#### Parametric Spectral Estimation

The most common model used in parametric spectral estimation is the rational function model used to describe ARMA systems

$$
H(z) = \frac{B(z)}{A(z)} = \frac{\sum_{k=0}^{q} b_k z^{-k}}{1 + \sum_{k=1}^{p} a_k z^{-k}}
$$

with the corresponding difference equation

$$
\mathbf{x}\left[n\right] = -\sum_{k=1}^{p} a_k \mathbf{x}\left[n-k\right] + \sum_{k=0}^{q} b_k \mathbf{w}\left[n-k\right]
$$

The power spectral density is

$$
G_{xx}\left(F\right) = G_{ww}\left(F\right)\left|H\left(F\right)\right|^2
$$

If the excitation w is a zero-mean white noise process then

$$
G_{xx}\left(F\right) = \sigma_{ww}^2 \left|H\left(F\right)\right|^2 = \sigma_{ww}^2 \left|\frac{B\left(e^{j2\pi F}\right)}{A\left(e^{j2\pi F}\right)}\right|^2
$$

The most common form of model is the AR model in which  $q = 0$  and  $b_0 = 1$  because it usually results in the fewest parameters needed to represent a process. In that case

$$
G_{xx}\left(F\right) = \frac{\sigma_{ww}^2}{\left|A\left(e^{j2\pi F}\right)\right|^2}
$$

The normal equations

$$
\mathbf{R}_{xx}\left[l\right] = -\sum_{k=1}^{p} a_p \left[k\right] \mathbf{R}_{xx} \left[l-k\right] , l = 1, 2, \cdots, p
$$

can be use to find the coefficients a.

Example

Let the autocorrelation of the response of a system to an applied white noise signal with variance  $\sigma_w^2$  = 6 have the following values

$$
R_{xx}\left[0\right] = 4 \ , \ R_{xx}\left[1\right] = 2 \ , \ R_{xx}\left[2\right] = 3
$$

Model the system producing this signal as an AR(2) system and graph the power spectral density of the signal.

The normal equations are

$$
\begin{bmatrix} R_{xx} \begin{bmatrix} 0 \end{bmatrix} & R_{xx} \begin{bmatrix} 1 \end{bmatrix} \begin{bmatrix} a_4 \begin{bmatrix} 1 \end{bmatrix} \end{bmatrix} = \begin{bmatrix} R_{xx} \begin{bmatrix} 1 \end{bmatrix} \end{bmatrix}
$$

$$
R_{xx} \begin{bmatrix} 1 \end{bmatrix} & R_{xx} \begin{bmatrix} 0 \end{bmatrix} \begin{bmatrix} a_4 \begin{bmatrix} 2 \end{bmatrix} \end{bmatrix} = \begin{bmatrix} R_{xx} \begin{bmatrix} 2 \end{bmatrix} \end{bmatrix}
$$

Example

$$
\begin{bmatrix} 4 & 2 \ 2 & 4 \end{bmatrix} \begin{bmatrix} a_4 \begin{bmatrix} 1 \end{bmatrix} \end{bmatrix} = \begin{bmatrix} 2 \ 3 \end{bmatrix} \Rightarrow \begin{bmatrix} a_4 \begin{bmatrix} 1 \end{bmatrix} \end{bmatrix} = \begin{bmatrix} -0.1667 \ -0.667 \end{bmatrix}
$$

$$
H(z) = \frac{1}{A(z)} = \frac{1}{1 - 0.1667z^{-1} - 667z^{-2}}
$$

The power spectral density is

$$
G_{xx}(F) = \frac{6}{\left|1 - 0.1667e^{-j2\pi F} - 0.667e^{-j4\pi F}\right|^2}
$$

$$
G_{xx}(F) = \frac{6}{1.473 - 0.111 \cos(2\pi F) - 1.333 \cos(4\pi F)}
$$

#### Example



#### MA Models

For an  $MA(q)$  system the transfer function is

$$
\mathbf{H}\left(z\right) = \mathbf{B}\left(z\right) = \sum_{k=0}^{q} b_k z^{-k}
$$

and the impulse response is  $h\left[n\right] = \sum b_k \delta\left[n - k\right]$ *k*=0 *q*  $\sum b_k \delta \big[ n-k \big]$ . So the

squared magnitude of the transfer function is

$$
H(z)H(z^{-1}) = B(z)B(z^{-1}) = D(z) = \sum_{k=-q}^{q} d_k z^{-k}
$$
  
where 
$$
\sum_{k=-q}^{q} d_k \delta[n-k] = h[n] * h[-n].
$$

#### MA Models

$$
(h[n] * h[-n])_{n=0} = b_0^2 + b_1^2 + \dots + b_q^2
$$
  
\n
$$
(h[n] * h[-n])_{n=1} = b_0b_1 + b_1b_2 + \dots + b_{q-1}b_q = (h[n] * h[-n])_{n=-1}
$$
  
\n
$$
(h[n] * h[-n])_{n=2} = b_0b_2 + b_1b_3 + \dots + b_{q-2}b_q = (h[n] * h[-n])_{n=-2}
$$
  
\n
$$
\vdots \qquad \vdots
$$
  
\n
$$
(h[n] * h[-n])_{n=m} = \sum_{k=0}^{q-|m|} b_kb_{k+m}, \ |m| \le q
$$

### MA Models

The autocorrelation of x is

$$
R_{xx} [m] = R_{ww} [m] * h[m] * h[-m] = \sigma_w^2 h[m] * h[-m]
$$
  
\n
$$
R_{xx} [m] = \sigma_w^2 \sum_{k=0}^{q-|m|} b_k b_{k+m} , |m| \le q
$$
  
\n
$$
R_{xx} [m] = \sigma_w^2 d_m
$$

Therefore

$$
G_{xx}\left(F\right) = \sum_{m=-\infty}^{\infty} R_{xx} \left[m\right] e^{-j2\pi F m} = \sum_{m=-q}^{q} R_{xx} \left[m\right] e^{-j2\pi F m}
$$