

a) If  $Ax=y$  has no solution, means  $A$  is singular  
(= does not have full rank). Hence,  $Ax=z$  will  
have no solution or an infinite number of solutions,  
depending on if  $z$  is in the range of  $A$ .  
It can never have a unique solution.

**TRUE** b)

**FALSE** c) A matrix may have 0 as an eigenvalue and  
still have a full set of linearly independent  
eigenvectors, and hence be diagonalizable.

**TRUE** d)  $n$  column vectors of length  $m$ . If  $n > m$ ,  
cannot be linearly independent.

**TRUE** e) If the cols of  $A$  are L.I., then  $A^T A$  is  
symmetric positive definite, i.e. all eigenvalues are  
positive. Singular values:  $\sqrt{\lambda}$  of eigs of  $A^T A$ .

2. a)  $Ax=0 \Rightarrow A^T Ax=0$   
 $A^T Ax=0 \Rightarrow x^T A^T Ax=0 \Leftrightarrow (Ax)^T Ax=0 \Rightarrow Ax=0$   
 $\therefore Ax=0 \Leftrightarrow A^T Ax=0$

For an  $m \times n$  matrix  $A$ :  $\text{rank}(A) + \text{nullity}(A) = n$   
 $\text{nullity}(A)$ : dimension of null space of  $A$ .

$A^T A$  is  $n \times n$  and  $\text{rank}(A^T A) + \text{nullity}(A^T A) = n$

$\Rightarrow \text{rank}(A) = \text{rank}(A^T A)$

b)  $A_{m \times n} = U_{m \times n} \Sigma_{n \times n} V_{n \times n}^T$ ,  $U = \begin{pmatrix} | & & | \\ u_1 & \dots & u_n \\ | & & | \end{pmatrix}$ ,  $V = \begin{pmatrix} | & & | \\ v_1 & \dots & v_n \\ | & & | \end{pmatrix}$

$u_1, \dots, u_n$ : and  $v_1, \dots, v_n$ : orthonormal vectors.

$\Sigma$ : diagonal matrix with singular values on diagonal

The SVD always exists. The singular values  $\sigma_i = \sqrt{\lambda_i}$  where  
 $\lambda_i, i=1, \dots, n$ , are eigenvalues of  $A^T A$ .

The number of positive singular values gives the rank of  $A$ .

3d). Double eigenvalue  $\lambda = -1$ .  
 Only one eigenvector,  $\underline{v} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

$$\lim_{t \rightarrow \infty} \frac{u_1(t)}{u_2(t)} = \frac{c_1 + c_2 t}{1/2 c_2} = \infty \quad \text{if } c_2 \neq 0, \text{ or if } c_2 = 0 \text{ and } c_1 \neq 0.$$



A stable improper node.

4. a)  $u$  is prey,  $v$  is predator.

b) critical pts:  $(u, v) = (0, 0)$  and  $(u, v) = (d/c, a/b)$

$$J(u, v) = \begin{pmatrix} a - bv - bu & -b \\ cv & cu - d \end{pmatrix}$$

At the strictly positive point:

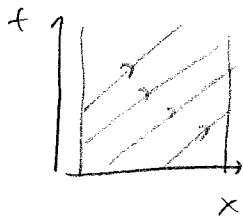
$$J(d/c, a/b) = \begin{pmatrix} 0 & -bd/c \\ ca/b & 0 \end{pmatrix}.$$

$$\text{Eigs: } \lambda^2 + \frac{ca}{b} \cdot \frac{bd}{c} = 0$$

$$\lambda = \pm i \sqrt{ad}.$$

For a linear system, this would be a neutrally stable center. This analysis does not determine the stability of the nonlinear system in this case. It can either be a spiral or a center.

5 d) contd.



$a > 0$ , const

Stencil:



The stencil is stable

$$(u_j^{n+1} = u_j^n - \Delta t a \frac{u_j^n - u_{j-1}^n}{\Delta x})$$

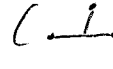
The stencil is unstable



$$(u_j^{n+1} = u_j^n - \Delta t a \frac{u_{j+1}^n - u_j^n}{\Delta x})$$

(Flip sign of  $a$  and this is reversed).

Unstable when stencil not compatible with propagation of information.

Alt: Can use this stencil (  ) but introduce diffusivity by averaging:  $u_j^{n+1} = \frac{1}{2}(u_{j+1}^n + u_{j-1}^n) - \Delta t a \frac{u_{j+1}^n - u_{j-1}^n}{2\Delta x}$  (Lax-Friedrich).

e) Lax-Equivalence theorem:

"The method is convergent if it is consistent and stable."

The theorem is useful because it is much easier to show that a method is consistent and stable than to directly show that it is convergent.

6.  $f(x) = \sum_{n=0}^{\infty} a_n P_n(x)$

Mult by  $P_m(x)$  and integrate:

$$\int_{-1}^1 f(x) P_m(x) dx = \sum_{n=0}^{\infty} a_n \int_{-1}^1 P_n(x) P_m(x) dx = \sum_{n=0}^{\infty} a_n \frac{2}{2n+1} \delta_{nm} = a_m \frac{2}{2m+1}$$

$$\Rightarrow a_m = \frac{2m+1}{2} \int_{-1}^1 f(x) P_m(x) dx.$$