

Downscaling applications in coastal research and paleoclimatology



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with contributions from

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Content

- **downscaling for wave hindcasts (1948 - present)**
- **downscaling for Mid-Holocene precipitation**
- **comments on Perfect Prog and Model Output Statistics**

Local wave hindcast for Helgoland area (German Bight)

Problem

- high-resolution wave statistics for the near-shore zones are needed for coastal protection and offshore engineering
- measurements are sparse (e.g. one location, 2km from Helgoland coast)
- existing hindcasts are too coarse and don't include shallow water effects

Approach

- simulate wave statistics with very high resolution model for several years
- validate
- use model as surrogate reality and fit statistical downscaling model, apply for several decades

(courtesy Lidia Gaslikova, Ralf Weisse)

Wave simulations

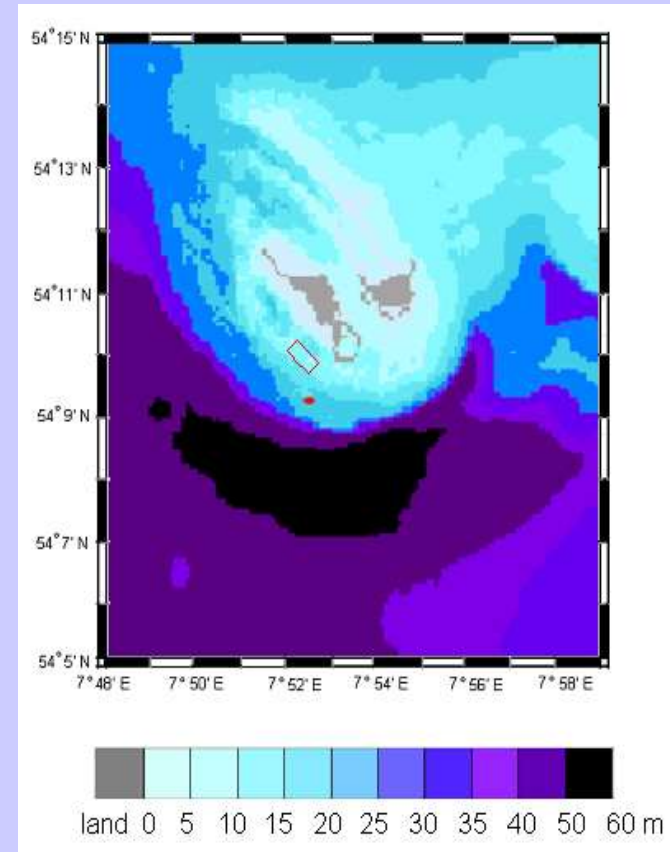
Low resolution wave climate from HIPOCAS

(HIndcast of dynamic Processes of the Ocean and Coastal AreaS of europe)

- dynamical spectral wave model (WAM), 5 km x 5 km, 3h output, covers North Sea
- forced with REMO winds 1948 - present

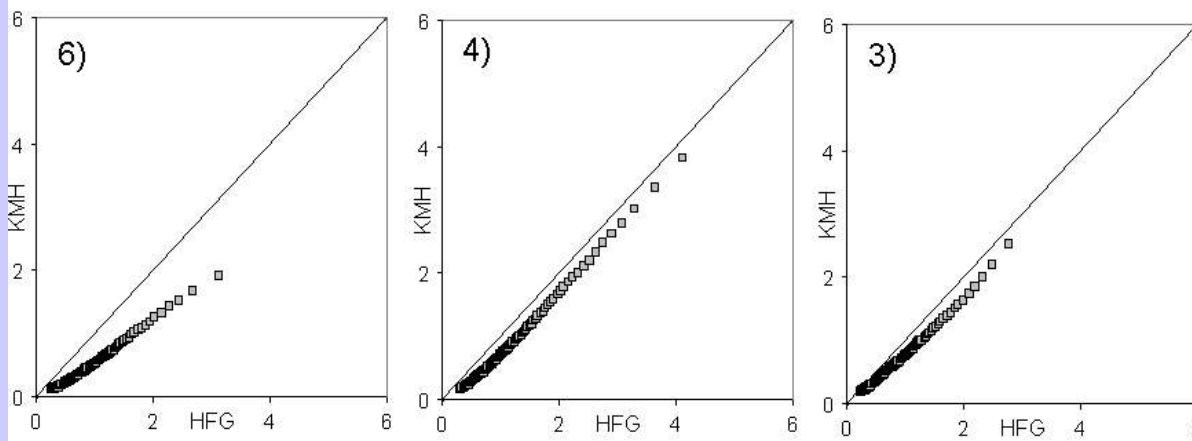
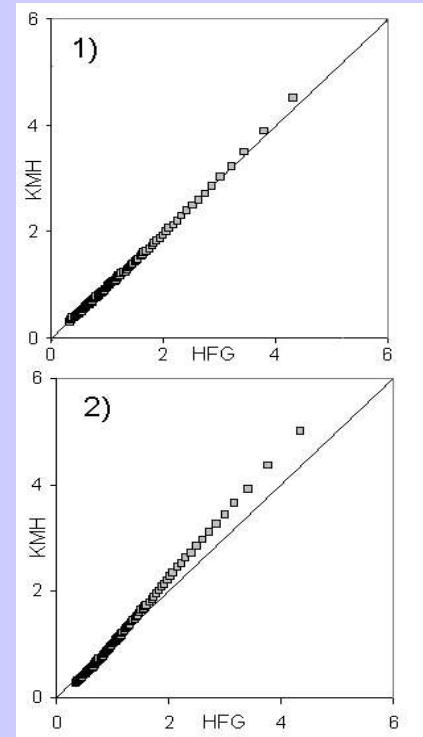
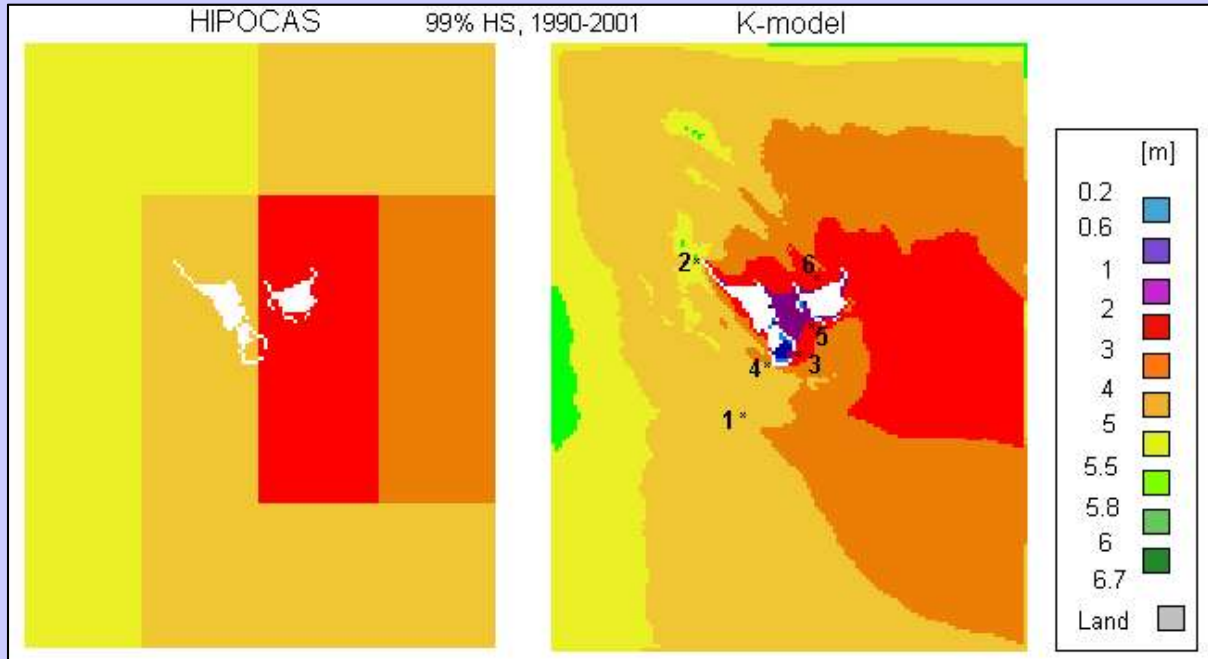
High resolution wave climate from K-model

- spectral, shallow water, bottom dissipation, wave breaking, non-linear dissipation, refraction caused by currents and changing depth
- input: HIPOCAS wave spectra at boundaries, 10m wind from REMO (1 hour), currents, water level variations
- output: 100 m, 1 h, significant wave heights peak period, peak direction, **validation OK**
- 1990-2001, 15 000 CPUh on NEC SX6



Added value from the K-model simulation

99 percentile of significant wave height



**5%-99% significant wave heights
HIPOCAS (x-axis),
K-model (y-axis)
1990-2001, 3-hourly**

(courtesy Lidia Gaslikova, Ralf Weisse)

Statistical downscaling

Predictand

high-resolution significant wave height field from HIPOCAS, 3 h

Methods

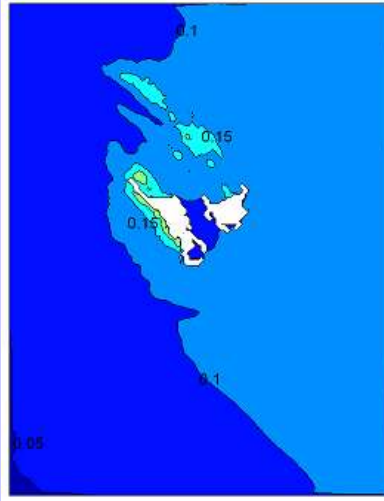
- linear regression separately for each K-Model grid point and for 8 wind segments
- PC-filtered CCA, 2 left, 2 right PCs, which explain 99.1 % and 98.3% of total variance
- analogs based on PC1 from HIPOCAS

Fitting/Validation

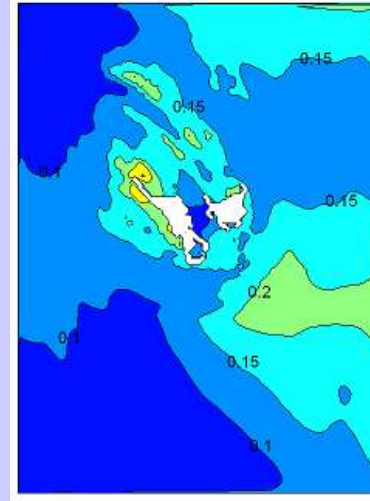
fitting period 1990-1994, validation period 1995-2001

Comparison of downscaling methods

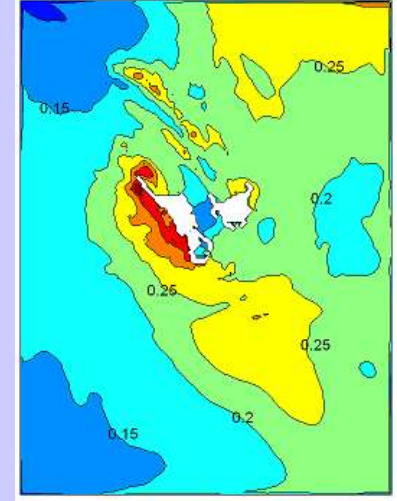
MLR



CCA



Analogs



RMSE between instantaneous significant wave heights from K-model and statistical methods

Brier skill score for 99%-tiles:

$$B = 1 - \frac{E((F - P)^2)}{E((R - P)^2)}$$

F – forecast (downscaling results)

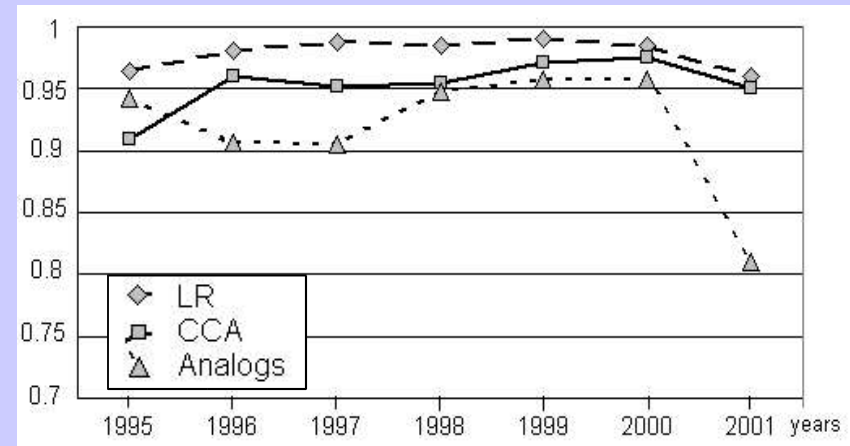
P – observations (K-model)

R – reference (HIPOCAS)

MLR performs best

MLR has more parameters than CCA

Analogs perform worst, despite 15 000 cases, no analog situations in 2001?



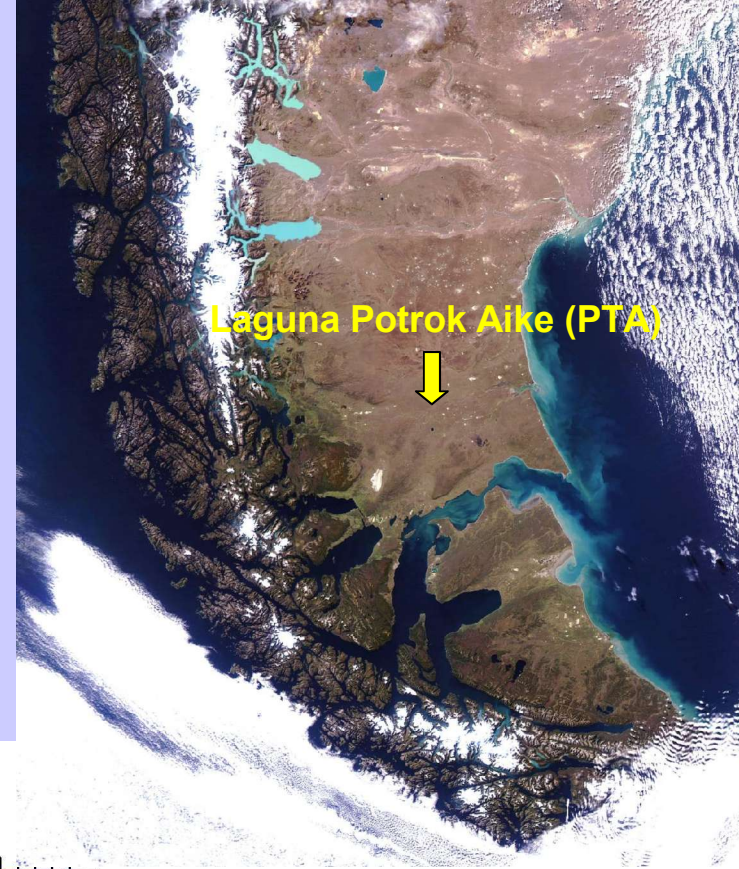
(courtesy Lidia Gaslikova, Ralf Weisse)

Mid-Holocene hydrological conditions in southern Patagonia

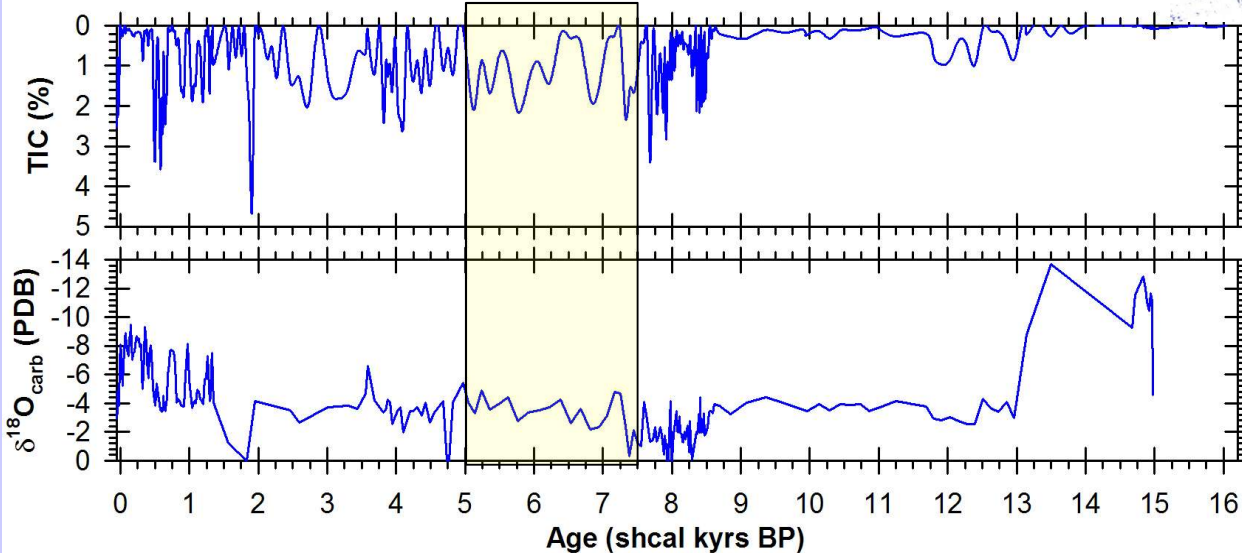
Total Inorganic Carbon (TIC) and $\delta^{18}\text{O}$ are proxies for lake level

lower lake levels in the Mid-Holocene compared to the pre-industrial period

hypothesis: changes are related to the position of the Southern Hemisphere Westerlies



simulated period



high
low

high
low



lake level



ECHO-G

ECHAM4 coupled with HOPE-G

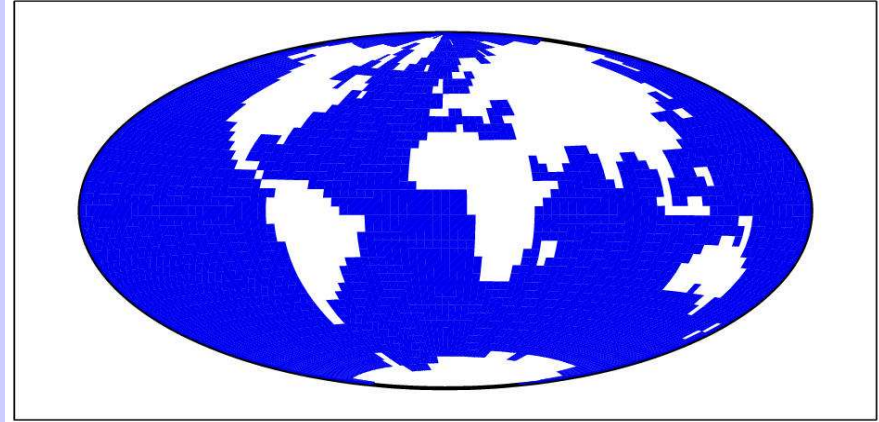
ECHAM4

- 19 atmospheric levels
- T30, approx. 3.8 x 3.8 degrees

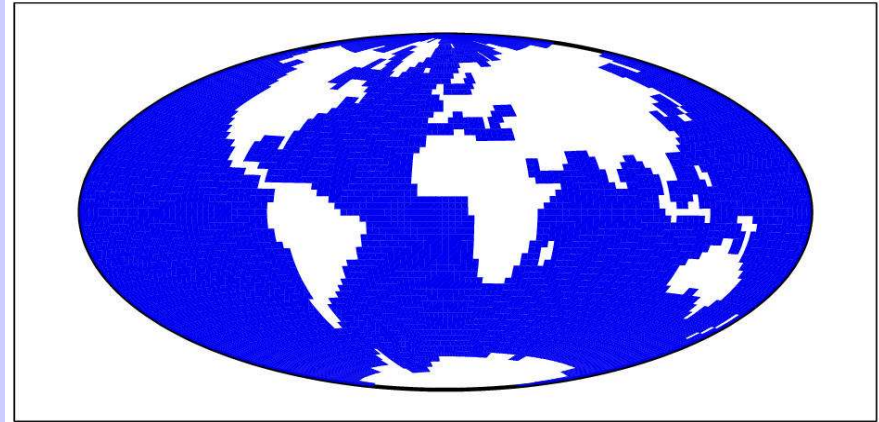
HOPE-G

- 20 ocean levels
- approx. 2.8 x 2.8 degrees
- dynamic-thermodynamic sea-ice model

T30 (ECHAM4)



T42 (HOPE-G)



Simulation speed of ~500 years/month on NEC SX6 8 processors

Model is flux corrected, with prescribed vegetation and land ice

MIDHOL Model Simulations



2. Forced Simulation with orbital forcing 7-4.5 ka BP - **ORB**
4. Forced simulation with orbital, solar und greenhouse gas (CO_2 , CH_4 , N_2O) 7-4.5kyr BP - **ORBSG**
6. Control simulation 7 ka BP - 500 years - internal variability

1. Control simulation 7 ka BP
CON7K
 - 2. Orbital forcing
 - 3. Orbital, solar, greenhouse gas

Also: Control simulation with pre-industrial conditions - 300 years –
CONPRE

<http://w3k.gkss.de/staff/wagner/midhol.html>

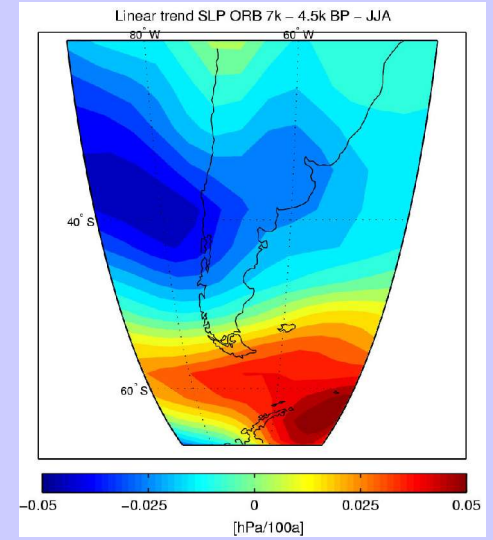
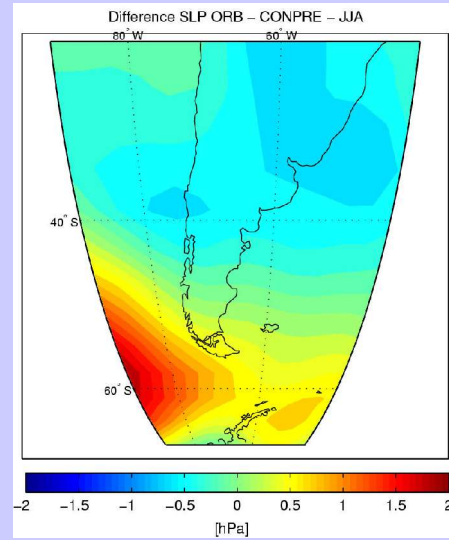
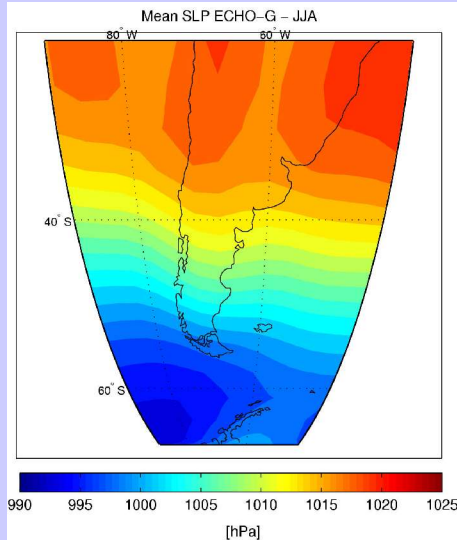
JJA circulation in ECHO-G

Pre-industrial
mean

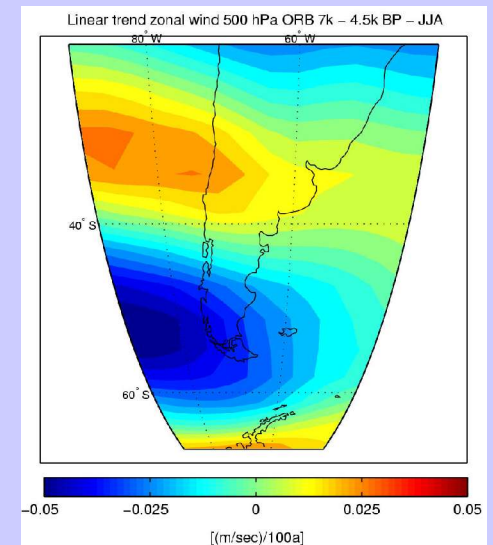
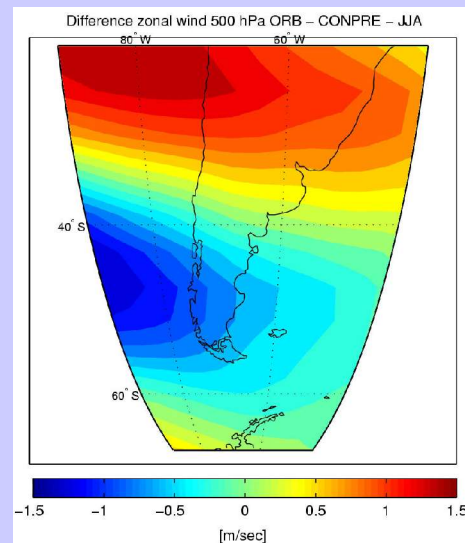
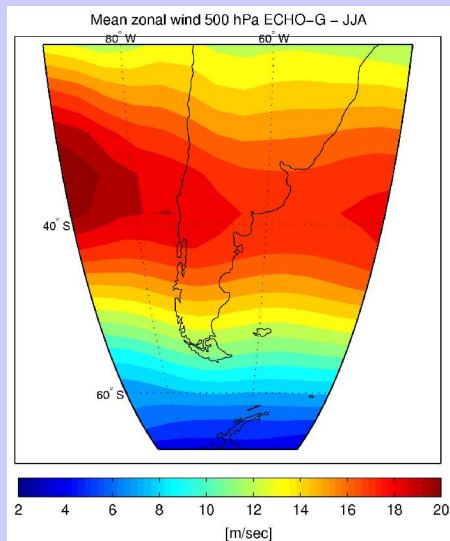
Mid-Holocene -
preindustrial

Mid-Holocene trends
(7 - 4.5 ky BP)

SLP



u500



(courtesy Sebastian Wagner)

Downscaling regression weights

predictand:
log of monthly VASCLIMO
JJA precip at Potrok Aike

final weights from
PC-filtered MLR
(in hPa^{-1} and s m^{-1})

upper row: separate
models for SLP and u500

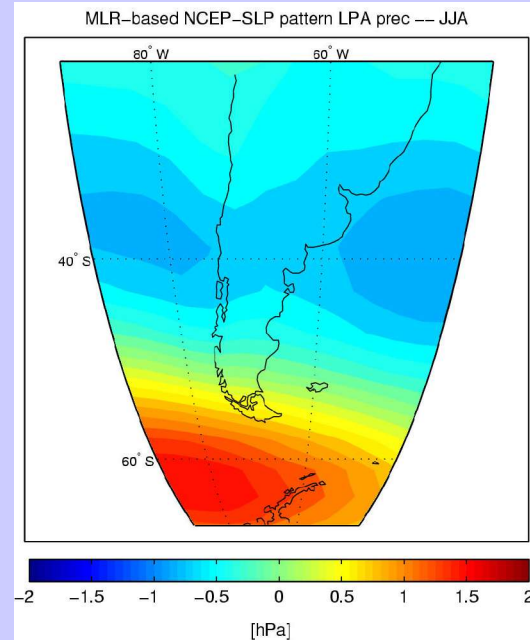
lower row: SLP and u500
as common predictor

skill combined JJA model:
 $r=0.60$, $\text{RE}=0.35$

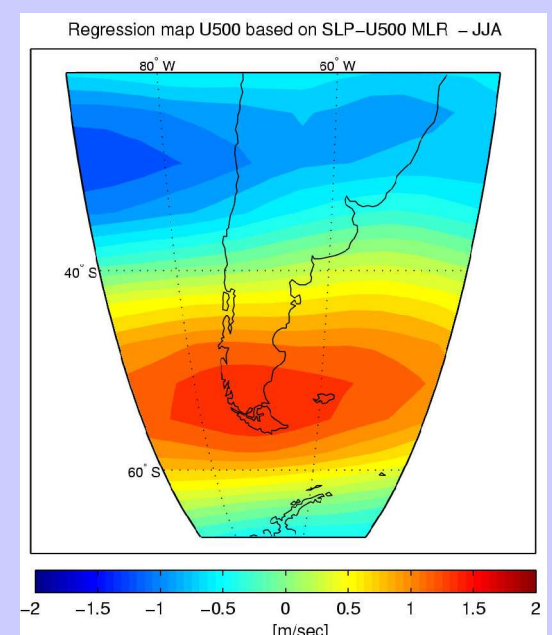
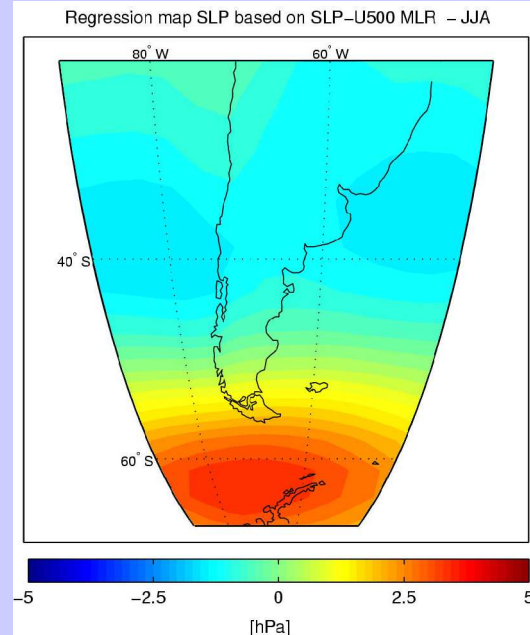
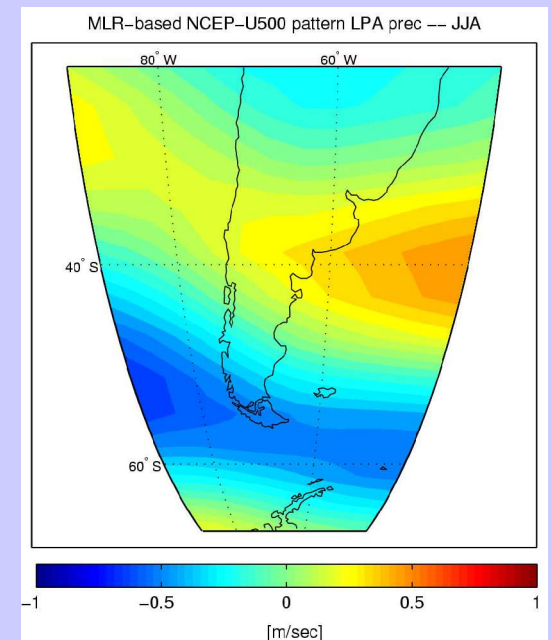
$[r=0.45, \text{RE}=0.19 \text{ (DJF)}]$

(courtesy Sebastian Wagner)

SLP



u500



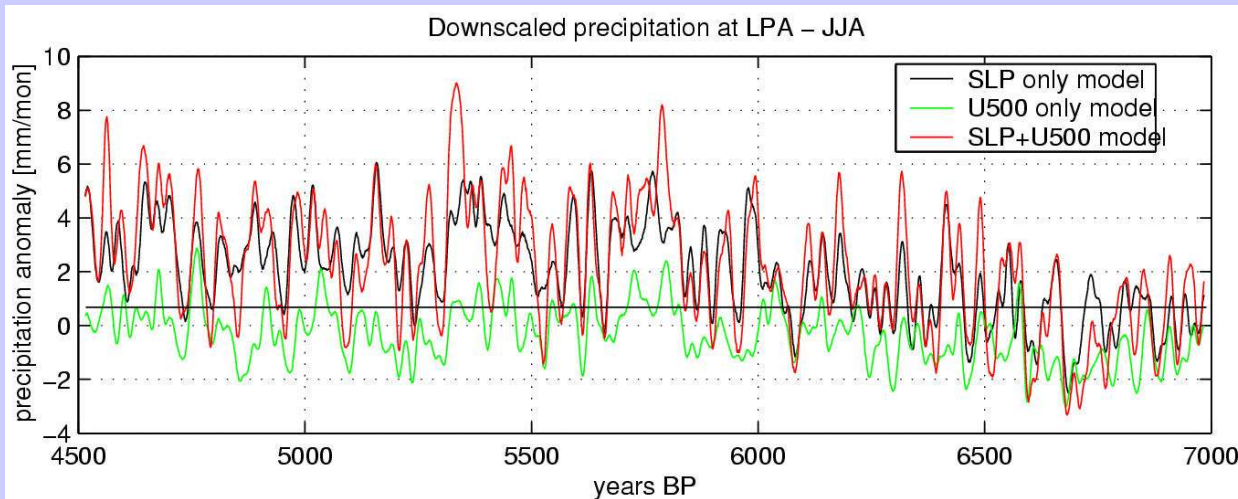
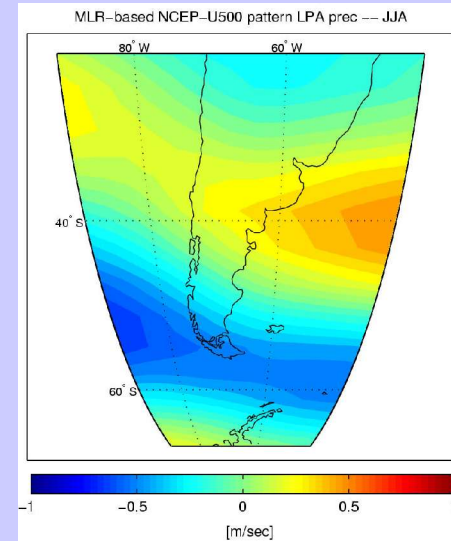
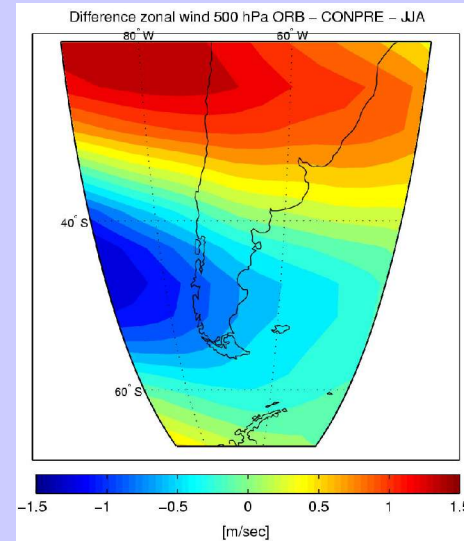
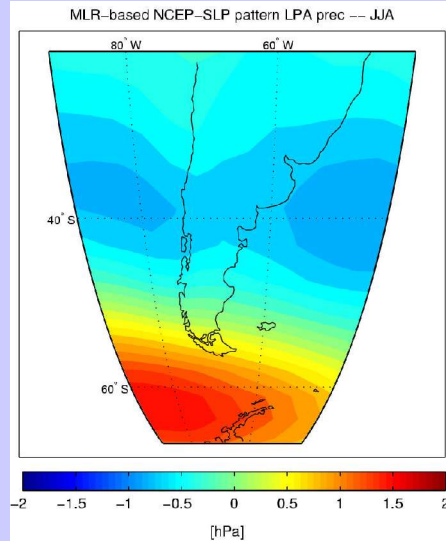
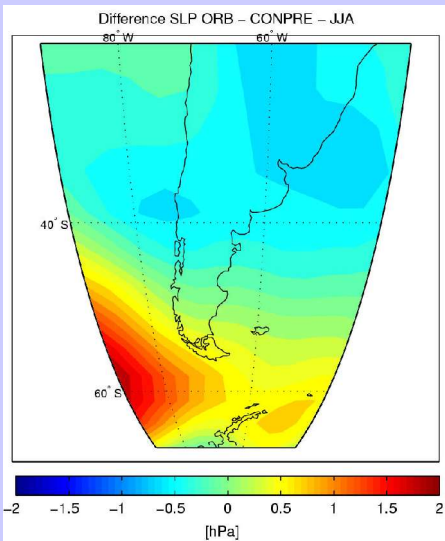
Mid-Holocene precipitation at Laguna Potrok Aike from DS

SLP

u500

Mid-Hol. - preindust. regression weights

Mid-Hol. - preindust. regression weights



est. precip model-dependent

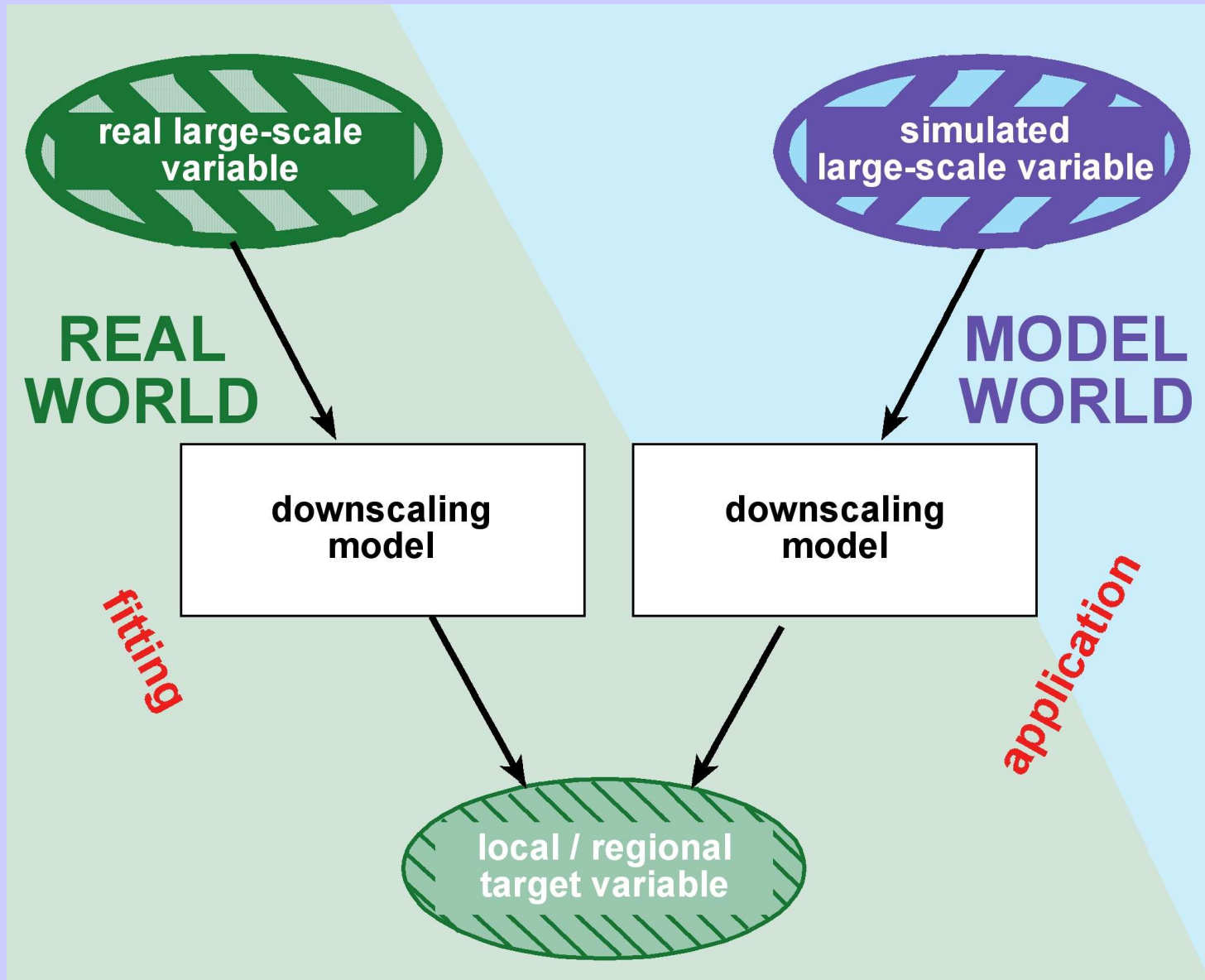
JJA not consistent with proxies

influence of areas that may not be physically linked to predictand

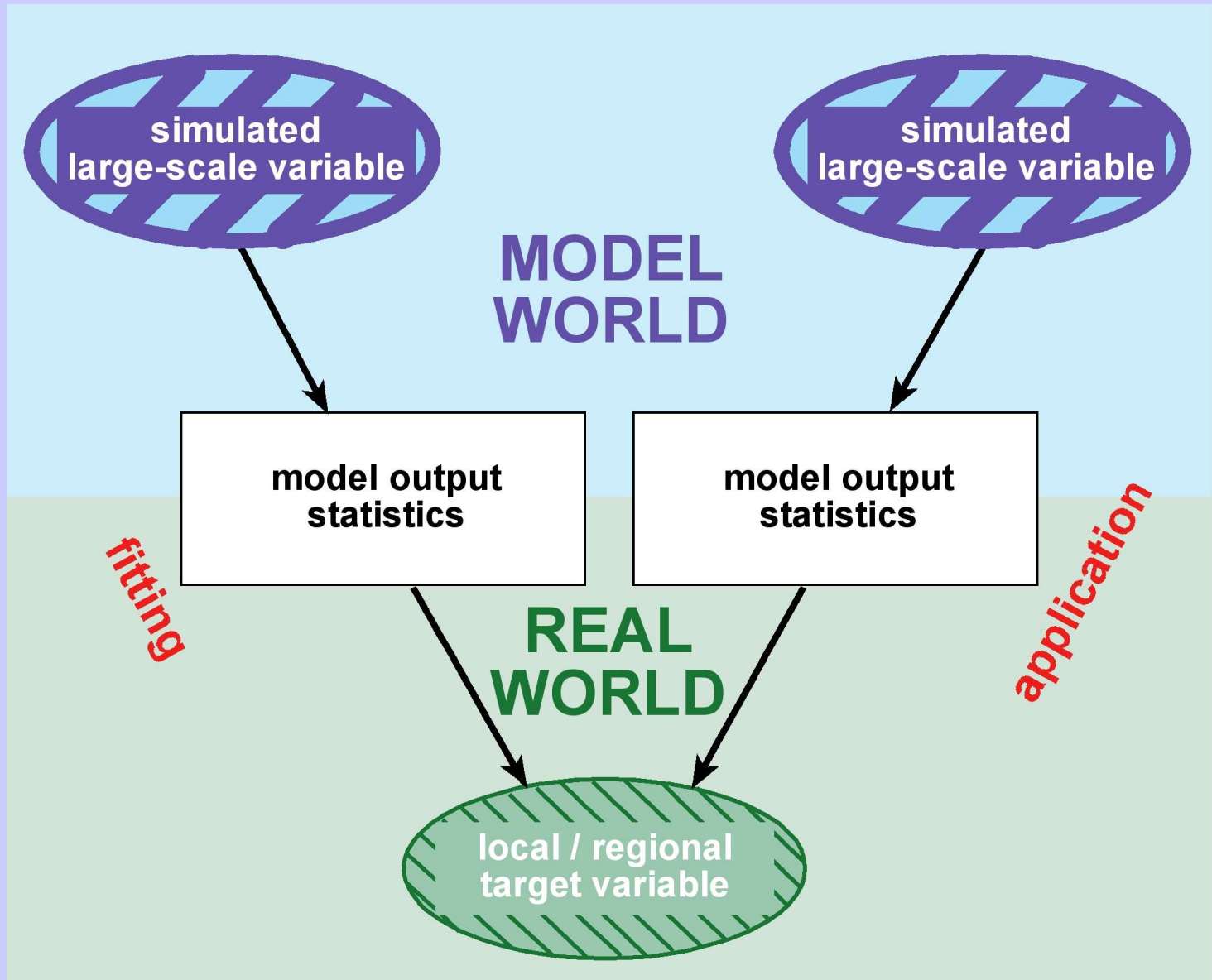
weights may be affected by interannual teleconnections

(courtesy Sebastian Wagner)

Perfect Prog(nosis) Downscaling



Model Output Statistics (MOS)



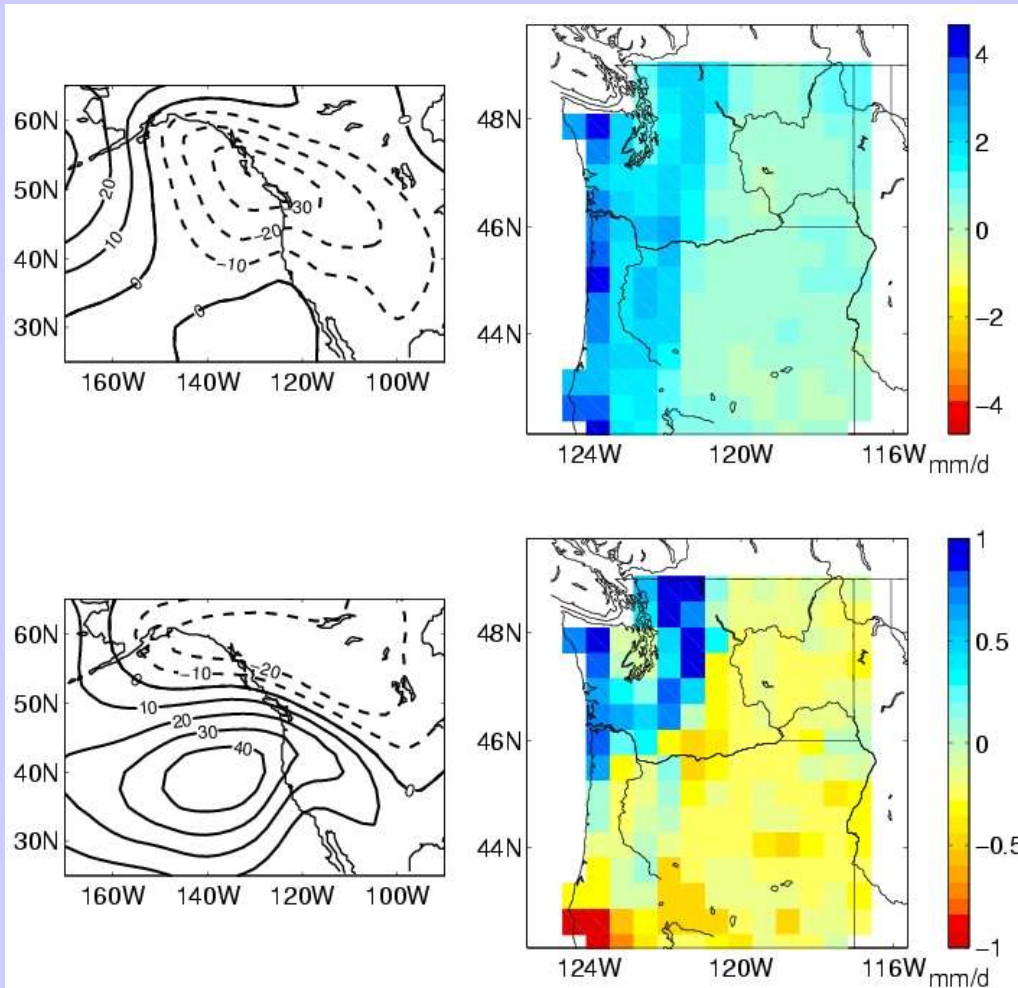
Coupled anomaly patterns (SVD) between DJF pressure (Z1000, NCEP) and daily precipitation (Perfect Prog downscaling)

geopot. height (Z1000)

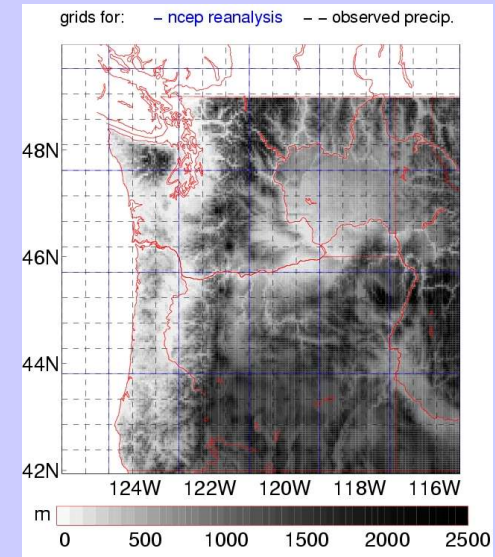
precipitation

topography

pair 1



pair 2

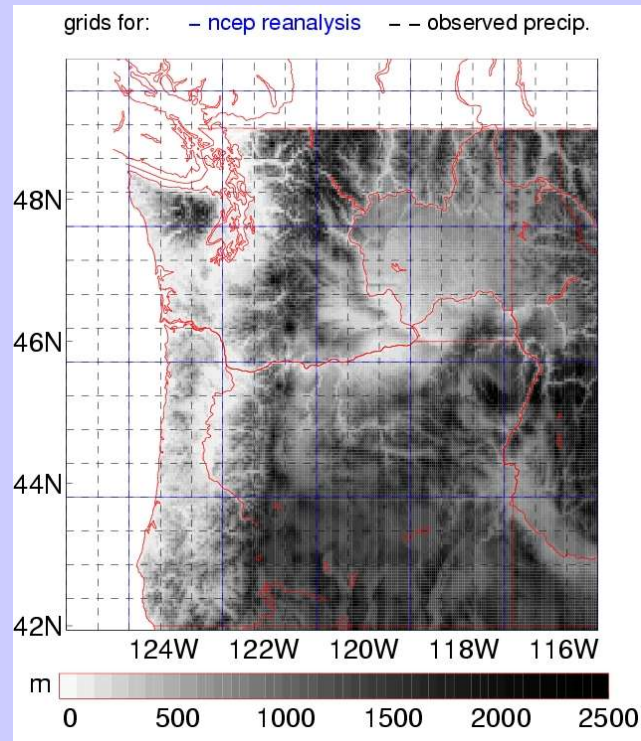
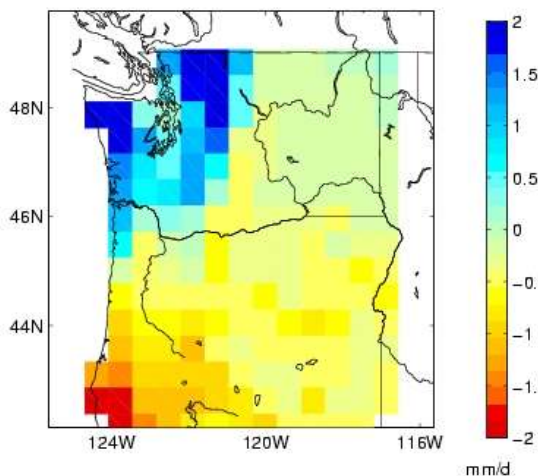
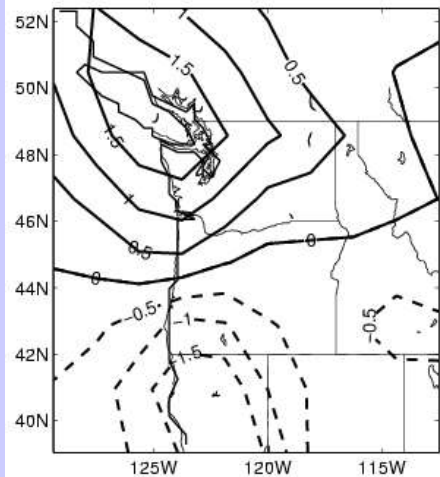
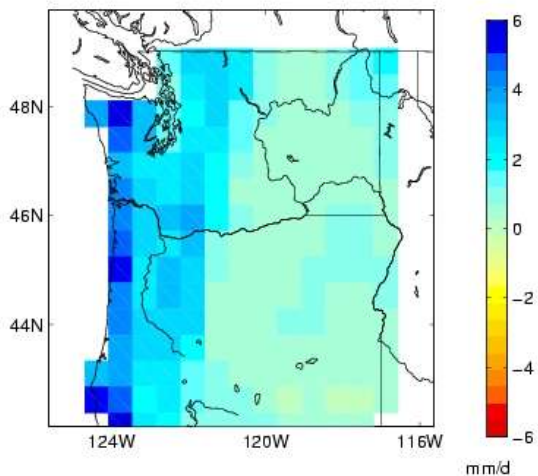
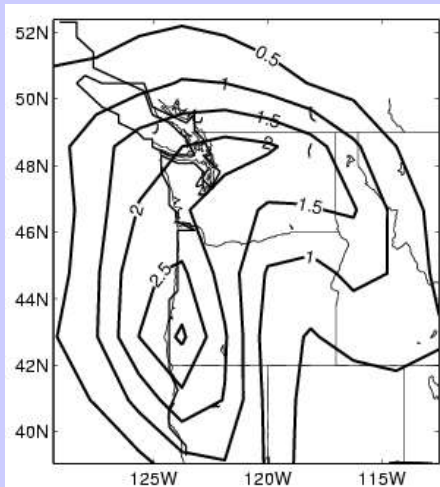


(Widmann and Bretherton, J. Climate 2000, Widmann et al., J. Climate, 2003)

Coupled anomaly patterns (SVD) between simulated (NCEP) and observed daily precipitation (DJF)

simulated precipitation
(NCEP reanalysis)

observations

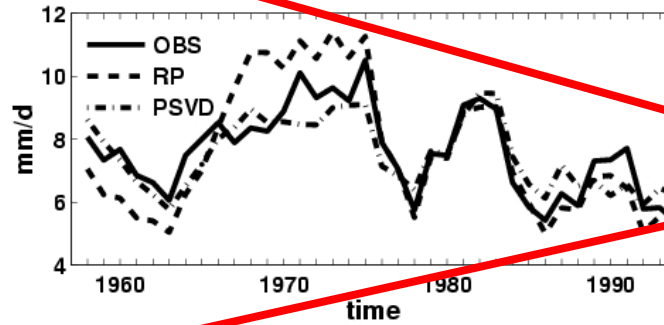
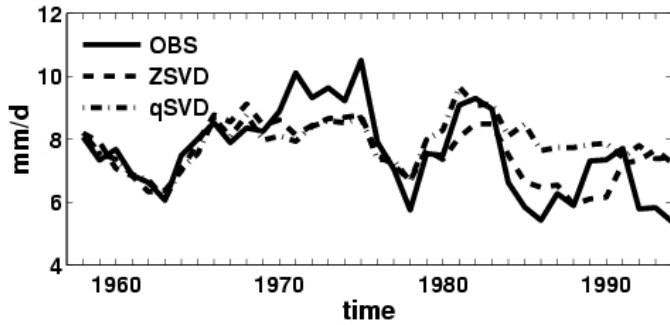


precipitation (3 y mean, DJF 1958-1998)

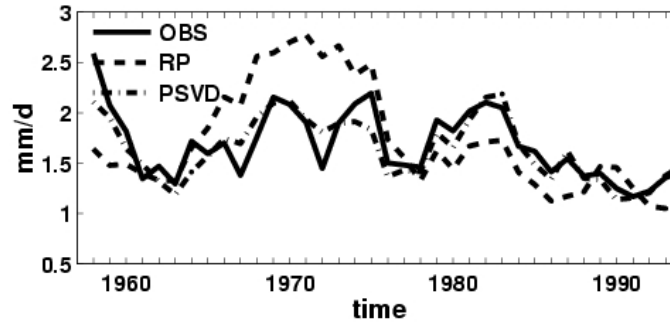
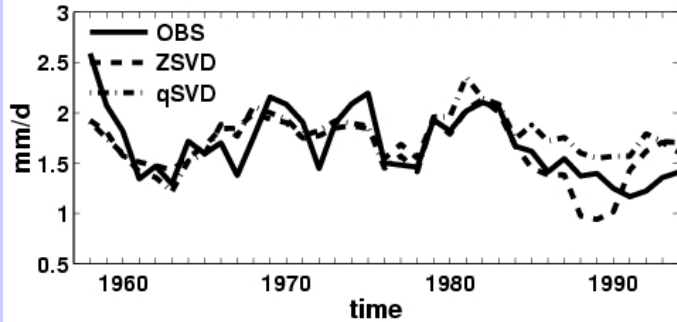
predictors: geopot. height (ZSVD)
or humidity (qSVD)

simulated precipitation
(RP and PSVD)

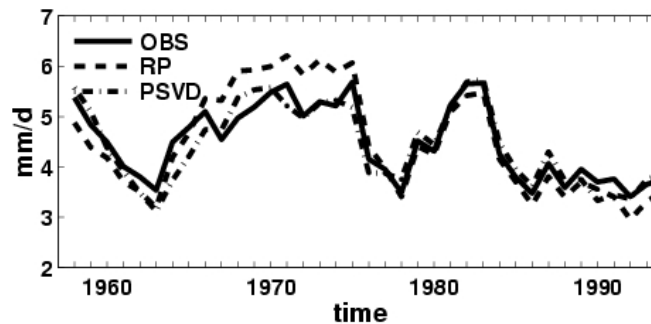
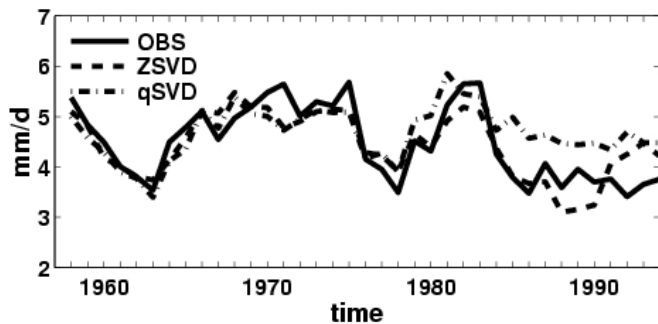
121.87 W, 46.67 N



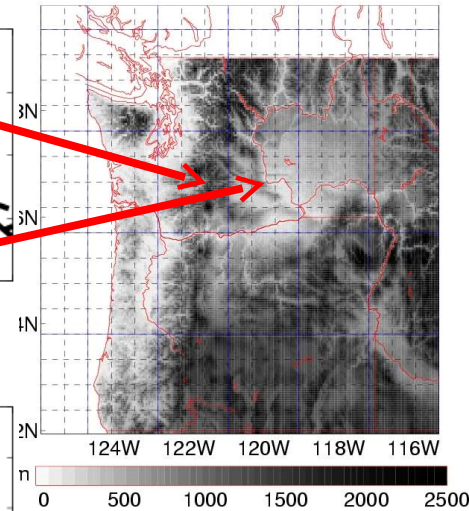
118.12 W, 46.67 N



OR and WA



grids for: - ncep reanalysis - - observed precip.



MOS applications and requirements

widely used in weather forecasting and partly in seasonal forecasting

potential advantages

- smaller danger of not taking relevant predictors into account
- the statistical model may be simpler, as many processes are still simulated rather than statistically modelled.
- extreme values may be better represented

challenge for application for future climate change and paleo application is the model fitting. One needs:

- GCM simulation in reanalysis mode
- or GCM simulation nudged towards reanalysis circulation states (nudging modules exist for ECHAM4/5)
- how stable are the relationships? (new issue if MOS is applied to long time scales)

Lessons and questions

DS for North Sea waves

individual regression for each grid cell work better than CCA and analogs

large number of model parameters seems beneficial and does in this case not lead to overfitting (what's the advantage of pattern-based methods?)

DS for Patagonia precipitation

DS is crucial in paleoclimatology for proxy-model consistency test

results are strongly model-dependent, but a smaller predictor area makes them stable (new results, not shown)

large predictor area can be problematic, because it increases the chance that predictor teleconnections that are not relevant for the downscaling problem affect the regression weights (overfitting, or time-scale dependent teleconnections)

MOS for climate change and paleo simulations

large potential for improvement

we need historic GCM simulations for fitting (nudged towards reanalyses)