

Newton's Method:  
The Basic Algorithm

univariate case

a simple problem :

Given a function

$$z \in \mathbb{R} \rightarrow f(z) \in \mathbb{R},$$

we would like to find a <sup>REAL</sup> root (or roots)  $Z$  :

$$f(Z) = 0.$$

Note  $z$  vs.  $Z$  and  $f(z)$  vs.  $f(Z) = 0$ .

the fundamental idea : improvement, or update

Given  $\tilde{z}$  (near  $Z$ ),

approximate  $f(z)$  near  $\tilde{z}$  as

$$f(z) \approx \tilde{f}(z) = f(\tilde{z}) + f'(\tilde{z})(z - \tilde{z});$$

Taylor series about  $\tilde{z}$

find root of  $\tilde{f}(z)$  :

$$\begin{aligned} \tilde{f}(\tilde{z}) = 0 &\Rightarrow 0 = f(\tilde{z}) + f'(\tilde{z})(\tilde{z} - \tilde{z}) \\ &\Rightarrow \tilde{z} = \tilde{z} - f(\tilde{z})/f'(\tilde{z}). \end{aligned}$$

the conceit :  $|\tilde{z} - Z| \downarrow < |\tilde{z} - Z|$  ;  
linear equation(s) easy to solve (efficiently).

iteration : an example

Consider

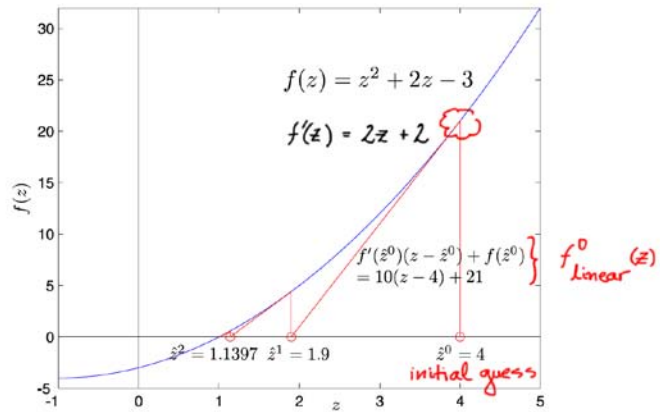
$$z^2 + 2z = 3 \quad (z = -3, z = 1)$$

$$\downarrow$$
$$z^2 + 2z - 3 = 0$$

$$\downarrow$$
$$f(z) = 0$$
$$f(z) = z^2 + 2z - 3$$

Note :  $f'(z) = 2z + 2$ .

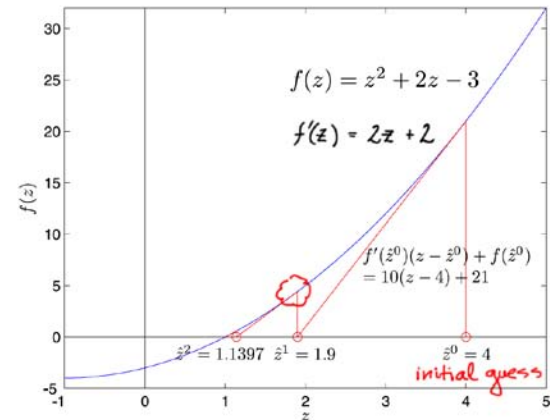
1<sup>st</sup> iteration :



$$f_{\text{linear}}^0(z) \equiv f'(z^0)(z - z^0) + f(z^0) = 10(z - 4) + 21$$

$$f_{\text{linear}}^0(z^1) \equiv 10(z^1 - 4) + 21 = 0 \Rightarrow z^1 = 1.9 = 4 - \frac{f(z^0)}{f'(z^0)}$$

2<sup>nd</sup> iteration :



$$f_{\text{linear}}^1(z) \equiv f'(z^1)(z - z^1) + f(z^1) = 5.8(z - 1.9) + 4.41$$

$$f_{\text{linear}}^1(z^2) = 0$$

$$z^2 = 1.1397 \dots$$

the algorithm :

**Algorithm 1** Newton algorithm with storage of intermediate approximations

```

k ← 0
while |f(ẑk)| > tol
    ẑk+1 ← ẑk - f(ẑk) / f'(ẑk)
    k ← k + 1
end
Z ← ẑk
    
```

$$Z \approx \hat{z}^k$$

**Algorithm 2** Newton algorithm without storage *and increment form*

```

ẑ ← ẑ0
while |f(ẑ)| > tol
    δẑ ← -f(ẑ) / f'(ẑ)
    ẑ ← ẑ + δẑ
end
Z ← ẑ
    
```

*current iterate* } *one iteration*

stopping criterion : at termination

If  $f(z)$  is smooth about  $Z$ ,  $f'(Z) \neq 0$  "0"

$$f(\hat{z}) = f(Z + (\hat{z} - Z))$$

$$\approx \underbrace{f(Z)}_0 + f'(Z)(\hat{z} - Z) + \dots$$

*h.o.t.* ↑

$$\hat{z} - Z = f(\hat{z}) / f'(Z)$$

$$\Rightarrow |\hat{z} - Z| \leq \text{tol} / |f'(Z)|$$

*sensitivity*

$$\Rightarrow \text{choose tol} = \text{tol}_Z \cdot |f'(Z)|$$

*estimate, say f'(ẑ)*

Hence  $|f(\hat{z})| (\leq \text{tol})$  is a natural error estimator.

Convergence rate:

Define error in  $k^{\text{th}}$  iterate as  $\epsilon^k = \hat{z}^k - Z$ ; after  $k$  iterations

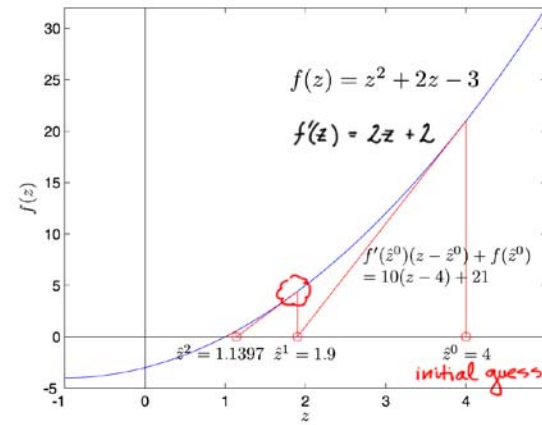
then, if

- (i)  $f(z)$  is smooth (e.g., the second derivative exists), \*\*
- \*0 (ii)  $f'(Z) \neq 0$  (i.e., the derivative at the root is nonzero), and \*
- (iii)  $|\epsilon^0|$  (the error of our initial guess) is sufficiently small, \*\*\*

we achieve quadratic convergence:

$$\epsilon^{k+1} \sim (\epsilon^k)^2 \left( \frac{1}{2} \frac{f''(Z)}{f'(Z)} \right)$$

evidence: simple example



k	$\hat{z}^k$	$\hat{z}^k - Z$ ( $Z = 1$ )
iteration	approximation	number of correct digits
0	4	0
1	1.9	1
2	1.1...	1
3	1.004...	3
4	1.000005...	6
5	1.000000000006...	12

(Note here  $\frac{1}{2} \frac{f''(Z)}{f'(Z)} = \frac{1}{4}$ .)

**IF** Newton's method converges,  $\hat{z}$  converges very rapidly to a root  $Z$ .

proof: Recall

$\epsilon^k \equiv \hat{z}^k - Z$ ; k<sup>th</sup> iterate exact solution

then

$\hat{z}^{k+1} = \hat{z}^k - \frac{f(\hat{z}^k)}{f'(\hat{z}^k)}$  Newton update

$Z + \epsilon^{k+1} = Z + \epsilon^k - \frac{f(Z + \epsilon^k)}{f'(Z + \epsilon^k)}$

Taylor series

$\epsilon^{k+1} = \epsilon^k - \frac{f(Z) + \epsilon^k f'(Z) + \frac{1}{2}(\epsilon^k)^2 f''(Z) + \dots}{f'(Z) + \epsilon^k f''(Z) + \dots}$  root smoothness

\*0:  $|f'(Z)| \leq \epsilon f'(Z)$

$\epsilon^{k+1} = \epsilon^k - \epsilon^k \frac{f'(Z) + \frac{1}{2}\epsilon^k f''(Z) + \dots}{f'(Z)(1 + \epsilon^k \frac{f''(Z)}{f'(Z)} + \dots)}$  if  $f'(Z) \neq 0$

$\epsilon^{k+1} = \epsilon^k - \epsilon^k \frac{f'(Z)(1 + \frac{1}{2}\epsilon^k \frac{f''(Z)}{f'(Z)} + \dots)}{f'(Z)(1 + \epsilon^k \frac{f''(Z)}{f'(Z)} + \dots)}$  h.o.t. for SMALL  $\epsilon^k$

now consider Taylor series of

$$g(y) \equiv \frac{1}{1+y}$$

about  $y=0$ :

$$\begin{aligned} g(y+\delta) &= g(y) + g'(y)\delta + \dots \\ &= \frac{1}{1+y} + \left(\frac{-1}{(1+y)^2}\right)\delta + \dots \\ &= 1 - \delta + \dots \quad \text{h.o.t. for } \delta \text{ small} \end{aligned}$$

hence

$$\frac{1}{1 + \underbrace{\epsilon^k \frac{f''(z)}{f'(z)}}_{f'(z)} \delta} = 1 - \epsilon^k \frac{f''(z)}{f'(z)} + \dots$$

thus

$$\begin{aligned} \epsilon^{k+1} &= \epsilon^k - \epsilon^k \left(1 + \frac{1}{2} \epsilon^k \frac{f''(z)}{f'(z)}\right) \left(1 - \epsilon^k \frac{f''(z)}{f'(z)}\right) + \dots \\ &= \epsilon^k \left(1 + \frac{1}{2} \epsilon^k \frac{f''}{f'} - \epsilon^k \frac{f''}{f'} + O((\epsilon^k)^2)\right) = -\epsilon^k + \frac{1}{2}(\epsilon^k)^2 \frac{f''}{f'} + \dots \end{aligned}$$

and

$$\begin{aligned} \epsilon^{k+1} &= \cancel{\epsilon^k} - \cancel{\epsilon^k} + \frac{1}{2}(\epsilon^k)^2 \frac{f''(z)}{f'(z)} + \dots \\ \epsilon^{k+1} &= \frac{1}{2} \frac{f''(z)}{f'(z)} (\epsilon^k)^2 + \dots \quad \text{QUADRATIC CONVERGENCE} \end{aligned}$$

Note if  $f'(z) = 0$ , then "0"  
 $\epsilon^{k+1} = \frac{1}{2} \epsilon^k$  — much slower convergence.

Newton's Method:  
The Basic Algorithm  
 multivariate case

example (n=2): 2 equations in 2 unknowns

$$\begin{cases} z_1^2 + 2z_2^2 = 22 \\ 2z_1^2 + z_2^2 = 17 \end{cases} \Rightarrow z = \begin{pmatrix} z_1 \\ z_2 \end{pmatrix}$$

Introduce

$$\begin{pmatrix} f_1(z_1, z_2) \\ f_2(z_1, z_2) \end{pmatrix} \equiv \begin{pmatrix} z_1^2 + 2z_2^2 - 22 \\ 2z_1^2 + z_2^2 - 17 \end{pmatrix};$$

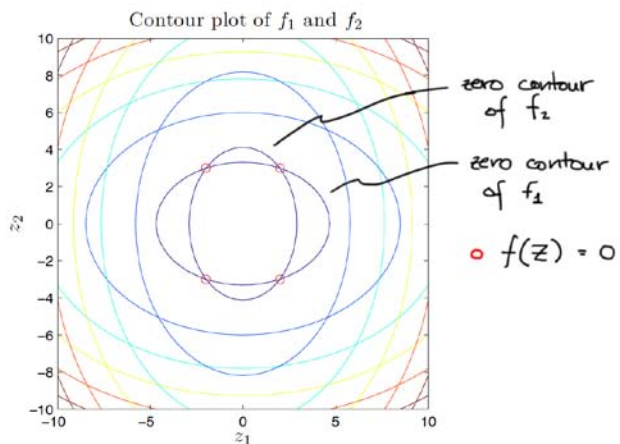
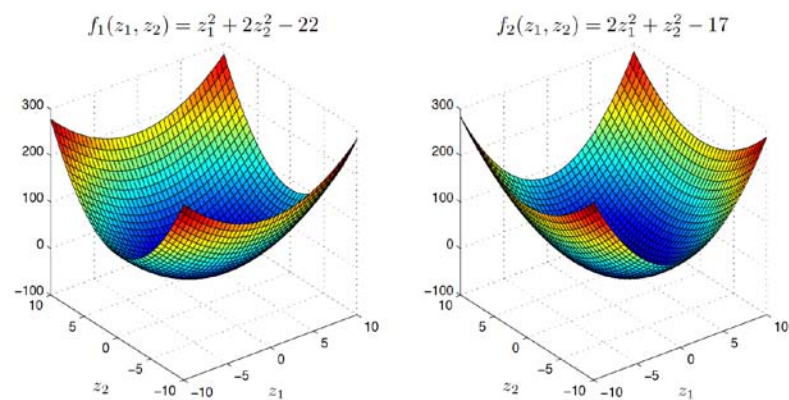
$$\begin{pmatrix} f_1(z) \\ f_2(z) \end{pmatrix} \quad \begin{array}{l} \text{a vector of} \\ \text{functions} \end{array} \quad \begin{array}{l} \text{(vector)} \\ | \\ z \end{array}$$

then

$$f(z) = 0 \quad \text{2 equations in 2 unknowns}$$

$$f\left(\begin{pmatrix} z_1 \\ z_2 \end{pmatrix}\right) = 0$$

$$\left. \begin{aligned} f_1\left(\begin{pmatrix} z_1 \\ z_2 \end{pmatrix}\right) = 0 &\Rightarrow z_1^2 + 2z_2^2 - 22 = 0 \\ f_2\left(\begin{pmatrix} z_1 \\ z_2 \end{pmatrix}\right) = 0 &\Rightarrow 2z_2^2 + z_1^2 - 17 = 0 \end{aligned} \right\}$$



the general case:

Define

$z$ : an  $n$ -vector,  $(z_1, \dots, z_n)^T \in \mathbb{R}^n$

$f$ : an  $n$ -vector of functions of  $z$

$$(f_1(z), f_2(z), \dots, f_n(z))^T$$

then we wish to find a root  $z$  such that

$$f(z) = 0 \Leftrightarrow \left. \begin{aligned} f_1((z_1, \dots, z_n)^T) &= 0 \\ f_2((z_1, \dots, z_n)^T) &= 0 \\ &\vdots \\ f_n((z_1, \dots, z_n)^T) &= 0 \end{aligned} \right\}$$

the Newton update:

Recall for  $n=1$ : to find  $\hat{z}^{k+1}$

$$f_{\text{linear}}^k(z) \equiv f'(\hat{z}^k)(z - \hat{z}^k) + f(\hat{z}^k);$$

$$f_{\text{linear}}^k(\hat{z}^{k+1}) = f'(\hat{z}^k) \underbrace{(\hat{z}^{k+1} - \hat{z}^k)}_{\delta z^k} + f(\hat{z}^k) = 0$$

⇓

$$\begin{aligned} \delta z^k &= -f(\hat{z}^k)/f'(\hat{z}^k) \\ \hat{z}^{k+1} &= \hat{z}^k + \delta z^k \end{aligned}$$

Now for general  $n$ :

$$f_{i,\text{linear}}^k(z) \equiv \frac{\partial f_i}{\partial z_1} \Big|_{\hat{z}^k} (z_1 - \hat{z}_1^k) + \frac{\partial f_i}{\partial z_2} \Big|_{\hat{z}^k} (z_2 - \hat{z}_2^k) + \dots + \frac{\partial f_i}{\partial z_n} \Big|_{\hat{z}^k} (z_n - \hat{z}_n^k) + f_i(\hat{z}^k)$$

for  $1 \leq i \leq n$ ;

then

$$f_{1,\text{linear}}^k(\hat{z}^{k+1}) \equiv \frac{\partial f_1}{\partial z_1} \Big|_{\hat{z}^k} (\hat{z}_1^{k+1} - \hat{z}_1^k) + \frac{\partial f_1}{\partial z_2} \Big|_{\hat{z}^k} (\hat{z}_2^{k+1} - \hat{z}_2^k) + \dots + \frac{\partial f_1}{\partial z_n} \Big|_{\hat{z}^k} (\hat{z}_n^{k+1} - \hat{z}_n^k) + f_1(\hat{z}^k) = 0,$$

$$f_{2,\text{linear}}^k(\hat{z}^{k+1}) \equiv \frac{\partial f_2}{\partial z_1} \Big|_{\hat{z}^k} (\hat{z}_1^{k+1} - \hat{z}_1^k) + \frac{\partial f_2}{\partial z_2} \Big|_{\hat{z}^k} (\hat{z}_2^{k+1} - \hat{z}_2^k) + \dots + \frac{\partial f_2}{\partial z_n} \Big|_{\hat{z}^k} (\hat{z}_n^{k+1} - \hat{z}_n^k) + f_2(\hat{z}^k) = 0,$$

$$f_{n,\text{linear}}^k(\hat{z}^{k+1}) \equiv \frac{\partial f_n}{\partial z_1} \Big|_{\hat{z}^k} (\hat{z}_1^{k+1} - \hat{z}_1^k) + \frac{\partial f_n}{\partial z_2} \Big|_{\hat{z}^k} (\hat{z}_2^{k+1} - \hat{z}_2^k) + \dots + \frac{\partial f_n}{\partial z_n} \Big|_{\hat{z}^k} (\hat{z}_n^{k+1} - \hat{z}_n^k) + f_n(\hat{z}^k) = 0;$$

hence

$$J_{\text{linear}}^k(\hat{z}^{k+1}) \equiv \begin{bmatrix} \frac{\partial f_1}{\partial z_1} & \frac{\partial f_1}{\partial z_2} & \dots & \frac{\partial f_1}{\partial z_n} \\ \frac{\partial f_2}{\partial z_1} & \frac{\partial f_2}{\partial z_2} & \dots & \frac{\partial f_2}{\partial z_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial z_1} & \frac{\partial f_n}{\partial z_2} & \dots & \frac{\partial f_n}{\partial z_n} \end{bmatrix} \Big|_{\hat{z}^k} \begin{bmatrix} (\hat{z}_1^{k+1} - \hat{z}_1^k) \\ (\hat{z}_2^{k+1} - \hat{z}_2^k) \\ \vdots \\ (\hat{z}_n^{k+1} - \hat{z}_n^k) \end{bmatrix} + \begin{bmatrix} f_1(\hat{z}^k) \\ f_2(\hat{z}^k) \\ \vdots \\ f_n(\hat{z}^k) \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$J(\hat{z}^k) \quad \delta \hat{z}^k \quad f(\hat{z}^k)$

or

$$J(\hat{z}^k) \delta \hat{z}^k = -f(\hat{z}^k);$$

$n$  equations in  $n$  unknowns  
**LINEAR**

then

$$\hat{z}^{k+1} = \hat{z}^k + \delta \hat{z}^k$$

Note

$$J(z) \equiv \begin{bmatrix} \frac{\partial f_1}{\partial z_1} & \frac{\partial f_1}{\partial z_2} & \dots & \frac{\partial f_1}{\partial z_n} \\ \frac{\partial f_2}{\partial z_1} & \frac{\partial f_2}{\partial z_2} & \dots & \frac{\partial f_2}{\partial z_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial z_1} & \frac{\partial f_n}{\partial z_2} & \dots & \frac{\partial f_n}{\partial z_n} \end{bmatrix} \Big|_z$$

is the  $n \times n$  Jacobian matrix:

$$\text{univariate } \underbrace{\frac{\partial f_i}{\partial z_1}}_{f'} \longrightarrow \text{multivariate } J_{ij} = \frac{\partial f_i}{\partial z_j}, \quad 1 \leq i, j \leq n.$$

the algorithm

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**Algorithm 3** Multivariate Newton algorithm without storage

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$\hat{z} \leftarrow \hat{z}^0$

while  $\|f(\hat{z})\| > \text{tol}$

$J(\hat{z})\delta\hat{z} = -f(\hat{z})$  {Solve the linearized system for  $\delta\hat{z}$ .}  $J \setminus (-f)$

$\hat{z} \leftarrow \hat{z} + \delta\hat{z}$

end

$Z \leftarrow \hat{z}$   $Z \approx \hat{z}$

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