- •Material covered in previous week and this week
 - Chapter 1, Nayfeh and Balachandran (1995, 2006)
 - ➤ Dissipation, Attracting Sets
 - > Lyapunov function based stability analyses
 - Chapter 2, Nayfeh and Balachandran (1995, 2006)
 - Local stability analyses- Linearization
- •Material to be covered today and next week
 - Chapter 2, Nayfeh and Balachandran (1995, 2006)
 - Fixed points of maps and differential equation systems
 - **❖** Local stability analyses

❖ Fixed Points of Maps (Section 2.2, N&B)

Here, we consider fixed points of the map

$$\mathbf{x}_{k+1} = \mathbf{F}(\mathbf{x}_k; \mathbf{M}) \tag{2.2.1}$$

A fixed point x_0 of this map satisfies the condition

$$\mathbf{x}_0 = \mathbf{F}^m(\mathbf{x}_0; \mathbf{M}_0) \text{ for all } m \in \mathcal{Z}$$
 (2.2.2)

where $M = M_0$ is the value of the vector of control parameters. We note that an orbit of a map initiated at a fixed point of the map is the fixed point itself. Moreover, the fixed points of a map are examples of invariant sets.

2.23. Consider the following three-dimensional map of Klein, Baier, and Rössler (1991):

$$x_{n+1} = \alpha - \alpha y_n^2 + dz_n$$

$$y_{n+1} = x_n + \beta + \gamma z_n$$

$$z_{n+1} = y_n$$

- (a) Examine if this map is dissipative in each of the following cases: (i) d < 1, (ii) d = 1, and (iii) d > 1.
- (b) Determine the fixed points of this map and discuss their stability.

❖ Fixed Points of Continuous-Time Systems (Section 2.1, N&B)

In the case of the autonomous system

$$\dot{\mathbf{x}} = \mathbf{F}(\mathbf{x}; \mathbf{M}) \tag{2.1.1}$$

the fixed points are defined by the vanishing of the vector field; that is,

$$\mathbf{F}(\mathbf{x}; \mathbf{M}) = 0 \tag{2.1.2}$$

A location in the state space where this condition is satisfied is called a singular point. At such a point, the integral curve of the vector field F corresponds to the point itself. Also, an orbit of a fixed point is the fixed point itself. Fixed points are also called stationary solutions, critical points, constant solutions, and sometimes steady-state solutions. Physically, a fixed point corresponds to an equilibrium position of a system. Further, fixed points are examples of invariant sets of (2.1.1).

2.25. Consider the following speed—control system investigated by Fallside and Patel (1965):

$$\dot{x}_1 = x_2$$

$$\dot{x}_2 = K_d x_2 - x_1 - G x_1^2 \left(-\frac{x_2}{K_d} + x_1 + 1 \right)$$

(a) For $K_d = -1$ and G = 6, determine the fixed points and their stability.

❖ Fixed Points of Continuous-Time Systems (Section 2.1, N&B): Linearization near an equilibrium solution

In the case of the autonomous system

$$\dot{\mathbf{x}} = \mathbf{F}(\mathbf{x}; \mathbf{M}) \tag{2.1.1}$$

the fixed points are defined by the vanishing of the vector field; that is,

$$\mathbf{F}(\mathbf{x}; \mathbf{M}) = 0 \tag{2.1.2}$$

Let the solution of (2.1.2) for $M = M_0$ be x_0 , where $x_0 \in \mathbb{R}^n$ and $M_0 \in \mathbb{R}^m$. To determine the stability of this equilibrium solution, we superimpose on it a small disturbance y and obtain

$$\mathbf{x}(t) = \mathbf{x}_0 + \mathbf{y}(t) \tag{2.1.3}$$

Substituting (2.1.3) into (2.1.1) yields

$$\dot{\mathbf{y}} = \mathbf{F}(\mathbf{x}_0 + \mathbf{y}; \mathbf{M}_0) \tag{2.1.4}$$

We note that the fixed point $x = x_0$ of (2.1.1) has been transformed into the fixed point y = 0 of (2.1.4). Assuming that F is at least C^2 , expanding (2.1.4) in a Taylor series about x_0 , and retaining only linear terms in the disturbance leads to

$$\dot{\mathbf{y}} = \mathbf{F}(\mathbf{x}_0; \mathbf{M}_0) + D_{\mathbf{x}} \mathbf{F}(\mathbf{x}_0; \mathbf{M}_0) \mathbf{y} + O(\parallel \mathbf{y} \parallel^2)$$

or

$$\dot{\mathbf{y}} \approx D_{\mathbf{x}} \mathbf{F}(\mathbf{x}_0; \mathbf{M}_0) \mathbf{y} \equiv A \mathbf{y}$$
 (2.1.5)

where A, the matrix of first partial derivatives, is called the **Jacobian matrix**. If the components of F are

❖ Fixed Points of Continuous-Time Systems (Section 2.1, N&B): Linearization near an equilibrium solution

$$\dot{\mathbf{y}} = \mathbf{F}(\mathbf{x}_0; \mathbf{M}_0) + D_{\mathbf{x}} \mathbf{F}(\mathbf{x}_0; \mathbf{M}_0) \mathbf{y} + O(\|\mathbf{y}\|^2)$$

or

$$\dot{\mathbf{y}} \approx D_{\mathbf{x}} \mathbf{F}(\mathbf{x}_0; \mathbf{M}_0) \mathbf{y} \equiv A \mathbf{y}$$
 (2.1.5)

where A, the matrix of first partial derivatives, is called the **Jacobian matrix**. If the components of F are

$$F_1(x_1, x_2, \dots, x_n), F_2(x_1, x_2, \dots, x_n), \dots, F_n(x_1, x_2, \dots, x_n),$$

then

$$A = \begin{bmatrix} \frac{\partial F_1}{\partial x_1} & \frac{\partial F_1}{\partial x_2} & \dots & \frac{\partial F_1}{\partial x_n} \\ \frac{\partial F_2}{\partial x_1} & \frac{\partial F_2}{\partial x_2} & \dots & \frac{\partial F_2}{\partial x_n} \\ & & & & & \\ & & & & & \\ \vdots & & & & & \\ \frac{\partial F_n}{\partial x_1} & \frac{\partial F_n}{\partial x_2} & \dots & \frac{\partial F_n}{\partial x_n} \end{bmatrix}$$

Next, we show that the eigenvalues of the constant matrix A provide information about the local stability of the fixed point x_0 . We say local because we have considered a small disturbance and linearized the vector field.

The solution of (2.1.5) that passes through the initial condition $y_0 \in \mathbb{R}^n$ at time $t_0 \in \mathbb{R}$ can be expressed as

$$\mathbf{y}(t) = e^{(t-t_0)A} \mathbf{y}_0 \tag{2.1.6}$$

❖ Fixed Points of Continuous-Time Systems (Section 2.1, N&B): Linearization near an equilibrium solution

If the eigenvalues λ_i of the matrix A are distinct, then there exists a matrix P such that $P^{-1}AP = D$, where D is a diagonal matrix with entries $\lambda_1, \lambda_2, \dots, \lambda_n$; that is,

If the eigenvalues are complex, then the matrix P will also be complex. The columns of the matrix P are the right eigenvectors $\mathbf{p}_1, \mathbf{p}_2, \cdots, \mathbf{p}_n$ of the matrix A corresponding to the eigenvalues $\lambda_1, \lambda_2, \cdots, \lambda_n$; that is, $P = [\mathbf{p}_1 \ \mathbf{p}_2 \ \cdots \ \mathbf{p}_n]$. Hence,

$$AP = [A\mathbf{p}_1 \ A\mathbf{p}_2 \ \cdots \ A\mathbf{p}_n] = [\lambda_1\mathbf{p}_1 \ \lambda_2\mathbf{p}_2 \ \cdots \ \lambda_n\mathbf{p}_n] = PD$$

Consequently,

$$D = P^{-1}AP$$

Introducing the transformation y = Pv into (2.1.5), we obtain

$$P\dot{\mathbf{v}} = AP\mathbf{v}$$
 or $\dot{\mathbf{v}} = D\mathbf{v}$

Hence,

$$\mathbf{v} = e^{(t-t_0)D} \mathbf{v}_0$$

where $\mathbf{v}_0 = \mathbf{v}(t_0) = P^{-1}\mathbf{y}_0$. In terms of \mathbf{y} , this solution becomes

$$\mathbf{y}(t) = Pe^{(t-t_0)D}P^{-1}\mathbf{y}_0 \tag{2.1.7}$$

The matrix $e^{(t-t_0)D}$ is a diagonal matrix with entries $e^{(t-t_0)\lambda_i}$. Hence, the eigenvalues of A are also known as the **characteristic exponents** associated with F at $(\mathbf{x}_0, \mathbf{M}_0)$.

❖ Fixed Points of Continuous-Time Systems (Section 2.1, N&B): Linearization near an equilibrium solution

If the eigenvalues of A are not distinct, then there exists a matrix P such that $P^{-1}AP = J$ is a Jordan canonical form with off-diagonal entries; that is,

where ϕ represents a matrix with zero entries and

In writing the matrix J, we have assumed that A has k distinct eigenvalues. Further, let the (algebraic) multiplicity of the mth eigenvalue λ_m be n_m . Then, the matrix J_m corresponding to the eigenvalue λ_m differs from the diagonal matrix D due to the presence of the elements 1 above the diagonal elements. In this case, the columns \mathbf{p}_i of the matrix P are the generalized eigenvectors corresponding to the eigenvalues λ_i of the matrix A. There are n_m generalized eigenvectors corresponding to the eigenvalue λ_m . These vectors are the nonzero solutions of

$$(A - \lambda_m \mathbf{I}) \mathbf{p} = 0, (A - \lambda_m \mathbf{I})^2 \mathbf{p} = 0, \dots, (A - \lambda_m \mathbf{I})^{n_m} \mathbf{p} = 0$$

For an $n \times n$ matrix with n distinct eigenvalues, the generalized eigenvectors are also the eigenvectors of the matrix. The components of \mathbf{v} have terms of the form $t^k e^{(t-t_0)\lambda_i}$, where the integer k depends on the multiplicity n_i of the eigenvalue λ_i .

❖ Fixed Points of Continuous-Time Systems (Section 2.1, N&B): Classification of equilibrium solutions

When all of the eigenvalues of A have nonzero real parts, the corresponding fixed point is called a hyperbolic fixed point, irrespective of the values of the imaginary parts; otherwise, it is called a nonhyperbolic fixed point.

There are three types of hyperbolic fixed points: sinks, sources, and saddle points. If all of the eigenvalues of A have negative real parts, then all of the components of the disturbance y decay in time, and hence x approaches the fixed point x_0 of (2.1.1) as $t \to \infty$. Therefore, the fixed point x_0 of (2.1.1) is asymptotically stable according to Section 1.4.2. An asymptotically stable fixed point is called a sink. If the matrix A associated with a sink has complex eigenvalues, the sink is also called a stable focus. On the other hand, if all of the eigenvalues of the matrix A associated with a sink are real, the sink is also called a stable node. A sink is stable in forward time (i.e., $t \to \infty$) but unstable in reverse time (i.e., $t \to -\infty$). Further, all sinks qualify as attractors.

If one or more of the eigenvalues of A have positive real parts, some of the components of y grow in time, and x moves away from the fixed point x_0 of (2.1.1) as t increases. In this case, the fixed point x_0 is said to be unstable. When all of the eigenvalues of A have positive real parts, x_0 is said to be a source. If the matrix A associated with a source has complex eigenvalues, the source is also called an unstable focus. On the other hand, if all of the eigenvalues of the matrix A associated with a source are real, the source is also called an unstable node. A source is unstable in forward time but stable in reverse time. Because trajectories move away from a source in forward time, the source is an example of a repellor.

When some, but not all, of the eigenvalues have positive real parts while the rest of the eigenvalues have negative real parts, the associated fixed point is called a saddle point. Because a saddle point is unstable in both forward and reverse times, some authors call it a nonstable fixed point (e.g., Parker and Chua, 1989).

Fixed Points of Continuous-Time Systems (Section 2.1, N&B): Classification of equilibrium solutions

Next, we address nonhyperbolic fixed points. A nonhyperbolic fixed point is unstable if one or more of the eigenvalues of A have positive real parts. If some of the eigenvalues of A have negative real parts while the rest of the eigenvalues have zero real parts, the fixed point $x = x_0$ of (2.1.1) is said to be neutrally or

marginally stable. If all of the eigenvalues of A are purely imaginary and nonzero, the corresponding fixed point is called a **center**.

Example: Duffing oscillator revisited

$$\dot{x}_1 = x_2$$
 (1.4.1)

$$\dot{x}_2 = -x_1 + x_1^3 - 2\mu x_2 \tag{1.4.2}$$

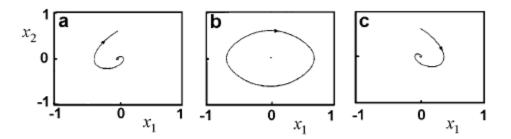


Fig. 2.1.1 Phase portraits in the vicinity of the origin of (1.4.1) and (1.4.2): (a) $\mu = -0.4$, (b) $\mu = 0$, and (c) $\mu = 0.4$.

Example: Duffing oscillator revisited

$$\dot{x}_1 = x_2$$
 (1.4.1)

$$\dot{x}_2 = -x_1 + x_1^3 - 2\mu x_2 \tag{1.4.2}$$

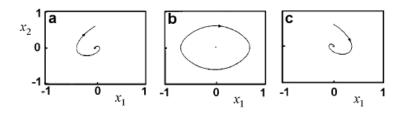


Fig. 2.1.1 Phase portraits in the vicinity of the origin of (1.4.1) and (1.4.2): (a) $\mu=-0.4$, (b) $\mu=0$, and (c) $\mu=0.4$.

Example 2.1. For illustration, we consider the classification of the fixed points (0,0), (-1,0), and (1,0) of (1.4.1) and (1.4.2). In the vicinity of a fixed point, we obtain the following system after linearization:

$$\dot{\mathbf{y}} = \begin{bmatrix} 0 & 1\\ -1 + 3x_1^2 & -2\mu \end{bmatrix} \mathbf{y} \tag{2.1.8}$$

Hence, the eigenvalues of the Jacobian matrix are

$$\lambda_1 = -\mu - \sqrt{\mu^2 - 1 + 3x_1^2}$$
 and $\lambda_2 = -\mu + \sqrt{\mu^2 - 1 + 3x_1^2}$ (2.1.9)

For all three fixed points, both of the eigenvalues have nonzero real parts when $\mu \neq 0$. Hence, all three fixed points are hyperbolic fixed points.

In the vicinity of the fixed point (0,0), (2.1.8) and (2.1.9) become

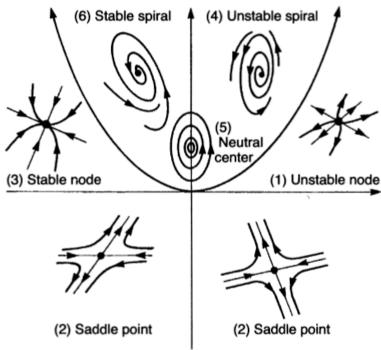
$$\dot{\mathbf{y}} = \begin{bmatrix} 0 & 1 \\ -1 & -2\mu \end{bmatrix} \mathbf{y} \tag{2.1.10}$$

and

$$\lambda_1 = -\mu - \sqrt{\mu^2 - 1}$$
 and $\lambda_2 = -\mu + \sqrt{\mu^2 - 1}$ (2.1.11)

respectively. We conclude from (2.1.11) that the fixed point (0,0) is a center when $\mu=0$, an unstable node when $\mu\leq -1$, an unstable focus when $-1<\mu<0$, a stable focus when $0<\mu<1$, and a stable node when $\mu\geq 1$. In Figures 2.1.1a–c, we show phase portraits in the vicinity of the origin of the x_2-x_1 space when the origin is an unstable focus, a center, and a stable focus, respectively. A positive orbit spirals away from a neighborhood of the unstable focus in Figure 2.1.1a, and a positive orbit spirals into the stable focus in Figure 2.1.1c. The orbit of Figure 2.1.1b, which corresponds to a periodic solution, closes on itself.

- ❖ Fixed Points of Continuous-Time Systems (Section 2.1, N&B): Classification of equilibrium solutions Planar Systems
- ❖ Different types of solutions of a planar linear system of differential equations are shown below. The trace of the Jacobian Matrix is plotted along the x-axis and the determinant of the Jacobian matrix is plotted on the y-axis. Along the parabolic boundary, the two eigenvalues have zero imaginary parts, since the square of the trace = four times the determinant.



- •Material to be covered next week
 - Chapter 2, Nayfeh and Balachandran (1995, 2006)