

# Homework "Signal theory: Part 1" solutions

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## 1 Introduction

### Homework

#### Reading assignment

- notes Leo part 1, sections 1.1–4
- If needed, follow this MATLAB tutorials.

## 2 Signals and systems

### Homework

#### Reading assignment

- section 1.1 (classification of signals), 1.2 (classification of systems), and chapter 2 (representation of signals and systems) from A. Oppenheim and A. Willsky, *Signals and Systems*, Prentice Hall, 1996

### Problems

- **Periodicity in discrete-time**

1. When is the signal  $f(t) = a \cos(\omega t + \phi)$ ,  $t \in \mathbb{Z}$  periodic?
2. Is  $a \cos(\omega t + \phi)$ ,  $t \in \mathbb{Z}$  periodic when  $\phi = 0$  and
  - (a)  $\omega = 2\pi/12$ ,
  - (b)  $\omega = 8\pi/31$ ,
  - (c)  $\omega = 1/6$ ?

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#### SOLUTION

1. The discrete-time signal  $f$  is periodic if there is a natural number  $T$  such that  $f(t) = f(t + T)$ , i.e.,

$$a \cos(\omega t + \phi) = a \cos(\omega(t + T) + \phi), \quad \text{for all } t \in \mathbb{Z}.$$

This is the case if  $\omega T = 2\pi k$  for some integer  $k$ . Therefore,  $\omega = 2\pi k/T$ , i.e., the discrete-time signal  $f(t) = a \cos(\omega t + \phi)$  is periodic if the frequency  $\omega$  is  $\pi$  times a rational number.

2.
  - (a) Yes, because  $\omega = 2\pi \times 1/12$ .
  - (b) Yes, because  $\omega = 2\pi \times 4/31$ .
  - (c) No, because  $\omega = 2\pi \times \frac{1}{12\pi}$  and  $\frac{1}{12\pi}$  is not a rational number.

- **Relation between impulse and step functions**

- Find relations between the impulse  $\delta$  and step  $s$  functions.

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SOLUTION

- \* in discrete-time

$$\delta(t) = s(t) - s(t-1), \quad s(t) = \delta(t) + \delta(t-1) + \dots = \sum_{\tau=-\infty}^t \delta(\tau)$$

- \* in continuous-time

$$\delta(t) = \frac{d}{dt}s(t), \quad s(t) = \int_{-\infty}^t \delta(\tau) d\tau$$

- **System classification**

- Give specific examples of:

- \* linear static system
- \* nonlinear static system
- \* linear time-invariant dynamical systems
  - finite impulse response (FIR)
  - infinite impulse response (IIR)
  - scalar
  - multivariable
- \* linear time-varying dynamical systems
- \* nonlinear time-invariant dynamical systems
- \* nonlinear time-varying dynamical systems

- A solution is given in this document.

SOLUTION

- **Peak and RMS values**

Find the peak and RMS values of  $x(t) := a \cos(\omega t + \phi)$ , for  $t \geq 0$ .

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SOLUTION

The peak value is  $\max_t |x(t)| = a$ .

The RMS value is by definition

$$\sqrt{\lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t x^2(\tau) d\tau}.$$

Since  $x$  is a periodic function, we need to compute the integral over one period  $[0, T]$

$$\sqrt{\frac{1}{T} \int_0^T x^2(\tau) d\tau}.$$

Using the identities  $2 \cos(t) = 1 + \cos(2t)$  and  $\int \cos(t) = \sin(t)$ , we have

$$\sqrt{\frac{1}{T} \int_0^T (a \cos(\omega t + \phi))^2 dt} = \sqrt{\frac{a^2}{2} + \frac{1}{2T} \sin(2\omega t + \phi) \Big|_0^T} = \frac{a}{\sqrt{2}}.$$

- **Response of 1st and 2nd order LTI system**

Find analytically the response of 1st and 2nd order linear time-invariant autonomous systems.

- Solution using Laplace transform  
See, this document.

SOLUTION

- Solution using a state space representation

SOLUTION

An autonomous linear time-invariant system has a state space representation  $\dot{x} = Ax$ ,  $y = Cx$ . In the 1st order case,  $x(t) \in \mathbb{R}$ ,  $A \in \mathbb{R}$ , and  $C \in \mathbb{R}$ . In the 2nd order case,  $x(t) \in \mathbb{R}^2$ ,  $A \in \mathbb{R}^{2 \times 2}$ , and  $C \in \mathbb{R}^{1 \times 2}$ . All responses  $y$  of the autonomous system can be parameterized by the initial state vector  $x(0)$  (initial condition) by  $y(t) = Ce^{At}x(0)$ , where  $e^{At}$  is the matrix exponential. The computation of the matrix exponential (and therefore a specific response of the system) can be done by a change of bases transformation that brings the system in a modal form (diagonal  $A$  matrix). See pages 8–12 in this document for details.

- **Multiple poles**

Consider the autonomous system represented by a difference equation

$$y(t+2) - 2ay(t+1) + a^2y(t) = 0.$$

(Its characteristic polynomial has both roots equal to  $\lambda = a$ .)

1. Show that both  $y(t) = a^t$  and  $y(t) = ta^t$  are solutions.
2. Find the trajectory generated from the initial conditions  $y(0) = 1$  and  $y(1) = 0$ .

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SOLUTION

1. To check  $y(t) = a^t$ , we have

$$a^{t+2} - 2a \cdot a^{t+1} + a^2 \cdot a^t = a^{t+2} - 2a^{t+2} + a^{t+2} = 0.$$

Thus,  $y(t) = a^t$  is a solution. Checking  $y(t) = ta^t$  we have

$$\begin{aligned} (t+2)a^{t+2} - 2a(t+1)a^{t+1} + a^2ta^t &= (t+2)a^{t+2} - 2(t+1)a^{t+2} + ta^{t+2} \\ &= (t+2-2t-2+1)a^{t+2} = 0. \end{aligned}$$

Thus,  $y(t) = ta^t$  is also a solution.

2. A second-order linear time-invariant autonomous system has a two dimensional solution set (two degrees of freedom in its general solution). This implies that two linearly independent solutions can form a basis set for the solutions space. We have found two solutions  $a^t$  and  $ta^t$ . It is easy to prove that these two solutions are linear independent. Because we can't find two constant  $c_1$  and  $c_2$  at least one of which is nonzero to satisfy

$$c_1a^t + c_2ta^t = 0$$

for all  $t = 0, 1, 2, \dots$ . Thus, any solution  $y(t)$  can be expressed as a linear combination of the basis set of solutions  $a^t$  and  $ta^t$ .

$$y(t) = c_1a^t + c_2ta^t,$$

where  $c_1$  and  $c_2$  are constant. Using the conditions  $y(0) = 1$  and  $y(1) = 0$ , we find  $c_1$  and  $c_2$

$$\begin{aligned} y(0) = c_1a^0 = 1 &\implies c_1 = 1 \\ y(1) = c_1a + c_2a = a + c_2a = 0 &\implies c_2 = -1. \end{aligned}$$

Thus, the trajectory is

$$y(t) = a^t - ta^t.$$

- $(A, B, C, D) \mapsto$  **impulse response**

Find the impulse response of the linear time-invariant system

$$\mathcal{B}(A, B, C, D) := \{ (u, y) \mid \text{there is } x, \text{ such that } \sigma x = Ax + Bu, y = Cx + Bu \}.$$

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SOLUTION

The impulse response  $h$  of a dynamical system is the response of the system under zero initial conditions and input that is a delta function. For a discrete-time linear time invariant system given by a state space represented, we have

$$\begin{aligned} h(0) &= Cx(0) + D\delta(0) = D & x(1) &= Ax(0) + B\delta(0) = B \\ h(1) &= Cx(1) + D\delta(1) = CB & x(2) &= Ax(1) + B\delta(1) = AB \\ h(2) &= Cx(2) + D\delta(2) = CAB & x(3) &= Ax(2) + B\delta(2) = A^2B \\ h(3) &= Cx(3) + D\delta(3) = CA^2B & x(4) &= Ax(3) + B\delta(3) = A^3B \\ &\vdots & &\vdots \\ h(t) &= Cx(t) + D\delta(t) = CA^{t-1}B & x(t+1) &= Ax(t) + B\delta(t) = A^tB \end{aligned}$$

- $(A, B, C, D) \mapsto$  **transfer function**

Find the transfer function of a linear time-invariant system given by a state space representation  $\mathcal{B}(A, B, C, D)$ .

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SOLUTION

Applying the Laplace transform  $\mathcal{L}$  on the state equation, we have

$$\mathcal{L}(\dot{x}) = \mathcal{L}(Ax + Bu) \implies sX = AX + BU \implies (sI - A)X = BU \implies X = (sI - A)^{-1}BU.$$

Then, using the output equation

$$\mathcal{L}(y) = \mathcal{L}(Cx + Du) \implies Y = CX + DU \implies Y = \underbrace{(C(sI - A)^{-1}B + D)}_H U.$$

Therefore, the transfer function of  $\mathcal{B}(A, B, C, D)$  is

$$H(s) = C(sI - A)^{-1}B + D.$$

### 3 Representations of LTI systems

#### Homework

#### Additional reading

Chapters 1 (behavioral models) and 4 (state-space representation) from

<http://wwwhome.math.utwente.nl/~poldermanjw/onderwijs/DISC/mathmod/book.pdf>

#### Problems

- **Matrix representation of the convolution operation**

Find a matrix representation of the discrete-time convolution operation.

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SOLUTION

The discrete-time convolution operation (with a kernel  $h$ ) is the map  $u \mapsto y$ , defined by

$$y(t) = \sum_{\tau=0}^t h(\tau)u(t-\tau).$$

Convolution is a linear operation, so that it has a matrix representation  $y = Hu$ , where the matrix  $H$  depends on the kernel  $h$ . In order to find explicitly the matrix  $H$ , we write the convolution formula as a system of linear equations:

$$\begin{aligned} y(0) &= h(0)u(0) \\ y(1) &= h(0)u(1) + h(1)u(0) \\ y(2) &= h(0)u(2) + h(1)u(1) + h(2)u(0) \\ &\vdots \\ y(t) &= h(0)u(t) + h(1)u(t-1) + \dots + h(t)u(0) \end{aligned}$$

This shows that  $H$  is low-triangular with equal elements on the main and sub diagonals:

$$H = \begin{bmatrix} h(0) & & & & \\ h(1) & h(0) & & & \\ h(2) & h(1) & h(0) & & \\ \vdots & \vdots & \ddots & \ddots & \\ h(t) & h(t-1) & \dots & h(1) & h(0) \end{bmatrix}.$$

- **Matrix representation of the discrete Fourier transform**

Find a matrix representation of the discrete Fourier transform.

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SOLUTION

Consider a finite sequence

$$u = (u(0), \dots, u(T-1)).$$

The discrete Fourier transform is the map  $u \mapsto U$ , defined by

$$U(k) = \sum_{\tau=0}^{T-1} w^{k\tau} u(\tau), \quad \text{for } k = 0, 1, \dots, T-1, \quad \text{where } w := e^{-i\frac{2\pi}{T}}.$$

The discrete Fourier transform is a linear transformation, so that it has a matrix representation  $U = Fu$ , where

$$F = \begin{bmatrix} w^0 & w^0 & \dots & w^0 \\ w^0 & w^1 & \dots & w^{T-1} \\ w^0 & w^2 & \dots & w^{2(T-1)} \\ \vdots & \vdots & & \vdots \\ w^0 & w^{T-1} & \dots & w^{(T-1)^2} \end{bmatrix}.$$

- **Prediction using a model**

(separate document "exercise autonomous models" with problems and solutions)

## 4 Stochastic models

### Homework

### Reading assignment

- notes Leo part 1 sections 1.5–1.8 and notes Leo part 2

## Problems

- **Wiener-Khintchine theorem**

For a discrete-time signal  $y$ , let

- $\phi_y := |F(y)|^2$ , where  $F(y)$  be a Fourier transform of  $y$ , and
- $r_y := \sum_{t=1}^T y(t)y(t-\tau)$ .

Show that  $\phi_y = F(r_y)$ .

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SOLUTION

The proof

$$\phi_y = F(y)F^*(y) = F(y)F(\text{rev}(y)) = F(y \star \text{rev}(y)) = F(r_y)$$

is based on the following properties of the Fourier transform

- \*  $F(y \star y) = F(y)F(y)$ ,
- \*  $F(\text{rev}(y)) = F^*(y)$ ,
- \*  $y \star \text{rev}(y) = r_y$ ,

Which are easy to show using the definitions of the Fourier transform and the convolution.

## 5 Least-squares estimation

### Homework

#### Reading assignment

- notes Leo part 3, sections 3.1–3.3

## Problems

- **Orthogonality principle for least-squares estimation**

Show that

1.  $\hat{x}$  being a least squares approximate solution of the system  $Ax = b$ , and
2.  $\hat{x}$  being such that  $b - A\hat{x}$  is orthogonal to the span of the columns of  $A$ ,

are equivalent. (This result is known as the orthogonality principle for least squares approximation.)

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SOLUTION

- \* Let  $\hat{x}$  be a least squares approximate solution of the system  $Ax = b$ . Assuming that  $A$  is full column rank,  $\hat{x}$  is unique and is given by  $\hat{x} = (A^\top A)^{-1}A^\top b$ . We have to show that

$$A^\top (b - A\hat{x}) = 0. \quad (*)$$

Indeed,

$$A^\top (b - A\hat{x}) = A^\top (I - A(A^\top A)^{-1}A^\top)b = (A^\top - A^\top) b = 0.$$

- \* Let  $\hat{x}$  being such that  $b - A\hat{x}$  is orthogonal to the span of the columns of  $A$ , i.e.,  $(*)$  holds. Then, assuming that  $A$  is full column rank,  $A^\top A$  is invertible and  $\hat{x} = (A^\top A)^{-1}A^\top b$ . This proves that  $\hat{x}$  is a least squares approximate solution of the system  $Ax = b$ .

- **Weighted least-squares approximate solution**

For a given positive definite matrix  $W \in \mathbb{R}^{m \times m}$ , define the weighted 2-norm

$$\|e\|_W = e^\top W e.$$

The weighted least-squares approximation problem is

$$\text{minimize over } \hat{x} \in \mathbb{R}^n \quad \|A\hat{x} - b\|_W. \quad (\text{WLS})$$

When does a solution exist and when is it unique? Under the assumptions of existence and uniqueness, derive a closed form expression for the least squares approximate solution.

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**SOLUTION**

Since  $W$  is a symmetric positive definite matrix, it has a factorization  $W = CC^\top$ , where  $C$  is an  $m \times m$  full rank matrix. We can re-write the weighted least-squares approximation problem as an equivalent standard least-squares approximation problem for a system of linear equations  $A'x = b'$ , where

$$A' = CA \quad \text{and} \quad b' = Cb.$$

At this point we can use existing results: 1) a solution always exists, 2) it is unique if and only if the matrix is full column rank (f.c.r.). Since  $C$  is full rank,  $A'$  is f.c.r. if and only if  $A$  is f.c.r. In this case the unique weighted least-squares approximate solution is

$$\hat{x} = (A^\top W A)^{-1} A^\top W b.$$