

Bayesian methods for high resolution downscaling of regional climate model data

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Outline

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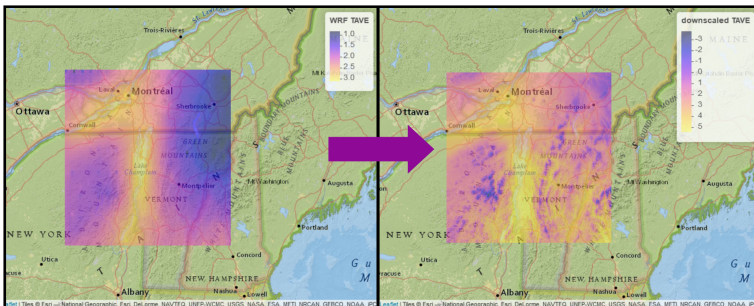
Introduction

The need for high resolution, accurate climate products

- Resolution of global climate models (GCMs) and even regional climate models (RCMs) are far too coarse to model local climate
- Climate model projections are biased, which is especially problematic for predicting future extreme weather events [1], [2]
- **Fine scale, accurate climate projections are critical for local climate impact assessments**

Research contribution

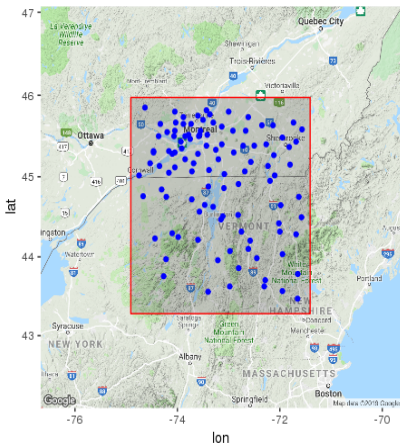
- 1 High-resolution (1km) downscaling and bias correction of WRF daily average temperature (TAVE) and precipitation (PRECIP) output using Bayesian spatial hierarchical modeling
 - Bias correction based on historical, observed weather station data
- 2 Refining WRF output to better capture extreme climate events



Downscaling adds value to climate model projections

Station Data

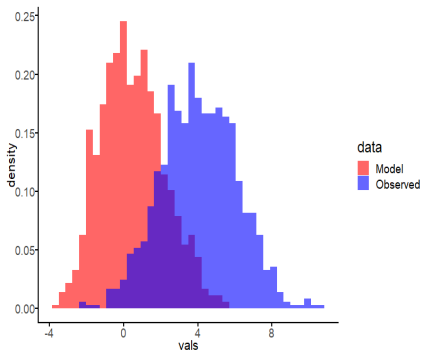
Source: Global Historical Climate network



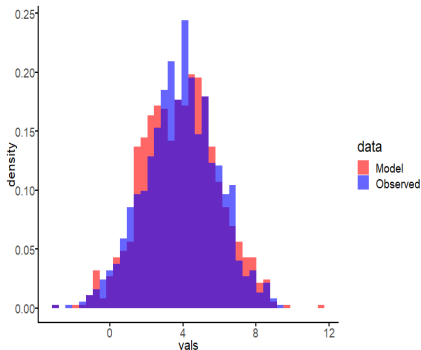
GHCN weather stations in the study domain

Bias correction

- **Goal:** Adjust model distribution to match that of historical data
- Quantify model and station distribution overlap with **Perkins skill score** [3]
- Method used for this study: quantile mapping

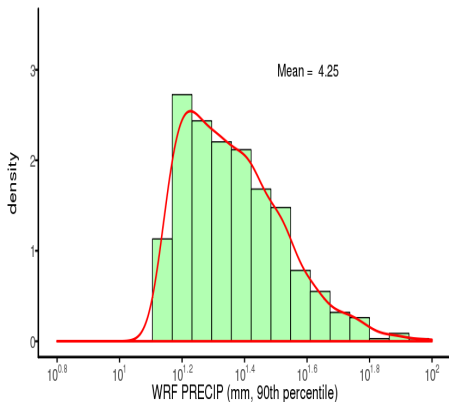


Before bias correction (skill score = 0.34)



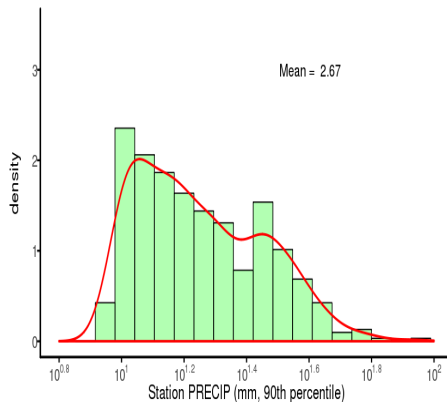
After bias correction (skill score = 0.97)

WRF vs stations: extreme PRECIP



WRF PRECIP, 90th percentile, 1980-1994
(1 station)

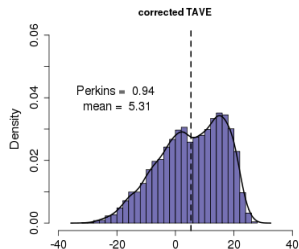
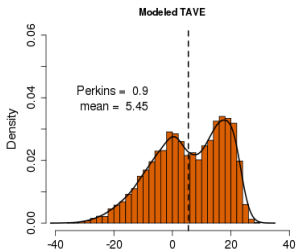
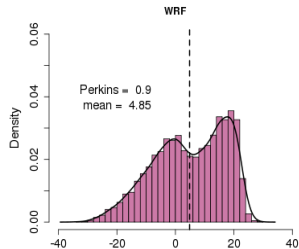
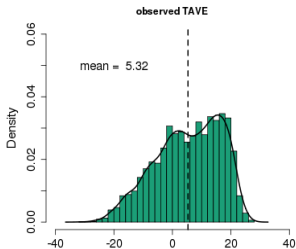
Skill score = 0.56



Station PRECIP, 90th
percentile, 1980-1994 (1 station)



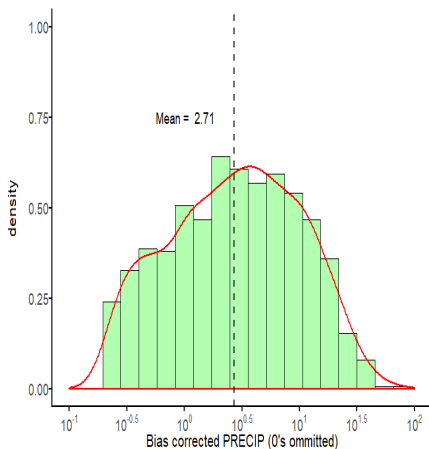
Preliminary results: TAVE



1980-1994 annual TAVE for 1 station: (turquoise), nearest WRF gridpoint (magenta),
predictions from model (orange), bias-corrected predictions (purple)

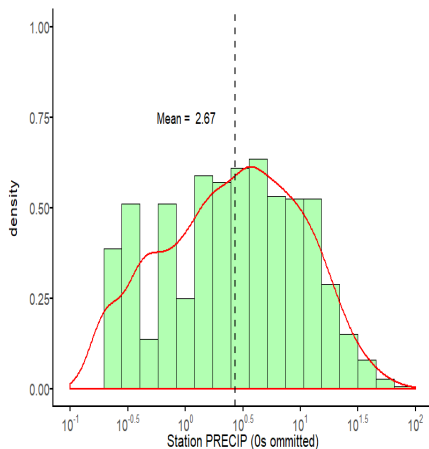


Bias correcting WRF PRECIP



Bias corrected daily PRECIP, 1980-89
(log scale, zeros omitted)

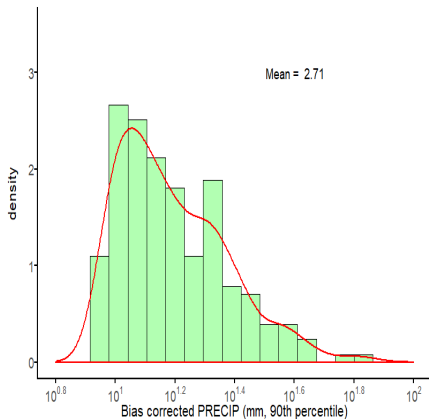
Skill score = 0.96



Station daily PRECIP,
1980-89 (log scale, zeros omitted)

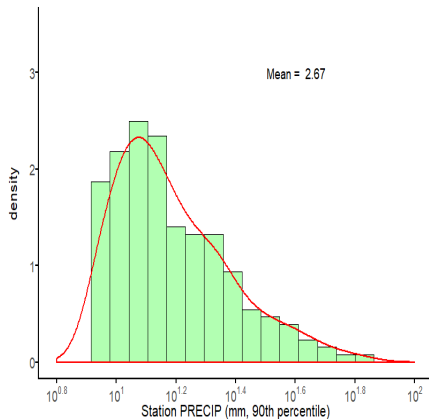


Bias correcting WRF extreme PRECIP



Bias corrected WRF PRECIP,
90th percentile, 1980-89 (log scale)

Skill score = 0.66



Station PRECIP, 90th
percentile, 1980-89 (log scale)



Conclusion

Conclusion

- Raw WRF data exhibits some bias for both temperature and precipitation
- WRF performs better in capturing temperature than precipitation, especially extreme precipitation
- Bayesian model works adequately for temperature but may need some adjustment for modeling precipitation
- Quantile mapping for bias correction corrects mean and increases skill score

Future work

Future work

- Compare results among other methods of bias correction (e.g. distribution mapping)
- Adjust Bayesian spatial model to better capture precipitation
- Refine WRF output to better capture extreme events using extreme value theory



Thank you!



Cerulean warbler



References I

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