Linear Algebra HW2

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- 1 [1.4] ex. 11.
- (a) $A = C + C^T$

A is symmetric because $(C+C^T)^T=C^T+(C^T)^T=C^T+C=C+C^T=A$.

(b) $B = C - C^T$

B is skew-symmetric because $(C-C^T)^T=C^T-(C^T)^T=C^T-C=-(C-C^T)=-B.$

(c) $D = C^T C$

D could be non-symmetric because the transpose of D is $(C^TC)^T = C^T(C^T)^T$ which does not equal C^TC unless C is symmetric, which it is not by definition.

(d) $E = C^T C - CC^T$

E could be non-symmetric for the same reason as D, as $(C^TC)^T \neq CC^T$ and $(CC^T)^T \neq C^TC$ in general. Hence, E does not necessarily equal its transpose.

(e) $F = (I + C)(I + C^T)$

F could be non-symmetric. The transpose of F is $(I+C)^T(I+C^T)^T=(I+C^T)(I+C)$ which is the same as F, but that does not mean F is symmetric since the multiplication of two symmetric matrices is not necessarily symmetric.

(f) $G = (I + C)(I - C^T)$

G could be non-symmetric. The transpose of G is $(I - C^T)^T (I + C)^T = (I - C)(I + C^T)$, which is not necessarily equal to G since C is not symmetric.

Thus, only A is necessarily symmetric and B is necessarily skew-symmetric, while D, E, F, and G could be non-symmetric.

2 [1.5] ex. 14.

U has the form:

$$U = \begin{bmatrix} u_{11} & u_{12} & \dots & u_{1n} \\ 0 & u_{22} & \dots & u_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & u_{nn} \end{bmatrix}$$

R has the form:

$$R = \begin{bmatrix} r_{11} & r_{12} & \dots & r_{1n} \\ 0 & r_{22} & \dots & r_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & r_{nn} \end{bmatrix}$$

The product T = UR will result in a matrix where the element at position (i, j) is computed as:

$$t_{ij} = \sum_{k=1}^{n} u_{ik} r_{kj}$$

For T to be upper triangular, $t_{ij} = 0$ for all i > j.

- 1. When i > j, since U is upper triangular, for any k < i, $u_{ik} = 0$. Also, since R is upper triangular, for any k > j, $r_{kj} = 0$. Thus, all terms in the sum $\sum_{k=1}^{n} u_{ik} r_{kj}$ will be zero since either u_{ik} or r_{kj} will be zero, or both. Hence, $t_{ij} = 0$ for all i > j, confirming T is upper triangular.
- 2. For the diagonal elements t_{ij} , the sum simplifies to just one term:

$$t_{jj} = \sum_{k=1}^{n} u_{jk} r_{kj} = u_{jj} r_{jj}$$

since for all k < j, $r_{kj} = 0$, and for all k > j, $u_{jk} = 0$. Therefore, the diagonal of T will be the product of the diagonals of U and R, i.e., $t_{jj} = u_{jj}r_{jj}$.

Hence, T is upper triangular and the diagonal elements of T are the products of the corresponding diagonal elements of U and R.

3 [1.5] ex. 24.

(a) If A is row equivalent to B, and B is row equivalent to C, then A is row equivalent to C.

Proof: Being row equivalent means that one matrix can be transformed into another by a sequence of elementary row operations. These operations can be represented by multiplication by elementary matrices. If A is row equivalent to B, there exists a finite sequence of elementary matrices E_1, E_2, \ldots, E_k such that:

$$B = E_k E_{k-1} \dots E_1 A$$

Similarly, if B is row equivalent to C, there exists a finite sequence of elementary matrices F_1, F_2, \ldots, F_m such that:

$$C = F_m F_{m-1} \dots F_1 B$$

Combining these two, we get:

$$C = F_m F_{m-1} \dots F_1 E_k E_{k-1} \dots E_1 A$$

Since the product of elementary matrices is also an elementary matrix (as the set of elementary matrices is closed under multiplication), this shows that A can be transformed into C by a sequence of elementary row operations. Therefore, A is row equivalent to C.

(b) Any two nonsingular $n \times n$ matrices are row equivalent.

Proof: A nonsingular matrix is one that is invertible. If we have two nonsingular matrices A and B, each can be transformed into the identity matrix I using a series of elementary row operations (since they are invertible):

$$I = E_A \cdot A$$

$$I = E_B \cdot B$$

Here, E_A and E_B are products of elementary matrices corresponding to the series of row operations that transform A and B into I, respectively. Since both A and B can be transformed into the identity matrix, they can be transformed into each other by the sequence of operations that first transforms A into I and then I into B:

$$B = E_B \cdot I = E_B \cdot E_A^{-1} \cdot A$$

 $E_B \cdot E_A^{-1}$ is again a product of elementary matrices, which proves that A and B are row equivalent.

4 [1.6] ex. 11.

For the matrix

$$A = \begin{bmatrix} A_{11} & A_{12} \\ 0 & A_{22} \end{bmatrix}$$

where A_{11} and A_{22} are nonsingular $n \times n$ matrices, the inverse of A, if it exists, can be found using the formula for the inverse of a block matrix. The inverse of A is given by

$$A^{-1} = \begin{bmatrix} A_{11}^{-1} & -A_{11}^{-1} A_{12} A_{22}^{-1} \\ 0 & A_{22}^{-1} \end{bmatrix}$$

(a) The matrix A is nonsingular because both A_{11} and A_{22} are nonsingular. Therefore, A^{-1} exists and has the block structure shown above.

To determine the block C which corresponds to the upper-right block of A^{-1} , we can multiply A by A^{-1} and set the result equal to the identity matrix:

$$AA^{-1} = \begin{bmatrix} A_{11} & A_{12} \\ 0 & A_{22} \end{bmatrix} \begin{bmatrix} A_{11}^{-1} & C \\ 0 & A_{22}^{-1} \end{bmatrix} = I$$

Multiplying out the matrices gives:

$$\begin{bmatrix} A_{11}A_{11}^{-1} \ A_{11}C + A_{12}A_{22}^{-1} \\ 0 \ A_{22}A_{22}^{-1} \end{bmatrix} = \begin{bmatrix} I \ 0 \\ 0 \ I \end{bmatrix}$$

(b) For the resulting product to be the identity matrix, the upper-right block must be zero. Therefore, we get the equation:

$$A_{11}C + A_{12}A_{22}^{-1} = 0$$

Solving for C gives:

$$C = -A_{11}^{-1}A_{12}A_{22}^{-1}$$

Thus, the complete inverse of A is:

$$A^{-1} = \begin{bmatrix} A_{11}^{-1} - A_{11}^{-1} A_{12} A_{22}^{-1} \\ 0 & A_{22}^{-1} \end{bmatrix}$$

$5 \quad [Matlab \ ex] \ ex. 5.$

Part (a):

Generate a 6×6 matrix A with entries as the floor of 10 times random numbers between 0 and 1. Generate a vector b with entries as the floor of 20 times random numbers between 0 and 1, subtracted by 10. Solve for x in Ax = b.

Part (b):

Make A singular by setting its third column to be a linear combination of the first two columns. Compute the RREF of the augmented matrix $\begin{bmatrix} A & b \end{bmatrix}$ and discuss the number of solutions.

Part (c):

Generate a new vector c as A times a new random vector. Find the RREF of $[A \ c]$ to explore the solutions of Ax = c.

Part (d):

Identify the solution associated with the free variable $x_3 = 0$ from the RREF found in part (c). Compute the residual vector c - Ax to check the solution.

Part (e):

Compute the null space of A. Determine a vector z in the null space with the third entry equal to 1.

Part (f):

Form a new solution v as w + 3z. Explain that v is a solution to Ax = c and use MATLAB to compute the residual c - Av. Discuss how the general solution of Ax = c can be expressed in terms of w and z.

```
Command Window
                                                  ---- [Matlab ex] ex. 5. (a) -----
     Solution x:
-2.006307594534478
-4.065268509691770
5.177200508420720
0.041086749285032
-3.727788369876075
3.323164918970450
     Solution from reduced row echelon form:
-2.006289308176101
-4.065260065268065
5.177215189973418
0.041095890410959
-3.72777777777778
3.323170731707317
     Difference:
1.0e-04 *
            0.182863583768622
0.004444237049483
0.146814526980066
0.091411259266275
0.105920982966801
0.058127368673944
    The system Ax = b has infinitely many solutions.

[Matlab ex] ex. 5. (c) =

Reduced Row Echelon Form of [A c]:

1 0 4 0 0 0 -40

0 1 3 0 0 0 -30

0 0 0 1 1 0 0 -7

0 0 0 0 0 1 0 -10

0 0 0 0 0 0 1 2

0 0 0 0 0 0 0 0
     Matlab ex] ex. 5. (d)
Residual vector for v:
-532 -529 -571 -571 -524 -531 -524
-573 -568 -612 -612 -570 -569 -564
-573 -568 -612 -612 -570 -569 -564
-693 -477 -513 -513 -482 -480 -475
-412 -410 -442 -442 -414 -406 -406
-700 -700 -754 -754 -750 -692 -691
      ----- [Matlab ex] ex. 5. (e) ----- Solution vector z with x3=1:
                                                            [Matlab ex] ex. 5. (f)
 General solution for Ax = c is given by w + k^*z, where k is any scalar.
```

Fig. 1. Results of [Matlab ex] ex. 5.



Fig. 2. Codes of [Matlab ex] ex. 5.

6 [Matlab ex] ex. 6(a) to 6(d).

Part (a):

This part involves determining the adjacency matrix A for the given graph. An adjacency matrix is a square matrix used to represent a finite graph. The elements of the matrix indicate whether pairs of vertices are adjacent or not in the graph.

Part (b):

Once we have the adjacency matrix A, we compute A^2 . In graph theory, the entry in the i-th row and j-th column of A^2 represents the number of walks of length 2 from vertex V_i to vertex V_j . We specifically look at the walks from V_1 to V_7 , V_4 to V_5 , to V_6 , and V_8 to V_3 .

Part (c):

Similarly, computing A^4 , A^6 , and A^8 gives us the number of walks of lengths 4, 6, and 8, respectively, between any two vertices i and j. The script focuses on the number of such walks from V_1 to V_7 .

Part (d):

This part is similar to part (c), but for walks of odd lengths 3, 5, and 7. Again, the entry in the *i*-th row and *j*-th column of A^n represents the number of walks of length n from vertex V_i to vertex V_j , with the specific walks from V_1 to V_7 being of interest.

The conjectures made at the end of the script seem to predict whether there are an even or odd number of walks of certain lengths between the vertices V_1 and V_7 . The MATLAB code will print the number of walks of these lengths, and the conjectures are based on whether the sum of the number of walks of certain lengths is even or odd.

This explanation corresponds to the parts of the MATLAB code provided and assumes that the graph structure does not change. If the graph structure changes, the adjacency matrix A must be updated accordingly to reflect the correct connections between the vertices.

Adjacen	ev matr	ix A:	[Mat					
0	1 0	0	1	0 1 0	0	0	0	
1		1	0	1	0	0	0	
0	1	0	0	0	1	0	0	
1	0	0	0	0	0	1	1	
0	0 1 0	0 0 0 1	0	0	0 0 0	0	0	
0	0	0	1	0	1	0	1	
0	0	0	0	0 0 0 0 0	0	1	0	
Number (of wal)	cs of]	length	2 for	each p	pair of	f vert	ices:
1								
0								
			[Mat	lab ex	ex.	6. (=)	
A^4:	0	6	0	6	0	7	0	
0	12	0	7	0	7	0	7	
6	0	7	0	6	0	7	0	
0	7	0	7	0	6	0		
6	0	6	0	7	0	7	0	
0	0 7 0	7	6	6 0 7 0 7	7	0	6	
7	7	0	0 7 0 7 0 6 0	0	6 0 7 0 6	12	0	
0	,	0		0		0	,	
A^6:								
33	0	32	0	32	0	40	0	
0	57	0	40	0	40	0	40	
32	57 0 40	33	0	32	0	40	0	
0	40	0	33	0	32	0	32	
32	0	32	0	33	0	40	0	
0	40	0	32	0	33	0	32	
40	40	40	3.2	40	3.3	40 0 40 0 40 0 57	33	
0	40	0	32	Ü	32		55	
A^8:								
170	0 291	169	0	169	0	217 0 217 0 217	0	
0	291	0	217	0	217	0	217	
169	0 217 0 217	170 0 169	0	169	0	217	0	
0	217	0	170	0	169	0	169	
169	0	169	0	170	0	217	0	
0	217	0	169	0	170	0	169	
217	217 0 217	217	0	217	1.00	291	0 170	
0	211							
A^3:			[Mat	lab ex] ex.	6. (d) ===	
A3:	4	0	3	0 4 0 2	2	0	2	
4	0	0 4 0	0	4			0	
0	4	0	2	0	0 3 0	0	2	
3	0	2	0	2	0	4	0	
0	4	0	2	0	2	0	3	
2	0	2 0 3	0	2	0	3 0 4 0 4	0	
0	0 4 0 3	0	4	0			4	
2	0	2	0	3	0	4	0	
A^5:								
0	19	0 19 0 13	14	0 19 0 13	13	0	13 0	
19	0	19	0	19	0	21	0	
0	19	0	13	0	14	0	13	
14	0		13	13	0 13	19	0	
0	19	0 14	13	13	13	0	14	
13	0	0	0	13	10	19	0 19	
13	21	13	19	14	13 0 19 0	19	19	
A^7:								
0	97	0	73 0 72 0	0 97 0	72	0	72	
97	0	97	0	97	0	120	0	
0	97	0	72	0	73	0 97	72	
73	0	72			70		73	
72	97 0	73	0	72	0	97	0	
0	120	0	97	0 72 0 73	97	0	97	
72	0	72	0	73	0	97 0 97	0	
Number o	of walk	s of l	ength	2:				
1								
0								
0								
Number o	of polic	o ne .	angth.	4 -				
7	-r walk	or 1	ength	**				
Number o	of walk	s of l	ength	6:				
40 Number	of walk	s of 1	ength	8:				
217								
Number o	of walk	s of 1	ength	3:				
Number o	of walk	s of 1	ength	5:				
Number o	of walk	s of l	ength	7:				
Conjecti 0 0	ire for	walks	of ev	en len	gth fr	om V1	to V 7:	
~								
1								

Fig. 3. Results of [Matlab ex] ex. 6(a) to 6(d).

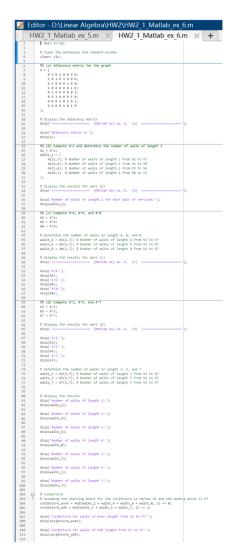


Fig. 4. Codes of [Matlab ex] ex. 6(a) to 6(d).