

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 λ of A .
5. _____ If $p(t)$ is a polynomial and \mathbf{v} is an eigenvector of A with associated eigenvalue λ , 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)

P1. (15 points) Find B so that $B^2 = A$ where

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

P2. (15 points) Find B so that $B^H B = A$ where A is from (1).

P3. (15 points) Find the best rank 2 approximation to A from (1) with respect to $\|\cdot\|_F$.

P4. (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.

- P1. 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 $\mathbb{C}^{n \times n}$.

- P2. Let A be a real $m \times n$ matrix and let $A^\dagger = V\Sigma^\dagger U^T$, where $A = U\Sigma V^T$ where U is $m \times m$, V is $n \times n$, both unitary, Σ is $m \times n$ and Σ^\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 $A\mathbf{x} = \mathbf{b}$.

P3. 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.

- P4. Use the SVD to show that any square matrix A can be written as $A = UP$ where U is unitary and P is Hermitian.