MA 138 – Calculus 2 with Life Science Applications Linear Systems: Theory (Section 11.1)

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Lecture 39

Equilibria and Stability

- In Section 8.2 we already encountered the concepts of equilibria and stability, when we discussed ordinary DEs. Both concepts can be extended to systems of DEs.
- We now restrict ourselves to the case

$$\frac{d\mathbf{x}}{dt} = A\mathbf{x} \quad \text{with} \quad A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \quad \mathbf{x} = \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix}.$$
We say that a point $\hat{\mathbf{x}} = \begin{bmatrix} \hat{x}_1 \\ \hat{x}_2 \end{bmatrix}$ is an **equilibrium point** of our given system of linear DEs whenever $A\hat{\mathbf{x}} = \mathbf{0}.$

It follows from results in Section 9.2 that if det A ≠ 0, then (0,0) is the only equilibrium of our given system of linear DEs. If det A = 0, then there will be other equilibria.

- If we start a system of DEs at an equilibrium, it remains there at all later times.
- This does not mean that if the system is in equilibrium and is perturbed by a small amount, it will return to the equilibrium.
- Whether or not a solution will return to an equilibrium after a small perturbation is addressed by the stability of the equilibrium.
- In the case when the matrix A has two real and distinct eigenvalues, the solution of our given system of linear DEs is given by

$$x(t)=c_1e^{\lambda_1t}\mathbf{v}_1+c_2e^{\lambda_2t}\mathbf{v}_2,$$

where \mathbf{v}_1 and \mathbf{v}_2 are the eigenvectors corresponding to the eigenvalues λ_1 and λ_2 of A and the constants c_1 and c_2 depend on the initial condition.

■ Knowing the solution will allow us to study the behavior of the solutions as t → ∞ and thus address the question of stability, at least when the eigenvalues are distinct.

Classification of Equilibria

Case 1: A has two distinct real nonzero eigenvalues λ_1 and λ_2

- Both eigenvalues are negative: The equilibrium (0,0) is globally stable, since the solution will approach the equilibrium (0,0) regardless of the starting point. We call (0,0) a sink or a stable node.
- The eigenvalues have opposite signs: Unless we start in the direction of the eigenvector associated with the negative eigenvalue, the solution will not converge to the equilibrium (0,0). We say that the equilibrium (0,0) is unstable and call (0,0) a saddle point.
- **3.** Both eigenvalues are positive: The solution will not converge to (0,0) unless we start at (0,0). We say that the equilibrium (0,0) is unstable, and we call (0,0) a source or an unstable node.

- **Case 2:** A has two complex conjugate eigenvalues $\lambda_{1,2} = \alpha \pm i\beta$
 - 1. $\alpha < 0$: Starting from any point other than (0,0), solutions spiral into the equilibrium (0,0). For this reason, the equilibrium (0,0) is called a **stable spiral**. When we plot solutions as functions of time, they show oscillations. The amplitude of the oscillations decreases over time. We therefore call the oscillations **damped**.
 - α > 0: Starting from any point other than (0,0), the solutions spiral out from the equilibrium (0,0). For this reason, we call the equilibrium (0,0) an unstable spiral. When we plot solutions as functions of time, we see that the solutions show oscillations as before, but this time their amplitudes are increasing.
 - α = 0: Solutions spiral around the equilibrium (0,0), but neither approach nor move away from the equilibrium (since the amplitude of the solutions does not change). The equilibrium (0,0) is called a neutral spiral or a center. The solutions form closed curves.

Example 8

(0,0) globally stable equilibrium; sink or a stable node

$$\frac{d}{dt} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} -3 & \sqrt{2} \\ \sqrt{2} & -2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

$$\lambda_1 = -1$$
 $\lambda_2 = -4$



Example 9 (0,0) unstable equilibrium; saddle point

$$\frac{d}{dt} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 2 & -2 \\ 2 & -3 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$
$$\lambda_1 = 1 \qquad \lambda_2 = -2$$



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Example 10 (0,0) unstable node; source

$$\frac{d}{dt} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 3 & 2 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

$$\lambda_1 = 1$$
 $\lambda_2 = 4$

Example 11 (0,0) stable spiral; negative real part



Example 12 (0,0) unstable spiral; positive real part



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Example 13 (0,0) neutral spiral or center; no real part

$$\frac{d}{dt} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 3 & 2 \\ -5 & -3 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

$$\lambda_{1,2}=\pm i$$





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IK Math

Summary of Stability at (0,0)

The relationships between the characteristic polynomial, eigenvalues, trace, and determinant of a 2 × 2 matrix A are given by

$$det(A - \lambda I_2) = \lambda^2 - trace(A)\lambda + det(A) = 0$$
$$trace(A) = Re(\lambda_1) + Re(\lambda_2) \qquad det(A) = \lambda_1 \lambda_2.$$

Also, the discriminant of the quadratic equation $det(A - \lambda I_2) = 0$ is

$$\Delta = [\operatorname{trace}(A)]^2 - 4 \operatorname{det}(A)$$

Thus the condition $\Delta = 0$ (\equiv repeated eigenvalue) describes the parabola det(A) = 1/4[trace(A)]² in the trace-det plane.

■ Theorem: The origin (0,0) of a system of two linear, homogeneous DEs with constant coefficients is a stable equilibrium ⇔ the real parts of both eigenvalues are negative ⇔ det(A) > 0 and trace(A) < 0.</p>

The stability properties of the equilibrium at the origin can be summarized graphically in terms of the determinant and the trace of the matrix A in the trace-det plane:

