





Improving statistically downscaled daily data for hydrology: diagnosis and opportunities

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ABSTRACT

Incorporation of climate change information into long-term evaluations of water and energy resources A primary challenge facing resource managers in accommodating climate change is determining the range and uncertainty in regional and local climate projections. Since global climate models (GCMs) produce output at a spatial scale incompatible with local impact assessment, different techniques have evolved to downscale GCM output so locally important climate features are expressed in the projections. Recent research has compared skill using two statistical downscaling methods, the constructed analogues (CA) and the bias correction and spatial downscaling (BCSD). The CA method uses daily GCM output and the BCSD monthly. We evaluate the downscaled precipitation and temperature (from the NCEP/NCAR reanalysis and a GCM) for the late 20th century against observations, and compare projections with both methods. We include an assessment of the GCM biases, and present a new method for correcting for GCM biases in a hybrid method combining the most important characteristics of both methods.

Downscaling with BCSD

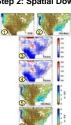
Step 1: Bias-Correction



- At each grid cell for "training" period develop monthly CDFs of P, T for
- GCM
- Observations (aggregated to GCM scale)
- Obs are from Maurer et al. [2002]
- Use quantile mapping to ensure monthly statistics (at GCM scale) match
- Apply same quantile mapping to "projected" period



Step 2: Spatial Downscaling

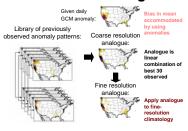


- Use bias-corrected monthly
- GCM output ①

 Aggregate obs to GCM scale ②
- Calculate P,T factors relative to coarse-scale climatology
 = ① (P) ③ = ① - ② (T)
- Interpolate factors to 1/8° grid
- Apply to fine-scale climatology (6) = (4)-(5)(P) (6) = (4)+(5)(T)

Daily Values from rescaled historical values

Downscaling with Constructed Analogues



Comparing BCSD and CA

Characteristics in Common

- . 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
- Capable of producing daily output

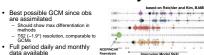
Fundamental Differences

- CA uses daily GCM data; BCSD uses monthly w/random resampling to produce daily values
- CA (like a "perfect prog" forecast system) develops relationships between coarse- and fine-scale observations, and applies them to GCM output
- BCSD (like a "MOS" type forecast) relates GCM output directly to observations, explicitly correcting for systematic GCM biases based on historic GCM performance
- CA corrects mean bias (using anomalies) but not – spatial GCM biases
- variability biases

Observational Baseline

- The "truth" against which BCSD and CA simulations are compared is hydrologic model output driven by gridded observed meteorology from 4037 4000.
- See Maurer and Hidalgo, 2008, for details on intercomparison

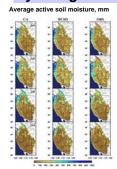
Surrogate GCM output: NCEP/NCAR Reanalysis

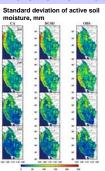


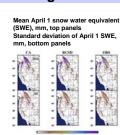
- . 1950-1976 used to "train" downscaling
- CA: coarse obs to fine (1/8°) obs
 BCSD: coarse reanalysis to fine (1/8°) obs
- 1977-1999 used to assess
- Shift in PDO in 1976-77, late 20th century warming
- Warmer, wetter in later period over Western U.S.

COMPARING HYDROLOGIC SIMULATIONS WITH CA AND BCSD

Hydrologic States: simulation of water storage in soil and snow pack







Conclusions, part 1

- - CA.
 End-of-season snow
 accumulation also appears to be
 plausibly reproduced by both
 BCSD and CA
 - Where they differ from Observations (for example, April soil moisture in the Pacific Northwest), BCSD and CA tend to differ in similar ways.
- Hydrologic states appear to be recovered well by either downscaling method.

Streamflow: Statistics at Selected Sites in the Western U.S.

Streamflow Gauges Used in this Study

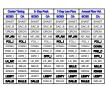
Number	Gauge	Name
1	SHAST	Sacramento River at Shasta Dam
2	SAC_B	Sacramento River at Bend Bridge
3	OROVI	Feather River at Oroville
4	NF_AM	North Fork American River at North Fork Dam
5	FOL_I	American River at Folsom Dam
6	CONSU	Cosumnes River at Michigan Bar
7	PRD_C	Mokelumne River at Pardee
8	DPR_I	Tuolumne River at New Don Pedro
9	LK_MC	Merced River at Lake McClure
10	MILLE	San Joaquin River at Millerton Lake
11	KINGS	Kings River - Pine Flat Dam
12	LESFY	Colorado River at Lees Ferry
13	DALLE	Columbia River at The Dalles



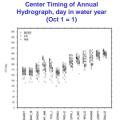




Same as table to right, but bold/highlighted indicates downscaled distribution of 22 values differs from the observed distribution, based on a Kolmogorov-Smirnov 2-sample test (at p=0.05

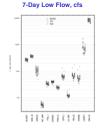


Streamflow statistics for 22 water years: 1978-1999





3-Day Peak Flow, cfs



Conclusions, part 2

- Center timing, a feature driven by temperature more than precipitation, shows correspondence with observations for CA at more locations than for BCSD
- This reflects the successful translation of large-scale daily skill in Reanalysis temperatures by CA
- For precipitation-driven daily statistics of low and high flows, BCSD shows correspondence with observations at more locations that CA.
- The differences in precipitation-driven differences could be explained by:
 -The random resampling of daily sequences in BCSD may bear more resemblance to observations than large-scale CA's daily downscaled Reanalysis anomalies
 -Observed interannual variability is preserved by bias correction in BCSD, but not in CA

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References
Maure, E. P. and H.G. Hadaps, 2008, Utility of daily vs. monthly large-scale climate data: an intercomparison of two statistical downscaling methods, Hydrology and Earth
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