Producing Dynamically Downscaled Information for Climate Change Projection in Arizona



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Our goal is to downscale climate model data to an appropriate resolution for hydrological applications.



1. We rely on General Circulation Models (GCMs) to estimate future climate variables



2. These projections are downscaled using either statistical or dynamical methods



3. Downscaled atmospheric fields force the VIC hydrologic model.

4. The outcomes will be used to generate water management data for drought planning, scenarios, modeling, agricultural, tribal activities, etc.



There are basically two approaches to downscale coupled climate model projections :

Statistical Downscaling

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Statistical Downscaling

These methods assume a relationship between largescale atmospheric variables (predictors) and local climate variables (predictands).

- Pro : Cheap and computationally efficient.
- Pro : Can use many different scenarios, model runs.
- Pro : Easily transferable to other regions.

- Con : Requires long and reliable observation data.
- Con : Depends on choice of predictors.
- Con : Assumes stationarity of predictor-predictand relationship.
- Con : Cannot account for feedbacks.



Statistical Downscaling

It is important to clarify that the Reclamation Data is Bias Corrected, so the observed climatological mean is matched in the historical data.



Raw GCM climatology for the three selected models (1950-2000)

"Reclamation" statistically downscaled climatology (1950-2000)

The second downscaling approach is dynamical downscaling.

Dynamical Downscaling

- Pro : Produces responses based on physically consistent processes.
- Pro : Captures feedbacks.
- Pro : Can model changes that have never been observed in historical record.
- Pro : Useful where topographic controls are important.

- Con : Requires significant computational power.
- Con : Limited amounts of models / runs / timescales.
- Con : Dependent on GCM boundary forcing.
- Con : Problems with drifting of large-scale climate.

Use the regional model as a "magnifying lens" to create higher resolution data with identical parameterization options as used for real time WRF UA forecasts...



Regional Model Grid (35km grid spacing)



GCM not only provides lateral boundary conditions. It is also used to force the interior of the model via a spectral nudging approach...



This helps maintain the appropriate variability in model fields at upper levels and at large scales.

We are performing different types of Dynamical Downscaling at the UofA using WRF:



Future Projections (IPCC) Using HadCM3 1968-2079

Science Question: How will climate in the Southwestern United States change due to global warming?

Methodology: We use WRF with spectral nudging to downscale 113 (1968-2081) years of SRES A2 data from three "well performing" IPCC models.



We chose to downscale the three well performing IPCC AR4 models that best represent the historical precipitation and temperature climatology in the Southwest and upper atmosphere circulation patterns in the Northern Hemisphere. These also have different GCM precipitation projections for Arizona (i.e. include "wet" and "dry" models)



Dominguez et al (2009)

Monthly average precipitation from IPCC models during the previous century





Historical average of simulations (sres_20c3m) 1970-2000

(Francina Dominguez)

Preliminary analysis of results...

1968-2000 monthly climatology shows that WRF represents the timing and intensity of the Monsoon more realistically than the raw model.





Preliminary analysis of results...

1968-2000 June July and August precipitation climatology of WRF downscaled UKMO-HadCM3 data show a much more realistic spatial representation of the North American Monsoon than the raw model.



Preliminary results of future precipitation show that the 2001-2040 climatology has a generalized higher precipitation – Particularly in July, as compared to the 1968-2000 monthly climatology. Similar results at river basin scale (e.g.Salt, Verde)



We have talked about the climatological analysis, now let's look at the interannual variability...

Connections between seasonal precipitation variability and large scale teleconnections



Statistical analysis methods:

Precipitation normalization

- Standard Precipitation Index(SPI): Accounts for non-normal distribution of precipitation

Spatial pattern recognition

 Rotated Empirical Orthogonal Function Analysis(REOF)

Relationship of dominant modes of precipitation variability to large-scale forcing factors

 Regression Analysis between precipitation modes and SST, geopotential height What we look for to assess natural variability in downscaled climate data

GCM: need to capture the SST forcing and associated large-scale circulation pattern

- RCM: need to capture the regional precipitation variability
- Expectations for dynamically downscaled output:
 - Capture Pacific forced interannual variability, which must be seen in the dominant regional precipitation patterns, associated sea surface temperatures and large scale circulation



Typical winter precipitation anomaly during El Nino year

Source: Climate Prediction Center







Fig. 14. Idealized relationship of monsoon ridge position and midlevel moisture transport to Pacific SSTs at monsoon onset.

Early summer teleconnection patterns (late June, early July)

(Castro et al., 2001)



RCM with dynamically downscaled GCM data is able to capture:

- Seasonal precipitation variability (winter and summer)
- Large-scale forcing corresponds to the dominant precipitation pattern
 - ENSO pattern
 - Stationary patterns in the atmospheric circulation both in winter and summer
 - Quasi-geostationary Rossby wave (different driving mechanisms for winter and summer)

The regional model is adding substantial value to the representation of the interannual variability of the driving global model.

Conclusions

- 1. Dynamical Downscaling of HADCM3 data has been finalized. Current working on two more IPCC AR4 models
- 2. Climatological analyses show clear improvements when compared to raw GCM data.
- 3. In the dynamically downscaled WRF HADCM3 interannual variability is well captured for both summer and winter seasons.

Also did not have time to discuss warm season seasonal forecasts—but similar promising results!