

Center for Western Weather and Water Extremes SCRIPPS INSTITUTION OF OCEANOGRAPHY

How do Spectrally Vast AR Thwart Attempts to Skillfully Forecast their Continental Precipitation?

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Modeling and Methods Session 1

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Outline

- Why do the Largest AR Scales Present a Problem to Limited Area Models?
- Results from Storm-Scale Configuration Tests
- Forecast Vapor Transport Structure Errors
- Why do the Scales Responsible for "Local Precipitation Response" Present an Additional Problem?
- Investigating Simulated Local Response in 2 NWP Systems
- Can Observed Linear Orographic Response Suggest Where to Invest Model Improvement?



AR Storm Scales Present a Problem for LAM





**Spectral Analysis from Miguez-Macho et al., J. Geophys Res. 2004

Models Investigated



44°N - 42°N - 40°N - 38°N - 38°N - 34°N - 316°N - 316°

Selected Physics:

- YSU BL
- Noah LSM
- MYJ Sfc
- Thomspon New MP
- Goddard SW
- RRTM LW
- GD 3D Cumulus (9 km only)

15 Oceanic AR and 10 Landfalling AR were simulated with 2 WRF configurations, both driven by GEFS 9.0.1 CTL member reforecast (GFSRe). Forecasts were run up to 7 days lead time with 24 hr lag.

West-WRF: 9 km / 3 km by 1-way nesting, 60 vertical levels, topographic wind correction.

WRF-ARWS: Identical to West-WRF except outer domain extent is smaller. Nested domains (used for precipitation verification) are identical.



West-WRF Verification Methods 1



West-WRF Domain (9km Resolution)

Location of CalWater dropsondes used in study of forecast errors

15 Calwater flights completed AR Core transects during CalWater 2014 / 2015.

All transects crossed a moderate strength (IVT > 500 kg m⁻¹ s⁻¹) core and were more than 1° from model boundaries.

Use these observations to investigate forecast accuracy at storm scales ($\Delta x > 80$ km)



Performance in Capturing Structure



Distance from Dropsonde Analyzed AR Core (km)

Left to Right: Partial IVT contours (kg m⁻² s⁻¹ - black) and θ_e (K - blue dashed) from Observations, GFSRe, West-WRF, WRF-ARWS

In panels showing model Partial IVT, quantity is model – obs. Negative contours are dotted.

AR Precip. at Local Scale a Challenge for LAM and GCM







**Idealized Wavenumber Spectra from Skamarock, Mon. Wea. Rev., 2004.

Direct Measurement of the Storm-Local Scale Relationship



Storm-total BUF (IVT Proxy, X Axis) is strong predictor of Storm-total rainfall (Y axis).

- BUF is determined primarily by storm scales (dx > 50 km)
- Rainfall response to BUF (slope, intercept, R²) is controlled by local dx < 10 km scales.
- Model type determines whether BC, dynamics or physics most influences the error in the response.
- Least Squares can be used to derive a linear model Y = F(X) for observations and forecasts.



**Scatterplot from Ralph et al., J. Hydromet., 2013.

West-WRF Validation Methods 2

Start Date	Start Time	Duration	Special Obs
12/10/2014	1500	32	
02/06/2015	0400	27	CalWater IOP
02/08/2015	0900	25	CalWater IOP
12/09/2015	1300	26	
12/20/2015	1400	47	
01/17/2016	0400	25	
01/28/2016	1700	32	
03/05/2016	2200	33	FIRO Soundings
03/09/2016	0800	42	FIRO Soundings
03/12/2016	1500	37	

10 "Moderate" or stronger AR that made landfall in the Russian River Watershed are used to build a database for verification.

The NOAA Coastal ARO is used to investigate forcing and precipitation response.

Forecasts are generated for lead times up to 7 days every 24 hr.

West-WRF, WRF-ARWS GFSRe is used for comparison



Which model simulates the forcing-response relationship at local scale?



Normalized 2-dim. Error (e_{xv}): ARO - Model

Lead Time	West-WRF	GFS
12 - 59	1.285	12.68
60 - 107	3.524	10.50
108 - 155	9.275	18.33

 $e_{xy} = \frac{E\left[\left(X - X_0\right)^2 + \left(Y - Y_o\right)^2\right]}{2\sqrt{V[X_0]V[Y_0]}}$



Linearizing the Response Relationship

Normalized Error in ST Precip:

 $e_{y} = \frac{E\left[\left(f\left(X\right) - Y_{o}\right)^{2}\right]}{V[Y_{o}]}$

If $F_o($); F() derived from linear LS fit obs. and modeled BUF-Prcp at ARO, then

$$\mathcal{C}e_{ypr} = \frac{E\left[\left(F_{o}\left(X\right) - Y_{0}\right)^{2}\right]}{\mathcal{C}e_{y}} - 1$$
$$\mathcal{C}e_{ypf} = \frac{E\left[\left(F\left(X_{o}\right) - Y_{0}\right)^{2}\right]}{\mathcal{C}e_{y}} - 1$$

are the reduction in forecast ST Precip. by "perfect" local response (pr) and "perfect" storm scale forcing, respectively.

Error Measure		Forecast Lead Time (hr)			
		12 – 59	60 – 107	108 - 155	
e _y	West-WRF	0.442	1.224	2.092	
	GFSRe	15.753	24.094	41.176	
$e_{\it ypf}$	West-WRF		-60.7%	-79.7%	
	GFSRe	-33.7%	-35.3%	-25.0%	
e _{ypr}	West-WRF		-43.4%	-51.4%	
	GFSRe	-95.3%	-94.4%	-86.8%	



Summary and Conclusions

- Forecasting AR Requires Special Attention in Constructing LAM Domain.
- LAM (West-WRF) are able to forecast AR as accurately as GNWP at large scales up to 7 day lead times.
- If Storm-Scale forcing is as accurate, the reduction in dynamics and representativeness errors in high-res LAM offer big improvement in local scale precip. forecasting
- The dominance of local response relationship in driving GFSRe precip. errors was verified by a linearized model
- Linearized model demonstrated that West-WRF can be tuned for better precip. forecasts as well, at both local and storm scales.



NWP Performance at Storm Scales



Only Sondes for which IVT \ge 250 kg m⁻¹ s⁻¹ used to compute BSS



NWP Performance at Storm Scales





QPF Deterministic Skill During Landfalling AR



Validated Against NCEP Stage-IV 24 hr QPE. Models Linearly Interpolated to Stage IV Grid.

