

## Exam 2

To avoid any confusion, unless specified otherwise, vector spaces are complex vector spaces, inner-products are complex inner-products, and matrices are complex matrices. The standard inner product is  $\langle \mathbf{u}, \mathbf{v} \rangle = \mathbf{v}^H \mathbf{u} = \sum_{i=1}^n \bar{v}_i u_i$ .

### Part I: True/False

Each problem is points for a total of 50 points. (5 points each.)

You do not need to justify the answers here, this is unlike the quizzes.

1. \_\_\_\_\_ If  $U$  is unitary, then  $U$  is itself unitarily diagonalizable. This means there is a unitary  $V$  so that  $U = VDV^H$  where  $D$  is diagonal.
2. \_\_\_\_\_ For any diagonalizable matrix  $A$ , one can use Gram-Schmidt to find an orthogonal basis consisting of eigenvectors.
3. \_\_\_\_\_ The collection of rank  $k$   $n \times n$  matrices is a subspace of  $\mathbb{R}^{n \times n}$ , for  $k < n$ .
4. \_\_\_\_\_ If  $A$  is unitary, then  $|\lambda| = 1$  for all eigenvalues  $\lambda$  of  $A$ .
5. \_\_\_\_\_ If  $p(t)$  is a polynomial and  $\mathbf{v}$  is an eigenvector of  $A$  with associated eigenvalue  $\lambda$ , then  $p(A)\mathbf{v} = p(\lambda)\mathbf{v}$ .
6. \_\_\_\_\_ If  $A$  and  $B$  are both  $n \times n$  and  $\mathcal{B}$  is a basis for  $\mathbb{C}^n$  consisting of eigenvectors for both  $A$  and  $B$ , then  $A$  and  $B$  commute.
7. \_\_\_\_\_ Any matrix  $A$  can be written as a weighted sum of rank 1 matrices..
8. \_\_\_\_\_ For all Hermitian matrices  $A$ , there is a matrix  $B$  so that  $B^H B = A$ .
9. \_\_\_\_\_ There are linear maps  $L : \mathbb{R}^5 \rightarrow \mathbb{R}^4$  such that  $\dim(\ker(L)) = 2 = \dim(\text{rng}(L))$ .
10. \_\_\_\_\_ If  $A$  is invertible, then  $ABA^{-1} = B$ .

## Part II: Computational (60 points)

**Problem 1.** (15 points) Find  $B$  so that  $B^2 = A$  where

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

**Problem 2.** (15 points) Find  $B$  so that  $B^H B = A$  where  $A$  is from (1).

**Problem 3.** (15 points) Find the best rank 2 approximation to  $A$  from (1) with respect to  $\|\cdot\|_F$ .

**Problem 4.** (15 points) Let

$$A = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix},$$

find the characteristic polynomial and all eigenvalues, both real and complex. Explain why  $A$  is diagonalizable and compute  $A^{2020}$ . Note, I do not ask you to diagonalize  $A$ .

### Part III: Theory and Proofs (45 points; 15 points each)

Pick three of the following four options. If you try all four, I will grade the first three, so if this is not what you intend, then just do three, or at least make it clear which I should grade.

**Problem 1.** Let  $S$  be a fixed invertible  $n \times n$  matrix. Let  $U$  be the set of  $n \times n$  matrices that are diagonalized by  $S$ , that is  $A = SD_AS^{-1}$  for some diagonal matrix  $A$ . Either prove that that  $U$  is a subspace of  $\mathbb{C}^{n \times n}$  or show that  $U$  is not a subspace of  $\mathbb{C}^{n \times n}$ .

**Problem 2.** Let  $A$  be a real  $m \times n$  matrix and let  $A^\dagger = V^T \Sigma^\dagger U$ , where  $A = U \Sigma V^T$  where  $U$  is  $m \times m$ ,  $V$  is  $n \times n$ , both unitary,  $\Sigma$  is  $m \times n$  and  $\Sigma^\dagger$  is  $n \times m$  have the form

$$\Sigma = \begin{bmatrix} \sigma_1 & & & & \\ & \ddots & & & \\ & & \sigma_r & & \\ & & & 0 & \\ & & & & \ddots \\ & & & & & 0 \end{bmatrix} \quad \text{and} \quad \Sigma^\dagger = \begin{bmatrix} \sigma_1^{-1} & & & & \\ & \ddots & & & \\ & & \sigma_r^{-1} & & \\ & & & 0 & \\ & & & & \ddots \\ & & & & & 0 \end{bmatrix}$$

with  $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_r > 0$ .

Show:  $\hat{\mathbf{x}} = A^\dagger \mathbf{b}$  is a least-squares solution to  $A\mathbf{x} = \mathbf{b}$ .

Previously we used  $\hat{\mathbf{x}} = (A^T A)^{-1} A^T \mathbf{b}$  for our least-squares solution, but we had the restriction that the columns of the "data" matrix  $A$  were independent, this guarantees that  $\text{NS}(A) = \text{NS}(A^T A) = \{\mathbf{0}\}$ . It is not hard to see that  $A^\dagger = (A^T A)^{-1} A^T$  if  $A$  has linear independent columns.

Review the comments about [Topic 5 DQ 2 in the Class Notes](#). Particularly point (2.) concerning what it means to be a least-squares solution to  $A\mathbf{x} = \mathbf{b}$ .

**Problem 3.** Prove that any complex inner-product  $\langle \cdot, \cdot \rangle_V$  on a complex vector space  $V$ , there is a basis  $\mathcal{U} = \{\mathbf{u}_1, \dots, \mathbf{u}_n\}$  so that

$$\langle \mathbf{x}, \mathbf{y} \rangle_V = [\mathbf{y}]_{\mathcal{U}}^H [\mathbf{x}]_{\mathcal{U}}$$

In other words for any finite dimensional inner-product space, there is a choice of basis, so that with respect to that basis, the inner-product is represented by the standard inner-product.

Here, in case you need it, is the [definition of an inner-product](#). All the notation here is as I always use it in my notes.



**Problem 4.** Use the SVD to show that any square matrix  $A$  can be written as  $A = UP$  where  $U$  is unitary and  $P$  is Hermitian.