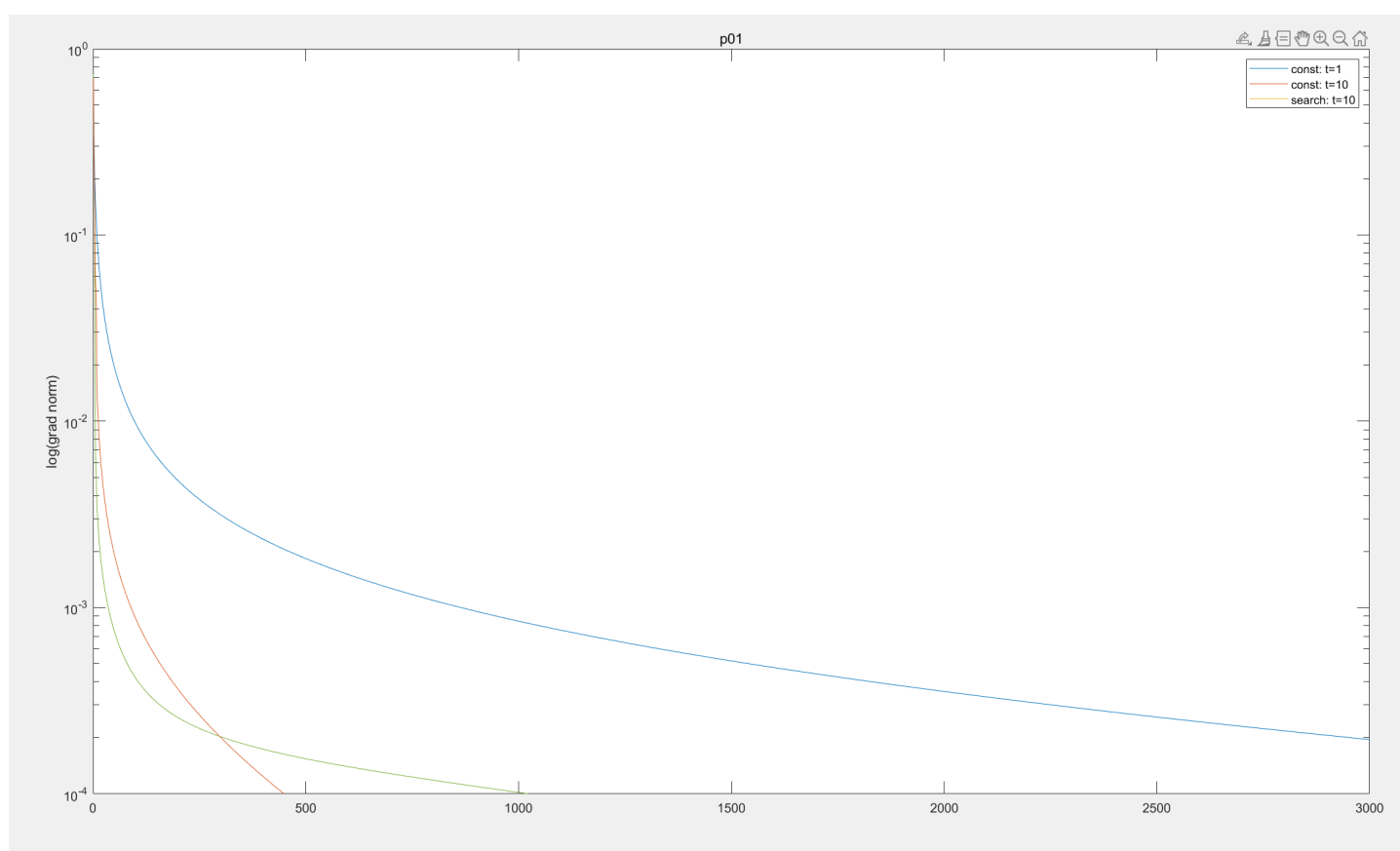


Assignment10

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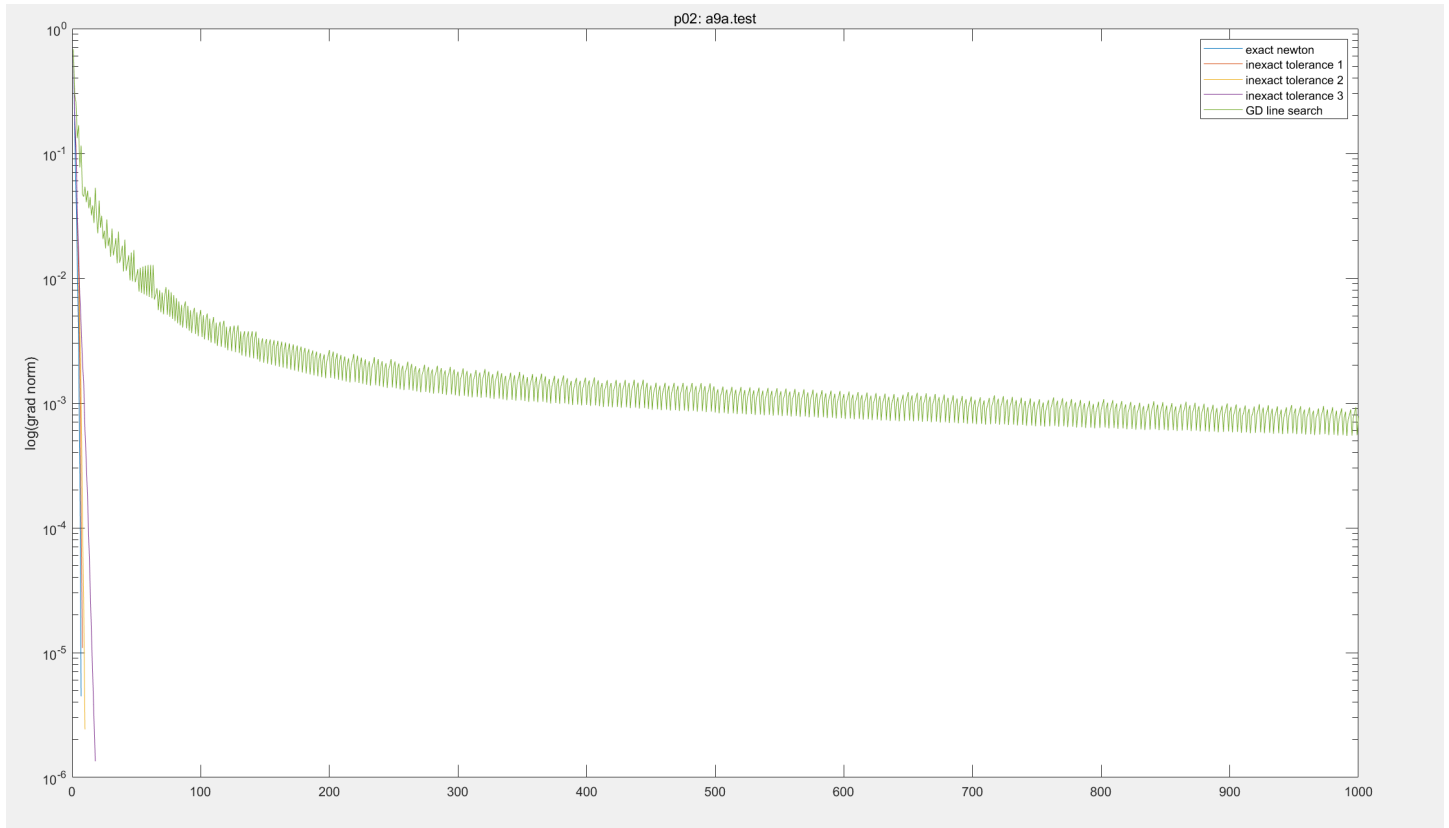
1. 问题1

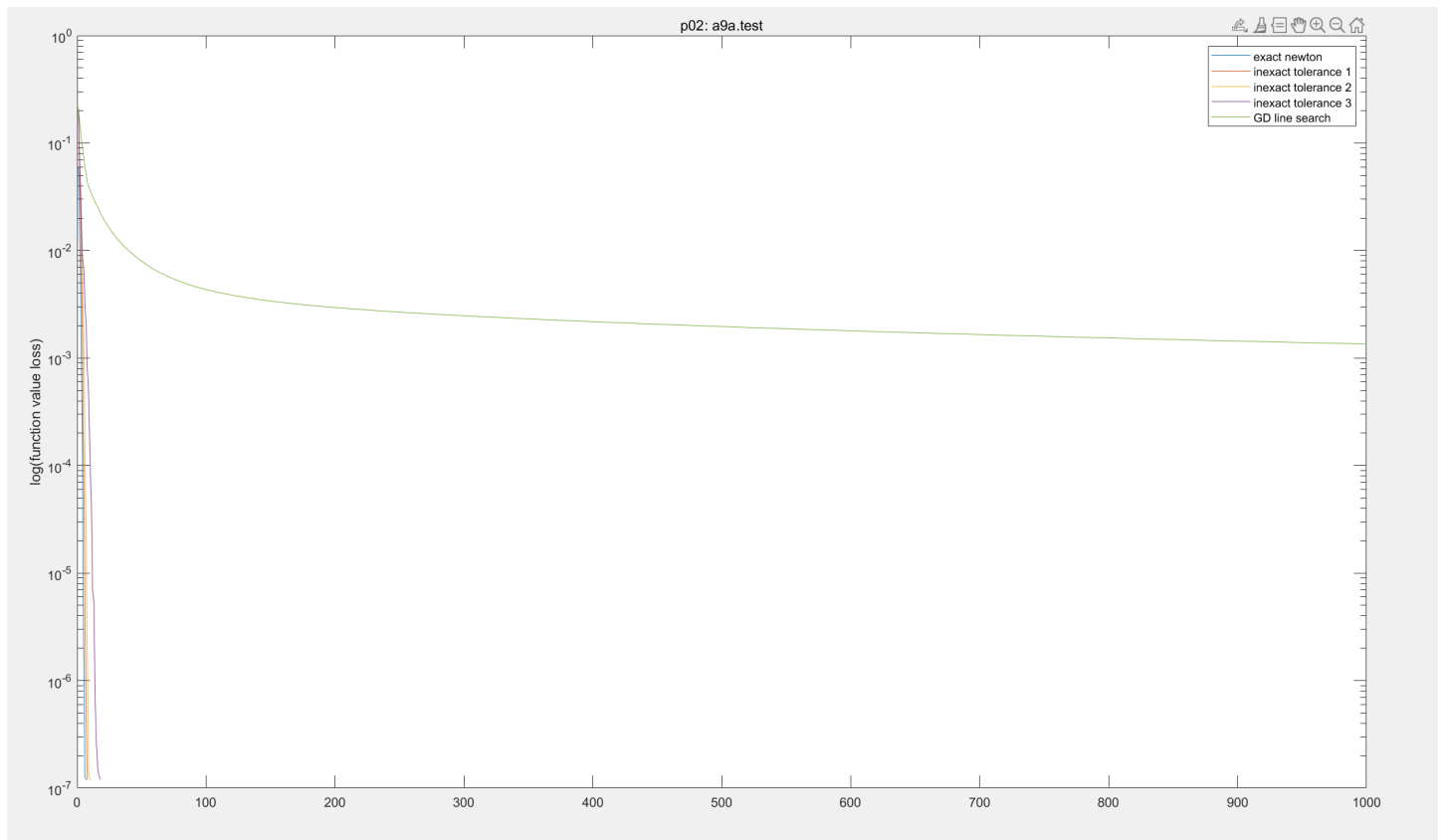
1. 当 $\text{const_step_size} = 1$, 有 $\text{Iteration} = 3000$; $\text{grad_norm} = 1.799591\text{e-}04$; 即算法迭代 3000 步后尚未收敛;
2. 当 $\text{const_step_size} = 10$, 有 $\text{Iteration} = 904$; $\text{grad_norm} = 9.994044\text{e-}05$; 即算法迭代 904 次后收敛;
3. 当 Armijo line search 采用 $\alpha=0.3$, $\beta=0.6$, $t_0=10.0$, 有 $\text{Iteration} = 425$; $\text{grad_norm} = 9.964404\text{e-}05$; 即算法迭代 425 次后收敛;
4. 综上所述, 对于随机生成的数据, 当 const_step_size 变大, 收敛速率变快; 保持初始 step_size 相同的条件下, 采用 line_search , 能进一步加快收敛速率.
5. 对 grad_norm 作 semilogy 图(共三条曲线)如下:



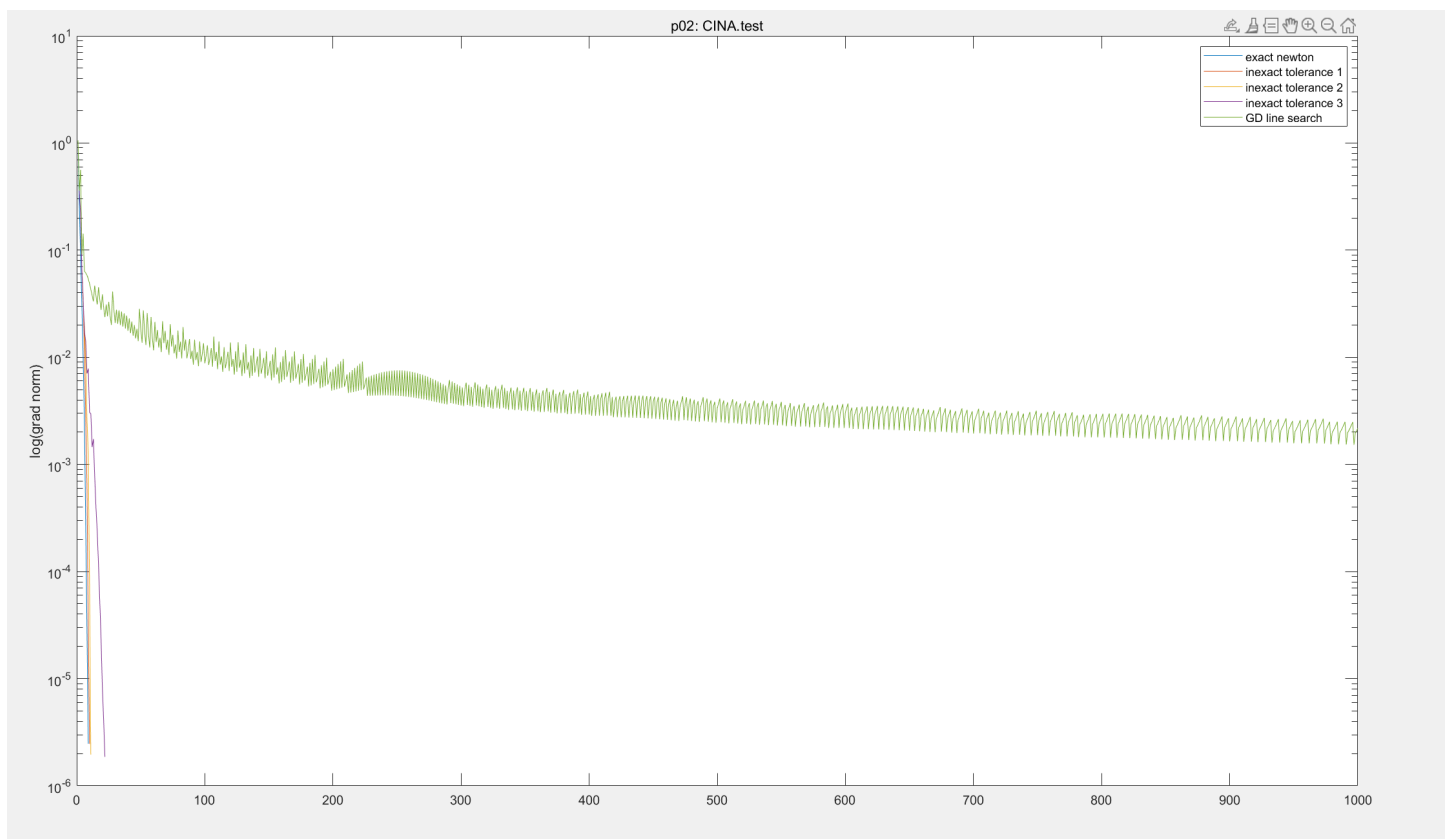
2. 问题2

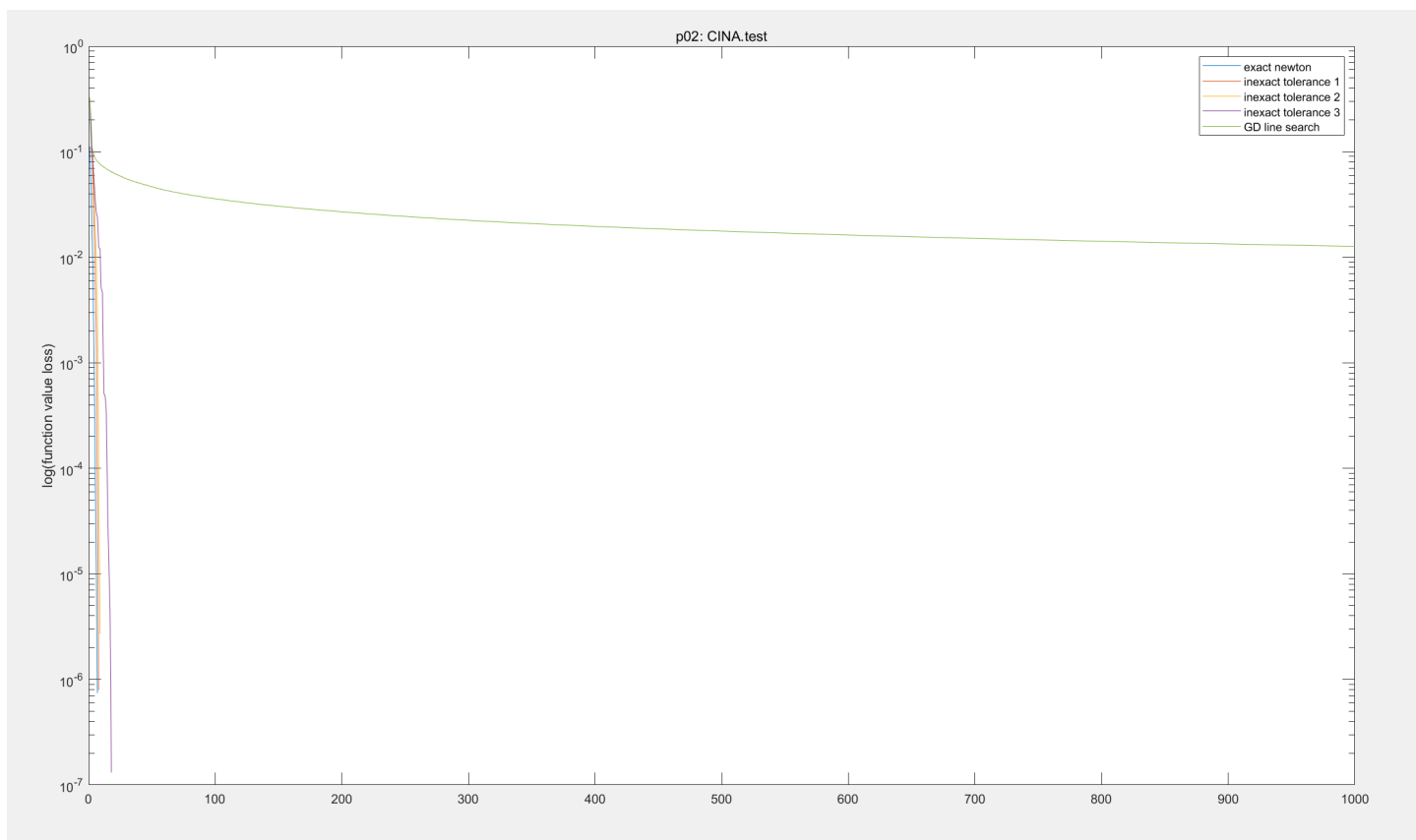
1. 对于 `newton` 和 `inexact_newton` , 不管哪个数据集, 参数都设置为: **$\alpha=0.3$; $\beta=0.6$; $t_0=1.0$** ;
2. 对于 `GD_line_search` , 对于三个数据集, **t_0 均设置为 5**;
3. 在上述参数下, 解得: `a9a.test` 的最优值为 **0.318797** , `CINA.test` 的最优值为 **0.159307** , `ijcnn1.test` 的最优值为 **0.195658** ;
4. `a9a.test` 的 `grad_norm` 和 `function_loss` 作 `semilogy` 图如下(共 5 条曲线):



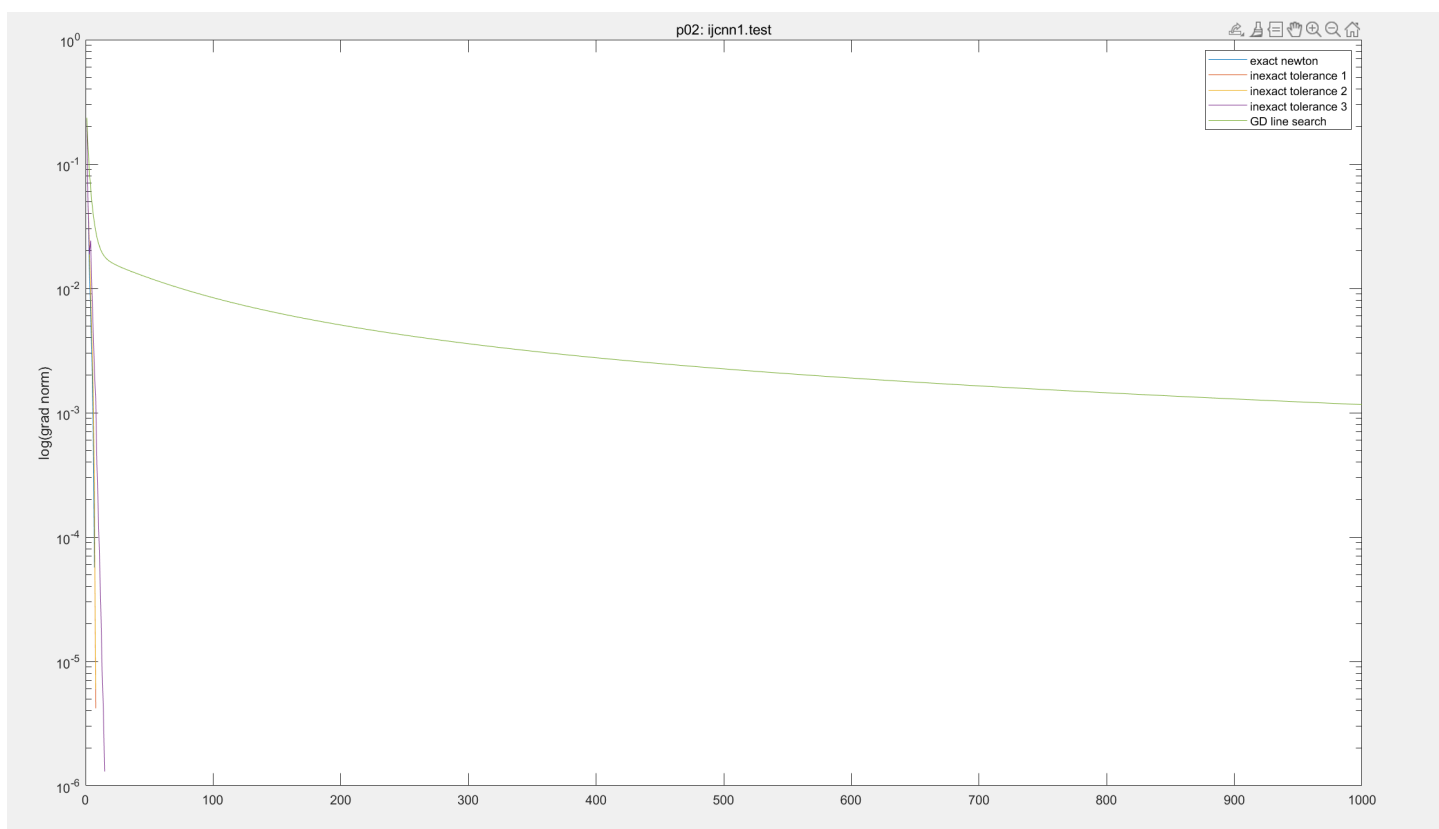


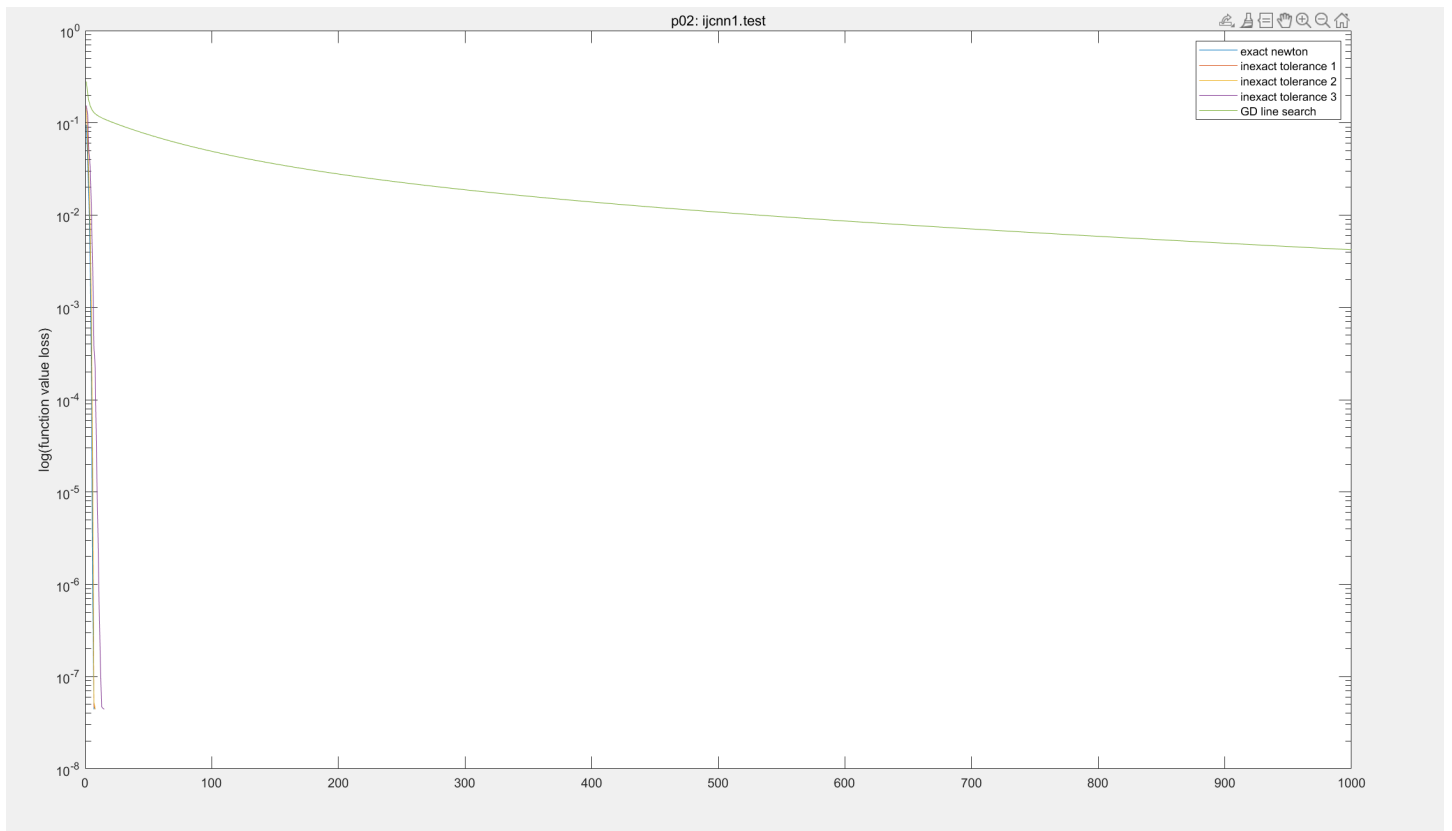
5. `CINA.test` 的 `grad_norm` 和 `function_loss` 作 `semilogy` 图如下(共 5 条曲线):



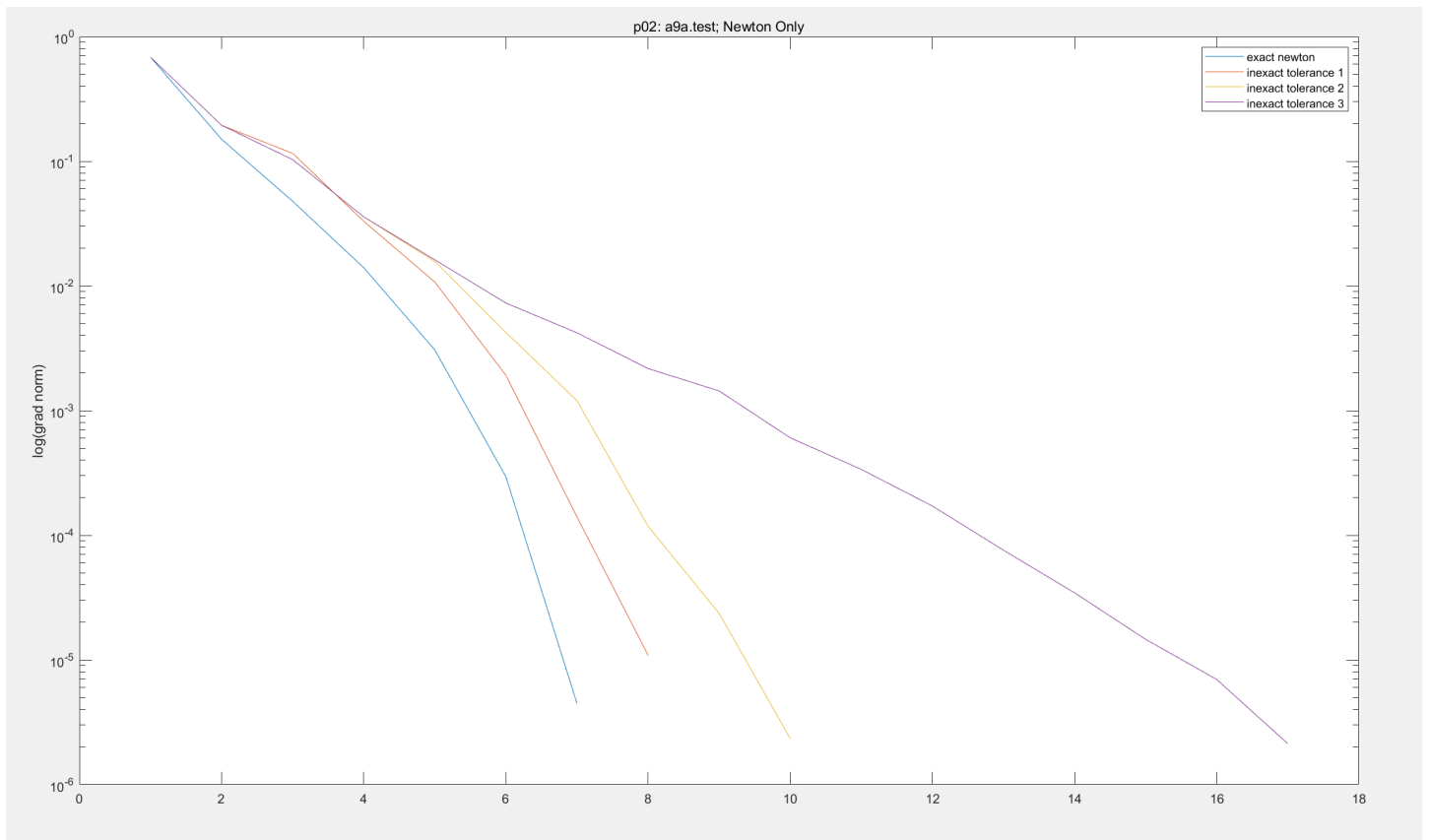


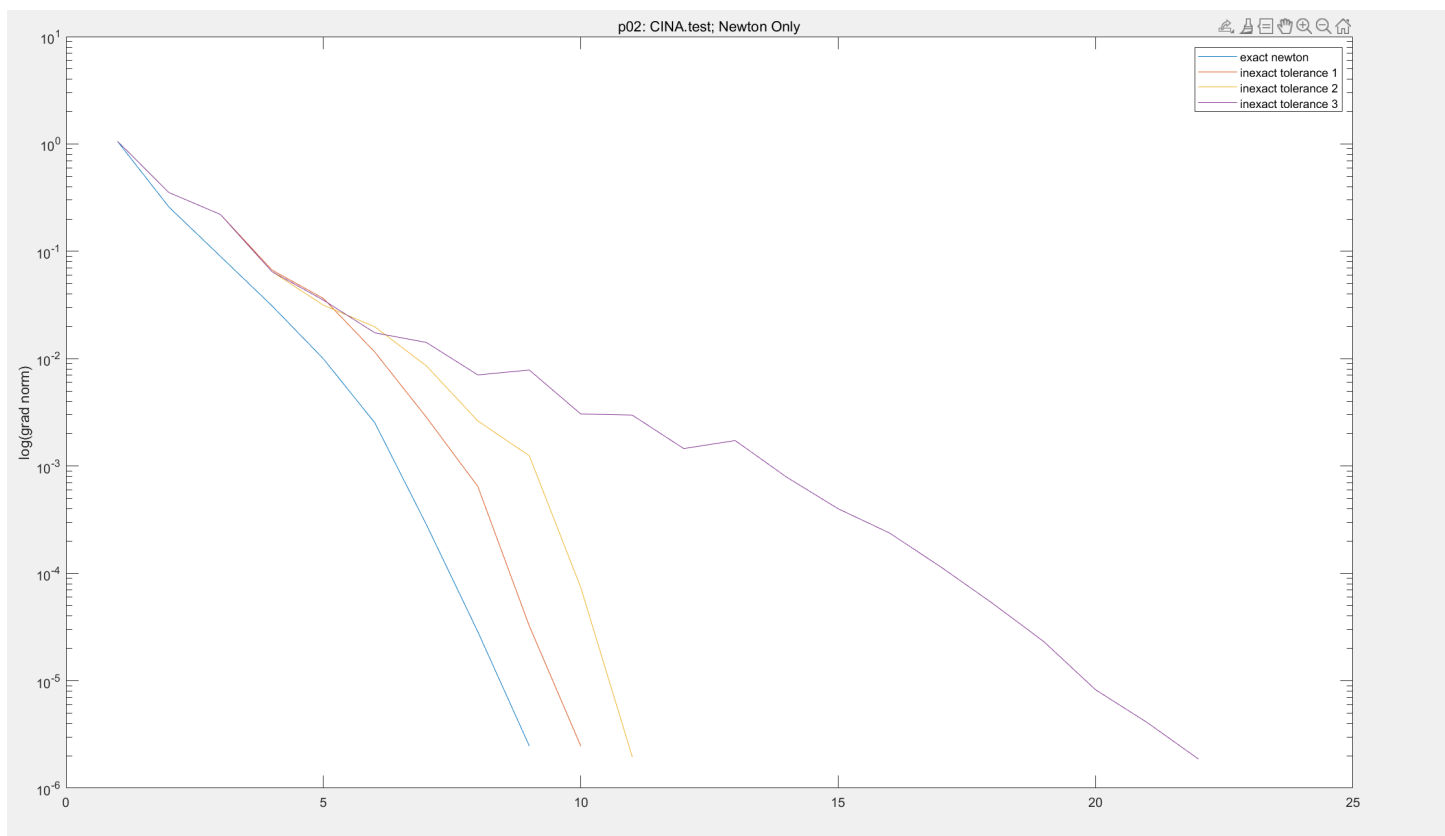
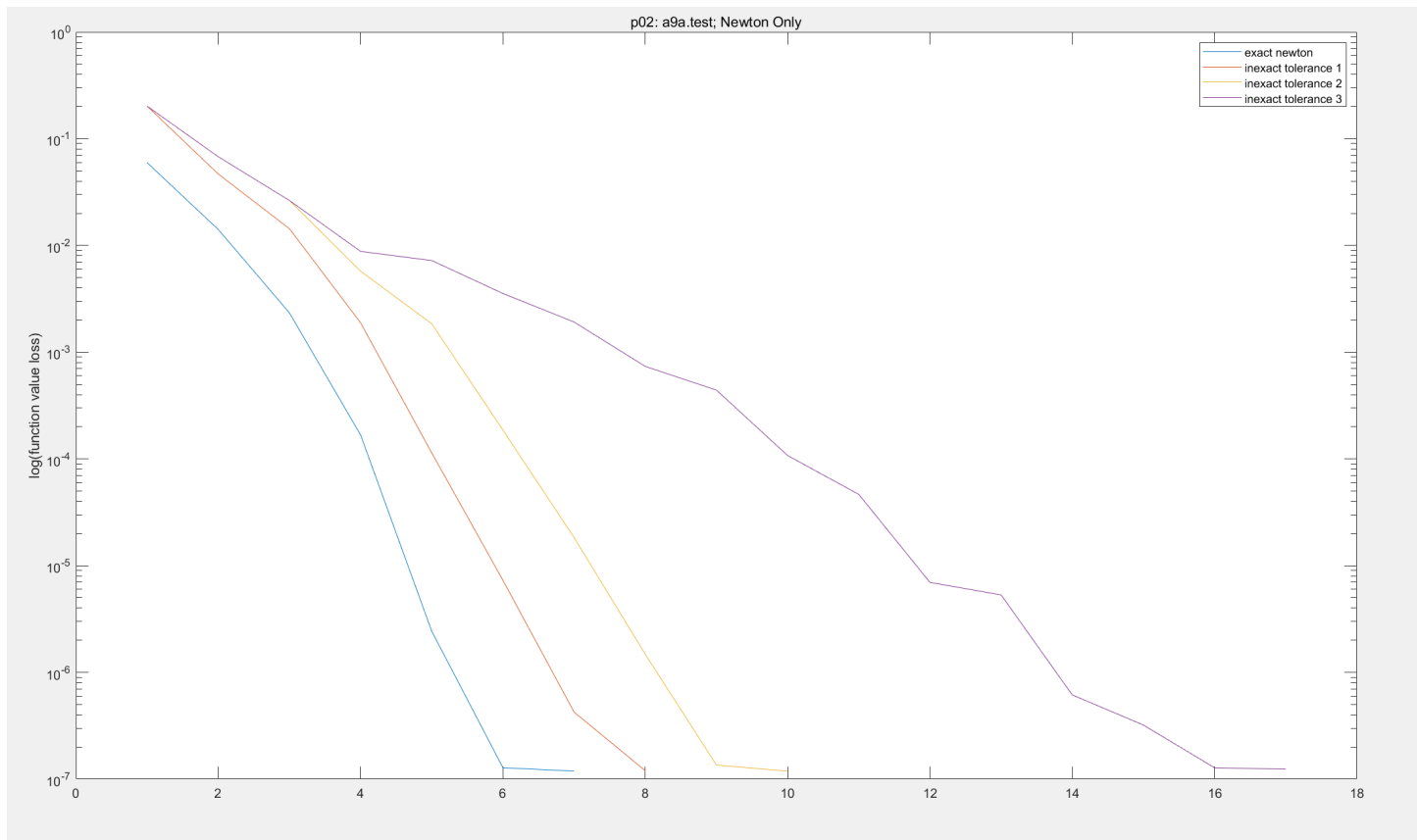
6. `ijcnn1.test` 的 `grad_norm` 和 `function_loss` 作 `semilogy` 图如下(共 5 条曲线):

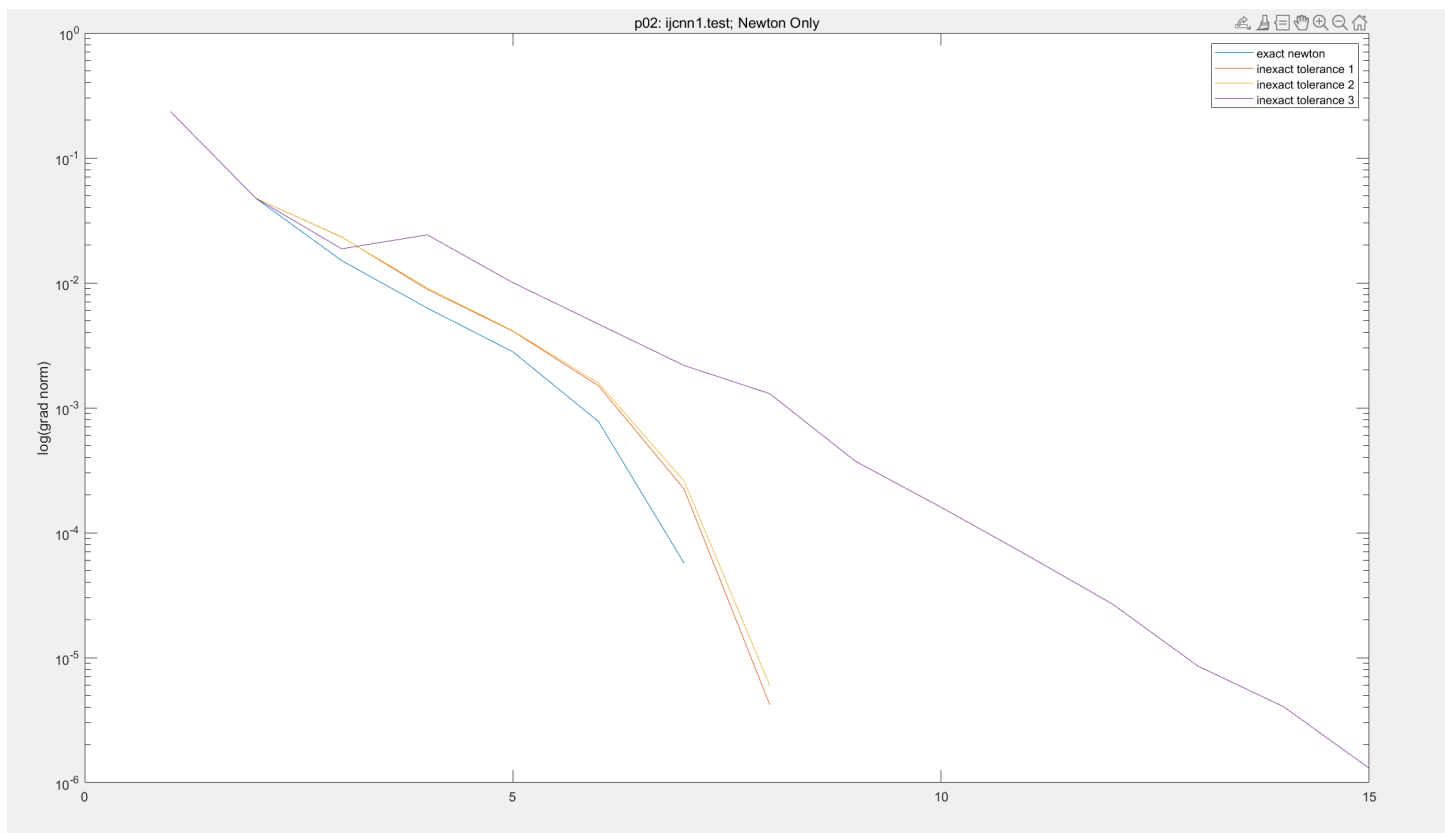
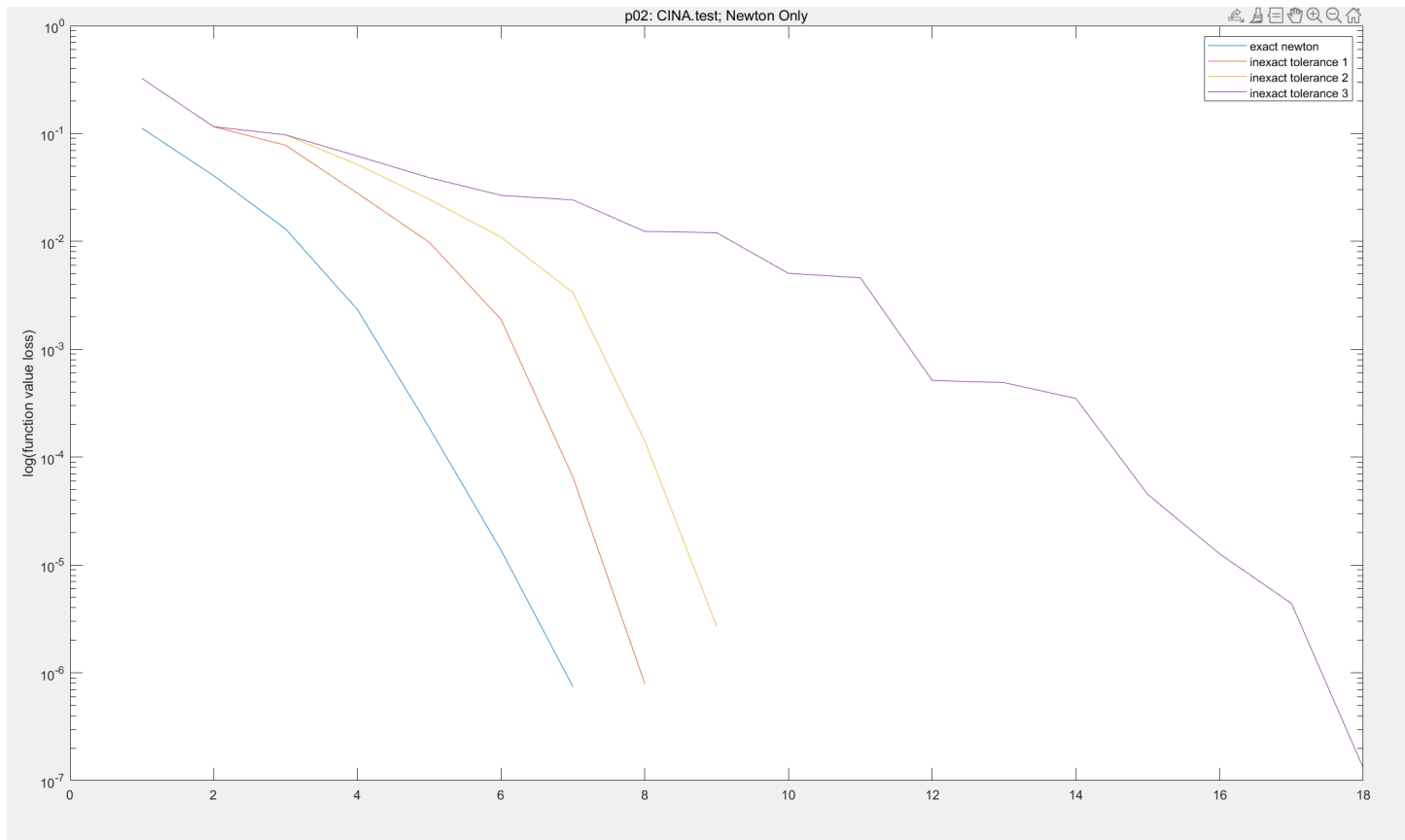


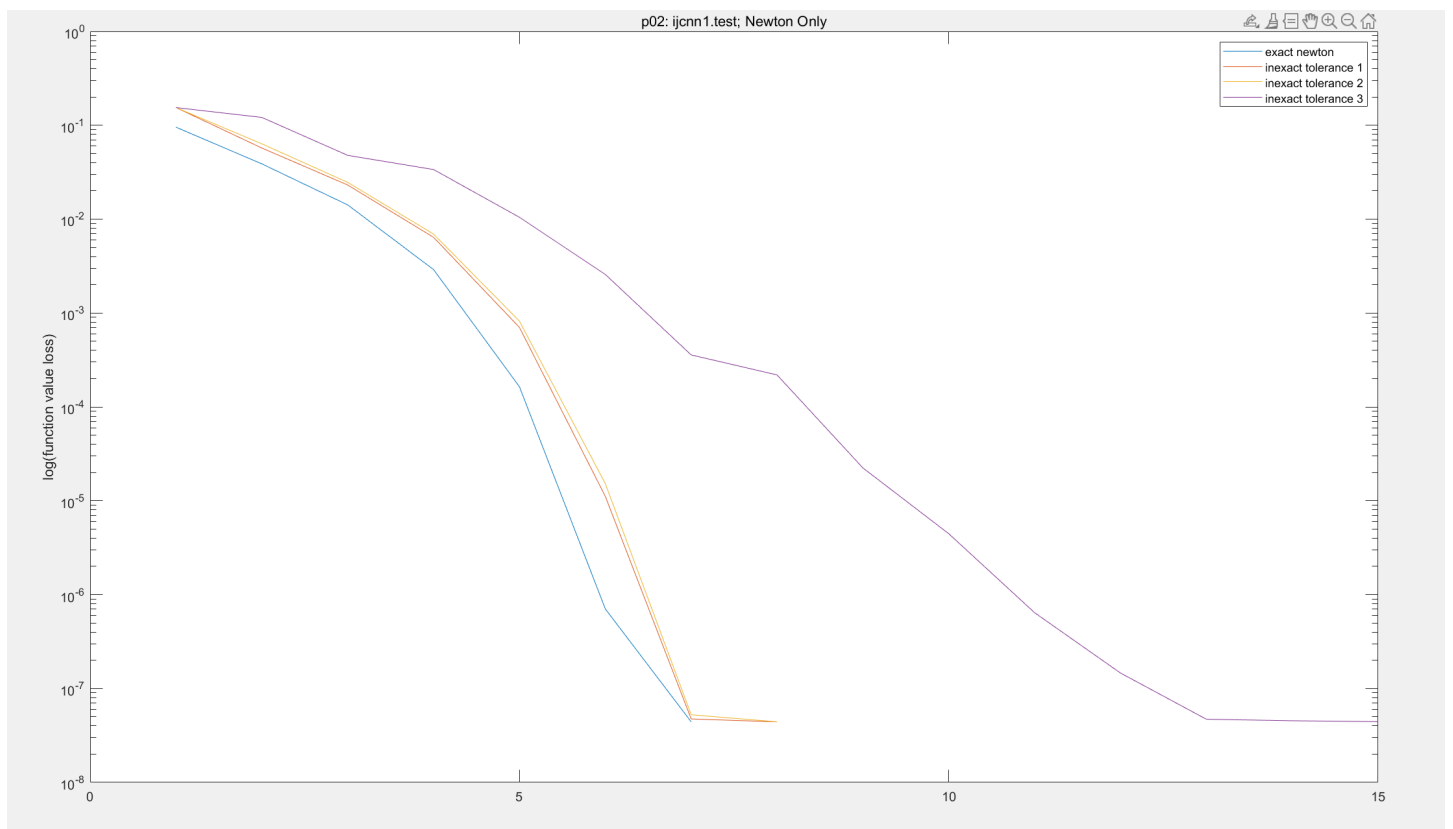


5. 比较下来可以发现, GD_line_search 的收敛速率显著慢于 Newton/inexact_newton, 且 a9a 和 CINA 两个数据集上, `grad norm` 在 GD_line_search 的过程中有持续的小幅震荡.
6. 进一步比较可以发现, **exact_newton** 的收敛速率最快, **inexact_newton** 随着 **tolerance** 的逐渐**放松**(由 tolerance1 -> tolerance 3), **收敛速率逐渐变慢**, 但变慢程度不大, 额外需要的更多步数在5步以内. 具体图像如以下6张图(每个数据集各两张).









3. 源代码

1. 工具函数合集:

- `gradient.m`; `Hessian.m`; `l_value.m`;
- `backtracking.m`; `backtracking_with_input_d.m`;
- `N_direction.m`;

```

1 % gradient.m
2 function [y] = gradient(m, n, A, b, x)
3 p = 1 ./ (1 + exp(-b.*(A*x)));
4 e = ones(m,1);
5 y = -A' * (b.*(e-p))/m + 1/(50*m)*x;
6
7 % Hessian.m
8 function H=Hessian(m,n,A,b,x)
9 e = ones(m,1);
10 p = 1 ./ (1+exp(-b.*(A*x)));
11 I = eye(n,n);
12 D_P = spdiags(p,[0],m,m);
13 % D_P = diag(p,0);
14 D_e_p = spdiags(e-p,[0],m,m);
15 % D_e_p = diag(e-p,0);
16 H= 1/m * A'*(D_P * D_e_p)*A + 1/(50*m)*I;

```



```

17
18 % l_value.m
19 function y = l_value(m,n,A,b,x)
20 p=1./(1+exp(-b.*(A*x)));
21 e = ones(m,1);
22 y = -e'*log(p)/m + 1/(100*m)*(norm(x,2)^2);
23
24 % backtracking.m
25 function t = backtracking(grad,m,n,A,b,x,t0,alpha,beta)
26 t = t0;
27 while l_value(m,n,A,b,x) - l_value(m,n,A,b,x-t*grad) ...
28     < alpha*t*(norm(grad,2)^2)
29     t = beta * t;
30 end
31
32 % backtracking_with_input_d.m
33 function t = backtracking_with_input_d(...
34     grad,d,m,n,A,b,x,t0,alpha,beta)
35 t = t0;
36 while l_value(m,n,A,b,x+t*d) - l_value(m,n,A,b,x) ...
37     > alpha*t*grad'*d
38     t = beta * t;
39 end
40
41 % N_direction.m
42 function d=Ndirection(m,n,A,b,g,x)
43 e = ones(m,1);
44 p = 1 ./ (1+exp(-b.*(A*x)));
45 I = eye(n,n);
46 D1 = spdiags(p,[0],m,m);
47 D2 = spdiags(e-p,[0],m,m);
48 H = 1/m * A'*(D1*D2)*A + 1/(50*m)*I;
49 d = -inv(H)*g;
50

```

2. 问题1 的源代码

```

1 % GD_const.m
2 function grad_norm = GD_const(t)
3 fprintf("GD_const: t = %.2f\n",t)
4 rand('seed', 21307140003);
5 m = 500; n = 1000;
6 A = randn(m,n); b = sign(rand(m,1)-0.5);
7
8 x = zeros(n,1);

```

```

9  grad_norm = zeros(3000);
10
11  for i=1:3000
12      grad = gradient(m,n,A,b,x);
13      x = x-t*grad;
14      grad_norm(i) = norm(grad,2);
15      fprintf("Iteration = %d; grad_norm = %.6e;\n", ...
16              i,grad_norm(i));
17      if grad_norm(i) < 1e-4
18          break
19      end
20  end
21
22  fprintf("GD_const_t=%.2f; Num_Iter = %d\n", t, i);
23
24
25  % GD_linesearch.m
26  function grad_norm = GD_linesearch(alpha, beta, t0)
27  fprintf("GD_line_search: alpha=%.1f, beta=%.1f, t0=%.1f\n",...
28          alpha, beta, t0)
29  rand('seed', 21307140003);
30  m=500; n=1000;
31  A=randn(m,n); b=sign(rand(m,1)-0.5);
32
33  x = zeros(n,1);
34  grad_norm = zeros(3000);
35
36  for i = 1:3000
37      grad = gradient(m,n,A,b,x);
38      ng = norm(grad,2);
39      t = backtracking(grad,m,n,A,b,x,t0,alpha,beta);
40      x = x - t*grad;
41      grad_norm(i) = ng;
42      fprintf("Iteration = %d; grad_norm = %.6e;\n",...
43              i, grad_norm(i));
44      if grad_norm(i) < 1e-4
45          break
46      end
47  end
48
49  fprintf("GD_line_search_t=%.2f; Num_Iter = %d\n", t, i);
50
51
52  % main_1.m
53  grad_norm_1 = GD_const(1);
54  semilogy(grad_norm_1);
55  hold on;

```

```

56
57 grad_norm_2 = GD_const(10);
58 semilogy(grad_norm_2);
59 hold on;
60
61 grad_norm_3 = GD_linesearch(0.3, 0.6, 10);
62 semilogy(grad_norm_3);
63 hold off;
64
65 title('p01')
66 ylabel('log(grad norm)')
67 legend('const: t=1', 'const: t=10', 'search: t=10');

```

3. 问题2 的源代码

```

1 % newton.m
2 function [grad_norm, l_value_loss] = newton(dataset, alpha, beta, t0)
3 fprintf("exact_newton: alpha=%.1f, beta=%.1f, t0=%.1f\n",...
4         alpha, beta, t0)
5 [b,A] = libsvmread(dataset);
6 [m,n] = size(A);
7 mu = 1e-2/m;
8
9 x = zeros(n,1);
10 grad_norm = [];
11 l_value_loss = [];
12
13 grad=gradient(m,n,A,b,x);
14 ng = norm(grad,2);
15 while ng >= 1e-6
16     d = Ndirection(m,n,A,b,grad,x);
17     t = backtracking_with_input_d(grad,d,m,n,A,b,x,t0,alpha,beta);
18     x = x + t*d;
19     grad_norm = [grad_norm,ng];
20     l_value_loss = [l_value_loss, l_value(m,n,A,b,x)];
21     grad = gradient(m,n,A,b,x);
22     ng = norm(grad, 2);
23 end
24 l_value_loss = l_value_loss - 0.195658;
25 fprintf("l(x*) = %.6f\n", l_value(m,n,A,b,x))
26
27
28 % inexact_newton.m
29 function [grad_norm, l_value_loss] = inexact_newton(...
30         dataset, alpha, beta, t0, tolerance)

```

```

31 fprintf("inexact_newton: alpha=%.1f, beta=%.1f, t0=%.1f\n",...
32         alpha, beta, t0)
33 [b,A] = libsvmread(dataset);
34 [m,n] = size(A);
35
36 x = zeros(n,1);
37 grad_norm = [];
38 l_value_loss = [];
39
40 grad = gradient(m,n,A,b,x);
41 ng = norm(grad, 2);
42 while ng >= 1e-6
43     H = Hessian(m,n,A,b,x);
44     d = myCG(H, grad, ng, 1000, tolerance);
45     t = backtracking_with_input_d(grad,d,m,n,A,b,x,t0,alpha,beta);
46     x = x + t*d;
47     grad_norm=[grad_norm,ng];
48     l_value_loss = [l_value_loss, l_value(m,n,A,b,x)];
49     grad = gradient(m,n,A,b,x);
50     ng = norm(grad, 2);
51 end
52 l_value_loss = l_value_loss - 0.195658;
53
54
55 % GD_linesearch_2.m
56 function [grad_norm, l_value_loss] = GD_linesearch_2(...
57         dataset, alpha, beta, t0)
58 fprintf("GD_line_search: alpha=%.1f, beta=%.1f, t0=%.1f\n",...
59         alpha, beta, t0)
60
61 [b,A] = libsvmread(dataset);
62 [m,n] = size(A);
63 mu = 1e-2/m;
64
65 x=zeros(n,1);
66 grad_norm = zeros(1000);
67 l_value_loss = zeros(1000);
68
69 for i = 1:1000
70     grad = gradient(m,n,A,b,x);
71     ng = norm(grad,2);
72     t = backtracking(grad,m,n,A,b,x,t0,alpha,beta);
73     x = x - t*grad;
74     grad_norm(i) = ng;
75     l_value_loss(i) = l_value(m,n,A,b,x);
76     fprintf("Iteration = %d; grad_norm = %.6e;\n",...
77             i, grad_norm(i));

```

```

78     if grad_norm(i) < 1e-4
79         break
80     end
81 end
82 l_value_loss = max(l_value_loss - 0.195658, zeros(1000));
83 fprintf("GD_line_search_t=%.2f; Num_Iter = %d\n", t, i);
84
85
86 % main_2.m
87 dataset = 'ijcnn1.test';
88 % a9a:  $l(x^*) = 0.318797$ ; GD_linesearch_t0 = 5
89 % CINA:  $l(x^*) = 0.159307$ ; GD_linesearch_t0 = 5
90 % ijcnn1:  $l(x^*) = 0.195658$ ; GD_linesearch_t0 = 50
91
92 [grad_norm_exact, l_value_loss_exact] = newton(...
93     dataset, 0.3, 0.6, 1);
94
95 tolerance_1 = @(ng) min([0.5, ng]);
96 tolerance_2 = @(ng) min([0.5, sqrt(ng)]);
97 tolerance_3 = @(ng) 0.5;
98 [grad_norm_1, l_value_loss_1] = inexact_newton(...
99     dataset, 0.3, 0.6, 1, tolerance_1);
100 [grad_norm_2, l_value_loss_2] = inexact_newton(...
101     dataset, 0.3, 0.6, 1, tolerance_2);
102 [grad_norm_3, l_value_loss_3] = inexact_newton(...
103     dataset, 0.3, 0.6, 1, tolerance_3);
104
105 [grad_norm_gd, l_value_loss_gd] = GD_linesearch_2(...
106     dataset, 0.3, 0.6, 50);
107
108 semilogy(grad_norm_exact)
109 hold on
110 semilogy(grad_norm_1)
111 hold on
112 semilogy(grad_norm_2)
113 hold on
114 semilogy(grad_norm_3)
115 hold on
116 semilogy(grad_norm_gd)
117 hold off
118 title(sprintf('p02: %s', dataset))
119 ylabel('log(grad norm)')
120 legend('exact newton', 'inexact tolerance 1',...
121 'inexact tolerance 2', 'inexact tolerance 3', 'GD_line_search')
122
123 figure
124

```

```
125 semilogy(l_value_loss_exact)
126 hold on
127 semilogy(l_value_loss_1)
128 hold on
129 semilogy(l_value_loss_2)
130 hold on
131 semilogy(l_value_loss_3)
132 hold on
133 semilogy(l_value_loss_gd)
134 hold off
135 title(sprintf('p02: %s', dataset))
136 ylabel('log(function value loss)')
137 legend('exact newton', 'inexact tolerance 1',...
138 'inexact tolerance 2', 'inexact tolerance 3', 'GD_line_search')
```