

Projected 21st Century changes in extreme windwave events

Alberto Meucci¹, Ian R. Young¹, Mark Hemer², Ebru Kirezci¹, Roshanka Ranasinghe^{3,4,5} alberto.meucci@unimelb.edu.au

¹Department of Infrastructure Engineering, University of Melbourne

²CSIRO Ocean and Atmosphere, Hobart, Tasmania

³Department of Water Science and Engineering, IHE-Delft, Delft, The Netherlands

⁴Coastal and Offshore Engineering, Deltares, The Netherlands

⁵Water Engineering and Management, Faculty of Engineering Technology, University of Twente











Background

Design Sea State:

Typically defined as the maximum significant wave height which can be expected over an N year period.







Collaroy (NSW) 2016 storm

(Hinkel et al., 2014)

In 2010, 290 million
people worldwide lived
below the 100-year
flood level and US\$9600
billion of assets were
exposed to inundation

1 in 100 years significant wave height Extreme Value Analysis (EVA)

The missing piece

Extreme wind-waves



The missing piece

Extreme wind-waves

Past uncertainties:

Wave model

Atmospheric model

Observations

Statistical



The missing piece

Extreme wind-waves

Past uncertainties:

Wave model

Atmospheric model

Observations

Statistical

Future uncertainties:

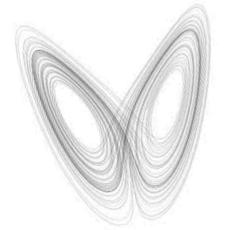
Emissions scenarios

GCMs



Ensemble approach to EVA

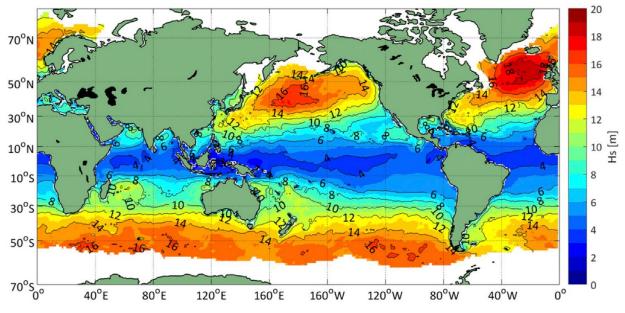
Equivalent of 750 years dataset





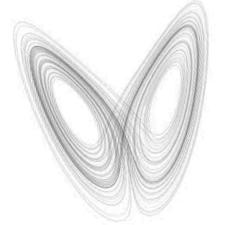
(Lorenz, 1965; Molteni et al., 1996)

(Breivik et al., 2013, 2014; Meucci et al., 2018)



Ensemble approach to EVA

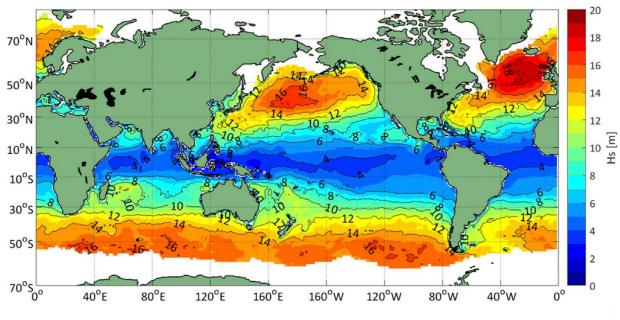
Equivalent of 750 years dataset

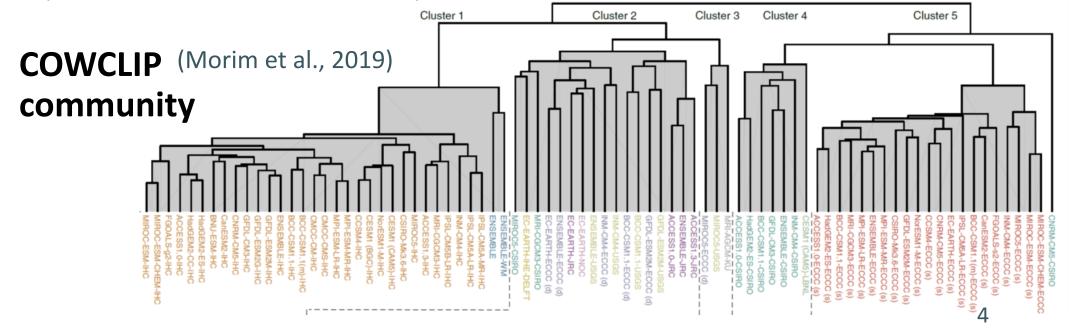




(Lorenz, 1965; Molteni et al., 1996)

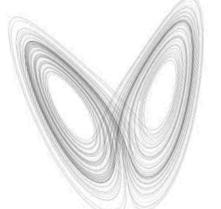
(Breivik et al., 2013, 2014; Meucci et al., 2018)

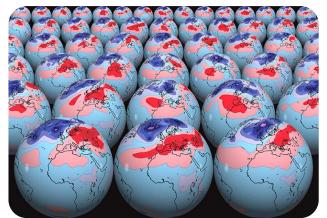




Ensemble approach to EVA

Equivalent of 750 years dataset





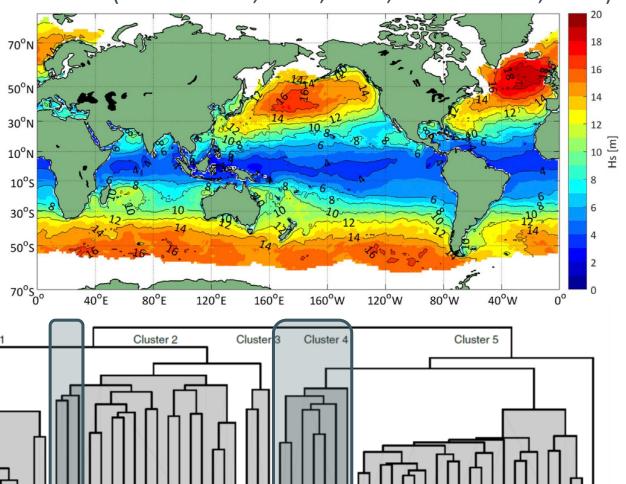
(Lorenz, 1965; Molteni et al., 1996)

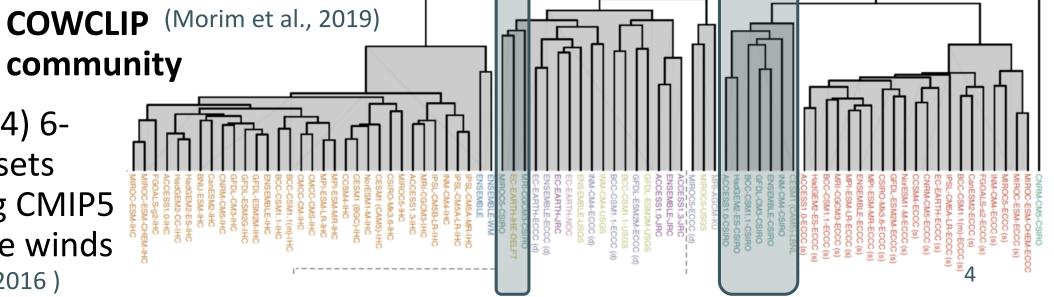
community

WWIII (v3.14) 6hourly datasets forced using CMIP5 GCM surface winds

(Hemer et al., 2016)







Selection of extremes

1979-2005 2081-2100 RCP4.5 2081-2100 RCP8.5

ACCESS1.0

BCC-CSM1.1

GFDL-CM3

HadGEM2-ES

INMCM4

MIROC5

MRI-CGCM3

Selection of extremes

1979-2005

2081-2100 RCP4.5

2081-2100 RCP8.5

(Lopatoukhin et al., 2000)

ACCESS1.0

BCC-CSM1.1

GFDL-CM3

HadGEM2-ES

INMCM4

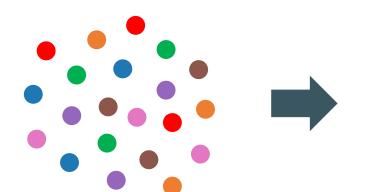
MIROC5

MRI-CGCM3

 $Z_{m} = \frac{H_{s,m} - \mu_{m}^{hist}}{\sigma_{m}^{hist}}$

(Aarnes et al., 2017)

peaks over 90th percentile threshold for each model -- 48h storm independence



1000 highest peaks

Selection of extremes

1979-2005

2081-2100 RCP4.5

2081-2100 RCP8.5

(Lopatoukhin et al., 2000)

ACCESS1.0

BCC-CSM1.1

GFDL-CM3

HadGEM2-ES

INMCM4

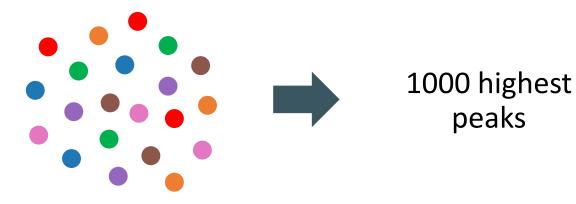
MIROC5

MRI-CGCM3

 $Z_{m} = \frac{H_{s,m} - \mu_{m}^{hist}}{\sigma_{m}^{hist}}$

(Aarnes et al., 2017)

peaks over 90th percentile threshold for each model -- 48h storm independence



Representative time interval

(Breivik et al., 2013, 2014)

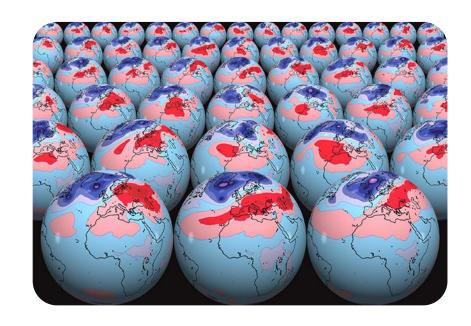
Historical dataset 1979-2005:

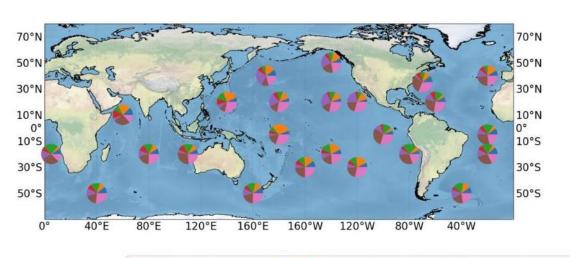
 T_{eq} = 27 years · 365 · 4 hindcasts a day · 6h · 7 GCMs = 189 years

Future projection dataset 2081-2100:

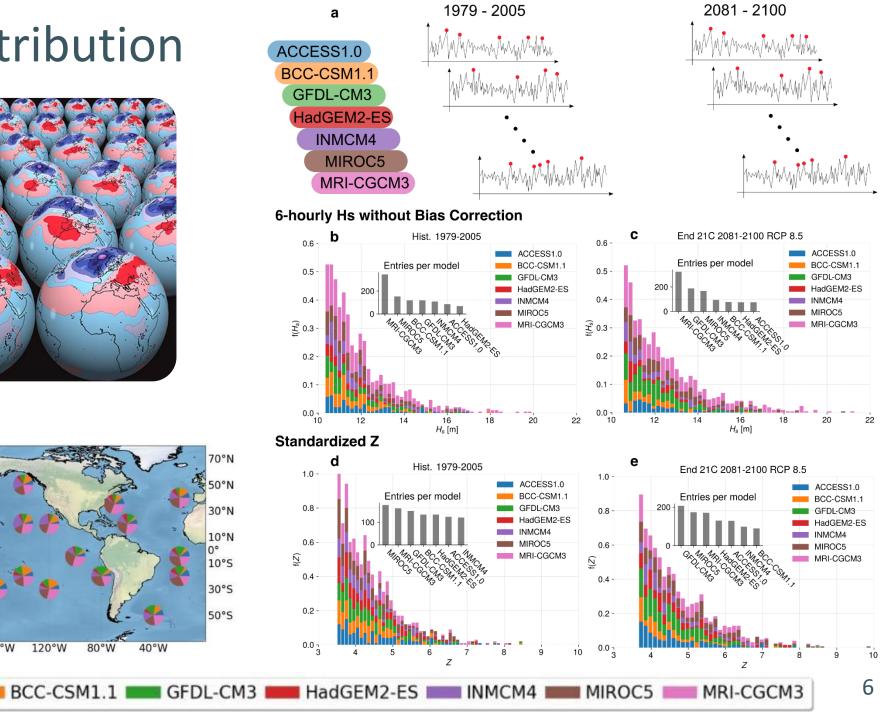
 T_{eq} = 20 years · 365 · 4 hindcasts a day · 6h · 7 GCMs = 140 years

Models contribution

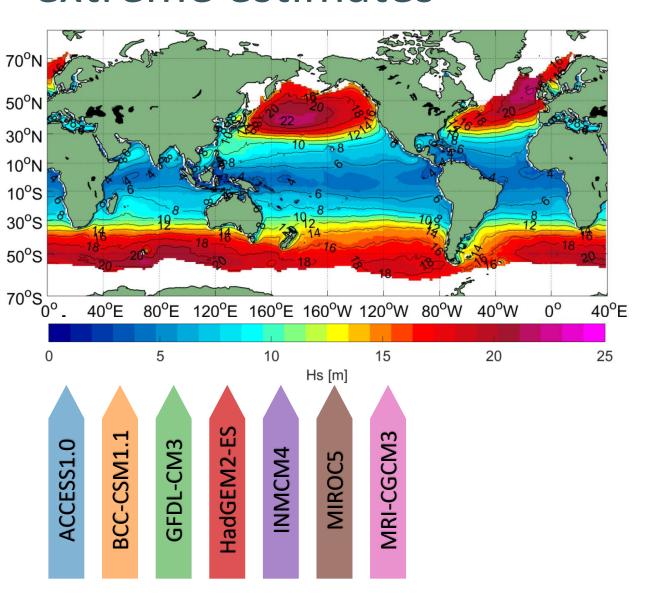




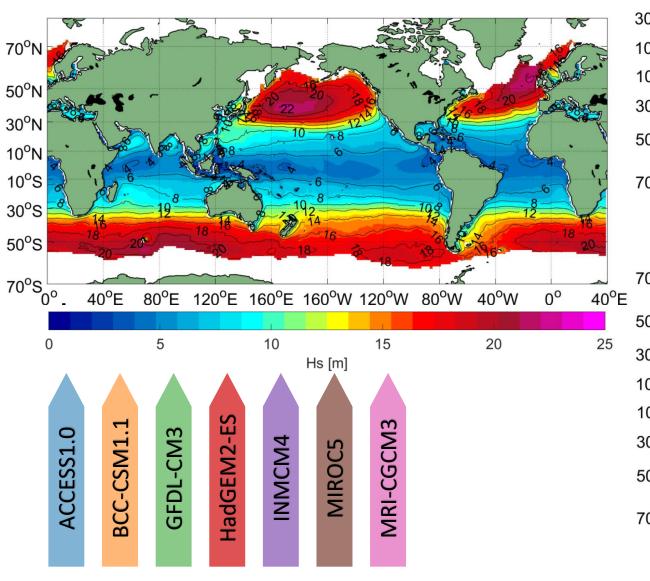
ACCESS1.0

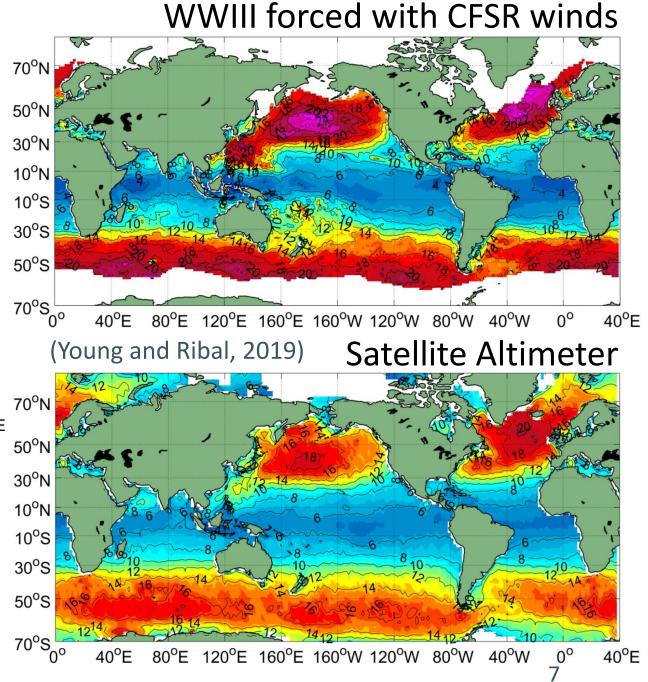


Historical dataset extreme estimates



Historical dataset extreme estimates

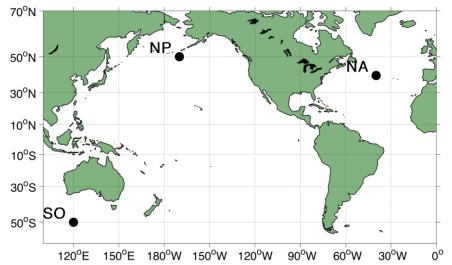


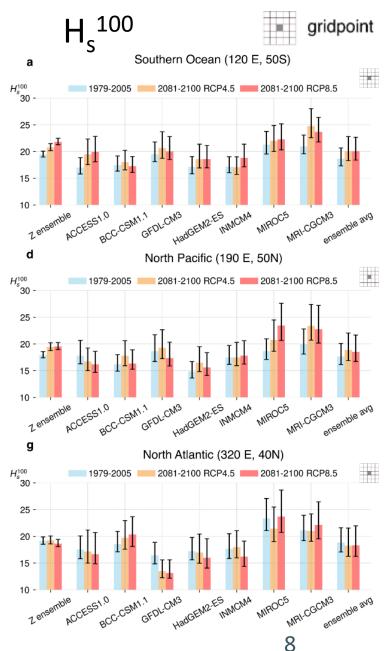


Confidence levels

Bootstrap estimates

on the 1000 peaks obtained from the ensemble pooling technique

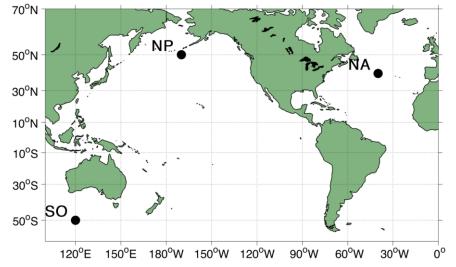




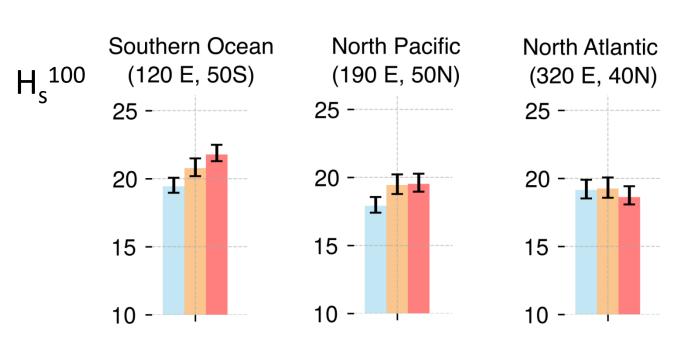
Confidence levels

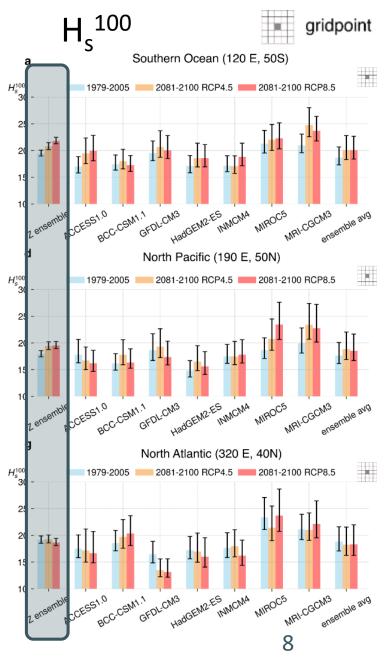
Bootstrap estimates

on the 1000 peaks obtained from the ensemble pooling technique



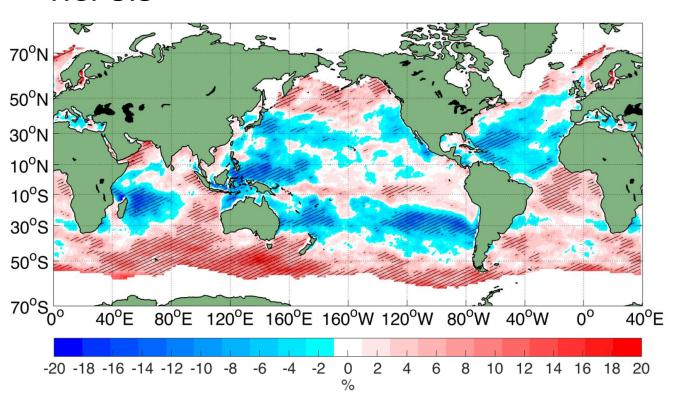






Projected changes in extreme wind-waves (H_s¹⁰⁰) 2081–2100 - 1979–2005

RCP8.5

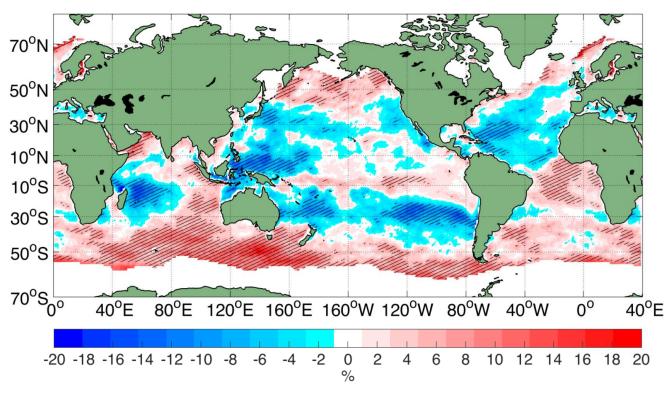


Projected changes in extreme wind-waves (H_s¹⁰⁰)

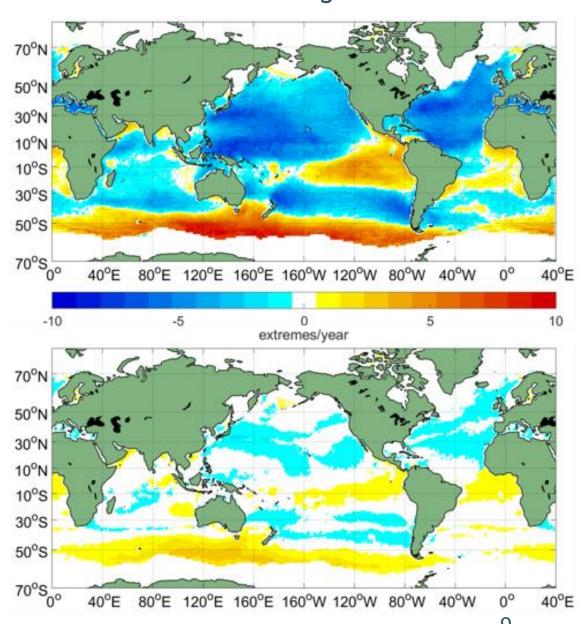
2081-2100 - 1979-2005

90th perc.





99.7th perc.



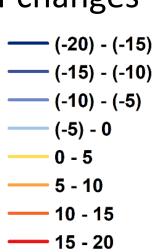
Changes along global coastlines

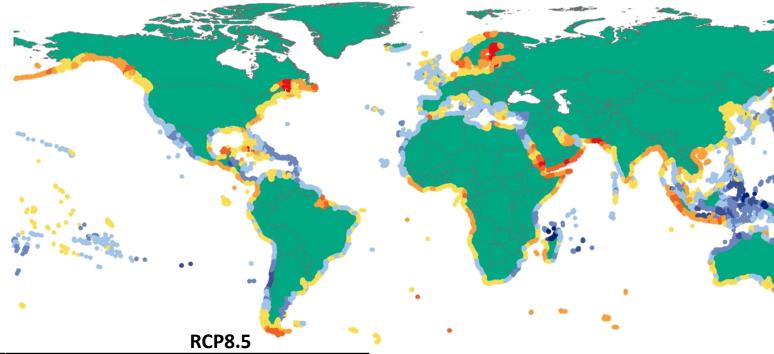
Percentage of changes

in H_s^{100}

10% to 15%

15% to 20%





6.31

2.23

R	C	P	4		5
	•		_	•	-

33,053

7,737

	Coastline	Coastline	Coastline	Coastline
(% change)	length	length	length	length
	(km)	(%)	(km)	(%)
-20% to -15%	9,643	0.89	7,399	0.69
-15% to -10%	13,130	1.22	25,281	2.34
-10% to -5%	69,208	6.42	120,625	11.18
-5% to 0%	277,810	25.76	285,227	26.45
0% to 5%	499,537	46.32	365,741	33.91
5% to 10%	168,420	15.62	182,163	16.89

3.06

0.72

68,087

24,015

DIVA dataset locations

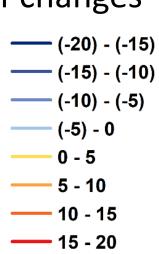
 ΔH_s^{100} at the closest offshore grid point

Changes along global coastlines

Percentage of changes

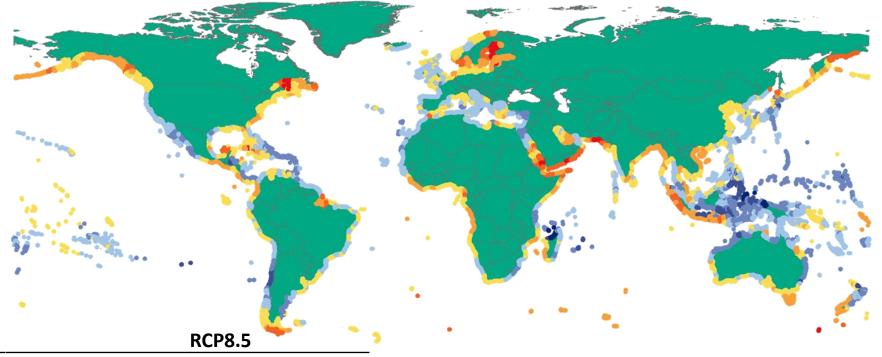
in $H_s^{\ 100}$

15% to 20%



RCP4.5

7,737



2.23

	Coastline	Coastline	Coastline	Coastline
(% change)	length	length	length	length
	(km)	(%)	(km)	(%)
-20% to -15%	9,643	0.89	7,399	0.69
-15% to -10%	13,130	1.22	25,281	2.34
-10% to -5%	69,208	6.42	120,625	11.18
-5% to 0%	277,810	25.76	285,227	26.45
0% to 5%	499,537	46.32	365,741	33.91
5% to 10%	168,420	15.62	182,163	16.89
10% to 15%	33.053	3.06	68.087	6.31

0.72

24,015

DIVA dataset locations

 ΔH_s^{100} at the closest offshore grid point

Stationarity

- Stationarity
- Inhomogeneous datasets

- Stationarity
- Inhomogeneous datasets
- Tropical Cyclones still not correctly reproduced by GCMs

- Stationarity
- Inhomogeneous datasets
- Tropical Cyclones still not correctly reproduced by GCMs

Potential

(Hemer et al. 2013; Wang et al. 2014; Aarnes et al. 2017; Morim et al. 2019)

Results are consistent with previous studies

- Stationarity
- Inhomogeneous datasets
- Tropical Cyclones still not correctly reproduced by GCMs

Potential

(Hemer et al. 2013; Wang et al. 2014; Aarnes et al. 2017; Morim et al. 2019)

- Results are consistent with previous studies
- Inter models low correlation guarantees independence

- Stationarity
- Inhomogeneous datasets
- Tropical Cyclones still not correctly reproduced by GCMs

Potential

(Hemer et al. 2013; Wang et al. 2014; Aarnes et al. 2017; Morim et al. 2019)

- Results are consistent with previous studies
- Inter models low correlation guarantees independence
- Possibility to synthesize an equivalent time series of duration longer than the simulation period

- Stationarity
- Inhomogeneous datasets
- Tropical Cyclones still not correctly reproduced by GCMs

Potential

(Hemer et al. 2013; Wang et al. 2014; Aarnes et al. 2017; Morim et al. 2019)

- Results are consistent with previous studies
- Inter models low correlation guarantees independence
- Possibility to synthesize an equivalent time series of duration longer than the simulation period
- Increased dataset reduces confidence intervals

• Higher resolutions are needed



Higher resolutions are needed

 Ensemble approach to TC areas with increasing model resolution



Higher resolutions are needed

 Ensemble approach to TC areas with increasing model resolution

 Still many uncertainties are characterizing observations of extremes



Higher resolutions are needed

 Ensemble approach to TC areas with increasing model resolution

 Still many uncertainties are characterizing observations of extremes

 Improved GCMs and additional models may allow use of Direct Return level Estimates



References

Aarnes, O. J., Reistad, M., Breivik, Ø., Bitner-Gregersen, E., Ingolf Eide, L., Gramstad, O., ... & Vanem, E. (2017). Projected changes in significant wave height toward the end of the 21st century: Northeast Atlantic. Journal of Geophysical Research: Oceans, 122(4), 3394-3403.

Breivik, \emptyset ., Aarnes, O. J., Bidlot, J.-R., Carrasco, A., and Saetra, \emptyset . (2013). Wave extremes in the northeast Atlantic from ensemble forecasts. *Journal of Climate*, 26(19):7525–7540.

Breivik, Ø., Aarnes, O. J., Abdalla, S., Bidlot, J. R., & Janssen, P. A. (2014). Wind and wave extremes over the world oceans from very large ensembles. *Geophysical Research Letters*, 41(14), 5122-5131.

Breivik, Ø. and Aarnes, O.J., (2017). Efficient bootstrap estimates for tail statistics. *Natural Hazards and Earth System Sciences*, 17(3), p.357.

Coles, S., Bawa, J., Trenner, L., & Dorazio, P. (2001). *An introduction to statistical modeling of extreme values* (Vol. 208). London: Springer.

Déqué, M. (2007). Frequency of precipitation and temperature extremes over France in an anthropogenic scenario: Model results and statistical correction according to observed values. Global and Planetary Change, 57(1-2), 16-26.

Hemer, M. A., Fan, Y., Mori, N., Semedo, A., & Wang, X. L. (2013). Projected changes in wave climate from a multi-model ensemble. Nature climate change, 3(5), 471.

Hemer, M. A., & Trenham, C. E. (2016). Evaluation of a CMIP5 derived dynamical global wind wave climate model ensemble. Ocean Modelling, 103, 190-203.

Lopatoukhin, L., Rozhkov, V., Ryabinin, V., Swail, V., Boukhanovsky, A., and Degtyarev, A. (2000). Estimation of extreme wind wave heights. Technical report, WMO. Lorenz, E. N. (1965). A study of the predictability of a 28-variable atmospheric model. Tellus, 17(3), 321-333.

prediction system: Methodology and validation. Quarterly journal of the royal meteorological society, 122(529), 73-119.

Morim, J., Hemer, M., Wang, X. L., Cartwright, N., Trenham, C., Semedo, A., ... & Erikson, L. (2019). Robustness and uncertainties in global multivariate wind-wave climate projections. *Nature Climate Change*, *9*(9), 711-718.

Taylor, K. E., Stouffer, R. J., & Meehl, G. A. (2012). An overview of CMIP5 and the experiment design. Bulletin of the American Meteorological Society, 93(4), 485-498.

Wang, X. L., Feng, Y., & Swail, V. R. (2014). Changes in global ocean wave heights as projected using multimodel CMIP5 simulations. *Geophysical Research Letters*, 41(3), 1026-1034.

Young, I. R., Zieger, S., & Babanin, A. V. (2011). Global trends in wind speed and wave height. *Science*, 332(6028), 451-455.

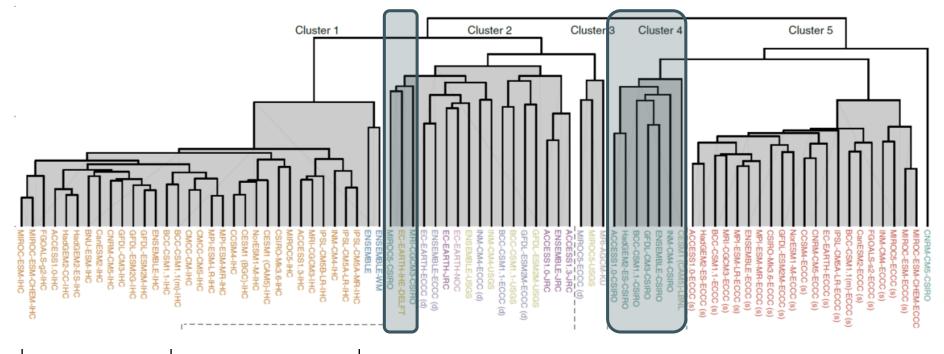
Young, I. R., & Ribal, A. (2019). Multiplatform evaluation of global trends in wind speed and wave height. *Science*, *364*(6440), 548-552.

Supplementary material

Dataset

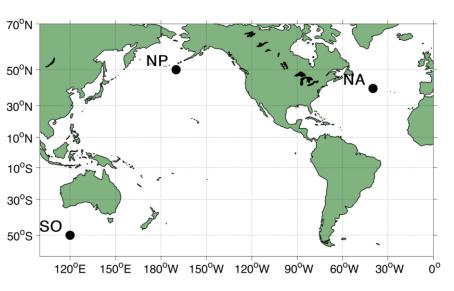
WWIII (v3.14) 6-hourly datasets forced using CMIP5 GCM surface winds

(Hemer et al., 2016)

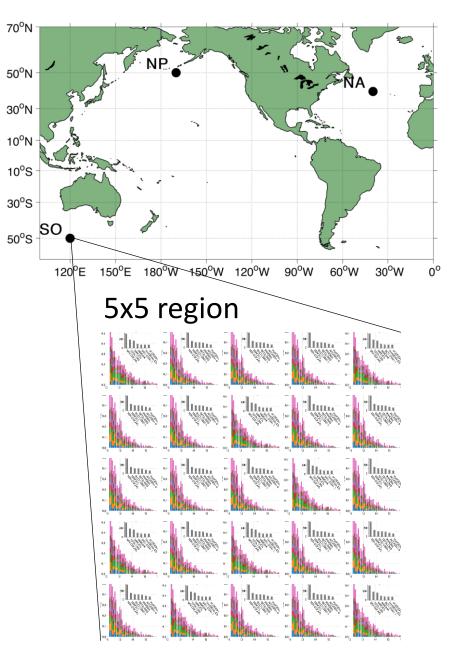


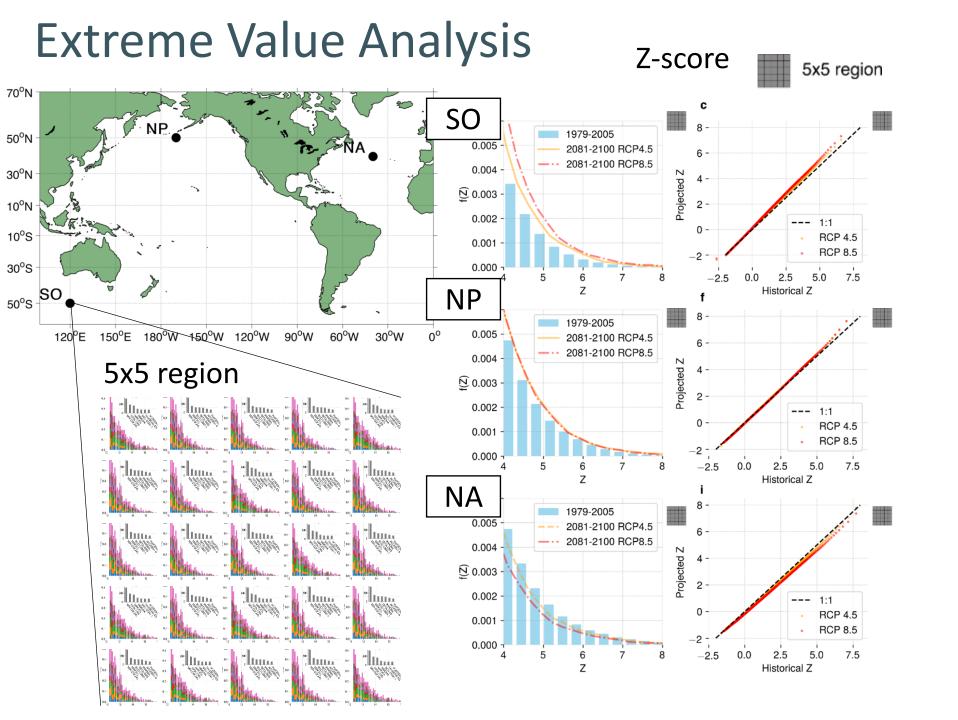
	CMIP	ATM	WAVE		Period		
	phase	lon x lat [°]	lon x lat [°]	Δt			
ACCESS1.0	5	1.88 x 1.25	1.0 x 1.0	6h	1979-2005	2081-2100	
BCC-CSM1.1	5	2.8 x 2.8	1.0 x 1.0	6h	1979-2005	2081-2100	RCP4.5
GFDL-CM3	5	2.5 x 2.0	1.0 x 1.0	6h	1979-2005	2081-2100	RCP4.5
HadGEM2-ES	5	1.88 x 1.25	1.0 x 1.0	6h	1979-2005	2081-2100	
INMCM4	5	2.0 x 1.25	1.0 x 1.0	6h	1979-2005	2081-2100	RCP8.5
MIROC5	5	1.4 x 1.4	1.0 x 1.0	6h	1979-2005	2081-2100	
MRI-CGCM3	5	1.1 x 1.1	1.0 x 1.0	6h	1979-2005	2081-2100	15
			-		-		

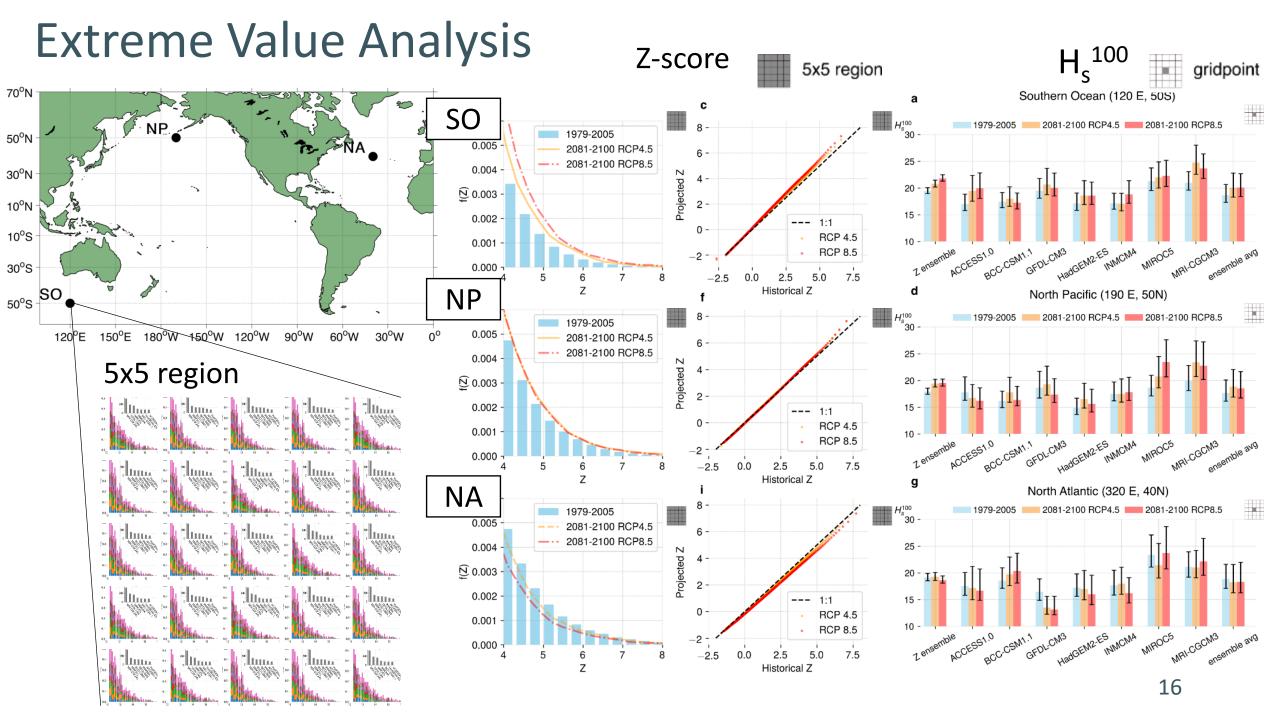
Extreme Value Analysis



Extreme Value Analysis







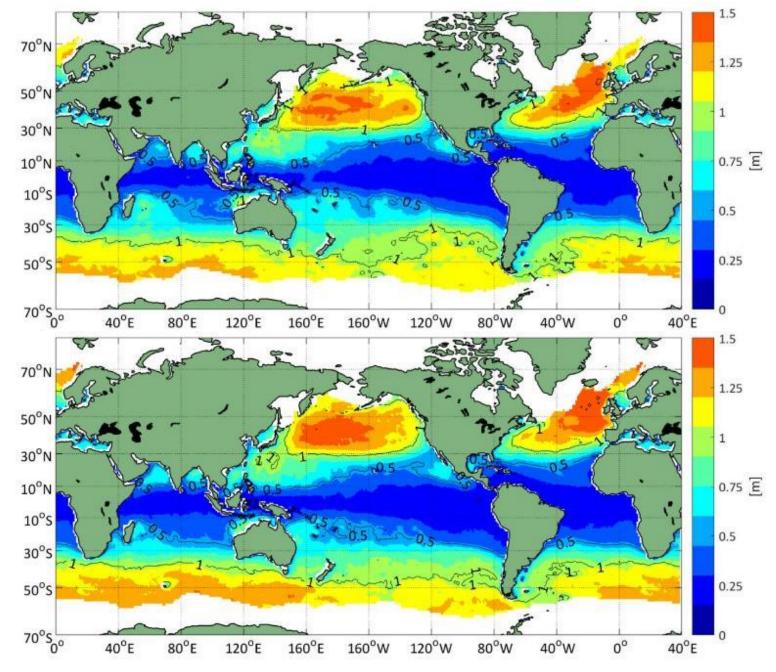
Confidence levels

1979-2005

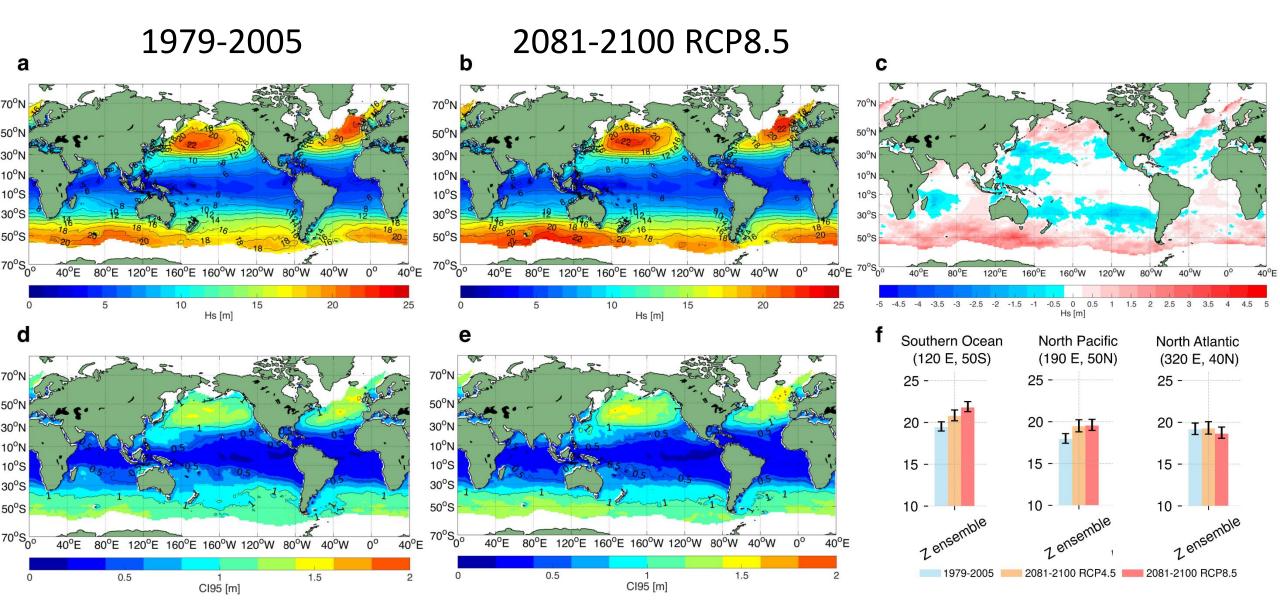
Bootstrap estimates

on the 1000 peaks obtained from the ensemble pooling technique

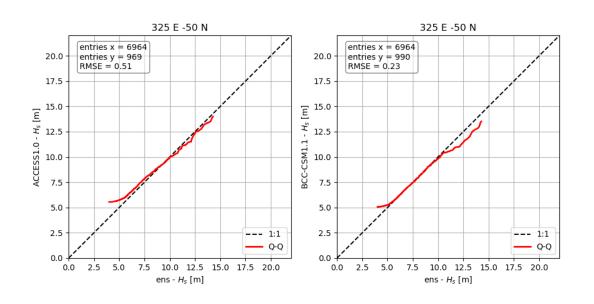
2081-2100 RCP8.5



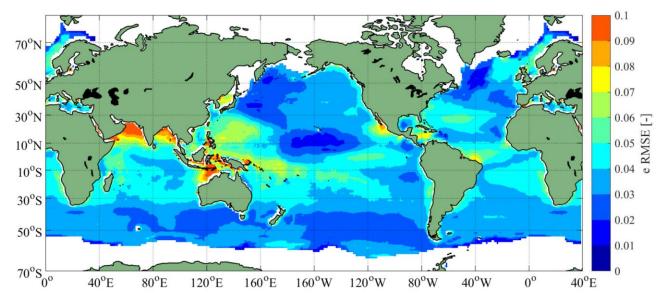
Confidence levels

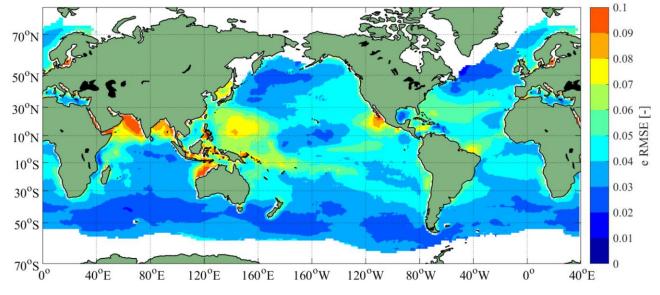


Similarity test between distribution of extremes

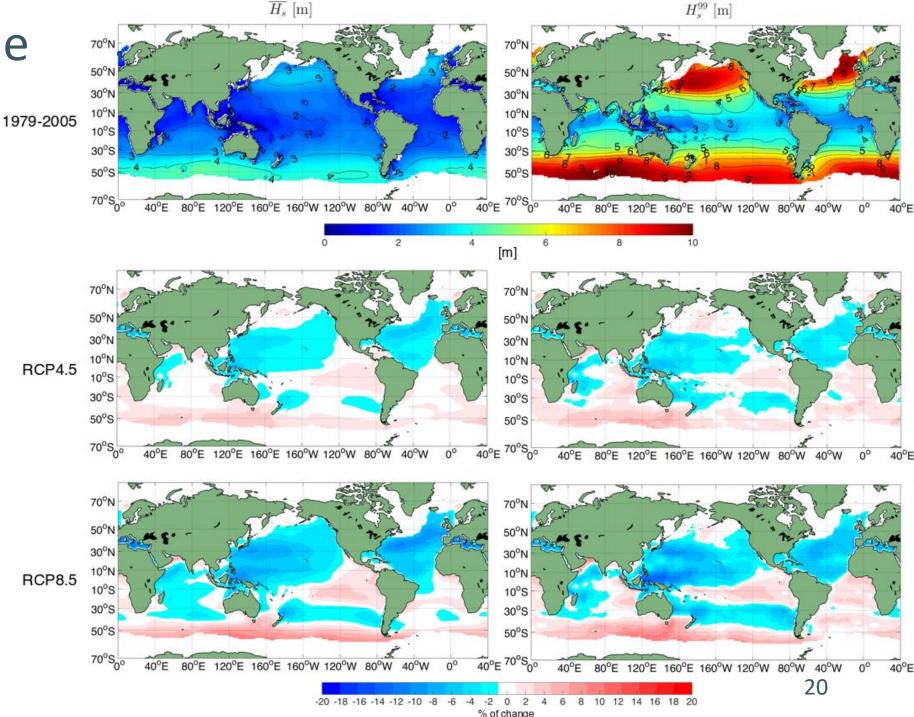


$$e_{RMSE} = 1 - \frac{\left| \frac{\sum RMSE_m}{n} - H_S^{100} \right|}{H_S^{100}}$$



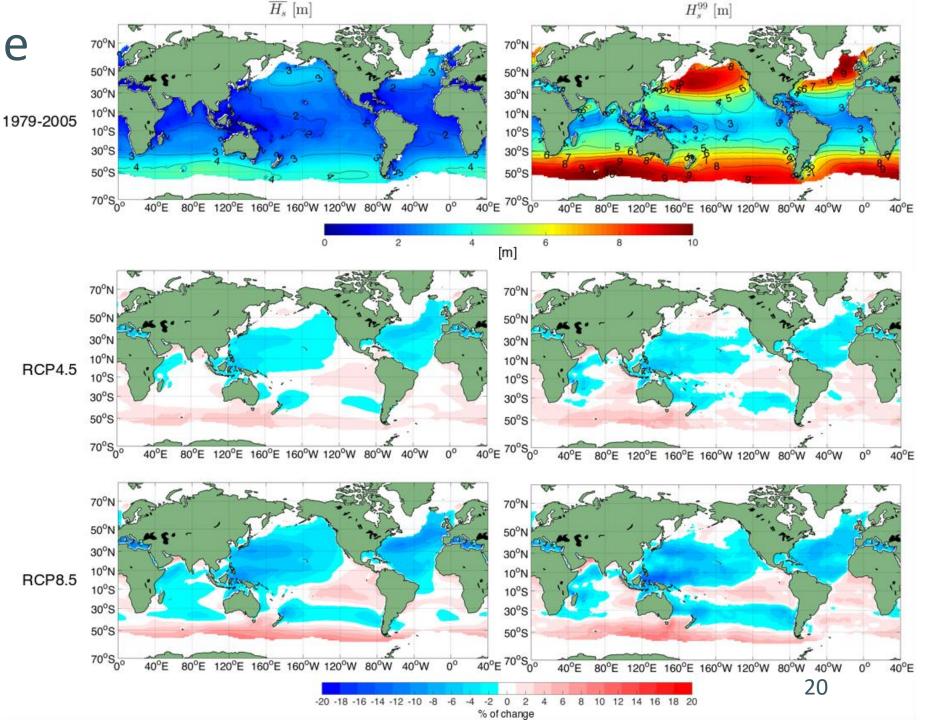


Model ensemble performance

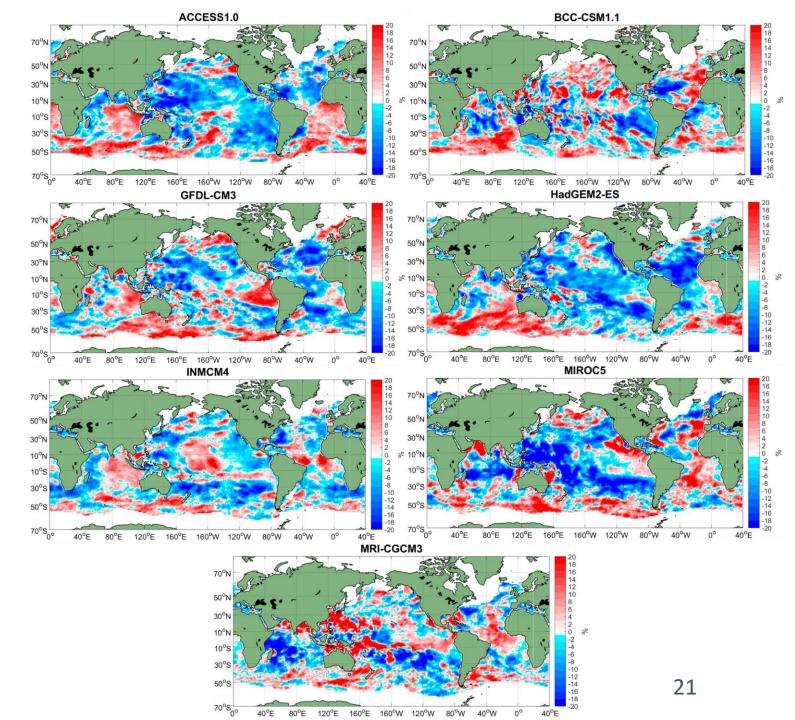


Model ensemble performance

Comparable to total multi-model ensemble in Morim et al., (2019)



Instability of single model projected extreme changes



Independent and Identically Distributed (i.i.d) data

