

Projecting climate change impacts on hydrology: the potential role of daily GCM output

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Precipitation. mm/d

P Intensity, mm/d

Our Goal: Improve methods for projecting climate change impacts to watershed scale Our Focus: Downscaling climate model output to capture changes in hydrology

Our Methods:

1 – Downscale precipitation and temperature over the Western U.S. using two different techniques

2 - Drive a hydrology model with each, and compare their performance

3 – Develop an improved method

What we downscale: NCEP/NCAR Reanalysis

- · Reanalysis represents the best possible GCM since obs are assimilated Should show max differentiation in methods
- T62 (~1.9°) resolution, comparable to GCMs
- Full period daily and monthly data available
- 1950-1976 used to "train" downscaling
- 1977-1999 used to assess, used as a "changed climate" for projections Shift in PDO in 1976-77, late 20th century warming
- · Change in data sources to Reanalysis in 1979
- Warmer, wetter in later period over Western U.S.

Step 1: Bias-Correction

At each grid cell, use quantile mapping to match monthly statistics (at GCM scale)

Step 2: Spatial Downscaling

Downscaling Methods

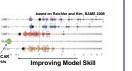
Bias Correction/Spatial Downscaling (BCSD)



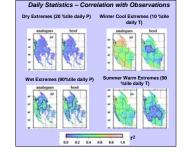
- · Calculate anomalies relative to coarse-scale climatology Interpolate anomalies to 1/8°
- grid · Apply to 1/8° climatology

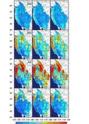
Common Characteristics of BCSD and CA

- · Both provide spatially continuous (gridded) downscaled fields
- · Observed spatial and temporal climate structure maintained
- · Automated and efficient: can be used for ensembles of GCMs
- · Capable of downscaling long transient GCM runs
- spatial GCM biases - variability biases



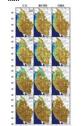
Downscaled Meteorology and Derived Hydrology





Evapotransipration.

mm/d



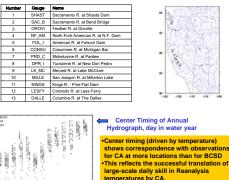
Active soil moisture,

mm



Annual P cycle captured with both methods, CA-higher T daily skills, BCSD -better rainfall intensity Mean, seasonal cycles and interannual variability of soil moisture are reasonably reproduced by both BCSD and CA. End-of-season snow accumulation also appears to be plausibly reproduced by both BCSD and CA Where BCSD or CA differ from Observations (e.g., April soil moisture in the Pacific Northwest), they differ in similar ways. Hydrologic states appear to be recovered well by either downscaling method.

Streamflow Simulations: 22 Years







Combining Downscaling Methods

Problematic Biases at large scale

Daily precipitation probabilities at reanalysis grid point at 37,1422, -110,625;



Step 1 from BCSD applied to

New streamflows generated

Bias correction at large scale

solves problems with peak

flows and annual volumes

 Problems remain at low flows BCCA outperforms both CA

and BCSD for most measures

daily reanalysis precip

•CA applied (without

anomalizing)

Drizzle bias (January shown here) Obs shows 40% of days with zero precip Reanalysis never has zero precip

Solution: Bias Correct before CA method (BCCA)

Gauge in bold face and highlighted indicates downscaled distribution of 22 values differs from the observed distribution, compared with a Kolmogoro Smirnov 2-sample test (at p=0.05).

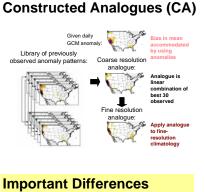
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BCBD	CA	BOCA	8080	CA	BOCA	8080	CA	BCCA	8080	CA	BCCA
SHAST	SHAST	SHAST	SHAST	SHAST	SHAST	SHAST	SHAST	SHAST	SHAST	SHAST	SHAST
SAC_B	SAC_B	SAC_B	SAC_B	SAC_B	SAC_B	SAC_B	SAC_B	SAC_B	SAC_B	SAC_B	SAC_B
OROVI	OROVI	OROVI	OROVI	OROVI	OROVI	OROVI	OROVI	OROVI	OROVI	OROVI	OROVI
NF_AM	NF_AM	NF_AM	NF_AM	NF_AM	NF_AM	HF_AM	NF_AM	HF_AM	NF_AM	NF_AM	NF_AM
FOLJ	FOLJ	FOLJ	FOL_I	FOLJ	FOL_I	FOLJ	FOL_I	FOLJ	FOL_I	FOLJ	FOL_I
CONSU	CONSU	CONSU	CONSU	CONSU	CONSU	CONSU	CONSU	CONSU	CONSU	CONSU	CONSU
PRD_C	PRD_C	PRD_C	PRD_C	PRD_C	PRD_C	PRD_C	PRD_C	PRD_C	PRD_C	PRD_C	PRD_C
DPR.J	DPR_I	DPR_I	DPR.I	DPRU	DPR_I	DPR.J	DPR_I	DPRJ	DPR_I	DPRJ	DPR.J
LICINC	LK_MC	LK_MC	LK_MC	LK_MC	LK_MC	LK_MC	LK_MC	LK_MC	LK_MC	LK_MC	LK_MC
MILLE	MLLE	MILLE	MILLE	MLLE	MILLE	MILLE	MILLE	MILLE	MILLE	MILLE	MILLE
KINGS	KINGS	KINGS	KINGS	KINGS	KINGS	KINGS	KINGB	KINGS	KINGS	KINGIS	KINGS
LESFY	LESFY	LESPY	LESFY	LEOFY	LESFY	LEBFY	LEGFY	LESFY	LEOFY	LESFY	LESFY
DALLE	DALLE	DALLE	DALLE	DALLE	DALLE	DALLE	DALLE	DALLE	DALLE	DALLE	DALLE

- Daily large-scale skill can be successfully downscaled to local scales
- · Anomalizing is not adequate for coping with large-scale biases
- Explicit bias correction solves many problems, but post-processing needed

Acknowledgement

This project was funded by the California State Energy Resources Conservation and Development Commission, as part of the 2008 Scenarios Impact and Adaptation Study

Maure: E. P. and H.G. Hidago, 2008, Willy of taily vs. monthly large-scale climate data: an intercomparison of two statistica downscaling methods. Hydrology and Earth System Sciences Vol. 12, 651-653 Rechter, T., and J. Km, 2008. How Will Do Couplet Models Simulae Today's Climate? Buil, Amer. Meteor. Soc., 69, 303–311. Wood, A.W., and D. Learmaner, 2008. A Test Bed for New Second Hydrology Forcesting Approaches Into Western Unhad States. Buil, Amer. Meteor. Soc., 67, 1699–1712.



Important Differences Between BCSD and CA

- · CA uses daily GCM data: BCSD uses monthly w/random resampling to produce daily values BCSD explicitly corrects for systematic GCM biases based on historic
- GCM performance CA corrects mean bias (using anomalies) but not

NCEPINCAR