

Exercise P8: On page 12 of the textbook, equation (2.4) says $(AB)^* = B^*A^*$. Prove this by showing the matrix entries are equal.

Solution:

Suppose that A is an $m \times n$ matrix and B is a $n \times l$ matrix. Considering the following $(AB)^*$ using the entry definition of matrix-matrix multiplication defined in equation (1.5) we know the following,

$$ab_{i,j} = \sum_{k=1}^n a_{i,k}b_{k,j}.$$

Taking the adjoint of AB to get AB^* just requires us to swap the indices i, j in the left hand side,

$$(ab)_{j,i}^* = \sum_{k=1}^n a_{i,k}b_{k,j}.$$

Now we can consider B^*A^* and apply the same formula,

$$(b)^*(a)_{i,j}^* = \sum_{k=1}^n (b)_{i,k}^*(a)_{k,j}^*.$$

Note that the adjoint operation can be described with the following,

$$b_{i,j} = (b)_{j,i}^*,$$

$$a_{i,j} = (a)_{j,i}^*.$$

Thus by substitution the following is true,

$$\sum_{k=1}^n a_{i,k}b_{k,j} = \sum_{k=1}^n (b)_{k,j}^*(a)_{i,k}^*.$$

Therefore $(AB)^* = B^*A^*$.

Exercise P9: On page 21 of the textbook, equation (3.10) gives a formula of the ∞ -norm of an $m \times n$ matrix. Let a_i denote the i^{th} row of A and prove it,

$$\|A\|_{\infty} = \max_{1 \leq i \leq m} \|a_i\|_1$$

Solution:

Suppose an $m \times n$ matrix A . Now consider the set $\{x \in \mathbb{C}^n : \max_{1 \leq j \leq n} |x_j| = 1\}$ and note that the set of vectors Ax satisfy,

$$\|Ax\|_{\infty} = \|a_i^*x\|_{\infty} \leq \max_{1 \leq i \leq n} \|a_i^*x\|_{\infty}.$$

Choosing x such that all the entries have the property that $|x_i| = 1$ we can maximize each inner product $\|a_i\|_1$. Note that this turns the inner product into the same operation as the one-norm for each a_i . Therefore the product $\|A\|_{\infty}$ attains the upper bound and we get the following,

$$\|A\|_{\infty} = \max_{1 \leq i \leq m} \|a_i\|_1.$$

Exercise 2.6: If u and v are m -vectors, the matrix $A = I + uv^*$ is known as the 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 the $\text{null}(A)$.

Solution:

Suppose that A is nonsingular and u and v are m -vectors such that $A = I + uv^*$. Consider the following equation for some scalar α ,

$$(I + uv^*)(I + \alpha uv^*) = I.$$

Expanding the product,

$$\begin{aligned}(I + uv^*)(I + \alpha uv^*) &= I, \\ \alpha uv^* + uv^* + uv^* \alpha uv^* + I &= I, \\ \alpha uv^* + uv^* + \alpha uv^* uv^* &= 0.\end{aligned}$$

Note that v^*u is a scalar, so the following applies,

$$\begin{aligned}\alpha uv^* + uv^* + \alpha u(v^*u)v^* &= 0, \\ \alpha uv^* + uv^* + \alpha(v^*u)uv^* &= 0, \\ (\alpha + 1 + \alpha(v^*u))uv^* &= 0.\end{aligned}$$

When $uv^* = 0$ we get the trivial case where $A = I$. Solving the other factor for α ,

$$\begin{aligned}\alpha + 1 + \alpha(v^*u) &= 0, \\ \alpha + \alpha(v^*u) &= -1, \\ \alpha(1 + (v^*u)) &= -1, \\ \alpha &= \frac{-1}{(1 + (v^*u))}.\end{aligned}$$

Thus A has an inverse of the form $I + \alpha uv^*$ when $\alpha = \frac{-1}{(1+(v^*u))}$ and $v^*u \neq -1$.

To show when A is singular, consider all u and v such that $u^*v = -1$. Implicit in this consideration is the fact that $u \neq 0$ and $v \neq 0$. Evaluating Au we get the following,

$$Au = (I + uv^*)u = Iu + uv^*u = u + (-1)u = 0.$$

Since $u \neq 0$, A must be singular. Furthermore substituting any vector in the $\text{span}(u)$ we also get 0 so by definition $\text{span}(u) = \text{Null}(A)$.

Exercise 3.2: Let $\|\cdot\|$ denote any norm on \mathbb{C}^m and also the induced matrix norm on $\mathbb{C}^{m \times m}$. Show that $\rho(A) \leq \|A\|$, where $\rho(A)$ is the spectral radius of A , i.e., the largest absolute value

$|\lambda|$ of an eigenvalue λ of A .

Solution:

By definition $\|A\|_{m \times m}$ is the supremum of the ratios, $\|Ax\|_m/\|x\|_m$. Now note that by the definition of the eigenvalue, for some eigenvector v and the corresponding eigenvalue λ we know the following.

$$Av = \lambda v.$$

Taking the norm of both sides we get,

$$\|Av\|_m = \|\lambda v\|_m.$$

Applying the linearity of vector norms, and solving for $|\lambda|$,

$$\begin{aligned}\|Av\|_m &= |\lambda| \|v\|_m, \\ \frac{\|Av\|_m}{\|v\|_m} &= |\lambda|.\end{aligned}$$

Consider some \hat{v} such that $|\lambda|$ is maximized and we get that,

$$\rho(A) = \max |\lambda| = \frac{\|A\hat{v}\|_m}{\|\hat{v}\|_m}.$$

Thus by definition $\rho(A)$ is contained in the set of all ratios $\|Ax\|_m/\|x\|_m$, where $\|A\|_{m \times m}$ is the supremum, therefore we know

$$\rho(A) \leq \|A\|.$$

Exercise 3.3: Vector and matrix p -norms are related by various inequalities, often involving the dimensions m or n . For each of the following, verify the inequality and give an example of a nonzero vector or matrix (for general m, n) for which the equality is achieved. In this problem x is an m -vector and A is an $m \times n$ matrix.

$$1. \|x\|_\infty \leq \|x\|_2$$

Solution:

Consider the definition of the vector ∞ -norm,

$$\|x\|_\infty = \max_{1 \leq i \leq m} |x_i|.$$

Clearly we can replace the absolute value operator by squaring then square-rooting the max term. Doing so we get the following,

$$\|x\|_\infty = \max_{1 \leq i \leq m} \sqrt{(x_i)^2}.$$

If we add the square of the remaining x_i terms it must be the case that we produce a sum larger than just the maximum x_i term, thus

$$\|x\|_\infty = \max_{1 \leq i \leq m} \sqrt{(x_i)^2} \leq \sqrt{\sum_{i=1}^m (x_i)^2} = \|x\|_2.$$

For an example consider the vector $\hat{x} = [2, 1]$ and note that $\|\hat{x}\|_\infty = 2 \leq \sqrt{5} = \|\hat{x}\|_2$

2. $\|x\|_2 \leq \sqrt{m}\|x\|_\infty$

Solution:

Consider the definition of the vector 2-norm.

$$\|x\|_2 = \sqrt{\sum_{i=1}^m (x_i)^2}.$$

Replacing every x_i in the sum with the $\max_{1 \leq i \leq m} x_i$ we get the following,

$$\|x\|_2 \leq \sqrt{\sum_{i=1}^m (\max_{1 \leq i \leq m} x_i)^2} = \sqrt{m(\max_{1 \leq i \leq m} x_i)^2} = \sqrt{m} \max_{1 \leq i \leq m} |x_i| = \sqrt{m}\|x\|_\infty.$$

For an example we can once again consider $\hat{x} = [2, 1]$ and note that $\|\hat{x}\|_2 = \sqrt{5} \leq \sqrt{2} * 2 = \sqrt{m}\|\hat{x}\|_\infty$.

Exercise 4.3: Write a MATLAB program which, given a real 2x2 matrix A , plots the right singular vectors v_1 , and v_2 in the unit circle and also the left singular vectors u_1 and u_2 in the appropriate ellipse, as in Figure 4.1. Apply your program to the matrix (3.7) and also to the 2x2 matrices in Exercise 4.1

Solution:

Code:

```
function [v,u, U] = vismat(A)
% This function takes a 2x2 matrix A,
% and returns the svd visualization.
```

```
%%%% ERROR CHECKING
```

```
DimensionCheck = size(A);
if (DimensionCheck(1) ~= DimensionCheck(2))
    |(DimensionCheck(1) + DimensionCheck(2) ~= 4)
```

```

        error('A is Unexpected Size')
    end

[U, S, V] = svd(A);
Vstar = V';

%% Plotting input space vectors v_1 and v_2
theta = 0:pi/50:2*pi;
xCircle = cos(theta);
yCircle = sin(theta);
plot(xCircle, yCircle, 'Color', 'red');
hold on
plot([Vstar(1,1) 0], [Vstar(2,1) 0], 'Color', 'red')
plot([Vstar(1,2), 0], [Vstar(2,2), 0], 'Color', 'red')

%% Plotting output space vectors sigma_1u_1 and sigma_2u_2
%% Computing the matrix rotation
if U(1,1) > 0
    alpha = -acos(dot(U(:,1), [0; 1])) + pi/2;
else
    alpha = pi/2 + acos(dot(U(:,1), [0; 1]));
end

%% Parameterizing the unit ellipse formed in the output space.
xEllipse = S(1,1)*cos(theta)*cos(alpha) - S(2,2)*sin(theta)*sin(alpha);
yEllipse = S(1,1)*cos(theta)*sin(alpha) + S(2,2)*sin(theta)*cos(alpha);
plot(xEllipse, yEllipse, 'Color', 'blue');
hold on
plot([S(1,1)*U(1,1) 0], [S(1,1)*U(2,1) 0], 'Color', 'blue')
plot([S(2,2)*U(1,2) 0], [S(2,2)*U(2,2) 0], 'Color', 'blue')

v = Vstar;
u = U*S;

```

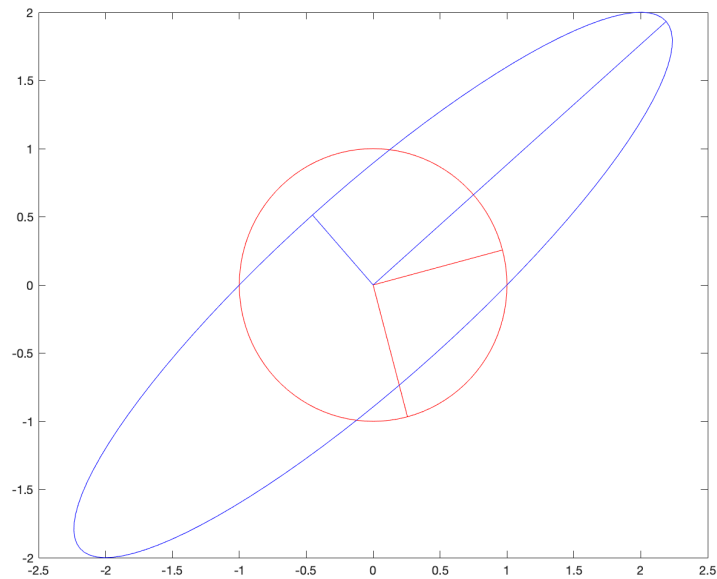
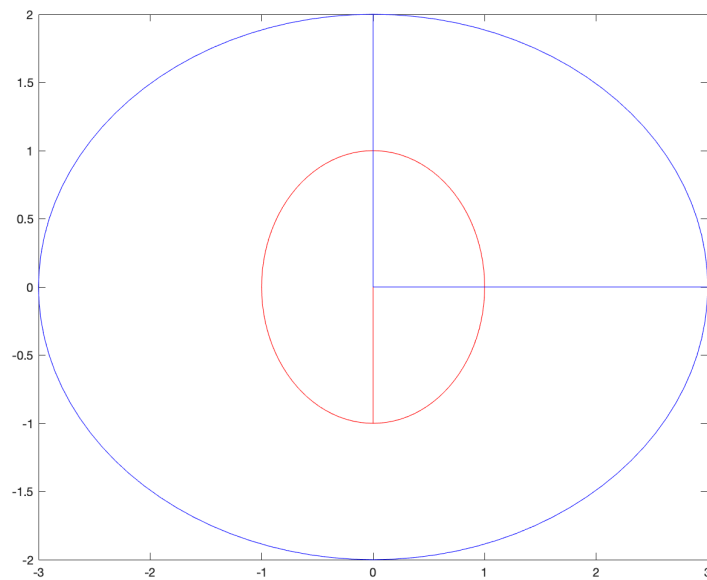
Figure 1: Vismat() with matrix from 3.7 $\begin{bmatrix} 1 & 2 \\ 0 & 2 \end{bmatrix}$ Figure 2: Vismat(A) with matrix from 4.1 $\begin{bmatrix} 3 & 0 \\ 0 & -2 \end{bmatrix}$ 

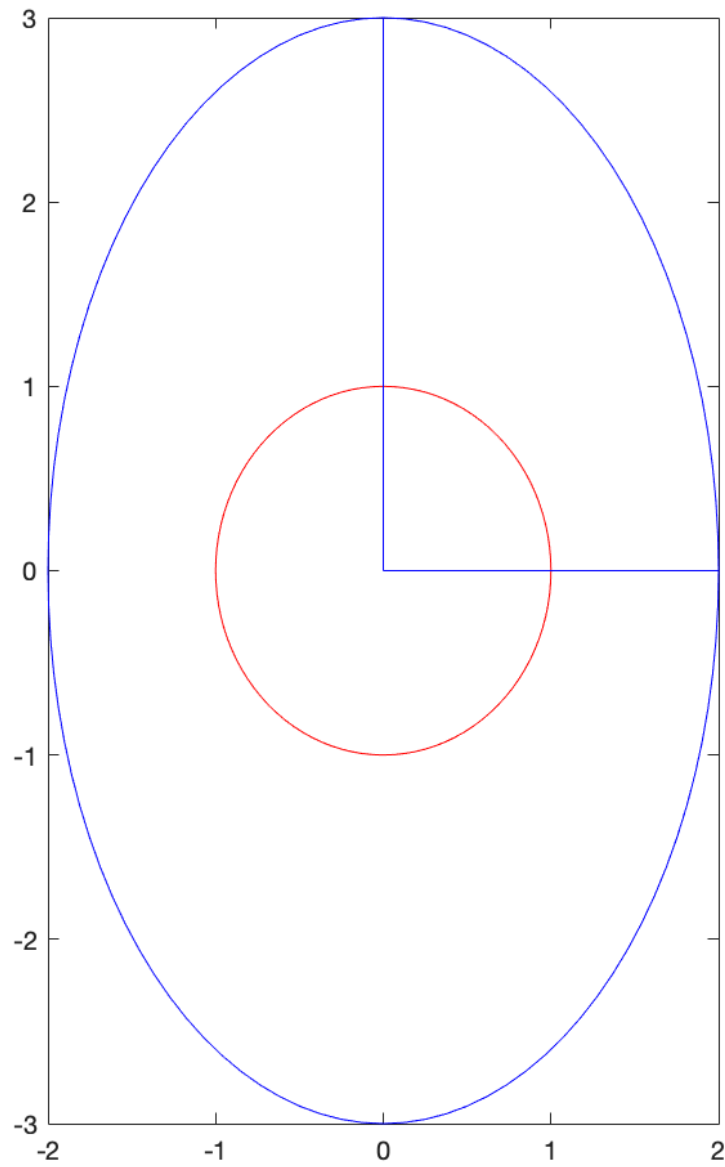
Figure 3: Vismat(A) with matrix from 4.1 $\begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix}$ 

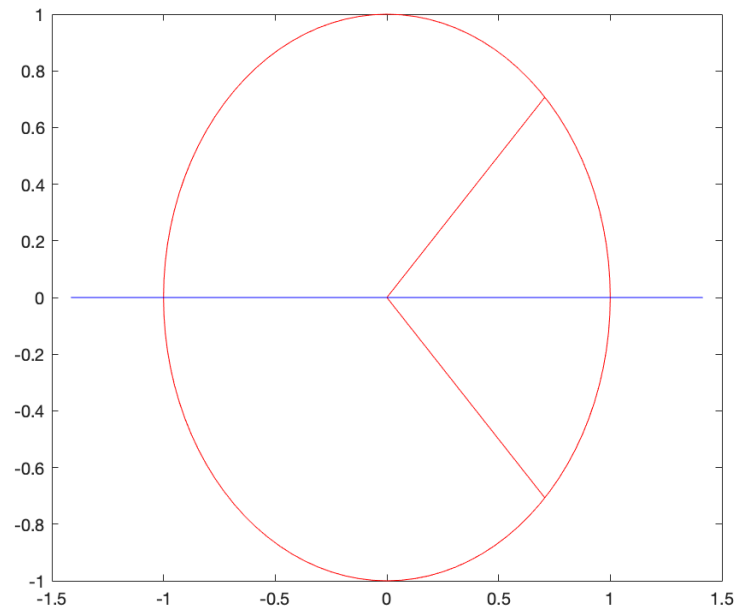
Figure 4: $\text{Vismat}(A)$ with matrix from 4.1 $\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$ 

Figure 5: Vismat(A) with matrix from 4.1 $\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$ 