





$$Q_i = \sum_{j=1}^m h_j q_j \begin{cases} i = 1, 2, \dots, l \\ j = 1, 2, \dots, m \end{cases} \quad (1)$$

$$\left. \begin{matrix} \\ \\ \end{matrix} \right\} i - j + 1 = 1, 2, \dots, n$$

where  $Q_i$  is the basin outlet section of each period discharge in m<sup>3</sup>/s;

$h_j$  is the rainfall in each period in mm;

$q_{i-j+1}$  is the ordinate of the unit hydrograph in each period in m<sup>3</sup>/s;

$i$  is the number of periods for the basin outlet section flow hydrograph;

$j$  is the net number of rainfall periods; and

In essence, the unit hydrograph is the characteristic of watershed concentration in the form of discharge hydrograph, i.e. concentration curve (d.johnstone, 1949). The calculation method of unit hydrograph confluence takes the basin as a whole and assumes that the net rain is uniformly distributed over the whole basin, without considering the inhomogeneity within the system; the basin confluence system is a linear time-invariant system, and at the same time, it is viewed that the net rainfall and the formation of the flow process are in agreement to superposition relationship. Therefore, the essential characteristics of the unit hydrograph are lumped resistance, linearity, and time invariance.

Conceptually, the unit hydrograph is a linear time-invariant basin system with a convergent flow curve. However, the physical mechanism of the watershed conflux is not considered by the method used to derive the unit hydrograph (Ramirez, 2000). The principle used to calculate the unit hydrograph is based on the system input (rainfall), which is converted using the unit hydrograph to determine the system response output (outlet section flow process), where the error is minimized. The traditional methods of derivation are as follows.

Analytical approach: the basin outlet section of the surface runoff is  $Q_1, Q_2, \dots, Q_l$ , the

rainfall process is  $h_1, h_2, \dots, h_m$ , where Eqn. (1) comprises  $q_1, q_2, \dots, q_n$  unknown linear

algebraic equations. The solutions of the equations can be obtained using a unit hydrograph.

$$q_i = \frac{Q_i}{h_1} \sum_{j=1}^m \frac{h_j q_j}{10^{i-j+1}} \begin{cases} i = 1, 2, \dots, n \\ j = 1, 2, \dots, m \end{cases} \quad (2)$$

where  $n$  is the number of periods of the unit hydrograph and  $n = l - m + 1$ .

In theory, there are no errors with the analytical approach when using the rainfall runoff measurements. This approach can obtain the correct answer if the watershed conflux conforms to a



64 linear time-invariant system (when the convergence of watershed system meets multiple  
65 proportions and superposition assumption, for a linear system). However, (1) the data  
66 measurements have errors and, (2) rainfall-runoff system is not a linear time-invariant system. Due  
67 to the accumulation of errors, unreasonable solutions often occur. For example, the unit  
68 hydrograph may be irregular or appear to be negative.

69 (b) Trial and error: the unit hydrograph is assumed to be  $q'_i$ . The flow  $Q'$  determined by  
70 the unit hydrograph is then compared with the measured flow  $Q$ . The unit hydrograph  $q'_i$  is  
71 produced when the error between  $Q'$  and  $Q$  satisfies certain error.

72 The trial and error method proposed by Collins (Collins, 1939) uses an iterative strategy. For  
73 a period of uneven net rainfall, the computational convergence is fast if the time period is large.  
74 However, the Collins iterative method has the following disadvantages: (1) the iterations are based  
75 on the initial unit hydrograph, but there is no strict method for selecting the initial unit hydrograph;  
76 and (2) trial-error approach does not necessarily converge to a solution.

77 (c) Least squares method: the measured surface runoff is assumed to be  $Q$  and the error is  
78 as follows:  $\varepsilon = Q - Q'$ . If  $\sum \varepsilon^2 = \sum (Q - Q')^2 \rightarrow \min$ , we try to convert Eqn. (1) into a  
79 normal equations system where the number of equations is equal to the variables. The optimal  
80 estimation of  $q'_i$  can be solved using the least squares method. The theory of the least squares  
81 method is better, but the results are sometimes fluctuating or negative.

82 There are many methods for determining unit hydrographs, e.g., the Z transform method and  
83 the harmonic analysis method (Dodge, 1973). A previous study (Hanson and Johnson 1964 )  
84 classified and compared the usual unit hydrograph calculation methods.

85 In recent decades, the use of probability distribution functions (pdfs) to develop synthetic unit  
86 hydrographs (SUH) has received much attention because of its similar properties to unit  
87 hydrographs. First, the type of unit hydrograph needs to be subjected to mathematical analysis.  
88 Typical functions, such as a parabola P-III (Yuan, et al., 1991) (Zhai and Li, 2004) curve, can be  
89 used to describe the unit line and a mathematical model of the unit hydrograph can be established.  
90 A previous study (Bhunya, et al., 2007) explored the potential of using four popular pdfs, i.e.,  
91 two-parameter Gamma, three-parameter Beta, two-parameter Weibull, and one-parameter  
92 Chi-square distribution, for deriving a SUH. The Gamma functions are the most widely used  
93 functions (Singh, 2005, 2009; Bhunya, et al., 2003). This approach aims to determine the  
94 relationship between each variable in a unit hydrograph, which facilitates a more in-depth analysis  
95 of a unit hydrograph.

96 In the present study, we used Gamma functions to describe a unit hydrograph and determined  
97 why a unit hydrograph may follow this distribution. The Gamma function parameters were



98 optimized using a genetic algorithm. Finally, the unit hydrograph was obtained using the Collins  
99 iterative method.

100 For a specific basin, the confluence time of a flood is relatively stable and can be determined  
101 according to the flood datas. Therefore, the use of gamma function to derive unit hydrograph is  
102 only in the trial calculation of parameters  $\beta$  and  $k$ . The unit hydrograph expressed by gamma  
103 function based on the combination of these parameters is optimal, which is compared with the use  
104 of P-III function to adjust its statistical parameters. The principle of hydrological frequency  
105 calculation by line fitness of numbers (mean, variation coefficient  $C_v$  and skewness coefficient  $C_s$ )  
106 is very similar. Meanwhile, the time interval  $t$ , parameters and values of gamma function can  
107 parameterize the regional characteristics of river basin confluence, which has the important  
108 significance for this study.

## 109 1 Mathematical model and method

### 110 1.1 Mathematical model

111 In spatial or temporal physical entropy-based modeling of hydrology and water resources, the  
112 cumulative distribution function (CDF) of a design variable (e.g., a flux or a discharge) is  
113 analyzed in terms of its concentration (e.g., stage of flow) (Cui, et al., 2012). The theory of  
114 composition proposed by Zhang (2003) provides a model and a uniform calculation method for  
115 studying the composition of things. This theory considers the analysis of three concepts, i.e., the  
116 general set, the distribution function, and the degree of complexity. This theory is also considered  
117 the most highly approved principle followed by random systems, i.e., the entropy principle.

118 The variable  $x$  is continuous and random, and can be viewed as a general set of flag  
119 variables. If the pdf  $f(x)$  of  $x$  agrees with the following function:

$$120 \quad f(x) = \frac{\beta^k}{(k-1)!} x^{k-1} e^{-\beta x}, x > 0 \quad (3)$$

121 Then the pdf follows a Gamma distribution, where  $\beta$  and  $k$  are shape and scale parameters.  
122 This is one of the famous Pearson pdfs, which is known as a Pearson type III distribution. The  
123 curve has a peak with a left-right asymmetry. In nature, many phenomena follow this distribution.  
124 In China, hydrological studies often use the Pearson type III distribution to simulate hydrological  
125 data series, because it has a greater than or equal to zero lower bound on the variable requirements  
126 and its elasticity is greater than the normal distribution (Ye and Xia, 2002). This choice is based  
127 on experience, but it lacks a theoretical justification.

128 Using entropy theory, a previous study (Zhang, 2003) described the physical form of this  
129 distribution. By analyzing the structure of Eqn. (3), is not difficult to show that it has a negative  
130 exponential distribution, which is a part of the exponential function, and it also has the  
131 characteristics of a power function in a Pareto-family distribution. The exponential distribution



132 corresponds to the constraints on the invariant algebraic average of the flag variables, while the  
 133 power function corresponds to the constraints on the invariant geometric mean. It may be  
 134 speculated that the constraints on the Gamma distribution are the fixed algebraic average and  
 135 geometric means of the variables.

136 In this study,  $f(x)$  is the pdf of a positively defined random variable, i.e.,

$$137 \int_0^{\infty} f(x) \quad (4)$$

138  $u$  represents the algebraic average of variables, thus

$$139 u = \int_0^{\infty} xf(x)dx \quad (5)$$

140 while  $v$  is the geometric mean of the random variable  $x$ , which can be expressed as the  
 141 algebraic average of the logarithm, i.e.,

$$142 v = \int_0^{\infty} \ln xf(x)dx \quad (6)$$

143 The entropy of the random variable  $x$  can be written as

$$144 H = - \int_0^{\infty} f(x) \ln xf(x)dx \quad (7)$$

145 Given the constraints in Eqns. (4), (5), and (6), the Lagrange method can be used to estimate  
 146 the distribution function  $F$  based on the maximum entropy to determine the distribution  
 147 function. Thus,  $F$  is defined as follows:

$$148 F = - \int_0^{\infty} f \ln f dx + C_1 \left( \int_0^{\infty} f dx - 1 \right) + C_2 \left( \int_0^{\infty} xf dx - u \right) + C_3 \left( \int_0^{\infty} \ln xf dx - v \right) \quad (8)$$

149 Where,  $C_1$ ,  $C_2$ , and  $C_3$  are undetermined constants. The entropy principle demands that the  
 150 value of  $F$  is maximal. The partial derivative of  $f(\bullet)$ , i.e., the partial derivative is 0, can be  
 151 obtained using Eqn (8). The results are as follows.

$$152 f(x) = \exp(C_1 - 1 + C_2x + C_3 \ln x) \quad (9)$$

153 This formula can be used to obtain the distribution function. It is the product of the power  
 154 function and the exponential function, and its form is identical to a Gamma function.

155 Hydrological data are random variables that exceed zero. If the hydrological processes are  
 156 stationary, then the algebraic average and geometric mean of the hydrological characteristics  
 157 variable can be approximated as a fixed value. For example, a mean basin annual runoff is  
 158 basically stable (the algebraic average is constant), so the probability of a major flood occurring is

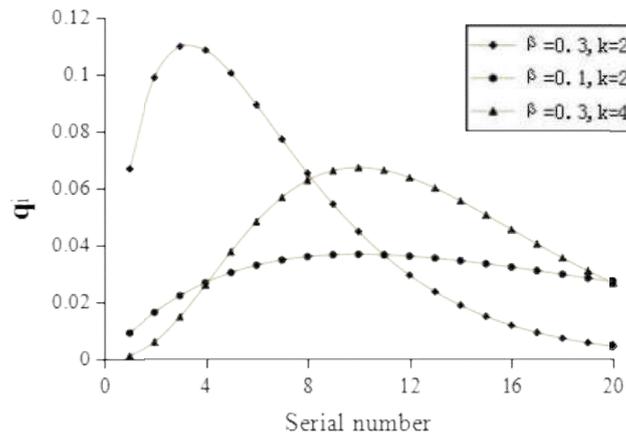


159 small, and most floods are close to the normal value of the accumulated years (the geometric mean  
 160 is constant). However, the uncertainty of a different type of flood occurring each time is maximal.  
 161 Thus, the complexity of the outcome is maximized. This is consistent with the following: “In a  
 162 generalized (objective, system, sampling experiment), if the algebraic average and geometric  
 163 mean of the variables (values of statistical indicators) are constant and the complexity is maximal,  
 164 then we can conclude that the probability (the percentage) of the flag value (all the values of the  
 165 variables) for each individual must obey a Gamma distribution (Pearson type III distribution)  
 166 (Zhang, 2003).”

167 At the same time, a lot of practical experiences showed that Gamma function can reflect the  
 168 characteristics of flood probability. In the view of this, the unit line  $q_i$  is defined as Gamma  
 169 function.

$$q_i = \frac{\beta^k i^{k-1} e^{-\beta i}}{(k-1)!}, i = 1, 2, \dots, n \quad (10)$$

171 When the parameters of  $\beta$  and  $k$  value were different types, the line type of  $q_i$  was  
 172 different, as shown in Fig. 1.



173

174

Fig. 1 The change of function under different parameters

175 **1.2 Method**

176 **1.2.1 Genetic algorithms**

177 Genetic Algorithms (GAs) is an effective global search method, which simulates natural  
 178 selection and genetic mechanism. This method of searching the optimal solution of the problem  
 179 through natural evolutionary process has been applied in many problems such as function



180 optimization and combinatorial optimization. Genetic algorithm can automatically acquire and  
181 accumulate the knowledge of search space in the search process, and adaptively control the search  
182 process to find the best solution(Davis,1991;Michalewicz,1996). The genetic algorithm regards a  
183 family of randomly generated feasible solutions as the parent population, takes fitness function  
184 (objective function or one of its transformation forms) as the measurement of the ability of the  
185 parent individual to adapt to the environment, generates the offspring individual through selection  
186 and hybridization, and then mutates the latter, eliminating the fittest and the fittest, so that the  
187 individual adapts to the environment through repeated evolutionary iterations. With the continuous  
188 improvement of ability, excellent individuals keep approaching the optimum point(Yuan,2002).  
189 After several generations, the algorithm converges to the best individual. The best individual in a  
190 group is likely to be the optimal or approximate optimal solution of the problem.

191 As a new random search and optimization method to simulate biological evolution, genetic  
192 algorithm has been widely used in the field of optimization(Chen,1996;Li,2009). The parameter  
193 optimization of many empirical formulas of hydrological models is essentially based on the global  
194 optimization ability of genetic algorithm.

### 195 1.2.2 Collins iteration method

196 Iterative method is a mathematical process to solve the problem by finding approximate  
197 solutions that meet the restrictive conditions from an initial value. Iterative algorithm is also a  
198 basic method to solve problems by computer. It makes use of the characteristic of fast computing  
199 speed and suitable for repetitive operation, so that the computer can repeat a set of instructions (or  
200 steps). When the instructions (or steps) are executed, a new value of the variable will be derived  
201 from the original value of the variable. Assuming that we want to derive an approximate solution,  
202 we should determine an initial value, an iteration function and a restriction condition according to  
203 the actual situation and data firstly, until the absolute value of the initial value and the calculated  
204 approximation value is less than a certain value. That is to say, we find the exact desired value.

## 205 2 Approach used to determined the unit hydrograph

206 The overall calculation process is divided into two parts: (1) the parameters of the unit  
207 hydrograph are optimized using the genetic algorithm, so the initial unit hydrograph can be  
208 calculated; and (2) the final unit hydrograph is calculated using the Collins iterative method.

### 209 2.1 Calculation of the initial unit hydrograph using a genetic algorithm

210 (a) Parametrization of Gamma Function

211 For simplicity, Eqn. (10) is transformed as follows

$$212 \begin{cases} q_i = \frac{1}{\Gamma(x)} \cdot x^{x-1} \cdot e^{-x} & i = 1, 2, \dots, n-1 \\ q_n = q^2 = 0 \end{cases} \quad (11)$$



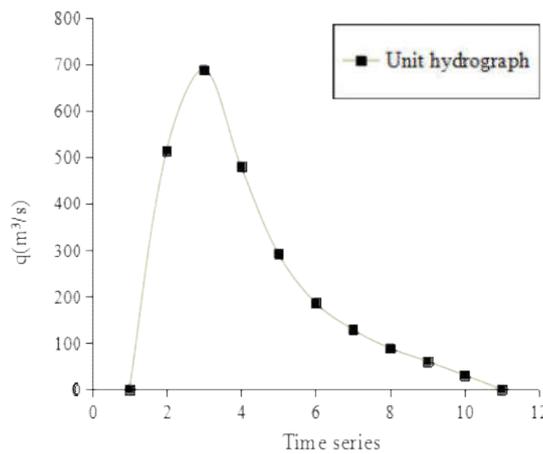
213 Where,  $x_1$ ,  $x_2$ , and  $x_3$  are the constants of the unit hydrograph  $q_i$ , which is the argument  
 214 of the problem for the genetic algorithm.  $i$  is the argument of the unit hydrograph, which is a  
 215 period number.  $i$  is the known number, which is processed by the genetic algorithm.  $n$  is the  
 216 time period of the unit hydrograph, which is defined according to the actual engineering problem.

217 The variables  $x_1$ ,  $x_2$ , and  $x_3$  are in a range of [0, 5]. The chromosome is coded as a floating  
 218 point value.

219 If one chromosome is  $v^i = [1.7450, 8.7014, 1.5042]$  and  $n = 11$  ( $n$  is the time period of  
 220 the unit hydrograph), the unit hydrograph obtained using the constant given above is as follows.

$$\begin{cases} q_i = \frac{1.7450^{8.7014}}{(8.7014-1) \cdot (i+1.5042)} \cdot e^{(8.7014-1) \cdot (-1.7450/(i+1.5042))} & i = 2, 3, \dots, 10 \\ q_1 = q_{11} = 0 \end{cases} \quad (12)$$

222 Using Eqn. (12), the results obtained for the unit hydrograph are [0, 570, 688, 563, 356, 187, 223  
 86, 35, 13, 5, 0], as shown in Fig. 2.



224

225 **Fig. 2** Unit hydrograph  $q_i$

226 (b) Determination of objective function

227 According to the basic principle used to derive the unit hydrograph, the objective function of  
 228 the genetic algorithm can be expressed as follows:

$$\max : \varphi(q(x_1, x_2, x_3)) = \frac{1}{(Q'(q_i) - Q)^2} \quad (13)$$

229



230 Where  $Q_i(q)$  is the converging flow obtained by the unit hydrograph  $q_i(x_1, x_2, x_3)$  in a

231 basin, and  $Q$  is the measured flow in the basin outlet section.

232 Physical interpretation of the objective function Eqn. (13)

233 A set of parameters,  $x_1^*$ ,  $x_2^*$ , and  $x_3^*$ , are searched based on the following conditions. The

234 inverse square of the difference between  $Q_i(q)$  and  $Q$  is maximal. To avoid computations if

235 the objective function value is too small, the expansion coefficient  $M$  is introduced. The value

236 of  $M$  is determined according to the specific situation. Eqn. (13) is converted into the following

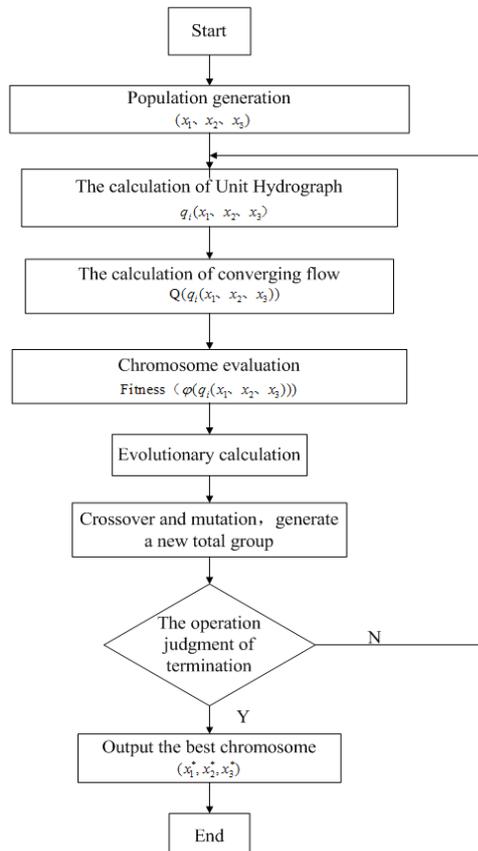
237 form.

$$\max : \varphi(q_i(x_1, x_2, x_3)) = \frac{M}{(Q_i(q) - Q)^2} \quad (14)$$

238

239 (c) Optimization parameters

240 We use the steps shown in Figure 3 of genetic algorithm to optimize the unit lineparameters:



241



242 **Fig. 3** The processes map that was used the genetic algorithm to optimize the parameters

243 The optimum parameters  $(x_1^*, x_2^*, x_3^*)$  are obtained by the above steps, and then the unit line  
 244 is calculated by the optimum parameters, as follows:

$$245 \begin{cases} q_i = \frac{x_1^* x_2^*}{(x_2^* x - 1)^2} \cdot (i + x_3^*)^{\frac{x_1^* - 1}{x_2^*}} \cdot e^{-x_3^* \cdot (i + x_3^*)}, i = 2, \dots, n - 1 \\ q_1 = q_n = 0 \end{cases} \quad (15)$$

246 **2.2 Calculation of the final unit hydrograph using the iterative method**

247 Collins iteration method is used to calculate the final unit hydrograph. Firstly, each period of  
 248 the net rainfall runoff process is calculated by unit hydrograph  $q_i^*$ . Meanwhile, the maximum net

249 rainfall  $h_{max}$  and its runoff process  $Q(q_i^*, h_{max})$  are determined, and overall net rainfall total

250 runoff  $\sum Q(q_i^*, h_{max})$  is calculated; Secondly, another unit hydrograph

251  $q_i' = \frac{Q - \sum Q(q_i^*, h_{max})}{h_{max}}$  is deduced, and new  $q_i'$  is deduced continuously by  $q_i^*$  according to

252 the restriction condition of  $\varepsilon = |q_i^* - q_i'| \leq ErrorExcepted$  until the error between the two

253 units meets the requirement, then the final unit hydrograph  $q_i = q_i'$ .

254 **3 Examples**

255 **Example 1**

256 In Table 1, the data were taken from a previous study (Zhuang and Lin, 1986).

257 **Table 1** The calculations of example 1

258 R: unit hydrograph; Q: measured discharge;  $Q'$ ,  $q'$ : discharge and unit hydrograph of trial and error method;  
 259 discharge and unit hydrograph of GACIM

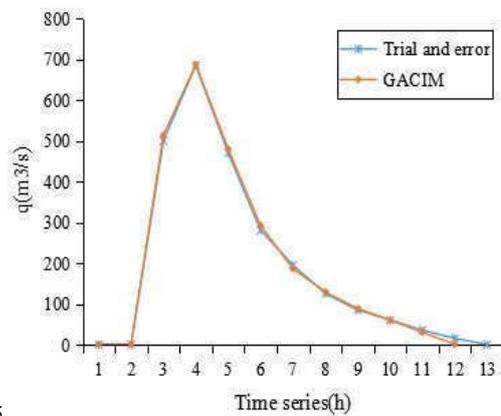
Time series(h)	R(mm)	The measured runoff Q(m <sup>3</sup> /s)	The trial and error method		GACIM	
			$Q'$ (m <sup>3</sup> /s)	$q'$ (m <sup>3</sup> /s)	$Q''$ (m <sup>3</sup> /s)	$q''$ (m <sup>3</sup> /s)
①	②	③	④	⑤	⑥	⑦
0		0	0	0	0	0
6	3.8	0	0	0	0	0
12	3.9	50	190	500	195	514
18	0	252	455	685	461	687
24	27.3	662	446	470	450	480
30	2.9	1700	1650	280	1700	292



36	2210	2200	195	2210	186
42	1630	1610	125	1630	129
48	1020	981	85	1020	88
54	650	669	60	650	60
60	440	433	35	440	30
66	290	288	15	290	0
72	190	195	0	190	
78	100	113		100	
84	40	51		9	
90	0	4		0	

260

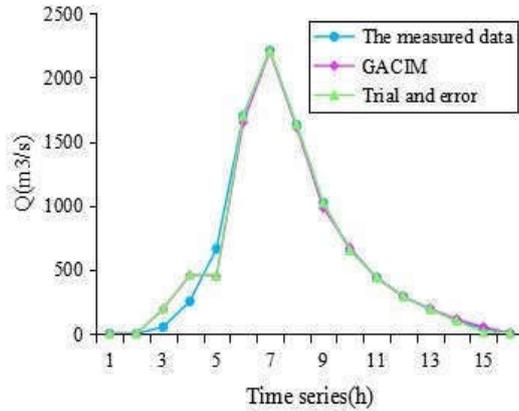
261 The unit hydrographs determined using the two methods are shown in columns (5) and (7) in  
 262 Table 1. A comparison of the unit hydrographs is shown in Figure 4. The flow processes  
 263 calculated using the unit hydrographs are shown in columns (4) and (6) in Table 1. A comparison  
 264 between the calculated flow process and the measured flow is shown in Figure 5.



265

266

Fig. 4 Example1: Unit hydrographs ascertained by two methods



267

268

**Fig. 5** Example1: The flow process of outletsection

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The actual hydrological data in Table 1 show that the period number for the runoff was 5 and the period number for the flow process in the outlet section was 15. According to the theory of the unit hydrograph, the period number for the unit hydrograph should be  $15 - 5 + 1 = 11$ . Using the GACIM (genetic algorithm and the Collins iterative method), the period number for the unit hydrograph was 11. Using the trial and error method, the period number for the unit hydrograph was 12. To further consider the performance of the two methods, we compared the flow process in the outlet section and the results obtained using the two unit hydrographs (measured value and calculated value). A statistical analysis of the results is shown in Table2.

277

**Table 2** The error statistics of example 1

Project Method	GACIM	The trial and error
The error of flood peak(m³/s)	0	10
The maximum error of discharge(m³/s)	212	216
The average absolute error of discharge(m³/s)	37.31	46.19
The total error of flood peak discharge(m³/s.h)	-111	-51
The relative error of flood peak discharge (%)	-1.20	-0.55

278

**279 Example 2**

280

In Table 3, the data were taken from a previous study (Li and Zheng, 1982).

281

**Table 3** The calculations of example 2

Time series(h)	R(mm)	The measured runoff Q(m³/s)	The trial and error method	GACIM
----------------	-------	-----------------------------	----------------------------	-------



	$Q'$ (m <sup>3</sup> /s)	$q'$ (m <sup>3</sup> /s)	$Q''$ (m <sup>3</sup> /s)	$q''$ (m <sup>3</sup> /s)		
①	②	③	④	⑤	⑥	⑦
0		0	0	0	0	0
6	15.3	97	96	63	97	49
12	7.4	214	215	110	214	119
18	5.8	304	308	124	304	149
24		371	374	143	371	128
30		294	294	76	294	85
36		190	202	41	190	47
42		123	120	30	123	23
48		80	80	22	80	10
54		49	52	12	49	4
60		30	22	0	19	0
66		15	7		4	
72		0	0		0	

282

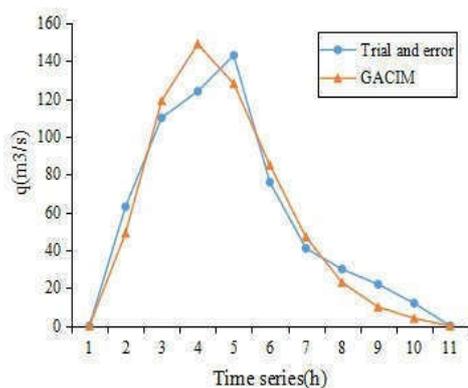
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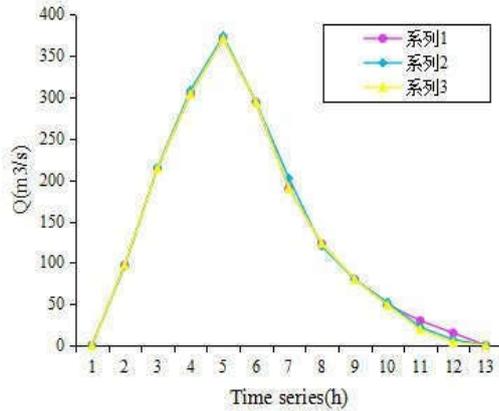
The unit hydrographs determined using the two methods are shown in columns (5) and (7) in Table 3. A comparison of the unit hydrograph is shown in Figure 6. The flow processes calculated using the unit hydrographs are shown in columns (4) and (6) in Table 3. A comparison of the calculated flow process and the measured flow is shown in Figure 7.



287

288

Fig. 6 Example2: Unit hydrographs ascertained by two methods



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290

Fig. 7 Example2: The flow process of outletsection

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Figure 4 shows that GACIM was significantly better than the trial and error method in terms of the shape of the curve. We also compared the flow process in the outlet section and the data obtained using the two unit hydrographs. A statistical analysis of the results is shown in Table4.

294

Table 4 The error statistics of example2

Project Method	GACIM	The trial and error
The error of flood peak(m3/s)	0	-3
The maximum error of discharge(m3/s)	11	-12
The average absolute error of discharge(m3/s)	1.69	-0.23
The total error of flood peak discharge(m3/s.h)	22	-3
The relative error of flood peak discharge (%)	1.25	-0.17

295

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From Table 2 and Table 4, the calculation accuracy of BGACM is obviously better than that of trial-and-error method in most projects. Although the total error of flood volume is larger than that of trial-and-error method, the relative error of flood volume is only 1.2% and 1.25%, so it does not affect the application of actual projects.

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Figure 6 and 7 show that the flow processes of the two unit hydrographs were similar. However, a comparison of the shapes of the unit hydrograph showed that the continuity and smoothness of GACIM were better than the trial and error method. The GACIM method conformed better with the features of a time-invariant system.

304

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It can be seen that BGACM method is better at simulating river basin confluence process, which depends on the physical mechanism of the algorithm, while trial-and-error method pays more attention to the balance of total flood volume. This is the respective characteristics and advantages of the two algorithms exactly.



#### 308 4 Discussion

309 The present study used a combination of a genetic algorithm and the Collins iterative method  
310 (GACIM) for determining a unit hydrograph. The method and implementation steps were  
311 described, while examples and analyses were used to demonstrate the scientificity, reliability and  
312 practicability of this method. The outcomes of this study are discussed below.

##### 313 (a) Difference between GACIM and other methods

314 In principle, GACIM is based on composition theory and it describes the physical mechanism  
315 and process of flood confluence using mathematical equations. Using the basic concept of the unit  
316 hydrograph and a genetic algorithm as a mathematical tool, this method can be used to simulate  
317 the flood confluence process.

318 Therefore, the simulation of the convergence process is more accurate with GACIM.

319 Other methods for calculating unit hydrographs include the analysis method, least squares  
320 method, and the trial and error method. These methods are more focused on the unit hydrograph as  
321 an outlet flow process and they fit the measured flow precisely, but they ignore the composition  
322 and structure of the unit hydrograph itself. Example 2 shows that GACIM performed better at  
323 simulating the basin confluence process, whereas other methods paid more attention to the balance  
324 of the total flood volume.

##### 325 (b) Genetic operator design issues

326 A genetic algorithm is a very useful optimization tool. Its biggest advantage is that it has  
327 wide adaptability and unlimited problem space, so it can handle many different constraints. This  
328 strategy uses a penalty factor. This is because the genetic algorithm method delivers exhaustive  
329 engineering accuracy if the population is sufficiently large.

330 There are two types of genetic algorithm, i.e., the standard genetic algorithm (crossover and  
331 mutation) and evolutionary computing (selection). A genetic algorithm simulates the  
332 recombination of genes to create new offspring in each generation, whereas evolutionary  
333 computation is a population process that updates each generation.

334 In this study, a genetic algorithm was used to optimize the parameters of the Gamma function  
335 and the unit hydrograph was calculated according to the law of basin confluence. Thus, the  
336 parameters were generated by a genetic algorithm. Therefore, the design of the genetic operators is  
337 related directly to whether reasonable generation parameters could be obtained.

338 A genetic algorithm has two components: crossover and mutation. Crossover is the main  
339 genetic operation that generates new individuals, but it also maintains the relative stability of the  
340 population at the same time. However, the variation is a basic calculation and the main effect is to  
341 produce a new gene from the population, which provides new information for the population.

342 In general, the initial population of the genetic algorithm is generated in the value space.  
343 Crossover and mutation are performed in the value space. In the present study, the value of the  
344 Gamma function was in a certain range. Initially, we could not define a reasonable space. If the  
345 value space is too large, a bigger population must be used to meet the needs of the individual



346 distribution density. However, this greatly reduces the computational speed. If the value space is  
 347 too small, it might not meet the parameter combination required for the engineering precision. To  
 348 solve this problem, we observed the following principles during the design of the genetic operator.

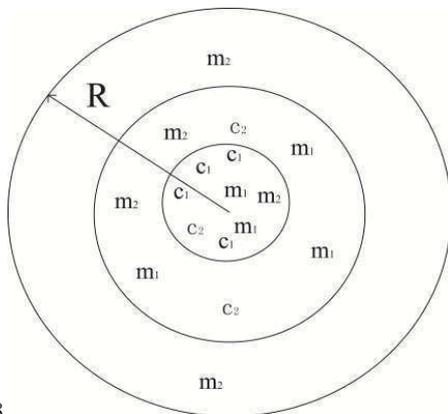
349 (i) The crossover operator was determined where a new individual was generated at random  
 350 in the space  $[0, a]$  ( $a > 0$ ).

351 (ii) The random expansion value space of the mutation operator was  $[0, a]$  and its amplitude  
 352 was random. For the floating point coding mutation operator, the following program was  
 353 implemented in MATLAB:

- 354 1) numMut=round(size(parent,1)\*Ops/2); to calculate the number of variations
- 355 2) numPop=size(parent,1); to calculate the size of the population
- 356 3) numPara=size(parent,2); to calculate the number of parameters
- 357 4) for j=1:numPara
- 358 5) for i=1:numMut
- 359 6) a=round(rand\*(numPop-2)+2); select a male parent
- 360 7) parent(a,numPara) = parent(a,numPara)\*(1+rand/gen); generate a new generation
- 361 8) end
- 362 9) end.

363 The parameters of the offspring chromosome were calculated during step 7) of this program,  
 364 where the variation in the amplitude was related to the number of evolutionary passages. The  
 365 variation in the amplitude declined gradually with increasing passage numbers.

366 Figure 8 illustrates the crossover operator and mutation operator with the passage of  
 367 evolutionary time.



368  
 369 **Fig. 8** The sketch map of the continuation of parameters valuespace  
 370

371 In the first generation, the filial generation caused by the crossover operator was still in the  
 372 initial parameter space, which corresponds to the inner loop in Figure 8. However, the filial  
 373 generation caused by the mutation operator was beyond this range and it expanded to the second



374 ring. In the second generation, the filial generation of the crossover operator was extended to the  
375 second ring while the filial generation of the mutation operator was extended to the third ring. The  
376 expanding amplitude of the adjacent ring decreased with the passage of evolutionary time. This  
377 method was repeated until the predetermined evolutionary algebra was completed.

378 In Examples 1 and 2, the initial values of the parameters were [0 5]. The two sets of  
379 optimized parameters were as follows.

380 Example 1: [1.7450, 8.7014, 1.5042]

381 Example 2: [1.6052, 7.9209, 0.30007]

382 These two examples demonstrate the design rationality and the validity of the genetic  
383 operator.

384 (c) Research methods for hydrological analysis and calculation

385 The factors that affect hydrological phenomena are very complex. There is still no accurate  
386 understanding of the causal relationships among hydrologic phenomena. It is considered that  
387 hydrological phenomena involve certainty and randomness, which form the basis of hydrological  
388 research. Therefore, causal analysis and probabilistic statistics are the main methods used for  
389 hydrological analysis and calculation. In practical applications, causal analysis is confined mostly  
390 to qualitative analysis. Quantitative problems demand empirical statistical relationships based on  
391 actual observational data.

392 Based on the theory of composition, the distribution function in statistical physics has been  
393 extended to hydrology as a non-physics field. Thus, hydrological systems can be viewed as a  
394 generalized collection. The regularities of hydrological phenomena have been simulated using  
395 distribution functions. Distribution functions and functional relationships have been determined  
396 using observation data, which generally means that objective laws are formalized.

397 The present study was a preliminary attempt to investigate the quantitative relationships  
398 among hydrological phenomena based on the theory of composition and its distribution function.  
399 The author believes that this theory could be a new approach to exploring hydrological rules.

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