



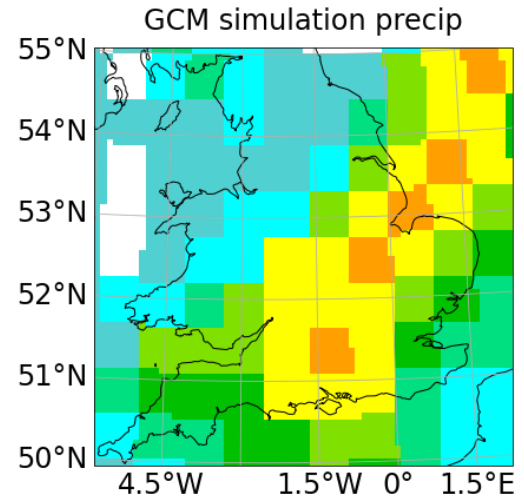
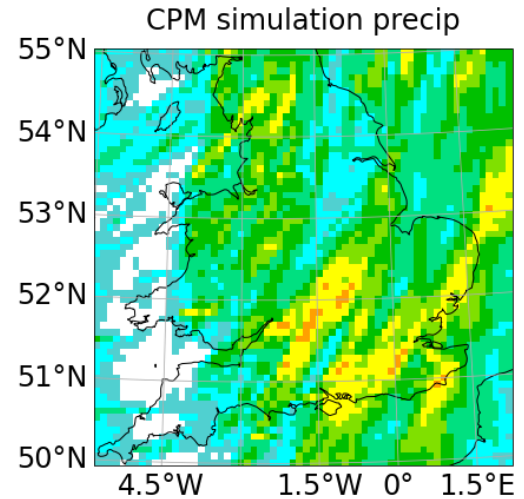
UK Research
and Innovation

Machine learning emulation of a local- scale UK climate model

Henry Addison, Elizabeth Kendon, Suman Ravuri,
Laurence Aitchison, Peter Watson

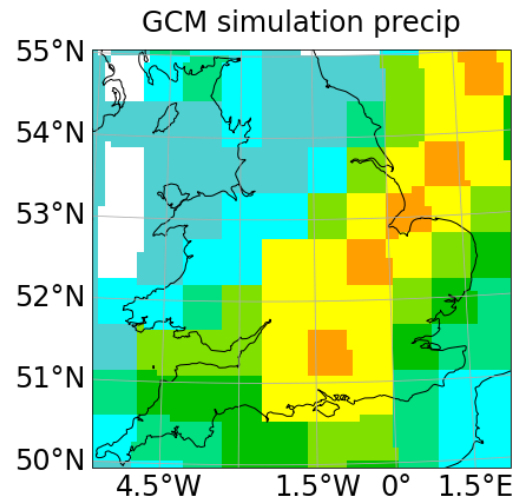
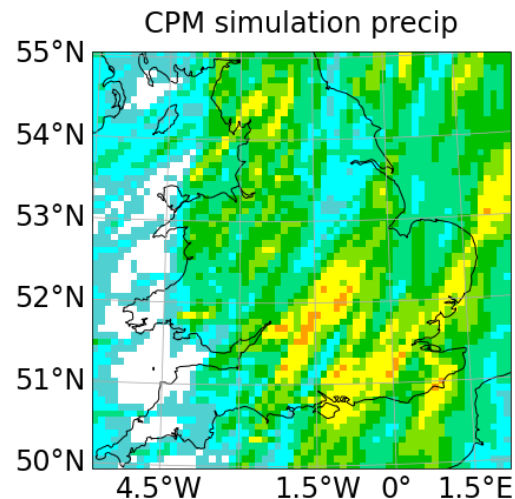
The problem

- **Can we learn hi-res precipitation from coarse, GCM climate variables?**
- High-resolution climate simulations are expensive
- New probabilistic ML methods could emulate hi-res, CPM simulator



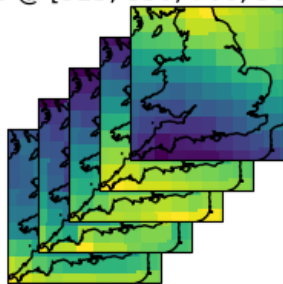
Data: Met Office UKCP18

- **Low resolution**
Global Climate Model (**GCM**) @ 60km
- **High resolution**
Convection-Permitting Model (**CPM**) @ 2.2km
(using 8.8km)
- Daily

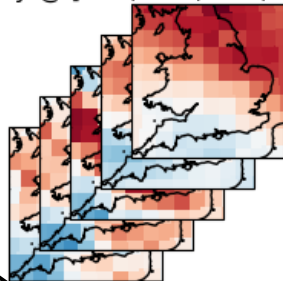


Input
Lo-res climate variables

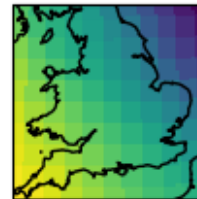
temp @ [925, 850, 700, 500, 250]



vorticity @ [925, 850, 700, 500, 250]

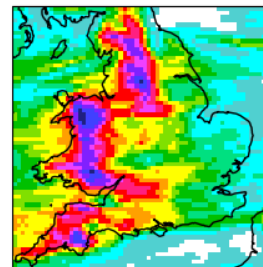
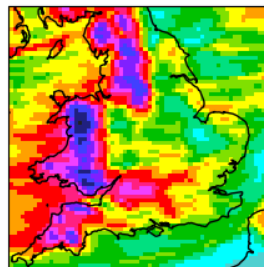
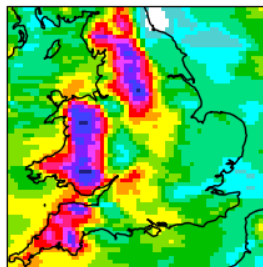


psl



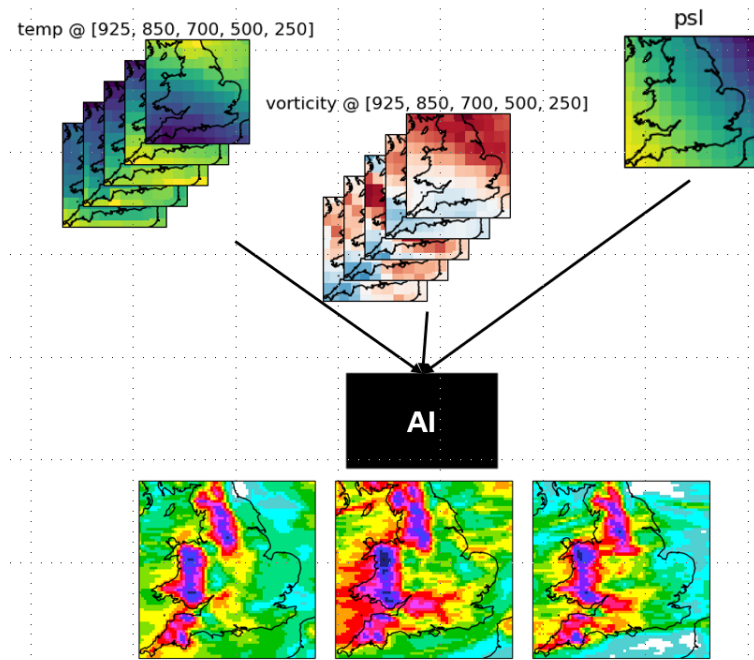
AI

Output
Hi-res precipitation



Approach

- Training: coarsened CPM variables → hi-res CPM precip
- Evaluating: coarsened CPM variables or GCM variables → hi-res CPM-like precip
- **Make sure to use variables which are well-represented in GCMs and cause rainfall**
 - Wind
 - Temperature
 - Pressure



Diffusion models



Teddy bears swimming at the Olympics 400m Butterfly event.



A cute corgi lives in a house made out of sushi.

From Figure 1 of Saharia, Chitwan, et al. "Photorealistic text-to-image diffusion models with deep language understanding." *Advances in Neural Information Processing Systems* 35 (2022)

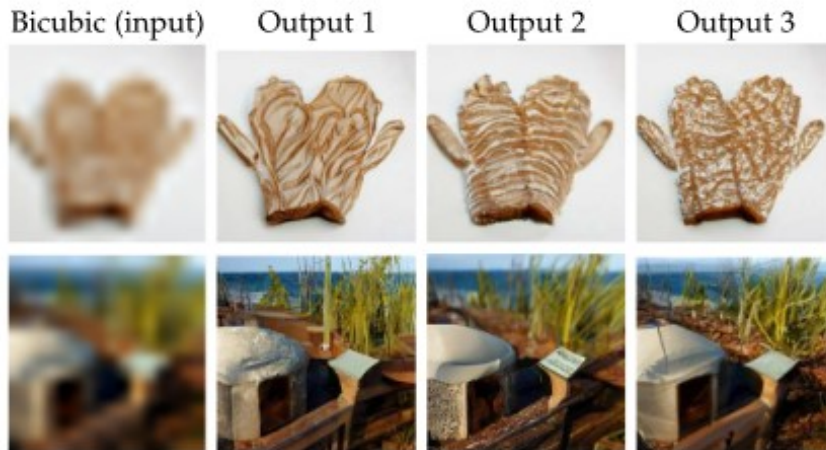
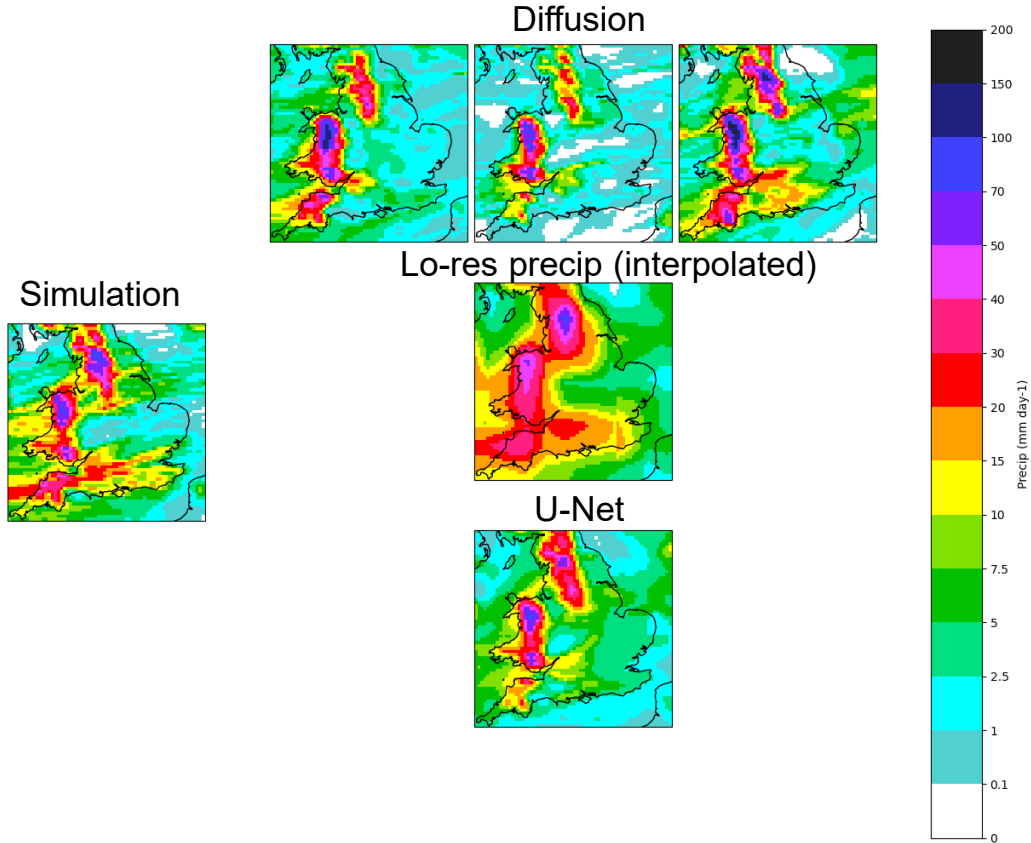


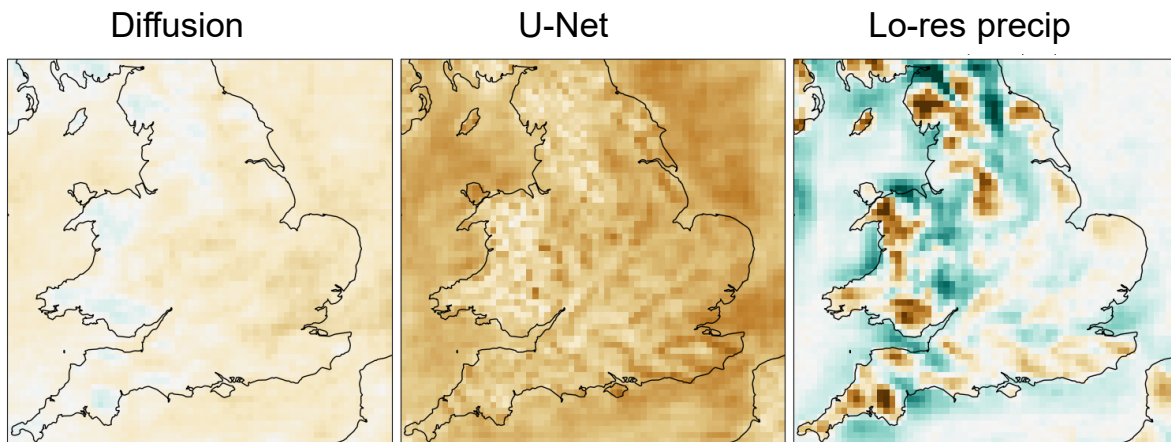
Fig. 5. Three samples from SR3 applied to ImageNet test images ($16 \times 16 \rightarrow 256 \times 256$), demonstrating SR3 diversity.

Saharia, Chitwan, et al. "Image super-resolution via iterative refinement." *IEEE Transactions on Pattern Analysis and Machine Intelligence* (2022).

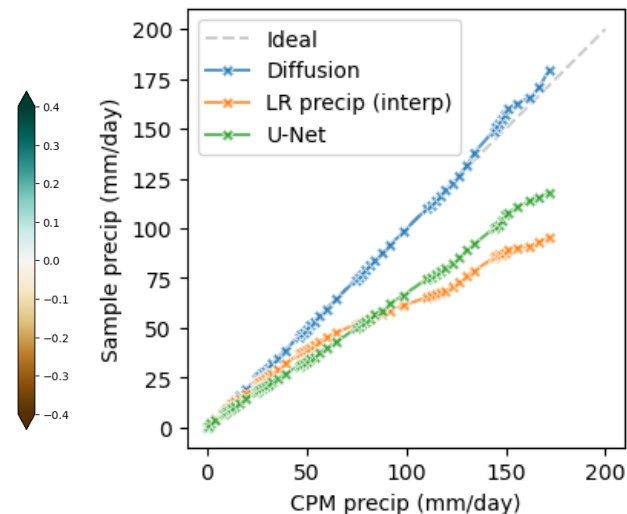
Coarsened CPM \rightarrow 8.8km CPM rainfall



Are we capturing the distribution of rainfall?



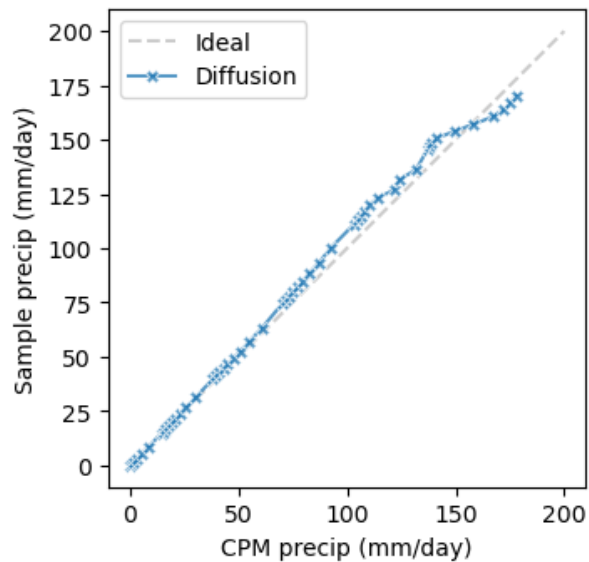
Normalized mean bias



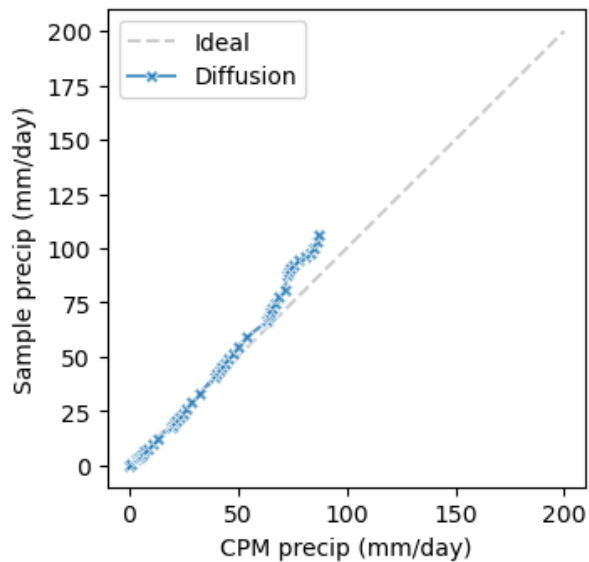
QQ plot

10th to 99.99999th centile

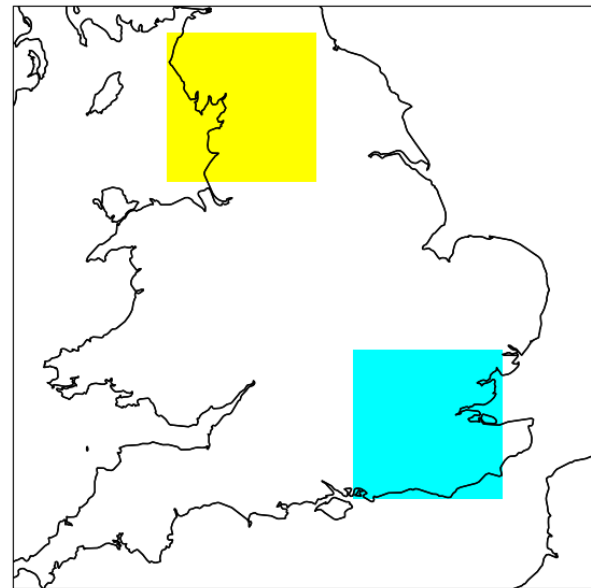
Subregions



NW Winter

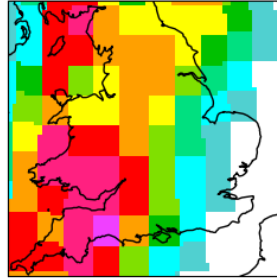


SE Summer

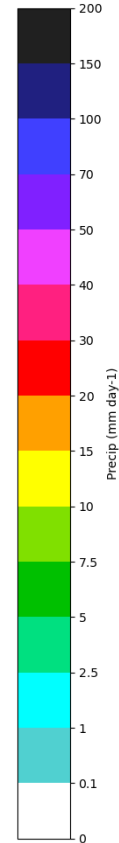
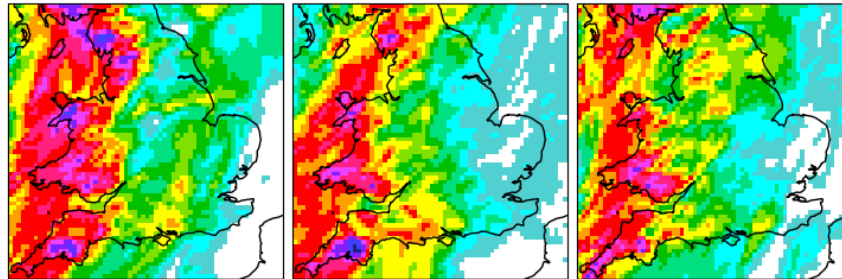


GCM → 8.8km CPM rainfall

Simulation

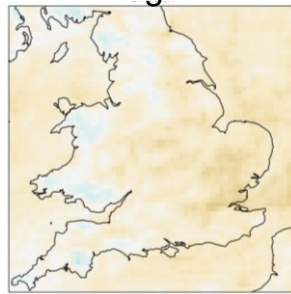


Diffusion



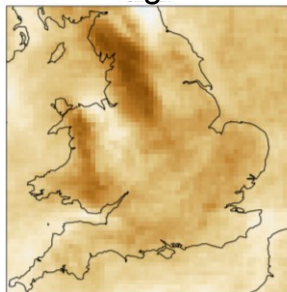
GCM → 8.8km CPM rainfall

Original

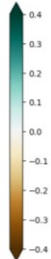
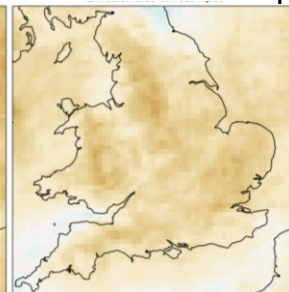


CPM driven

Original

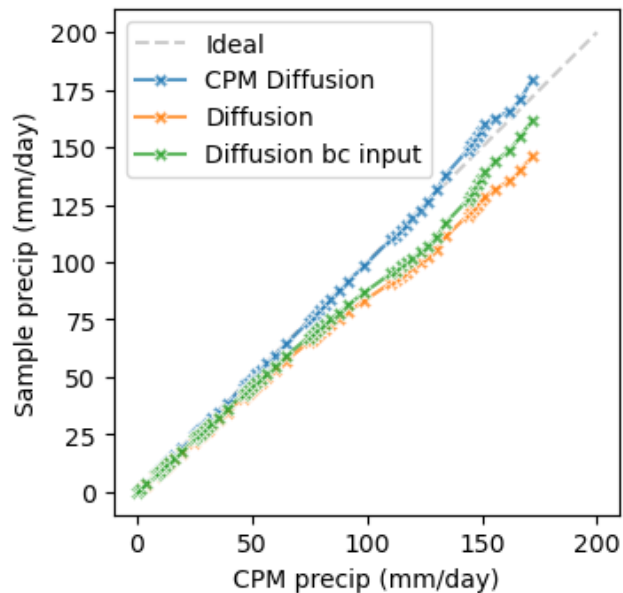


Bias corrected inputs



GCM driven

Normalized mean bias



QQ plot

10th to 99.9999th centile

Future work

- Evaluation
 - Spatial structure
 - Conditioning
 - Probabilistic element
- Generalizing to (large ensembles of) other climate models, time periods, locations
- More extreme Extremes: 1-in-100 years
- Sub-daily frequency and temporal sequences (video)
- Flood modelling applications

Summary

Using SOTA diffusion model to emulate Met Office's hi-res UK climate model

- Reproduces realistic spatial structure and variability of rainfall
- Coarsened CPM driven: good match in mean and quantiles
- GCM driven: extra bias correction of inputs required

Any questions or suggestions? `henry.addison@bristol.ac.uk`

Key references

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