

## 1 Order and Rate of Convergence

**1.1**  $x^{(k)} = 2 + 0.2^k$

Clearly, the algorithm converges to 2.

$$\lim_{k \rightarrow \infty} \frac{\|x^{(k+1)} - x^*\|}{\|x^{(k)} - x^*\|} = \lim_{k \rightarrow \infty} \frac{0.2^{k+1}}{0.2^k} = 0.2$$

The algorithm converge at least linearly.

$$\lim_{k \rightarrow \infty} \frac{\|x^{(k+1)} - x^*\|}{\|x^{(k)} - x^*\|^p} = \frac{0.2^{k+1}}{0.2^{kp}} = \frac{0.2}{0.2^{k(p-1)}} = \infty$$

Linear convergence.

**1.2**  $x^{(k)} = 1 + \left(\frac{4}{k}\right)^k$

$$\begin{aligned} \lim_{k \rightarrow \infty} x^{(k)} &= 1 + \lim_{k \rightarrow \infty} \left(\frac{4}{k}\right)^k = 1 \\ \lim_{k \rightarrow \infty} \frac{\|x^{(k+1)} - x^*\|}{\|x^{(k)} - x^*\|} &= \frac{(4/(k+1))^{(k+1)}}{(4/k)^k} = 0 \end{aligned}$$

The algorithm converge at least linearly.

$$\lim_{k \rightarrow \infty} \frac{\|x^{(k+1)} - x^*\|}{\|x^{(k)} - x^*\|^p} = \frac{(4/(k+1))^{(k+1)}}{(4/k)^{pk}} = \infty$$

Linear convergence.

**1.3**  $x^{(k)} = 2^{-2^k}$

$$\begin{aligned} \lim_{k \rightarrow \infty} x^{(k)} &= \lim_{k \rightarrow \infty} 2^{-2^k} = 0 \\ \lim_{k \rightarrow \infty} \frac{\|x^{(k+1)} - x^*\|}{\|x^{(k)} - x^*\|} &= \frac{2^{-2^{(k+1)}}}{2^{-2^k}} = 0 \end{aligned}$$

The algorithm converge at least linearly.

$$\lim_{k \rightarrow \infty} \frac{\|x^{(k+1)} - x^*\|}{\|x^{(k)} - x^*\|^p} = \frac{2^{-2^{(k+1)}}}{2^{-2^{pk}}} = \infty$$

Linear convergence.

**1.4**  $x^{(k+1)} = \frac{1}{2}x^{(k)} + \frac{8}{x^{(k)}}$

Suppose the algorithm converges, we have

$$\begin{aligned} x^* &= \frac{1}{2}x^* + \frac{8}{x^*} \\ x^* &= \pm 4 \end{aligned}$$

The problem gives  $x^{(0)} \leq 0$ , we have  $x^* = -4$ .

$$\lim_{k \rightarrow \infty} \frac{\|x^{(k+1)} - x^*\|}{\|x^{(k)} - x^*\|^2} = \lim_{k \rightarrow \infty} \frac{\|\frac{1}{2}x^{(k)} + \frac{8}{x^{(k)}} + 4\|}{\|x^{(k)} + 4\|^2} = \lim_{k \rightarrow \infty} \frac{1}{\|2x^{(k)}\|} = \frac{1}{8}$$

The algorithm converges at order 2.

## 2 Convergence of SD

$$f(\mathbf{x}) = x_1^2 + 2x_2^2$$

Starting at  $\mathbf{x}^{(0)} = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$

### 2.1 Recurrence Relation

We verify that using numeric example for the first 5 iterations.

$$\begin{aligned} x^{(1-5)} &= \begin{pmatrix} 2 & 0.6667 & 0.2222 & 0.0741 & 0.0247 \\ 1 & -0.3333 & 0.1111 & -0.0370 & 0.0123 \end{pmatrix} \\ x_k^{(1-5)} &= \begin{pmatrix} 2 & 0.6667 & 0.2222 & 0.0741 & 0.0247 \\ 1 & -0.3333 & 0.1111 & -0.0370 & 0.0123 \end{pmatrix} \end{aligned}$$

which are equivalent. Similarly,

$$\begin{aligned} f^{(1-4)}/9 &= \begin{pmatrix} 0.6667 & 0.0741 & 0.0082 & 9.1111\text{e-}04 \end{pmatrix} \\ f^{(2-5)} &= \begin{pmatrix} 0.6667 & 0.0741 & 0.0082 & 9.1449\text{e-}04 \end{pmatrix} \end{aligned}$$

They are roughly equal. The error probably coming from round-off for division during the numeric computation.

*See attached code for the validation of first 100 iteration.*

### 2.2 $\mathbf{x}^*$ and Convergence Ratio

$$\begin{aligned} \lim_{k \rightarrow \infty} x^{(k)} &= \lim_{k \rightarrow \infty} \frac{1}{3^k} \begin{pmatrix} 2 \\ (-1)^k \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \\ \lim_{x \rightarrow \infty} \frac{\|x^{(k+1)} - x^*\|}{\|x^{(k)} - x^*\|} &= \lim_{x \rightarrow \infty} \frac{\left\| \begin{pmatrix} \frac{2}{3^{k+1}} \\ \frac{-1^{k+1}}{3^{k+1}} \end{pmatrix} \right\|}{\left\| \begin{pmatrix} \frac{2}{3^k} \\ \frac{-1^k}{3^k} \end{pmatrix} \right\|} = \frac{1}{3} \end{aligned}$$

## 3 SD / NR

$$f(\mathbf{x}) = x_2(x_1 - 2)^2 + 10(x_2 - 1)^2$$

with  $x^{(0)} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ .

### 3.1 SD

- $k = 1$

$$d = \begin{pmatrix} 2 \\ -1 \end{pmatrix}$$

$$x^{(1)} = \begin{pmatrix} 2\alpha + 1 \\ 1 - \alpha \end{pmatrix}$$

$$\phi = 10\alpha^2 - (2\alpha - 1)^2(\alpha - 1)$$

Solve for optimal  $\alpha$

$$\alpha = 0.1460$$

$$x^{(1)} = \begin{pmatrix} 1.2920 \\ 0.8540 \end{pmatrix}$$

- $k = 2$

$$d = \begin{pmatrix} 1.2093 \\ 2.4186 \end{pmatrix}$$

$$x^{(2)} = \begin{pmatrix} 1.2093\alpha + 1.2920 \\ 2.4186\alpha + 0.8540 \end{pmatrix}$$

$$\phi = 10(2.4186\alpha - 0.1460)^2 + (1.2093\alpha - 0.7080)^2(2.4186\alpha + 0.8540)$$

Solve for optimal  $\alpha$

$$\alpha = 0.0653$$

$$x^{(2)} = \begin{pmatrix} 1.3710 \\ 1.0120 \end{pmatrix}$$

### 3.2 NR

1.  $k = 1$

$$\begin{aligned}
 H_f &= \begin{pmatrix} 2 & -2 \\ -2 & 20 \end{pmatrix} \\
 g &= \begin{pmatrix} -2 \\ 1 \end{pmatrix} \\
 d &= \begin{pmatrix} 1.0556 \\ 0.0556 \end{pmatrix} ca \\
 x^{(1)} &= \begin{pmatrix} 1.0556 \alpha + 1 \\ 0.0556 \alpha + 1 \end{pmatrix} \\
 \phi &= (1.0556 \alpha - 1)^2 (0.0556 \alpha + 1) + 0.0309 \alpha^2 \\
 \alpha &= 0.9230 \\
 x^{(1)} &= \begin{pmatrix} 1.9743 \\ 1.0513 \end{pmatrix}
 \end{aligned}$$

2.  $k = 2$

$$\begin{aligned}
 H_f &= \begin{pmatrix} 2.1026 & -0.0514 \\ -0.0514 & 20 \end{pmatrix} \\
 g &= \begin{pmatrix} -0.0540 \\ 1.0263 \end{pmatrix} \\
 d &= \begin{pmatrix} 0.0244 \\ -0.0512 \end{pmatrix} x^{(2)} = \begin{pmatrix} 0.0244 \alpha + 1.9743 \\ 1.0513 - 0.0512 \alpha \end{pmatrix} \\
 \phi &= 10 (0.0512 \alpha - 0.0513)^2 - (0.0512 \alpha - 1.0513) (0.0244 \alpha - 0.0257)^2 \\
 \alpha &= 1.0017 \\
 x^{(2)} &= \begin{pmatrix} 1.9988 \\ 0.9999 \end{pmatrix}
 \end{aligned}$$

# Prob 2 Code

```
syms x1 x2 alpha k
```

```
f = x1^2 + 2*x2^2
```

$$f = x_1^2 + 2x_2^2$$

```
g = gradient(f)
```

$$g = \begin{pmatrix} 2x_1 \\ 4x_2 \end{pmatrix}$$

```
x = [x1;x2]
```

$$x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

```
x_val = [2;1]
```

$$x_{\text{val}} = \begin{matrix} 2 \times 1 \\ 2 \\ 1 \end{matrix}$$

```
size = 100
```

```
size = 100
```

```
g_val = subs(g, x, x_val)
```

$$g_{\text{val}} = \begin{pmatrix} 4 \\ 4 \end{pmatrix}$$

```
all_x = zeros(2, size + 1)
```

$$all\_x = \begin{matrix} 2 \times 101 \\ \begin{matrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \dots \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{matrix} \end{matrix}$$

```
all_f = zeros(1, size + 1)
```

$$all\_f = \begin{matrix} 1 \times 101 \\ \begin{matrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \dots \end{matrix} \end{matrix}$$

```
all_x(:, 1) = x_val
```

$$all\_x = \begin{matrix} 2 \times 101 \\ \begin{matrix} 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \dots \end{matrix} \end{matrix}$$

```
1 0 0 0 0 0 0 0 0 0 0 0 0
```

```
all_f(1) = subs(f, x, x_val)
```

```
all_f = 1x101
```

```
6 0 0 0 0 0 0 0 0 0 0 0 0...
```

```
threshold = 0.001
```

```
threshold = 1.0000e-03
```

```
echo off;
for i = 2:size+1
    d = - g_val;
    x_new = x_val + alpha * d;

    phi = subs(f, x, x_new);

    alpha_vals = solve([gradient(phi) == 0, alpha >= 0], alpha);
    phi_vals = double(subs(phi, alpha, alpha_vals));
    [phi_vals, alpha_idx] = sort(phi_vals);

    alpha_val = alpha_vals(alpha_idx(1));

    x_val = subs(x_new, alpha, alpha_val);
    g_val = subs(g, x, x_val);

    all_x(:, i) = x_val;
    all_f(i) = subs(f, x, x_val);
    % norm(g_val), threshold *
end
echo on;
```

```
syms k
```

```
xk = 1/3^k * [2; (-1)^k]
```

```
xk =
```

$$\begin{pmatrix} \frac{2}{3^k} \\ \frac{(-1)^k}{3^k} \end{pmatrix}$$

```
xks = double(subs(xk, 0:size))
```

```
xks = 2x101
```

```
2.0000 0.6667 0.2222 0.0741 0.0247 0.0082 0.0027 0.0009 ...
1.0000 -0.3333 0.1111 -0.0370 0.0123 -0.0041 0.0014 -0.0005
```

```
sym(xks(:, 1:5))
```

```
ans =
```

$$\begin{pmatrix} 2 & 0.6667 & 0.2222 & 0.0741 & 0.0247 \\ 1 & -0.3333 & 0.1111 & -0.0370 & 0.0123 \end{pmatrix}$$

```
sym(all_x(:, 1:5))
```

```
ans =
```

$$\begin{pmatrix} 2 & 0.6667 & 0.2222 & 0.0741 & 0.0247 \\ 1 & -0.3333 & 0.1111 & -0.0370 & 0.0123 \end{pmatrix}$$

```
% compare the computed x's from sd and the equation
all(all(all_x - xks < 0.0001))
```

```
ans = logical
      1
```

```
sym(all_f(1:4) / 9)
```

```
ans = (0.6667 0.0741 0.0082 9.1111e-04)
```

```
sym(all_f(2:5))
```

```
ans = (0.6667 0.0741 0.0082 9.1449e-04)
```

```
% compare the recurrence relation of f
all(all_f(1:size-1)/9 - all_f(2:size) < 0.0001)
```

```
ans = logical
      1
```

## Prob 3 - SD Code

```
syms x [1 2]
syms alpha
```

```
f = x2 * (x1 - 2)^2 + 10 * (x2 - 1)^2
```

$$f = x_2 (x_1 - 2)^2 + 10 (x_2 - 1)^2$$

```
g = gradient(f, x)
```

$$g = \begin{pmatrix} x_2 (2x_1 - 4) \\ 20x_2 + (x_1 - 2)^2 - 20 \end{pmatrix}$$

```
x = transpose(x)
```

$$x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

```
x_val = [1;1]
```

$$x_{\text{val}} = \begin{matrix} 2 \times 1 \\ 1 \\ 1 \end{matrix}$$

```
g_val = subs(g, x, x_val)
```

$$g_{\text{val}} = \begin{pmatrix} -2 \\ 1 \end{pmatrix}$$

```
threshold = 0.001
```

```
threshold = 1.0000e-03
```

```
for i = 1:2
    d = - g_val
    x_new = x_val + alpha * d

    phi = subs(f, x, x_new)

    alpha_vals = solve([gradient(phi) == 0, alpha >= 0], alpha)
    phi_vals = double(subs(phi, alpha, alpha_vals))
    [phi_vals, alpha_idx] = sort(phi_vals)

    alpha_val = alpha_vals(alpha_idx(1))

    x_val = subs(x_new, alpha, alpha_val)
```



```
g_val = subs(g, x, x_val)
```

```
end
```

```
d =
```

$$\begin{pmatrix} 2 \\ -1 \end{pmatrix}$$

```
x_new =
```

$$\begin{pmatrix} 2\alpha + 1 \\ 1 - \alpha \end{pmatrix}$$

$$\text{phi} = 10\alpha^2 - (2\alpha - 1)^2(\alpha - 1)$$

```
alpha_vals =
```

$$\begin{pmatrix} 0.1460 \\ 2.8540 \end{pmatrix}$$

```
phi_vals = 2x1
```

```
0.6412
```

```
40.3588
```

```
phi_vals = 2x1
```

```
0.6412
```

```
40.3588
```

```
alpha_idx = 2x1
```

```
1
```

```
2
```

```
alpha_val = 0.1460
```

```
x_val =
```

$$\begin{pmatrix} 1.2920 \\ 0.8540 \end{pmatrix}$$

```
g_val =
```

$$\begin{pmatrix} -1.2093 \\ -2.4186 \end{pmatrix}$$

```
d =
```

$$\begin{pmatrix} 1.2093 \\ 2.4186 \end{pmatrix}$$

```
x_new =
```

$$\begin{pmatrix} 1.2093\alpha + 1.2920 \\ 2.4186\alpha + 0.8540 \end{pmatrix}$$

$$\text{phi} = 10(2.4186\alpha - 0.1460)^2 + (1.2093\alpha - 0.7080)^2(2.4186\alpha + 0.8540)$$

```
alpha_vals = 0.0653
```

```
phi_vals = 0.4018
```

```
phi_vals = 0.4018
```

```
alpha_idx = 1
```

```
alpha_val = 0.0653
```

```
x_val =
```

$$\begin{pmatrix} 1.3710 \\ 1.0120 \end{pmatrix}$$

```
g_val =
```

$$\begin{pmatrix} -1.2731 \\ 0.6366 \end{pmatrix}$$

## Prob 3 - NR Code

```
syms x [1 2]
syms alpha
```

```
f = x2 * (x1 - 2)^2 + 10 * (x2 - 1)^2
```

$$f = x_2 (x_1 - 2)^2 + 10 (x_2 - 1)^2$$

```
g = gradient(f, x)
```

$$g = \begin{pmatrix} x_2 (2x_1 - 4) \\ 20x_2 + (x_1 - 2)^2 - 20 \end{pmatrix}$$

```
h = hessian(f)
```

$$h = \begin{pmatrix} 2x_2 & 2x_1 - 4 \\ 2x_1 - 4 & 20 \end{pmatrix}$$

```
x = transpose(x)
```

$$x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

```
x_val = [1;1]
```

```
x_val = 2x1
      1
      1
```

```
for i = 1:2
    h_val = subs(h, x, x_val)
    g_val = subs(g, x, x_val)

    d_val = - inv(h_val) * g_val

    x_new = x_val + alpha * d_val

    phi = subs(f, x, x_new)

    alpha_vals = solve([gradient(phi) == 0, alpha >= 0], alpha)
    phi_vals = double(subs(phi, alpha, alpha_vals))
    [phi_vals, alpha_idx] = sort(phi_vals)

    alpha_val = alpha_vals(alpha_idx(1))
```

```

    x_val = subs(x_new, alpha, alpha_val)
    g_val = subs(g, x, x_val)
end

```

```

h_val =

$$\begin{pmatrix} 2 & -2 \\ -2 & 20 \end{pmatrix}$$

g_val =

$$\begin{pmatrix} -2 \\ 1 \end{pmatrix}$$

d_val =

$$\begin{pmatrix} 1.0556 \\ 0.0556 \end{pmatrix}$$

x_new =

$$\begin{pmatrix} 1.0556\alpha + 1 \\ 0.0556\alpha + 1 \end{pmatrix}$$

phi = (1.0556  $\alpha$  - 1)2 (0.0556  $\alpha$  + 1) + 0.0309  $\alpha$ 2
alpha_vals = 0.9230
phi_vals = 0.0270
phi_vals = 0.0270
alpha_idx = 1
alpha_val = 0.9230
x_val =

$$\begin{pmatrix} 1.9743 \\ 1.0513 \end{pmatrix}$$

g_val =

$$\begin{pmatrix} -0.0540 \\ 1.0263 \end{pmatrix}$$

h_val =

$$\begin{pmatrix} 2.1026 & -0.0514 \\ -0.0514 & 20 \end{pmatrix}$$

g_val =

$$\begin{pmatrix} -0.0540 \\ 1.0263 \end{pmatrix}$$

d_val =

$$\begin{pmatrix} 0.0244 \\ -0.0512 \end{pmatrix}$$

x_new =

$$\begin{pmatrix} 0.0244\alpha + 1.9743 \\ 1.0513 - 0.0512\alpha \end{pmatrix}$$

phi = 10 (0.0512  $\alpha$  - 0.0513)2 - (0.0512  $\alpha$  - 1.0513) (0.0244  $\alpha$  - 0.0257)2
alpha_vals =

$$\begin{pmatrix} 586.2227 \\ 1.0017 \end{pmatrix}$$

phi_vals = 2×1

```

```

103 ×
    3.0670
    0.0000
phi_vals = 2×1
103 ×
    0.0000
    3.0670
alpha_idx = 2×1
    2
    1
alpha_val = 1.0017
x_val =
    (1.9988)
    (0.9999)
g_val =
    (-0.0024)
    (-0.0012)

```