

Assignment 1

Machine Learning Course

Linear Algebra

Alliance Group

Nima Goodarzi, Amin Haratian, Mobina Sedaghat

18-June-2021

Exercise 1

$Ax = b$, unique solution $\implies A$ is non-singular

To solve this problem, we use proof by contradiction, so we assume that A is singular hence, A is non-invertible (the columns are dependent) . Then, there would be a non-zero vector in its null space, call this vector z

$$Az = 0 \rightarrow A(x - y) = 0$$

because A is not invertible $\rightarrow x \neq y \diamond$

$$\text{Also } Ax - Ay = 0 \rightarrow Ax = Ay \diamond \diamond$$

According to the hypothesis of the problem $Ax = b$, so by looking at $\diamond \diamond$ therefore $Ay = b$. This is implying that $x = y$ which is in contradiction with \diamond . accordingly, the proposition has been proven.

Now, we need to look at the problem the other way around:

A is non-singular $\implies Ax = b$, unique solution

$$Ax = b$$

$$A^{-1}(Ax) = A^{-1}b$$

$$(A^{-1}A)x = A^{-1}b$$

$$Ix = A^{-1}b$$

$$x = A^{-1}b$$

Exercise 2

We know that $E = e(I_m)$. So we just need to find the value of e_1 as if $E_1 E = e_1(e(I))$ which means $e_1(e(A)) = A$.

First scenario: we assume e is a elementary row operation which multiplies the r row of A by the non-zero scalar c . Now we define e_1 as a type I elementary row operation which multiplies the r row of A by the non-zero scalar $\frac{1}{c}$. So:

$$\begin{aligned} a) r \neq i : e_1(e(A))_{ij} &= e_1(A_{ij}) = A_{ij} \\ b) r = i : e_1(e(A))_{rj} &= \frac{1}{c}e(A)_{rj} = \frac{1}{c}.c(A_{rj}) = A_{rj} \\ &\rightarrow e_1(e(A)) = A \end{aligned}$$

Second scenario: we assume that e is a elementary row operation which replaces the r^{th} row of A with $r^{\text{th}} + c.s^{\text{th}}$. Now we let e_1 be a type II elementary row operation which replaces the r^{th} row of A with $r^{\text{th}} + (-c).s^{\text{th}}$. for each i , $1 \leq i \leq m$, and each j , $1 \leq j \leq n$:

$$\begin{aligned} a) r \neq i : e_1(e(A))_{ij} &= e_1(A_{ij}) = A_{ij} \\ b) r = i : e_1(e(A))_{ry} &= e_1(e(A))_{ry} = e(A)_{ry} + (-c)e(A_{sj}) = A_{rj} + cA_{sj} - cA_{sj} = A_{ij} \\ &\rightarrow e_1(e(A)) = A \end{aligned}$$

Third scenario: we assume e is a elementary row operation which replaces the r^{th} row of A with the s^{th} row. Now we define that e_1 is a Type III row operation which replaces the s^{th} row with the r^{th} row. So:

$$\begin{aligned} a) i \neq r, s : e_1(e(A_{ij})) &= e_1(A_{ij}) = A_{ij} \\ b) i = r : e_1(e(A_{ij})) &= e_1(A_{sj}) = A_{rj} \\ c) i = s : e_1(e(A_{sj})) &= e_1(A_{rj}) = A_{sj} \\ &\rightarrow e_1(e(A)) = A \end{aligned}$$

Exercise 3

Part 1

$$\begin{bmatrix} a_1 & c_1 & 0 \\ e_2 & a_2 & c_2 \\ 0 & e_3 & a_3 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix} = \begin{bmatrix} a_1 v_1 + c_1 v_2 \\ e_2 v_1 + a_2 v_2 + c_2 v_3 \\ e_3 v_2 + a_3 v_3 \end{bmatrix}$$

By looking at the mentioned equation, we can deduct the following algorithm:

$$v_n = e_n v_{n-1} + a_n v_n + c_n v_{n+1}$$

```
1 import numpy as np
2 # import time as time
3
4
5 #Since the array's range in the algorithms differs from the ones in the
   code, employed equations are different as well.
6
7
8
9 def TridMult(a, c, e, v):
10     answer=np.zeros(len(a))
11     c=np.append(c,0)
12     v=np.append(v,0)
13     e=np.append(e,0)
14     e=np.insert(e,0,0,0)
15     # tic=time.time()
16     for i in range(len(a)):
17         answer[i]=(a[i]*v[i]+c[i]*v[i+1]+e[i]*v[i-1])
18     # toc=time.time()
19     print('answer= ', answer)
20     # print('time= ', toc-tic)
21
22
23
24
25 while True:
26     a=np.array([int(j) for j in input("Enter your a array, for instance:
       1,2,3,4,...: \n").split(',')])
27     e=np.array([int(j) for j in input("Enter your e array:
       \n").split(',')])
28     c=np.array([int(j) for j in input("Enter your c array:
       \n").split(',')])
29     v=np.array([int(j) for j in input("Enter your v array:
       \n").split(',')])
30     # a=np.ones(4000000)
31     # c=np.ones(3999999)
32     # v=np.ones(4000000)
33     # e=np.ones(3999999)
```

```

34     if len(a) == len(v) and len(a) == len(e)+1 and len(a) == len(c)+1:
35         break
36     else:
37         print("Your given arrays are not valid. Check your input arrays
38             again.")
39 TridMult(a,c,e,v)

```

time complexity : $\mathcal{O}(5n)$

Part 2

A=LU

$$\begin{bmatrix} a_1 & c_1 & 0 \\ e_2 & a_2 & c_2 \\ 0 & e_3 & a_3 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ \beta_2 & 1 & 0 \\ 0 & \beta_3 & 1 \end{bmatrix} \begin{bmatrix} \alpha_1 & c_1 & 0 \\ 0 & \alpha_2 & c_2 \\ 0 & 0 & \alpha_3 \end{bmatrix}$$
$$\begin{bmatrix} a_1 & c_1 & 0 \\ e_2 & a_2 & c_2 \\ 0 & e_3 & a_3 \end{bmatrix} = \begin{bmatrix} \alpha_1 & c_1 & 0 \\ \alpha_1\beta_2 & \beta_2c_1 + \alpha_2 & c_2 \\ 0 & \alpha_2\beta_3 & \beta_3c_2 + \alpha_3 \end{bmatrix}$$

$$a_1 = \alpha_1$$

$$\beta_2 = \frac{e_2}{\alpha_1}, \alpha_2 = a_2 - \beta_2c_1$$

$$\beta_3 = \frac{e_3}{\alpha_2}, \alpha_3 = a_3 - \beta_3c_2$$

By looking at the values of α and β which are calculated above, we can see that such pattern has been formed:

$$\beta_n = \frac{e_n}{\alpha_{n-1}}$$

$$\alpha_n = a_n - \beta_nc_{n-1}$$

Part 3

```
1 import numpy as np
2 # import time as time
3
4
5
6 def LUdecomp(a,c,e):
7     beta=np.zeros(len(c))
8     alpha=np.zeros(len(a))
9     alpha[0]=a[0]
10    # tic=time.time()
11    for i in range (len(e)):
12        beta[i]=e[i]/alpha[i]
13        alpha[i+1]=a[i+1]-beta[i]*c[i]
14    # toc=time.time()
15    print('alpha= ', alpha, '\n', 'beta= ', beta)
16    # print('time= ', toc-tic)
17
18
19
```

```

20
21 while True:
22     a=np.array([int(j) for j in input("Enter your a array, for instance:
23         1,2,3,4,...: \n").split(',')]])
24     e=np.array([int(j) for j in input("Enter your e array:
25         \n").split(',')]])
26     c=np.array([int(j) for j in input("Enter your c array:
27         \n").split(',')]])
28     # a=np.ones(800000)*5
29     # c=np.ones(799999)*3
30     # e=np.ones(799999)*4
31     if len(a) == len(e)+1 and len(a) == len(c)+1:
32         break
33     else:
34         print("Your given arrays are not valid. Check your input arrays
35             again.")
36
37 LUdecomp(a,c,e)

```

Part 4

time complexity : $\mathcal{O}(3n)$

Part 5

$$Ax = B \rightarrow A = LU$$

$$LUx = B \rightarrow Ly = B, Ux = y$$

$$\begin{bmatrix} 1 & 0 & 0 \\ \beta_2 & 1 & 0 \\ 0 & \beta_3 & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} B_1 \\ B_2 \\ B_3 \end{bmatrix}$$

$$B_1 = y_1$$

$$B_2 = B_2y_1 + y_2 \rightarrow y_2 = B_2 - B_2y_1$$

$$B_3 = B_3y_2 + y_3 \rightarrow y_3 = B_3 - B_3y_2$$

Once again we see that a pattern has emerged:

$$y_n = B_n - B_ny_{n-1}, n \neq 0$$

Also:

$$\alpha_3x_3 = y_3 \rightarrow x_3 = \frac{y_3}{\alpha_3}$$

$$\alpha_2x_2 = y_2 \rightarrow x_2 = \frac{y_2 - c_2x_3}{\alpha_2}$$

$$\alpha_1x_1 = y_1 \rightarrow x_1 = \frac{y_1 - c_1x_2}{\alpha_1}$$

Now if we reverse a,y,c, and x:

$$x_1 = \frac{y_1}{\alpha_1}$$

$$x_2 = \frac{y_2 - c_1x_1}{\alpha_2}$$

$$x_3 = \frac{y_3 - c_2x_2}{\alpha_3}$$

According to the calculations done above the following pattern would be the right algorithm:

$$x_n = \frac{y_n - c_{n-1}x_{n-1}}{\alpha_n}, n \neq 0$$

Each calculated answer needs to be reversed once again.

Part 6

```
1 import numpy as np
2 # import time as time
3
4
5 def Linsolver(alpha, beta, c, B):
6     y=np.ones(len(B))
7     y[0]=B[0]
8     # tic=time.time()
9     for i in range(len(B)-1):
10         y[i+1]=B[i+1]-beta[i]*y[i]
11     # print('y=', y)
12     y=y[:-1]
13     c=c[:-1]
14     alpha=alpha[:-1]
15     x=np.ones(len(B))
16     x[0]=y[0]/alpha[0]
17     for i in range(len(y)-1):
18         x[i+1]=(y[i+1]-c[i]*x[i])/alpha[i+1]
19     x=x[:-1]
20     # toc=time.time()
21     print('x=', x)
22     # print('time= ', toc-tic)
23
24 while True:
25
26     alpha=np.array([int(j) for j in input("Enter your alpha array, for
27         instance: 1,2,3,4,...: \n").split(',')])
28     beta=np.array([int(j) for j in input("Enter your beta array:
29         \n").split(',')])
30     c=np.array([int(j) for j in input("Enter your c array:
31         \n").split(',')])
32     B=np.array([int(j) for j in input("Enter your right-hand-side array,
33         B: \n").split(',')])
34     # alpha=np.ones(10000000)
35     # beta=np.ones(9999999)
36     # B=np.ones(10000000)
37     # c=np.ones(9999999)
38     if len(alpha) == len(beta)+1 and len(alpha) == len(c)+1 and len(B)
39         == len(alpha):
40         break
41     else:
42         print("Your given arrays are not valid. Check your input arrays
43             again.")
44 Linsolver(alpha, beta, c, B)
```

time complexity : $\mathcal{O}(5n)$

Part 7

$A = RR^T$, R is a lower triangular matrix.

$$R = \begin{bmatrix} \alpha_1 & 0 & 0 \\ \beta_1 & \alpha_2 & 0 \\ \epsilon_1 & \beta_2 & \alpha_3 \end{bmatrix}, R^T = \begin{bmatrix} \alpha_1 & \beta_1 & \epsilon_1 \\ 0 & \alpha_2 & \beta_2 \\ 0 & 0 & \alpha_3 \end{bmatrix}$$

$$RR^T = \begin{bmatrix} \alpha_1^2 & \alpha_1\beta_1 & \alpha_1\epsilon_1 \\ \alpha_1\beta_1 & \alpha_2^2 + \beta_1^2 & \alpha_2\beta_2 \\ \alpha_1\epsilon_1 & \alpha_2\beta_2 & \alpha_3^2 + \beta_2^2 \end{bmatrix} = \begin{bmatrix} a_1 & c_1 & 0 \\ c_1 & a_2 & c_2 \\ 0 & c_2 & a_3 \end{bmatrix}$$

$$\alpha_1\epsilon_1 = 0 \rightarrow \epsilon_1 = 0$$

$$\alpha_1^2 = a_1 \rightarrow \alpha_1 = \sqrt{a_1}$$

$$\alpha_1\beta_1 = c_1 \rightarrow \beta_1 = \frac{c_1}{\alpha_1}$$

$$\alpha_2^2 + \beta_1^2 = a_2 \rightarrow \alpha_2 = \sqrt{a_2 - \beta_1^2}$$

$$\alpha_2\beta_2 = c_2 \rightarrow \beta_2 = \frac{c_2}{\alpha_2}$$

$$\alpha_3 = \sqrt{a_3 - \beta_2^2}$$

According to the calculations above, we can extract the following equations:

$$\alpha_1 = \sqrt{a_1}$$

$$\beta_n = \frac{c_n}{\alpha_n}$$

$$\alpha_n = \sqrt{a_n - \beta_{n-1}^2}, n \neq 1$$

Part 8

```

1
2 import numpy as np
3 # import time as time
4
5
6
7
8 def Rdecomp (MD, OD):
9     alpha=np.ones(len(MD))
10    beta=np.ones(len(OD))
11    alpha[0]=(MD[0])**0.5
12    # tic=time.time()
13    for i in range(len(OD)):
14        beta[i]= OD[i]/alpha[i]
15        alpha[i+1] = (MD[i+1]-beta[i]**2)**0.5
16    # toc=time.time()
17    print('alpha= ', alpha, '\n', 'beta= ', beta)
18    # print('time= ', toc-tic)
19
20
21
22 while True:
23     # MD=np.array([int(j) for j in input("Enter your positive symetirc
24     definite main diagonal, for instance: 1,2,3,4,...: \n").split(',')])
25     # OD=np.array([int(j) for j in input("Enter your positive symetirc
26     definite outer diagonal: \n").split(',')])
27     MD=np.ones(32000000)*32
28     OD=np.ones(31999999)*8
29     if len(MD) == len(OD)+1:
30         break
31     else:
32         print("Your given arrays are not valid. Check your input arrays
33         again.")
34
35 try:
36     Rdecomp(MD, OD)
37 except:
38     print("Given matrix is not posibble to be decomposed to R and R
39     transposed")

```

time complexity : $\mathcal{O}(4n)$