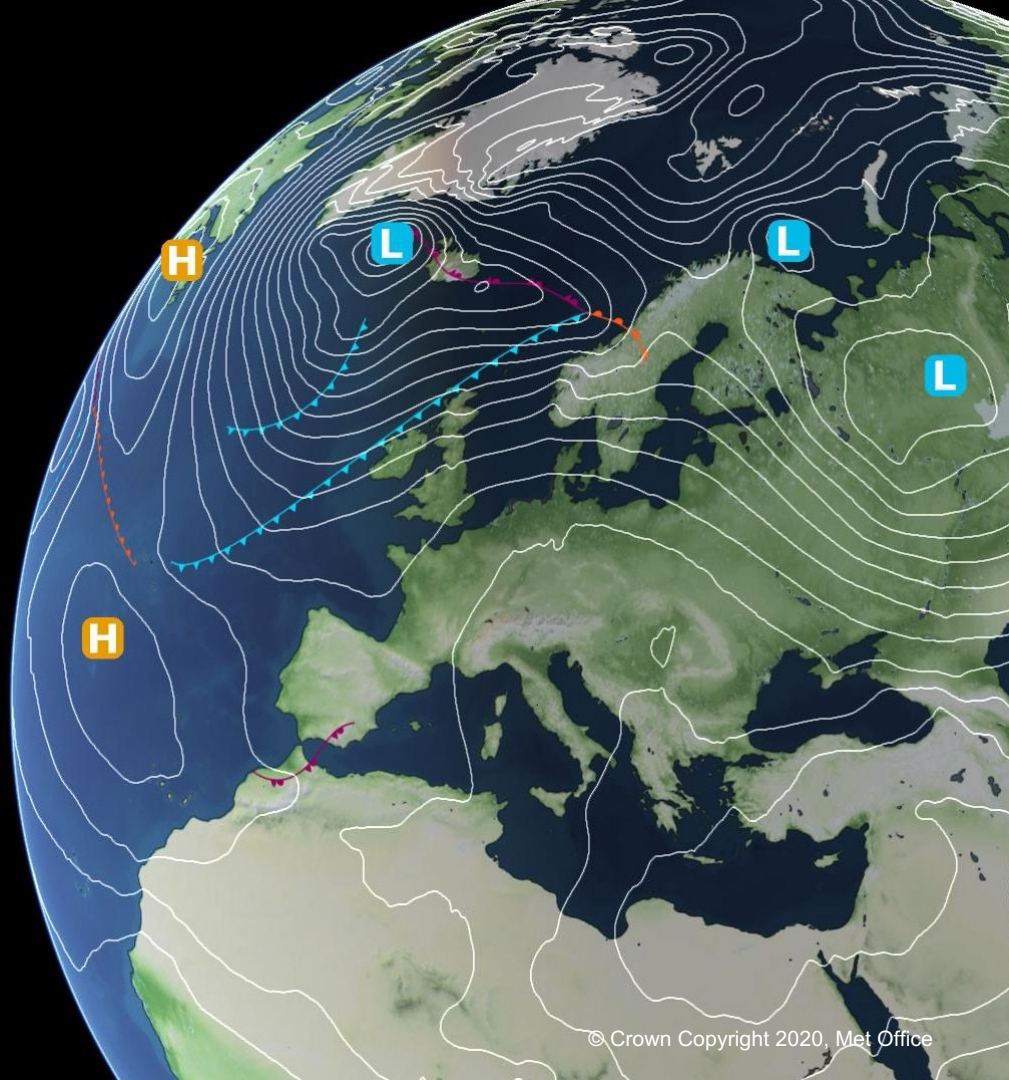


# Future directions for convection-permitting climate modelling at the Met Office

Chris Short

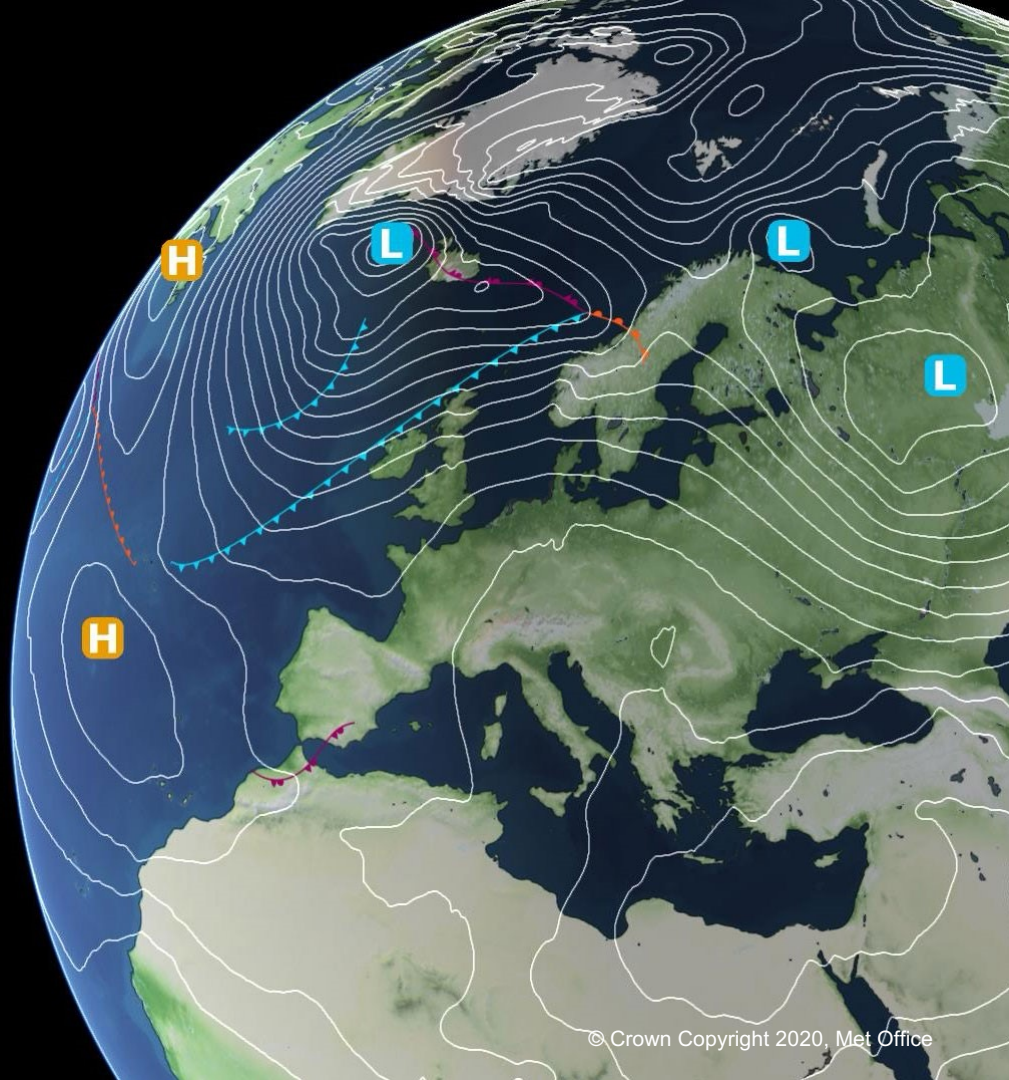
2024 BoM Annual R&D Workshop

11/09/2024



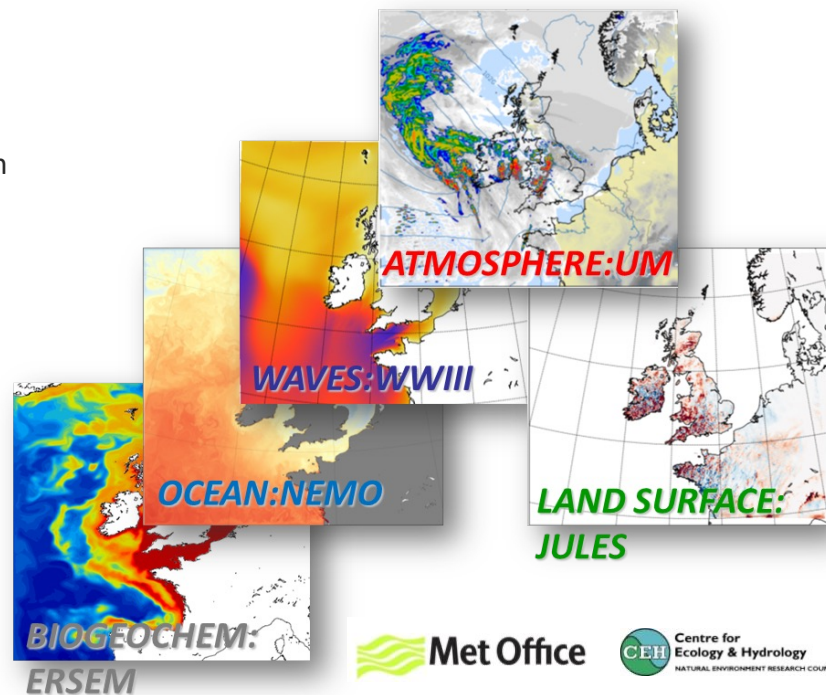
# Adding environmental complexity to UK climate projections

Alex Arnold, Segolene Berthou, Juan Castillo,  
Chris Short, Simon Tucker



# Motivation

- Current set of UK national climate projections (UKCP18) included a 12-member 2.2 km CPM for the first time
  - Atmosphere coupled to land, SSTs imposed as daily means from driving GCM
- Aim to run climate change experiments with the Regional Environmental Prediction (REP) system
  - Fully coupled: atmosphere, land, ocean (with explicit tides), waves, rivers
- Key topics to investigate:
  - The importance of processes missing from atmosphere-only simulations for regional climate change projections (e.g. coupled system feedbacks, tidal mixing, river outflows)
  - The impact of climate change on multi-hazard compound events
  - The interdependence of compound marine and land heat waves





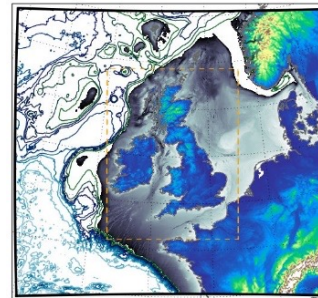
# UK climate runs with the Regional Coupled Suite

## Configuration:

- Atmosphere/land: RAL3.3.1
- Ocean/waves: CO8
- Explicit tides
- Rivers: River Flow Model (RFM)

## Resolution:

Atmosphere/land: 4.4km -> 2.2km  
Ocean: 1.5km  
Waves: 3km -> 1.5km



## 3x timeslice runs (RCP8.5)

→ UKCP18 2.2 km projections (standard member)

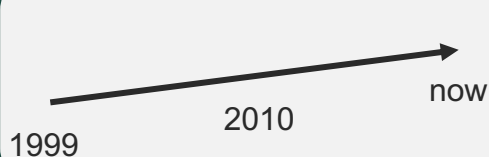
→ Ocean AMM15 spin-up

→ Coupled run

1990-2000

2065-2075

Alex Arnold

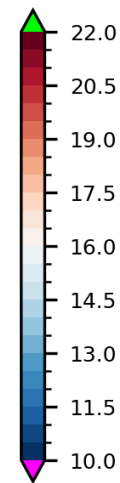
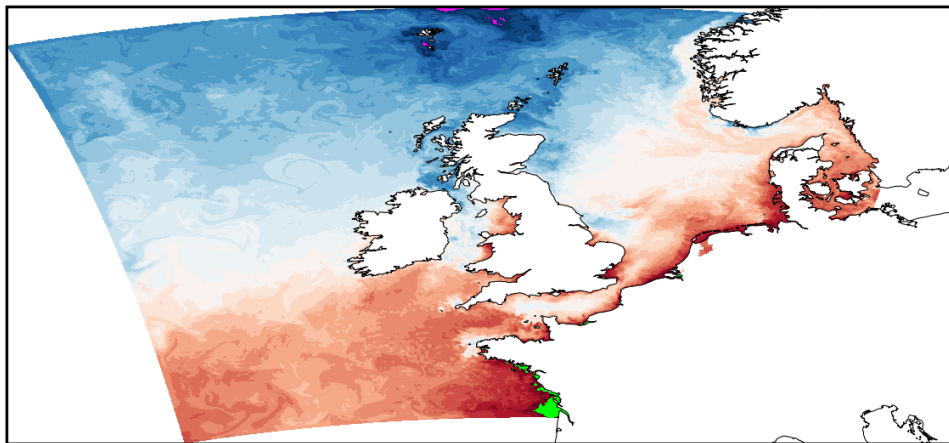


## Hindcast run

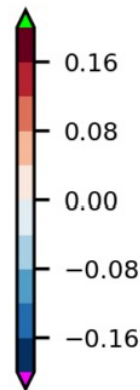
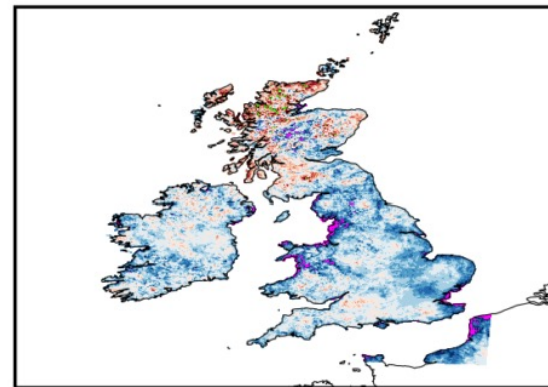
ERA-5 boundaries (atmosphere & waves)  
GloSea5 boundaries (ocean)

# Current status and future directions

- Runs are underway:
  - Hindcast atm-only: ~4.5 yrs in
  - Hindcast cpld: ~3.5 yrs in
  - Present-day atm-only: ~2.5 yrs in
  - Present-day cpld: ~0.5 yrs in
  - Future atm-only: ~2 yrs in
  - Future cpld: ~0.5 yrs in
- This will be the first look at the REP in UK climate mode:
  - Opportunities for inclusion in next generation of UK climate information



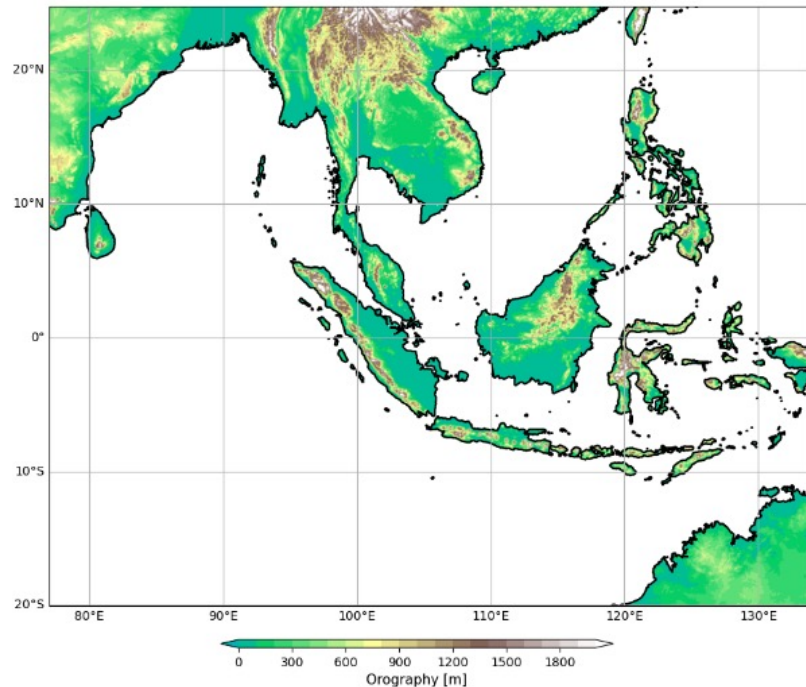
Coupled - Atm-only



DJF mean daily tasmin

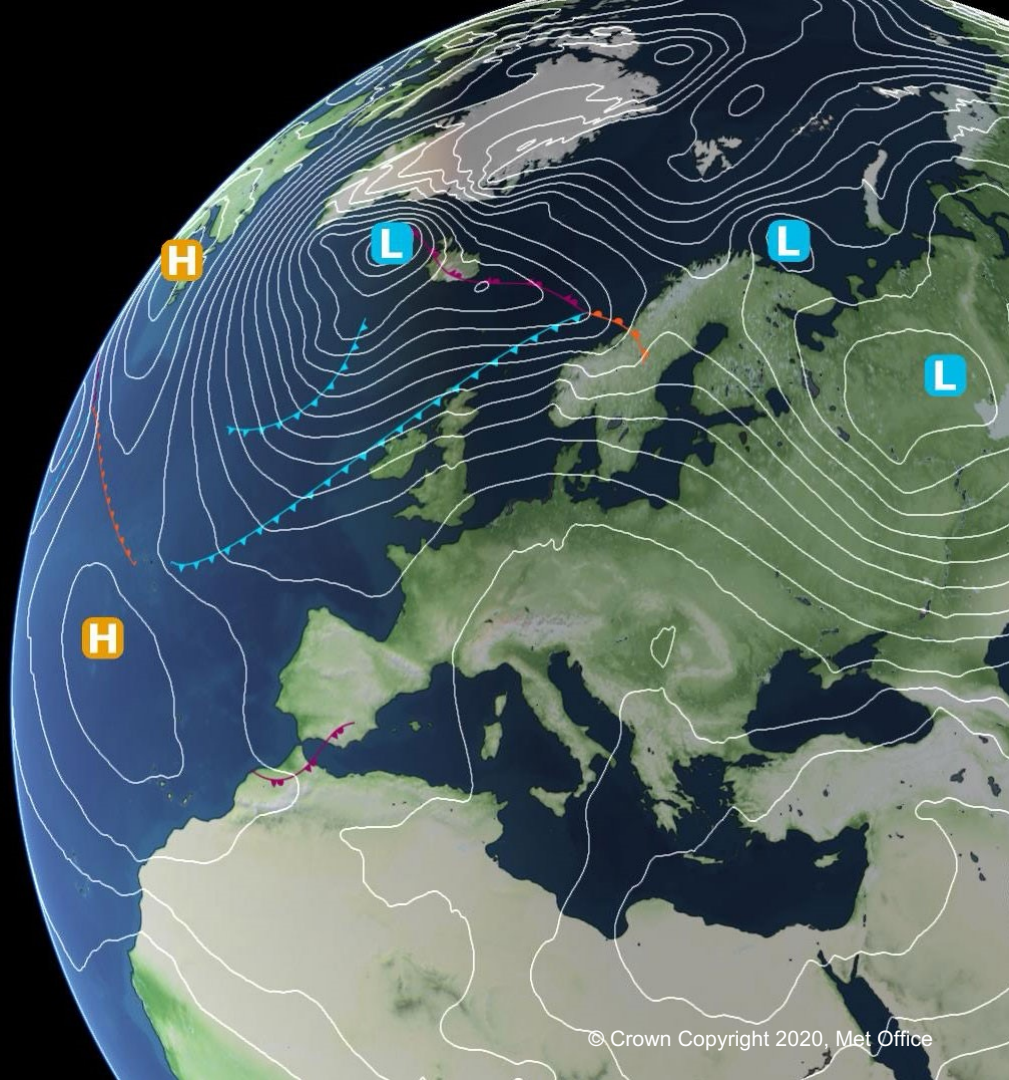
# Advert: Coupled Maritime Continent runs

- Recently completed a pair of simulations over the Maritime Continent:
  - **Atmosphere-only:** 4.4 km atmosphere (RAL3.1/3.2) with hourly LBCs from ERA5 and daily mean SSTs prescribed from an ocean-only run of the 1/12 deg NOC regional ocean model (forced by Drakkar reanalysis)
  - **Coupled:** same as above but with the regional atmosphere fully coupled to the 1/12 deg NOC regional ocean model. Ocean LBCs and ICs from NOC.
- NOC 1/12 deg regional ocean model is shelf-enabled with explicit tides
- 10 years long 2002 – 2012
- Access to data available upon request
  - Email me if interested: [christopher.short@metoffice.gov.uk](mailto:christopher.short@metoffice.gov.uk)



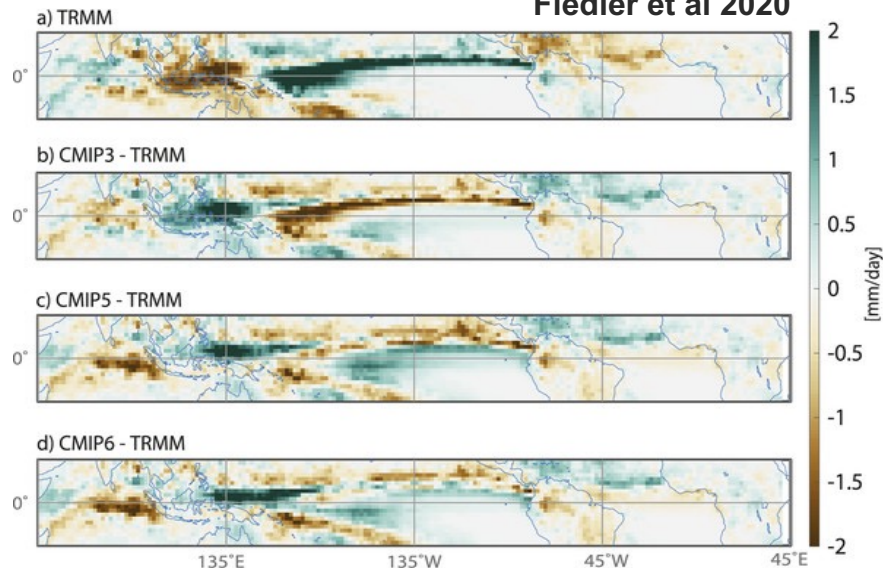
# Preparing for km-scale global climate models of the future

Calum Scullion, Chris Short + the wider K-  
Scale team

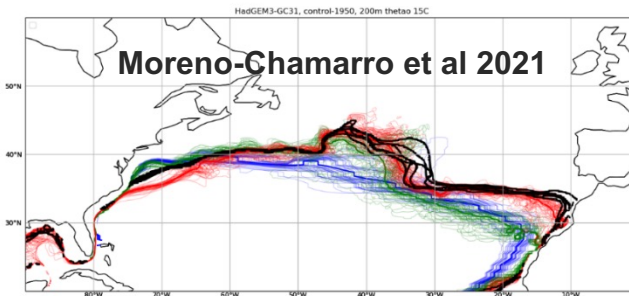


- Can km-scale global models offer a step-change over conventional CMIP-class models?
  - Local impacts: explicit convection improves representation of precipitation extremes – critical for adaptation
  - Upscale impacts: improved representation of tropical precipitation affects mid-latitude climate via global teleconnections
  - Different cloud distributions and properties, modifying feedbacks and impacting climate sensitivity
  - Ocean mesoscale important for regional climate change too
- Challenges:
  - Computationally expensive
  - Huge data volumes
  - Convection not fully resolved
  - Tuning

Fiedler et al 2020

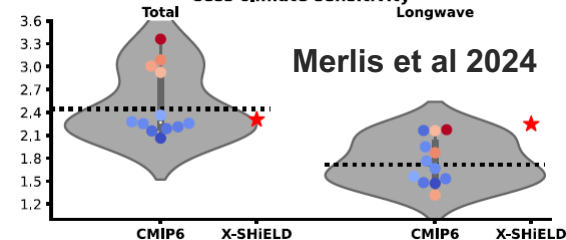


Moreno-Chamorro et al 2021



Cess climate sensitivity  
Longwave

Merlis et al 2024

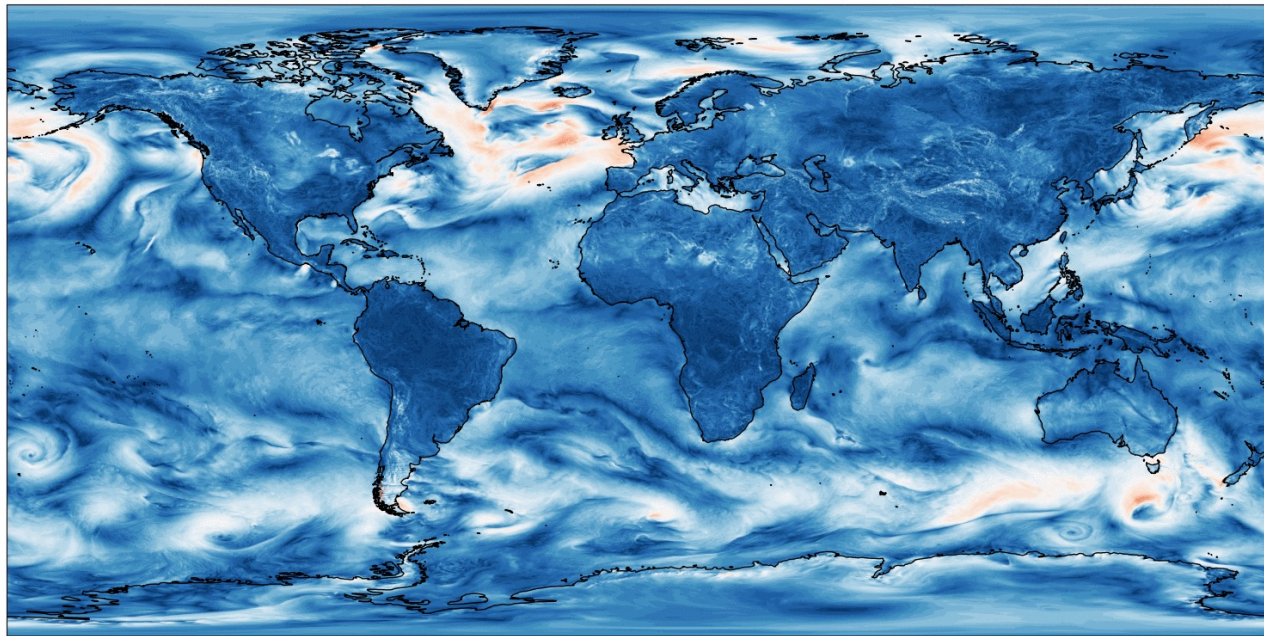




# Current status and future directions

- N2560 global model with RAL3.2 and GAL9 ancils now runs!
- Next steps:
  - Design year-long atm-only and atm-only +4K SST runs to contribute to DYAMOND v3 and HighResMIP2
  - Potentially a year-long run with ocean coupling for DYAMOND v3 too
  - Can we extend to multi-decadal on new HPC for full AMIP and AMIP +4K SST runs?!

N2560 L70 80km - RAL3 WIND at T+01 20171230T0000Z

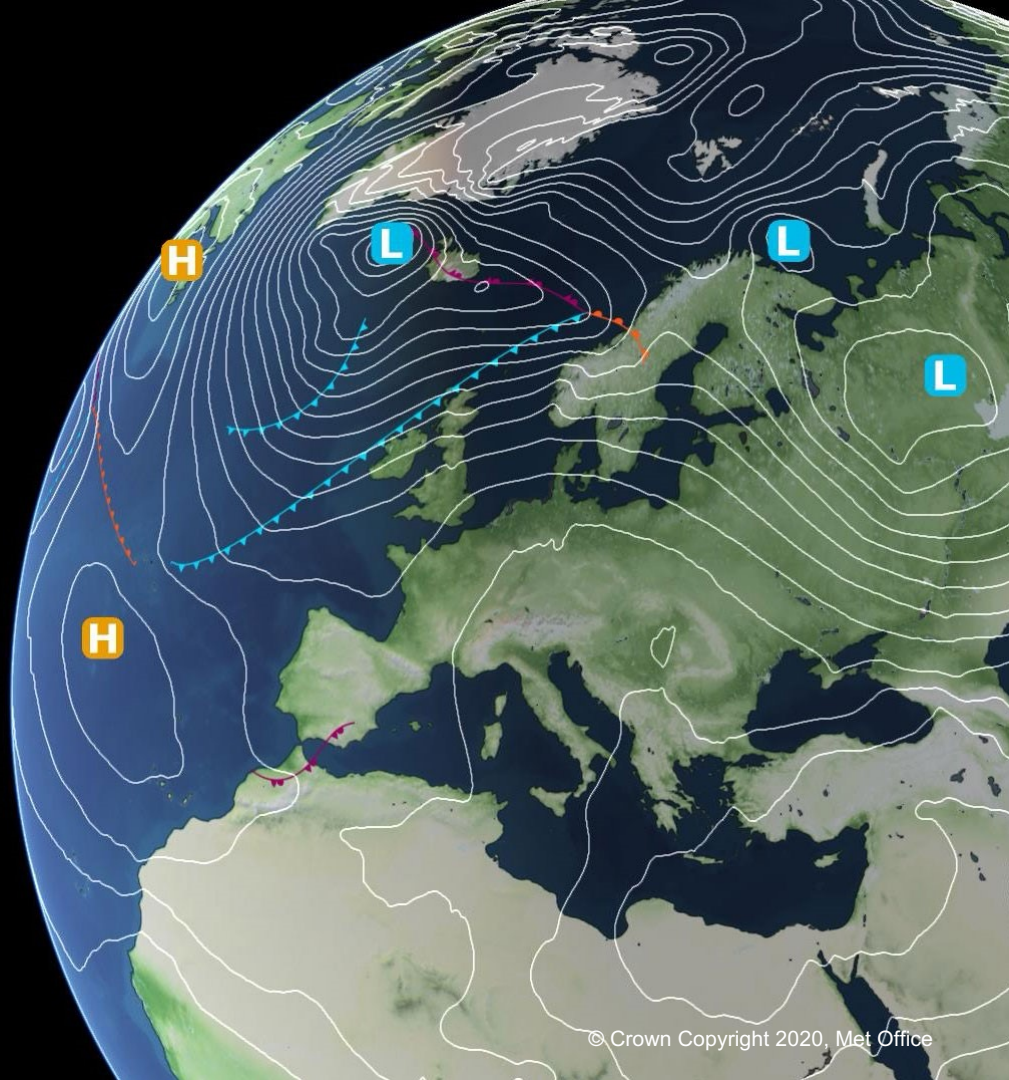


DYAMOND v3 protocol: <https://www.researchsquare.com/article/rs-4458164/v1>  
HighResMIP2 protocol: <https://egusphere.copernicus.org/preprints/2024/egusphere-2024-2582/>

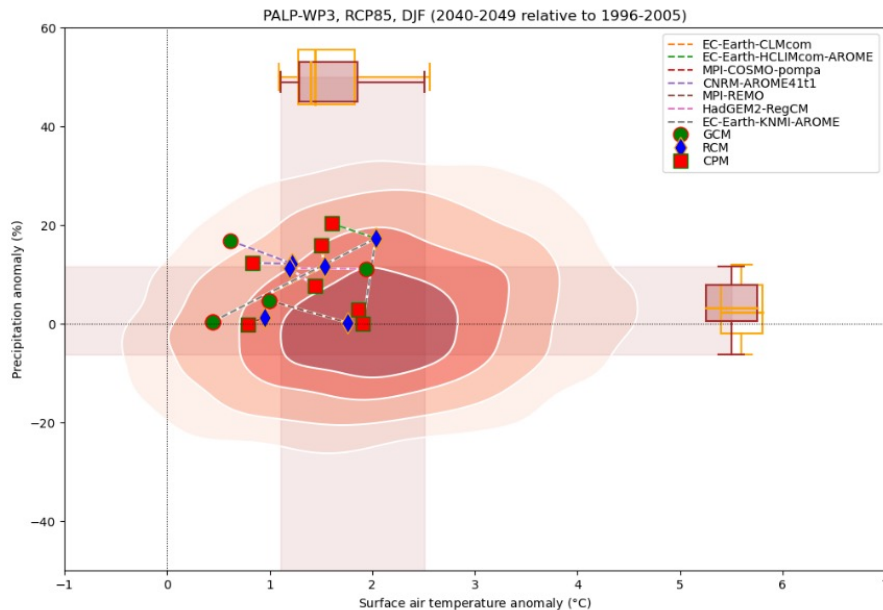
Huw Lewis

# Emulation of CPMs using machine-learning techniques

Lizzie Kendon, Henry Addison, Antoine Doury, Ben Booth, Samuel Somot, Peter Watson, Erika Coppola, Jose Manuel Gutierrez, James Murphy, Calum Scullion

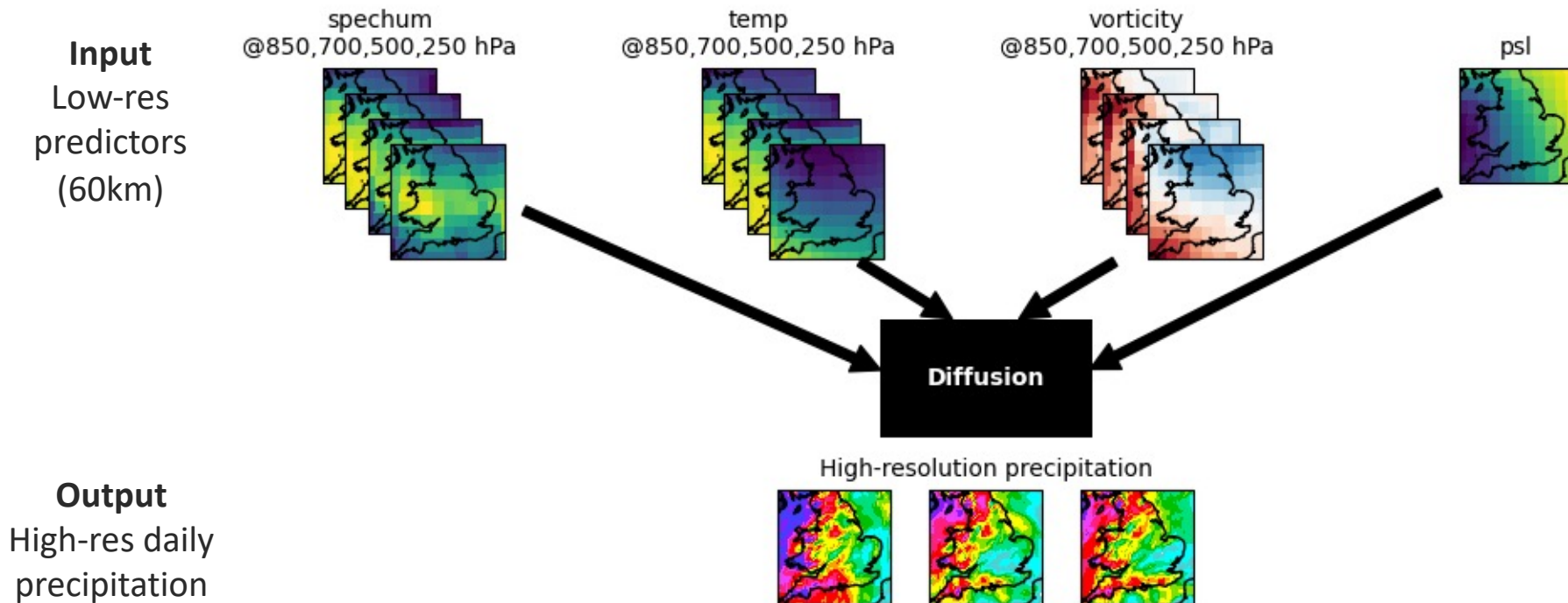


- Key potential benefits of ML down-scaling:
  - Augment high resolution regional climate simulations at a fraction of the cost
  - Provide a more comprehensive sampling of uncertainty (downscaling other GCMs/scenarios) to support decision making
  - Downscale a much larger set of GCMs (not just those for which LBC data is available)
  - Allow rapid production of local climate information
- See “*Potential for machine learning to augment regional climate simulations in provision of local climate change information*”, Kendon et al 2024, submitted to BAMS.



Kendon et al 2024

# Emulating UKCP Local precipitation with ML



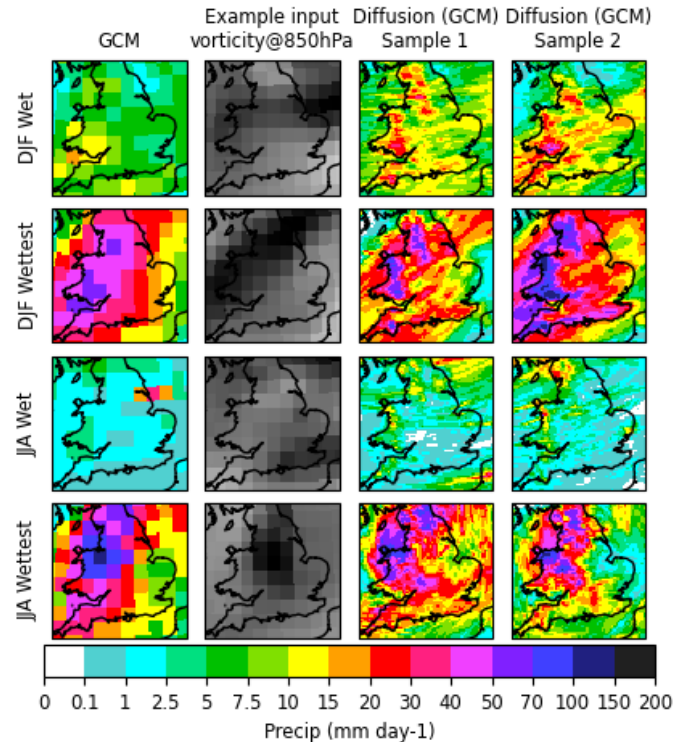
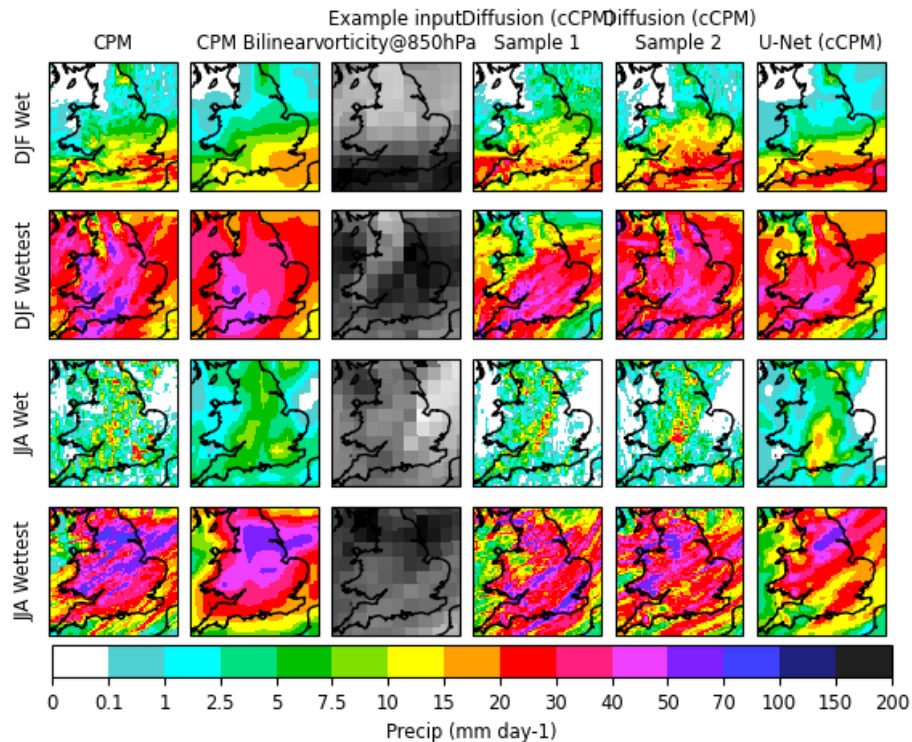


# Example output

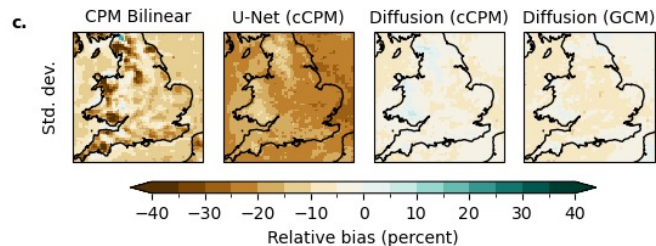
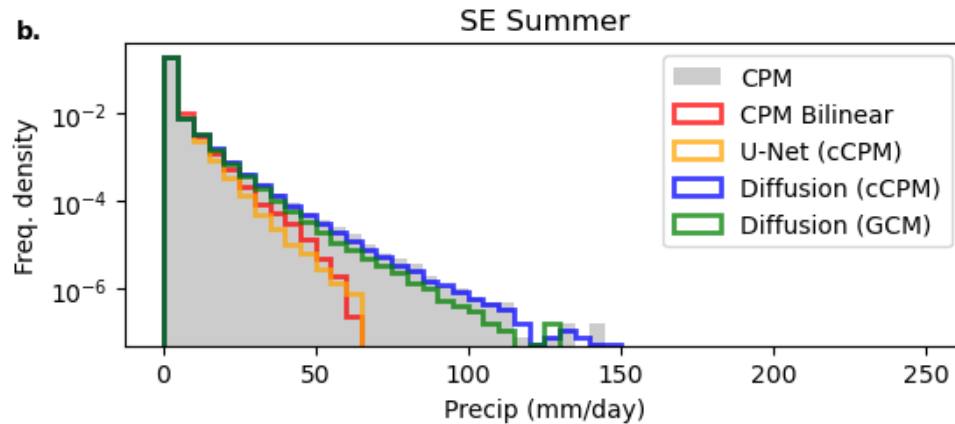
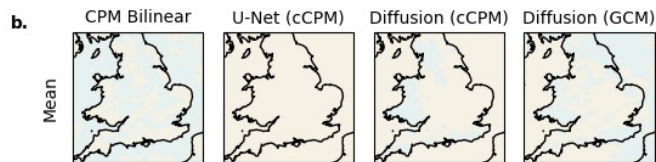
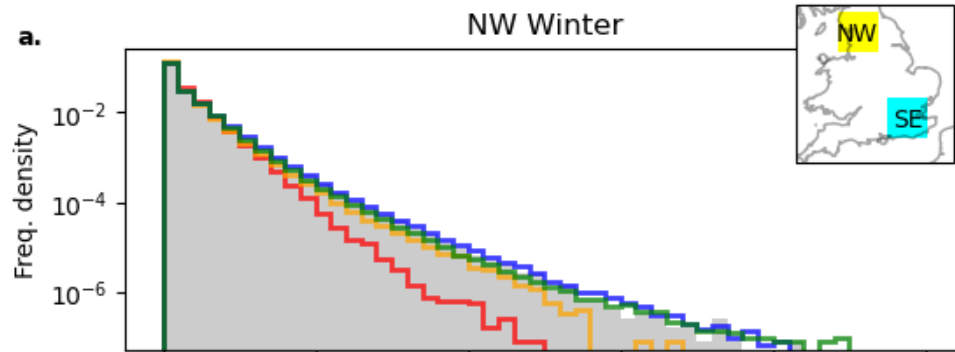
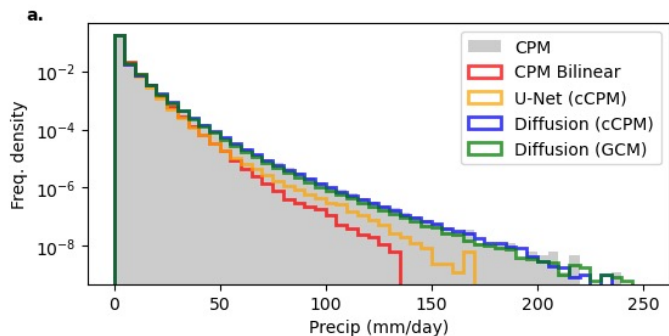
Target Coarsened  
CPM

Samples using coarse-  
grained CPM as input

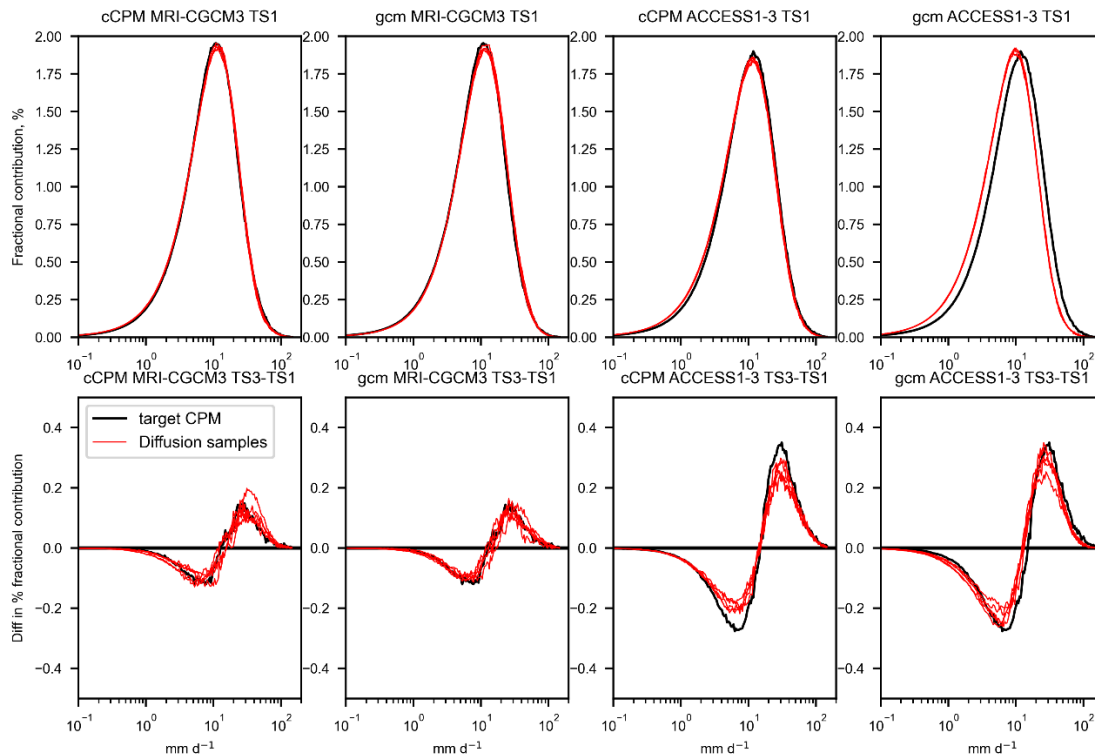
Samples using  
inputs from GCM



# Skill for emulating daily precipitation



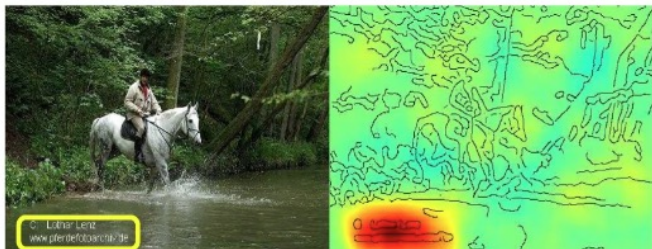
# Transferability to other GCMs



Kendon et al 2024

# Future directions

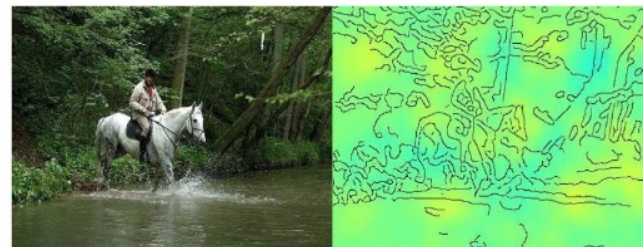
- Several key areas for future work:
  - Can ML approaches give reliable predictions for extremes?
  - Transferability to other GCMs
  - Multi-variate emulation: required for risk assessments of compound hazards and impact metrics such as drought and heat stress
  - Evaluation of ML skill:
    - Weather and climate knowledge required to define what “good” looks like
    - Develop open benchmark datasets to assess ML added value (e.g. compared to classical statistical down-scaling)
    - Interpretability – right answers for right reasons?



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present



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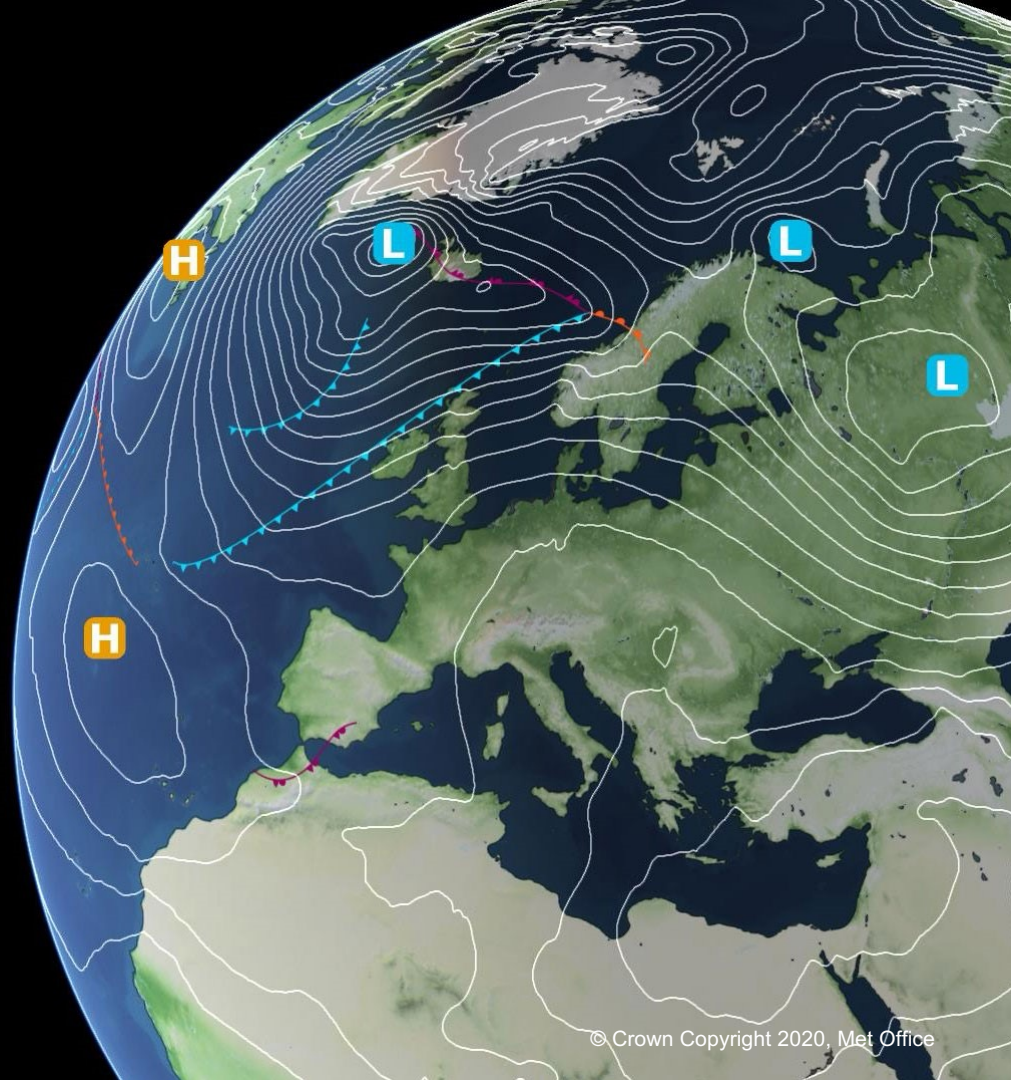


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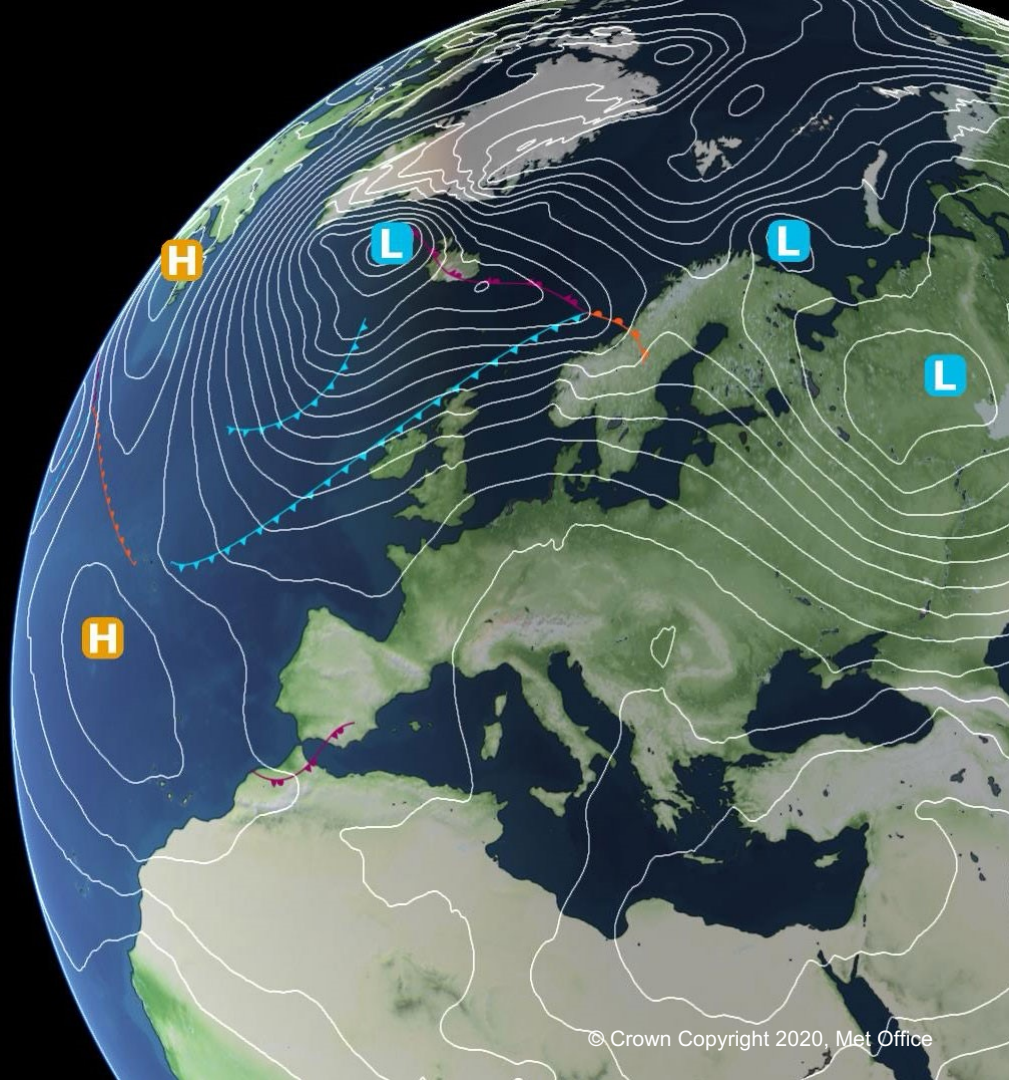
Lapuschkin et al 2019



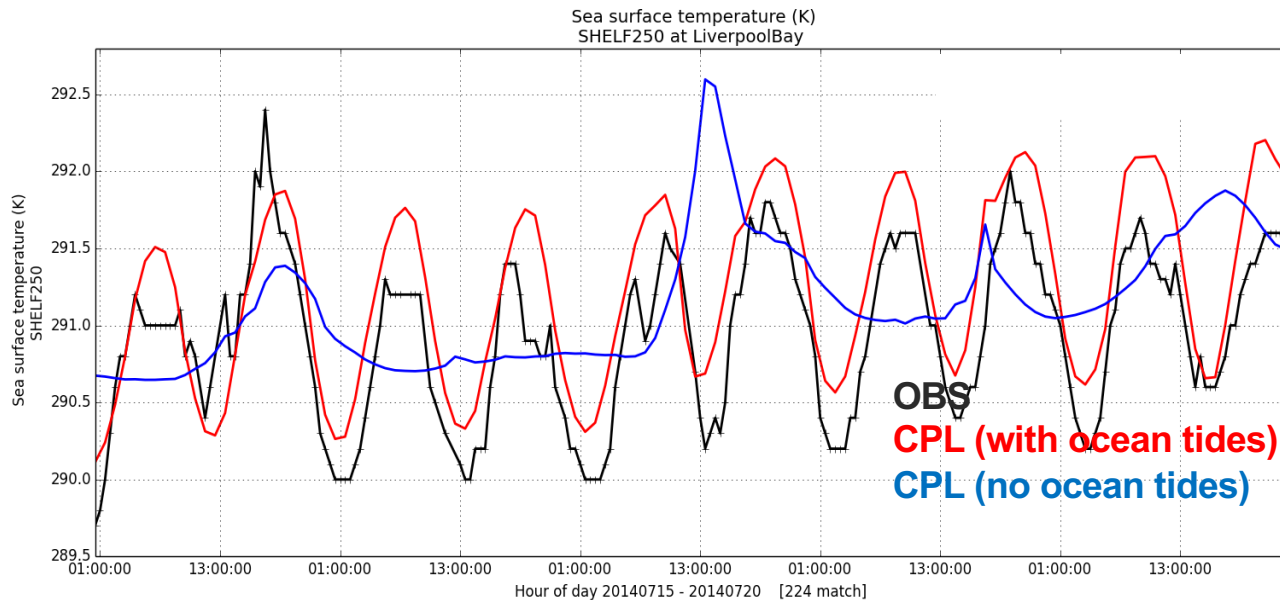
Thank you!



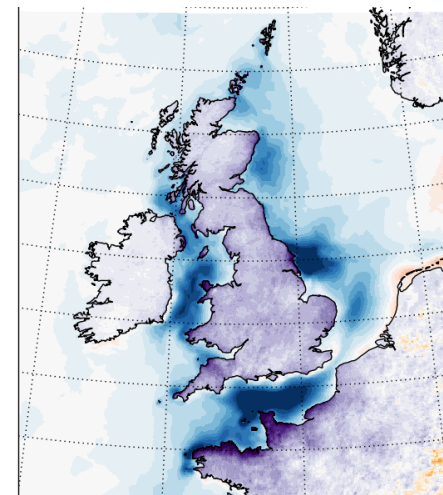
# Spare slides



# Improved understanding – REP as a ‘laboratory’



TIDE – NO\_TIDE

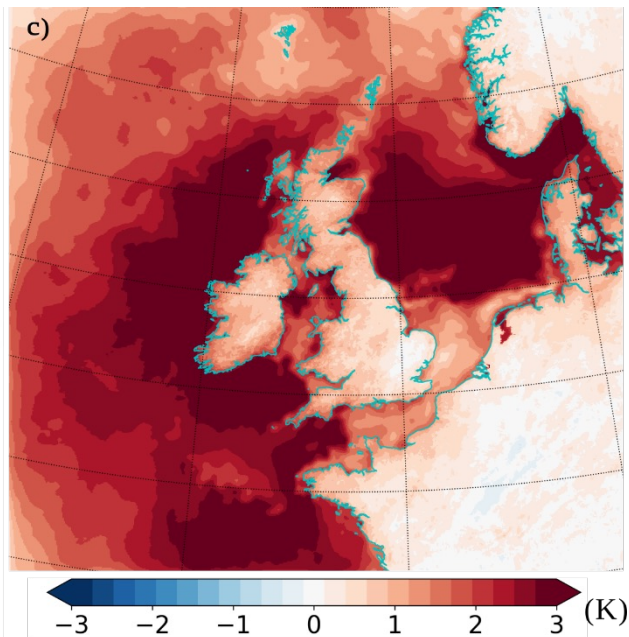


DIFF MEAN Air temperature (K) [Land]

DIFF MEAN Air temperature (K) [Sea]

- Tidal mixing reduces ocean stratification in the North West shelf
- Tides dampened the 2018 heatwave by 0.3K on average, locally 1K

Effect of the marine heatwave on air temperature on June 19<sup>th</sup>-26<sup>th</sup>



Warmest June on record in the UK (+0.9°C), of which 0.6°C came from the marine heatwave feedback on the atmosphere

Warmer temperatures advected by sea breezes (peak in the early afternoon)

Sea breezes increased, despite weaker land/sea contrast and increased rainfall by 20% for a week



## Evaluation of ML skill

- Weather and climate knowledge needs to be employed to define what “good” looks like.
- Realism encompasses full spatial and temporal structure
- Visualisation of extreme cases important
- Benchmark important for assessing ML added value e.g. compared to classical statistical downscaling
- Use available RCM/CPM simulations as “truth” for future climate
- Interpretable AI – right answers for right reasons?

❖ Key area where climate science community can contribute

### Some suggested evaluation metrics

Metric	Variable(s)
Seasonal mean	All
Daily (and potentially sub-daily) distribution	All
Wet day frequency and intensity	Precipitation
Joint distribution of intensity v duration	Precipitation, temperature
Joint distribution of intensity v spatial extent	Precipitation, temperature
Temporal correlation structure	All
Wet/dry or warm/cold spells	Precipitation, temperature
Diurnal cycle	Precipitation, temperature
Extremes on seasonal, daily (and potentially sub-daily) timescales (including 10y+ return period events)	All
Long-term stability	All
Temperature-precipitation scaling	Precipitation (and humidity)
Daily distribution by weather regime	All

# Implications for CPM/RCM ensemble design

- For climate emulation: training and test data for future periods need to come from future climate model simulations
    - Hybrid approach with ML augmenting RCM/CPM simulations
  - ML to provide uncertainty information, i.e. reducing need for large ensembles of dynamical simulations
    - In near future potential to help fill CORDEX GCM/RCM/CPM matrix
  - Using ML can mine a wider range of GCMs for which expensive LBC data are not available
  - Focus of dynamical simulations to provide the best training data for ML emulators
- Implications:
    - Dynamical simulations at the highest possible resolution and/or greatest complexity affordable
    - Work to address existing biases in CPMs key priority for coordinated community activity
    - Prioritise simulations of high-end cases, i.e. dynamically downscaling high climate sensitivity models/scenarios, and also extreme weather events
    - Training data needs to span the range of plausible climate and weather conditions
    - Focus coordination on spatial domains (difficulties in spatial transferability)