Chapter 03.04
Newton-Raphson Method of Solving a Nonlinear Equation

After reading this chapter, you should be able to:

1. derive the Newton-Raphson method formula,
2. develop the algorithm of the Newton-Raphson method,
3. use the Newton-Raphson method to solve a nonlinear equation, and
4. discuss the drawbacks of the Newton-Raphson method.

Introduction

Methods such as the bisection method and the false position method of finding roots of a nonlinear equation \( f(x) = 0 \) require bracketing of the root by two guesses. Such methods are called bracketing methods. These methods are always convergent since they are based on reducing the interval between the two guesses so as to zero in on the root of the equation.

In the Newton-Raphson method, the root is not bracketed. In fact, only one initial guess of the root is needed to get the iterative process started to find the root of an equation. The method hence falls in the category of open methods. Convergence in open methods is not guaranteed but if the method does converge, it does so much faster than the bracketing methods.

Derivation

The Newton-Raphson method is based on the principle that if the initial guess of the root of \( f(x) = 0 \) is at \( x_i \), then if one draws the tangent to the curve at \( f(x_i) \), the point \( x_{i+1} \) where the tangent crosses the \( x \)-axis is an improved estimate of the root (Figure 1). Using the definition of the slope of a function, at \( x = x_i \)

\[
f'(x_i) = \tan \theta = \frac{f(x_i) - 0}{x_i - x_{i+1}},
\]

which gives

\[
x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)}
\] (1)
Equation (1) is called the Newton-Raphson formula for solving nonlinear equations of the form $f(x) = 0$. So starting with an initial guess, $x_i$, one can find the next guess, $x_{i+1}$, by using Equation (1). One can repeat this process until one finds the root within a desirable tolerance.

**Algorithm**

The steps of the Newton-Raphson method to find the root of an equation $f(x) = 0$ are

1. Evaluate $f'(x)$ symbolically
2. Use an initial guess of the root, $x_i$, to estimate the new value of the root, $x_{i+1}$, as
   
   \[ x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)} \]

3. Find the absolute relative approximate error $|\varepsilon_a|$ as
   
   \[ |\varepsilon_a| = \left| \frac{x_{i+1} - x_i}{x_{i+1}} \right| \times 100 \]

4. Compare the absolute relative approximate error with the pre-specified relative error tolerance, $\varepsilon_s$. If $|\varepsilon_a| > \varepsilon_s$, then go to Step 2, else 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.

![Figure 1](#) Geometrical illustration of the Newton-Raphson method.
Example 1
A trunnion has to be cooled before it is shrink fitted into a steel hub.

![Figure 2 Trunnion to be slid through the hub after contracting.](image)

The equation that gives the temperature \( T_f \) to which the trunnion has to be cooled to obtain the desired contraction is given by

\[
\begin{align*}
0 &= -0.50598 \times 10^{-10} T_f^3 + 0.38292 \times 10^{-7} T_f^2 + 0.74363 \times 10^{-4} T_f + 0.88318 \times 10^{-2} \\
&= f(T_f)
\end{align*}
\]

Use the Newton-Raphson method of finding roots of equations to find the temperature \( T_f \) to which the trunnion has to be cooled. Conduct three iterations to estimate the root of the above equation. Find the absolute relative approximate error at the end of each iteration and the number of significant digits at least correct at the end of each iteration.

**Solution**

\[
\begin{align*}
f(T_f) &= -0.50598 \times 10^{-10} T_f^3 + 0.38292 \times 10^{-7} T_f^2 + 0.74363 \times 10^{-4} T_f + 0.88318 \times 10^{-2} \\
&= -1.51794 \times 10^{-10} T_f^2 + 0.76584 \times 10^{-7} T_f + 0.74363 \times 10^{-4} \\
f'(T_f) &= 0.50598 \times 10^{-10} T_f^2 + 0.74363 \times 10^{-4} T_f + 0.88318 \times 10^{-2}
\end{align*}
\]

Let us assume the initial guess of the root of \( f(T_f) = 0 \) as \( T_{f,0} = -100 \).

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

\[
T_{f,1} = T_{f,0} - \frac{f(T_{f,0})}{f'(T_{f,0})}
\]

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

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

\[
= -100 - (28.058)
\]
= −128.06

The absolute relative approximate error \( |\varepsilon_a| \) at the end of Iteration 1 is

\[
|\varepsilon_a| = \left| \frac{T_{f,1} - T_{f,0}}{T_{f,1}} \right| \times 100
\]

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

\[
= 21.910\%
\]

The number of significant digits at least correct is 0, as you need an absolute relative approximate error of less than 5% for one significant digit to be correct in your result.

**Iteration 2**

The estimate of the root is

\[
T_{f,2} = T_{f,1} - \frac{f(T_{f,1})}{f'(T_{f,1})}
\]

\[
= -128.06 - \left( -\frac{-0.50598 \times 10^{-10}(-128.06)^3 + 0.38292 \times 10^{-7} - (128.06)^2}{1.5179 \times 10^{-10}(-128.06)^2 + 0.76584 \times 10^{-7}(-128.06)} + 0.74363 \times 10^{-4} \right)
\]

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

\[
= -128.06 - (0.69625)
\]

\[
= -128.75
\]

The absolute relative approximate error \( |\varepsilon_a| \) at the end of Iteration 2 is

\[
|\varepsilon_a| = \left| \frac{T_{f,2} - T_{f,1}}{T_{f,2}} \right| \times 100
\]

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

\[
= 0.54076\%
\]

The number of significant digits at least correct is 1.

**Iteration 3**

The estimate of the root is

\[
T_{f,3} = T_{f,2} - \frac{f(T_{f,2})}{f'(T_{f,2})}
\]

\[
= -128.75 - \left( -\frac{-0.50598 \times 10^{-10}(-128.75)^3 + 0.38292 \times 10^{-7} - (-128.75)^2}{1.5179 \times 10^{-10}(-128.75)^2 + 0.76584 \times 10^{-7}(-128.75)} + 0.74363 \times 10^{-4} \right)
\]

\[
= -128.75 - \frac{-0.50598 \times 10^{-10}(-128.75)^3 + 0.38292 \times 10^{-7} - (-128.75)^2}{1.5179 \times 10^{-10}(-128.75)^2 + 0.76584 \times 10^{-7}(-128.75)} + 0.74363 \times 10^{-4}
\]
The absolute relative approximate error $|e_a|$ at the end of Iteration 3 is

$$|e_a| = \left| \frac{T_{f,3} - T_{f,2}}{T_{f,3}} \right| \times 100$$

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

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

Hence the number of significant digits at least correct is given by the largest value of $m$ for which

$$|e_a| \leq 0.5 \times 10^{2-m}$$

$$3.5086 \times 10^{-4} \leq 0.5 \times 10^{2-m}$$

$$7.0173 \times 10^{-4} \leq 10^{2-m}$$

$$\log(7.0173 \times 10^{-4}) \leq 2 - m$$

$$m \leq 2 - \log(7.0173 \times 10^{-4}) = 5.1538$$

So

$$m = 5$$

The number of significant digits at least correct in the estimated root $-128.75$ is 5.

**Drawbacks of the Newton-Raphson Method**

1. Divergence at inflection points

If the selection of the initial guess or an iterated value of the root turns out to be close to the inflection point (see the definition in the appendix of this chapter) of the function $f(x)$ in the equation $f(x) = 0$, Newton-Raphson method may start diverging away from the root. It may then start converging back to the root. For example, to find the root of the equation

$$f(x) = (x - 1)^3 + 0.512 = 0$$

the Newton-Raphson method reduces to

$$x_{i+1} = x_i - \frac{(x_i^3 - 1)^3 + 0.512}{3(x_i - 1)^2}$$

Starting with an initial guess of $x_0 = 5.0$, Table 1 shows the iterated values of the root of the equation. As you can observe, the root starts to diverge at Iteration 6 because the previous estimate of 0.92589 is close to the inflection point of $x = 1$ (the value of $f'(x)$ is zero at the inflection point). Eventually, after 12 more iterations the root converges to the exact value of $x = 0.2$. 

\[
\begin{align*}
  x_0 & = 5.0 \\
  x_1 & = 3.88 \\
  x_2 & = 2.31 \\
  x_3 & = 1.51 \\
  x_4 & = 1.29 \\
  x_5 & = 1.26 \\
  x_6 & = 1.26 \\
  x_7 & = 1.26 \\
  x_8 & = 1.26 \\
  x_9 & = 1.26 \\
  x_{10} & = 1.26 \\
  x_{11} & = 1.26 \\
  x_{12} & = 1.26 \\
  x_{13} & = 1.26 \\
  x_{14} & = 1.26 \\
  x_{15} & = 1.26 \\
  x_{16} & = 1.26 \\
  x_{17} & = 1.26 \\
  x_{18} & = 1.26 \\
  x_{19} & = 1.26 \\
  x_{20} & = 1.26 \\
  x_{21} & = 1.26 \\
  x_{22} & = 1.26 \\
  x_{23} & = 1.26 \\
  x_{24} & = 1.26 \\
  x_{25} & = 1.26 \\
  x_{26} & = 1.26 \\
  x_{27} & = 1.26 \\
  x_{28} & = 1.26 \\
  x_{29} & = 1.26 \\
  x_{30} & = 1.26 \\
\end{align*}
\]
Table 1  Divergence near inflection point.

<table>
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<tr>
<th>Iteration Number</th>
<th>$x_i$</th>
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<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>8</td>
<td>-12.831</td>
</tr>
<tr>
<td>9</td>
<td>-8.2217</td>
</tr>
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<td>10</td>
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<td>11</td>
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<td>15</td>
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<td>16</td>
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<td>17</td>
<td>0.19924</td>
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<td>18</td>
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</table>

Figure 3  Divergence at inflection point for $f(x) = (x-1)^3 = 0$.

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 \( x_0 = 0.02 \), division by zero occurs (Figure 4). For an initial guess close to 0.02 such as \( x_0 = 0.01999 \), one may avoid division by zero, but then the denominator in the formula is a small number. For this case, as given in Table 2, even after 9 iterations, the Newton-Raphson method does not converge.

### Table 2 Division by near zero in Newton-Raphson method.

| Iteration Number | \( x_i \)   | \( f(x_i) \) | \( |\varepsilon| \)% |
|-----------------|-------------|--------------|-----------------|
| 0               | 0.019990    | \(-1.60000 \times 10^{-6}\) | —               |
| 1               | -2.6480     | 18.778       | 100.75          |
| 2               | -1.7620     | -5.5638      | 50.282          |
| 3               | -1.1714     | -1.6485      | 50.422          |
| 4               | -0.77765    | -0.48842     | 50.632          |
| 5               | -0.51518    | -0.14470     | 50.946          |
| 6               | -0.34025    | -0.042862    | 51.413          |
| 7               | -0.22369    | -0.012692    | 52.107          |
| 8               | -0.14608    | -0.0037553   | 53.127          |
| 9               | -0.094490   | -0.0011091   | 54.602          |

**Figure 4** Pitfall of division by zero or a near 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 (Figure 5 and Table 3).
Figure 5 Oscillations around local minima for $f(x) = x^2 + 2$.

| Iteration Number | $x_i$  | $f(x_i)$ | $|\epsilon|\%$ |
|------------------|--------|----------|-------------|
| 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      |

4. Root jumping
In some case 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 for solving the equation $\sin x = 0$ if you choose $x_0 = 2.4\pi = (7.539822)$ as an initial guess, it converges to the root of $x = 0$ as shown in Table 4 and Figure 6. However, one may have chosen this as an initial guess to converge to $x = 2\pi = 6.2831853$. 
Table 4  Root jumping in Newton-Raphson method.

| Iteration Number | \( x_i \)     | \( f(x_i) \) | \( |e_i| \\% \) |
|------------------|----------------|--------------|---------------|
| 0                | 7.539822       | 0.951        |               |
| 1                | 4.462          | -0.969       | 68.973        |
| 2                | 0.5499         | 0.5226       | 711.44        |
| 3                | -0.06307       | -0.06303     | 971.91        |
| 4                | 8.376 \times 10^{-4} | 8.375 \times 10^{-5} | 7.54 \times 10^4 |
| 5                | -1.95861 \times 10^{-13} | -1.95861 \times 10^{-13} | 4.28 \times 10^{10} |

Figure 6  Root jumping from intended location of root for \( f(x) = \sin x = 0 \).

Appendix A. What is an inflection point?

For a function \( f(x) \), the point where the concavity changes from up-to-down or down-to-up is called its inflection point. For example, for the function \( f(x) = (x-1)^3 \), the concavity changes at \( x=1 \) (see Figure 3), and hence \((1,0)\) is an inflection point.

An inflection points MAY exist at a point where \( f''(x)=0 \) and where \( f'''(x) \) does not exist. The reason we say that it MAY exist is because if \( f''(x)=0 \), it only makes it a possible inflection point. For example, for \( f(x) = x^4 - 16 \), \( f''(0) = 0 \), but the concavity does not change at \( x=0 \). Hence the point \((0,-16)\) is not an inflection point of \( f(x) = x^4 - 16 \).

For \( f'(x) = (x-1)^3 \), \( f''(x) \) changes sign at \( x=1 \) (\( f''(x) < 0 \) for \( x<1 \), and \( f''(x) > 0 \) for \( x>1 \)), and thus brings up the Inflection Point Theorem for a function \( f(x) \) that states the following.

"If \( f'(c) \) exists and \( f''(c) \) changes sign at \( x=c \), then the point \((c,f(c))\) is an inflection point of the graph of \( f \)."
Appendix B. Derivation of Newton-Raphson method from Taylor series

Newton-Raphson method can also be derived from Taylor series. For a general function $f(x)$, the Taylor series is

$$f(x_{i+1}) = f(x_i) + f'(x_i)(x_{i+1} - x_i) + \frac{f''(x_i)}{2!}(x_{i+1} - x_i)^2 + \ldots$$

As an approximation, taking only the first two terms of the right hand side,

$$f(x_{i+1}) \approx f(x_i) + f'(x_i)(x_{i+1} - x_i)$$

and we are seeking a point where $f(x) = 0$, that is, if we assume

$$f(x_{i+1}) = 0,$$

$$0 \approx f(x_i) + f'(x_i)(x_{i+1} - x_i)$$

which gives

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

This is the same Newton-Raphson method formula series as derived previously using the geometric method.

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### NONLINEAR EQUATIONS

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