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Empirical Methods HA 3

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Problem 1

$$\beta_{mle} = \begin{pmatrix} 2.5339 \\ -0.0323 \\ 0.1157 \\ -0.3540 \\ 0.0798 \\ -0.4094 \end{pmatrix};$$

Problem 2

$$\beta_{mleqn} = \begin{pmatrix} 2.5339 \\ -0.0323 \\ 0.1157 \\ -0.3540 \\ 0.0798 \\ -0.4094 \end{pmatrix};$$

Problem 3

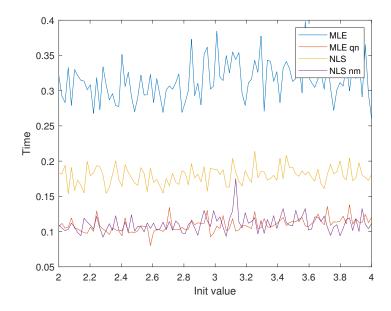
$$\beta_{nls} = \begin{pmatrix} 2.5126 \\ -0.0384 \\ 0.1141 \\ -0.2796 \\ 0.0676 \\ -0.3698 \end{pmatrix}$$

Problem 4

$$\beta_{nlsnm} = \begin{pmatrix} 2.5126 \\ -0.0384 \\ 0.1141 \\ -0.2796 \\ 0.0676 \\ -0.3698 \end{pmatrix};$$

Problem 5

By looking on the graph we may conclude that MLE via the Nelder-Mead is the slowest method, then comes the NLS (lsqonlin) method. And MLE via a quasi-Newton and NLS via Nelder-Mead are the fastest ones.



```
function [L, J] = likelihood1(y,x,b)
s = size(x);
k1 = y-exp(x*b);
k2 = kron(k1,ones(1,s(2)));
z1 = k2.*x;
J = sum(z1,1)'/s(1);
L1 = -exp(x*b)+y.*x*b;
L = sum(L1)/s(1);
end
```

```
function [L, J] = likelihood2(y,X,b)
s = size(X);
k1 = y-exp(X*b);
k2 = kron(k1,ones(1,s(2)));
z1 = k2.*X;
L1 = -exp(X*b)+y.*X*b;
L = norm(sum(L1));
J = sum(z1,1)'/(2*L);
end
```

```
_{1} function [a, b, c] = mle_{est}(init, y, X)
_{2} b<sub>-</sub>mle = init;
 l = @(b) - likelihood1(y, X, b);
4 opt1 = optimset('Display', 'Final', 'TolF', 1e-8, 'TolX', 1e-8);
exit = 2;
  maxit = 0;
  tic
  while (exit~=1)&&(maxit<100)
       [b_mle, fval, exit] = fminsearch(l, b_mle, opt1);
       maxit = maxit+1;
10
11 end
a = toc;
b = exit;
c = 1;
15 end
```

```
function [a, b, c] = mle_qn_est(init,y,X)
b_mle_init = init;

opt2 = optimset('Display','Final');

l_z = @(b) likelihood2(y,X,b);

tic

[b_mle_qn, fval, exit] = fminunc(l_z, b_mle_init, opt2);

a = toc;

b = exit;

c = 2;

end
```

```
_{1} function [a, b, c] = nls_{n}m_{est}(init, y, X)
 b_nls_m = init;
  n ls_f = @(b)sum((y-exp(X*b)).^2);
4 opt4 = optimset('Display', 'Final', 'TolF', 1e-16, 'TolX', 1e-16);
  exit2 = 2;
  \max it2 = 0;
  tic
  while (exit2~=1)&&(maxit2<100)
       [b_nls_nm, fval, exit2] = fminsearch(nls_f, b_nls_nm, opt4);
       \max it2 = \max it2+1;
10
  \operatorname{end}
a = toc;
b = exit2;
c = 4;
15 end
```

```
M=load ('hw3', '-mat');
X = M.X;
  y = M.y;
  clear M;
6
  % Problem 1
  b_{-}mle = [3;0;0;0;0;0];
  l = @(b) - likelihood1(y, X, b);
  options = optimset('Display', 'Final', 'TolF', 1e-8, 'TolX', 1e-8);
  exit1 = 2;
  \max it1 = 0;
12
  while (exit1~=1)\&\&(maxit1<100)
13
       [b_mle, fval, exit1]=fminsearch(l, b_mle, options);
       \max it1 = \max it1+1;
15
  end
16
  fprintf('Iter: %d, Likelihood f value: %f \n', maxit1, likelihood1(y, X
      , b_{-}mle));
  b_mle
18
19
  % Problem 2
  b_{mle_{init}} = [3;0;0;0;0;0];
  options2 = optimset('Display', 'final');
  1_z = 0(b) \text{ likelihood } 2(y, X, b);
  b_mle_qn = fminunc(l_z, b_mle_init, options2);
  fprintf('Norm of Likelihood f value: %f \n', likelihood2(y,X,
     b_mle_qn);
  b_mle_qn
26
27
  % Problem 3
28
  b_n ls_i nit = [3;0;0;0;0;0];
29
  options3 = optimoptions ('lsqnonlin', 'Display', 'Final', 'TolF', 1e-16, '
     TolX', 1e-16);
  f=0(b)y-\exp(X*b);
```

```
b_n = lsqnonlin(f, b_n = lsqno
        fprintf('Obj Function: %f \n',sum(f(b_nls).^2));
33
        b_nls
34
       % Problem 4
36
        b_nls_nm = [3;0;0;0;0;0];
37
        n ls_f = 0(b) sum((y-exp(X*b)).^2);
38
        options4 = optimset('Display', 'Final', 'TolF', 1e-16, 'TolX', 1e-16);
39
        exit2 = 2;
40
        \max_{i=0}
41
        while (exit 2 = 1) \&\& (maxit 2 < 100)
                     [b_nls_nm, fval, exit2]=fminsearch(nls_f, b_nls_nm, options4);
43
                     \max it2 = \max it2+1;
44
        end
45
        fprintf('Iter: %d, Obj Function: %f \n', maxit2, nls_f(b_nls_nm));
        b_nls_nm
47
48
       % Problem 5
50
        range = linspace(2,4);
        tests = zeros(length(range)*4,4);
        for k = 1: length (range)
                     b_{init} = [range(k); 0; 0; 0; 0; 0; 0];
54
                     [x1,x2,x3] = mle_est(b_init,y,X);
55
                     [x4,x5,x6] = mle_qn_est(b_init,y,X);
56
                     [x7,x8,x9] = nls_est(b_init,y,X);
57
                     [x10, x11, x12] = nls_nm_est(b_init, y, X);
58
                     tests((k-1)*4+1,1) = range(k);
59
                     tests((k-1)*4+1,2) = x1;
60
                     tests((k-1)*4+1,3) = x2;
61
                     tests((k-1)*4+1,4) = x3;
62
                     tests((k-1)*4+2,1) = range(k);
63
                     tests((k-1)*4+2,2) = x4;
                     tests((k-1)*4+2,3) = x5;
65
                     tests((k-1)*4+2,4) = x6;
66
```

```
tests((k-1)*4+3,1) = range(k);
67
       tests((k-1)*4+3,2) = x7;
68
       tests((k-1)*4+3,3) = x8;
69
       tests((k-1)*4+3,4) = x9;
70
       tests((k-1)*4+4,1) = range(k);
71
       tests((k-1)*4+4,2) = x10;
72
       tests((k-1)*4+4,3) = x11;
73
       tests((k-1)*4+4,4) = x12;
74
  end
75
  z1 = find(tests(:,4) == 1);
76
  z2 = find(tests(:,4) == 2);
  z3 = find(tests(:,4) == 3);
  z4 = find(tests(:,4) == 4);
79
  mle_tests = tests(z1,:);
80
  mleqn_tests = tests(z2,:);
  nls\_tests = tests(z3,:);
82
  nlsnm\_tests = tests(z4,:);
83
  plot(range, mle_tests(:,2), range, mleqn_tests(:,2), range, nls_tests
      (:,2), range, nlsnm_tests(:,2));
  xlabel('Init value');
  ylabel('Time');
  legend('MLE', 'MLE qn', 'NLS', 'NLS nm');
```