

VECTOR SPACE

- Composed of a set V and the two operations of vector addition and scalar multiplication
- Must satisfy the 8 axioms in Medici
- **Subspace of V** is a subset $W \subseteq V$ that is itself a vector space
 - o Has same operations of addition and multiplication as V
 - o **Theorem:** a non-empty subset W is a subspace of V iff $cx + y \in W$ when $x, y \in W$
- **5.3 Prop. 1:** if S is a subset of vector space V , then $\text{span}S$ is a subspace of V

DIMENSION

- Number of basis vectors for a vector space
 - o **Fundamental Theorem – the number of vectors in a linearly independent list of vectors in V cannot exceed $\dim V$**
- **Existence of Bases Theorem** (6.4 Theorem V) – any LI list of vectors can be extended to a basis for V
 - o (6.4 Theorem VI) – if U is a subspace of n -dimensional V , then $\dim U \leq n$
 - $\dim U = n$ iff $U = V$
 - o (6.4 Theorem VII) – any spanning set for V contains a basis for V
 - o (6.4 Theorem VIII) – if $\dim V = n$, then 1) any set of n vectors that is LI is a basis for V , and 2) any set of n vectors that spans V is a basis for V

RANK

- **Column Space** – span of columns of A $\{A\mathbf{x} \mid \mathbf{x} \in \mathbb{R}\}$
 - o Can write matrix vector product $A\mathbf{x}$ as LC of columns in matrix
 - o Can write matrix product AB as a matrix where each column is a LC of columns in A
 - $\text{col } AB \subseteq \text{col } A$
 - **Lemma** – if each column in C is an LC of the columns of A , then there exists a matrix B such that $C = AB$
- **Row Space** – span of rows of A (= span of columns of $A^T = \{A^T\mathbf{x} \mid \mathbf{x} \in \mathbb{R}\}$)
 - o Row operations change column space, don't change row space
- **Rank** – $\text{rank } A = \dim \text{col } A$ (number of LI columns of A)
- **Theorem** – A is invertible iff $\text{col } A = n\mathbb{R}$
- **Theorem** – A is invertible iff $\text{rank } A = n$
- **Rank Theorem** – for any matrix A : $\dim \text{col } A = \dim \text{row } A = \text{rank } A$

- (7.4 Property II) – $\text{rank } A = \text{rank } A^T$
 - o **Theorem** – $\text{rank } A = \text{rank } A^T = \text{rank } A^T A = \text{rank } A A^T$
- (7.4 Property III) – $\text{rank } AB \leq \min\{\text{rank } A, \text{rank } B\}$
- **Theorem** – $\text{rank } A \leq \text{rank } [A|b]$, and $Ax = b$ has a solution *iff* $\text{rank } A = \text{rank } [A|b]$
- **Full Row/Column Rank** (7.4 Theorem IV and V) – $\text{rank } A = \text{number of rows/columns}$
 - o All rows/columns are LI
- **7.2 Prop 1 Theorem** – if B has full column rank, then $\text{rank } A = \text{rank } BA$
 - o Multiplying a matrix on the left by a matrix with full column rank doesn't change its rank
- **Theorem** – if C has full row rank, then $\text{rank } A = \text{rank } AC$
 - o Multiplying a matrix on the right by a matrix with full row rank doesn't change its rank
- **Theorem** – A has full column rank *iff* $A^T A$ is invertible, A has full row rank *iff* $A A^T$ is invertible

NULL SPACE

- **Null Space** – set of solutions to the homogeneous system $Ax = 0$
 - o $\{x \mid Ax = 0\}$
- **Nullity** – $\text{nullity } A = \dim \text{null } A$
- **Rank-Nullity Theorem** (7.4 Theorem II) – $\text{nullity } A = n - \text{rank } A$

COORDINATES

- **Theorem** – $[cx + y]_a = c[x]_a + [y]_a$
- **8.4 Theorem I** – coordinates preserve linear independence
- **Change of Basis Matrix** – $[x]_b = P_{ba}[x]_a$ where columns of P_{ba} are coordinate vectors of basis vectors for a WRT basis b
 - o **Theorem** – Inverse of matrix is P_{ab}

DETERMINANT

- **A determinant function** – unique, alternating, multilinear
 - o Any alternating multilinear function f satisfies $f(A) = (\det A)f(I)$

- **THE Determinant** – unique, alternating, multilinear function on the rows of matrix A, where its value on the identity matrix I is 1
 - o Determinant of a 2x2 matrix $\det \begin{bmatrix} a & b \\ c & d \end{bmatrix} = ad - bc$
 - o If A has equal rows, then $\det A = 0$
 - o If A has an entire row of zeros, then $\det A = 0$
 - o If rows of A are LD, then $\det A = 0$
- **ij minor** of an nxn matrix A – omit the ith row and jth column of A (**A_{ij}**)

$$\det A = \sum_{j=1}^n (-1)^{i+j} a_{ij} \det A_{ij}$$
- **Theorem** - expanding along the rows
 - o Can also expand along columns

Theorem VI. INVERTIBILITY THEOREM: Let $A \in \mathbb{R}^n$. Then A is invertible if and only if $\det A \neq 0$.
- If A is not full rank, $\det A = 0$
- If $\det A \neq 0$, rows of A are LI
- **Product Theorem** – if A and B be nxn matrices, then $\det AB = \det A \det B$
- **Theorem** – if A is invertible then $\det A^{-1} = (\det A)^{-1}$
- **Adjoint of A** – nxn matrix whose ij entry is $(-1)^{i+j} \det A_{ji}$
- **Theorem** – $A(\text{adj } A) = (\det A)I$
 - o if A is invertible, $A^{-1} = (1/\det A)\text{adj } A$
- **Transpose Theorem** – $\det A^T = \det A$
- **Cramer's Rule** – solution to $Ax = b$ is: $x_i = \frac{\det B_i}{\det A}$
 - o B_i is same as A but replace ith column of A by b

EIGENVALUES AND EIGENVECTORS

- **Eigenvector of A (nxn matrix):** a NON-ZERO vector where $Ax = \lambda x$
 - o λ is an **eigenvalue** of A corresponding to x
 - o **Eigenspace corresponding to eigenvalue λ :** subspace containing set of all eigenvectors corresponding to that eigenvalue (together with the zero vector)
 - o $E_\lambda(A) = \{x \in \mathbb{R}^n \mid Ax = \lambda x\} = \text{null}(\lambda I - A)$
- **10.2 Prop. 1:** the intersection of two different eigenspaces is the zero vector
- **Characteristic polynomial of A:** $p_A(\lambda) = \det(\lambda I - A)$
 - o $c_A(\lambda) = \lambda^n - (\text{tr } A)\lambda^{n-1} + \dots + (-1)^n \det A$
 - o Degree of exactly n

- As a result: an $n \times n$ matrix A can have at most n distinct eigenvalues
- **TO SOLVE FOR EIGENVALUES AND EIGENVECTORS:**
 - λ is an eigenvalue *iff* $\lambda I - A$ is non-invertible *iff* $\det(\lambda I - A) = 0$ *iff* $p_A(\lambda) = 0$
 - *iff* λ is a ROOT of characteristic polynomial
 - For eigenvector/eigenspaces: $E_\lambda(A) = \text{null}(\lambda I - A)$

DIAGONALIZATION

- A and B ($n \times n$) are similar if there exists an invertible $n \times n$ matrix S such that $A = SBS^{-1}$
 - Same determinant, rank, characteristic polynomial, eigenvalues, and trace
- **Diagonal Matrix:** all entries except for diagonal are 0
- **Diagonalizable Matrix:** $n \times n$ matrix A similar to an $n \times n$ diagonal matrix D ($A = SDS^{-1}$)
 - Diagonal entries of D are eigenvalues
 - S is change of basis matrix from eigenbasis to standard basis
 - **Diagonalization Theorem:** A is diagonalizable *iff* A 's eigenvectors are a basis for \mathbb{R}^n
- **10.3 Theorem II:** if A is an $n \times n$ matrix with exactly n distinct eigenvalues, then $\det A$ is the product of the eigenvalues and the $\text{tr} A$ is the sum of the eigenvalues
- **MULTIPLICITIES:**
 - **Algebraic of λ :** number of times λ appears as a root of $p_A(\lambda)$
 - **Geometric of λ :** $\dim E_\lambda(A)$
 - **Multiplicity Theorem:** $1 \leq \text{geometric} \leq \text{algebraic}$
- **CONDITIONS FOR DIAGONALIZATION:**
 - Diagonalization Theorem
 - **10.4 Theorem III:** if A ($n \times n$) has n distinct eigenvalues, then A is diagonalizable
 - **10.5 Theorem VI (Diagonalization Test):** A ($n \times n$) is diagonalizable *iff* the geometric and algebraic multiplicities of each eigenvalue are equal

LINEAR DIFFERENTIAL EQUATIONS WITH CONSTANT COEFFICIENTS

Theorem: Let A be a $n \times n$ diagonalizable matrix with eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$ (not necessarily distinct). Let $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ be a basis for \mathbb{R}^n consisting of eigenvectors of A . If $\mathbf{x}_0 = c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \dots + c_n \mathbf{v}_n$, then the system $\mathbf{x}' = A\mathbf{x}$, $\mathbf{x}(0) = \mathbf{x}_0$ has solution

$$\mathbf{x}(t) = e^{\lambda_1 t}(c_1 \mathbf{v}_1) + e^{\lambda_2 t}(c_2 \mathbf{v}_2) + \dots + e^{\lambda_n t}(c_n \mathbf{v}_n)$$