

Remember to write your name! You are allowed to use a calculator. You are not allowed to use the textbook, your notes, the internet, or your neighbor. To receive full credit on a problem you must show **sufficient justification for your conclusion** unless explicitly stated otherwise.

Name: _____

1. (30 points) If the statement is **always true** mark “TRUE”; if it is possible for the statement to be false then mark “FALSE.” If the statement seems neither true nor false but rather incoherent, raise your hand. No justification is necessary. **Students in 4720 can pick 5 out of 6 questions to answer. Students in 5720 must answer all.**

_____ (a) If a matrix \mathbf{A} is normal then the eigenvalues of the perturbed matrix $\mathbf{A} + \mathbf{E}$ are all within a distance $\|\mathbf{E}\|_3$ of the eigenvalues of \mathbf{A} .

False This is the Bauer-Fike theorem. You can use any p -norm. But in the 3-norm the condition of the eigenvector basis is not necessarily 1, as it would be in the 2-norm.

_____ (b) Let λ be a simple eigenvalue of \mathbf{A} with left and right eigenvectors \mathbf{y} and \mathbf{x} , each of which are 2-norm unit vectors, and let $\mathbf{A} + \mathbf{E}$ be the perturbed matrix where $\|\mathbf{E}\|_2 = \epsilon$. True or False: The perturbed matrix will have an eigenvalue μ within a distance of approximately $\epsilon/|\mathbf{y}^*\mathbf{x}|$ from λ , for small-enough ϵ .

True

_____ (c) Let \mathbf{A} be a diagonalizable matrix with approximate eigenvalue/eigenvector pair (μ, \mathbf{x}) . True or false: (μ, \mathbf{x}) is an exact eigenvalue/eigenvector pair for a perturbed matrix $\mathbf{A} + \mathbf{E}$ where $\|\mathbf{E}\|_2 \leq \|\mathbf{A}\mathbf{x} - \mu\mathbf{x}\|_2$.

False This would be true for *normal* matrices, but the correct statement for non-normal matrices includes the condition number of the eigenvector basis.

_____ (d) Suppose that \mathbf{A} is $n \times n$ with LU factorization $\mathbf{PA} = \mathbf{LU}$. True or false: The matrix $\mathbf{UP}^T\mathbf{L}$ has the same eigenvalues as \mathbf{A} .

True The matrices are similar:

$$\mathbf{UP}^T\mathbf{L} = \mathbf{L}^{-1}\mathbf{PAP}^T\mathbf{L}.$$

\mathbf{L} is always invertible (even if \mathbf{A} is not) because it is lower triangular with ones on the diagonal.

_____ (e) Let λ and \mathbf{x} be an eigenvalue/eigenvector pair for \mathbf{A} . True or false: The matrix $\mathbf{A} - \lambda\mathbf{x}\mathbf{x}^*/\|\mathbf{x}\|_2^2$ has eigenvector \mathbf{x} with eigenvalue 0.

True

_____ (f) Let \mathcal{S} be a nontrivial subspace that is invariant under a square matrix \mathbf{A} . True or False: There is an eigenvector of \mathbf{A} in \mathcal{S} .

True

2. (20 points) Suppose that you are given one eigenvalue/eigenvector pair of an $n \times n$ matrix \mathbf{A} . Explain how you can reduce the problem of finding the remaining eigenvalues of \mathbf{A} to finding the eigenvalues of an $(n-1) \times (n-1)$ matrix. Show explicitly how to construct the $(n-1) \times (n-1)$ matrix. Hint: Start by constructing an invertible matrix \mathbf{X} whose first column is the eigenvector.

Let \mathbf{X} be an invertible matrix whose first column is the eigenvector. Then

$$\mathbf{X}^{-1}\mathbf{A}\mathbf{X} = \left[\begin{array}{c|c} \lambda & * \\ \hline \mathbf{0} & \mathbf{B} \end{array} \right].$$

\mathbf{A} is similar to the RHS, which is block-upper triangular. The eigenvalues of \mathbf{A} are therefore λ together with the eigenvalues of \mathbf{B} , which is $(n-1) \times (n-1)$. Kudos if you used a unitary similarity transform rather than just an invertible \mathbf{X} .

3. Computing the SVD of a real $m \times n$ matrix \mathbf{A} requires computing the eigenvalues and eigenvectors of $\mathbf{A}^T\mathbf{A}$ and $\mathbf{A}\mathbf{A}^T$.
- (a) Let \mathbf{P} and \mathbf{Q} be real orthogonal matrices of size $m \times m$ and $n \times n$ respectively, and let $\mathbf{B} = \mathbf{P}\mathbf{A}\mathbf{Q}$. Show that the singular values of \mathbf{B} are the same as the singular values of \mathbf{A} .

The singular values of \mathbf{A} are the square roots of the eigenvalues of $\mathbf{A}^T\mathbf{A}$, and similarly for the singular values of \mathbf{B} . Note that

$$\mathbf{B}^T\mathbf{B} = \mathbf{Q}^T\mathbf{A}^T\mathbf{A}\mathbf{Q}$$

so $\mathbf{B}^T\mathbf{B}$ is (orthogonally-)similar to $\mathbf{A}^T\mathbf{A}$, and they therefore have the same eigenvalues.

- (b) Let \mathbf{v} be an eigenvector of $\mathbf{B}^T\mathbf{B}$. How is it related to the corresponding eigenvector of $\mathbf{A}^T\mathbf{A}$?

The above analysis shows that if \mathbf{v} is an eigenvector of $\mathbf{B}^T\mathbf{B}$, then $\mathbf{Q}\mathbf{v}$ is an eigenvector of $\mathbf{A}^T\mathbf{A}$.

- (c) It is possible to choose \mathbf{P} and \mathbf{Q} such that \mathbf{B} is bi-diagonal (nonzeros immediately above the diagonal). Prove that $\mathbf{B}^T\mathbf{B}$ and $\mathbf{B}\mathbf{B}^T$ are tridiagonal (you may cite any relevant theorem from class).

The banded-matrix-multiplication theorem shows that multiplying a lower-bidiagonal and an upper-bidiagonal matrix yields a tridiagonal matrix.

4. (20 points)

- **5720 Only** Let \mathbf{A} be an $n \times n$ diagonalizable matrix with eigenvalues satisfying $\lambda_1 = \dots = \lambda_k$ with $|\lambda_k| > |\lambda_{k+1}| \geq \dots \geq |\lambda_n|$. Show that the vectors generated by the power method will converge to an eigenvector of \mathbf{A} (under standard assumptions on the starting vector).

Let the eigenvectors of \mathbf{A} be $\mathbf{v}_1, \dots, \mathbf{v}_n$, and the initial vector for the power method be $\mathbf{x}_0 = c_1\mathbf{v}_1 + \dots + c_n\mathbf{v}_n$. Assume that c_1, \dots, c_k are not all zero. Then

$$\mathbf{A}^p \mathbf{x}_0 = c_1 \lambda_1^p \mathbf{v}_1 + \dots + c_k \lambda_1^p \mathbf{v}_k + c_{k+1} \lambda_{k+1}^p \mathbf{v}_{k+1} + \dots + c_n \lambda_n^p \mathbf{v}_n$$

$$\mathbf{A}^p \mathbf{x}_0 = \lambda_1^p (c_1 \mathbf{v}_1 + \dots + c_k \mathbf{v}_k) + c_{k+1} \lambda_{k+1}^p \mathbf{v}_{k+1} + \dots + c_n \lambda_n^p \mathbf{v}_n$$

If you normalize then the coefficients of $\mathbf{v}_{k+1}, \dots, \mathbf{v}_n$ will decay to 0 as $p \rightarrow \infty$ so that $\mathbf{A}^p \mathbf{x}_0 / \|\mathbf{A}^p \mathbf{x}_0\|$ will converge to

$$\frac{c_1 \mathbf{v}_1 + \dots + c_k \mathbf{v}_k}{\|c_1 \mathbf{v}_1 + \dots + c_k \mathbf{v}_k\|}.$$

This vector is an eigenvector of \mathbf{A} because

$$\mathbf{A}(c_1 \mathbf{v}_1 + \dots + c_k \mathbf{v}_k) = \lambda_1 (c_1 \mathbf{v}_1 + \dots + c_k \mathbf{v}_k).$$

- **4720 Only** The basic shifted QR algorithm is

$$\mathbf{A}_{m-1} - \rho \mathbf{I} = \mathbf{Q}_m \mathbf{R}_m, \quad \mathbf{A}_m = \mathbf{R}_m \mathbf{Q}_m + \rho \mathbf{I}, \quad \mathbf{A}_0 = \mathbf{A}.$$

Show that \mathbf{A}_m is orthogonally similar to \mathbf{A}_{m-1} (you may assume everything is real).

$$\mathbf{A}_{m-1} = \mathbf{Q}_m \mathbf{R}_m + \rho \mathbf{I} \Rightarrow \mathbf{Q}_m^T \mathbf{A}_{m-1} \mathbf{Q}_m = \mathbf{R}_m \mathbf{Q}_m + \rho \mathbf{I} = \mathbf{A}_m$$