

# Class Progress

Basics of Linux, gnuplot, C

Visualization of numerical data

## **Roots of nonlinear equations**

(Midterm 1)

Solutions of systems of linear equations

Solutions of systems of nonlinear equations

Monte Carlo simulation

Interpolation of sparse data points

Numerical integration

(Midterm 2)

Solutions of ordinary differential equations

# General Problem of Nonlinear Equation Roots

Find  $x$  for which  $f(x)=0$

Examples:

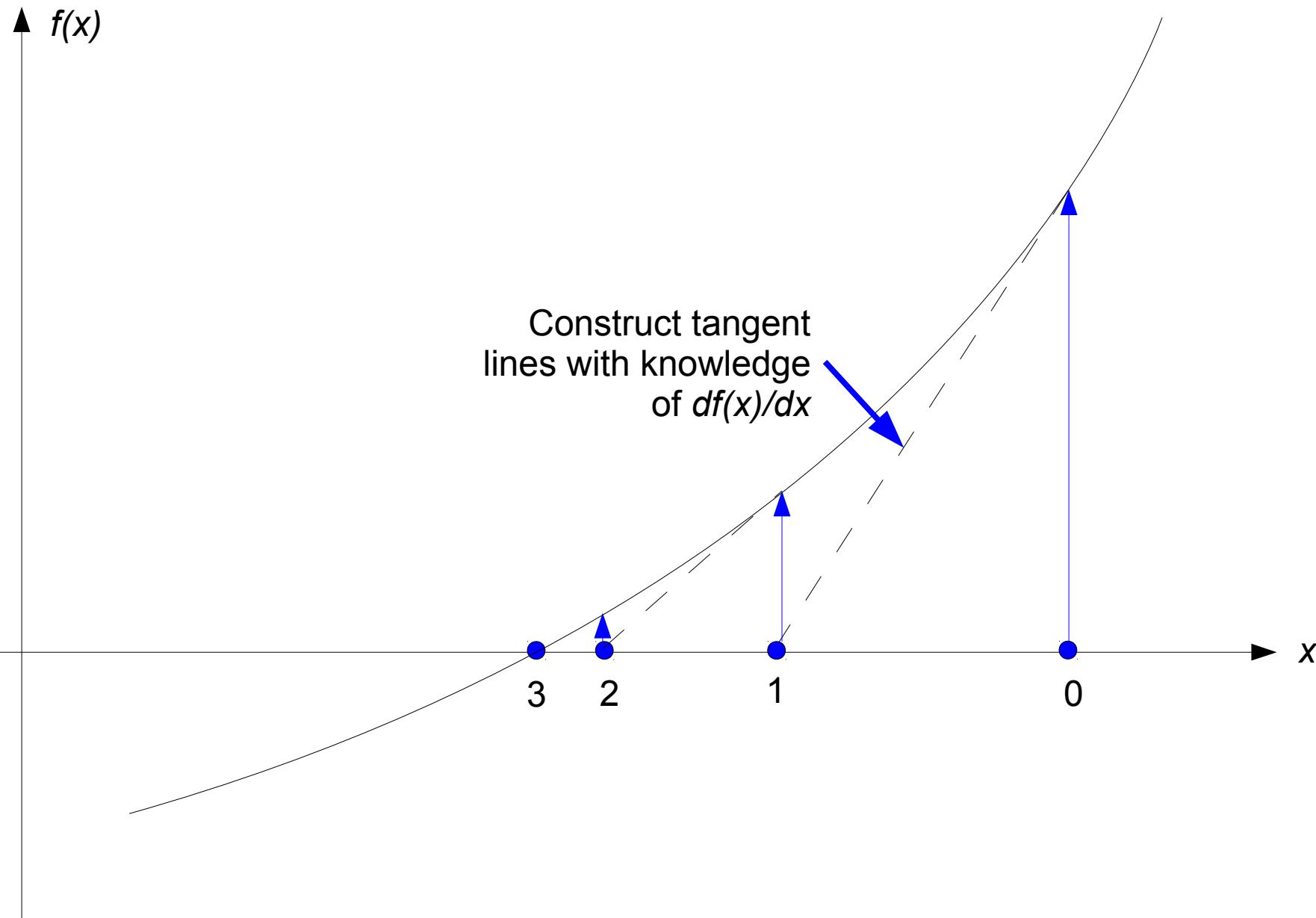
$$x^6 + 11x^5 - x^4 + 14x^3 - 23x^2 + 7x - 3 = 0$$

$$\frac{e^{k_1 x} + e^{-k_2 x}}{8} - \frac{1}{4} = 0$$

$$\sin(3x) - \cos(5x) - \sqrt{7x} + 13 = 0$$

Algorithms for finding roots are iterative, that is, they start at one or two initial guess points,  $x_0$  or  $x_{-1}$  and  $x_0$ , and then iterate to give a series of successive values of  $x_i$ . Hopefully the sequence of  $x_i$  will converge on the root!

# Newton-Raphson Algorithm for Root Finding



# Newton-Raphson Algorithm

given  $f(x)$  and an initial guess  $x_0$

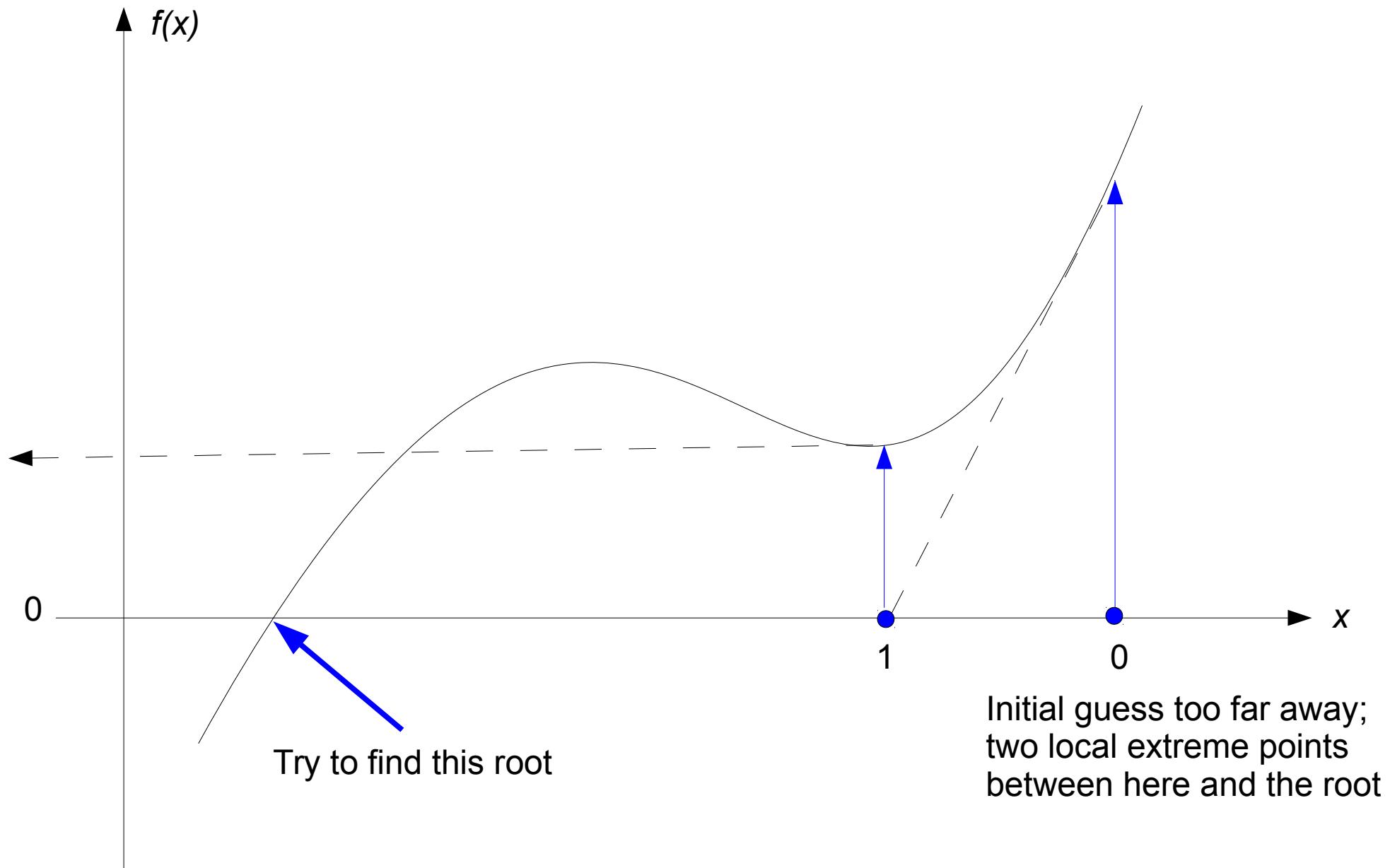
$$\frac{f(x_i)}{x_i - x_{i+1}} = f'(x_i)$$

where  $x_i$  is the trial value of  $x$  after iteration  $i$

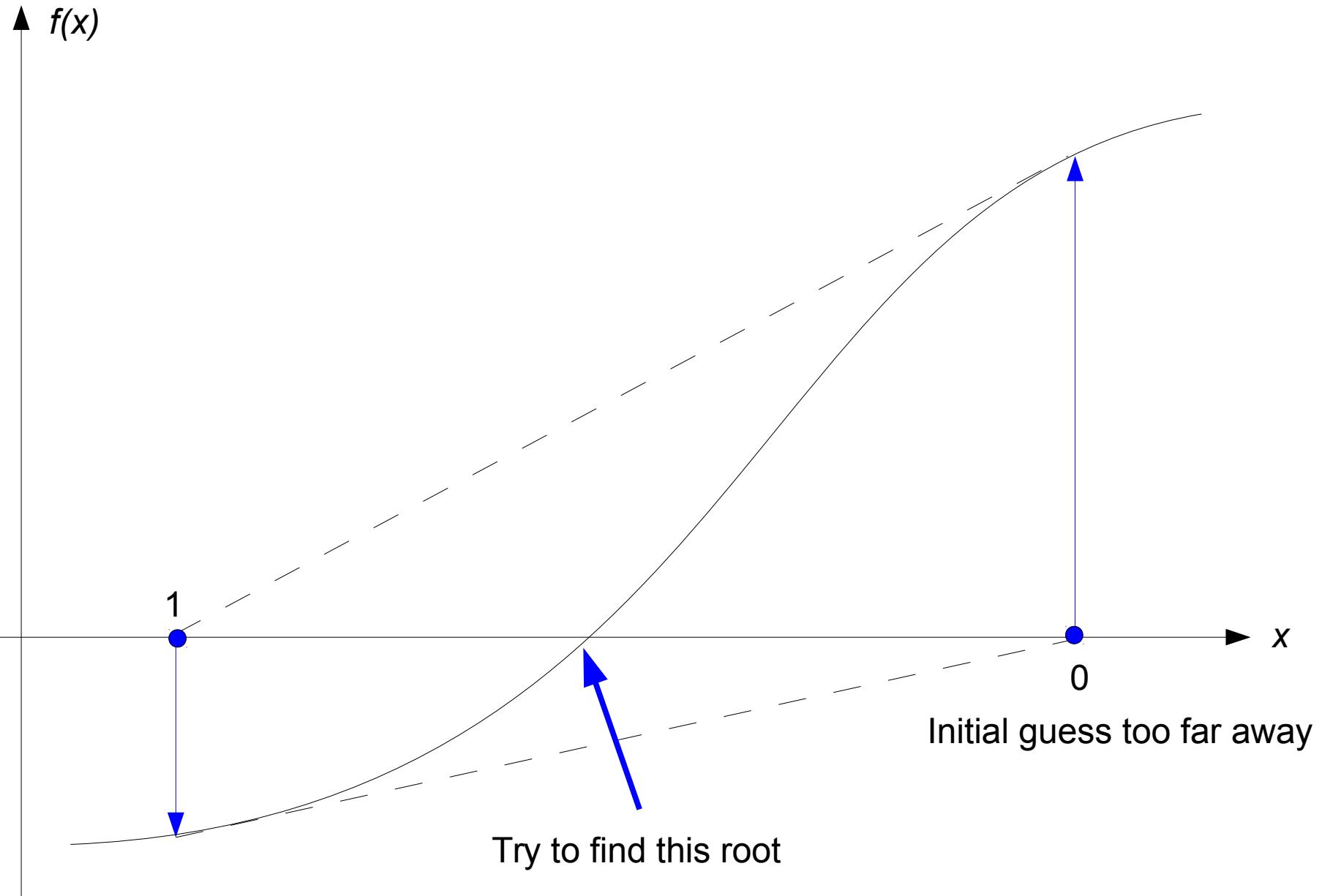
$$x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)}$$

But what about local extreme points when  $f'(x_i)$  is near zero?

# Divergence of Newton-Raphson Algorithm



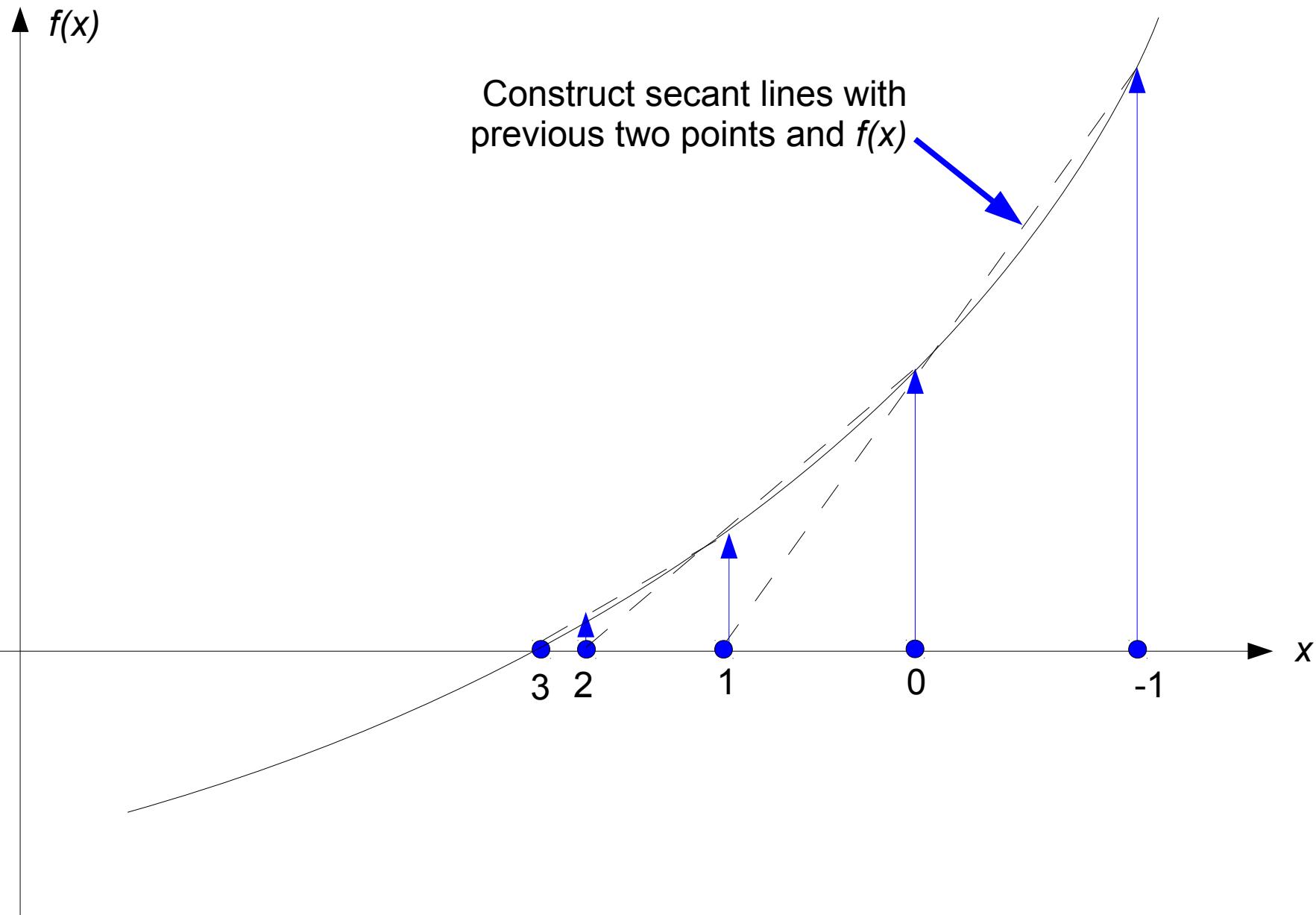
# Oscillation of Newton-Raphson Algorithm



# Newton-Raphson Algorithm

- Requires coding of both the function being solved and its derivative
- Requires any single point close to a root to start
- Minimal storage requirements for each iteration
- Usually will converge if starting point is sufficiently close to root, but also easily divergent for sharply nonlinear functions around local extrema
- Fastest convergence rate

# Secant Algorithm for Root Finding



# Secant Algorithm

given  $f(x)$  and initial guesses  $x_0$  and  $x_{-1}$

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

where  $x_i$  is the trial value of  $x$  after iteration  $i$

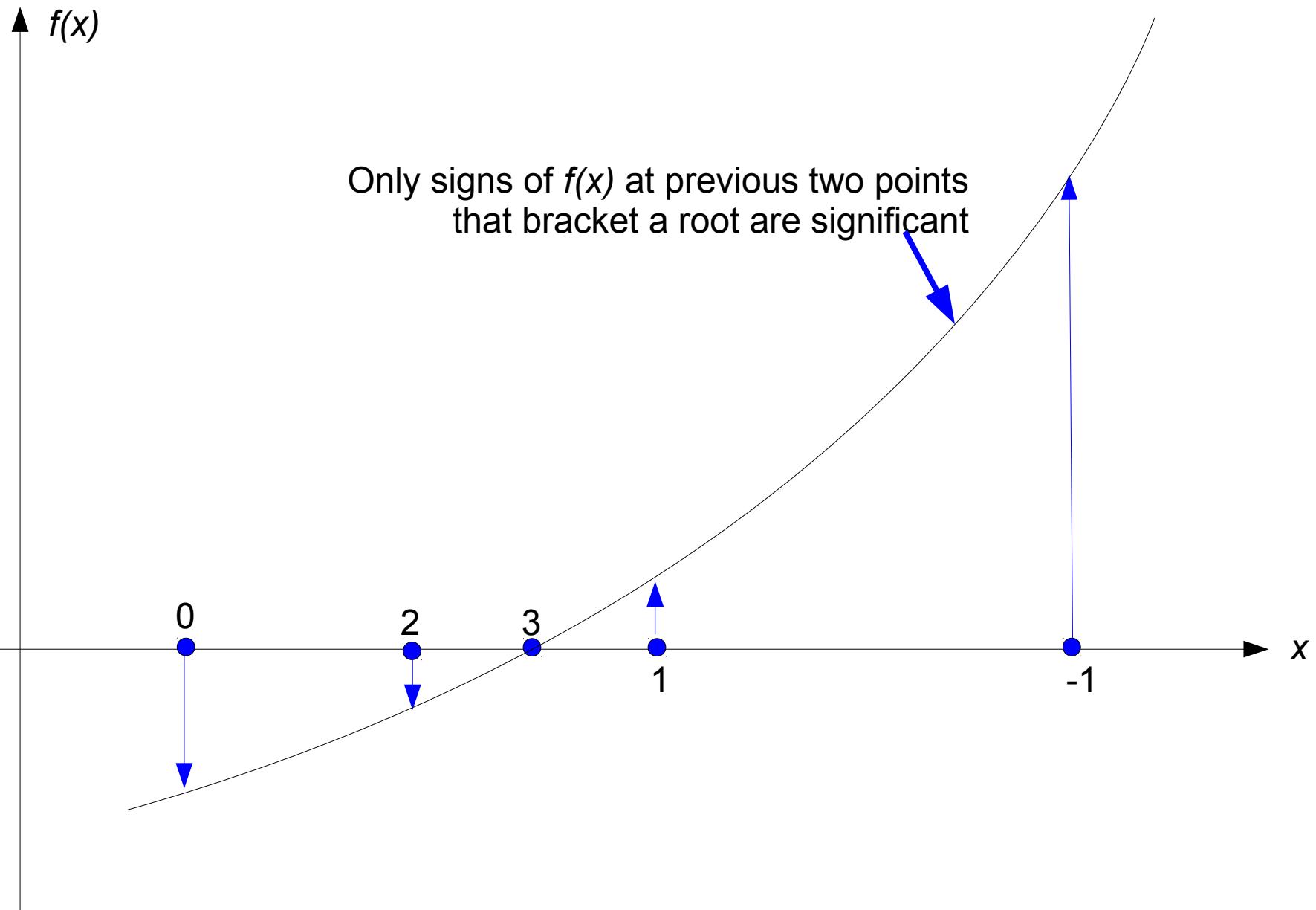
$$x_{i+1} = x_i - \frac{x_i - x_{i-1}}{f(x_i) - f(x_{i-1})} \cdot f(x_i)$$

But what about when  $f(x_{i-1}) - f(x_i)$  is near zero?

# Secant Algorithm

- Requires any two points close to a root to start
- Coding of derivative not required
- Some small additional storage requirements over Newton-Raphson, significant when systems of nonlinear equations are to be solved
- Usually will converge if starting points are sufficiently close to root
- Sharply nonlinear functions, or starting points too far from root, can cause divergence
- Fast convergence rate; useful if function is expensive to evaluate

# Bisection Algorithm for Root Finding



# Bisection Algorithm

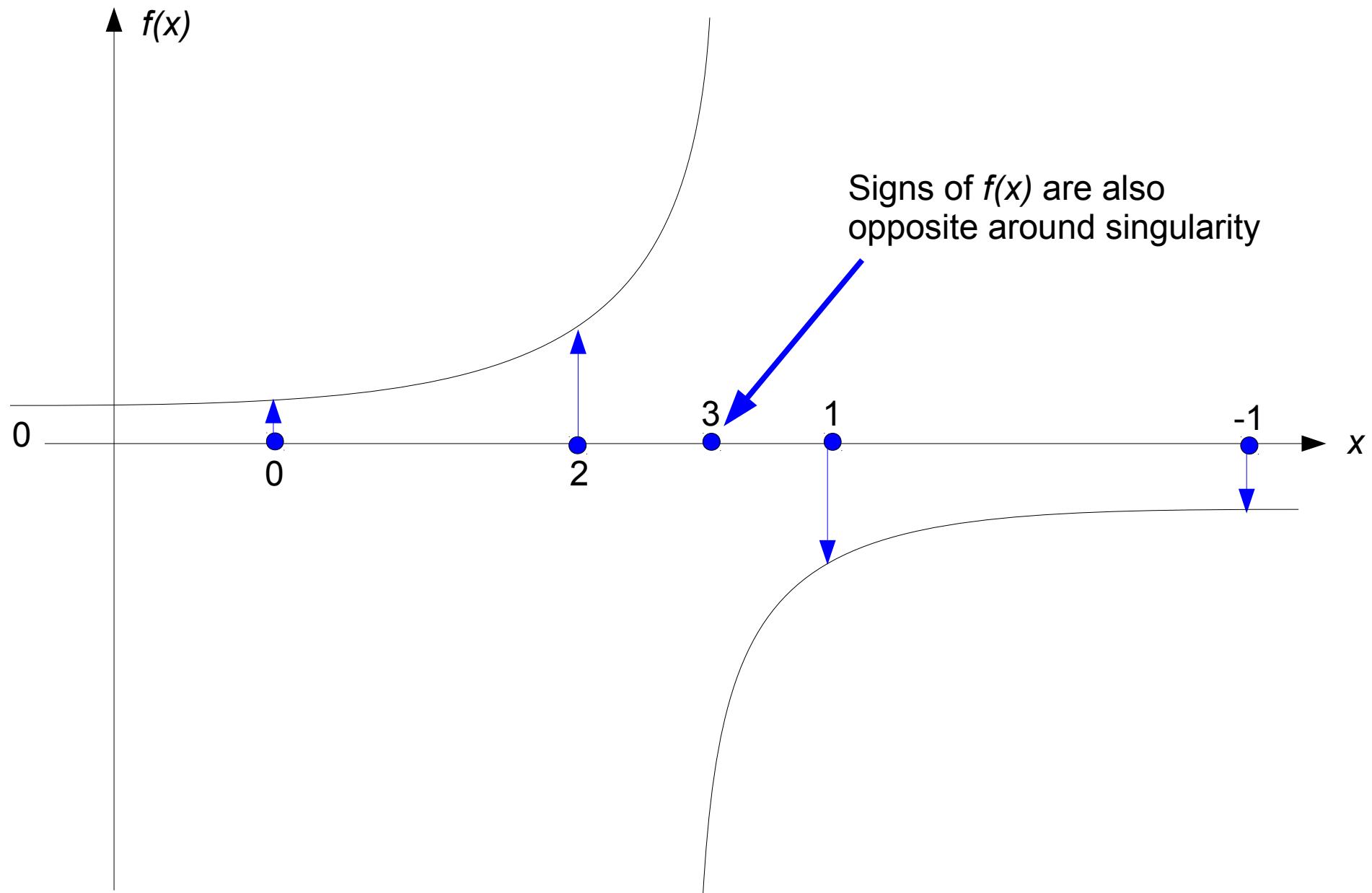
given  $f(x)$  and initial guesses  $x_0$  and  $x_{-1}$  that **must bracket the root**

$$x_{i+1} = \frac{x_i + x_{i-1}}{2}$$

where  $x_i$  is the trial value of  $x$  after iteration  $i$

For the two points used in the next iteration, retain  $x_{i+1}$  and either  $x_i$  or  $x_{i-1}$  such that the root remains bracketed

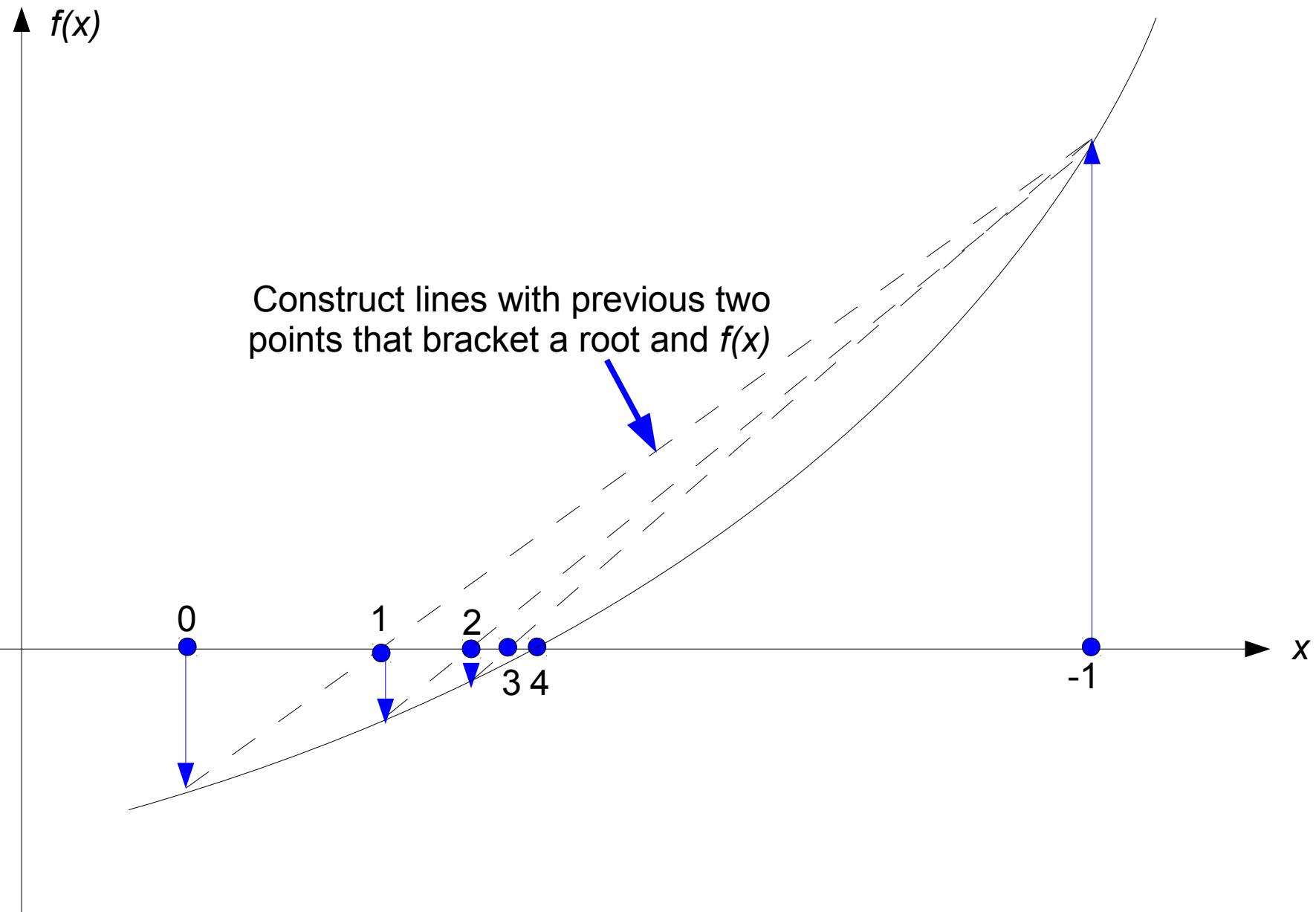
# Bisection Algorithm Stuck at Singularities



# Bisection Algorithm

- Requires two points that bracket a root to start, not necessarily very close to root
- Must converge on a root or a singularity, even for sharply nonlinear functions
- May be trapped by a singularity in the function
- Slowest convergence rate

# Regula Falsi Algorithm for Root Finding



# Regula Falsi Algorithm

given  $f(x)$  and initial guesses  $x_0$  and  $x_{-1}$  that **must bracket the root**

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

where  $x_i$  is the trial value of  $x$  after iteration  $i$

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

Look familiar? This is the same expression as for the secant algorithm!

For the two points used in the next iteration, retain  $x_{i+1}$  and

either  $x_i$  or  $x_{i-1}$  such that the root remains bracketed

Another version of this expression that is computationally more balanced is

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

# Regula Falsi Algorithm

- Requires two points that bracket a root to start
- New trial point is calculated identically with the Secant Algorithm
- As algorithm proceeds, the two active points maintain bracketing, so almost always will converge to a root
- Moderate convergence rate
- Good first choice algorithm for real world functions with unknown nonlinearity

# Root Finding Algorithms

Different vehicles are appropriate for different road surfaces; different root finding algorithms are appropriate for different function “surfaces”



*Robustness*



Bisection



*Convergence rate*



Regula Falsi

Secant

Newton-Raphson

# Convergence Rates of Root Finding Algorithms

given  $f(x)$  where  $r$  is the desired root, so  $f(r)=0$   
and the sequence of  $x$  values determined by the algorithm is  $x_0, x_1, x_2, \dots, x_n$

For Bisection:  $|x_{n+1} - r| \leq C |x_n - r|$

For Secant:  $|x_{n+1} - r| \leq C |x_n - r|^\alpha$  where  $\alpha \approx 1.62$

For Newton-Raphson:  $|x_{n+1} - r| \leq C |x_n - r|^2$

For Regula Falsi:  $|x_{n+1} - r| \leq C |x_n - r|^\alpha$  where  $1 < \alpha < 1.62$

# Convergence Rate of Newton-Raphson

$$\text{New error } (x_{n+1} - r) = \left[ x_n - \frac{f(x_n)}{f'(x_n)} \right] - r = \frac{(x_n - r)f'(x_n) - f(x_n)}{f'(x_n)}$$

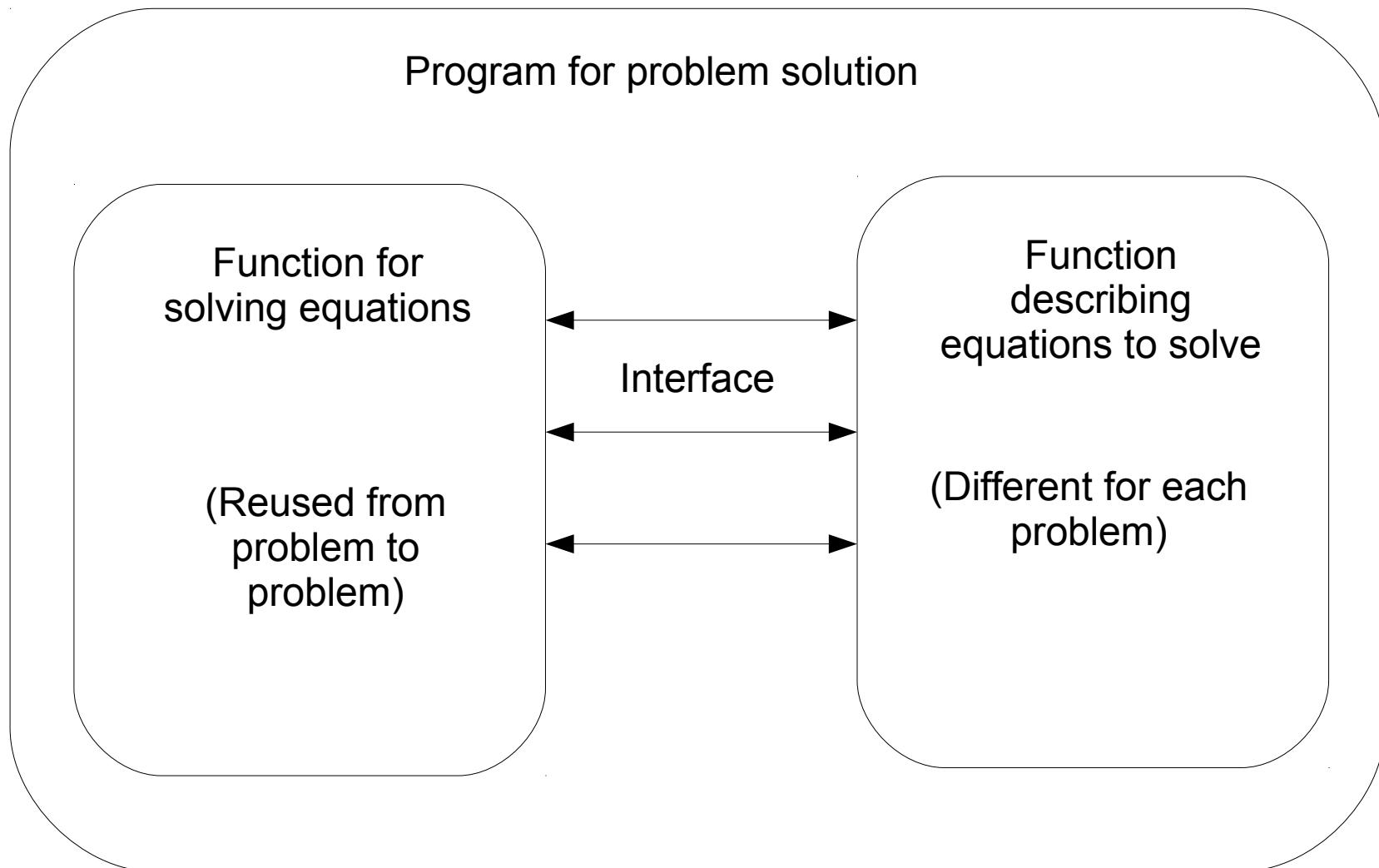
$$\text{From Taylor series } f(r) = 0 = f(x_n) - (x_n - r)f'(x_n) + \frac{(x_n - r)^2}{2}f''(\xi_n)$$

$$\text{or } (x_n - r)f'(x_n) - f(x_n) = \frac{(x_n - r)^2}{2}f''(\xi_n)$$

where  $\xi_n$  lies between  $r$  and  $x_n$

$$\text{plug into new error expression: } (x_{n+1} - r) = \frac{f''(\xi_n)}{2f'(x_n)}(x_n - r)^2$$

# Reusing Program Code



# Newton-Raphson Function in roots.c

```
int newton_raphson(double (*func) (double x), double (*deriv) (double x), double
x, int nmax, double tol, double min) {
    double fx, fp, d;
    int n;

    fx = (*func) (x);
    printf("0 %.8g %.8g\n", x, fx);
    n = 1;
    while (n <= nmax) {
        fp = (*deriv) (x);
        if (fabs(fp) < min) {
            fprintf(stderr, "newton_raphson: Derivative too small\n");
            return(2);
        }
        d = fx / fp;
        x = x - d;
        fx = (*func) (x);
        printf("%d %.8g %.8g\n", n, x, fx);
        if (fabs(d) < tol) break;
        n++;
    }
    if (n > nmax) {
        fprintf(stderr, "newton_raphson: Iteration limit of %d reached\n", nmax);
        return(1);
    }
    return(0);
}
```



# Bisection Function in roots.c

```
int bisection(double (*func) (double x), double a, double b, int nmax, double tol) {
    double fa, fb, c, fc, d;
    int n;

    fa = (*func)(a);
    fb = (*func)(b);
    if (fa * fb > 0.0) {
        fprintf(stderr, "bisection: Root not bracketed by initial points\n");
        return(2);
    }
    printf("0 %.8g %.8g\n", b, fb);
    printf("1 %.8g %.8g\n", a, fa);
    n = 2;
    while (n <= nmax) {
        d = 0.5 * (b - a);
        if (fabs(d) < tol) break;
        c = a + d;
        fc = (*func)(c);
        printf("%d %.8g %.8g\n", n, c, fc);
        if (fa * fc < 0.0) {
            b = c;
            fb = fc;
        }
        else {
            a = c;
            fa = fc;
        }
        n++;
    }
    if (n > nmax) {
        fprintf(stderr, "bisection: Iteration limit of %d reached\n", nmax);
        return(1);
    }
    return(0);
}
```

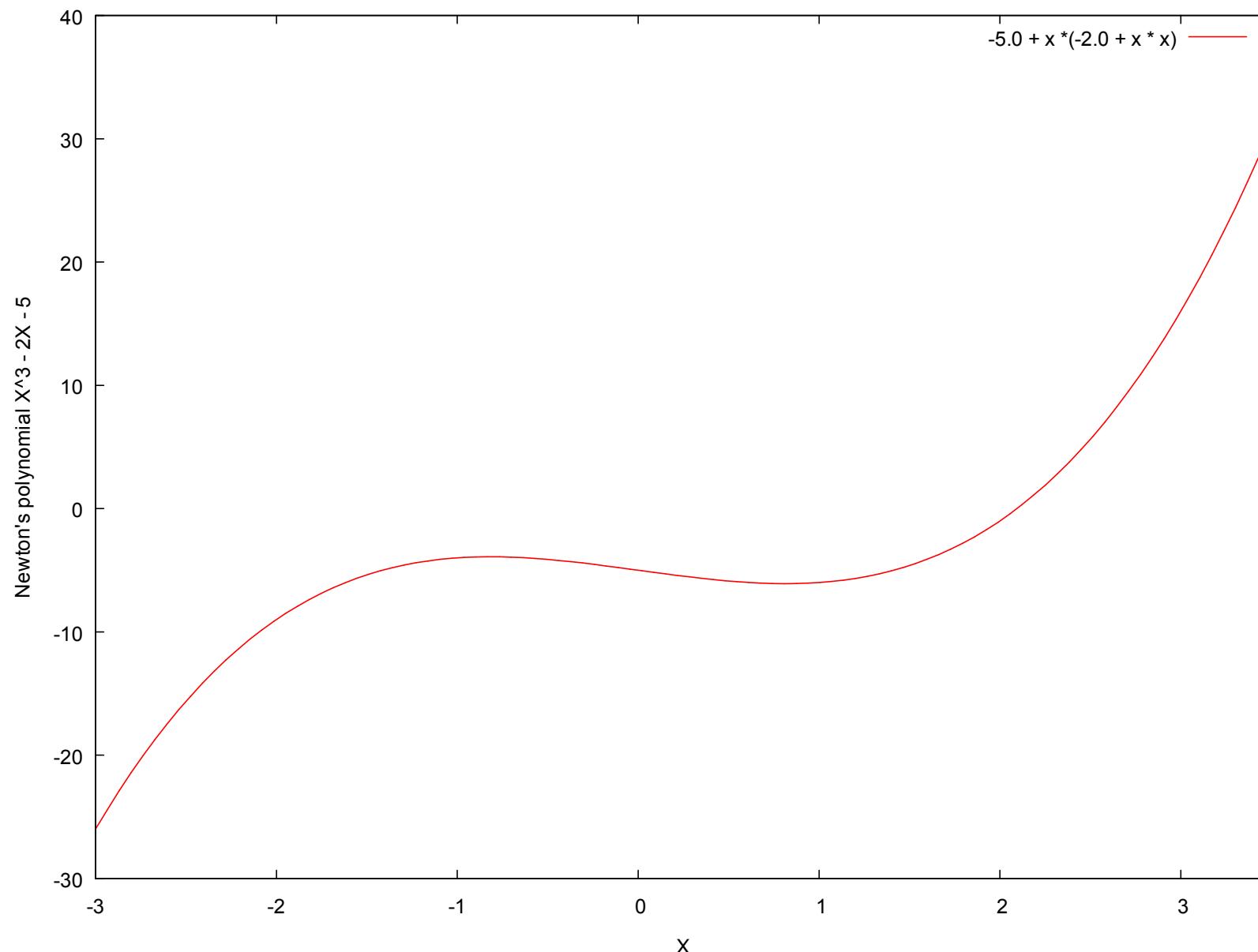


# Header file roots.h

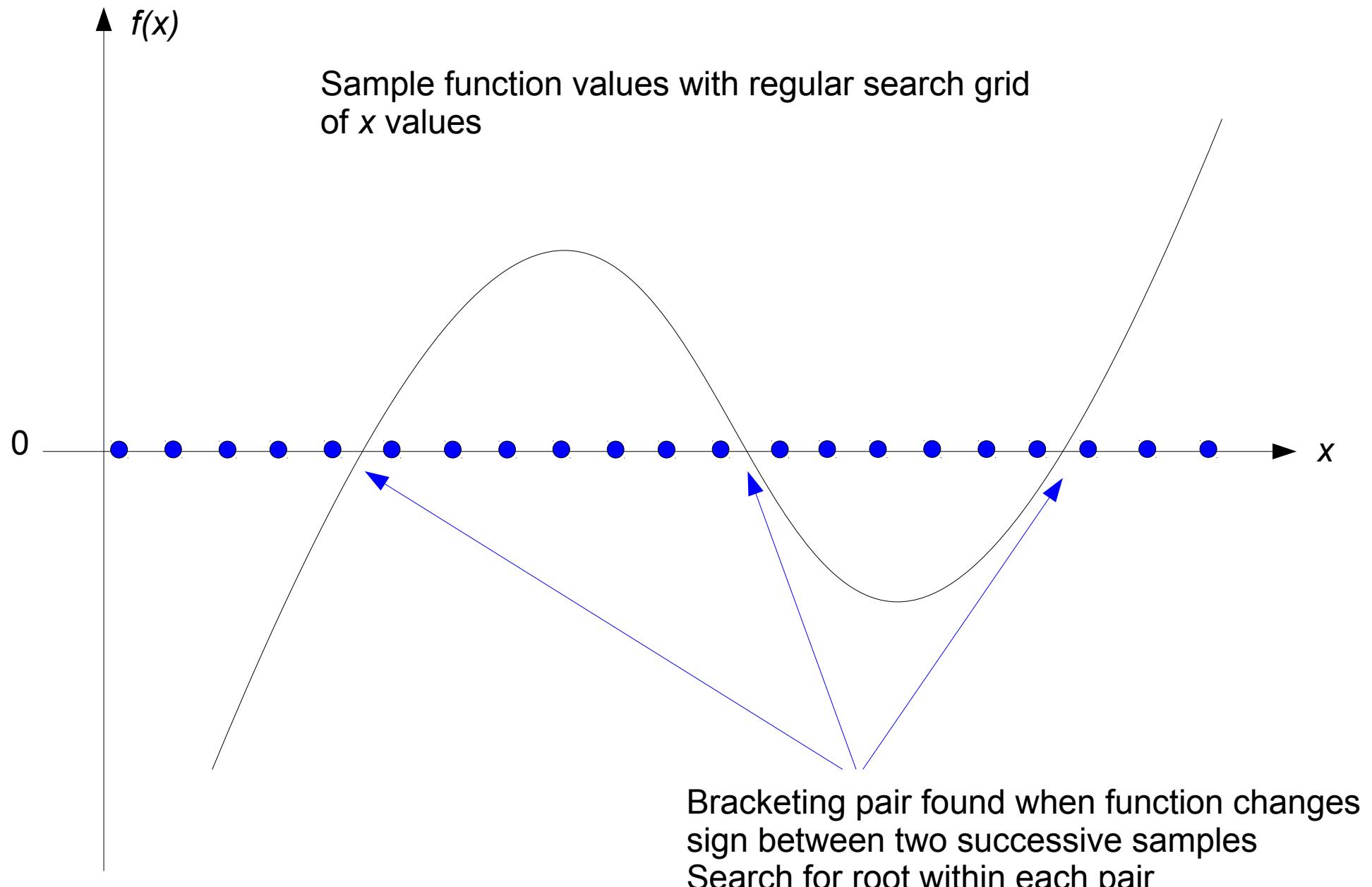
```
int newton_raphson(double (*func) (double x), double (*deriv) (double x), double  
x, int nmax, double tol, double min);  
int regula_falsi(double (*func) (double x), double a, double b, int nmax, double  
tol);  
int bisection(double (*func) (double x), double a, double b, int nmax, double tol);
```

# Newton's Polynomial for Root Finding

See Cheney & Kincaid Computer Problem 3.2.5

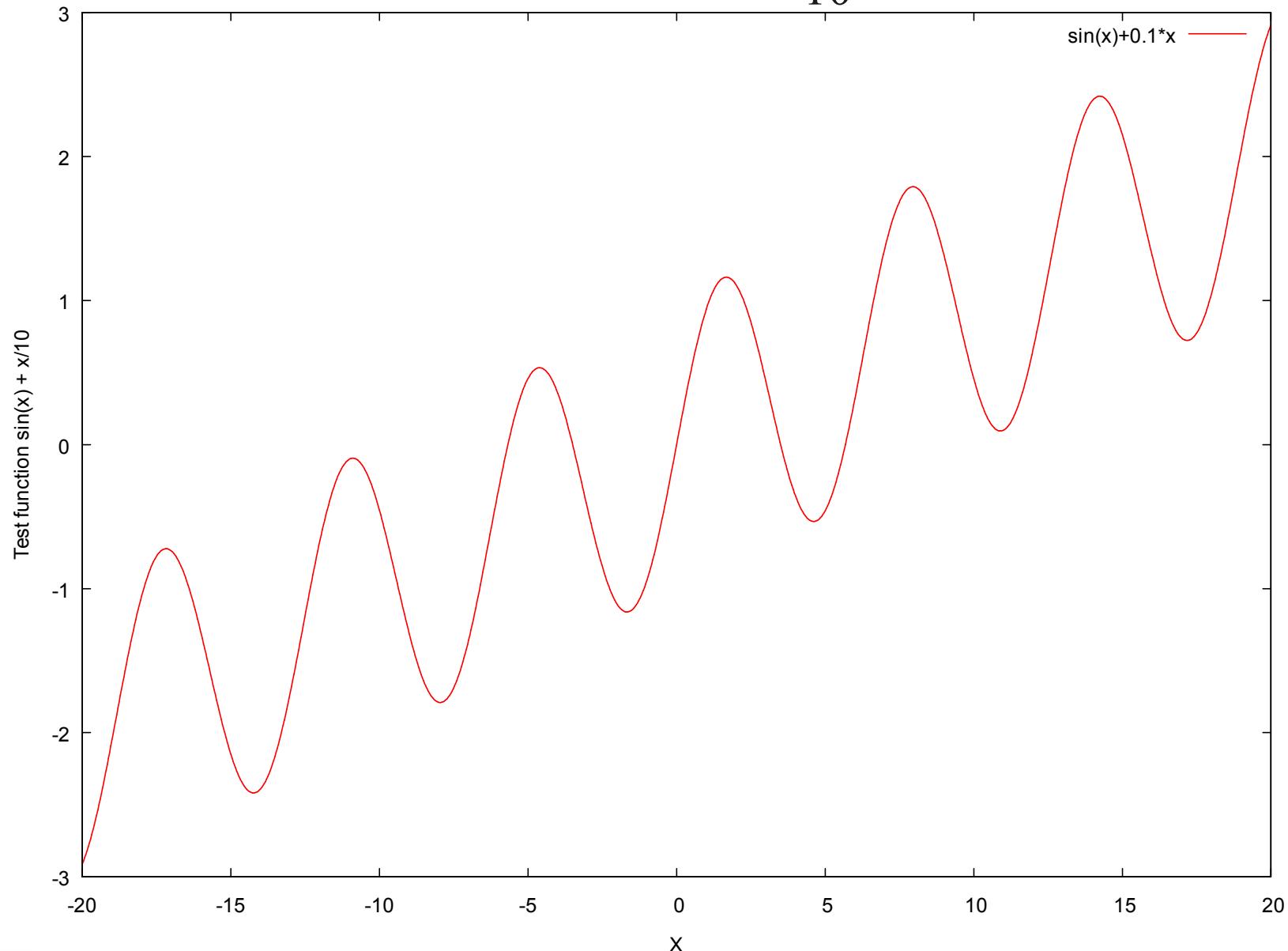


# Finding Initial Pairs of Bracketing Points



# Test Function for Root Finding

$$f(x) = \sin(x) + \frac{x}{10}$$



# Bisection Functions in sweep\_roots.c

```
int bisection(double (*func)(double x), double a, double b, int nmax, double tol) {
    ... (same as in roots.c except without printf() calls)
}

int sweep_bisection(double (*func)(double x), double xstart, double xstop, double
xinc, int nmax, double tol) {
    double a,b,fa,fb;
    int retval;

    xstop = xstop + (xinc * 0.5);
    a = xstart;
    fa = (*func)(a);
    b = a + xinc;
    while (((xinc > 0.0) && (b < xstop)) || ((xinc < 0.0) && (b > xstop))) {
        fb = (*func)(b);
        if ((fa * fb) < 0.0) { /*root bracketed, converge with bisection*/
            retval = bisection(func,a,b,nmax,tol);
            if (retval > 0) return(retval);
        }
        a = b;
        fa = fb;
        b = b + xinc;
    }
    return(0);
}
```

# Newton-Raphson Functions in sweep\_roots.c

```
int newton_raphson(double (*func) (double x), double (*deriv) (double x), double x, int
nmax, double tol, double min) {
    ... (same as in roots.c except without printf() calls)
}

int sweep_newton_raphson(double (*func) (double x), double (*deriv) (double x), double
xstart, double xstop, double xinc, int nmax, double tol, double min) {
    double a,b,fa,fb;
    int retval;

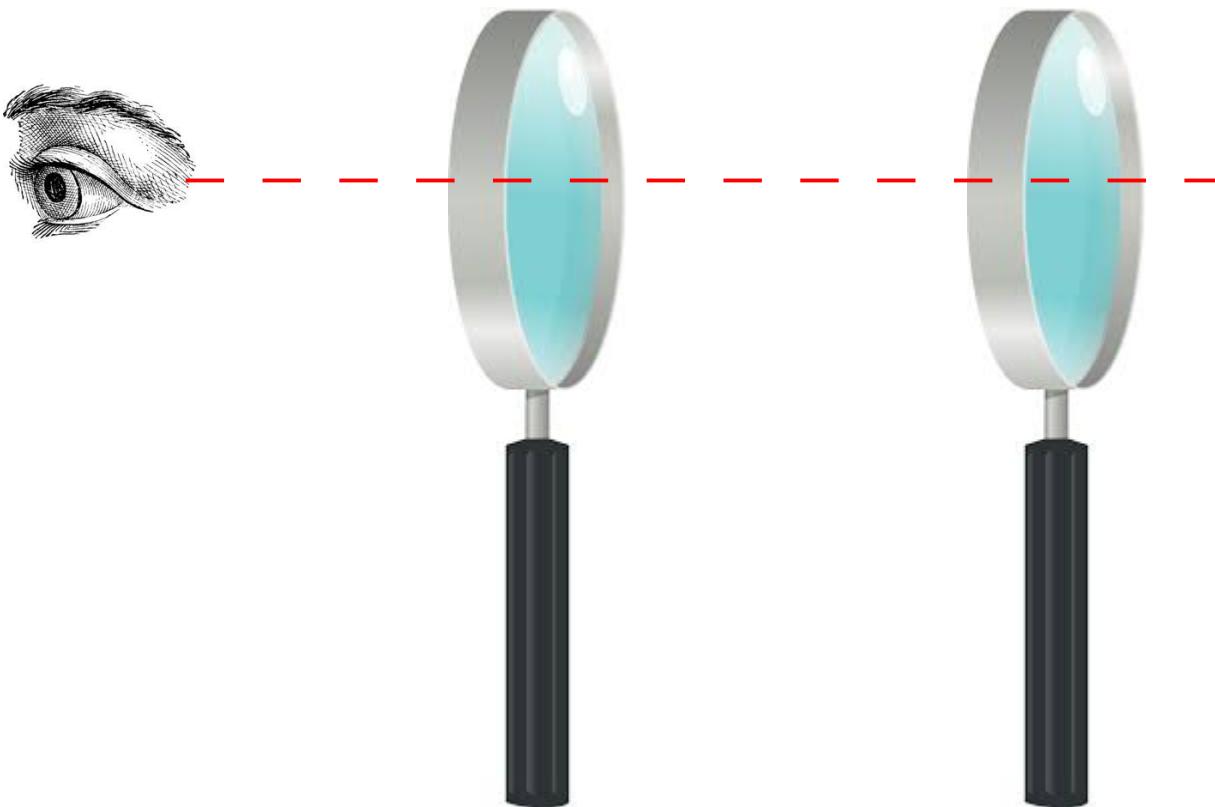
    xstop = xstop + (xinc * 0.5);
    a = xstart;
    fa = (*func) (a);
    b = a + xinc;
    while (((xinc > 0.0) && (b < xstop)) || ((xinc < 0.0) && (b > xstop))) {
        fb = (*func) (b);
        if ((fa * fb) < 0.0) { /*root bracketed, converge with newton_raphson*/
            if (fabs(fa) < fabs(fb)) {
                retval = newton_raphson(func,deriv,a,nmax,tol,min);
            }
            else {
                retval = newton_raphson(func,deriv,b,nmax,tol,min);
            }
            if (retval > 0) return(retval);
        }
        a = b;
        fa = fb;
        b = b + xinc;
    }
    return(0);
}
```



# Header file sweep\_roots.h

```
int sweep_newton_raphson(double (*func) (double x), double (*deriv) (double x), double xstart, double xstop, double xinc, int nmax, double tol, double min);  
int sweep_regula_falsi(double (*func) (double x), double xstart, double xstop, double xinc, int nmax, double tol);  
int sweep_bisection(double (*func) (double x), double xstart, double xstop, double xinc, int nmax, double tol);
```

# Solving Physics Problems Numerically

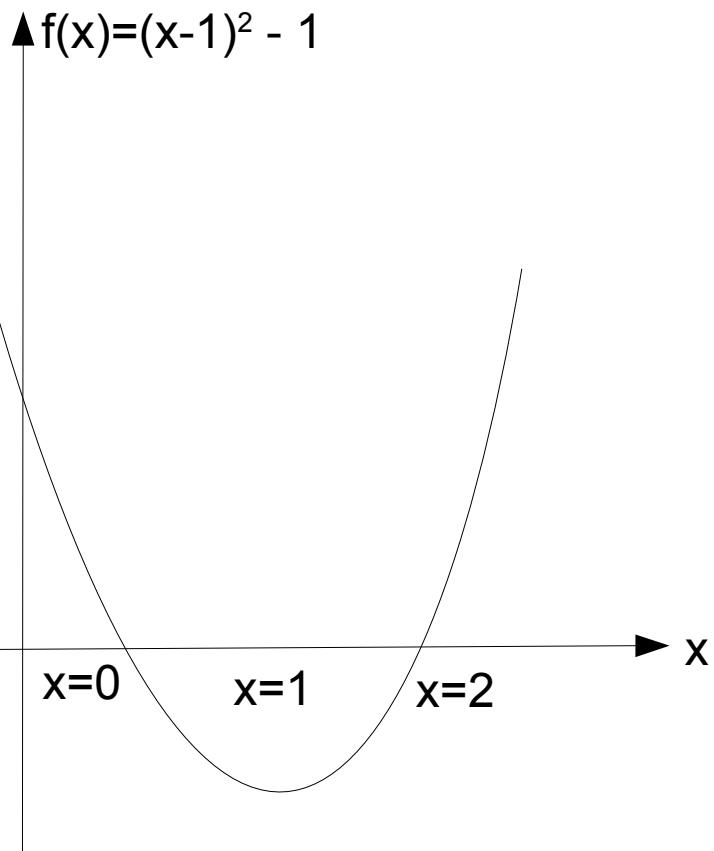


A blackboard filled with various physics equations, including:

$$\begin{aligned} D &= \frac{1}{2}mv^2 & E_k &= \frac{1}{2}mv^2 & p &= mv & D &= \frac{1}{2}mv^2 & E_k \\ |\vec{p}|^2 &= \frac{\Delta E}{\Delta t} & \vec{p} = \frac{\Delta E}{\Delta t} & \epsilon = \propto |\vec{p}|^2 & \vec{p} = \frac{\Delta E}{\Delta t} & \epsilon = \propto |\vec{p}|^2 & \vec{p} = \frac{\Delta E}{\Delta t} & \epsilon = \propto |\vec{p}|^2 & \vec{p} = \frac{\Delta E}{\Delta t} & \epsilon = \propto |\vec{p}|^2 \\ W &= PT & F &= G \frac{m_1 m_2}{r^2} & W &= PT & F &= G \frac{m_1 m_2}{r^2} & W &= PT & F &= G \frac{m_1 m_2}{r^2} \\ H &= U + PV & G &= \sigma \frac{A}{L} & F &= m \ddot{x} & H &= U + PV & G &= \sigma \frac{A}{L} & F &= m \ddot{x} \\ B & f = \frac{(c+v)}{(c-v)} & f = \frac{(c+v)}{(c-v)} & F = qv \times B & f = \frac{(c+v)}{(c-v)} & f = \frac{(c+v)}{(c-v)} & F = qv \times B & f = \frac{(c+v)}{(c-v)} & f = \frac{(c+v)}{(c-v)} & F = qv \times B \\ f & = \frac{(c+v)}{(c-v)} & f & = \frac{(c+v)}{(c-v)} & F = qv \times B & f = \frac{(c+v)}{(c-v)} & f = \frac{(c+v)}{(c-v)} & F = qv \times B & f = \frac{(c+v)}{(c-v)} & f = \frac{(c+v)}{(c-v)} & F = qv \times B \\ V & = J/C & V & = J/C \\ A + B & = A + B & A + B & = A + B & A + B & = A + B & A + B & = A + B & A + B & = A + B & A + B = A + B \\ \sigma_x \sigma_p & \geq \frac{\hbar}{2} & \sigma_x \sigma_p & \geq \frac{\hbar}{2} \\ B & = v \cdot \frac{1}{c} E & E_k & = \frac{1}{2}mv^2 & p & = mv & B & = v \cdot \frac{1}{c} E & E_k & = \frac{1}{2}mv^2 & p & = mv \\ \vec{p}^2 &= \frac{\Delta E}{\Delta t} & \vec{p} = \frac{\Delta E}{\Delta t} & \epsilon = \propto |\vec{p}|^2 & \vec{p} = \frac{\Delta E}{\Delta t} & \epsilon = \propto |\vec{p}|^2 & \vec{p} = \frac{\Delta E}{\Delta t} & \epsilon = \propto |\vec{p}|^2 & \vec{p} = \frac{\Delta E}{\Delta t} & \epsilon = \propto |\vec{p}|^2 & \vec{p} = \frac{\Delta E}{\Delta t} & \epsilon = \propto |\vec{p}|^2 \\ W &= PT & F &= G \frac{m_1 m_2}{r^2} & W &= PT & F &= G \frac{m_1 m_2}{r^2} & W &= PT & F &= G \frac{m_1 m_2}{r^2} \\ H &= U + PV & G &= \sigma \frac{A}{L} & F &= m \ddot{x} & H &= U + PV & G &= \sigma \frac{A}{L} & F &= m \ddot{x} \\ B & f = \frac{(c+v)}{(c-v)} & f = \frac{(c+v)}{(c-v)} & F = qv \times B & f = \frac{(c+v)}{(c-v)} & f = \frac{(c+v)}{(c-v)} & F = qv \times B & f = \frac{(c+v)}{(c-v)} & f = \frac{(c+v)}{(c-v)} & F = qv \times B \\ f & = \frac{(c+v)}{(c-v)} & f & = \frac{(c+v)}{(c-v)} & F = qv \times B & f = \frac{(c+v)}{(c-v)} & f & = \frac{(c+v)}{(c-v)} & F = qv \times B & f = \frac{(c+v)}{(c-v)} & f = \frac{(c+v)}{(c-v)} & F = qv \times B \\ V & = J/C & V & = J/C \\ A + B & = A + B & A + B & = A + B & A + B & = A + B & A + B & = A + B & A + B & = A + B & A + B = A + B \end{aligned}$$

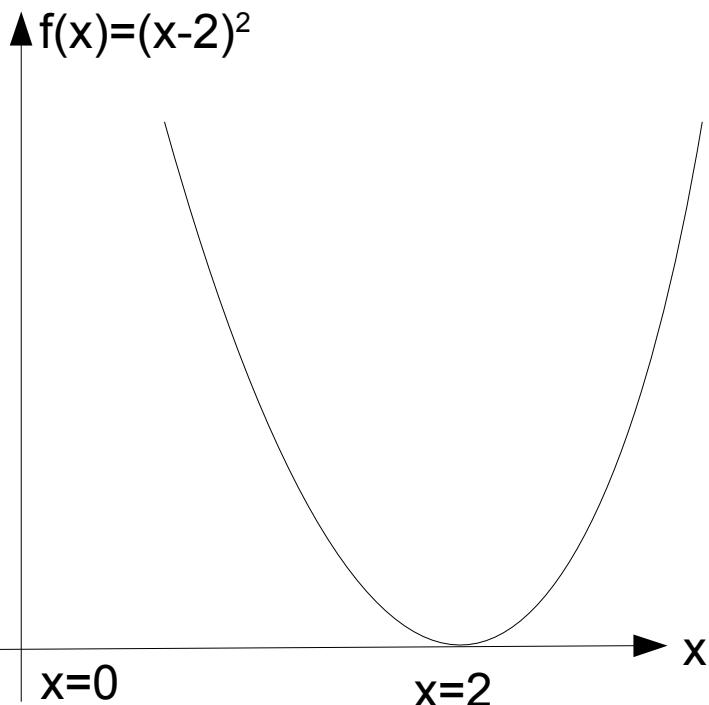
We have studied the limitations of the lens of computer hardware through which we must look for numerical results. What about the limitations of our numerical algorithms?

# Behavior of N-R Algorithm at Multiple Roots



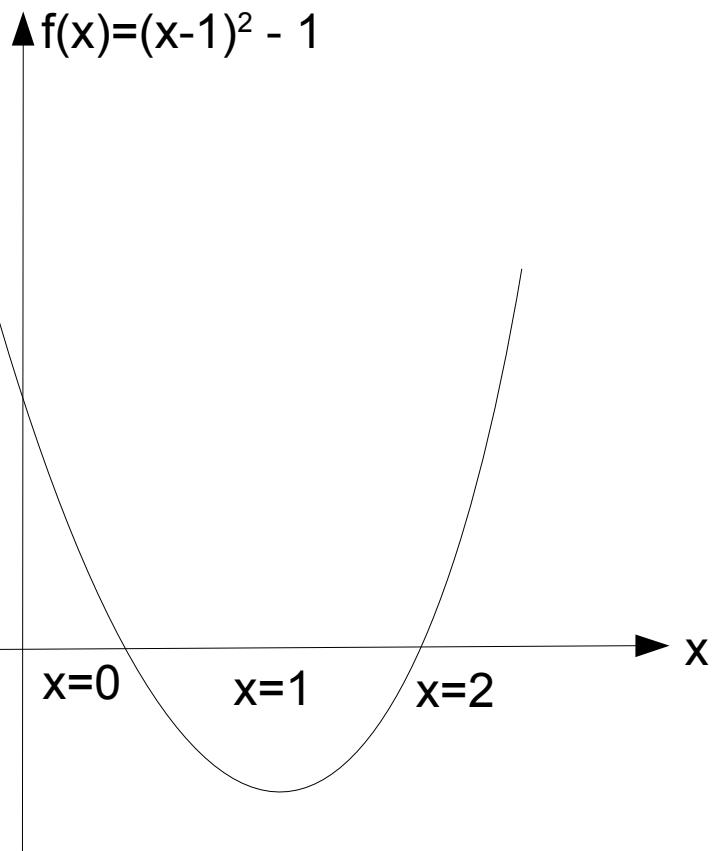
0	3	3
1	2.25	0.5625
2	2.025	0.05062499999999982
3	2.000304878048781	0.0006098490481859375
4	2.00000046461147	9.292229690631253e-08
5	2.00000000000001	1.77635683940025e-15
6	2	0

# Behavior of N-R Algorithm at Multiple Roots



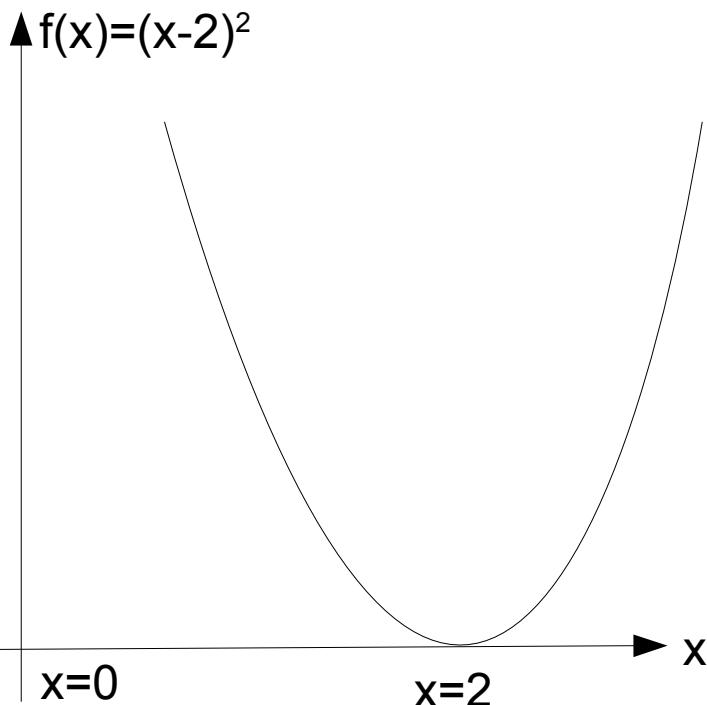
0	3	1
1	2.5	0.25
2	2.25	0.0625
3	2.125	0.015625
4	2.0625	0.00390625
5	2.03125	0.0009765625
6	2.015625	0.000244140625
7	2.0078125	6.103515625e-05
8	2.00390625	1.52587890625e-05
9	2.001953125	3.814697265625e-06
10	2.0009765625	9.5367431640625e-07
11	2.00048828125	2.384185791015625e-07
12	2.000244140625	5.960464477539062e-08
13	2.0001220703125	1.490116119384766e-08
14	2.00006103515625	3.725290298461914e-09
15	2.000030517578125	9.313225746154785e-10
16	2.000015258789062	2.328306436538696e-10
17	2.000007629394531	5.820766091346741e-11
18	2.000003814697266	1.455191522836685e-11
19	2.000001907348633	3.637978807091713e-12
20	2.000000953674316	9.094947017729282e-13
21	2.000000476837158	2.273736754432321e-13
22	2.000000238418579	5.684341886080801e-14
23	2.00000011920929	1.4210854715202e-14
24	2.000000059604645	3.552713678800501e-15
25	2.000000029802322	8.881784197001252e-16
26	2.000000014901161	2.220446049250313e-16
27	2.000000007450581	5.55115123125783e-17
28	2.00000000372529	1.387778780781446e-17
29	2.000000001862645	3.469446951953614e-18
30	2.000000000931323	8.673617379884035e-19

# Behavior of Secant Algorithm at Multiple Roots



0	4	8
1	3	3
2	2.4	0.9599999999999997
3	2.117647058823529	0.2491349480968859
4	2.018691588785047	0.03773255306140229
5	2.00102933607823	0.002059731689221219
6	2.000009526032265	1.905215527485849e-05
7	2.00000004900199	9.80039785057011e-09
8	2.00000000000024	4.707345624410664e-14

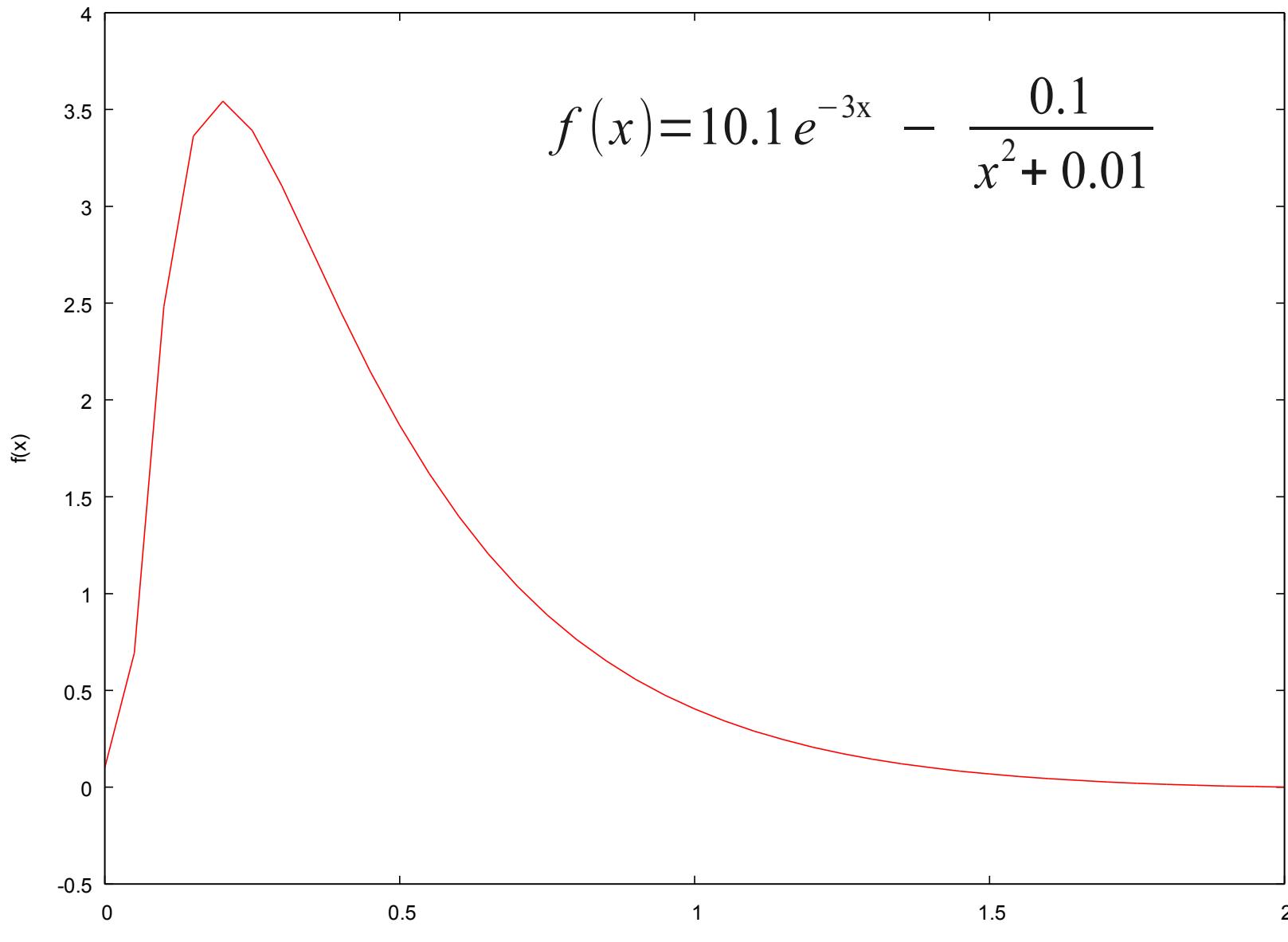
# Behavior of Secant Algorithm at Multiple Roots



```
0 4 4
1 3 1
2 2.666666666666667 0.4444444444444443
3 2.4 0.1599999999999999
4 2.25 0.0625
5 2.153846153846154 0.02366863905325441
6 2.095238095238095 0.009070294784580518
7 2.058823529411765 0.003460207612456775
8 2.036363636363637 0.001322314049586794
9 2.02247191011236 0.0005049867440979699
10 2.0138888888888889 0.0001929012345678999
11 2.008583690987125 7.367975096244472e-05
12 2.005305039787798 2.814344715012285e-05
13 2.00327868852459 1.074979844127896e-05
14 2.002026342451874 4.106063732267847e-06
15 2.001252348152787 1.568375895788087e-06
16 2.0007399380805 5.990664148993789e-07
17 2.000478354460655 2.288229900287754e-07
18 2.00029563932003 8.740260754764651e-08
19 2.000182715147086 3.338482497454056e-08
20 2.000112924171419 1.275186849068122e-08
21 2.000069790976027 4.870780334769662e-09
22 2.000043133195307 1.860472537393645e-09
23 2.00002665778074 7.106372739678463e-10
24 2.000016475414562 2.714392850029487e-10
25 2.000010182366178 1.036805809874421e-10
26 2.000006293048384 3.960245796558146e-11
27 2.000003889317794 1.512679290309838e-11
28 2.00000240373059 5.777920747782671e-12
29 2.000001485587204 2.206969341769987e-12
30 2.000000918143385 8.429872760069362e-13
31 2.000000567443819 3.219924872671856e-13
32 2.000000350699567 1.229901860977224e-13
33 2.000000216744252 4.697807072287988e-14
34 2.000000133955314 1.794402625504285e-14
35 2.000000082788937 6.854008169606801e-15
36 2.000000051166377 2.617998171864294e-15
37 2.00000003162256 9.999863089231212e-16
38 2.000000019543817 3.819607746096342e-16
39 2.000000012078743 1.458960299180093e-16
40 2.000000007465074 5.572732821864115e-17
41 2.000000004613669 2.128594576142821e-17
42 2.000000002851404 8.130507314050401e-18
43 2.000000001762265 3.105577932316062e-18
```

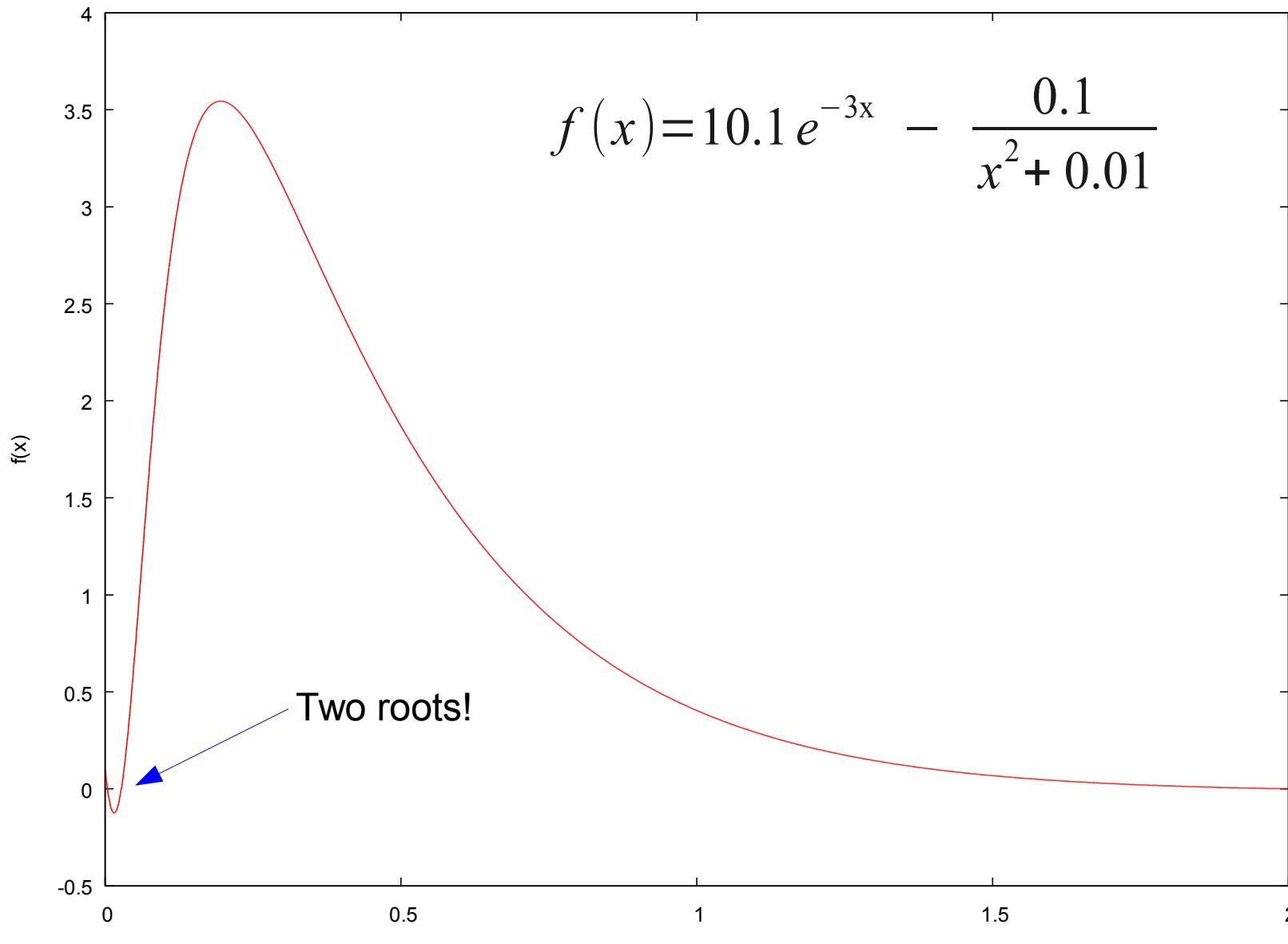
# Pathological Function from Hopkins and Phillips

x step size = 0.05

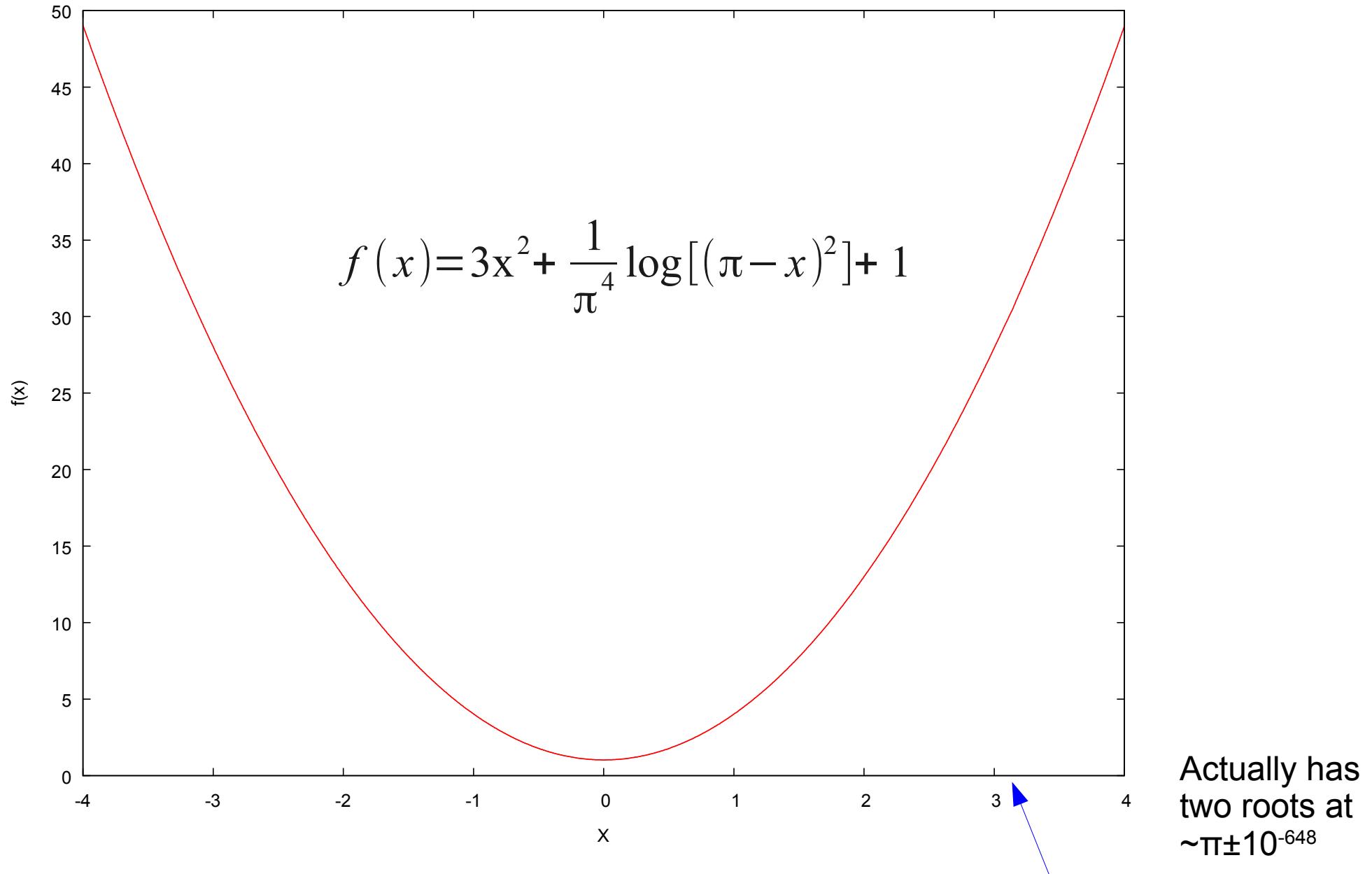


# Pathological Function from Hopkins and Phillips

x step size = 0.001



# Pathological Function Example from Numerical Recipes



# Pathological Function Code

Note use of 'long double' data type

```
#include <stdio.h>
#include <stdlib.h>
#include <math.h>
#define PI 3.14159265358979323846264338327950288L

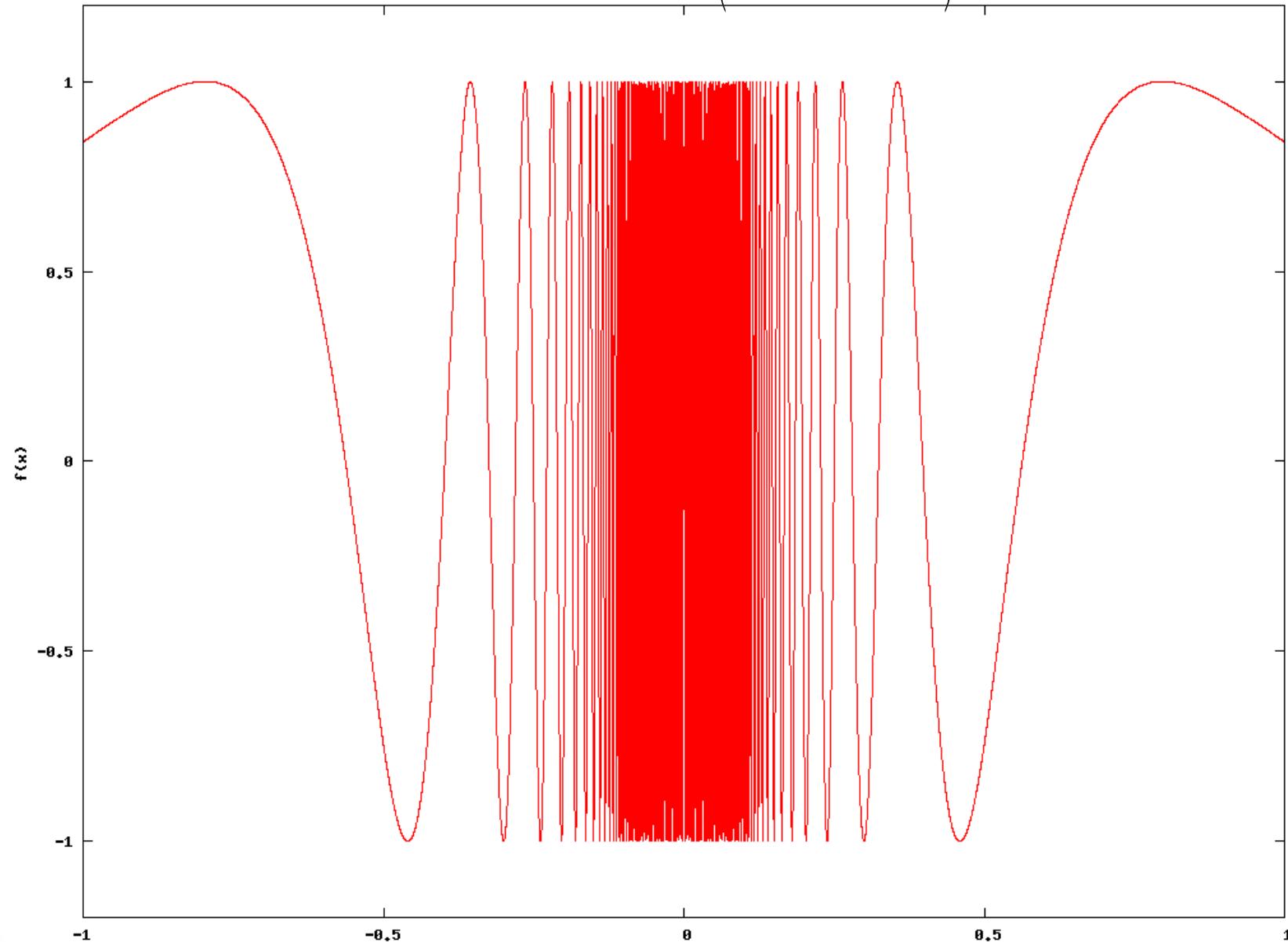
int main() {
    long double d, x, f;
    d = -10.0e-647L;
    while (d < 10.005e-647L) {
        x = PI + d;
        f = 3.0L*x*x + logl(d*d) / (PI*PI*PI*PI) + 1.0L;
        printf("d = %.6Le      f = %.6Le\n", d, f);
        d = d + 0.01e-647L;
    }
    exit(0);
}
```

# Pathological Function Code Output

$d = -1.000000e-646$	$f = 6.813257e-02$
$d = -9.990000e-647$	$f = 6.811203e-02$
$d = -9.980000e-647$	$f = 6.809146e-02$
$\dots$	
$d = -4.000000e-648$	$f = 2.042725e-03$
$d = -3.900000e-648$	$f = 1.522900e-03$
$d = -3.800000e-648$	$f = 9.895727e-04$
$d = -3.700000e-648$	$f = 4.420212e-04$
$d = -3.600000e-648$	$f = -1.205336e-04$
$d = -3.500000e-648$	$f = -6.989371e-04$
$d = -3.400000e-648$	$f = -1.294108e-03$
$d = -3.300000e-648$	$f = -1.907048e-03$
$\dots$	
$d = 3.200000e-648$	$f = -2.538851e-03$
$d = 3.300000e-648$	$f = -1.907048e-03$
$d = 3.400000e-648$	$f = -1.294108e-03$
$d = 3.500000e-648$	$f = -6.989371e-04$
$d = 3.600000e-648$	$f = -1.205336e-04$
$d = 3.700000e-648$	$f = 4.420212e-04$
$d = 3.800000e-648$	$f = 9.895727e-04$
$d = 3.900000e-648$	$f = 1.522900e-03$
$d = 4.000000e-648$	$f = 2.042725e-03$
$\dots$	

# Pathological Function with a Huge Number of Roots

$$f(x) = \sin\left(\frac{1}{x^2 + 0.001}\right)$$



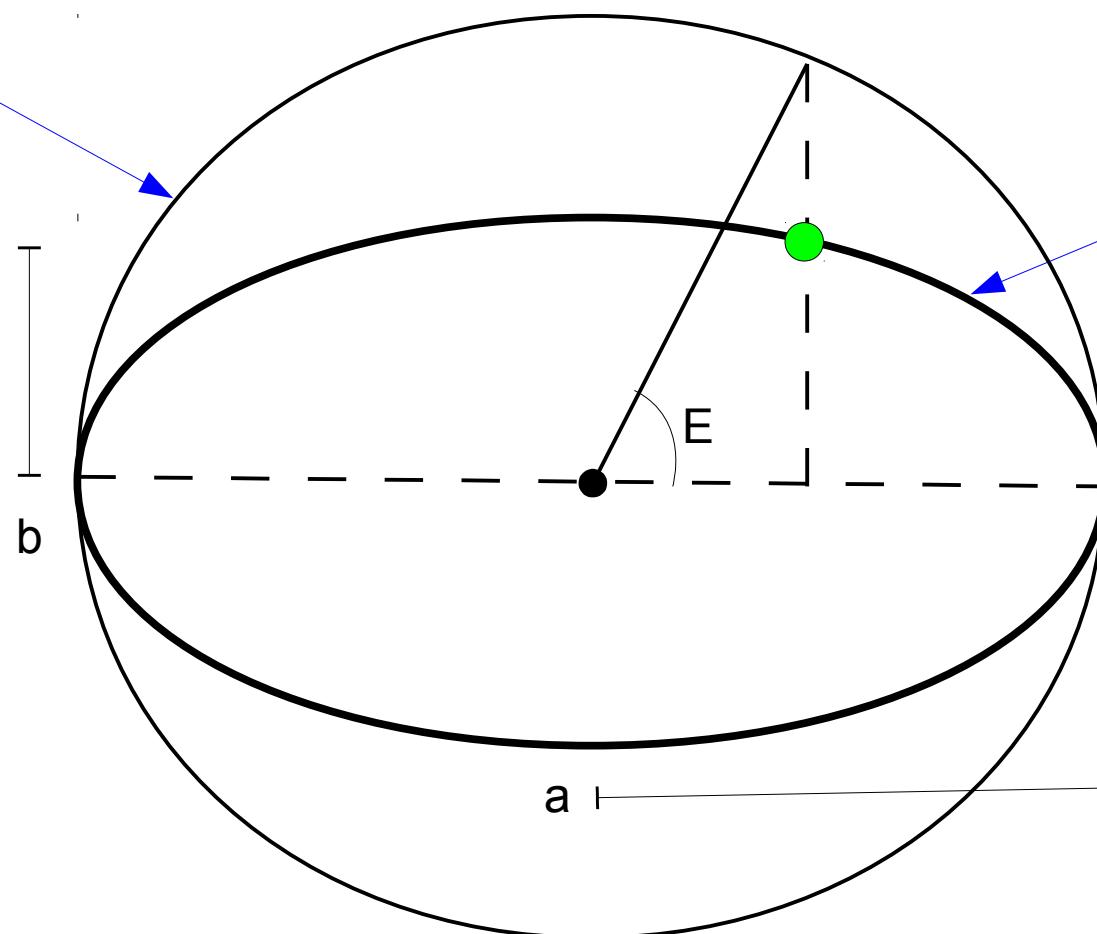
# Kepler's Equation

$$M = E - \epsilon \cdot \sin(E)$$

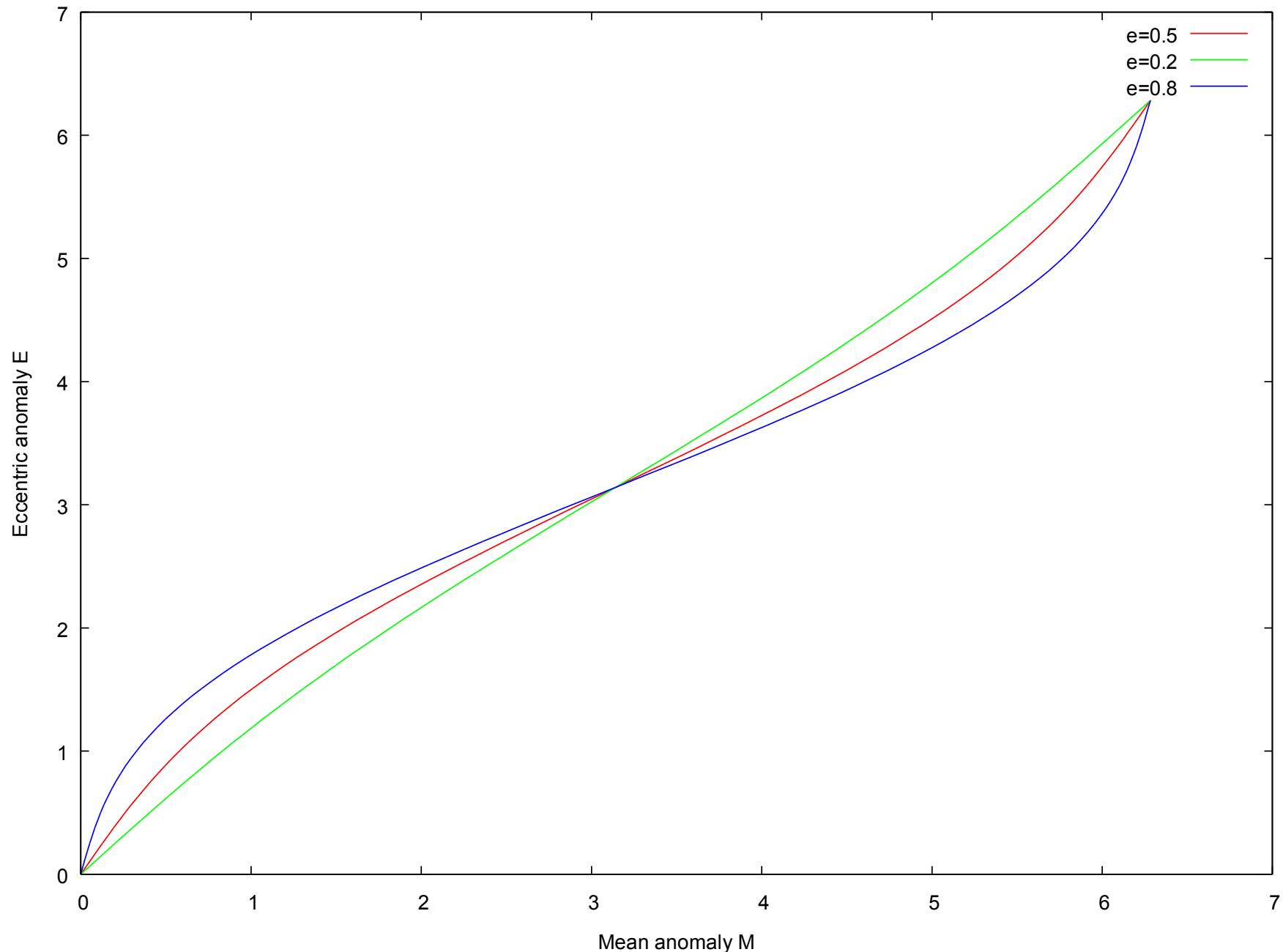
$$M = \text{Mean anomaly} = 2\pi \frac{t}{T} \quad E = \text{Eccentric anomaly} \quad \epsilon = \text{Eccentricity} = \sqrt{\left(1 - \frac{b^2}{a^2}\right)}$$

Auxiliary circle,  
radius  $a$

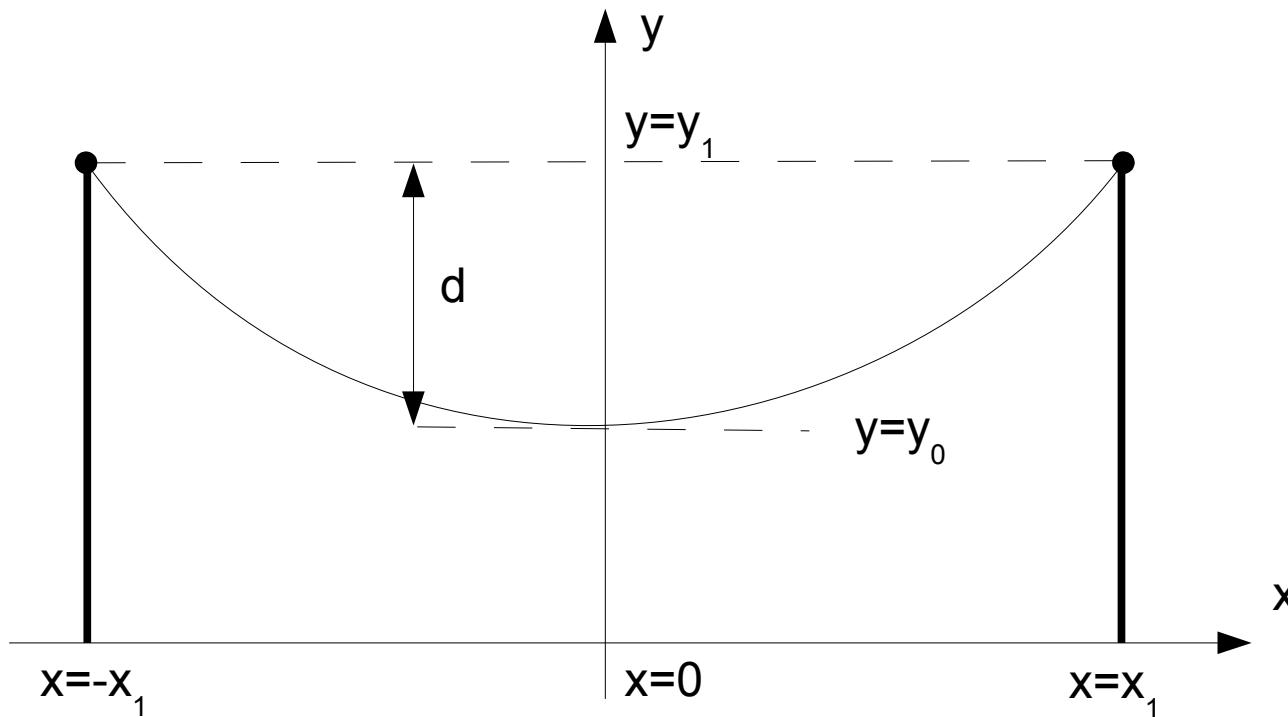
Planet in  
elliptical  
orbit with  
period  $T$



# Kepler's Equation Solutions



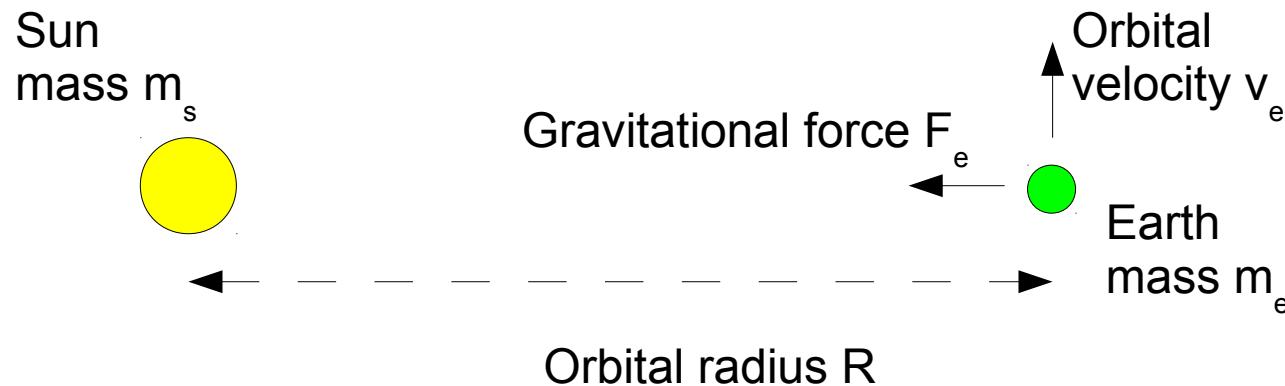
# Catenary Profile of a Suspended Line



$$\text{elevation } y(x) = \lambda \cosh\left(\frac{x}{\lambda}\right) - \lambda + y_0$$

$$\text{length } L = 2\lambda \sinh\left(\frac{x_1}{\lambda}\right)$$

# Sun-Earth Orbital System

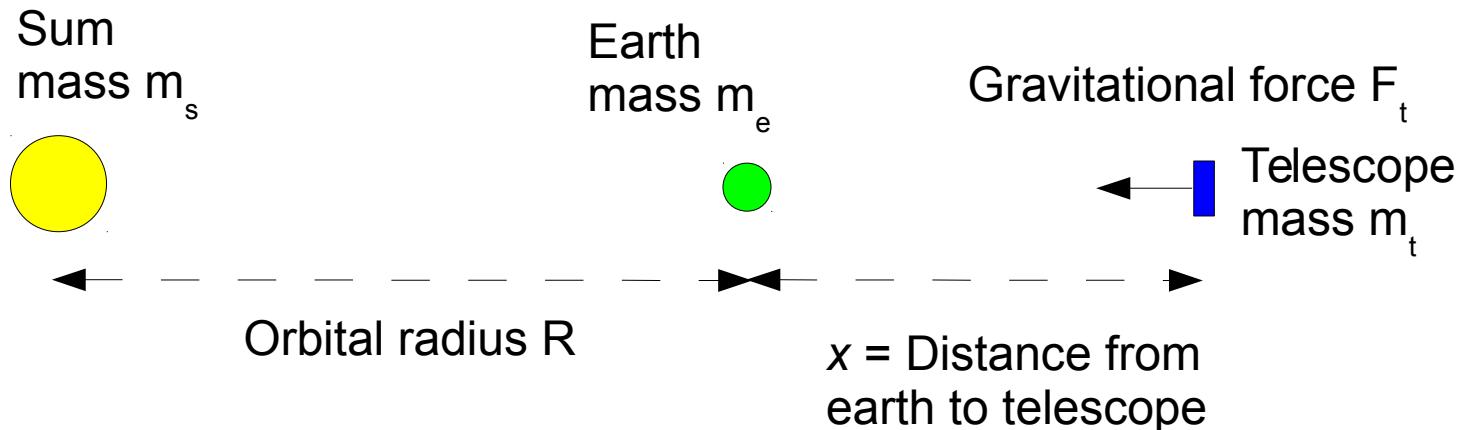


$$F_e = G \frac{m_s m_e}{R^2}$$

$$F_e = \frac{m_e v_e^2}{R} = \frac{4\pi^2 R m_e}{T_e^2}$$

$$T_e^2 = \frac{4\pi^2 R^3}{G m_s}$$

# Sun-Earth Orbital System with JWST

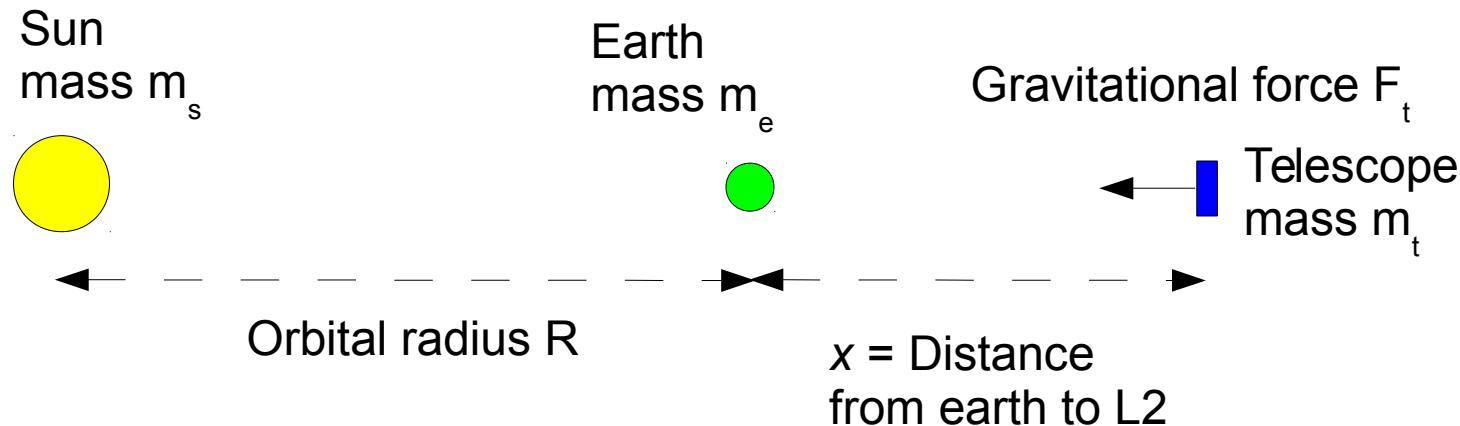


$$F_t = G \frac{m_s m_t}{(R+x)^2} + G \frac{m_e m_t}{x^2}$$

$$F_t = \frac{4\pi^2(R+x)m_t}{T_t^2}$$

$$T_t^2 = \frac{4\pi^2(R+x)}{\frac{G m_s}{(R+x)^2} + \frac{G m_e}{x^2}}$$

# Sun-Earth Orbital System with JWST at L2



At L2 point  $T_e^2 = T_t^2$

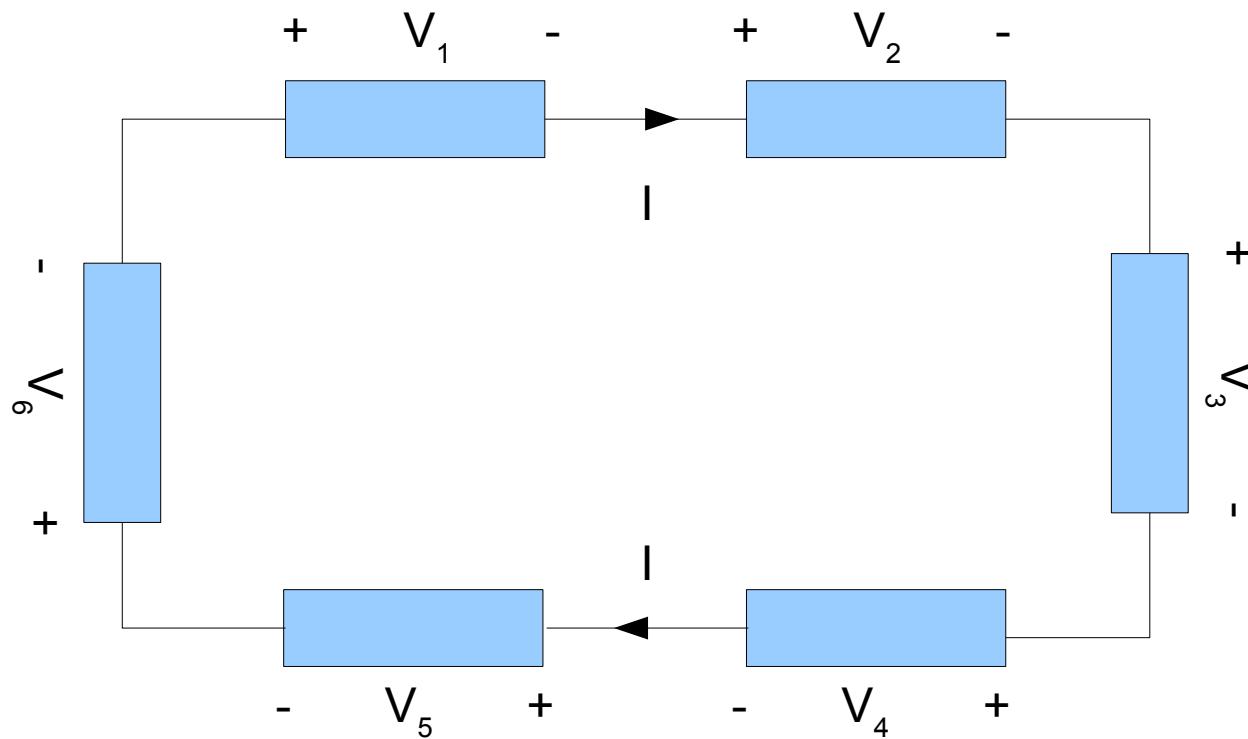
$$\frac{4\pi^2 R^3}{G m_s} = \frac{4\pi^2 (R+x)}{\frac{G m_s}{(R+x)^2} + \frac{G m_e}{x^2}}$$

# Kirchoff's Voltage Law

Electrical circuit elements in a loop carrying current I

If all the reference polarities of all potential drops across loop elements are in the same loop direction, then the sum of all potential drops around the loop = 0

With the relationship between I and  $V_j$  for each circuit element, the current I can be found by solving for the root of an equation



$$\sum V_j = 0$$

# Example of Kirchoff's Voltage Law

$$V_{PS} + V_R + V_{LED} = 0$$

Note: Polarity of element voltage drop is determined by reference direction around loop

