

Correcting biases in climate simulations using unsupervised image-to-image-translation networks

James Fulton, Ben Clarke
Gabriele Hegerl, Friederike Otto

[Abstract](#)

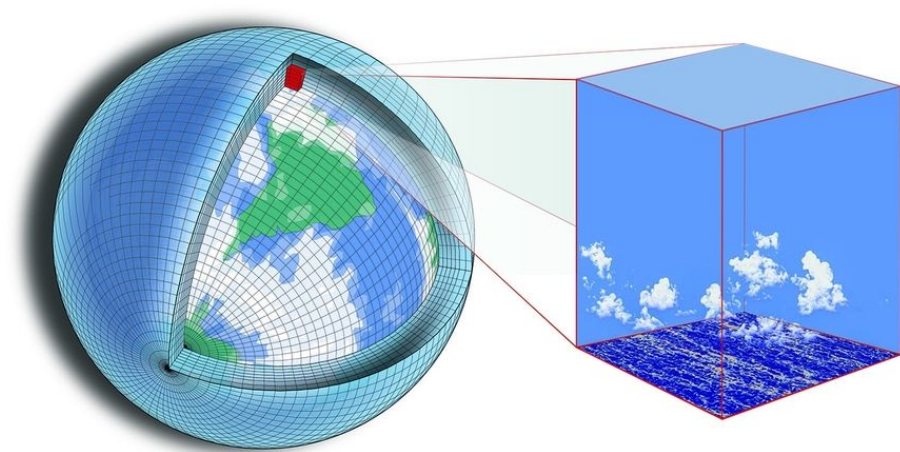


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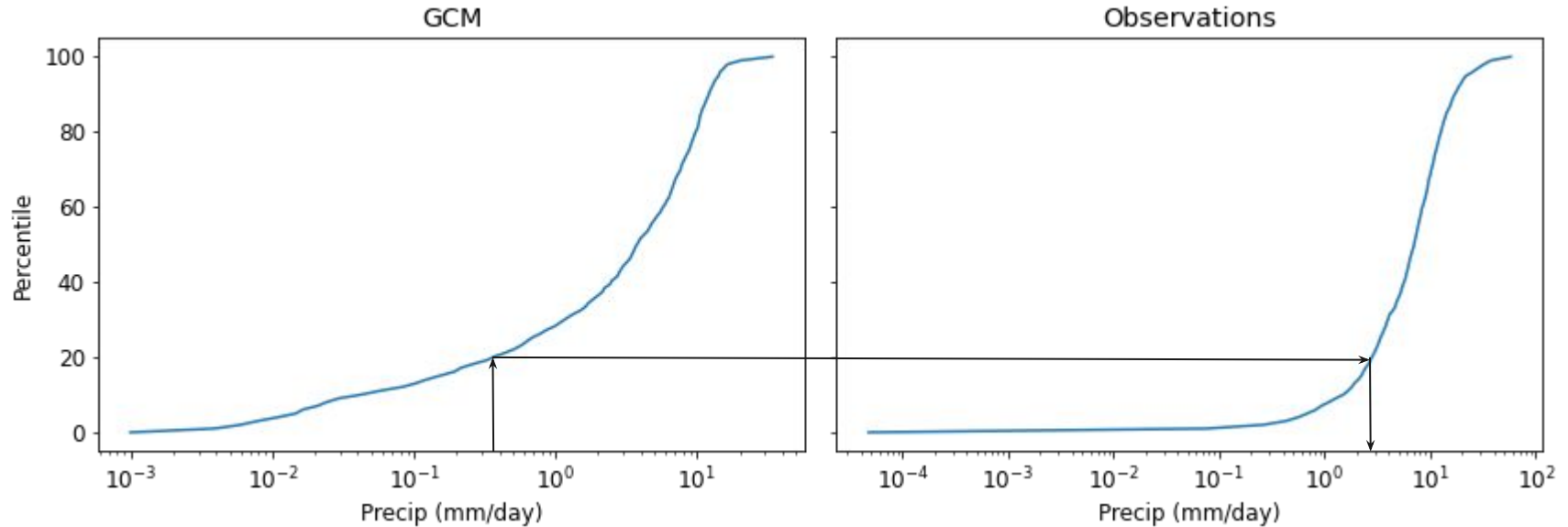
Motivation - imperfect global circulation models

- GCMs aren't perfect representations of the earth system
- Current outputs are important for policy relevant research
- **We need to make best use of the simulation results via post-processing**



<https://news.mit.edu/2018/new-climate-modeling-alliance-clima-1212>

The status-quo: Quantile mapping



Unsupervised image to image translation

Deep learning networks learn mapping between image domains without supervision

E.g. Deep learning architectures UNIT [1], CycleGAN [2], AlignFlow [3], etc.

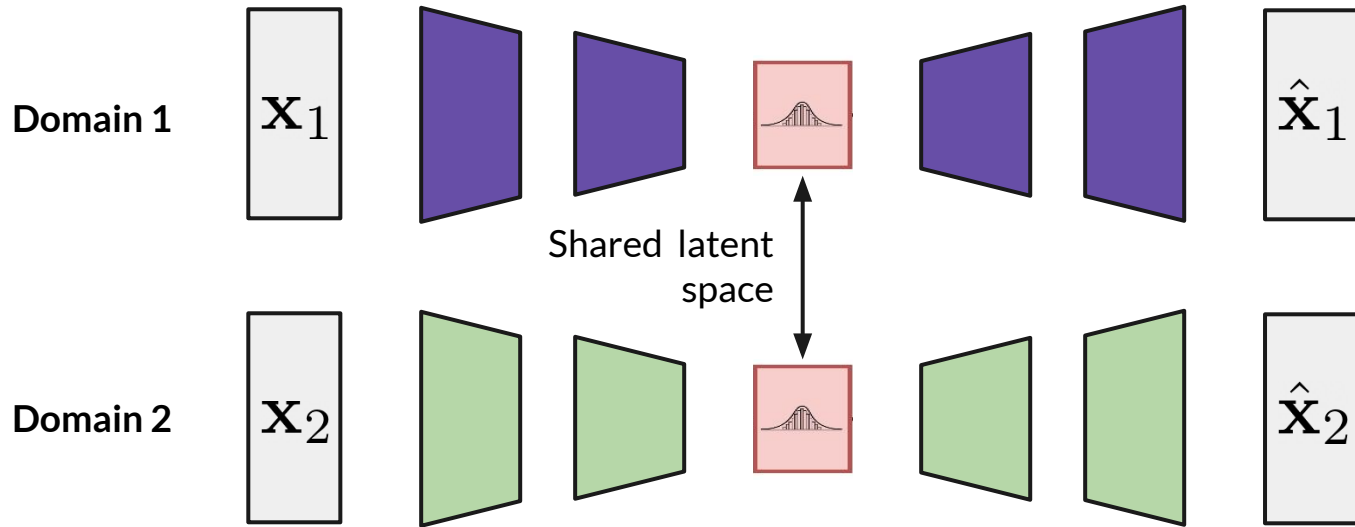


Example image translation from [2]

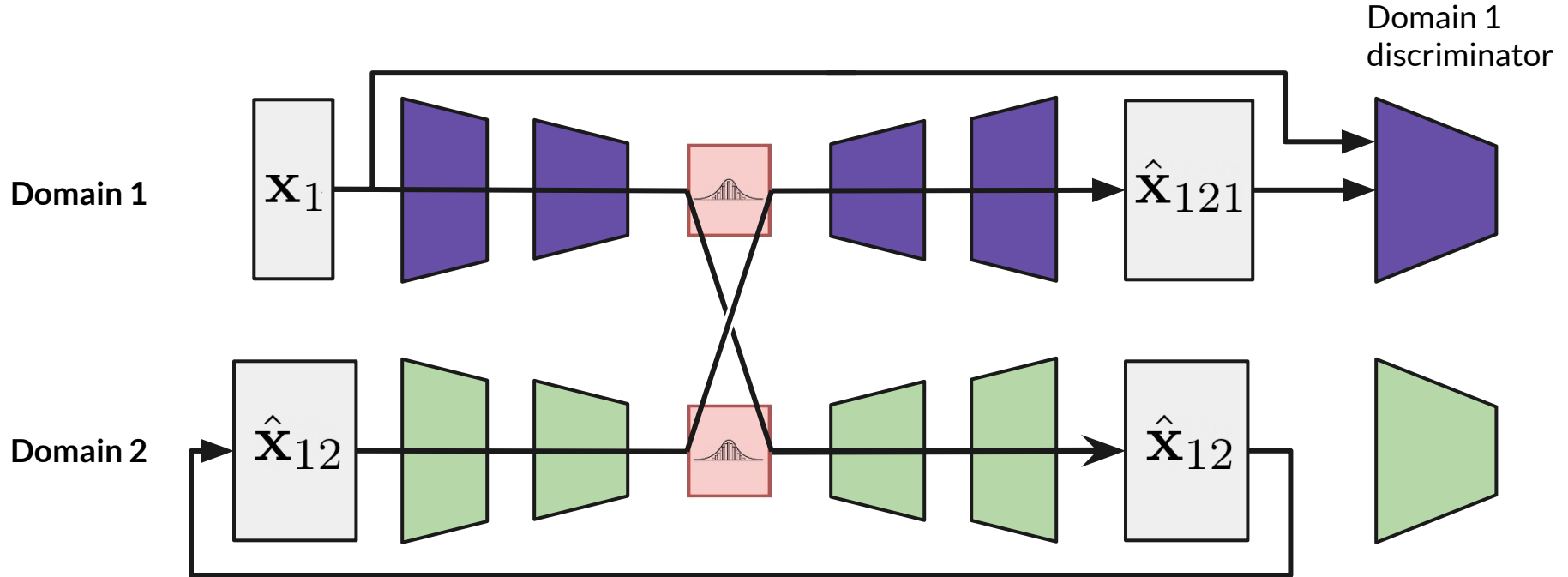
GTA → Cityscapes

- This ‘bias correction’ is performed without corresponding input-output pairs
- Long GCM simulations can’t be used to create corresponding pairs to the observations

UNIT [1] - UNsupervised Image to image Translation network

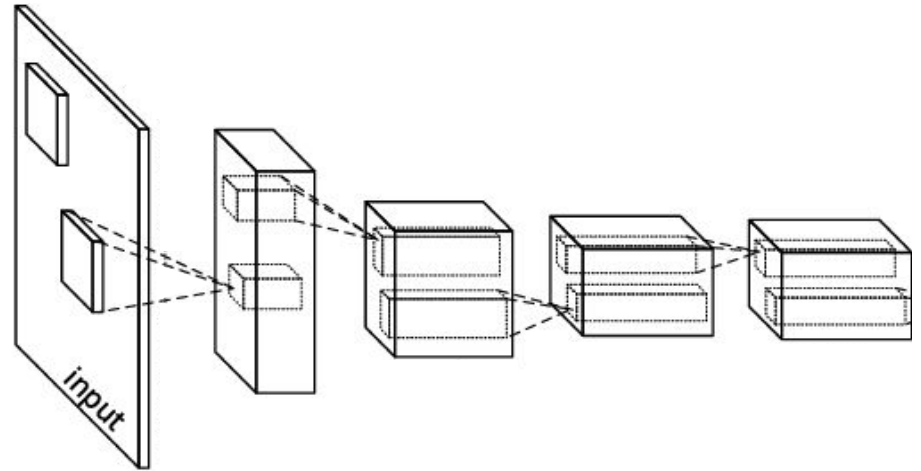


UNIT [1] - UNsupervised Image to image Translation network



UNIT advantages

- Convolutional layers means translation simultaneously across all variables and large scales
- `Sees' cross-variable relations
- `Sees' spatial relations
- Helpful for compound extreme events



Data sources and extent

Physical variables

- Daily min, mean, and max temperature
- Daily total precipitation
- Z500 geopotential height

Data sources

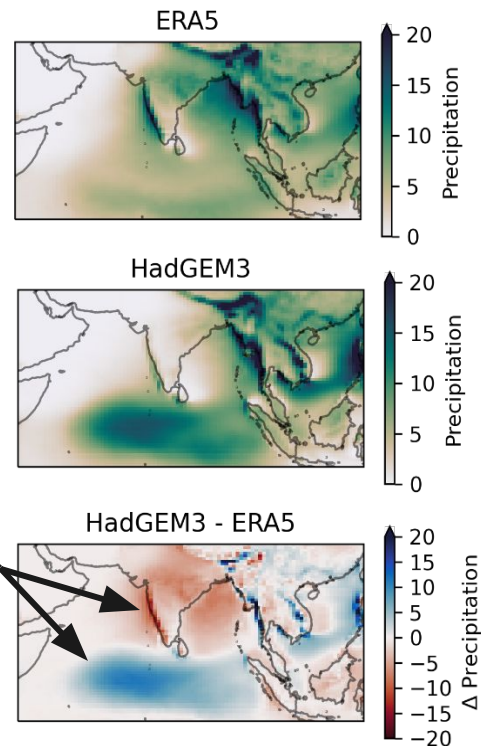
- HadGEM3
 - C20C+ archive
 - Historical recreation scenario
- ERA5

Spatial extent

- South Asian monsoon region
 - 8°S - 30°N, 44°E - 121°E
- Has large known biases in precipitation

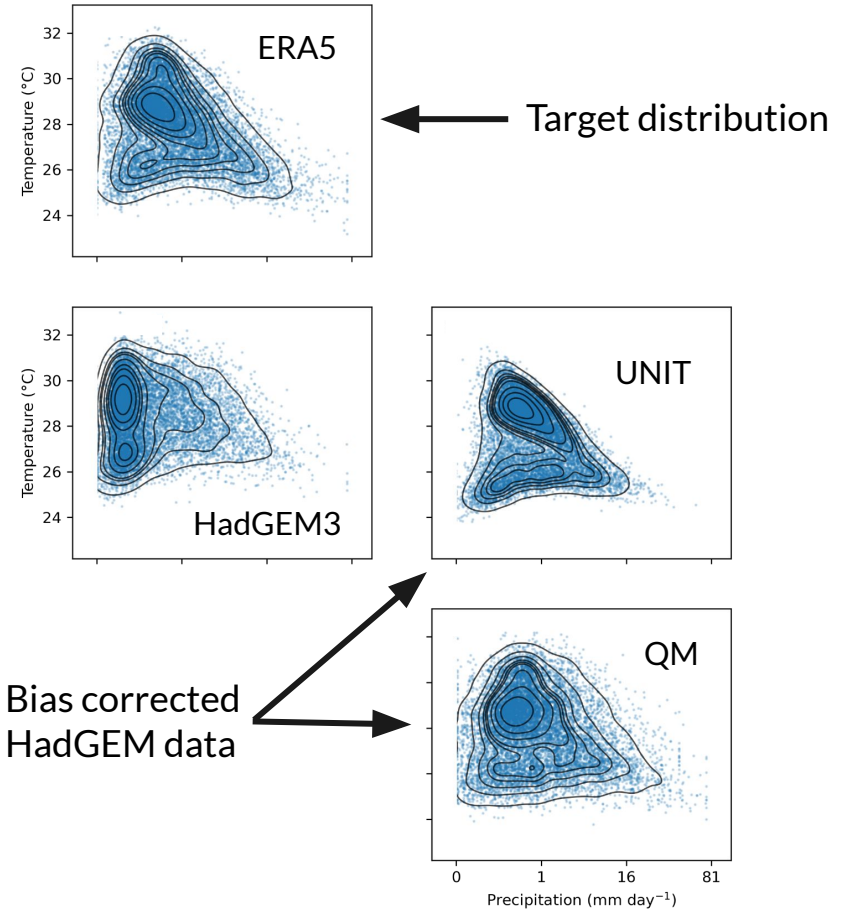
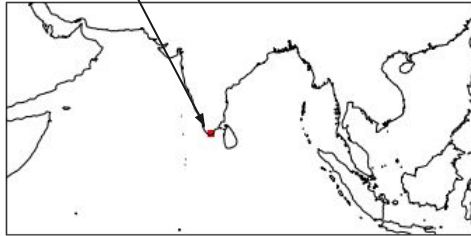
Time period

- Jan 1979 - Dec 2013
- Approx. 12 000 days



Cross-variable biases

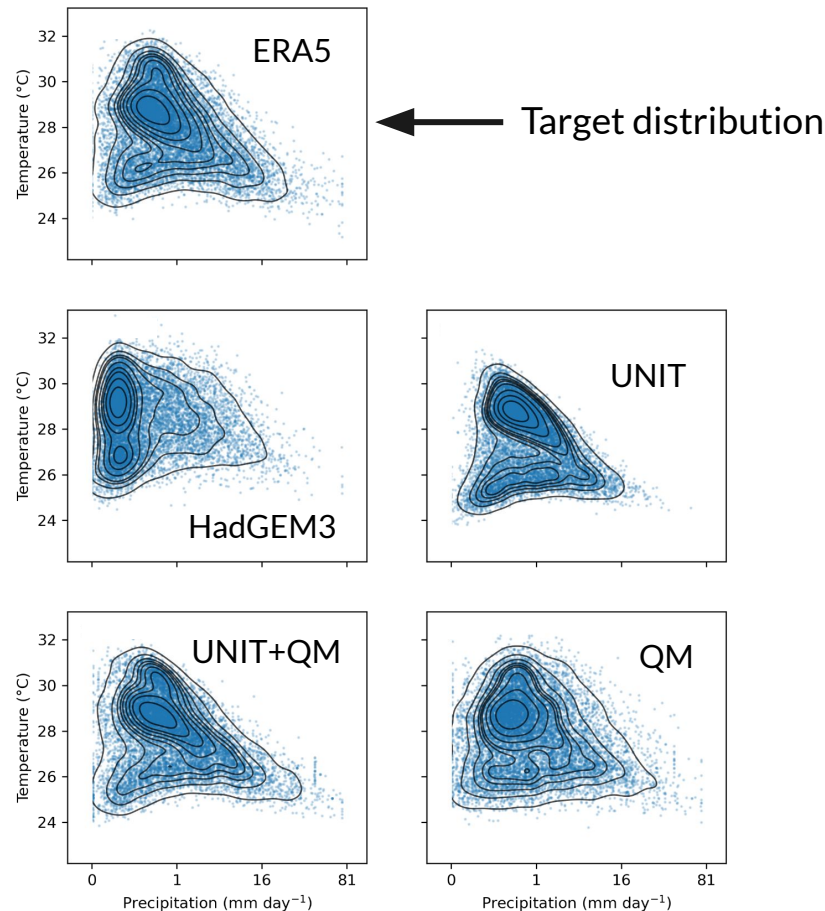
- Joint distribution sampled from single spatial location



UNIT+QM

- Apply UNIT transform
- Then apply quantile mapping

Best match to target →



More results

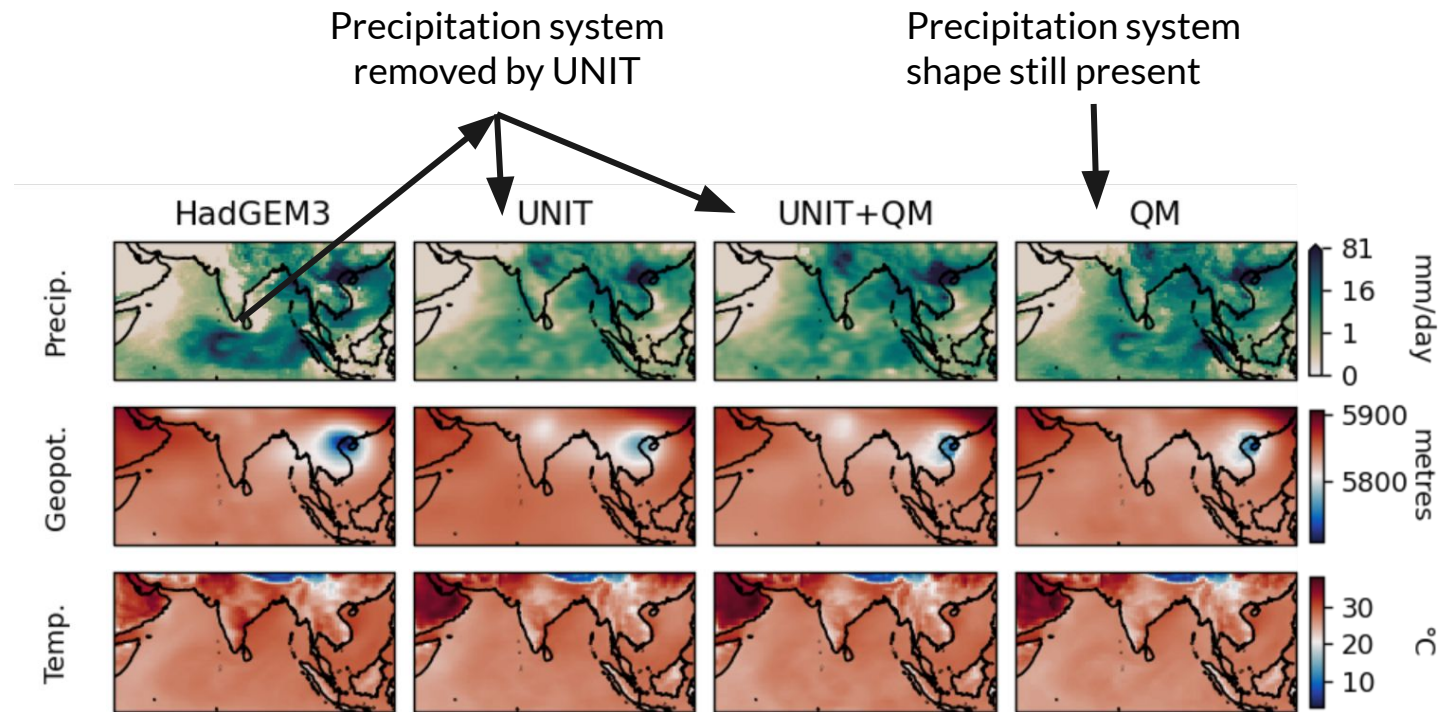
UNIT+QM:

- Maps 1D distributions like QM
- Maps cross-variable distributions better than QM
- Maps spatial correlations better than QM

Submitted article:

Bias correcting climate model simulations using unpaired image-to-image translation networks submitted

Artificial Intelligence for the Earth Systems (AIES) - American Meteorological Society



References

1. Liu, M.Y., Breuel, T. and Kautz, J., 2017. *Unsupervised image-to-image translation networks*.
2. Zhu, J.Y., Park, T., Isola, P. and Efros, A.A., 2017. *Unpaired image-to-image translation using cycle-consistent adversarial networks*.
3. Grover, A., Chute, C., Shu, R., Cao, Z. and Ermon, S., 2020, April. *Alignflow: Cycle consistent learning from multiple domains via normalizing flows*.

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