

# Optimasi Tanpa Kendala: Satu Peubah

Pertemuan ke-4

STA1373- Optimisasi Statistika

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# Outline

## Elimination Method

- Search with fixed step size
- Exhaustive Search
- Dichotomous/ Binary Search
- **Fibonacci Search**
- Golden Section Search
- Simplex Search

## Interpolation Method

- Requiring no derivatives: **Quadratic interpolation**
- Requiring derivatives: Cubic, Direct root (Newton, Quasi-Newton, Secant)

# Pengantar: Optimasi Tak Linier

Optimasi tak linier adalah pencarian dan penentuan nilai optimum (maksimum dan atau minimum) dari suatu persamaan nonlinier, yang secara umum dapat dituliskan sebagai berikut:

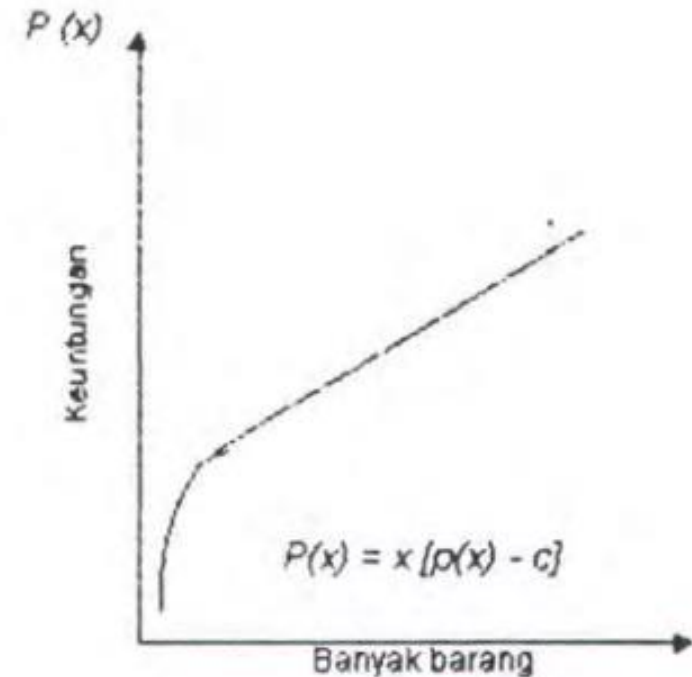
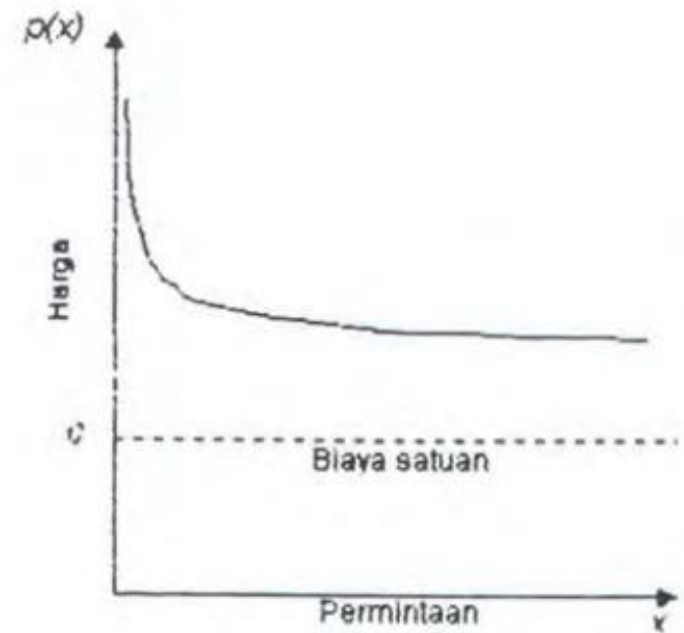
$$f(x^*) = \min_x f(x), x \in \mathbb{R}$$

Dimana  $f(x)$  merupakan fungsi nonlinier yang dapat melibatkan  $n$  buah peubah.

# Contoh permasalahan

## Masalah produk campuran dan elastisitas harga

Penentuan fungsi keuntungan pada perusahaan besar tidak hanya dipengaruhi oleh jenis produk saja, melainkan dapat juga dipengaruhi oleh elastisitas harga di mana banyaknya barang yang dapat dijual berbanding terbalik dengan harganya. Hal ini menjadikan kurva harga permintaan menjadi tidak linear.



# NUMERICAL EVALUATION OF ROOTS OF EQUATIONS

- Metode yang sebelumnya dipelajari adalah metode berbasis iteratif, Dimana algoritma pencarian titik optimum adalah sbb:

**Remarks.** For all iterative procedures, care should be taken, especially, of the following:

- (i) In choosing the initial value.
- (ii) In choosing an algorithm that converges.
- (iii) Rate of convergence should be known.
- (iv) Stopping criterion should be given in advance.

We may use one of the following stopping criteria:

- (a) Stop when the gradient vector  $g_i$  is small, i.e.,  $|g_i| < \epsilon$  for a given  $\epsilon$ .
- (b) Components of iterates do not change. That is,

$$\max_j |j\text{th component of } g_i| < \epsilon.$$

- (c) The function values do not change much. That is, for a given  $\epsilon$ ,

$$|f(\mathbf{x}_n) - f(\mathbf{x}_{n+1})| < \epsilon$$

at the  $n$ th stage.

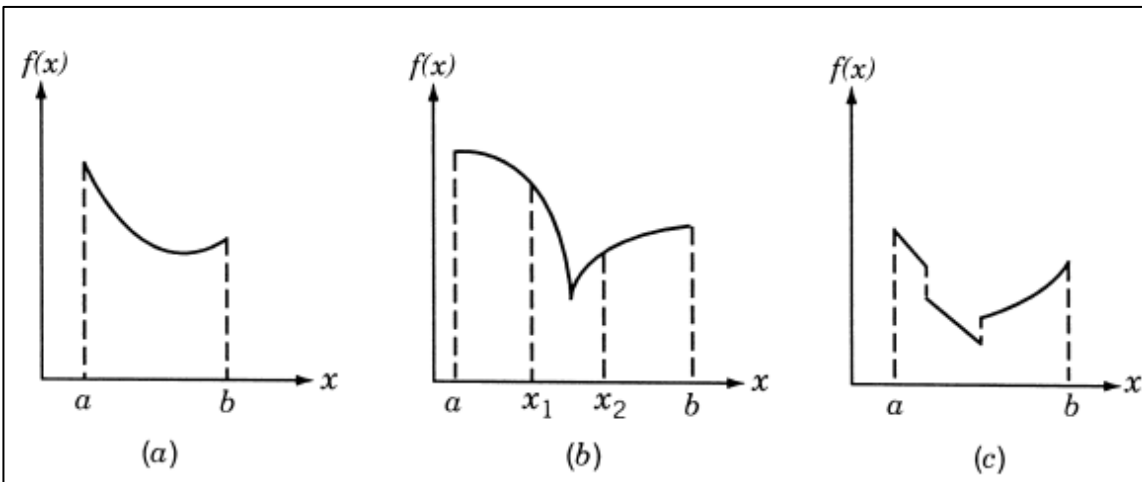
# DIRECT SEARCH METHOD

When the objective function to be optimized is either too complicated or is unavailable for direct computation in a closed mathematical form, optimum-seeking procedures using search strategies can be used. Sometimes, the functional relationship can be estimated from available data for the problem and the estimated relationship can then be used for the purpose of optimization. **However, such procedures may not always be feasible, and direct search procedures are used.**

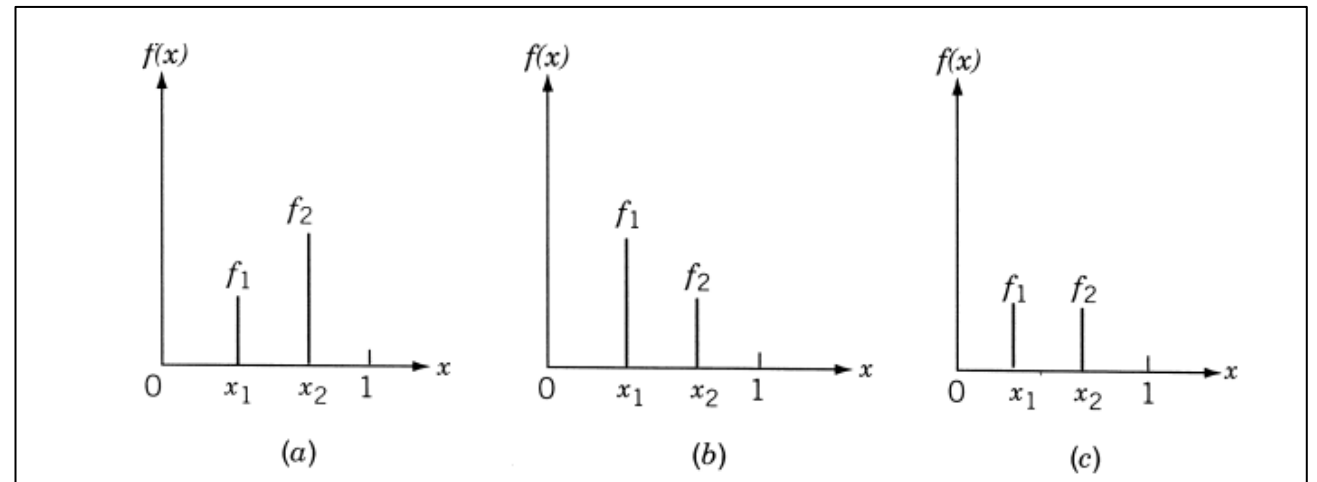
# Pengantar: Optimasi Tanpa Kendala Satu Peubah

- Fungsi yang akan dioptimumkan merupakan fungsi  $f(x)$  unimodal.
- Fungsi unimodal adalah fungsi yang hanya memiliki satu puncak (maksimum) atau lembah (minimum) dalam interval tertentu.
- Oleh karena itu syarat perlu dan cukup agar penyelesaian  $x = x^*$  menjadi optimal (maksimum global) adalah:

$$\frac{df}{dx} = 0 \text{ pada } x = x^*$$



**Figure 5.4** Unimodal function.



**Figure 5.5** Outcome of first two experiments: (a)  $f_1 < f_2$ ; (b)  $f_1 > f_2$ ; (c)  $f_1 = f_2$ .

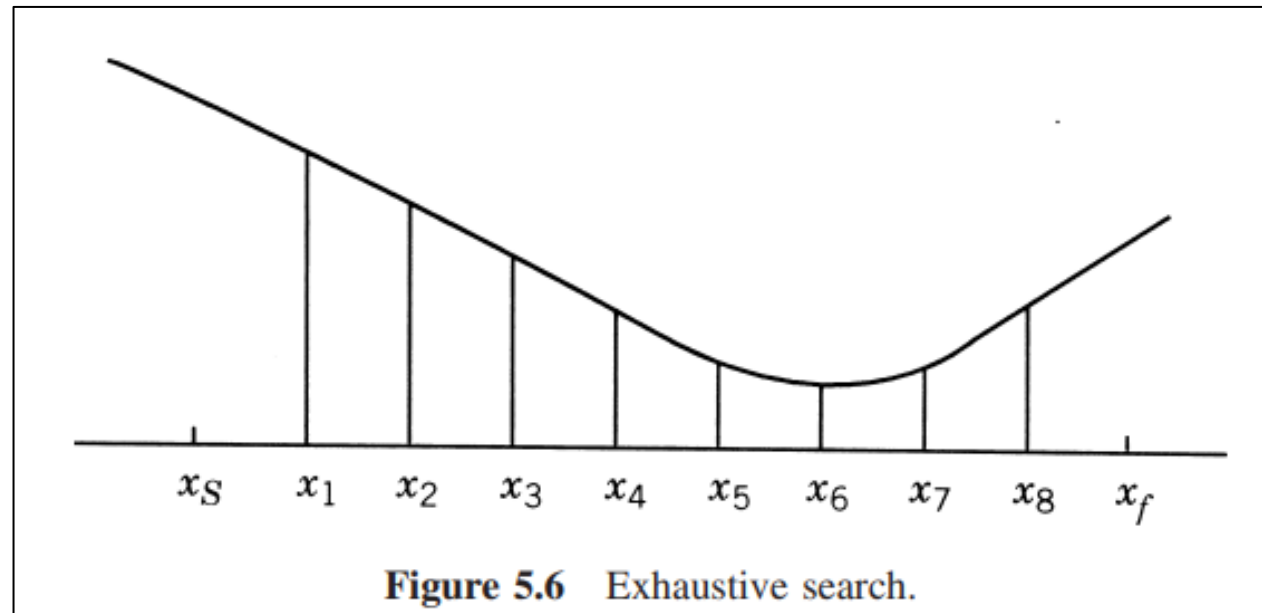
# SEARCH WITH FIXED STEP SIZE

- The most elementary approach for such a problem is to use a fixed step size and move from an initial guess point in a favorable direction (positive or negative).
- Algorithm:
  1. Start with an initial guess point, say,  $x_1$ .
  2. Find  $f_1 = f(x_1)$ .
  3. Assuming a step size  $s$ , find  $x_2 = x_1 + s$ .
  4. Find  $f_2 = f(x_2)$ .
  5. If  $f_2 < f_1$ , and if the problem is one of minimization, the assumption of unimodality indicates that the desired minimum cannot lie at  $x < x_1$ . Hence the search can be continued further along points  $x_3, x_4, \dots$  using the unimodality assumption while testing each pair of experiments. This procedure is continued until a point,  $x_i = x_1 + (i - 1)s$ , shows an increase in the function value.
  6. The search is terminated at  $x_i$ , and either  $x_{i-1}$  or  $x_i$  can be taken as the optimum point.
  7. Originally, if  $f_2 > f_1$ , the search should be carried in the reverse direction at points  $x_{-2}, x_{-3}, \dots$ , where  $x_{-j} = x_1 - (j - 1)s$ .
  8. If  $f_2 = f_1$ , the desired minimum lies in between  $x_1$  and  $x_2$ , and the minimum point can be taken as either  $x_1$  or  $x_2$ .
  9. If it happens that both  $f_2$  and  $f_{-2}$  are greater than  $f_1$ , it implies that the desired minimum will lie in the double interval  $x_{-2} < x < x_2$ .



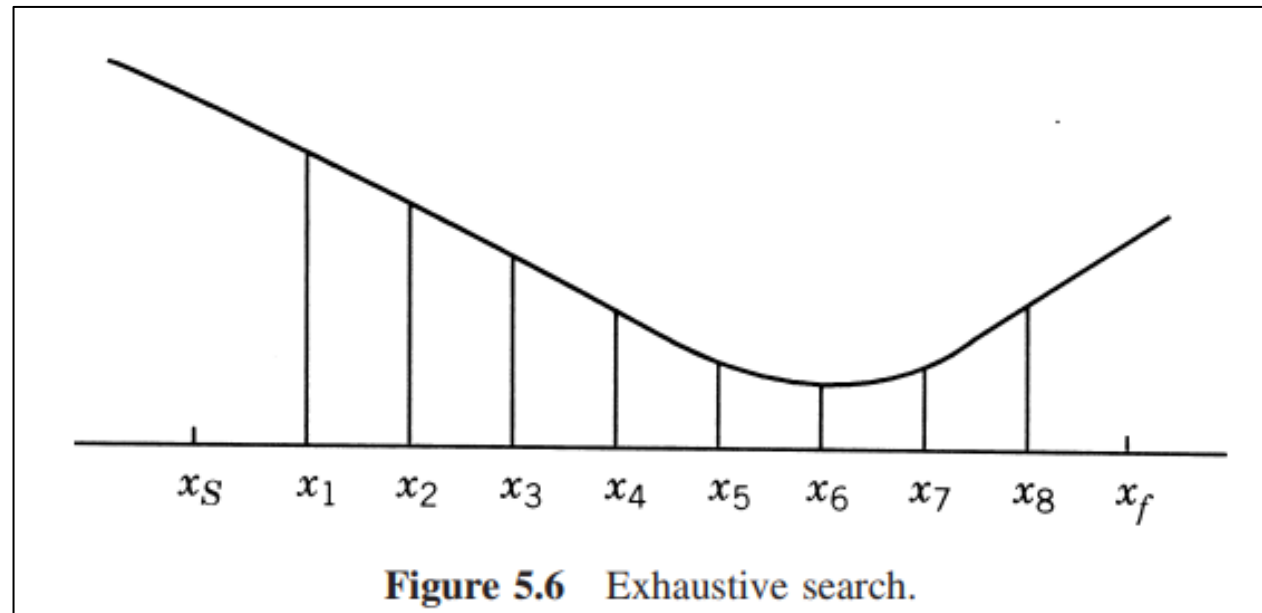
# EXHAUSTIVE SEARCH

The exhaustive search method can be used to solve problems where the interval in which the optimum is known to lie is finite. Let  $x_s$  and  $x_f$  denote, respectively, the starting and final points of the interval of uncertainty. The exhaustive search method consists of evaluating the objective function at a predetermined number of equally spaced points in the interval  $(x_s, x_f)$ , and reducing the interval of uncertainty using the assumption of unimodality.



# DICHOTOMOUS SEARCH

The exhaustive search method can be used to solve problems where the interval in which the optimum is known to lie is finite. Let  $x_s$  and  $x_f$  denote, respectively, the starting and final points of the interval of uncertainty. The exhaustive search method consists of evaluating the objective function at a predetermined number of equally spaced points in the interval  $(x_s, x_f)$ , and reducing the interval of uncertainty using the assumption of unimodality.



# FIBONACCI SEARCH METHOD

As stated earlier, the *Fibonacci method* can be used to find the minimum of a function of one variable even if the function is not continuous. This method, like many other elimination methods, has the following limitations:

1. The initial interval of uncertainty, in which the optimum lies, has to be known.
2. The function being optimized has to be unimodal in the initial interval of uncertainty.
3. The exact optimum cannot be located in this method. Only an interval known as the *final interval of uncertainty* will be known. The final interval of uncertainty can be made as small as desired by using more computations.
4. The number of function evaluations to be used in the search or the resolution required has to be specified beforehand.

# FIBONACCI SEARCH METHOD

- Pada metode ini kita harus menentukan lebih dulu dua bilangan sebagai interval dimana nilai minimum akan kita cari pada interval tersebut.
- Setiap iterasi pada metode ini memerlukan deret Fibonacci, dimana ditentukan sebagai berikut:

$$F_{v+1} = F_v + F_{v-1}$$

Dimana  $F_0$  dan  $F_1 = 1$  ;  $v = 1, 2, 3, \dots$ , dst

Deret Fibonacci adalah: 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89, ....

# FIBONACCI SEARCH METHOD

Algoritma:

- Menentukan interval awal yaitu  $a_1$  dan  $b_1$
- Memilih jumlah iterasi “n”
- Menentukan  $\lambda_1$  dan  $\mu_1$  dan  $k=1$  dan  $\varepsilon = 0.01$

nilai bawah  $\lambda_k = a_k + \frac{F_{n-k-1}}{F_{n-k+1}} (b_k - a_k) \quad k = 1, \dots, n-1.$

nilai atas  $\mu_k = a_k + \frac{F_{n-k}}{F_{n-k+1}} (b_k - a_k) \quad k = 1, \dots, n-1.$

- Menghitung  $F(\lambda_k)$  dan  $F(\mu_k)$

Jika  $F(\lambda_k) > F(\mu_k)$  maka interval yang baru  $[\lambda_k, b_k]$

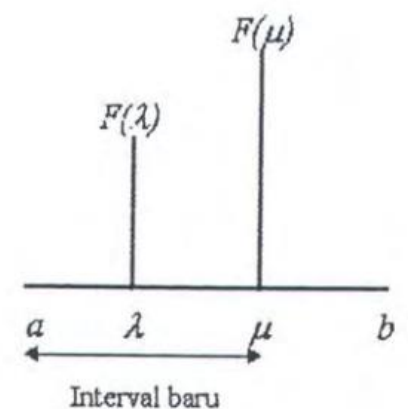
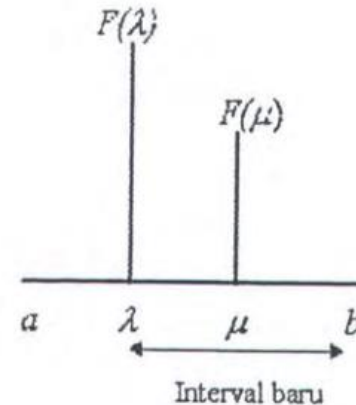
Jika  $F(\lambda_k) \leq F(\mu_k)$  maka interval yang baru  $[a_k, \mu_k]$ ,

Keterangan:

$a_k$  = batas bawah

$b_k$  = batas atas

n = jumlah evaluasi fungsi



# FIBONACCI SEARCH METHOD

**Result:** Estimate an approximate value of  $x^*$ , and determine  $f(x^*)$  and  $R$

```
1 Define  $f(x)$ ; // Define the objective function
2 Initialize  $x_l, x_r$  and  $n$ ;
3 Calculate  $L_0 \leftarrow x_r - x_l$ ; // Initial interval of uncertainty
4 Calculate  $L_i \leftarrow \frac{F_{n-2}}{F_n} L_0$ ;
5 Calculate  $R_1 \leftarrow \frac{L_i}{L_0}$ ; // Initial reduction ratio
6 for  $i \leftarrow 2$  to  $n$  by 1 do
7   if  $L_i \geq \frac{L_0}{2}$  // Compare whether  $x_1 < x_2$ 
8     then
9       Assign  $x_1 \leftarrow x_r - L_i$ ;
10      Assign  $x_2 \leftarrow x_l + L_i$ ;
11    else
12      Assign  $x_1 \leftarrow x_l + L_i$ ;
13      Assign  $x_2 \leftarrow x_r - L_i$ ;
14    end if
15    Calculate  $f_1 \leftarrow f(x_1)$ ;
16    Calculate  $f_2 \leftarrow f(x_2)$ ;
```

```
17   if  $f_1 < f_2$  then
18     Assign  $x_r \leftarrow x_2$ ;
19     Assign  $F_i \leftarrow \frac{F_{n-i}}{F_{n-(i-2)}} L_0$ ; // New interval of uncertainty
20     generated
21   else if  $f_1 > f_2$  then
22     Assign  $x_l \leftarrow x_1$ ;
23     Assign  $F_i \leftarrow \frac{F_{n-i}}{F_{n-(i-2)}} L_0$ ; // New interval of uncertainty
24     generated
25   else
26     Assign  $x_l \leftarrow x_1$ ;
27     Assign  $x_r \leftarrow x_2$ ;
28     Assign  $F_i \leftarrow \frac{F_{n-i}}{F_{n-(i-2)}} (x_r - x_l)$ ; // New interval of uncertainty
29     generated
30   end if
31   Assign  $L_0 \leftarrow x_r - x_l$ ;
32   Assign  $R \leftarrow \frac{L_i}{R_1}$ ; // New reduction ratio
33 end for
34 if  $f_1 \leq f_2$  then
35   Return  $x^* \leftarrow x_1$ ;
36   Return  $f(x^*) \leftarrow f(x_1)$ ;
37   Return  $R$ ;
38 else
39   Return  $x^* \leftarrow x_2$ ;
40   Return  $f(x^*) \leftarrow f(x_2)$ ;
41   Return  $R$ ;
42 end if
```

# FIBONACCI SEARCH METHOD

- Contoh:

Misalkan diketahui sebuah fungsi tak linear sebagai berikut:

$$f(x) = x^5 - 5x^3 - 20x + 5$$

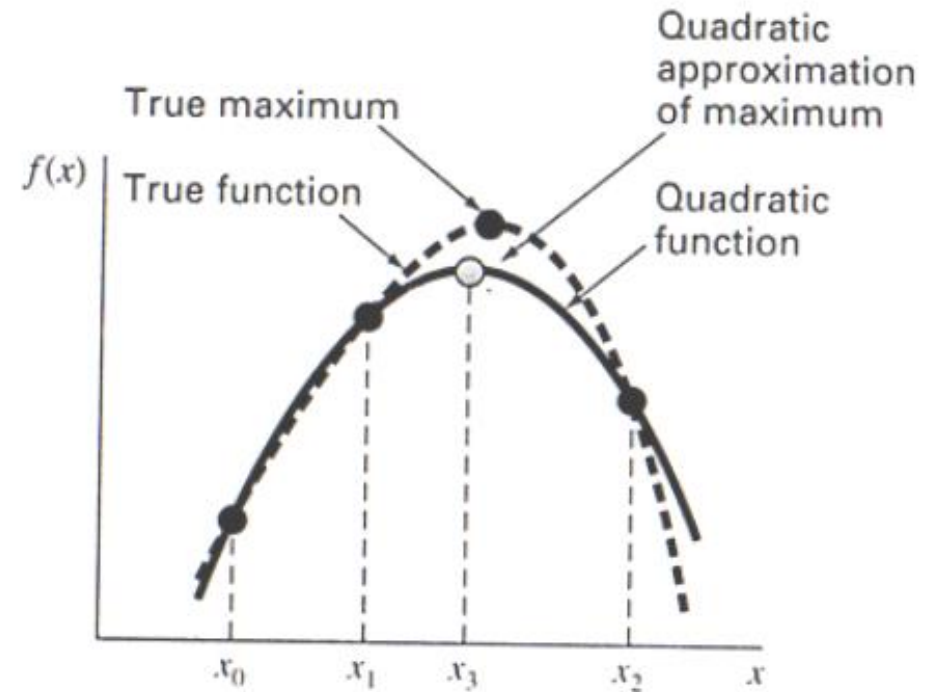
Fibonacci search method akan digunakan untuk mencari nilai  $x^*$  minimum. Misalkan  $n=25$  dan interval awal adalah  $[-2.5; 2.5]$ .

Maka sintaks R untuk menyelesaikan masalah tersebut adalah:

Terlampir pada file: Fibonacci.R

# QUADRATIC INTERPOLATION

- Metode interpolasi kuadratik melibatkan tiga titik tebakan awal dalam mencari nilai optimum.
- Penggunaan polinomial orde-dua menghasilkan pendekatan cukup baik terhadap bentuk  $f(x)$  di dekat titik optimum, sehingga Metode interpolasi kuadrat dapat digunakan untuk melakukan optimasi secara numerik





# QUADRATIC INTERPOLATION

- Algoritma

```
Result: Estimate an approximate value of  $x^*$  and determine  $f(x^*)$ 
1 Define  $f(x)$ ; // Define the objective function
2 Define  $f_1(x, y)$ ; // Define the first order forward divided
  difference  $f[x, y]$ 
3 Define  $f_2(x, y, z)$ ; // Define the second order forward divided
  difference  $f[x, y, z]$ 
4 Define  $nt(seq, n)$ ; // Function to find out the nearest value to
  a number,  $n$  from a list,  $seq$ 
5 Define  $ft(seq, n)$ ; // Function to find out the furthest value to
  a number,  $n$  from a list,  $seq$ 
6 Define  $maxim\_fv(seq)$ ; // Function to find out the element from
  the  $seq$  that has highest value of  $f(x)$ 
7 Initialize the starting experimental point  $x$ , the discrete step size  $s$ , the
  maximum step size  $m$  and the tolerance  $\epsilon$ ;
8 Calculate  $f_x \leftarrow f(x)$ ;
9 Calculate  $f_s \leftarrow f(x + s)$ ;
10 if  $f_x < f_s$  then
11   Calculate  $x_0 \leftarrow x - s$ ; // Set  $x_0$ 
12   Calculate  $x_1 \leftarrow x$ ; // Set  $x_1$ 
13   Calculate  $x_2 \leftarrow x + s$ ; // Set  $x_2$ 
14 else
15   Calculate  $x_0 \leftarrow x$ ; // Set  $x_0$ 
16   Calculate  $x_1 \leftarrow x + s$ ; // Set  $x_1$ 
17   Calculate  $x_2 \leftarrow x + 2s$ ; // Set  $x_2$ 
18 end if
19 while TRUE do
```

```
20   Calculate  $x_t \leftarrow \frac{f_2(x_0, x_1, x_2)(x_0, x_1) - f_1(x_0, x_1)}{2f_2(x_0, x_1, x_2)}$ ; // Calculate the
    approximate minimizer
21   Calculate  $x_n \leftarrow nt((x_0, x_1, x_2), x_t)$ ; // Picks the point from
     $(x_0, x_1, x_2)$  which is the nearest to  $x_t$ 
22   Calculate  $x_f \leftarrow ft((x_0, x_1, x_2), x_t)$ ; // Picks the point from
     $(x_0, x_1, x_2)$  which is the furthest to  $x_t$ 
23   if  $(f_2(x_0, x_1, x_2) > 0) \ \& \ (|x_t - x_n| > m)$  then
24     Remove  $x_f$  from  $(x_0, x_1, x_2)$ ;
25     Take a step of size  $m$  towards the direction of descent from the point
      with the lowest value;
26   else if  $f_2(x_0, x_1, x_2) < 0$  then
27     Remove  $x_n$  from  $(x_0, x_1, x_2)$ ;
28     Take a step of size  $m$  towards the direction of descent from the point
      with the lowest value;
29   else
30     if  $|x_t - x_n| < \epsilon$  // Check for terminating condition
31       then
32         Break;
33         Return  $x^* \leftarrow \frac{x_t + x_n}{2}$ ;
34         Return  $f(x^*)$ ;
35     else
36       Replace  $maxim\_fv((x_0, x_1, x_2))$  from  $(x_0, x_1, x_2)$  with  $x_t$ ;
37     end if
38   end if
39 end while
```

# QUADRATIC INTERPOLATION

- Contoh:

Misalkan diketahui sebuah fungsi tak linear sebagai berikut:

$$f(x) = x^4 - 2x^2 + \frac{1}{4}$$

Quadratic Interpolation akan digunakan untuk mencari nilai minimum  $x^*$  dari fungsi tersebut.

Maka sintaks R untuk menyelesaikan masalah tersebut adalah:

Terlampir pada file: Newton Quadratic.R

# NEWTON'S METHOD

- Metode Newton (atau seringkali disebut dengan metode Newton–Raphson) memerlukan fungsi tujuan tanpa kendala dalam interval yang menjadi perhatian dan mempunyai derivasi pertama maupun keduanya. Metode ini banyak pula dikembangkan untuk memecahkan permasalahan optimasi multi variabel. Metode Newton seringkali dipandang sebagai metode untuk mencari akar dari suatu fungsi.
- Pada metode ini pertama kita menentukan nilai awal ( $x_0$ ), selanjutnya dari nilai awal ini, sebuah garis singgung dapat diperluas dari titik ( $x_0, f(x_0)$ ). Titik dimana garis singgung memotong garis sumbu x biasanya menunjukkan sebuah taksiran perbaikan dari harga x. Metode Newton dapat diturunkan berdasarkan interpretasi geometri k atau sebuah metode alternatif yang didasarkan pada deret Taylor.

# NEWTON'S METHOD

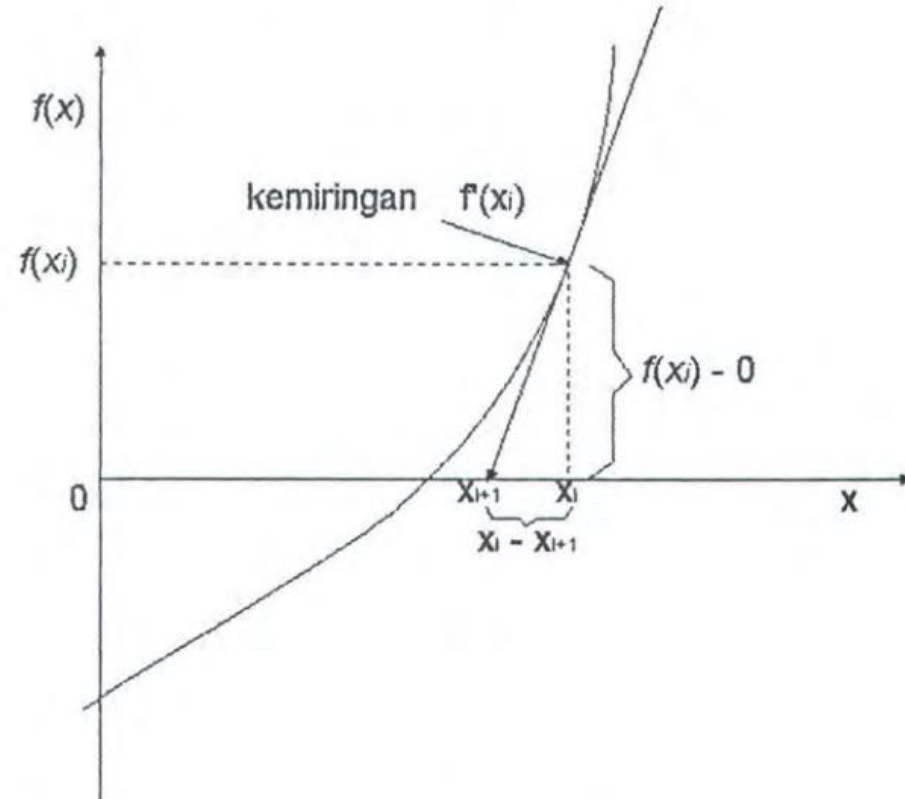
$$F(x) = f(x^{(k)}) + f'(x^{(k)})(x - x^{(k)}) + \frac{1}{2}f''(x^{(k)})(x - x^{(k)})^2$$

$$F'(x) = f'(x^{(k)}) + f''(x^{(k)})(x^* - x^{(k)}) = 0$$

$$x^* = x^{(k)} - \frac{f'(x^{(k)})}{f''(x^{(k)})}$$

$$x^{(k+1)} = x^{(k)} - \frac{f'(x^{(k)})}{f''(x^{(k)})}$$

$$|x^{(k+1)} - x^{(k)}| < \varepsilon$$



# NEWTON'S METHOD

- Algoritma:

```
Result: Estimate an approximate value of the root  $x^*$  and determine  $f(x^*)$ 
1 Define  $f(x)$ ; // Define the objective function
2 Define  $fprime(x) \leftarrow \frac{df}{dx}(x)$ ; // Define the first derivative of
  the objective function
3 Define  $fprime2(x) \leftarrow \frac{d^2f}{dx^2}(x)$ ; // Define the second derivative of
  the objective function
4 Initialize the initial starting point  $x_j$ , close to  $x^*$ , and the tolerance,  $\epsilon$ ;
5 while TRUE do
6   Assign  $f' \leftarrow fprime(x_j)$ ; // First derivative of  $f(x)$  at  $x_j$ 
7   Assign  $f'' \leftarrow fprime2(x_j)$ ; // Second derivative of  $f(x)$  at  $x_j$ 
8   Calculate  $x_{j+1} \leftarrow x_j - \frac{f'}{f''}$ ; // Update the next iterate
9   if  $|x_{j+1} - x_j| \leq \epsilon$  // Checking the termination condition
10  then
11    Break;
12    Return  $x^* \leftarrow x_{j+1}$ ;
13    Return  $f(x^*)$ ;
14  else
15    Assign  $x_j \leftarrow x_{j+1}$ ;
16  end if
17 end while
```

# NEWTON'S METHOD

- Contoh

Misalkan diketahui sebuah fungsi tak linear sebagai berikut:

$$f(x) = x^3 + x^2 - 1$$

Newton method akan digunakan untuk mencari nilai akar  $x^*$  dari fungsi tersebut.

Maka sintaks R untuk menyelesaikan masalah tersebut adalah:

Terlampir pada file: Newton Rapshon.R

# REFERENSI

- B. S. Everitt. 1987. Introduction to Optimization Methods and their Application in Statistics. Chapman and Hall, New York
- Rao, Singiresu S Engineering Optimization Theory and and Practice, 4th ed., pages 273:279
- <https://indrag49.github.io/Numerical-Optimization/solving-one-dimensional-optimization-problems.html>
- <https://rpubs.com/aaronsc32/newton-raphson-method>
- <https://github.com/brunasqz/NonlinearOpMethods/blob/versionII/R/quadraticinterpolation.R>