# **Chapter 8**

## **Bifurcations**

#### 8.1 What Are Bifurcations?

One of the important questions you can answer by mathematically analyzing a dynamical system is how the system's long-term behavior depends on its parameters. Most of the time, you can assume that a slight change in parameter values causes only a slight quantitative change in the system's behavior too, with the essential structure of the system's phase space unchanged. However, sometimes you may witness that a slight change in parameter values causes a drastic, qualitative change in the system's behavior, with the structure of its phase space topologically altered. This is called a *bifurcation*, and the parameter values at which a bifurcation occurs are called the *critical thresholds*.

Bifurcation is a qualitative, topological change of a system's phase space that occurs when some parameters are slightly varied across their critical thresholds.

Bifurcations play important roles in many real-world systems as a switching mechanism. Examples include excitation of neurons, pattern formation in morphogenesis (this will be discussed later), catastrophic transition of ecosystem states, and binary information storage in computer memory, to name a few.

There are two categories of bifurcations. One is called a *local bifurcation*, which can be characterized by a change in the stability of equilibrium points. It is called local because it can be detected and analyzed only by using localized information around the equilibrium point. The other category is called a *global bifurcation*, which occurs when non-local features of the phase space, such as limit cycles (to be discussed later), collide with equilibrium points in a phase space. This type of bifurcation can't be characterized just by using localized information around the equilibrium point. In this textbook, we focus only on

the local bifurcations, as they can be easily analyzed using the concepts of linear stability that we discussed in the previous chapters.

Local bifurcations occur when the stability of an equilibrium point changes between stable and unstable. Mathematically, this condition can be written down as follows:

Local bifurcations occur when the eigenvalues  $\lambda_i$  of the Jacobian matrix at an equilibrium point satisfy the following:

For discrete-time models:  $|\lambda_i| = 1$  for some i, while  $|\lambda_i| < 1$  for the rest.

For continuous-time models:  $Re(\lambda_i) = 0$  for some i, while  $Re(\lambda_i) < 0$  for the rest.

These conditions describe a critical situation when the equilibrium point is about to change its stability. We can formulate these conditions in equations and then solve them in terms of the parameters, in order to obtain their critical thresholds. Let's see how this analysis can be done through some examples below.

#### 8.2 Bifurcations in 1-D Continuous-Time Models

For bifurcation analysis, continuous-time models are actually simpler than discrete-time models (we will discuss the reasons for this later). So let's begin with the simplest example, a continuous-time, first-order, autonomous dynamical system with just one variable:

$$\frac{dx}{dt} = F(x) \tag{8.1}$$

In this case, the Jacobian matrix is a  $1 \times 1$  matrix whose eigenvalue is its content itself (because it is a scalar), which is given by dF/dx. Since this is a continuous-time model, the critical condition at which a bifurcation occurs in this system is given by

$$\operatorname{Re}\left(\frac{dF}{dx}\right)\Big|_{x=x_{\text{eq}}} = \frac{dF}{dx}\Big|_{x=x_{\text{eq}}} = 0.$$
 (8.2)

Let's work on the following example:

$$\frac{dx}{dt} = r - x^2 \tag{8.3}$$

The first thing we need to do is to find the equilibrium points, which is easy in this case. Letting dx/dt=0 immediately gives

$$x_{\rm eq} = \pm \sqrt{r},\tag{8.4}$$

which means that equilibrium points exist only for non-negative r. The critical condition when a bifurcation occurs is given as follows:

$$\frac{dF}{dx} = -2x\tag{8.5}$$

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$$\frac{dF}{dx}\Big|_{x=x_{\text{eq}}} = \pm 2\sqrt{r} = 0 \tag{8.6}$$

$$r = 0 ag{8.7}$$

Therefore, now we know a bifurcation occurs when r=0. Moreover, by plugging each solution of Eq. (8.4) into dF/dx = -2x, we know that one equilibrium point is stable while the other is unstable. These results are summarized in Table 8.1.

Table 8.1: Summary of bifurcation analysis of  $dx/dt = r - x^2$ .

Equilibrium point	r < 0	0 < r
$x_{\rm eq} = \sqrt{r}$	doesn't exist	stable
$x_{\rm eq} = -\sqrt{r}$	doesn't exist	unstable

There is a more visual way to illustrate the results. It is called a bifurcation diagram. This works only for systems with one variable and one parameter, but it is still conceptually helpful in understanding the nature of bifurcations. A bifurcation diagram can be drawn by using the parameter being varied as the horizontal axis, while using the location(s) of the equilibrium point(s) of the system as the vertical axis. Then you draw how each equilibrium point depends on the parameter, using different colors and/or line styles to indicate the stability of the point. Here is an example of how to draw a bifurcation diagram in Python:

#### Code 8.1: bifurcation-diagram.py

```
from pylab import *
def xeq1(r):
   return sqrt(r)
def xeq2(r):
   return -sqrt(r)
domain = linspace(0, 10)
plot(domain, xeq1(domain), 'b-', linewidth = 3)
```

```
plot(domain, xeq2(domain), 'r--', linewidth = 3)
plot([0], [0], 'go')
axis([-10, 10, -5, 5])
xlabel('r')
ylabel('x_eq')
show()
```

The result is shown in Fig. 8.1, where the blue solid curve indicates a stable equilibrium point  $x_{\rm eq} = \sqrt{r}$ , and the red dashed curve indicates an unstable equilibrium point  $x_{\rm eq} = -\sqrt{r}$ , with the green circle in the middle showing a neutral equilibrium point. This type of bifurcation is called a *saddle-node bifurcation*, in which a pair of equilibrium points appear (or collide and annihilate, depending on which way you vary r).

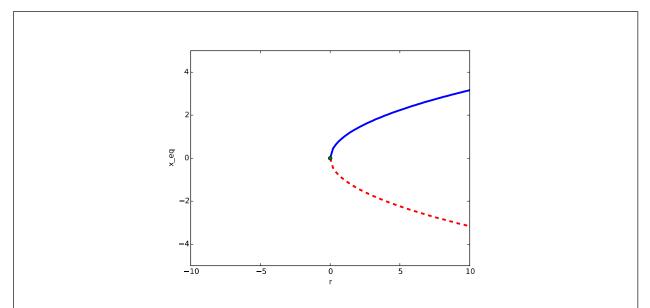


Figure 8.1: Visual output of Code 8.1, showing a bifurcation diagram of a saddle-node bifurcation, obtained from Eq (8.3).

Each vertical slice of the bifurcation diagram for a particular parameter value depicts a phase space of the dynamical system we are studying. For example, for r=5 in the diagram above, there are two equilibrium points, one stable (blue/solid) and the other unstable (red/dashed). You can visualize flows of the system's state by adding a downward arrow above the stable equilibrium point, an upward arrow from the unstable one to the stable one, and then another downward arrow below the unstable one. In this way, it is

clear that the system's state is converging to the stable equilibrium point while it is repelling from the unstable one. If you do the same for several different values of r, you obtain Fig. 8.2, which shows how to interpret this diagram.

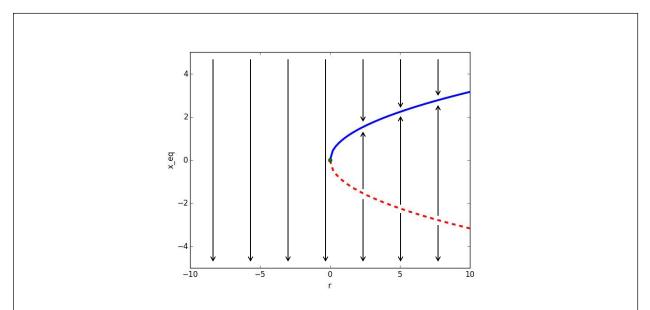


Figure 8.2: How to interpret a bifurcation diagram. Each vertical slice of the diagram depicts a phase space of the system for a particular parameter value.

There are other types of bifurcations. A *transcritical bifurcation* is a bifurcation where one equilibrium point "passes through" another, exchanging their stabilities. For example:

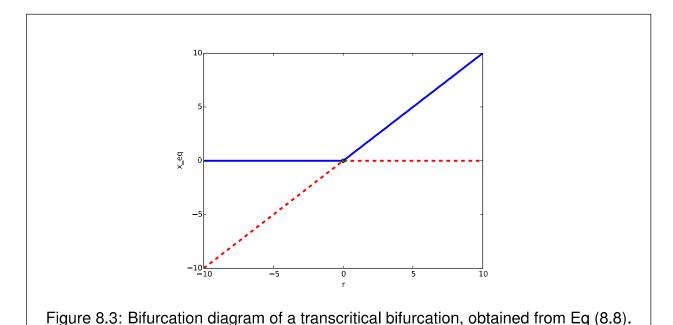
$$\frac{dx}{dt} = rx - x^2 \tag{8.8}$$

This dynamical system always has the following two equilibrium points

$$x_{\rm eq} = 0, r, \tag{8.9}$$

with the exception that they collide when r=0, which is when they swap their stabilities. Its bifurcation diagram is shown in Fig. 8.3.

Another one is a *pitchfork bifurcation*, where an equilibrium point splits into three. Two of these (the outermost two) have the same stability as the original equilibrium point, while the one between them has a stability opposite to the original one's stability. There are two types of pitchfork bifurcations. A *supercritical pitchfork bifurcation* makes a stable



equilibrium point split into three, two stable and one unstable. For example:

$$\frac{dx}{dt} = rx - x^3 \tag{8.10}$$

This dynamical system has the following three equilibrium points

$$x_{\rm eq} = 0, \pm \sqrt{r},\tag{8.11}$$

but the last two exist only for  $r \geq 0$ . You can show that  $x_{\rm eq} = 0$  is stable for r < 0 and unstable for r > 0, while  $x_{\rm eq} = \pm \sqrt{r}$  are always stable if they exist. Its bifurcation diagram is shown in Fig. 8.4.

In the meantime, a *subcritical pitchfork bifurcation* makes an unstable equilibrium point split into three, two unstable and one stable. For example:

$$\frac{dx}{dt} = rx + x^3 \tag{8.12}$$

This dynamical system has the following three equilibrium points

$$x_{\rm eq} = 0, \pm \sqrt{-r},$$
 (8.13)

but the last two exist only for  $r \leq 0$ . Its bifurcation diagram is shown in Fig. 8.5.

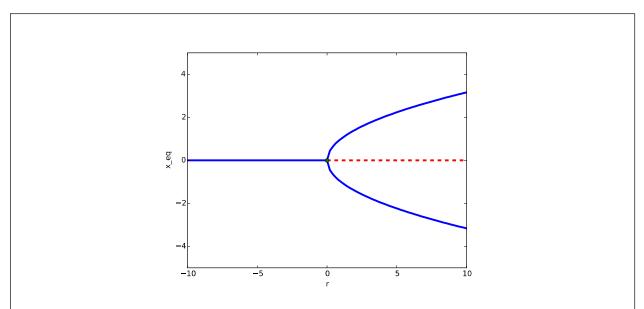


Figure 8.4: Bifurcation diagram of a supercritical pitchfork bifurcation, obtained from Eq (8.10).

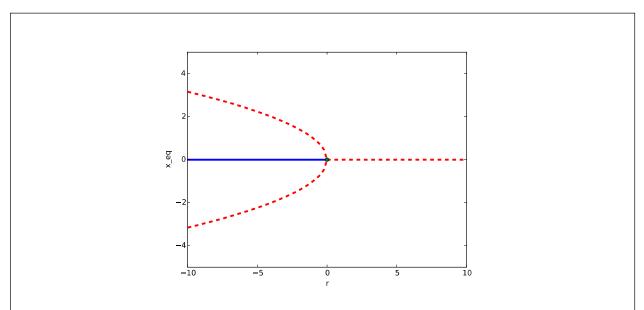


Figure 8.5: Bifurcation diagram of a subcritical pitchfork bifurcation, obtained from Eq (8.10).

These bifurcations can arise in combined forms too. For example:

$$\frac{dx}{dt} = r + x - x^3 \tag{8.14}$$

This dynamical system has three equilibrium points, which are rather complicated to calculate in a straightforward way. However, if we solve dx/dt=0 in terms of r, we can easily obtain

$$r = -x + x^3$$
, (8.15)

which is sufficient for drawing the bifurcation diagram. We can also know the stability of each equilibrium point by calculating

$$\operatorname{Re}\left(\frac{dF}{dx}\right)\Big|_{x=x_{\text{eq}}} = 1 - 3x^2,\tag{8.16}$$

i.e., when  $x^2 > 1/3$ , the equilibrium points are stable, otherwise they are unstable. The bifurcation diagram of this system is shown in Fig. 8.6.

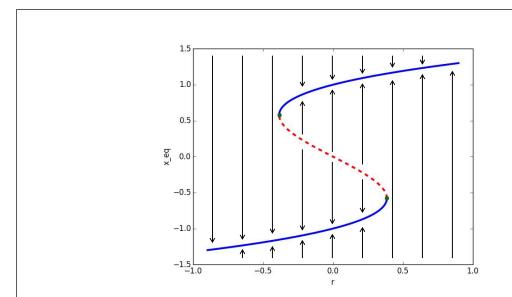


Figure 8.6: Bifurcation diagram showing hysteresis, obtained from Eq (8.14). Arrows are added to help with interpretation.

This diagram is a combination of two saddle-node bifurcations, showing that this system has *hysteresis* as its dynamical property. Hysteresis is the dependence of a system's

output (asymptotic state in this case) not only on the input (parameter r in this case) but also on its history. To understand what this means, imagine that you are slowly changing r from -1 upward. Initially, the system's state stays at the stable equilibrium at the bottom of the diagram, which continues until you reach a critical threshold at  $r\approx 0.4$ . As soon as you cross this threshold, the system's state suddenly jumps up to another stable equilibrium point at the top of the diagram. Such a sudden jump in the system's state is often called a *catastrophe*. You get upset, and try to bring the system's state back to where it was, by reducing r. However, counter to your expectation, the system's state remains high even after you reduce r below 0.4. This is hysteresis; the system's asymptotic state depends not just on r, but also on where its state was in the immediate past. In other words, the system's state works as a memory of its history. In order to bring the system's state back down to the original value, you have to spend extra effort to reduce r all the way below another critical threshold,  $r\approx -0.4$ .

Such hysteresis could be useful; every bit (binary digit) of computer memory has this kind of bifurcation dynamics, which is why we can store information in it. But in other contexts, hysteresis could be devastating—if an ecosystem's state has this property (many studies indicate it does), it takes a huge amount of effort and resources to revert a deserted ecosystem back to a habitat with vegetation, for example.

<u>Exercise 8.1</u> Conduct a bifurcation analysis of the following dynamical system with parameter r:

$$\frac{dx}{dt} = rx(x+1) - x \tag{8.17}$$

Find the critical threshold of r at which a bifurcation occurs. Draw a bifurcation diagram and determine what kind of bifurcation it is.

Exercise 8.2 Assume that two companies, A and B, are competing against each other for the market share in a local region. Let x and y be the market share of A and B, respectively. Assuming that there are no other third-party competitors, x + y = 1 (100%), and therefore this system can be understood as a one-variable system. The growth/decay of A's market share can thus be modeled as

$$\frac{dx}{dt} = ax(1-x)(x-y),\tag{8.18}$$

where x is the current market share of A, 1-x is the size of the available potential customer base, and x-y is the relative competitive edge of A, which can be

rewritten as x - (1 - x) = 2x - 1. Obtain equilibrium points of this system and their stabilities.

Then make an additional assumption that this regional market is connected to and influenced by a much larger global market, where company A's market share is somehow kept at p (whose change is very slow so we can consider it constant):

$$\frac{dx}{dt} = ax(1-x)(x-y) + r(p-x)$$
 (8.19)

Here r is the strength of influence from the global to the local market. Determine a critical condition regarding r and p at which a bifurcation occurs in this system. Draw its bifurcation diagram over varying r with a=1 and p=0.5, and determine what kind of bifurcation it is.

Finally, using the results of the bifurcation analysis, discuss what kind of marketing strategy you would take if you were a director of a marketing department of a company that is currently overwhelmed by its competitor in the local market. How can you "flip" the market?

### 8.3 Hopf Bifurcations in 2-D Continuous-Time Models

For dynamical systems with two or more variables, the dominant eigenvalues of the Jacobian matrix at an equilibrium point could be complex conjugates. If such an equilibrium point, showing an oscillatory behavior around it, switches its stability, the resulting bifurcation is called a *Hopf bifurcation*. A Hopf bifurcation typically causes the appearance (or disappearance) of a *limit cycle* around the equilibrium point. A limit cycle is a cyclic, closed trajectory in the phase space that is defined as an asymptotic limit of other oscillatory trajectories nearby. You can check whether the bifurcation is Hopf or not by looking at the imaginary components of the dominant eigenvalues whose real parts are at a critical value (zero); if there are non-zero imaginary components, it must be a Hopf bifurcation.

Here is an example, a dynamical model of a nonlinear oscillator, called the *van der Pol oscillator*:

$$\frac{d^2x}{dt^2} + r(x^2 - 1)\frac{dx}{dt} + x = 0 ag{8.20}$$

This is a second-order differential equation, so we should introduce an additional variable