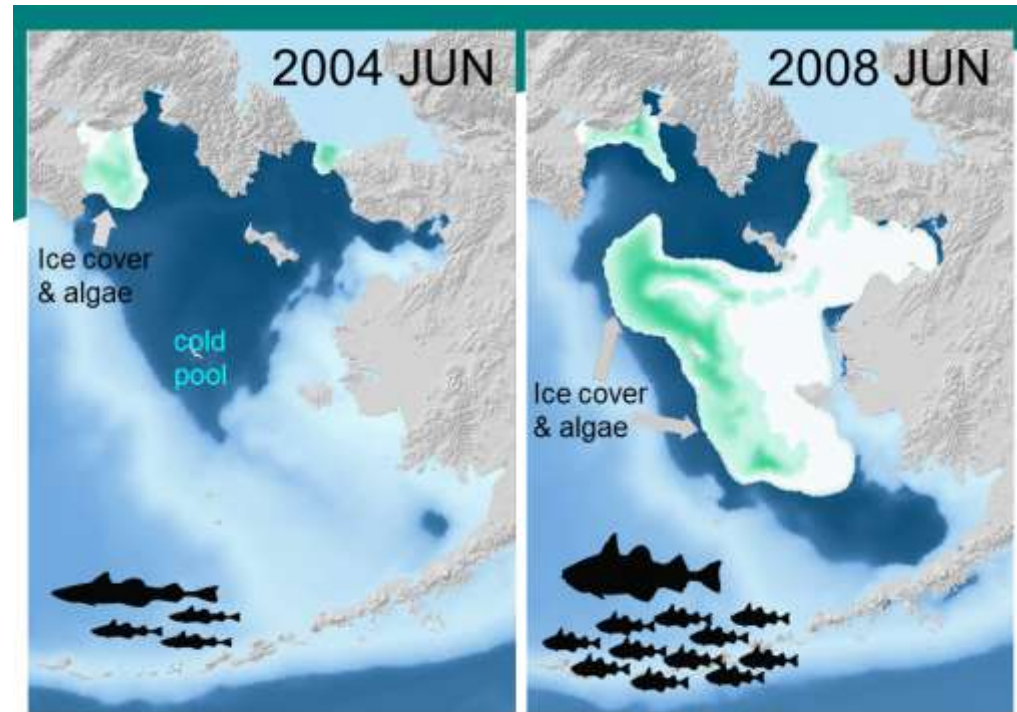
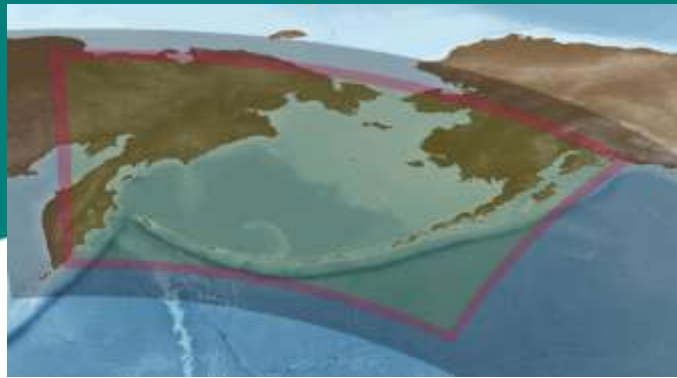


Bering ROMS/NPZ (Regional Oceanographic model with nutrients and plankton dynamics)

- 3-D, 10km² resolution ocean model (10 or 30 vertical layers)
- Developed with NSF/North Pacific Research Board (BSIERP)
- AFSC/PMEL partnership funded by IEA, ACLIM, FATE, MAPP
- Significant advances in ice modeling, ice plankton





Bering ROMS/NPZ (Regional Oceanographic model with nutrients and plankton dynamics)

Products

- 48-year hindcast (1970-2018) forced by measured conditions (IEA)
- 9-month forecast (annual) forced by CFS forecasts (MAPP)
- Forecasts to 2100 forced by IPCC model outputs (FATE, ACLIM)

Uses

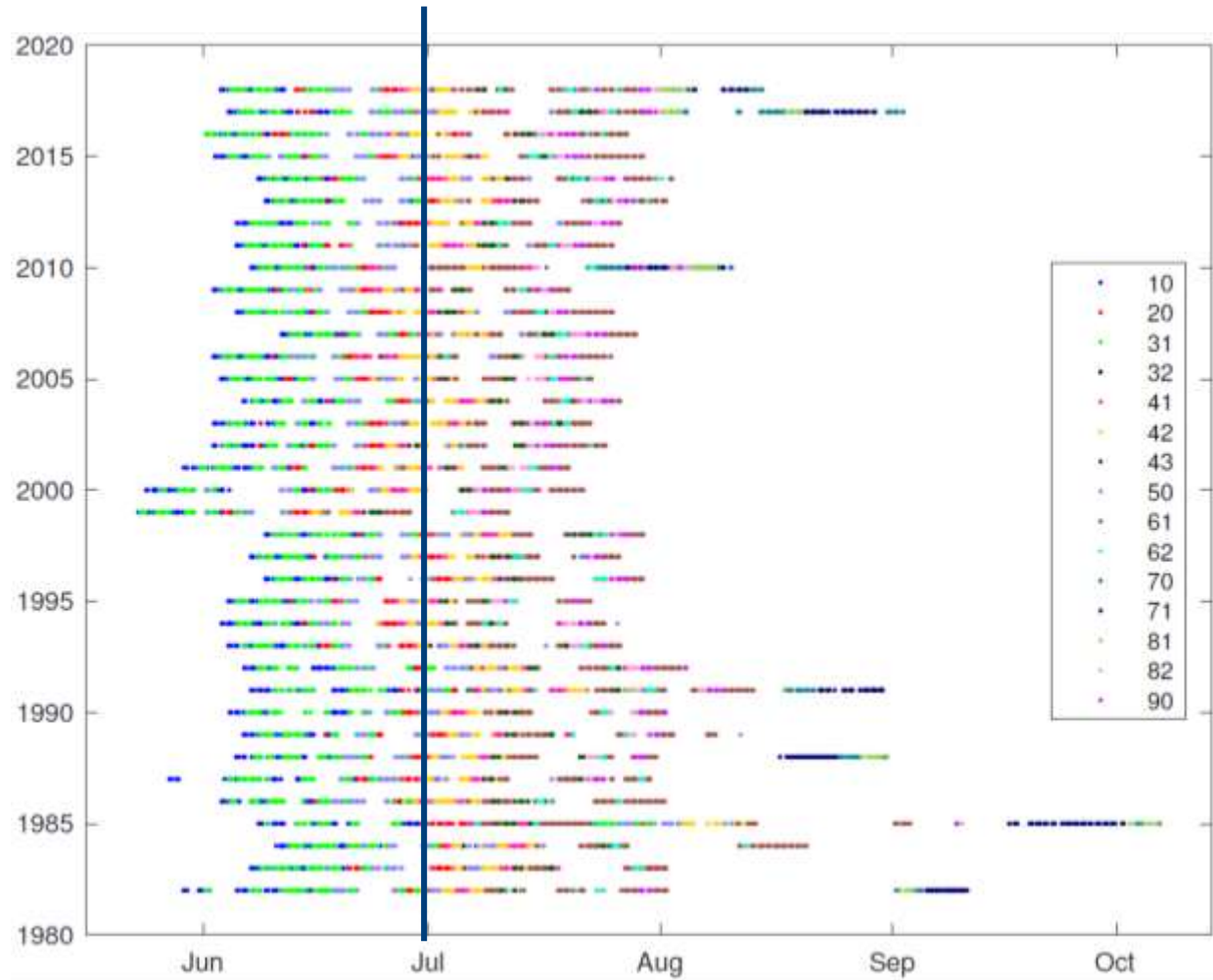
- Forcing for biological models
 - CEATTLE, others
- Rapid Climate Assessment
- EFH predictive maps
- VAST distribution maps

Focus today

- *Has current NBS bottom temperature happened before? (hindcast)*
- *Is it the “new normal”? (IPCC forecasts)*
- *Implications for CEATTLE*



BTS survey dates by stratum





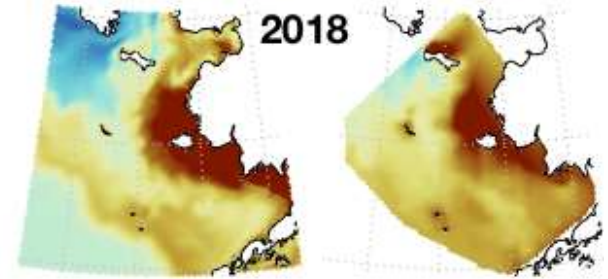
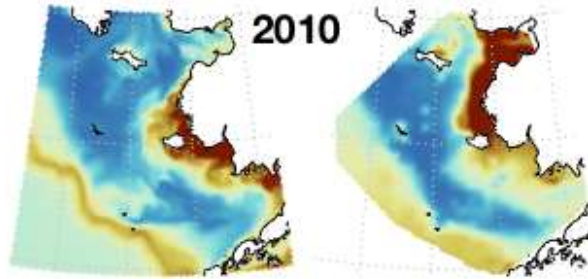
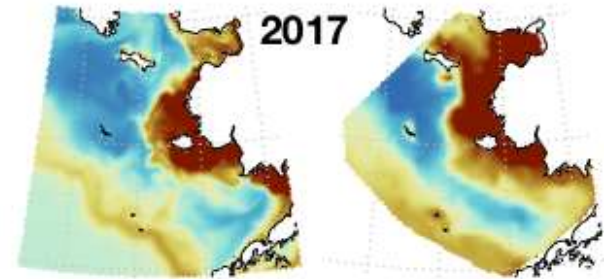
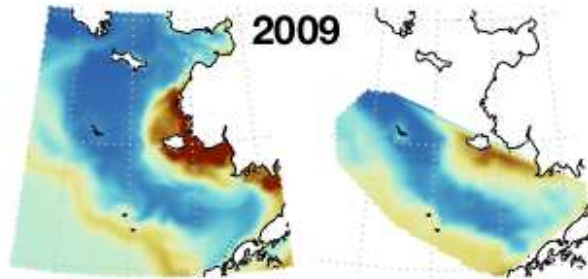
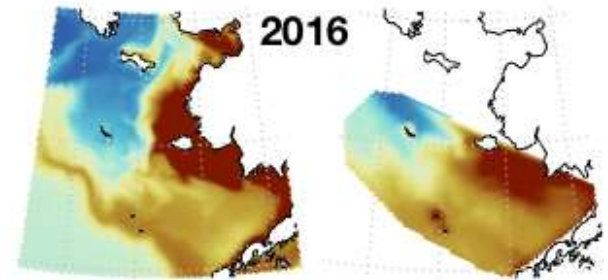
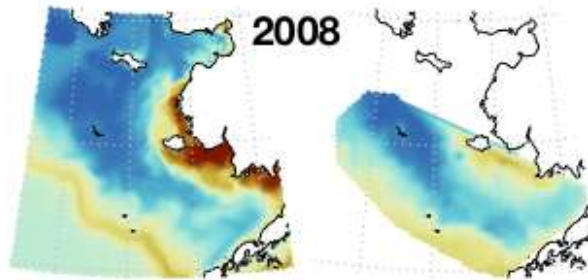
Bottom temperature comparison

Jul 1 model

BTS data

Jul 1 model

BTS data





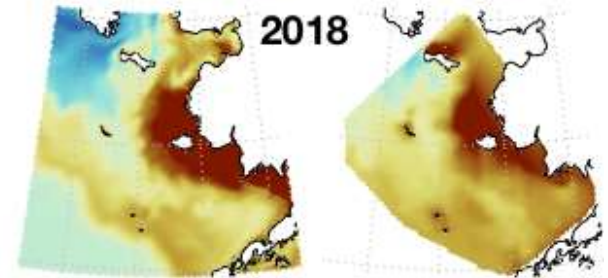
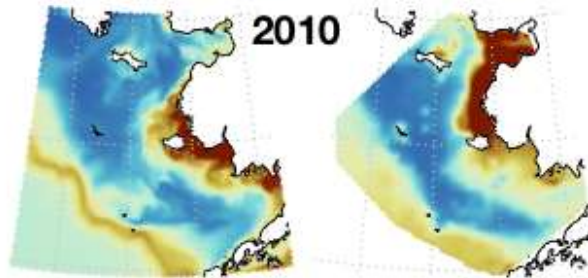
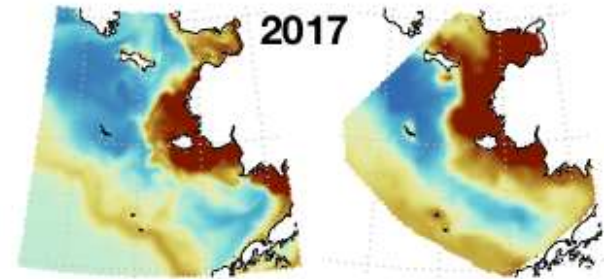
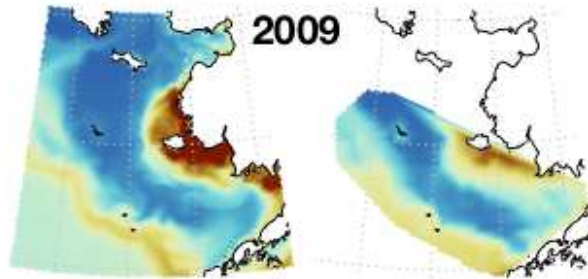
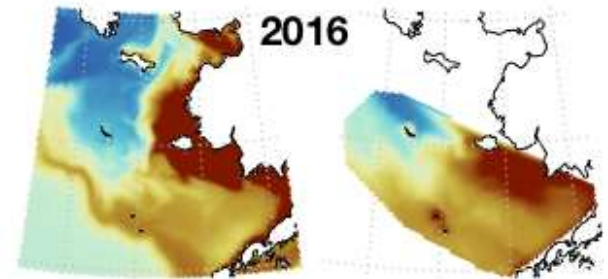
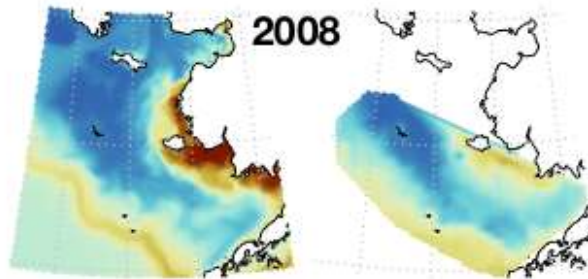
Bottom temperature comparison

Jul 1 model

BTS data

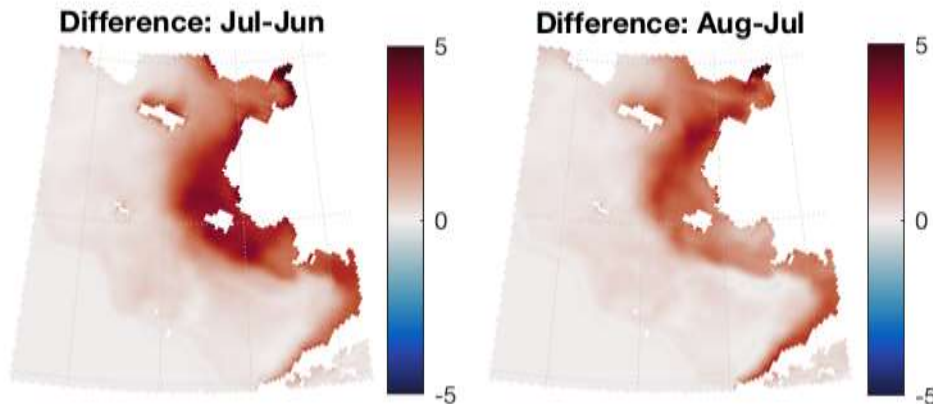
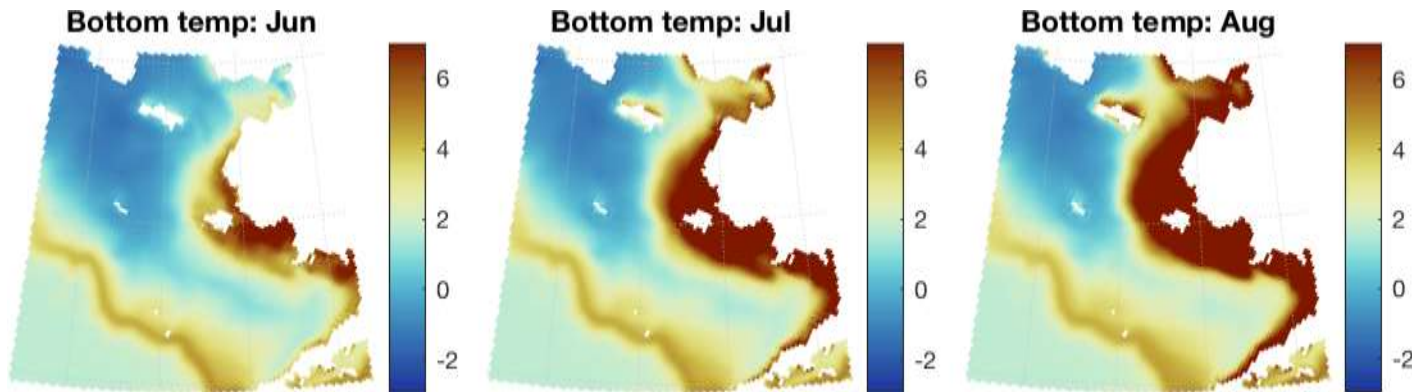
Jul 1 model

BTS data





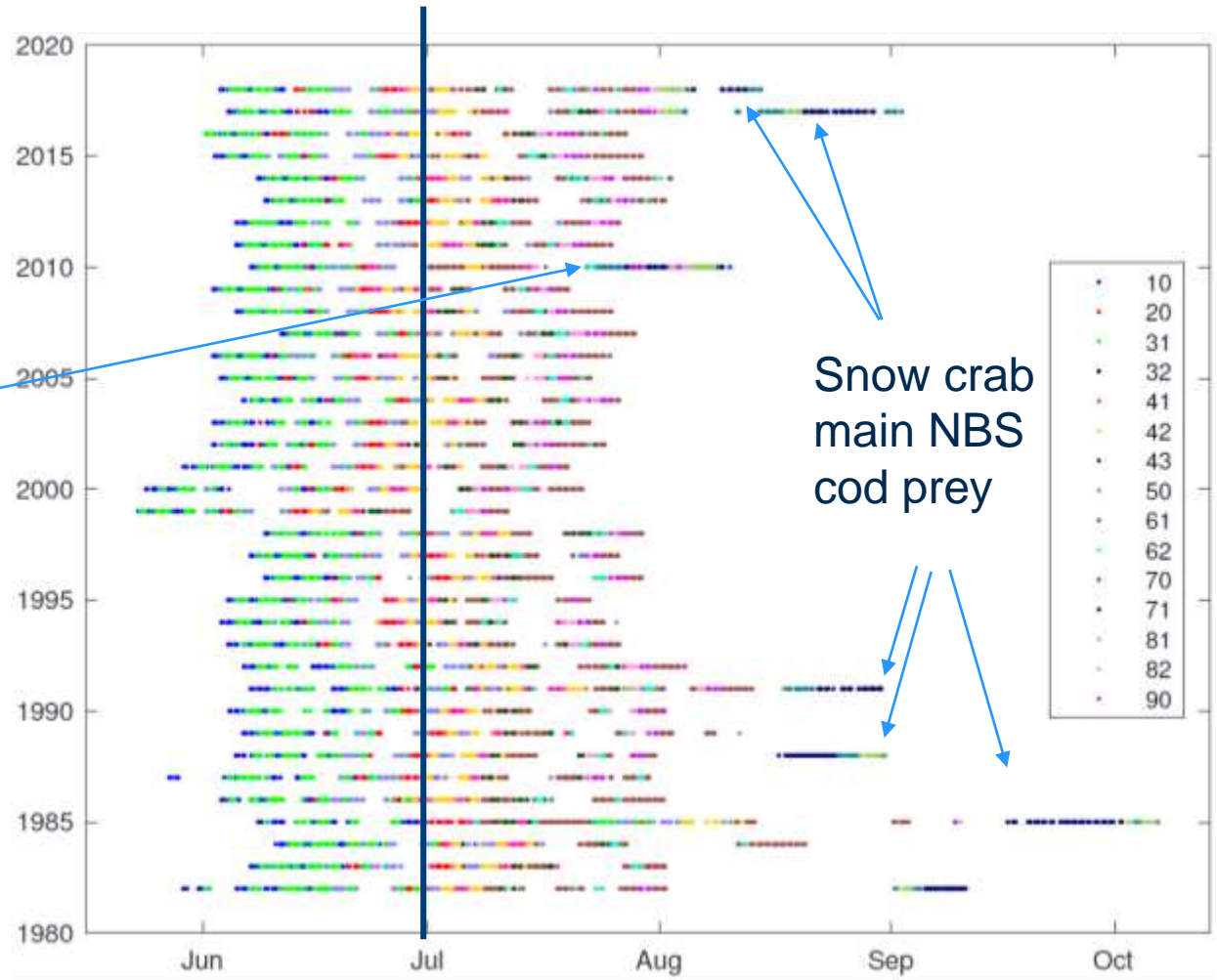
Bottom temperature comparison





BTS survey dates by stratum

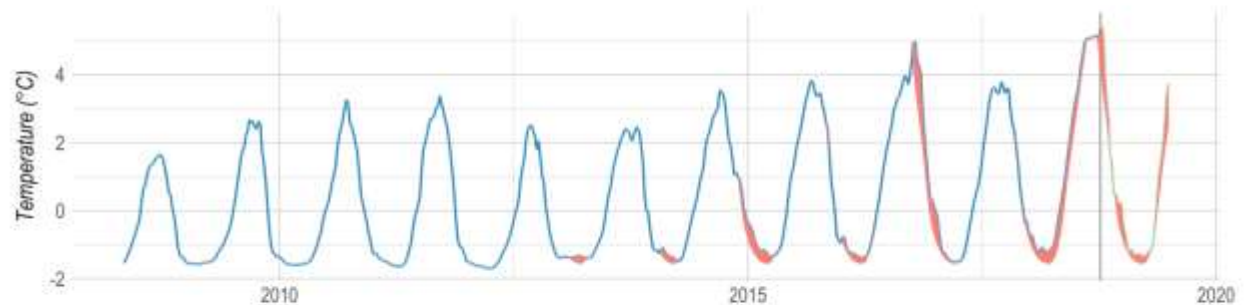
Juvenile flatfish, other benthos main NBS cod prey



MARINE HEATWAVE ANALYSIS

NEBS Bottom Temperature

Marine heatwave analysis based on downscaled ROMSNPZ hindcast and 1970-2000 climatology.



ROMSNPZ: K. Kearney, A. Hermann, K. Aydin, 2018
Heatwave analysis: K. Holsman, 2018, based on Hobday et al. (2016)
Data source: NOAA PMEL, AFSC REEM Program, IEA, MAPP Bering Seasons

KIRSTIN HOLSMAN

KELLY KEARNEY

KERIM AYDIN

AL HERMANN

STEVE BARBEAUX

2018

ARTICLE

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OPEN

Longer and more frequent marine heatwaves over the past century

Eric C.J. Oliver^{1,2,3}, Markus G. Donat^{4,5}, Michael T. Burrows⁶, Pippa J. Moore⁷, Dan A. Smale^{8,9}, Lisa V. Alexander^{4,5}, Jessica A. Benthuyzen¹⁰, Ming Feng¹¹, Alex Sen Gupta^{4,5}, Alistair J. Hobday¹², Neil J. Holbrook^{2,13}, Sarah E. Perkins-Kirkpatrick^{4,5}, Hillary A. Scannell^{14,15}, Sandra C. Straub⁹ & Thomas Wernberg⁹

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A hierarchical approach to defining marine heatwaves



Alistair J. Hobday^{4,16}, Lisa V. Alexander^{16,17}, Sarah E. Perkins-Kirkpatrick^{16,17}, Dan A. Smale^{16,18}, Sandra C. Straub¹⁹, Eric C.J. Oliver¹⁶, Jessica A. Benthuyzen²⁰, Michael T. Burrows²¹, Markus G. Donat^{16,17}, Ming Feng²², Neil J. Holbrook²³, Pippa J. Moore²⁴, Hillary A. Scannell²⁵, Alex Sen Gupta^{16,17}, Thomas Wernberg¹⁹

¹CSIRO Oceans and Atmosphere, Hobart, Tasmania 7000, Australia

²ARC Centre of Excellence for Climate System Science, The University of New South Wales, Sydney, Australia

³Climate Change Research Centre, The University of New South Wales, Sydney, Australia

⁴Marine Biological Association of the United Kingdom, The Laboratory, Citadel Hill, Plymouth PL1 2PB, UK

⁵UWA Oceans Institute and School of Plant Biology, The University of Western Australia, Crawley, 6009 Western Australia, Australia

⁶Institute for Marine and Antarctic Studies, University of Tasmania, Hobart, Australia

⁷Australian Institute of Marine Science, Townsville, Queensland, Australia

⁸Department of Zoology, Scottish Association for Marine Science, Scottish Marine Institute, Oban, Argyll PA37 9QA, Scotland, UK

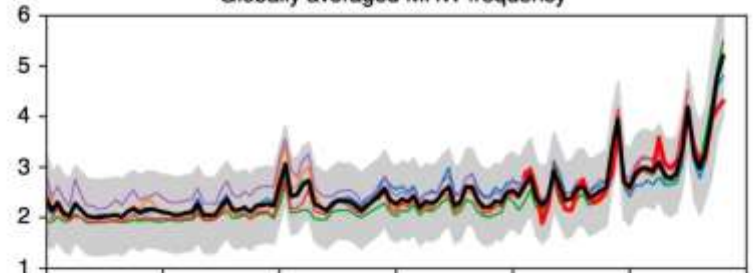
⁹CSIRO Oceans and Atmosphere, Perth, Western Australia, Australia

¹⁰Institute of Biological, Environmental and Rural Sciences, Aberystwyth University, Aberystwyth SY23 3DA, UK

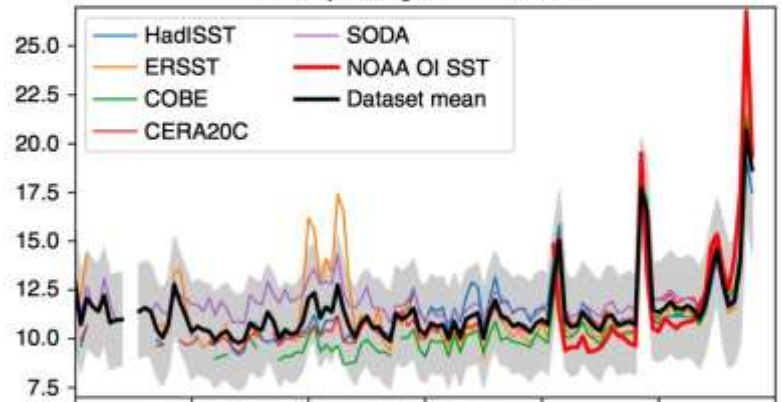
¹¹School of Oceanography, University of Washington, Seattle, WA, USA

¹²NOAA/Pacific Marine Environmental Laboratory, Seattle, WA, USA

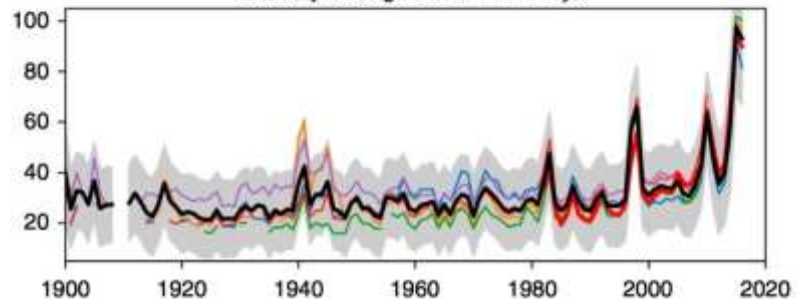
b Globally averaged MHW frequency

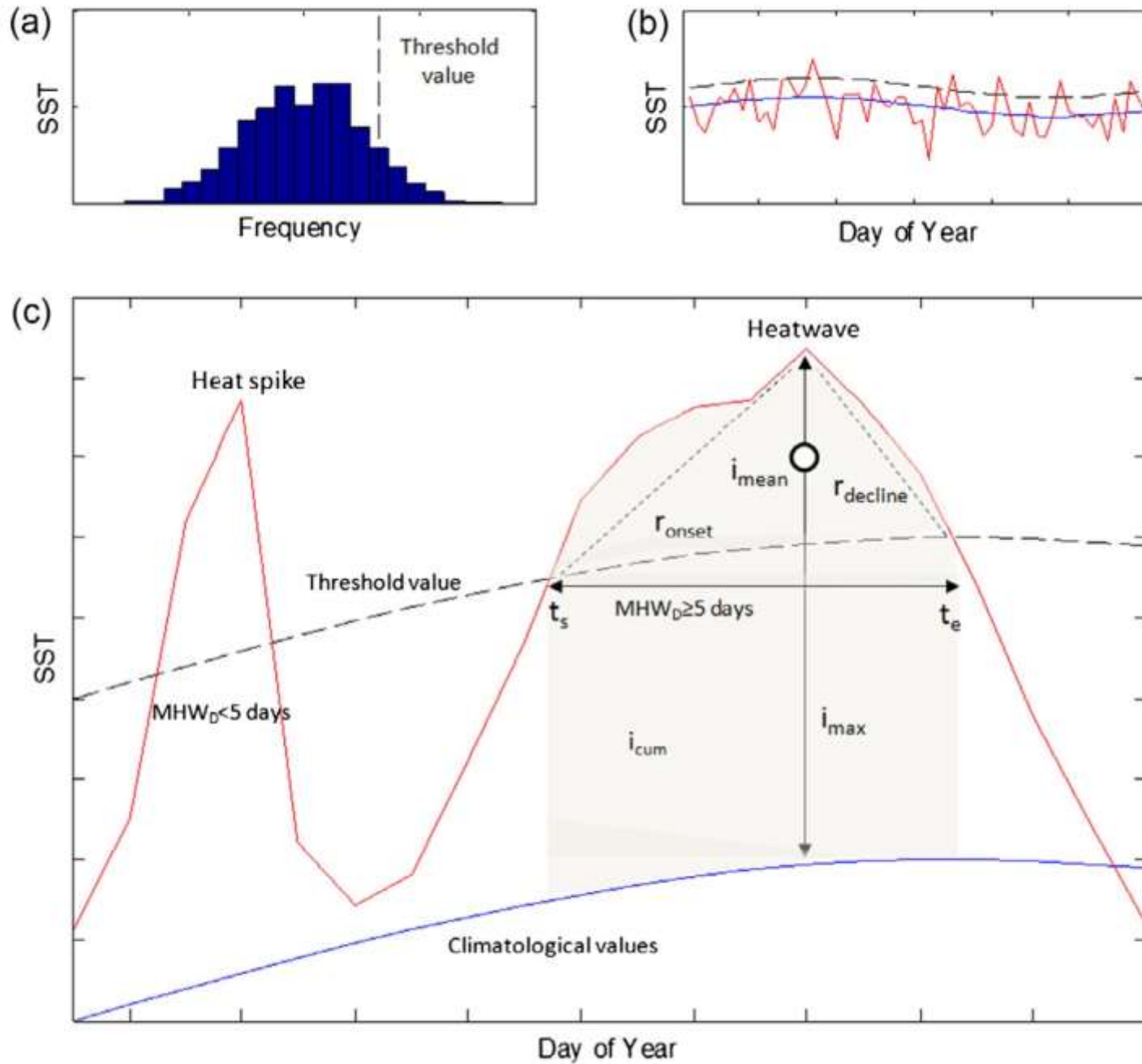


d Globally averaged MHW duration



f Globally averaged total MHW days

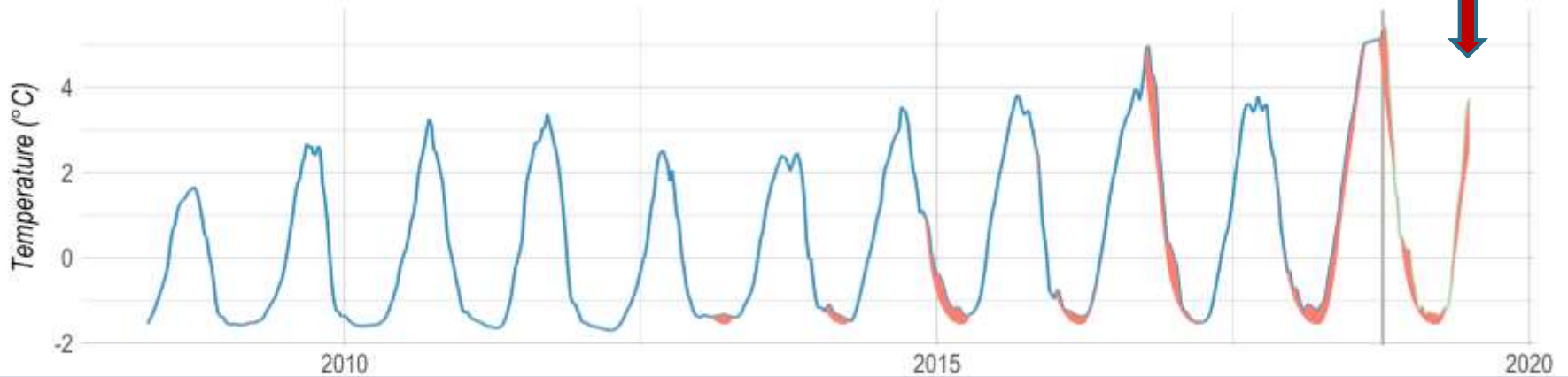




NEBS Bottom Temperature

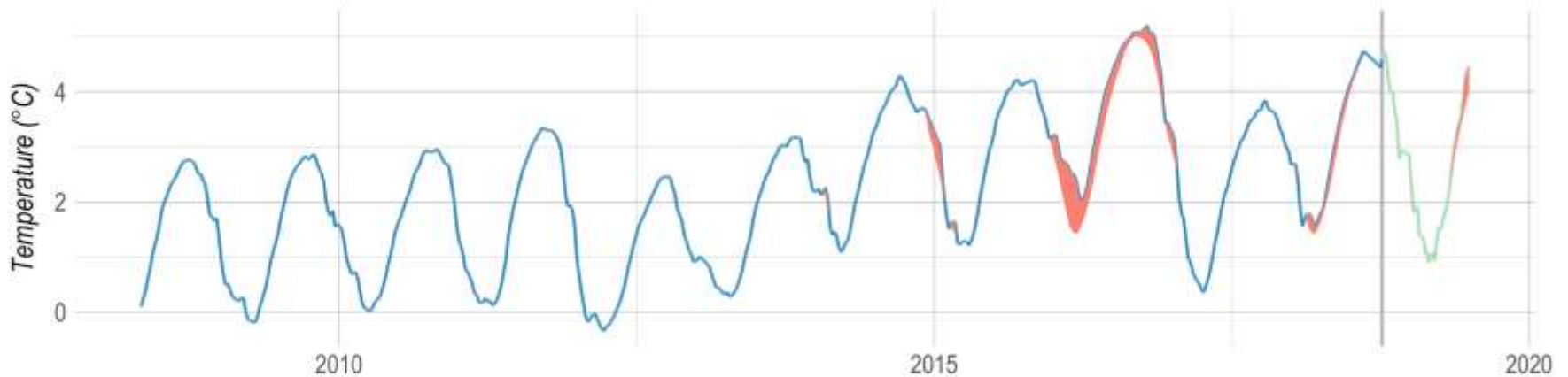
Marine heatwave analysis based on downscaled ROMSNPZ hindcast and 1970-2000 climatology.

Continued HW
Expected in 2019



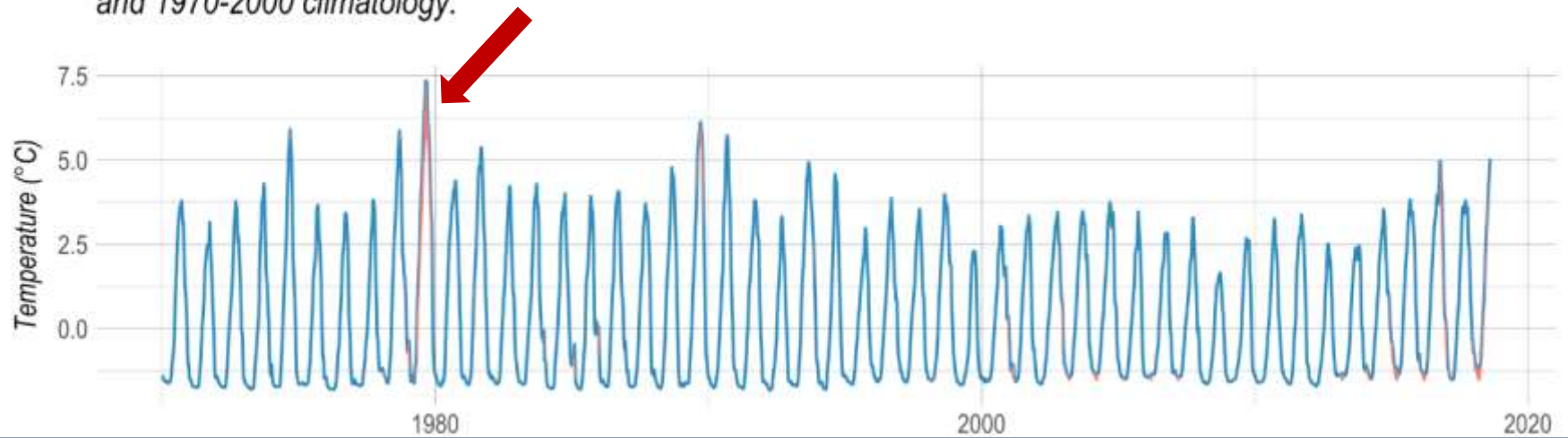
EBS Bottom Temperature

Marine heatwave analysis based on downscaled ROMSNPZ hindcast and 1970-2000 climatology.



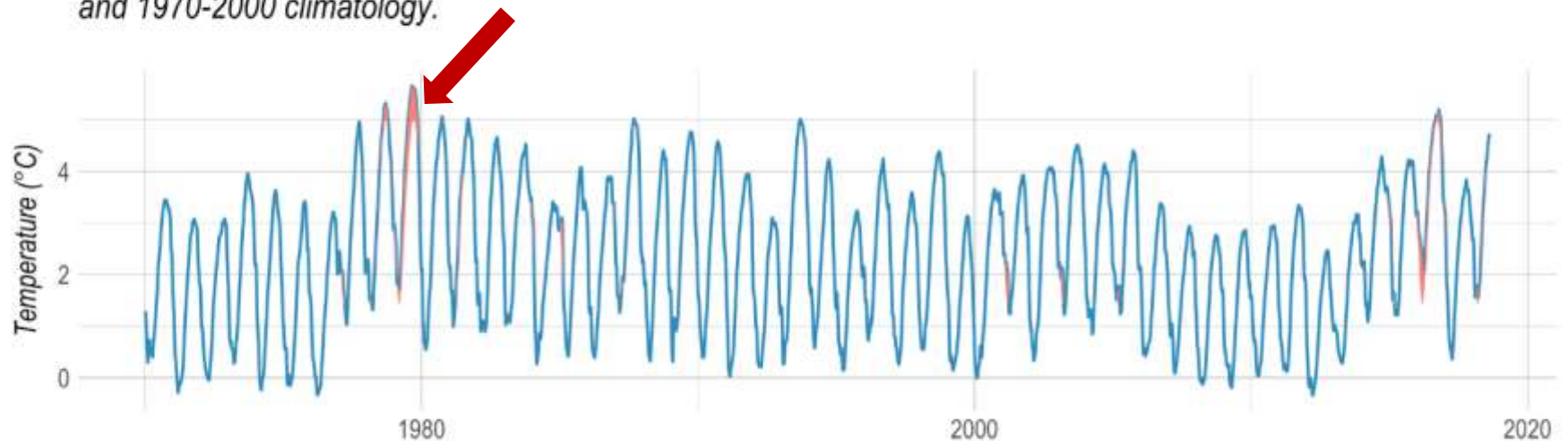
NEBS Bottom Temperature

Marine heatwave analysis based on downscaled ROMSNPZ hindcast and 1970-2000 climatology.



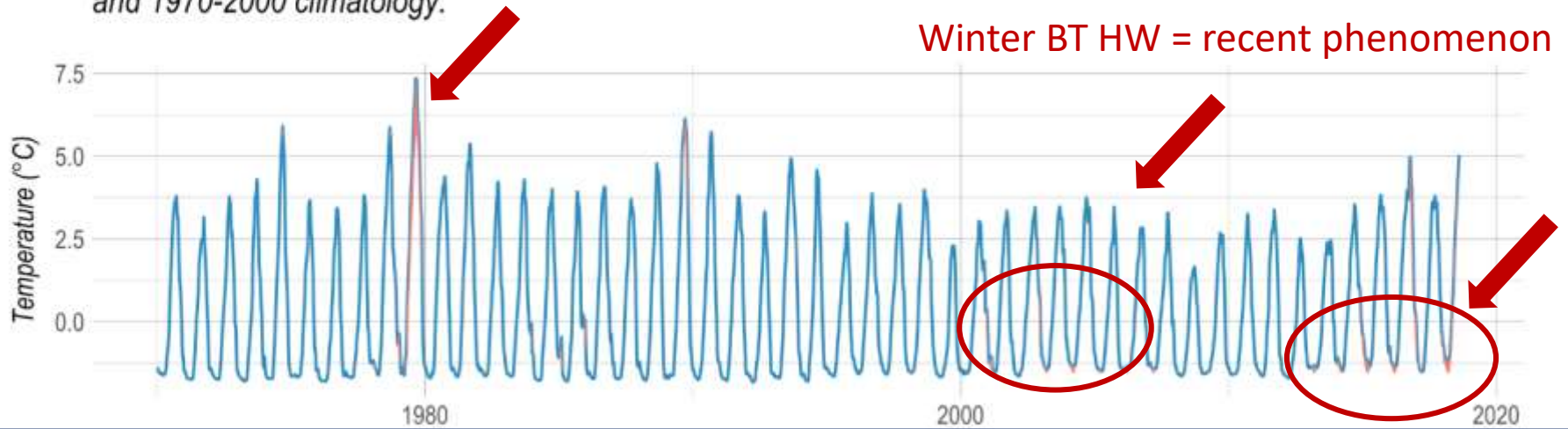
EBS Bottom Temperature

Marine heatwave analysis based on downscaled ROMSNPZ hindcast and 1970-2000 climatology.



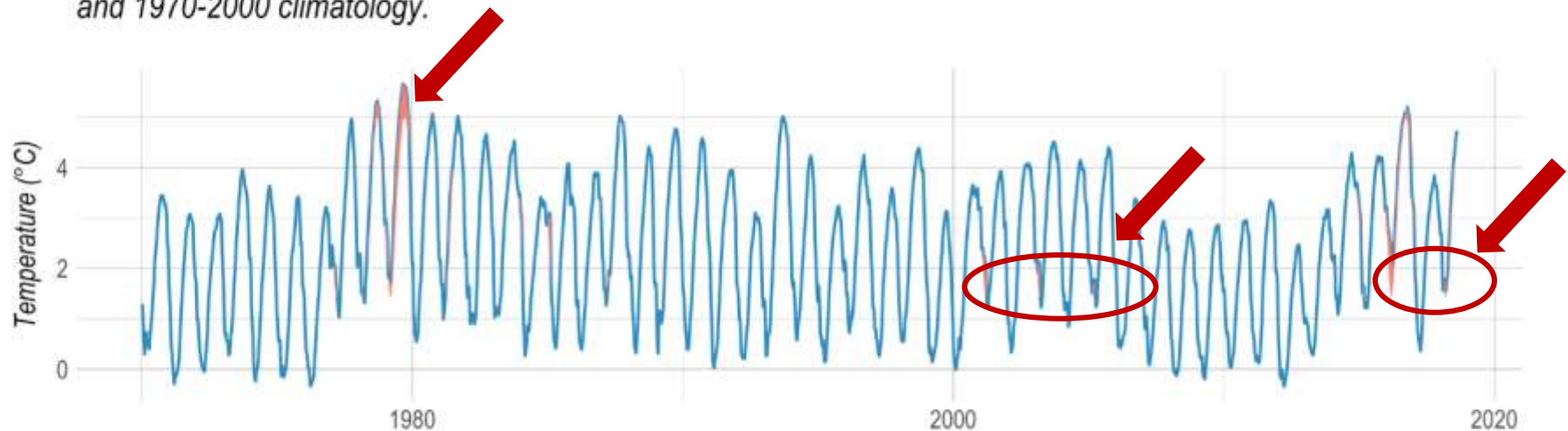
NEBS Bottom Temperature

Marine heatwave analysis based on downscaled ROMSNPZ hindcast and 1970-2000 climatology.



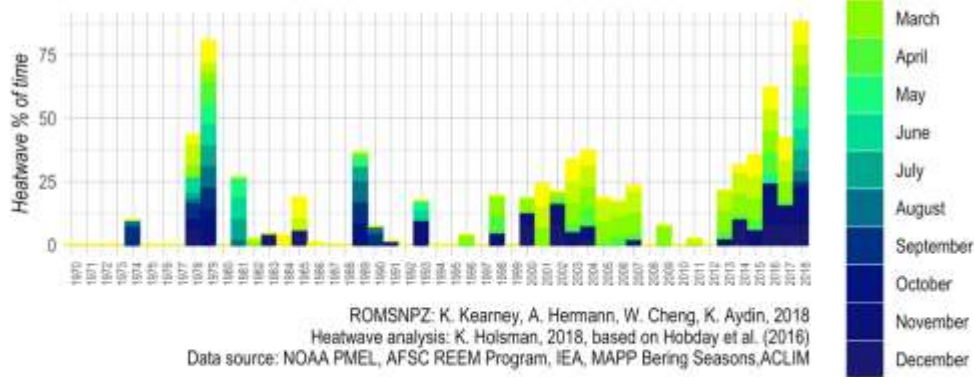
EBS Bottom Temperature

Marine heatwave analysis based on downscaled ROMSNPZ hindcast and 1970-2000 climatology.



Frequency NEBS Bottom Temperature

Marine heatwave analysis based on downscaled ROMSNPZ hindcast + projections, and 1970-2000 climatology.

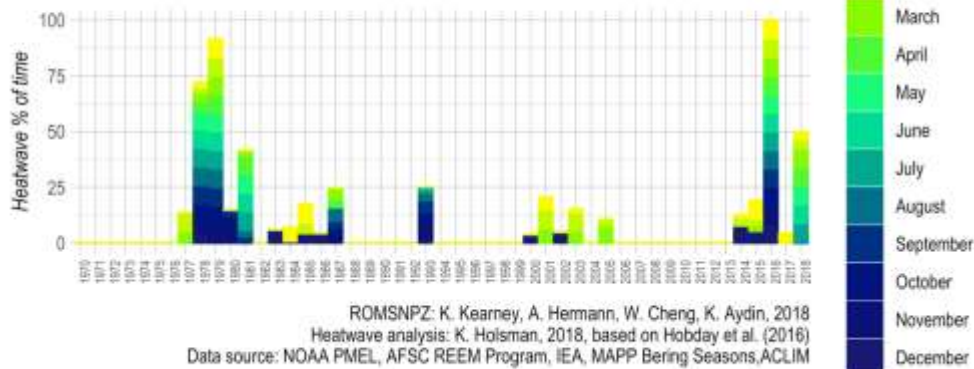


Overwinter survival during winter heat waves?

- Warm prevents direct physiological mortality, BUT
- Warm **with** prey promotes growth
- Warm **without** prey promotes starvation

Frequency EBS Bottom Temperature

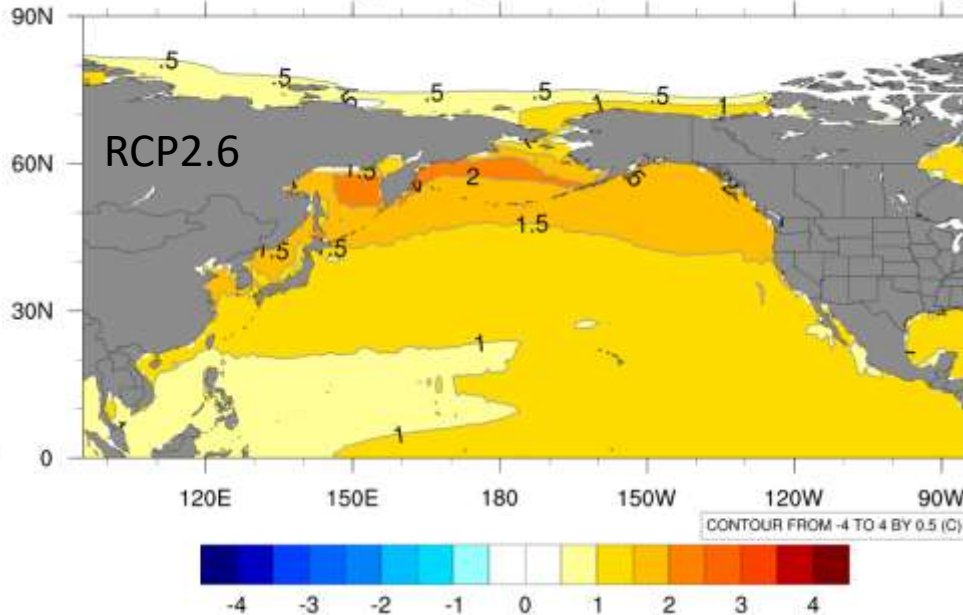
Marine heatwave analysis based on downscaled ROMSNPZ hindcast + projections, and 1970-2000 climatology.



CMIP5 ENSMN Annual SST anomaly ($^{\circ}\text{C}$) (2050 to 2099) - (1956 to 2005)

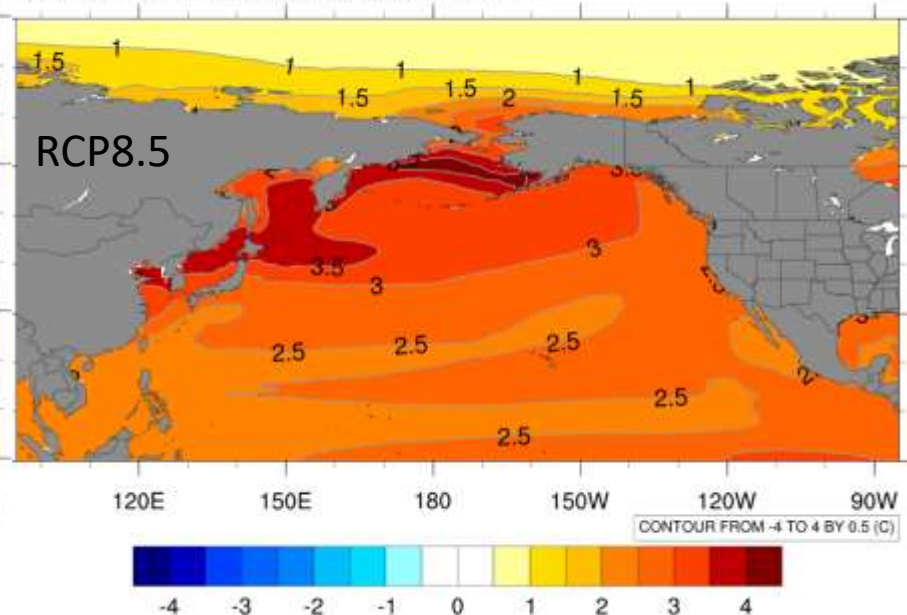
CO2 mitigation scenario

CMIP5 ENSMN RCP2.6 anomaly (2050-2099)-(1956-2005)



High baseline scenario ("Business as usual")

C CMIP5 ENSMN RCP8.5 anomaly (2050-2099)-(1956-2005)

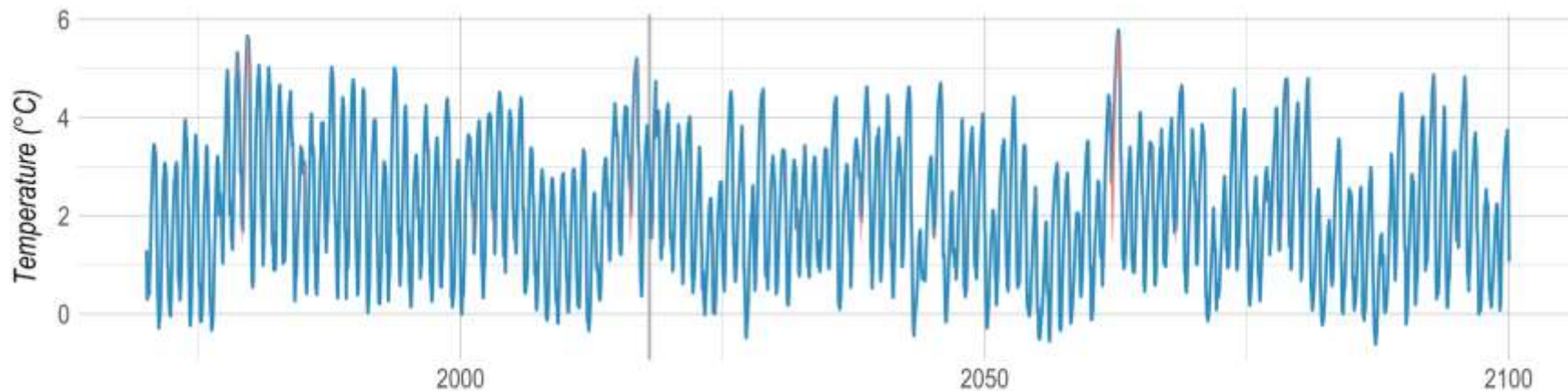


Projection data from CMIP5 (Taylor et al., 2012) avail. at: www.esrl.noaa.gov/psd/ipcc/ocn

Modified from Fig. 6.2 Holsman et al. 2018 [in] Barange et al. (Eds.) 2018. *Impacts of climate change on fisheries and aquaculture*. TP 627.

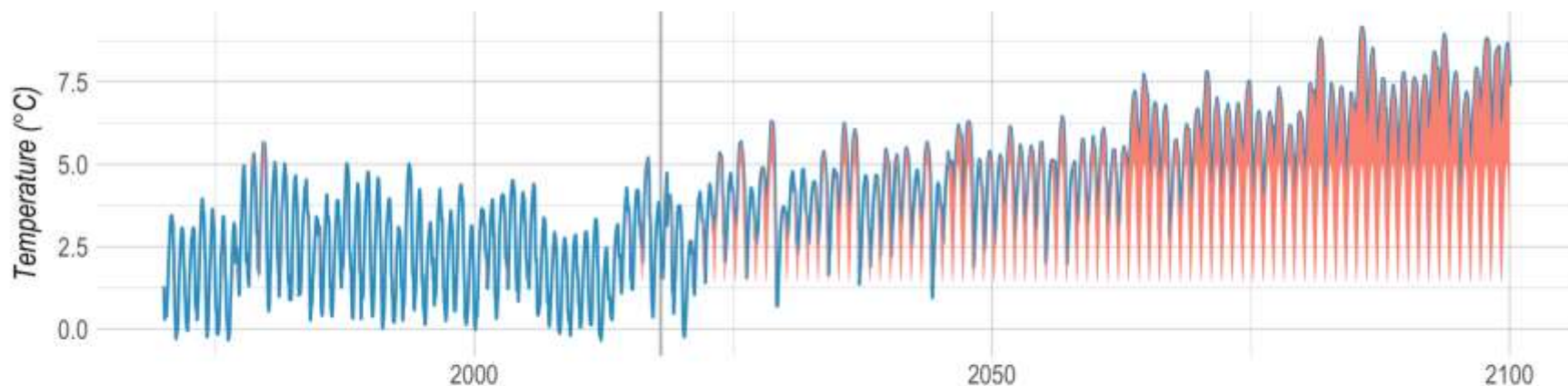
GFDL_rcp45 BT

Marine heatwave analysis based on downscaled ROMSNPZ hindcast and 1970-2000 climatology.



MIROC_rcp85 BT

Marine heatwave analysis based on downscaled ROMSNPZ hindcast and 1970-2000 climatology.



ROMSNPZ: K. Kearney, A. Hermann, W. Cheng, K. Aydin, 2018

Heatwave analysis: K. Holsman, 2018, based on Hobday et al. (2016)

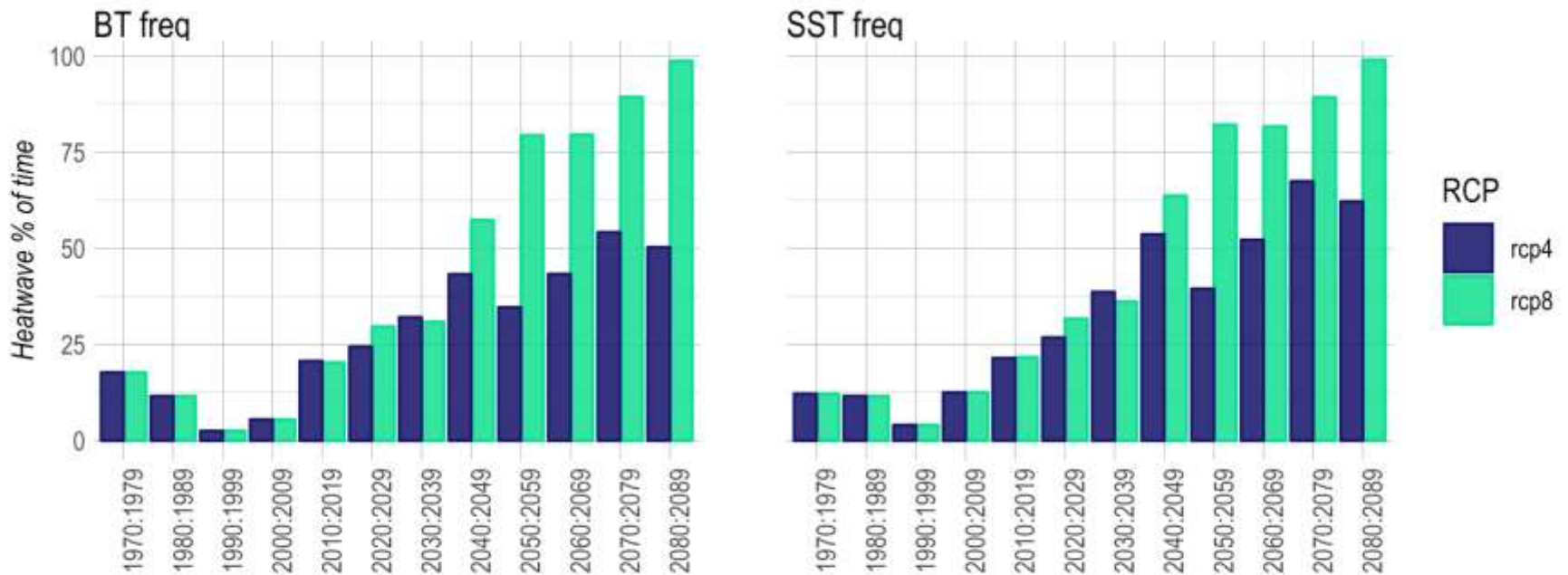
Data source: NOAA PMEL, AFSC REEM Program, IEA, MAPP Bering Seasons, ACLIM

Marine heatwaves will likely increase in frequency and duration

Heatwaves
Now ~ 21% of the time
2050 ~ 30-77% of the time
2100 ~ 60-90% of the time

Duration

Marine heatwave analysis based on downscaled ROMSNPZ hindcast + projections, and 1970-2000 climatology.

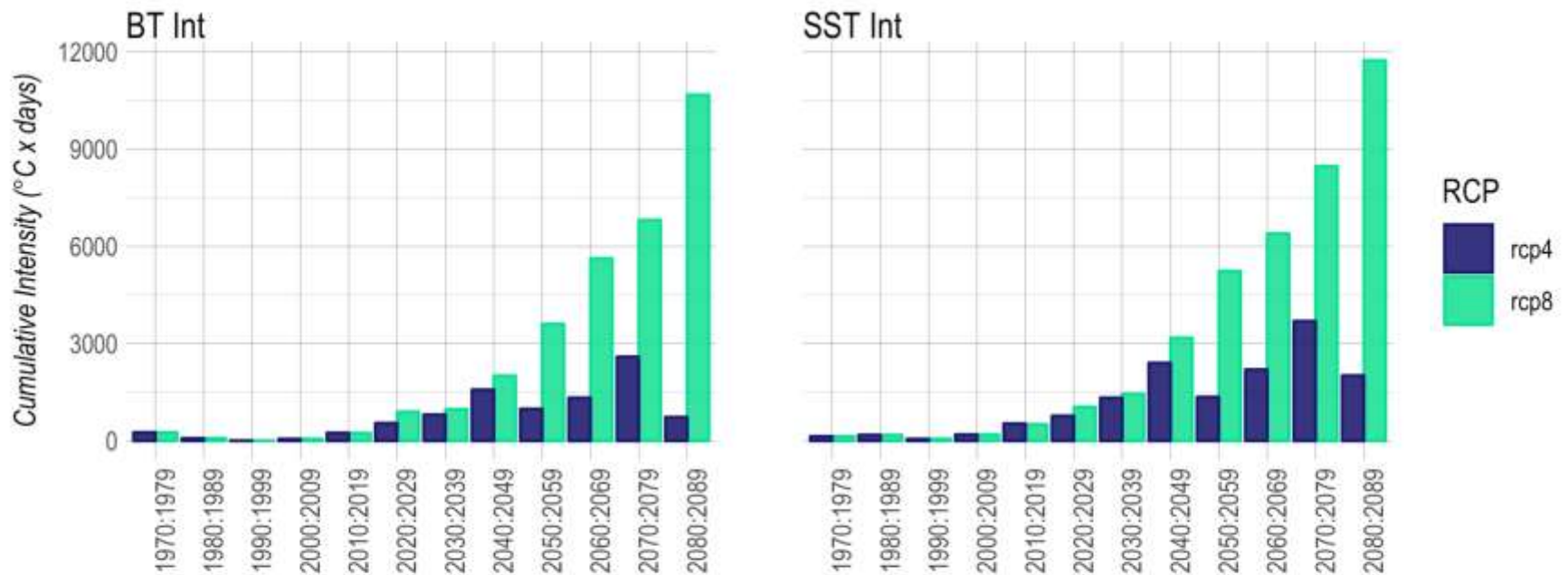


ROMSNPZ: K. Kearney, A. Hermann, W. Cheng, K. Aydin, 2018
Heatwave analysis: K. Holsman, 2018, based on Hobday et al. (2016)
Data source: NOAA PMEL, AFSC REEM Program, IEA, MAPP Bering Seasons, ACLIM

Marine heatwaves will likely increase in intensity

Intensity

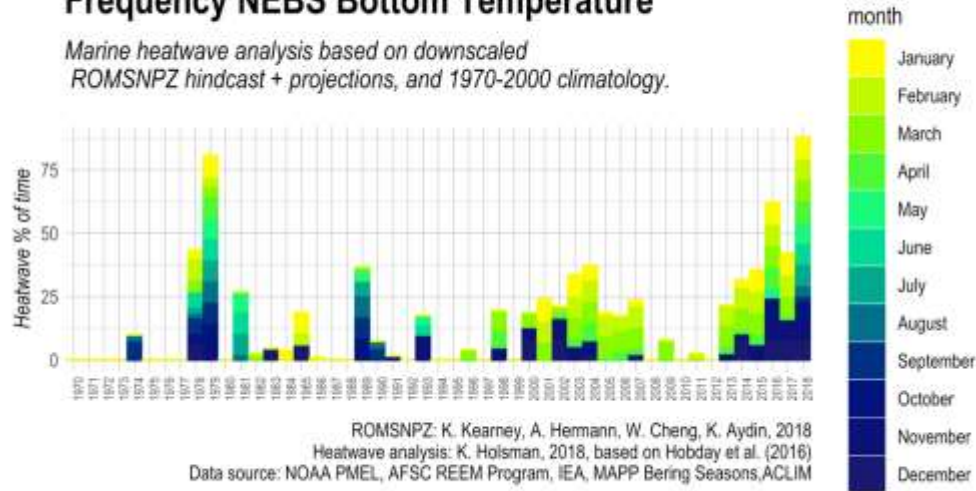
Marine heatwave analysis based on downscaled ROMSNPZ hindcast + projections, and 1970-2000 climatology.



ROMSNPZ: K. Kearney, A. Hermann, W. Cheng, K. Aydin, 2018
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Data source: NOAA PMEL, AFSC REEM Program, IEA, MAPP Bering Seasons, ACLIM

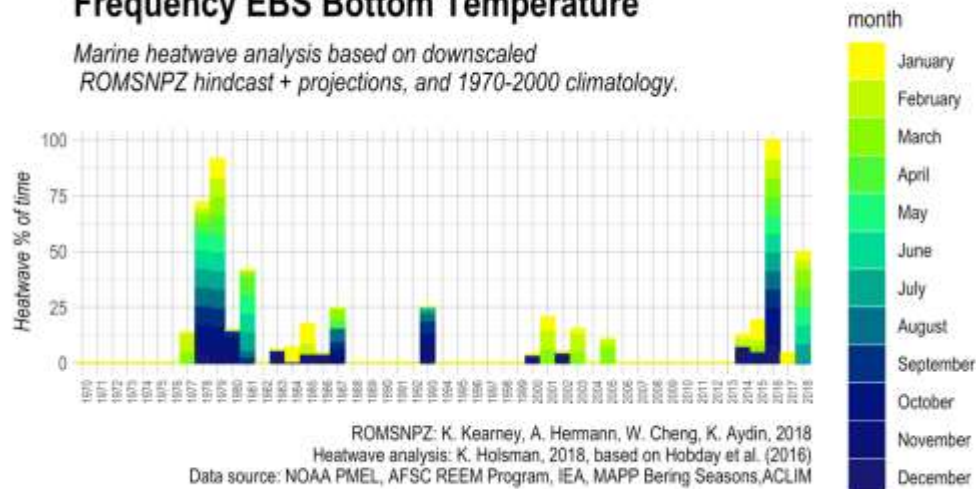
Frequency NEBS Bottom Temperature

Marine heatwave analysis based on downscaled
ROMSNPZ hindcast + projections, and 1970-2000 climatology.



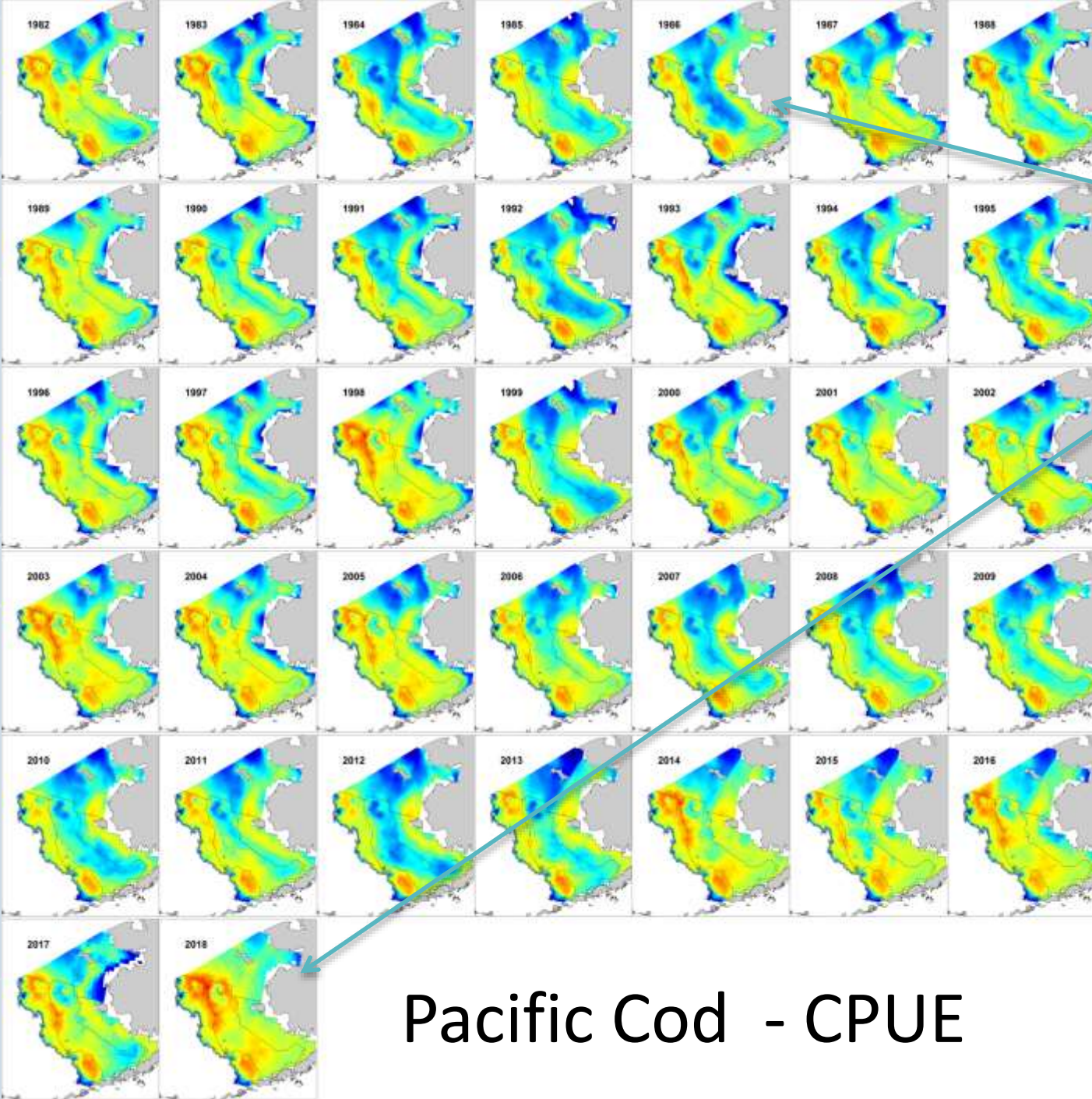
Frequency EBS Bottom Temperature

Marine heatwave analysis based on downscaled
ROMSNPZ hindcast + projections, and 1970-2000 climatology.



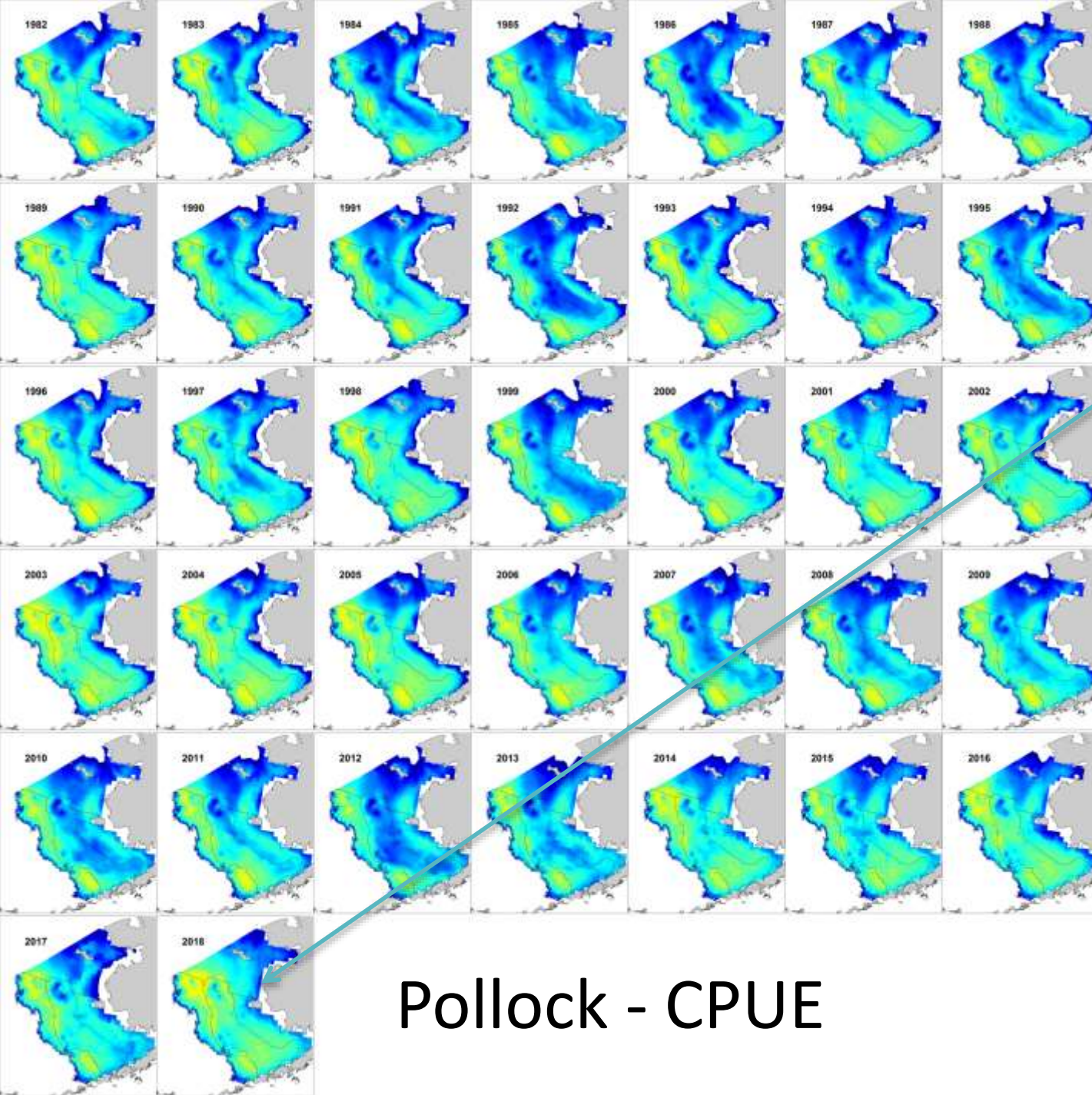
EFH PREDICTED DISTRIBUTIONS (ROOPER)

- Same approach as for 2017 EFH
 - Generalized additive models
 - ROMS hindcasted temperatures
 - Other variables: depth, slope, current speed, tidal current speed, sediment size
- Maps present log-transformed predictions – $\log(\text{CPUE})$
 - Not scaled the same
- Graphs and calculations are on back-transformed data



- Distribution influenced by cold pool
- Bottom temperature most important variable in model
- Consistently ~ 74% of the population is inside the survey area
- No trend over time

Pacific Cod - CPUE



- Distribution influenced by cold pool
- Bottom temperature most important variable in model
- Declining proportion of the population inside survey area
- ~ 77% in 2018

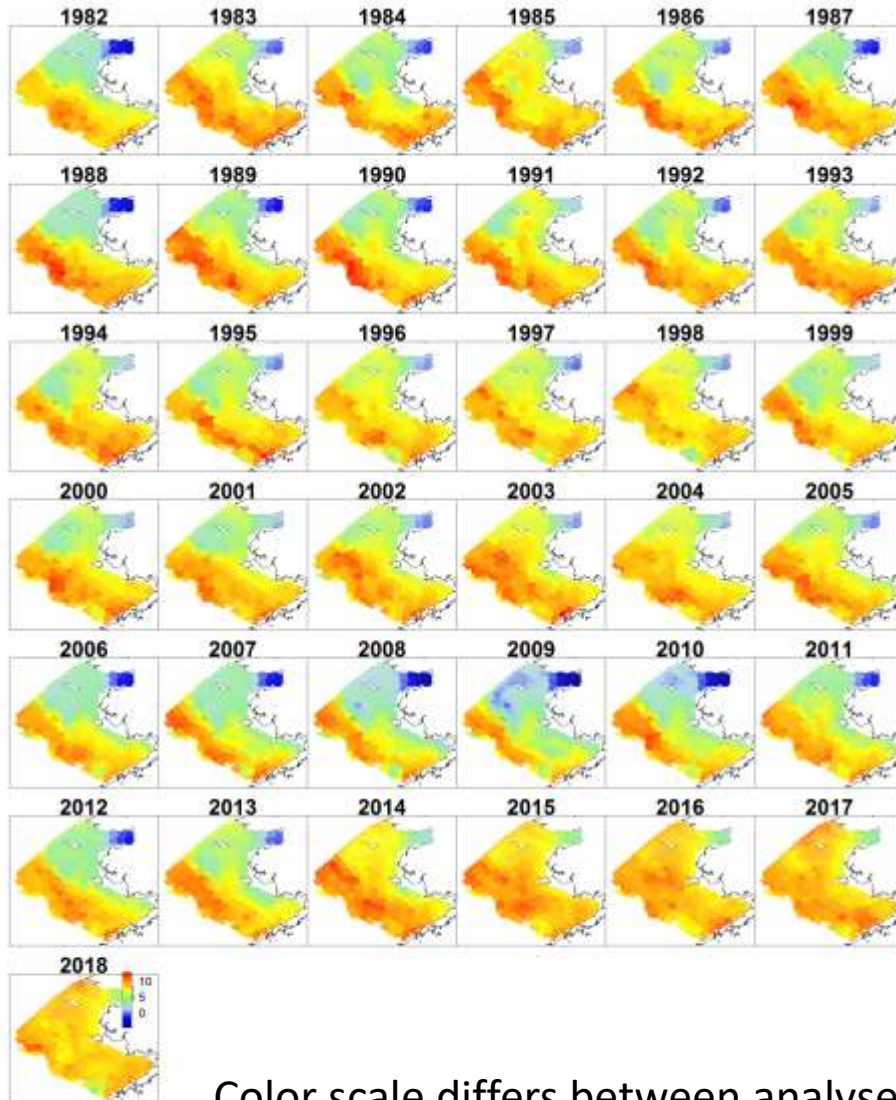
Pollock - CPUE

VAST MODEL (THORSON)

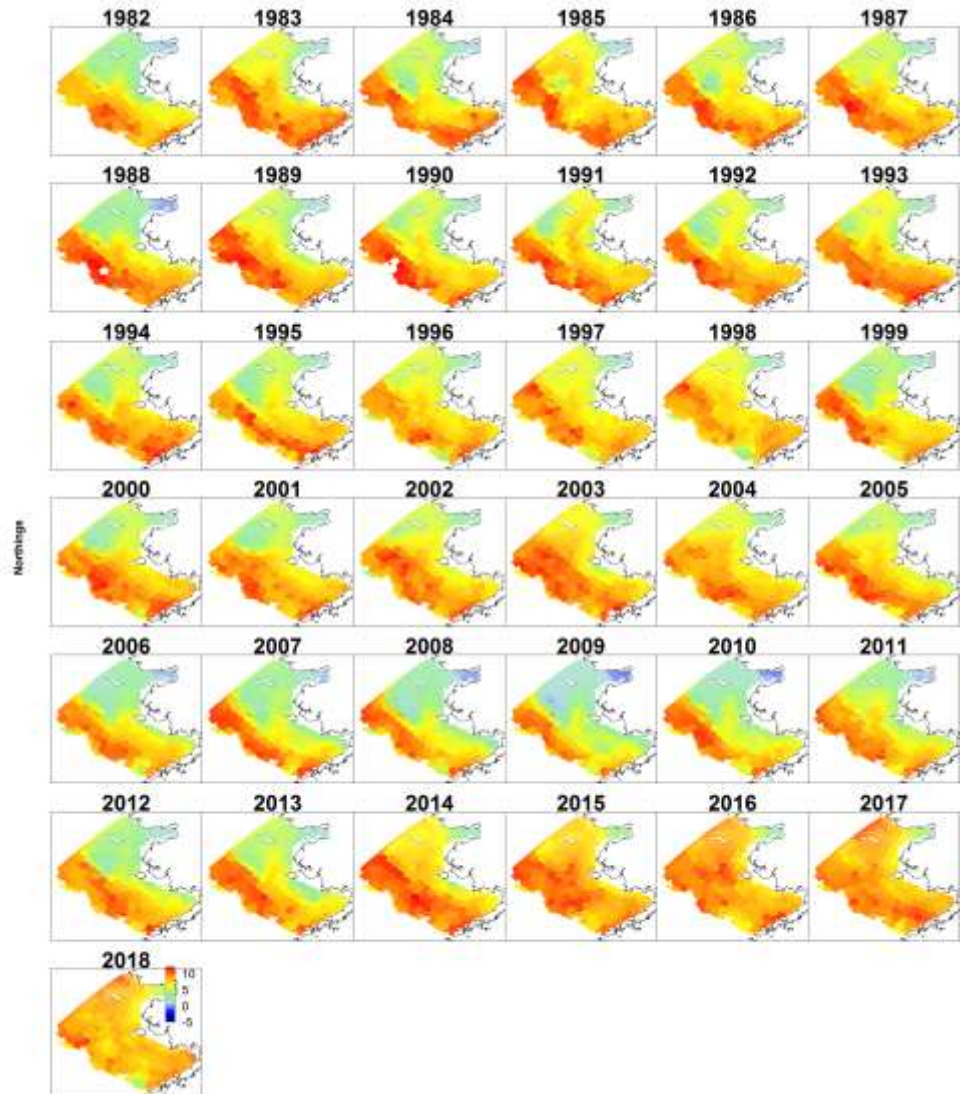
- Multivariate spatio-temporal model that estimates variation in population density across space and time for multiple sizes, ages, and/or species based on survey and fishery data
- Used in ecosystem and stock assessments primarily in the North Pacific and South Africa

Pollock distribution - Comparing with vs. without temperature

Without temperature



With temperature



Color scale differs between analyses

Eastings

Eastings

Pollock distribution - Comparing with vs. without temperature

Conclusions

1. Including temperature has relatively little impact on relative biomass in NBS vs. total

