

# **4th VALUE Training School**

## **Validation of Regional Climate Change Projections**

### **Model Output Statistics and Bias Correction**

Douglas Maraun

Wegener Center for Climate and Global Change

Rationale

MOS in Climate Science

Delta Change Approaches

Distributionwise MOS

Pairwise MOS

# Rationale

MOS in Climate Science

Delta Change Approaches

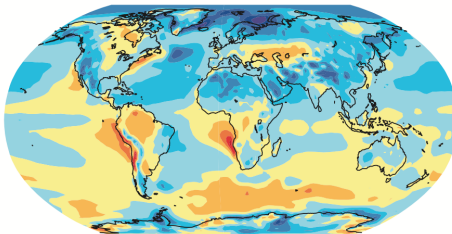
Distributionwise MOS

Pairwise MOS

# Temperature and precipitation biases

CMIP5, multi-model mean

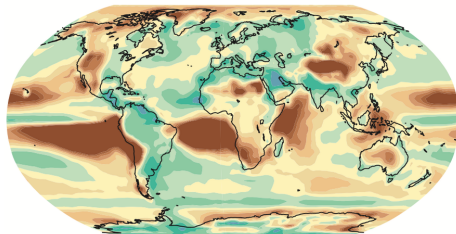
(b) Multi Model Mean Bias



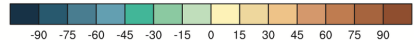
(°C)



(d) Multi Model Mean of Relative Error



(%)

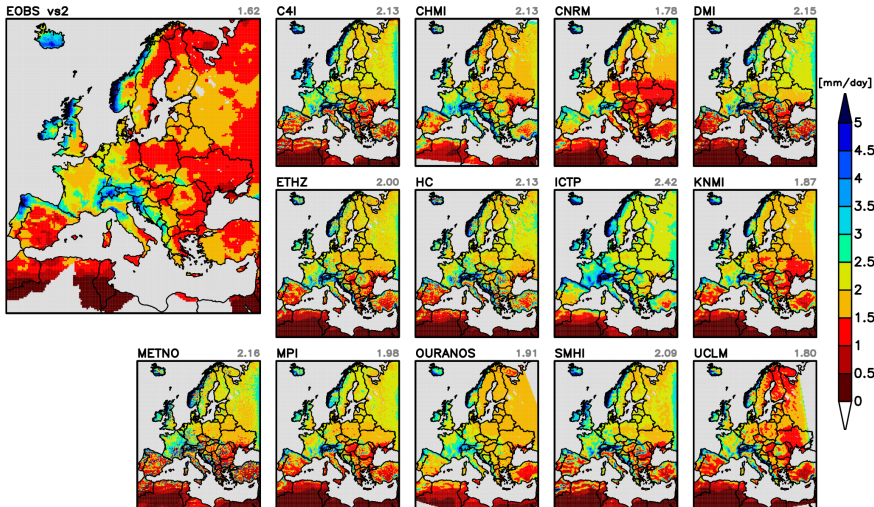


Impact modellers often desire to “correct” these biases.

*Flato et al., IPCC AR5, 2013*

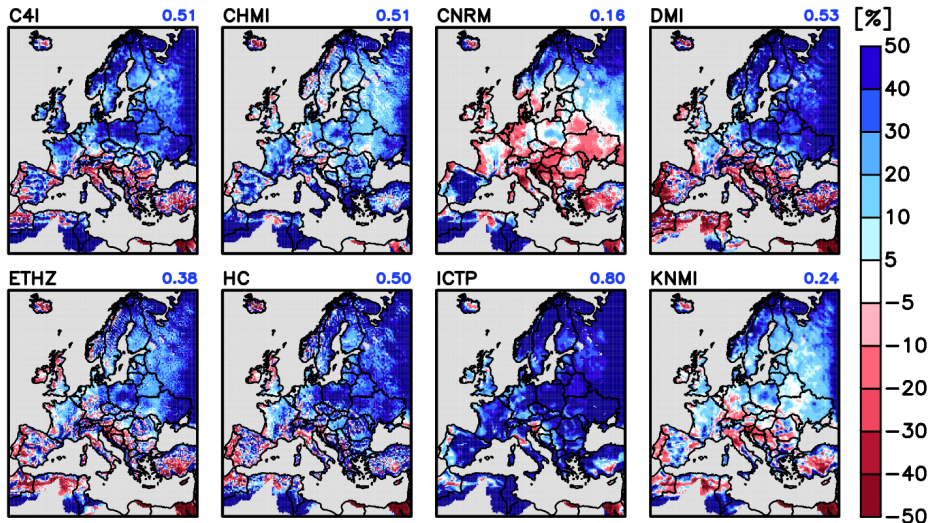
# Also RCMs are biased

Mean annual precipitation (1961–2000) [mm/day]



*Kotlarski, unpublished*

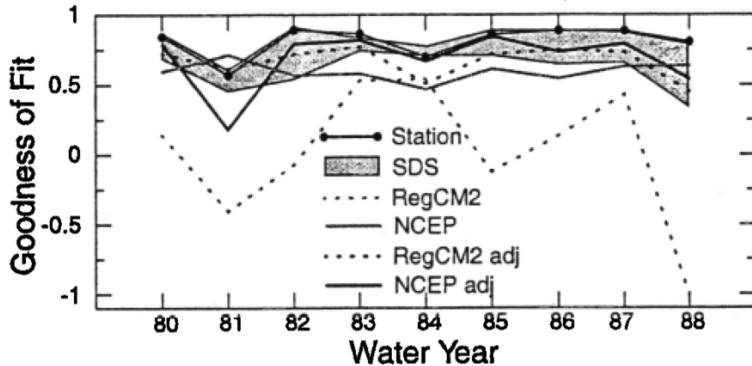
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*Kotlarski, unpublished*

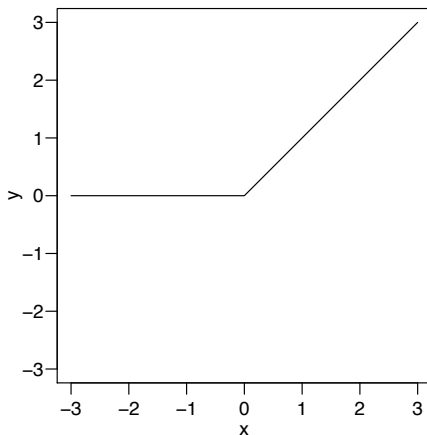
# Example: Driving Hydrological Model

Goodness of fit of hydrological model simulation driven with different input data



Wilby et al., *Geophys. Res. Lett.*, 2000

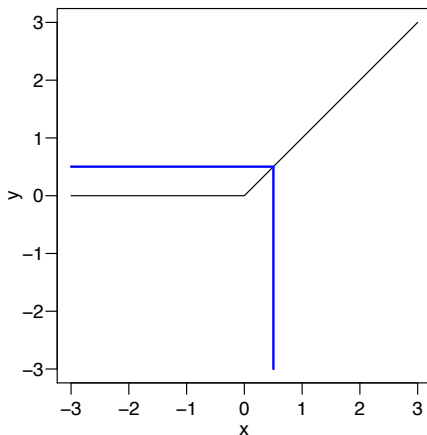
# MMEs do not Solve the Bias Problem



black: transfer function; blue: true value; red shading: model ensemble; red line: ensemble mean.

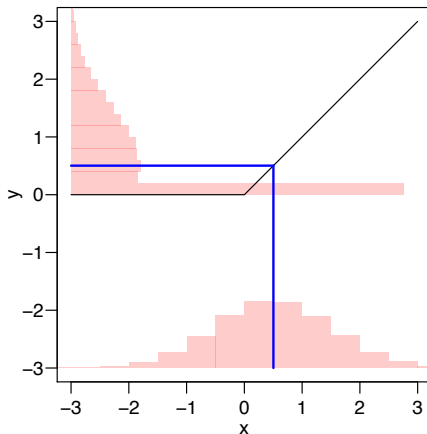


# MMEs do not Solve the Bias Problem



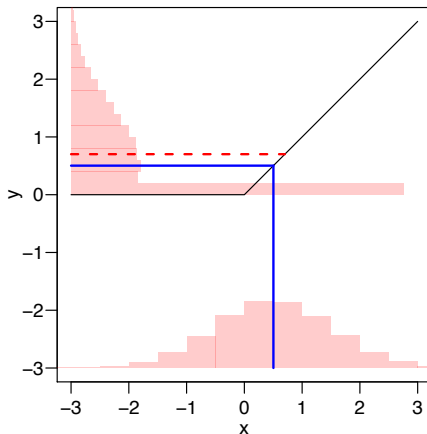
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# NWP vs. Climate Projections?

What is different between the two?

# NWP vs. Climate Projections?

What is different between the two?

- ▶ time scales of calibration and prediction/projection
- ▶ extrapolation requirements
- ▶ validation data availability
- ▶ **synchronicity**

# Pairwise vs. distributionwise calibration

## **Pairwise (NWP)**

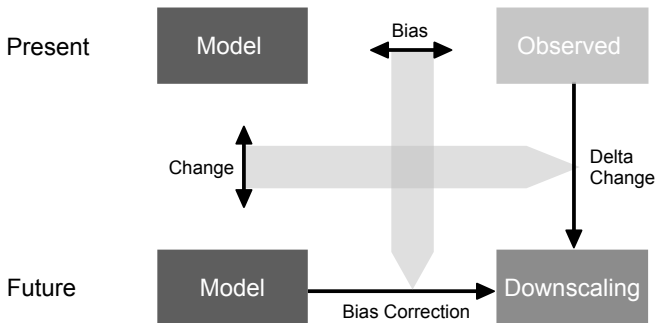
- ▶ simulated and observed series in sync.
- ▶ regression model possible
- ▶ skill evaluation possible

## **Distributionwise (typically climate)**

- ▶ independent climate simulation
- ▶ no simple regression (with time as index) possible
- ▶ skill evaluation difficult
- ▶ typically homogeneous!

# Delta change vs. bias correction

Do we take series/parameters from obs or model?



*Maraun et al., Rev. Geophys., 2010*



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# “Delta” Method

shifting/rescaling observed rainfall by change “factors”

Temperature:

$$y_{i+T}^f = x_{\text{obs},i}^p + \bar{y}_{\text{mod}}^f - \bar{x}_{\text{mod}}^p$$

Precipitation:

$$y_{i+T}^f = x_{\text{obs},i}^p \frac{\bar{y}_{\text{mod}}^f}{\bar{x}_{\text{mod}}^p}$$

Note: this method

- ▶ does not use the RCM time series, but only the climate change signal;
- ▶ does not account for dynamical changes;

# Change Factor Weather Generators (WGs)

- ▶ Calibrate WG in present day climate against observations;
- ▶ derive changes of WG parameters from RCM;
- ▶ run the WG with changed parameters.

Note: as the delta change method, these WGs do not use the RCM time series, but only the climate change signal

*e.g., Kilsby et al., Env. Mod. Soft., 2007*

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# Direct Method

correcting regional climate model output by a correction term or scaling factor

Temperature:

$$y_i^f = x_{\text{mod},i}^f + \bar{y}_{\text{obs}}^p - \bar{x}_{\text{mod}}^p$$

Precipitation:

$$y_i^f = x_{\text{mod},i}^f \frac{\bar{y}_{\text{obs}}^p}{\bar{x}_{\text{mod}}^p}$$

$y_i^f$ : corrected simulated future series

$x_{\text{mod},i}^f$ : uncorrected simulated future series

$\bar{y}_{\text{obs}}^p$ : observed present mean

$\bar{x}_{\text{mod}}^p$ : simulated present mean

# Correction of mean and variance

$$y_i^f = \frac{x_{\text{mod},i}^f - \bar{x}_{\text{mod}}^p}{sd_{\text{mod}}^p / sd_{\text{obs}}^p} + \bar{x}_{\text{obs}}^p$$

$sd_{\text{mod}}^p$  present simulated variance

$sd_{\text{obs}}^p$  present observed variance

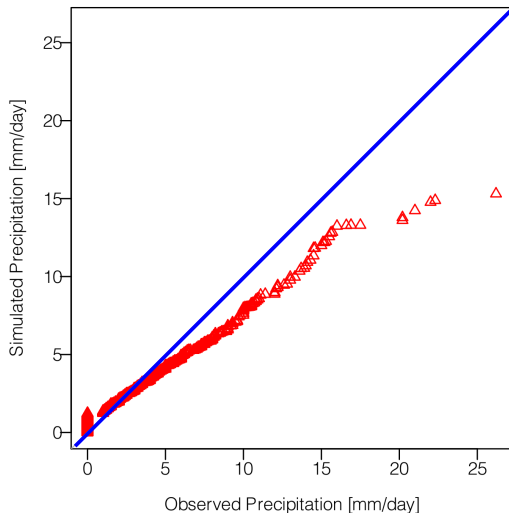
# Quantile (-Quantile) mapping

quantile dependent mapping

$$PDF(y_i^f) = T_{cal}(PDF(x_{mod,i}^f))$$

$x$  modelled grid box rainfall;  $y$  observed (grid box) rainfall

# Quantile mapping

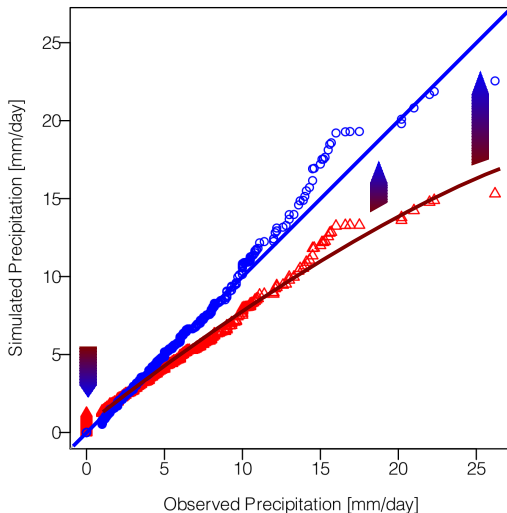


- ▶ adjust intensities;
- ▶ adjust number of wetdays.

*e.g., Piani et al., T.A.C., 2010; Maraun, J. Climate, 2013*



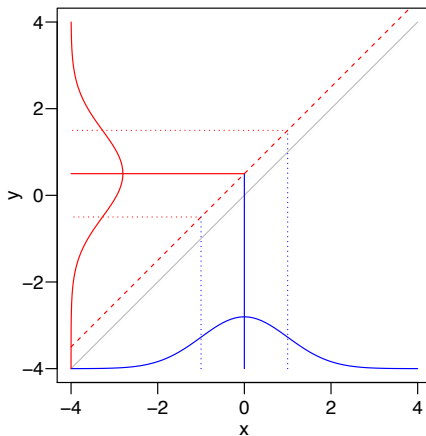
# Quantile mapping



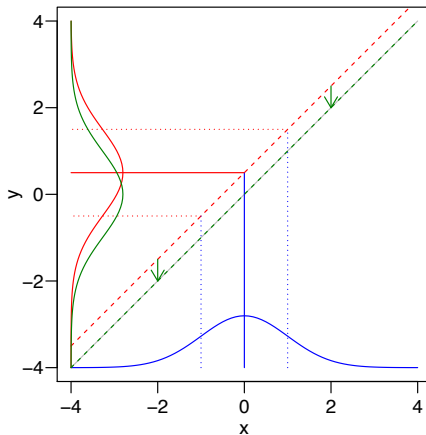
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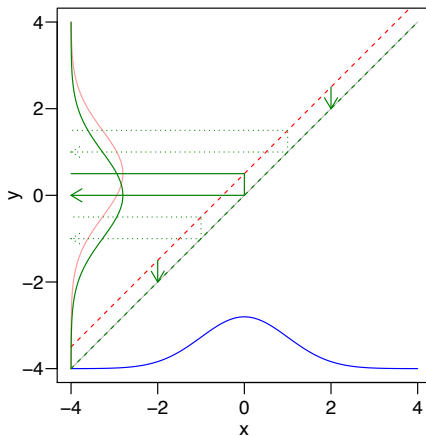
# Correction of mean



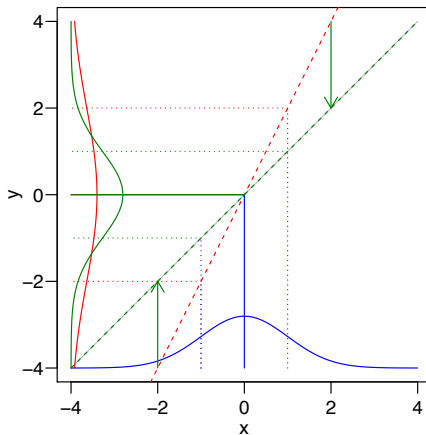
# Correction of mean



# Correction of mean

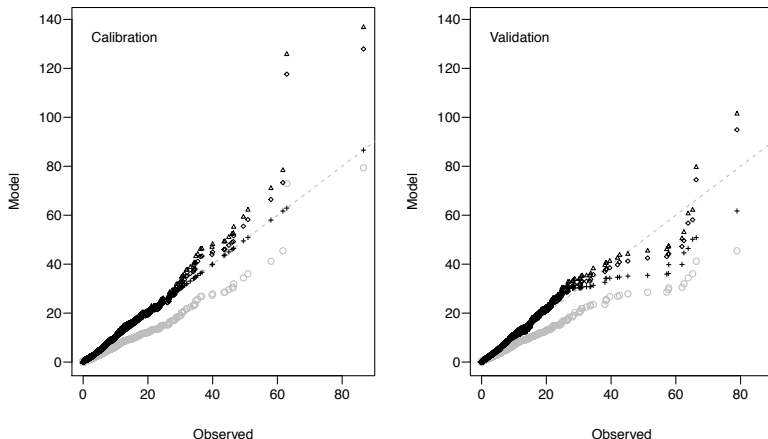


# Correction of variance



# Comparison of methods

Example 1: daily summer precipitation at Clausthal-Zellerfeld-Erbprinzentanne (Id: 862)



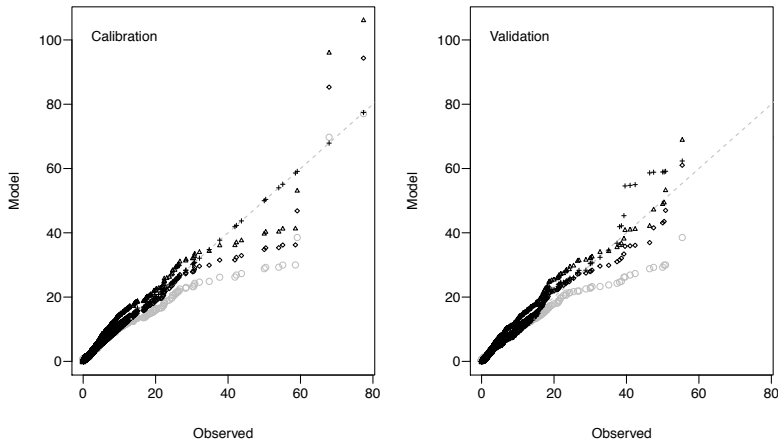
Top: Clausthal-Zellerfeld-Erbprinzentanne (Id: 862), bottom: Eckertal (Id: 1117)

Left: Calibration period, right: validation period

Grey circles: uncorrected RCM, triangles: scaling, diamonds: parametric quantile mapping, crosses: non-parametric quantile mapping.

# Comparison of methods

## Example 2: daily summer precipitation at Eckertal (Id: 1117)



Left: Calibration period, right: validation period

Grey circles: uncorrected RCM, triangles: scaling, diamonds: parametric quantile mapping, crosses: non-parametric quantile mapping.

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## Pairwise MOS · Multipredictor methods

# Pairwise MOS

## MOS as in Forecasting:

- ▶ re-analysis driven RCM, potentially spectrally nudged;
- ▶ synchronicity;
- ▶ predictors other than the variable itself are possible.

But: can only correct RCM biases!

(Discuss: how could one define pairs other than by temporal indexing?)

*Themessl et al., Int. J. Climatol., 2010*

## Pairwise MOS · Stochastic methods

# Stochastic model output statistics

## Idea: regression model in pairwise MOS

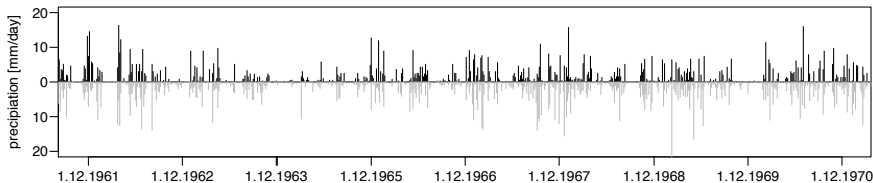
observed precipitation = corrected RCM precipitation + noise

Requirement for any regression model:

- ▶ a day to day correspondence between predictor and predictand.

# Perfect boundary setting

Cambridge, CEE England (Station Id. 454)



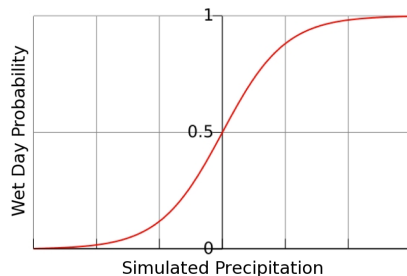
Synchronicity ensured by ERA40 boundaries + nudging

# Wet day probabilities

## Logistic Regression

$$p_i = \frac{\exp(a + bx_i)}{1 + \exp(a + bx_i)}$$

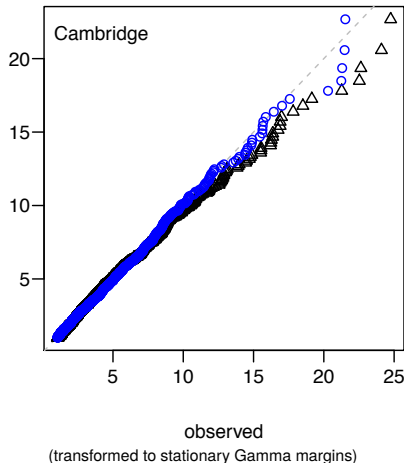
$p_i$ : wet day probability  
 $x_i$ : simulated precipitation  
 $a, b$ : parameters



Wong et al., J Climate, 2014.

# Biases are effectively corrected (DJF)

Cambridge, CEE England (Station Id. 454)

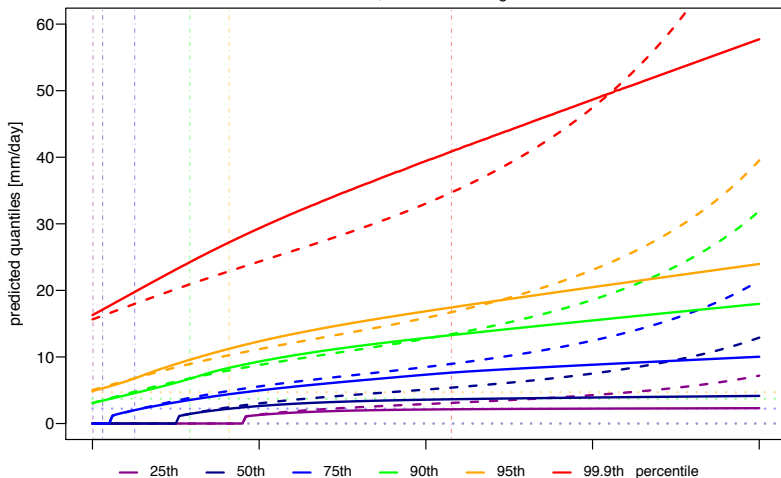


- VGLM mixture model
- VGLM gamma model

# Predicted quantiles (DJF)

Cambridge, CEE England (Station Id. 454)

solid: VGLM mixture; dashed: VGLM gamma

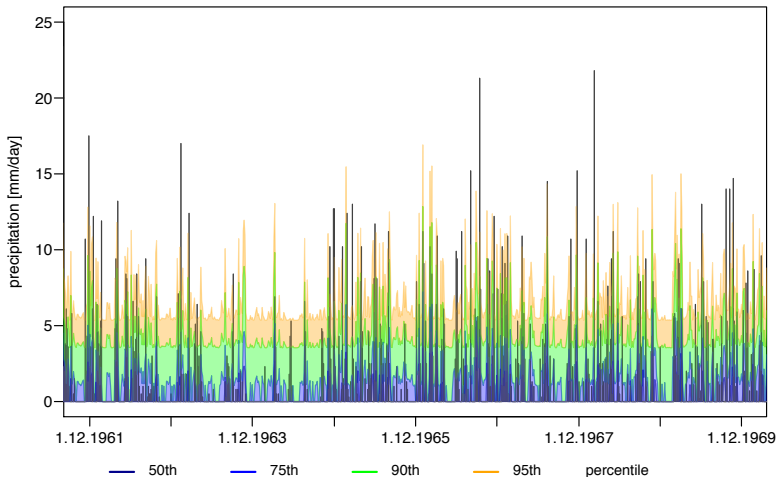


Wong et al., J Climate, 2014.



# Example series (DJF)

Cambridge, CEE England (Station Id. 454)



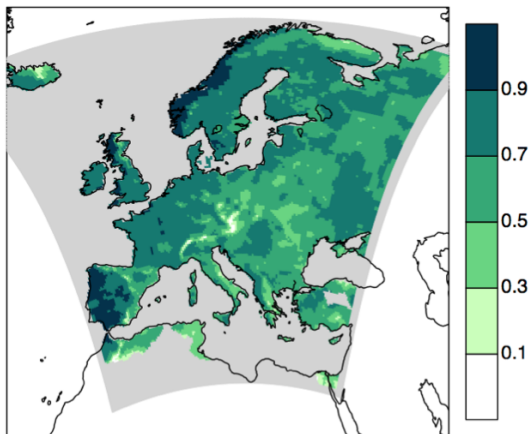
color: predicted quantiles; black: observations

Wong et al., J Climate, 2014.

## Pairwise MOS · Location correcting methods

# Do RCMs correctly represent location?

Correlation between observed/modelled seasonal winter precipitation

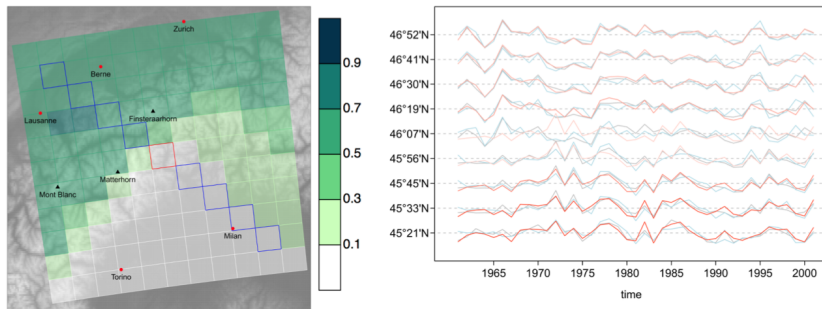


*Maraun & Widmann, HESS, 2015*

# An example from the Alps

Left: correlation between observed precipitation at Domodossola (red square) and simulated precipitation.

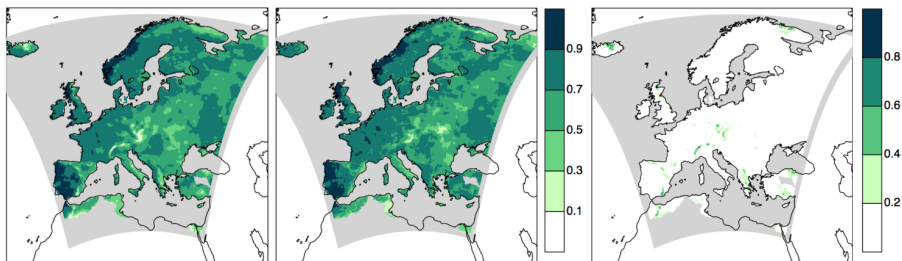
right: time series; blue: observed; red: model at the same grid box; grey: model at optimal grid box



*Maraun & Widmann, HESS, 2015*

# Non-local bias correction

Correlation between observed/modelled precipitation. Left: local correction; centre: non-local correction; right: improvement.



*Maraun & Widmann, HESS, 2015*