RESEARCH ARTICLES

J. Sci. Soc. Thailand, 5 (1979) 73-87

THE WATER STORAGE PROBLEM WITH INDEPENDENT NORMAL INFLOWS

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(Received 1 March 1979)

Summary

Assuming that the annual inflows to the reservoir under consideration are distributed as independent normal variables and the annual outflows are equal to the sample mean of the inflows, the distributions of the water content in various years, the maximum amounts of water in surplus and in deficit, and the storage capacity were investigated under the condition that the reservoir allows neither spillage nor emptiness during its lifetime. The water contents in various years were found to be distributed as normal variables; the maximum amount of water in surplus or in deficit has a probability mass at zero and a probability density function elsewhere, which was fitted by the Type I Pearson curve; and the storage capacity was fitted by the Type III curve. All the fittings were assessed by using the Chi-square test of goodness of fit.

Introduction

The planning and design of a reservoir always involves the determination of its storage capacity, which depends upon the conditions imposed. Suppose that the annual flows to the reservoir in different years, denoted by X_k , come from the same population distribution. Suppose furthermore that during a lifetime of n years, the annual outflows are constants equal to the sample mean \overline{X}_n :

$$\overline{X}_{n} = \sum_{k=1}^{n} X_{k}/n \tag{1}$$

The problem is to determine the reservoir size or the reservoir storage capacity in order that during the lifetime the reservoir never spills and never runs dry.

Let S_i^* denote the storage of the reservoir or the water content in the ith year with reference to the initial level, for i = 0, 1, ..., n, where $S_0^* = 0$, then by means of the continuity equation

$$S_i^* - S_{i-1}^* = X_i - \overline{X}_n$$

it follows that

$$S_i^* = \sum_{k=1}^i (X_i - \overline{X}_n)$$

Let $S_i = \sum_{k=1}^{1} X_k$, the partial sum of X_k ; it can be seen that

$$S_i^* = S_i - \frac{i}{n} S_n, i = 1, 2, ..., n$$
 (2)

In this form, S_i is called the adjusted partial sum.

Define

$$M_n^* = \max (0, S_1^*, ..., S_n^*)$$

 $m_n^* = \min (0, S_1^*, ..., S_n^*)$
(3)

and

$$R_n^* = M_n^* - m_n^*,$$

 M_n^* is called the adjusted surplus, m_n^* the adjusted deficit, and R_n^* the adjusted range. To satisfy the stated conditions, the storage capacity of the reservoir should be optimally equal to the adjusted range R_n^* . The random variables M_n^* and m_n^* are then the maximum amounts of water in surplus and in deficit over n years, respectively. The random variables S_i^* , M_n^* , m_n^* and R_n^* , illustrated in Fig. 1 are therefore of important interest in the field of water resources development.

Hurst¹ collected a large amount of statistical material relating to water levels and other phenomena, and came to the conclusion that the observed adjusted range has a mean proportional to n^c, where c varies from 0.69 to 0.80, with a mean of 0.729 and a standard deviation of 0.092. This finding, commonly referred to as the Hurst phenomenon, has been a stimulant for a great deal of research. Recently, it has been considered to be related to the rescaled adjusted range².

Feller³ derived the asymptotic distribution of R_n^* and its asymptotic mean and variance. For independent variables of zero mean and unit variance, the results are

$$E(R_n^*) \doteq (\pi n/2)^{1/2}$$
 (4)

$$Var(R_n^*) \doteq \pi(\pi/6-1/2)n \tag{5}$$

Solari and Anis⁴ introduced the random variable U_n^* called the maximum of adjusted partial sums which is defined by

$$U_n = \max (S_1^*, ..., S_n^*)$$

and derived the exact formulas for its first two moments as follows:

$$E(U_n^*) = \frac{1}{2} \left(\frac{n}{2\pi}\right)^{1/2} \sum_{i=1}^{n-1} i^{-1/2} (n-i)^{-1/2}$$
 (6)

$$E(U_n^{*2}) = \frac{1}{6} \left\{ \frac{n^2 - 1}{n} + \frac{\sqrt{n}}{2\pi} \sum_{i=2}^{n-1} \sum_{j=1}^{i-1} i(2i - n) \left[(n - i)j^3 (i - j)^3 \right]^{-1/2} \right\}$$
 (7)

for $n \ge 2$

Salas-La Cruz⁵ obtained the expected value of the adjusted range for the case of independent standard normal variables. The result was:

$$E(R_n^*) = \left(\frac{n}{2\pi}\right)^{1/2} \sum_{i=1}^{n-1} \frac{2(n-i)^{1/2}}{ni^{1/2}}$$
 (8)

Later, Boes and Salas-La $Cruz^6$ and Salas-La Cruz and $Boes^7$ derived the expected value of R_n^* for the case where the outflow is only a fraction of the sample mean \overline{X}_n . They also investigated the asymptotic behaviour of $E(R_n^*)$ in connection with the Hurst phenomenon.

Sutabutr⁸, based on computed values of the Pearson criterion from simulated samples, suggested the use of the Type III curve of the Pearson system to fit the distribution of R_n^* .

As seen before, the random variables adjusted surplus, adjusted deficit, and adjusted range are very important in the storage problem. An attempt is made in this study to obtain their distributions when X_k are distribution as independent normal variables.

Expected Values of M_n^* , m_n^* and R_n^*

It should be noted that for $n=1,\ \overline{X}_n=X_1,$ hence $S_1^*=0,$ and consequently $M_1^*=m_1^*=R_i^*=0.$ This case is excluded from the following analysis. For $n\geq 2,$ the sequence $Y_1,\ Y_2,\ \ldots,\ Y_n$ where $Y_k=X_k-\overline{X}_n$ is exchangeable $^6;$ thus the following equation 9 holds

$$E(M_n^*) = \sum_{i=1}^{n} E(V_i)/i$$
 (9)

where

$$V_i = \max (0, S_i^*) = |S_i^*| / 2 + S_i^* / 2$$
 (10)

Observe that since

$$m_n^* = \min (0, S_1^*, ..., S_n^*)$$

= -max (0, -S_1^*, ..., -S_n^*)

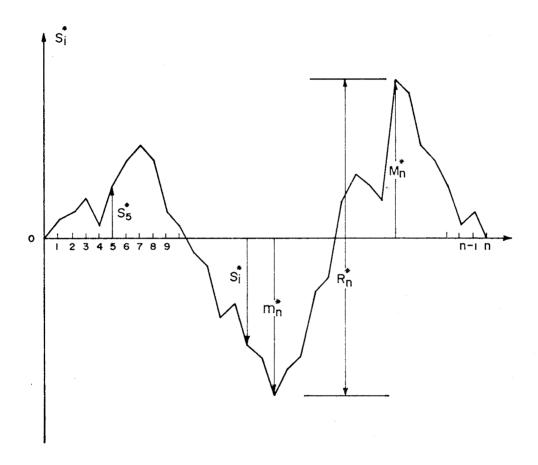


Fig. 1. Definition of the adjusted partial sum S_i^* , adjusted surplus M_n^* , adjusted deficit m_n^* . and adjusted range R_n^* .

it follows that

$$E(m_n^*) = -\sum_{i=1}^{n} E(W_i) /i$$
 (11)

where

where
$$W_{i} = \max_{i} (0, -S_{i}^{*}) = |S_{i}^{*}| / 2 - S_{i}^{*} / 2$$
From the definition of S_{i}^{*} , (12)

$$E(S_i^*) = 0$$

thus,
$$E(V_i) = E(W_i) = E(|S_i^*|) /2$$
 (13)

and from (9) and (11),

$$E(M_n^*) = -E(m_n^*) \tag{14}$$

$$E(R_n^*) = E(M_n^*) - E(m_n^*)$$

$$E(M_n^*) = -E(m_n^*) \tag{14}$$
 The definition of R_n^* leads to:
$$E(R_n^*) = E(M_n^*) - E(m_n^*)$$
 or
$$E(R_n^*) = 2E(M_n^*) \tag{15}$$

Equations (14) and (15) are abtained using the fact that $E(S_i^*) = 0$, which in turn results from the condition $E(Y_k) = 0$, k = 1, ..., n, where $Y_k = X_k - \overline{X}_n$ is called the net input in the kth year.

These equations state that the expected values of the adjusted surplus and adjusted deficit have the same magnitude, and the expected value of the adjusted range is double of the adjusted surplus, provided that the expected value of the net inputs in different years are identically equal to zero.

For the case of independent normal inflows with mean μ and standard deviation σ .

$$E(S_i) = \sum_{k=1}^{1} E(X_k) = i\mu,$$

and

$$Var(S_i) = \sum_{k=1}^{i} Var(X_k) = i\sigma^2$$

The expected value of S_i^* is zero, and the variance can be written, using (2), as

$$Var(S_i^*) = Var(S_i) + (i/n)^2 Var(S_n) - 2i Cov(S_i, S_n) /n$$

Since

$$Cov(S_i, S_n) = i\sigma^2,$$

$$Var(S_i^*) = i(n-i)\sigma^2/n.$$
 (16)

It should be noted that \overline{X}_n is a normal variable, so is Y_k , and therefore S_i^* is a normal variable. The knowledge of the expected value and variance of S_i thus completely determines the distribution of S_i.

In summary, the storage of the reservoir in the ith year, $i \le n$, is distributed as a normal variable with zero mean, and variance given by (16). The expected value of |S_i | can be easily shown to be

$$E(|S_i^*|) = (2/\pi)^{1/2} [i(n-i)/n]^{1/2} \sigma$$

Substituting this expresssion into (13) yields

$$E(V_i) = E(W_i) = (2\pi)^{-1/2} [i(n-i)/n]^{1/2} \sigma$$
 (17)

The expected values of M_n^* and m_n^* are obtained using (9) and (11), respectively and the expression of $E(V_i)$ and $E(W_i)$ in (17):

$$E(M_n^*) = -E(m_n^*)$$

$$= (2\pi)^{-1/2} \sigma \sum_{i=1}^n \left(\frac{n-i}{ni}\right)^{1/2}$$
(18)

and the expected value of R_n is given by (15)

$$E(R_n^*) = (2/\pi)^{1/2} \sigma \sum_{i=1}^n \left(\frac{n-i}{ni}\right)^{1/2}$$
 (19)

Second Moments of M_n, m_n and R_n

For n = 2,

$$M_2^* = \max (0, S_1^*, S_2^*) = V_1$$

because $S_2^* = 0$. Thus
 $E(M_2^{*2}) = E(V_1^2) = E(S^{*2}) / 2 = \sigma^2 / 4$ (20)
Similarly, $m_2^* = -W_1$, and
 $E(m_2^{*2}) = E(W_1^2) = E(S_1^{*2}) / 2 = \sigma^2 / 4$ (21)

$$E(m_2^{*2}) = E(W_1^2) = E(S_1^{*2}) / 2 = \sigma^2 / 4$$
 (21)

For $n \ge 3$, the second moments of M_n^* and m_n^* can be obtained from the formula of Solari and Anis⁴, by noting that the random variable $U_n^* = \max (S_1^*, S_2^*, ..., S_{n-1}^*, S_n^*)$

$$U_n^* = \max (S_1^*, S_2^*, ..., S_{n-1}^*, S_n^*)$$

$$U_n^* = \max(S_1^*, S_2^*, ..., S_{n-1}^*, 0) = M_n^*$$
 (22)

corresponding to the condition $S_n^* = 0$ becomes $U_n^* = \max_1 (S_1^*, S_2^*, ..., S_{n-1}^*, 0) = M_n^*$ The second moment of M_n^* is then derived from (7)

$$E(M_n^{*2}) = \frac{\sigma^2}{6} \left\{ \frac{n^2 - 1}{n} + \frac{\sqrt{n}}{2\pi} \sum_{i=1}^{n-1} \sum_{j=1}^{i-1} i(2i - n) \left[(n - i)j^3 (i - j)^3 \right]^{-1/2} \right\}$$
 (23)

for $n \geq 3$.

Equation (23) obviously can apply to the second moment of m_n^* for normal variables. Based on the identification between U_n^* and M_n^* , one can show that the expected value of M_n^* derived in (18) is equivalent to the expected value of U_n^* obtained by Solari and Anis⁴ in (6) for the case $\sigma = 1.0$.

The exact formula for the second moment of R_n^* can be obtained for n = 2. It is clear that $R_2^* = S_1^*$, thus

$$E(R_2^{*2}) = \sigma^2 / 2 \tag{24}$$

For $n \ge 3$, the Monte Carlo method was used to compute the variance of R_n^* from simulated samples. It was found that for a size of 5,000, the values of $E(M_n^*)$

M 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1	the state of the s	The second secon
n	From approximate formula	Simulated values
. 4	0.3589	0.3672
6	0.5071	0.5063
8	0.6553	0.6491
10	0.8035	0.8145
15	1.1740	1.1738
20	1.5445	1.5530
25	1.9150	1.9097
30	2.2855	2.2863
35	2.5660	2.6596
40	3.0265	3.0198
45	3,3970	3.3985
50	3.7675	3.7662

TABLE I: VARIANCE OF R_n^* FOR THE CASE OF INDEPENDENT NORMAL VARIABLES $(\sigma = 1.0)$

computed in this method were close to those computed from the exact formula of (23), thus for each n, a sample of size equal to 5,000 was used. From the computed values, the variance of R_n^* was found to be approximated by the following equation:

$$Var(R_n^*) = (0.0741n + 0.0625) \sigma^2$$
 (25)

A comparison between the values of Var (R_n^*) given by (25) and those computed from simulated samples is shown in Table I for the case $\sigma = 1.0$. It should be noted that for n approaching infinity, Var (R_n^*) given by (25) approaches 0.0741n. σ^2 , the asymptotic value given by Feller³.

Third and Fourth Moments of Mn and mn

The third and fourth moments of M₂ and m₂ can be shown to be given by

$$E(M_2^{*3}) = -E(m_2^{*3}) = (4\pi)^{-1/2} \sigma^3$$
 (26)

and
$$E(M_2^{*4}) = E(m_2^{*4}) = 3\sigma^4 / 8.$$
 (27)

For $n \ge 3$, the Monte Carlo method was used to compute the third and fourth moments of M_n^* . The computed values of these moments can be approximated by

$$\mu_3 = 0.03077 \text{ (n-1)}^{1.43572}.\sigma^3$$
 (28)

and
$$\mu_4 = 0.05923 \, (n-1)^{1.89966} \cdot \sigma^4$$
 (29)

where μ_3 and μ_4 denote the third and fourth central moments of M_n^* , respectively. The constants in (28) and (29) were obtained by means of the non-linear least squares method. It should be noted that (28) and (29) also apply to the third and fourth central moments of $-m_n^*$. A comparison between the simulated values and the values computed from the approximate formulas for these moments is made in Table II.

		μ_3	μ_{4}		
n	Approximate formula	Simulated values	Approximate formula	Simulated values	
4	0.14899	0.14481	0.47743	0.47922	
6	0.31023	0.32797	1.25994	1.23965	
8	0.50290	0.54035	2.38750	2.39696	
10	0.72142	0.73380	3.84841	3.83481	
15	1.36044	1.38171	8.9083 2	8.896 0 4	
20	2.10908	2.09347	15.91265	15.86346	
25	2.94956	2.94701	24.80150	24.83179	
30	3.87039	3.86099	35.5037 8	35.52419	
35	4.86336	4.87649	48.06576	48.07655	
40	5 .92 222	5.90418	62.37755	62.36756	
45	7.04204	7.03930	78.44182	78.46499	
50	8.21882	8.20401	96.23752	96.19436	

TABLE II: THIRD AND FOURTH CENTRAL MOMENTS OF M_n^*FOR THE CASE OF INDEPENDENT NORMAL VARIABLES ($\sigma = 1.0$)

Distribution of M_n

Having obtained the first four moments of M_n^* , the moments ratios β_1 , β_2 and the Pearson criterion κ can be computed from the equations

$$\beta_{1} = \mu_{3}^{2}/\mu_{2}^{3}$$

$$\beta_{2} = \mu_{4}/\mu_{2}^{2}$$

$$K = \frac{\beta_{1} (\beta_{2} + 3)^{2}}{4(2\beta_{2} - 3\beta_{1} - 6) (4\beta_{2} - 3\beta_{1})}$$

and

where μ_r , r = 2, 3, 4 are the rth central moments of the random variable.

The results are listed in Table III. For these values of β_1 , β_2 and κ , the most appropriate frequency curve to be selected in approximating the distribution of M_n^* is the Type I curve of the Pearson system^{10,11}. The general equation of the Type I curve, also referred to as the beta distribution, is

TABLE III: NUMERICAL VALUES OF THE MOMENT RATIOS AND OF THE PEARSON CRITERION

n	$oldsymbol{eta}_1$	$oldsymbol{eta_2}$	κ
5	0.34649	2.96955	-0.25882
10	0.42126	3.34278	-0.60521
15	0.44183	3.42851	-0.78652
20	0.44789	3.44436	-0.82205
25	0.44872	3.43680	-0.79315
30	0.44735	3.42036	-0.74523
35	0.44501	3.40050	- 0.69571
40	0.44225	3.37960	-0.65032
45	0.43933	3.35879	-0.61042
50	0.43638	3.33846	-0. 57559

$$f(x) = \frac{b^{1-p-q}}{B(p,q)} (a_1 + x)^{p-1} (a_2 - x)^{q-1}, a_1 \le x \le a_2$$
 (30)

where

p, q = two parameters,

 a_1 , a_2 = two terminals,

$$b = a_1 + a_2,$$

and B(p, q) is the beta function of p, q, defined by

$$B(p,q) = \int_{0}^{1} t^{p-1} (1-t)^{q-1} dt$$

Among the three cases listed by Pearson and Johnson¹², only the first can apply. For this, the procedure of determining p, q, a_1 and a_2 is as follows. Letting

$$\tau = \frac{6(\beta_2 - \beta_1 + 1)}{3\beta_1 - 2\beta_2 + 6}$$

$$\varepsilon = \frac{\tau^2}{4 + \frac{1}{4} \beta_1 (\tau + 2)^2 / (\tau + 1)}$$

it can be shown that $\tau = p+q$ and $\varepsilon = pq$, so that p and q are the roots of the quadratic

$$Z^2 - \tau Z + \varepsilon = 0 \tag{31}$$

In this case $\mu_3 > 0$, p is then the smaller or the two roots. The value of b can be expressed¹¹ as

$$b = \tau [\mu_2 (\tau + 1)/\epsilon]^{1/2}$$
(32)

Locating the origin of the coordinates at the mode of the frequency curve allows a_1 and a_2 to be determined from

$$\frac{a_1}{p-1} = \frac{a_2}{q-1} = \frac{b}{p+q-2} = \frac{b}{r-2}$$
 (33)

The determination of p and q from (31) and a_1 and a_2 from (33) constitutes the process of fitting the distribution of M_n . The fitted curve can be located using the knowledge that the distance from the mode to the mean (expected value) is determined by

$$\delta = \frac{bp}{p+q} - a_1 \tag{34}$$

For any value of n, the distance \mathcal{E} from the start of the curve to the mean, bq/(p+q), is quarter than the expected value $\nu_1 = E(M_n^*)$ of M_n^* . This implies that the frequency curve starts at a point to the left of the initial or zero level (Fig. 2). The area under the curve between the start $-a_1$ and the initial level $X_0 = \delta - \nu_1$ then represents the probability mass of the adjusted surplus at zero:

$$Pr = Prob (M_n^* = 0) = I_{x_0} (p,q) = \frac{1}{B(p,q)} \int_0^{x_0} t^{p-1} (1-t)^{q-1} dt$$
 (35)

where
$$x_0 = \frac{a_1 + x_0}{b}$$
 (36)

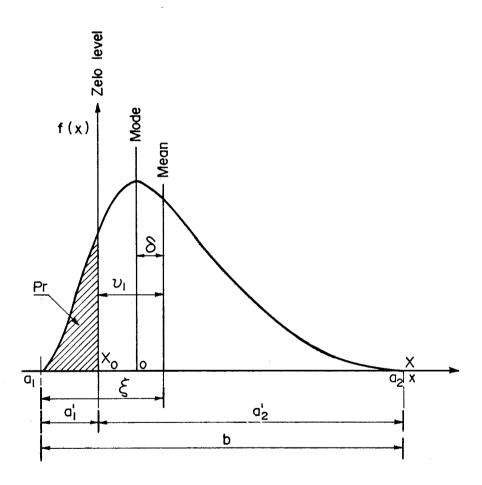


Fig. 2. Graphical representation of symbols used in describing the distribution of \mathbf{M}_{n}^{*} ,

The integral I_{x_0} (p,q) is the incomplete beta function which was tabulated and first edited by Karl Pearson in 1954 (see Ref. 12). With the origin at the inital level, the probability density function of M_n^* is deduced from (30) as

$$f(x) = C(a_1' + x)^{p-1} (a_2' - x)^{q-1}, \ 0 < x \ll a_2'$$
 in which
$$C = \frac{b^{1-p-q}}{B(p,q)}$$
 (37)

$$a'_1 = a_1 + x_0$$

 $a'_2 = a_2 + x_0$

At zero, M_n^* has a probability mass given by (35). Table IV lists the values of the parameters and other related constants describing the distribution of M_n^* .

TABLE IV: CHARACTERISTICS OF THE DISTRIBUTION OF THE ADJUSTED SURPLUS FOR INDEPENDENT NORMAL VARIABLES ($\sigma = 1.0$)

n	10	20	30	40	50
р	3.84795	4.32210	4.09144	3.81221	3.57050
q	16.09100	22.00731	19.52161	16.66923	14.48116
b	12.01064	20.69107	23.53264	24.68230	25.39018
\mathbf{a}_{1}	1.90678	2.82529	3.36601	3.75575	4.06586
-	10.10385	17.86577	20.16661	20.92653	21.32420
$\overset{\mathrm{a}_{2}}{oldsymbol{\delta}}$	0.41111	0.57124	0.71150	0.83836	0.95605
Ċ	0.43544×10^{-16}	0.44539×10^{-28}	0.36867×10^{-26}	0.96788×10^{-23}	0.58324×10^{-20}
X_{o}	-0.98377	-1.64657	-2.13685	-2.54132	-2.89154
Χo	0.07685	0.05697	0.05223	0.04920	0.04625
Pr	0.06430	0.03459	0.02441	0.01893	0.01535
a_1	0.92301	1.17872	1.22916	1.21443	1.17431
a;	11.08762	19.51233	22.30345	23.46783	24.21585

TABLE V: COMPUTED VALUES OF χ^2

n	10	20	30	40	50	
χ2	23.682	16.722	16.117	25.189	31.872	

Finally, a Chi-square test of goodness of fit was used. The computed values of the Chi-square statistic shown in Table V, indicate that fitting the distribution of M_n^* by the Type I curve is acceptable.

Distribution of mn

For normal variables, the random variable $-m_n^*$ has the same distribution as

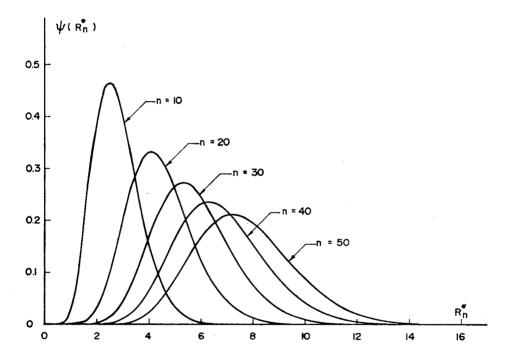


Fig. 3. Approximate probability density functions of R_n^* for n = 10 (10) 50

 M_n^* ; consequently the distribution of m_n can be readily deduced from that of M_n^* . Thus m_n has a probability mass at zero and a probability density function g defined as follows:

Prob
$$(m_n^*) = I_{x_0}(p,q)$$

 $g(m_n^*) = f(-m_n^*), -a_2' \le m_n^* < 0$

 $g(m_n^*) = f(-m_n^*), -a_2' \le m_n^* < 0$ in which I_{x_0} (p,q) and f are given by (35) and (37), respectively.

Distribution of R_n*

Following suggestion of Sutabutr8, the Type III curve was used to fit the distribution of R_n^* . From the definition, $R_n^* \ge 0$, the two-parameter Pearson Type III curve, which is commonly known as the two-parameter gamma distribution, can be used. The probability density function of R_n^* is then written as:

$$\psi(R_n^*) = \beta^{-\alpha}(R_n^*)^{\alpha-1} e^{-R_n^*/\beta} / \Gamma(\alpha), R_n^* \ge 0$$
(38)

where $\Gamma(\alpha)$ denotes the gamma function of α . The two parameters α and β are determined from the expected value and variance of R_n by

$$\alpha = [E(R_n^*)]^2 / Var (R_n^*)$$

$$\beta = Var (R_n^*) / E(R_n^*)$$

TABLE VI: RESULTS FROM FITTING THE DISTRIBUTION OF R.

n	α	β	χ²	
 10	9.68608	0.28802	5.267	
20	12.73857	0.34820	5.785	
30	14.19927	0.40120	17.129	
40	15.09630	0.44775	9.818	
50	05.71841	0.48958	16.351	

NOTES: Significance level = 0.05

Number of degrees of freedom = 29

Critical value = 42.557Sample size used = 1.000

Numerical values of these parameters given in Table VI. Also shown in Table VI are the computed values of the Chi-squre statistic at various n. These values indicate the acceptability of the fitting. The approximate probability density functions of R_n^* for n = 10(10)50 are shown in Fig. 3 for the case $\sigma = 1.0$.

Summary and Conclusions

For a reservoir allowing neither spillage nor emptiness during its design lifetime of n years where the outflows in different years are equal to the sample mean over n years of the annual inflows, the storage capacity was found to be optimally equal to the adjusted range. Under the above conditions, the adjusted partial sum, the adjusted surplus and the adjusted deficit represent the water content of the reservoir, the maximum amount of water in surplus, and the maximum amount of water in deficit, respectively. For the case where the inflows are distributed as independent normal variables, the water contents in different years were found to be distributed as normal variables; the maximum amount of water in surplus or in deficit was found to have a probability mass at zero and a probability density function elsewhere, which was fitted by the Type I curve; and the storage capacity was found to be fitted by the Type III curve.

List of Symbols

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The following symbols have been used in this paper:
            = terminal of the beta distribution;
a_1, a_2
            = terminals of the distribution of the adjusted surplus;
            = distance between two terminals of the beta distribution;
B(p,q)
            = beta function of p and q;
\mathbf{C}
            = constant in the distribution of the adjusted surplus;
Cov (. . .)
            = covariance:
E(.)
            = expected value:
f
            = probability density function of the adjusted surplus:
            = probability density function of the adjusted deficit;
            = incomplete beta function of p and q;
I_{x_0} (p,q)
            = adjusted deficit (maximum amount of water in deficit over n
m_n
                years);
M_n
            = adjusted surplus (maximum amount of water in surplus over
                n years);
            = reservoir lifetime in years;
n
             = parameters of the beta distribution;
p,q
            = probability mass at zero;
Pr
R_n^*
            = adjusted range (reservoir storage capacity);
S_{i}
            = partial sum;
            = adjusted partial sum (water content in the ith year);
\mathbf{U}_{\mathbf{n}}^{r}
            = maximum of adjusted partial sums;
            = positive part of S_i^*;
V_i
Var(.)
            = variance;
            = negative part of S_i^*;
W_i
            = annual inflow in year k;
X_k
X_n
            = sample mean;
```

= net input in year k; Y_{k} = shape parameter of the gamma distribution; a β = scale parameter of the gamma distribution; β_1, β_2 = moment ratios: $\Gamma(.)$ = gamma function: δ = distance from the mode to the mean of the beta distribution; = pq; = Pearson criterion; = mean of normal variables: u = central moment of order r: $\mu_{\rm r}$ ψ = probability density function of R_n; = standard deviation of normal variables: σ T = p+q.

Acknowledgements

This paper is based partly upon the Dissertation by H.N. Phien for the Degree of Doctor of Technical Science at the Asian Institute of Technology which was conducted under the supervision of the other two authors. The comments made by Professor Vujica Yevjevich of Colorado State University as external examiner are greatly appreciated.

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บทกัดย่อ

ได้ศึกษาสภาพการเปลี่ยนแปลงปริมาณน้ำในอ่างเก็บน้ำในปีต่าง ๆ การเปลี่ยนแปลงของจำนวนสูงสุด ของปริมาณน้ำที่เกินหรือต่ำกว่าระดับเริ่มต้น รวมทั้งการเปลี่ยนแปลงของปริมาณความสามารถในการเก็บกัก ภายในช่วงอายุการใช้งานของอ่างเก็บน้ำซึ่งถือว่าไม่มีการแห้งหรือล้นเลย การศึกษานี้มีข้อสมมุติฐานว่าปริมาณ น้ำที่ใหลลงอ่างแต่ละบีมีการเปลี่ยนแปลงในแบบตัวแปรอิสระของการแจกแจงทางสถิติของ Gauss และ ปริมาณน้ำที่ใหลออกจากอ่างมีค่าเท่ากับค่าเฉลี่ยของตัวอย่างข้อมูลของปริมาณน้ำที่ใหลลงอ่าง

ได้พบว่าปริมาณน้ำที่เก็บไว้ในอ่างในบี้ต่าง ๆ เปลี่ยนแปลงตามการแจกแจงทางสถิติของ Gauss ส่วนปริมาณสูงสุดของน้ำที่ขาดและเกินนั้นมีค่า probability mass ที่ศูนย์ ส่วนที่อื่นมีค่าเบ็น probability แบบต่อเนื่องซึ่งสามารถอธิบายโดยใช้เส้นโค้ง Pearson แบบที่ 1 สำหรับปริมาณความสามารถในการเก็บ กักสามารถอธิบายโดยใช้เส้นโค้ง Pearson แบบที่ 3 การใช้เส้นโค้ง Pearson อธิบายนี้เราได้ประเมิน ความเหมาะสมโดยอาศัยการทดสอบทางสถิติซึ่งเรียกว่า Chi-square