Performance of a zero-inertia model for irrigation with rapidly varied inflow discharges

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Abstract: The zero-inertia model is widely used for simulating surface flow in irrigation systems. This model is accurate when inflow discharge is constant. However, simulation of irrigation systems with rapidly varied inflow discharge is needed due to the development of real time control irrigation technology. Hence, the objective of this study is to validate the zero-inertia model with rapidly varied inflow discharge. For this purpose, twenty-three border irrigation tests at a range of inflow changes on different field slopes and roughness coefficients were conducted. Then, the sensitivity analyses of bed slope, infiltration parameters, and roughness coefficient were examined. The results indicate that the zero-inertia model predictions are in good agreement with field data in both advance/recession times and flow depths. The infiltration parameters were the most sensitive input variable of the zero-inertia model. The input variables have a more considerable impact on the recession phase than the advance phase.

Keywords: zero-inertia model, rapidly varied inflow, border irrigation, sensitivity analysis **DOI:** 10.25165/j.ijabe.20201302.5228

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1 Introduction

Surface irrigation is still the most common method for irrigating crops across the world. Numerical simulation of surface irrigation flow can provide the necessary tools to simulate, design and evaluate irrigation systems. Since the 1970s, with the popularization of computer application, the numerical simulation model of surface irrigation has developed rapidly. At present, there are mainly four mathematical models.

The full hydrodynamic model is recognized as the accurate description of the water flow in surface irrigation^[1]. It is based on Saint–Venant long wave equations, which are two separate hyperbolic partial differential equations and include a mass conservation equation and a momentum equation. Because these equations are hard to solve algebraically in the purest form, numerical solution methods, such as the finite difference method^[2,3], the method of characteristics^[4], the finite element

method^[5] and the finite volume method^[6], were used to obtain the numerical solution. Although the full hydrodynamic model can provide an accurate simulation, sometimes it is numerically sensitive, and existing a numerical oscillation^[7]. Additionally, the complexity also limits its wide application. For these reasons, Strelkof and Katopodes^[8] introduced the zero-inertia model based on the full hydrodynamic model. Removing the inertial, or acceleration, terms in the momentum equation simplifies the calculation and increases the operation speed as well as robustness^[1,9], and the result is still accurate^[10]. The kinematic-wave model further simplifies Saint-Venant long wave equations by assuming that the water surface slope deviates little from the bed slope. Some experiments verified the model can be used to simulate the advance and recession curves^[11]. However, the kinematic-wave model accuracy is limited by the field slope and soil textures^[10]. The volume balance model is the simplest model. The momentum equation is neglected, and the amount of irrigation water usually approximated by flow and infiltration depth at the key points of field, the shape factor and the advance distance^[12]. However, the volume balance model is rarely used, because it is hard to guarantee its accuracy.

The zero-inertia model has been widely used for simulation in surface irrigation systems and proved to be accurate when inflow discharge is constant^[13]. However, with the development of automatic irrigation technology, such as the self-adaptive control of surface irrigation^[14], inflow discharge is not always constant, and the inflow discharge may also be suddenly adjusted. Saint-Venant long wave Equations is not strictly fit to rapidly varied unsteady flow^[15]. Szydtowski and Zima^[16] investigated rapidly varied open channel flow due to dam-break effect and pointed out that Saint-Venant long wave Equations can simulate dam-break water flow although they cannot well describe the flow near the local

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hydraulic effects. However, the surface irrigation system is more complex and microscopic than open channel transportation system, which means that the errors caused by rapidly varied inflow are likely to amplify. In this case, the applicability of zero-inertia model, which is derived from the simplification of Saint-Venant Equations need to be verified by field measurements. At present, there are few studies on the surface irrigation system with rapidly varied inflow, and related field experiments are also lacking. Hence, the objectives of this study were as follows: (1) to validate the applicability of the zero-inertia model (23 border irrigation experiments were conducted for this purpose); and (2) to analyze the sensitivity of system inputs to provide recommendations for future experimental measurement in rapidly varied inflow irrigation system.

2 Materials and methods

2.1 Field measurements

2.1.1 Study area

Experiments were carried out in a wheat field at the CAS Ecological Agricultural Experiment Station in the town of Nanpi in Hebei Province, China (116°40′E, 38°06′N). The region is in a warm temperate zone and has a semi-humid monsoon climate. The annual rainfall ranges from 500 mm to 600 mm, and generally occurs during the summer (about 73% of the whole year). The annual evaporation is between 1500 mm and 1800 mm. The groundwater depth is below 2 m. The surface soil is a silt loam, and the dry bulk density is 1.48 g/cm³ at depth of 1 m. The farming in study area is carried out according to the local perennial crop. Winter wheat is the main crop and planted in October, and generally harvested in next June.

2.1.2 Experiment design and measurement

A total of 23 border irrigation tests were conducted. The tested borders were divided into two groups (Group A: A1 to A12 and Group B: B1 to B11). Group A was carried out in 2016, and each border in this group was designed to be 80 m in length and 3.7 m in width. Group B was carried out in 2018, and each border in this group was designed to be 100 m in length and 3 m in width. Similar to the border used by local farmers, all test borders were closed-ended. In order to be comprehensive, the inflow discharge for borders A1 to A3 and for borders B1 to B4 was constant. And the inflow discharge for borders A4 to A12 and for borders B5 to B11 was rapidly varied.

To better cover the wide range of irrigation practices, the inflow rate ranged between 2-7 L/s·m^[17,18]. Hence, the inflow rates within or slightly larger than this range were arranged at the study area. The irrigation pipe system of Group A is designed with a valve to regulate inflow rate. However, it is very difficult to control the inflow rate with the valves. This resulted in inconsistent inflow rate for almost every border. In order to make the experiment better controlled, an improvement was implemented in Group B. Three independent irrigation pipe systems were used, and the inflow rate is fixed in each of them at 2.4 L/s·m, 4.17 L/s·m and 6.5 L/s·m. During the irrigation events, the irrigation pipe systems were all opened to ensure that the inflow rates were stable. When the inflow discharge needs to be changed, the corresponding irrigation pipe was linked in.

According to local farming practices, the machine levels the land while sowing. There is no professional machine to level the land, so the field is not well leveled. For each border, longitudinal slopes were measured with 5 m interval. All border slopes were generally downward, except a few intervals in some borders are inclined upward.

Soil was classified as a silt loam based on particle size (65.53% silt, 21.75% sand, and 12.72% clay on average). Irrigation inflow rates were measured by electromagnetic flow meters (accuracy of $\pm 1.5\%$). The infiltration parameters *k* and *a* of the Kostiakov Equation were derived by the soil moisture before and after irrigation at two observation points, and the equations can be written as Equations (1) and (2). The roughness coefficient of border was determined by the zero-inertia model and the trial and error approach^[17,19], since it is difficult to measure or calculate it directly. The different values of roughness coefficient were fed into the zero-inertia model. Then, the simulated advance and recession trajectories of the model were compared with the observed data, and when a good fit between them was achieved, the actual roughness coefficient was established.

$$\alpha = \frac{\ln(z_1 / z_2)}{\ln(t_1 / t_2)}$$
(1)

$$k = \frac{z_1}{t_1^{\alpha}} \tag{2}$$

where, z_1 and z_2 are the cumulative infiltration of observation points 1 and 2 respectively, m; t_1 and t_2 are the infiltration times of observation points 1 and 2 respectively, s.

2.2 Computational schemes

The zero-inertia model are composed of a mass conservation equation and a simplified momentum equation, and these equations can be written as:

$$\frac{\partial q}{\partial x} + \frac{\partial h}{\partial t} + \frac{\partial z}{\partial t} = 0$$
(3)

$$\frac{\partial h}{\partial x} = s_0 - s_f \tag{4}$$

where, x is the distance from the upper end of the channel, m; t is the time from the beginning of inflow, s; q is the discharge per unit width, m³/(s·m); h is flow depth, m; z is the cumulative infiltration, m; s_0 is the border slope; s_f is the friction slope, and it can be written as:

$$s_{\rm f} = \frac{Q^2 n^2}{A^2 R^{\frac{4}{3}}} \approx \frac{V^2 n^2}{h^{\frac{4}{3}}}$$
(5)

where, Q is volume flow rate, m^3/s ; A is the cross-sectional area of flow, m^2 ; n is roughness coefficient.

Kostiakov Equation is the most frequently used infiltration equation in the numerical simulation of surface irrigation:

$$Z = kt^{\alpha} \tag{6}$$

where, $k (m/s^{\alpha})$ and α are empirical coefficients.

The numerical solution follows the finite difference method, similar to that presented by Mcclymont^[1]. The time-space computation grid is a rectangular grid system as depicted in Figure 1. The time step is invariant, and the space step is variable. The subscript referencing J, M, L and R represent the variable's grid cell representation and the basic equations (Equations (3) and (4)) can be transformed into finite difference form and written as:

$$\begin{bmatrix} \theta q_{\rm R} + (1-\theta)q_{\rm M} - \theta q_{\rm L} - (1-\theta)q_{\rm J} \end{bmatrix} dt + \begin{bmatrix} \varphi h_{\rm R} + (1-\varphi)h_{\rm L} - \varphi h_{\rm M} - (1-\varphi)h_{\rm J} \end{bmatrix} dx$$
(7)
+
$$\begin{bmatrix} \varphi z_{\rm R} + (1-\varphi)z_{\rm L} - \varphi z_{\rm M} - (1-\varphi)z_{\rm J} \end{bmatrix} dx = R_{C} \begin{bmatrix} \theta q_{\rm R} + (1-\theta)q_{\rm M} - \theta q_{\rm L} - (1-\theta)q_{\rm J} \end{bmatrix} + \begin{bmatrix} -s_0 + \theta \varphi s_{f\rm R} + (1-\theta)\varphi s_{f\rm M} + \theta (1-\varphi)s_{f\rm L} + (1-\theta)(1-\varphi)s_{f\rm J} \end{bmatrix} dx = R_{M}$$
(8)

where, θ and φ are the space-averaging and time-averaging coefficients, and these values are typically equal to $0.6^{[1]}$; $R_{\rm C}$ and $R_{\rm M}$ are residuals from the continuity and momentum equations respectively, and these values should theoretically be equal to zero.





The solution is implicit, so the basic equations should be solved simultaneously for all cells at each time step. The following is illustrated by an example of the calculation of period of t_{k-1} (known) to t_k (unknown). Assuming that the number of cells at t_k is N (the number of unknown points at t_k is N+1). Each call can establish two equations, together with the boundary conditions, so the total number of equations is 2N+2. There are two variables at each unknown point (*q* and *h*), so the total number of variables is also 2N+2. The double-sweep method described by Walker and Skogerboe^[20] and Mcclymont^[1] was used to solve these implicit equations.

2.3 Statistical analysis

The accuracy of the modified model could be evaluated by the coefficient of determination (R^2) , root mean square error (RMSE)

and coefficient of residual mass (CRM)^[21]. R^2 varies from zero to one, and the larger it is, the more valuable simulation results are. The RMSE has minimum value of zero, and the closer it gets to zero, the more accurate the simulation results are. CRM can be negative or positive, and it indicates the underestimation or overestimation amounts of predictions. R^2 , RMSE and CRM can be written as:

$$R^{2} = 1 - \frac{\sum_{i=1}^{n} (t_{p} - t_{m})_{i}^{2}}{\sum_{i=1}^{n} (t_{m} - \overline{t})_{i}^{2}}$$
(9)

RMSE =
$$\sqrt{\frac{\sum_{i=1}^{n} (t_p - t_m)_i^2}{n}}$$
 (10)

$$CRM = \frac{\sum_{i=1}^{n} (t_p - t_m)_i}{\sum_{i=1}^{n} t_m}$$
(11)

where, t_p is the calculated advance/recession time, min; t_m is the measured advance/recession time, min; \overline{t} is the mean of the measured advance/recession time, min; n is the number of stations along the border length.

3 Results and discussion

3.1 Experimental data

Table 1 provides the longitudinal slopes of every border each 5 m. Although the average slope of each border was about 0.002, the field was not very flat. This indicates that local traditional farming methods did not perform very well in field levelling. Table 2 provides details of the inflow rate, inflow change time, cut-off time, Kostiakov infiltration parameters and roughness coefficient. These were used as input parameters of the zero-inertia model for each irrigation event to simulate the advance trajectories, recession trajectories and flow depths.

Table 1 Observational slope of each border	Table 1	Observational slope of each border
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Border									Avera	age slop	e of eac	h 5 m									Average
number	0-5	5-10	10-15	15-20	20-25	25-30	30-35	35-40	40-45	45-50	50-55	55-60	60-65	65-70	70-75	75-80	80-85	85-90	90-95	95-100	slope
Al	0.0020	0.0008	0.0054	-0.0010	0.0010	0.0030	-0.0014	0.0026	0.0018	0.0016	0.0100	-0.0020	0.0042	0.0018	0.0060	0.0062					0.0026
A2	0.0044	0.0032	0.0010	0.0000	0.0004	0.0006	0.0014	0.0018	0.0070	-0.0030	0.0034	0.0042	0.0046	-0.0044	0.0118	-0.0022					0.0021
A3	-0.0106	0.0028	0.0052	-0.0018	0.0020	-0.0004	0.0042	0.0022	0.0014	0.0024	0.0072	-0.0020	0.0014	0.0046	0.0088	0.0005					0.0017
A4	-0.0020	-0.0024	0.0102	-0.0038	0.0042	0.0046	-0.0040	0.0036	0.0050	0.0034	-0.0002	0.0052	-0.0038	0.0044	0.0096	-0.0066		-			0.0017
A5	0.0020	0.0030	0.0020	-0.0034	0.0040	0.0078	0.0000	0.0016	0.0016	-0.0008	0.0038	0.0066	-0.0028	0.0024	0.0050	-0.0010		-			0.0020
A6	0.0024	0.0006	0.0064	0.0034	0.0008	0.0032	0.0018	0.0008	0.0014	0.0046	0.0040	0.0002	0.0012	0.0010	0.0038	0.0046		-			0.0025
A7	0.0000	0.0018	0.0016	0.0012	0.0040	0.0030	0.0012	0.0000	0.0058	-0.0004	0.0034	0.0048	-0.0016	0.0060	0.0058	0.0020		-			0.0024
A8	0.0004	0.0114	-0.0024	0.0022	0.0020	0.0010	0.0006	0.0034	0.0050	0.0058	-0.0034	0.0032	0.0046	0.0004	0.0038	0.0014		-			0.0025
A9	-0.0006	0.0012	0.0030	-0.0008	0.0020	0.0044	-0.0020	0.0030	0.0038	0.0022	-0.0040	0.0038	-0.0005	0.0030	0.0024	0.0010		-			0.0014
A10	-0.0016	0.0024	0.0030	-0.0025	0.0017	0.0020	-0.0008	0.0024	0.0026	-0.0012	0.0038	-0.0016	0.0012	0.0042	0.0058	-0.0024		-			0.0012
A11	-0.0014	0.0008	0.0052	-0.0010	0.0012	0.0088	0.0062	-0.0018	0.0014	0.0004	0.0082	0.0032	0.0008	0.0056	0.0022	0.0040		-		-	0.0027
A12	-0.0042	0.0014	0.0012	0.0000	-0.0048	0.0088	0.0074	-0.0004	0.0002	-0.0004	0.0048	0.0048	0.0036	-0.0014	0.0050	0.0080		-		-	0.0021
B1	0.0040	0.0015	0.0010	0.0017	0.0018	0.0010	0.0017	0.0015	0.0021	0.0010	0.0010	0.0011	0.0010	0.0010	0.0010	0.0010	0.0020	0.0020	0.0020	0.0030	0.0016
B2	0.0016	0.0013	0.0013	0.0039	0.0014	0.0001	0.0017	0.0033	0.0019	0.0005	0.0005	0.0005	0.0005	0.0005	0.0007	0.0007	0.0021	0.0046	0.0038	0.0042	0.0018
B3	0.0024	0.0027	0.0015	0.0021	0.0034	0.0010	0.0021	0.0011	0.0030	0.0007	0.0005	0.0005	0.0008	0.0005	0.0002	0.0006	0.0005	0.0041	0.0045	0.0043	0.0018
B4	0.0025	0.0013	0.0010	0.0031	0.0010	0.0017	0.0011	0.0008	0.0026	0.0005	0.0013	0.0015	0.0011	0.0018	0.0015	0.0015	0.0044	0.0047	0.0037	0.0030	0.0020
В5	0.0018	0.0010	0.0019	0.0020	0.0025	0.0009	0.0021	0.0011	0.0013	0.0002	0.0001	0.0002	0.0001	0.0002	0.0000	0.0003	0.0013	0.0030	0.0045	0.0057	0.0015
B6	0.0023	0.0004	0.0017	0.0031	0.0020	0.0021	0.0006	0.0010	0.0029	0.0013	0.0006	0.0003	0.0000	0.0004	0.0003	0.0013	0.0017	0.0029	0.0029	0.0045	0.0016
B7	0.0009	0.0019	0.0009	0.0029	0.0018	0.0011	0.0007	0.0023	0.0025	0.0010	0.0002	0.0011	0.0003	0.0003	0.0017	0.0022	0.0033	0.0026	0.0017	0.0032	0.0016
B8	0.0013	0.0009	0.0028	0.0010	0.0026	0.0001	0.0008	0.0026	0.0039	0.0011	0.0011	0.0015	0.0005	0.0007	0.0010	0.0020	0.0023	0.0039	0.0015	0.0023	0.0017
B9	0.0009	0.0011	0.0001	0.0027	0.0031	0.0004	0.0018	0.0016	0.0005	0.0019	0.0023	0.0003	0.0019	0.0001	0.0013	0.0029	0.0049	0.0051	0.0017	0.0028	0.0019
B10	0.0017	0.0013	0.0038	0.0005	0.0013	0.0001	0.0016	0.0014	0.0030	0.0009	0.0010	0.0006	0.0033	0.0010	0.0030	0.0050	0.0020	0.0057	0.0031	0.0023	0.0021
B11	0.0024	0.0044	0.0020	0.0005	-0.0010	0.0010	0.0020	0.0009	0.0015	0.0019	0.0009	0.0002	-0.0017	0.0030	-0.0020	0.0007	0.0055	0.0020	0.0029	0.0032	0.0015

Table 2 Observational data of border irrigation tests, Kostiakov parameters and Roughness coefficient

Dordor		Initial	First chan	ge inflow	Second cha	inge inflow	Cut-off	Kostiakov para	Roughness		
nı	umber	inflow rate /l s ⁻¹ ·m ⁻¹	Time of flow rate varying/min	ime of flow rate Inflow rate after Time of flow rate Inflow rate after varying/min varying/l $s^{-1} \cdot m^{-1}$ varying/min varying/l $s^{-1} \cdot m^{-1}$		Inflow rate after varying/l s ⁻¹ ·m ⁻¹	distance ratio	$k (\mathrm{mm} \cdot \mathrm{min}^{-\alpha}) \qquad \alpha$		coefficient n	
Al	constant	3.00	_	_	_	_	0.90	6.651	0.69	0.1	
A2	inflow	5.38	_	_	_	_	0.85	7.176	0.71	0.1	
A3	borders	6.40	_	_	_	_	0.80	6.549	0.69	0.1	
A4		3.04	6.0	4.20	14.4	5.71	0.85	6.568	0.69	0.1	
A5		2.79	6.0	4.09	15.3	7.02	0.80	6.924	0.70	0.1	
A6		3.10	5.4	5.55	11.4	7.16	0.80	6.782	0.70	0.1	
A7	varied	3.09	5.4	3.73	13.8	4.80	0.85	6.534	0.69	0.1	
A8	inflow	3.11	6.0	4.12	13.8	7.16	0.80	6.747	0.70	0.1	
A9	borders	2.42	6.6	5.91	14.4	7.33	0.80	7.567	0.72	0.1	
A10		2.97	6.0	4.04	14.4	8.37	0.80	6.305	0.67	0.1	
A11		3.16	5.4	4.91	13.2	5.44	0.85	6.927	0.70	0.1	
A12		2.93	6.0	5.22	12.6	7.14	0.80	6.466	0.68	0.1	
B1		2.40	—	_	—	_	0.90	13.466	0.45	0.14	
B2	constant	4.17	—	_	—	_	0.85	14.100	0.45	0.14	
B3	borders	6.50	—	_	—	_	0.75	13.942	0.45	0.14	
B4		4.17	—	_	—	_	0.85	15.585	0.42	0.14	
B5		6.50	10.2	4.17	_	_	0.85	14.259	0.45	0.14	
B6		6.50	10.2	2.40	14.7	4.17	0.85	13.322	0.46	0.14	
B7	varied	6.50	10.8	2.40	19.8	4.17	0.85	14.331	0.42	0.14	
B8	inflow	2.40	23.4	4.17	_	_	0.85	14.143	0.46	0.14	
B9	borders	2.40	22.8	6.50	33.6	4.17	0.85	15.674	0.42	0.14	
B10		2.40	27.6	6.50	39.0	4.17	0.85	13.731	0.48	0.14	
B11		2.40	23.4	6.50	34.2	4.17	0.85	13.942	0.45	0.14	

3.2 Validation

The applicability of the zero-inertia model in the border irrigation system with rapidly varied inflow was validated by comparing the predicted advance time, recession time and flow depth with observed field study data.

The predicted advance and recession trajectories by the zero-inertia model were compared with the field measured data and the result are presented in Figures 2 and 3. The results showed that the simulations were in good agreement with field data.

For details, in borders with constant inflow discharge (borders A1 to A3 and borders B1 to B4), the relative error was mostly within 10% for simulated advance times and 15% for recession times, and the average relative error was 6.36% and 9.46% for advance and regression times, respectively. In borders with varied inflow discharge, the relative error was mostly within 15% for simulated advance times and 20% for recession times, and the

average relative error was 7.24% and 10.32% for advance and regression times, respectively. The relative errors were similar to other results of the border irrigation simulation. For examples, Zhang et al.^[22] simulated the border irrigation system by Roe finite-volume method and improved hybrid numerical method, the average relative error of simulated advance and recession times was 4.9% and 7.38% for the improved hybrid numerical method, 6.88% and 9.45% for the Roe finite-volume method. Mahdizadeh Khasraghi et al.^[13] simulated the cultivated closed-end border irrigation system by the zero-inertia model and full hydrodynamic model in the SIRMOD software. Except for the abnormal experimental data (ROTH-16 test), the average relative error of simulated advance and recession times was 7.49% and 14.09% for the zero-inertia model, 7.49% and 15.36% for the full hydrodynamic model. Therefore, the simulation results in this study are acceptable.



Figure 2 Comparison of simulated and measured advance and recession trajectories for border A12, border B11



Figure 3 Observed versus predicted advance and recession times for the total data of borders with Zero-inertia model

Several statistics indicators were computed to further verify the accuracy of the zero-inertia model. The results showed that the zero-inertia model have a fair degree of accuracy (Table 3). In the irrigation events with constant inflow discharge, the values of R^2 were between 0.993 and 0.999 for simulated advance times, and between 0.874 and 0.940 for simulated recession times. And in the irrigation events with rapidly varied inflow, the values of R^2 were between 0.960 and 0.998 for simulated advance times, and between 0.852 and 0.961 for simulated recession times. Although the values of R^2 for the irrigation events with rapidly varied inflow was slightly smaller, it still showed a high correlation of the zero-inertia model with the measured data^[23]. The values of RMSE and CRM have the same tendency. In the irrigation events with constant inflow, the values of RMSE were between 0.496 and 1.793 min for simulated advance times, and between 1.852 and 15.757 min for simulated recession times respectively. The values of CRM were between -0.032 and 0.043 for simulated advance times, between -0.104 and 0.029 for simulated recession times respectively. And in the irrigation events with rapidly varied inflow, the values of CRM were between -0.103 and 0.047 for simulated advance times, between -0.103 and 0.014 for simulated recession times respectively. In the irrigation events with rapidly varied inflow, the values of RMSE were between 0.729 min and 2.317 min for simulated advance times, between 4.343 min and 17.280 min for simulated recession times respectively. These results agreed with other researches of the constant-discharge surface irrigation simulation. Mahdizadeh Khasraghi et al.^[13] used SIRMOD software to simulate 22 sets of border irrigation with constant inflow discharge, and the average value of RMSE was 1.11 min for simulated advance times and 10.93 min for simulated recession times. Sayari et al.^[24] used the SIRMOD software and a zero-inertia finite element model to simulate the flow in surface irrigation with constant inflow

discharge, and the RMSE values of simulated recession times were between 2.80 min and 12.58 min for the zero-inertia finite element model, between 2.26 min and 17.89 min for SIRMOD. Hence, the zero-inertia model is satisfactory to simulate the advance and recession phase in border irrigation with rapidly varied inflow discharges.

Border		A	dvance tin	ne	Recession time					
N	lumber	R^2	$R^2 \frac{RMSE}{/min}$ CRM R^2		R^2	RMSE /min	CRM			
A1	constant	0.999	0.496	-0.004	0.904	1.852	0.017			
A2	inflow	0.993	0.616	0.035	0.908	2.298	0.009			
A3	borders	0.994	0.515	-0.006	0.893	5.438	-0.046			
A4		0.960	1.963	-0.103	0.866	7.681	0.014			
A5		0.983	1.275	-0.062	0.873	6.933	-0.044			
A6		0.990	0.781	-0.035	0.892	6.282	-0.041			
A7	varied	0.992	0.862	-0.040	0.878	4.343	-0.062			
A8	inflow	0.992	0.830	-0.052	0.865	4.792	-0.088			
A9	borders	0.984	1.180	-0.036	0.852	5.303	-0.003			
A10		0.980	1.351	-0.020	0.909	6.203	0.007			
A11		0.967	1.634	-0.091	0.894	5.510	-0.038			
A12		0.988	0.898	-0.030	0.900	6.471	0.004			
B1		0.995	1.793	0.043	0.874	11.506	-0.022			
B2	constant	0.998	0.630	-0.018	0.930	14.391	-0.104			
B3	borders	0.993	0.937	0.010	0.878	14.849	-0.095			
B4		0.996	1.000	-0.032	0.940	15.757	0.029			
B5		0.997	0.729	0.031	0.917	11.698	-0.014			
B6		0.995	1.066	0.036	0.918	13.718	-0.103			
B7	varied	0.996	0.904	-0.030	0.961	15.138	-0.080			
B 8	inflow	0.995	1.407	-0.018	0.889	17.280	-0.090			
B9	borders	0.993	1.307	0.047	0.932	15.318	-0.022			
B10		0.985	2.317	-0.061	0.897	17.248	-0.016			
B11		0.998	0.740	-0.001	0.877	16.261	-0.029			

Table 3 Goodness of fit of Zero-inertia model

As water enters the border from the upper boundary, the flow depth fluctuates near the upper boundary. The flow depth observation points are designed at a quarter distance (20 m in Field A and 25 m in Field B) from the upper boundary of the border. Flow depths were measured by capacitance water level logger (resolution: 1 mm). The observation time interval of Field A (in 2016) was too long (about 2 min) to understand how the water

depth changes with the change inflow. Therefore, the observation time interval of Field B (in 2018) was set to 30 s. The comparisons of simulated and measured flow depths are shown in Figures 4 and 5. The simulated flow depths are closely matched with the experimental data with small relative error values. At the same time, Figure 5 also shows that the flow depth obviously changes with the variable inflow.



Figure 4 Comparison of simulated and measured flow depths for border A7, and border A11





3.3 Sensitivity analysis

Because of the uncertainty of the model input parameters (bed slope, Kostiakov infiltration parameters (α and k), roughness coefficient), systematic sensitivity analyses must be conducted to investigate the sensitivity of the performance to changes in these input parameters. The sensitivity analysis of α was only conducted because there is a certain correlation between α and $k^{[25]}$. Based on the input parameters of B5 (irrigation event with rapidly decreased inflow) or B8 (irrigation event with rapidly increased inflow), change only one input parameter equal (±50%) while keeping all other parameters unchanged to perform sensitivity analysis^[26]. The sensitivity index (SI) was used for this purpose, and it can indicate the percentage change in the output resulting from an input parameter changing^[26,27].

$$SI = \frac{100}{n} \sum_{i=1}^{n} (\frac{t_n - t_p}{t_p})_i \cdot \Delta^{-1}$$
(12)

where, t_p is the simulated advance/recession time, min; t_n is the new value of simulated advance/recession time with a changed input parameter, min; Δ is the absolute value of change in the input parameter; *n* is the number of stations along the border length.

The sensitivity of the zero-inertia model in B5 (rapidly decreased inflow) and B8 (rapidly increased inflow) is illustrated in Figure 6. In general, the infiltration parameter (α) was the most sensitive input variable of the zero-inertia model. In addition, the effect of model input (bed slope in particular) on the recession

phase was more than the advance phase. For -50% change in border slope, the output of advance times increased by 10.3% in B5 and 8.4% in B8, and that of recession times increased by 94.5% in B5 and 81.1% in B8. These results agreed with the research of Maheshwari et al.^[27] who studied the sensitivity analysis of slope for six border irrigation models in the border irrigation systems with constant inflow discharges. Their results also showed that the effect of the slope change on the recession phase is much greater than the advance phase.



Figure 6 Sensitivity analysis of the zero-inertia model

4 Conclusions

The zero-inertia model is a suitable and simple tool to simulate surface irrigation event with constant inflow discharge. Nevertheless, this model performance with rapidly varied inflow discharge was not verified. In this study, the zero-inertia model was solved using the finite difference method to verify the model with rapidly varied inflow discharge. The simulation results were compared with the experimental data. The evaluation of the model in simulating the advance and recession phases revealed that the zero-inertia model is also a good model for the surface irrigation system with rapidly varied inflow discharge. Sensitivity analysis of the input variables indicated that the infiltration parameter (α) was more influential than bed slope and roughness coefficient. Therefore, more attention should be paid to the measurement of infiltration parameters in rapidly varied inflow irrigation system. Moreover, for all input, especially bed slope, the recession phase was more sensitive than the advance phase.

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