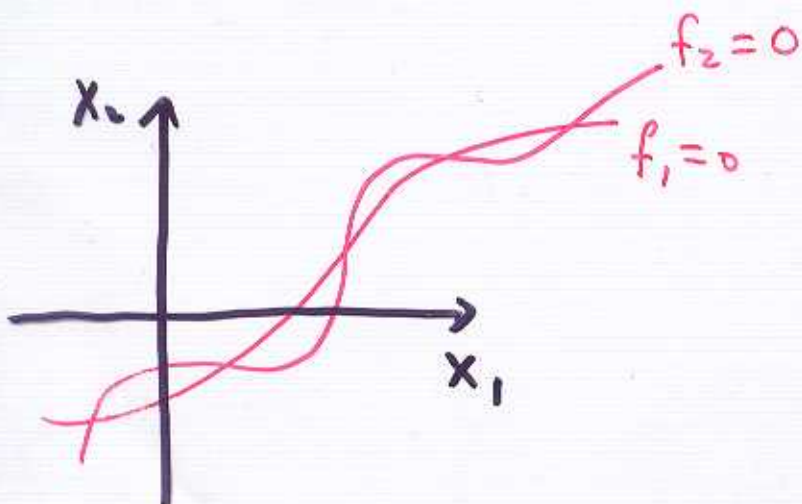
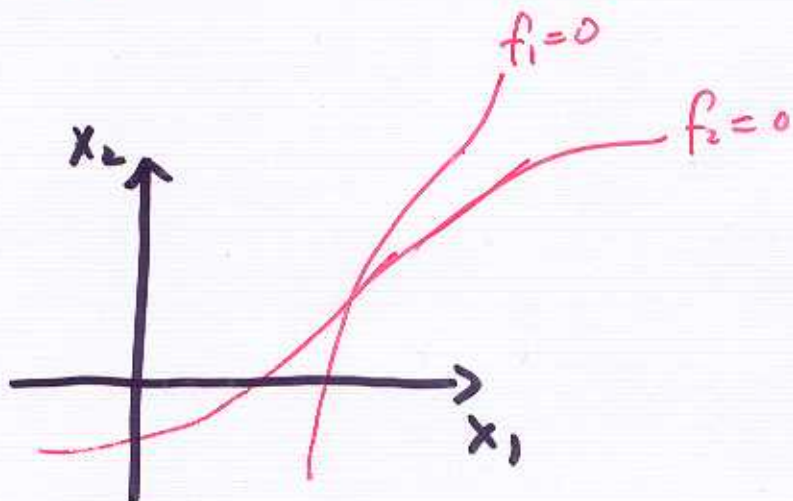
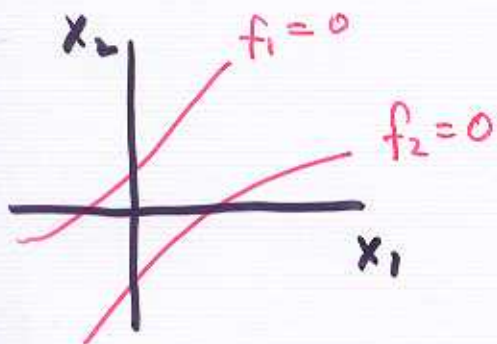


Systems of non-linear eqns.

$$\left. \begin{array}{l} f_1(x_1, \dots, x_n) = 0 \\ f_2(x_1, \dots, x_n) = 0 \\ \vdots \\ f_n(x_1, \dots, x_n) = 0 \end{array} \right\} \rightarrow \underline{F}(\underline{x}) = \underline{0} \text{ --- (I)}$$

- Whether (I) has solns., and how many?

Example:



Picard iteration:

• Decompose $F = Ax + H(x)$

↑
linear

↑
nonlinear part

• Perform iteration

$$x_{i+1} = -A^{-1} H(x_i)$$

$$x = A^{-1} F - A^{-1} H$$

$$F(x) = 0$$

⇒ This is of the form $x_{i+1} = G(x_i)$

• Assume soln. x^* of $F(x^*) = 0 \Leftrightarrow x^* = G(x^*)$

• Suppose $\|G'(x)\| \leq \gamma < 1$ for $\|x - x^*\| < \beta$

where $G' \equiv \begin{bmatrix} \frac{\partial G_1}{\partial x_1} & \frac{\partial G_1}{\partial x_2} & \dots \\ \vdots & \vdots & \vdots \end{bmatrix}$,

iteration converges if $\|x_0 - x^*\| < \beta$

pf $\frac{G(x_0) - G(x^*)}{x_0 - x^*} = G'(c)$ c between x^* & x_0

$$x_{i+1} = G(x_i) \quad \|G(x_0) - G(x^*)\| = \|G'(c)\| \|x_0 - x^*\|$$

$$\|x_1 - x^*\| = \|G'(c)\| \|x_0 - x^*\|$$

$$\|x_1 - x^*\| \leq \gamma \|x_0 - x^*\|$$

$$\|x_2 - x^*\| \leq \gamma \|x_1 - x^*\| < \gamma^2 \|x_0 - x^*\|$$

$$\Rightarrow \|x_n - x^*\| \leq \gamma^n \|x_0 - x^*\|$$

\Rightarrow linear convergence

Fixed Point Iteration

$$\vec{x}^{(n+1)} = \vec{g}(\vec{x}^{(n)}) \quad \text{with fixed point } \vec{\alpha}: \\ \vec{\alpha} = \vec{g}(\vec{\alpha})$$

with slight modifications, contraction mapping theorem from 1-D generalizes directly to multi-dimensions.

\Rightarrow The absolute values are replaced with norms.

Error:

$$\vec{\alpha} - \vec{x}^{(n+1)} = \vec{g}(\alpha) - \vec{g}(\vec{x}^{(n)})$$

$$= \begin{pmatrix} \frac{\partial g_1}{\partial x_1} & \dots & \frac{\partial g_1}{\partial x_n} \\ \vdots & & \vdots \\ \frac{\partial g_n}{\partial x_1} & \dots & \frac{\partial g_n}{\partial x_n} \end{pmatrix} (\vec{g}'^{(n)}) (\vec{\alpha} - \vec{x}^{(n)})$$

$\vec{g}'^{(n)}$ is on line joining $\vec{\alpha}$ & $\vec{x}^{(n)}$

$$\|\vec{\alpha} - \vec{x}^{(n+1)}\| \leq \|G(\vec{g}'^{(n)})\| \|\vec{\alpha} - \vec{x}^{(n)}\|$$

Jacobian G is analog to g' from 1-D case.

Convergence if $\|G\| < 1$.

For a domain to have contraction map, the domain must be convex, all lines connecting two points in domain must lie entirely in domain.

Summary:

Let $\vec{\alpha}$ be a fixed point of $\vec{g}(\vec{x})$. assume that \vec{g} is cont. diff. in a neighborhood of $\vec{\alpha}$ & $\|G\| < 1$.
For \vec{x}_0 sufficiently close to $\vec{\alpha}$, $\vec{x}^{(n+1)} = \vec{g}(\vec{x}^{(n)})$ will converge to $\vec{\alpha}$.

Converting root finding to fixed pt. iteration:

$$\vec{x} = \vec{x} + A \cdot \vec{f}(\vec{x}) \equiv \vec{g}(\vec{x}) \quad A: m \times m \text{ matrix (of constants, for example)}$$

Jacobian $G = I + A \cdot \vec{F}(\vec{x})$, \vec{F} is Jacobian of \vec{f}

- Recall that $\|G(\alpha)\| < 1$ for convergence
- Higher order of convergence if $\|G(\alpha)\| = 0$

$$\|G(\alpha)\| = \|I + A \vec{F}(\vec{\alpha})\| = 0$$

$$\therefore A = -\vec{F}^{-1}(\vec{\alpha}) \quad \text{or in the iteration}$$

$$A = -\vec{F}^{-1}(\vec{x}^n)$$

$$\therefore \vec{x}^{n+1} = \vec{x}^n - \vec{F}^{-1}(\vec{x}^n) \cdot \vec{f}(\vec{x}^n)$$

\Rightarrow Newton's method for systems

Numerically, $\vec{x}^{n+1} = \vec{x}^n + \vec{\delta}^{n+1}$

where

$$\vec{F}(\vec{x}^n) \vec{\delta}^{n+1} = -\vec{f}(\vec{x}^n)$$

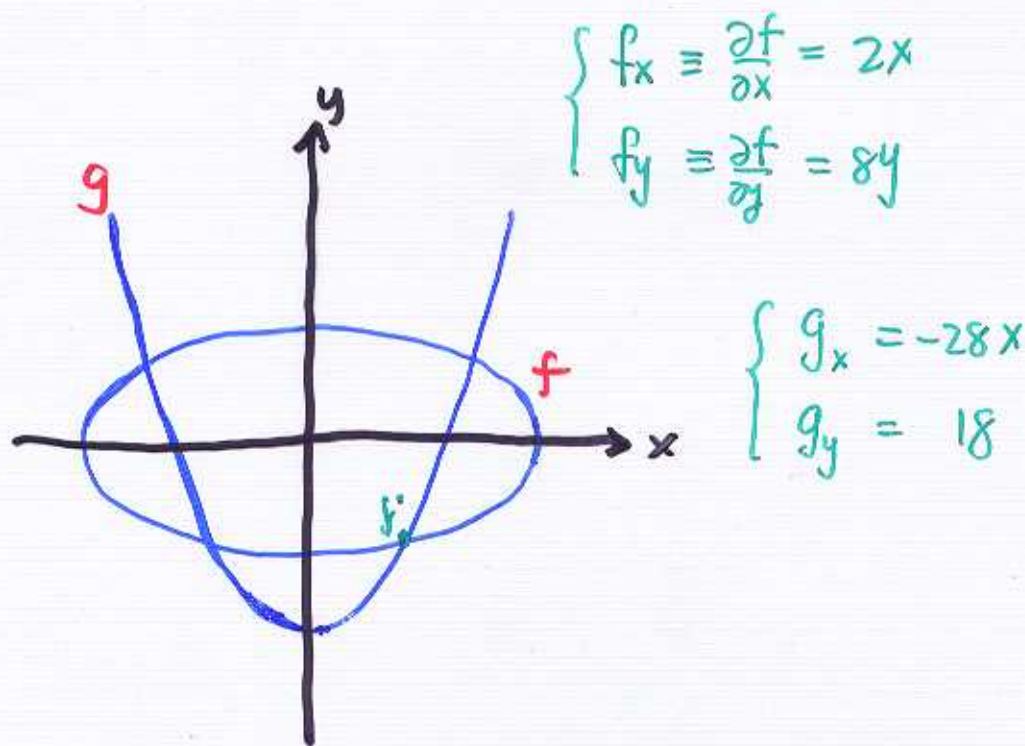
Use linear solver to find $\vec{\delta}^{n+1}$

- Recall that convergence results follow from fixed pt. thm.
- Computationally Expansive step for each iteration: Soln. of linear system of equations $\sim \mathcal{O}(n^3)$
- A lot less iterations as we get closer to the fixed pt. \Rightarrow Jacobian is almost const.

Example:

$$f(x, y) \equiv x^2 + 4y^2 - 9 = 0$$

$$g(x, y) \equiv 18y - 14x^2 + 45 = 0$$



Newton's method:

$$\begin{cases} x^{n+1} = x^n + \delta_x^{n+1} \\ y^{n+1} = y^n + \delta_y^{n+1} \end{cases}$$

$$\delta = \text{rhs}/A$$

$$\begin{bmatrix} f_x^n & f_y^n \\ g_x^n & g_y^n \end{bmatrix} \begin{bmatrix} \delta_x^{n+1} \\ \delta_y^{n+1} \end{bmatrix} = - \begin{bmatrix} f^n \\ g^n \end{bmatrix}$$

$$\begin{bmatrix} 2x^n & 8y^n \\ -28x^n & 18 \end{bmatrix} \begin{bmatrix} \delta_x^{n+1} \\ \delta_y^{n+1} \end{bmatrix} = - \begin{bmatrix} x^n^2 + 4y^n^2 - 9 \\ 18y^n - 14x^n^2 + 45 \end{bmatrix}$$

$$(x^0, y^0) = (1, -1)$$

$$x^1 = 1 + \delta_x^1$$

$$y^1 = -1 + \delta_y^1$$

$$\begin{bmatrix} 2 & -8 \\ -28 & 18 \end{bmatrix} \begin{bmatrix} \delta_x^1 \\ \delta_y^1 \end{bmatrix} = - \begin{bmatrix} -4 \\ 49 \end{bmatrix}, \quad \begin{bmatrix} \delta_x^1 \\ \delta_y^1 \end{bmatrix} = \begin{bmatrix} 0.1702\dots \\ -0.4574\dots \end{bmatrix}$$

$$(x^2, y^2) = (1.202158829, -1.376760321)$$

$$(x^3, y^3) = (1.203165807, -1.374083487)$$

$$(x^4, y^4) = (1.203166963, -1.374085342)$$

⋮

Example: $f(x) = x^2 + 1 = 0$

$x = a + bi$ Real Imag.

$f(a, b) = \underline{a^2 - b^2 + 1} + \underline{2ab}i$

$\Rightarrow f(a, b) = \text{Re}[f(x = a + bi)] = a^2 - b^2 + 1$

$g(a, b) = \text{Im}[f(x = a + bi)] = 2ab$

$f_a = 2a$ $f_b = -2b$

$g_a = 2b$ $g_b = 2a$

$a^{i+1} = a^i + \delta a^{i+1}$

$b^{i+1} = b^i + \delta b^{i+1}$

$$\begin{bmatrix} f_a^i & f_b^i \\ g_a^i & g_b^i \end{bmatrix} \begin{bmatrix} \delta a^{i+1} \\ \delta b^{i+1} \end{bmatrix} = - \begin{bmatrix} a^2 - b^2 + 1 \\ 2ab \end{bmatrix}$$

$a^0 = 1$ $b^0 = -\frac{1}{2}$ $\rightarrow \begin{cases} a^1 = 1 - \frac{9}{10} = \frac{1}{10} \\ b^1 = -\frac{1}{2} + \frac{1}{20} = -\frac{9}{20} \end{cases}$

$\rightarrow \begin{cases} a^2 = -0.09862 \\ b^2 = -0.437652 \end{cases}$

⋮

Table 7.7. The Modified Newton's Method for Solving the Nonlinear System [7.65] with $n = 5$

k	$\ \alpha - x^{(k)}\ $	Ratio
0	3.58E - 1	
1	3.42E - 2	0.096
2	7.03E - 3	0.206
3	6.83E - 4	0.097
4	1.54E - 4	0.225
5	2.09E - 5	0.136
6	2.30E - 6	0.110
7	5.68E - 7	0.247
8	5.53E - 8	0.097
9	7.24E - 9	0.131

Newton's method (7.62) and the modified Newton's method (7.71)

$$x^{(k+1)} = x^{(k)} - A^{-1}F(x^{(k)})$$

are examples of fixed point iteration:

$$x^{(k+1)} = g(x^{(k)}), \quad k \geq 0$$

where $g(x)$ is a vector function

$$g(x) = \begin{bmatrix} g_1(x_1, \dots, x_n) \\ \vdots \\ g_n(x_1, \dots, x_n) \end{bmatrix}$$

There is a general theory for such fixed-point iteration methods involving a strengthened form of Theorem 3.4.2; but we do not consider it here [e.g., see Atkinson (1989, §2.10)].

MATLAB PROGRAM. We give a MATLAB program for the Newton method applied to the solution of a systems of two nonlinear equations in two unknowns:

$$\begin{aligned} F_1(x_1, x_2) &= 0 \\ F_2(x_1, x_2) &= 0 \end{aligned}$$

```
function solution = newton_sys(x_init, err_tol, max_iterates)
%
% The calling sequence for newton_sys.m is
```

math.uiowa.edu/
ftp/atkinson/
ENA_Materials

```

% solution = newton_sys(x_init,err_tol,max_iterates)
% This solves a pair of two nonlinear equations in two unknowns,
%           f(x) = 0
% with f(x) a column vector of length 2. The definition of f(x)
% is to be given below in the function named fsys; and you
% also need to give the Jacobian matrix for f(x) in the
% function named deriv_fsys.
%
% x_init is a vector of length 2, and it is an initial guess
% at the solution.
%
% The parameters err_tol and max_iterates are upper limits on
% the desired error in the solution and the maximum number of
% iterates to be computed.

% Initialization.
x0 = zeros(2,1);
for i=1:2
    x0(i) = x_init(i);
end

error = inf;
it_count = 0;

% Begin the main loop.
while (error > err_tol & it_count < max_iterates)
    it_count = it_count + 1;
    rhs = fsys(x0);
    A = deriv_fsys(x0);
    delta = A\rhs;
    x1 = x0 - delta;
    error = norm(delta,inf);
    % The following statement is an internal print to show
    % the course of the iteration. It and the pause
    % statement following it can be commented out.
    [it_count x1' error]
    pause
    x0 = x1; ← swapping
end

% Return with the solution.
solution = x1;
if it_count == max_iterates
    disp(' ')
    disp('*** Your answers may possibly not ...

```

gthened
§2.10).

plied to

```

%                               satisfy your error tolerance.')
```

```

end

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
function f_val = fsys(x)
%
% The equations being solved are
%   x(1)^2 + 4*x(2)^2 - 9 = 0
%   18*x(2) - 14*x(1)^2 + 45 = 0

f_val = [x(1)^2+4*x(2)^2-9, 18*x(2)-14*x(1)^2+45]';

function df_val = deriv_fsys(x)
%
% This defines the Jacobian matrix for the function
% given in fsys

df_val = [2*x(1), 8*x(2); -28*x(1), 18];

```

PROBLEMS

- Find the remaining three solutions of the system (7.52). Note that symmetry can be used to reduce the number of solutions that must be found computationally.
- Find all solutions to the following systems, using Newton's method (7.55–7.56).
 - $x^2 + y^2 = 4$, $x^2 - y^2 = 1$
 - $x^2 + 4y^2 = 4$, $y = x^2 - 0.4x - 1.96$
 - $x^2 + y^2 = 1$, $2y = 2x^3 + x + 1$
 - $x + y - 2xy = 0$, $x^2 + y^2 - 2x + 2y + 1 = 0$
- Find all solutions to the system

$$x^2 + xy^3 = 9, \quad 3x^2y - y^3 = 4$$

Use each of the initial guesses $(x_0, y_0) = (1.2, 2.5)$, $(-2, 2.5)$, $(-1.2, -2.5)$, $(2, -2.5)$. Observe the root to which each of the iterations converges and the number of iterates computed. Comment on your results.

- Solve the nonlinear system (7.65) for $n = 10, 15, 20$, to the full accuracy of your computer. Compare the needed number of iterates. Graph the solutions x for each n , as in Figure 7.10.
- Use the modified Newton's method (7.71) to solve for the roots of (7.52), presenting results as in Table 7.7. Use $A = F'(x^{(0)})$.
- Repeat Problem 5 for the systems given in Problem 2.

Example:

consider an initial value problem:

$$\frac{\partial y}{\partial t} = f(x, y) \quad y(0) = y_0$$

Use backward Euler:

$$\frac{y_{k+1} - y_k}{\Delta t} = f(x_{k+1}, y_{k+1})$$

$$y_{k+1} = y_k + \Delta t \cdot f(x_{k+1}, y_{k+1})$$

Apply Newton method:

- y_k known
- need to solve for $y_{k+1} \equiv u$

$$F(u) \equiv u - \Delta t \cdot f(x_{k+1}, u) - y_k = 0$$

$$u_{i+1} = u_i - \frac{F(u_i)}{F'(u_i)}$$

$$= u_i - \frac{u_i - \Delta t \cdot f(x_{k+1}, u_i) - y_k}{1 - \Delta t \cdot f_u}$$

EXTRA

Notes

$$\|x - x^*\| < \beta$$

$$\|x - x^*\| = \sqrt{\sum_i (x_i - x_i^*)^2}$$

$$\begin{aligned} \vec{\alpha} - \vec{x}^{(n+1)} &= \vec{g}(\alpha) - \vec{g}(\vec{x}^{(n)}) \\ &= \begin{pmatrix} \frac{\partial g_1}{\partial x_1} & \frac{\partial g_1}{\partial x_2} \\ \frac{\partial g_2}{\partial x_1} & \frac{\partial g_2}{\partial x_2} \end{pmatrix} (\vec{\xi}_n) \cdot (\vec{\alpha} - \vec{x}^{(n)}) \end{aligned}$$

first comp.

$$\begin{aligned} \alpha_1 - x_1 &= \frac{\partial g_1}{\partial x_1} (\alpha_1 - x_1^{(n)}) \\ &\quad + \frac{\partial g_1}{\partial x_2} (\alpha_2 - x_2^{(n)}) \end{aligned}$$

converting root finding prob

→ fixed point problem.

⇒ root of $f(x)$

$$\Rightarrow x = x + c \cdot f(x)$$

$$\vec{F} = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \vdots & & \vdots \end{pmatrix}$$

$$\vec{F} = \frac{\partial f}{\partial \vec{x}}$$

$$\vec{F}^{-1} = \frac{1}{\frac{\partial f}{\partial \vec{x}}}$$

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

$$\vec{x}_{n+1} = \vec{x}_n - \vec{F}^{-1}(x_n) \cdot \vec{f}(\vec{x}_n)$$

$$\begin{bmatrix} 2x^n & 8y^n \\ -28x^n & 18 \end{bmatrix} \begin{bmatrix} \delta x \\ \delta y \end{bmatrix} = - \begin{bmatrix} f^n \\ g^n \end{bmatrix}$$

$$\begin{bmatrix} \delta x \\ \delta y \end{bmatrix} = - \begin{bmatrix} & \\ & \end{bmatrix}^{-1} \begin{bmatrix} f^n \\ g^n \end{bmatrix}$$

$$\star (x^0, y^0) = (1, -1)$$

$$\begin{bmatrix} \delta x^1 \\ \delta y^1 \end{bmatrix} = - \begin{bmatrix} 2 & -8 \\ -28 & 18 \end{bmatrix}^{-1} \begin{bmatrix} f^0 \\ g^0 \end{bmatrix}$$