

MAT1120 - Assignment 2

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PROBLEM 1

Assume that $U = [\mathbf{u}_1, \dots, \mathbf{u}_n]$ is an $(n \times n)$ -matrix. Verify that U is **semi-orthogonal** if and only if $\mathbf{u}_j \neq \mathbf{0}$ for $j = 1, \dots, n$ and all the \mathbf{u}_j 's are orthogonal to each other. In other words, $\mathbf{u}_j \cdot \mathbf{u}_i = 0$ when $j \neq i$.

Definition. A quadratic matrix U is called **semi-orthogonal** if $U^T U$ is a **diagonal** matrix with only positive entries along the main diagonal.

Definition. A matrix U is called **diagonal** if U has non-zero entries only along the main diagonal.

We now want to show that the condition is both sufficient and necessary for the matrix to be **semi-orthogonal**.

Let us first form the matrix $B = U^T U$.

$$B = \begin{pmatrix} \mathbf{u}_1 \\ \vdots \\ \mathbf{u}_n \end{pmatrix} [\mathbf{u}_1 \quad \dots \quad \mathbf{u}_n] = \begin{bmatrix} \mathbf{u}_1 \cdot \mathbf{u}_1 & \dots & \mathbf{u}_1 \cdot \mathbf{u}_n \\ \vdots & \ddots & \vdots \\ \mathbf{u}_n \cdot \mathbf{u}_1 & \dots & \mathbf{u}_n \cdot \mathbf{u}_n \end{bmatrix}.$$

Assume now that each column in U is orthogonal to all the other columns. We can then directly see that all the entries in B that are not along the main diagonal must be zero. It is now sufficient to show that the entries along the main diagonal must be strictly positive. Let $\mathbf{u} \in U$ be an arbitrary vector from U .

$$\mathbf{u} \cdot \mathbf{u} = u_1^2 + \dots + u_n^2.$$

Each term is squared, thus $\mathbf{u} \cdot \mathbf{u}$ is strictly positive.

If we instead were to assume that each column in U was **not** orthogonal to the other columns, we can directly from the matrix B see that we get non-zero entries that are not along the main diagonal, which contradict the condition that B must be a diagonal matrix. We can then conclude that U is **semi-orthogonal** if and only if the columns of U are orthogonal, and not the zero vector.

Further assume that U is **semi-orthogonal** and let

$$\mathbf{u}'_j = \frac{1}{\mathbf{u}_j \cdot \mathbf{u}_j} \mathbf{u}_j, \quad j = 1, \dots, n.$$

Argue that U is invertible and that

$$U^{-1} = [\mathbf{u}'_1 \dots \mathbf{u}'_n]^T.$$

Let us look at the columns of U , $\{\mathbf{u}_1, \dots, \mathbf{u}_n\}$. If the set of columns is an orthogonal set of non-zero vectors then the set is linearly independent. Since we assumed that U is **semi-orthogonal**, then it follows that the set of columns is linearly independent.

By the Invertible Matrix Theorem, U is invertible.

We now want to show that the inverse of U is as given above. Let us check if $UU^{-1} = I$.

$$UU^{-1} = [\mathbf{u}_1 \quad \dots \quad \mathbf{u}_n] \begin{pmatrix} \mathbf{u}'_1 \\ \vdots \\ \mathbf{u}'_n \end{pmatrix} = \begin{bmatrix} \mathbf{u}_1 \cdot \mathbf{u}'_1 & \dots & \mathbf{u}_1 \cdot \mathbf{u}'_n \\ \vdots & \ddots & \vdots \\ \mathbf{u}_n \cdot \mathbf{u}'_1 & \dots & \mathbf{u}_n \cdot \mathbf{u}'_n \end{bmatrix}$$

Let $\mathbf{x}, \mathbf{y}' \in \{\mathbf{u}_1, \dots, \mathbf{u}_n\}$.

- $\mathbf{x} = \mathbf{y} \Rightarrow \mathbf{x} \cdot \mathbf{y}' = 1$
- $\mathbf{x} \neq \mathbf{y} \Rightarrow \mathbf{x} \cdot \mathbf{y}' = 0$

We have that $\mathbf{x} = \mathbf{y}$ only along the main diagonal, therefore $UU^{-1} = I$.

PROBLEM 2

Divide $[0, \pi]$ into 3 intervals of length $\frac{\pi}{3}$ and let t_1, t_2 and t_3 be the midpoints of these intervals. Let

$$C_3 = \begin{bmatrix} 1 & \cos(t_1) & \cos(2t_1) \\ 1 & \cos(t_2) & \cos(2t_2) \\ 1 & \cos(t_3) & \cos(2t_3) \end{bmatrix} \quad S_3 = \begin{bmatrix} \sin(t_1) & \sin(2t_1) & \sin(3t_1) \\ \sin(t_2) & \sin(2t_2) & \sin(3t_2) \\ \sin(t_3) & \sin(2t_3) & \sin(3t_3) \end{bmatrix}$$

Plugging in the values of t_1, t_2 and t_3 gives:

$$C_3 = \begin{bmatrix} 1 & \frac{\sqrt{3}}{2} & \frac{1}{2} \\ 1 & 0 & -1 \\ 1 & -\frac{\sqrt{3}}{2} & \frac{1}{2} \end{bmatrix} \quad S_3 = \begin{bmatrix} \frac{1}{2} & \frac{\sqrt{3}}{2} & 1 \\ 1 & 0 & -1 \\ \frac{1}{2} & -\frac{\sqrt{3}}{2} & 1 \end{bmatrix}$$

These two matrices are **semi-orthogonal** because $C_3^T C_3$ and $S_3^T S_3$ are diagonal matrices with only positive entries along the main diagonal.

$$C_3^T C_3 = \begin{bmatrix} 3 & 0 & 0 \\ 0 & \frac{3}{2} & 0 \\ 0 & 0 & \frac{3}{2} \end{bmatrix} \quad S_3^T S_3 = \begin{bmatrix} \frac{3}{2} & 0 & 0 \\ 0 & \frac{3}{2} & 0 \\ 0 & 0 & 3 \end{bmatrix}$$

Based on this we know that the inverses of these matrices are given by $U^{-1} = [\mathbf{u}'_1 \quad \dots \quad \mathbf{u}'_n]^T$. This gives us the inverses:

$$C_3^{-1} = \begin{bmatrix} \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\ \frac{\sqrt{3}}{3} & 0 & -\frac{\sqrt{3}}{3} \\ \frac{1}{3} & -\frac{2}{3} & \frac{1}{3} \end{bmatrix} \quad S_3^{-1} = \begin{bmatrix} \frac{1}{3} & \frac{2}{3} & \frac{1}{3} \\ \frac{\sqrt{3}}{3} & 0 & -\frac{\sqrt{3}}{3} \\ \frac{1}{3} & -\frac{1}{3} & \frac{1}{3} \end{bmatrix}$$

PROBLEM 3

According to my python program both the matrices C and S are **semi-orthogonal**, with the following inverses:

$$C^{-1} = \begin{bmatrix} 0.125 & 0.125 & 0.125 & 0.125 & 0.125 & 0.125 & 0.125 & 0.125 \\ 0.245 & 0.208 & 0.139 & 0.049 & -0.049 & -0.139 & -0.208 & -0.245 \\ 0.231 & 0.096 & -0.096 & -0.231 & -0.231 & -0.096 & 0.096 & 0.231 \\ 0.208 & -0.049 & -0.245 & -0.139 & 0.139 & 0.245 & 0.049 & -0.208 \\ 0.177 & -0.177 & -0.177 & 0.177 & 0.177 & -0.177 & -0.177 & 0.177 \\ 0.139 & -0.245 & 0.049 & 0.208 & -0.208 & -0.049 & 0.245 & -0.139 \\ 0.096 & -0.231 & 0.231 & -0.096 & -0.096 & 0.231 & -0.231 & 0.096 \\ 0.049 & -0.139 & 0.208 & -0.245 & 0.245 & -0.208 & 0.139 & -0.049 \end{bmatrix}$$

$$S^{-1} = \begin{bmatrix} 0.049 & 0.139 & 0.208 & 0.245 & 0.245 & 0.208 & 0.139 & 0.049 \\ 0.096 & 0.231 & 0.231 & 0.096 & -0.096 & -0.231 & -0.231 & -0.096 \\ 0.139 & 0.245 & 0.049 & -0.208 & -0.208 & 0.049 & 0.245 & 0.139 \\ 0.177 & 0.177 & -0.177 & -0.177 & 0.177 & 0.177 & -0.177 & -0.177 \\ 0.208 & 0.049 & -0.245 & 0.139 & 0.139 & -0.245 & 0.049 & 0.208 \\ 0.231 & -0.096 & -0.096 & 0.231 & -0.231 & 0.096 & 0.096 & -0.231 \\ 0.245 & -0.208 & 0.139 & -0.049 & -0.049 & 0.139 & -0.208 & 0.245 \\ 0.125 & -0.125 & 0.125 & -0.125 & 0.125 & -0.125 & 0.125 & -0.125 \end{bmatrix}$$

Multiplying the matrices with the respective inverses yields the identity matrix in both cases.

PROBLEM 4

Let $\mathcal{F}([0, 2\pi], \mathbb{R})$ be the vector space consisting of all real functions defined on $[0, \pi]$ and let \mathcal{W}_c be its subspace that is spanned by the set

$$\mathcal{C} = \{1, \cos(t), \cos(2t), \dots, \cos(7t)\}$$

We showed in the previous problem that C is semi-orthogonal. It then follows that C is invertible, which means that the columns of C necessarily has to be linearly independent. This, in turn, means that \mathcal{C} is a basis for \mathcal{W}_C . There is a result regarding signal spaces that tells us that it is sufficient to show that this holds for the discrete values $t = t_1, \dots, t_8$

PROBLEM 5

Let $T_c : \mathcal{W}_C \rightarrow \mathbb{R}^8$ be defined by:

$$T_c(h) = (h(t_1), \dots, h(t_8))$$

for $h \in \mathcal{W}_C$, and let $\epsilon = \{\mathbf{e}_1, \dots, \mathbf{e}_8\}$.

Verify that C is the matrix of T_c with respect to the bases \mathcal{C} and ϵ , and show that T_c is an isomorphism.

C is a matrix with those properties if $T_c(\mathbf{x})$ maps $[\mathbf{x}]_{\mathcal{C}}$ to $[T_c(\mathbf{x})]_{\epsilon}$.

Let h be any arbitrary vector in \mathcal{W}_C with respect to the basis \mathcal{C}

$$h = \begin{bmatrix} \lambda_1 \\ \vdots \\ \lambda_8 \end{bmatrix}$$

Multiplying from the left by the matrix C , we get

$$Ch = \begin{bmatrix} 1 & \dots & \cos(7t_1) \\ \vdots & \dots & \vdots \\ 1 & \dots & \cos(7t_8) \end{bmatrix} \begin{bmatrix} \lambda_1 \\ \vdots \\ \lambda_8 \end{bmatrix} = [(h(t_1), \dots, h(t_8))]_{\epsilon}$$

Since h was any arbitrary vector from \mathcal{W}_C , this means that C is the matrix of T_c with respect to the bases \mathcal{C} and ϵ .

Since \mathcal{C} is a basis for \mathcal{W}_C , and ϵ is a basis for \mathbb{R}^8 it follows that T_c is a bijective transformation.

Since T_c is linear, this means T_c is an isomorphism transformation. The matrix for T_c^{-1} is C^{-1} .

PROBLEM 6

Let $g \in \mathcal{F}([0, \pi], \mathbb{R})$ and set $\mathbf{y} = (g(t_1), \dots, g(t_8)) \in \mathbb{R}^8$. In addition, let

$$g^c = T_c^{-1}(\mathbf{y}) \in \mathcal{W}_C$$

Show that g^c is an 8-midpoint interpolation of g on the interval $[0, \pi]$ that satisfy $[g^c]_C = C^{-1}\mathbf{y}$.

g^c is a function in \mathcal{W}_C that is defined in such a way that g^c interpolate g in the midpoints t_1, \dots, t_8 . This means that g^c is an 8-midpoint interpolation of g on $[0, \pi]$ since,

$$Cg^c = CT_c^{-1}(\mathbf{y}) = CC^{-1}\mathbf{y} = \mathbf{y}$$

$$T_c(g^c) = C[g^c]_C = y \Leftrightarrow [g^c]_C = C^{-1}\mathbf{y}$$

PROBLEM 7

Argue that V_l and V_o are subspaces of V .

We want to show that both V_l and V_o are closed under addition, multiplication by a scalar, and also that the zero-function is both spaces. Ill show the calculations for V_l , the same is proved for V_o similarly. Let $f_1, f_2 \in V_l$, and set $f_3 = f_1 + f_2$. We now want to show that $f_3 \in V_l$.

$$f_3(-t) = (f_1 + f_2)(-t) = f_1(-t) + f_2(-t) = f_1(t) + f_2(t) = f_3(t)$$

Thus V_l is closed under addition. Now, let $f_3 = cf_1, c \in \mathbb{R}$. We want to show that $f_3 \in V_l$

$$f_3(-t) = (cf_1)(t) = cf_1(-t) = cf_1(t) = f_3(t) \in V_l.$$

We now want to show that if $f \in V$, then $f_l \in V_l$ and $f_o \in V_o$ as well as $f = f_l + f_o$. Assume $f \in V$. We then have:

$$f_l = \frac{1}{2}(f + T(f))$$

$$f_o = \frac{1}{2}(f - T(f))$$

$$f_l + f_o = \frac{1}{2}2f = f$$

To show $f_l \in V_l$ and $f_o \in V_o$.

$$[T(f_l)](t) = \frac{1}{2}([T(f)](t) + [T(T(f))](t))$$

$$= \frac{1}{2}(f(t) + f(-t)) = f_l(t)$$

Thus $f_l \in V_l$.

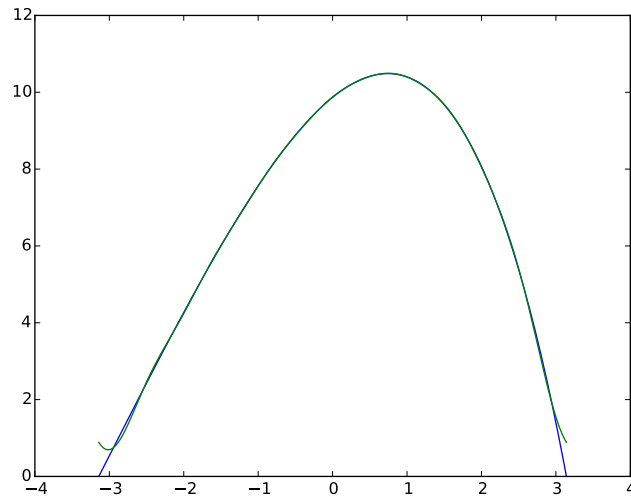
$$[T(f_o)](t) = \frac{1}{2}(f(-t) - f(t))$$

$$= f_o(-t) = -f_o(t).$$

Thus $f_o \in V_o$.

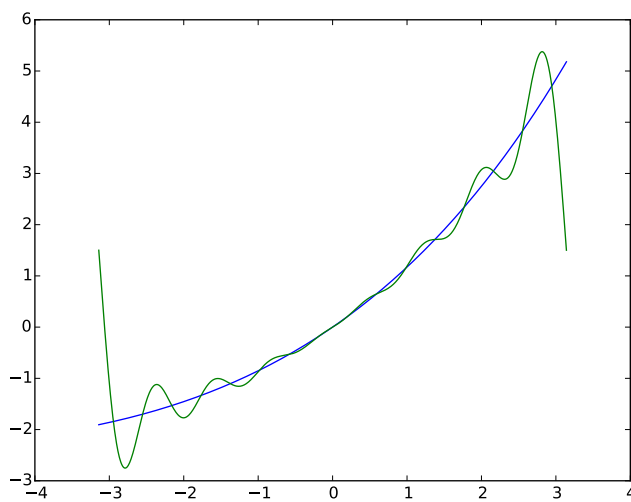
PROBLEM 8

This is a 16-midpoint interpolation of the function $f(t) = (\pi^2 - t^2)e^{t/2\pi}$, $t \in [-\pi, \pi]$



PROBLEM 9

This is a 16-midpoint interpolation of the function $f(t) = te^{t/2\pi}$, $t \in [-\pi, \pi]$



APPENDIX

```
1 # This is the code used for the second mandatory assignment in
2 # MAT1120 – Linear Algebra
3 # at the University of Oslo.
4
5 import numpy as np
6
7 def is_semi_orthogonal(matrix):
8     """ Tests whether a matrix is semi-orthogonal
9         based on the definition given in the
10         assignment text
11
12         Assumes a square matrix (n x n),
13         raises an exception if not square.
14     """
15     if matrix.shape[0] != matrix.shape[1]:
16         raise Exception('Matrix is not square')
17
18     B = matrix.T.dot(matrix) # Forms the matrix to be tested
19     eps = 1.0e-12
20     B[np.abs(B) < eps] = 0 # Makes all entries below a certain threshold zero
21     try:
22         B = B.getA()
23     except:
24         pass
25
26     for i in range(matrix.shape[0]):
27         for j in range(matrix.shape[0]):
28             if i == j:
29                 if B[i][j] <= 0:
30                     return False
31             else:
32                 if abs(B[i][j]) > eps:
33                     return False
34     return True
35
36 def find_inverse(matrix):
37     """ Finds and returns the inverse of a matrix with the properties
38         given in the assignment text.
39
40         Assumes a semi-orthogonal matrix,
41         raises an exception if the matrix
42         is not orthogonal.
43     """
44     if not is_semi_orthogonal(matrix):
45         raise Exception("Matrix is not orthogonal")
46
47
48     matrix_inv = np.zeros((matrix.shape[0], matrix.shape[0]))
49     index = 0
50     for i in matrix.T: # Iterating over columns instead of rows, hence the transpose
51         matrix_inv[index] = transform_vector(i)
```

```

52         index += 1
53
54     eps = 1.0e-12
55     matrix_inv[np.abs(matrix_inv) < eps] = 0
56     return matrix_inv
57
58 def transform_vector(vec):
59     """ Transforms the given vector u_j into
60     u_j' as given in problem 1
61     """
62     try:
63         vec = vec.getA()
64     except:
65         pass
66     return (1./vec.dot(vec.T))*vec
67
68 def print_matrix(matrix):
69     try:
70         matrix = matrix.getA()
71     except:
72         pass
73     for i in matrix:
74         for j in i:
75             print("%7.3f" % (j), end="")
76         print()
77
78 def matrix_to_tex(matrix, filename):
79     outfile = open(filename+".tex", 'w')
80     outfile.write("\\begin{bmatrix}" + "\\n")
81     body = ""
82     for i in matrix:
83         for j in i:
84             body += "%7.3f" % (j) + "&"
85         body = body[:-1]
86         body += " \\\\" + "\\n"
87     outfile.write(body)
88     outfile.write("\\end{bmatrix}")
89
90 def calc_vector_y(function, n):
91     """ Calculates the vector y_o/y_l using a function f_o/f_l, and n midpoints
92     This is used for exercise 8 and 9
93     """
94     t = np.pi/16 + (np.linspace(1, 8, 8)-1)*np.pi/8
95     return function(t)
96
97 if __name__ == "__main__":
98
99     t1 = np.pi / 6; t2 = np.pi / 2; t3 = 5 * np.pi / 6
100     A = np.matrix([[1, np.cos(t1), np.cos(2*t1)], [1, np.cos(t2), np.cos(2*t2)], [1, np.
101     cos(t3), np.cos(2*t3)]])
102     B = np.matrix([[1, 2, 3], [4, 5, 6], [7, 8, 9]])
103     print_matrix(find_inverse(A))
104
105     def test_function(t):

```



```
105         return 2*t
106
107     calc_yo(test_function , 8)
```

```

1 # The code used for problem 3
2 # in the second mandatory assignment
3 # in MAT1120 at the University of Oslo
4
5 from assignment2 import *
6 import numpy as np
7 # Initializing matrices and t values
8
9 n = 8
10 C = np.zeros((n, n))
11 S = np.zeros((n, n))
12 t = np.pi/16 + (np.linspace(1, 8, 8)-1)*np.pi/8
13
14 for i in range(n):
15     for j in range(n):
16         C[i][j] = np.cos((j)*t[i])
17         S[i][j] = np.sin((j+1)*t[i])
18
19 # Verifying that the matrices C and S are semi-orthogonal:
20
21 print("Is C semi-orthogonal? " , is_semi_orthogonal(C))
22 print("Is S semi-orthogonal? " , is_semi_orthogonal(S))
23
24 # Finding the inverse matrices of C and S:
25
26 print("Inverse of C: ")
27 print_matrix(find_inverse(C))
28 print("Inverse of S: ")
29 print_matrix(find_inverse(S))
30
31 matrix_to_tex(find_inverse(C) , "C_matrix")
32 matrix_to_tex(find_inverse(S) , "S_matrix")

```

```

1 # This is the code used for the second mandatory assignment in
2 # MAT1120 – Linear Algebra
3 # at the University of Oslo.
4
5 import numpy as np
6 import matplotlib.pyplot as plt
7 from assignment import *
8
9 # Initializing and calculating matrices C and S
10 n = 8
11 C = np.zeros((n, n))
12 S = np.zeros((n, n))
13 t = np.pi/16 + (np.linspace(1, 8, 8)-1)*np.pi/8
14
15 for i in range(n):
16     for j in range(n):
17         C[i][j] = np.cos((j)*t[i])
18         S[i][j] = np.sin((j+1)*t[i])
19
20 def f(t):
21     # This is the only line that differs between Problem 8 and 9.
22     return (np.pi**2 - t**2)*np.exp(t/(2*np.pi))
23
24 def f_l(t):
25     return 0.5*(f(t) + f(-t))
26
27 def f_o(t):
28     return 0.5*(f(t) - f(-t))
29
30 def f_midpoint(t, yc, ys):
31     c = np.cos
32     s = np.sin
33     return yc[0] + yc[1]*c(t) + yc[2]*c(2*t) + yc[3]*c(3*t) + yc[4]*c(4*t)\
34           + yc[5]*c(5*t) + yc[6]*c(6*t) + yc[7]*c(7*t)\
35           + ys[0]*s(t) + ys[1]*s(2*t) + ys[2]*s(3*t) + ys[3]*s(4*t) + ys[4]*s(5*t)\
36           + ys[5]*s(6*t) + ys[6]*s(7*t) + ys[7]*s(8*t)
37
38
39
40
41
42
43 time_values = np.linspace(-np.pi, np.pi, 1000)
44 yl = calc_vector_y(f_l, n)
45 yo = calc_vector_y(f_o, n)
46 yc = find_inverse(C).dot(yl)
47 ys = find_inverse(S).dot(yo)
48
49 plt.plot(time_values, f(time_values), time_values, f_midpoint(time_values, yc, ys))
50 plt.savefig("Problem8.pdf")

```