

Math 225 Introductory Matrix Theory Review Sheet

1 Linear Equations

- Definitions: linear equation, system of linear equations, solution, solution set, row equivalent, augmented matrix, pivot position, pivot column, column vector, vector equation, homogeneous,
- Elementary row operations
- Row Echelon Form and Reduced Row Echelon Form; the Row Reduction Algorithm
- General solutions of matrix equations: parametric descriptions and parametric vector descriptions
- Linear combinations and Span; linear independence
- Correspondence between systems of linear equations, vector equations, and matrix equations
- Applications of linear systems: economics, network flow, chemistry

2 Matrix Algebra

- Definitions: diagonal entries, main diagonal, diagonal matrix, transpose
- Basic matrix operations: addition, scalar multiplication, matrix multiplication
- Inverses: the 2×2 formula and the general method: $[A|I] \rightarrow [I|A^{-1}]$; properties
- The Invertible Matrix Theorem (see also Section 4.6)
- Applications: economics - $(I - C)\vec{x} = \vec{d}$

3 Determinant

- Cofactors
- Definition of determinant using cofactor expansion; simplified computation for triangular matrices
- How determinant changes with row operations; how to compute determinant using Row Reduction Algorithm

- A is invertible if and only if $\det A \neq 0$
 - $\det A^T = \det A$, $\det AB = (\det A)(\det B)$
 - $A^{-1} = \frac{1}{\det A} \text{adj } A$
 - Relationship between determinant and volume
- section Vector Spaces
- Examples of vector spaces: \mathbb{R}^n , \mathbb{P}_n , \mathbb{P} , $\{\vec{0}\}$, etc.
 - Definition of subspace (3 properties)
 - Linear independence, spanning sets, and bases
 - A span is always a subspace
 - Null Space, Column Space, Row Space: How to compute them and a basis for each
 - Dimension; $\text{rank} = \dim \text{Col } A = \dim \text{Row } A = n - \dim \text{Nul } A$ for $A m \times n$

4 Eigenvectors and Eigenvalues

- Definitions: eigenvector, eigenvalue, eigenspace, characteristic equation, characteristic polynomial, multiplicity of eigenvalues
- Eigenvalues of a triangular matrix
- A is invertible if and only if 0 is not an eigenvalue
- Distinct eigenvalues implies linearly independent eigenvectors
- Similar matrices: definition, effect on characteristic polynomial
- Diagonalization: need n linearly independent eigenvectors - then you can form P and D so $A = PDP^{-1}$

5 Orthogonality and Inner Product

- Definitions: inner/dot product, length, unit vector, distance, orthogonal, orthogonal complement W^\perp , orthogonal set, orthogonal basis
- $(\text{Row } A)^\perp = \text{Nul } A$, $(\text{Col } A)^\perp = \text{Nul } A^T$
- Orthogonality implies linear independence
- Orthogonal projection:

$$\hat{y} = \text{proj}_{\text{Span}\{\vec{u}_1, \dots, \vec{u}_k\}} \vec{y} = \frac{\vec{y} \cdot \vec{u}_1}{\vec{u}_1 \cdot \vec{u}_1} \vec{u}_1 + \dots + \frac{\vec{y} \cdot \vec{u}_k}{\vec{u}_k \cdot \vec{u}_k} \vec{u}_k$$

- Orthogonal component $\vec{z} = \vec{y} - \hat{y} = \vec{y} - \text{proj}_W \vec{y}$
- Orthonormal set/basis - simplifies orthogonal projection
- An $m \times n$ matrix U has orthonormal columns if and only if $U^T U = I$; if $U^{-1} = U^T$, U is an *orthogonal* matrix
- Least squares problem/solutions: $A^T A \vec{x} = A^T \vec{b}$
- Section 6.6 will not be on the test.