

# More Fun with Roots: the Secant Method

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MTH 499-02

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- Implementation and examples

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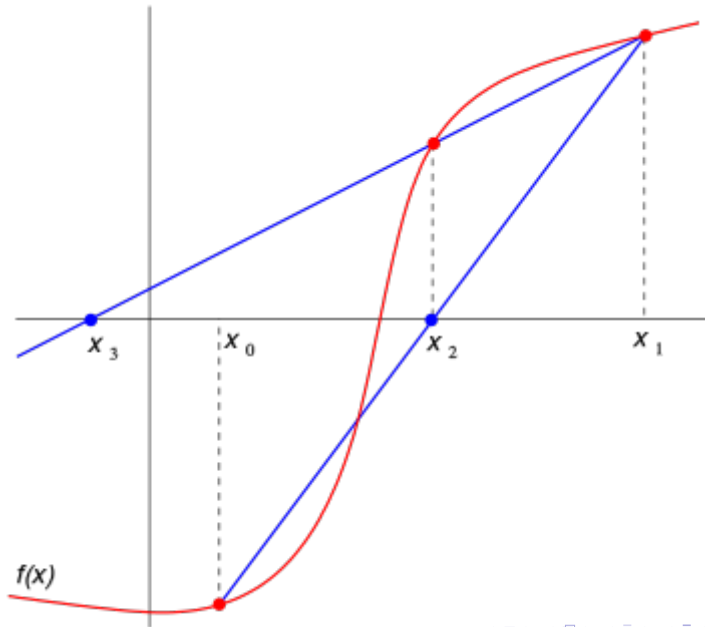
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$$x_{n+1} = x_n - f(x_n) \frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})}$$



```

1: function SECANT( $f, x_0, x_1, M, \delta, \epsilon$ )
2:    $xvals_0 \leftarrow x_0; xvals_1 \leftarrow x_1$ 
3:    $fvals_0 \leftarrow f(xvals_0); fvals_1 \leftarrow f(xvals_1)$ 
4:   output 0,  $xvals_0, fvals_0$ 
5:   output 1,  $xvals_1, fvals_1$ 
6:   for  $k = 2$  to  $M$  do
7:     if  $|fvals_{k-2}| > |fvals_{k-1}|$  then
8:        $xvals_{k-2} \leftrightarrow xvals_{k-1}; fvals_{k-2} \leftrightarrow fvals_{k-1}$ 
9:        $xvals_k \leftarrow xvals_{k-2} - fvals_{k-2} \cdot \frac{xvals_{k-1} - xvals_{k-2}}{fvals_{k-1} - fvals_{k-2}}$ 
10:       $xvals_{k-1} \leftarrow xvals_{k-2}$ 
11:       $fvals_{k-1} \leftarrow fvals_{k-2}$ 
12:       $fvals_k \leftarrow f(xvals_k)$ 
13:   return  $xvals, fvals$ 

```

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- Using the definition of the secant method, we have

$$\begin{aligned} e_{n+1} &= x_{n+1} - r \\ &= \frac{f(x_n)x_{n-1} - f(x_{n-1})x_n}{f(x_n) - f(x_{n-1})} - r. \end{aligned}$$

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- We can substitute  $x_n - r = e_n$  from the definition of our error and then simplify to give us

$$e_{n+1} = \frac{f(x_n)e_{n-1} - f(x_{n-1})e_n}{f(x_n) - f(x_{n-1})}.$$

- We can then remove a factor of  $e_n e_{n-1}$  and insert a factor of  $(x_n - x_{n-1})$  to give

$$e_{n+1} = \left[ \frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})} \right] \cdot \left[ \frac{f(x_n)/e_n - f(x_{n-1})/e_{n-1}}{x_n - x_{n-1}} \right] \cdot e_n e_{n-1}. \quad (1)$$

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- Remember this equation (or write it down if you can't memorize this). It's important.
- We wish to bound  $f(x_n)$  and  $f(x_{n-1})$  in order to estimate our error in terms of a constant. So, first, note that Taylor's theorem gives us

$$\begin{aligned} f(x_n) &= f(r + e_n) \\ &= f(r) + e_n f'(r) + \frac{1}{2} e_n^2 f''(r) + \mathcal{O}(e_n^3). \end{aligned}$$

- We are searching for a zero to the function, so clearly  $f(r) = 0$ . Thus,

$$\frac{f(x_n)}{e_n} = f'(r) + \frac{1}{2}e_n f''(r) + \mathcal{O}(e_n^2).$$

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- If we change the index from  $n$  to  $n - 1$ , we see that

$$\frac{f(x_{n-1})}{e_{n-1}} = f'(r) + \frac{1}{2}e_{n-1} f''(r) + \mathcal{O}(e_{n-1}^2).$$

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- Subtracting the previous two equations yields

$$\frac{f(x_n)}{e_n} - \frac{f(x_{n-1})}{e_{n-1}} = \frac{1}{2}(e_n - e_{n-1})f''(r) + \mathcal{O}(e_{n-1}^2).$$



- Since  $x_n - x_{n-1} = e_n - e_{n-1}$ , we can divide each side of the previous equation by  $x_n - x_{n-1}$  and get

$$\frac{\frac{f(x_n)}{e_n} - \frac{f(x_{n-1})}{e_{n-1}}}{x_n - x_{n-1}} \approx \frac{1}{2}f''(r).$$

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- Note that the quantity contained within the first set of brackets in Equation (1) can be rewritten as

$$\frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})} \approx \frac{1}{f'(r)}.$$

- For those who were imprudent, remember that Equation (1) was

$$e_{n+1} = \left[ \frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})} \right] \cdot \left[ \frac{f(x_n)/e_n - f(x_{n-1})/e_{n-1}}{x_n - x_{n-1}} \right] \cdot e_n e_{n-1}.$$

- Thus, we have shown that

$$e_{n+1} \approx \frac{1}{2} \cdot \frac{f''(r)}{f'(r)} e_n e_{n-1} = C e_n e_{n-1}.$$

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- Note that this is similar to the error in Newton's method, which was

$$e_{n+1} = \frac{1}{2} \cdot \frac{f''(\xi_n)}{f'(x_n)} e_n^2 \approx \frac{1}{2} \frac{f''(r)}{f'(r)} e_n^2 = C e_n^2.$$

- Code!