

Find the **optimum** (maximum or minimum) of an **objective function**  $f$

## **Application examples:**

**Process Design:** Find the operating conditions of temperature, pressure and catalyst concentration to maximize the yield of a chemical reactor.

**Resource Allocation:** For a production process, find the amounts of various inputs that minimizes the total manufacturing cost.

# Objective Function

In general, the objective function depends on several variables:  $f(x_1, x_2, \dots, x_n)$ , e.g. the yield of the reactor may depend on temperature, pressure, and catalyst concentration. These variables are termed the **control variables**.

In most realistic optimization problems, the control variables are **constrained**, e.g. the reactor may be able to operate only within a specific temperature range, constrained by the thermodynamics of the reaction and reactor vessel safety considerations.

The objective function may be determined from fundamental considerations, e.g. by writing a model for the process based on conservation balances, or it may be determined empirically, e.g. by curve-fitting a function to process data.

- ▶ Unconstrained Optimization
- ▶ Constrained Optimization
- ▶ Linear Programming

A maximization problem can be converted into a minimization problem in terms of the negative of the objective function, and vice versa.

# Minimum of a function of one variable

The necessary and sufficient condition for  $f$  to have a minimum at  $x=x^*$  is:

$$h(x^*) = f'(x^*) = \left. \frac{df}{dx} \right|_{x=x^*} = 0$$

and

$$h'(x^*) = f''(x^*) = \left. \frac{d^2f}{dx^2} \right|_{x=x^*} > 0$$

e.g.  $f(x)=x^2 - x \Rightarrow h(x) = 2x - 1, h'(x) = 2$

$\Rightarrow x^* = 1/2$  is root of  $h(x)$ , and a minimum of  $f(x)$ .

Note: The problem of minimization of the function  $f(x)$  has been transformed to that of finding the root of the function  $h(x) = f'(x)$ .

The root of a function  $h(x)$  is the value of  $x$  for which  $h(x)=0$ . Numerical methods may be necessary to find the root.

# Bisection Method

**Method:** At each step, the search interval is reduced by a factor of 2.

**Principle:** A function has at least one root in an interval in which it changes sign.

**Problem:** Find root of  $h(x)$  in interval  $[a,b]$ .

**Procedure:** Evaluate  $h(a)$ ,  $h(b)$  and  $h(c)=(a+b)/2$

If  $h(a)h(c) < 0 \Rightarrow x^* \in [a,c]$

If  $h(b)h(c) < 0 \Rightarrow x^* \in [c,b]$

Continue procedure until convergence criteria is satisfied,

e.g.  $h(x) < 10^{-6}$

**Advantages:**

Simple to implement

**Disadvantages:**

Slow.

Not useful when the function depends on more than one variable.

## Example 1

Find minimum of  $f(x) = x^4/4 + 4x^3/3 - 10x + 1$  in the interval  $[1,2] \Leftrightarrow$   
find root of  $h(x) = x^3 + 4x^2 - 10$  in the interval  $[1,2]$ .

Step 1:  $h(1)=-5$ ,  $h(2)=14$ ,  $h(1.5)=2.375 \Rightarrow x^* \in [1,1.5]$

Step 2:  $h(1.25)=-1.796 \Rightarrow x^* \in [1.25,1.5]$

Converges to  $x^* = 1.365$  after twelve steps.

# Newton-Raphson Method

Iterative Sequence:

$$x_{k+1} = x_k - \frac{h(x_k)}{h'(x_k)} = x_k - \frac{f'(x_k)}{f''(x_k)} \quad k = 0, 1, \dots$$

Requires an initial estimate  $x_0$ .

## **Advantages**

Rapid convergence if initial estimate is good.

## **Disadvantages**

Requires good initial estimate, or convergence can be slow or may even diverge.

Requires evaluation of derivatives.

## Example 2

$$h(x) = x^3 + 4x^2 - 10, x_0 = 1.5$$

$$x_1 = x_0 - \frac{h(x_0)}{h'(x_0)} = 1.5 - \frac{h(1.5)}{h'(1.5)} = 1.5 - \frac{2.375}{18.75} = 1.3733$$

$$x_2 = x_1 - \frac{h(x_1)}{h'(x_1)} = 1.3733 - \frac{0.1343}{16.645} = 1.3653$$

Converges to  $x^* = 1.365$  in 2 steps.



# Minimum of a function of several variables

Minimize  $f(\mathbf{x})$ .

$f$  is a scalar function of the control variables  $\mathbf{x}=[x_1, x_2, \dots, x_n]$

If  $f$  has a minimum at  $\mathbf{x}=\mathbf{x}^* \Rightarrow \nabla f|_{\mathbf{x}=\mathbf{x}^*} = 0$ .

## Newton's Method

At minimum (using Taylor series expansion for  $\nabla f$ ):

$$\begin{aligned}\nabla f(\mathbf{x}^{k+1}) &= \nabla f(\mathbf{x}^k) + \mathbf{H}(\mathbf{x}^k) \cdot \Delta \mathbf{x}^k = 0 & \mathbf{H} &= \nabla \nabla f \\ \Rightarrow \mathbf{H}(\mathbf{x}^k) \cdot \Delta \mathbf{x}^k &= -\nabla f(\mathbf{x}^k) \\ \Rightarrow \Delta \mathbf{x}^k &= \mathbf{x}^{k+1} - \mathbf{x}^k = -\mathbf{H}^{-1}(\mathbf{x}^k) \cdot \nabla f(\mathbf{x}^k)\end{aligned}$$

## Example 3 — slide 1 of 2

$$f(x) = (x_1 - 2)^4 + (x_1 - 2)^2 x_2^2 + (x_2 + 1)^2 \quad \mathbf{x}^0 = [1 \ 1]$$

$$\nabla f = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \end{bmatrix} = \begin{bmatrix} 4(x_1 - 2)^3 + 2(x_1 - 2)x_2^2 \\ 2(x_1 - 2)^2 x_2 + 2(x_2 + 1) \end{bmatrix}$$

$$\begin{aligned} \mathbf{H} &= \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2^2} \end{bmatrix} \\ &= \begin{bmatrix} 12(x_1 - 2)^2 + 2x_2^2 & 4(x_1 - 2)x_2 \\ 4(x_1 - 2)x_2 & 2(x_1 - 2)^2 + 2 \end{bmatrix} \end{aligned}$$

## Example 3 — slide 2 of 2

$$\nabla f(\mathbf{x}^0) = \begin{bmatrix} -6 \\ +6 \end{bmatrix} \quad \mathbf{x}^0 = [1 \ 1]$$

$$\mathbf{H}(\mathbf{x}^0) = \begin{bmatrix} 14 & -4 \\ -4 & 4 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 14 & -4 \\ -4 & 4 \end{bmatrix} \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \end{bmatrix} = \begin{bmatrix} +6 \\ -6 \end{bmatrix}$$

$$\Rightarrow \Delta x_1 = 0 \quad \Delta x_2 = -1.5$$

#	$x_1$	$x_2$
0	+1.000	+1.000
1	+1.000	-0.500
2	+1.391	-0.696
3	+1.746	-0.949
4	+1.986	-1.048
5	+1.999	-1.000

# Method of Steepest Descent

Minimize  $f(\mathbf{x})$ .

$$\mathbf{x}^{k+1} = \mathbf{x}^k - t \nabla f(\mathbf{x}^k)$$

$t$  is the minimum of  $g(t)=f(\mathbf{x}^{k+1})$ , or the root of  $g'(t)$ .

Does not require evaluation of second derivatives.

## Example

$$f(x) = (x_1 - 2)^4 + (x_1 - 2)^2 x_2^2 + (x_2 + 1)^2 \quad \mathbf{x}^0 = [1 \ 1]$$

$$\nabla f = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \end{bmatrix} = \begin{bmatrix} 4(x_1 - 2)^3 + 2(x_1 - 2)x_2^2 \\ 2(x_1 - 2)^2 x_2 + 2(x_2 + 1) \end{bmatrix}$$

$$\mathbf{x}^1 = \mathbf{x}^0 - t \nabla f(\mathbf{x}^0) = \begin{bmatrix} 1 \\ 1 \end{bmatrix} - t \begin{bmatrix} -6 \\ +6 \end{bmatrix} = \begin{bmatrix} 1 + 6t \\ 1 - 6t \end{bmatrix}$$

$$g(t) = f(\mathbf{x}^1) = 2(1 - 6t)^4 + (2 - 6t)^2$$

Find the value of  $t$  for which  $g'(t)=0$  by a root finding method,

$$t \approx 0.25 \Rightarrow \mathbf{x}^1 = [1+6(0.25) \ 1-6(0.25)] = [2.5, -0.5].$$

Continuing one finds the same solution as before.