



INDIAN INSTITUTE OF SCIENCE

# **STOCHASTIC HYDROLOGY**

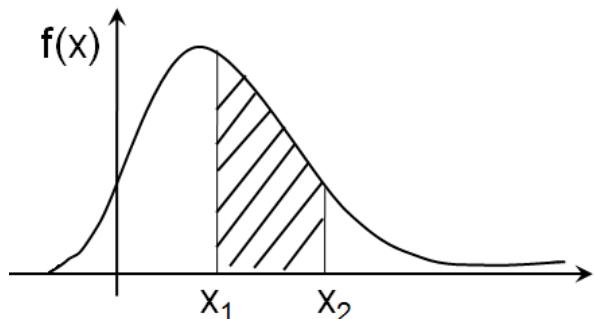
Lecture -40

Course Instructor : Prof. P. P. MUJUMDAR

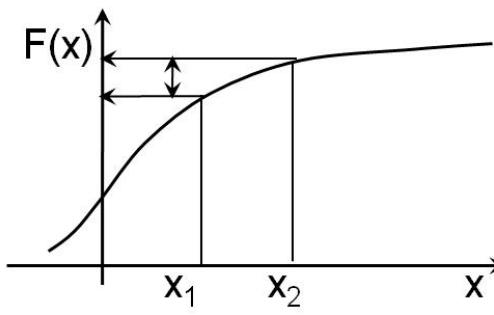
Department of Civil Engg., IISc.

# Summary of the Course

- Concept of a random variable
- Discrete and continuous random variables
- Probability mass function, density function and cumulative distribution functions



PDF



CDF

$$\begin{aligned} f(x) &\geq 0 \\ \int_{-\infty}^{\infty} f(x) dx &= 1 \end{aligned}$$

# Summary of the Course (contd.)

- Bivariate distributions, Joint pmf and pdf
- Marginal density functions

$$g(x) = \int_{-\infty}^y f(x, y) dy$$

$$F(x) = \int_{-\infty}^x g(x) dx$$

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy = 1$$
$$f(x, y) \geq 0$$

- Conditional distributions

$$g(x/y) = \frac{f(x, y)}{h(y)} \quad h(y) > 0$$

# Summary of the Course (contd.)

- Independent random variables

$$g(x/y) = \frac{f(x,y)}{h(y)} \quad h(y) > 0$$

$$g(x) = \frac{f(x,y)}{h(y)}$$

$$f(x,y) = g(x).h(y)$$

- Functions of random variables

- Moments of a distribution  $\mu_n^0 = \int_{-\infty}^{\infty} x^n f(x) dx$

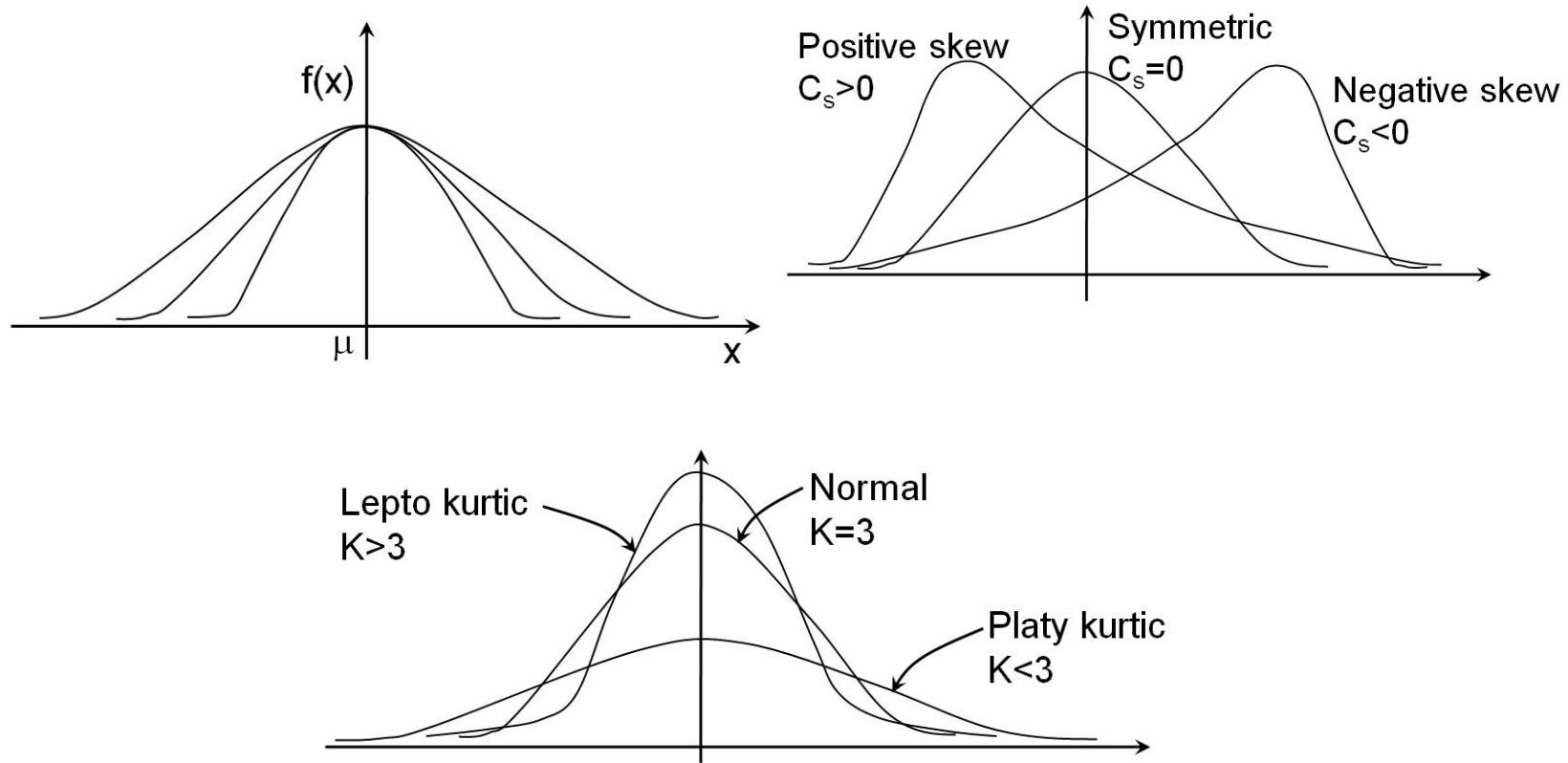
- Expected value

$$\mu = E(X) = \int_{-\infty}^{\infty} x f(x) dx$$

$$\mu_n = \int_{-\infty}^{\infty} (x - \mu)^n f(x) dx$$

# Summary of the Course (contd.)

- Measures of central tendency, dispersion, symmetry and peakedness

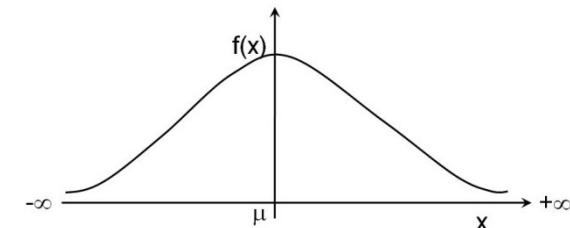


# Summary of the Course (contd.)

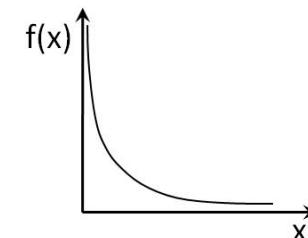
- Normal distribution

$$Z = \frac{X - \mu}{\sigma}$$

$$F(z) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^z e^{-z^2/2} dz \quad -\infty < z < +\infty$$



- Central limit theorem
- Log-normal distribution
- Exponential Distribution
- Gamma Distribution
- Extreme Value Distributions
  - Extreme Value Type-I Distribution  
(Gumbel's Extreme Value Distribution)
  - Extreme Value Type-III Minimum Distribution  
(Weibull's Distribution)



# Summary of the Course (contd.)

- Parameter estimation
  - Method of matching points
  - Method of moments

Equate the first ‘m’ moments of the population to the sample estimates of the first ‘m’ moments

Results in ‘m’ equations; solve to get the ‘m’ unknown parameters of the distribution

- Method of maximum likelihood

$$L = f(x_1; \theta_1; \theta_2 \dots \theta_m) \times f(x_2; \theta_1; \theta_2 \dots \theta_m) \times f(x_n; \theta_1; \theta_2 \dots \theta_m)$$

$$= \prod_{i=1}^n f(x_i, \theta_1, \dots, \theta_m)$$

$$\frac{\partial L}{\partial \theta_i} = 0 \quad \forall i$$

# Summary of the Course (contd.)

- Jointly Distributed Random Variables
- Covariance

$$\begin{aligned}\mu_{1,1} &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (x - \mu_x)(y - \mu_y) f(x, y) dx dy \\ &= E[(x - \mu_x)(y - \mu_y)] \\ s_{X,Y} &= \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{n-1}\end{aligned}$$

- Correlation coefficient

$$r_{X,Y} = \frac{s_{X,Y}}{s_X s_Y}$$

# Summary of the Course (contd.)

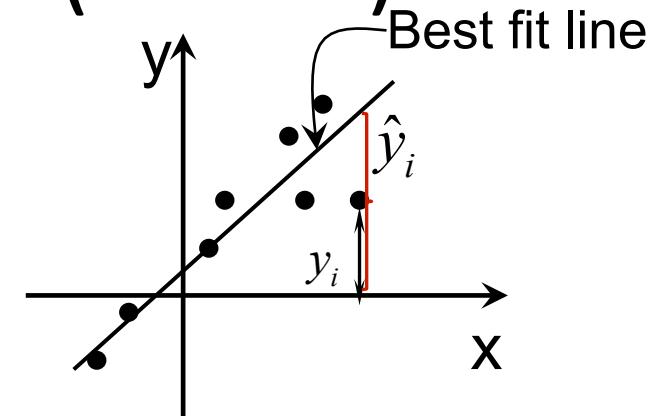
## Simple Linear Regression

$(x_i, y_i)$  are observed values

$\hat{y}_i$  is predicted value of  $x_i$

$$\hat{y}_i = a + bx_i$$

Error,  $e_i = y_i - \hat{y}_i$       Estimate the parameters  $a, b$  such that the square error is minimum



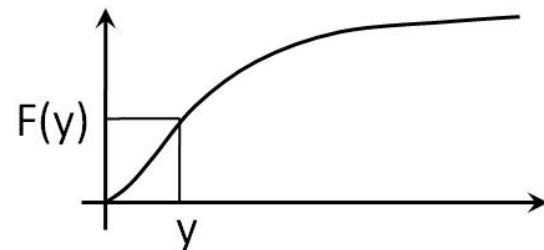
# Summary of the Course (contd.)

- Data Generation

Randomly picked up values of  $F(y)$  follow a uniform distribution  $u(0, 1)$

$$F(y) = \int_{-\infty}^y f(y) dy$$

$$F(y) = R_u = \int_{-\infty}^y f(y) dy$$



Gamma distribution

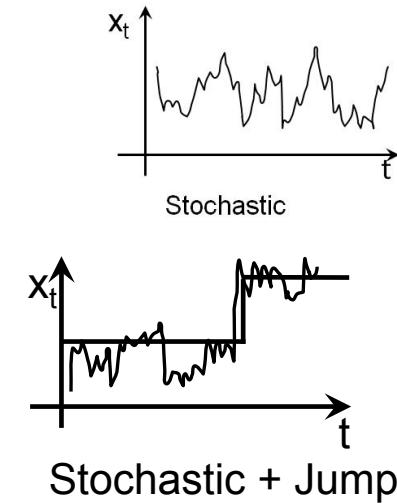
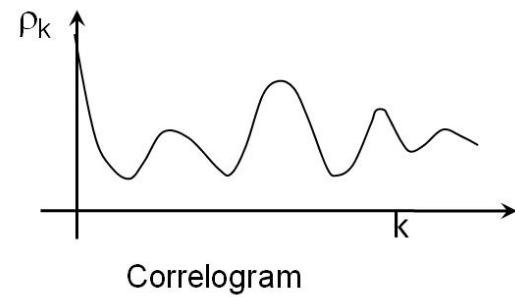
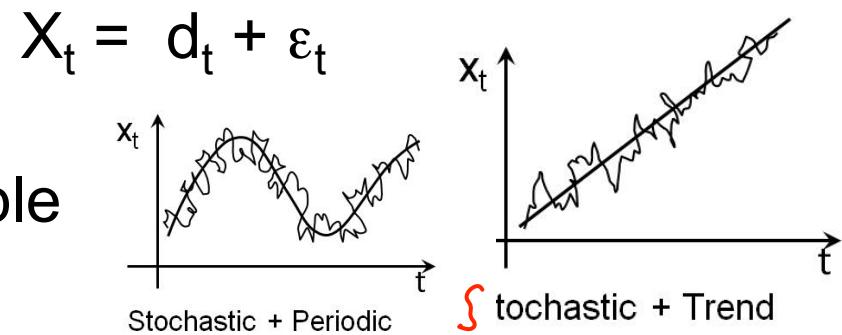
$$y = \frac{-\sum_{i=1}^{\eta} \ln R_{u_i}}{\lambda} \quad (\text{for integer values of } \eta)$$

Normal distribution: Given  $R_N$ , data is generated by

$$y = \sigma R_N + \mu$$

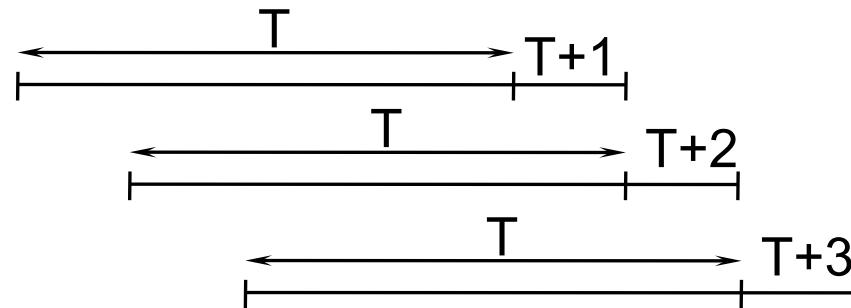
# Summary of the Course (contd.)

- Time series analysis
  - Realization ; Ensemble
  - Stationarity
  - Auto covariance Function
  - Auto correleation and correlogram



# Summary of the Course (contd.)

- Data Extension & Forecasting
  - Moving average
  - Double moving average



- Data Generation – Uncorrelated Data

# Summary of the Course (contd.)

- Data Generation – Serially Correlated Data
  - First order Markov Model
    - Annual flow generation

$$X_{t+1} = \mu_x + \rho_1 (X_t - \mu_x) + u_{t+1} \sigma_x \sqrt{1 - \rho_1^2}$$

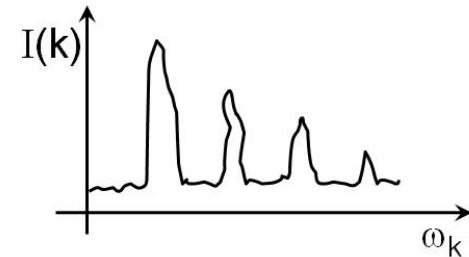
- First order Markov model with non-stationarity
  - Thoma Fiering model for monthly and seasonal flow generation

$$X_{i,j+1} = \mu_{j+1} + \rho_j \frac{\sigma_{j+1}}{\sigma_j} (X_{ij} - \mu_j) + t_{i,j+1} \sigma_{j+1} \sqrt{1 - \rho_j^2}$$

# Summary of the Course (contd.)

- Frequency domain analysis
  - Spectral density
  - Test for significance of periodicities
  - Removing periodicities
  - Standardizing the data

$$Z_t = \frac{(X_t - \bar{X}_i)}{S_i}$$



$$I(k) = \frac{N}{2} [\alpha_k^2 + \beta_k^2]$$
$$\omega_k = \frac{2\pi k}{N}$$

# Summary of the Course (contd.)

- ARIMA models

$$X_t = \phi_1 X_{t-1} + \phi_2 X_{t-2} + \dots + \phi_p X_{t-p} + \theta_1 e_{t-1} + \theta_2 e_{t-2} + \dots + \theta_q e_{t-q} + e_t$$

$\{e_t\}$  is the residual series

Assumptions :  $\{e_t\}$  has zero mean with uncorrelated terms

$$\hat{X}_{t+1} = \sum_{j=1}^p \phi_j X_{t-j} + \sum_{j=1}^q \theta_j e_{t-j}$$

- Partial Auto Correlation function

Calculation of Partial Auto Correlations:  $\phi_p$  is the PAC of order p  
(Yule Walker equations)

$$P_p * \phi_p = \rho_p$$

# Summary of the Course (contd.)

## Box Jenkins Time series models

Differencing:  $Y_t = X_t' = X_t - X_{t-1}$

$X_t'$  is First order differencing

$X_t'' = X_t' - X_{t-1}'$

$X_t''$  is Second order differencing

Operator ‘B’:

The effect of operator ‘B’ is to shift the argument to that one step behind.

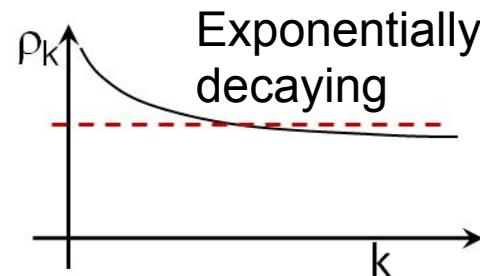
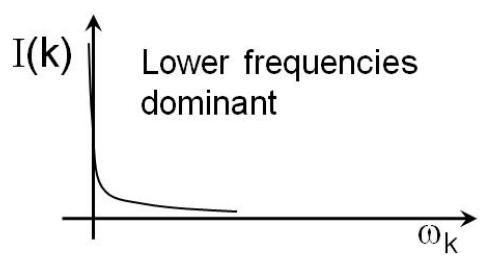
$$BX_t = X_{t-1}$$

$$BX_{t-1} = X_{t-2}$$

# Summary of the Course (contd.)

- Behavior of AR and MA process

AR(2) process

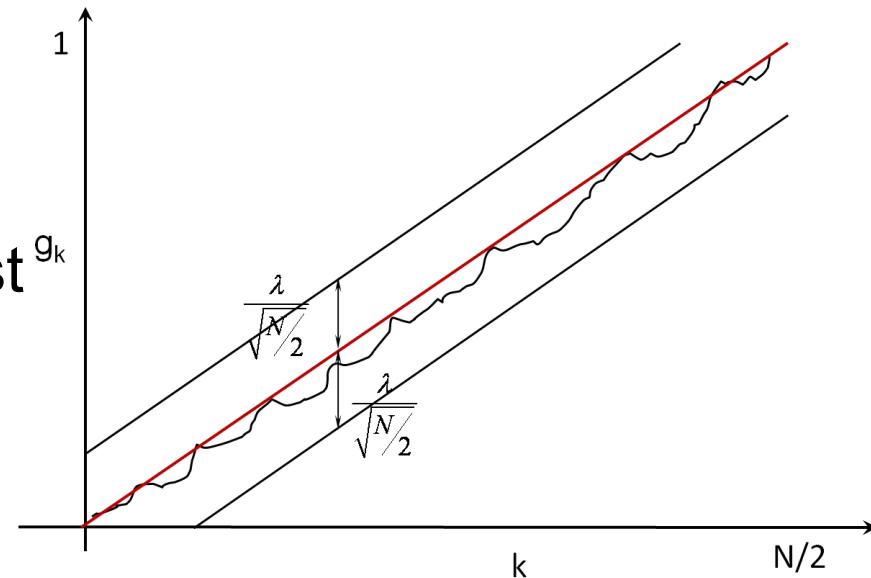


- Parameter estimation
  - Marquadt's algorithm
  - Matlab function “armax”  
 $m = \text{armax}(\text{data}, \text{orders})$
- Model selection
  - Maximum likelihood rule
  - Mean square error

# Summary of the Course (contd.)

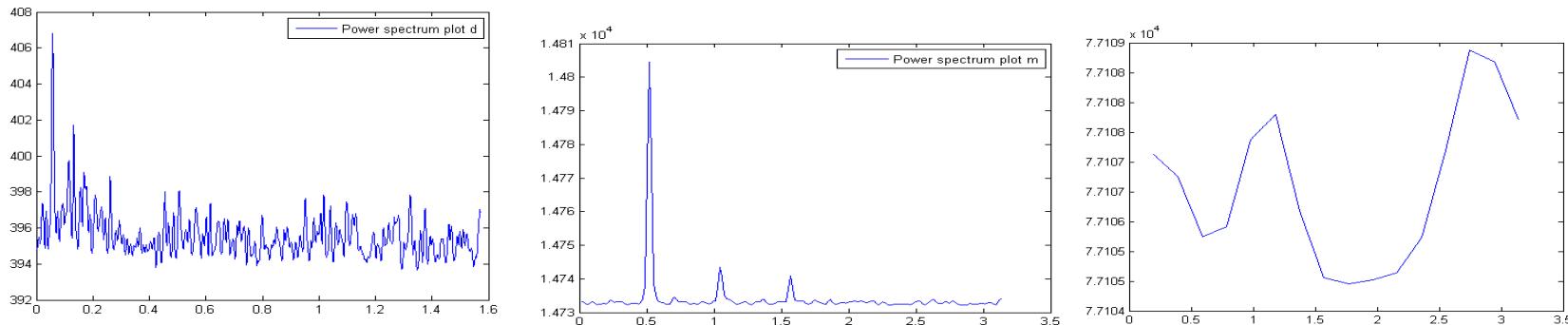
- Model testing/validation
  - Significance of residual mean
  - Significance of periodicities
    - Cumulative periodogram test or Bartlett's test
  - White noise test
    - Whittle's test
    - Portmanteau test

$$\eta(e) = \frac{N^{1/2} \bar{e}}{\hat{\rho}^{1/2}}$$



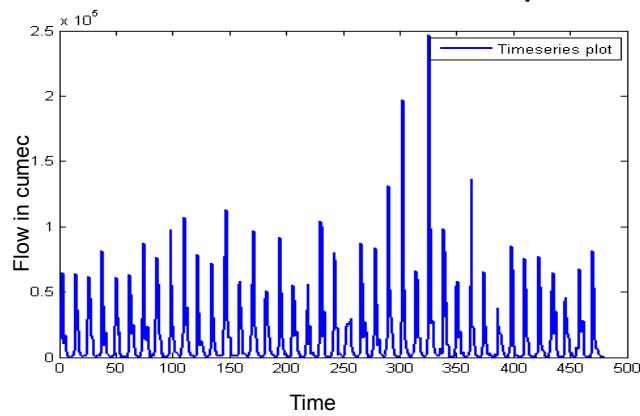
# Summary of the Course (contd.)

- Use of ARMA models for data generation and one-time step ahead forecasting
- Case studies 1 and 2
  - Daily, monthly and annual rainfall
  - Annual streamflow
  - Monthly streamflow

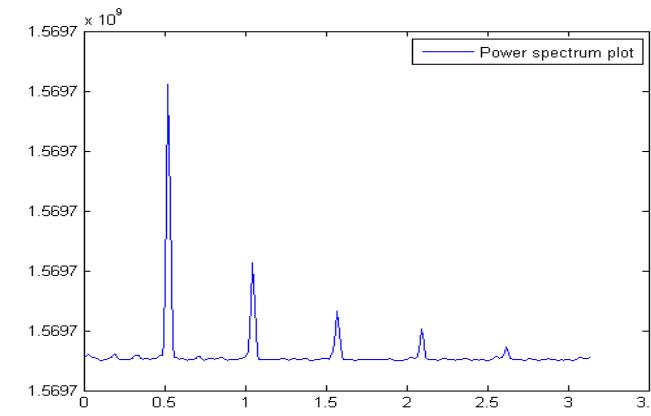


# Summary of the Course (contd.)

- Case study -3: Monthly streamflows at KRS reservoir
  - Plots of Time series, Correlogram, Partial Autocorrelation function and Power spectrum
  - Standardization to remove periodicities
  - Candidate ARMA models : (contiguous and non-contiguous)
    - Log Likelihood
    - Mean square error



Monthly stream flow data



Power spectrum

# Summary of the Course (contd.)

Markov chain

$$P[X_t/X_{t-1}, X_{t-2}, \dots, X_0] = P[X_t/X_{t-1}]$$

Transition Probability Matrix(TPM):

$$\sum_{j=1}^m P_{ij} = 1 \quad \forall i \quad \hat{P}_{ij} = \frac{n_{ij}}{\sum_{j=1}^m n_{ij}}$$

$$p^{(n)} = p^{(0)} \times P^n \quad - \text{ Steady state Markov chains}$$

# Summary of the Course (contd.)

- Introduction to frequency analysis
  - Extreme events
- Recurrence interval
- Return period
  - Expected value of recurrence interval
- $P[X \geq x_T] = p = 1/T$
- Probability that a  $T$  year return period event will occur at least once in  $N$  years is

$$1 - \left(1 - \frac{1}{T}\right)^N$$

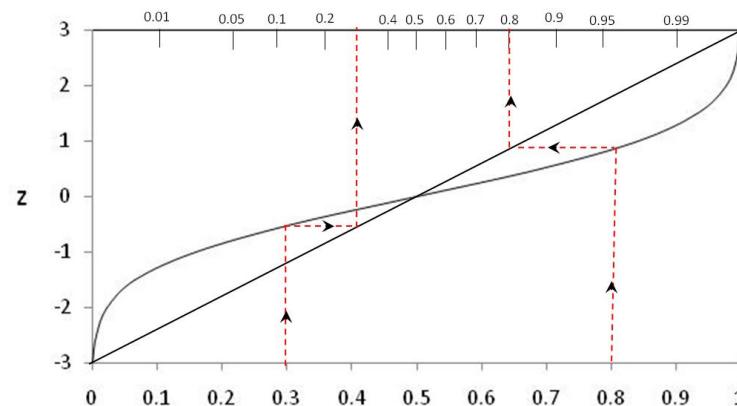
# Summary of the Course (contd.)

- Frequency factors  $x_T = \bar{x} + K_T s$ 
  - Normal distribution
  - Gumbel's Extreme Value Type-I distribution
$$K_T = -\frac{\sqrt{6}}{\pi} \left\{ 0.5772 + \ln \left[ \ln \left( \frac{T}{T-1} \right) \right] \right\}$$
  - Log Pearson Type-III distribution
- Probability plotting

Method to check whether a probability distribution fits a set of data or not.

# Summary of the Course (contd.)

- Probability paper construction
  - Mathematical construction
  - Graphical construction
    - Normal distribution
- Plotting position
  - Weibull's formula



$$p(X \geq x_m) = \frac{m}{n+1}$$

# Summary of the Course (contd.)

- Goodness of fit
  - Chi-square test

$$\chi^2_{data} = \sum_{i=1}^k \frac{(N_i - E_i)^2}{E_i} \quad \chi^2_{data} < \chi^2_{1-\alpha, k-p-1}$$

- Kolmogorov-Smirnov test

k - no. of class intervals  
p – no. of parameters

$$\Delta = \text{maximum } |P(x_i) - F(x_i)|$$

$$\Delta < \Delta_0$$

# Summary of the Course (contd.)

- IDF relationship
  - Procedure for creating IDF curves

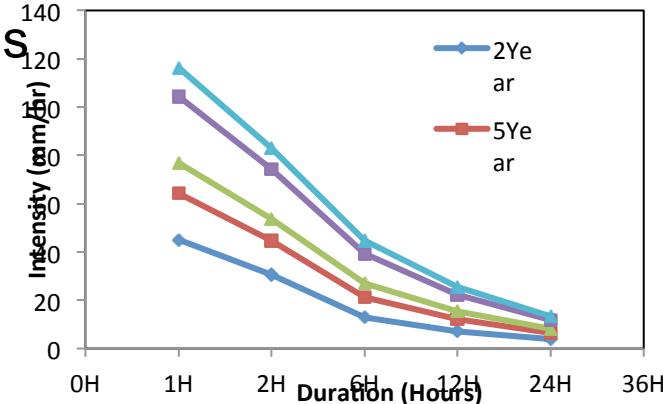
Step 1: Preparation of annual maximum data series

Step 2: Fitting the probability distribution

Step 3: Determining the rainfall depths

$$x_T = \bar{x} + K_T s$$

$$x_T = F^{-1} \left( \frac{T-1}{T} \right)$$



- Empirical equations for IDF relationships

# Summary of the Course (contd.)

- Multiple linear regression

$$y = \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \dots + \beta_p x_p$$

y is dependent variable,  
 $x_1, x_2, x_3, \dots, x_p$  are independent variables and  
 $\beta_1, \beta_2, \beta_3, \dots, \beta_p$  are unknown parameters

$$y_i = \sum_{j=1}^p \beta_j x_{i,j} \quad Y_{(n \times 1)} = X_{(n \times p)} \times B_{(p \times 1)}$$

# Summary of the Course (contd.)

## Principal Component Analysis

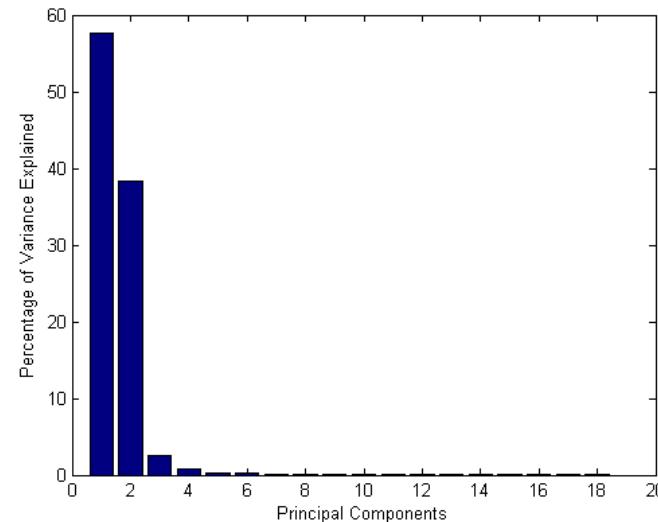
PCA is a way of identifying patterns in the data

$$Z = X A$$

X is nxp matrix of n observations  
on p variables

Z is nxp matrix of n values for  
each of p components

A is pxp matrix of coefficients  
defining the linear  
transformation



# Summary of the Course (contd.)

- Multivariate stochastic models
  - Cross correlation

$$r_{j,h}(k) = \frac{\sum_{i=1}^n (x_{j,i} - \bar{x}_j)(x_{h,i+k} - \bar{x}_h)}{(n-k)s_j s_h}$$

- Two site Markov model

$$X_{h,t} = \bar{x}_h + r_{j,h}(0) \frac{s_h}{s_j} (X_{j,t} - \bar{x}_j) + u_t s_h \sqrt{1 - r_{j,h}^2(0)}$$

# Summary of the Course (contd.)

- Multivariate stochastic models
  - Multisite Markov model

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

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  $\sqrt{1 - \rho_j^2(1)}$

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

# Summary of the Course (contd.)

- Multivariate stochastic models
  - Matalas model

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

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

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

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

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

$M_0$  is the cross-correlation matrix (size  $p \times p$ ) of lag zero

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

$M_1$  is the cross-correlation matrix (size  $p \times p$ ) of lag one

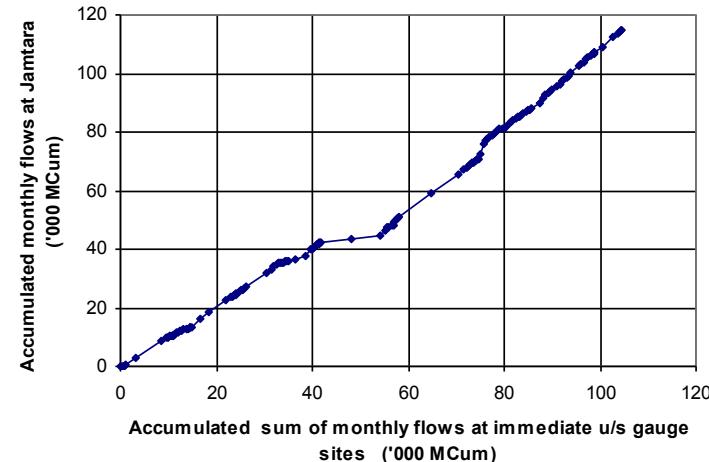
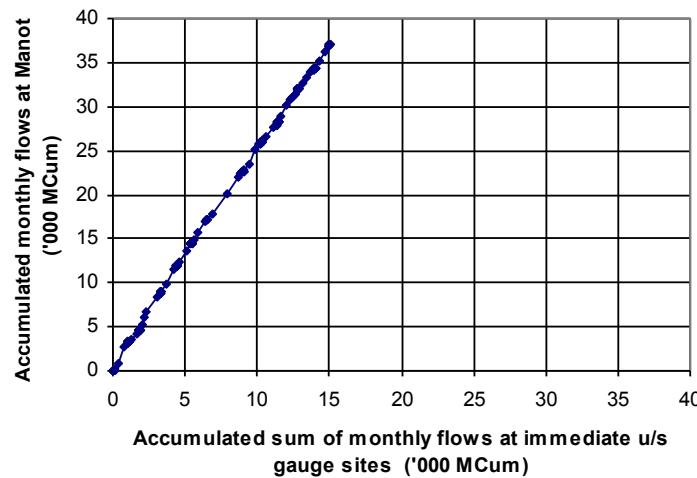
# Summary of the Course (contd.)

- Data consistency checks
  - Filling the missing data for an example basin
    - Monsoon period
    - Non-monsoon period
  - Statistical analysis of data
  - Specific flow

The specific flow is expressed as flow volume per unit area of the catchment

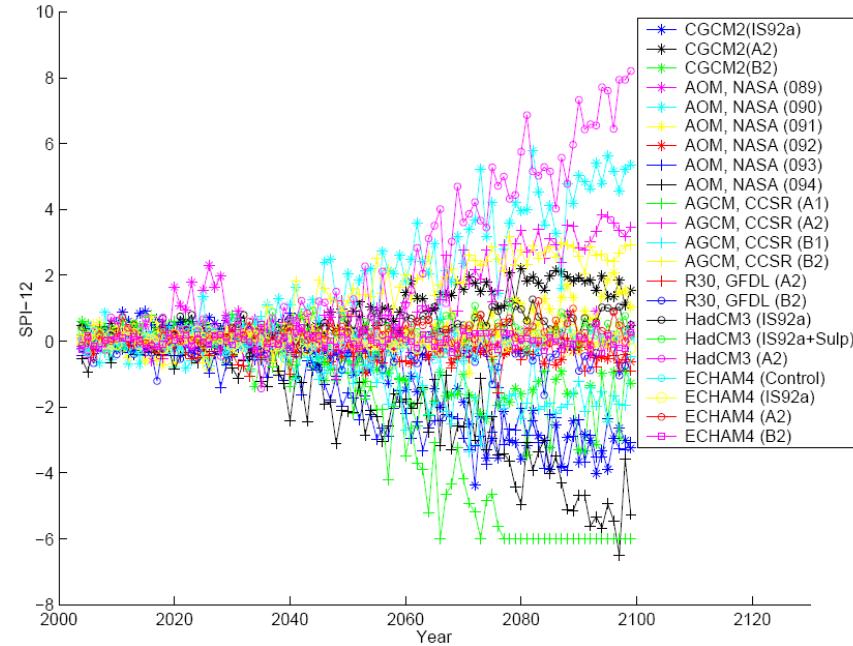
# Summary of the Course (contd.)

- Data consistency checks
  - Double Mass Curve



# Summary of the Course (contd.)

- Data representation through box plots
- Normalisation of flow data
- Assessment of Climate Change Impacts
- Downscaling



# Acknowledgment

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