

#### 5.4. Stability of first-order linear system

Motivation: A solution needs to be stable in order to be useful in practice. The U.S. missile defense system is not yet stable.

Consider

$$\frac{d\vec{x}}{dt} = A\vec{x}, \quad \vec{x}(0) = \vec{C}. \quad (1)$$

where  $A$  is an  $n \times n$  matrix of constants, with  $n$  distinct eigenvalues. The solution formula is

$$\vec{x}(t) = \alpha_1 \vec{a}_1 e^{\lambda_1 t} + \alpha_2 \vec{a}_2 e^{\lambda_2 t} + \dots + \alpha_n \vec{a}_n e^{\lambda_n t}.$$

**Theorem 1.** If the real parts of all the eigenvalues of the coefficient matrix  $A$  are (strictly) negative, then any solution to (1) goes to zero as  $t \rightarrow +\infty$ .

**Theorem 2.** If one or more eigenvalues of  $A$  have positive real parts, then some solutions of (1) go to infinity as  $t \rightarrow +\infty$ .

*Proofs:* They follow from the solution formula if all eigenvalues are distinct. Otherwise, solutions are like  $t^m e^{\lambda t}$  which also go to zero if the real part of  $\lambda$  is negative, or go to infinity if the real part of  $\lambda$  is positive.

Now let us consider a perturbation of (1)

$$\frac{d\vec{x}}{dt} = A\vec{x} + R(t, \vec{x}). \quad (2)$$

Suppose

$$\|R(t, \vec{x})\| \leq \alpha \|\vec{x}\|, \quad \text{on } \{t \geq 0, \|\vec{x}\| < H\}. \quad (3)$$

for some constants  $\alpha$  and  $H > 0$ . Then

**Theorem 3.** If the real parts of all the eigenvalues of  $A$  are (strictly) negative, and (3) holds for a suitably small  $\alpha$ , then the zero solution of (1) is asymptotically stable; *i.e.*, all solutions of (2) with small initial data go to zero as  $t \rightarrow +\infty$ .

**Theorem 4.** If one or more eigenvalues of  $A$  have positive real parts, then the zero solution is not stable, provided that (3) holds for a suitably small  $\alpha$ .

What if one eigenvalue has zero real part and all others have negative real parts? This is called the *critical case*, and is where *bifurcation* occurs. We will discuss these issues in the next section. We provide some concrete stability examples below.

**Examples 1.** Consider

$$\begin{cases} \frac{dx_1}{dt} = \lambda x_1 \\ \frac{dx_2}{dt} = \mu x_2. \end{cases} \quad (4)$$

Suppose that  $\lambda < 0$ ,  $\mu < 0$ . Then all solutions go to zero as  $t \rightarrow +\infty$ . Add perturbation  $R(t, x_1, x_2) : \|R(t, \vec{x})\| < \varepsilon \|\vec{x}\|$ ,

$$\begin{cases} \frac{dx_1}{dt} = \lambda x_1 + R_1(t, x_1, x_2) \\ \frac{dx_2}{dt} = \mu x_2 + R_2(t, x_1, x_2), \end{cases}$$

where  $\varepsilon < \min(|\lambda|, |\mu|)$ , the zero solution is stable: all solutions  $x(t) \rightarrow 0$  as  $t \rightarrow +\infty$ .

**2.** For (4) again, but  $\lambda < 0 < \mu$ . Zero is still a solution. But it is not stable since the initially nearby solution

$$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ \alpha e^{\mu t} \end{pmatrix},$$

where  $\alpha$  is small, grows to infinity.

**3.** Consider now for  $\beta \neq 0$ , the system

$$\begin{cases} \frac{dx_1}{dt} = \beta x_2 \\ \frac{dx_2}{dt} = -\beta x_1. \end{cases}$$

Differentiating the first equation and using the second equation we find

$$\frac{d^2 x_1}{dt^2} + \beta^2 x_1 = 0.$$

We can therefore find the solution formula

$$\begin{cases} x_1 = x_1^0 \cos(\beta t) + x_2^0 \sin(\beta t) \\ x_2 = -x_1^0 \sin(\beta t) + x_2^0 \cos(\beta t). \end{cases}$$

Introduce  $\rho(t) = (x_1^2 + x_2^2)^{\frac{1}{2}}$ , then  $\rho(t) = ((x_1^0)^2 + (x_2^0)^2)^{\frac{1}{2}}$ . See Figure 5.2 for the phase portrait of the solutions. This solution is however unstable to perturbations of the form

$$\begin{pmatrix} 0 \\ \alpha x_2 \end{pmatrix},$$

where  $\alpha > 0$ , because then the equation has the matrix

$$\begin{pmatrix} 0 & \beta \\ -\beta & \alpha \end{pmatrix},$$

one of whose eigenvalues has positive real part.

**Notes 1.** The stability of a nonzero solution  $w(t)$  can be transformed to the stability of the zero solution to the equation for  $v(t) \equiv u(t) - w(t)$ .

**2.** A general nonlinear system

$$\frac{d\vec{x}}{dt} = \vec{F}(t, \vec{x})$$

may be approximated by (2) just as a curve can be approximated by its tangent lines.

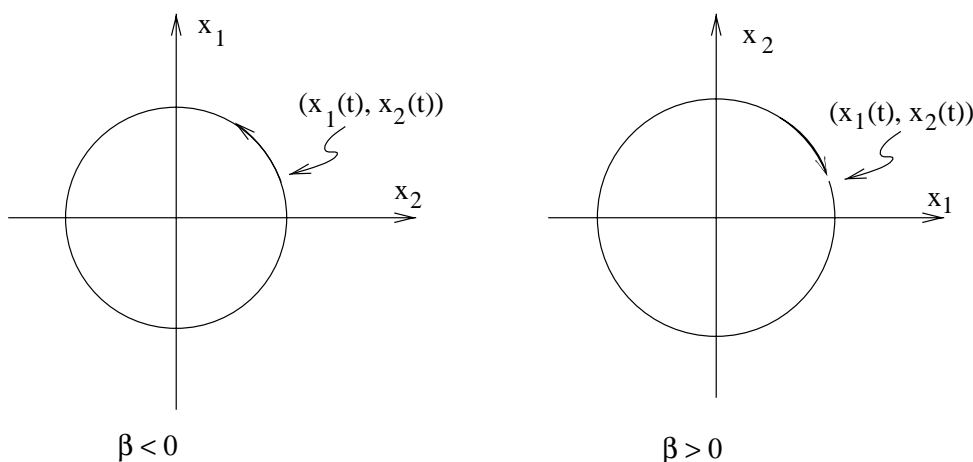


Figure 5.1. Any solution of Example 3 traces a circle.

### 5.5. Hopf bifurcations and example.

Motivation: Bifurcation theory is used in many life sciences, ecological systems, weather system, fluid, chaos, and turbulence.

Consider

$$\frac{d^2u}{dt^2} + (u^2 - \lambda) \frac{du}{dt} + u = 0. \tag{5}$$

It has the solution  $u = 0$ . Let us consider the linearized equation

$$\frac{d^2u}{dt^2} - \lambda \frac{du}{dt} + u = 0.$$

When we try solutions of the form  $u = e^{\mu t}$ , we find

$$\mu^2 - \lambda u + 1 = 0.$$

For  $\lambda < 0$ , both roots have negative real parts, so zero solution is stable. For  $\lambda > 0$ , both roots have positive real parts, so zero solution is unstable. At  $\lambda = 0$ , the roots are purely imaginary  $\mu = i, -i$  and the linearized equation has periodic solutions  $u = e^{it} = \cos t + i \sin t$ , or  $u = e^{-it} = \cos t - i \sin t$ . Both the real and imaginary parts are real solutions  $u(t) = \cos t$  or  $\sin t$ .

We write equation (5) in vector form by introducing  $u_1 = u, u_2 = u'$  :

$$\begin{cases} u_1' &= u_2 \\ u_2' &= (\lambda - u_1^2)u_2 - u_1. \end{cases}$$

Or

$$\vec{u}(t) = \begin{pmatrix} 0 & 1 \\ -1 & \lambda \end{pmatrix} \cdot \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} + \begin{pmatrix} 0 \\ -u_1^2 u_2 \end{pmatrix}.$$

This nonlinear system has nonzero periodic solution near  $\lambda = 0$  :

$$\lambda = \frac{\varepsilon^2}{4} + O(\varepsilon^3), \quad u_1(t) = \varepsilon \cos(\omega t) + O(\varepsilon^3), \quad \omega = 1 + O(\varepsilon^3).$$

We will derive this expansion in perturbation theory next semester. For now we have a bifurcation diagram, see Figure 5.2, and we state a general bifurcation theorem called Hopf bifurcation.

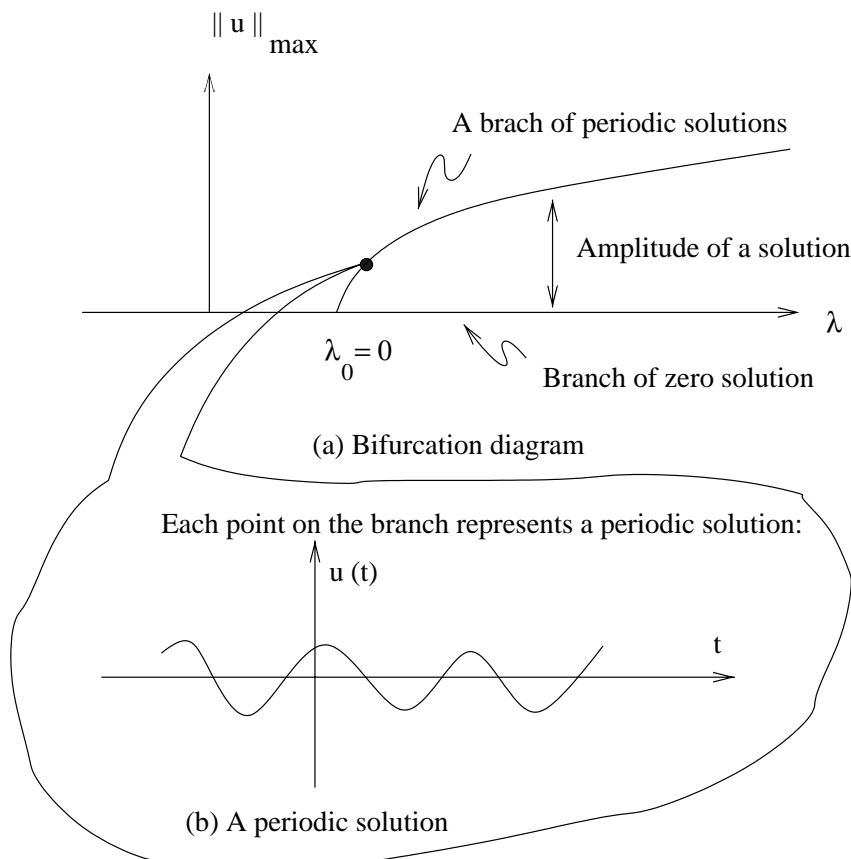


Figure 5.2. Hopf bifurcation diagram.

**Theorem**(Hopf Bifurcation). Suppose the  $n \times n$  matrix  $A(\lambda)$  has eigenvalues  $\mu_j = \mu_j(\lambda), (j = 1, 2, \dots, n)$ , and that for  $\lambda = \lambda_0$ ,  $\mu_1(\lambda_0) = i\beta$ ,  $\mu_2(\lambda_0) = -i\beta$  and  $\text{Re } \mu_j(\lambda_0) \neq 0$  for all  $j > 2$ . Suppose further that  $\text{Re } (\mu_1'(\lambda_0)) \neq 0$ . Then the system of differential equations

$$\frac{du}{dt} = A(\lambda)u + f(u)$$

with  $f(0) = 0, f(u)$  a smooth function of  $u$ , has a branch (continuum) of periodic solutions emanating from  $u = 0, \lambda = \lambda_0$ .

(The direction of bifurcation is not determined by the Hopf Bifurcation Theorem, but must be calculated by a local power series expansion (See Keener)).

We plan to do serious perturbation theory next semester, where we can understand how a mathematician's perturbation and calculation helps locating the position of the 9th planet of the solar system.

### 5.6. Another bifurcation example.

See Keener p.478: Nonlinear Eigenvalue problems.

Consider the *elastica equation* (a.k.a. Euler column)

$$\begin{cases} y'' + (\lambda - \frac{1}{2} \int_0^1 (y')^2 ds)y = 0 \\ y(0) = y(1), \end{cases} \quad (6)$$

where  $\lambda$  is a parameter.

We see that the integral  $\int_0^1 (y')^2(s)ds$  is a number. So let us introduce the number  $\mu = \lambda - \frac{1}{2} \int_0^1 (y')^2(s)ds$ . Then equation (6) becomes

$$y'' + \mu y = 0, \quad y(0) = y(1), \quad (7)$$

which has solutions

$$y(x) = A \sin(n\pi x), \quad \text{for } \mu = n^2\pi^2. \quad (8)$$

These solutions produce

$$\mu = \lambda - \frac{1}{2} \int_0^1 (y')^2(s) = \lambda - \frac{1}{2} \int_0^1 (An\pi)^2 \cos^2(n\pi x)dx = \lambda - \frac{1}{4}(An\pi)^2.$$

To satisfy (6), we need this  $\mu$  to be the same as in (8); that is

$$\frac{A_n^2}{4} = \frac{\lambda}{n^2\pi^2} - 1.$$

Thus we find many branches of solutions besides the zero solution, See Figure 5.3.

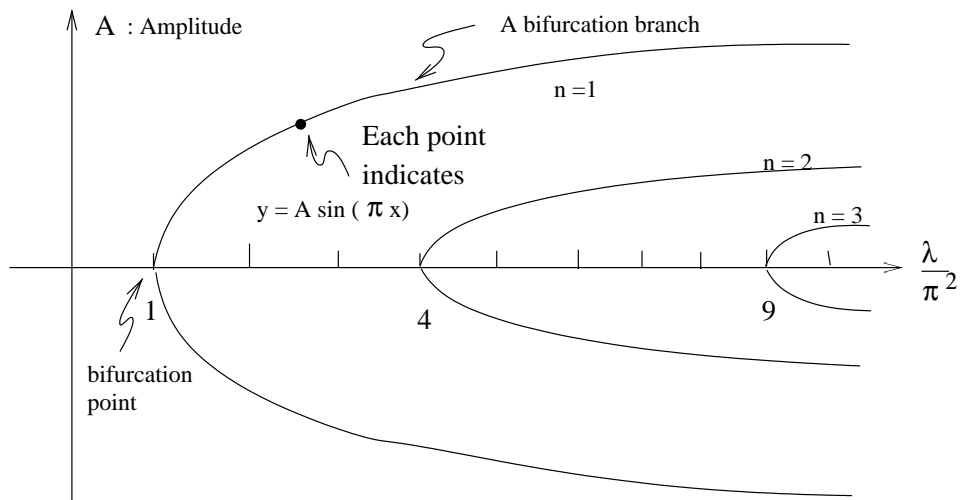


Figure 5.3. A nonlinear eigenvalue bifurcation diagram.