

# Numerical Analysis for Non-Linear Optimization

The notes for this week have been taken from > Kelley, W. G. (2010). The theory of differential equations. Springer.

## First-Order Linear Equations

In the next week, we will study equation of the form:

$$x'(t) = f(t, x(t))$$

where  $f : (a, b) \times (c, d) \rightarrow \mathbb{B}$  is continuous,  $-\infty \leq a < b \leq \infty$  and  $-\infty \leq c, d \leq \infty$ .

## Variation of Constants Formula

Consider

$$\begin{cases} x' = p(t)x + q(t) \\ x(t_0) = x_0 \end{cases} \quad (1.5)$$

where  $p, q : (a, b) \rightarrow \mathbb{R}$  are continuous functions,  $-\infty \leq a < b \leq \infty$ ,  $t_0 \in (a, b)$  and  $x_0 \in \mathbb{R}$ .

Then for  $t \in (a, b)$ , the solution of the above equation is given by

$$x(t) = e^{\int_{t_0}^t p(\tau) d\tau} x_0 + e^{\int_{t_0}^t p(\tau) d\tau} \int_{t_0}^t e^{-\int_{t_0}^s p(\tau) d\tau} q(s) ds,$$

## Example

Consider  $p(\tau) = \tau$  and  $q(\tau) = \sin(\tau)$ . First, let's write a function to compute the solution of the above equation.

```
def phi(p, q, x0, t0, t):
    tau, s = var('tau s')
    # Compute the exponential integrating factor
    exp_int_p = exp(integrate(p(tau), (tau, t0, t)))

    # First term
    term1 = exp_int_p * x0

    # Inner integral inside the second term
    inner_exp = exp(-integrate(p(tau), (tau, t0, s)))
    inner_integrand = inner_exp * q(s)
    inner_integral = integrate(inner_integrand, (s, t0, t))
```

```
# Second term
term2 = exp_int_p * inner_integral

return term1 + term2
```

Then, let's define the adequate parameters:

```
t, t0 = var('t t0')

assume( t0 > 0)
assume(t > t0)

p(t) = -2
q(t) = 140 + 20 * exp(-2*t)
x0 = 40
x = phi(p, q, x0, 0, t)
print(x.expand())
```

Finally, let's verify that the given function is a solution of the initial value problem.

```
diff(x,t).expand()

(p(t)*x + q(t)).expand()

x(t=0)
```

## Autonomous Equations

The equation

$$x' = f(x) \tag{1.6}$$

is called autonomous because  $f$  doesn't depend explicitly on  $t$ . We assume that  $f : \mathbb{R} \rightarrow \mathbb{R}$  is a continuously differentiable function.

### Theorem 1.8

If  $x$  is a solution of the autonomous differential equation (1.6) on an interval  $(a, b)$ , where  $-\infty \leq a < b \leq \infty$ , then for any constant  $c$ , the function  $y$  defined by  $y(t) := x(t - c)$ , for  $t \in (a + c, b + c)$ , is a solution of (1.6) on  $(a + c, b + c)$ .

### Definition 1.9

If  $f(x_0) = 0$  we say that  $x_0$  is an **equilibrium point** for the differential equation (1.6). If, in addition, there is a  $\delta > 0$  such that

$$f(x) \neq 0 \quad \text{for } |x - x_0| < \delta, \quad x \neq x_0,$$

then we say  $x_0$  is an **isolated equilibrium point**.

**Example (1.10)**

Consider the equation  $x' = -2(x - 70)$ .

- $x = 70$  is the only equilibrium point.
- Any solution is in the form  $x(t) = De^{-2t} + 70$ .
- We can verify that  $x(t - c)$  is a solution.

**Definition 1.11**

Let  $\phi$  be a solution of (1.6) with maximal interval of existence  $(\alpha, \omega)$ . Then the set

$$\{\phi(t) : t \in (\alpha, \omega)\}$$

is called an orbit for the differential equation (1.6).

Note that the orbits for

$$x' = -2(x - 70)$$

are the images of the sets

$$(-\infty, 70), \quad \{70\}, \quad (70, \infty).$$

under the solution  $x(t)$ .

```
var('x t x0')
f(x) = -2 * (x - 70)

def G(x0):
    return 70 + (x0 - 70) * exp(-2 * t)

phi = G(x0)

phi.diff(t).expand()
f(phi).expand()

phi(t = 0)

# Initial conditions
x0_values = [40, 50, 60, 70, 80, 90, 100]
solutions = [ phi(x0=x0) for x0 in x0_values]

# Slope field in the (t, x) plane
slope_field = plot_slope_field(
    f, (t, 0, 5), (x, 40, 100), color='lightgray')
```

```

# Plot solution curves
solution_plots = sum(
    [plot(sol, (t, 0, 3),) for sol, x0 in zip(solutions, x0_values)])

# Equilibrium line
equilibrium_line = plot(
    70, (t, 0, 5),
    color='black', linestyle='--', legend_label='Equilibrium x = 70')

# Show the complete plot
show(slope_field + solution_plots + equilibrium_line, figsize=6)

```

### Theorem 1.12

Assume that  $f : \mathbb{R} \rightarrow \mathbb{R}$  is continuously differentiable. Then two orbits of (1.6) are either disjoint sets or are the same set.

### Example 1.13 (*Logistic Growth*)

The logistic law of population growth (Verhulst [52], 1838) is

$$N' = rN \left(1 - \frac{N}{K}\right),$$

where  $N$  is the number of individuals in the population,  $r(1 - N/K)$  is the *per capita growth rate* that declines with increasing population, and  $K > 0$  is the *carrying capacity* of the environment.

What are the orbits of the differential equation in this case?

```

var('t r K')
N = function('N')(t)
Np = r*N*(1-N/K)
factor(diff(Np, t).subs({diff(N,t):Np}))

# Define variables
var('t N r K N0')

f = r * N * (1-N/K)

# phi = (K*N0*exp(r*t))/(K + N0*(exp(r*t) - 1))
def G(N0, K, r):
    return (K*N0*exp(r*t))/(K + N0*(exp(r*t) - 1))

phi = G(N0, K, r)

phi.diff(t).factor()

f(N = phi).factor()

phi(t=0)

```

```

# Initial conditions
import numpy as np

K = 100
r = 0.1
NO_values = np.arange(1, 2*K, 10)

solutions = [ G(NO, K, r) for NO in NO_values]

T = 110

# Slope field in the (t, x) plane
slope_field = plot_slope_field(f(r=r, K =K), (t, 0, T), (N, 1, 2*K), color='lightgray')

# Plot solution curves
solution_plots = sum([plot(sol, (t, 0, T),) for sol, NO in zip(solutions, NO_values)])

# Equilibrium line
equilibrium_line = plot(K, (t, 0, T), color='black', linestyle='--', legend_label=f'N = {K}')

# Show the complete plot
show(slope_field + solution_plots + equilibrium_line, figsize=6)

```

**Definition 1.15** We say that an equilibrium point  $x_0$  of the differential equation (1.6) is *stable* provided given any  $\epsilon > 0$  there is a  $\delta > 0$  such that whenever  $|x_1 - x_0| < \delta$  it follows that the solution  $\phi(\cdot, x_1)$  exists on  $[0, \infty)$  and  $|\phi(t, x_1) - x_0| < \epsilon$ , for  $t \geq 0$ .

If, in addition, there is a  $\delta_0 > 0$  such that  $|x_1 - x_0| < \delta_0$  implies that  $\lim_{t \rightarrow \infty} \phi(t, x_1) = x_0$ , then we say that the equilibrium point  $x_0$  is *asymptotically stable*. If an equilibrium point is not stable, then we say that it is *unstable*.

For the differential equation  $N' = rN(1 - N/K)$  the equilibrium point  $N_1 = 0$  is unstable and the equilibrium point  $N_2 = K$  is asymptotically stable.

## Potential Energy Function

**Definition 1.16** We say that  $F$  is a *potential energy function* for the differential equation (1.6) provided  $f(x) = -F'(x)$ .

**Theorem 1.17** If  $F$  is a potential energy function for (1.6), then  $F(x(t))$  is strictly decreasing along any nonconstant solution  $x$ . Also,  $x_0$  is an equilibrium point of (1.6) iff  $F'(x_0) = 0$ . If  $x_0$  is an isolated equilibrium point of (1.6) such that  $F$  has a local minimum at  $x_0$ , then  $x_0$  is asymptotically stable.

### Example 1.18

Find the potential function for  $x' = -2(x - 70)$ .

```

var('u')

f = -2*(u-70)

F = -integrate(f, u, 0, x)
F

tc = 70
f(u=tc)

X = function('X')(t)
ode = diff(X, t) == f(u=X)
ode

var('X0')
phi = desolve(ode, X, [0, X0])

F(x=phi).diff(t).expand()

(-f(u=phi)^2).expand()

```

## Generalized Logistic Equation

Suppose  $p, q$  are continuous, and  $x$  is the solution of

$$x' = -p(t)x + q(t) \quad (1.8)$$

with  $x(t) \neq 0$  on  $I$ .

Then  $y(t) = \frac{1}{x(t)}$ ,  $t \in I$  is a solution of the *generalized linear equation*:

$$y' = (p(t) - q(t)y) y \quad (1.9)$$

**Theorem 1.19:** If  $y_0 \neq 0$  and

$$\frac{1}{y_0} + \int_{t_0}^t q(s) e^{\int_{t_0}^s p(\tau) d\tau} ds \neq 0, \quad t \in I,$$

then the solution of the IVP

$$y' = (p(t) - q(t)y) y, \quad y(t_0) = y_0, \quad t_0 \in I \quad (1.10)$$

is given by

$$y(t) = \frac{e^{\int_{t_0}^t p(\tau) d\tau}}{\frac{1}{y_0} + \int_{t_0}^t q(s) e^{\int_{t_0}^s p(\tau) d\tau} ds}. \quad (1.11)$$

**Corollary 1.20** If  $y_0 \neq 0$  and

$$\frac{1}{y_0} - \frac{1}{K} + \frac{1}{K} e^{\int_{t_0}^t p(s) ds} \neq 0, \quad t \in I,$$

then the solution of the IVP

$$y' = p(t) \left[ 1 - \frac{y}{K} \right] y, \quad y(t_0) = y_0 \quad (1.13)$$

is given by

$$y(t) = \frac{e^{\int_{t_0}^t p(s) ds}}{\frac{1}{y_0} - \frac{1}{K} + \frac{1}{K} e^{\int_{t_0}^t p(s) ds}}. \quad (1.14)$$

**Theorem 1.21** Assume  $p : [t_0, \infty) \rightarrow [0, \infty)$  is continuous and  $\int_{t_0}^{\infty} p(t) dt = \infty$ . Let  $y(t)$  be the solution of the IVP (1.13) with  $y_0 > 0$ , then  $y(t)$  exists on  $[t_0, \infty)$ . Also if  $0 < y_0 < K$ , then  $y(t)$  is nondecreasing with  $\lim_{t \rightarrow \infty} y(t) = K$ . If  $y_0 > K$ , then  $y(t)$  is nonincreasing with  $\lim_{t \rightarrow \infty} y(t) = K$ .

## Bifurcation

In the context of differential equations, a bifurcation refers to a qualitative change in the behavior of a system as a parameter is varied. More specifically, it occurs when a small smooth change made to the value of a parameter causes a sudden ‘bifurcation’ or splitting in the structure of the system’s solutions — such as the number or stability of equilibrium points.

**Example 1.22**  $x' = \lambda(x - 1)$

```
var("t x")

for l in [-1,0,1]:

    f = 1 * (x-1)

    g = Graphics()

    g += plot_slope_field(f, (t, -1, 1), (x, -1, 3), color='blue')

    g += plot( 1 , (t, -1, 1), color = 'red')

    g.show()

var("t x")

for l in [-1,0,1]:
```

```

a = sqrt(abs(l))

f = 1 + x^2

g = Graphics()

g += plot_slope_field(f, (t, -1, 1), (x, -a-1, a+1), color='blue')

if l == 0 :
    g += plot(0, (t,-1,1), color = "red")

if l < 0:
    g += plot(-a, (t,-1,1), color = "red")
    g += plot(a, (t,-1,1), color = "red")

g.show()

```