

ME 310

Numerical Methods

Finding Roots of Nonlinear Equations

These presentations are prepared by

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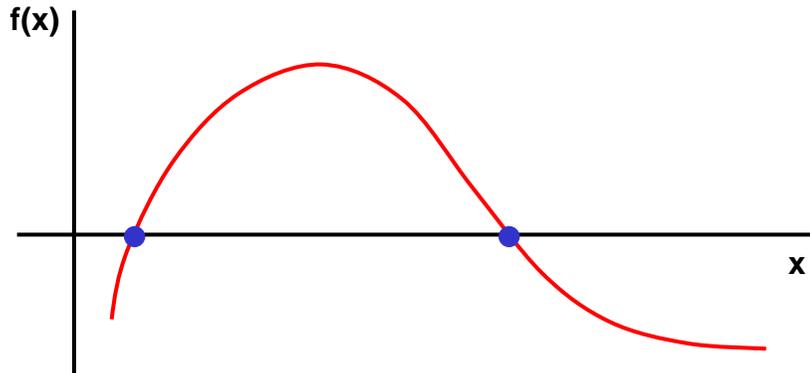
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Part 2. Finding Roots of Equations

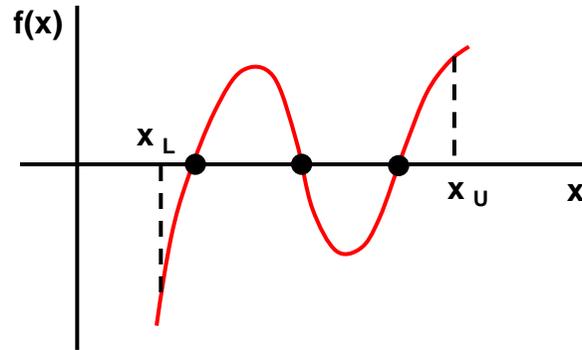
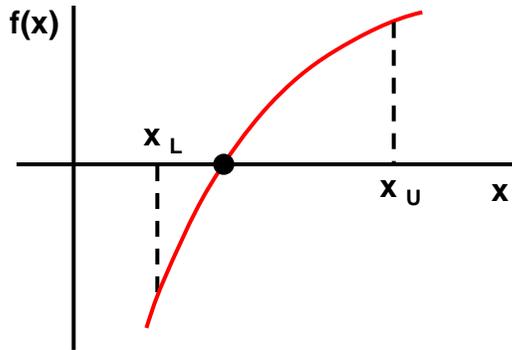


$f(x)$ is given

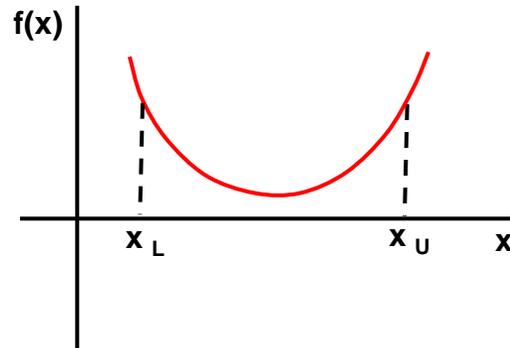
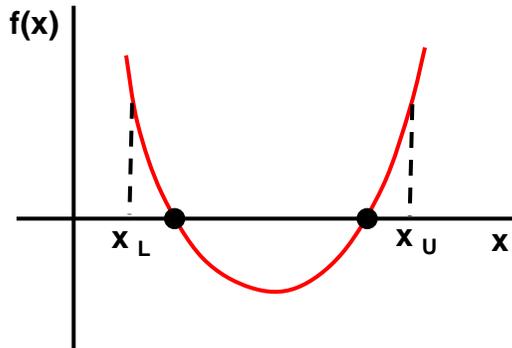
$$f(x_r) = 0 \rightarrow x_r = ?$$

- Bracketing Methods (Need two initial estimates that will bracket the root. Always converge.)
 - Bisection Method
 - False-Position Method
- Open Methods (Need one or two initial estimates. May diverge.)
 - Simple One-Point Iteration
 - Newton-Raphson Method (Needs the derivative of the function.)
 - Secant Method

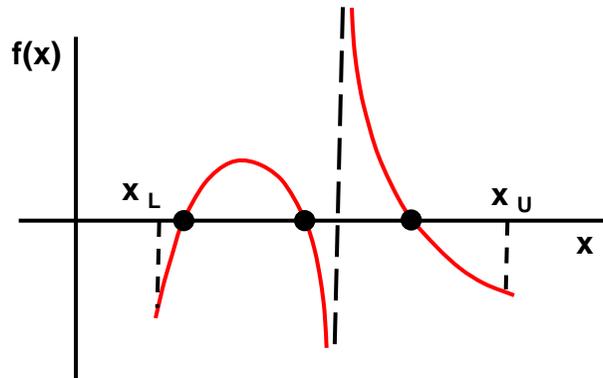
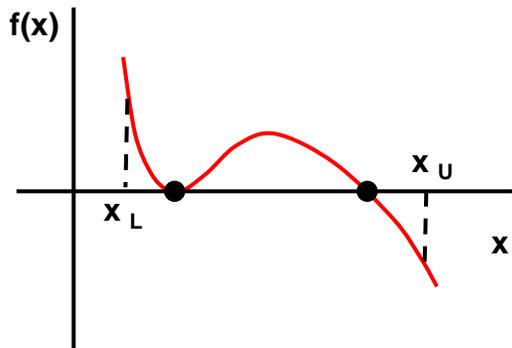
Ch 5. General Idea of Bracketing Methods



Rule 1: If $f(x_L) \cdot f(x_U) < 0$
than there are
odd number of roots

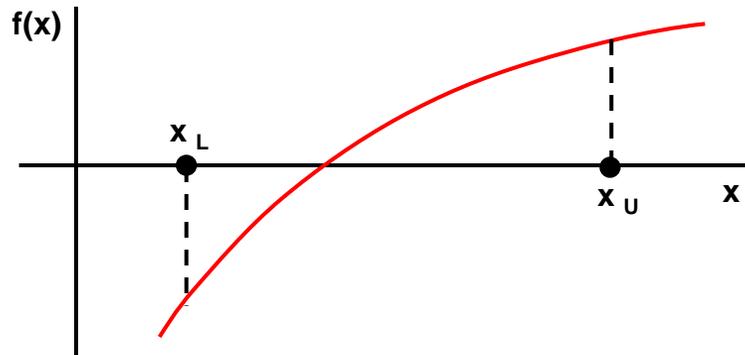


Rule 2: If $f(x_L) \cdot f(x_U) > 0$
than there are
i) even number of roots
ii) no roots

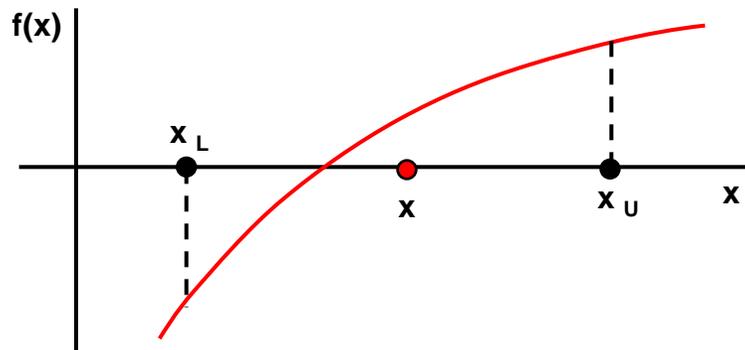


Violations:
i) multiple roots
ii) discontinuities

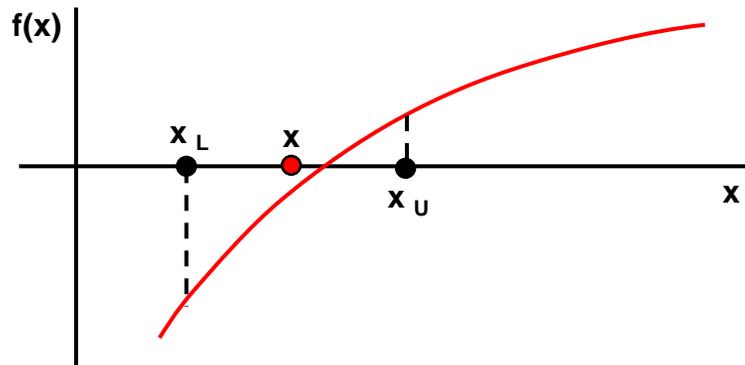
Bisection Method



- Start with two initial guesses, x_{LOWER} and x_{UPPER} .
- They should bracket the root, i.e. $f(x_L) * f(x_U) < 0$



- Estimate the root as the midpoint of this interval. $x = (x_L + x_U)/2$
- Determine the interval which contains the root
if $f(x_L) * f(x) < 0$ root is between x_L and x
else root is between x and x_U



- Estimate a new root in this new interval
- Iterate like this and compute the error after each iteration.
- Stop when the specified tolerance is reached.

Bisection Method (cont'd)

- It always converge to the true root (**but be careful** about the following)
- $f(x_L) * f(x_U) < 0$ is true if the interval has odd number of roots, not necessarily one root.

Example 6: Find the square root of 11.

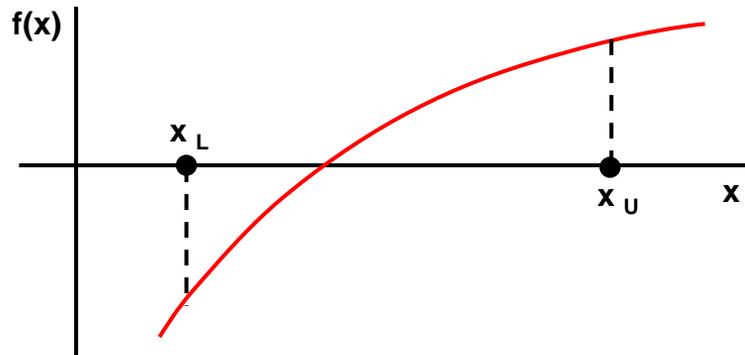
$$x^2 = 11 \rightarrow f(x) = x^2 - 11 \quad (\text{note that the exact solution is } 3.31662479)$$

Select initial guesses: $3^2=9 < 11$, $4^2=16 > 11$ $\rightarrow x_L = 3, x_U = 4$

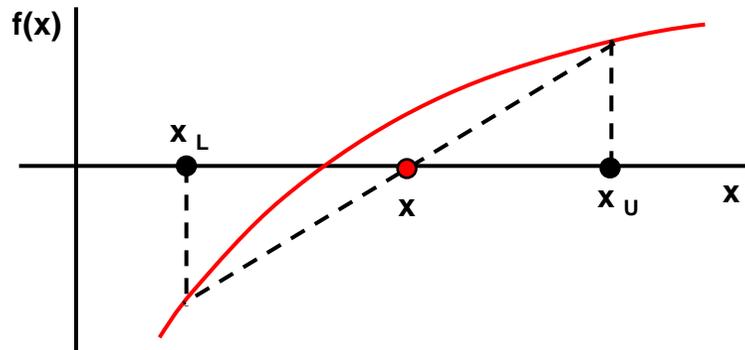
Iteration no.	x	f(x)	$ \varepsilon_t $ %	$ \varepsilon_a $ %
1	3.5	1.25	5.53	-----
2	3.25	-0.4375	2.01	7.69
3	3.375	0.390625	1.76	3.70
4	3.3125	-0.02734375	0.12	1.89
5	3.34375	0.180664062	0.82	0.93
6	3.328125	0.076416015	0.35	0.47

- Errors do not drop monotonically but oscillate.
- $\varepsilon_a > \varepsilon_t$ at each step, which is good. It means using ε_a is conservative.
- ε_a can also be estimated as $\varepsilon_a = (x_U - x_L) / (x_U + x_L)$. This can be used for the 1st iteration too.

False-Position Method



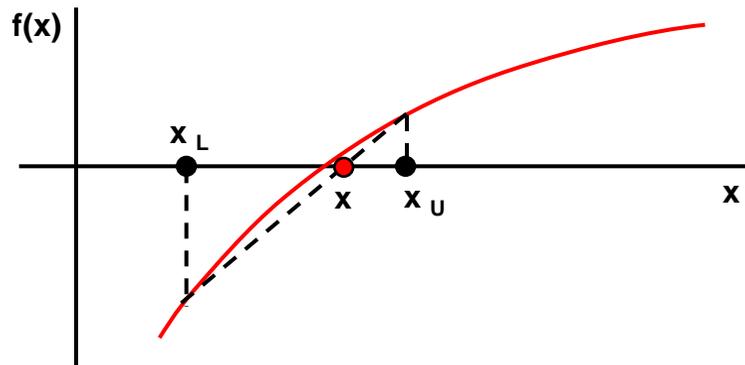
- Start with two initial guesses, x_{LOWER} and x_{UPPER} .
- They should bracket the root, i.e. $f(x_L) * f(x_U) < 0$



- Estimate the root using similar triangles.

$$x = x_U - \frac{f(x_U)(x_L - x_U)}{f(x_L) - f(x_U)}$$

- Determine the interval which contains the root
if $f(x_L) * f(x) < 0$ root is between x_L and x
else root is between x and x_U



- Estimate a new root in this new interval
- Iterate like this and compute the error after each iteration.
- Stop when the specified tolerance is reached.

False-Position Method (cont'd)

- It always converge to the true root.
- $f(x_L) * f(x_U) < 0$ is true if the interval has odd number of roots, not necessarily one root.
- **Generally converges faster than the bisection method** (See page 127 for an exception).

Example 7: Repeat the previous example (Find the square root of 11).

Iteration no.	x	f(x)	$ \epsilon_t $ %	$ \epsilon_a $ %
1	3.28571429	-0.1040816	0.932	-----
2	3.31372549	-0.0192234	0.087	0.845
3	3.31635389	-0.0017969	0.0082	0.079
4	3.31659949	-0.0001678	0.00076	0.0074
5	3.31662243	-0.0000157	0.00007	0.00069
6	3.31662457	-0.0000015	0.00001	0.00006

- Errors drop monotonically. Converges faster than the bisection method.
- $\epsilon_a > \epsilon_t$ at each step, which is good. It means ϵ_a is conservative.

About bracketing methods

- A plot of the function is always helpful.
 - to determine the number of all roots, if there are any.
 - to determine whether the roots are multiple or not.
 - to determine whether the method converges to the desired root.
 - to determine the initial guesses.
- Incremental search technique can be used to determine the initial guesses.
 - Start from one end of the region of interest.
 - Evaluate the function at specified intervals.
 - If the sign of the function changes, then there is a root in that interval.
 - Select your intervals small, otherwise you may miss some of the roots. But if they are too small, incremental search might become too costly.
 - Incremental search, just by itself, can be used as a root finding technique with very small intervals (not efficient).

Fortran Code for Bisection Method

```
PROGRAM BISECTION  ! Calculates the root of a function
```

```
INTEGER :: iter, maxiter
```

```
REAL(8) :: xL, xU, x, fxL, fxU, fx, tol, error
```

```
READ(*,*) xL, xU, tol, maxiter
```

```
DO iter = 1, maxiter
```

```
  x = (xL + xU) / 2  ! Only this line changes for the False-Position Method
```

```
  fx = FUNC (x)      ! Call a subroutine to calculate the function.
```

```
  error = (xU - xL) / (xU + xL)*100  ! See page 121
```

```
  WRITE(*,*) iter, x, fx, error
```

```
  IF (error < tol) STOP
```

```
  fxL = FUNC (xL)
```

```
  fxU = FUNC (xU)
```

```
  IF(fxL * fx < 0) THEN
```

```
    xU = x
```

```
  ELSE
```

```
    xL = x
```

```
  ENDIF
```

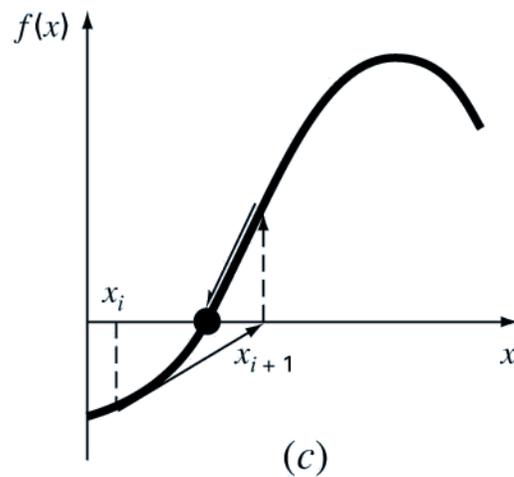
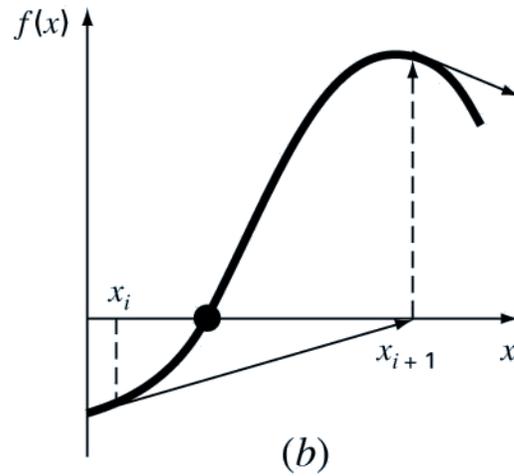
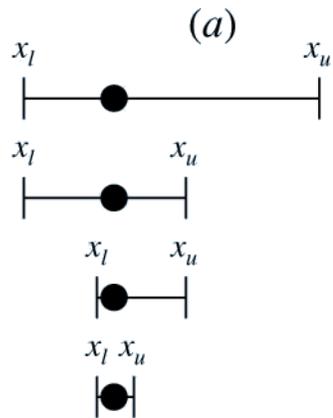
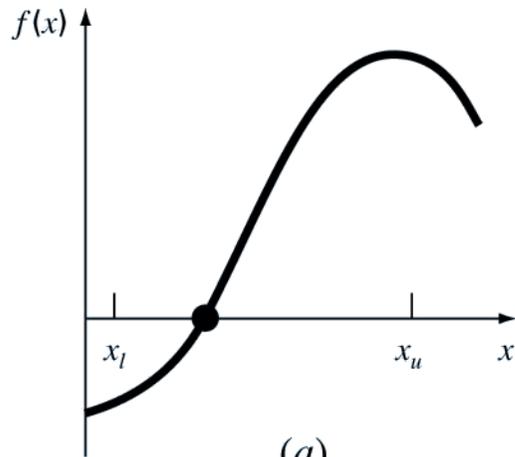
```
ENDDO
```

```
END PROGRAM BISECTION
```

Exercise 5: Write a C program for the Bisection method and implement the following improvements

- Check if the initial guesses bracket the root or not.
- Read the true value if it is known and calculate ε_t in addition to ε_a .
- Check for the cases of $f(x_L)=0$ or $f(x_U)=0$

Ch 6. Open Methods



- (a) Bracketing methods always converge.
- (b) Open methods may diverge
- (c) or converge very rapidly.

Simple One-Point Iteration

- Put the original formulation of $f(x) = 0$ into a form of $x=g(x)$.
- Many possibilities are possible
 - $\ln(x) - 3x + 5 = 0 \quad \rightarrow \quad x = [\ln(x) + 5] / 3 \quad \text{or} \quad x = e^{3x-5}$
 - $\cos(x) = 0 \quad \rightarrow \quad x = x + \cos(x)$
- Start with an initial guess x_0
- Calculate a new estimate for the root using $x_1 = g(x_0)$
- Iterate like this. General formula is $x_{i+1} = g(x_i)$
- Converges if $|g'(x)| < 1$ in the region of interest (Easier to see graphically in the coming slides).

Simple One-Point Iteration (cont'd)

Example 8: Repeat the previous example (Find the square root of 11). Start with $x_0 = 3$.

(a) $x^2 - 11 = 0 \rightarrow x = x + x^2 - 11$

i	x_i	$x_{i+1}=g(x_i)$	$ \epsilon_t \%$
0	3	1	70
1	1	-9	371
2	-9	61	1739
3	61	3771	113600
Diverges			

(b) $11 - x^2 = 0 \rightarrow x = (4x + 11 - x^2)/4$

i	x_i	$x_{i+1}=g(x_i)$	$ \epsilon_t \%$
0	3	3.5	5.53
1	3.5	3.1875	3.89
2	3.1875	3.39746094	2.44
3	3.39746094	3.26177573	1.65
4	3.26177573	3.35198050	1.07
5	3.35198050	3.29303718	0.71
Converges			

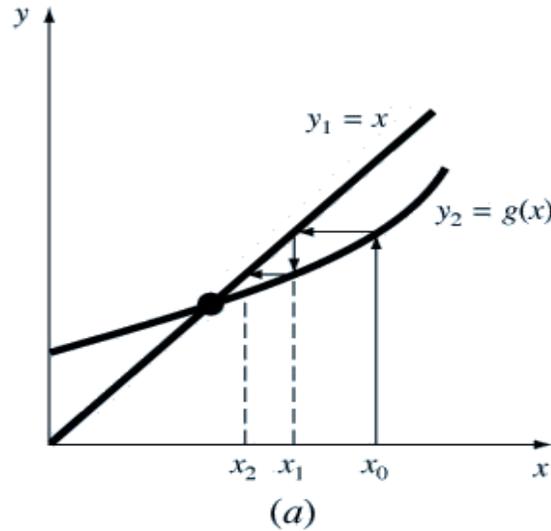
- Selection of $g(x)$ is important. Note that we did not use the trivial one, $x = \sqrt{11}$.
- If the method converges, convergence is linear. That is the relative error at each iteration is roughly proportional to the half of the previous error. This is easier to see for ϵ_t .

Exercise 6: Show that (a) violates the convergence criteria $|g'(x)| < 1$.

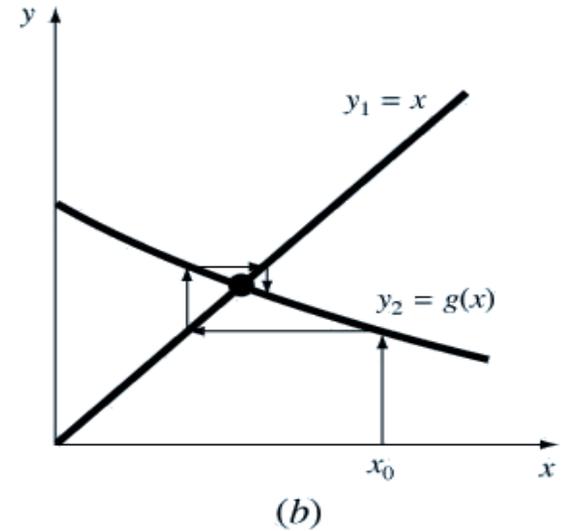
Simple One-Point Iteration (cont'd)

- We can use the two-curve graphical method to check convergence $|g'(x)| < 1$.

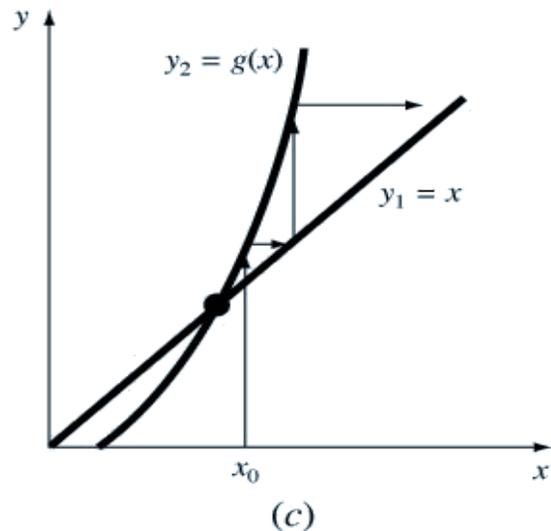
(a) Monotone convergence



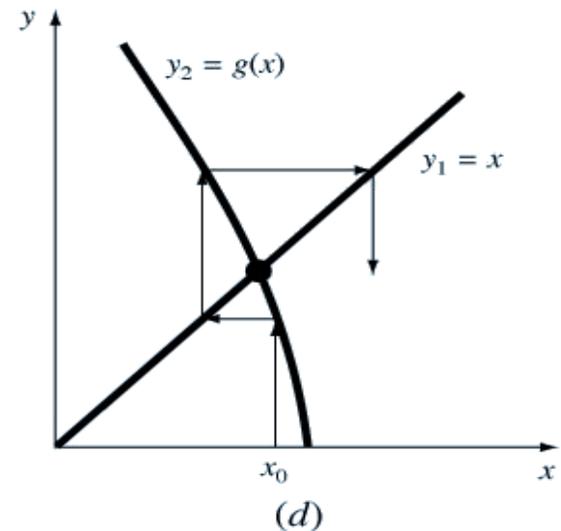
(b) Spiral convergence



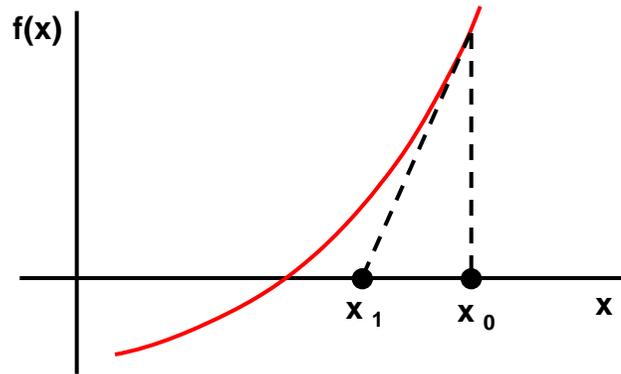
(c) Monotone divergence



(d) Spiral divergence

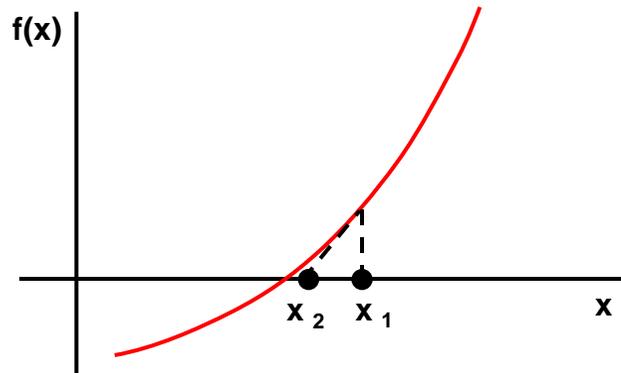


Newton-Raphson Method



- Start with an initial guess x_0 and calculate x_1 as

$$f'(x_0) = \frac{f(x_0) - 0}{x_1 - x_0} \rightarrow x_1 = x_0 - \frac{f(x_0)}{f'(x_0)}$$



- Continue like this until you reach the specified tolerance or maximum number of iterations.

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

Newton-Raphson Method (cont'd)

Example 9: Repeat the previous example (Find the square root of 11). Start with $x_0 = 3$.

$$f(x) = x^2 - 11 = 0, \quad f'(x) = 2x \quad \rightarrow \quad x_{i+1} = x_i - (x^2 - 11)/2x$$

iter	x	f(x)	$ \varepsilon_t $ %
0	3	-2	9.55
1	3.33333333	0.11111111	0.50
2	3.31666667	0.0002778	0.00126
3	3.31662479	0.0000000	0.00000

Selection of the initial guess affects the convergence a lot.

Exercise 7: NR method is quite sensitive to the starting point. Try to find the first positive root of $\sin(x)$ (which is 1.570796) starting with a) $x_0 = 1.5$, b) $x_0 = 1.7$, c) $x_0 = 1.8$, d) $x_0 = 1.9$ (They all converge to different roots).

Derivation of NR Method from Taylor Series

$$\mathbf{f}(\mathbf{x}_{i+1}) = \mathbf{f}(\mathbf{x}_i) + \mathbf{f}'(\mathbf{x}_i)(\mathbf{x}_{i+1} - \mathbf{x}_i) + \frac{\mathbf{f}''(\xi)}{2!}(\mathbf{x}_{i+1} - \mathbf{x}_i)^2$$

Use first order approximation $\rightarrow \mathbf{f}(\mathbf{x}_{i+1}) \approx \mathbf{f}(\mathbf{x}_i) + \mathbf{f}'(\mathbf{x}_i)(\mathbf{x}_{i+1} - \mathbf{x}_i)$

To find the root, $\mathbf{f}(\mathbf{x}_{i+1})$ should be zero $\rightarrow \mathbf{0} \approx \mathbf{f}(\mathbf{x}_i) + \mathbf{f}'(\mathbf{x}_i)(\mathbf{x}_{i+1} - \mathbf{x}_i)$

Solve for \mathbf{x}_{i+1} $\rightarrow \mathbf{x}_{i+1} = \mathbf{x}_i - \frac{\mathbf{f}(\mathbf{x}_i)}{\mathbf{f}'(\mathbf{x}_i)}$

Error Analysis

If we use the complete Taylor Series the result would be exact.

$$\mathbf{f}(\mathbf{x}_e) = \mathbf{0} = \mathbf{f}(\mathbf{x}_i) + \mathbf{f}'(\mathbf{x}_i)(\mathbf{x}_e - \mathbf{x}_i) + \frac{\mathbf{f}''(\xi)}{2!}(\mathbf{x}_e - \mathbf{x}_i)^2 \quad \mathbf{x}_e \text{ is the exact root}$$

Subtract this from the 3rd equation $\rightarrow \mathbf{0} = \mathbf{f}'(\mathbf{x}_i)(\mathbf{x}_e - \mathbf{x}_{i+1}) + \frac{\mathbf{f}''(\xi)}{2}(\mathbf{x}_e - \mathbf{x}_i)^2$

Note that $E_{t,i} = (\mathbf{x}_e - \mathbf{x}_i)$ and $E_{t,i+1} = (\mathbf{x}_e - \mathbf{x}_{i+1})$ $\rightarrow \mathbf{0} = \mathbf{f}'(\mathbf{x}_i)E_{t,i+1} + \frac{\mathbf{f}''(\xi)}{2}E_{t,i}^2$

For convergence both \mathbf{x}_i and ξ should approach to \mathbf{x}_e $\rightarrow E_{t,i+1} \approx \frac{\mathbf{f}''(\mathbf{x}_e)}{2\mathbf{f}'(\mathbf{x}_e)}E_{t,i}^2$

This is quadratic convergence. That is the error at each iteration is roughly proportional to the square of the previous error. (See page 141 for a nice example).

Exercise 8: We showed that the NR can be derived from a 1st order Taylor series expansion. Derive a new method using 2nd order Taylor series expansion. Compare the convergence of this method with NR. Comment on the practicality of this new method.

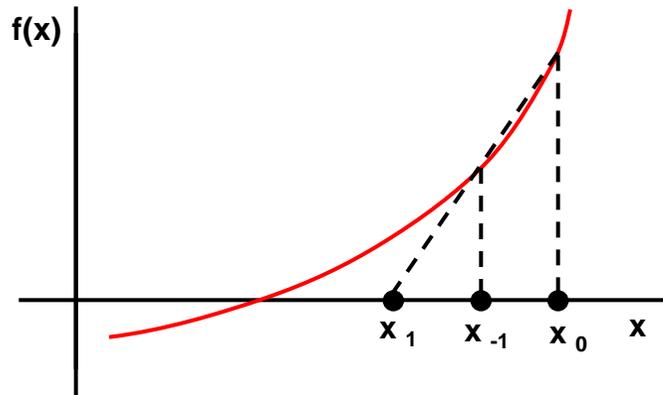
Exercise 9: NR method can be seen as Simple One-Point Iteration method with $g(x) = x_i - f(x_i) / f'(x_i)$. Using the convergence criteria of the Simple One-Point Iteration Method, derive a convergence criteria for the NR Method.

Difficulties with the NR Method (page 144)

- Need to know the derivative of the function.
- May diverge around inflection points.
- May give oscillations around local minimums or maximums.
- Zero or near zero slope is a problem, because f' is at the denominator.
- Convergence may be slow if the initial guess is poor.

Exercise 9.1 : Use the NR method to locate one of the roots of $f(x) = x(x-5)(x-6)+3$ starting with $x_0 = 3.5$ (NR will oscillate around the local minimum).

Secant Method

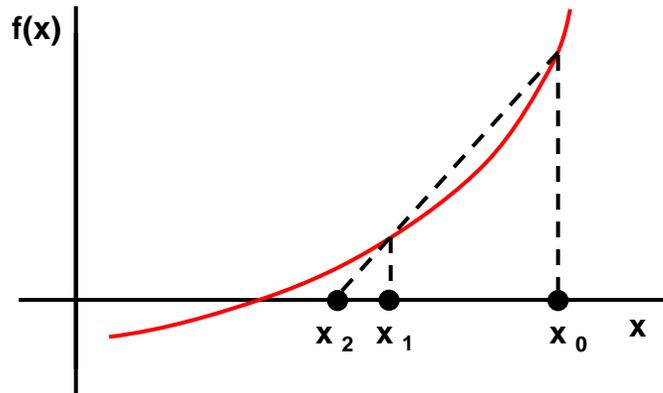


- Start with two initial guess x_{-1} and x_0 .

$$f'(x_0) \approx \frac{f(x_{-1}) - f(x_0)}{x_{-1} - x_0}$$

Use this in the equation of NR method.

$$x_1 = x_0 - \frac{f(x_0)(x_{-1} - x_0)}{f(x_{-1}) - f(x_0)}$$



- Continue like this until you reach the specified tolerance or maximum number of iterations.

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

Secant Method (cont'd)

Example 10: Repeat the same example (Find the square root of 11).

Start with $x_{-1} = 2, x_0 = 3$

$$f(x) = x^2 - 11 = 0$$

iter	x	f(x)	$ \varepsilon_t $ %
-1	2	-7	-311
0	3	-2	-160
1	3.4	0.56	2.51
2	3.3125	-0.0273438	0.12
3	3.31657356	-0.0003398	0.0015
4	3.31662482	0.0000002	0.0000

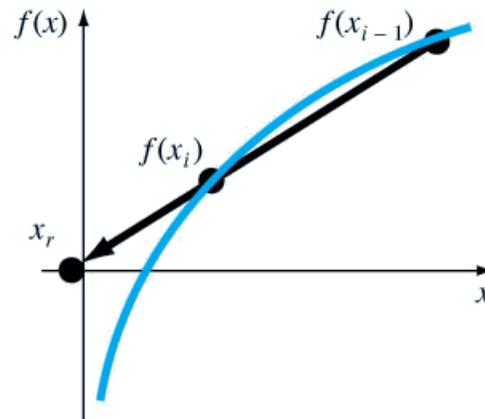
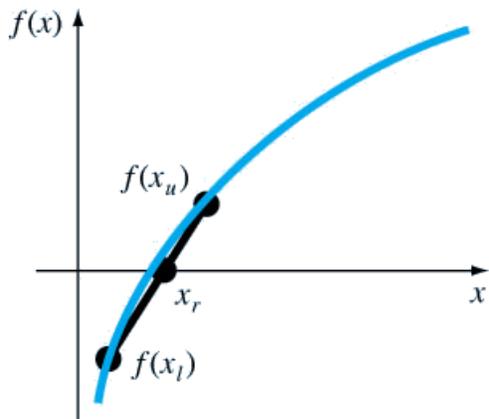
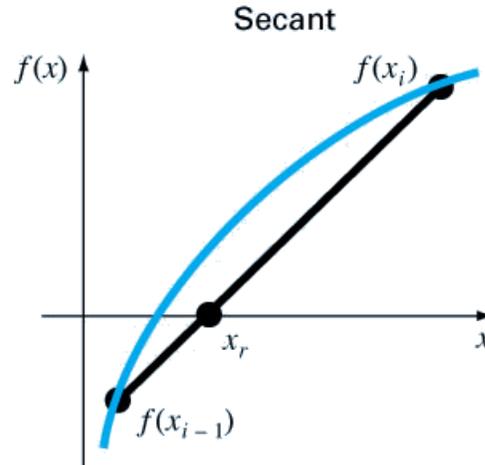
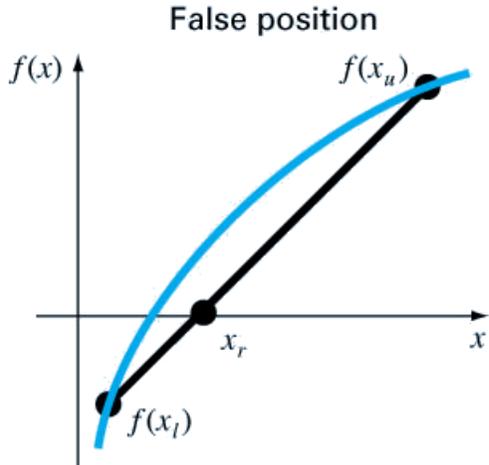
Selection of the initial guess affects the convergence a lot.

Exercise 10: There is no guarantee for the secant method to converge. Try to calculate the root of $\ln(x)$ starting with (a) $x_{-1} = 0.5$ and $x_0 = 4$, (b) $x_{-1} = 0.5$ and $x_0 = 5$. Part (a) converges, but not part (b)

Secant vs. False Position

Secant:
$$\mathbf{x}_{i+1} = \mathbf{x}_i - \frac{\mathbf{f}(\mathbf{x}_i)(\mathbf{x}_{i-1} - \mathbf{x}_i)}{\mathbf{f}(\mathbf{x}_{i-1}) - \mathbf{f}(\mathbf{x}_i)}$$

False-Position:
$$\mathbf{x} = \mathbf{x}_U - \frac{\mathbf{f}(\mathbf{x}_U)(\mathbf{x}_L - \mathbf{x}_U)}{\mathbf{f}(\mathbf{x}_L) - \mathbf{f}(\mathbf{x}_U)}$$



First iterations of both methods are the same.

Second iterations are different in terms of how the previous estimates are replaced with the newly calculated root.

- False-position Method drops one of previous estimates so that the remaining ones bracket the root.
- Secant Method always drops the oldest estimate.

Special Treatment of Multiple Roots

- At even multiple roots, bracketing methods can not be used at all.
- Open methods still work but
 - $f'(x)$ also goes to zero at a multiple root. Possibility of division by zero for Secant and NR. $f(x)$ will reach zero faster than $f'(x)$, therefore use a zero-check for $f(x)$ and stop properly.
 - they converge slowly (linear instead of quadratic convergence).
- Some modifications can be made for speed up.

i) If you know the multiplicity of the root NR can be modified as

$$x_{i+1} = x_i - m \frac{f(x_i)}{f'(x_i)} \quad m=2 \text{ for a double root, } m=3 \text{ for a triple root, etc.}$$

ii) Another alternative is to define a new function $u(x)=f(x)/f'(x)$ and use it in the formulation of NR Method

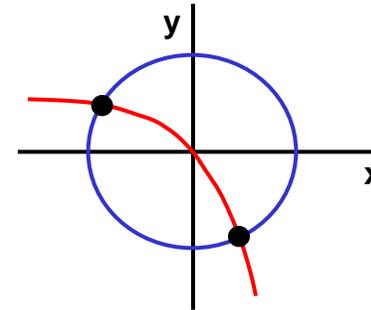
$$x_{i+1} = x_i - \frac{u(x_i)}{u'(x_i)} = \frac{f(x_i)f'(x_i)}{[f'(x_i)]^2 - f(x_i)f''(x_i)} \quad \text{You need to know } f''(x)$$

(Similar modifications can be made for the Secant Method. See the book for details.)

Solving System of Nonlinear Equations

Using Simple One-Point Iteration Method

Solve the following system of equations

$$\begin{aligned}x^2 + y^2 &= 4 \\e^x + y &= 1\end{aligned}$$


Put the functions into the form $x=g_1(x,y)$, $y=g_2(x,y)$

$$x = g_1(y) = \ln(1 - y)$$

$$y = g_2(x) = -\sqrt{4 - x^2}$$

Select a starting values for x and y , such as $x_0=0.0$ and $y_0=0.0$. They don't need to satisfy the equations. Use these values in g functions to calculate new values.

$$x_1 = g_1(y_0) = 0$$

$$y_1 = g_2(x_1) = -2$$

$$x_2 = g_1(y_1) = 1.098612289$$

$$y_2 = g_2(x_2) = -1.67124236$$

$$x_3 = g_1(y_2) = 0.982543669$$

$$y_3 = g_2(x_3) = -1.74201261$$

$$x_4 = g_1(y_3) = 1.00869218$$

$$y_4 = g_2(x_4) = -1.72700321$$

The solution is converging to the exact solution of $x=1.004169$, $y=-1.729637$

Exercise 11: Solve the same system but rearrange the equations as $x=\ln(1-y)$ $y = (4-x^2)/y$ and start from $x_0=1$ $y_0=-1.7$. Remember that this method may diverge.

Solving System of Nonlinear Equations

Using Newton-Raphson Method

Consider the following general form of a two equation system

$$\begin{aligned} \mathbf{u}(\mathbf{x}, \mathbf{y}) &= \mathbf{0} \\ \mathbf{v}(\mathbf{x}, \mathbf{y}) &= \mathbf{0} \end{aligned}$$

Write 1st order TSE for these equations

$$\mathbf{u}_{i+1} = \mathbf{u}_i + \frac{\partial \mathbf{u}_i}{\partial \mathbf{x}} (\mathbf{x}_{i+1} - \mathbf{x}_i) + \frac{\partial \mathbf{u}_i}{\partial \mathbf{y}} (\mathbf{y}_{i+1} - \mathbf{y}_i)$$

$$\mathbf{v}_{i+1} = \mathbf{v}_i + \frac{\partial \mathbf{v}_i}{\partial \mathbf{x}} (\mathbf{x}_{i+1} - \mathbf{x}_i) + \frac{\partial \mathbf{v}_i}{\partial \mathbf{y}} (\mathbf{y}_{i+1} - \mathbf{y}_i)$$

To find the solution set $\mathbf{u}_{i+1} = 0$ and $\mathbf{v}_{i+1} = 0$. Rearrange

$$\frac{\partial \mathbf{u}_i}{\partial \mathbf{x}} \mathbf{x}_{i+1} + \frac{\partial \mathbf{u}_i}{\partial \mathbf{y}} \mathbf{y}_{i+1} = -\mathbf{u}_i + \frac{\partial \mathbf{u}_i}{\partial \mathbf{x}} \mathbf{x}_i + \frac{\partial \mathbf{u}_i}{\partial \mathbf{y}} \mathbf{y}_i$$

$$\frac{\partial \mathbf{v}_i}{\partial \mathbf{x}} \mathbf{x}_{i+1} + \frac{\partial \mathbf{v}_i}{\partial \mathbf{y}} \mathbf{y}_{i+1} = -\mathbf{v}_i + \frac{\partial \mathbf{v}_i}{\partial \mathbf{x}} \mathbf{x}_i + \frac{\partial \mathbf{v}_i}{\partial \mathbf{y}} \mathbf{y}_i$$

$$\begin{bmatrix} \frac{\partial \mathbf{u}_i}{\partial \mathbf{x}} & \frac{\partial \mathbf{u}_i}{\partial \mathbf{y}} \\ \frac{\partial \mathbf{v}_i}{\partial \mathbf{x}} & \frac{\partial \mathbf{v}_i}{\partial \mathbf{y}} \end{bmatrix} \begin{Bmatrix} \mathbf{x}_{i+1} \\ \mathbf{y}_{i+1} \end{Bmatrix} = \begin{Bmatrix} -\mathbf{u}_i + \frac{\partial \mathbf{u}_i}{\partial \mathbf{x}} \mathbf{x}_i + \frac{\partial \mathbf{u}_i}{\partial \mathbf{y}} \mathbf{y}_i \\ -\mathbf{v}_i + \frac{\partial \mathbf{v}_i}{\partial \mathbf{x}} \mathbf{x}_i + \frac{\partial \mathbf{v}_i}{\partial \mathbf{y}} \mathbf{y}_i \end{Bmatrix}$$

The first matrix is called the Jacobian matrix.

Solve for \mathbf{x}_{i+1} , \mathbf{y}_{i+1} and iterate.

Can be generalized for n simultaneous equations

Solving System of Eqns Using NR Method (cont'd)

Solve for x_{i+1} and y_{i+1} using Cramer's rule (ME 210)

$$x_{i+1} = x_i - \frac{u_i \frac{\partial v_i}{\partial y} - v_i \frac{\partial u_i}{\partial y}}{\frac{\partial u_i}{\partial x} \frac{\partial v_i}{\partial y} - \frac{\partial u_i}{\partial y} \frac{\partial v_i}{\partial x}}, \quad y_{i+1} = y_i - \frac{v_i \frac{\partial u_i}{\partial x} - u_i \frac{\partial v_i}{\partial x}}{\frac{\partial u_i}{\partial x} \frac{\partial v_i}{\partial y} - \frac{\partial u_i}{\partial y} \frac{\partial v_i}{\partial x}}$$

The denominator is the determinant of the Jacobian matrix $|J|$.

Example 11: Solve the same system of equations $u = x^2 + y^2 - 4 = 0$ stating with $x_0=1, y_0=1$
 $v = e^x + y - 1 = 0$

$$\frac{\partial u}{\partial x} = 2x, \quad \frac{\partial u}{\partial y} = 2y, \quad \frac{\partial v}{\partial x} = e^x, \quad \frac{\partial v}{\partial y} = 1 \rightarrow |J| = 2x - 2ye^x$$

$$x_{i+1} = x_i - \frac{u_i - 2y_i v_i}{|J|}, \quad y_{i+1} = y_i - \frac{2xv_i - e^{x_i} u_i}{|J|}$$

$i=0, x_0=1$	$y_0=1$	$u_0=-2$	$v_0=2.718282$	$ J_0 =-3.436564$
$i=1, x_1=-1.163953$	$y_1=3.335387$	$u_1=8.479595$	$v_1=2.647636$	$ J_1 =-4.410851$
$i=2, x_2=-3.245681$	$y_2=1.337769$	$u_2=8.324069$	$v_2=0.376711$	$ J_2 =-6.595552$
$i=3, x_3=-2.136423$	$y_3=0.959110$	$u_3=1.484197$	$v_3=0.077187$	$ J_3 =-4.499343$

Looks like it is converging to the root in the 2nd quadrant $x \approx -1.8, y \approx 0.8$.

Exercise 12: Can you start with $x_0=0$ and $y_0=0$?

Exercise 13: Try to find starting points that will converge to the solution in the 4th quadrant.