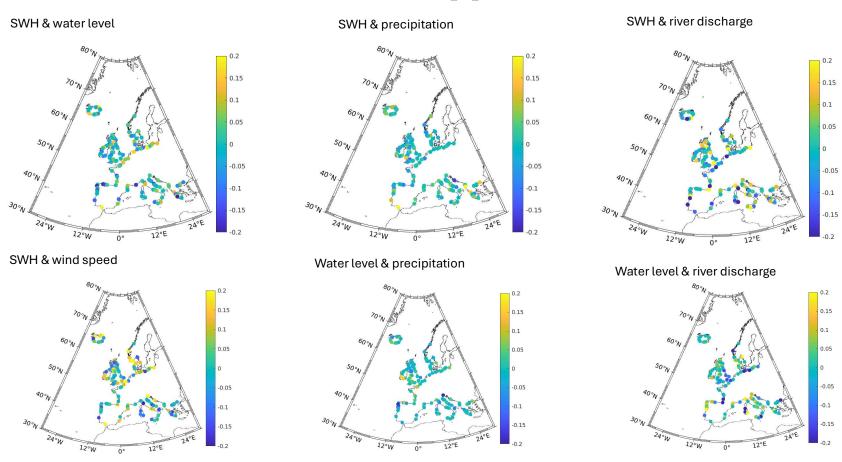






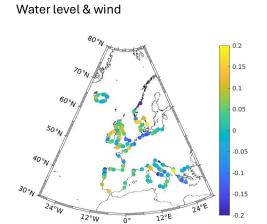
Coupling parameter (Gumbel copula)

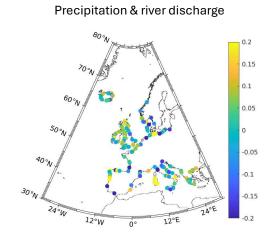


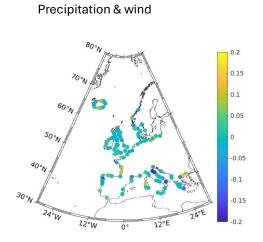


Change of coupling parameter (Gumbel copula)

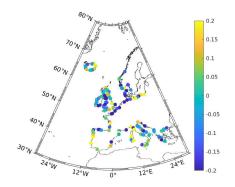


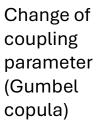






River discharge & wind







Summary

- •tsEVA 2.0 extends univariate nonstationary EVA to **multivariate extremes**, combining tsEVA with copula theory to model evolving dependencies
- •Supports block-maxima (GEV) and POT (GPD) approaches, with time-varying copulas capturing both marginal and dependency non-stationarity; includes built-in goodness-of-fit metrics and joint return period estimation
- •Application on different dataset indicated importance of considering non-stationarity both at the marginal as well as a dependency parameter highlighting the importance of accounting for **dynamic compound hazards**

Directions for future work:

- •Integrate **climate model projections** to assess future non-stationary joint extremes and support **risk-informed adaptation strategies**
- •Use ML approaches in the implementation of tsEVA

