COM-514: Mathematical Foundations of Signal Processing

Fall 2019

Homework #1 - Due date: 18^{th} October 2019

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Solution 1: Bases and Matrix Representation of Linear Operators Part (a)

(i) In order to have span($\{\psi_0(t)\}$) = span($\{\varphi_0(t)\}$), we need $\psi_0(t)$ to be a constant, say $\psi_0(t) = c$ with $c \in \mathbb{R} \setminus \{0\}$, and we want $\psi_0(t)$ to have unit norm.

$$\|\psi_0(t)\| := \sqrt{\langle \psi_0(t), \psi_0(t) \rangle} = \sqrt{\int_{-1}^1 c^2 dt} = \sqrt{[c^2 x]_{-1}^1} = c\sqrt{2} = 1 \Longleftrightarrow c = \frac{1}{\sqrt{2}}$$
 (1)

(ii) Let's check that $\varphi_1(t) \perp \psi_0(t)$:

$$\langle \varphi_1(t), \psi_0(t) \rangle = \int_{-1}^1 \frac{1}{\sqrt{2}} t dt = \frac{1}{\sqrt{2}} \left[\frac{t^2}{2} \right]_{-1}^1 = 0$$
 (2)

Now, let $\psi_1(t) = b\varphi_1(t)$ for $b \in \mathbb{R} \setminus \{0\}$ such that $\psi_1(t)$ has unit norm.

$$\|\psi_1(t)\| := \sqrt{\langle \psi_1(t), \psi_1(t) \rangle} = \sqrt{\int_{-1}^1 b^2 t^2 dt} = \sqrt{\left[b^2 \frac{t^3}{3}\right]_{-1}^1} = b\sqrt{\frac{2}{3}} = 1 \iff b = \sqrt{\frac{3}{2}}$$
 (3)

(iii) Finally, let $\psi_2(t) = at^2 + bt + c$ for $a, b, c \in \mathbb{R}$ with $a \neq 0$ in order to span all the polynomials up to order 2. We need $\psi_2(t) \perp \psi_0(t)$ and $\psi_2(t) \perp \psi_1(t)$ in order to form a bases:

$$\langle \psi_2(t), \psi_0(t) \rangle = \frac{1}{\sqrt{2}} \int_{-1}^1 at^2 + bt + cdt = \frac{1}{\sqrt{2}} \left[a \frac{t^3}{3} + b \frac{t^2}{2} + ct \right]_{-1}^1 = \frac{1}{\sqrt{2}} (a \frac{2}{3} + 2c) = 0 \iff c = \frac{-a}{3}$$
(4)

$$\langle \psi_2(t), \psi_1(t) \rangle = \sqrt{\frac{3}{2}} \int_{-1}^1 at^3 + bt^2 + ct dt = \sqrt{\frac{3}{2}} \left[a \frac{t^4}{4} + b \frac{t^3}{3} + c \frac{t^2}{2} \right]_{-1}^1 = \sqrt{\frac{3}{2}} b \frac{2}{3} = 0 \iff b = 0$$
 (5)

Finally we want an orthonormal bases so we need $\|\psi_2(t)\| = 1$. Applying the results in (4), (5), we get:

$$\|\psi_2(t)\| := \sqrt{\langle \psi_2(t), \psi_2(t) \rangle} = \sqrt{\int_{-1}^1 \left(at^2 - \frac{a}{3} \right)^2 dt} = \sqrt{\left[\frac{a^2 t^5}{5} - \frac{2}{3} \frac{a^2 t^3}{3} + \frac{a^2 t}{9} \right]_{-1}^1}$$
 (6)

$$=\sqrt{\frac{a^28}{45}} = 1 \Longleftrightarrow a = \frac{3\sqrt{10}}{4} \tag{7}$$

So $\Psi = \left\{ \frac{1}{\sqrt{2}}, \sqrt{\frac{3}{2}}t, \frac{3\sqrt{10}}{4}t^2 - \frac{\sqrt{10}}{4} \right\}$ is an orthonormal bases for H.

Part (b)

(iv) Let $\Gamma_{\Phi}: \mathbb{R}^3 \to \mathbb{R}^3$ be the matrix that performs differentiation on the expansion coefficients of Φ . Then, as we are using the usual base for polynomials, we know that $\frac{d}{dt}\Phi = \{0, 1, 2t\}$, and thus

$$\Gamma_{\Phi} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{bmatrix} \tag{8}$$

(v) (a) Let $A: H \to H$ be the differentiation operator,

$$\Gamma_{\Psi} = \Psi^* A \Psi = \begin{bmatrix} \langle A\psi_0, \psi_0 \rangle & \langle A\psi_1, \psi_0 \rangle & \langle A\psi_2, \psi_0 \rangle \\ \langle A\psi_0, \psi_1 \rangle & \langle A\psi_1, \psi_1 \rangle & \langle A\psi_2, \psi_1 \rangle \\ \langle A\psi_0, \psi_2 \rangle & \langle A\psi_1, \psi_2 \rangle & \langle A\psi_2, \psi_2 \rangle \end{bmatrix} = \begin{bmatrix} \langle \frac{d}{dt}\psi_0, \psi_0 \rangle & \langle \frac{d}{dt}\psi_1, \psi_0 \rangle & \langle \frac{d}{dt}\psi_2, \psi_0 \rangle \\ \langle \frac{d}{dt}\psi_0, \psi_1 \rangle & \langle \frac{d}{dt}\psi_1, \psi_1 \rangle & \langle \frac{d}{dt}\psi_2, \psi_1 \rangle \\ \langle \frac{d}{dt}\psi_0, \psi_2 \rangle & \langle \frac{d}{dt}\psi_1, \psi_2 \rangle & \langle \frac{d}{dt}\psi_2, \psi_2 \rangle \end{bmatrix}$$

$$(9)$$

since Ψ defines an orthonormal bases.

(b) Numerically, we have that

$$\Gamma_{\Psi} = \begin{bmatrix}
\langle 0, \frac{1}{\sqrt{2}} \rangle & \langle \sqrt{\frac{3}{2}}, \frac{1}{\sqrt{2}} \rangle & \langle \frac{3\sqrt{10}}{2}t, \frac{1}{\sqrt{2}} \rangle \\
\langle 0, \sqrt{\frac{3}{2}}t \rangle & \langle \sqrt{\frac{3}{2}}, \sqrt{\frac{3}{2}}t \rangle & \langle \frac{3\sqrt{10}}{2}t, \sqrt{\frac{3}{2}}t \rangle \\
\langle 0, \frac{3\sqrt{10}}{4}t^{2} - \frac{\sqrt{10}}{4} \rangle & \langle \sqrt{\frac{3}{2}}, \frac{3\sqrt{10}}{4}t^{2} - \frac{\sqrt{10}}{4} \rangle & \langle \frac{3\sqrt{10}}{2}t, \frac{3\sqrt{10}}{4}t^{2} - \frac{\sqrt{10}}{4} \rangle \end{bmatrix} = \begin{bmatrix} 0 & \sqrt{3} & 0 \\ 0 & 0 & \sqrt{15} \\ 0 & 0 & 0 \end{bmatrix} \tag{10}$$

Solution 2: Norms of Oblique Projections

(i)

Fact 1. If $\Pi: \mathcal{H} \to \mathcal{H}$ is an orthogonal projection in a Hilbert space \mathcal{H} ,

$$\|\Pi x\| \le \|x\|, \quad \forall x \in \mathcal{H} \tag{11}$$

Using Fact 1, we have that,

$$\|\Pi\| := \max_{x:\|x\|=1} \|\Pi x\| \le \max_{x:\|x\|=1} \|x\| = 1 \tag{12}$$

 Π is an orthogonal projection iff $\Pi\Pi = \Pi$ and $\Pi^T = \Pi$. Mixing both properties, we have that $\Pi^T\Pi = \Pi\Pi = \Pi$. Let's prove that $I - \Pi$ is also an orthogonal projection

$$(I - \Pi)^T = I^T - \Pi^T = I - \Pi \tag{13}$$

$$(I - \Pi)(I - \Pi) = II - \Pi I - I\Pi + \Pi \Pi = I - 2\Pi + \Pi = I - \Pi$$
(14)

By idempotence (so this also holds for oblique projections) and definition of the spectral norm,

$$\|\Pi x\| = \|\Pi(\Pi x)\| \le \|\Pi\| \|\Pi x\| \iff \|\Pi\| \ge 1$$
 (15)

So putting together (12), (15) we get that $\|\Pi\| = 1$. Moreover, since $(I - \Pi)$ is also an orthogonal projection as proved before,

$$\|\Pi\| = \|I - \Pi\| = 1 \tag{16}$$

(ii)

$$||x||^2 + ||y||^2 := \langle Pu, Pu \rangle + \langle (I - P)u, (I - P)u \rangle \tag{17}$$

$$= \langle Pu, Pu \rangle + \langle u, (I-P)u \rangle - \langle Pu, (I-P)u \rangle \tag{18}$$

$$= \langle Pu, Pu \rangle + \langle (I - P)u, u \rangle^* - \langle (I - P)u, Pu \rangle^*$$
(19)

$$= \langle Pu, Pu \rangle^* + \langle u, u \rangle^* - \langle Pu, u \rangle^* - \langle u, Pu \rangle^* + \langle Pu, Pu \rangle^*$$
(20)

where I only used fundamental properties of the inner product and in the last equality, I use the fact that a norm is real (so $||x||^2 = \langle x, x \rangle = \langle x, x \rangle^*$).

By rearranging terms in (20), we get that

$$||x||^2 + ||y||^2 = ||u|| - (\langle Pu, u \rangle^* - \langle Pu, Pu \rangle^*) - (\langle u, Pu \rangle^* - \langle Pu, Pu \rangle^*)$$
(21)

$$= ||u|| - (\langle u, Pu \rangle - \langle Pu, Pu \rangle) - \langle (I - P)u, Pu \rangle^*$$
(22)

$$= ||u|| - \langle (I - P)u, Pu \rangle - \langle Pu, (I - P)u \rangle \tag{23}$$

$$= ||u|| - (\langle Pu, (I-P)u \rangle^* + \langle Pu, (I-P)u \rangle) \tag{24}$$

$$:= ||u|| - 2\mathbb{R}e\langle x, y\rangle \tag{25}$$

(iii) In the case that x = 0,

$$||Pu|| := ||x|| = 0 \le ||I - P|| \tag{26}$$

which follows from non-negativity of norms.

Let P' := I - P. We have that

$$P'^{2} = (I - P)(I - P) = I - P - P + P^{2} = I - P := P'$$
(27)

so P' is an oblique projection.

If y := (I - P)u = 0, we have that u = Pu, and hence

$$||Pu|| = ||u|| = 1 \le ||I - P|| \tag{28}$$

where the inequality holds since the norm of any projection is greater or equal than one as proved in (15).

(iv)

$$||w||^2 := ||\tilde{x} + \tilde{y}||^2 := \langle \tilde{x} + \tilde{y}, \tilde{x} + \tilde{y} \rangle = \langle \tilde{x}, \tilde{x} + \tilde{y} \rangle + \langle \tilde{y}, \tilde{x} + \tilde{y} \rangle$$
(29)

$$= \langle \tilde{x}, \tilde{x} \rangle^* + \langle \tilde{y}, \tilde{x} \rangle^* + \langle \tilde{x}, \tilde{y} \rangle^* + \langle \tilde{y}, \tilde{y} \rangle^* = ||y||^2 + \langle y, x \rangle^* + \langle x, y \rangle^* + ||x||^2$$
(30)

$$= ||y||^2 + \langle x, y \rangle + \langle x, y \rangle^* + ||x||^2 = ||x||^2 + ||y||^2 + 2\mathbb{R}e\langle x, y \rangle$$
(31)

$$= ||u||^2$$
 (32)

where the last step follows from (25). Moreover, since ||u|| = 1 by construction, ||w|| = 1.

(v)

$$(I - P)w := (I - P)\frac{\|y\|}{\|x\|}x + (I - P)\frac{\|x\|}{\|y\|}y$$
(33)

$$= \frac{\|y\|}{\|x\|} x - \frac{\|y\|}{\|x\|} Px + \frac{\|x\|}{\|y\|} y - \frac{\|x\|}{\|y\|} Py \qquad (P^2 = P \Rightarrow Px := PPu = Pu := x)$$

(34)

$$= \frac{\|x\|}{\|y\|} y - \frac{\|x\|}{\|y\|} Py := \frac{\|x\|}{\|y\|} (I - P)u - \frac{\|x\|}{\|y\|} P(I - P)u$$
(35)

$$= \frac{\|x\|}{\|y\|} u - \frac{\|x\|}{\|y\|} Pu - \frac{\|x\|}{\|y\|} Pu + \frac{\|x\|}{\|y\|} Pu$$
 (P² = P)

$$= \frac{\|x\|}{\|y\|} (I - P)u := \frac{\|x\|}{\|y\|} y \tag{37}$$

So using (37), we have that

$$\|(I - P)w\| = \frac{\|x\|}{\|y\|} \|y\| = \|x\| := \|Pu\|$$
(38)

So taking equations (32), (38) and by definition of the spectral norm,

$$||Pu|| = ||(I - P)w|| \le ||I - P|| \, ||w|| = ||I - P||$$
(39)

Applying again the definition of spectral norm and given that ||u|| = 1,

$$||I - P|| \ge ||Pu|| \ge ||P|| \, ||u|| = ||P|| \tag{40}$$

which concludes the proof.

(vi) Take P' := I - P as in (27), where it was proved that P' is an oblique projection so (40) holds and we have that

$$||I - P|| := ||P'|| \le ||I - P'|| := ||I - (I - P)|| = ||P||$$

$$\tag{41}$$

By mixing (40) and (41), we have that ||I - P|| = ||P||.

Solution 3: DFT Matrix

1. This matrices have coefficients

$$A_{k,n} = e^{-j\frac{2\pi kn}{N}} \quad ; \quad B_{n,k} = \frac{1}{N}e^{j\frac{2\pi kn}{N}}$$
 (42)

$$\begin{bmatrix} X_1 \\ \vdots \\ X_N \end{bmatrix} = \begin{bmatrix} & A & \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_N \end{bmatrix} \quad ; \quad \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} & B & \end{bmatrix} \begin{bmatrix} X_1 \\ \vdots \\ X_N \end{bmatrix}$$
(43)

with $n, k \in \{1, 2, \dots, N\}$.

2. Note that equality (b) is trivial to prove, since the coefficients of the matrices A and B as expressed in (42) satisfy $A_{k,n} = NB_{n,k}^*$, so $A = NB^*$.

If equality (a) is satisfied, then $AB = BA = I_{N \times N}$. Let's start by proving that A is a left-inverse of B.

$$(AB)_{i,j} = \frac{1}{N} \sum_{l=0}^{N-1} e^{-j\frac{2\pi il}{N}} e^{j\frac{2\pi lj}{N}}$$

$$\tag{44}$$

So here we can distinguish two cases. On the one hand, if i = j, sum simplifies to

$$(AB)_{i,j} = \frac{1}{N} \sum_{l=0}^{N-1} 1 = 1 \tag{45}$$

On the other hand, if $i \neq j$,

$$(AB)_{i,j} = \frac{1 - (e^{j\frac{2\pi N}{N}})^{j-i}}{1 - e^{j\frac{2\pi(j-i)}{N}}} = \frac{1 - 1^{j-i}}{1 - e^{j\frac{2\pi(j-i)}{N}}} = 0$$

$$(46)$$

so A is a left-inverse of B. Now, let's see if it's actually a proper inverse.

$$(BA)_{i,j} = \frac{1}{N} \sum_{l=0}^{N-1} e^{-j\frac{2\pi l j}{N}} e^{j\frac{2\pi i l}{N}} = (AB)_{j,i} = \mathbb{1}_{\{i=j\}}$$

$$(47)$$

and hence we can conclude that $A = B^{-1}$.

3. Let $b_k(n) := \frac{1}{N} e^{j\frac{2\pi kn}{N}}$. If $b_k := [b_k(0), \dots, b_k(n)]^T$ is an eigenvector of C, then $Cb_k = \lambda_k b_k$. Let \mathcal{A}_k be the sequence resulting of applying the Inverse Discrete Fourier Transform to the sequence α_n for $k, n \in \{0, 1, \dots, N-1\}$.

$$(Cb_k)_0 = \sum_{l=0}^{N-1} b_k(l)\alpha_l = \alpha_0 b_k(0) + \alpha_1 b_k(1) + \cdots$$
(48)

$$:= \mathcal{A}_k = N\mathcal{A}_k b_k(0) \tag{49}$$

so it seems $\lambda_k = N \mathcal{A}_k$.

Note that $b_k(n+1) = b_k(n)e^{j\frac{2\pi k}{N}}$ and that $b_k(N) := \frac{1}{N}e^{j2\pi k} = \frac{1}{N} = b_k(0)$.

$$(Cb_k)_1 = \sum_{l=0}^{N-1} b_k(l)\alpha_{(1-l) \mod N} = \alpha_{N-1}b_k(0) + \alpha_0b_k(1) + \alpha_1b_k(2) + \cdots$$
 (50)

$$= \alpha_{N-1}b_k(0) + e^{j\frac{2\pi k}{N}}(\alpha_0 b_k(0) + \alpha_1 b_k(1) + \cdots)$$
(51)

$$= \alpha_{N-1}b_k(0) + e^{j\frac{2\pi k}{N}}N\mathcal{A}_kb(0) - e^{j\frac{2\pi k}{N}}\alpha_{N-1}b(N-1) = \alpha_{N-1}b_k(0) + N\mathcal{A}_kb(1) - \alpha_{N-1}b(N)$$
(52)

$$= N\mathcal{A}_k b(1) \tag{53}$$

The general form of $(Cb_k)_i$ is

$$(Cb_k)_i = \sum_{l=0}^{N-1} b_k(l)\alpha_{(i-l) \mod N}$$
 (54)

Let's finish the proof by induction on i. We can do so since by the circular property of the matrix, we can express $(Cb_k)_{i+1}$ in terms of $(Cb_k)_i$.

I already proved the base cases i=0,1 before. Now suppose by induction hypothesis that $(Cb_k)_i=N\mathcal{A}_kb_k(i)$ holds for the i-th coefficient and let $\beta=(i+1) \mod N=0$. Note that the sequence $\alpha_{(i+1-l) \mod N}$ is $\alpha_{\beta},\alpha_{\beta+1},\ldots,\alpha_{\beta+N-1}$ and the sequence $\alpha_{(i-l) \mod N}$ is $\alpha_{\beta+1},\alpha_{\beta+2},\ldots,\alpha_{\beta+N-1},\alpha_{\beta}$.

$$= \alpha_{\beta}b_k(0) + \alpha_{\beta+1}b_k(1) + \alpha_{\beta+2}b_k(2) + \cdots$$

$$(56)$$

$$= \alpha_{\beta} b_k(0) + e^{j\frac{2\pi k}{N}} (\alpha_{\beta+1} b_k(0) + \alpha_{\beta+2} b_k(1) + \cdots)$$
(57)

$$= \alpha_{\beta} b_k(0) + e^{j\frac{2\pi k}{N}} N \mathcal{A}_k b_k(i) - e^{j\frac{2\pi k}{N}} \alpha_{\beta} b(N-1)$$
 By induction hypothesis (58)

$$= \alpha_{\beta} b_k(0) + e^{j\frac{2\pi k}{N}} N \mathcal{A}_k b_k(i) - \alpha_{\beta} b(N)$$
(59)

$$= N\mathcal{A}_k b_k(i) e^{j\frac{2\pi k}{N}} = N\mathcal{A}_k b_k(i+1)$$

$$\tag{60}$$

4. By the findings of the previous section, we know that

$$CB = C \begin{bmatrix} | & & | \\ b_0 & \dots & b_{N-1} \\ | & & | \end{bmatrix} = N \begin{bmatrix} | & & | \\ \mathcal{A}_0 b_0 & \dots & \mathcal{A}_{N-1} b_{N-1} \\ | & & | \end{bmatrix}$$
 (61)

Since $A = B^{-1}$, AB = I so

$$ACB = N\operatorname{diag}(\mathcal{A}) = N \begin{bmatrix} \mathcal{A}_0 & 0 \\ & \ddots \\ 0 & \mathcal{A}_{N-1} \end{bmatrix}$$
 (62)

5. Given that we can run FFT in $\mathcal{O}(n \log_2 n)$, using the analytical result of the operation ACB obtained in (62) is computationally much cheaper than computing the direct matrix product in case of not having a circular matrix. In this case, we would require $\mathcal{O}(n^3)$ operations (or in the best case $\mathcal{O}(n^{2.373})$, which is the asymptotically fastest known algorithm to perform $n \times n$ matrix multiplication).

Numerically, for N=1024, which is a common number of DFT coefficients (power of 2 so fft performs better) the comparison is shown in Table 1.

| Algorithm | Asymptotic number of operations |
|---|------------------------------------|
| FFT $(C \text{ circulant})$ | 10,240 |
| Matrix multiplication Efficient matrix multiplication | $1,073,741,824 \\ \sim 13,913,673$ |

Table 1: Comparison with non-circulant C