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%matplotlib notebook  
import sympy as sp
```

# Homework #4

These problems cover the material in Lecture 6.

## Problem #1

Show that if  $P$  is an orthogonal projector, then  $I - 2P$  is unitary. Prove this algebraically, and give a geometric interpretation.

### Solution

If  $P$  is an orthogonal projector, then  $P = P^*$ .

Let  $A = I - 2P$ . For  $A$  to be unitary, then  $A^* = A^{-1}$  therefore to show that  $A$  is unitary then we must show that  $A^*A = I$  and this is as below;

$$A^*A = (I - 2P)^*(I - 2P) \quad (1)$$

$$= (I^* - 2P^*)(I - 2P) \quad (2)$$

$$= (I - 2P^*)(I - 2P) \quad (3)$$

$$= I - 2P - 2P^* + 4P^*P \quad (4)$$

But  $P = P^*$ ;

$$\Rightarrow A^*A = I - 4P + 4P^2 \quad (5)$$

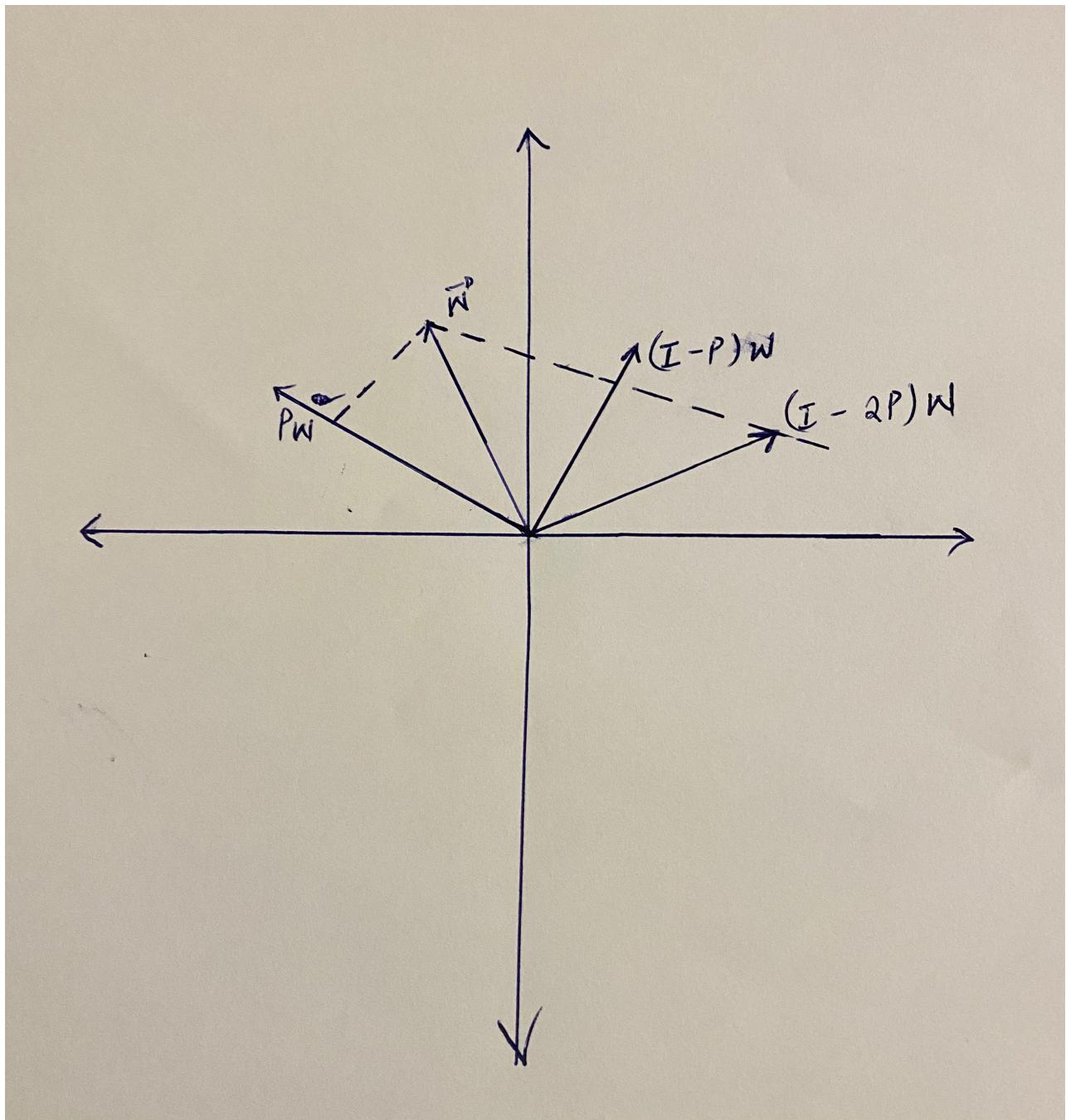
$$= I - 4P + 4P \quad (6)$$

$$= I \quad (7)$$

Therefore  $I - 2P$  is unitary since  $A^*A = (I - 2P)^*(I - 2P) = I$ .

### Geometric interpretation

Suppose vector  $\mathbf{w} \in \text{range}(P)$  and we have that;



$$(I - 2P)\mathbf{w} = \mathbf{w} - 2P\mathbf{w} \quad (8)$$

$$= (I - P)\mathbf{w} - P\mathbf{w} \quad (9)$$

Therefore the geometric interpretation of this is that  $(I - 2P)\mathbf{w}$  is the reflection of vector  $\mathbf{w}$  along the space orthogonal to  $P\mathbf{w}$

## Problem #2

Given  $A \in \mathcal{C}^{m \times n}$  with  $m \geq n$ , show that  $A^* A$  is non-singular if and only if  $A$  has full rank.

### Solution

Let  $C = A^* A$  be a non-singular matrix. This means that  $\det(C) = \det(A^* A) \neq 0$  implying that matrix  $C$  has full rank. The adjoint of  $C$  is then given by;

$$C^* = (A^* A)^* = A^* A \quad (10)$$

$$\Rightarrow \det(C^*) = \det(A^* A) \neq 0 \quad (11)$$

This then implies that  $C^*$  also has full rank.

Since  $A^* A$  is hermitian and  $\det(A^* A) \neq 0$ , then  $\det(A) \neq 0$  thus  $A$  has full rank.

Therefore  $A^* A$  is non singular if and only if  $A$  has full rank.

## Problem #3

The singular vectors and values of a matrix  $A$  satisfy

$$A\mathbf{v}_i = \sigma_i \mathbf{u}_i \quad (12)$$

When we write this in matrix form, we obtain the SVD of  $A$ , given by

$$A = U\Sigma V^* \quad (13)$$

By convention, we order the singular values in the diagonal of  $\Sigma$  in decreasing order so that

$$\Sigma_{11} \geq \Sigma_{22} \geq \Sigma_{33} \geq \dots \geq \Sigma_{rr} > 0 \quad (14)$$

where  $r$  is the rank of  $A$ . Show that it in fact doesn't matter how we arrange the singular vectors into columns of  $U$  and  $V$  and  $\Sigma$ , as long as we are consistent with the ordering of the vectors within  $U$  and  $V$ .

### Solution

To show that it doesn't matter how we arrange the singular vectors into columns of  $U$  and  $V$  with their corresponding singular values in  $\Sigma$  as long as we are consistent, we can use a

permutation matrix say  $P$  to show this.

Assuming  $A$  is an  $m \times n$  matrix with  $m \geq n$  and that it has full rank that is  $\text{rank}(A) = r = n$ ,  $P$  will then be a  $n \times n$  matrix with exactly one entry of 1 in each row and each column and 0s elsewhere.

Let

$$U = \begin{pmatrix} u_{11} & u_{12} & \cdots & u_{1r} \\ u_{21} & u_{22} & \cdots & u_{2r} \\ \vdots & \vdots & \ddots & \vdots \\ u_{m1} & u_{m2} & \cdots & u_{mr} \end{pmatrix}, \quad \Sigma = \begin{pmatrix} \Sigma_{11} & 0 & \cdots & 0 \\ 0 & \Sigma_{22} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \Sigma_{rr} \end{pmatrix} \quad V = \begin{pmatrix} v_{11} & v_{12} \\ v_{21} & v_{22} \\ \vdots & \vdots \\ v_{r1} & v_{r2} \end{pmatrix}$$

If I want to interchange  $\Sigma_{11}$  with  $\Sigma_{22}$ , then  $P = \begin{pmatrix} 0 & 1 & 0 & \cdots & 0 \\ 1 & 0 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & \vdots \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 1 \end{pmatrix}$

The diagonal elements  $\Sigma_{11}$  and  $\Sigma_{22}$  can be interchanged as shown below;

$$P \begin{pmatrix} 0 & 1 & 0 & \cdots & 0 \\ 1 & 0 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & \vdots \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 1 \end{pmatrix} \Sigma \begin{pmatrix} \Sigma_{11} & 0 & \cdots & 0 \\ 0 & \Sigma_{22} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \Sigma_{rr} \end{pmatrix} P \begin{pmatrix} 0 & 1 & 0 & \cdots & 0 \\ 1 & 0 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & \vdots \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 1 \end{pmatrix} = \begin{pmatrix} \Sigma_{22} & 0 \\ 0 & \Sigma_{11} \\ \vdots & \vdots \\ 0 & 0 \end{pmatrix}$$

Since consistency is a key factor we as well have to interchange the columns of  $U$  with respect to  $\Sigma_{11}$  and  $\Sigma_{22}$  as shown below;

$$U \begin{pmatrix} u_{11} & u_{12} & \cdots & u_{1r} \\ u_{21} & u_{22} & \cdots & u_{2r} \\ \vdots & \vdots & \ddots & \vdots \\ u_{m1} & u_{m2} & \cdots & u_{mr} \end{pmatrix} P \begin{pmatrix} 0 & 1 & 0 & \cdots & 0 \\ 1 & 0 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & \vdots \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 1 \end{pmatrix} = \begin{pmatrix} u_{12} & u_{11} & \cdots & u_{1r} \\ u_{22} & u_{21} & \cdots & u_{2r} \\ \vdots & \vdots & \ddots & \vdots \\ u_{m2} & u_{m1} & \cdots & u_{mr} \end{pmatrix}$$

Lastly we have to interchange the rows of  $V^*$  with respect to  $\Sigma_{11}$  and  $\Sigma_{22}$  as shown below

$$\begin{array}{c} \left( \begin{array}{cccccc} 0 & 1 & 0 & \cdots & 0 \\ 1 & 0 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & \\ \vdots & \vdots & \ddots & & \vdots \\ 0 & 0 & 0 & \cdots & 1 \end{array} \right) \quad P \\ V^* \end{array} \quad \begin{pmatrix} \bar{v}_{11} & \bar{v}_{21} & \cdots & \bar{v}_{r1} \\ \bar{v}_{12} & \bar{v}_{22} & \cdots & \bar{v}_{r2} \\ \vdots & \vdots & \vdots & \vdots \\ \bar{v}_{1r} & \bar{v}_{2r} & \cdots & \bar{v}_{rr} \end{pmatrix} = \begin{pmatrix} \bar{v}_{12} & \bar{v}_{22} & \cdots & \bar{v}_{r2} \\ \bar{v}_{11} & v_{21} & \cdots & \bar{v}_{r1} \\ \vdots & \vdots & \vdots & \vdots \\ \bar{v}_{1r} & \bar{v}_{2r} & \cdots & \bar{v}_{rr} \end{pmatrix}$$

We notice then that to re-arrange the singular vectors into columns of  $U$  and  $V$  with their corresponding singular values in  $\Sigma$  consistently  $A$  becomes;

$$A = (UP)(P\Sigma P)(PV^*) \quad (15)$$

$$= UP^2\Sigma P^2V^* \quad (16)$$

But  $P^2 = I$ ,  $\Rightarrow A = U\Sigma V^*$ . This therefore shows that it doesn't matter how we arrange the singular vectors into columns of  $U$  and  $V$  with their corresponding singular values in  $\Sigma$  as long as we are consistent.

## Problem #4

Let  $P \in \mathcal{C}^{m \times m}$  be a non-zero projector. Show that  $\|P\|_2 \geq 1$  with equality if and only if  $P$  is an orthogonal projector.

**Hint:** Use the definition of a project presented in the introduction to Lecture 6.

## Solution

Since  $P$  is a non-zero projector, then  $\|P\|_2 \neq 0$  and  $P^2 = P$ .

Suppose  $\mathbf{v} \in \text{range}(P)$  and  $\mathbf{v} = P\mathbf{x}$  for some  $\mathbf{x}$ , then  $\mathbf{v}$  lies exactly on its own shadow, and applying the projector on  $\mathbf{v}$  results in;

$$P\mathbf{v} = P(P\mathbf{x}) = P^2\mathbf{x} = P\mathbf{x} = \mathbf{v} \quad (17)$$

since

$$P\mathbf{v} = \mathbf{v} \quad (18)$$

$$\Rightarrow \|P\mathbf{v}\|_2 = \|\mathbf{v}\|_2 \quad (19)$$

Using the Cauchy–Schwarz inequality,  $\|P\mathbf{v}\|_2 \leq \|P\|_2 \|\mathbf{v}\|_2$  the above equation becomes;

$$\|P\|_2 \|\mathbf{v}\|_2 \geq \|\mathbf{v}\|_2 \quad (20)$$

$$\Rightarrow \|P\|_2 \geq 1 \quad (21)$$

If  $P$  is an orthogonal projector, then  $P = P^*$ .

Let  $P = U\Sigma V^*$  be the SVD form of  $P$  where  $U$  and  $V$  are the left and right singular vectors respectively,  $UU^* = VV^* = I$  and  $\Sigma$  is a diagonal matrix of singular values  $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_m > 0$ .

From Lecture 3, we know that  $\|P\|_2 = \max_{1 \leq i \leq m} \sigma_i = \sigma_1$ .

Therefore  $\|P\|_2 = 1$  if and only if  $\sigma_1 = 1$  and  $\sigma_1 = 1$  iff  $\Sigma = I$ .

We also know that;

$$P^2 = P \quad (22)$$

$$\Rightarrow U\Sigma V^* U\Sigma V^* = U\Sigma V^* \quad (23)$$

$$V^* U\Sigma = I \quad (24)$$

This means that  $\Sigma = I$  iff  $V^* U = I$  i.e.  $V = U$  which also means that  $\Sigma = I$  if and only iff  $P$  is an orthogonal projector.

Therefore  $\|P\|_2 = 1$  if and only if  $P$  is an orthogonal projector.

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## Problem #5

Suppose both  $P_1$  and  $P_2$  are orthogonal projectors onto the subspace  $S$ . Show that  $P_1 = P_2$ .

**Hint:** Show that  $\|(P_1 - P_2)\mathbf{x}\|_2^2 = 0$ .

### Solution

Suppose  $\mathbf{x} \in \mathbb{R}^m$ , then if  $P_1$  and  $P_2$  are orthogonal projectors onto the subspace  $S$ ,  $P_1\mathbf{x}, P_2\mathbf{x} \subset S$  and  $(I - P_1)\mathbf{x}, (I - P_2)\mathbf{x} \subset S^\perp$ . Also  $P_i^2 = P_i^T = P_i$  for  $i = 1, 2$ .

To show that  $P_1 = P_2$ , we must show that  $\|(P_1 - P_2)\mathbf{x}\|_2^2 = 0$  and this is as below;

$$\|(P_1 - P_2)\mathbf{x}\|_2^2 = (P_1\mathbf{x} - P_2\mathbf{x})^T(P_1 - P_2)\mathbf{x} \quad (25)$$

$$= (P_1\mathbf{x})^T(P_1 - P_2)\mathbf{x} - (P_2\mathbf{x})^T(P_1 - P_2)\mathbf{x} \quad (26)$$

$$= \mathbf{x}^T P_1 (P_1 - P_2) \mathbf{x} - \mathbf{x}^T P_2 (P_1 - P_2) \mathbf{x} \quad (27)$$

$$= \mathbf{x}^T P_1 (I - P_2) \mathbf{x} - \mathbf{x}^T P_2 (I - P_1) \mathbf{x} \quad (28)$$

$$= \mathbf{x}^T P_1 (I - P_2) \mathbf{x} + \mathbf{x}^T P_2 (I - P_1) \mathbf{x} \quad (29)$$

Since  $S$  and  $S^\perp$  are complementary subspaces,  $P_1(I - P_2)\mathbf{x} = P_2(I - P_1)\mathbf{x} = 0$ .

$$\Rightarrow \|(P_1 - P_2)\mathbf{x}\|_2^2 = 0 \quad (30)$$

Therefore  $P_1 = P_2$  and thus the projection onto  $S$  is unique.

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## Problem #6

Let  $U$  be the matrix of singular vectors of the matrix  $A$ . Show that  $P = UU^T$  is an orthogonal projector onto  $\text{range}(A)$ . Assume that the columns of  $A$  are linearly independent.

### Solution

If we assume that the columns of matrix  $A$  are linearly independent and that  $m \geq n$ , then  $\text{range}(A)$  is spanned by all columns of  $A$ . Thus  $A$  is full rank with  $\text{rank}(A) = n$ .

Let  $A = U\Sigma V^T$  be the reduced SVD form of an  $m \times n$  matrix  $A$  where  $U$  is an  $m \times n$  matrix of orthonormal left singular vectors,  $V$  is an  $n \times n$  matrix of orthonormal right singular vectors and  $\Sigma$  is  $n \times n$  diagonal matrix of singular values. We then have that  $V^T V = I$ , and  $\Sigma^T = \Sigma$

If  $P$  is an orthogonal projection onto  $\text{range}(A)$ , then it is given by;

$$P = A(A^T A)^{-1} A^T \quad (31)$$

$$= U\Sigma V^T [(U\Sigma V^T)^T U\Sigma V^T]^{-1} (U\Sigma V^T)^T \quad (32)$$

$$= U\Sigma V^T [V\Sigma U^T U\Sigma V^T]^{-1} V\Sigma U^T \quad (33)$$

$$= U\Sigma V^T [V\Sigma^2 V^T]^{-1} V\Sigma U^T \quad (34)$$

$$= U\Sigma V^T V^{-T} \Sigma^{-2} V^{-1} V\Sigma U^T \quad (35)$$

$$= U\Sigma V^T V \Sigma^{-2} V^T V\Sigma U^T \quad (36)$$

$$= U\Sigma \Sigma^{-2} \Sigma U^T \quad (37)$$

$$\therefore P = UU^T \quad (38)$$

In [ ]: