Linking Atmospheric and Watershed Models

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Statistical DownScaling (SDS) Dynamical DownScaling (DDS)

Day 0-8 Statistical DownScaling (SDS) Ensemble Streamflow Prediction (ESP)



Day 0

Linking atmospheric and watershed models

DAY 0

Global-scale model –

National Centers for Environmental Prediction/National Center for Atmospheric Research Reanalysis (NCEP)



NCEP







January Air Temperature Anomalies

July Precipitation Anomalies





Linking atmospheric and watershed models

Compare SDS and DDS output by using it to drive a distributed hydrologic model





Hydrologic Model

Precipitation Runoff Modeling System (PRMS)

[distributed –parameter, physicallybased watershed model]

Implemented in:

The Modular Modeling System (MMS)

[A set of modeling tools to enable a user to selectively couple the most appropriate algorithms]



Hydrologic Model -- PRMS

- Distributed capabilities provided by partitioning watershed into Modeling Response Units

- Basin and MRU delineation, characterization and parameterization done using the GIS Weasel

- No parameter calibration performed on GIS weasel generated parameters

 Calibration focused on WB parameters affecting PET and precipitation distribution

- Other model parameters based on parameter sets from comparable basins

Hydrologic Model -- PRMS

Need to be able to distribute from a station or a grid point to each MRU

Input Data Sets:

- 1. Station Data
- 2. NCEP
- 3. SDS
- 4. DDS



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XYZ Methodology

Distributes a single mean value from a group of stations (or a model grid node) to each modeling unit within a basin.



XYZ Methodology One predictor (Z) example for predicting daily PRCP Mean station elevation (Z) For each day solve for y-intercept 1. VS. mean station PRCP $intercept = PRCP_{sta} - slope^*Z_{sta}$ where **PRCP**_{sta} is mean station PRCP and Z_{sta} is mean station elevation intercept slope is monthly value from MLRs Slope from MLR

2. $PRCP_{mru} = slope^*Z_{mru} + intercept$ where $PRCP_{mru}$ is PRCP for your modeling response unit Z_{mru} is mean elevation of your modeling response unit

Hydrologic Model -- PRMS

Input Data Sets:

- 1. Station Data
- 2. NCEP
- 3. SDS

4. DDS







Snowmelt **Dominated**

Snowmelt Dominated

792km²

inated

922km



Rainfall **Dominated**

3626km²



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DDS

52 km grid node spacing





DDS





Hydrologic Model -- PRMS

Station Input Data Sets:

- Stations used to calibrate PRMS (Best-Sta)
 - All Stations within RegCM2 buffered area (excluding Best-Sta)





All-Sta, DDS, and NCEP need a bias correction



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Linking atmospheric and watershed models DAY 0-8

Use PRMS to produce 9-day forecasts of runoff using SDS and ESP

- 100 SDS ensembles





Snowmelt Dominated

Compare ESP and SDS 9-day forecasts of runoff every 5 days

Rainfall

Dominated

BASINS

Snowmelt Dominated

792km²

Snowmelt Dominated

526km²

922km²

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3626km²

Monthly Mean Precipitation (measured and SDS)



Monthly Mean Maximum Temperature (measured and SDS)







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Month



OBSERVED DISCHARGE AUTOCORRELATION

- Day+8 correlation
 - Alapaha = 0.75
 - Animas = 0.84
 - Carson = 0.78
 - Cle Elum = 0.51

Observed Discharge Autocorrelation





Compare SDS and ESP Forecasts Perfect Model Scenario

-Nash Sutcliffe Goodness of Fit

- Measure of deterministic forecast skill

-Ranked Probability Score

-Measure of probabilistic forecast skill

-Forecasts are increasingly penalized as more probability is assigned to event categories further removed from the actual outcome

-Ensemble Spread

- Range in forecasts



Nash Sutcliffe Goodness of Fit

Measure of deterministic forecast skill



Ranked Probability Score

Measure of probabilistic forecast skill





Ensemble Spread

Range in forecasts









ESP

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 Super-ensemble approach to watershed modeling

Run hydrologic models in ensemble mode to provide probablistic forecasts of streamflow and estimates of forecast uncertainty

Super Ensemble Approach to Watershed Modeling













 Super-ensemble approach to watershed modeling

Run hydrologic models in ensemble mode to provide probablistic forecasts of streamflow and estimates of forecast uncertainty

 Physically based watershed model that needs limited calibration

