# Study of a dynamic downscaling scheme for quantitative precipitation forecasting

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Abstract Dynamic downscaling is the best way to get high resolution quantitative precipitation forecasts. The Weather Research and Forecasting Model (WRF) was utilized as a dynamic downscaling tool in this study. The influence of the horizontal resolutions and the model domains on precipitation forecasts has been analysed to establish an optimized dynamic downscaling scheme. Three precipitation events over Xijiang basin, China, were simulated with different horizontal resolutions and domains. The results indicate that both the horizontal resolution and model domain have an influence on the precipitation forecast. However, the correlation between high precipitation forecast accuracy and the high horizontal resolution or the large model domain were not very strong. Comprehensive consideration of the results shows that the accuracy of forecast is best when the horizontal resolution is 20-km. Although the model domain size has no significant influence on the precipitation forecast, a larger domain may improve the stability of forecasting.

Key words WRF model; dynamic downscaling; horizontal resolution; model domain; Xijiang Basin

## INTRODUCTION

China is located in the eastern part of Asia and has complicated topography, and its climate varies greatly from region to region. The South of China is greatly affected by the monsoon climate and tropical cyclones, and flood disasters take place frequently. Flood forecasting, as one of the non-structural flood protection measures, is capable of reducing the risk of flood disaster effectively. Precipitation is the most important atmospheric input to hydrological models, and is always acquired by gauge-based precipitation measurements. However, utilizing observed precipitation data as input to hydrological models would limit the flood lead time. Compared with traditional methods (Lu *et al.*, 2006), using quantitative precipitation forecasts with a certain accuracy as input can be an effective way to provide longer flood lead times.

Although the global numerical weather prediction models have skills in predicting large-scale circulation features, meso- and micro-scale weather systems, which are the major factor for the occurrence of rainstorms, cannot be predicted effectively. Also, existing global models typically run at a coarse resolution due to the limitations of computing resources. In order to meet the requirements of flood forecasting, the downscaling methods should be utilized to increase the horizontal resolution. Downscaling techniques that have been widely used are the dynamic downscaling method and statistical downscaling method. Compared with the statistical downscaling method, dynamic downscaling techniques are more physically based. The customary practice of dynamic downscaling method is to utilize mesoscale regional atmospheric models within a global model to simulate local climate. Atmospheric dynamic processes and physical processes have been taken into consideration (Li *et al.*, 2009). Regional atmospheric models run at finer horizontal resolution than global-scale models, and thus provide a more accurate output with higher temporal and spatial resolution.

The Weather Research and Forecasting (WRF) model is a mesoscale numerical weather prediction (NWP) and atmospheric simulation system. The WRF dynamics solver integrates the compressible, non-hydrostatic Euler equations. Geographical position can be easily located and multiple horizontal resolutions can be set as well. Research has indicated that utilizing the WRF model as a dynamic downscaling tool with a suitable dynamic downscaling scheme is capable of forecasting precipitation with higher horizontal resolution for different regions in China (Wang *et al.*, 2008; Wei *et al.*, 2011). Both the horizontal resolution and domain size have a significant influence on the accuracy of weather forecasts. So it is of great importance to set proper

resolutions and domains when the WRF model is implemented. The Xijiang River is the major tributary of the Pearl River, South China. However, application of the WRF model over Xijiang basin has been investigated less. We implemented the WRF model as a dynamic downscaling tool to analyse the influence of horizontal resolution and domain size on the simulations over Xijiang basin in this paper. An optimized dynamic downscaling scheme for Xijiang basin was established. Higher accuracy of quantitative precipitation forecasting with higher temporal and spatial resolutions for Xijiang basin are provided as input to hydrological models.

#### DATA AND METHODOLOGY

#### Data

The output of the Global Forecast System (GFS) model is used as input to the WRF model in this paper. The global forecast data we have used are at a horizontal resolution of  $1^{\circ} \times 1^{\circ}$  with a frequency of 6 hourly. The output of different downscaling schemes has been validated by using automatic precipitation stations over Xijiang basin.

Several rainstorms were selected according to the average precipitation amount over Xijiang basin. The occurrence time of these rainstorms are shown in Table 1.

**Table 1** Occurrence time of selected rainstorms over Xijiang basin

Rainstorm	Occurrence time	
Storm 1	5–6 June 2005	
Storm 2	21–22 June 2005	
Storm 3	12–13 June 2008	

## Methodology

The WRF model is used to simulate rainstorms over Xijiang basin. In this study, all domains are centred at 23.8N, 109.2W. All runs utilize the same physical functions on a Lambert conformal projection. The vertical structure of the model includes 28 layers covering the whole troposphere. The WRF single-moment 3-class microphysics parameterization (Hong and Lim, 2006) and the Kain-Fritsch cumulus parameterization (Kain, 2004) which are related to the precipitation simulation, have being used on all domains for all runs. Other physical options are left as the recommended default settings. Simulations were conducted for three periods corresponding to the occurrence of three selected typical rainstorms: from 0000 UTC 5th June 2005 to 0000 UTC 6th June 2005 (Storm 1); from 0000 UTC 21th June 2005 to 0000 UTC 22th June 2005 (Storm 2); from 0000 UTC 12th June 2008 to 0000 UTC 13th June 2008 (Storm 3).

#### Experimental design of horizontal resolution sensitivity

To analyse the influence of horizontal resolution on the precipitation forecast, the domain is set to be the same with different horizontal resolutions. The number of grids at different resolutions is shown in Table 2.

Table 2 The designations of simulations with different horizontal resolutions

Horizontal resolution	Number of grids	
30 km	50×36	
25 km	60×44	
20 km	77×54	
15 km	100×72	
10 km	150×108	
5 km	300×216	

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## Experimental design of model domain sensitivity

The horizontal resolution was set as 20 km for the test. Four different domain sizes were utilized to simulate the selected rainstorms (Fig. 1). The Xijiang basin is included under different domain sizes from D01 to D04.

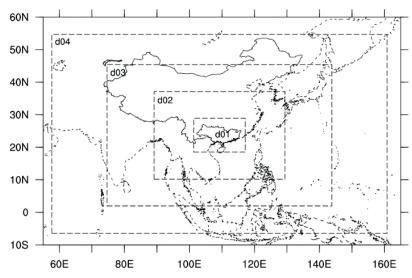


Fig. 1 The designations of simulations with different domain sizes.

## **Evaluation methodology**

Evaluation of the WRF sensitivity experiments was performed using a variety of statistical verification techniques. To facilitate comparisons, the output of WRF was bilinearly interpolated into the automatic precipitation stations over Xijiang basin. The quantitative accuracy of simulations is assessed by the mean error (ME), root mean square error (RMSE) and correlation coefficient (R), which are defined as:

$$ME = \frac{1}{N} \sum_{i=1}^{N} (p_i - o_i)$$
 (1)

$$RMSE = \left[\frac{1}{N}\sum_{i=1}^{N}(p_i - o_i)^2\right]^{1/2}$$
 (2)

$$R = \frac{\sum_{i=1}^{N} (p_i - \bar{p})(o_i - \bar{o})}{\left[\sum_{i=1}^{N} (p_i - \bar{p})^2 \cdot \sum_{i=1}^{N} (o_i - \bar{o})^2\right]^{1/2}}$$
(3)

where N is the total number of verification grid point, p and o are the model simulation and observed values, and  $\bar{p}$  and  $\bar{o}$  are the mean forecasted and observed precipitation.

To verify the accuracy of the spatial distribution of precipitation forecasting at different precipitation thresholds over Xijiang basin, the threat score (*TS*) was computed. *TS* is used to measure the skill of predicting the area of precipitation for a certain threshold. *TS* is defined as:

$$TS = \frac{a}{a+b+c} \tag{4}$$

where a is the number of points where both the forecasted and observed amount of precipitation are in a certain threshold, b is the number of points where the forecasted amount of precipitation is beyond the threshold while the observed is under the threshold, c is the number of points where forecasted amount of precipitation is under the threshold while the observed is beyond the threshold.

Daily rainfall rates at stations are classified into five grades of intensity for the computation of threat scores: light rain (0.1 mm/d  $\leq R \leq$  9.9 mm/d), moderate rain (10 mm/d  $\leq R \leq$  24.9 mm/d), large rain (25 mm/d  $\leq R \leq$  49.9 mm/d), heavy rain (50 mm/d  $\leq R \leq$  99.9 mm/d) and hard rain (R > 100 mm/d).

#### RESULTS AND DISCUSSION

#### Influence of horizontal resolution on simulations

Figure 2 indicates that all simulations at different horizontal resolutions are capable of forecasting this precipitation process. However, the location and intensity of the rainstorm centre are not the same when the horizontal resolutions are different. When the horizontal resolution is 30 km, the simulated intensity of the rainstorm centre is much lower than observed. The intensity and area of the rainstorm centre increases as the horizontal resolution increases. The simulated location of the rainstorm centre also gets much closer to observed. The simulation is best when the horizontal resolution is 20 km. Further increase of the horizontal resolution leads to the occurrence of false rainstorm centres.

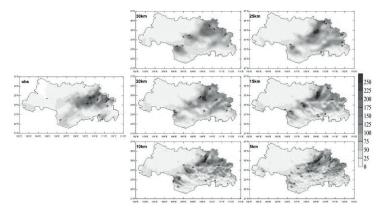


Fig. 2 Comparison of observed and simulated precipitation distribution at different horizontal resolution for Storm 3

Table 3 presents the threat scores of simulations with different horizontal resolutions and indicates that no obvious tendency has been observed as the horizontal resolution increases. The influence of horizontal resolution is different for each selected rainstorm at different thresholds.

For large rain, the TS is the best when the horizontal resolution is 30 km for the simulation of Storm 1 and Storm 2. The best simulation of Storm 3 occurs at the large rain threshold when the horizontal resolution is 5 km.

The result of TS at thresholds  $\geq 50$  mm is totally different from the results at the large rain threshold. TS increases as the horizontal resolution gets higher. The highest TS occurs when the horizontal resolution is 10 km for Storm 1 and 15 km for Storm 2.

The TS of Storm 3 at heavy rain threshold is best when the horizontal resolution is 30 km, and worst when the horizontal resolution is 10 km. The highest TS of Storm 3 at a threshold  $\geq$ 100 mm appears at a horizontal resolution of 20 km.

It is obvious that the horizontal resolution has a significant influence on simulations. According to the results, the TS at different thresholds is more stable when the horizontal resolution is 20 km.

Rainstorm	Threshold	Horizontal resolution						
		30km	25km	20km	15km	10km	5km	
Storm 1	25~49.9mm	0.533	0.467	0.2	0.13	0.133	0.267	
	≥50mm	0.227	0.545	0.727	0.773	0.864	0.545	
Storm 2	25~49.9mm	0.3	0.15	0.2	0.25	0.15	0.3	
	≥50mm	0.783	0.870	0.870	0.913	0.870	0.670	
Storm 3	25~49.9mm	0.237	0.184	0.263	0.263	0.184	0.342	
	50~99.9mm	0.540	0.428	0.444	0.381	0.210	0.333	
	>100mm	0.564	0.340	0.723	0.606	0.649	0.660	

Table 3 Threat Scores of each simulation at different horizontal resolution

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#### Influence of domain size on simulations

The simulated distribution of precipitation has been compared with the observed precipitation distribution for Storm 3 (Fig. 3). The intensity decreases from north-east to south-west for all runs, which the tendency is the same as we have observed. When the domain is D01, the intensity of rainstorm centre is lower than observed. The intensity, location and area of rainstorm centre are almost the same when the domain is D02, D03 and D04.

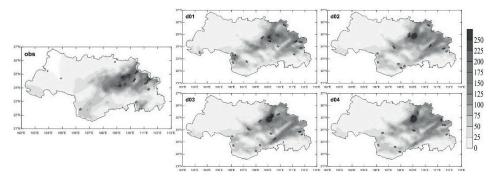


Fig. 3 Comparison of observed and simulated precipitation at different domain size for Storm 3.

According to the results shown in Table 4, domain size also has an influence on simulation. When the domain size is D01, the simulations are always the worst. The mean error of Storm 2 is 7.95, almost twice higher than the results of other domains. The mean error of Storm 3 is also much higher than for the other domains. *R* of Storm 2 and Storm 3 is also lower than others.

When domain size increases to D02, the simulations are not stable. The results are strongly affected by the storm we have selected. The ME of Storm 1 with D02 is -8.6. However, the ME of Storm 2 is 3.48 and for Storm 3 is -1.88.

The results of D03 and D04 are almost the same. Although a larger domain size would not raise forecast accuracy effectively, a larger domain may improve stability of forecast.

Rainstorm	Evaluation method	Domain size				
		D01	D02	D03	D04	
	ME	-6.54	-8.6	-4.9	-4.3	
Storm 1	RMSE	31.3	31.4	32.5	32.6	
	R	0.27	0.34	0.31	0.34	
	ME	7.95	3.48	3.64	4.45	
Storm 2	RMSE	44.0	40.4	40.2	40.6	
	R	0.36	0.41	0.41	0.41	
Storm 3	ME	-8.3	-1.88	-0.59	-2.7	
	RMSE	54.4	57.2	61.0	57.1	
	R	0.64	0.58	0.50	0.58	

Table 4 Statistical evaluation of simulations at different domain size

#### DISCUSSION

From the results, it is obvious that higher horizontal resolution does not lead to higher accuracy. A similar conclusion was found by Shi *et al.* (2012) utilizing WRF to analysis the influence of horizontal resolution (27 km, 9 km) on precipitation and temperature forecasts; the results were better when the horizontal resolution was 27 km. The best horizontal resolution is 20 km for Xijiang basin in our study. Chen and Wang (2006) also found that WRF with a horizontal resolution of 20 km can simulate rainstorms well in Huaihe River basin, indicating that WRF was capable of forecasting precipitation at such a resolution too. When the horizontal resolution is higher than 20 km, we found that the magnitude of precipitation forecast is higher than observed and the results tend to be chaotic. Deng *et al.* (2012) pointed out that higher horizontal resolution

always leads to a higher false alarm rate, when they compared the results of WRF with different horizontal resolutions (45, 30, 20 and 10 km) on simulating rainstorms in Huaihe River Basin.

The domain size also has an influence on the accuracy of precipitation forecast. We found the best domain size includes most parts of East Asia, and takes topography into consideration. The best domain for Xijiang basin in this research is  $77 \times 54$  points at a horizontal resolution of 20 km. Such a large domain is able to simulate large-scale weather systems and the results are stable. Olsson *et al.* (2013) also pointed out that a large domain can capture the larger scales of the sub-synoptic forcing. However, a larger domain always lead to higher computing time. So, we should balance the model accuracy and efficiency when choosing a simulation domain for the WFR model.

### **CONCLUSION**

In this study, the WRF model was utilized as a dynamic downscaling tool to analyse the influence of horizontal resolution and domain size on precipitation forecasting over Xijiang basin. Several typical rainstorms were simulated under different downscaling schemes. An optimized dynamic downscaling scheme has been established. The conclusions are drawn as follows:

- (a) Higher horizontal resolution may improve the spatial distribution forecast of rainstorms. However, a threshold of horizontal resolution exists. Continuous raise of horizontal resolution may lead to lower accuracy of precipitation forecast. The threshold of WRF model is 20-km over Xijiang basin.
- (b) Topography should be considered. Although D02 is larger than D01, the simulations are more chaotic for the left margin is close to Qinghai-Tibet. The results are almost the same when the domain is D03 and D04. But larger domain size always lead to longer computing time. D03 is more likely to be used on simulations.
- (c) Overall, an optimized downscaling scheme is established, which the horizontal resolution is 20-km and domain size is D03. This optimized scheme has a significant guiding meaning and practical application value for the quantitative precipitation forecast and flood prevention over Xijiang basin.

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