Lecture 42 : Review for Final Exam, Part III

Math 3013 Oklahoma State University

January 31, 2022

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Lecture 41 : Review for Final Exam, Part III

Final Exam :

- Section 62663 (MWF, 10:30pm): Friday, December 10, 10:00am - 11:50am
- Section 62667 (MWF, 1:30pm): Wednesday, December 8, 10:00am - 11:50am

See posts on the Math 3013 Canvas homepage for solutions to exams (both midterm exams and sample exams).

Review for Final - Part III (Material covered since the 2nd exam)

11. Eigenvectors and Eigenvalues

The **Eigenvector/Eigenvalue Problem** for an $n \times n$ matrix **A** is the problem of finding non-zero vectors (eigenvectors) **v** and numbers (eigenvalues) λ such that

$$\mathbf{A}\mathbf{v} = \lambda\mathbf{v}$$

- the eigenvalues of **A** are the solutions of $det(\mathbf{A} \lambda \mathbf{I}) = 0$
- The eigenvectors with eigenvalue λ are solutions of (A – λI) x = 0
- the algebraic multiplicity of an eigenvalue r : the number of factors of (λ − r) in the characteristic polynomial p_A (λ) ≡ det (A − λI)
- ► geometric multiplicity of an eigenvalue r : the dimension of the r-eigenspace = dim (NullSp (A - rI))

12. Diagonalization of Square Matrices

Def. An $n \times n$ matrix **A** is **diagonalizable** if there exists an invertible $n \times n$ matrix **C** and a diagonal $n \times n$ matrix **D** such that

 $\mathbf{C}^{-1}\mathbf{A}\mathbf{C} = \mathbf{D}$

An n × n matrix A is diagonalizable if and only if it has n linearly independent eigenvectors v₁,..., v_n, in which case

$$\mathbf{C} = \begin{bmatrix} \uparrow & \cdots & \uparrow \\ \mathbf{v}_1 & \cdots & \mathbf{v}_n \\ \downarrow & \cdots & \downarrow \end{bmatrix} \quad \text{and} \quad \mathbf{D} = \begin{bmatrix} \lambda_1 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \lambda_n \end{bmatrix}$$

where λ_i is the eigenvalue corresponding to the eigenvector \mathbf{v}_i .

- If an n × n matrix A has n distinct eigenvalues then it diagonalizable.
- An n × n matrix A is diagonalizable if and only if the algebraic multiplicity of each eigenvalue r coincides with its geometric multiplicity.

▶ If
$$\mathbf{A}^T = \mathbf{A}$$
, then \mathbf{A} is diagonalizable.

13. Orthogonal Decomposition of a Vector w.r.t. a Subspace

► Let W be a subspace of ℝⁿ, the orthogonal complement of W is the subspace

$$W_{\perp} \equiv \{ \mathbf{v} \in \mathbb{R}^n \mid \mathbf{v} \cdot \mathbf{w} = 0 \text{ for all } \mathbf{w} \in W \}$$

Let v ∈ ℝⁿ and let W be a subspace of ℝⁿ. The orthogonal decomposition of v w.r.t. W is the unique splitting

$$\mathbf{v} = \mathbf{v}_W + \mathbf{v}_\perp$$

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where $\mathbf{v}_{W} \in W$ and $\mathbf{v}_{\perp} \in W_{\perp}$.

• If $\{\mathbf{b}_1, \ldots, \mathbf{b}_k\}$ is a basis for W, and

$$\mathbf{B} = \begin{bmatrix} \leftarrow & \mathbf{b}_1 & \rightarrow \\ & \vdots & \\ \leftarrow & \mathbf{b}_k & \rightarrow \end{bmatrix}$$

then

$$W_{\perp} = \mathit{NullSp}\left(\mathbf{B}
ight)$$

If {b₁,..., b_k} is a basis for W and {b_{k+1},..., b_n} is a basis for W_⊥, then {b₁,..., b_k, b_{k+1},..., b_n} is a basis for ℝⁿ.
 If v ∈ ℝⁿ, then

$$\mathbf{v} = c_1 \mathbf{b}_1 + \dots + c_k \mathbf{b}_k + c_{k+1} \mathbf{b}_{k+1} + \dots + c_n \mathbf{b}_n$$

= $(c_1 \mathbf{b}_1 + \dots + c_k \mathbf{b}_k) + (c_{k+1} \mathbf{b}_{k+1} + \dots + c_n \mathbf{b}_n)$
= $\mathbf{v}_W + \mathbf{v}_\perp$

To get the coefficients $c_1, \ldots, c_k, c_{k+1}, \ldots, c_n$, one calculates the coordinate vector of **v** w.r.t. the basis { $\mathbf{b}_1, \ldots, \mathbf{b}_k, \mathbf{b}_{k+1}, \ldots, \mathbf{b}_n$ }

$$\left[\begin{array}{ccccc} \uparrow & \uparrow & \uparrow \\ \mathbf{b}_1 & \cdots & \mathbf{b}_k & \mathbf{b}_{k+1} & \cdots & \mathbf{b}_n \\ \downarrow & & \downarrow & \downarrow \end{array} \right]$$

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$$\begin{bmatrix} 1 & 0 & \cdots & 0 & | & c_1 \\ 0 & \ddots & & \vdots & | & \vdots \\ \vdots & & \ddots & 0 & | & \vdots \\ 0 & \cdots & 0 & 1 & | & c_n \end{bmatrix}$$

14. Orthonormal Bases

An orthonormal basis for a subspace W is a basis $\{\mathbf{b}_1, \dots, \mathbf{b}_k\}$ such that

$$\mathbf{b}_i \cdot \mathbf{b}_j = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{if } i \neq j \end{cases}$$

If $\{\mathbf{b}_1, \dots, \mathbf{b}_k\}$ is an orthonormal basis for W then not only can every vector $\mathbf{w} \in W$ be expressed as

$$\mathbf{w} = c_1 \mathbf{b}_1 + \cdots + c_k \mathbf{v}_k$$

but also the each coefficient c_i can be determined by

$$c_i = \mathbf{b}_i \cdot \mathbf{w}$$

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15. The Gram-Schmidt Process (for constructing an orthonormal basis)

Suppose $\{\mathbf{b}_1, \ldots, \mathbf{b}_k\}$ is a basis for a subspace W. Then an orthogonal basis $\{\mathbf{o}_1, \ldots, \mathbf{o}_k\}$ for W can be constructed via the following algorithm:

$$\mathbf{o}_{1} = \mathbf{b}_{1}$$

$$\mathbf{o}_{2} = \mathbf{b}_{2} - \frac{\mathbf{o}_{1} \cdot \mathbf{b}_{2}}{\mathbf{o}_{1} \cdot \mathbf{o}_{1}} \mathbf{o}_{1}$$

$$\vdots$$

$$\mathbf{o}_{k} = \mathbf{b}_{k} - \frac{\mathbf{o}_{1} \cdot \mathbf{b}_{k}}{\mathbf{o}_{1} \cdot \mathbf{o}_{1}} \mathbf{o}_{1} - \dots - \frac{\mathbf{o}_{k-1} \cdot \mathbf{b}_{k}}{\mathbf{o}_{k-1} \cdot \mathbf{o}_{k-1}} \mathbf{o}_{k-1}$$

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Once an orthogonal basis $\{\mathbf{o}_1, \ldots, \mathbf{o}_k\}$ has been constructed, an **orthonormal basis** $\{\mathbf{n}_1, \ldots, \mathbf{n}_k\}$ for W can be constructed by simply rescaling the orthogonal basis vectors so that they have unit length:

$$\mathbf{n}_{1} = \frac{\mathbf{o}_{1}}{\|\mathbf{o}_{1}\|} = \frac{\mathbf{o}_{1}}{\sqrt{\mathbf{o}_{1} \cdot \mathbf{o}_{1}}}$$
$$\mathbf{n}_{2} = \frac{\mathbf{o}_{2}}{\|\mathbf{o}_{2}\|} = \frac{\mathbf{o}_{2}}{\sqrt{\mathbf{o}_{2} \cdot \mathbf{o}_{2}}}$$
$$\vdots$$
$$\mathbf{n}_{k} = \frac{\mathbf{o}_{k}}{\|\mathbf{o}_{k}\|} = \frac{\mathbf{o}_{k}}{\sqrt{\mathbf{o}_{k} \cdot \mathbf{o}_{k}}}$$

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