

# WeeklyNote

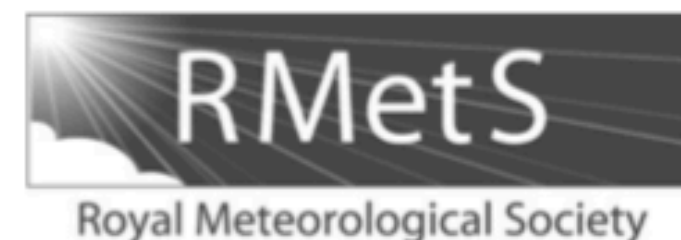
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# **Statistical downscaling of general circulation model outputs to precipitation – part 1: calibration and validation**

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# Results

1. 地表降水速率被认为是研究降水的最具代表的变量（列表中除7月不包含，其余月份均包含）；
2. 湿度变量（相对湿度和比湿）是大气水汽的代表性变量（12个月中有7个月均包含[February, March, May, September, October, November, December]）；
3. 在7月中，被选中的变量仅有风速和850hPa地势高度——这表明这些变量在一定程度上也可解释降水过程；
4. 表格中被选中的格点基本围绕{4, 4}；因此在选择变量时，不仅要考虑其和观测降水的相关性还需考虑时间尺度上的相关性。

Table 2. Final sets of potential predictors for each calendar month.

Month	Potential variables used in the model with grid locations
January	Surface precipitation rate {(3,3),(4,4)} 1000 hPa specific humidity {(3,3),(3,4),(4,4)} 850hPa meridional wind {(2,6),(3,5),(3,6)} 850hPa relative humidity {(1,2)} 2 m specific humidity {(3,3),(3,4)}
February	Surface precipitation rate {(3,4),(4,4),(4,5)}
March	Surface precipitation rate {(3,3),(3,4),(3,5),(4,3),(4,4),(4,5),(4,6)}
April	850hPa relative humidity {(4,3),(4,4)} Surface precipitation rate {(4,3)}
May	Surface precipitation rate {(4,4),(5,5)} 850hPa geopotential height {(4,3)}
June	Surface precipitation rate {(3,2),(3,3),(4,2),(4,3),(4,4),(4,5)} Mean sea level pressure {(4,3),(5,3)} 850hPa zonal wind {(2,4)} Surface pressure{(4,3),(5,3),(5,4)}
July	850hPa zonal wind {(1,3),(1,4)} 850hPa geopotential height {(4,3),(4,4),(4,5)}
August	Surface precipitation rate {(4,3),(5,4),(5,5)}
September	Surface precipitation rate {(2,1),(2,2),(3,2),(3,3),(3,5),(4,2),(4,3),(4,4),(4,5)} 850hPa relative humidity {(3,3)} 700hPa relative humidity {(3,4)}
October	Surface precipitation rate {(3,2),(4,2),(4,3),(4,4)} 850hPa relative humidity {(4,3)} 700hPa geopotential height {(1,1)}
November	850hPa relative humidity {(3,2),(3,3)} Surface precipitation rate {(4,3),(4,5)}
December	Surface precipitation rate {(2,1),(3,2),(4,3),(4,4),(5,5)} 850hPa relative humidity {(3,2)}

hPa, atmospheric pressure in hectopascal; the locations are given within brackets (see Figure 1).

# Results

1. 该降尺度模型具有捕捉月降水特征和强度的能力。
2. 在降水较小时被高估，在降水高值区域则又被低估。

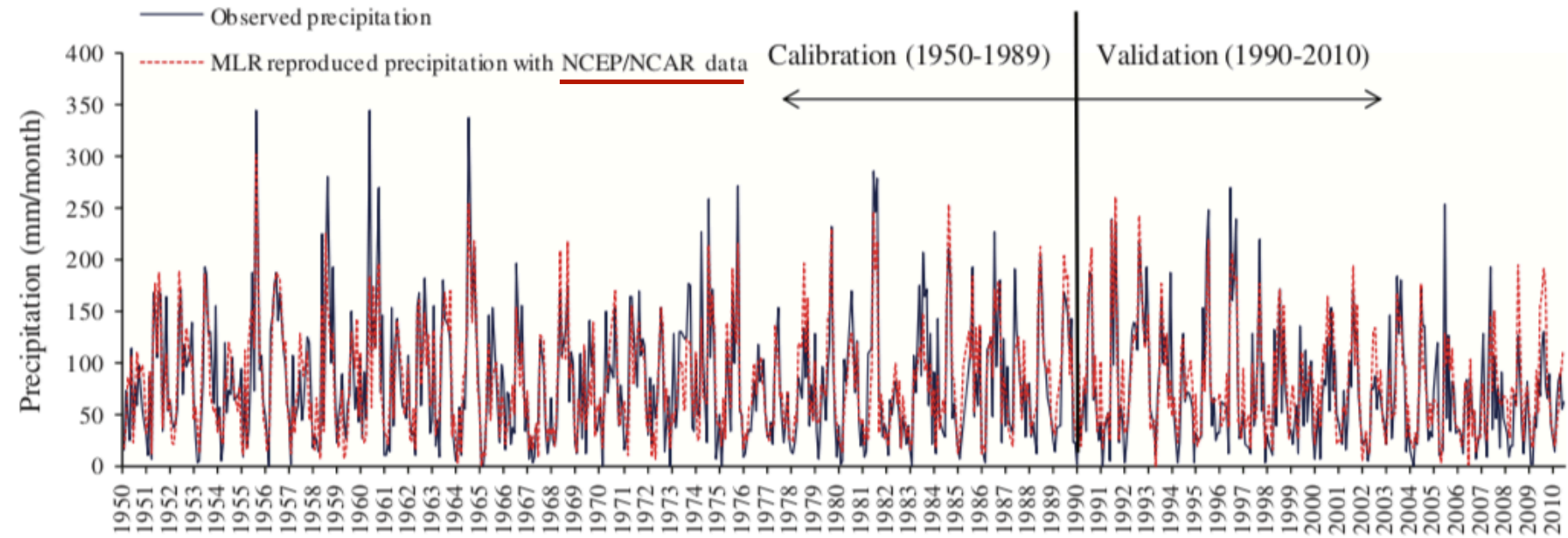
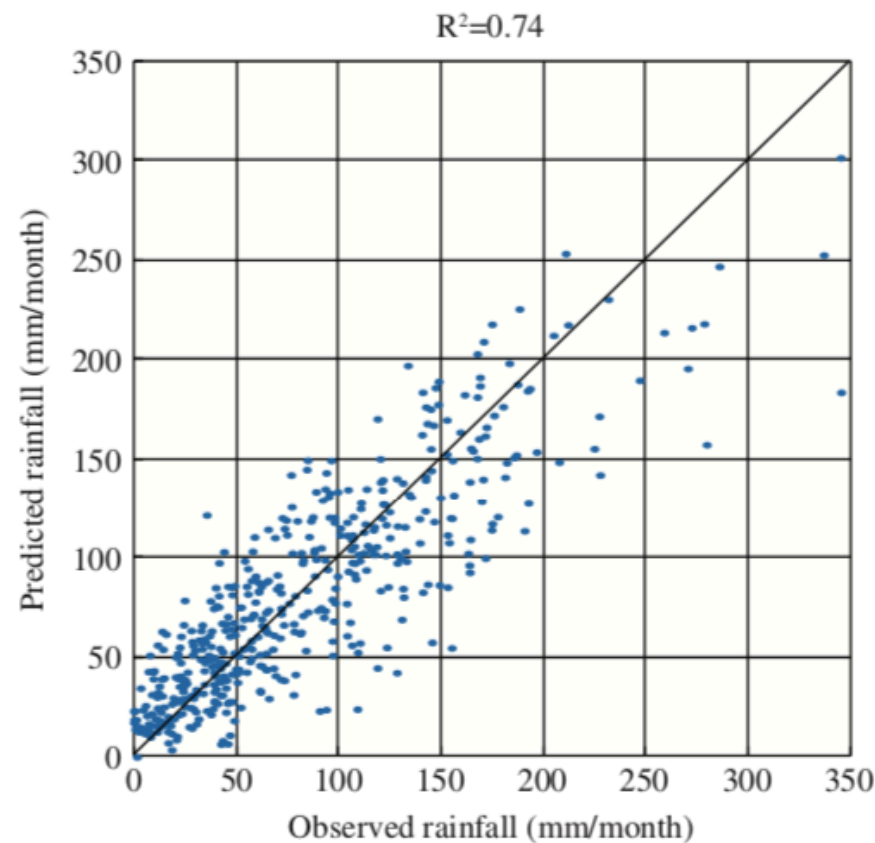
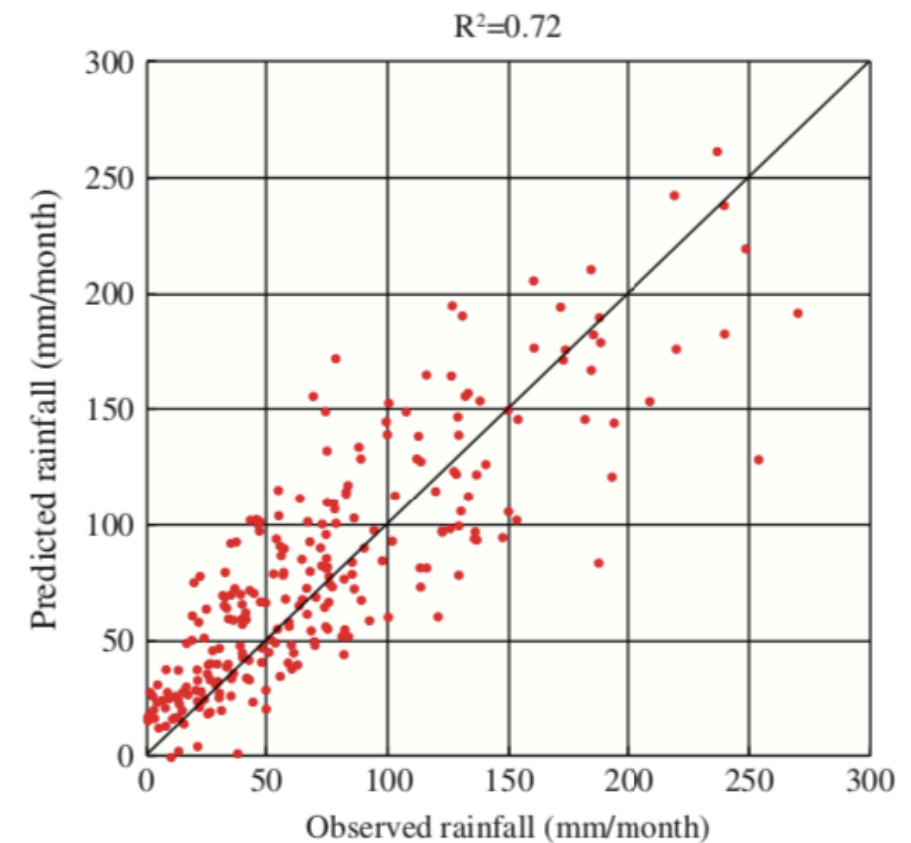


Figure 2. Observed and Model<sub>(NCEP/NCAR)</sub> reproduced monthly precipitation (1950 to 2010).



(a) Calibration



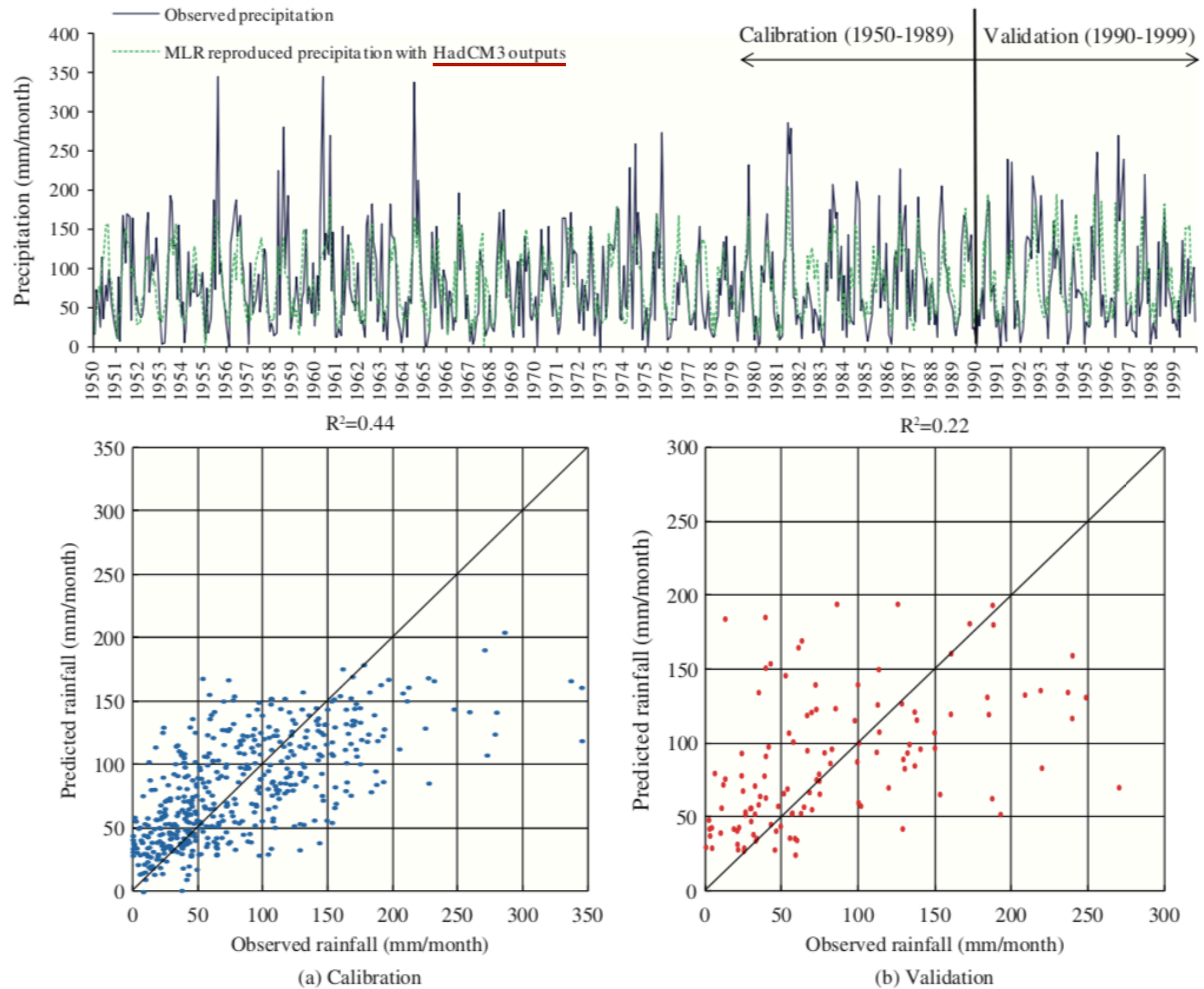
(b) Validation

Figure 3. Scatter plots of observed and Model<sub>(NCEP/NCAR)</sub> reproduced monthly precipitation for calibration (1950–1989) and validation (1990–2010).



# Results

1. 该降尺度模型并未较好的重现降水的高值部分；
2. 并且，该模型对观测数据的再现不如使用NCEP/NCAR数据作为输出的模型好；
3. 然而，该模型还是可以大致捕捉了观测数据的降水特征；



1. 训练集和验证集中，均表现为在降水较小时被高估，在降水高值区域则又被低估；
2. 这是由于降水观测数据的方差要比GCM或再分析资料的方差大得多，因此在降尺度过程中，模型只能观测数据方差的部分，而无法表达高值和低值。

# Results

1. 两个模型在训练集和验证集均对观测的平均值捕捉的很好；
2. 然而两个模型均无法正确捕捉到观测数据的标准差和方差；
3. 相较而言，使用NCEP/NCAR数据作为输出的降尺度模型表现更好。

Table 3. Performances of downscaling models in calibration and validation.

Statistic	Calibration (1950–1989)			Validation (1990–2010)/(1990–1999) <sup>a</sup>			
	Observations	Model <sub>(NCEP/NCAR)</sub>	Model <sub>(HadCM3)</sub>	Observations		Model <sub>(NCEP/NCAR)</sub>	Model <sub>(HadCM3)</sub>
				1990–2010	1990–1999		
Avg	81.8	82.0	81.7	73.3	<b>81.8</b>	81.0	<b>87.6</b>
Std	61.7	53.2	41.1	56.9	<b>64.3</b>	51.9	<b>44.5</b>
$C_v$	0.75	0.65	0.50	0.78	<b>0.79</b>	0.64	<b>0.51</b>
NSE		0.74	0.44			0.70	<b>0.17</b>
SANS		0.66	0.26			0.61	<b>−0.20</b>
$R^2$		0.74	0.44			0.72	<b>0.22</b>

Avg, average of monthly precipitation in mm;  $C_v$ , coefficient of variation; NSE, Nash–Sutcliffe efficiency;  $R^2$ , coefficient of determination; Std, standard deviation of monthly precipitation in mm; SANS, Seasonally Adjusted Nash–Sutcliffe efficiency. <sup>a</sup>Bold italicized values in the table refer to period 1990–1999.

Table 4. Seasonal performances of downscaling models.

Model	Statistic	Calibration (1950–1989)				Validation (1990–2010)/(1990–1999) <sup>a</sup>			
		Season				Season			
		Summer	Autumn	Winter	Spring	Summer	Autumn	Winter	Spring
Observed	Avg	40.7	73.7	125.1	87.7	42.9/( <b>44.3</b> )	54.1/( <b>57.0</b> )	119.4/( <b>136.1</b> )	78.3/( <b>89.8</b> )
Model <sub>(NCEP/NCAR)</sub>		40.7	73.7	125.1	87.7	49.2	57.8	132.5	85.1
Model <sub>(HadCM3)</sub>		40.3	73.8	125.1	87.8	( <b>44.9</b> )	( <b>78.8</b> )	( <b>128.3</b> )	( <b>98.5</b> )
Observed	Std	33.7	58.8	64.5	53.5	41.0/( <b>46.8</b> )	43.1/( <b>46.5</b> )	61.2/( <b>66.3</b> )	48.4/( <b>55.1</b> )
Model <sub>(NCEP/NCAR)</sub>		26.0	46.6	54.1	43.9	29.8	33.1	54.1	41.7
Model <sub>(HadCM3)</sub>		15.6	34.4	26.7	30.5	( <b>12.7</b> )	( <b>39.0</b> )	( <b>30.0</b> )	( <b>42.0</b> )
Observed	$C_v$	0.83	0.80	0.52	0.61	0.96/( <b>1.06</b> )	0.80/( <b>0.82</b> )	0.51/( <b>0.49</b> )	0.62/( <b>0.61</b> )
Model <sub>(NCEP/NCAR)</sub>		0.64	0.63	0.43	0.50	0.61	0.57	0.41	0.49
Model <sub>(HadCM3)</sub>		0.39	0.47	0.21	0.35	( <b>0.28</b> )	( <b>0.49</b> )	( <b>0.23</b> )	( <b>0.43</b> )
Model <sub>(NCEP/NCAR)</sub>	NSE	0.60	0.63	0.70	0.67	0.42	0.75	0.58	0.64
Model <sub>(HadCM3)</sub>		0.16	0.34	0.17	0.33	( <b>0.12</b> )	( <b>−0.58</b> )	( <b>−0.20</b> )	( <b>−0.15</b> )
Model <sub>(NCEP/NCAR)</sub>	$R^2$	0.60	0.63	0.70	0.67	0.45	0.71	0.63	0.65
Model <sub>(HadCM3)</sub>		0.16	0.34	0.17	0.33	( <b>0.13</b> )	( <b>0.04</b> )	( <b>0.00</b> )	( <b>0.09</b> )

Avg, average of monthly precipitation in mm;  $C_v$ , coefficient of variation; NSE, Nash–Sutcliffe efficiency;  $R^2$ , coefficient of determination; Std, standard deviation of monthly precipitation in mm. <sup>a</sup>Bold italicized values in brackets in the table refer to period 1990–1999.

# Results

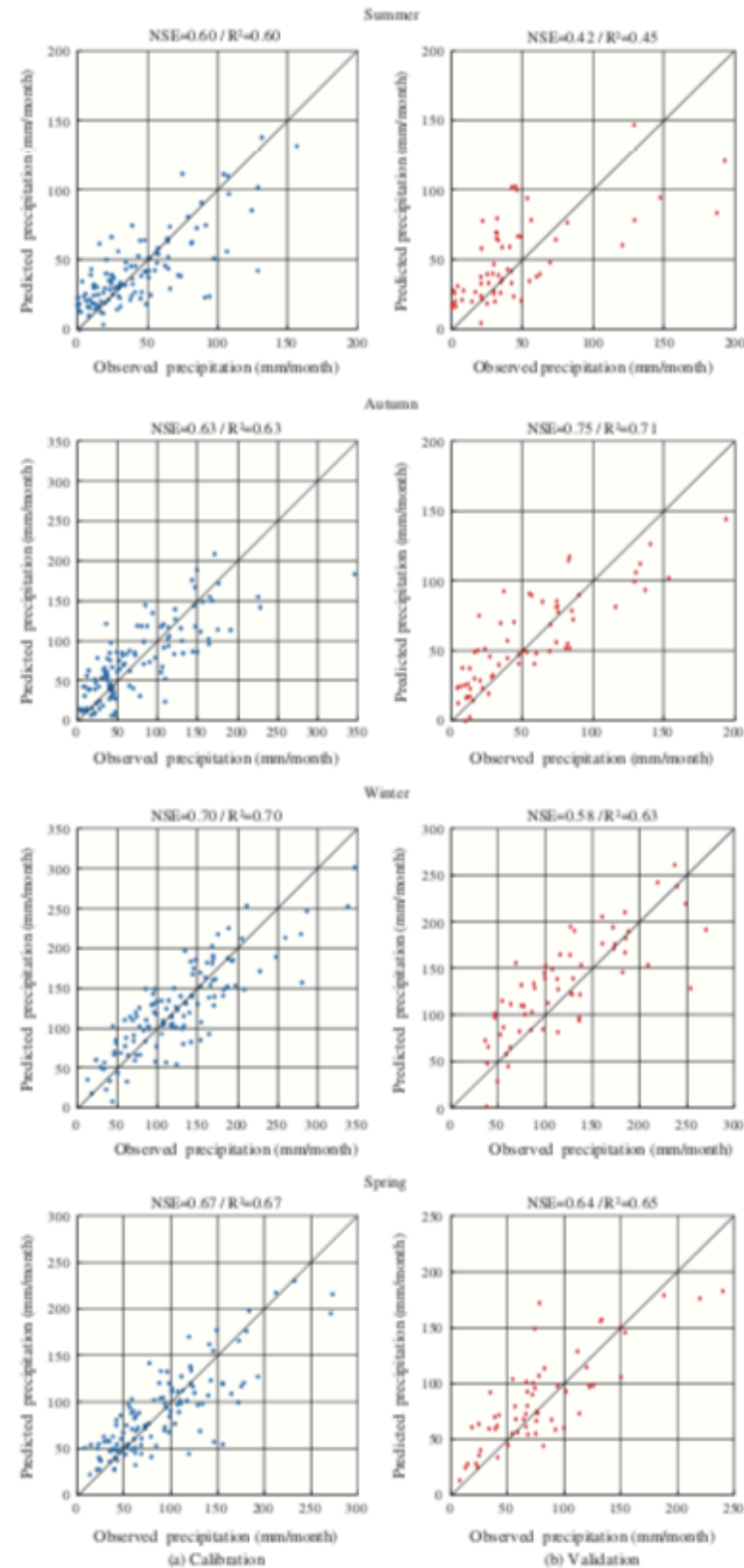


Figure 6. Seasonal scatter plots of observed and Model(NCEP/NCAR) reproduced monthly precipitation for calibration (1950–1989) and validation (1990–2010).

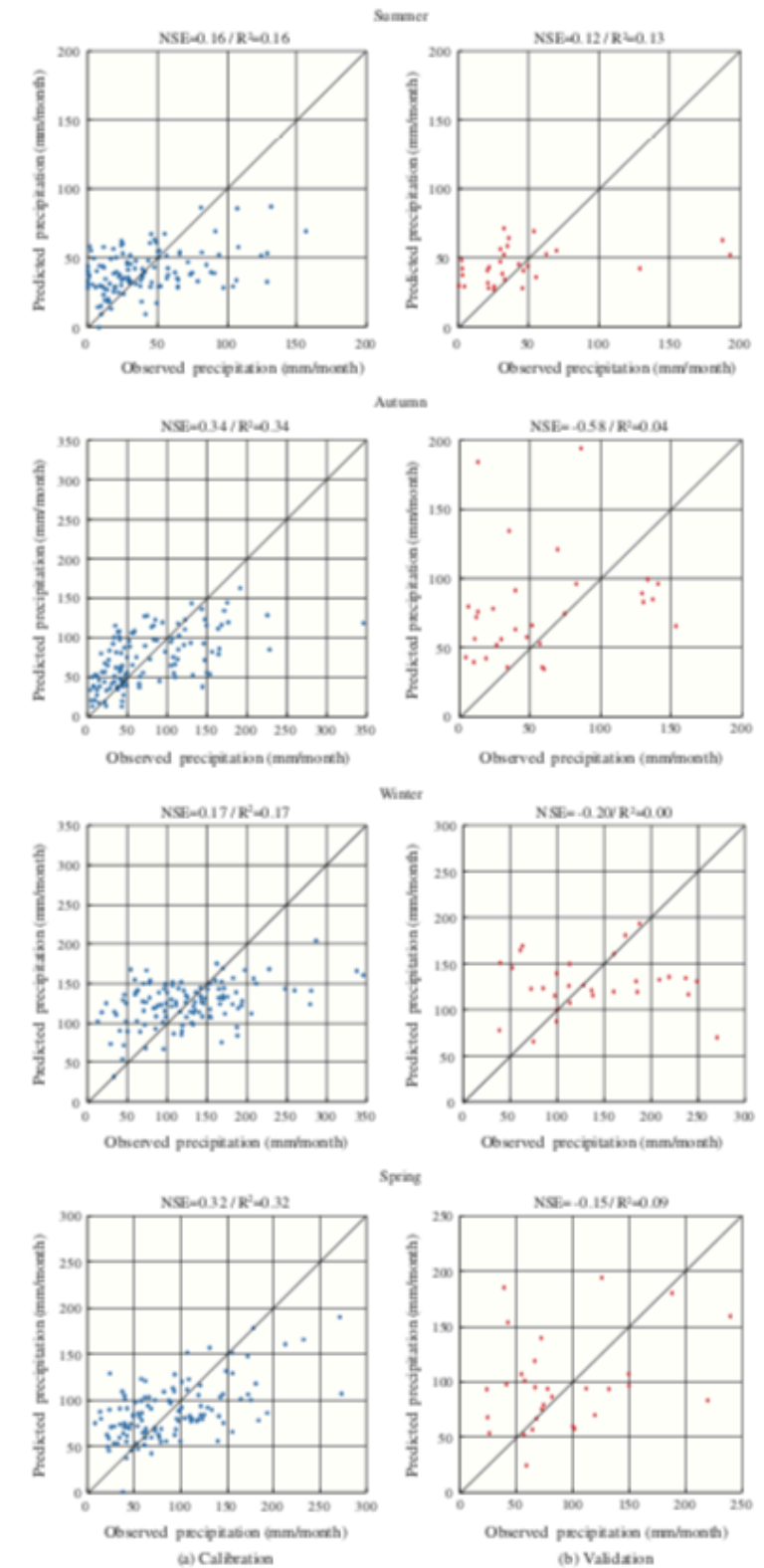


Figure 7. Seasonal scatter plots of observed and Model(HadCM3) reproduced monthly precipitation for calibration (1950–1989) and validation (1990–1999).



# Results

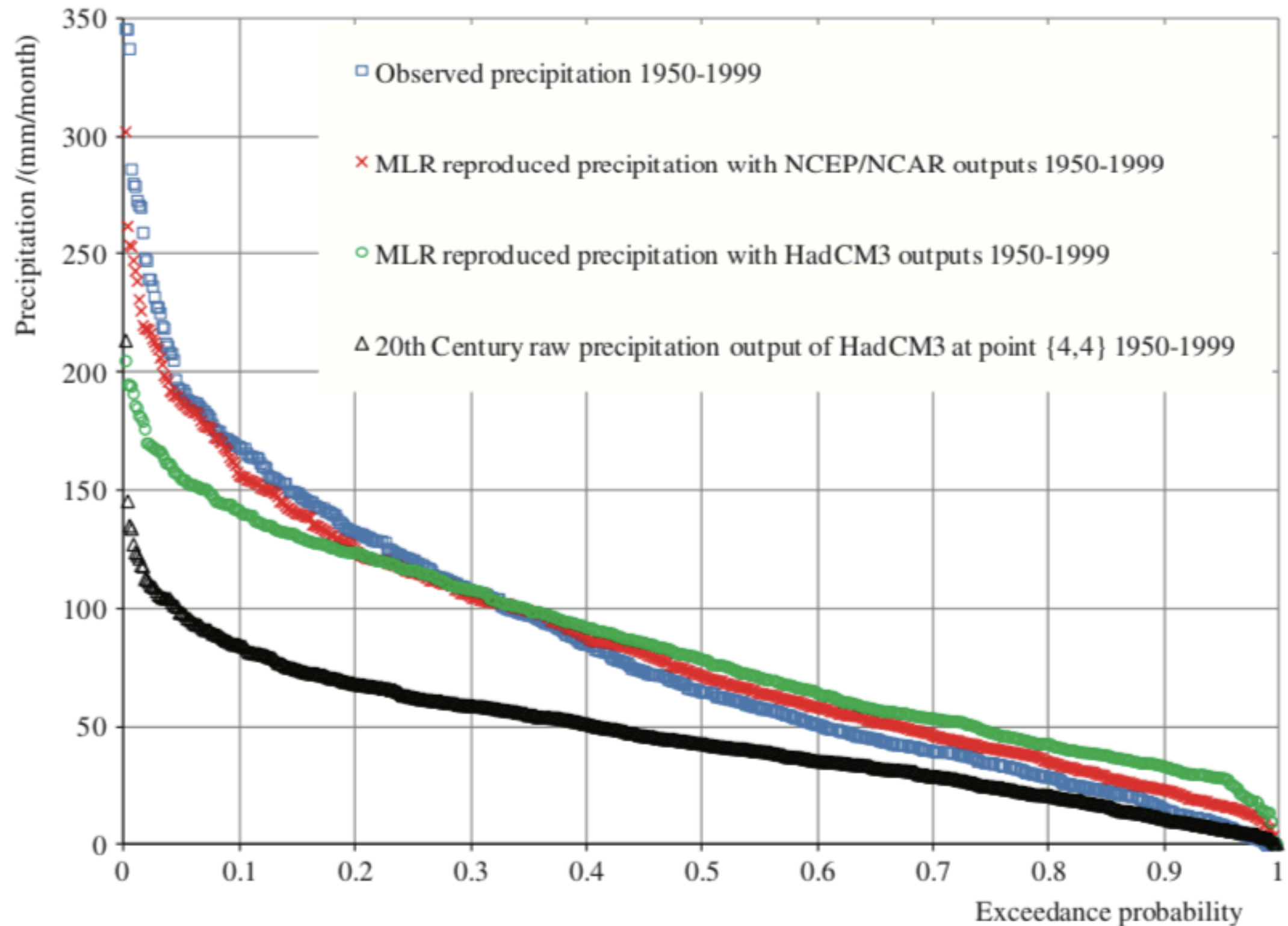


Figure 8. Precipitation probability exceedance curves (1950–1999).



# Conclusions

1. 对于全年的降水，降水速率被认为是最可用于解释观测降水数据的变量（除去7月），湿度、地势高度、平均海平面高度地表压力、和风速都在整个时间段上和观测降水变量有较高的相关性；
2. 使用NCEP/NCAR作为输出的降尺度模型在训练集和验证集上均表现良好，而使用HadCM3的模型表现较差；
3. HadCM3输出和观测数据相差较大(There was a quality mismatch between the NCEP/NCAR reanalysis and HadCM3 outputs, over the period 1950–1999)。

谢谢