

# Math 544, Exam 3 Information

## Exam 3 will be based on:

- Sections 4.1 - 4.5, 4.7, and 5.2, 5.3.
- The corresponding assigned homework problems  
(see <http://www.math.sc.edu/~boylan/SCCourses/math5443/544.html>).  
**At minimum, you need to understand how to do the homework problems.**
- Lecture notes: 10/29 - 11/26.
- Quizzes: 7 - 10.

## Topic List (not necessarily comprehensive):

**You will need to know how to define vocabulary words/phrases defined in class.**

**§4.1: The eigenvalue problem for  $2 \times 2$  matrices:** Definition and computation of eigenvalues and eigenvectors for  $2 \times 2$  matrices.

**§4.2: Determinants and the eigenvalue problem:** Definition and computation of determinants of matrices  $A \in \text{Mat}_{n \times n}(\mathbb{R})$ . Computation of determinants by expansion across rows or down columns using **minors** and **cofactors**. What is the **minor** and **cofactor** associated to a matrix entry  $(a_{ij})$  of  $A$ ? Properties of determinants, for example:

- $\det(AB) = \det(A)\det(B)$ .
- $A \in \text{Mat}_{n \times n}(\mathbb{R})$  is singular  $\iff \det(A) = 0$ .
- If  $A$  is invertible, then  $\det(A^{-1}) = \frac{1}{\det(A)}$ .

What is the determinant of a **triangular** matrix?

**§4.3: Elementary operations and determinants:** Important properties:  $\det(A) = \det(A^T)$ ,  $\det(cA) = c^n \det(A)$  for  $c \neq 0$  in  $\mathbb{R}$ . Effects of elementary row **and** column operations on the computation of a determinant:

- Interchanging two rows or two columns changes the sign of the determinant.
- If the row operation

$$R_i \mapsto \frac{1}{k}R_i, \quad k \neq 0$$

transforms matrix  $A$  into matrix  $B$ , then  $\det(A) = k\det(B)$ . (In effect, you are “factoring”  $k$  out of the  $i$ th row of  $A$ .) Similarly, if the column operation

$$C_i \mapsto \frac{1}{k}C_i, \quad k \neq 0$$

transforms matrix  $A$  into matrix  $B$ , then  $\det(A) = k\det(B)$ .

- A row operation of the form

$$R_i \mapsto R_i + kR_j, \quad k \neq 0, \quad i \neq j$$

does nothing to the determinant. Similarly, a column operation of the form

$$C_i \mapsto C_i + kC_j, \quad k \neq 0, \quad i \neq j$$

does nothing to the determinant.

**§4.4: Eigenvalues and the characteristic polynomial:** The definition and computation of the **eigenvalues** of a matrix  $A \in \text{Mat}_{n \times n}(\mathbb{R})$ . i.e., computation of the **characteristic polynomial**  $p(t) = \det(A - tI_n)$ ; the **algebraic multiplicity** of an eigenvalue  $\lambda$  is the number of times the factor  $(t - \lambda)$  occurs in the characteristic polynomial  $p(t)$ .

- If  $\lambda$  is an eigenvalue of  $A$  and  $k \geq 0$  is an integer, then  $\lambda^k$  is an eigenvalue of  $A^k$ .
- If  $\lambda$  is an eigenvalue of  $A$  and  $\alpha \in \mathbb{R}$ , then  $\lambda + \alpha$  is an eigenvalue of  $A + \alpha I_n$ .
- If  $A$  is invertible and  $\lambda$  is an eigenvalue of  $A$ , then  $\frac{1}{\lambda}$  is an eigenvalue of  $A^{-1}$ .
- If  $\lambda$  is an eigenvalue of  $A$ , then it is also an eigenvalue of  $A^T$ .
- A matrix  $A$  has 0 as one of its eigenvalues if and only if it is singular.

What are the eigenvalues of a triangular matrix?

**§4.5: Eigenspaces and eigenvectors:** The definition and computation of the eigenvectors of a matrix  $A \in \text{Mat}_{n \times n}(\mathbb{R})$ . If  $\lambda$  is an eigenvalue of  $A \in \text{Mat}_{n \times n}(\mathbb{R})$ , then the **eigenspace** associated to  $\lambda$  is  $E_\lambda = \text{Null}(A - \lambda I)$  and the **geometric multiplicity** of  $\lambda$  is the dimension of  $E_\lambda$  (i.e., the nullity of  $A - \lambda I$ ). The relationship between algebraic and geometric multiplicities is

$$1 \leq \text{geometric mult.}(\lambda) \leq \text{algebraic mult.}(\lambda).$$

Definition of a **defective** matrix: a matrix  $A$  is defective if  $A$  has at least one eigenvalue whose geometric mult. is **strictly less than** its algebraic mult. i.e., there is an eigenvalue  $\lambda$  with

$$\text{geom. mult.}(\lambda) < \text{alg. mult.}(\lambda).$$

**Important fact:** Eigenvectors associated to distinct eigenvalues are linearly independent. As a consequence, if  $A \in \text{Mat}_{n \times n}(\mathbb{R})$  is not defective, then  $A$  has  $n$  linearly independent eigenvectors, and these eigenvectors form a basis for  $\mathbb{R}^n$ . In particular, if  $A$  has  $n$  distinct eigenvalues, then  $A$  is not defective.

**§4.7: Similarity transformations and diagonalization:** Matrices  $A$  and  $B \in \text{Mat}_{n \times n}(\mathbb{R})$  are **similar** if there is an invertible matrix  $S$  for which

$$B = S^{-1}AS.$$

A matrix  $A$  is **diagonalizable** if it is similar to a diagonal matrix  $D$ . If  $A$  and  $B$  are similar, they have the same:

- characteristic polynomial:  $p_A(t) = p_B(t)$ . **However, the converse is not true:** If  $p_A(t) = p_B(t)$ , then it is not always true that  $A$  and  $B$  are similar.
- eigenvalues and algebraic multiplicities (**but the corresponding eigenvectors are typically different!** If  $B = S^{-1}AS$  (so  $A$  and  $B$  are similar) and if  $\vec{x}$  is an eigenvector of  $B$  associated to  $\lambda$  (so  $B\vec{x} = \lambda\vec{x}$ ), then  $S\vec{x}$  is an eigenvector of  $A$  associated to  $\lambda$  (so  $A(S\vec{x}) = \lambda(S\vec{x})$ )).

**Criterion for diagonalizability:** The diagonalizability of  $A$  is equivalent to

- $A$  has  $n$  linearly independent eigenvectors (the maximum possible).
- $A$  is not defective (i.e., the geometric and algebraic multiplicities agree for all eigenvalues of  $A$ ). So a matrix  $A$  is either defective or diagonalizable.

If  $A$  is diagonalizable, then there is an invertible matrix  $S$  and a diagonal matrix  $D$  for which

$$D = S^{-1}AS.$$

How do you find the matrices  $S$  and  $D$ ?

- Compute the eigenvalues of  $A$  and their algebraic multiplicities. Suppose that the **distinct** eigenvalues of  $A$  are  $\lambda_1, \dots, \lambda_k$ .
- Compute bases  $\mathcal{B}_1, \dots, \mathcal{B}_k$  for the eigenspaces  $E_{\lambda_1}, \dots, E_{\lambda_k}$ . The dimension of  $E_{\lambda_i}$  is the geometric multiplicity of  $\lambda_i$ . If for all  $i$ ,

$$\text{alg. mult.}(\lambda_i) = \text{geom. mult.}(\lambda_i),$$

then  $A$  is diagonalizable.

- If  $A$  is diagonalizable, form the set  $\mathcal{B} = \{\vec{w}_1, \dots, \vec{w}_n\}$  consisting of all the basis vectors for the eigenspaces of  $A$ . Then the invertible matrix  $S$  which diagonalizes  $A$  is

$$S = (\vec{w}_1 \mid \vec{w}_2 \mid \dots \mid \vec{w}_n).$$

So we have

$$D = S^{-1}AS,$$

where  $D$  is a diagonal matrix with diagonal entry  $(D)_{ii} = \lambda_i$  and  $\lambda_i$  is the eigenvalue of  $A$  associated to the eigenvector  $\vec{w}_i$ :  $A\vec{w}_i = \lambda_i\vec{w}_i$ .

If  $A$  is diagonalizable, and  $k \geq 0$  is an integer, how can you compute  $A^k$ ? Here's how:  $A$  diagonalizable implies that for some invertible matrix  $S$ ,  $D = S^{-1}AS$  is diagonal. We then have  $D^k = (S^{-1}AS)^k = S^{-1}D^kS$ . Moving the  $S$ 's to the left side, we obtain  $SD^kS^{-1} = A^k$ . So if you know  $S$  and  $S^{-1}$  (it is easy to compute  $D^k$  if  $D$  is diagonal), you can compute  $A^k$ .

**Orthogonal matrices:** Their definition and basic properties:

- $Q \in \text{Mat}_{n \times n}(\mathbb{R})$  is orthogonal if and only if its rows and columns form orthonormal bases for  $\mathbb{R}^n$ .
- If you rearrange the rows or columns of an orthogonal matrix, the resulting matrix is still orthogonal.
- If  $Q$  is orthogonal, then:
  - $\forall \vec{x} \in \mathbb{R}^n, \|Q\vec{x}\| = \|\vec{x}\|$ . (Multiplication by  $Q$  preserves length)
  - $\forall \vec{x}, \vec{y} \in \mathbb{R}^n, Q\vec{x} \cdot Q\vec{y} = \vec{x} \cdot \vec{y}$ . (Multiplication by  $Q$  preserves the angle between vectors.)
  - $\det(Q) = \pm 1$ .
- $A \in \text{Mat}_{n \times n}(\mathbb{R})$  is symmetric if and only if it is **orthogonally diagonalizable**. i.e.,  $\exists Q$ , orthogonal, such that  $Q^{-1}AQ = D$  is diagonal.

**§5.2: Vector spaces:** The definition of vector space (a set  $V$  and a scalar field  $F$  together with an addition operation on  $V$  and a scalar multiplication operation); in particular, the ten vector space axioms: 2 closure axioms, 4 axioms for vector addition, 4 axioms for scalar multiplication. Examples of vector spaces:  $\text{Mat}_{m \times n}(\mathbb{R}), P_n$ . Check whether a set  $V$  together with an addition and scalar multiplication is or is not a vector space.

**§5.3: Subspaces:** Determination of whether or not certain subsets of  $\text{Mat}_{m \times n}(\mathbb{R})$  or  $P_n$  are subspaces.