

SSB
**Signals and Systems in
Bioengineering**

Department of Bioengineering

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Chapter 1

Spaces

Exercise 1.0.1. *Demonstration of the Cauchy-Schwartz inequality*

If $\mathbf{x} = \mathbf{y} = 0$ the demonstration is trivial. Let us consider the following inequality

$$0 \leq \|x - \alpha y\|^2 = \|x\|^2 - 2\operatorname{Re}\langle \mathbf{x}, \alpha \mathbf{y} \rangle + |\alpha|^2 \|y\|^2 \quad (1.1)$$

e and let us compute the value of the scalar α that minimizes the norm $\|\mathbf{x} - \alpha \mathbf{y}\|^2$ by deriving with it with respect α ,

$$\frac{d\|\mathbf{x} - \alpha \mathbf{y}\|^2}{d\alpha} = -2\operatorname{Re}\langle \mathbf{x}, \mathbf{y} \rangle + 2\alpha \|y\|^2 = 0 \Rightarrow \quad (1.2)$$

$$\alpha = \frac{\operatorname{Re}\langle \mathbf{x}, \mathbf{y} \rangle}{\|y\|^2} \quad (1.3)$$

and by replacing

$$0 \leq \|x - \alpha y\|^2 = \|x\|^2 - 2\frac{\operatorname{Re}\langle \mathbf{x}, \alpha \mathbf{y} \rangle}{\|y\|^2} + \frac{\operatorname{Re}\langle \mathbf{x}, \alpha \mathbf{y} \rangle}{\|y\|^2} \quad (1.4)$$

$$= \|x\|^2 - \frac{\operatorname{Re}\langle \mathbf{x}, \alpha \mathbf{y} \rangle}{\|y\|^2} \Rightarrow \quad (1.5)$$

$$|\operatorname{Re}\langle \mathbf{x}, \mathbf{y} \rangle| \leq \|\mathbf{x}\| \|\mathbf{y}\|, \quad (1.6)$$

If $\mathbf{y} = \alpha \mathbf{x}$ then

$$\langle \mathbf{x}, \mathbf{y} \rangle = \langle \mathbf{x}, \alpha \mathbf{x} \rangle = \alpha \langle \mathbf{x}, \mathbf{x} \rangle = \alpha \|x\|^2. \quad (1.7)$$

Since $\|\mathbf{y}\| = |\alpha| \|x\| \Rightarrow \|\mathbf{x}\| = \|\mathbf{y}\| / \alpha$ he have

$$\langle \mathbf{x}, \mathbf{y} \rangle = \alpha \|x\| \frac{\|\mathbf{y}\|}{\alpha} = \|x\| \|\mathbf{y}\|, \quad (1.8)$$

q.e.d

Remark:

$$\begin{aligned}\|x - y\|^2 &= \langle \mathbf{x} - \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle \\ &= \langle \mathbf{x}, \mathbf{x} \rangle - \langle \mathbf{x}, \mathbf{y} \rangle - \langle \mathbf{y}, \mathbf{x} \rangle + \langle \mathbf{y}, \mathbf{y} \rangle \\ &= \|x\|^2 - 2\operatorname{Re}\langle \mathbf{x}, \mathbf{y} \rangle + \|y\|^2\end{aligned}$$

For a real space

$$\|x - y\|^2 = \|x\|^2 - 2\langle \mathbf{x}, \mathbf{y} \rangle + \|y\|^2 \quad (1.9)$$

From the Cauchy-Schwartz inequality it is possible to prove the triangular inequality,

$$\begin{aligned}\|\mathbf{x} + \mathbf{y}\|^2 &= \langle \mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y} \rangle \\ &= \langle \mathbf{x}, \mathbf{x} \rangle + 2\langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{y} \rangle \\ &\leq \langle \mathbf{x}, \mathbf{x} \rangle + 2\|\mathbf{x}\|\|\mathbf{y}\| + \langle \mathbf{y}, \mathbf{y} \rangle \\ &= (\|\mathbf{x}\| + \|\mathbf{y}\|)^2 \Rightarrow\end{aligned} \quad (1.10)$$

$$\|\mathbf{x} + \mathbf{y}\| \leq \|\mathbf{x}\| + \|\mathbf{y}\| \quad (1.11)$$

For a space of real functions defined in the interval $[a, b]$ the Cauchy-Schwartz inequality is the following:

$$\left[\int_a^b f(t)g(t)dt \right]^2 \leq \int_a^b f^2(t)dt \int_a^b g^2(t)dt \quad (1.12)$$

Exercise 1.0.2. (*Ex.2.1-1, pag. 121*)

Consider the vector $x = [1, 2, 3, 4, 5, 6]^T$. Calculate the matrix l_p , $d_p(\mathbf{x}, \mathbf{0})$ for $p = 1, 2, 4, 10, 100, \infty$. Comment on the fact that $d_p(\mathbf{x}, \mathbf{0}) \rightarrow \max(x_i)$ when $p \rightarrow \infty$.

Exercise 1.0.3. (*Ex.2.1-2, pag. 121*) (**Solved**)

Let \mathbf{X} be an arbitrary set. Show that the following function

$$d(x, y) = \begin{cases} 0, & x = y \\ 1, & x \neq y \end{cases} \quad (1.13)$$

is a metric function.

$d(x, y)$ is a metric function if and only if

i) $d(x, y) \geq 0, \forall x, y \in \mathbf{X}$, which is true,

ii) $d(x, y) = 0$ if and only if $x = y$, which is also true and finally

iii) the triangular inequality must hold. If $x = y = z$ the proof is trivial. If

$(x = y) \wedge (y \neq z)$ or $(x \neq y) \wedge (y = z)$ then $d(x, y) + d(y, z) = 1 + 0 = d(x, z)$.

If $x \neq y \neq z$, $d(x, y) + d(y, z) = 1 + 1 = 2 > 1 = d(x, z)$ q.e.d.

Exercise 1.0.4. (Ex.2.1-4, pag. 121)(**Solved**)
Triangular Inequality

- For $x, y \in \mathbb{R}$ the following inequality holds

$$|x + y| \leq |x| + |y|. \quad (1.14)$$

What is the condition for the equality?

Demonstration:

$$|x + y|^2 = x^2 + 2xy + y^2 \leq x^2 + 2|x||y| + y^2 = (|x| + |y|)^2 \Rightarrow |x + y| \leq |x| + |y| \text{ q.e.d.}$$

- Prove the triangular inequality for $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$.

$$\|\mathbf{x} + \mathbf{y}\| \leq \|\mathbf{x}\| + \|\mathbf{y}\|. \quad (1.15)$$

where $\|\cdot\|$ is the Euclidean norm.

Suggestion: Use the fact of $\sum_{i=1}^n x_i y_i \leq \|x\| \|y\|$ (Cauchy-Schwartz inequality.)

Demonstration:

$$\begin{aligned} (\|\mathbf{x} + \mathbf{y}\|)^2 &= \langle \mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y} \rangle = \langle \mathbf{x}, \mathbf{x} \rangle + 2\langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{y} \rangle \leq \\ &(\|\mathbf{x}\|)^2 + 2(\|\mathbf{x}\|)(\|\mathbf{y}\|) + (\|\mathbf{y}\|)^2 = ((\|\mathbf{x}\|) + (\|\mathbf{y}\|))^2 \Rightarrow (\|\mathbf{x} + \mathbf{y}\|) \leq \\ &(\|\mathbf{x}\|)(\|\mathbf{y}\|) \text{ q.e.d.} \end{aligned}$$

Exercise 1.0.5. (Ex.2.1-5, pag. 122)(**Solved**)
 Let (\mathbf{X}, d) be a metric space. Show that

$$g(x, y) = \frac{d(x, y)}{1 + d(x, y)} \quad (1.16)$$

is a metric function \mathbf{X} .

Prove:

First let us check if the properties of the metric function hold.

i) $g(x, y) = g(y, x)$: is true;

ii) $\frac{d}{1+d} \geq 0 \Rightarrow d \geq 1 + d \Rightarrow 0 \geq 1$, is true;

iii) $\frac{d(x, y)}{1 + d(x, y)} = 0 \Rightarrow \{d(x, y) = 0 \text{ ou } 1 + d(x, y) = \infty\} \Rightarrow x = y$ because

$d(x, y)$ is a metric function.

iv) $g(x, z) \leq g(x, y) + g(y, z)$:

Let us apply the function $g(x)$ to both sides of $d(x, z) \leq d(x, y) + d(y, z)$, and the result is $g(x, z) \leq g(d(x, y) + d(y, z))$. Now, it is enough to prove that $g(a + b) \leq g(a) + g(b)$:

$$\frac{a + b}{1 + a + b} \leq \frac{a}{1 + a} + \frac{b}{1 + b} = \frac{a + b + 2ab}{(1 + a)(1 + b)} = \frac{a + b + 2ab}{1 + a + b + ab}. \quad (1.17)$$

which is equivalent to,

$$\frac{(a+b)}{(1+a+b)} \leq \frac{(a+b)+2ab}{(1+a+b)+ab} \Rightarrow \text{para } n = a+b, c = 2ab \text{ e } 1+a+b,$$

$$\frac{n}{d} \leq \frac{n+2c}{d+c} \Rightarrow$$

$\Rightarrow nd+nc \leq nd+2cd \Rightarrow n \leq 2d \Rightarrow n \leq 2(n+1) \Rightarrow n \leq 2n+2 \Rightarrow n+2 \geq 0$,
which is always true. Therefore, ,

$$\frac{d(x, z)}{1+d(x, y)} \leq g(d(x, y) + d(y, z)) \leq g(d(x, y) + g(d(y, z))) \quad (1.18)$$

$$g(d(x, y) + g(d(y, z))) = \frac{d(x, y)}{1+d(x, y)} + \frac{d(y, z)}{1+d(y, z)} \quad (1.19)$$

$\Rightarrow \frac{d(x, y)}{1+d(x, y)}$ is a metric function (q.e.d).

Exercise 1.0.6. (Ex.2.1-8, pag. 122)

Show that in a the metric space (R^n, d_p) the function $d_p(\mathbf{x}, \mathbf{y})$ decreases with p , that is, $d_p(\mathbf{x}, \mathbf{y}) \geq d_q(\mathbf{x}, \mathbf{y})$ if $p \leq q$. Suggestion: Compute the derivative with respect to p and show it is less or equal to zero. Use the inequality of the a logarithmic sum. This inequality states that for non negative sequences a_1, a_2, \dots, a_n and b_1, b_2, \dots, b_n ,

$$\sum_{i=1}^n a_i \log(a_i/b_i) \geq \left[\sum_{i=1}^n a_i \right] \log \left[\frac{\sum_{i=1}^n a_i}{\sum_{i=1}^n b_i} \right] \quad (1.20)$$

Make $b_i = 1$ and $a_i = |x_i - y_i|^p$. Use also the fact of, for non negative sequences $\{\alpha_i\}$ such that $\sum_{i=1}^n \alpha_i = 1$ the maximum of

$$\sum_{i=1}^n \alpha_i \log \alpha_i \quad (1.21)$$

is zero.

Exercise 1.0.7. (Ex.2.1-20, pag. 123)(**Solved**)

Show that, if $\{x_n\}$ is a sequence where

$$d(x_{n+1}, x_n) < Cr^n \quad (1.22)$$

for $0 \leq r < 1$, than $\{x_n\}$ is a Cauchy sequence.

A sequence is a Cauchy sequence if for any $\epsilon > 0$, there is a $N > 0$ such that $d(x_n, x_m) < \epsilon$ for all $n, m > N$.

Let $n > m$, the distance

$$\begin{aligned}
 d(x_n, x_m) &< d(x_n, x_{n-1}) + d(x_{n-1}, x_m) \\
 &< d(x_n, x_{n-1}) + d(x_{n-1}, x_{n-2}) + \dots + d(x_{m+1}, x_m) \\
 &= Cr^{n-1} + Cr^{n-2} + \dots + Cr^{m+1} \\
 &= C(r^{n-1} + r^{n-2} + \dots + r^{m+1}) \\
 &= Cr^m \frac{1 - r^{n-m}}{1 - r} \\
 &< Cr^m / (1 - r) = \epsilon \Rightarrow \\
 N &= \text{ceil}(\log((1 - r)\epsilon / C) / \log(r))
 \end{aligned} \tag{1.23}$$

which means, when $n, m > N$, $d(x_n, x_m) < \epsilon$

Exercise 1.0.8. (Ex.2.1-21, pag. 123) (**Solved**)

Let $p_n = (x_n, y_n, z_n) \in R^3$. Show that if $\{p_n\}$ is a Cauchy sequence with the following metric

$$d(p_j, p_k) = \sqrt{(x_j - x_k)^2 + (y_j - y_k)^2 + (z_j - z_k)^2}, \tag{1.24}$$

then $\{x_n\}$, $\{y_n\}$ e $\{z_n\}$ are also Cauchy sequences with the following metric $d(x_j, x_k) = |x_j - x_k|$.

Demonstration:

A sequence is called Cauchy sequence if for every $\epsilon > 0$, there is a $N > 0$ such that $d(x_n, x_m) < \epsilon$ for any $m, n > N$.

$p_n = (x_n, y_n, z_n) \in R^3$ is a Cauchy sequence with the metric:

$$d(p_j, p_k) = \sqrt{(x_j - x_k)^2 + (y_j - y_k)^2 + (z_j - z_k)^2} \tag{1.25}$$

Therefore, for $j, k > N$,

$d(p_j, p_k) = \sqrt{\delta_x^2 + \delta_y^2 + \delta_z^2} \leq \epsilon \Rightarrow \delta_x^2 + \delta_y^2 + \delta_z^2 \leq \epsilon^2 \Rightarrow 0 \leq \delta_x^2 \leq \epsilon^2 - \delta_y^2 - \delta_z^2 \leq \epsilon^2 \Rightarrow |\delta_x| = |x_j - x_k| \leq \epsilon, \forall j, k > N$, which means, each sequence x_n, y_n and z_n are Cauchy sequences under the norm $d(x, y) = |x - y|$, q.e.d.

Exercise 1.0.9. (Ex.2.3-30, pag. 124)

Let S the set of all solutions of the following differential equation defined in $C^3[0, \infty]$

$$\frac{d^3x}{dt^3} + b \frac{d^2x}{dt^2} + c \frac{dx}{dt} + dx = 0. \tag{1.26}$$

Show that S is a linear subspace of $C^3[0, \infty]$.

Exercise 1.0.10. (Ex.2.3-33, pag. 124) (**Solved**)
Show that in a normed linear space

$$|\|x\| - \|y\|| \leq \|x - y\|. \quad (1.27)$$

If $x = y$ the demonstration is trivial. Let $\|x\| > \|y\|$. The goal is to prove that

$$\begin{aligned} \|x\| - \|y\| &< \|x - y\| \Rightarrow \\ \|x\| &< \|y\| + \|x - y\| \end{aligned} \quad (1.28)$$

and making $e = x - y \Rightarrow x = y + e$ we obtain

$$\|e + y\| < \|y\| + \|e\| \quad (1.29)$$

which is true because results directly from the triangular inequality. If $\|x\| < \|y\|$ the demonstration is similar.

Exercise 1.0.11. (Ex.2.3-34, pag. 124) (**Solved**)
Prove the convexity of the norm function.

Demonstration:

By definition, a scalar function $f : X \times X \rightarrow R$ is a convex function if and only if $f(\alpha x + (1 - \alpha)y) \leq \alpha f(x) + (1 - \alpha)f(y)$ with $0 \leq \alpha \leq 1$.

The norm is $\|\alpha x + (1 - \alpha)y\| \leq \|\alpha x\| + \|(1 - \alpha)y\| = |\alpha|\|x\| + |1 - \alpha|\|y\| = \alpha\|x\| + (1 - \alpha)\|y\|$ q.e.d.

Exercise 1.0.12. (Ex.2.4-40, pag. 124)

Compute the inner product $\langle \mathbf{f}, \mathbf{g} \rangle$ using the definition

$$\langle \mathbf{f}, \mathbf{g} \rangle = \int_0^1 f(t)g(t)dt. \quad (1.30)$$

1. $f(t) = t^2 + 2t$, $g(t) = t + 1$.
2. $f(t) = e^{-t}$, $g(t) = t + 1$.
3. $f(t) = \cos(2\pi t)$, $g(t) = \sin(2\pi t)$.

Exercise 1.0.13. (Ex.2.4-41, pag. 125)

Compute the inner product based on the Eucliden norm using a vector operation,

$$1. x = [1, 2, -3, 4]^T, y = [2, 3, 4, 1]^T.$$

$$2. x = [2, 3]^T, y = [1, -2]^T.$$

Exercise 1.0.14. (Ex.2.5-43, pag. 125) (**Solved**)

Show that for an induced norm $\|\cdot\|$ over a real vector space:

1. The parallelogram law is true:

$$\|x + y\|^2 + \|x - y\|^2 = 2\|x\|^2 + 2\|y\|^2. \quad (1.31)$$

Illustrate graphically the parallelogram law.

Demonstration:

$$\|x + y\|^2 = \langle \mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y} \rangle = \langle \mathbf{x}, \mathbf{x} \rangle + 2\langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{y} \rangle = \|x\|^2 + 2\langle \mathbf{x}, \mathbf{y} \rangle + \|y\|^2 \text{ and}$$

$$\|x - y\|^2 = \langle \mathbf{x} - \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle = \langle \mathbf{x}, \mathbf{x} \rangle - 2\langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{y} \rangle = \|x\|^2 - 2\langle \mathbf{x}, \mathbf{y} \rangle + \|y\|^2.$$

Summing both previous equations, leads to:

$$\|x + y\|^2 + \|x - y\|^2 = 2\|x\|^2 + 2\|y\|^2 \quad (1.32)$$

q.e.d.

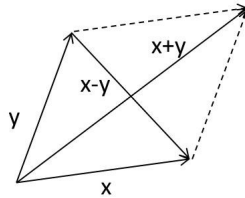


Figure 1.1: Parallelogram law.

2. Show that

$$\langle \mathbf{x}, \mathbf{y} \rangle = \frac{\|x + y\|^2 - \|x - y\|^2}{4}, \quad (1.33)$$

the polarization identity.

Demonstration:

$$\|x + y\|^2 - \|x - y\|^2 = 4\langle \mathbf{x}, \mathbf{y} \rangle \Rightarrow \langle \mathbf{x}, \mathbf{y} \rangle = \frac{\|x + y\|^2 - \|x - y\|^2}{4} \quad (1.34)$$

q.e.d.

Exercise 1.0.15. (Ex.2.6-44, pag. 125)

For the following inner product operation

$$\langle \mathbf{f}, \mathbf{g} \rangle = \int_0^1 f(t)g(t)dt \quad (1.35)$$

show that the Cauchy-Schwartz inequality is true for

1. (Solved) $f(t) = e^{-t}$, $g(t) = t + 1$.

Demonstration:

$$\|f\| = \int_0^1 \varepsilon^{-2t} dt = -\frac{1}{2}(\varepsilon^{-2t})\Big|_0^1 = -\frac{1}{2}(\varepsilon^{-2} - 1) = \frac{1}{2}(1 - \varepsilon^{-2})$$

$$\|g\| = \int_0^1 (t+1)^2 dt = \int_0^1 (t^2 + 2t + 1) dt = \left(\frac{t^3}{3} + t^2 + t\right)\Big|_0^1 = \frac{1}{3} + 1 + 1 = \frac{7}{3}$$

$$\langle \mathbf{f}, \mathbf{g} \rangle = \int_0^1 (t+1)\varepsilon^{-t} dt = \int_0^1 (t\varepsilon^{-t} + \varepsilon^{-t}) dt = -t\varepsilon^{-t}\Big|_0^1 + \int_0^1 \varepsilon^{-t} + \int_0^1 \varepsilon^{-t} = -\varepsilon^{-1} + 2 = 2 - \varepsilon^{-1}$$

$$\|f\|\|g\| = \frac{7}{6}(2 - \varepsilon^{-2}) = 1.0088$$

$$|\langle \mathbf{f}, \mathbf{g} \rangle| = |2 - \varepsilon^{-1}| = 0.8964 \Rightarrow \|f\|\|g\| > |\langle \mathbf{f}, \mathbf{g} \rangle|$$

q.e.d

2. $f(t) = e^{-t}$, $g(t) = -5e^{-t}$

Exercise 1.0.16. (Ex.2.6-45, pag. 125) (**Solved**)

Show that the following inequality is true:

$$-1 \leq \frac{\langle \mathbf{x}, \mathbf{y} \rangle}{\|\mathbf{x}\|_2 \|\mathbf{y}\|_2} \leq 1 \quad (1.36)$$

Demonstration:

Cauchy-Schwartz inequality: $\|\mathbf{x}\|\|\mathbf{y}\| \geq |\langle \mathbf{x}, \mathbf{y} \rangle|$

$$\text{Se } \langle \mathbf{x}, \mathbf{y} \rangle > 0 \Rightarrow \frac{\langle \mathbf{x}, \mathbf{y} \rangle}{\|\mathbf{x}\|\|\mathbf{y}\|} \leq 1$$

$$\text{Se } \langle \mathbf{x}, \mathbf{y} \rangle < 0 \Rightarrow -\frac{\langle \mathbf{x}, \mathbf{y} \rangle}{\|\mathbf{x}\|\|\mathbf{y}\|} \leq 1 \Rightarrow \frac{\langle \mathbf{x}, \mathbf{y} \rangle}{\|\mathbf{x}\|\|\mathbf{y}\|} \geq -1 \Rightarrow -1 \leq \frac{\langle \mathbf{x}, \mathbf{y} \rangle}{\|\mathbf{x}\|\|\mathbf{y}\|} \leq 1$$

Exercise 1.0.17. (Ex.2.7-47, pag. 125)

Let $x_1(t) = 3t^2 - 1$, $x_2(t) = 5t^3 - 3t$, $x_3(t) = 2t^2 - t$ and $\langle \mathbf{f}, \mathbf{g} \rangle = \int_{-1}^1 f(t)g(t)dt$. Compute the angle of each pairwise combination of these functions and identify the pairs that are orthogonal.

Exercise 1.0.18. (Ex.2.7-48, pag. 125)(Solved)

Let

$$\begin{aligned}x_1 &= [1, 2, 4, -2]^T \\x_2 &= [5, -2, -3, 1]^T \\x_3 &= [1, 2, 1, 2]^T.\end{aligned}$$

Compute the angle between these vectors using the Euclidean inner product and identify the pairs that are orthogonal.

$$\begin{aligned}\|x_1\| &= \sqrt{\langle x_1, x_1 \rangle} = \sqrt{1 + 4 + 16 + 4} = \sqrt{25} = 5 \\ \|x_2\| &= \sqrt{\langle x_2, x_2 \rangle} = \sqrt{25 + 4 + 9 + 1} = \sqrt{39} \\ \|x_3\| &= \sqrt{\langle x_3, x_3 \rangle} = \sqrt{1 + 4 + 1 + 4} = \sqrt{10}\end{aligned}$$

$$\cos(\theta_{1,2}) = \frac{\langle x_1, x_2 \rangle}{\|x_1\| \|x_2\|} = \frac{5 - 4 - 12 - 2}{5\sqrt{39}} \rightarrow \theta_{1,2} = \arccos\left(\frac{-23}{5\sqrt{39}}\right) = 137,4417$$

$$\cos(\theta_{1,3}) = \frac{\langle x_1, x_3 \rangle}{\|x_1\| \|x_3\|} = \frac{1 + 4 + 4 - 4}{5\sqrt{10}} = \frac{1}{\sqrt{10}} \rightarrow \theta_{1,3} = 71,5651$$

$$\cos(\theta_{2,3}) = \frac{\langle x_2, x_3 \rangle}{\|x_2\| \|x_3\|} = \frac{5 - 4 - 3 + 2}{\sqrt{390}} = \frac{0}{\sqrt{390}} \rightarrow \theta_{2,3} = \pi/2$$

The vectors x_2 e x_3 are orthogonal.

Exercise 1.0.19. Show that $f(t) = \sin(t)$ and $g(t) = \cos(t)$ are orthogonal in $L^2(-\pi, \pi)$.

Exercise 1.0.20. (solved)

Let V the space spanned by the following functions $p_1(x) = \cos(x)/\sqrt{\pi}$ and $p_2(x) = \sin(x)/\sqrt{\pi}$ in $L^2(-\pi, \pi)$.

1. Show that $\{p_1, p_2\}$ is an orthonormal basis .

To be an orthonormal basis its elements need to be pairwise orthogonal

with unit norm, that is, $\langle e_i, e_j \rangle = \begin{cases} 1 & i = j \\ 0 & i \neq j \end{cases}$

$\langle e_1, e_2 \rangle = \langle e_2, e_1 \rangle = 0$, como mostrado no exercicio anterior.

$$\begin{aligned}\langle e_1, e_1 \rangle &= \frac{1}{\pi} \int_{-\pi}^{\pi} \cos^2(t) dt = \frac{1}{2\pi} \int_{-\pi}^{\pi} (1 + \cos(2t)) dt = \frac{1}{2\pi} \int_{-\pi}^{\pi} 1 dt + \\ &\frac{1}{4\pi} \int_{-\pi}^{\pi} 2\cos(2t) dt = \frac{2\pi}{2\pi} + \frac{1}{4\pi} \sin(2t) \Big|_{-\pi}^{\pi} = 1 + 0 = 1\end{aligned}$$

$$\begin{aligned}\langle e_2, e_2 \rangle &= \frac{1}{\pi} \int_{-\pi}^{\pi} \sin^2(t) dt = \frac{1}{2\pi} \int_{-\pi}^{\pi} (1 - \cos(2t)) dt = \frac{1}{2\pi} \int_{-\pi}^{\pi} 1 dt - \\ &\frac{1}{4\pi} \int_{-\pi}^{\pi} 2\cos(2t) dt = \frac{2\pi}{4\pi} - \frac{1}{4\pi} \sin(2t) \Big|_{-\pi}^{\pi} = 1 - 0 = 1\end{aligned}$$

2. Compute the projection of $f(x) = x$ into V .

$$f_0 = \langle f, e_1 \rangle e_1 + \langle f, e_2 \rangle e_2$$

$$\begin{aligned} \langle f, e_1 \rangle &= \frac{1}{\sqrt{\pi}} \int_{-\pi}^{\pi} f(x) \cos(x) dx = \frac{1}{\sqrt{\pi}} \int_{-\pi}^{\pi} x \cos(x) dx \\ &= \frac{1}{\sqrt{\pi}} \left[x \sin(x) \Big|_{-\pi}^{\pi} - \int_{-\pi}^{\pi} \sin(x) dx \right] = 0 - \frac{1}{\sqrt{\pi}} \cos(x) \Big|_{-\pi}^{\pi} = 0 \\ \langle f, e_2 \rangle &= \frac{1}{\sqrt{\pi}} \int_{-\pi}^{\pi} f(x) \sin(x) dx = \frac{1}{\sqrt{\pi}} \int_{-\pi}^{\pi} x \sin(x) dx \\ &= \frac{1}{\sqrt{\pi}} \left[-x \cos(x) \Big|_{-\pi}^{\pi} + \int_{-\pi}^{\pi} \cos(x) dx \right] = \\ &= \frac{1}{\sqrt{\pi}} \left[2\pi + \sin(x) \Big|_{-\pi}^{\pi} \right] = \frac{2\pi}{\sqrt{\pi}} = 2\sqrt{\pi} \\ f_0 &= \langle f, e_2 \rangle e_2 = 2\sqrt{\pi} \frac{\sin(x)}{\sqrt{\pi}} = 2\sin(x) \end{aligned}$$

3. Consider W the space spanned by $\phi(x) = 1$ and $\psi = \begin{cases} 1, & 0 \leq x \leq 1/2; \\ -1, & 1/2 \leq x \leq 1. \end{cases}$ in the interval $[0, 1]$. The functions $\phi(x)$ and $\psi(x)$ are called scale and wavelet functions respectively.

(a) Show that these functions are orthogonal in $L^2(0, 1)$.

$$\begin{aligned} \langle \phi, \psi \rangle &= \int_0^1 \phi(x) \psi(x) dx = \int_0^{1/2} 1 dx - \int_{1/2}^1 1 dx = x \Big|_0^{1/2} - x \Big|_{1/2}^1 \\ &= 1/2 - (1 - 1/2) = 0 \end{aligned}$$

(b) Compute the projection of $f(x) = x$ in W .

$$\begin{aligned} \langle f, \phi \rangle &= \int_0^1 f(x) dx = \int_0^1 x dx = \frac{x^2}{2} \Big|_0^1 = 1/2 \\ \langle f, \psi \rangle &= \int_0^{1/2} x dx - \int_{1/2}^1 x dx = \frac{x^2}{2} \Big|_0^{1/2} - \frac{x^2}{2} \Big|_{1/2}^1 = \frac{1}{8} - \left[\frac{1}{2} - \frac{1}{8} \right] = -\frac{1}{4} \\ f_1(x) &= \langle f, \phi \rangle \phi(x) + \langle f, \psi \rangle \psi(x) = 1/2 \phi(x) - 1/4 \psi(x) = \\ &\begin{cases} 1/4 & 0 \leq x < 1/2 \\ 3/4 & 1/2 \leq x < 1 \end{cases} \end{aligned}$$

Exercise 1.0.21. (Ex. 2.7-49, pag. 126)

Show that a set of nonzero vectors $\{p_1, p_2, \dots, p_m\}$ that are mutually orthogonal, so that, $\langle p_i, p_j \rangle = 0, i \neq j$, is linearly independent. (Orthogonality implies linear independence.)

Exercise 1.0.22. (Ex.2.12-57, pag. 126)

Show that if V and W are subspaces of a vector space S then their intersection, $V \cap W$, is also a subspace.

Exercise 1.0.23. (Ex.2.12-64, pag. 127)(**Solved**)

If v is a vector, show that the matrix which projects onto $\text{span}(v)$ is

$$P_v = \frac{vv^H}{v^Hv}. \quad (1.37)$$

Demonstration: By the orthogonality principle

$$\begin{cases} \langle \mathbf{x} - \alpha \mathbf{v}, \mathbf{v} \rangle = 0 \\ \mathbf{y} = \alpha \mathbf{v} \end{cases} \quad \begin{cases} v^T x = \alpha v^T v \Rightarrow \alpha = \frac{v^T x}{v^T v} \\ \mathbf{y} = v \frac{v^T x}{v^T v} = \frac{v^T x}{v^T v} \mathbf{x} = P_v x \end{cases}$$

Exercise 1.0.24. (Ex.2.13-68, pag. 127)(**Solved**)

Let A be a matrix which can be factored as

$$A = U\Sigma V^H$$

where $U^H U = I$, $V^H V = I$. Σ is a real diagonal matrix. This factorization is called single value decomposition (SVD).

Let $A = [p_1, p_2, \dots, p_n]$ and $V = \text{span}(p_1, p_2, \dots, p_n)$ and the inner product $\langle \mathbf{x}, \mathbf{y} \rangle = \mathbf{y}^T \mathbf{x}$, between $v = P_A x$ where $P_A = A(A^H A)^{-1} A^H$. Show that $P_A = P_U$.

Demonstration:

$$\begin{aligned} P_A &= (U\Sigma V^H)((U\Sigma V^H)^H(U\Sigma V^H))^{-1}(U\Sigma V^H)^H = \\ &= (U\Sigma V^H)(V\Sigma^H U^H U\Sigma V^H)^{-1}(V\Sigma^H U^H) = \\ &= (U\Sigma V^H)(V\Sigma^H \Sigma V^H)^{-1}(V\Sigma^H U^H) = \\ &= (U\Sigma V^H)(V(\Sigma^H \Sigma)^{-1} V^H)(V\Sigma^H U^H) = \\ &= U\Sigma(\Sigma^H \Sigma)^{-1} \Sigma^H U^H = U\Sigma \Sigma^{-1} \Sigma^{-H} \Sigma^H U^H = I \\ P_U &= U(U^H U)^{-1} U^H = I \end{aligned}$$

q.e.d.

Exercise 1.0.25. (Ex.3.8-3, pag. 216)

Consider the following data sets $x = \{2, 2.5, 3, 5, 9\}$ $e y = \{-4.2, -5, 2, 1, 24.3\}$.

1. Display graphically these data sets.
2. Compute the linear regression parameters ($y = ax + b$) that better fits the data under the least square (LS) criterion.

3. Assuming that the first and last observations are more accurate formulate the estimation process under the weight LS approach, by specifying the weight matrix. Compute the new regression parameters, a and b .

Exercise 1.0.26. (Ex.3.8-(4-7), pag. 216)

Formulate the regression problem for the following models:

$$\begin{aligned}y_i &= a_0 + a_1x_i + a_2x_i^2 \\z_i &= ax_i + by_i + c \\y_i &= ce^{ax_i} \\y_i &= ax_i^b\end{aligned}$$

Exercise 1.0.27. (Ex.3.8-10, pag. 217)

Consider the following definition of inner product of matrices

$$\langle \mathbf{X}, \mathbf{Y} \rangle = \text{trace}(\mathbf{X}\mathbf{Y}^H).$$

The goal is to approximate the matrix \mathbf{Y} with the following linear combination,

$$\mathbf{Y} = c_1\mathbf{X}_1 + c_2\mathbf{X}_2 + \dots + c_m\mathbf{X}_m + \mathbf{E}.$$

Using the orthogonality principle, determine a set of normal equations that can be used to find $\{c_1, c_2, \dots, c_m\}$ that minimize the induced norm of \mathbf{E} .

Exercise 1.0.28. (Ex.3.9-12, pag. 217)

For the data sequence, $\{1, 1, 2, 3, 5, 8, 13\}$,

1. Write the data matrix and the Grammian $\mathbf{A}^H\mathbf{A}$ using (i) covariance, and (ii) the autocorrelation methods. Assume $m = 2$.
2. This sequence will be used to train a linear predictor. The desired signal $d(t)$ is the value of $x(t)$, and the data used are the two prior samples,

$$x(t) = a_1x(t-1) + a_2x(t-2) + e(t),$$

where $e(t)$ is the prediction error. Determine the least squares coefficients for the predictor using the covariance and autocorrelation methods.

3. Determine the minimum least-squares error for both methods.

Exercise 1.0.29. (Ex.3.15-30, pag. 223)(**Solved**)

Using the projection theorem, solve the finite dimensional problem

$$\hat{\mathbf{x}} = \arg \min_x \mathbf{x}^H Q \mathbf{x}$$

subject to $A\mathbf{x} = \mathbf{b}$

where $\mathbf{x} \in C^n$, Q is a positive-definite symmetric matrix, and A is an $m \times n$ matrix with $m < n$.

Demonstration:

$$A\mathbf{x} = \mathbf{b} \Rightarrow \begin{bmatrix} \mathbf{y}_1^H \\ \mathbf{y}_2^H \\ \vdots \\ \mathbf{y}_m^H \end{bmatrix} \mathbf{x} = \mathbf{b} \Rightarrow \begin{cases} \langle \mathbf{x}, \mathbf{y}_1 \rangle = b_1 \\ \langle \mathbf{x}, \mathbf{y}_2 \rangle = b_2 \\ \vdots \\ \langle \mathbf{x}, \mathbf{y}_m \rangle = b_m \end{cases}$$

where \mathbf{y}_i^H are the lines of A , $\hat{\mathbf{x}} = [\mathbf{y}_1, \mathbf{y}_2, \dots, \mathbf{y}_m] \mathbf{c} = A^H \mathbf{c}$ and it is assumed that the solution belongs to the subspace spanned by the vectors \mathbf{y}_i , that is, $\hat{\mathbf{x}} \in V = \text{span}(\mathbf{y}_1, \dots, \mathbf{y}_m)$.

Let $\mathbf{x} = \hat{\mathbf{x}} + \varepsilon$ be the system of equations described above:

$$\begin{cases} \langle \hat{\mathbf{x}} + \varepsilon, \mathbf{y}_1 \rangle_Q = b_1 \\ \langle \hat{\mathbf{x}} + \varepsilon, \mathbf{y}_2 \rangle_Q = b_2 \\ \vdots \\ \langle \hat{\mathbf{x}} + \varepsilon, \mathbf{y}_m \rangle_Q = b_m \end{cases} \Rightarrow \begin{cases} \langle \hat{\mathbf{x}}, \mathbf{y}_1 \rangle_Q + \langle \varepsilon, \mathbf{y}_1 \rangle_Q = b_1 \\ \langle \hat{\mathbf{x}}, \mathbf{y}_2 \rangle_Q + \langle \varepsilon, \mathbf{y}_2 \rangle_Q = b_2 \\ \vdots \\ \langle \hat{\mathbf{x}}, \mathbf{y}_m \rangle_Q + \langle \varepsilon, \mathbf{y}_m \rangle_Q = b_m \end{cases} \Rightarrow A Q \hat{\mathbf{x}} = \mathbf{b}$$

where it was assumed that $\langle \varepsilon, \mathbf{y}_1 \rangle_Q = \langle \varepsilon, \mathbf{y}_2 \rangle_Q = \dots = \langle \varepsilon, \mathbf{y}_m \rangle_Q = 0$ because $\hat{\mathbf{x}} \in V$.

By the dual approximation $\hat{\mathbf{x}} \in V \Rightarrow \hat{\mathbf{x}} = A^T \mathbf{c}$.

$$\begin{cases} A Q \hat{\mathbf{x}} = \mathbf{b} \\ \hat{\mathbf{x}} = A^T \mathbf{c} \end{cases} \Rightarrow A Q A^T \mathbf{c} = \mathbf{b} \Rightarrow \mathbf{c} = (A Q A^T)^{-1} \mathbf{b} \Rightarrow \hat{\mathbf{x}} = A^T (A Q A^T)^{-1} \mathbf{b}$$

Exercise 1.0.30. Compute the angle between signals t and t^2 , defined in the interval $[-1, 1]$

Exercise 1.0.31. Let $g(x, y) = e^{d(x, y)}$ where $d(x, y)$ is a metric function. Is $g(x, y)$ a metric function? Why?

Chapter 2

Transforms

Exercise 2.0.32. Verifique que as funes $\phi_k(n)$ e $\phi_r(n)$ so ortogonais em qualquer intervalo de dimenso N , por exemplo, $[n_0, n_0 + N - 1]$ e que formam um conjunto completo, isto , apenas a sequencia nula ortogonal a todas as funes de base.

$$\begin{aligned}\langle \phi_{\mathbf{k}}(\mathbf{n}), \phi_{\mathbf{r}}(\mathbf{n}) \rangle &= \sum_{n=n_0}^{n_0+N-1} \phi_k(n) \phi_r^*(n) = \frac{1}{N} \sum_{n=0}^{N-1} e^{j\frac{2\pi}{N}kn} e^{-j\frac{2\pi}{N}rn} \\ &= \frac{1}{N} \sum_{n=0}^{N-1} e^{j\frac{2\pi}{N}(k-r)n} = \frac{1}{N} \begin{cases} N, & k = r; \\ \frac{1-e^{j2\pi(k-r)}}{1-e^{-j\frac{2\pi}{N}(k-r)}}, & k \neq r. \end{cases} \\ &= \begin{cases} 1, & k = r; \\ 0, & k \neq r. \end{cases} = \delta(k-r) \end{aligned} \quad (2.1)$$

Estas funes, de periodo $N, N/2, N/3, \dots, 1, 0$, para $k = 1, 2, \dots, N, 0$ respectivamente, podem ento ser utilizadas para representar sequencias discretas de periodo $T \leq N$ atravs da seguinte combinao linear

$$\tilde{x}(n) = \sum_{k=0}^N X(k) \phi_k(n) = \sum_{k=0}^{N-1} X(k) e^{j\frac{2\pi}{N}kn} \quad (2.2)$$

em que $X(k)$ so os coeficientes da combinao linear.

Exercise 2.0.33. Deduza a equao que permite calcular os coeficientes $X(k)$ a partir de $\tilde{x}(n)$, equao de anlise. Sugesto: Calcule $\langle \phi_{\mathbf{r}}(\mathbf{n}), \sum_{\mathbf{k}=0}^N \mathbf{X}(\mathbf{k}) \phi_{\mathbf{k}}(\mathbf{n}) \rangle$

Exercise 2.0.34. Verifique a seguinte igualdade,

$$\frac{1}{N} \sum_{n=0}^{N-1} e^{j\frac{2\pi}{N}kn} = \begin{cases} 1, & \text{se } k = mN; \\ 0, & \text{caso contrrio.} \end{cases} \quad (2.3)$$

Exercise 2.0.35. Considere a seguinte sequncia peridica (onda quadrada) de perodo $N = 10$, $x(n) = \dots 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, \dots$ e o sistema descrito pela equao: $y(n) = 0.8y(n-1) + 0.2x(n)$.

1. Calcule a DFS do sinal peridico $x(n)$.
2. Qual a sada e a respectiva DFS?

Exercise 2.0.36. Considere o sistema linear de 1 ordem $y(n) = ay(n-1) + x(n)$ com $a \in [0, 1[$ e $x(n)$ peridico de perodo N . Determine a resposta impulsiva do filtro FIR com uma resposta forada igual do filtro apresentado.

Exercise 2.0.37. Considere as seguintes sequncias peridicas de perodo $N = 4$,

$$\begin{aligned}\tilde{x}(n) &= \dots 0, 2, 0, 1, \dots \\ \tilde{y}(n) &= \dots -1, 0, 1, 0, \dots\end{aligned}$$

1. Determine os coeficientes da srie de Fourier discreta de ambas as sequncias.
2. Usando os coeficientes que calculou na alnea anterior, calcule a convoluo peridica das sequncias.
3. Considere a sequncia $\tilde{z}(n) = \tilde{x}(n) + \tilde{y}(n)$. Determine a sua convoluo com $\tilde{x}(n)$.

Exercise 2.0.38. Considere a sequncia $\tilde{w}(n) = \tilde{x}(n) + \tilde{y}(n)$. Sabendo que $\tilde{x}(n)$ tem perodo N e que $\tilde{y}(n)$ tem perodo M :

1. Mostre que $\tilde{w}(n)$ tem perodo NM
2. Determine $\tilde{W}(K)$ em funo de $\tilde{X}(k)$ e de $\tilde{Y}(k)$.

Exercise 2.0.39. Calcular a DFT das seguintes sequncias:

1. $x(n) = \delta(n)$.
2. $x(n) = \delta(n - n_0)$ para $0 \leq n_0 \leq N$.
3. $x(n) = a(n)$ para $0 \leq n \leq N - 1$.

Exercise 2.0.40. Dada a sequncia de durao finita $x(n) = \{1, 2, 3\}$ com $0 \leq n \leq 3$, desenhar $x((-n)_4)$.

Exercise 2.0.41. Pretende-se determinar o espectro de um sinal analógico usando uma DFT de 1024 pontos. Se o sinal for amostrado a 10kHz, qual a resolução que se obtém em frequência?

Exercise 2.0.42. A DFT de uma sequência de duração finita corresponde a amostras da sua transformada Z no círculo de raio unitário. Por exemplo, a DFT de ordem 10 da sequência $x(n)$ de comprimento 10, corresponde a 10 amostras de $X(z)$ igualmente espaçadas. Pretende-se determinar a sequência $x_1(n)$ que tem como DFT as amostras de $X(z)$ em $z = 0.5e^{j\frac{2\pi}{10}k + \frac{\pi}{10}}$.

Exercise 2.0.43. Uma sequência de duração finita tem a seguinte DFT: $X(k) = \{4, 3, 2, 1, 0, 1, 2, 3\}$, para $0 \leq k < 8$. Determine os 16 pontos da DFT da sequência $y(n)$ obtida da seguinte forma

$$y(n) = \begin{cases} x(n/2), & n \text{ par;} \\ 0, & n \text{ mpar.} \end{cases} \quad (2.4)$$

Exercise 2.0.44. Considere a sequência de duração finita $x(n)$ com $x(n) = 0$ para $n \in Z \setminus [0, N-1]$. Pretende-se determinar M amostras de $X(z)$ igualmente espaçadas sobre o círculo de unitário com $M < N$. Uma das amostras está em $z = 1$. Como obter as M amostras de $X(z)$ usando uma DFT de M pontos de uma sequência de M pontos obtida a partir de $x(n)$?

Exercise 2.0.45. Considere duas sequências de duração finita $x(n)$ e $y(n)$, nulas para $n < 0$ e $\begin{pmatrix} x(n) = 0 & n \geq 8 \\ y(n) = 0 & n \geq 20 \end{pmatrix}$. As DFTs de 20 pontos de ambas as sequências são multiplicadas e calculada a transformada inversa do resultado, $r(n)$. Indique quais os pontos de $r(n)$ que correspondem a pontos que se obteriam pela convolução linear de $x(n)$ com $y(n)$.

Exercise 2.0.46. Pretende-se realizar um filtro FIR de ordem 50 com a técnica de overlap-save: i) sobreposição de V amostras nas secções de entrada, ii) das saídas de cada secção devem-se retirar M amostras que se devem juntar para obter a saída desejada, filtrada. Assuma que os segmentos de entrada têm 100 amostras de comprimento e que a DFT tem dimensão 128. Além disso, assuma que a sequência de saída da convolução circular indexada do ponto 0 ao ponto 127.

1. Determine V .
2. Determine M .
3. Determine o índice do início e do fim dos M pontos extraídos, ou seja, quais dos 128 pontos da convolução circular deverão ser extraídos para juntar aos resultados da secção anterior.

Exercise 2.0.47. Usando a definio, determine a DFT das seguintes sequncias (caso no existam diga a razo).

1. $x(n) = 0.5^n u(n)$
2. $x(n) = 0.5^{|n|}$
3. $x(n) = 2^n u(-n)$
4. $x(n) = 0.5^n u(-n)$
5. $x(n) = 2^{|n|}$
6. $x(n) = 3(8)^{|n|} \cos(0.1\pi n)$

Exercise 2.0.48. Calcule a DFT das seguintes sequncias:

1. $x(n) = [1, 0, -1, 0]$
2. $x(n) = [j, 0, j, 1]$
3. $x(n) = [1, 1, 1, 1, 1, 1, 1, 1]$
4. $x(n) = \cos(0.25\pi n)$ para $n = 0, \dots, 7$.
5. $x(n) = 0.9^n$ para $n = 0, \dots, 7$.

Exercise 2.0.49. Sejam $X(k) = [1, j, -1, -j]$ e $H(k) = [0, 1, -1, 1]$ as DFTs de duas sequncias $x(n)$ e $h(n)$ respectivamente. Usando as propriedades da DFT (sem calcular explicitamente $x(n)$ e $y(n)$) determine as DFTs das seguintes sequncias:

1. $x((n-1)_4)$
2. $x((n+3)_4)$
3. $y(n) = h(n) \otimes x(n)$, em que \otimes significa convoluao circular.
4. $(-1)^n x(n)$
5. $j^n x(n)$
6. $x((-n)_4)$
7. $x((2-n)_4)$

Exercise 2.0.50. Duas sequncias finitas $h(n)$ e $x(n)$ tm as seguintes DFTs, $X(k) = DFT(x) = [1, -2, 1, -2]$ e $H(k) = DFT(h) = [1, j, 1, -j]$. Seja $y(n)$ uma sequncia de dimenso 4, obtida atravs da convoluo circular de $h(n)$ com $x(n)$, i.e., $y(n) = h(n) \otimes x(n)$. Sem calcular explicitamente $h(n)$ e $x(n)$,

1. determine $DFT(x((n-1)_4))$ e $DFT(h((n+2)_4))$ onde os deslocamentos so circulares.
2. determine os valores $y(0)$ e $y(1)$.

Exercise 2.0.51. Seja $X(k)$ a DFT de $x(n)$ e sejam as seguintes sequncias:

$$s(n) = [x(0), \dots, x(N-1), x(0), \dots, x(N-1)] \quad (2.5)$$

$$y(n) = [x(0), \dots, x(N-1), \underbrace{0, \dots, 0}_{N \text{ zeros}}]. \quad (2.6)$$

1. Mostre que $S(2k) = Y(2k) = X(k)$, para $k = 0, \dots, N-1$.
2. Mostre que $S(2k+1) = 0$, para $k = 0, \dots, N-1$.
3. Se $y(n) = s(n)w(n)$, com $w(n) = [\underbrace{1, \dots, 1}_N, \underbrace{0, \dots, 0}_N]$, determine uma expresso (frmula de interpolao) para $Y(2k+1)$.

Exercise 2.0.52. Um problema que existe em muitos pacotes de clculo cientfico (como o MatLab, p.ex.) o facto de apenas se ter acesso a ndices positivos. Suponha que quer determinar a DFT da seguinte sequncia: $x(n) = [0, 1, 2, 3, \underbrace{4}_{x(0)}, 5, 6, 7]$, e o algoritmo apenas lhe permite calcular a DFT de $[x(0), x(1), \dots, x(N-1)]$. Como resolveria o problema?

Exercise 2.0.53. Seja $X(k) = DFT(x(n))$ com $n, k = 0, \dots, N-1$. Determine a relao entre $X(k)$ e as seguintes DFTs,

1. $DFT(x^*(n))$
2. $DFT(x((-n)_N))$
3. $DFT(Re(x(n)))$
4. $DFT(Im(x(n)))$
5. Aplique os resultados anteriores sequncia $x(n)$ tal que $DFT(x(n)) = [1, 2, 3, 4, 5, 6, 7, 8]$.

Exercise 2.0.54. Seja $X(k) = DFT(x(n))$ com $n, k = 0, \dots, N - 1$. Determine a relação entre $X(k)$ e as seguintes DFTs,

1. $IDFT(X^*(k))$
2. $IDFT(X((-k)_N))$
3. $IDFT(Re(X(k)))$
4. $IDFT(Im(X(k)))$
5. Aplique os resultados anteriores à sequência $X(k)$ em que $IDFT(X(k)) = [1, 2j, -2j, 0, 1, -j, j]$.

Exercise 2.0.55. Seja $x(n)$ uma sequência periódica cujo período $[1, -2, \underbrace{3}_{x(0)}, -45, -6]$. Esta sequência é a entrada de um sistema linear e invariante no tempo cuja resposta impulsiva $h(n) = 0.8^{|n|}$. Determine um período da sequência de saída $y(n)$.

Exercise 2.0.56. Considere uma DFT de 9 pontos.

1. Desenhe o diagrama de fluxo com decimação no tempo usando uma decomposição 3×3 .
2. Quantas multiplicações complexas por potências de $e^{j2\pi/N}$ são necessárias?
3. É possível realizar todos os cálculos usando apenas um vetor (in place)?

Exercise 2.0.57. O cálculo da DFT exige geralmente a realização de produtos de números complexos. Considere o produto, $(a + jb)(c + jd) = (ac - bd) + j(bc + ad)$. Nesta forma a multiplicação complexa requer 4 multiplicações e 2 somas reais. Verifique que a multiplicação complexa pode também ser efetuada com 3 multiplicações e 5 somas reais: $[(a - b)d + (c - d)a] + j[(a - b)d + (c + d)b]$.

Exercise 2.0.58. Considere as seguintes sequências de comprimento 8:

$$\begin{aligned} x(n) &= [3, 0, 2, 0, 1, 0, -2, 0] \\ y(n) &= [2, 0, 1, 0, 2, 0, 1, 0]. \end{aligned} \quad (2.7)$$

Resolva as aléguas seguintes usando o menor número de operações possível.

1. Determine a DFT de $x(n)$ usando o algoritmo FFT.
2. Determine a DFT de $y(n)$ usando o algoritmo FFT.

3. Determine a sequência $z(n)$ resultante da convolução circular de ordem 8 de $x(n)$ com $y(n)$.

Exercise 2.0.59. Construa o gráfico de fluxo do algoritmo FFT de base 2 com 16 pontos e decimação no tempo. Indique os fatores multiplicativos diferentes de 1. Indique as amostras das sequências de entrada e de saída. Determine o número de multiplicações e somas reais necessárias.

Exercise 2.0.60. Considere a sequência complexa $x(n)$ de comprimento N que satisfaz a seguinte condição de simetria: $x(n) = -x((n * N/2)_N)$ com $0 \leq n \leq N - 1$ em que N par.

1. Mostre que a DFT de $X(k) = 0$ quando k par.
2. Indique como calcular os termos ímpares da DFT de $x(n)$ usando apenas uma DFT de $N/2$ pontos com alguns cálculos adicionais.

Exercise 2.0.61. Considere uma sequência $x(n)$ de comprimento N cuja DFT $X(k)$ com $k \in [0, N - 1]$. O algoritmo seguinte calcula as amostras pares da DFT, para N par, usando uma única DFT de $N/2$ pontos.

- Crie a sequência $y(n) = \begin{cases} x(n) + x(n + N/2), & 0 \leq n \leq N/2 - 1; \\ 0, & \text{caso contrário.} \end{cases}$
- Calcule a DFT de $N/2$ pontos de $y(n)$, $Y(k)$ com $k \in [0, N/2 - 1]$.
- As amostras pares de $X(k)$ são $X(k) = Y(k/2)$ para $k \in [0, N - 1]$.

1. Mostre que o algoritmo apresentado produz os resultados pretendidos.
2. Considere agora que a sequência $y(n)$ criada a partir de $x(n)$ por:

$$y(n) = \begin{cases} \sum_{r=-\infty}^{\infty} x(n + rM), & 0 \leq n \leq M - 1; \\ 0, & \text{caso contrário.} \end{cases} \quad (2.8)$$

Determine a relação entre a DFT de M pontos $Y(k)$ e a transformada de Fourier de $x(n)$. Mostre que a relação anterior é um caso particular desta.

3. Desenvolva um algoritmo semelhante ao exposto para calcular as amostras ímpares de $X(k)$ usando apenas uma DFT de $N/2$ pontos.

Exercise 2.0.62. Considere as sequências $x(n)$ de comprimento L e $h(n)$ de comprimento P . Pretende-se calcular a convolução linear destas duas sequências, $y(n) = x(n) * h(n)$.

1. Qual o comprimento de $y(n)$?
2. Quantas multiplicaes reais so necessrias para calcular directamente $y(n)$?
3. Defina um procedimento para usar uma DFT na determinao das amostras de $y(n)$. Determine a menor dimenso das transformadas directa e inversa em termos de L e P .
4. Assuma que $L = P = N/2$ em que N uma potncia de 2 e a dimenso da DFT. Determine uma expresso para o nmero de multiplicaes reais necessrias para calcular $y(n)$ usando o mtodo da alnea anterior com FFT de base 2. Com base nessa expresso qual o menor valor de N para o qual o mtodo da FFT requer menos multiplicaes reais que o clculo directo da convoluo.

Exercise 2.0.63. A entrada e a sada de um sistema linear e invariante no tempo satisfazem a seguinte equao s diferenas:

$$y(n) = \sum_{k=1}^N a_k y(n-k) + \sum_{k=0}^M b_k x(n-k) \quad (2.9)$$

Suponha de que dispe de um programa para calcular DFTs de sequncias de comprimento $N = 2^v$. Descreva um procedimento que utilize esse programa e que permita calcular: $H(e^{j\frac{2\pi}{512}k})$ para $k \in [0, 511]$.

Chapter 3

Random Variables

1. Consider the following filter

$$H(z) = -\frac{1}{1 - 0.5z^{-1}} \quad (3.1)$$

For a zero mean white noise input, $\epsilon(n)$, with variance $\sigma^2 = 1$, what is the mean of the output signal, $\mu_y = E[y(n)]$?

- a) $\mu_y = \sqrt{2}$.
- b) $\mu_y = 1$.
- c) $\mu_y = -\sqrt{2}$.
- d) None

2. Consider the following filter

$$H(z) = -\frac{1}{1 - 0.25z^{-1}} \quad (3.2)$$

For a zero mean white noise input, $\epsilon(n)$, with variance $\sigma^2 = 1$, what is *power density spectrum* (PDS) of the output signal?

- a) $1/(1.0625 - 0.5\cos(\omega))$.
- b) $4/(1.25 - \cos(\omega))$.
- c) $4/(1.25 - 0.25\cos(\omega))$.
- d) None

3. Let $x(n)$ be a stationary random process with *Power Density Spectrum* (PDS) $P_x(\omega)$ and let $y(n) = x(n) + c$, where c is a constant. What is the PDS of y , $P_y(\omega)$?

- a) $P_y(\omega) = P_x(\omega) + 2\pi c(c + 2\mu_x)\delta(\omega)$.
 - b) $P_y(\omega) = P_x(\omega) + 2\pi c^2\delta(\omega)$.
 - c) $P_y(\omega) = P_x(\omega) + c$.
 - d) None
4. Let us consider the adaptive filter that uses the LMS algorithm where $w_0 = [0, 0, 0, 0]^T$, $x = y = [\dots, -1, -1, \dots, -1, 1]$ and be a stationary random process with *Power Density Spectrum* (PDS) $P_x(\omega)$ and let $y(n) = x(n) + c$, where c is a constant. What is the PDS of y , $P_y(\omega)$?
- a) $P_y(\omega) = P_x(\omega) + 2\pi c(c + 2\mu_x)\delta(\omega)$.
 - b) $P_y(\omega) = P_x(\omega) + 2\pi c^2\delta(\omega)$.
 - c) $P_y(\omega) = P_x(\omega) + c$.
 - d) None