

Reservoir Flood Season Segmentation and Risk–benefit Cooperative Decision of Staged Flood Limited Water Level

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Abstract

Investigation of seasonal inconsistency between water consumption and rainfall variation is important for more efficient use of floodwater resources. Adjustment of the flood limited water level (FLWL) is an effective way to improve the floodwater use efficiency. Flood season segmentation provides the basis for determining the FLWL and tapping the potential use of foodwater resources. Compromise between the beneft of foodwater use and food control is crucial to FLWL decision. We use the circular distribution method for food season segmentation and the relative frequency method for verifcation. We select the performances of water supply and hydropower generation as the beneft index and the extreme risk rate as the risk index. On the basis of the game theory, we establish a multiobjective cooperative decision-making model and obtain a Nash negotiation solution of staged FLWL. An optimal scheme is determined according to the fuzzy pattern recognition theory. When the risk and beneft are equally valued, the resulting FLWLs of the optimal scheme are 129.0 m, and 128.5 m for a selected reservoir in the pre-food season and the post-food season, respectively. By adjusting the preference values of the risk and beneft indexes, we determine the optimal FLWL scheme under diferent preferences to risk and beneft for each stage.

Keywords Flood season segmentation · Extreme risk rate · Multi-objective cooperative decision · Nash negotiation · Adjustment of flood limited water level

1 Introduction

Reservoir operation is signifcant for management of water resources and sustainable socioeconomic developments (Tan et al. [2017;](#page-16-0) Xu et al. [2021\)](#page-16-1). The control of food limited water level (FLWL) is a signifcantly productive method of reservoir operation.

FLWL is an important index to compromise the risk of food and the beneft of foodwater use. It is mainly determined by reservoir regulation based on the biggest

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foods. However, in actual operation, most of the foods are medium or small. This means that the storage capacity, which is originally vacated for large foods, is only used to store the medium or small foods, and thus the rest of the food control capacity is wasted. In China, the foods in most watersheds show a strong seasonal variation due to the monsoon climate, varying signifcantly throughout the whole food season. If the entire food season operates with an annually fxed FLWL, it inevitably causes a low FLWL, which would lead to waste of floodwater resources. To solve these problems, the entire food season should be better divided into several sub-seasons according to the seasonal variations of food. On the basis of food season segmentation, FLWL of each stage is determined respectively. It cannot only ensure the food control safety in the food season but also improve the beneft to a greater extent.

The study of food season segmentation (Jiang et al. [2019\)](#page-15-0) has gone through a process of in-depth research. Advances primarily involve the techniques of statistical method, genesis analysis method (Singh et al. [2005\)](#page-15-1), fuzzy clustering method (Ju et al. 2020), fractal theory (Li et al. 2018), relative frequency and directional statistics method (Cunderlik et al. [2004](#page-15-4)), and so on. Owing to the seasonality of foods, the beginning and end of the food season have the characteristics of periodic change. The circular distribution method (Villarini [2016](#page-16-2)) is a method to analyze the phenomenon of periodic change and fnd its internal law. It has the characteristics of simple and convenient calculation and fexible analysis. It also takes the food magnitude into account, and its results are more accurate than other methods. Besides, the relative frequency method is used to recheck the results of the circular distribution method. Although its result cannot be as precise as the other method down to a day, it is simple and requires less data. Using 62-year daily discharge data, we select the circular distribution method and the relative frequency method to segment the food season in this study.

On the premise of meeting the downstream flood control safety, the efficiency of foodwater use has improved through adjustment of FLWL (Huang et al. [2017](#page-15-5); Liu et al. [2019](#page-15-6)). There have been many studies on how to realize efective control of FLWL. The staged control of FLWL is proved to be a more efficient measure. The widely used methods (Jiang et al. [2015](#page-15-7)) mainly include fuzzy set analysis method (Ruan and Singh [2008\)](#page-15-8) and multi-objective optimization method (Xie et al. [2018](#page-16-3)) etc. Researchers have made extraordinary achievements in producing suitable operating methods (Ma et al. [2020;](#page-15-9) Mo et al. [2022;](#page-15-10) Wu and Zhong [2018](#page-16-4)), whereas most of the existing studies do not separately consider FLWL at each stage. Generally, in the pre-food season, to meet the storage requirement of an incoming food, the decision-makers would mainly focus on food safety control and thus prefer risk to beneft, resulting in a lower FLWL. Conversely, in the post-food season, the decision-makers would prefer beneft to risk as the concern of water supply increases, expecting a higher FLWL. Therefore, FLWL in diferent stages should be considered respectively to compromise the beneft and risk.

In this paper, on the basis of food segmentation, we establish a set of beneft and risk indexes of FLWL control following the game theory and a multi-objective cooperative decision-making model of risk and beneft. Taking Xianghongdian (XHD) reservoir as an example, we choose FLWL in the main food season as the original design with a focus on the pre-food and post-food seasons according to the goal of food control and design of food hydrograph. Starting with the Nash negotiation theory, we obtain the optimal FLWL scheme under diferent preferences between risk and beneft in each stage.

2 The Division of Flood Season

2.1 Circular Distribution Method

2.1.1 Concept

Circular distribution method transforms the periodic data into linear through trigonometric function transformation. Treating the flood events as a vector, it calculates the vector eigen-values to determine the flood season segmentation (Chen et al. [2010\)](#page-15-11).

2.1.2 Computing Method

Assuming that the length of the calculation period is *T*, the occurrence time of the food peak of the *i*-th flood sample is D_i , and the magnitude is q_i , the equations of the coordinate value (x_i, y_i) of flood occurrence time are:

$$
(x_i, y_i) = \begin{cases} (\cos \alpha_i, \sin \alpha_i) & \text{without considering the magnitude of flood} \\ (q_i \cos \alpha_i, q_i \sin \alpha_i) & \text{considering the magnitude of flood} \end{cases}
$$
 (1)

and

$$
(\overline{x}, \overline{y}) = \left(\sum_{i=1}^{N} x_i / N, \sum_{i=1}^{N} y_i / N\right)
$$
 (2)

where *N* is the sample size, and $\alpha_i = D_i \frac{2\pi}{T}$ is the occurrence time of the *i*-th flood.

The concentration period $\bar{\alpha}$ and concentration *r* are respectively:

$$
\overline{\alpha} = \begin{cases}\n\arctan \overline{y}/\overline{x} & \overline{x} > 0, \ \overline{y} > 0 \\
2\pi + \arctan \overline{y}/\overline{x} & \overline{x} > 0, \ \overline{y} < 0 \\
\pi + \arctan \overline{y}/\overline{x} & \overline{x} < 0 \\
\pi/2 & \overline{x} = 0, \ \overline{y} > 0 \\
3\pi/2 & \overline{x} = 0, \ \overline{y} < 0 \\
\text{Uncertainty} & \overline{x} = 0, \ \overline{y} = 0\n\end{cases}
$$
\n(3)

and

$$
r = \begin{cases} \sqrt{\overline{x}^2 + \overline{y}^2} & \text{without considering the magnitude of flood} \\ \sqrt{\overline{x}^2 + \overline{y}^2} & \text{considering the magnitude of flood} \end{cases}
$$
(4)

where \overline{q} is the mean flow value of *N* samples, and the concentration day corresponding to the concentration period $\overline{\alpha}$ is $D_i = \overline{\alpha} \frac{T}{2\pi}$.

Concentration *r* describes the central tendency of α_i in the circular distribution with $0 \le r \le 1$. When *r* is closer to 1, it means that the flood occurrence time is more concentrated in a specific area, not vice-versa. *s* is the standard deviation of α_i . The relationship between *r* and *s* is:

$$
s = \sqrt{-2\ln r} \tag{5}
$$

where *s* is an index indicating the discrete trend of the flood occurrence time. Then the starting and ending dates D_1 and D_2 of the main flood season are, respectively:

$$
D_1 = \frac{\overline{\alpha} - s}{2\pi}T\tag{6}
$$

$$
D_2 = \frac{\overline{\alpha} + s}{2\pi}T\tag{7}
$$

2.2 Relative Frequency Method

The entire food season is divided into *k* periods, and the amount of the maximum food in each period is counted. The equation for the relative frequency of each period is:

$$
RF_k = \frac{b_k}{n} \tag{8}
$$

where RF_k is the relative frequency of the *k*-th period, b_k is the number of annual maximum foods in the *k*-th period, and *n* is the number of years of fow data.

If each period is 10 days, the length of each period of a month is diferent. To make each period has the same length, the period needs to be adjusted by multiplying a coefficient. For the last ten-day of a longer month (e.g., Jan. with 31 days), the last 10-day period should be multiplied by 10/11; and for the last 10-day of February, it should be multiplied by a coefficient of $10/8$ ($10/9$ for a leap year). When the algebraic sum *s* of the initial frequency RF_k is not equal to the total of the adjusted RF_k ' s', generally each RF_k ' is multiplied by the coefficient s/s' , that is, the adjusted relative frequency is:

$$
RF'_{k} = RF_{k} \cdot \frac{s}{s'}
$$
 (9)

3 Analysis of Risk and Benefit

The food season staged operation is to take diferent FLWL control scheme and schedule strategy for each segmentation. To raise the FLWL for improving the benefts of water conservancy while ensuring the food control safety for the project and the downstream, we frst conduct a cooperative decision-making analysis of risk and beneft.

3.1 Risk Analysis

During reservoir operation, events that are not conducive to the dam safety, threaten the flood control safety of downstream, and affect the benefit of reservoir will happen. The occurrence probability and loss due to these events is called food control risk (Zhou and Guo [2014\)](#page-16-5). The flood control risk rate in the extreme state is referred to as the extreme risk rate. With the FLWL Z_0 as the initial water level, the design flood hydrograph of each frequency is calculated for flood routing. Suppose that Z_m is the highest water level of flood

routing for the design flood hydrograph at a certain frequency. When Z_m is exactly equal to a selected safe level, the frequency is the extreme risk rate P_{lim} . We take the design flood level Z_d as the selected safe level. The extreme risk rate can be expressed by the following equation:

$$
P_{\lim} = P\{Z_m \ge Z_d\} \tag{10}
$$

3.2 Benefit Analysis

Under the premise of ensuring the food control safety, the FLWL can be raised as much as possible. Therefore, the utilization beneft of hydropower generation and water supply are improved accordingly.

3.2.1 Water Supply Benefit

As the FLWL rises, the added water supply capacity mainly comes from the increased storage of abandoned water. In dry years, there is no water to store after the food season, and the benefts of FLWL adjustment can be clearly refected. On the contrary, in wet years, the reservoir can be reserved to the normal water level even if FLWL is not raised. Raising the FLWL will only bring about the hydropower generation beneft caused by the increased water head. Therefore, the multi-year average water supply, which can be increased by raising FLWL, should be deducted from wet years. The added water supply volume can be obtained as:

$$
\Delta V_2 = \Delta V_1 \times (1 - \lambda) \tag{11}
$$

where ΔV_2 is the increased water supply storage, and ΔV_1 is the increased water storage capacity. According to the curve between water level and capacity, the increased water storage under different FLWL can be obtained. λ is the proportion of wet years.

The increased water supply beneft brought about by the adjustment of the FLWL is:

$$
B_{\Delta} = \Delta V_2 \times \delta \tag{12}
$$

where B_Λ is the increased water supply benefit, and δ is the water supply benefit of per cubic meter of water.

3.2.2 Hydropower Generation Benefit

Added water capacity and increased water head cause the improvement of hydropower generation beneft. The beneft brought by the increased water capacity is calculated by the average water consumption rate of per unit electric energy, which is obtained by counting the average annual hydropower generation water consumption and annual average hydropower generation in recent years. The additional electricity generated by the increased water capacity is expressed as:

$$
E_1 = \Delta V_2 / \varphi \tag{13}
$$

where E_1 is the additional electricity, ΔV_2 is the added water supply storage, and φ is the average water consumption rate of per unit electricity.

The beneft generated by increased water head is expressed by the following equation:

$$
E_2 = 9.81 \cdot \eta \cdot Q_T \cdot \Delta H \cdot \Delta T \tag{14}
$$

where E_2 is the additional electricity generated by increased water head, η is the efficiency of the hydropower station, $η \in (0,1)$, Q_T is the maximum hydropower generation flow, Δ*H* is the increased water head caused by the adjustment of the FLWL, and ΔT is the calculation time length, here refers to the time length for beneft analysis.

4 Design of Multi‑objective Cooperative Decision‑making Model

4.1 Establishment of Multi‑objective Cooperative Decision‑making Model

In the efficient use of reservoirs, flood season operation is a typical multi-objective decision-making problem. Assuming that the decision-making system has *n* alternative schemes to choose from, the pros and cons of *n* schemes are judged according to the characteristic values of *m* indexes. Then the eigenvalue matrix of *m* indexes of each scheme is:

$$
X = \begin{bmatrix} x_{11} & x_{12} & \cdots & x_{1m} \\ x_{21} & x_{22} & \cdots & x_{2m} \\ \vdots & \vdots & \ddots & \vdots \\ x_{n1} & x_{n2} & \cdots & x_{nm} \end{bmatrix} = \{x_{ij}\}
$$
 (15)

where x_{ij} is the eigenvalue of the *j*-th index of the *i*-th scheme, $1 \le i \le n$ and $1 \le j \le m$.

In the actual decision-making, indexes can be divided into three categories: one is beneft index, the second is the risk index, and the third is an intermediate index. Due to the different units among these indexes, each index must be normalized.

For the beneft index, the equation of the relative membership degree of the evaluation index is (Fang et al. [2019\)](#page-15-12):

$$
r_{ij\max} = \frac{x_{ij} - x_{i\min}}{x_{i\max} - x_{i\min}}
$$
(16)

For the risk index, the equation is:

$$
r_{ij\,\text{min}} = \frac{x_{i\,\text{max}} - x_{ij}}{x_{i\,\text{max}} - x_{i\,\text{min}}}
$$
\n⁽¹⁷⁾

`For the intermediate index, the equation is

$$
r_{ijmid} = \begin{cases} 1 - \frac{x_i' - x_{ij}}{x_i^*} & x_{ij} < x_i' \\ 1 & x_i' \le x_{ij} \le x_i' \\ 1 - \frac{x_{ij} - x_i'}{x_i^*} & x_{ij} > x_i'' \end{cases} \tag{18}
$$

where $[x_i, x_i]$ is the best stable interval of x_{ij} . In Eqs. ([16](#page-5-0)) and [\(17\)](#page-5-1), x_{imax} and x_{imin} are respectively the largest and smallest eigenvalues in the *j*-th (1≤*j*≤m) index set, and $x_i^* = \max\{x_i' - x_{i\min}, x_{i\max} - x_i''\}.$

Then, we obtain the relative superior membership degree matrix *R*:

$$
R = \begin{bmatrix} r_{11} & r_{12} & \cdots & r_{1m} \\ r_{21} & r_{22} & \cdots & r_{2m} \\ \vdots & \vdots & \ddots & \vdots \\ r_{n1} & r_{n2} & \cdots & r_{nm} \end{bmatrix} = \{r_{ij}\}
$$
 (19)

Therefore, *m* objectives can be divided into *p* beneft indexes and (*m-p*) risk indexes. Then the problem can be described as:

$$
\max F(u_1(x), u_2(x))
$$

s.t. $g_k(x) \ge 0, k = 1, 2, \dots, T$ (20)

where $F(x)$ is the objective function, and T is the number of constraint conditions. The multi-attribute utility functions $u_1(x)$ and $u_2(x)$ are:

$$
u_1^j(x) = \omega_1 r_{1j} + \omega_2 r_{2j} + \dots + \omega_p r_{pj}
$$
 (21)

$$
u_2^j(x) = \omega_{p+1} r_{(p+1)j} + \omega_{p+2} r_{(p+2)j} + \dots + \omega_m r_{mj}
$$
 (22)

A decision maker can thus obtain the utility winning function by assigning weights to the multiple indexes. Based on the above, we construct a multi-objective cooperative decision-making model for reservoir operation in food seasons.

4.2 Solution of Multi‑objective Cooperative Decision‑making Model

We consider the problem of cooperative decision-making of two index sets, beneft index set DM_1 and risk index set DM_2 . Let $N = \{1,2\}$ be the sequence number of index sets. For any $i \in \mathbb{N}$, there are k_i indexes in the *i*-th decision indexes set.

$$
f_i(x_1, x_2) = (f_{i1}(x_1, x_2), f_{i2}(x_1, x_2), \cdots, f_{ik_i}(x_1, x_2)) \in G_i
$$
\n(23)

where $G_i \subset R^{k_i}$ is the reachable target set, $x_i \in \Omega_i \subset R^{m_i}$ is the decision vector, R^{k_i} and R^{m_i} are k_i -dimensional and m_i -dimensional real vector spaces, respectively. The winning function of the index set is represented by the utility function. The utility function of the *i*-th indicator set is:

$$
u_i(x_1, x_2) = u_i(f_i(x_1, x_2))
$$
\n(24)

A pair of utility functions is mapped into a subset *E* of the two-dimensional Euclidean space R^2 , and $E \subset R^2$ represents the joint reachable utility set.

$$
E = \left\{ \left(u_1(f_1), u_2(f_2) \right) \in R^2 \middle| f_i(x_1, x_2) \in G_i, x_i \in \Omega_i, i = 1, 2 \right\}
$$
 (25)

If *E* is a closed bounded convex set, $d = (d_1, d_2) \in E$ is the current point. The relationship between *E* and *d* is called a multi-objective negotiation decision-making problem. If the decision maker assigns $d_1 > d_2$, it means that the index set DM_1 has greater initiative in the process of negotiation, and the fnal negotiation result will also be biased towards the index set DM_1 and vice versa. Suppose that $h(E,d) \in E$ is the solution of (E,d) , let $h^*(E,d) = u^*$, for all $u \in E$, $u > d$ and $u \neq u^*$,

$$
\left(u_1^* - d_1\right)\left(u_2^* - d_2\right) > \left(u_1 - d_1\right)\left(u_2 - d_2\right) \tag{26}
$$

At this time, $h^*(E,d) = u^*$ becomes the Nash negotiation solution to the multi-objective problem.

If no feasible scheme is exactly the same as the Nash negotiation solution, the fuzzy pattern recognition theory is used to determine the closest, which is the optimal.

5 Application

5.1 Overview of the Study Reservoir

XHD reservoir on the West source of the Pi River Basin, which is one of the upper tributaries of the Huai River in China, is selected as the study object. The river basin is an area afected by the subtropical monsoon climate and frequently subjected to rainstorms and foods. The average annual precipitation in the Pi River basin is 1334 mm. XHD reservoir is a large size, 1 year regulating reservoir with a watershed area of 1400 km^2 . The Pi River basin and the location of XHD reservoir are shown as Fig. [1.](#page-7-0)

The annual average runoff of the reservoir is 33.1 m^3 /s. The main task of the reservoir is food control, and the comprehensive utilization of water supply and hydropower generation are taken into account. The design food control level is 139.10 m, the corresponding storage capacity is 2.228 billion m^3 , and the check flood control level is 143.60 m. The maximum hydropower generation flow of the power station is 200 m^3 /s.

Fig. 1 Maps of the Pi River basin and XHD reservoir location

The food season is from May 1st to October 1st, and the designed FLWL is 125.0 m. There is 500 million $m³$ of flood control capacity above 125.0 m that is responsible for downstream food control and peak reduction tasks. When the reservoir water level exceeds 132.60 m, the discharge tunnel is fully opened, with a maximum discharge volume of 618 m^3/s .

5.2 Segmentation of Flood Season

The restored daily infow of the reservoir during the food season of years 1956–2017 was used as the basic data for calculation. That is, the sample size $N=62$ years, and the length of the calculation period $T=154$ days.

5.2.1 Segmentation of Flood Season by Circular Distribution Method

The samples were taken according to different time intervals $t(t=1, 3, 5, 7, 10, 15)$ to stage the flood season. Firstly, the maximum discharge q_t , $(L=1, 2,..., n)$ of t days is selected from the *L*-th-year daily discharge in the flood season. Then, various indexes of flood situation are calculated by Eqs. (1) (1) (1) – (7) (7) . Table [1](#page-7-0) and Fig. 1 denote the results of segmentation of diferent sampling intervals.

The segmentation is obtained mainly based on the results of the maximum *t*-day discharge $(t=1, 3, 5)$ considering the flood magnitude. Table [1](#page-8-0) indicated that the calculated main food season is from June 10 to August 17.

Figure [2](#page-9-0) shows the occurrence time of the maximum discharge, which are expressed in angle. The distribution of data points considering and ignoring the magnitude of the food is expressed by the coordinate value of the occurrence time multiplied by the magnitude. Obviously, the points considering the magnitude are concentrated in the period from June 10 to August 17.

Sampling intervals t/d	Whether to con- sider the flood level	\mathbf{r}	$\overline{\alpha}$	\overline{D}	Pre-flood season	Main flood season	Post-flood season
1	N		0.352 2.580	7/2	$5/1 - 5/17$	$5/18 - 8/7$	$8/8 - 10/1$
	Y	0.392	3.072 7/14		$5/1 - 6/9$	$6/10 - 8/17$	$8/18 - 10/1$
3	N	0.497	2.366 6/27		$5/1 - 5/28$	5/29-7/26	$7/27 - 10/1$
	Y		0.418 2.838 7/9		$5/1 - 6/9$	$6/10 - 8/17$	$8/18 - 10/1$
5	N	0.230	2.453	6/29	$5/1 - 5/17$	$5/18 - 8/10$	$8/11 - 10/1$
	Y	0.270	3.233	7/18	$5/1 - 6/9$	$6/10 - 8/17$	$8/18 - 10/1$
7	N	0.437	2.038 6/19		$5/1 - 5/17$	$5/18 - 7/21$	$7/22 - 10/1$
	Y	0.349	2.703	7/5	$5/1 - 5/30$	$5/31 - 8/10$	$8/11 - 10/1$
10	N	0.438	1.978 6/17		$5/1 - 5/16$	5/17-7/19	$7/20 - 10/1$
	Y	0.359	2.553 7/2		$5/1 - 5/26$	$5/27 - 8/6$	$8/7 - 10/1$
15	N	0.539	1.969	6/17	$5/1 - 5/20$	$5/21 - 7/15$	$7/16 - 10/1$
	Y		0.456 2.362 6/27		$5/1 - 5/26$	5/27-7/28	$7/29 - 10/1$

Table 1 The results of food season segmentation by circular distribution method

Based on the above, combining the actual situation of the reservoir, the main food season of the reservoir is gained as June 10—August 17.

5.2.2 Segmentation of Flood Season by the Relative Frequency Method

The whole food season is divided into 15 periods (October 1 is classifed as the late September). In each period, the occurrence times of daily maximum discharge over the years 1956–2017 are counted to calculate the relative frequency. The results are shown in the following table.

It can be seen from the Table [2](#page-10-0) that the relatively high frequency is mainly concentrated in the frst ten days of June to the middle of August, which is consistent with the food season segmentation results by the circular distribution method.

5.3 Analysis of Risk and Benefit

5.3.1 Risk Analysis

Several typical food hydrographs are selected in the pre-food season and post-food season severally, and the food hydrographs with diferent design frequencies are derived. Considering the food situation in the reservoir area, the values of frequency *P* are selected as 0.01%, 0.10%, 0.20%, 1%, 2%, 3.30%, 5%, 10% and 20% respectively. For the selection of the dynamic domain of FLWL, considering that the original FLWL of the reservoir is 125.0 m, and the normal water level is 130.0 m, the FLWLs start at 125.0 m and increase at a step of 0.5 m to a maximum of 130.0 m.

The food hydrograph that is most unfavorable for the reservoir food control is selected. According to the design results of the reservoir, in the pre-food season, the design food peak is much smaller than that in the main food season, but the food volume is large and the peaks are dense. The flood in 1974 with peak discharge of $3,938.84 \text{ m}^3/\text{s}$ is selected as the typical. In the post-food season, the design food is lighter than that in the main food season. While compared with the pre-food season, it has a high peak, large volume and

many fluctuations. The flood in 2005 with peak discharge of $7,942.69 \text{ m}^3/\text{s}$ is chosen as the typical.

Then, taking Z_i of each scheme as the starting water level, the time interval is 3 h, flood regulation is carried out according to the predetermined dispatching rules. The relationship between the highest flood regulation water level $Z_m(Z_i)$ and the corresponding frequency $P(Z_i)$, *i*=1, 2, …, *n* is obtained.

Taking the designed food control level (139.10 m) as the selected safe level, the frequency when the food regulation upper water level starting from each FLWL is exactly equal to the selected safe level is obtained. The results of diferent schemes and corresponding extreme risk rates are displayed in Table [3](#page-11-0).

Figure 3 presents the relationship curve between different FLWLs Z_i and the corresponding extreme risk rate P_{lim} .

5.3.2 Benefit Analysis

XHD reservoir supplies water mainly including industrial water, agricultural water, domestic water and ecological water, of which 70% is used for agricultural water. According to relevant research, the reservoir water supply beneft generated by per cubic metre water is 35.40 yuan (Zheng [2020\)](#page-16-6). The hydropower generation operation is subject to the food control and irrigation operation. According to the actual operation data of the hydropower station (Yu 2010), the average water consumption rate per unit electric energy was 9.786 m³.

In the pre-food season, to meet the storage requirements of the incoming food, the reservoir needs to empty its water in advance. The beneft of raising the FLWL only considers the hydropower generation, which positively infuenced by the increase of average water head. Equation [\(14\)](#page-5-2) is used to calculate the increased hydropower generation beneft.

In the post-food season, the reservoir begins to store water to improve the utilization beneft. Therefore, the benefts include not only the hydropower generation beneft caused by increased water head and volume, but also the water supply beneft. According to the

measured hydrological data of the watershed over the years (Hu [2015\)](#page-15-13), the proportion of wet years that can meet the water supply needs without raising the FLWL is 25%. Deducting the proportion of these years, the actual increased water supply beneft by raising the FLWL can be obtained. Then Eq. [\(12\)](#page-4-0) is used to calculate the water supply beneft. Next, Using Eqs. ([13](#page-4-1)) and ([14](#page-5-2)), the hydropower generation beneft caused by increased water head and volume can be calculated. Table [4](#page-12-0) presents the beneft results in the pre-food season and the post-food season, respectively.

5.4 Nash Negotiation Solution

The value of each index is standardized by Eqs. [\(16\)](#page-5-0) and [\(17\)](#page-5-1). Considering that the reservoir focuses on food control and water supply, supplemented by hydropower generation, the weight of the two beneft indexes, namely water supply and hydropower generation, is (0.6, 0.4). The utility payoff function values u_1 and u_2 are calculated by Eqs. ([21](#page-6-0)) and ([22](#page-6-1)). Table [5](#page-12-1) shows the solution of payoff function values of each scheme.

(b) Extreme risk rate under different flood limited water levels in the post-flood season

Water level (m)	Pre-flood season	Post-flood season						
	Hydropower generation benefit (10^4 kW-h)	Increased water supply benefit	Increased hydropower generation benefit					
		(10^8 yuan)	Water volume benefit (10^4 kW-h)	Water head benefit (10^4 kW-h)				
125.0	0.000	0.000	0.000	0.000				
125.5	96.052	8.031	231.839	87.675				
126.0	192.104	16.129	465.593	175.350				
126.5	288.157	24.307	701.647	263.025				
127.0	384.209	32.577	940.384	350.700				
127.5	480.261	40.874	1179.886	438.375				
128.0	576.313	49.237	1421.305	526.051				
128.5	672.365	57.680	1665.023	613.726				
129.0	768.418	66.229	1911.807	701.401				
129.5	864.470	74.937	2163.189	789.076				
130.0	960.522	83.879	2421.315	876.751				

Table 4 Hydropower generation beneft and water supply beneft of diferent schemes in the pre-food season and post-food season

Firstly, it is assumed that the winning function values of the two index sets are both 0, that is, the current point $d = (0,0)$. In order to obtain the joint reachable utility set *E* of the two index sets, the winning function values (u_1, u_2) of each flood season are marked, as shown in Fig. [4](#page-13-0) below. For u_1 and u_2 , the maximum value of the product should appear in the range where the values of the two variables are closest, corresponding to the lines of AB and BC in Fig. [4.](#page-13-0) Then the problem of Nash negotiation is transformed into the following nonlinear programming problems. The product value is optimized during the two lines respectively, and the objective function is as follows.

For pre-flood season:

$$
\begin{aligned}\n\max \ u_1 \cdot u_2 \\
s.t. \ 0 \le u_2 \le 1 \\
0.927u_1 + u_2 \le 1.285 \\
1.581u_1 + u_2 \le 1.722\n\end{aligned}
$$

For post-flood season:

$$
\max u_1 \cdot u_2
$$

s.t. $0 \le u_1 \le 1$, $0 \le u_2 \le 1$
 $0.842u_1 + u_2 \le 1.252$
 $1.493u_1 + u_2 \le 1.683$

Based on the above, the Nash negotiation solution u_1 in the pre-flood season is 0.918 and u_2 is 0.669. In the post-flood season, u_1 is 0.658 and u_2 is 0.874. Since the two negotiation solutions are not equal to u_1 and u_2 of each scheme, it is necessary to adopt the fuzzy pattern recognition model and grid schedule theory to calculate the membership degree and decide the optimal scheme. Table [6](#page-13-1) presents the membership degree of each scheme, and then the optimal water level can be determined.

The membership degree is the largest when the water level in the pre-food season is 129.0 m and in the post-flood season is 128.5 m. When $d = (0,0)$, the FLWLs of the optimal scheme are 129.0 m and 128.5 m for the pre-food season and the post-food season, respectively.

The water level at diferent sub-seasons of the food season should be considered for specifc situations. The food control safety is mainly considered in the pre-food season,

Takis & Themostomp degree of each beneme in the pre-hood beabon and poor hood beabon										
Water level /m					125.0 125.5 126.0 126.5 127.0 127.5 128.0 128.5 129.0 129.5 130.0					
Pre-flood season 0.082 0.152 0.225 0.301 0.382 0.473 0.577 0.709 0.895 0.820 0.331										
Post-flood season 0.342 0.424 0.509 0.597 0.690 0.791 0.908 0.956 0.777 0.528 0.126										

Table 6 Membership degree of each scheme in the pre-food season and post-food season

Decision preference Pre-flood season				Post-flood season				
	Equally valued Risk-biased				Equally valued Benefit-biased			
(d_1,d_2)	(0,0)		$(0,0.1)$ $(0,0.3)$ $(0,0.5)$ $(0,0)$				$(0.1,0)$ $(0.3,0)$ $(0.5,0)$	
Flood limited water level	129.0		129.0 128.5 128.0		128.5	128.5	129.0	129.5

Table 7 Optimal flood limited water level under different decision preference

so that the decision preference is targeting on risk assessment. In the post-food season, to ensure water supply, decision preference will focus on beneft assessment. A decisionmaker would change the preference by changing the current point. Table [7](#page-14-0) presents the specific results. It can be deduced that the more d_2 increases, the lower the FLWL, and the better the flood control safety can be guaranteed. On the contrary, the more d_1 increases, the higher the FLWL, and the more utilization benefts can be improved.

6 Conclusions

- 1. We use the circular distribution method and relative frequency method to calculate the flood segmentation. The main flood season of XHD reservoir is gained as June 10— August 17. The results of circular distribution method are verifed by relative frequency method, and the verifcation results are consistent.
- 2. We establish the index sets of risk and beneft respectively and use the Nash negotiation theorem to solve the utility payoff function of the two index sets, and, based on the game theory, we establish a multi-objective cooperative decision-making model.
- 3. Taking XHD reservoir as an example, we obtain an optimal FLWL control scheme for sub-seasons. The resulting FLWLs of the optimal scheme are 129.0 m, and 128.5 m in the pre-food season and the post-food season, respectively in the condition that the risk and beneft are equally valued. By adjusting the current point to refect decisionmakers' preference, the decision-makers' expectations for risk and beneft in diferent sub-seasons can be realized through this study.

In the future, we should deeply analyze the characteristics of rainstorm and food in combination with the hydrological and meteorological characteristics of the basin where the reservoir is located. Based on the real-time rainfall and fow data, the FLWL of each stage more in line with the reality could be achieved.

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Availability of Data and Materials The datasets used or analysed during the current study are available from the corresponding author on reasonable request.

Declarations

Ethics Approval All authors kept to the Ethical Responsibilities of Authors.

Consent to Participate All of the authors consent to participate in the relevant research content in this paper.

Consent for Publication All the authors have approved the submission and consented for the publication.

Competing Interests The authors declare that they have no confict of interest.

References

- Chen L, Guo SL, Yan BW, Liu P, Fang B (2010) A new seasonal design food method based on bivariate joint distribution of food magnitude and date of occurrence. Hydrolog Sci J 55(8):1264–1280. [https://](https://doi.org/10.1080/02626667.2010.520564) doi.org/10.1080/02626667.2010.520564
- Cunderlik JM, Ouarda TBMJ, Bobée B (2004) On the objective identifcation of food seasons. Water Resour Res 40(1):1520. <https://doi.org/10.1029/2003WR002295>
- Fang YH, Zheng XL, Peng H, Wang H, Xin J (2019) A new method of the relative membership degree calculation in variable fuzzy sets for water quality assessment. Ecol Indic 98:515–522. [https://doi.org/10.](https://doi.org/10.1016/j.ecolind.2018.11.032) [1016/j.ecolind.2018.11.032](https://doi.org/10.1016/j.ecolind.2018.11.032)
- Hu JL (2015) The study on flood period stage and limited water levels of Meishan Reservoir in Anhui province. Dissertation, Central China Normal University (in Chinese)
- Huang XF, Chen YQ, Lin J, Fang GH, Qu XP, Zhu LX (2017) Research on beneft of reservoir food resources utilization based on the dynamic control of limited water level. Desalin Water Treat 79:214– 220. <https://doi.org/10.5004/dwt.2017.20857>
- Jiang H, Wang ZZ, Ye A, Liu KL, Wang XH, Wang LJ (2019) Hydrological characteristic-based methodology for dividing food seasons: an empirical analysis from China. Environ Earth Sci 78(14). [https://doi.](https://doi.org/10.1007/s12665-019-8392-z) [org/10.1007/s12665-019-8392-z](https://doi.org/10.1007/s12665-019-8392-z)
- Jiang HY, Yu ZB, Mo CX (2015) Reservoir Flood Season Segmentation and Optimal Operation of Flood-Limited Water Levels. J Hydrol Eng 20(9):05014035. [https://doi.org/10.1061/\(asce\)he.1943-5584.](https://doi.org/10.1061/(asce)he.1943-5584.0001151) [0001151](https://doi.org/10.1061/(asce)he.1943-5584.0001151)
- Ju B, Yu YC, Zhang FH, Lei XH, You FH (2020) Flood season partition and food limited water level determination for cascade reservoirs downstream jinshajiang river. IOP Conference Series: Earth Environ Sci 569(1).<https://doi.org/10.1088/1755-1315/569/1/012005>
- Li Q, Yu HF, Hu LL, Yu WG, Chi YF, Li H (2018) Study on the stage of food season in typhoon afected area - a case study in Tingxia. IOP Conf Ser Earth Environ Sci 189(5). [https://doi.org/10.1088/1755-](https://doi.org/10.1088/1755-1315/189/5/052033) [1315/189/5/052033](https://doi.org/10.1088/1755-1315/189/5/052033)
- Liu GJ, Qin H, Shen Q, Tian R, Liu YQ (2019) Multi-objective optimal scheduling model of dynamic control of food limited water level for cascade reservoirs. Water 11(9):1836. [https://doi.org/10.3390/](https://doi.org/10.3390/w11091836) [w11091836](https://doi.org/10.3390/w11091836)
- Ma C, Xu R, He W, Xia JJ (2020) Determining the limited water level of early food season by combining multi-objective optimization scheduling and copula joint distribution function: A case study of three gorges reservoir. Sci Total Environ.<https://doi.org/10.1016/j.scitotenv.2020.139789>
- Mo CX, Deng J, Lei XB, Ruan YL, Lai SF, Sun GK, Xing ZX (2022) Flood season staging and adjustment of limited water level for a multi-purpose reservoir. Water (Switzerland) 14(5). [https://doi.org/10.3390/](https://doi.org/10.3390/w14050775) [w14050775](https://doi.org/10.3390/w14050775)
- Ruan Y, Singh VP (2008) Multiple duration limited water level and dynamic limited water level for food control, with implications on water supply. J Hydrol 354(1–4):160–170. [https://doi.org/10.1016/j.jhydrol.](https://doi.org/10.1016/j.jhydrol.2008.03.003) [2008.03.003](https://doi.org/10.1016/j.jhydrol.2008.03.003)
- Singh VP, Wang SX, Zhang L (2005) Frequency analysis of nonidentically distributed hydrologic food data. J Hydrol 307(1–4):175–195.<https://doi.org/10.1016/j.jhydrol.2004.10.029>
- Tan QF, Wang X, Liu P, Lei XH, Cai SY, Wang H, Ji Y (2017) The dynamic control bound of food limited water level considering capacity compensation regulation and food spatial pattern uncertainty. Water Resour Manag 31(1):143–158.<https://doi.org/10.1007/s11269-016-1515-3>
- Villarini G (2016) On the seasonality of fooding across the continental United States. Adv Water Resour 87:80–91.<https://doi.org/10.1016/j.advwatres.2015.11.009>
- Wu X, Zhong P (2018) The application and research of dynamic control the flood limited water level of wuqiangxi reservoir. MATEC Web Conf. <https://doi.org/10.1051/matecconf/201824601055>
- Xie AL, Liu P, Guo SL, Zhang XQ, Jiang H, Yang G (2018) Optimal design of seasonal food limited water levels by jointing operation of the reservoir and foodplains. Water Resour Manag 32(1):179–193. <https://doi.org/10.1007/s11269-017-1802-7>
- Xu B, Huang X, Mo R, Zhong PA, Lu QW, Zhang HW, Si W, Xiao JF, Sun Y (2021) Integrated real-time food risk identifcation, analysis, and diagnosis model framework for a multireservoir system considering temporally and spatially dependent forecast uncertainties. J Hydrol. [https://doi.org/10.1016/j.](https://doi.org/10.1016/j.jhydrol.2021.126679) [jhydrol.2021.126679](https://doi.org/10.1016/j.jhydrol.2021.126679)
- Yu CX (2010) Different period flood analysis and dispatch opinion for the Xianghongdian Reservoir. Jianghuai Water Resour Sci and Technol 2010(1):19–21 (in Chinese)
- Zheng HR (2020) Analysis and evaluation of food resources utilization in the river basin. Dissertation, Hohai University (in Chinese)
- Zhou YL, Guo SL (2014) Risk analysis for food control operation of seasonal food-limited water level incorporating infow forecasting error. Hydrol Sci J 59(5):1006–1019. [https://doi.org/10.1080/02626667.2014.](https://doi.org/10.1080/02626667.2014.901515) [901515](https://doi.org/10.1080/02626667.2014.901515)

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