

ÉCOLE POLYTECHNIQUE FÉDÉRALE DE LAUSANNE

School of Computer and Communication Sciences

Handout 8
Homework 5

Advanced Digital Communications
October 16, 2009

PROBLEM 1. (Amplitude-limited functions) Sometimes it is important to generate baseband waveforms with bounded amplitude. This problem explores pulse shapes that can accomplish this

- (a) Find the Fourier transform of $g(t) = \text{sinc}^2(2Wt)$. Show that $g(t)$ is bandlimited to $f \leq W$ and sketch both $g(t)$ and $\hat{g}(f)$. [*Hint.* Recall that multiplication in the time domain corresponds to convolution in the frequency domain.]
- (b) Let $u(t)$ be a continuous real \mathcal{L}_2 function baseband-limited to $f \leq W$ (i.e. a function such that $u(t) = \sum_k u(kT)\text{sinc}(\frac{t}{T} - k)$, where $T = \frac{1}{2W}$). Let $v(t) = u(t) * g(t)$. Express $v(t)$ in terms of the samples $\{u(kT); k \in \mathcal{Z}\}$ of $u(t)$ and the shifts $\{g(t - kT); k \in \mathcal{Z}\}$ of $g(t)$. [*Hint.* Use your sketches in part (a) to evaluate $g(t) * \text{sinc}(\frac{t}{T})$.]
- (c) Show that if the T -spaced samples of $u(t)$ are nonnegative, then $v(t) \geq 0$ for all t .
- (d) Explain why $\sum_k \text{sinc}(\frac{t}{T} - k) = 1$ for all t .
- (e) Using (d), show that $\sum_k g(\frac{t}{T} - k) = c$ for all t and find the constant c . [*Hint.* Use the hint in (b) again.]
- (f) Now assume that $u(t)$, as defined in part (b), also satisfies $u(kT) \leq 1$ for all $k \in \mathcal{Z}$. Show that $v(t) \leq 2$ for all t .
- (g) Allow $u(t)$ to be complex now, with $|u(kT)| \leq 1$. Show that $v(t) \leq 2$ for all t .

PROBLEM 2. (Orthogonal sets) The function $\text{rect}(\frac{t}{T})$ has the very special property that it, plus its time and frequency shifts, by kT and $\frac{j}{T}$, respectively, form an orthogonal set. The function $\text{sinc}(\frac{t}{T})$ has this same property. We explore other functions that are generalizations of $\text{rect}(\frac{t}{T})$ and which, as you will show in parts (a)–(d), have this same interesting property. For simplicity, choose $T = 1$.

These functions take only the values 0 and 1 and are allowed to be nonzero only over $[-1, 1]$ rather than $[-\frac{1}{2}, \frac{1}{2}]$ as with $\text{rect}(\frac{t}{T})$. Explicitly, the functions considered here satisfy the following constraints:

$$p(t) = p^2(t) \quad \text{for all } t \quad (0/1 \text{ property}); \tag{1}$$

$$p(t) = 0 \quad \text{for } |t| > 1; \tag{2}$$

$$p(t) = p(-t) \quad \text{for all } t \quad (\text{symmetry}); \tag{3}$$

$$p(t) = 1 - p(t - 1) \quad \text{for } 0 \leq t \leq 1/2. \tag{4}$$

Note: because of property (3), condition (4) also holds for $\frac{1}{2} < t \leq 1$.

Note also that $p(t)$ at the single points $t = \pm\frac{1}{2}$ does not affect any orthogonality properties, so you are free to ignore these points in your arguments.

- (a) Show that $p(t)$ is orthogonal to $p(t - 1)$.

Hint. Evaluate $p(t)p(t - 1)$ for each $t \in [0, 1]$ other than $t = \frac{1}{2}$.

- (b) Show that $p(t)$ is orthogonal to $p(t - k)$ for all integer $k \neq 0$
- (c) Show that $p(t)$ is orthogonal to $p(t - k)e^{j2\pi mt}$ for integer $k \neq 0$ and $m \neq 0$.
- (d) Show that $p(t)$ is orthogonal to $p(t)e^{j2\pi mt}$ for integer $m \neq 0$.
Hint. Evaluate $p(t)e^{j2\pi mt} + p(t - 1)e^{j2\pi m(t-1)}$.
- (e) Let $h(t) = \hat{p}(t)$ where $\hat{p}(f)$ is the Fourier transform of $p(t)$. If $p(t)$ satisfies properties (1) – (4), does it follow that $h(t)$ has the property that it is orthogonal to $h(t - k)e^{j2\pi mt}$ whenever either the integer k or m is nonzero?

Note: almost no calculation is required in this problem.

PROBLEM 3. Consider estimating the real zero-mean scalar x from:

$$\mathbf{y} = \mathbf{h}x + \mathbf{w}$$

where $\mathbf{w} \sim \mathbf{N}(0, \frac{N_0}{2}\mathbf{I})$ is uncorrelated with x and \mathbf{h} is a fixed vector in \mathcal{R}^n .

- (a) Consider the scaled linear estimate $\mathbf{c}^t \mathbf{y}$ (with the normalization $\|\mathbf{c}\| = 1$):

$$\hat{x} = \mathbf{a} \mathbf{c}^t \mathbf{y} = (\mathbf{a} \mathbf{c}^t \mathbf{h})x + \mathbf{a} \mathbf{c}^t \mathbf{z} \quad (5)$$

Show that the constant a that minimizes the mean square error $(x - \hat{x})^2$ is equal to

$$\frac{\mathbb{E}[x^2]|\mathbf{c}^t \mathbf{h}|}{\mathbb{E}[x^2]|\mathbf{c}^t \mathbf{h}|^2 + \frac{N_0}{2}} \quad (6)$$

- (b) Calculate the minimal mean square error (denoted by MMSE) of the linear estimate in (5) (by using the value of a in (6)). Show that

$$\frac{\mathbb{E}[x^2]}{\text{MMSE}} = 1 + \text{SNR} = 1 + \frac{\mathbb{E}[x^2]|\mathbf{c}^t \mathbf{h}|^2}{\frac{N_0}{2}} \quad (7)$$

For every fixed linear estimator \mathbf{c} , this shows the relationship between the corresponding SNR and MMSE (of an appropriately scaled estimate).

- (c) In particular, relation (7) holds when we optimize over all \mathbf{c} leading to the best linear estimator. Find the value of vector \mathbf{c} (with the normalization $\|\mathbf{c}\| = 1$) by minimizing the MMSE derived in part (b). Compute optimal MMSE.

Hint. Use Cauchy–Schwarz inequality.

PROBLEM 4. (Linear Estimation) Consider the additive noise model given below,

$$Y_1 = X + Z_1 \quad (8)$$

$$Y_2 = X + Z_2. \quad (9)$$

Let $X, Y_1, Y_2, Z_1, Z_2 \in \mathcal{C}$, i.e. they are complex random variables. Moreover, assume X, Z_1 and Z_2 are zero mean and Z_1 and Z_2 are independent of X .

- (a) Assume the following: $\mathbb{E}[|X|^2] = \mathcal{E}_x$, $\mathbb{E}[|Z_1|^2] = \mathbb{E}[|Z_2|^2] = 1$ and $\mathbb{E}[Z_1 Z_2^*] = 0$. Given Y_1, Y_2 find the best minimum mean squared error linear estimator \hat{X} , where the optimization criterion is $\mathbb{E}[|X - \hat{X}|^2]$.
- (b) If $\mathbb{E}[Z_1 Z_2^*] = \frac{1}{\sqrt{2}}$, what is the best MMSE linear estimator of X ?
- (c) If $\mathbb{E}[Z_1 Z_2^*] = 1$, what is the best MMSE linear estimator of X ?