

3. Quadratic and Cubic Search for a Minimum

Background for Search Methods

An approach for finding the minimum of $f(x)$ in a given interval is to evaluate the function many times and search for a local minimum. To reduce the number of function evaluations it is important to have a good strategy for determining where $f(x)$ is to be evaluated. Two efficient bracketing methods are the [golden ratio](#) and [Fibonacci](#) searches. To use either bracketing method for finding the minimum of $f(x)$, a special condition must be met to ensure that there is a proper minimum in the given interval.

Definition (Unimodal Function) The function $f(x)$ is unimodal on $I = [a, b]$, if there exists a unique number $p \in I$ such that

$f(x)$ is decreasing on $[a, p]$,
and
 $f(x)$ is increasing on $[p, b]$.

Minimization Using Derivatives

Suppose that $f(x)$ is unimodal over $[a, b]$ and has a unique minimum at $x = p$. Also, assume that $f'(x)$ is defined at all points in (a, b) . Let the starting value p_0 lie in (a, b) . If $f'(p_0) < 0$, then the minimum point p lies to the right of p_0 . If $f'(p_0) > 0$, then the minimum point p lies to the left of p_0 .

Background for Bracketing the Minimum

Our first task is to obtain three test values,

- (1) $p_0, p_1 = p_0 + h$, and $p_2 = p_0 + 2h$,
so that
(2) $f(p_0) > f(p_1)$, and $f(p_1) < f(p_2)$.

Suppose that $f'(p_0) < 0$; then $p_0 < p$ and the step size h should be chosen positive. It is an easy task to find a value of h so that the three points in (1) satisfy (2). Start with $h = 1$ in formula (1) (provided that $a + 1 < b$); if not, take $h = \frac{1}{2}$, and so on.

Case (i) If (2) is satisfied we are done.

Case (ii) If $f(p_0) > f(p_1)$, and $f(p_1) > f(p_2)$, then $p_2 < p$.

We need to check points that lie farther to the right. Double the step size and repeat the process.

Case (iii) If $f(p_0) < f(p_1)$, we have jumped over p and h is too large.

We need to check values closer to p_0 . Reduce the step size by a factor of $\frac{1}{2}$ and repeat the process.

When $f'(p_0) > 0$, the step size h should be chosen negative and then cases similar to (i), (ii), and (iii) can be used.

Quadratic Approximation to Find p

Finally, we have three points (1) that satisfy (2). We will use quadratic interpolation to find p_{\min} , which is an approximation to p . The Lagrange polynomial based on the nodes in (1) is

$$(3) \quad Q(x) = Y_0 \frac{(x-p_1)(x-p_2)}{2h^2} - Y_1 \frac{(x-p_0)(x-p_2)}{h^2} + Y_2 \frac{(x-p_0)(x-p_1)}{2h^2},$$

where $Y_i = f(p_i)$ for $i = 1, 2, 3$.

The derivative of $Q(x)$ is

$$(4) \quad Q'(x) = Y_0 \frac{(2x-p_1-p_2)}{2h^2} - Y_1 \frac{(2x-p_0-p_2)}{h^2} + Y_2 \frac{(2x-p_0-p_1)}{2h^2}.$$

Solving $Q'(x) = 0$ in the form $Q(p_0 + h_{\min}) = 0$ yields

$$(5) \quad 0 = Y_0 \frac{(2(p_0 + h_{\min}) - p_1 - p_2)}{2h^2} - Y_1 \frac{(4(p_0 + h_{\min}) - 2p_0 - 2p_2)}{2h^2} + Y_2 \frac{(2(p_0 + h_{\min}) - p_0 - p_1)}{2h^2}.$$

Multiply each term in (5) by $2h^2$ and collect terms involving h_{\min} :

$$0 = h_{\min}(2Y_0 - 4Y_1 + 2Y_2) + Y_0(2p_0 - p_1 - p_2) - Y_1(4p_0 - 2p_0 - 2p_2) + Y_2(2p_0 - p_0 - p_1)$$

$$0 = h_{\min}(2Y_0 - 4Y_1 + 2Y_2) + Y_0(2p_0 - p_1 - p_2) - Y_1(2p_0 - 2p_2) + Y_2(p_0 - p_1)$$

$$0 = h_{\min}(2Y_0 - 4Y_1 + 2Y_2) - 3hY_0 + 4hY_1 - hY_2$$

$$0 = -h_{\min}2(2Y_1 - Y_0 - Y_2) + h(4Y_1 - 3Y_0 - Y_2)$$

This last quantity is easily solved for h_{\min} :

$$h_{\min} = \frac{h(4Y_1 - 3Y_0 - Y_2)}{2(2Y_1 - Y_0 - Y_2)}.$$

The value $p_{\min} = p_0 + h_{\min}$ is a better approximation to p than p_0 . Hence we can replace p_0 with p_{\min} and repeat the two processes outlined above to determine a new h and a new h_{\min} . Continue the iteration until the desired accuracy is achieved. In this algorithm the derivative of the objective function $f(x)$ was used implicitly in (4) to locate the minimum of the interpolatory quadratic. The reader should note that the subroutine makes no explicit use of the derivative.

Cubic Approximation to Find p

We now consider an approach that utilizes functional evaluations of both $f(x)$ and $f'(x)$. An alternative approach that uses both functional and derivative evaluations explicitly is to find the minimum of a third-degree polynomial that interpolates the objective function $f(x)$ at two points. Assume that $f(x)$ is unimodal and differentiable on $[a, b]$, and has a unique minimum at $x = p$. Let $p_0 = a$. Any good step size h can be used to start the iteration. The [Mean Value Theorem](#) could be used to obtain $f'(c) = \frac{f(a+h) - f(a)}{h}$ and if $a+h$ was just to the right of the minimum, then the slope $f'(a)$ might be twice $f'(c)$ which would mean that $\frac{1}{2}f'(a) = \frac{f(a+h) - f(a)}{h}$ we do not know how much further to the right $c = a+h$ lies, so we can imagine that $f(a+h)$ is close to $f(b)$ and estimate h with the formula:

$$h = \frac{2(f[b] - f[a])}{f'[a]}.$$

Thus $p_1 = p_0 + h$. The cubic approximating polynomial $P[x]$ is expanded in a Taylor series about $x = p_2$ (which is the abscissa of the minimum). At the minimum we have $P'[p_2] = 0$, and we write $P[x]$ in the form:

$$(6) \quad P[x] = \frac{\alpha}{h^3}(x-p_2)^3 + \frac{\beta}{h^2}(x-p_2)^2 + f[p_2],$$

and

$$(7) \quad P'[x] = \frac{3\alpha}{h^3} (x - p_z)^2 + \frac{2\beta}{h^2} (x - p_z).$$

The introduction of $h = p_1 - p_0$ in the denominators of (6) and (7) will make further calculations less tiresome. It is required that $P[p_0] = f[p_0]$, $P[p_1] = f[p_1]$, $P'[p_0] = f'[p_0]$, and $P'[p_1] = f'[p_1]$. To find p_z we define:

$$(8) \quad p_z = p_0 + \gamma h,$$

and we must go through several intermediate calculations before we end up with γ .

Use use (6) to obtain

$$f[p_1] - f[p_0] = \frac{\alpha (p_1 - p_z)^3 - \alpha (p_0 - p_z)^3}{h^3} + \frac{\beta (p_1 - p_z)^2 - \beta (p_0 - p_z)^2}{h^2}$$

Then use (8) to get

$$f[p_1] - f[p_0] = \alpha \gamma^3 + \frac{\alpha (-h\gamma - p_0 + p_1)^3}{h^3} + \frac{\beta (-h\gamma - p_0 + p_1)^2}{h^2} - \beta \gamma^2$$

Then substitute $h = p_1 - p_0$ and we have

$$(9) \quad F = f[p_1] - f[p_0] = \alpha (3\gamma^3 - 3\gamma + 1) + \beta (1 - 2\gamma)$$

Use use (7) to obtain

$$f'[p_1] - f'[p_0] = \frac{3\alpha (p_1 - p_z)^2 - 3\alpha (p_0 - p_z)^2}{h^3} + \frac{2\beta (p_1 - p_z) - 2\beta (p_0 - p_z)}{h^2}$$

$$h (f'[p_1] - f'[p_0]) = \frac{3\alpha (p_1 - p_z)^2 - 3\alpha (p_0 - p_z)^2}{h^2} + \frac{2\beta (p_1 - p_z) - 2\beta (p_0 - p_z)}{h}$$

Then use (8) to get

$$h (f'[p_1] - f'[p_0]) = \frac{3\alpha (-h\gamma - p_0 + p_1)^2 - 3h^2\alpha\gamma^2}{h^2} + \frac{2\beta (-h\gamma - p_0 + p_1) + 2h\beta\gamma}{h}$$

Then substitute $h = p_1 - p_0$ and we have

$$(10) \quad G = h (f'[p_1] - f'[p_0]) = 3\alpha (1 - 2\gamma) + 2\beta$$

Finally, use (7) and write

$$h f'[p_0] = h \left(\frac{3\alpha (p_0 - p_z)^2}{h^3} + \frac{2\beta (p_0 - p_z)}{h^2} \right)$$

Then use (8) to get

$$(11) \quad h f'[p_0] = 3\alpha \gamma^2 - 2\beta \gamma$$

Now we will use the three nonlinear equations (9), (10), (11) listed below in (12). The order of determining the variables will be F, G, α, γ (the variable β will be eliminated).

$$F = \alpha (3\gamma^3 - 3\gamma + 1) + \beta (1 - 2\gamma)$$

$$(12) \quad \begin{aligned} G &= 3\alpha(1-2\gamma) + 2\beta \\ hf'[p_0] &= 3\alpha\gamma^2 - 2\beta\gamma \end{aligned}$$

First, we will find α which is accomplished by combining the equation in (12) as follows:

$$G - 2(F - hf'[p_0]) = 3\alpha(1-2\gamma) + 2\beta - 2(\alpha(3\gamma^2 - 3\gamma + 1) + \beta(1-2\gamma) - (3\alpha\gamma^2 - 2\beta\gamma))$$

Straightforward simplification yields $G - 2(F - hf'[p_0]) = \alpha$, therefore α is given by

$$(13) \quad \alpha = G - 2(F - hf'[p_0]).$$

Second, we will eliminate β by combining the equation in (12) as follows, multiply the first equation by γ and add it to the third equation

$$\begin{aligned} G\gamma &= 3\alpha\gamma - 6\alpha\gamma^2 + 2\beta\gamma \\ hf'[p_0] &= 3\alpha\gamma^2 - 2\beta\gamma \\ G\gamma + hf'[p_0] &= 3\alpha\gamma - 3\alpha\gamma^2 \end{aligned}$$

which can be rearranged in the form

$$G\gamma - 3\alpha\gamma + 3\alpha\gamma^2 + hf'[p_0] = 0$$

Now the quadratic equation can be used to solve for γ

$$\gamma = \frac{(G - 3\alpha) - \sqrt{(G - 3\alpha)^2 - 12hf'[p_0]}}{-6\alpha}$$

It will take a bit of effort to simplify this equation into its computationally preferred form.

$$\begin{aligned} \frac{(G - 3\alpha) - \sqrt{(G - 3\alpha)^2 - 12hf'[p_0]}}{-6\alpha} &= \frac{(G - 3\alpha) + \sqrt{(G - 3\alpha)^2 - 12hf'[p_0]}}{(G - 3\alpha) + \sqrt{(G - 3\alpha)^2 - 12hf'[p_0]}} \frac{(G - 3\alpha) - \sqrt{(G - 3\alpha)^2 - 12hf'[p_0]}}{-6\alpha} \\ \frac{(G - 3\alpha) - \sqrt{(G - 3\alpha)^2 - 12hf'[p_0]}}{-6\alpha} &= \frac{12hf'[p_0]}{((G - 3\alpha) + \sqrt{(G - 3\alpha)^2 - 12hf'[p_0]})(-6\alpha)} \\ \frac{(G - 3\alpha) - \sqrt{(G - 3\alpha)^2 - 12hf'[p_0]}}{-6\alpha} &= \frac{-2hf'[p_0]}{(G - 3\alpha) + \sqrt{(G - 3\alpha)^2 - 12hf'[p_0]}} \end{aligned}$$

Hence,

$$(14) \quad \gamma = \frac{-2hf'[p_0]}{(G - 3\alpha) + \sqrt{(G - 3\alpha)^2 - 12hf'[p_0]}}$$

Therefore, the value of p_2 is found by substituting the calculated value of γ in (14) into the formula $p_2 = p_0 + \gamma h$. To continue the iteration process, let $h = p_2 - p_1$ and replace p_0 and p_1 with p_1 and p_2 , respectively, in formulas (12), (13), and (14). The algorithm outlined above is **not** a bracketing method. Thus determining stopping criteria becomes more problematic. One technique would be to require that $|f'(p_n)| < \epsilon$, since $f'(p) = 0$.

Example 1. Find the minimum of the function $f[x] = x^2 - \sin[x]$ on the interval $[0, 1]$ using the quadratic search method.

Solution 1.

Algorithm (Cubic Search for a Minimum). To numerically approximate the minimum of $f(x)$ on the interval $[a, b]$ by using a "cubic interpolative" search. Proceed with the method only if $f(x)$ is a unimodal function on the interval $[a, b]$.

Start with $p_0 = a$, $p_1 = b$, and $h_0 = \frac{2(f[p_1] - f[p_0])}{f'[p_0]}$ and use the following sequence steps in the iteration:

$$F = f[p_{k+1}] - f[p_k]$$

$$G = h_k (f'[p_{k+1}] - f'[p_k])$$

$$\alpha = G - 2(F - h_k f'[p_k])$$

$$\gamma = \frac{-2 h_k f'[p_k]}{(G - 3\alpha) + \sqrt{(G - 3\alpha)^2 - 12 h_k \alpha f'[p_k]}}$$

$$p_{k+2} = p_k + \gamma h_k$$

$$h_{k+1} = p_{k+2} - p_{k+1}$$

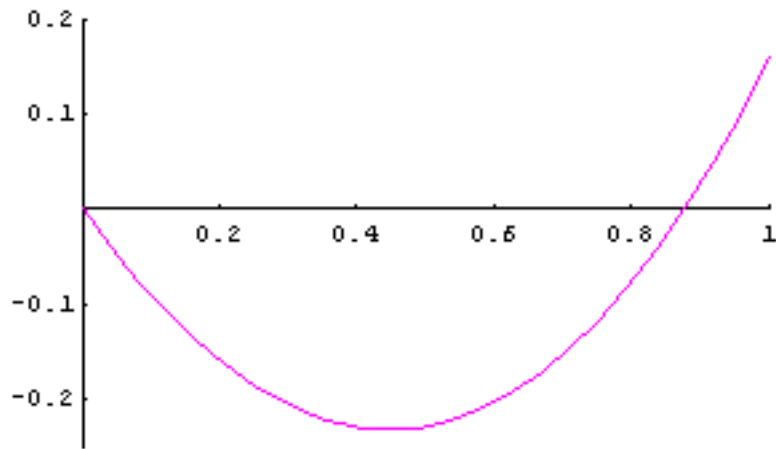
for $k = 0, 1, 2, \dots, n$.

Example 2. Find the minimum of $f(x) = \frac{1}{2} + x^5 - \frac{4}{5}x$ on the interval $[0, 1]$ using the quadratic search method.

Solution 2.

Example 1. Find the minimum of the function $f[x] = x^2 - \sin[x]$ on the interval $[0, 1]$ using the quadratic search method.

Solution 1.



$$y = f[x] = x^2 - \sin[x]$$

Set $p_0 = a$, and $p_1 = b$, and $h = \frac{b-a}{2}$.

$$f[x] = x^2 - \sin[x]$$

$$h = 0.5$$

$$p_0 = 0.$$

$$p_1 = 0.5$$

$$p_2 = 1.$$

$$Y_0 = f[p_0] = f[0.00000000] = 0.$$

$$Y_1 = f[p_1] = f[0.50000000] = -0.229426$$

$$Y_2 = f[p_2] = f[1.00000000] = 0.158529$$

We must decide whether to use p_0 , p_1 , p_2 or change the step size to $\frac{h}{2}$ or $2h$ and recompute some of these values.

```

f[x] = x2 - Sin[x]
h = 0.5
p0 = 0.
p1 = 0.5
p2 = 1.

Y0 = f[p0] = f[ 0.00000000 ] = 0.
Y1 = f[p1] = f[ 0.50000000 ] = -0.229426
Y2 = f[p2] = f[ 1.00000000 ] = 0.158529

```

The step size has not changed and we now compute $p_{\min} = p_0 + H_{\min}$.

```

f[x] = x2 - Sin[x]
h = 0.5
p0 = 0.
p1 = 0.5
p2 = 1.

Y0 = f[p0] = f[ 0.00000000 ] = 0.
Y1 = f[p1] = f[ 0.50000000 ] = -0.229426
Y2 = f[p2] = f[ 1.00000000 ] = 0.158529

pmin =  $\frac{h (4 Y_1 - 3 Y_0 - Y_2)}{2 (2 Y_1 - Y_0 - Y_2)}$ 
pmin = h (4 Y1 - 3 Y0 - Y2) / 2 (2 Y1 - Y0 - Y2)
pmin = 0.5(-1.07623) / (2(-0.61738))
pmin = 0.5(-1.07623) / (-1.23476)

pmin = p0 + Hmin
pmin = 0. + (0.435806)
pmin = 0.435806
Ymin = f[pmin] = f[ 0.43580575 ] = -0.232214

```

We use the "latest three values" and change the step size and continue the iteration with the following three values used in the interpolating quadratic.

```

f[x] = x2 - Sin[x]
h = 0.0641943
p0 = 0.435806
p1 = 0.5
p2 = 0.564194

```

```

Y0 = f[p0] = f[ 0.43580575 ] = -0.2322143232
Y1 = f[p1] = f[ 0.50000000 ] = -0.2294255386
Y2 = f[p2] = f[ 0.56419425 ] = -0.2164199625

```

The last three values will be labeled p_3, p_4, p_5 by the subroutine. The list of computations obtained by using the Quadraticsearch subroutine are:

```

p0 = 0.00000000, f[ 0.00000000 ] = -0.2322143232
p1 = 0.50000000, f[ 0.50000000 ] = -0.2294255386
p2 = 1.00000000, f[ 1.00000000 ] = 0.1585290152
p3 = 0.43580575, f[ 0.43580575 ] = -0.2322143232
p4 = 0.50000000, f[ 0.50000000 ] = -0.2294255386
p5 = 0.56419425, f[ 0.56419425 ] = -0.2164199625
q5,0 = 0.46790287, f[ 0.46790287 ] = -0.2320824604
q5,1 = 0.45185431, f[ 0.45185431 ] = -0.2324621759
p6 = 0.45016827, f[ 0.45016827 ] = -0.2324655749
p7 = 0.45185431, f[ 0.45185431 ] = -0.2324621759
p8 = 0.45354035, f[ 0.45354035 ] = -0.2324518503
q8,0 = 0.45101129, f[ 0.45101129 ] = -0.2324647410
q8,1 = 0.45058978, f[ 0.45058978 ] = -0.2324653743
q8,2 = 0.45037903, f[ 0.45037903 ] = -0.2324655287
q8,3 = 0.45027365, f[ 0.45027365 ] = -0.2324655653
q8,4 = 0.45022096, f[ 0.45022096 ] = -0.2324655735
q8,5 = 0.45019462, f[ 0.45019462 ] = -0.2324655750
p9 = 0.45018361, f[ 0.45018361 ] = -0.2324655752

```

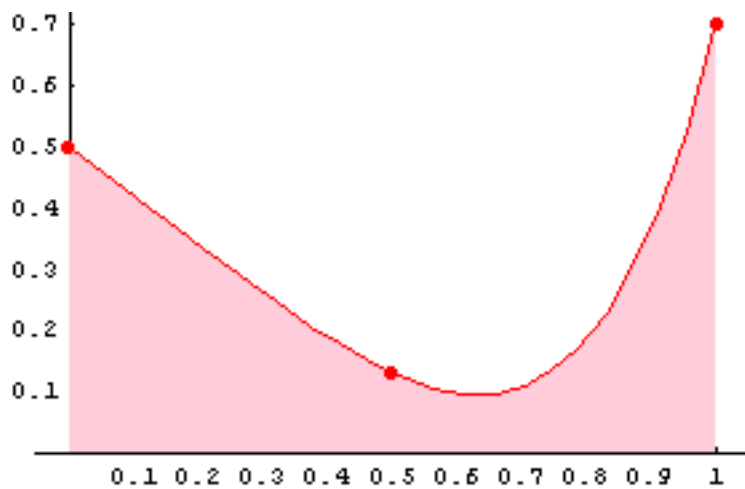
Let us compare these answers with *Mathematica's* subroutine FindMinimum.


```
f[x] = x2 - Sin[x]  
{-0.232466, {x → 0.450184}}
```

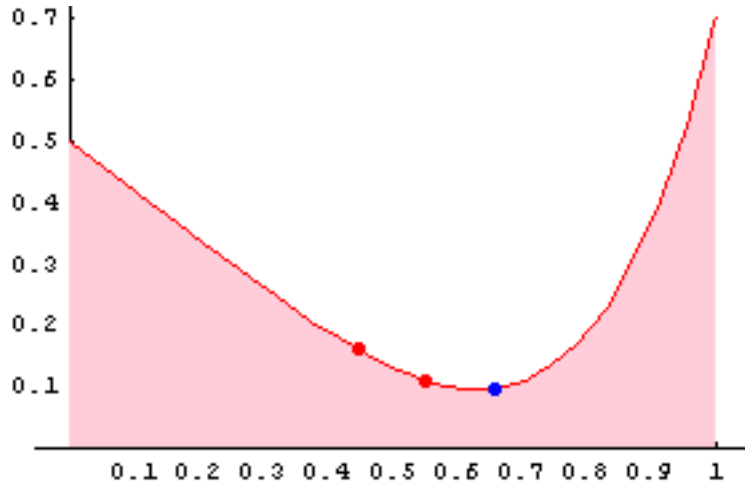
```
p = 0.45018361  
f[p] = -0.23246558
```

Example 2. Find the minimum of $f(x) = \frac{1}{2} + x^5 - \frac{4}{5}x$ on the interval $[0, 1]$ using the quadratic search method.

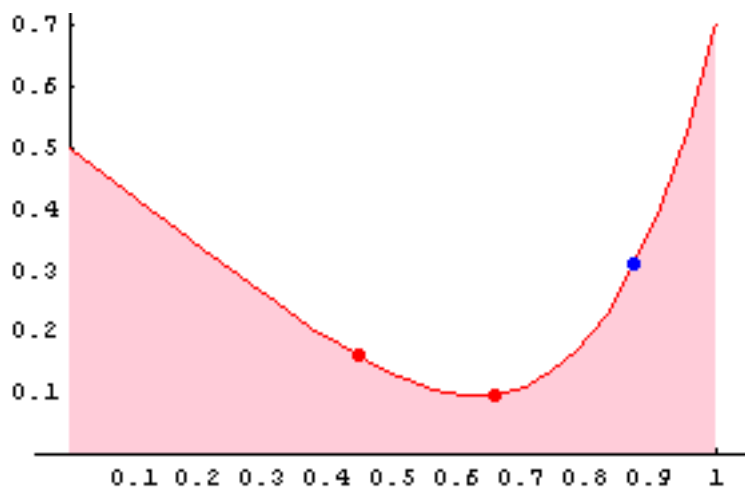
Solution 2.



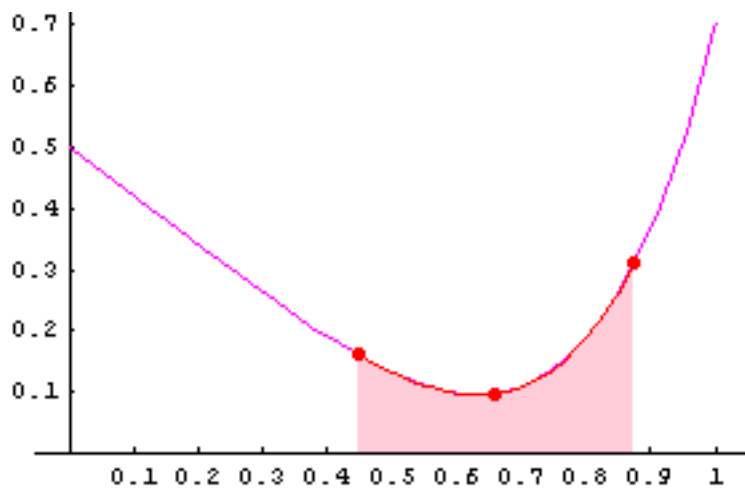
$$\begin{aligned} p_0 &= 0.00000000, & f[0.00000000] &= 0.5000000000 \\ p_1 &= 0.50000000, & f[0.50000000] &= 0.1312500000 \\ p_2 &= 1.00000000, & f[1.00000000] &= 0.7000000000 \end{aligned}$$



$$q_{5,0}q_{5,0} = 0.66000000, \quad f[0.66000000] = 0.0972332576$$



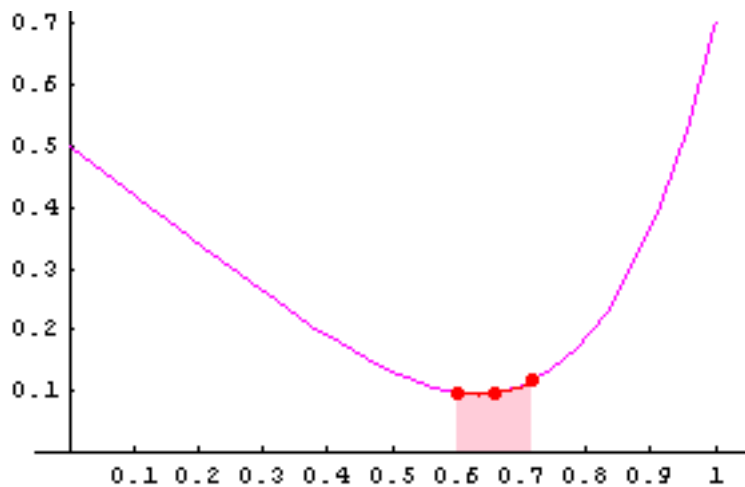
$$q_5, lq_5, 1 = 0.87333333, \quad f[0.87333333] = 0.3093759954$$



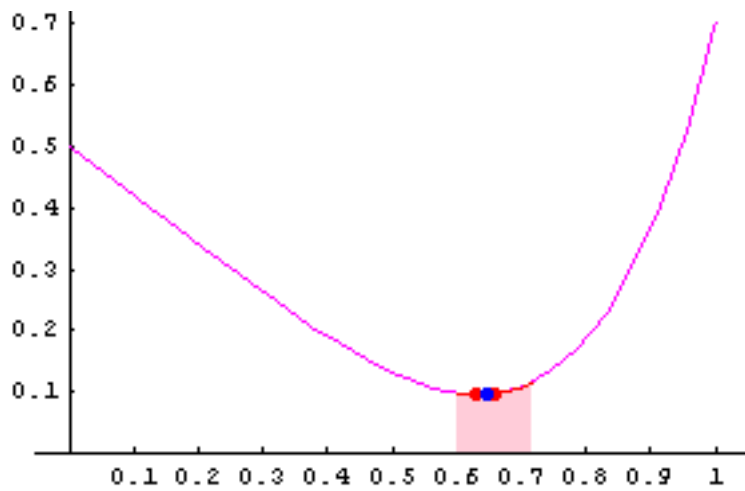
$$p_3 = 0.44666667, \quad f[0.44666667] = 0.1604460919$$

$$p_4 = 0.66000000, \quad f[0.66000000] = 0.0972332576$$

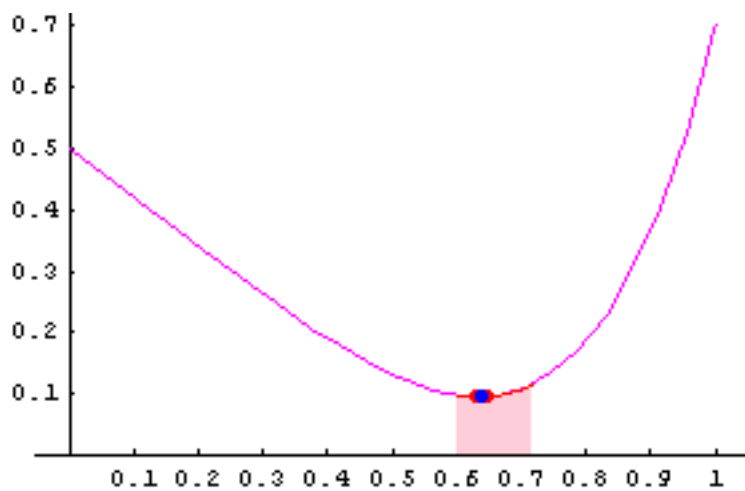
$$p_5 = 0.87333333, \quad f[0.87333333] = 0.3093759954$$



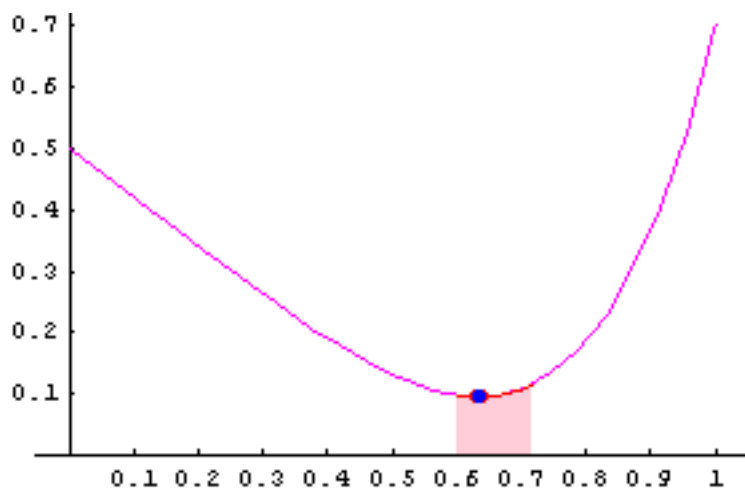
$p_6 = 0.60230784, \quad f[0.60230784] = 0.0974207564$
 $p_7 = 0.66000000, \quad f[0.66000000] = 0.0972332576$
 $p_8 = 0.71769216, \quad f[0.71769216] = 0.1162568207$



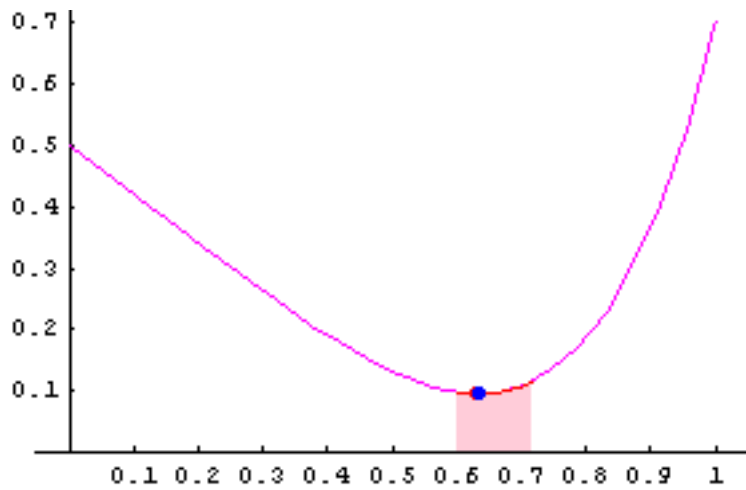
$q_{11,0} = 0.64585850, \quad f[0.64585850] = 0.0956926487$



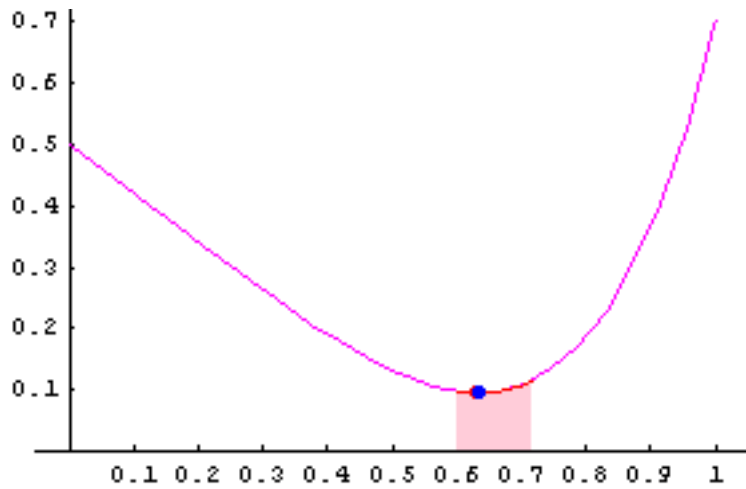
$q_{11,1} = 0.63878775, \quad f[0.63878775] = 0.0953309183$



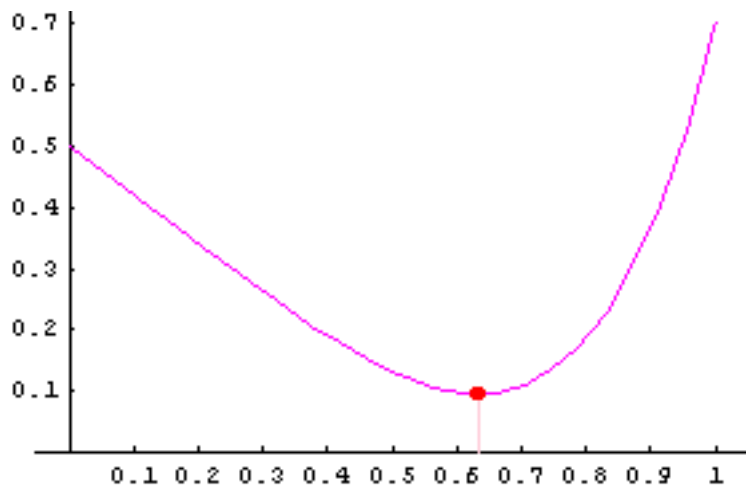
$$q_{11,2} = 0.63525237, \quad f[0.63525237] = 0.0952483362$$



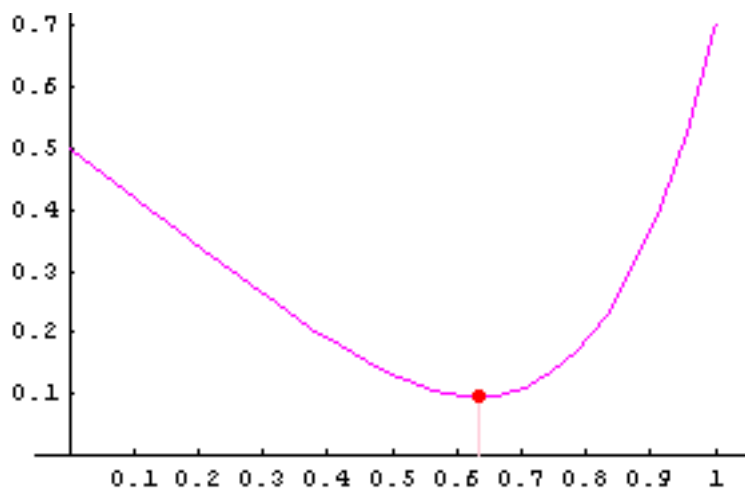
$$q_{11,3} = 0.63348468, \quad f[0.63348468] = 0.0952311433$$



$$q_{11,4} = 0.63260084, \quad f[0.63260084] = 0.0952285129$$



$p_9 = 0.63171699, \quad f[0.63171699] = 0.0952298377$
 $p_{10} = 0.63260084, \quad f[0.63260084] = 0.0952285129$
 $p_{11} = 0.63348468, \quad f[0.63348468] = 0.0952311433$



$p_{12} = 0.63245496, \quad f[0.63245496] = 0.0952284595$
 $f[x] = \frac{1}{2} - \frac{4x}{5} + x^5$