Newton-Raphson Method

Mechanical Engineering Majors

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Newton-Raphson Method

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Newton-Raphson Method

\[ x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)} \]

\([x_i, f(x_i)]\)

**Figure 1** Geometrical illustration of the Newton-Raphson method.
Derivation

\[ \tan(\alpha) = \frac{AB}{AC} \]

\[ f'(x_i) = \frac{f(x_i)}{x_i - x_{i+1}} \]

\[ x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)} \]

Figure 2 Derivation of the Newton-Raphson method.

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Algorithm for Newton-Raphson Method
Step 1

Evaluate $f'(x)$ symbolically.
Step 2

Use an initial guess of the root, $x_i$, to estimate the new value of the root, $x_{i+1}$, as

$$x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)}$$
Step 3

Find the absolute relative approximate error $|\varepsilon_a|$ as

$$|\varepsilon_a| = \frac{x_{i+1} - x_i}{x_i} \times 100$$
Step 4

Compare the absolute relative approximate error with the pre-specified relative error tolerance \( \epsilon_s \).

| Is \( |\epsilon_a| > \epsilon_s \) ? |
|-----------------------------------|
| Yes                              |
| Go to Step 2 using new estimate of the root. |
| No                               |
| Stop the algorithm               |

Also, check if the number of iterations has exceeded the maximum number of iterations allowed. If so, one needs to terminate the algorithm and notify the user.
Example 1

A trunnion has to be cooled before it is shrink fitted into a steel hub.

The equation that gives the temperature $x$ to which the trunnion has to be cooled to obtain the desired contraction is given by the following equation.

$$f(x) = -0.50598 \times 10^{-10} x^3 + 0.38292 \times 10^{-7} x^2 + 0.74363 \times 10^{-4} x + 0.88318 \times 10^{-2} = 0$$
Use the Newton-Raphson method of finding roots of equations

a) To find the temperature $x$ to which the trunnion has to be cooled. Conduct three iterations to estimate the root of the above equation.

b) Find the absolute relative approximate error at the end of each iteration, and

c) the number of significant digits at least correct at the end of each iteration.
Example 1 Cont.

\[ f(x) = -0.50598 \times 10^{-10} x^3 - 0.38292 \times 10^{-7} x^2 + 0.74363 \times 10^{-4} x + 0.88318 \times 10^{-2} = 0 \]

**Figure 4** Graph of the function \( f(x) \).
Example 1 Cont.

\[ f(x) = -0.50598 \times 10^{-10} x^3 - 0.38292 \times 10^{-7} x^2 + 0.74363 \times 10^{-4} x + 0.88318 \times 10^{-2} = 0 \]

\[ f'(x) = -1.51794 \times 10^{-10} x^2 + 0.76584 \times 10^{-7} x + 0.74363 \times 10^{-4} \]

**Iteration 1**
The estimate of the root is

**Initial guess:** \[ x_0 = -100 \]

\[ f(-100) = -0.50598 \times 10^{-10} (-100)^3 - 0.38292 \times 10^{-7} (-100)^2 + 0.74363 \times 10^{-4} (-100) + 0.88318 \times 10^{-2} \]

\[ = 1.8290 \times 10^{-3} \]

\[ f''(-100) = -1.51794 \times 10^{-10} (-100)^2 + 0.76582 \times 10^{-7} (-100) + 0.74363 \times 10^{-4} \]

\[ = 6.5187 \times 10^{-5} \]
Example 1 Cont.

The iteration formula is:

\[ x_1 = x_0 - \frac{f(x_0)}{f'(x_0)} \]

Using the given values:

\[ x_1 = 100 - \frac{1.8290 \times 10^{-3}}{6.5187 \times 10^{-5}} \]

\[ x_1 = -128.06 \]

The absolute relative approximate error is:

\[ |\varepsilon_a| = \left| \frac{x_1 - x_0}{x_1} \right| \times 100 \]

\[ = \left| \frac{-128.0582 - (-100)}{-128.0582} \right| \times 100 \]

\[ = 21.910\% \]

The number of significant digits at least correct is 0.

**Figure 5** Graph of estimated root after Iteration 1.
Example 1 Cont.

**Iteration 2**
The estimate of the root is

\[ f(-128.06) = -0.50598 \times 10^{-10}(-128.06)^3 - 0.38292 \times 10^{-7}(-128.06)^2 + 0.74363 \times 10^{-4}(-128.06) \]

\[ + 0.88318 \times 10^{-2} \]

\[ = 4.3214 \times 10^{-5} \]

\[ f'(-128.06) = -1.5179 \times 10^{-10}(-128.06)^2 + 0.76584 \times 10^{-7}(-128.06) + 0.74363 \times 10^{-4} \]

\[ = 6.2067 \times 10^{-5} \]

\[ x_2 = x_1 - \frac{f(x_1)}{f'(x_1)} \]

\[ x_2 = -128.06 - \frac{4.3214 \times 10^{-5}}{6.2067 \times 10^{-5}} \]

\[ x_2 = -128.75 \]
Example 1 Cont.

The absolute relative approximate error is

\[
|\epsilon_a| = \left| \frac{x_2 - x_1}{x_2} \right| \times 100
\]

\[
= \left| \frac{-128.75 - (-128.06)}{-128.75} \right| \times 100
\]

\[
= 0.54076\%
\]

The number of significant digits at least correct is 1.

**Figure 6** Graph of estimated root after Iteration 2.
Example 1 Cont.

**Iteration 3**
The estimate of the root is

\[ f(-128.75) = -0.50598 \times 10^{-10}(-128.75)^3 - 0.38292 \times 10^{-7}(-128.75)^2 + 0.74363 \times 10^{-4}(-128.75) + 0.88318 \times 10^{-2} = 2.8002 \times 10^{-8} \]

\[ f'(-128.7544) = -1.5179 \times 10^{-10}(-128.75)^2 + 0.76582 \times 10^{-7}(-128.75) + 0.74363 \times 10^{-4} = 6.9186 \times 10^{-5} \]

\[ x_3 = x_2 - \frac{f(x_2)}{f'(x_2)} \]

\[ x_3 = -128.75 - \frac{2.8002 \times 10^{-8}}{6.9186 \times 10^{-5}} = -128.75 \]
Example 1 Cont.

The absolute relative approximate error is

$$\left| \varepsilon_a \right| = \left| \frac{x_3 - x_2}{x_3} \right| \times 100$$

$$= \left| \frac{-128.75 - (-128.75)}{-128.75} \right| \times 100$$

$$= 3.5086 \times 10^{-4} \%$$

The number of significant digits at least correct is 5.

**Figure 7** Graph of estimated root after Iteration 3.
Advantages and Drawbacks of Newton Raphson Method

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Advantages

- Converges fast (quadratic convergence), if it converges.
- Requires only one guess
Drawbacks

1. **Divergence at inflection points**
   Selection of the initial guess or an iteration value of the root that is close to the inflection point of the function \( f(x) \) may start diverging away from the root in the Newton-Raphson method.

For example, to find the root of the equation \( f(x) = (x - 1)^3 + 0.512 = 0 \).

The Newton-Raphson method reduces to \[ x_{i+1} = x_i - \frac{(x_i^3 - 1)^3 + 0.512}{3(x_i - 1)^2} \].

Table 1 shows the iterated values of the root of the equation.

The root starts to diverge at Iteration 6 because the previous estimate of 0.92589 is close to the inflection point of \( x = 1 \).

Eventually after 12 more iterations the root converges to the exact value of \( x = 0.2 \).
Drawbacks – Inflection Points

Table 1 Divergence near inflection point.

<table>
<thead>
<tr>
<th>Iteration Number</th>
<th>( x_i )</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>5.0000</td>
</tr>
<tr>
<td>1</td>
<td>3.6560</td>
</tr>
<tr>
<td>2</td>
<td>2.7465</td>
</tr>
<tr>
<td>3</td>
<td>2.1084</td>
</tr>
<tr>
<td>4</td>
<td>1.6000</td>
</tr>
<tr>
<td>5</td>
<td>0.92589</td>
</tr>
<tr>
<td>6</td>
<td>−30.119</td>
</tr>
<tr>
<td>7</td>
<td>−19.746</td>
</tr>
<tr>
<td>18</td>
<td>0.2000</td>
</tr>
</tbody>
</table>

Figure 8 Divergence at inflection point for 
\[ f(x) = (x-1)^3 + 0.512 = 0 \]
2. **Division by zero**
   
   For the equation
   
   \[ f(x) = x^3 - 0.03x^2 + 2.4 \times 10^{-6} = 0 \]
   
   the Newton-Raphson method reduces to
   
   \[ x_{i+1} = x_i - \frac{x_i^3 - 0.03x_i^2 + 2.4 \times 10^{-6}}{3x_i^2 - 0.06x_i} \]
   
   For \( x_0 = 0 \) or \( x_0 = 0.02 \), the denominator will equal zero.

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**Figure 9** Pitfall of division by zero or near a zero number
3. Oscillations near local maximum and minimum

Results obtained from the Newton-Raphson method may oscillate about the local maximum or minimum without converging on a root but converging on the local maximum or minimum.

Eventually, it may lead to division by a number close to zero and may diverge.

For example, for $f(x) = x^2 + 2 = 0$ the equation has no real roots.
Drawbacks – Oscillations near local maximum and minimum

**Table 3** Oscillations near local maxima and mimima in Newton-Raphson method.

| Iteration Number | $x_i$ | $f(x_i)$ | $|\varepsilon_a|/\%$ |
|------------------|------|---------|------------------|
| 0                | -1.0000 | 3.00 | |
| 1                | 0.5 | 2.25 | 300.00 |
| 2                | -1.75 | 5.063 | 128.571 |
| 3                | -0.30357 | 2.092 | 476.47 |
| 4                | 3.1423 | 11.874 | 109.66 |
| 5                | 1.2529 | 3.570 | 150.80 |
| 6                | -0.17166 | 2.029 | 829.88 |
| 7                | 5.7395 | 34.942 | 102.99 |
| 8                | 2.6955 | 9.266 | 112.93 |
| 9                | 0.97678 | 2.954 | 175.96 |

**Figure 10** Oscillations around local minima for $f(x) = x^2 + 2$. 

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4. **Root Jumping**

In some cases where the function $f(x)$ is oscillating and has a number of roots, one may choose an initial guess close to a root. However, the guesses may jump and converge to some other root.

For example

$f(x) = \sin x = 0$

Choose

$x_0 = 2.4\pi = 7.539822$

It will converge to $x = 0$

instead of $x = 2\pi = 6.2831853$

**Figure 11** Root jumping from intended location of root for $f(x) = \sin x = 0$
Additional Resources

For all resources on this topic such as digital audiovisual lectures, primers, textbook chapters, multiple-choice tests, worksheets in MATLAB, MATHEMATICA, MathCad and MAPLE, blogs, related physical problems, please visit

http://numericalmethods.eng.usf.edu/topics/newton_raphson.html
THE END

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