

Detection and Modeling of Phosphorus in Water Quality Monitoring Programs

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Abstract—Understanding the impact of phosphorus is essential to ensure water quality. Excessive nutrient loading, primarily phosphorus (P), is one of the vital reason for eutrophic. In this paper a survey concerning the concentration of phosphorus in a reservoir of northeast China was carried out. The raw data was collected monthly and a 19data sets monthly were collected from January 2011 to July 2012. The purpose of this investigation is to understand how phosphorus would affect the average annual water quality. A specific empirical model is proposed to show the dynamics of phosphorus level variation in reservoir. The model was presented to show water quality levels in response to various phosphorus levels. The simulation results can be an important reference for the evaluation of nutrient loads.

Keywords—water quality monitoring; phosphorus; water temperature; reservoir.

I. INTRODUCTION

The restoration of reservoirs serving as municipal water supplies and drinking water, is recognized as one of the most widespread problems in water quality management. Eutrophication of water quality has become one of the water quality issues in reservoirs. Phosphorus (P) is regarded as a primary or contributing nutrient to freshwater eutrophication.

Some research has been focused on water quality pollution caused by phosphorus. M. Taheriyounet al. proposed a new method called entropy-based fuzzy synthetic evaluation method, that obtained the uncertainties and randomness in the raw data and developed EFEI to determine the trophic status in the Satarkhan Reservoir [1]. Vera Istva'novic et al. developed a program that complemented lake-specific ecological criteria within the range existing and concluded that the simple objectives are the good choice in the initial stages of eutrophication management [2]. Dale M. et al. applied eutrophication programs to three different applications, including the Seepage Lake, BATHTUB, and WILMS Model to verify specific empirical models Model [3]. Simulation results show that the reduction of external P load would reduce near-surface TP. Josef S. Kardos et al. examined the impact of water quality model uncertainty on the nutritional trading scheme, and designed a scheme that would be implemented watershed TMDL for Nontidal Passaic River in New Jersey [4]. This paper showed how the uncertainty of water quality model affects the results on potential trades between the

sewage treatment plants. M. Tarvainen et al. estimated phosphorus release with a bioenergetics model, that made use of data from roach diet and growth, and estimated the phosphorus release from different age roach groups to determine whether phosphorus released by the fish can determine the observed increase in the phosphorus concentration in the summer [5]. Dennis Trollet et al. tested Taihu Lake sediment nitrogen content and the phosphorus changing with vertical and horizontal variations [6]. Stepwise linear regression analysis showed that 91% of the horizontal variability in the sediment total phosphorus concentration and 65% of the vertical variation was due to the combination of carbon concentration and sediment manganese.

In this paper Fuzzy Inference System was developed to assess trophic evaluation of reservoirs. Five water quality parameters of total phosphorus (TP), chlorophyll a, secchi depth (SD), hypolimnetic oxygen (HO%), and total nitrogen (TN) were monitored. The raw data was collected monthly and a 19 data set monthly was collected from January 2011 to July 2012. The purpose of this investigation is to understand how phosphorus would affect the reservoir eutrophication level. The simulation results can be an important reference for the evaluation of nutrient loads. This result indicated that the methodology adopted in this study was basically an attractive alternative, offering a relatively fast algorithm to assess trophic evaluation of reservoirs.

The remaining sections of this paper are organized as follows. Section 2 presents the developed Fuzzy Inference System and the entropy weighing method. Section 3 presents MATLAB simulation is detailed. The paper concludes with Section 4.

II. MATERIALS AND METHODS

Data used for the reservoir eutrophication level in northeast China are monthly records from January 2011 to July 2012. A 19 data sets monthly were collected from January 2011 to July 2012. Five water quality parameters used in this study analysis are total phosphorus (TP), chlorophyll a, secchi depth (SD), hypolimnetic oxygen (HO%), and total nitrogen (TN). See Table 1 for specific values of the parameter sampling.

In this study, Fuzzy Inference System was developed to assess trophic evaluation of reservoirs. The assessment criteria is established from the OECD eutrophication

program (modified from Vollenweider and Kerekes, OECD, 1982)[7]. Table 2 presents the assessment criteria in detail.

Let U be an evaluation factor set based on water quality parameters. U be represented as follows:

$$U = \{u_i\}, i = 1, 2, \dots, n \quad (1)$$

Where u_i is the i^{th} water quality parameter

TABLE I. WATER QUALITY DATA DURING JANUARY 2011- JULY 2012(MG L⁻¹)

Parameters	2011											
	1	2	3	4	5	6	7	8	9	10	11	12
chlorophyll a	6.0	4.0	1.0	1.0	8.0	5.0	2.0	9.0	12.0	3.0	1.0	1.0
TP	10	40	50	70	40	40	30	30	10	70	10	<DL
TN(mg)	3.81	4.62	5.18	3.81	3.01	3.65	2.68	2.52	2.96	3.57	3.46	3.05
SD	1	1	3	2	1	2	2	2	2	3	3	3
HO	8.4	6.5	8.0	8.1	10.1	7.0	8.1	8.5	8.5	7.0	7.9	9.6
Parameters	2012											
	1	2	3	4	5	6	7					
chlorophyll a	6.0	2.0	2.0	1.0	1.0	6.0	16					
TP	40	40	30	90	30	50	40					
TN(mg)	3.23	3.24	3.06	3.95	4.48	3.22	3.67					
SD	3	2	3	2	2	2	1					
HO	9.8	8.8	10.4	5.4	10.5	8.6	9.1					

TABLE II. CLASSIFICATION PROGRAM OF TROPHIC STATUS

Water quality indicator	Oligotrophic		Mesotrophic		Eutrophic	
	range	mean	range	mean	range	mean
Chlorophyll a	0.8-3.4	1.7	3.0-7.4	4.7	6.7-31	14.3
TP	4.9-13.3	8.0	14.5-49	26.7	48-189	84.4
TN	371-1181	661	485-1170	753	861-4081	1875
SD(m)	5.9-16.5	9.9	2.4-7.4	4.2	1.5-4.0	2.45
HO	>80%	80%	10-80%	45%	<10%	10%

Let V be an evaluation criteria set based on the OECD eutrophication program as shown in Table 2. This study defined three levels of eutrophic, mesotrophic and oligotrophic as fuzzy sets.

$$V = \{v_{ij}\}, i = 1, 2, \dots, n; j = 1, 2, \dots, m \quad (2)$$

Where v_{ij} is the criteria mean value of the i^{th} assessment parameter at the j^{th} level.

The fuzzy relation R between U and V is given by

$$R = \begin{pmatrix} r_{11} & \dots & r_{1m} \\ \vdots & \ddots & \vdots \\ r_{n1} & \dots & r_{nm} \end{pmatrix} \quad (3)$$

Eq.(4) demonstrates the membership function for the water quality assessment parameter, the smaller the better such as chlorophyll a, TP and TN.

$$r_{i1}(u_i) = \begin{cases} 1 & u_i \leq v_{i1} \\ (u_i - v_{i2}) / (v_{i1} - v_{i2}) & v_{i1} < u_i < v_{i2} \\ 0 & u_i \geq v_{i2} \end{cases}$$

$$r_{i2}(u_i) = \begin{cases} 0 & u_i \leq v_{i1} \\ (u_i - v_{i1}) / (v_{i2} - v_{i1}) & v_{i1} < u_i \leq v_{i2} \\ (u_i - v_{i3}) / (v_{i2} - v_{i3}) & v_{i2} < u_i < v_{i3} \\ 0 & u_i \geq v_{i3} \end{cases}$$

$$r_{i3}(u_i) = \begin{cases} 0 & u_i \leq v_{i2} \\ (u_i - v_{i2}) / (v_{i3} - v_{i2}) & v_{i2} < u_i < v_{i3} \\ 1 & u_i \geq v_{i3} \end{cases} \quad (4)$$

For the water quality assessment parameter, the bigger the better such as SD and HO, the membership functions are

$$r_{i1}(u_i) = \begin{cases} 1 & u_i \geq v_{i1} \\ (u_i - v_{i2}) / (v_{i1} - v_{i2}) & v_{i2} < u_i < v_{i1} \\ 0 & u_i \leq v_{i2} \end{cases}$$

$$r_{i2}(u_i) = \begin{cases} 0 & u_i \leq v_{i1} \\ (u_i - v_{i1}) / (v_{i2} - v_{i1}) & v_{i2} < u_i \leq v_{i1} \\ (u_i - v_{i3}) / (v_{i2} - v_{i3}) & v_{i3} < u_i < v_{i2} \\ 0 & u_i \leq v_{i3} \end{cases}$$

$$r_{i3}(u_i) = \begin{cases} 0 & u_i \geq v_{i2} \\ (u_i - v_{i2}) / (v_{i3} - v_{i2}) & v_{i3} < u_i < v_{i2} \\ 1 & u_i \leq v_{i3} \end{cases} \quad (5)$$

In order to evaluate the impact of different parameters that could affect the reservoir eutrophication level, the entropy weighing method is introduced to calculate the water parameters weight related on eutrophication.

Let X denote the matrix of l samples for five water quality parameters as shown in Eq. (6).

$$X = \begin{bmatrix} X_{11} & \cdots & X_{1n} \\ \vdots & \ddots & \vdots \\ X_{l1} & \cdots & X_{ln} \end{bmatrix} \quad (6)$$

Where X_{ij} represents the i^{th} sample for the j^{th} parameter ($i = 1, 2, \dots, l; j = 1, 2, \dots, n$).

The elements of the evaluation matrix can be normalized by maximum criterion as shown in Eq. (7).

$$r_{ij} = \frac{X_{ij}}{\max\{X_{ij}\}_j} \quad (7)$$

The weight vector $W = \{w_i\}, i = 1, 2, \dots, n$ is defined as[7]:

$$w_i = (1 - E_i) / \sum_{i=1}^n (1 - E_i) \quad (8)$$

Where

$$E_i = -C \sum_{j=1}^n (P_{ji} \ln P_{ji}) \quad (9)$$

$$P_{ji} = r_{ji} / \sum_{i=1}^n r_{ji} \quad (10)$$

III. SIMULATION

Based on the original monitoring data (Table 1) including 19 data set monthly ($l=19$) and 5 parameters ($m=5$). The weight vector of chlorophyll a, TP, TN SD, and HO using Eq. 8 is calculated as follows:

$$W = [0.59 \ 0.26 \ 0.03 \ 0.10 \ 0.02] \quad (11)$$

The aggregation was obtained by

$$B = W \times R \quad (12)$$

where $W = \{w_i\}, i = 1, 2, \dots, n$.

A Evaluation result is through maximizing element in matrix $B = \{b_i\}, i = 1, 2, \dots, m$ as shown in Eq. (13).

$$\text{Level} = \max(b_i), i = 1, 2, \dots, m \quad (13)$$

Table 3 presents the membership values calculated by evaluation matrix based on three different trophic status levels and the evaluation result of trophic level by maximum criterion.

January, March, April, October, November and December during low temperature season were assessed as oligotrophic level since the low temperatures lead to low chlorophyll a concentration. June and July were mesotrophic due to increase both total phosphorous and chlorophylla. August and September were eutrophic due to the high concentration of both total phosphorous and chlorophylla. But May different from both April and June was belong to eutrophic mainly because of total phosphorous increasing suddenly. The sudden increase in TP was due to farmland fertilizer flowing into reservoir in the spring season. In 2012, reservoir management was enhanced, this situation does not occur again and May was assessed as oligotrophic.

TABLE III. MONTH LEVEL OF TROPHIC STATUS

Month	1	2	3	4	5	6	7	8	9	10
E	0.13	0.10	0.21	0.32	0.34	0.14	0.14	0.41	0.60	0.32
M	0.00	0.66	0.10	0.10	0.30	0.77	0.75	0.24	0.05	0.30
O	0.76	0.20	0.61	0.61	0.20	0.02	0.02	0.25	0.25	0.34
trophic level	O	M	O	O	E	M	M	E	E	O
Month	11	12	1	2	3	4	5	6	7	
E	0.13	0.13	0.13	0.13	0.13	0.39	0.10	0.24	0.73	
M	0.00	0.00	0.70	0.20	0.24	0.02	0.24	0.51	0.20	
O	0.82	0.84	0.02	0.60	0.55	0.59	0.62	0.20	0.07	
trophic level	O	O	M	O	O	O	O	M	E	

O=oligotrophic, M=mesotrophic and E= eutrophic

IV. CONCLUSION

In this paper Fuzzy Inference System was developed to assess trophic evaluation of reservoirs. The assessment criteria is established from the OECD eutrophication program (modified from Vollenweider and Kerekes, OECD, 1982). Five water quality parameters of total phosphorous (TP), chlorophyll a, secchi depth (SD), hypolimnetic oxygen (HO%), and total nitrogen (TN) were monitored. In order to evaluate the impact of different parameters that could affect the reservoir eutrophication level, the entropy weighing method is introduced to calculate the water parameters weight related on eutrophication. The simulation results can be an important reference for the evaluation of nutrient loads.

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