

Math 211

Lecture #30

Solutions of Systems and Stability

November 5, 2003

Multiplicities

A an $n \times n$ matrix

- Distinct eigenvalues $\lambda_1, \dots, \lambda_k$.
- The characteristic polynomial is

$$p(\lambda) = (\lambda - \lambda_1)^{q_1} (\lambda - \lambda_2)^{q_2} \cdot \dots \cdot (\lambda - \lambda_k)^{q_k}.$$

- The *algebraic multiplicity* of λ_j is q_j .
- The *geometric multiplicity* of λ_j is d_j , the dimension of the eigenspace of λ_j .

Properties of Multiplicities

- $q_1 + q_2 + \cdots + q_k = n$.
- $1 \leq d_j \leq q_j$.
- There are d_j linearly independent exponential solutions corresponding to λ_j .
- If $d_j = q_j$ for all j we have n linearly independent exponential solutions.
- If $d_j < q_j$ we can use our **Proposition**.

Generalized Eigenvectors

Definition: If λ is an eigenvalue of A and $[A - \lambda I]^p \mathbf{v} = \mathbf{0}$ for some integer $p \geq 1$, then \mathbf{v} is called a *generalized eigenvector* associated with λ .

Theorem: If λ is an eigenvalue of A with algebraic multiplicity q , then there is an integer $p \leq q$ such that $\text{null}([A - \lambda I]^p)$ has dimension q .

- For each generalized eigenvector \mathbf{v} we can compute $e^{tA} \mathbf{v}$.
- We can find q linearly independent solutions associated with the eigenvalue λ .

Procedure for Solving $\mathbf{x}' = A\mathbf{x}$

- Find the eigenvalues and their **algebraic multiplicities**.
- For each eigenvalue λ with algebraic multiplicity q find q linearly independent solutions associated with λ :
 - ♦ Find the smallest **integer** p such that $\text{null}([A - \lambda I]^p)$ has dimension q .
 - ♦ Find a basis $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_q$ of $\text{null}([A - \lambda I]^p)$.
 - ♦ For $j = 1, 2, \dots, q$ **compute** $\mathbf{x}_j(t) = e^{tA}\mathbf{v}_j$.
- **This** results in n linearly independent solutions.

Procedure for a Complex Eigenvalue

- If λ is complex of algebraic multiplicity q . Then $\bar{\lambda}$ also has multiplicity q .
 - ◆ Find the smallest integer p such that $\text{null}([A - \lambda I]^p)$ has dimension q .
 - ◆ Find a basis $\mathbf{w}_1, \mathbf{w}_2, \dots, \mathbf{w}_q$ of $\text{null}([A - \lambda I]^p)$.
 - ◆ For $j = 1, 2, \dots, q$ **compute** $\mathbf{z}_j(t) = e^{tA}\mathbf{w}_j$.
 - ◆ Compute $\mathbf{x}_j(t) = \text{Re}(\mathbf{z}_j(t))$ and $\mathbf{y}_j(t) = \text{Im}(\mathbf{z}_j(t))$.
- This results in $2q$ linearly independent real solutions corresponding to the eigenvalues λ and $\bar{\lambda}$.

Stability

Autonomous system $\mathbf{x}' = \mathbf{f}(\mathbf{x})$ with an equilibrium point at \mathbf{x}_0 .
The basic question is: What happens to *all solutions* that start near \mathbf{x}_0 as $t \rightarrow \infty$?

- \mathbf{x}_0 is *stable* if for every $\epsilon > 0$ there is a $\delta > 0$ such that a solution $\mathbf{x}(t)$ with $|\mathbf{x}(0) - \mathbf{x}_0| < \delta \Rightarrow |\mathbf{x}(t) - \mathbf{x}_0| < \epsilon$ for all $t \geq 0$.
 - ♦ Every solution that starts close to \mathbf{x}_0 stays close to \mathbf{x}_0 .
 - ♦ In dimension 2 centers and sinks are stable.

- \mathbf{x}_0 is *asymptotically stable* if it is *stable* and there is an $\eta > 0$ such that if $\mathbf{x}(t)$ is a solution with $|\mathbf{x}(0) - \mathbf{x}_0| < \eta$, then $\mathbf{x}(t) \rightarrow \mathbf{x}_0$ as $t \rightarrow \infty$.
 - ◆ Every solution that starts close to \mathbf{x}_0 approaches \mathbf{x}_0 .
 - ◆ In $d = 2$ sinks are asymptotically stable, centers are not.
 - ◆ \mathbf{x}_0 is called a *sink*.
- \mathbf{x}_0 is *unstable* if there is an $\epsilon > 0$ such that for any $\delta > 0$ there is a solution $\mathbf{x}(t)$ with $|\mathbf{x}(0) - \mathbf{x}_0| < \delta$ with the property that there are values of $t > 0$ such that $|\mathbf{x}(t) - \mathbf{x}_0| > \epsilon$.
 - ◆ There are solutions starting arbitrarily close to \mathbf{x}_0 that move away from \mathbf{x}_0 .
 - ◆ In $d = 2$ sources and saddles are unstable.

Dimension 2

- Sinks are *asymptotically stable*.
 - ◆ The eigenvalues have negative real part.
- Sources are unstable.
 - ◆ The eigenvalues have positive real part.
- Saddles are unstable.
 - ◆ One eigenvalue has positive real part.
- Centers are *stable* but not asymptotically stable.
 - ◆ The eigenvalues have real part = 0.

Stability Theorem

Theorem: Let A be an $n \times n$ real matrix.

- **Suppose** the real part of every eigenvalue of A is negative. Then $\mathbf{0}$ is an **asymptotically stable** equilibrium point for the system $\mathbf{x}' = A\mathbf{x}$.
- Suppose A has at least one eigenvalue with positive real part. Then $\mathbf{0}$ is an **unstable** equilibrium point for the system $\mathbf{x}' = A\mathbf{x}$.

Notice that if there are eigenvalues with real part equal to 0, no conclusion is made.

Examples

- Suppose the dimension is 2 and $T^2 - 4D = 0$.
 - ♦ $T < 0 \Rightarrow$ sink. $T > 0 \Rightarrow$ source.
- $\mathbf{y}' = A\mathbf{y}$,

$$A = \begin{pmatrix} -2 & -18 & -7 & -14 \\ 1 & 6 & 2 & 5 \\ 2 & 2 & -3 & 0 \\ -2 & -8 & -1 & -6 \end{pmatrix}.$$

- ♦ A has eigenvalues -1 , -2 , & $-1 \pm i$.
- ♦ $\mathbf{0}$ is asymptotically stable.

Proposition

Proposition: Suppose that A is an $n \times n$ matrix, λ is a number, and \mathbf{v} is a vector.

1. If $[A - \lambda I]\mathbf{v} = \mathbf{0}$, then $e^{tA}\mathbf{v} = e^{\lambda t}\mathbf{v}$.
2. If $[A - \lambda I]^2\mathbf{v} = \mathbf{0}$, then $e^{tA}\mathbf{v} = e^{\lambda t}(\mathbf{v} + t[A - \lambda I]\mathbf{v})$.
3. If $[A - \lambda I]^k\mathbf{v} = \mathbf{0}$, then

$$e^{tA}\mathbf{v} = e^{\lambda t} \left(\mathbf{v} + t[A - \lambda I]\mathbf{v} + \frac{t^2}{2!}[A - \lambda I]^2\mathbf{v} + \cdots + \frac{t^{k-1}}{(k-1)!}[A - \lambda I]^{k-1}\mathbf{v} \right).$$