CIE 6010/MDS 6118 Midterm Exam: Part 2

Problem Solving in Matlab: CIE 25 points / MDS 35 points

Take-home. Due 5:00pm, November 9, 2018

— Room 511, Dao Yuan Building —

No discussions are allowed except possibly with the instructor or TAs. Please include this page as the cover page for your submission.

Name (print): _	
Student Number:	
Course Number:	

Your Matlab function should have the following interface:

```
function [x,hist,iter] = myphase(A,B,d,x0,tol,maxit,itGN)
% Solve the phase retrieval least squares problem
         \min f(x) = 0.5 * norm(|(A + iB)x|^2 - d)^2
% input:
응
      A, B, d = real-valued data
          x0 = initial quess
         tol = tolerance
       maxit = maximum number of iterations in total
        itGN = min number of Gauss-Newton iterations before
응
               possibly switching to Newton iterations
% output:
응
           x = computed solution
        hist = history vector for c_k = norm(g_k)/(1+f_k)
응
        iter = total number of iterations taken
```

A copy of my code myphase.m

```
function [ x,hist,iter ] = myphase(A,B,d,x0,tol,maxit,itGN )
% Solve the phase retrieval least squares problem:
       \min f(x) = 0.5 * norm(|(A + iB)x|^2 - d)^2
%
% Input:
    A, B, d = real-valued data
          x0 = initial guess
%
         tol = tolerance
%
        maxit = maximum number of iterations in total
        itGN = min number of Gauss-Newton iterations before
% Output:
%
            x = computed solution
%
         hist = history vector for c_k = norm(g_k) / (1 + f_k)
%
         iter = total number of iterations taken
x = x0;
hist = [];
itGN = itGN*5;
for iter = 1:maxit
    [J,r] = Jacobian_compute(A,B,d,x);
   nabla = J' * r;
   nabla2_fake = J' * J;
   f = norm(r)^2*0.5;
   gnrm = norm(nabla);
   crit = gnrm / (1 + f);
   fprintf('iter %i: f(x) = %9.2e \text{ gnrm} = %9.2e \text{ crit} = %9.2e \n',...
      iter, f, gnrm, crit)
   hist(iter) = crit;
   if crit <= tol, break;end</pre>
   if iter <= itGN</pre>
       m = nabla2_fake \ nabla;
       x = x - m;
   else
       S = 2*(r'.*A'*A+r'.*B'*B);
       nabla2 = nabla2_fake + S;
       m = nabla2 \ nabla;
       x = x - m;
   end
end
end
function [J,r] = Jacobian_compute(A,B,d,x)
Ax = A*x;
Bx = B*x;
r = Ax.^2 + Bx.^2-d;
J = 2*A.*Ax+2*B.*Bx;
end
```

MATLAB Screen Printout

```
Command Window
  >> test_phase
  (return for default) n = 1000
  Phase: n = 1000, m = 5000
  iter: 1 f(x) = 1.503e+10 gnrm = 5.426e+08 crit = 3.610e-02
  iter: 2 f(x) = 4.677e+09 gnrm = 1.606e+09 crit = 3.433e-01
           3 f(x) = 3.591e+08 gnrm = 2.010e+08 crit = 5.597e-01
  iter:
  iter: 4 f(x) = 1.588e+08 gnrm = 3.127e+07 crit = 1.970e-01
  iter: 5 f(x) = 1.227e+08 gnrm = 3.726e+07 crit = 3.038e-01 iter: 86 f(x) = 8.722e+06 gnrm = 1.634e-04 crit = 1.873e-11
  iter: 87 f(x) = 8.722e+06 gnrm = 1.262e-04 crit = 1.447e-11
  iter: 88 f(x) = 8.722e+06 gnrm = 9.595e-05 crit = 1.100e-11
  iter: 89 f(x) = 8.722e+06 gnrm = 7.376e-05 crit = 8.457e-12
  --- yzphase: iter 89 relerr = 1.359e-01 time 11.155 --
  Current plot held
         1 f(x) = 1.503e+10 gnrm = 5.426e+08 crit = 3.610e-02
  iter:
          2 f(x) = 4.677e+09 gnrm = 1.606e+09 crit = 3.433e-01
  iter: 3 f(x) = 3.591e+08 gnrm = 2.010e+08 crit = 5.597e-01
  iter: 4 f(x) = 1.588e+08 gnrm = 3.127e+07 crit = 1.970e-01 iter: 5 f(x) = 1.227e+08 gnrm = 3.726e+07 crit = 3.038e-01
  iter: 6 f(x) = 1.066e+08 gnrm = 3.733e+07 crit = 3.501e-01
  iter: 7 f(x) = 1.008e+08 gnrm = 4.149e+07 crit = 4.116e-01
  iter: 8 f(x) = 9.249e+07 gnrm = 4.637e+07 crit = 5.013e-01 iter: 9 f(x) = 8.114e+07 gnrm = 5.287e+07 crit = 6.517e-01
  iter: 10 f(x) = 7.360e+07 gnrm = 8.889e+07 crit = 1.208e+00
  iter: 11 f(x) = 7.499e+07 gnrm = 1.318e+08 crit = 1.758e+00
  iter: 12 f(x) = 4.128e+07 gnrm = 9.818e+07 crit = 2.378e+00
  iter: 13 f(x) = 1.312e+07 gnrm = 3.279e+07 crit = 2.499e+00
  iter: 14 f(x) = 8.852e+06 gnrm = 2.894e+06 crit = 3.270e-01
  iter: 15 f(x) = 8.732e+06 gnrm = 1.600e+05 crit = 1.833e-02
  iter: 16 f(x) = 8.724e+06 gnrm = 5.345e+04 crit = 6.127e-03
  iter: 17 f(x) = 8.723e+06 gnrm = 3.015e+04 crit = 3.457e-03
  iter: 18 f(x) = 8.722e+06 gnrm = 5.360e+02 crit = 6.145e-05
  iter: 19 f(x) = 8.722e+06 gnrm = 1.892e-02 crit = 2.169e-09
  iter: 20 f(x) = 8.722e+06 gnrm = 1.363e-06 crit = 1.563e-13
  --- yzphase: iter 20 relerr = 1.359e-01 time 3.51279 --
  iter 1: f(x) = 1.50e+10 \text{ gnrm} = 5.43e+08 \text{ crit} = 3.61e-02
  iter 2: f(x) = 4.68e+09 \text{ gnrm} = 1.61e+09 \text{ crit} = 3.43e-01
  iter 3: f(x) = 3.59e+08 gnrm = 2.01e+08 crit = 5.60e-01
  iter 4: f(x) = 1.59e+08 gnrm = 3.13e+07 crit = 1.97e-01
  iter 86: f(x) = 8.72e+06 \text{ gnrm} = 1.63e-04 \text{ crit} = 1.87e-11
  iter 87: f(x) = 8.72e+06 \text{ gnrm} = 1.26e-04 \text{ crit} = 1.45e-11
  iter 88: f(x) = 8.72e+06 \text{ gnrm} = 9.59e-05 \text{ crit} = 1.10e-11
  iter 89: f(x) = 8.72e+06 \text{ gnrm} = 7.38e-05 \text{ crit} = 8.46e-12
  --- myphase: iter 89 relerr = 1.359e-01 time 9.38611 ---
  Current plot held
  iter 1: f(x) = 1.50e+10 \text{ gnrm} = 5.43e+08 \text{ crit} = 3.61e-02
  iter 2: f(x) = 4.68e+09 \text{ gnrm} = 1.61e+09 \text{ crit} = 3.43e-01
  iter 3: f(x) = 3.59e+08 gnrm = 2.01e+08 crit = 5.60e-01
  iter 4: f(x) = 1.59e+08 \text{ gnrm} = 3.13e+07 \text{ crit} = 1.97e-01
  iter 5: f(x) = 1.23e+08 gnrm = 3.73e+07 crit = 3.04e-01
  iter 6: f(x) = 1.07e+08 \text{ gnrm} = 3.73e+07 \text{ crit} = 3.50e-01
  iter 7: f(x) = 1.01e+08 \text{ gnrm} = 4.15e+07 \text{ crit} = 4.12e-01
  iter 8: f(x) = 9.25e+07 \text{ gnrm} = 4.64e+07 \text{ crit} = 5.01e-01
  iter 9: f(x) = 8.11e+07 \text{ gnrm} = 5.29e+07 \text{ crit} = 6.52e-01
  iter 10: f(x) = 7.36e+07 \text{ gnrm} = 8.89e+07 \text{ crit} = 1.21e+00
  iter 11: f(x) = 7.50e+07 \text{ gnrm} = 1.32e+08 \text{ crit} = 1.76e+00
  iter 12: f(x) = 4.13e+07 \text{ gnrm} = 9.82e+07 \text{ crit} = 2.38e+00
  iter 13: f(x) = 1.31e+07 gnrm = 3.28e+07 crit = 2.50e+00
  iter 14: f(x) = 8.85e+06 gnrm = 2.89e+06 crit = 3.27e-01
  iter 15: f(x) = 8.73e+06 \text{ gnrm} = 1.60e+05 \text{ crit} = 1.83e-02
  iter 16: f(x) = 8.72e+06 \text{ gnrm} = 5.35e+04 \text{ crit} = 6.13e-03
  iter 17: f(x) = 8.72e+06 \text{ gnrm} = 2.40e+03 \text{ crit} = 2.75e-04
  iter 18: f(x) = 8.72e+06 \text{ gnrm} = 2.09e-01 \text{ crit} = 2.39e-08
  iter 19: f(x) = 8.72e+06 \text{ gnrm} = 9.81e-06 \text{ crit} = 1.12e-12
  --- myphase: iter 19 relerr = 1.359e-01 time 2.86407 ---
                                                     3
f_{\underline{x}} >>
```

Figure 1: Matlab Screen Printout for file test_phase.m

MATLAB Plots

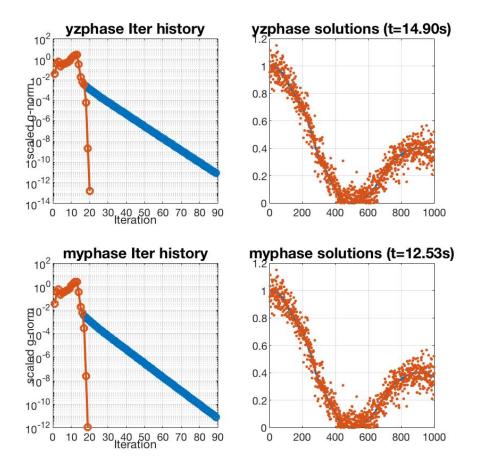


Figure 2: Matlab plots from file test_phase.m

Summary for Phase Retrieval MATLAB Problem

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I. INTRODUCTION

¹ This project is about solving a *phase retrieval least* squares problem given as follows:

$$\begin{aligned} \min_{x \in \mathbb{R}^n} & & f(x) \triangleq \frac{1}{2} \| r(x) \|^2 \\ \text{such that} & & r_i(x) = (a_i^{\mathrm{T}} x)^2 + (b_i^{\mathrm{T}} x)^2 - d_i, i = 1, 2, \dots, m, \end{aligned}$$

where A+iB are known matrix with $a_i^{\rm T}, b_i^{\rm T}$ denoting the i-th row of A and B, respectively, d is our observation vector with d_i denoting its i-th entry. Also define the matrices $E_i := a_i a_i^{\rm T} + b_i b_i^{\rm T}, \ i = 1, \ldots, m$.

After computations, we obtain that:

$$J^{\mathrm{T}}(x) \cdot J(x) = 2 \sum_{i=1}^{m} E_i(xx^{\mathrm{T}}) E_i$$
 (2a)

$$S(x) = 2\sum_{i=1}^{m} E_i \cdot r_i(x)$$
 (2b)

$$\nabla f(x) = S(x) \cdot x \tag{2c}$$

$$\nabla^2 f(x) = J^{\mathrm{T}}(x) \cdot J(x) + S(x) \tag{2d}$$

The Gauss-Newton direction and Newton direction at a given x is given by:

$$d_{\text{GN}} = -[J^{\text{T}}(x) \cdot J(x)]^{-1} \nabla f(x)$$
(3a)

$$d_{\text{Newton}} = -[\nabla^2 f(x)]^{-1} \nabla f(x) \tag{3b}$$

The suggested algorithmatic approach is listed as follows:

II. DECISIONS & ARRANGEMENTS ON COMPUTATION

- a) Choice on itGN: As n=1000, adopting Newton's method too soon is not a good idea. We decide to change the minimum iteration taken to swithch to Newton's method as itGN=15.
- b) Compute S(x) and $J^{\mathrm{T}}(x) \cdot J(x)$ smartly: We observe that the barrier for computation is due to the Hessian matrix and $J^{\mathrm{T}}(x) \cdot J(x)$. Pre-save the data matrix $\{E_i\}_{i=1,\ldots,m}$ is not a good idea, since it will occupy too much space such that MATLAB cannot handle it. Therefore we should not implement formula (2a) and (2b) directly but re-formulate and implement it in terms of A, B, d, x.
- c) Pre-set Function handles: Another barrier is that computing the Hessian and Jacobian matrix requires calling specific structures too many times. Thus we should pre-write the funtion handle for computing these values and therefore can speed up at least 20%.

¹This MATLAB code has been complied and examined in MATLAB2016b, MATLAB2017a version

Algorithm 1: Combination of Gauss-Newton and Newton's method to solve (1)

Input: Real-valued data A, B, d; initial guess x0; switching threshold itGN.

Output: solution to the least squares problem x

1 $x \leftarrow x0$;

2 for iter = 1, 2, ..., itGN **do**

 $x \leftarrow x + d_{GN};$

4 check termination criteria

5 end

6 for iter = itGN+1,...,maxit do

 $x \leftarrow x + d_{\text{Newton}};$

8 check termination criteria

9 end

10 return x;

III. OBSERVATIONS

- a) Local convergence rate: Although our conclusion tells us that the Newton and Gauss-Newton method on Least squares problem both have locally quadratic convergence rate in general, during the experiment, we find that Newton's method has much faster convergence rate than Gauss-Newton. From the plot we can see that the Gauss-Newton only allows a linear convergence rate, however, with each iteration being much less expansive than the Newton's method. Therefore we choose to implement Gauss-Newton method at the beginning, and when close to the optimal point, we change our method into Newton.
- b) Efficiency interpretation: As we can see, the convergence for the combination of G-N and Newton's method requires 19 iterations; while the pure G-N method requires 89 iterations. The reason is that the G-N method approximates Hessian with $J^{\rm T}J$, which sacrifice accuracy for calculation speed.
- c) Comments on usage of bsxfun: Although bsxfun speeds up the computation for Jacobian and Hessian matrix greatly, if we re-arrange the computation, we can achieve even faster speed than bsxfun by simply calling the dot product and matrix multiplication. The reason it that the bsxfun only provides element-wise operation, but the combination of element-wise operation and matrix product operation will perform better in general.