



INDIAN INSTITUTE OF SCIENCE

# **STOCHASTIC HYDROLOGY**

Lecture -35

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# Summary of the previous lecture

- Multivariate stochastic models
  - Multisite Markov model

$$X_{t+1} = EX_t + G\varepsilon$$

where

$X_t$  is a  $p \times 1$  vector of standardized values of the variable generated at time  $t$ ,

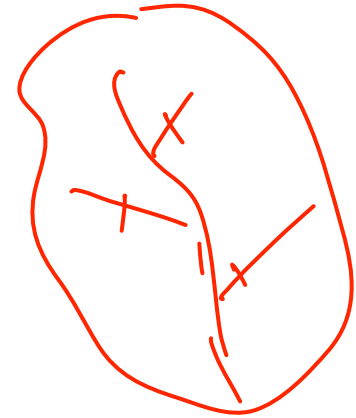
$E$  is a  $p \times p$  diagonal matrix whose  $j^{\text{th}}$  diagonal element is  $\rho_j(1)$ ,

$G$  is a  $p \times p$  diagonal matrix whose  $j^{\text{th}}$  diagonal element is  $\sqrt{1 - \rho_j^2(1)}$

$\varepsilon$  is a  $p \times 1$  vector of random variates

- Matalas model

$$X_{t+1} = AX_t + B\varepsilon_{t+1}$$



*P: no  
of sites*

# Multivariate Stochastic models

- Matalas (1967) has given a multisite normal generation model that preserves the mean, variance, lag one serial correlation, lag one cross-correlation and lag zero cross-correlation.

$$X_{t+1} = AX_t + B\epsilon_{t+1}$$

*Standardized data*

$$s_t = \frac{x_t - \bar{x}}{s}$$

where

$X_t$  and  $X_{t+1}$  are  $p \times 1$  vectors representing standardized data corresponding to  $p$  sites at time steps  $t$  and  $t+1$  resp.

Assumption is that the model is multivariate normal.

Ref.: Matalas, N.C. (1967) Mathematical assessment of synthetic hydrology, Water Resources Research 3(4):937-945

# Multivariate Stochastic models

$\varepsilon_{t+1}$  is  $N(0,1)$ ;  $p \times 1$  vector with  $\varepsilon_{t+1}$  independent of  $X_t$ .

A and B are coefficient matrices of size  $p \times p$ . B is assumed to be lower triangular matrix

$$M_0 = E \left[ X_t X_t' \right] \quad M_0 \text{ is the cross-correlation matrix (size } p \times p \text{) of lag zero}$$

$$m_0(i, j) = \frac{1}{n} \sum_{t=1}^n \left( \frac{Q_{i,t} - \bar{Q}_i}{s_i} \right) \left( \frac{Q_{j,t} - \bar{Q}_j}{s_j} \right)$$

# Multivariate Stochastic models

- The expectation of  $X_t X_{t-1}'$  is denoted by  $M_1$

$$M_1 = E [ X_t X_{t-1}' ]$$

If  $m_1(i, j)$  is a element of  $M_1$  matrix (size p x p) in the  $i^{\text{th}}$  row and  $j^{\text{th}}$  column,

$$m_1(i, j) = E [ X(i, t) X(j, t - 1) ]$$

Expected value of a matrix is matrix of expected values of individual elements

$$m_1(i, j) = \frac{1}{n-1} \sum_{t=2}^n x(i, t) x(j, t - 1)$$

n is no. of time periods

# Multivariate Stochastic models

$$m_1(i, j) = \frac{1}{n-1} \sum_{t=2}^n \left( \frac{Q_{i,t} - \bar{Q}_i}{s_i} \right) \left( \frac{Q_{j,t-1} - \bar{Q}_j}{s_j} \right)$$

Q is the original random variable before standardization e.g., stream flow

i.e.,  $m_1(i, j)$  represents lag one cross correlation between the data at sites  $i$  and  $j$ .

Therefore  $M_1$  is the cross-correlation matrix of lag one.

# Multivariate Stochastic models

Considering the model,

$$X_{t+1} = AX_t + B\varepsilon_{t+1}$$

Post multiplying with  $X_t'$  on both sides and taking the expectation, .

$$E[X_{t+1}X_t'] = AE[X_tX_t'] + \underbrace{BE[\varepsilon_{t+1}X_t']}_0$$

$$M_1 = AM_0 + 0$$

$\varepsilon_{t+1}$  and  $X_t'$  are independent

$$A = M_1M_0^{-1}$$

# Multivariate Stochastic models

Post multiplying with  $X_{t+1}'$  on both sides and taking the expectation,.

$$X_{t+1} = AX_t + B\mathcal{E}_{t+1}$$

$$X_{t+1}X_{t+1}' = AX_tX_{t+1}' + B\mathcal{E}_{t+1}X_{t+1}'$$

$$\underbrace{E[X_{t+1}X_{t+1}']}_{M_0} = AE[X_tX_{t+1}'] + BE[\mathcal{E}_{t+1}X_{t+1}']$$



# Multivariate Stochastic models

$$M_1 = E [X_t X_{t-1}']$$

$$M_1' = \left\{ E [X_t X_{t-1}'] \right\}'$$

$$= E \left[ \left\{ X_t X_{t-1}' \right\}' \right]$$

$$= E [X_{t-1} X_t']$$

or

$$M_1' = E [X_t X_{t+1}']$$

$(A B)'$   
 $B' A'$

# Multivariate Stochastic models

$$\begin{aligned}\varepsilon_{t+1}' X_{t+1}' &= \varepsilon_{t+1}' \{AX_t + B\varepsilon_{t+1}\}' \\ &= \varepsilon_{t+1}' X_t' A' + \varepsilon_{t+1}' \varepsilon_{t+1}' B'\end{aligned}$$

Taking expectation on both sides,

$$\begin{aligned}E[\varepsilon_{t+1}' X_{t+1}'] &= E[\varepsilon_{t+1}' X_t' A' + \varepsilon_{t+1}' \varepsilon_{t+1}' B'] \\ &= E[\varepsilon_{t+1}' X_t' A'] + E[\varepsilon_{t+1}' \varepsilon_{t+1}' B'] \\ &= 0 + IB' \\ &= B'\end{aligned}$$

Since  $\varepsilon_{t+1}$  has  
unit variance

# Multivariate Stochastic models

Substituting in the equation,

$$E \left[ X_{t+1} X_{t+1}' \right] = AE \left[ X_t X_t' \right] + BE \left[ \varepsilon_{t+1} \varepsilon_{t+1}' \right]$$

$$M_0 = AM_1 + BB'$$

$$M_0 = M_1 M_0^{-1} M_1' + BB'$$

$$BB' = M_0 - M_1 M_0^{-1} M_1'$$

$$A = M_1 M_0^{-1}$$

If  $C = BB'$

$$C = M_0 - M_1 M_0^{-1} M_1'$$

# Multivariate Stochastic models

- B does not have a unique solution.
- One method is to assume B to be a lower triangular matrix.

$$BB' = \begin{bmatrix} b(1,1) & 0 & 0 & \cdot & \cdot & 0 \\ b(2,1) & b(2,2) & 0 & \cdot & \cdot & 0 \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ b(p,1) & b(p,2) & \cdot & \cdot & \cdot & b(p,p) \end{bmatrix} \begin{bmatrix} b(1,1) & b(2,1) & \cdot & \cdot & \cdot & b(p,1) \\ 0 & b(2,2) & \cdot & \cdot & \cdot & b(p,2) \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ 0 & 0 & \cdot & \cdot & \cdot & b(p,p) \end{bmatrix}$$

$$C = \begin{bmatrix} c(1,1) & c(1,2) & c(1,3) & \cdot & \cdot & c(1,p) \\ c(2,1) & c(2,2) & c(2,3) & \cdot & \cdot & c(2,p) \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ c(p,1) & c(p,2) & \cdot & \cdot & \cdot & c(p,p) \end{bmatrix}$$

# Multivariate Stochastic models

- The diagonal elements of the B matrix are obtained as,

$$b(1,1) = c(1,1)^{1/2}$$

$$b(2,2) = \left\{ c(2,2) - b^2(2,1) \right\}^{1/2}$$

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$$b(k,k) = \left\{ c(k,k) - b^2(k,k-1) - b^2(k,k-2) - \dots - b^2(k,1) \right\}^{1/2}$$

These elements are obtained one by one, using also the expressions for the  $k^{\text{th}}$  row elements given in the next slide

# Multivariate Stochastic models

- The elements in the  $k^{\text{th}}$  row are obtained as,

$$b(k,1) = \frac{c(k,1)}{b(1,1)}$$

$$b(k,2) = \frac{c(k,2) - b(2,1)b(k,1)}{b(2,2)}$$

⋮

$$b(k,j) = \frac{c(k,j) - b(j,1)b(k,1) - b(j,2)b(k,2) \dots - b(j,j-1)b(k,j-1)}{b(j,j)}$$

# Example – 1

The annual flow in MCM at two sites P and Q is given below. Generate the first two values of data from these two sites.

Year	1	2	3	4	5	6	7	8	9	10
Annual flow at site P (MCM)	4946	7017	6653	6355	5908	5327	4548	3556	3852	5319
Annual flow at site Q (MCM)	5142	6240	5648	5977	6008	5045	4630	4604	4250	6182

Year	11	12	13	14	15	16	17	18	19
Annual flow at site P (MCM)	4631	5746	5111	5419	6060	7336	3736	3780	6034
Annual flow at site Q (MCM)	4703	6582	5461	5288	5440	7546	4634	4823	5577

# Example – 1 (Contd.)

Site	P	Q
Mean	5333	5462
Std.dev.	1125.1	823.5

$M_0$  matrix is cross correlation matrix of lag zero

$$M_0 = \begin{matrix} & \begin{matrix} P & Q \end{matrix} \\ \begin{matrix} P \\ Q \end{matrix} & \begin{bmatrix} r_{P,P}(0) & r_{P,Q}(0) \\ r_{Q,P}(0) & r_{Q,Q}(0) \end{bmatrix} \end{matrix}$$



# Example – 1 (Contd.)

$$r_{P,Q}(0) = \frac{\sum_{i=1}^n (x_{P,i} - \bar{x}_P)(x_{Q,i} - \bar{x}_Q)}{(n)S_P S_Q}$$

$$M_0 = \begin{bmatrix} 1 & 0.796 \\ 0.796 & 1 \end{bmatrix}$$

$M_1$  matrix is cross correlation matrix of lag one

$$M_1 = \begin{matrix} & \begin{matrix} P & Q \end{matrix} \\ \begin{matrix} P \\ Q \end{matrix} & \begin{bmatrix} r_{P,P}(1) & r_{P,Q}(1) \\ r_{Q,P}(1) & r_{Q,Q}(1) \end{bmatrix} \end{matrix}$$

## Example – 1 (Contd.)

$$r_{P,Q}(1) = \frac{\sum_{i=1}^n (x_{P,i} - \bar{x}_P)(x_{Q,i+1} - \bar{x}_Q)}{(n-1)s_P s_Q}$$

$$M_1 = \begin{bmatrix} 0.302 & 0.164 \\ 0.02 & -0.118 \end{bmatrix}$$

$$A = M_1 M_0^{-1}$$

$$M_0^{-1} = \begin{bmatrix} 2.73 & -2.17 \\ -2.17 & 2.73 \end{bmatrix}$$

# Example – 1 (Contd.)

$$A = M_1 M_0^{-1}$$

$$= \begin{bmatrix} 0.302 & 0.164 \\ 0.02 & -0.118 \end{bmatrix} \begin{bmatrix} 2.73 & -2.17 \\ -2.17 & 2.73 \end{bmatrix}$$

$$A = \begin{bmatrix} 0.47 & -0.21 \\ 0.31 & -0.37 \end{bmatrix}$$

# Example – 1 (Contd.)

$$C = M_0 - M_1 M_0^{-1} M_1'$$

$$M_0 = \begin{bmatrix} 1 & 0.796 \\ 0.796 & 1 \end{bmatrix} \quad M_1 = \begin{bmatrix} 0.302 & 0.164 \\ 0.02 & -0.118 \end{bmatrix}$$

$$M_0^{-1} = \begin{bmatrix} 2.73 & -2.17 \\ -2.17 & 2.73 \end{bmatrix} \quad M_1' = \begin{bmatrix} 0.302 & 0.02 \\ 0.164 & -0.118 \end{bmatrix}$$

$$C = \begin{bmatrix} 0.89 & 0.76 \\ 0.76 & 0.95 \end{bmatrix}$$

## Example – 1 (Contd.)

$$c(1,1) = 0.89, c(1,2) = c(2,1) = 0.76,$$

$$c(2,2) = 0.95$$

$$b(1,1) = c(1,1)^{1/2} = (0.89)^{1/2} = 0.94$$

$$b(2,1) = \frac{c(2,1)}{b(1,1)} = \frac{0.76}{0.94} = 0.81 \qquad b(k,1) = \frac{c(k,1)}{b(1,1)}$$

$$\begin{aligned} b(2,2) &= \left\{ c(2,2) - b^2(2,1) \right\}^{1/2} \\ &= \left\{ 0.95 - 0.81^2 \right\}^{1/2} = 0.54 \end{aligned}$$

## Example – 1 (Contd.)

$$b(1,1) = 0.94, b(2,1) = 0.81, b(2,2) = 0.54$$

$$B = \begin{bmatrix} 0.94 & 0 \\ 0.81 & 0.54 \end{bmatrix}$$

$$X_{t+1} = AX_t + B\mathcal{E}_{t+1}$$

$$\begin{bmatrix} x_{P,t+1} \\ x_{Q,t+1} \end{bmatrix} = \begin{bmatrix} 0.47 & -0.21 \\ 0.31 & -0.37 \end{bmatrix} \begin{bmatrix} x_{P,t} \\ x_{Q,t} \end{bmatrix} + \begin{bmatrix} 0.94 & 0 \\ 0.81 & 0.54 \end{bmatrix} \begin{bmatrix} \mathcal{E}_{P,t+1} \\ \mathcal{E}_{Q,t+1} \end{bmatrix}$$

## Example – 1 (Contd.)

$$\begin{bmatrix} x_{P,t+1} \\ x_{Q,t+1} \end{bmatrix} = \begin{bmatrix} 0.47 & -0.21 \\ 0.31 & -0.37 \end{bmatrix} \begin{bmatrix} x_{P,t} \\ x_{Q,t} \end{bmatrix} + \begin{bmatrix} 0.94 & 0 \\ 0.81 & 0.54 \end{bmatrix} \begin{bmatrix} \varepsilon_{P,t+1} \\ \varepsilon_{Q,t+1} \end{bmatrix}$$

The initial value of  $x_{P,0}$  and  $x_{Q,0}$  are considered as zero  
 $\varepsilon_{P,1} = -0.134$  and  $\varepsilon_{Q,1} = -0.268$

$$\begin{aligned} \begin{bmatrix} x_{P,1} \\ x_{Q,1} \end{bmatrix} &= \begin{bmatrix} 0.94 & 0 \\ 0.81 & 0.54 \end{bmatrix} \begin{bmatrix} -0.134 \\ -0.268 \end{bmatrix} \\ &= \begin{bmatrix} -0.126 \\ -0.254 \end{bmatrix} \end{aligned}$$

## Example – 1 (Contd.)

$$\begin{bmatrix} x_{P,2} \\ x_{Q,2} \end{bmatrix} = \begin{bmatrix} 0.47 & -0.21 \\ 0.31 & -0.37 \end{bmatrix} \begin{bmatrix} x_{P,1} \\ x_{Q,1} \end{bmatrix} + \begin{bmatrix} 0.94 & 0 \\ 0.81 & 0.54 \end{bmatrix} \begin{bmatrix} \varepsilon_{P,2} \\ \varepsilon_{Q,2} \end{bmatrix}$$

$$\varepsilon_{P,2} = 1.639 \text{ and } \varepsilon_{Q,2} = 0.134$$

$$\begin{aligned} \begin{bmatrix} x_{P,2} \\ x_{Q,2} \end{bmatrix} &= \begin{bmatrix} 0.47 & -0.21 \\ 0.31 & -0.37 \end{bmatrix} \begin{bmatrix} -0.126 \\ -0.254 \end{bmatrix} + \begin{bmatrix} 0.94 & 0 \\ 0.81 & 0.54 \end{bmatrix} \begin{bmatrix} 1.639 \\ 0.134 \end{bmatrix} \\ &= \begin{bmatrix} 1.543 \\ 1.449 \end{bmatrix} \end{aligned}$$



# Example – 1 (Contd.)

Generated annual flow at Site X :

Site	X	Y
Mean	5333	5462
Std.dev.	1125.1	823.5

$$Q_{X,1} = \bar{x}_X + x(X,1)s_X$$

$$Q_{X,1} = 5333 - 0.126*1125.1 = 5191.2 \text{ MCM}$$

$$Q_{X,2} = \bar{x}_X + x(X,2)s_X$$

$$Q_{X,2} = 5333 + 1.543*1125.1 = 7069 \text{ MCM}$$

Similarly at Site Y :

$$Q_{Y,1} = 5462 - 0.254*823.5 = 5252.8 \text{ MCM}$$

$$Q_{Y,2} = 5462 + 1.449*823.5 = 6655.3 \text{ MCM}$$

# **CASE STUDIES**