

Statistical signal processing 5CTA0

Estimation theory - linear signal models

Linear signal model

- Many signals can be modeled by linear signal models
- For this class of signal models, efficient estimators exist
- Signal model:

$$s(\theta) = \mathbf{H}\theta$$

\mathbf{H} is the so-called observation matrix

$$s_n(\theta) = \sum_{k=0}^{K-1} [H]_{n,k} \theta_k$$

Example I

- Polynomial of degree $K - 1$,

$$s_n(\boldsymbol{\theta}) = \theta_0 n^0 + \theta_1 n^1 + \cdots + \theta_{K-1} n^{K-1} \quad \text{for } 0 \leq n \leq N - 1$$

- Observation matrix:

$$\mathbf{H} = \begin{bmatrix} 0^0 & 0^1 & \cdots & 0^{K-1} \\ 1^0 & 1^1 & \cdots & 1^{K-1} \\ \vdots & \vdots & \ddots & \vdots \\ (N-1)^0 & \cdots & (N-1)^{K-1} \end{bmatrix}$$

Efficient estimator for linear models and additive Gaussian noise

- Observation model:

$$\mathbf{x} = \mathbf{H}\boldsymbol{\theta} + \mathbf{w} \quad \mathcal{N}(\mathbf{0}, \mathbf{C})$$

- Joint PDF:

$$p(\mathbf{x}, \boldsymbol{\theta}) = \frac{1}{|2\pi\mathbf{C}|^{\frac{1}{2}}} \exp\left(-\frac{1}{2}(\mathbf{x} - \mathbf{H}\boldsymbol{\theta})^T \mathbf{C}^{-1}(\mathbf{x} - \mathbf{H}\boldsymbol{\theta})\right)$$

- Efficient estimator:

$$g(\mathbf{x}) = (\mathbf{H}^T \mathbf{C}^{-1} \mathbf{H})^{-1} \mathbf{H}^T \mathbf{C}^{-1} \mathbf{x} = \cancel{\sigma^2} (\mathbf{H}^T \mathbf{I} \mathbf{H})^{-1} \mathbf{H}^T \cancel{\frac{1}{\sigma^2}} \mathbf{x}$$

$$\begin{bmatrix} \frac{1}{\sigma^2} & & \\ & \frac{1}{\sigma^2} & \\ & & \ddots \\ & & & \frac{1}{\sigma^2} \end{bmatrix} = \frac{1}{\sigma^2} \mathbf{I} = (\mathbf{H}^T \mathbf{H})^{-1} \mathbf{H}^T \mathbf{x}$$

Special case: white Gaussian noise

■ Covariance matrix:

■ Efficient estimator

$$C = \sigma^2 I = \begin{bmatrix} \sigma^2 & & \\ & \sigma^2 & \\ & & \ddots \\ & & & \sigma^2 \end{bmatrix}$$

$$g(\mathbf{x}) = (\mathbf{H}^T \mathbf{H})^{-1} \mathbf{H}^T \mathbf{x}$$

$$I(\theta) = \frac{\mathbf{H}^T \mathbf{H}}{\sigma^2} = \frac{N}{\sigma^2}$$

$$x_n = A + w_n$$

$$\underline{x} = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} A + \underline{w}$$

$$\sum_{n=0}^{N-1} 1 = N$$

Example II - Fourier analysis

- Signal composed of P sines and cosines with known frequency but unknown amplitudes
- Signal is embedded in white Gaussian noise
- Observation model

$$x_n = \sum_{p=0}^{P-1} a_p \cos \left(2\pi \frac{k_p}{N} n \right) + \sum_{p=0}^{P-1} b_p \sin \left(2\pi \frac{k_p}{N} n \right) + w_n$$

Example II - Fourier analysis



$$\mathbf{x} = \mathbf{H}\boldsymbol{\theta} + \mathbf{w}.$$

- Observation matrix

$$\mathbf{H} = [\mathbf{c}_{k_0} \quad \dots \quad \mathbf{c}_{k_{(P-1)}} \quad \mathbf{s}_{k_0} \quad \dots \quad \mathbf{s}_{k_{(P-1)}}]$$

$$\mathbf{c}_{k_p} = \left[\cos\left(2\pi \frac{k_p}{N} 0\right) \quad \cos\left(2\pi \frac{k_p}{N} 1\right) \quad \dots \quad \cos\left(2\pi \frac{k_p}{N} (N-1)\right) \right]^T$$

$$\mathbf{s}_{k_p} = \left[\sin\left(2\pi \frac{k_p}{N} 0\right) \quad \sin\left(2\pi \frac{k_p}{N} 1\right) \quad \dots \quad \sin\left(2\pi \frac{k_p}{N} (N-1)\right) \right]^T$$

- Parameter vector

$$\boldsymbol{\theta} = [a_0 \quad \dots \quad a_{(P-1)} \quad b_0 \dots b_{(P-1)}]^T$$

$$0 \leq k_p \leq N-1$$

$$\mathbf{H}^T \mathbf{H} = \frac{N}{2} \mathbf{I}$$

Example II - Fourier analysis

- Efficient estimator

$$\hat{\boldsymbol{\theta}} = \underbrace{(\mathbf{H}^T \mathbf{H})^{-1}}_{\frac{2}{N}} \mathbf{H}^T \mathbf{x} = \frac{2}{N} \begin{bmatrix} \mathbf{c}_{k_0}^T \\ \vdots \\ \mathbf{c}_{k_{(P-1)}}^T \\ \mathbf{s}_{k_0}^T \\ \vdots \\ \mathbf{s}_{k_{(P-1)}}^T \end{bmatrix} \mathbf{x}$$

$$\hat{a}_p = \frac{2}{N} \sum_{n=0}^{N-1} x_n \cos \left(2\pi \frac{k_p}{N} n \right) \quad \hat{b}_p = \frac{2}{N} \sum_{n=0}^{N-1} x_n \sin \left(2\pi \frac{k_p}{N} n \right)$$

Example II - Fourier analysis

- Signal composed of P cosines with known frequency but unknown amplitude and phase
- Observation model

$$x_n = \sum_{p=0}^{P-1} c_p \cos \left(2\pi \frac{k_p}{N} n - \varphi_p \right) + w_n$$

- Model is nonlinear in the unknown phase

Example II - Fourier analysis

■ Trigonometric identities

$$\cos(\alpha) \cos(\beta) = \frac{1}{2} (\cos(\alpha - \beta) + \cos(\alpha + \beta))$$

$$\sin(\alpha) \sin(\beta) = \frac{1}{2} (\cos(\alpha - \beta) - \cos(\alpha + \beta))$$

$$\cos(\alpha) \cos(\beta) + \sin(\alpha) \sin(\beta) = \cos(\alpha - \beta)$$

■ Transformed model

$$x_n = \sum_{p=0}^{P-1} \underbrace{c_p \cos(\varphi_p)}_{a_p} \cos\left(2\pi \frac{k_p}{N} n\right) + \underbrace{c_p \sin(\varphi_p)}_{b_p} \sin\left(2\pi \frac{k_p}{N} n\right) + w_n$$

$$c_p = \sqrt{a_p^2 + b_p^2} \quad \varphi_p = \arctan(b_p, a_p)$$