## New Dam Break Risk Assessment Method in Fuzzy Framework

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HS7.5 - Hydro-meteorological extremes and hazards: vulnerability, risk, impacts and mitigation

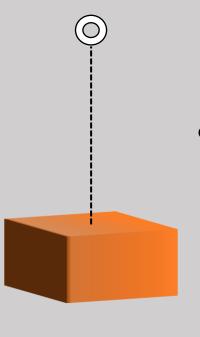
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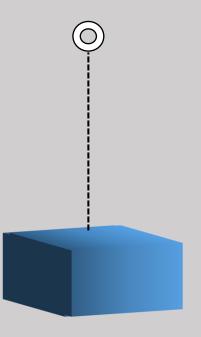
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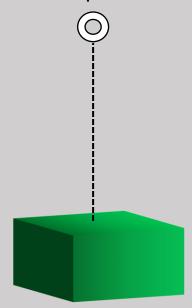
Dams are useful for mitigation of floods, and at the same time there is risk of dam failure due to floods, seismic hazards and factors such as ageing of dam material.



There is need to perform risk assessment of dams as their failure can have catastrophic consequences on establishments located in their downstream locations.

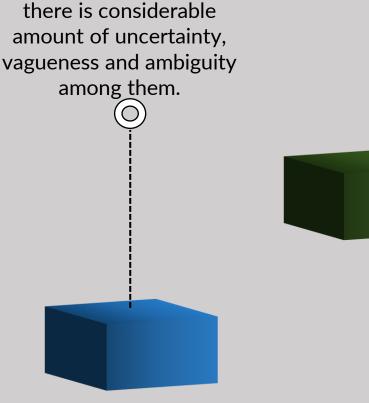


Effective risk analysis requires accounting for both failure probability of dam and dam break consequences.



INTRODUCTION

Fuzzy hierarchical model for risk assessment based on combination of static and There are numerous variable fuzzy set theory is factors which effect the proposed in the study. consequences, and due to



lack of data and knowledge



#### Risk = Likelihood x Consequences

The model synthetically evaluates the dam break consequences based on exposure and Hazard Severity. Hence, the dam break risk can be assessed based on three evaluation indices i.e., *likelihood of dam break flood*, *hazard severity*  $(S_D)$  *and degree of exposure*.

$$\mathbf{R}_{i} = f(\ell, \mathbf{S}_{D}, \mathbf{e}_{i})$$

Where.

 $\ell$  is the likelihood of occurrence of dam break flood,

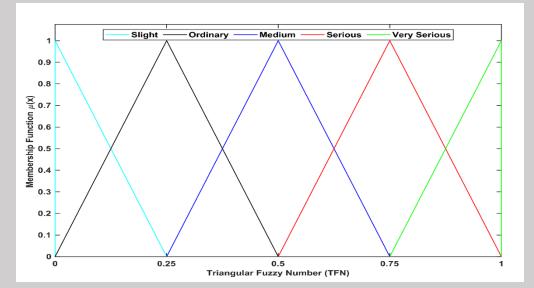
 $S_D$  is the Hazard severity

 $e_i$  is the degree of exposure for  $i^{th}$  influencing factor of dam break consequences.

#### Risk Indexes classification in fuzzy framework

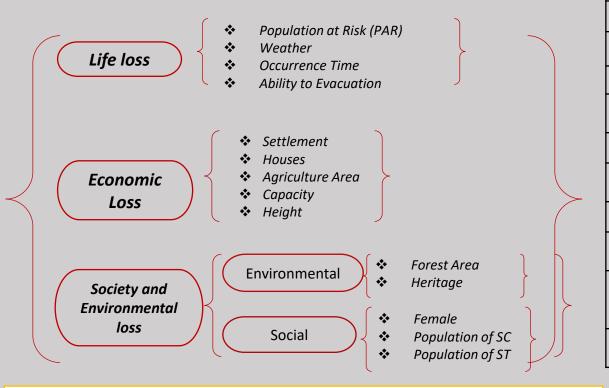
| <b>Evaluation Level</b> | Likelihood                         | Hazard Severity (S <sub>D</sub> ) | TFN                  |
|-------------------------|------------------------------------|-----------------------------------|----------------------|
| <b>Extremely Low</b>    | $>\frac{1}{10}$                    | < 0.3                             | < 0, 0, 0.25 >       |
| Low                     | $\frac{1}{10} - \frac{1}{100}$     | 0.3 -0.6                          | < 0, 0.25, 0.50 >    |
| Moderate                | $\frac{1}{100} - \frac{1}{1000}$   | 0.6 – 1.0                         | < 0.25, 0.50, 0.75 > |
| Severe                  | $\frac{1}{1000} - \frac{1}{10000}$ | 1.0 – 4.0                         | < 0.50, 0.75, 1 >    |
| <b>Extremely Severe</b> | $<\frac{1}{10000}$                 | > 4.0                             | < 0.75, 1, 1 >       |

Linguistic classification of grades of likelihood of flood and hazard severity and their corresponding TFNs





- ✓ To establish a hierarchical structure of influencing factors of dam break consequences, the exposure evaluation index is classified into 3 risk categories i.e., *life loss, economic loss and social and environmental influence*, which are main contents of dam break consequence assessment.
- ✓ Furthermore, each risk category is sub-divided into influencing factors of consequences called exposure indices or risk items



#### Dam break consequences

| Quantitative Data            |                    |                         |                      |                         |                        |
|------------------------------|--------------------|-------------------------|----------------------|-------------------------|------------------------|
| <b>Evaluation Level</b>      | Slight             | Ordinary                | Medium               | Serious                 | Extremely Serious      |
| Population at Risk (PAR)     | < 500              | 501 - 1000              | 1001 – 2000          | 2001 - 3000             | >3000                  |
| Households                   | < 100              | 101 - 200               | 201 - 300            | 301 - 450               | > 450                  |
| <b>Female Population</b>     | < 250              | 251 - 500               | 501 - 750            | 751 - 1000              | > 1000                 |
| Schedule Caste<br>Population | < 50               | 51 - 100                | 101 -200             | 201 - 300               | >300                   |
| Schedule Tribe<br>Population | < 50               | 51 - 100                | 101 -200             | 201 - 300               | >300                   |
| Capacity (Mm <sup>3</sup> )  | < 1                | 1 - 10                  | 10 - 100             | 100 - 1000              | > 1000                 |
| Height (m)                   | < 10               | 10 - 20                 | 20 - 30              | 30 - 60                 | > 60                   |
| Forests Area (Ha)            | < 50               | 50 - 100                | 100 - 150            | 150 - 200               | > 200                  |
| Agricultural Area<br>(Ha)    | < 100              | 100 -200                | 200 - 350            | 350 - 500               | > 500                  |
| Qualitative Data             |                    |                         |                      |                         |                        |
| Settlement                   | Few<br>Households  | Village                 | Town                 | City                    | Capital City           |
| Heritage                     | General            | Municipality            | State                | National                | World                  |
| Weather                      | Sunny Day          | Sprinkle                | Moderate<br>Rain     | Rainstorm               | Heavy<br>Storm         |
| Occurrence Time              | Daytime            | Working Day in<br>Night | Holidays at<br>Night | Working Days in Morning | Holidays in<br>Morning |
| Ability to<br>Evacuation     | Extremely<br>Sound | Sound                   | General              | Unsound                 | Terrible<br>Unsound    |

□ Construction of Fuzzy Pairwise Comparison matrix using Saaty scale of comparison: Let W be the pairwise comparison matrix then,

$$W = \begin{bmatrix} w_{ij} \end{bmatrix}_{n \times n} = \begin{bmatrix} \frac{w_i}{w_j} \end{bmatrix}_{n \times n} = \begin{bmatrix} w_{11} & w_{12} & w_{13} & \cdots & w_{1n} \\ w_{21} & w_{22} & w_{23} & \cdots & w_{2n} \\ w_{31} & w_{32} & w_{33} & \cdots & w_{3n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ w_{n1} & w_{n2} & w_{n3} & \cdots & w_{nn} \end{bmatrix}$$

■ Estimation of local weights of criterion using Eigen vector method: if W is known and w is unknown then w can be recovered using eigen vector method  $\begin{bmatrix} w_1 & w_2 & w_3 \end{bmatrix}$ 

**Consistency Check**: After obtaining  $\lambda_{\text{max}}$ , first Consistency Index (CI) and then consistency Ratio (CR) is determined as follows. If C.R. < 0.1 the inconsistencies are tolerable and reliable results are expected from F-AHP.

$$C.I. = \frac{\lambda_{max} - n}{n - 1}$$
  $C.R. = \frac{C.I.}{R.I._n}$   $n = \text{number of criteria}$ 

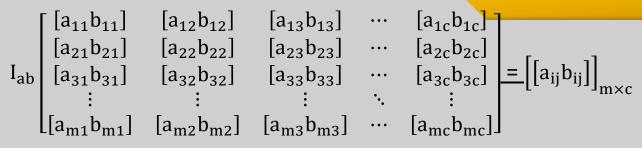
**Defuzzification:** Since the calculation so far involves fuzzy variables, the next step is to defuzzify the fuzzy weights to form meaningful local weights for analysis  $M_i = \frac{[Lw_i + Mw_i + Uw_i]}{2}$ 

## Fuzzy AHP for weight calculation

| Risk Category                         | Intermediate<br>Layer | Sub-Criterion /<br>Index Layer     | Weight                      |
|---------------------------------------|-----------------------|------------------------------------|-----------------------------|
| (Wi)                                  |                       | (W <sub>ij</sub> )                 | $(W_n = W_i \times W_{ij})$ |
| Life<br>0.7338                        |                       | PAR<br>0.5147                      | 0.42                        |
|                                       |                       | Weather<br>0.1495)                 | 0.1097                      |
|                                       |                       | Occurrence time 0.0532             | 0.0390)                     |
|                                       |                       | Ability to<br>evacuation<br>0.2826 | 0.2074                      |
| Economy<br>0.0995                     |                       | Settlement<br>0.3812               | 0.0379                      |
|                                       |                       | Houses<br>0.1229                   | 0.0122                      |
|                                       |                       | Agriculture area 0.0508            | 0.0050                      |
|                                       |                       | Capacity<br>0.2225                 | 0.0221                      |
|                                       |                       | Height<br>0.2225                   | 0.0221                      |
| Environmental<br>and Social<br>0.1667 | Environmental         | Forest area<br>0.7424              | 0.0618                      |
|                                       | (0.5)                 | Heritage<br>0.2576                 | 0.0214                      |
|                                       | Social (0.5)          | Female<br>0.15                     | 0.0125                      |
|                                       |                       | Population of SC 0.42              | 0.0350                      |
|                                       |                       | Population of ST 0.42)             | 0.0350                      |

#### Step 1: Interval Matrix

## Calculation of Degree of Exposure using VFS Theory



#### **Step 2: Bound Matrix**

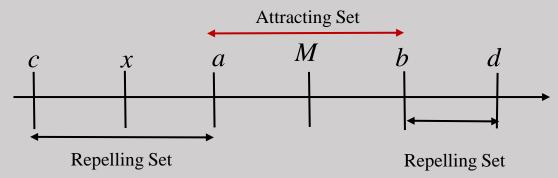
Next step is to construct the indicator variable interval matrix I<sub>cd</sub> also called extended interval matrix.

$$\begin{bmatrix} c_{ij}d_{ij} \end{bmatrix} \begin{cases} a_{ij-1}b_{ij+1,} & j-1>0, j When  $x$  is located to the right of  $M$ , then relative membership function can be membership function can be expressed as: 
$$I_{cd} = \begin{bmatrix} [c_{11}d_{11}] & [c_{12}d_{12}] & [c_{13}d_{13}] & \cdots & [c_{1c}d_{1c}] \\ [c_{21}d_{21}] & [c_{22}d_{22}] & [c_{23}d_{23}] & \cdots & [c_{2c}d_{2c}] \\ [c_{31}d_{31}] & [c_{32}d_{32}] & [c_{33}d_{33}] & \cdots & [c_{3c}d_{3c}] \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ [c_{m1}d_{m1}] & [c_{m2}d_{m2}] & [c_{m3}d_{m3}] & \cdots & [c_{mc}d_{mc}] \end{bmatrix} = \begin{bmatrix} [c_{ij}d_{ij}] \end{bmatrix}_{m \times c}$$
 
$$x \in [M,b]$$
 
$$x \in [a,M]$$$$

## Step 3: Point value matrix

$$M_{ij} = \begin{cases} \frac{a_{ij} + b_{ij}}{2} & j \neq 1, c \\ a_{ij} & j = 1 \\ b_{ij} & j = c \end{cases}$$

## Step 4: Relative Membership Degree (RMD)



When x is located to the right When x is located to the left then relative of M,

then relative membership function can be membership function can be

$$\mu_A(u) = 0.5 \times \left\{ 1 + \frac{(x-b)}{(M-b)} \right\}^{\beta},$$

$$x \in [M, b]$$

$$\mu_A(u) = 0.5 \times \left\{1 - \frac{(x-b)}{(d-b)}\right\}^{\beta},$$

$$x \in [b, d]$$

$$\mu_A(u) = 0.5 \times \left\{ 1 + \frac{(x-a)}{(M-a)} \right\}^{\beta}$$

$$x \in [a, M]$$

$$\mu_A(u) = 0.5 \times \left\{ 1 - \frac{(x-b)}{(d-b)} \right\}^{\beta}, \quad \mu_A(u) = 0.5 \times \left\{ 1 - \frac{(x-a)}{(c-a)} \right\}^{\beta}, \\
x \in [b, d] \quad x \in [c, a]$$

Where,  $\beta$  is index bigger than 0, usually we take it as 1, above become linear functions.

#### **Step 5: Risk Assessment Matrix**

The RMD matrix of the m indexes to the c levels can be calculated illustrated as follows:

$$\mu_{A}(u)_{m \times c} = \begin{bmatrix} \mu_{A}(u)_{11} & \mu_{A}(u)_{12} & \mu_{A}(u)_{13} & \cdots & \mu_{A}(u)_{1c} \\ \mu_{A}(u)_{21} & \mu_{A}(u)_{22} & \mu_{A}(u)_{23} & \cdots & \mu_{A}(u)_{2c} \\ \mu_{A}(u)_{31} & \mu_{A}(u)_{32} & \mu_{A}(u)_{33} & \cdots & \mu_{A}(u)_{3c} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \mu_{A}(u)_{m1} & \mu_{A}(u)_{m2} & \mu_{A}(u)_{m3} & \cdots & \mu_{A}(u)_{mc} \end{bmatrix}$$

# Step 6: Synthetic Relative Membership Degree (SRMD)

$$\mu'_{h} = \left\{ 1 + \left[ \frac{\sum_{i=1}^{m} \left[ \omega_{i} \left( 1 - \mu_{A}(x_{i})_{j} \right) \right]^{p}}{\sum_{i=1}^{m} \left[ \omega_{i} \left( \mu_{A}(x_{i})_{j} \right) \right]^{p}} \right]^{\frac{\alpha}{p}} \right\}^{-1}$$

SRMD can be calculated using variable fuzzy synthetic evaluation model.

#### **Step 7: Normalized SRMD**

$$\beta_j = \frac{\beta_j'}{\sum_{j=1}^c \beta_j'}$$
,  $(j = 1,2,3 \dots c)$ 

# Step 8: Level feature value (H) using rank feature values (RFV) equation (Chen, 1998)

$$H = (1,2,3\cdots c) \blacksquare \beta \qquad H = \sum_{j=1}^{c} \beta_j \times j \quad , (j = 1,2,3\cdots c)$$

## Step 9: Calculation of exposure degree by discrimination rule

$$1 < H < 1.5$$
 ,  $\in$  grade 1   
  $j - 0.5 < H < j + 0.5$  .  $\in$  grade  $j, j = 1, 2, 3 \cdots c$    
  $c - 0.5 < H < c$  , grade  $c$ 

#### Step 10: Calculation of Risk Rate (Type 1)

$$\begin{array}{c} \text{Likelihood} \rightarrow A_i \rightarrow < a_{i1} \text{ , } a_{i2}, a_{i3} > \\ \text{Hazard severity} \rightarrow B_i \rightarrow < b_{i1} \text{ , } b_{i2}, b_{i3} > \\ \text{Exposure} \rightarrow C_i \rightarrow < c_{i1} \text{ , } c_{i2}, c_{i3} > \\ \text{Risk rate} \rightarrow \text{g} = A \otimes B \otimes C \\ \text{g} \rightarrow < a_1b_1c_1 \text{ , } a_2b_2c_2, a_3b_3c_3 > \rightarrow < l_i \text{ , } m_i, u_i > \\ g_i = \frac{[Lw_i + Mw_i + Uw_i]}{3} \end{array}$$



## Step 10: Calculation of Risk Rate (Type 2)

#### Calculation of Risk Index using Static Fuzzy Set Theory

Risk rate 
$$\rightarrow g_{improved} = \sqrt[3]{A \otimes B \otimes C}$$

#### Step 11: Fuzzy Risk assessment matrix

Let  $V = \{V_1, V_2, V_3, V_4, V_5\}$  be the set of rating for each risk item (exposure index). By fuzzy relation on  $C \times V$ ; the fuzzy assessment matrix (M) for  $C \times V$  is established

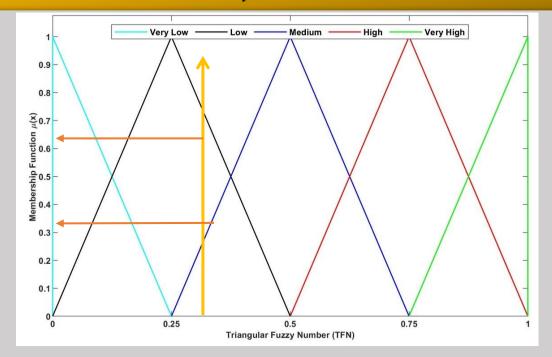
$$M = \begin{pmatrix} V_{1} & V_{2} & V_{3} & V_{4} & V_{5} \\ C_{1} & V(A, B_{\nu} C_{1}, 1) & V(A, B_{\nu} C_{1}, 2) & V(A, B_{\nu} C_{1}, 3) & V(A, B_{\nu} C_{1}, 4) & V(A, B_{\nu} C_{1}, 5) \\ \vdots & Step 13: Calculation of Risk Index usin \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & (Chen, 1998) \\ C_{N} & V(A, B_{N} C_{N}, 1) & V(A, B_{N} C_{N}, 2) & V(A, B_{N}$$

## Step 12: Evaluation of Aggregative Risk:

$$[R]_{1\times 5} = [W]_{1\times N} \times [M]_{N\times 5}$$

|W| = Weight vector of Sub criteria estimated using AHP and FAHP.

[M] = Fuzzy Risk assessment matrix



# Step 13: Calculation of Risk Index using RFV equation

$$RI = [1 \ 2 \ 3 \ 4 \ 5] \times [R]_{1 \times 5}$$

| L                       | 3 2 31/13                        |
|-------------------------|----------------------------------|
| <b>Evaluation Level</b> | Risk Index                       |
| Very Low                | 1.0 <ri< 1.5<="" th=""></ri<>    |
| Low                     | 1.5 < RI < 2.5                   |
| Moderate                | 2.5 < RI < 3.5                   |
| High                    | 3.5 <ri<4.5< th=""></ri<4.5<>    |
| Very High               | 4.5 <ri 5.0<="" <="" th=""></ri> |
|                         |                                  |

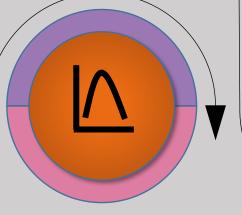


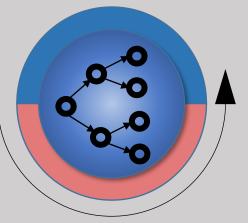
#### Methodology

PMP Estimation:

Design Hyetograph are constructed by time disaggregating 1-Day and 2-Day PMP.

Event Tree Analysis
TA was performed to
estimate likelihood of
dam break flood.

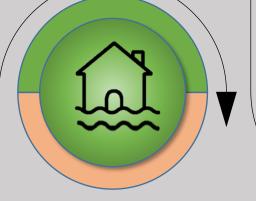


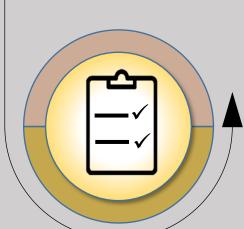


Flood Hydrograph:

Design Hyetograph is convoluted with 1h Unit Hydrograph to estimate

Flood Hydrograph entering the reservoir.





Risk index based on and

Fuzzy framework is

likelihood of flood,

consequences.

severity of flood and

**S** estimated utilizing

Risk Index:

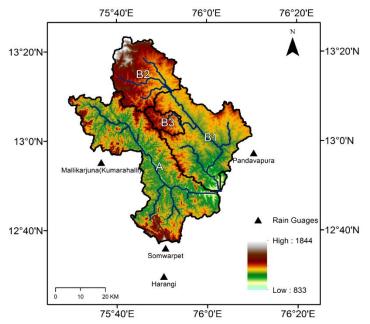
Dam Break Analysis

DBA is performed in 1D2D coupled HEC-RAS dam
break model to estimate
extent of Flood inundation,
maximum depth and flood
velocity at locations
downstream of dam.

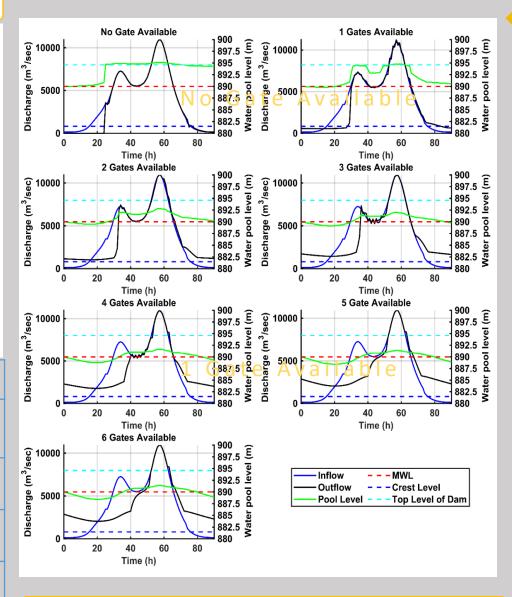




#### Catchment Plan of Hemavathy Dam.



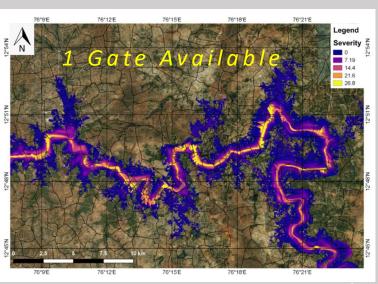
| River                   | Hemavathy                | Full Reservoir<br>Level   | 890.62 m           |
|-------------------------|--------------------------|---------------------------|--------------------|
| Longitude/<br>Latitude  | 76°03' E/<br>12°45' N    | Min.<br>Drawdown<br>Level | 872.33 m           |
| Capacity of<br>Spillway | 3624.55<br>cumecs        | No. of<br>Spillway Gates  | 6                  |
| Type of<br>Dam          | Earthen +<br>Gravity Dam | Type of<br>Spillway Gates | Radial             |
| Length of<br>Dam        | 4692 m                   | Size of<br>Spillway Gates | 10.67m x<br>15.24m |
| Height of<br>Dam        | 58.5 m                   | Shape of<br>Spillway      | Ogee               |
| Catchment<br>Area       | 2810 Sq. Km              | Length of<br>Spillway     | 94.5 m             |



Flood Routing Calculation corresponding to design flood based on each case of gate availability, ranging from 0 to 6. for Hemavathy reservoir project

#### Results





Flood inundation map showing maximum severity

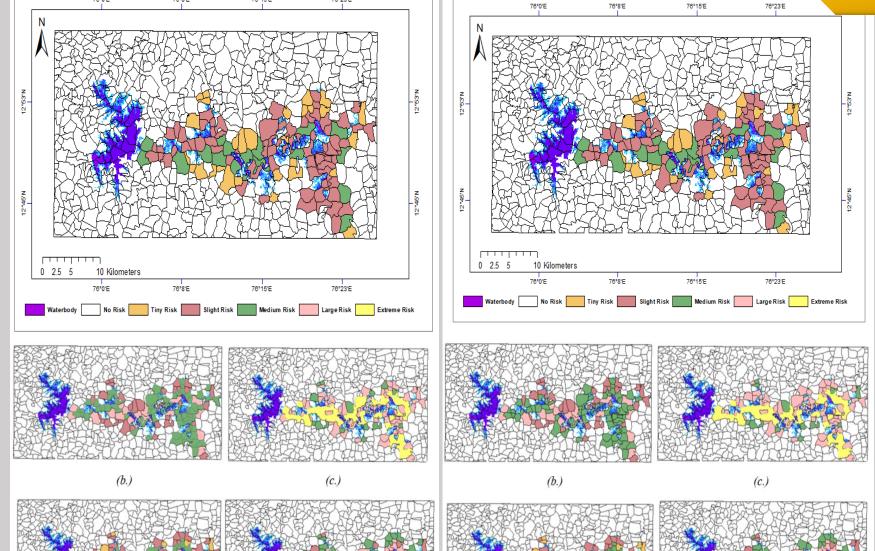


#### No Gate Available

(d.)

#### 1 Gate Available

# Village Wise Aggregative Risk Index (VW-ARI)

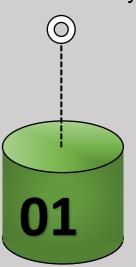


(d.)

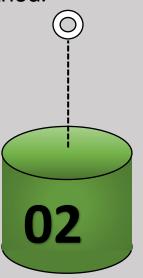
- a) Method 1: Conventional method for estimating ARI based on SFS theory (Type 1).
- b) Method 2: Modified Conventional method for estimating ARI based on SFS theory (Type 2).
- c) Method 3: ARI based VFS theory.
- d) Method 4: ARI based on integration of SFS and VFS theory (Type 1)
- e) Method 5: ARI based on integration of SFS and VFS theory (Type 2)

#### Conclusions

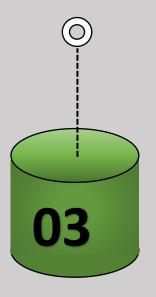
ARI based solely on VFS theory results in over estimation of Risk as compared to Risk based on integration of SFS and VFS theory.



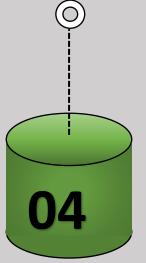
ARI based on Type 1 method results in under estimation of risk as compared to Risk based on Type 2 method.



ARI based on integration of SFS and VFS theory (Type 2) is preferred as the estimated risk is reasonable and it takes into account the influence of uncertainties / vagueness associated with linguistic classification of Risk Items.



Weights calculated using Fuzzy AHP are more reliable as they take into account uncertainty associated with subjective pairwise comparison matrix





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