

EE 4715 Array Processing

4. Wideband Models

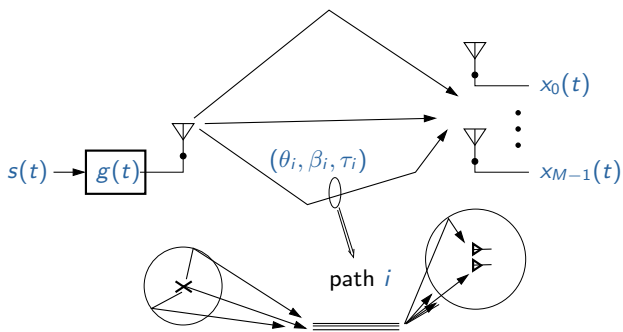
April 2022

Models

Wideband models are used in wireless communication, acoustics (audio, ultrasound, underwater, geophysics), and elsewhere. In this lecture we take a communications perspective.

- Physical channel
- Source model
- Receiver model

Multipath channel model



Jakes' multipath model

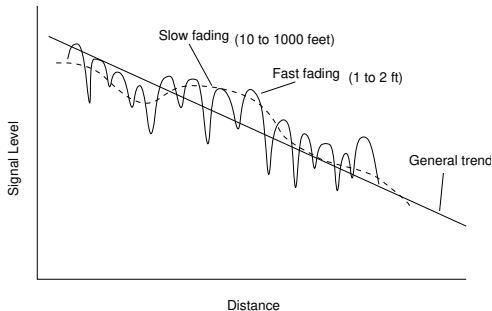
- scatterers local to mobile (nearby buildings, sensitive to motion)
- remote scatterers (remote buildings, hills: dominant scatterers)
- scatterers local to base (static effect)

Multipath channel model

$$\mathbf{x}(t) = \left[\sum_{i=1}^r \mathbf{a}(\theta_i) \beta_i g(t - \tau_i) \right] * s(t)$$

- $g(t)$ is a pulse shape function (where $s(t)$ contains dirac pulses)
- $\mathbf{a}(\theta)$ is the array response vector
- Scattering local to mobile causes fluctuations in path gains β_i :
 - Each ray has a (small) angular and delay spread: phase changes
 - Adding the rays in/out of phase creates major effects on the gains:
fading

Fading

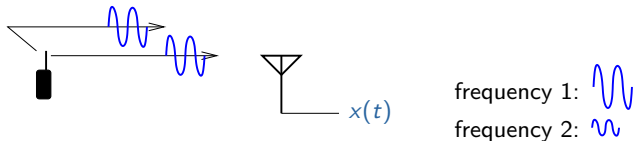


- **General trend:** $\approx 35 - 50$ dB / decade (path loss)
- **Slow fading:** Caused by shadowing.
- **Fast Fading:** Caused by local scatterers near mobile. Typically Rayleigh distributed

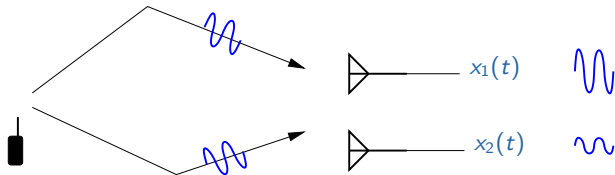
Fading

$\phi = \omega_0 T$: a phase can change due to changes in frequency or delay

- Time-selective fading (due to motion/Doppler)
- Frequency-selective fading (due to long delays)



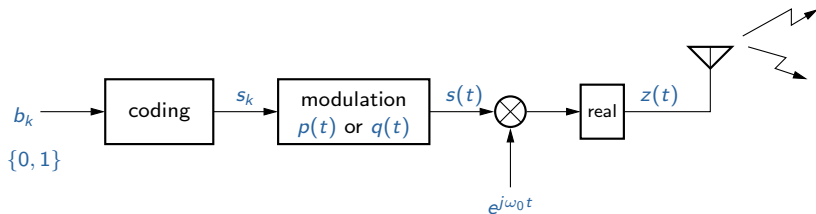
- Space-selective fading (large angles)



Typical channel parameter values

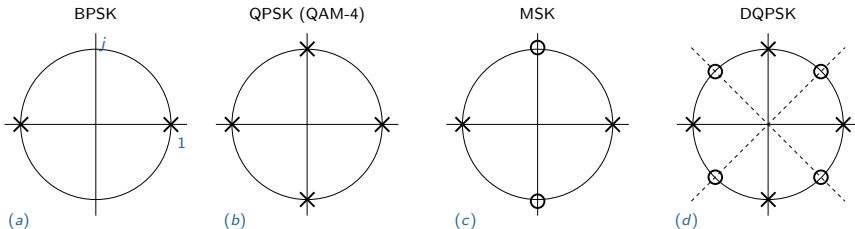
| Environment | delay spread | angle spread | Doppler spread |
|--------------------|-------------------|--------------|----------------|
| Flat rural (macro) | $0.5 \mu\text{s}$ | 1° | 190 Hz |
| Urban (macro) | $5 \mu\text{s}$ | 20° | 120 Hz |
| Hilly (macro) | $20 \mu\text{s}$ | 30° | 190 Hz |
| Mall (micro) | $0.3 \mu\text{s}$ | 120° | 10 Hz |
| Indoors (pico) | $0.1 \mu\text{s}$ | 360° | 5 Hz |

Signal modulation



- Digital alphabets
- Modulation
 - Linear modulation (e.g. with a raised-cosine pulse)
 - Phase modulation (not discussed here)

Digital constellations



| $b_k \in \{0, -1\}$ | $\rightarrow s_k$ chosen from (up to scaling): |
|---------------------|--|
| BPSK | $\{1, -1\}$ |
| QPSK (QAM-4) | $\{1, j, -1, -j\}$ |
| m -PSK | $\{1, e^{j2\pi/m}, e^{j2\pi 2/m}, \dots, e^{j2\pi(m-1)/m}\}$ |
| m -PAM | $\{\pm 1, \pm 3, \dots, \pm(2m-1)\}$ |
| m -PAM | $\{\pm 1, \pm 3, \dots, \pm(2m-1)\}$ |
| MSK | $\{1, -1\}, k \text{ even}; \{j, -j\}, k \text{ odd}$ |

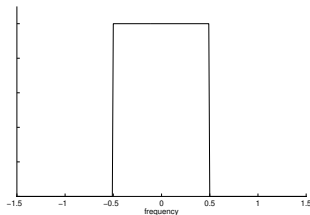
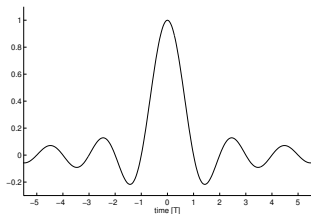
Linear modulation

- Amplitude is modulated (possibly complex):

$$s_{\delta}(t) = \sum_k s_k \delta(t - k) \quad s(t) = p(t) * s_{\delta}(t) = \sum_k s_k p(t - k)$$

- Optimum waveform is localized in time and frequency (contradiction)
- One possible choice: sinc pulse shape

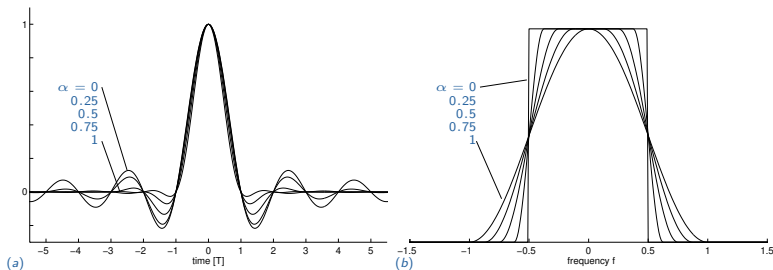
$$p(t) = \frac{\sin \pi t}{\pi t}, \quad P(f) = \begin{cases} 1, & |f| < \frac{1}{2} \\ 0, & \text{otherwise} \end{cases}$$



Linear modulation

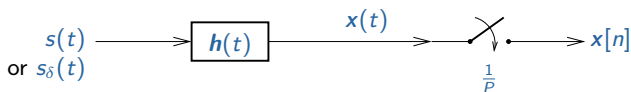
- Modification: *raised-cosine* pulseshapes

$$p(t) = \frac{\sin \pi t}{\pi t} \cdot \frac{\cos \alpha \pi t}{1 - 4\alpha^2 t^2}, \quad P(f) = \begin{cases} 1, & |f| < \frac{1}{2}(1 - \alpha) \\ \frac{1}{2} - \frac{1}{2} \sin\left(\frac{\pi}{\alpha}(|f| - \frac{1}{2})\right), & \frac{1}{2}(1 - \alpha) < |f| < \frac{1}{2}(1 + \alpha) \\ 0, & \text{otherwise} \end{cases}$$



α : excess bandwidth (or rolloff) parameter; common choice is $\alpha = 0.35$

FIR channel model



- We collect all temporal effects in $h(t)$
 - Filtering effects at the transmitter and the receiver
 - Propagation channel (array response, multipath delays)
 - Optional: pulse shape $g(t)$ for linear modulation
- Resulting data model (M antennas):

$$\mathbf{x}(t) = \mathbf{h}(t) * s(t), \quad \mathbf{x}(t) = \begin{bmatrix} x_1(t) \\ \vdots \\ x_M(t) \end{bmatrix}, \quad \mathbf{h}(t) = \begin{bmatrix} h_1(t) \\ \vdots \\ h_M(t) \end{bmatrix}$$

Oversampling

If the symbol rate is below the Nyquist rate, we can **oversample** beyond the symbol rate, by a factor P . Often, $P = 2$.

Define $P \times 1$ vectors spanning 1 symbol period:

$$\mathbf{x}[n] = \begin{bmatrix} x(n) \\ x(n + \frac{1}{P}) \\ \vdots \\ x(n + \frac{P-1}{P}) \end{bmatrix}, \quad \mathbf{h}[k] = \begin{bmatrix} h(k) \\ h(k + \frac{1}{P}) \\ \vdots \\ h(k + \frac{P-1}{P}) \end{bmatrix}$$

In the received signal, over 1 symbol period, the same L transmitted symbols play a role:

$$\mathbf{x}[n] = \mathbf{h}[n] * s[n] = \sum_{k=0}^{L-1} \mathbf{h}[k]s[n - k]$$

- This is the same as we had before, but now using sample vectors consisting of the P samples that fall within one sample period.

Matrix representation of convolution

Collect samples in a matrix $\mathbf{X} = [\mathbf{x}[0] \quad \mathbf{x}[1] \quad \cdots \quad \mathbf{x}[L-1]]$.

The convolution model $\mathbf{x}[n] = \mathbf{h}[n] * \mathbf{s}[n]$ gives

$$\mathbf{X} = \mathbf{H}\mathbf{S}$$

with

$$\mathbf{H} = [\mathbf{h}[0] \quad \mathbf{h}[1] \quad \cdots \quad \mathbf{h}[L-1]]$$

$$\mathbf{S} = \left[\begin{array}{ccc|ccc|ccc} s_0 & s_1 & \cdots & s_{L-1} & \cdots & s_{N_s-2} & s_{N_s-1} & & & \mathbf{0} \\ & s_0 & \cdots & \cdots & \cdots & \cdots & s_{N_s-2} & s_{N_s-1} & & \\ & & \ddots & s_1 & \cdots & \cdots & \cdots & \ddots & \ddots & \\ \mathbf{0} & & & s_0 & \cdots & \cdots & s_{N_s-L} & \cdots & s_{N_s-2} & s_{N_s-1} \end{array} \right]$$

Stacking

Create an augmented data matrix by stacking m samples of $\mathbf{x}[n]$:

$$\mathcal{X} = \begin{bmatrix} \mathbf{x}[0] & \mathbf{x}[1] & \ddots & \mathbf{x}[N-m] \\ \mathbf{x}[1] & \mathbf{x}[2] & \ddots & \vdots \\ \vdots & \vdots & \ddots & \mathbf{x}[N-2] \\ \mathbf{x}[m-1] & \vdots & \mathbf{x}[N-2] & \mathbf{x}[N-1] \end{bmatrix}.$$

Due to the convolution model we can factor \mathcal{X} as

$$\mathcal{X} = \mathcal{H}\mathcal{S} = \begin{bmatrix} \mathbf{0} & \boxed{H} \\ \vdots & \vdots \\ \boxed{H} & \vdots \\ \boxed{H} & \mathbf{0} \end{bmatrix} \begin{bmatrix} s_{m-1} & \ddots & s_{N-2} & s_{N-1} \\ \vdots & \ddots & \vdots & s_{N-2} \\ s_{-L+2} & s_{-L+3} & \ddots & \vdots \\ s_{-L+1} & s_{-L+2} & \ddots & s_{N-L-m+1} \end{bmatrix}$$

Space-time equalizer

A space-time equalizer is a vector \mathbf{w} which combines the rows of \mathcal{X} :

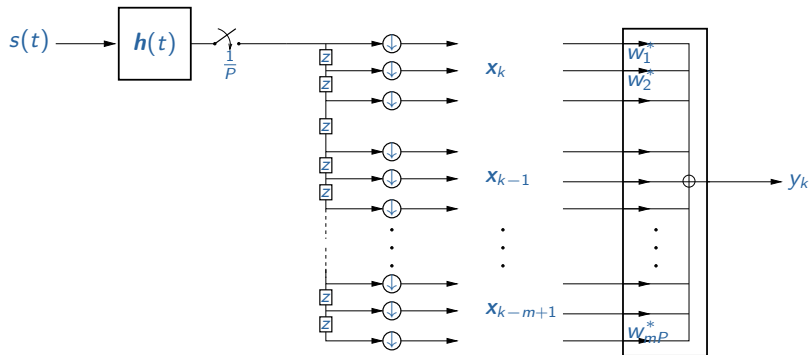
$$\mathbf{w}^H \mathcal{X} = [\hat{s}_{k_0} \hat{s}_{k_0+1} \cdots]$$

From the factorization $\mathcal{X} = \mathcal{H}\mathcal{S}$ we see that

$$\mathbf{w}^H \mathcal{H} = [0, \cdots, 0, 1, 0, \cdots, 0]$$

- \mathbf{w}^H is one row of a left inverse \mathcal{H}^\dagger of \mathcal{H} .
- We can choose which row of \mathcal{S} we reconstruct.

Space-time equalizer



Connection to the multiray model

The multiray propagation model is (for specular multipath)

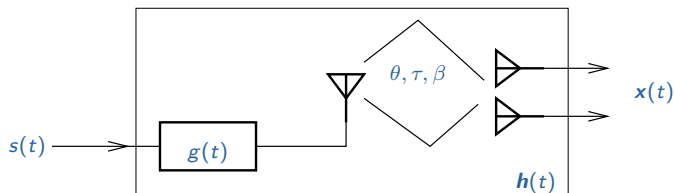
$$\mathbf{h}(t) = \begin{bmatrix} h_1(t) \\ \vdots \\ h_M(t) \end{bmatrix} = \sum_{i=1}^r \mathbf{a}(\theta_i) \beta_i g(t - \tau_i)$$

where $g(t)$: pulse shape function (e.g., raised-cosine),

θ_i : direction-of-arrival (DOA),

τ_i : path delay,

$\beta_i \in \mathbb{C}$: complex path attenuation (fading).



Connection to the multiray model

Collect the samples of $g(t)$ into a row vector

$$\mathbf{g} = [g(0) \quad g(\frac{1}{P}) \quad \cdots \quad g(L - \frac{1}{P})]$$

Similar for $g(t - \tau_i)$: gives $\mathbf{g}_i = [g(k - \tau_i)]_{k=0,1/P,\dots,L-1/P}$

The channel model can be written as

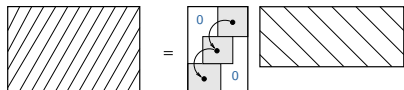
$$\mathbf{H}' = \begin{bmatrix} \text{---} \mathbf{h}_1 \text{---} \\ \vdots \\ \text{---} \mathbf{h}_M \text{---} \end{bmatrix} = \begin{bmatrix} \mathbf{a}_1 & \cdots & \mathbf{a}_r \end{bmatrix} \begin{bmatrix} \beta_1 & 0 \\ \cdots & \beta_r \\ 0 & \beta_r \end{bmatrix} \begin{bmatrix} \text{---} \mathbf{g}_1 \text{---} \\ \vdots \\ \text{---} \mathbf{g}_r \text{---} \end{bmatrix} = \mathbf{ABG}$$

where $\mathbf{a}_i = \mathbf{a}(\theta_i)$, and \mathbf{B} is diagonal (fading matrix)

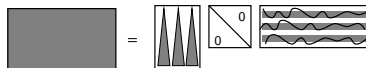
- \mathbf{H} and \mathbf{H}' are “the same”, but reorganized: \mathbf{H} is $MP \times L$ and \mathbf{H}' is $M \times LP$.

Summary of properties

$$\mathcal{X} = \mathcal{H}\mathcal{S}$$



$$H' = ABG$$



| Properties | | \mathcal{H} | \mathcal{S} |
|------------|------------------------|---|---|
| macro | matrix | block Hankel $\text{col}(\mathcal{H}) = \text{col}(\mathcal{X})$ | block Toeplitz $\text{row}(\mathcal{S}) = \text{row}(\mathcal{X})$ |
| | modulation temporal | cyclostationarity | constellation, non-Gaussian independence |
| parametric | temporal spatial | known $\mathbf{g}(\tau)$ known $\mathbf{a}(\theta)$ | training: known $\{s_k\}$ |