Supporting Information for

Enhancing Inverse Modeling in Groundwater Systems through Machine Learning: A Comprehensive Comparative Study

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Training data number	Optimal algorithms	С	Е	σ	MSE
	GA	18.640	6.0117E-03	0.398	6.2652E-02
200	DE	27.526	4.8503E-03	0.391	6.2498E-02
200	PSO	27.526	4.8503E-03	0.391	6.2498E-02
	SA	35.533	8.3451E-06	0.334	6.2499E-02
	GA	54.278	4.9071E-03	0.509	4.9246E-02
500	DE	39.979	3.0950E-03	0.867	4.9729E-02
300	PSO	48.596	3.5939E-03	0.706	4.9215E-02
	SA	32.241	5.5964E-03	0.615	4.8987E-02
	GA	23.296	4.2424E-03	0.724	4.3391E-02
1000	DE	40.680	3.9406E-03	0.585	4.3556E-02
1000	PSO	25.317	6.1069E-03	0.820	4.3510E-02
	SA	71.104	4.0023E-05	0.561	4.3777E-02
	GA	61.888	1.1828E-03	0.918	3.5188E-02
2000	DE	53.579	1.6516E-03	0.964	3.5137E-02
2000	PSO	50.431	9.4148E-04	0.921	3.5120E-02
	SA	50.307	9.0781E-03	1.033	3.5265E-02

Table S1 Optimal hyperparameters for MSVR by four metaheuristic algorithms

Note: The rows in bold represent the optimal hyperparameter configurations corresponding to the smallest MSE values.

 $\underline{Table \ S2 \ RMSE_{(All)} \ values \ of \ FC-DNN \ with \ different \ number \ of \ hidden \ layers}$

Training data number	Hidden layer number							
	1	2	3	4	5	6	7	
200	0.07588	0.05882	0.06870	0.17916	0.16125	0.13690	0.13340	
500	0.07050	0.04308	0.03788	0.03786	0.05824	0.09567	0.10229	
1000	0.05118	0.03571	0.02703	0.02732	0.02866	0.04213	0.07825	
2000	0.03936	0.02944	0.02090	0.02168	0.02580	0.03064	0.06887	

Note: The bold values represent the smallest MSE values among the considered seven hidden layer numbers.

Table S3 R_{AII}^2 values of FC-DNN with different number of hidden layers

Training data	Hidden layer number						
number	1	2	3	4	5	6	7
200	0.94140	0.96479	0.95197	0.67332	0.73539	0.80926	0.81890
500	0.94942	0.98111	0.98540	0.98541	0.96548	0.90685	0.89351
1000	0.97334	0.98703	0.99256	0.99240	0.99164	0.98194	0.93768
2000	0.98424	0.99118	0.99555	0.99522	0.99323	0.99045	0.95173

Note: The bold values represent the largest R_{All}^2 values among the considered seven hidden layer numbers.

Scenarios -		Metahe					
		GA	DE	PSO	SA	I ININ <i>P</i>	1
Scenario 1	$N_{\rm PC} = 100$	0.7844	0.5984	0.9423	0.7720	epoch=200	0.4895
	$N_{\rm PC} = 500$	0.8246	0.7639	0.6379	0.8980	epoch=1000	0.4748
	$N_{\rm PC} = 1000$	0.6659	0.6391	0.7127	0.8012		
Scenario 2	$N_{\rm PC}=100$	0.9554	0.5223	0.8785	0.6987	epoch=200	0.4317
	$N_{\rm PC} = 500$	0.6164	0.4925	1.0293	1.1549	epoch=1000	0.4271
	$N_{\rm PC} = 1000$	0.5389	0.5322	0.9686	0.6288		
Scenario 3	$N_{\rm PC} = 100$	0.5386	0.3892	0.5486	0.5647	epoch=200	0.3161
	$N_{\rm PC} = 500$	0.4339	0.4271	0.5762	0.5714	epoch=1000	0.2970
	$N_{\rm PC} = 1000$	0.4060	0.5042	0.6295	0.5558		
Scenario 4	$N_{\rm PC} = 100$	0.4436	0.3841	0.5723	0.6459	epoch=200	0.2749
	$N_{\rm PC} = 500$	0.4265	0.3971	0.3770	0.5654	epoch=1000	0.2328
	$N_{\rm PC} = 1000$	0.3653	0.3459	0.5367	0.5033		

Table S4. RMSE values of estimated log-permeability fields for the four metaheuristic algorithms and the TNNA algorithm under Scenario 1-4.



Fig.S1. Detailed architecture of a LeNet based CNN. The input matrix data are obtained according to Figure 2(c) and subjected to feature extraction through a sequence of two convolutional and pooling layers, subsequently connected to the output layer using a flatten layer and two fully connected layers.



Fig.S2. Detailed architecture of a ResNet based CNN. The input matrix data are obtained according to Figure 2(c). "Res Block-1" and "Res Block-2" are two different types of residual blocks used in this ResNet. Eight residual blocks in four stages are designed in this ResNet. "Stage i (j)" represents the jth residual block used in stage i.



Fig.S3. Performance of MSVR based surrogate models for the solute concentration and hydraulic head prediction. (a~d) are pair-wise comparisons based on surrogate models trained by 200, 500, 1000, and 2000 training samples, respectively.



Fig.S4. Performance of FC-DNN based surrogate models for the solute concentration and hydraulic head prediction. (a~d) are pair-wise comparisons based on surrogate models trained by 200, 500, 1000, and 2000 training samples, respectively.



Fig.S5. Performance of LeNet CNN based surrogate models for the solute concentration and hydraulic head prediction. (a~d) are pair-wise comparisons based on surrogate models trained by 200, 500, 1000, and 2000 training samples, respectively.



Fig.S6. Performance of ResNet CNN based surrogate models for the solute concentration and hydraulic head prediction. (a~d) are pair-wise comparisons based on surrogate models trained by 200, 500, 1000, and 2000 training samples, respectively.



Fig.S7 Spatial distributions of log-permeability field estimation results (row 1, 3, and 5 for N_{PC} =100, 500, and 1000, respectively) and absolute errors (row 2, 4, and 6 for N_{PC} =100, 500, and 1000, respectively) for Scenario 1, achieved by four metaheuristic algorithms.



Fig.S8 Spatial distributions of log-permeability field estimation results (row 1, 3, and 5 for N_{PC} =100, 500, and 1000, respectively) and absolute errors (row 2, 4, and 6 for N_{PC} =100, 500, and 1000, respectively) for Scenario 2, achieved by four metaheuristic algorithms.



Fig.S9 Spatial distributions of log-permeability field estimation results (row 1, 3, and 5 for N_{PC} =100, 500, and 1000, respectively) and absolute errors (row 2, 4, and 6 for N_{PC} =100, 500, and 1000, respectively) for Scenario 3, achieved by four metaheuristic algorithms.



Fig.S10 Spatial distributions of log-permeability field estimation results (row 1, 3, and 5 for N_{PC} =100, 500, and 1000, respectively) and absolute errors (row 2, 4, and 6 for N_{PC} =100, 500, and 1000, respectively) for Scenario 4, achieved by four metaheuristic algorithms.



Fig.S11. Spatial distributions log-permeability field estimation results and absolute errors for Scenario 1, achieved by the TNNA inversion algorithm.



Fig.S12. Spatial distributions log-permeability field estimation results and absolute errors for Scenario 2, achieved by the TNNA inversion algorithm.



Fig.S13. Spatial distributions log-permeability field estimation results and absolute errors for Scenario 3, achieved by the TNNA inversion algorithm.



Fig.S14. Spatial distributions log-permeability field estimation results and absolute errors for Scenario 4, achieved by the TNNA inversion algorithm.



Fig.S15. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (*t*=2 day) using the TNNA algorithm and four metaheuristic algorithms.



Fig.S16. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (*t*=4 day) using the TNNA algorithm and four metaheuristic algorithms.



Fig.S17. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (*t*=6 day) using the TNNA algorithm and four metaheuristic algorithms.





Fig.S18. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (*t*=8 day) using the TNNA algorithm and four metaheuristic algorithms.



Fig.S19. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (t=10 day) using the TNNA algorithm and four metaheuristic algorithms.



distributions of calibrated numerical simulation results and absolute errors for solute concentrations (t=12 day) using the TNNA algorithm and four metaheuristic algorithms.





Fig.S21. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (t=14 day) using the TNNA algorithm and four metaheuristic algorithms.



Fig.S22 Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (*t*=16 day) using the TNNA algorithm and four metaheuristic algorithms. (a) Calibrated simulation results for solute concentrations (t=18 day)



Fig.S23 Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (*t*=18 day) using the TNNA algorithm and four metaheuristic algorithms.





Fig.S24 Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (*t*=20 day) using the TNNA algorithm and four metaheuristic algorithms.



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Fig.S26. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (t=24 day) using the TNNA algorithm and four metaheuristic algorithms.





Fig.S27. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (t=26 day) using the TNNA algorithm and four metaheuristic algorithms.



Fig.S28. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (*t*=28 day) using the TNNA algorithm and four metaheuristic algorithms. (a) Calibrated simulation results for solute concentrations (t=30 day)



Fig.S29. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (t=30 day) using the TNNA algorithm and four metaheuristic algorithms.





Fig.S30. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (*t*=32 day) using the TNNA algorithm and four metaheuristic algorithms.



Fig.S31. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (*t*=34 day) using the TNNA algorithm and four metaheuristic algorithms. (a) Calibrated simulation results for solute concentrations (t=36 day)



Fig.S32. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (t=36 day) using the TNNA algorithm and four metaheuristic algorithms.





Fig.S33. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (t=38 day) using the TNNA algorithm and four metaheuristic algorithms.



Fig.S34. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (*t*=40 day) using the TNNA algorithm and four metaheuristic algorithms. (a) Calibrated simulation results for solute concentrations (*t*=42 day)



Fig.S35. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (t=42 day) using the TNNA algorithm and four metaheuristic algorithms.



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Fig.S37. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (*t*=46 day) using the TNNA algorithm and four metaheuristic algorithms. (a) Calibrated simulation results for solute concentrations (*t*=48 day)



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Fig.S42. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (*t*=56 day) using the TNNA algorithm and four metaheuristic algorithms.



Fig.S43. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (t=58 day) using the TNNA algorithm and four metaheuristic algorithms.



Fig.S44. Spatial distributions of calibrated numerical simulation results and absolute errors for solute concentrations (t=60 day) using the TNNA algorithm and four metaheuristic algorithms.