APPLICABILITY OF ARTIFICIAL NEURAL NETWORK MODEL FOR SIMULATION OF MONTHLY RUNOFF IN COMPARISON WITH SOME OTHER TRADITIONAL MODELS

Le Van Duc
University of Technology, VNU-HCM
(Manuscript Received on August 08th, 2008, Manuscript Revised May25th, 2009)

ABSTRACT: Artificial Neural Network (ANN) model along with Back Propagation Algorithm (BPA) has been applied in many fields, especially in hydrology and water resources management to simulate or forecast rainfall runoff process, discharge and water level - time series, and other hydrological variables. Several researches have recently been focusing to compare the applicability of ANN model with other theory-driven and data-driven approaches. The comparison of ANN with M5 model trees for rainfall-runoff forecasting, with ARMAX models for deriving flow series, with AR models and regression models for forecasting and estimating daily river flows have been carried out. The better results that were implemented by ANN model have been concluded. So, this research trend is continued for the comparison of ANN model with Tank, Harmonic, Thomas and Fiering models in simulation of the monthly runoffs at Dong Nai river basin, Viet Nam. The results proved ANN being the best choice among these models, if suitable and enough data sources were available.

Key words: artificial neural networks (ANNs); simulation; rainfall runoff model; monthly runoff; Tank model; Harmonic model; Thomas and Fiering model; Dong Nai river basin.

1. INTRODUCTION

Many available techniques for time series analysis assume linear relationships among variables. In the real world, however, temporal variations in data do not exhibit simple regularities and are difficult to analyze and predict accurately. It seems necessary that nonlinear models such as artificial neural networks, which are suited to complex nonlinear systems, be used for the analysis of real-world temporal data [1]. There are numerous applications of ANNs in the field of water resources: application of ANN for deriving the rainfall-runoff relationship [2, 3, 4]; for rainfall forecasting [5]; for river runoff forecasting [6]; for flood forecasting at the upper reach of the Red river basin, North Vietnam [7]. Other applications of ANNs include regional flood frequency analysis [8], regional drought analysis [9], and so on. Hsu [10] compared ANNs with traditional methods to model rainfall-runoff process. Chibanga [11] compared the performance of ANNs with that of multivariate ARMA model in application to the monthly inflow forecast and to historical record of river flow time series. Campolo [12] applied an ANN to forecast the flooding behaviour of the river Tagliamento using rainfall and water level as the only inputs. In continuity of this trend, in this paper, the simulation of monthly runoffs at Tri An and Phuoc Hoa stations, in Dong Nai river basin, Viet Nam were implemented by using ANN models and then the results were compared with those from Tank, Harmonic and Thomas & Fiering models.

2. THREE-LAYER FEEDFORWARD ARTIFICIAL NEURAL NETWORK MODEL

Where processes to be modeled are complex enough to be described mathematically, neural networks are considered to outperform the conventional, deterministic models most of
the time (www.stowa-nn.ihe.nl/Applications_Fuzzy_Logic.htm). And in other view of
Kolmogorov’s theorem [13] and Funahashi’s work [14], it is now universally held that a three-
layered ANN can serve as any continuous function approximator for as long as a sufficient
number of hidden neurons is used. Given a training set of input-output data, the most common
learning rule for multi-layer perceptrons is the back-propagation algorithm [1]. So, here, three-
layer ANN with ten hidden nodes, using BPA method with multi-input variables, one output
variable has been used to simulate the runoff process (Fig. 1). Details concerning to Back
Propagation Neural Network can be obtained in reference [2].

In hydrology, the values of input can be causal variables such as rainfall, temperature,
evaporation, water level, discharge, etc. at the previous or same period of output variable; at
the various locations within the considered rivers or basin. The values of output can be
hydrological responses such as runoff, water level, etc. The numbers of input nodes have
strictly to be satisfied with that of causal input variables in relation to output targets. The
relations between the output target variables and input variables in quantitative and causal
qualitative have to be selected as appropriate as possible based on the physical meanings and
relations of the problem to reduce the waste of computational and search time in training
process [2].

The procedure for application of three-layer ANN model for generation of the monthly
runoff at a station within the basin follows these steps:
- Select the input set of causal variables related to the target, monthly runoff, based on the
  physical meaning of the real world and/or using the correlation matrix for the study catchment
- Search for the appropriate structure of ANN: number of input variables, number of
  nodes in hidden layer, and the connection weights and biases by using the trial method and
  depending on the basis of the standards for statistic performances for training and testing
  stages.

3. HARMONIC ANALYSIS MODEL

The periodicity of physical vibrations can commonly be found in natural phenomena. The
water level in coastal river, seasonal rainfall and runoff, etc. may be thought as the
illustrations. Actually, runoff in a river includes three principal components, namely trend,
periodicity and randomness. The trend is due to human activity in their use of water such as
diversion and/or extraction of water for irrigation, water supply etc., by the channel-

![Fig 1. Three-layer feedforward ANN with one output variable](image-url)
construction, or deforestation within the basin. The trend component can be eliminated through adding the diversion to measured data. Then, the runoff can be investigated as a synthesis of periodicity and randomness. If the contributing level of randomness is not so high, then harmonic analysis through using the Fourier series can be applied to study the runoff phenomenon for generating the flow as described by the following equation:

\[ Q_t = \frac{1}{T} \sum_{i=1}^{T} F_i + \sum_{k=1}^{N} C_k \sin\left(\frac{2\pi}{T_k} t + \phi_k\right) \]  

(1)

Where, \( Q_t \) is the calculated monthly discharge; \( F_i \) is measured monthly discharge. The first term of RHS is the mean discharge; the second term of RHS describes the periodic sinusoidal vibrations of sub-harmonic components. The values of component number (N), period (\( T_k \)), amplitude (\( C_k \)) and phase (\( \phi_k \)) of each sub-harmonic component have to be determined during the modeling process. Various search technique can be applied to obtain the fittest solution, such as Genetic Algorithm (GA), other local search approaches, or even the trial approach. The GA or other search techniques requires the setting-up of algorithm and programming. The trial approach usually takes time and effort to get the fittest solution. Therefore, in present study, a proposed approach was done for searching the solution. The search method is as follows:

(i) Introduction of a combination of \( N \) sub-harmonic component with given set of \( T_k \) was started out. During this process, the basic period of 12 months due to seasonal rainfall phenomenon should firstly be considered. The standard deviation of historic monthly runoff data could be used as the amplitude to this component at the starting point.

(ii) The process was continued by searching for \( C_k \) and \( \phi_k \) such that the Root Mean Square Error (RMSE) of the simulated values compared with the measured data was minimum:

\[ \text{Min} \left\{ \sqrt{\frac{1}{T} \sum_{i=1}^{T} (F_i - Q_i)^2} \right\} \]  

(2)

\[ C_k \geq 0; \forall k = 1..N \]  

(3)

\[ Q_i \geq Q_{\text{min}}; \forall i = 1..T \]  

(4)

Constraints:

Where, \( Q_{\text{min}} \) being an absolutely minimum value of monthly runoff might occur at the considered station. This value being a positive number was assumed based on the historic monthly runoff series at that station (\( F_i \)) based on the statistical theory. The constraint (4) was added to avoid the calculated monthly runoff being negative (not reasonable due to the nature of monthly runoff) during the search process.

(iii) The combination of sub-harmonic component was changed step by step, by adding one more component of \( T_k \). Then, the repetition of calculation process can be restarted from step (ii).

This repetitive calculation process would be stopped whenever the value of RMSE was not able to decrease anymore.

An MS EXCEL program has been written for convenient adjustment of the calculated curve to obtain the better fit to the recorded one. The automatically graphical show and the
value of Root Mean Square Error (RMSE) for trial cases can be shown promptly during the solution search process.

4. THOMAS AND FIERING MODEL

The formula to simulate monthly runoff for Thomas and Fiering [15] model include two portions. The first is fixed portion. It is created from the historic data and their statistic characteristics. For monthly flow, the periodicity with basic period of year was considered. Then, the statistic characteristics of each month were used to create this fixed portion of the flow with zero value of normally distributed random number (u=0). The second portion of monthly runoff is the random one. It is used to describe the randomness of monthly runoff for each month whenever the generated monthly runoff is being calculated. The governing equation is as follows:

\[ Q(i, j) = Q_{av}(j) + b(j) \cdot (Q(i, j-1) - Q_{av}(j-1)) + S(j) \cdot [1 - (r(j))^2]^{1/2} \cdot t \]  

where, \( Q(i, j) \) is the discharge during the \( j^{th} \) month of the \( i^{th} \) water year in a generated sequence, reckoned from the start of synthesized sequence; \( Q_{av}(j) \) is mean historic monthly discharge during the \( j^{th} \) month (\( j: 1..12 \) for Jan. to Dec.); \( b(j) \) is the regression coefficient for estimating flow in the \( j^{th} \) month from \( (j-1)^{th} \) month; \( S(j) \) is the standard deviation of flows in the \( j^{th} \) month; \( r(j) \) is the correlation coefficient with flow in preceding month; \( t \) is the standard normal variable which changes its value each time the equation is applied. In Eq. (5), values of \( j \) can be applied from 2 to 12. In the case \( j=1 \), then \( j-1 \) is taken by using month December (\( j=12 \)) of the previous year. The statistic parameters can be found in reference [16].

5. TANK MODEL

Tank model was proposed by Prof. M. Sugawara, in Japan, in year 1956’s for simulation of flood flow. It was modified and improved, thereafter and applied in many places in the world. The procedure of simple Tank model is presented in Fig. 2.

![Fig 2. Sketch of Tank model](image-url)
Due to deformation of the river, channel storage and over-bank flow, the output of Tank model should be modified by introducing another Tank (linear or non-linear) and even the consideration of channel storage and flood plain storage effects should be taken. It may be recognized that the input variables of Tank model include rainfall, evaporation. These are the variables that directly cause the runoff. Beside, Tank model also consider other physical factors, such as the infiltration, drainage area, base flow and a mechanism to simulate the physical phenomenon of rainfall runoff. The theoretical aspects of Tank model can be referred to Tank program [9, 17, 18, 19, 20].

6. APPLICATION OF RAINFALL-RUNOFF MODELS

6.1. Dong Nai river basin

Dong Nai River (DNR) basin is located in the South of Viet Nam. Its inland river system, with a total length of 635 km, is the largest in the nation. The catchment area of DNR basin is 37,400 km² and the average annual flow that drains into the sea, is 31 billions cubic meters. This area is in the tropical monsoon zone with two distinct wet and dry seasons in a year. The wet season is from May to October and the dry season is from November to April [21]. Two runoff stations at Tri An and Phuoc Hoa were studied in this river basin. Tri An runoff is the inflow source of the Tri An reservoir. Its mean monthly runoff and standard deviation during the year are shown in Fig. 3.b. Other statistic parameters are shown in Fig. 3a, as well. Phuoc Hoa monthly runoff is one of the main sources that supply water for Ben Than treatment plant, and for irrigation of Lower Dong Nai river basin through Sai Gon river. Its mean monthly runoff and standard deviation and other statistic parameters are shown in Fig. 4b and Fig. 4a, respectively.

Thereinafter, the above mentioned models were utilized to simulate the monthly runoffs at Tri An and Phuoc Hoa stations. Then the comparison and analysis of their simulated results are carried out. And, the conclusion on ANN model is finally conducted.

6.2. Availability and arrangement of data for simulation

![Fig 3a](image-url)  **Statistic parameters**

**Tri An inflow (year 1978-1997)**

- **Correlation coefficient**, \( r(j) \);
- **Coefficient of the regression**, \( b(j) \);
- **Skewness** of monthly runoff at Tri An station

![Fig 3b](image-url)  **Statistic parameters**

**Tri An inflow (year 1978-1997)**

- **Mean**, \( Q_{av}(j) \) and standard deviation, \( S(j) \)
- **of monthly runoff at Tri An station**
6.2.1. Measured data
At Phuoc Hoa runoff station: available data from 1/1978 to 12/1992

6.2.2. Simulated monthly runoff using Tank model by Jica
At Tri An station: simulated data is available from 1/1978 to 12/1986
At Phuoc Hoa station: simulated data is available from 1/1978 to 12/1992

6.3. Segmentation of time periods for monthly runoff simulation
6.3.1. Calibration stage
The 108 months from 1/1978 to 12/1986 for both Tri An and Phuoc Hoa stations were used for training phase to search for the suitable structure of ANN model, including appropriate input set and reasonable connection weights and threshold values. This period was also used for calibrating phases for Harmonic and Thomas and Fiering models.

6.3.2. Validation stage
At Tri An runoff station: including 36 months from 1/1989 to 12/1991
These data were used for testing phase after the structure of ANN model had been completely trained. And, they were also used for Harmonic and Thomas & Fiering models for validation.

6.4. ANN model
Using trial method to search for the best combination of input set by comparing the performance of various alternatives based upon the statistic parameters. Finally, the input set of ANN model at two runoff stations were obtained.

6.4.1. Tri An monthly runoff
\[ Q_{ta}(t) = f[Q_{ta}(t-1), R_{dg}(t), R_{tt}(t), R_{dp}(t), R_{pl}(t), R_{dat}(t), R_{bl}(t), R_{dl}(t), R_{lk}(t)] \]
Rdla(t), Rdd(t), Rtp(t), Rxl(t)]  \tag{6}

Where, The output variable: Qta(t) is the monthly discharge in month t, at Tri An station;
The input variables:
Qta(t-1): monthly discharge in month t-1, at Tri An station. This previous monthly runoff
of the output variable was used here to simulate the base flow in month t during the dry
seasons.
Rdg(t), Rt(t), Rdp(t), Rp(t), Rdat(t), Rbl(t), Rdl(t), Rlk(t), Rdla(t), Rdd(t), Rtp(t), Rxl(t):
are respectively monthly rainfall at Dau Giay, Tue Trung, Dong Phu, Phuoc Long, Date, Bao
Loc, Di Linh, Lien Khuong, Da Lat, Don Duong, Ta Pao, Xuan Loc stations within DNR
basin.

So, there are 165 connection weights required to be trained to describe the structure
of ANN model to simulate suitably the rainfall runoff phenomenon at Tri An station.

6.4.2. Phuoc Hoa monthly runoff
Qph(t) = f[Qph(t-1), Qpl(t), Rpl(t), Rdp(t), Rbil(t)]  \tag{7}
Where, The output variable: Qph(t) is the monthly discharge in month t, at Phuoc Hoa
station;
The input variables:
Qph(t-1): monthly discharge in month t-1, at Phuoc Hoa station to simulate the base flow
in month t during the dry seasons.
Qpl(t): monthly discharge during the same month (t) at Phuoc Long station, upstream of
Phuoc Hoa station, was included in the input set to describe the correlation of monthly
discharge between two stations in the same month and the same river.
Rpl(t), Rdp(t), Rbil(t) are respectively monthly rainfall at Phuoc Long, Dong Phu and Binh
Long stations within the Be river catchment area.

Therefore, there were 71 connection weights needed to be trained to describe the structure
of ANN model to simulate suitably the rainfall runoff phenomenon at Phuoc Hoa station. The
comparison between recorded monthly runoff and simulated results for training and testing
stages at Tri An and Phuoc Hoa stations are presented in Fig. 5a and Fig. 5b. The values of
RMSE, Root Mean Square Error Mean (RMSEM), Mean Absolute Deviation (MAD), RMSES
(Root Mean Square Error over Standard Deviation) and Efficiency Index (EI) for training and validation stages for both stations obtained by ANN are shown in the Table 1, 2, 3 and 4.

6.5. Tank model

The simulated monthly runoffs at Tri An and Phuoc Hoa stations, that were calculated using Tank model by Jica, were used for comparison with the results from other models.

Briefly, Tank model configuration used by Jica included 4 columns in horizontal direction. In each column, there were 4 tanks. Each tank, there were two holes, one for side and one for bottom to allow the water drain out to next tank or lower tank for simulation of runoff (including overland flow and infiltration). There almost were 64 parameters used to describe this model scheme, here. The simulated monthly runoffs by TANK model [21] for Tri An and Phuoc Hoa stations are presented in Fig. 6a and Fig. 6b. The statistic parameters for simulation by Tank model are also shown in the Table 1, 3 and 4.

6.6. Harmonic analysis

6.6.1. The simulation results at Tri An station

The $Q_{av}$ stands for the first term, $\frac{1}{T} \sum_{i=1}^{T} F_i$ in the RHS of Eq. 1 that is the average of historic monthly runoff, 542.01 m$^3$/s. After searching, all the sub-harmonic components were reached, and the simulated monthly runoff was shown in Fig. 7a. The bottom and middle levels of monthly runoff curves at these stations strongly have the characteristic of periodicity, during period of the dry season. The characteristic of periodicity and less randomness during these months is also revealed in Fig. 3b. During this period, the values of standard deviation of monthly runoff are very low. The top portions of monthly runoff curves depend on the rainfall. Therefore, they strongly have the characteristic of randomness during the rainy season. So, if the trial increases in amplitude of sub-harmonic components were implemented to catch these top portions of curves, then the values at the bottom will not satisfy the constraint (4), and even they may become negative, not reasonable.
6.6.2. The simulation results at Phuoc Hoa station

The mean discharge at Phuoc Hoa station $Q_{av}$ is 209.21 m³/s. After using EXCEL program searching for the expected combinations of sub-harmonic components, satisfying two constraints (3) and (4) and trying to gain the RMSE value of simulated monthly runoff as low as possible, the parameters of these sub-harmonic components are finally obtained and the simulated monthly runoff are presented in Fig. 7b. Similar to the Tri An case, the characteristics of periodicity revealed on the bottom and middle levels, and the tops of the monthly runoff curves revealed the characteristic of randomness.

6.7. Thomas and Fiering model [15]

As seen in Eq. 5, the monthly runoff in month $j$, in year $i$ depends on the mean monthly runoff in month $j$, $Q_{av}(j)$; the mean monthly runoff in month $j-1$, $Q_{av}(j-1)$; the regression
coefficient of monthly runoff in month $j$ with that in previous month, $b(j)$; the standard deviation of monthly runoff in month $j$, $S(j)$; the correlation coefficient of monthly runoff in month $j$ with that in previous month, $r(j)$. These values are constants for each historic time series data set. The simulated monthly runoff time series will depends on the starting monthly runoff and the random numbers, namely the standard normal variable, changing its value between 0 and 1. In this model, monthly rainfall data within the basin were not included.

Therefore, simulated monthly runoff may not follow closely recorded monthly runoff since the random characteristic of rainfall during the rainy season (from July to October) when standard deviation is high, as shown in Fig. 3b and Fig. 4b. Similar to Harmonic Analysis model, characteristics of periodicity revealed on bottom of monthly runoff curve during the dry season. So, its simulated values follow closely with the recorded ones. However, at top portions, its simulated values can not catch well with the recorded curve as shown in Fig. 8a and Fig. 8b.

6.8. Comparison and evaluation of simulated results of the ANN and other traditional model.

The performance statistics on resulting monthly runoffs by four models at both Tri An and Phuoc Hoa stations are shown in Tables 1, 2, 3 and 4. Compare these values and analyze the graphs in the above figures, the conclusions can be obtained as follows:

**Table 1.** Performance statistics on resulting monthly runoffs at Tri An, calibration stage

<table>
<thead>
<tr>
<th>Kind of Model</th>
<th>EI</th>
<th>RMSE (m³/s)</th>
<th>RMSEM (m³/s)</th>
<th>MAD (m³/s)</th>
<th>RMSES</th>
</tr>
</thead>
<tbody>
<tr>
<td>ANN</td>
<td>0.946</td>
<td>131.180</td>
<td>0.242</td>
<td>76.942</td>
<td>0.234</td>
</tr>
<tr>
<td>Tank (simulated by Jica)</td>
<td>0.833</td>
<td>197.947</td>
<td>0.365</td>
<td>145.486</td>
<td>0.353</td>
</tr>
<tr>
<td>Thomas &amp; Fiering</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Min</td>
<td>0.798</td>
<td>222.565</td>
<td>0.411</td>
<td>128.594</td>
<td>0.397</td>
</tr>
<tr>
<td>Random</td>
<td>0.826</td>
<td>235.883</td>
<td>0.435</td>
<td>144.966</td>
<td>0.420</td>
</tr>
<tr>
<td>Max</td>
<td>0.790</td>
<td>293.356</td>
<td>0.541</td>
<td>192.528</td>
<td>0.523</td>
</tr>
<tr>
<td>Harmonic Analysis</td>
<td>0.840</td>
<td>205.302</td>
<td>0.379</td>
<td>131.318</td>
<td>0.366</td>
</tr>
</tbody>
</table>

**Table 2.** Performance statistics on resulting monthly runoffs at Tri An, validation stage

<table>
<thead>
<tr>
<th>Model</th>
<th>EI</th>
<th>RMSE (m³/s)</th>
<th>RMSEM (m³/s)</th>
<th>MAD (m³/s)</th>
<th>RMSES</th>
</tr>
</thead>
<tbody>
<tr>
<td>ANN</td>
<td>0.902</td>
<td>155.498</td>
<td>0.294</td>
<td>104.069</td>
<td>0.295</td>
</tr>
<tr>
<td>Thomas &amp; Fiering</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Min</td>
<td>0.939</td>
<td>122.294</td>
<td>0.231</td>
<td>85.501</td>
<td>0.232</td>
</tr>
<tr>
<td>Random</td>
<td>0.925</td>
<td>147.899</td>
<td>0.279</td>
<td>109.436</td>
<td>0.280</td>
</tr>
<tr>
<td>Max</td>
<td>0.850</td>
<td>248.998</td>
<td>0.471</td>
<td>180.854</td>
<td>0.472</td>
</tr>
<tr>
<td>Harmonic Analysis</td>
<td>0.920</td>
<td>142.291</td>
<td>0.269</td>
<td>109.280</td>
<td>0.270</td>
</tr>
</tbody>
</table>
Table 3. Performance statistics on resulting monthly runoffs at Phuoc Hoa, calibration stage

<table>
<thead>
<tr>
<th>Model</th>
<th>EI</th>
<th>RMSE (m³/s)</th>
<th>RMSEM</th>
<th>MAD (m³/s)</th>
<th>RMSES</th>
</tr>
</thead>
<tbody>
<tr>
<td>ANN</td>
<td>0.965</td>
<td>45.260</td>
<td>0.216</td>
<td>27.270</td>
<td>0.198</td>
</tr>
<tr>
<td>Tank (simulated by Jica)</td>
<td>0.924</td>
<td>62.114</td>
<td>0.297</td>
<td>40.934</td>
<td>0.272</td>
</tr>
<tr>
<td>Thomas &amp; Fiering</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Min</td>
<td>0.854</td>
<td>82.676</td>
<td>0.395</td>
<td>47.762</td>
<td>0.362</td>
</tr>
<tr>
<td>Random</td>
<td>0.838</td>
<td>102.769</td>
<td>0.491</td>
<td>60.980</td>
<td>0.449</td>
</tr>
<tr>
<td>Max</td>
<td>0.802</td>
<td>128.376</td>
<td>0.614</td>
<td>75.033</td>
<td>0.561</td>
</tr>
<tr>
<td>Harmonic Analysis</td>
<td>0.837</td>
<td>83.080</td>
<td>0.397</td>
<td>51.691</td>
<td>0.363</td>
</tr>
</tbody>
</table>

Table 4. Performance statistics on resulting monthly runoffs at Phuoc Hoa, validation stage

<table>
<thead>
<tr>
<th>Model</th>
<th>EI</th>
<th>RMSE (m³/s)</th>
<th>RMSEM</th>
<th>MAD (m³/s)</th>
<th>RMSES</th>
</tr>
</thead>
<tbody>
<tr>
<td>ANN</td>
<td>0.971</td>
<td>41.062</td>
<td>0.175</td>
<td>25.863</td>
<td>0.165</td>
</tr>
<tr>
<td>Tank (simulated by Jica)</td>
<td>0.825</td>
<td>90.456</td>
<td>0.385</td>
<td>55.808</td>
<td>0.364</td>
</tr>
<tr>
<td>Thomas &amp; Fiering</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Min</td>
<td>0.823</td>
<td>91.157</td>
<td>0.388</td>
<td>50.621</td>
<td>0.367</td>
</tr>
<tr>
<td>Random</td>
<td>0.848</td>
<td>97.549</td>
<td>0.415</td>
<td>54.281</td>
<td>0.393</td>
</tr>
<tr>
<td>Max</td>
<td>0.833</td>
<td>115.783</td>
<td>0.493</td>
<td>65.328</td>
<td>0.466</td>
</tr>
<tr>
<td>Harmonic Analysis</td>
<td>0.741</td>
<td>104.706</td>
<td>0.446</td>
<td>62.895</td>
<td>0.422</td>
</tr>
</tbody>
</table>

- The simulated monthly runoffs for all models showed the good fitness at the bottom and the middle of the monthly runoff curves. And, some large deviations at the top level occurred due to the randomness caused by rainfall during rainy season.

- For simulation of monthly runoff at Tri An and Phuoc Hoa stations, all models may be used. ANN obtained highest fitness to the measured monthly runoff curves, except at Tri An station, in validation stage from 1989 to 1991. For validation stage at Tri An station, the simulated results by ANN not better than other models may be due to some changes in the causal relations between output monthly discharge and input monthly rainfalls, such as the changes in ground surface plant covers, or some human activities within the basin: deforestation, dam construction, etc. within the period from 1987 to 1988, or due to other random variables that may not be included in the input layer of ANN model. The fitness for Tank model is ranked the second. However, Both ANN and Tank models required much more data including both monthly discharges and rainfalls at various stations within the basin.

- The fitness obtained by Harmonic and Thomas & Fiering approaches was somewhat lower than two above-mentioned models but they did not require so much data for simulation. Just historic monthly discharge data at considered station were included. No rainfall data within the basin were required. Therefore, in the case missing or not enough data, especially rainfall data, and reasonable accuracy required, Harmonic Analysis is preferred. However, it has to be supported by Search Procedure for obtaining the fittest solution. Thomas & Fiering model also obtained the necessary fitness to the measured data. However, it required more data than Harmonic Analysis because of the requirement of statistic calculation. Besides, the
discharge time series simulated by Thomas & Fiering model depends on the randomness, each time it is created.

7. CONCLUSION

Since limit on the available data and resources, this research has been conducted just for two runoff stations. However, following the recent research trends, this research proved that the ANN model may continuously be confirmed its value and role for simulation of monthly runoff. It is clearly better than other traditional models, namely Tank, Thomas & Fiering, and Harmonic models. In order to get rather general conclusions on the capability and efficiency of ANN model for monthly runoff simulation, more additional researches should continuously be implemented for the Dong Nai river basin.

APPLICATION OF HARMONIC & THOMAS & FIERING WHEN APPL YING RESEARCH

Đồ Dòng Chay Thẳng Và So Sánh Kết Quả Với Các Mô Hình

TRUYỀN THÔNG KHÁC

Lê Văn Đức
Trường Đại Học Bách Khoa, DHQG-HCM

Talesce 1: Mô hình mạng thần kinh nhân tạo (ANN) được giải quyết liên truyền ngược (BPA) đã được áp dụng trong nhiều ngành, đặc biệt trong ngành thủy văn và quản lý nguồn nước, để mô phỏng hoặc dự báo tiến trình dòng chảy do mưa, chịu lưu lượng và mực nước theo thời gian và các biến thủy văn khác. Gần đây, một số nhà nghiên cứu tập trung so sánh năng lực áp dụng của mô hình ANN với những giải pháp dùng phương pháp lý thuyết hoặc phân tích dữ liệu khác. Nhiều nghiên cứu đã dẫn hành so sánh mô hình ANN với mô hình cây M5 để dự báo dòng chảy do mưa, so sánh với mô hình ARMAX để tìm chuỗi dòng chảy, so sánh với mô hình AR và mô hình hồi quy để dự báo và thẩm đoán dòng chảy ngày ở các sông. Kết luận so sánh cho thấy mô hình ANN đạt được kết quả tốt hơn. Vì vậy, để tiếp tục xu hướng nghiên cứu này, bài báo đã thực hiện việc so sánh mô hình ANN với các mô hình Tank, Harmonic và Thomas & Fiering khi áp dụng để mô phỏng dòng chảy tháng ở lưu vực Sông Đồng Nai, Việt Nam. Kết quả cho thấy, mô hình ANN là một phương án chọn lựa tốt nhất nếu có nguồn dữ liệu đầy đủ và thích hợp.

Tiêu kết: Mạng thần kinh nhân tạo (ANNs); mô phỏng; mô hình dòng chảy do mưa; dòng chảy tháng; mô hình TANK; mô hình điều hòa; mô hình Thomas và Fiering; lưu vực Sông Đồng Nai.

REFERENCE


