

## ***International Journal of Scientific Research and Reviews***

### **Evaluation of Rain Gauge Network using Maximum Information Minimum Redundancy Theory**

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#### **ABSTRACT**

Establishment and maintenance of a Rain Gauge Network (RGN) in any geographical region is required for planning, design and management of water resources projects. Setting up and maintaining a RGN is an evolutionary process, wherein a network is established early in the development of the geographical area; and the network is reviewed and upgraded periodically to arrive at the optimum network. This paper illustrates the use of entropy theory-based criterion, namely, Maximum Information Minimum Redundancy (MIMR) for evaluation of RGN of upper Bhima basin. The transinformation index obtained from marginal and conditional entropy values, adopting 2-parameter Normal/ Log-Normal (N2 and LN2) distributions are used for evaluation of RGN. The paper presents that the derived network consisting of seventeen raingauge stations with network density of 865 km<sup>2</sup> per gauging station is considered as the optimum RGN for upper Bhima basin, which satisfied the WMO recommended value of 600-900 km<sup>2</sup> per station.

**KEYWORDS:** Entropy, Network density, Raingauge, Transinformation index, Optimization

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## INTROUDCTION

Monitoring networks established in different river basins collect data on hydrometeorological parameters such as rainfall, streamflow and stage, water quality, sediment, etc required for water resources planning, design and management of water resources projects. Efficient information gathering system should assess the variability in hydrologic processes with spatial and temporal scales and reduce the costs of data collection<sup>1</sup>. Network optimization involves determining optimal monitoring stations and their siting at appropriate locations. This study exemplifies the use of MIMR theory for evaluation of RGN of upper Bhima basin.

Entropy method quantifies the relative information content for RGN, and has the advantage that the method needs only rainfall data for evaluation. The method facilitates network design by quantifying the marginal contribution of each data collection node to the overall information provided by the network using an index termed as marginal entropy<sup>2</sup>. The MIMR criterion maximizes the joint entropy of stations within the optimal set, and the transinformation between stations within and outside of the optimal set. Meanwhile, it insures that the optimal set contains minimum duplicated information<sup>3</sup>. In recent past, entropy method is widely applied by various researchers for evaluation of different hydrometeorological networks such as streamflow, rainfall, groundwater, water quality monitoring etc., to derive optimum networks.

Yang and Burn<sup>4</sup> applied entropy based approach to design stream gauge network for Pembina river basin in southern Manitoba, Canada. They have adopted a directional informational transfer index to identify the most priority station amongst seven stations in the network. Ozkul et al<sup>5</sup> applied the entropy technique for the assessment of water quality monitoring networks for Mississippi River in Louisiana. They concluded that the concept of entropy is a promising method as it quantitatively measures the information produced by network, which will improve the efficiency and cost-effectiveness of current monitoring programs. Markus et al<sup>6</sup> applied entropy and least square methods to evaluate the stream gauging network in the State of Illinois through an assessment of the transferring of information (transinformation) among gauging records for low, average, and high flow conditions. They expressed that the hybrid combination of the entropy and GLS measures of regional value of stations would certainly provide more insight in assessment of stream gauging program and the station rankings based on the combined method can preserve correlation with rankings using both entropy and GLS.

Sarlak and Sorman<sup>7</sup> applied entropy theory to evaluate the streamflow network for Kizihrmak basin. They compared the model uncertainty in streamflow network with

transinformation index using normal, log-normal and gamma distributions. They also suggested that the ranking of log-normal distribution could be considered for selection of stream gauge network for Kizihrmak basin. Yoo et al<sup>8</sup> compared the applications of mixed and continuous distributions in entropy theory to study about the rainfall variability and rainfall intermittency using transinformation index. Their study showed that the mixed distribution is suitable for evaluation of rain gauge network for Choongju basin, Korea. Jordan and Jason<sup>9</sup> analyzed the application of a novel, process-oriented stream gauging method for generating rating curves to establish a stream gauging network in the Whitewater River basin in south central Kansas. They described that the model was developed for the type of channels typically found in this watershed and designed to handle deep, narrow geomorphically stable channels with irregular platforms, and can model overbank flow over a vegetated floodplain. Ridolfi et al<sup>10</sup> applied entropy theory to the high density raingauge network of the urban area of Rome for evaluation of maximum information content achievable by a rainfall network for different sampling time intervals. In this paper, N2 and LN2 distributions are used for computation of entropy values for the raingauge stations under consideration. The methodology adopted in computation of entropy measures given by N2 and LN2 for evaluation of RGN of upper Bhima basin is briefly described in the ensuing sections.

## **METHODOLOGY**

The main objective of evaluating a data collection network is to identify the stations that are producing redundant or repeated information, which can be measured quantitatively by computing transinformation or the redundancy produced by each station in the network.

### ***Concept of Entropy***

A quantitative measure of the uncertainty associated with a probability distribution or the information content of the distributions termed Shannon entropy<sup>11</sup> can be mathematically expressed as:

$$H(X) = -k \sum p_i \ln(p_i) \quad \dots (1)$$

where  $H(X)$  is the entropy corresponding to the random variable  $X$ ;  $k$  is a constant that has value equal to one, when natural logarithm is taken; and  $p_i$  represents the probability of  $i^{\text{th}}$  event of random variable  $X$ .

### ***Marginal Entropy***

Marginal entropy for the discrete random variable  $X$  is defined as:

$$H(X) = - \sum_{i=1}^N p(X_i) \ln(p(X_i)) \quad \dots (2)$$

where  $p(X_i)$  is the probability of occurrence of  $X_i$ , computed by N2 and LN2 distributions, and N is the number of observations<sup>12</sup>. The marginal entropy  $H(X)$  indicates the amount of information or uncertainty that X has. If the variables X and Y are considered as independent, then the joint entropy  $[H(X, Y)]$  is equal to the sum of their marginal entropies defined by:

$$H(X, Y) = H(X) + H(Y) \quad \dots (3)$$

If the variables are stochastically dependent, then the joint entropy is less than its total entropy.

### ***Conditional Entropy***

Conditional entropy measures the entropy of a random variable Y, if one has already learned completely about the random variable X. For given X, the conditional entropy of Y is defined as:

$$H(Y|X) = H(X, Y) - H(X) \quad \dots (4)$$

Conditional entropy value becomes zero, if the value of one variable is completely determined by the value of other variable. If the variables are independent then  $H(Y|X) = H(Y)$ .

### ***Transinformation Index***

Transinformation index measures the redundant or mutual information between variables, and is computed from marginal and conditional entropy indices to derive an optimum network. Transinformation represents the amount of information, which is common to two stochastically dependent variables X and Y. The transinformation between X and Y is defined as:

$$T(X, Y) = H(X) + H(Y) - H(X, Y) = H(Y) - H(Y|X) \quad \dots (5)$$

For independent X and Y,  $T(X, Y) = 0$ . The value of  $T(X, Y)$  is also known as transinformation index<sup>13</sup>.

### ***Steps Involved in Computation of Entropy Measures***

In practice, the existing sampling sites of a RGN can be arranged in the order of information content. In the ordered list thus obtained, the first station is the one, where the highest uncertainty about the variable occurs, and the subsequent stations serve to reduce the uncertainty further. The steps involved in selecting the best combination of stations using entropy measures are as follows:

- i) Let the data collection network under review, consists of  $M$  monitoring stations. The data series of the variable of interest at each station  $(X_1, X_2, \dots, X_M)$  is represented by  $X_{ij}$ , where 'i' denotes the station identification number ( $i=1, 2, \dots, M$ ) and 'j' is for time period ( $j=1, 2, \dots, N$ ). The data length at all stations is assumed to be equal to  $N$ . The best fitted multivariate joint probability density function for the subset  $(X_1, X_2, \dots, X_M)$  of  $M$  monitoring stations is selected.
- ii) The marginal entropy of the variable  $H(X_i)$  ( $i=1, 2, \dots, M$ ) for each station is calculated. The station with the highest marginal entropy is denoted as the first priority station  $Pr(X_{Z_1})$ . This is the location, where the highest uncertainty occurs about the variable, and hence information-gain will be highest from the observations recorded at this site.
- iii) This station  $Pr(X_{Z_1})$  is coupled with every other  $(M-1)$  stations in the network to compute transinformation  $T(X_i, Pr(X_{Z_1}))$  with  $X_i \neq Pr(X_{Z_1})$ ,  $i=1, 2, \dots, M$ ; and to select that pair, which gives the least transinformation. The station that fulfils this condition is marked as the second priority location  $Pr(X_{Z_2})$ .
- iv) The pair  $(Pr(X_{Z_1}), Pr(X_{Z_2}))$  is coupled with every other  $(M-2)$  station in the network to select a triplet with the least transinformation  $T(X_i; Pr(X_{Z_1}), Pr(X_{Z_2}))$ . The same procedure is continued by successively considering combinations of three and more stations, and selecting the combination that produces the least transinformation. Finally, all  $M$  monitoring stations  $(X_1, X_2, \dots, X_M)$  can be ranked in priority order to get  $(Pr(X_{Z_1}), Pr(X_{Z_2}), \dots, Pr(X_{Z_M}))$ .
- v) It is possible to terminate the above process early, before carrying out for all  $M$  stations by selecting a particular threshold transinformation value as the amount of redundant information to be permitted in the network, such that sampling of the variable may be stopped at the stations that exceed the threshold to get optimum number of stations, which is less than  $M$ .

In the above procedure, the benefits for each combination of sampling sites are measured in terms of least transinformation or the highest conditional entropy produced by that combination. The above procedure helps to assess network configurations with respect to the existing stations. If new stations are to be added to the system, their locations may be selected again on the basis of the entropy measures by ensuring maximum gain of information with minimum redundancy. The

correlation coefficient of each monitoring station can be computed by using Eq. (6) based on transinformation index.

$$\text{Transinformation (T)} = -\frac{1}{2} \ln(1 - R^2) \quad \dots (6)$$

Where R represents the multiple correlation coefficient of  $X_i$  on  $\text{Pr}(X_{Z_1}), \text{Pr}(X_{Z_2}), \dots, \text{Pr}(X_{Z_M})$ .

## APPLICATION

The methodology detailed above has been applied to evaluate the RGN for upper Bhima basin upto Ujjani reservoir. The basin is located in the western part of Maharashtra between 18° 03' N to 19° 24' N latitude and 70° 20' E to 75° 18' E longitude. The geographical area of the basin is 14,712 km<sup>2</sup>. Of the total geographical area under study, 25 % is hilly and/or highly dissected, 55 % plateau and 20 % is remaining plain area<sup>14</sup>. From scrutiny of historical rainfall data, it is noted that twenty-five raingauge stations have concurrent data for the period 1971-2007; and considered in deriving the optimum RGN for the basin under study.

## RESULTS AND DISCUSSIONS

By applying the procedure detailed above, a computer program was developed and used to compute the transinformation index from marginal and conditional entropy values. The program identifies the first priority station based on marginal entropy given by N2 and LN2 distributions; computes the conditional entropy with reference to first priority station; and arranges the raingauge stations in order of priority based on transinformation index. Table 1 gives the model parameters and marginal entropy values given by N2 and LN2 distributions for twenty-five raingauge stations considered in the study.

From Table 1, it may be noted that the Khandala station is the first priority station based on marginal entropy value given by N2 distribution. Also, from Table 1, it may be noted that the Amboli station is the first priority station based on marginal entropy value, when LN2 is used. The priority station is coupled with other twenty-four raingauge stations individually to identify the next priority station in order and to compute the transinformation index. Tables 2 and 3 give the details on redundant information passed by each station based on transinformation index and its corresponding correlation coefficient values given by N2 and LN2 respectively.

Table 1: Parameters and marginal entropy values of raingauge stations given by N2 and LN2 distributions

S. No.	Raingauge station	Normal (N2)			Lognormal (LN2)		
		Parameters (m <sup>3</sup> /s)		Marginal entropy	Parameters (m <sup>3</sup> /s)		Marginal entropy
		Mean	Std. Dev.		Mean	Std. Dev.	
1	Askheda	855.8	265.6	7.001	6.707	0.303	0.225
2	Aundhe	1,808.4	652.1	7.899	7.443	0.341	0.343
3	Chaskman	722.3	202.9	6.732	6.543	0.288	0.174
4	Holkarpul	729.5	206.2	6.748	6.556	0.271	0.114
5	Katraj-tunnel	928.2	294.9	7.105	6.785	0.321	0.282
6	Khandala	4,312.9	969.7	<b>8.296</b>	8.344	0.233	0.038
7	Kiwale	950.5	277.4	7.044	6.813	0.309	0.245
8	Kolgaon	448.4	153.7	6.454	6.042	0.373	0.432
9	Koliye	1,971.5	868.0	8.185	7.512	0.374	0.436
10	Kurwandi	862.8	265.1	6.999	6.718	0.289	0.179
11	Lonikand	537.5	156.2	6.470	6.227	0.394	0.487
12	Pimpalgaon-Joga	1,076.8	277.4	7.044	6.951	0.250	0.031
13	Pimpalwandi	579.5	196.2	6.698	6.301	0.370	0.425
14	Ranjangaon	522.2	199.3	6.714	6.177	0.434	0.585
15	Shirur	523.1	186.6	6.648	6.187	0.416	0.543
16	Shive	1,089.3	482.5	7.598	6.869	0.563	0.845
17	Supa	437.3	183.0	6.629	5.968	0.534	0.792
18	Tembhurni	571.4	181.4	6.620	6.290	0.367	0.416
19	Whiram	2,181.7	651.4	7.898	7.645	0.300	0.215
20	Amboli	1,862.9	783.0	8.082	7.099	2.396	<b>2.293</b>
21	Chandoh	420.6	199.1	6.713	5.600	2.182	2.200
22	Kadus	715.2	250.1	6.941	6.510	0.368	0.419
23	Kathapur	487.0	197.6	6.705	6.098	0.453	0.628
24	Pabal	588.5	191.5	6.674	6.318	0.366	0.413
25	Sarola-Kasar	481.1	257.8	6.971	5.731	2.195	2.205

Table 2: Details of redundant information passed by each station together with correlation coefficient based on N2 distribution

Priority Rank	Raingauge station (based on priority rank)	Transinformation index (T)	Optimum redundant information passed $[(T/1.855)*100]$	Correlation coefficient $[1-Exp(-2T)]^{0.5}$
1	Khandala	...	...	...
2	Kathapur	0.002	0.1	0.063
3	Sarola-Kasar	0.064	3.5	0.347
4	Shive	0.121	6.5	0.464
5	Kolgaon	0.163	8.8	0.527
6	Koliye	0.265	14.3	0.641
7	Shirur	0.278	15.0	0.653
8	Pimpalgaon-Joga	0.283	15.3	0.657
9	Tembhurni	0.323	17.4	0.690
10	Amboli	0.416	22.4	0.752
11	Lonikand	0.455	24.5	0.773
12	Supa	0.552	29.8	0.818
13	Chandoh	0.605	32.6	0.838
14	Aundhe	0.660	35.6	0.856
15	Kiwale	0.693	37.4	0.866
16	Kurwandi	0.737	39.7	0.878
17	Katraj-tunnel	0.889	47.9	0.912
18	Pimpalwandi	0.964	52.0	0.924
19	Pabal	1.014	54.7	0.932
20	Kadus	1.150	62.0	0.949
21	Chaskman	1.210	65.2	0.955
22	Holkarpul	1.376	74.2	0.968
23	Ranjangaon	1.487	80.2	0.974
24	Askheda	1.581	85.2	0.979
25	Whiram	1.855	100.0	0.988



**Table 3: Details of redundant information passed by each station together with correlation coefficient based on LN2 distribution**

Priority Rank	Raingauge station (based on priority rank)	Transinformation index (T)	Optimum redundant information passed $[(T/2.241)*100]$	Correlation coefficient $[1-Exp(-2T)]^{0.5}$
1	Amboli	...	...	...
2	Shirur	0.000	0.000	0.003
3	Aundhe	0.001	0.027	0.035
4	Sarola-Kasar	0.003	0.143	0.080
5	Kolgaon	0.068	3.048	0.357
6	Koliye	0.114	5.078	0.451
7	Pimpalgaon-Joga	0.178	7.961	0.548
8	Shive	0.231	10.326	0.609
9	Tembhurni	0.273	12.169	0.648
10	Lonikand	0.304	13.543	0.675
11	Kathapur	0.485	21.647	0.788
12	Holkarpul	0.534	23.838	0.810
13	Supa	0.578	25.783	0.828
14	Chandoh	0.658	29.362	0.855
15	Pimpalwandi	0.759	33.882	0.884
16	Khandala	0.816	36.426	0.897
17	Kiwale	0.894	39.871	0.912
18	Pabal	0.911	40.634	0.916
19	Whiram	1.043	46.537	0.936
20	Askheda	1.189	53.034	0.952
21	Katraj-tunnel	1.217	54.297	0.955
22	Kurwandi	1.413	63.066	0.970
23	Ranjangaon	1.457	64.998	0.972
24	Chaskman	1.570	70.054	0.978
25	Chandoh	2.241	99.991	0.994

From Tables 2 and 3, it may be noted that Whiram and Chandoh stations provide about 100% redundant information based on transinformation index given by N2 and LN2 respectively; and considered as the reference station for computation of redundant information for other stations. Also, from Table 2, it may be noted that Kadus, Chaskman, Holkarpul, Ranjangaon and Askheda raingauge stations are providing more than 60% redundant information in addition to Whiram. From Table 3, it may be noted that Kurwandi, Ranjangaon and Chaskman raingauge stations are providing more than 60% redundant information in addition to Chandoh.

Based on MIMR criterion, it is identified that the raingauge stations such as Askheda, Chandoh, Chaskman, Kadus, Kurwandi, Holkarpul, Ranjangaon and Whiram are providing more than 60% redundancy, indicating possible discontinuation from the existing RGN in the basin. The study showed that the derived network consists of seventeen raingauge stations with 865 km<sup>2</sup> per gauging station are considered as the optimum RGN for the basin, which satisfies the WMO<sup>15</sup> recommended value of 600-900 km<sup>2</sup> per station for minimum density of raingauge network. The results are expected to be of assistance to stakeholders for decision making as regards RGN optimisation in the Upper Bhima basin.

## **CONCLUSIONS**

This paper presented an of entropy theory-based criterion of MIMR for evaluation of RGN of upper Bhima basin consisting of twenty-five raingauge stations. The study described that N2 and LN2 distributions are used to represent the data and then to compute the marginal and conditional entropy indices and the transinformation index. The results showed that Khandala and Amboli stations are the first priority stations with highest marginal entropy for upper Bhima basin, when N2 and LN2 are used. The results also showed that Whiram and Chandoh stations provided about 100% redundant information and hence considered as the reference station for computation of redundant information passed by other stations. From the results of transinformation index given by N2 and LN2 distributions, it may be noted that the amount of redundant information given by Askheda, Chandoh, Chaskman, Kadus, Kurwandi, Holkarpul, Ranjangaon and Whiram raingauge stations are more than 60%, proposed for possible discontinuation of the stations from the core network while optimizing RGN of upper Bhima basin upto Ujjani reservoir. Thus, the derived optimum network for Upper Bhima basin consists of seventeen stations with network density of 865 km<sup>2</sup> per gauging station satisfying the WMO recommended value of 600-900 km<sup>2</sup> per station. The results are expected to be of assistance to stakeholders for decision making as regards network optimisation in the basin.

## **ACKNOWLEDGEMENTS**

The authors are grateful to Dr. I.D. Gupta, Director, Central Water and Power Research Station, Pune, for his continuous encouragement and giving permission to publish this paper. Kind co-operation extended by Shri S. Govindan, Additional Director, is gratefully acknowledged. The authors are thankful to the Executive Engineer, Hydrometeorological Data Processing Division, Nashik, for making available the rainfall data of Upper Bhima basin.

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