

MA579H Scientific Computing

Newton's Method for Solving Systems of Nonlinear Equations

Systems of Nonlinear Equations

Consider a nonlinear system of equations given by

$$f_1(x_1, x_2, \dots, x_n) = 0$$

$$f_2(x_1, x_2, \dots, x_n) = 0$$

⋮

$$f_n(x_1, x_2, \dots, x_n) = 0$$

where f_1, f_2, \dots, f_n are real-valued functions of the independent variables x_1, \dots, x_n in \mathbb{R} .

Systems of Nonlinear Equations

Consider a nonlinear system of equations given by

$$f_1(x_1, x_2, \dots, x_n) = 0$$

$$f_2(x_1, x_2, \dots, x_n) = 0$$

⋮

$$f_n(x_1, x_2, \dots, x_n) = 0$$

where f_1, f_2, \dots, f_n are real-valued functions of the independent variables x_1, \dots, x_n in \mathbb{R} .

Let $U \subset \mathbb{R}^n$. Define $F : U \rightarrow \mathbb{R}^n$ by

$$F(x) := \begin{bmatrix} f_1(x) \\ \vdots \\ f_n(x) \end{bmatrix} \text{ for } x := \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \in \mathbb{R}^n.$$

Problem: Solve $F(x) = 0$.

Jacobian matrix

If the function $F(x)$ is differentiable on U , that is, if the functions $f_1(x), \dots, f_n(x)$ are differentiable on U then the Jacobian matrix $J(c)$ of F at $c \in U$ is given by

$$J(c) := \begin{bmatrix} \frac{\partial f_1(c)}{\partial x_1} & \dots & \frac{\partial f_1(c)}{\partial x_n} \\ \frac{\partial f_2(c)}{\partial x_1} & \dots & \frac{\partial f_2(c)}{\partial x_n} \\ \vdots & \vdots & \vdots \\ \frac{\partial f_n(c)}{\partial x_1} & \dots & \frac{\partial f_n(c)}{\partial x_n} \end{bmatrix} \in \mathbb{R}^{n \times n}.$$

Jacobian matrix

If the function $F(x)$ is differentiable on U , that is, if the functions $f_1(x), \dots, f_n(x)$ are differentiable on U then the Jacobian matrix $J(c)$ of F at $c \in U$ is given by

$$J(c) := \begin{bmatrix} \frac{\partial f_1(c)}{\partial x_1} & \dots & \frac{\partial f_1(c)}{\partial x_n} \\ \frac{\partial f_2(c)}{\partial x_1} & \dots & \frac{\partial f_2(c)}{\partial x_n} \\ \vdots & \vdots & \vdots \\ \frac{\partial f_n(c)}{\partial x_1} & \dots & \frac{\partial f_n(c)}{\partial x_n} \end{bmatrix} \in \mathbb{R}^{n \times n}.$$

Example: Let $F(x, y) = [e^{x+y} - 2 \quad \sin x]^\top \in \mathbb{R}^2$. Then

$$J(x, y) = \begin{bmatrix} \frac{\partial(e^{x+y}-2)}{\partial x} & \frac{\partial(e^{x+y}-2)}{\partial y} \\ \frac{\partial(\sin x)}{\partial x} & \frac{\partial(\sin x)}{\partial y} \end{bmatrix} = \begin{bmatrix} e^{x+y} & e^{x+y} \\ \cos x & 0 \end{bmatrix}.$$

Systems of Nonlinear Equations

Example 1: The system

$$e^{u+v} = 2$$

$$\sin u = 0$$

is the same as $F(u, v) = 0$ where $F(u, v) = [e^{u+v} - 2, \sin u]^\top$ and

$$J(u, v) = \begin{bmatrix} e^{u+v} & e^{u+v} \\ \cos u & 0 \end{bmatrix}.$$

Systems of Nonlinear Equations

Example 1: The system

$$e^{u+v} = 2$$

$$\sin u = 0$$

is the same as $F(u, v) = 0$ where $F(u, v) = [e^{u+v} - 2, \sin u]^\top$ and

$$J(u, v) = \begin{bmatrix} e^{u+v} & e^{u+v} \\ \cos u & 0 \end{bmatrix}.$$

Example 2: The system

$$x^2 + 0.25 = y$$

$$y^2 + 0.25 = x$$

is the same as $F(x, y) = 0$ where $F(x) = [x^2 - y + .25, -x + y^2 + .25]^\top$ and

$$J(x, y) = \begin{bmatrix} 2x - 1 & -1 \\ -1 & 2y \end{bmatrix}.$$

Newton's Method

If $F(x)$ is differentiable at $c \in U$ then $y = F(c) + J(c)(x - c)$ is a linear approximation of $y = F(x)$ at c in the sense that

$$F(x) = F(c) + J(c)(x - c) + \mathcal{O}(\|x - c\|^2)$$

Newton's Method

If $F(x)$ is differentiable at $c \in U$ then $y = F(c) + J(c)(x - c)$ is a linear approximation of $y = F(x)$ at c in the sense that

$$F(x) = F(c) + J(c)(x - c) + \mathcal{O}(\|x - c\|^2)$$

Hence an approximate solution of $F(x) = 0$ is obtained by solving the linear system $F(x_0) + J(x_0)(x - x_0) = 0$ which gives

$$x_1 = x_0 - J(x_0)^{-1} F(x_0).$$

Newton's Method

If $F(x)$ is differentiable at $c \in U$ then $y = F(c) + J(c)(x - c)$ is a linear approximation of $y = F(x)$ at c in the sense that

$$F(x) = F(c) + J(c)(x - c) + \mathcal{O}(\|x - c\|^2)$$

Hence an approximate solution of $F(x) = 0$ is obtained by solving the linear system $F(x_0) + J(x_0)(x - x_0) = 0$ which gives

$$x_1 = x_0 - J(x_0)^{-1} F(x_0).$$

This sets up **multivariate Newton's method** for solving $F(x) = 0$ with an initial guess x_0

$$x_{k+1} = x_k - J(x_k)^{-1} F(x_k), \quad k = 0, 1, \dots$$

Newton's Method

If $F(x)$ is differentiable at $c \in U$ then $y = F(c) + J(c)(x - c)$ is a linear approximation of $y = F(x)$ at c in the sense that

$$F(x) = F(c) + J(c)(x - c) + \mathcal{O}(\|x - c\|^2)$$

Hence an approximate solution of $F(x) = 0$ is obtained by solving the linear system $F(x_0) + J(x_0)(x - x_0) = 0$ which gives

$$x_1 = x_0 - J(x_0)^{-1} F(x_0).$$

This sets up **multivariate Newton's method** for solving $F(x) = 0$ with an initial guess x_0

$$x_{k+1} = x_k - J(x_k)^{-1} F(x_k), \quad k = 0, 1, \dots$$

The multivariate Newton iterations can be rewritten as

for $k = 0, 1, 2, \dots$

Solve $J(x_k)s = -F(x_k)$ for s

Set $x_{k+1} = x_k + s$

Example

Consider the system

$$v - u^3 = 0$$

$$u^2 + v^2 = 1$$

Here $F(u, v) = [v - u^3, u^2 + v^2 - 1]^\top$ and

$$J(u, v) = \begin{bmatrix} -3u^2 & 1 \\ 2u & 2v \end{bmatrix}.$$

Starting with $x_0 = [1, 2]^\top$, $F(x_0) = [1, 4]^\top$ and $s = [0, -1]^\top$ is a solution of $J(x_0)s = -F(x_0)$.

Example

Consider the system

$$v - u^3 = 0$$

$$u^2 + v^2 = 1$$

Here $F(u, v) = [v - u^3, u^2 + v^2 - 1]^\top$ and

$$J(u, v) = \begin{bmatrix} -3u^2 & 1 \\ 2u & 2v \end{bmatrix}.$$

Starting with $x_0 = [1, 2]^\top$, $F(x_0) = [1, 4]^\top$ and $s = [0, -1]^\top$ is a solution of $J(x_0)s = -F(x_0)$.

Therefore the first iteration gives $x_1 := [1, 2]^\top + [0, -1]^\top = [1, 1]^\top$.

Newton's Method

The iterations nearly converge after 7 steps to a solution (0.8636, 0.5636).

iteration	u	v
1	1	2
2	1	1
3	0.875	0.625
4	0.8290363482671175	0.5643491124260355
5	0.8260401081706523	0.5636197735028443
6	0.8260313577324098	0.5636241621316300
7	0.826031357654187	0.5636241621612585

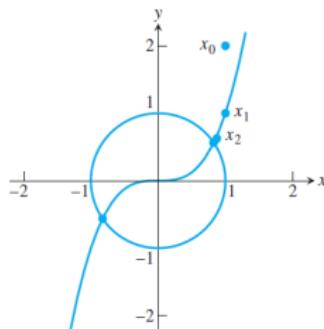


Figure : Figure shows the solutions of $F(x) = 0$ as dots on the circle and Newton iterations with $x_0 = (1, 2)$ converging to the solution (0.8636, 0.5636).