

# HW4

Yilun Zhang

October 26, 2015

## Exercise 1

Method 1, let  $f(x) = dx - 1$ , so the solution of  $f(x) = 0$  is  $1/d$ . The matlab code is

```
function [x l] = newroot(x0,a, maxiter, tol)
fold = a*x0-1;
fnew = fold-1;
x=x0;
iter=0;
while( abs(fnew-fold)>tol && iter< maxiter)
fold = fnew;
iter=iter+1;
x=x-(a*x-1)/a;
fnew=a*x+1;
end
l=iter;
end
```

Start value of both 0.3 and 1 both converge in 1 step, this is because

$$x_{k+1} = x_k - \frac{ex_k - 1}{e} = \frac{1}{e}$$

Method 2, let  $g(x) = 1/x - d$ , the update steps is

$$x_{k+1} = x_k + (1/x_k - d)x_k^2 = 2x_k - dx_k^2$$

Then if the start value is 0.3 it will converge to the true value, if the start value is 1, it will not converge.

## Exercise 2

**a**

$$\begin{aligned} x_{k+1} &= x_k - \frac{x_k^q}{qx_k^{q-1}} \\ &= x_k \left(1 - \frac{1}{q}\right) \end{aligned}$$

Therefore, it's Q-linearly converge.

**b**

Let  $f(x) = \|x\|_2^\beta$

$$\nabla f = \beta \|x\|_2^{\beta-2} x$$

$$\nabla^2 f = \beta \|x\|_2^{\beta-4} (\|x\|_2^2 I + (\beta - 2)xx')$$

Because  $(A + CBC')^{-1} = A^{-1} - A^{-1}C(B^{-1} + C'A^{-1}C)^{-1}C'A^{-1}$

$$(\nabla^2 f)^{-1} = \frac{1}{\beta \|x\|_2^{\beta-2}} \left( I - \frac{\beta - 2}{\beta - 1} \frac{1}{\|x\|_2^2} xx' \right)$$

The pure Newton method:

$$\begin{aligned} x_{k+1} &= x_k - (\nabla^2 f(x_k))^{-1} \nabla f(x_k) \\ &= x_k - \frac{1}{\beta \|x_k\|_2^{\beta-2}} \left( I - \frac{\beta - 2}{\beta - 1} \frac{1}{\|x_k\|_2^2} x_k x_k' \right) \beta \|x_k\|_2^{\beta-2} x_k \\ &= \frac{\beta - 2}{\beta - 1} x_k \end{aligned}$$

Then,  $x_k$  converges Q-linearly when  $\beta > 3/2$  globally. When  $\beta = 2$ , it converges in 1 step. When  $\beta \leq 1$ , it will not converge for start point  $x_0 \neq 0$ .

**c**

The Newton method:

$$\begin{aligned}
x_{k+1} &= x_k - \alpha_k (\nabla^2 f(x_k))^{-1} \nabla f(x_k) \\
&= x_k - \alpha_k \frac{1}{\beta \|x_k\|_2^{\beta-2}} \left( I - \frac{\beta-2}{\beta-1} \frac{1}{\|x_k\|_2^2} x_k x_k' \right) \beta \|x_k\|_2^{\beta-2} x_k \\
&= \left( 1 - \frac{\alpha_k}{\beta-1} \right) x_k
\end{aligned}$$

The Armijo condition:  $f(x_k + \alpha_k p_k) \leq f(x_k) + c_1 \alpha_k [\nabla f(x_k)] p_k$ . This implies

$$\left[ 1 - \left| 1 - \frac{\alpha_k}{\beta-1} \right|^\beta \right] \|x_k\|_2^\beta \leq c_1 \alpha_k \frac{\beta}{\beta-1} \|x_k\|_2^\beta$$

as long as  $\|x_k\|_2$  is not zero, we can see  $\alpha_k$  doesn't depend on  $k$ , then let  $\alpha_k \equiv \alpha$  where  $\alpha$  satisfies

$$\left[ 1 - \left| 1 - \frac{\alpha}{\beta-1} \right|^\beta \right] \leq c_1 \alpha \frac{\beta}{\beta-1}$$

Then,  $x_k$  converges Q-linearly when  $|1 - \frac{\alpha}{\beta-1}| < 1$  globally. When  $\frac{\alpha}{\beta-1} = 1$ , it converges in 1 step. When  $\beta \leq 1$ , it will not converge for start point  $x_0 \neq 0$ , because  $|1 - \frac{\alpha_k}{\beta-1}|$  always larger than 1. when  $\alpha > 0$  and  $\beta > 1$ , it will Q-linearly converge.

## Exercise 3

Prove this by induction.

For  $k=0, i=0$ , need to show  $D^1 q^0 = p^0$

$$D^1 q^0 = (D^0 + \frac{y^0 y^{0'}}{q^{0'} y^0}) q^0 = D^0 q^0 + \frac{y^0 y^{0'} q^0}{q^{0'} y^0} = p^0$$

Suppose  $k-1$  hold,

$$D^{k+1} q^i = (D^k + \frac{y^k y^{k'}}{q^{k'} y^k}) q^i = D^k q^i + \frac{y^k y^{k'} q^i}{q^{k'} y^k} = p^i$$

if  $i < k$

$$LHS = D^k q^i + \frac{y^k y^{k'} q^i}{q^{k'} y^k} = p^i + \frac{1}{q^{k'} y^k} (p^{k'} q_i - q^{k'} p^i)$$

since  $Q p^i = q^i$ , and  $Q$  is positive definite thus invertible.  $p^{k'} q_i - q^{k'} p^i = 0$ . So the LHS =  $p^i$

if  $i = k$

$$D^{k+1}q^k = (D^k + \frac{y^k y^{k'}}{q^{k'} y^k})q^k = D^k q^k + \frac{y^k y^{k'} q^k}{q^{k'} y^k} = p^k$$

therefore , we have  $D^{k+1}q^i = p^i$

Since  $D^n q^i = p^i$  and  $Q^{-1}q^i = p^i$  for  $i = 0, 1, \dots, n-1$ . Let  $X = [q^0, q^1, \dots, q^{n-1}]$ . Because  $\{q^i\}$  are linearly independent.  $X$  is full rank.  $(D^n - Q^{-1})X = 0$  implies  $D^n = Q^{-1}$ .

## Exercide 4

I already implemented BFGS in last homework, the matlab code:

```
function [x,fc,itc] = newton(obj,i,maxit,tol,qusi,eps)
c=0.0001;
x0=p00_start(i,p00_n(i));
n=p00_n(i);
[fc,gc,hc]=obj(i,x0);
[P,D]=eig(hc);
if(min(diag(D))<eps)
D=max(D,eps*eye(n));
hc=P*D*P';
end
g0=gc;
xc=x0;
itc=1;
H=inv(hc);
while(norm(gc) > tol*norm(g0) & itc <= maxit)
p=-H*gc;
alpha=1.0; xt=xc+alpha*p; ft=obj(i,xt);
fg= fc + c*alpha*(p'*gc);
cout=1;
while(ft > fg) % check Armijo condition
alpha=alpha*0.9;
```

```

fg= fc + c*alpha*(gc'*p);
xt=xc+alpha*p;
ft=obj(i,xt);
cout=cout+1;
if(cout>20)
break
end
end
xc=xt;
go=gc;
[fc,gc,hc]=obj(i,xc);
itc=itc+1;
if(qusi)
s=alpha*p;
y=gc-go;
pho=1/(y'*s);
H=(eye(n)-pho*s*y')*H*(eye(n)-pho*y*s')+pho*s*s';

else
[P,D]=eig(hc);
if(min(diag(D))<eps)
D=diag(D);
D(D<=1e-8)=eps;
D=diag(D);
hc=P*D*P'
H=inv(hc);
end
end
end
x=xc;

```

prob ID	$f^*$	number of iter
1	4.46500818473627	20001
2	0.242677406804875	412
3	1.12793276962393e-08	4
4	0.0032230846252832	20001
5	3.97089247824700e-14	6
6	6.37118858804940e-12	45
7	0.00228767657879530	3076
8	2.49997654797500e-06	59
9	0.489395214700814	16
10	25364766502.5965	20001
11	85822.2019034065	251
12	8.39437213309053e-20	13
13	8.42718143579277e-16	13
14	4.93038065763132e-32	2
15	1.63898347747199e-06	14859
16	1.81352665939531e-08	20001
17	7.79452679362206	20001
18	5.93799296903267e-14	13

Table 1: result of Newton method

end

Table 1 is the result of pure Newton method(with eigen value correction), Table 2 shows result of quasi newton.

We can see all problems converge but number 2, 5, 10, 16 return a NA. I guess this is because  $\rho = 1/y's$ , when  $y$  and  $s$  are both very small, it will cause overflow in calculation in Matlab.

prob ID	$f^*$	number of iter
1	4.75669872116614e-07	36
2	NaN	3
3	1.12793283165719e-08	3
4	8.23648565011059e-10	203
5	NaN	3
6	3.52702079374637e-18	7
7	0.00228767029059872	20
8	2.49997507553623e-06	8
9	0.489395214700777	8
10	NaN	8
11	85822.2162472798	18
12	32.8349996345640	7
13	1.80370537607437e-13	5
14	4.93038065763132e-32	2
15	1.67963844460425e-05	15
16	NaN	5
17	7.87479283486014	13
18	0.444444444430489	203

Table 2: result of quasi Newton method