

HW 1

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1. Let
- B
- be a
- 4×4
- matrix to which we apply the following operations:

1. double column 1,
2. halve row 3,
3. add row 3 to row 1,
4. interchange columns 1 and 4,
5. subtract row 2 from each of the other rows,
6. replace column 4 by column 3,
7. delete column 1 (so that the column dimension is reduced by 1).

- (a) Write this result as a product of eight matrices.
 (b) Write it again as a product ABC (same B) of three matrices.

a) Let $B = \begin{bmatrix} b_{11} & b_{12} & b_{13} & b_{14} \\ b_{21} & b_{22} & b_{23} & b_{24} \\ b_{31} & b_{32} & b_{33} & b_{34} \\ b_{41} & b_{42} & b_{43} & b_{44} \end{bmatrix}$

$$A_1 = \begin{bmatrix} 2 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$A_2 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & \frac{1}{2} & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$A_3 = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$A_4 = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix}$$

$$A_5 = \begin{bmatrix} 1 & -1 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & -1 & 0 & 1 \end{bmatrix}$$

$$A_6 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$A_7 = \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Then we have $A_5 A_3 A_2 B A_1 A_4 A_6 A_7$

$$\begin{aligned}
 b) \quad A &= A_5 A_3 A_2 = \begin{bmatrix} 1 & -1 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & -1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & \frac{1}{2} & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \\
 &= \begin{bmatrix} 1 & -1 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & -1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & \frac{1}{2} & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \\
 &= \begin{bmatrix} 1 & -1 & \frac{1}{2} & 0 \\ 0 & 1 & 0 & 0 \\ 0 & -1 & \frac{1}{2} & 0 \\ 0 & -1 & 0 & 1 \end{bmatrix}
 \end{aligned}$$

$$\begin{aligned}
 C &= A_1 A_4 A_6 A_7 = \begin{bmatrix} 2 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \\
 &= \begin{bmatrix} 0 & 0 & 0 & 2 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix} \\
 &= \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix}
 \end{aligned}$$

Results from Matlab:

Set up B:

Interchange columns 1 and 4:				
B =				
0.8909	0.1493	0.8143	0.1966	0.5046
0.9593	0.2575	0.2435	0.2511	0.2511
0.5472	0.8407	0.9293	0.6160	0.3080
0.1386	0.2543	0.3500	0.4733	0.4733
Subtract row 2 from each of the other rows:				
0.2535	0.3121	1.0354	0.4104	
0.2511	0.2575	0.2435	1.9186	
0.3080	0.4204	0.4646	0.5472	
0.4733	0.2543	0.3500	0.2772	
Double column 1:				
0.2535	0.3121	1.0354	0.4104	
0.2511	0.2575	0.2435	1.9186	
0.3080	0.4204	0.4646	0.5472	
0.4733	0.2543	0.3500	0.2772	
Halve row 3:				
0.2535	0.3121	1.0354	0.4104	
0.2511	0.2575	0.2435	1.9186	
0.0569	0.1629	0.2211	-1.3714	
0.4733	0.2543	0.3500	0.2772	
Add row 3 to row 1:				
0.2535	0.3121	1.0354	0.4104	
0.2511	0.2575	0.2435	1.9186	
0.0569	0.1629	0.2211	-1.3714	
0.4733	0.2543	0.3500	0.2772	
Replace column 4 by column 3:				
0.2535	0.3121	1.0354	1.0354	
0.2511	0.2575	0.2435	0.2435	
0.0569	0.1629	0.2211	0.2211	
0.2222	-0.0032	0.1065	-1.6413	
0.2535	0.3121	1.0354	1.0354	
0.2511	0.2575	0.2435	0.2435	
0.0569	0.1629	0.2211	0.2211	
0.2222	-0.0032	0.1065	0.1065	

Delete column 1:

B =

0.3121	1.0354	1.0354
0.2575	0.2435	0.2435
0.1629	0.2211	0.2211
-0.0032	0.1065	0.1065

Compute matrix A:

A =

1.0000	-1.0000	0.5000	0
0	1.0000	0	0
0	-1.0000	0.5000	0
0	-1.0000	0	1.0000

Compute matrix C:

C =

0	0	0
1	0	0
0	1	1
0	0	0

Result for ABC:

ans =

0.3121	1.0354	1.0354
0.2575	0.2435	0.2435
0.1629	0.2211	0.2211
-0.0032	0.1065	0.1065

3.

2.1. Show that if a matrix A is both triangular and unitary, then it is diagonal.

Suppose A is triangular and unitary

Corollary: We claim if A is a nonsingular $m \times m$ triangular matrix then A^{-1} is also a triangular matrix of same kind.

proof. Suppose A is upper-triangular

Note one way to find inverse of a triangular matrix is by back substitution.

$$\Rightarrow \left[\begin{array}{cccc|ccc} a_{11} & a_{12} & \cdots & a_{1m} & 1 & 0 & 0 & \cdots & 0 \\ 0 & a_{22} & & & 0 & 1 & 0 & & \\ \vdots & \vdots & & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & a_{mm} & 0 & 0 & 0 & \cdots & 1 \end{array} \right]$$

Where we use row $m-1$ minus $(\text{row } m/a_{mm}) \cdot \alpha$ for the first operation

Then we use row $m-2$ minus $(\text{row } m-1/a_{m-1,m-1}) \cdot \beta$ for the second

Then we keep doing similar operation.

Note the upper triangular structure is reserved by these operations.

Thus we conclude A^{-1} is also upper triangular.

(Note: lower-triangular use similar arithmetic).



Since A is also unitary we have:

$A^* = A^{-1} \rightarrow A^*$ is also triangular with the same kind

Since the transpose is also triangular we have: A' is diagonal.



2.2. The Pythagorean theorem asserts that for a set of n orthogonal vectors $\{x_i\}$,

$$\left\| \sum_{i=1}^n x_i \right\|^2 = \sum_{i=1}^n \|x_i\|^2.$$

(a) Prove this in the case $n = 2$ by an explicit computation of $\|x_1 + x_2\|^2$.

(b) Show that this computation also establishes the general case, by induction.

a) Claim: For a set of 2 orthogonal vectors: $\{x_1, x_2\}$

$$\|x_1 + x_2\|^2 = \|x_1\|^2 + \|x_2\|^2$$

$$\text{Note } \|x_1 + x_2\|^2 = (x_1 + x_2)^*(x_1 + x_2)$$

Note inner product is bilinear so we have:

$$(x_1^* + x_2^*)(x_1 + x_2) = x_1^* x_1 + x_2^* x_1 + x_1^* x_2 + x_2^* x_2$$

Since x_1 and x_2 are orthogonal we have:

$$x_2^* x_1 = x_1^* x_2 = 0$$

$$\begin{aligned} \text{Thus } (x_1^* + x_2^*)(x_1 + x_2) &= x_1^* x_1 + x_2^* x_2 \\ &= \|x_1\|^2 + \|x_2\|^2 \end{aligned}$$

Thus the theorem is true in the case of $n=2$. \square

b) Base case: $n=2$

We have already proven $n=2$ case in part a).

Induction Step:

Suppose theorem is true for $n=k$.

$$\begin{aligned} \Rightarrow \left\| \sum_{i=1}^{k+1} x_i \right\|^2 &= \left\| \sum_{i=1}^k x_i + x_{k+1} \right\|^2 \\ &= \left(\sum_{i=1}^k x_i + x_{k+1} \right)^* \left(\sum_{i=1}^k x_i + x_{k+1} \right) \end{aligned}$$

Note inner product is bilinear, we have:

$$\begin{aligned} \left\| \sum_{i=1}^{k+1} x_i \right\|^2 &= \left(\sum_{i=1}^k x_i \right)^* \left(\sum_{i=1}^k x_i \right) + x_{k+1}^* \left(\sum_{i=1}^k x_i \right) \\ &\quad + \left(\sum_{i=1}^k x_i \right)^* x_{k+1} + x_{k+1}^* x_{k+1} \end{aligned}$$

Note since all the vectors are orthogonal

$$\Rightarrow x_{k+1}^* \left(\sum_{i=1}^k x_i \right) = \left(\sum_{i=1}^k x_i \right)^* x_{k+1} = 0$$

$$\begin{aligned} \Rightarrow \left\| \sum_{i=1}^{k+1} x_i \right\|^2 &= \sum_{i=1}^k \|x_i\|^2 + \|x_{k+1}\|^2 \\ &= \sum_{i=1}^{k+1} \|x_i\|^2 \end{aligned}$$

We have proven the statement by induction \square

2.6. If u and v are m -vectors, the matrix $A = I + uv^*$ is known as a *rank-one perturbation of the identity*. Show that if A is nonsingular, then its inverse has the form $A^{-1} = I + \alpha uv^*$ for some scalar α , and give an expression for α . For what u and v is A singular? If it is singular, what is $\text{null}(A)$?

① Suppose u and v are non-zero m -vectors

Let A be nonsingular and $A = I + uv^*$

Then note:

$$(I + \alpha uv^*)(I + uv^*) = I + \alpha uv^* + uv^* + \alpha uv^*uv^* \\ = I + (\alpha + 1 + \alpha v^*u)uv^*$$

Then note:

$$u \cdot v^* = \begin{bmatrix} u_1 \\ \vdots \\ u_m \end{bmatrix} [v_1, \dots, v_m] \quad v^* u = [v_1, \dots, v_m] \begin{bmatrix} u_1 \\ \vdots \\ u_m \end{bmatrix}$$

matrix associativity and the fact v^*u is scalar

$$= \begin{bmatrix} u_1 v_1 & u_1 v_2 & \dots & u_1 v_m \\ \vdots & \vdots & & \vdots \\ u_m v_1 & u_m v_2 & \dots & u_m v_m \end{bmatrix} = v_1 u_1 + v_2 u_2 + \dots + v_m u_m$$

Then note $u_i v_j + 1$ is the coefficient when we calculate the determinant of A .

Since A is nonsingular we have $I + v^*u \neq 0$

Thus for $\alpha + 1 + \alpha v^*u$ we can choose α , s.t.

$$\alpha + 1 + \alpha v^*u = 0 \Rightarrow \alpha = -\frac{1}{1 + v^*u}$$

Then we have $(I + \alpha uv^*) \cdot A = I \Rightarrow A^{-1} = I + \alpha uv^*$

where $\alpha = -\frac{1}{1 + v^*u}$

② A is singular if $I + v^*u = 0$

We have proven this statement from part ①.

③ If A is singular then: $I + v^*u = 0$

Let $w \in \text{null}(A)$

$$\Rightarrow Aw = (I + uv^*)w = 0 \\ = w + uv^*w \quad \text{matrix associativity and the fact } v^*w \text{ is scalar} \\ = w + (v^*w)u \\ \Rightarrow w + (v^*w)u = 0$$

$$w = -(v^*w)u$$

Note v^*w is a scalar

$$\Rightarrow w \in \text{span}(u)$$

$$\Rightarrow \text{Null}(A) = \text{span}(u)$$