

Topics: Random processes, spectral estimation, and eigen-image analysis

Spring 2010 Final: Problem 5 (power spectrum and MMSE prediction)

Let $X = [x_1, \dots, x_N]$ be a $P \times N$ matrix formed by P dimensional column vectors, $x_n \in \mathfrak{R}^P$ where $N < P$. We will assume that each column vector is an independent multivariate Gaussian random vector with distribution $N(0, R)$, where R is a positive definite and symmetric matrix.

Furthermore, let $X = U\Sigma V^t$ be the singular value decomposition of X , where $\Sigma_{i,i} \geq \Sigma_{j,j}$ when $i < j$.

a) Write a simple matrix expression for the sample covariance, \hat{R} .

Solution:

$$\hat{R} = \frac{1}{N} X X^t$$

b) Let $\hat{R} = E\Lambda E^t$ be the eigen decomposition of the sample covariance matrix, where E is the orthonormal transform with eigenvectors as columns and Λ is the diagonal matrix of eigen values. (Without loss of generality assume that the eigenvalues are ordered from largest to smallest so that $\Lambda_{i,i} \geq \Lambda_{j,j}$ when $i < j$.)

How many non-zero eigenvalues does the matrix \hat{R} contain?

Solution:

There are N non-zero eigenvalues.

c) Specify the eigenvectors and eigenvalues of \hat{R} in terms of the SVD of X .

Solution:

$$\begin{aligned} \hat{R} &= \frac{1}{N} X X^t \\ &= \frac{1}{N} U \Sigma V^t (U \Sigma V^t)^t \\ &= \frac{1}{N} U \Sigma V^t V \Sigma U^t \\ &= \frac{1}{N} U \Sigma^2 U^t \end{aligned}$$

$$\text{So, } E = U \text{ and } \Lambda = \frac{1}{N} \Sigma^2$$

d) In some application, you can only use two numbers to specify each vector, x_n . So each vector must be approximated by

$$x_n \approx a_n e_1 + b_n e_2$$

where $e_1 \in \mathbb{R}^P$ and $e_2 \in \mathbb{R}^P$ are two orthonormal vectors, and a_n and b_n are two scalar values used to specify each vector, x_n .

What is the best choice of e_1 and e_2 ?

Solution:

The vectors e_1 and e_2 should be the first and second columns of U , so that e_1 is the singular vector corresponding to the largest singular value Σ_{11} , and e_2 is the singular vector corresponding to the second largest singular value Σ_{22} .

Spring 2010 Exam 1: Problem 3 (power spectrum and MMSE prediction)

Let y_n be a wide-sense stationary, jointly Gaussian, zero-mean, discrete-time random process. Then we know that the minimum mean squared error (MSEE) predictor has the form

$$\begin{aligned}\hat{y}_n &= E[y_n | y_k \text{ for } k < n] \\ &= \sum_{i=1}^{\infty} h_{n,i} y_{n-i}\end{aligned}$$

for some scalar constants $h_{n,i}$.

a) How are the functions $h_{n,i}$ and $h_{k,i}$ related for all n , k , and i ? Provide a precise justification for your answer.

Solution:

$h_{n,i} = h_{k,i}$, because Y_n is a stationary random process.

b) Consider the function $z_n = y_{-n}$. What is the MMSE predictor for z_n ? Provide a precise justification for your answer.

Solution:

Notice that:

$$\begin{aligned}R_z(k) &= E[Z_n Z_{n+k}] \\ &= E[Y_{-n} Y_{-n-k}] \\ &= E[Y_{m+k} Y_m], \text{ using } m = -n - k \\ &= E[Y_m Y_{m+k}] \\ &= R_y(k)\end{aligned}$$

$\Rightarrow Y_n$ and Z_n have the same distribution.

$\Rightarrow Y_n$ and Z_n have the same predictor.

$$\hat{z}_n = \sum_{i=1}^{\infty} h_i z_{n-i}$$

c) What is the autocorrelation of $x_n = y_n - \hat{y}_n$. Provide a precise justification for your answer.

Solution:

$$R_X(k) = \delta_k \sigma_x^2 \text{ (see proof in notes)}$$

d) Derive an expression for the power spectrum of the random process, y_n .

Solution:

$$S_x(e^{j\omega}) = \frac{\sigma_x^2}{|1 - H(e^{j\omega})|^2}$$

Spring 2010 Exam 1: Problem 1 (power spectrum and IIR filters)

Consider the following 2-D discrete-time linear system.

$$y(m, n) = x(m, n) + ay(m - 1, n) + by(m, n - 1) - aby(m - 1, n - 1)$$

where a and b are scalar constants.

a) Compute the transfer function $H(z_1, z_2)$ for the system.

Solution:

$$\begin{aligned} Y(z_1, z_2) (1 - az_1^{-1} - bz_2^{-1} + abz_1^{-1}z_2^{-1}) &= X(z_1, z_2) \\ H(z_1, z_2) &= \frac{Y(z_1, z_2)}{X(z_1, z_2)} = \frac{1}{(1 - az_1^{-1})(1 - bz_2^{-1})} \end{aligned}$$

b) Compute the impulse response $h(m, n)$ for the system.

Solution:

$$h(m, n) = (a^m u(m)) (b^n u(n))$$

c) For what values of a and b is the system stable.

Solution:

$$|a| < 1, \text{ and } |b| < 1$$

d) Compute $S_y(e^{j\mu}, e^{j\nu})$, the power spectrum of $y(m, n)$, when $x(m, n)$ is a set of i.i.d. $\mathcal{N}(0, \sigma^2)$ random variables.

Solution:

$$\begin{aligned} S_x(e^{j\mu}, e^{j\nu}) &= \sigma^2 \\ S_y(e^{j\mu}, e^{j\nu}) &= \sigma^2 |H(e^{j\mu}, e^{j\nu})|^2 \\ &= \frac{\sigma^2}{|1 - ae^{-j\mu}|^2 |1 - be^{-j\nu}|^2} \end{aligned}$$