

VERIFICATION OF SOIL MOISTURE SIMULATING ACCURACY ON DRY-LAND WINTER WHEAT AND SPRING MAIZE FIELD BY EPIC MODEL ON THE LOESS PLATEAU OF CHINA

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ABSTRACT

EPIC model is an effective tool to simulate soil water which is the key factor to influence the crop production on the Loess Plateau of China. So there is a practical meaning to evaluate the simulation results of soil water. In Loess Plateau amount of rainfall varies greatly through out the year. So it's annually and monthly distribution has a great significance for the development of crop production and recovery of soil moisture in different layers. In this study, the accuracy of simulated monthly soil moisture was assessed, based on measured monthly soil moisture (0-2m soil layer) data from 1987 to 1996 at Changwu Agricultural Station. Results showed that RRMSE (Relative Root Mean Square Error) between simulated and measured soil moisture (0-2m soil layer) in winter wheat field and spring maize field was 2.8% and -0.2%, respectively, the value for RMSE (Root Mean Square Error) was 0.023m/m and 0.015 m/m, respectively. The accuracy of simulated soil moisture influenced by the precipitation amount of simulated years were lower in extreme rainfall years (extreme rainy years and extreme drought years) than in other rainfall years. Therefore, the modified EPIC model predicted well soil water in different soil layers and provided basis for the EPIC users to research the law of soil moisture changes in arid cereal land at Changwu arid-plateau. If reasonable crop database, soil database and meteorology database are built up into the EPIC model, the accuracy of simulated soil moisture will be increased.

Key words: Soil moisture, winter wheat, spring maize, EPIC model.

INTRODUCTION

The Environmental Policy Integrated Climate Model (EPIC, formerly known as the Erosion Productivity Impact Calculator) is one of the predominant crop models which can be used to simulate the process of long-term changing of soil water resource and crop productivity (Williams *et al.*, 1989). However, there is a necessity to justify EPIC model when it is applied to a specific situation which is not same as its birth-place, Black Land Research Center of America (Rosenberg *et al.*, 1992; Mearns *et al.*, 1999; Carbone *et al.*, 2003; Easterling *et al.*, 2003; Liu *et al.*, 2007a, 2007 b).

Few studies found to estimate its simulation accuracy of soil water, though great progress had been made on its ability to simulate soil water. Aiming to evaluate the effect of soil and water resources on crop production, the EPIC model built up at Black Land Research Center and named it Erosion Productivity Impact Calculator (Williams *et al.*, 1989). After it was build up, many efforts had been made to improve its capacity to simulate soil water. Jones *et al.*, (1991) improved the crop growth model, one sub-model of EPIC model, by modifying the growth and distribution of crop's roots, and made its simulation results of root distribution of crop more reasonable. Williams *et al.*, (1995) brought three erosion equations (MUSS, MUST and MUSI) into EPIC model. Renard (1997) took RUSLE equation into EPIC model. These equations brought out a higher accuracy for EPIC model to simulate the runoff. Roloff *et al.*, (1998) took an equation, which was called Baier-Robertson and was used to calculate the potential evaporation of plant, into the EPIC model. Williams *et al.*, (2000) added an infiltration equation (Green and Ampt) into EPIC model. All these efforts made to EPIC model improved its capacity to simulate soil water; however, few research reports were founded to evaluate its simulation results of soil moisture.

EPIC model has been modified and applied by lots of scientists, since it was introduced into China. Wang *et al.*, (2002) researched potential productivity of spring maize and winter wheat on the Loess Plateau of China and their results showed that spring maize yield was precisely simulated by modified EPIC model. Li *et al.*, (2004a, 2004b, 2004c and 2005) presented the frame and theory of EPIC model from 2004 to 2005, and they built up the basis foundation for the application of this model on the Loess Plateau of China. While Chen *et al.*, (2006) showed that the law of yield change of winter wheat and alfalfa were predicted well by EPIC model. Chun *et al.*, (2007) studied

alfalfa growth by EPIC model in Beijing city, and this study showed that biomass of alfalfa in different growth period was estimated well by EPIC model.

However, few researches were reported to evaluate the simulation results of soil water on the Loess Plateau of China. Li *et al.*, (2004a, 2004b and 2004c) evaluated the simulation results of total soil water year by year and their research indicted annually total soil water (in 0-6m soil) were simulated well by EPIC model. However, model still showed lack in its accuracy to simulate monthly soil water and its distribution of soil water in different soil layers. So there are still some uncertainties for the simulation results of EPIC model for soil water on the Loess Plateau of China. For example, research about alfalfa in Beijing city of China by Chun *et al.*, (2007) showed that simulated soil water did not agree well with the observed value because the parameters of EPIC model were not modified based on the local situation in his research. It showed the need of necessary amendment and evaluation based on local or regional situation by modification of EPIC parameters, before using it to research the soil water, which is a key factor of crop production for this region.

The objective of this paper was to evaluate the simulation results of monthly soil water and its distribution in different soil layers, using measured data of fixed long-term experiments (1987 to 1996) at Changwu Agriculture Station on the Loess Plateau of China.

MATERIALS AND METHODS

Site description and long-term experiment

In China, the Loess Plateau located in the upper-middle reaches of the Yellow River, and bordered on Taihang Mountain in the east, Reyue-Helan Mountain in the west, Qinling Mountain in the south and Yinsan Mountain in the north (100.90°-114.51° E and 33.70°-41.26°N), covers a total area of 62.85×104km², with its elevation ranging within 1200-1600m above sea level, and its loess cover mainly ranging from 30 to 80m thick. This region is a transitional zone between the southeastern humid monsoon climate and the northwestern continental dry climate with its Annual precipitation ranging from 200 to 750mm, its annual mean temperature varying between 8.6 and 13.5 8C, and its frost-free period ranging from 185 to 210 days.

Shilipu village (36°02'N, 104°25'E), the middle region of the Loess Plateau, is located in Changwu county of Shaanxi province China. It is a semi-humid continental monsoon climate zone and is a representative rain-fed area of China. Its mean annual precipitation is 578.5mm, its mean temperature is 9.1°C, and its mean frost-free period is 171d. Its predominant soil type is heilu soil (silt content is about65-75%, clay content is about 18-25%). Field capacity of this soil (by weight) is 22%, and its wilting point is 8%. Its primary crops are winter wheat and spring maize.

Field experiments with mono winter wheat and mono spring maize were carried out at Shilipu Village from 1987 to 1996. Fertilizer for winter wheat and spring maize were as following. Pure N was 120kg/hm²; pure P₂O₅ was 60 kg/hm²; sTable manure (organic matter content was 44.29g/kg) was 75000kg/hm². Winter wheat was sown in late-September and its grain yield was measured in mid-June each year. Spring maize was sown in late-April and its grain yield was measured in late-September each year. Soil water in 0-2m soil was measured by core break method (Bennie *et al.*, 1987) on the 20th day of each month from 1987 to 1996. Soil water content was measured (gravimetrically) for each soil sample by the oven-drying method (Blake and Hartge, 1986).

According to the statistical precipitation data from 1987 to 1996, its mean annual precipitation was 560mm in Changwu meteorological station. Based on this mean annual precipitation, all years were divided into different categories as in Table 1. From 1987 to 1996, the appearance probability of extreme rainy years was the same as it of extreme drought years and both of them were 10%; this probability of other years (rainy years, normal years or drought years) was 80%. Considering for a longer period, from 1957 to 2008, the appearance probability was less than 5% either for the extreme rainy years or for the extreme drought years but was more than 90% for other years. All years are classified into following categories as in the Table 1 (rainy years, normal years and drought years).

Crop parameters

Based on crop parameters and other related parameters, the EPIC model can be adopted to calculate the uptakes of soil water and nutrients by crop, estimate the impacts of temperature, water, nutrients (N, P and k), air and salt stresses on crop biomass accumulation and crop yield, and examine the process of crop growth by daily step. In this study, relevant crop parameters (Table 2) were modified based on the measured and published data (Wu *et al.*, 2009; Wang *et al.*, 2008; Ma *et al.*, 2008; Jiang, 2009).

Table 1. Different rainfall years of Changwu meteorological station from 1987 to 1996.

Type of rain fall years	Annual precipitation	Year
Extreme rainy years	>700mm	1988
Rainy years	600-700mm	1990 and 1996
Normal years	500-600mm	1987, 1989, 1992, 1993
Drought years	400-500mm	1991 and 1994
Extreme drought years	<400mm	1995

Soil data

In the EPIC model, soil data was stored in the file named SOIL_DATA. In this study, 6m soil was divided into 9 layers (Table 3) based on the measured data and soil survey books published in China. Its mean bulk density was 1.32g/cm³ with mean field capacity was 0.27m/m and mean wilting point was 0.13m/m. Water and wind erosion were not taken into account in the EPIC model.

Tab. 2 Important revised vegetal parameters of winter wheat and spring maize in the EPIC model

Parameters	Winter wheat	Spring maize
WA	30	40
HI	0.4	0.5
TB	15.5	25.5
TG	0.0	8.0
DMLA	6.0	6.0
DLAI	0.91	0.80
RLAD	1.0	1.0
RBMD	1.0	1.0
GSI	0.007	0.007
HMX	1.2	2.0
RDMX	4.0	4.0
FRST1	15.20	5.01
FRST2	25.50	15.95
RWPC1	0.40	0.40
RWPC2	0.20	0.20

Acronyms: WA: Energy to biomass conversion; HI: Harvest index; TB: Optimal temperature for plant growth; TG: Minimum temperature for plant growth; DMLA: Maximum lead area index; DLAI: When leaf area index starts; RLAD: Leaf area index decline rate; RBMD: Biomass-energy ratio decline rate; GSI: Maximum stomatal conductance; HMX: Maximum crop height; RDMX: Maximum root depth; FRST1: Frost days 1; FRST2: Frost days 2; RWPC1: Root weight /biomass portioning coefficient 1; RWPC2: Root weight /biomass portioning coefficient 2.

Table 3. Important physical and chemical parameters of *heilu* soil.

layer number	1	2	3	4	5	6	7	8	9
Layer depth (m)	0.1	0.5	1	1.5	2	3	4	5	6
Field capacity (m/m)	0.28	0.28	0.28	0.28	0.28	0.27	0.27	0.27	0.26
Wilting point (m/m)	0.10	0.10	0.11	0.11	0.12	0.13	0.13	0.13	0.14
Bulk density (g/cm ³)	1.30	1.30	1.31	1.31	1.32	1.32	1.33	1.33	1.33
PH	8.2	8.2	8.2	8.3	8.4	8.3	8.2	8.2	8.3
Action exchange capacity (cmol/kg)	9.6	9.6	9.9	12.2	8.6	9.7	6.5	4.5	3.5
Calcium carbonate (%)	8	8	6.8	7.0	14.2	11.9	0	0	0
Phosphorus (ppm)	4	4	3.0	3.0	2	2	0	0	0
Initial nitrate concentration (ppm)	30	50	50	40	30	20	0	0	0
Organic nitrogen (ppm)	613	613	452	553	400	480	0	0	0
Organic carbonate (%)	0.62	0.62	0.45	0.55	0.45	0.58	0	0	0

Meteorological data

The meteorological data (Table 4), including daily maximum temperature, daily minimum temperature, daily precipitation, daily wind speed, daily relative humidity and daily sunshine hours, were obtained from Changwu meteorological station. In this study, sunshine radiation, needed by EPIC model, was translated from sunshine hours, recorded at Changwu meteorological station, based on equations suggested by Li *et al.*, (2004a).

Table 4. Monthly statistical meteorological parameters at Changwu station, Shaanxi province China.

Items	Jan.	Feb.	Mar.	Apr.	May.	Jun.	Jul.	Aug.	Sep.	Oct.	Nov.	Dec.
WSPD	2.02	2.23	2.56	2.77	2.54	2.39	2.49	2.32	2.01	1.93	2.05	1.9
TMX	1.14	3.96	10.07	16.94	21.75	26.22	27.73	26.3	20.54	14.82	8.1	2.58
TMN	-9.87	-6.45	-0.69	4.81	9.16	13.32	17.03	16.08	10.96	4.97	-1.88	-7.88
SKWS	1.26	0.85	0.72	1.01	0.80	0.72	0.50	0.72	0.84	0.90	1.10	1.03
SKRF	1.67	1.35	1.25	2.3	2.78	2.54	1.73	2.45	1.78	2.05	2.48	2.36
SDWS	1.21	1.17	1.23	1.3	1.17	1.01	1.03	1.11	1.05	1.1	1.26	1.16
SDRF	1.53	1.92	4.03	7.54	8.36	8.72	12.43	14.8	10.19	6.6	4.68	2.2
SDMX	4.17	5.17	5.39	5.3	4.69	3.96	3.6	3.78	4.31	4.41	4.7	4.35
SDMN	3.53	3.92	3.39	3.65	3.41	2.87	2.33	2.67	3.29	3.88	3.9	3.8
RHUM	0.59	0.61	0.63	0.63	0.65	0.64	0.73	0.77	0.8	0.78	0.72	0.64
RAD	7.0	8.1	9.9	12.3	14.1	14.8	14.1	13.0	9.6	7.9	6.9	6.4
PW W	0.29	0.27	0.32	0.34	0.36	0.35	0.43	0.38	0.52	0.48	0.43	0.21
PW D	0.05	0.06	0.11	0.15	0.17	0.19	0.22	0.21	0.21	0.15	0.08	0.03
PRCP	5.5	7.4	22.7	42	53.3	57.5	104.	102	92.9	51.9	20.8	3.7
ETMX	15.7	21.3	24.7	30.9	33.9	36.9	37.6	36	34.3	28.7	20.9	17.1
ETMN	-22	-20.3	-12.5	-8.8	-1.7	4.0	9.0	7.3	-0.5	-8.5	-16.7	-25.2
DAYW	0.42	0.27	0.38	0.78	0.4	0.11	0.11	0.11	0.13	0.16	0.36	0.2
DAYP	2.18	2.40	4.47	5.71	6.56	6.89	8.71	7.98	9.16	7.04	4.11	1.27
AEMX	8.47	12.9	19.6	25.9	29.2	32.22	33.04	31.67	27.41	21.69	15.81	9.86
AEMN	-16	-13.6	-7.2	-2.2	2.1	7.4	12.2	10.8	4.6	-2.3	-9.1	-14.5

Acronyms: WSPD: Wind velocity; TMX: Maximum daily air temperature; TMN: Minimum daily air temperature; SKWS: Monthly skew coefficient for daily drought stress; SKRF: Monthly skew coefficient for daily precipitation; SDWS: Monthly standard deviation for daily drought stress; SDRF: Monthly standard deviation of daily precipitation; SDTMX: Monthly standard deviation of daily maximum temperature; SDTMN: Monthly standard deviation of daily minimum temperature; RHUM: Relative humidity; RAD: Solar radiation; PW|W: Monthly probability of wet day after wet; PW|D: Monthly probability of wet day after dry; PRCP: Precipitation; ETMX: Maximum evapotranspiration; ETMN: Minimum evapotranspiration; DAYW: Wind day; DAYP: Precipitation day; AEMX: Maximum single applied volume allowed; AEMN: Minimum single applied volume allowed.

Management parameters

All other management parameters and equipment were built up into the EPIC model based on location situations. Two crop systems (mono winter wheat and mono spring maize) were built up into the ROTATION file; parameters of equipment for planting, harvest and fertilizers, were modified based on local situations; three kinds of fertilizer (urea, pure P₂O₅ and Table manual) were added into the EPIC model by the filed named FERMAN.

RESULTS AND ANALYSIS

Soil water in mono winter wheat field

In 1987 to 1996, the mean simulated monthly soil moisture in 0-2m soil depth was 0.207m/m and the mean measured monthly soil moisture in 0-2m soil depth was 0.201m/m (Table 5). The correlation coefficient between simulated and measured soil moisture was 0.888 (p<0.01), with Root Mean Square Errors (RMSE) was 0.023m/m and Relative Root Mean Square Errors (RRMSE) was 11.4%.

Table 5. Comparison of simulated and measured soil moisture in different soil layers in winter wheat field on Changwu arid-plateau. ** means p<0.01.

	0-0.1m	0.1-0.5m	0.5-1.0m	1.0-1.5m	1.5-2.0m
Measured (m/m)	0.234	0.192	0.204	0.189	0.187
Simulated (m/m)	0.234	0.195	0.217	0.199	0.188
Relative error (%)	0.0	1.6	6.4	5.3	0.5
Correlation coefficient	0.884**	0.901**	0.921**	0.820**	0.916**
RMSE(m/m)	0.032	0.023	0.025	0.025	0.012
RRMSE (%)	14	12	12	13	6
Regression equation	$y = 0.85x + 0.06$	$y = 0.81x + 0.04$	$y = 1.14x - 0.02$	$y = 1.13x - 0.02$	$y = 1.09x - 0.01$

Statistical values among different years were significantly different for simulated and observed soil moisture in 0-2m soil depth. Between simulated and measured soil moisture, the values of correlation coefficients, relative errors, RMSE and RRMSE were 0.860(p<0.05), 7.4%, 0.027m/m and 17%, respectively in extreme drought years; 0.919(p<0.01), 3.9%, 0.022 m/m and 12%, respectively in drought years; 0.926(p<0.01), 0.2%, 0.023 m/m and 11% in normal years; 0.874(p<0.01), 6.6%, 0.028m/m and 14% in rainy years; 0.802(p<0.05), 2.6%, 0.029m/m and 13% in extreme rainy years.

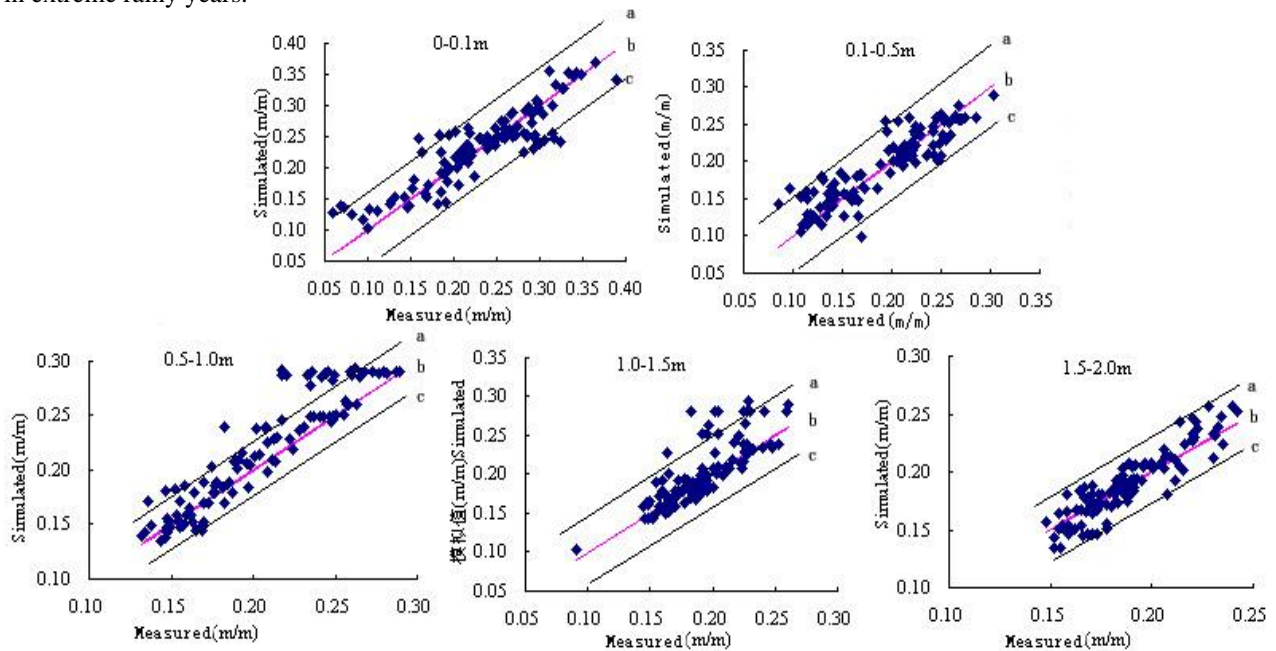


Fig.1. Comparison of simulated and measured soil moisture in different soil layers in winter wheat field on Changwu arid-plateau.

In mono winter wheat field, soil moisture was predicted well generally by EPIC model comparing with measured data (Fig. 1). The measured and simulated winter wheat yields uniformly distributed near the line of y=x, and the intercepts for their regression equations were near to 0, the regression index was near to 1. Most simulated values (97% in average) were scattered between the line $y = x + STx$ and the line $y = x - STx$ (Fig. 1). X

was the measured soil moisture, ST_x was the standard deviation of measured soil moisture and y is the simulated soil moisture.

The distribution of simulated and measured soil moisture was different in different soil layers (Fig. 1). In 0-0.1m, 0.1-0.5m and 1.5-2.0m soil layers, comparing with 0.5-1.0m and 1.0-1.5m soil layers, there was a less distance between the value of regression index and 1. There were more values departure from the line $y=x$ in 0.5-1.0m and 1.0-1.5 soil layers than in other three layers.

Soil moisture was generally predicted well by EPIC model, comparing with measured data. a is the line $y = x + ST_x$; b is the line $y=x$; c is the line $y=x-ST_x$; y is the simulated soil moisture, x is the observed soil moisture, ST_x is the standard deviation of observed soil moisture; 0-0.1m, 0.1-0.5m, 0.5-1.0m, 1.0-1.5m and 1.5-2.0m are the different soil layers.

Soil water in mono spring maize field

In 1987 to 1996, the mean simulated monthly soil moisture in 0-2m soil was 0.215m/m and the mean measured monthly soil moisture in 0-2m soil was 0.216m/m (Table 6). The Correlation coefficients between the simulated and the measured soil moisture was 0.943 ($p<0.01$) with RMSE was 0.015m/m and RRMSE was 7%.

Table 6. Comparison of simulated and measured soil moisture in different soil layers in spring maize field on Changwu arid-plateau. * means $p<0.05$; **means $p<0.01$.

	0-0.1m	0.1-0.5m	0.5-1.0m	1.0-1.5m	1.5-2.0m
Measured (m/m)	0.234	0.196	0.213	0.219	0.217
Simulated (m/m)	0.228	0.199	0.212	0.219	0.218
Relative error (%)	-2.6	1.5	-0.5	0.0	0.5
Correlation coefficient	0.940**	0.928**	0.920**	0.966**	0.963**
RMSE (m/m)	0.023	0.017	0.017	0.009	0.008
RRMSE (%)	10	9	8	4	4
Regression equation	$y = 0.88x + 0.02$	$y = 0.87x + 0.03$	$y = 0.96x + 0.01$	$y = 0.97x + 0.01$	$y = 1.06x - 0.01$

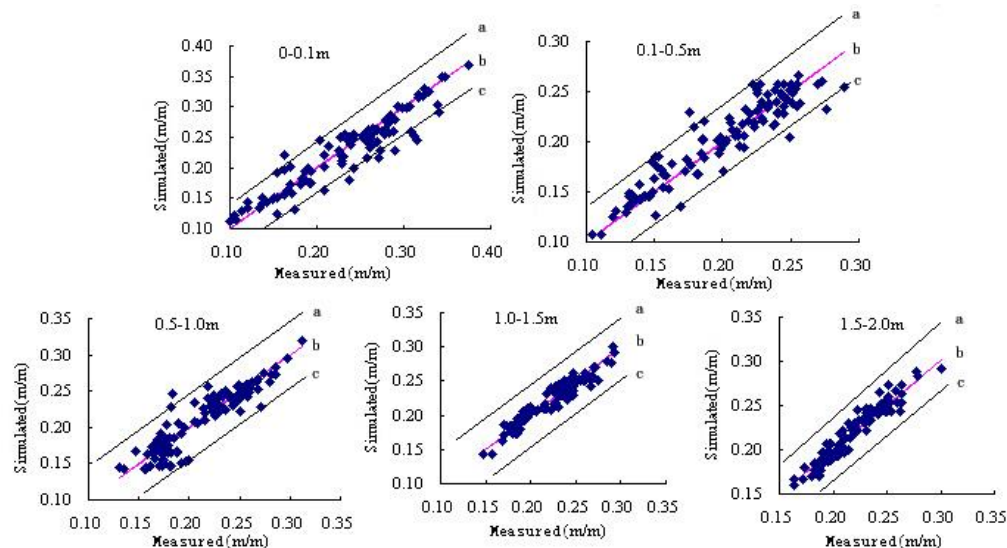


Fig. 2. Comparison of simulated and observed soil moisture in different soil layers in spring maize land on Cahngwu arid-plateau.

a is the line $y = x + ST_x$; b is the line $y=x$; c is the line $y=x-ST_x$; y is simulated soil moisture, x is the observed soil moisture, ST_x is the standard deviation of observed soil moisture; 0-0.1m, 0.1-0.5m, 0.5-1.0m, 1.0-1.5m and 1.5-2.0m are the different soil layers.

Statistical values among different rainfall years were different for simulated and observed soil moisture in 0-2m soil depth. Between simulated and measured soil moisture, the values of correlation coefficient, relative error, RMSE and RRMSE were 0.844($p<0.05$), -2.9%, 0.019m/m and 10%, respectively in extreme drought years; 0.909($p<0.01$), 0.0%, 0.012m/m and 6%, respectively in drought years; 0.966($p<0.01$), 0.4%, 0.009m/m and 4% in normal years; 0.885($p<0.01$), -1.1%, 0.015m/m and 7% in rainy years; 0.823($p<0.05$), -2.8%, 0.024m/m and 10% in extreme rainy years.

Soil moisture was estimated well generally, comparing with measured data, in mono spring maize field (figure 2). The measured and simulated soil moisture uniformly distributed near the line of $y=x$, and the interceptions for their regression equations were near to 0, their regression index were near to 1. Most simulated value (99% in average) distributed among the line $y = x + STx$ and the line $y = x - STx$ (figure 2). X was the measured soil moisture, STx was the standard deviation of measured soil moisture and y is the simulated soil moisture.

The distribution of simulated and measured soil moisture in mono spring maize field was different in different soil layers (Fig. 2). In 0.5-1.0m, 1.0-1.5m and 1.5-2.0m soil layers, comparing with that in 0-0.1m and 0.1-0.5m soil layers, there was a less distance between the value of regression index and 1. There were more values departure from the line of $y=x$ in 0-0.1m and 0.1-0.5 soil layers than in the other three layers.

DISCUSSION

This study showed that soil moisture were predicted well comparing with measured value in different soil layers, either in mono winter wheat field or in mono spring maize field (Table 5 and Table 6), though the mean simulated moisture was higher than mean measured value in winter wheat field and was lower than mean measured value in mono spring maize field. Mean monthly soil moistures in 0-2m depth soil were 0.207m/m and 0.201m/m for simulated and measured in mono winter wheat field, respectively. Mean monthly soil moistures in 0-2m depth soil were 0.215m/m and 0.216m/m for simulated and measured in mono spring maize field, respectively. The relation between simulated and observed soil moisture was significant; the significant level was lower than 0.01; and the RRMSE was lower than 0.03, both in mono winter wheat field and in mono spring maize field.

The accuracy of simulated soil moisture was higher in mono spring maize field than in mono winter wheat field. Correlation coefficient between simulated and measured soil moisture in 0-2m depth soil was higher and the value of RRMSE was lower in mono spring maize field compared with mono winter wheat field. In EPIC model, meteorological data, soil data and other data of control Table for mono winter wheat were the same as them for mono spring maize, so the difference of estimated accuracy of soil moisture between mono winter wheat field and mono spring maize field was produced by the difference of crop data of winter wheat and spring maize. Therefore, the estimated accuracy of soil moisture was relative to the crop data in EPIC model.

EPIC model satisfactorily simulated and measured soil moisture in different rain fall years, though it was influenced by the amount of annual rainfall and its accuracy were some lower in extreme rainfall years. The mean correlation coefficients between simulated and measured soil moisture in 0-2m soil depth, (Table 1) of the classified years (drought years, normal years and rainy years) were 0.880($p<0.01$) and 0.938($P<0.01$) in mono winter wheat field and mono spring maize field respectively. Its value of two extreme rainfall years (extreme rainy years and extreme drought years) were 0.831($P<0.05$) and 0.833($P<0.05$) in mono winter wheat field and mono spring maize field, respectively. The value of RMSE were generally lower in drought years, normal years and rainy years, compared with in extreme drought years and extreme rainy years. From 1987 to 1996, there were only one year (1988) was extreme rainy year and one year (1995) was extreme drought year, i.e. their appearance probability was 20%. Considering from a longer period (1957-2008), the appearance probability of extreme years, extreme rainy years or extreme drought years was lower (<10%). Though the accuracy of simulated soil moisture in extreme rainfall years was lower than other rainfall years, the significant level of correlation coefficient was lower than 0.05. Therefore the simulation results about soil moisture in arid land at Changwu arid plateau were generally reasonable.

The accuracy of estimated soil moisture will be higher, if reasonable soil database, crop database and meteorological database, based on the local or regional situation, are built up into the EPIC model. The distributions of simulated and measured soil moisture were more concentrated in mono spring maize field than it in mono winter wheat field (Fig. 1 and Fig. 2). These distributions in 1.0-1.5m and 1.5-2.0m soil layers were more concentrated than it in 0-0.1m and 0.1-0.5m soil layers, either in mono winter wheat field or in mono spring maize field. The depletion of soil water by crop root was the key factor to influence the soil moisture in 1.0-2.0 soil layers, i.e. 1.0-1.5m and 1.5-2.0m soil layers, and it was influenced by potential Evapo-transpiration, one crop parameters of the EPIC model. Evaporation was the key factor to influence the soil moisture in 0-0.5m soil layer, i.e. 0-0.1m and 0.1-0.5m soil layers, and it was influenced by the amount of sunshine radiation in meteorological database, the leaf area index in crop database and the equation to calculate soil moisture in the EPIC model. Benson *et al.*, (1992) pointed out that

the accuracy of simulation results of EPIC model were relative to the calculating equations chosen by the user of EPIC model, potential evapo-transpiration equation and soil moisture calculating equation. Roloff *et al.*, (1998) believed that the accuracy of simulated soil water was relative to the amount of precipitation, the amount of sunshine radiation and the thickness of soil layers. When we amended the EPIC model, we found that the other soil parameters, such as the sand content, the silt content, the loam content and the bulk density, could influence the accuracy of simulated soil water. Therefore, the reasonability of database for EPIC model was the key factor to influence the accuracy of simulation results.

Conclusion

- 1) The modified EPIC model can be used to estimate the soil moisture and then to research the law of soil water change of arid land in the Loess Plateau of China. Relative errors between simulated and observed soil moisture in 0-2m depth soil were 2.8% and -0.2% in mono winter wheat field and mono spring maize field, respectively. Correlation coefficients were 0.888 and 0.943, respectively, RMSE were 0.023m/m and 0.015m/m, respectively.
- 2) There was a varying accuracy of simulated soil moisture in different rainfall years. Comparing with the soil moisture in drought years, rainy years and normal years, there was a lower accuracy in extreme rainy years and extreme drought years. The highest accuracy appeared in normal years among all different rainfall years. Correlation coefficients between simulated and measured soil moisture in normal years were respectively 0.926 and 0.966 in mono winter wheat field and mono spring maize field; RMSE was 0.023m/m and 0.009m/m, respectively.
- 3) Crop parameters, amount of potential evaporation transpiration, soil parameters, equation to calculate soil moisture, and amount of rainfall can influence the accuracy of estimated soil moisture. Therefore, accuracy of estimated soil moisture will be increased, provided reasonable soil database, crop database and meteorological database, based on the local or regional situation, built up into the EPIC model.

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