

# Q1

## Output:

-----  
 Oscar Martinez    Take-Home 2: Question 1    Metrics III  
 -----

Weibull Model - The dependent variable is: days

Grad:	0.6363	LogL:	-2192.8495	Size:	1.000
Grad:	0.7267	LogL:	-1986.4985	Size:	1.000
Grad:	0.8283	LogL:	-1798.1316	Size:	1.000
Grad:	0.9357	LogL:	-1631.6029	Size:	1.000
Grad:	1.0336	LogL:	-1490.8103	Size:	1.000
Grad:	1.0907	LogL:	-1378.6446	Size:	1.000
Grad:	1.0617	LogL:	-1295.2310	Size:	1.000
Grad:	0.9140	LogL:	-1236.4924	Size:	1.000
Grad:	0.6690	LogL:	-1195.0893	Size:	1.000
Grad:	0.3942	LogL:	-1163.6355	Size:	1.000
Grad:	0.1454	LogL:	-1137.0129	Size:	1.000
Grad:	0.0591	LogL:	-1112.4269	Size:	1.000
Grad:	0.2248	LogL:	-1088.5531	Size:	1.000
Grad:	0.3648	LogL:	-1064.8436	Size:	1.000
Grad:	0.4930	LogL:	-1041.1644	Size:	1.000
Grad:	0.6224	LogL:	-1017.6633	Size:	1.000
Grad:	0.7635	LogL:	-994.7860	Size:	1.000
Grad:	0.9177	LogL:	-973.3826	Size:	1.000
Grad:	1.0525	LogL:	-954.7238	Size:	1.000
Grad:	1.0613	LogL:	-939.8889	Size:	1.000
Grad:	0.8367	LogL:	-928.8476	Size:	1.000
Grad:	0.4715	LogL:	-921.7429	Size:	1.000
Grad:	0.1384	LogL:	-918.9711	Size:	1.000
Grad:	0.0832	LogL:	-918.5399	Size:	1.000
Grad:	0.0493	LogL:	-918.5114	Size:	1.000
Grad:	0.0117	LogL:	-918.5101	Size:	1.000
Grad:	0.0015	LogL:	-918.5100	Size:	1.000
Grad:	0.0008	LogL:	-918.5100	Size:	1.000

Regressor	Coefficient	Std. Error	t-stat	Prob> t
con	-4.60591	0.80675	-5.70920	0.00000
vac	-0.05647	0.17883	-0.31577	0.75241
lap	-0.32446	0.17236	-1.88240	0.06079
beta	1.06906	0.07933	13.47682	0.00000

## Code:

```
1 % Computer Take-Home 2: Question 1
2 % Metrics III
3 %Oscar Martinez
4
5 %Diary
6 diary Q2_Output_Oscar_Martinez.txt
7
8 %Introduction
9 fprintf('_____\\n')
10 ;
11 fprintf('Oscar Martinez \\t Take-Home 2: Question 1 \\t Metrics III\\n');
12 fprintf('_____\\n')
13 ;
14 %Exam 2 Question 1:
15 data='mIsUp3'; bin='sold'; dep='days'; ind='vac lap';
16 theta=[-3; 0; 0; 1];
17 theta=weibull(data,dep,bin,ind,theta);
18
19 %closing output
20 diary off
```

## Q2

### Output:

```
-----
Oscar Martinez   Take-Home 2: Question 2   Metrics III
-----
```

lambda =

3.5700

Heckman-Lee Estimates of Tobit Model

The data set is: martins2

Probit Model - The dependent variable is: employ

The data set is: martins2

Grad:	0.4564	LogL:	-1386.3792	Size:	1.000
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Grad:	0.1241	LogL:	-1378.6496	Size:	1.000
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```

Grad:    0.0071    LogL: -1378.1894    Size:    1.000
Grad:    0.0015    LogL: -1378.1890    Size:    1.000
Grad:    0.0004    LogL: -1378.1890    Size:    1.000
Grad:    0.0001    LogL: -1378.1890    Size:    1.000

```

Regressor	Coefficient	Std. Error	t-stat	Prob> t
Con	-0.58845	0.17882	-3.29068	0.00101
edu	0.12722	0.00981	12.97394	0.00000
pexp	0.03854	0.01384	2.78353	0.00542
pexp2	-0.00108	0.00027	-4.00210	0.00006
pexpchd	-0.01351	0.00365	-3.69861	0.00022
pexpch2	0.00031	0.00010	2.95627	0.00315

OLS Coefficients - The dependent variable is: lwage2

Regressor	Coefficient	Std. Error	t-stat	Prob> t
Con	1.13123	0.09118	12.40620	0.00000
edu	0.10306	0.00408	25.28947	0.00000
pexp	0.00772	0.00749	1.03174	0.30237
pexp2	0.00012	0.00016	0.73849	0.46034
pexpchd	0.00424	0.00221	1.91628	0.05553
pexpch2	-0.00013	0.00007	-2.01065	0.04455
Ln(Sig)	-0.61670			
LogL	-1120.13392			

Heckman-Lee Coefficients - The dependent variable is: lwage2  
Uncorrected Standard Errors

Regressor	Coefficient	Std. Error	t-stat	Prob> t
Con	-1.78801	0.35442	-5.04484	0.00000
edu	0.25397	0.01818	13.97311	0.00000
pexp	0.08243	0.01142	7.21847	0.00000
pexp2	-0.00188	0.00028	-6.71127	0.00000
pexpchd	-0.01550	0.00317	-4.89216	0.00000
pexpch2	0.00030	0.00008	3.68844	0.00023
Mills	2.65040	0.31149	8.50882	0.00000
Ln(Sig)	0.67175			

Tobit Model - The dependent variable is: lwage2  
The data set is: martins2

Grad:	1.4521	LogL:	-3452.5899	Size:	0.500
Grad:	0.5195	LogL:	-3425.3926	Size:	0.250
Grad:	0.4281	LogL:	-3417.9811	Size:	0.500
Grad:	0.3289	LogL:	-3410.6604	Size:	0.250
Grad:	0.2992	LogL:	-3407.7637	Size:	0.250
Grad:	0.1011	LogL:	-3405.6837	Size:	0.250
Grad:	0.1490	LogL:	-3405.5492	Size:	0.500
Grad:	0.2158	LogL:	-3403.6044	Size:	0.250
Grad:	0.1635	LogL:	-3402.8549	Size:	0.250
Grad:	0.0792	LogL:	-3402.3696	Size:	0.250
Grad:	0.0753	LogL:	-3402.0902	Size:	0.250
Grad:	0.0300	LogL:	-3401.9125	Size:	0.250
Grad:	0.0348	LogL:	-3401.8469	Size:	0.500
Grad:	0.0551	LogL:	-3401.7197	Size:	0.250
Grad:	0.0422	LogL:	-3401.6572	Size:	0.250
Grad:	0.0209	LogL:	-3401.6216	Size:	0.250
Grad:	0.0185	LogL:	-3401.6008	Size:	0.250
Grad:	0.0075	LogL:	-3401.5972	Size:	0.500
Grad:	0.0241	LogL:	-3401.5809	Size:	0.250
Grad:	0.0131	LogL:	-3401.5727	Size:	0.250
Grad:	0.0102	LogL:	-3401.5684	Size:	0.250
Grad:	0.0049	LogL:	-3401.5660	Size:	0.250
Grad:	0.0045	LogL:	-3401.5645	Size:	0.250
Grad:	0.0017	LogL:	-3401.5639	Size:	0.500
Grad:	0.0057	LogL:	-3401.5629	Size:	0.250
Grad:	0.0030	LogL:	-3401.5624	Size:	0.250
Grad:	0.0024	LogL:	-3401.5621	Size:	0.250
Grad:	0.0011	LogL:	-3401.5620	Size:	0.250
Grad:	0.0011	LogL:	-3401.5619	Size:	0.500
Grad:	0.0018	LogL:	-3401.5618	Size:	0.250
Grad:	0.0013	LogL:	-3401.5618	Size:	0.250
Grad:	0.0007	LogL:	-3401.5617	Size:	0.250
Grad:	0.0006	LogL:	-3401.5617	Size:	0.250
Grad:	0.0002	LogL:	-3401.5617	Size:	0.500
Grad:	0.0007	LogL:	-3401.5617	Size:	0.250
Grad:	0.0004	LogL:	-3401.5617	Size:	0.250
Grad:	0.0003	LogL:	-3401.5617	Size:	0.250
Grad:	0.0001	LogL:	-3401.5617	Size:	0.250
Grad:	0.0001	LogL:	-3401.5617	Size:	0.250
Grad:	0.0000	LogL:	-3401.5617	Size:	0.500

Regressor	Coefficient	Std. Error	t-stat	Prob> t
Con	-0.94189	0.24209	-3.89069	0.00010
edu	0.21239	0.01499	14.16795	0.00000

```

pexp      0.07146      0.01841      3.88082      0.00011
pexp2     -0.00168      0.00036     -4.65441      0.00000
pexpchd   -0.01488      0.00478     -3.11657      0.00185
pexpch2    0.00031      0.00013      2.28637      0.02232
lnsig      0.47007      0.03788     12.40952      0.00000
Sigma      1.60011

```

Common Threshold Model (based on Transformed lwage)Tobit Model - The dependent variable  
The data set is: martins2

```

Grad:      0.0002      LogL: -3401.5617      Size:  0.250
Grad:      0.0001      LogL: -3401.5617      Size:  0.250
Grad:      0.0001      LogL: -3401.5617      Size:  0.250

```

Regressor	Coefficient	Std. Error	t-stat	Prob> t
Con	-0.94185	0.24209	-3.89052	0.00010
edu	0.21238	0.01499	14.16737	0.00000
pexp	0.07146	0.01841	3.88073	0.00011
pexp2	-0.00168	0.00036	-4.65424	0.00000
pexpchd	-0.01488	0.00478	-3.11637	0.00185
pexpch2	0.00031	0.00013	2.28622	0.02233
Sigma	1.60006	0.06061	26.39906	0.00000

-----Part A-----

---Confidence Interval---

(0.1830,0.2418)

-----Part B-----

---Variance-Covariance Matrix---

vc =

```

0.0586  -0.0019  -0.0035   0.0001   0.0001  -0.0000  -0.0033
-0.0019   0.0002   0.0001  -0.0000  -0.0000   0.0000   0.0005
-0.0035   0.0001   0.0003  -0.0000  -0.0000   0.0000   0.0003
0.0001  -0.0000  -0.0000   0.0000   0.0000  -0.0000  -0.0000
0.0001  -0.0000  -0.0000   0.0000   0.0000  -0.0000  -0.0001
-0.0000   0.0000   0.0000  -0.0000  -0.0000   0.0000   0.0000
-0.0033   0.0005   0.0003  -0.0000  -0.0001   0.0000   0.0037

```

Alpha	Beta	Delta	Gamma	Theta	Lambda	Sigma
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---Median Values---

MEDout =

2.0000    25.0000    6.0000    5.7000    298.8674

child      pexp          edu      lwage      wage

---Marginal Effect---

MargEff =

-3.4433

---Std Err---

MargSD =

1.3303

---Confidence Interval---

(-6.0507,-0.8360)

-----Part C-----

Tobit Model - The dependent variable is: wage2

The data set is: martins2

Grad:      0.0001      LogL: -5920.1049      Size:    1.000

Regressor	Coefficient	Std. Error	t-stat	Prob> t
Con	-13.90854	1.44897	-9.59891	0.00000
edu	2.00114	0.06246	32.03723	0.00000
pexp	0.38796	0.10927	3.55059	0.00039
pexp2	-0.00726	0.00217	-3.35333	0.00081
pexpchd	-0.04597	0.03170	-1.45009	0.14717
pexpch2	0.00068	0.00091	0.74116	0.45867
Sigma	10.33038	0.14686	70.34180	0.00000

---Confidence Interval---

(1.8787,2.1236)

-----Part D-----

gmwage =

3.8730

---Geometrically Scaled Wages---

Iteration Limit Exceeded in PD  
 Iteration Limit Exceeded in PD  
 Iteration Limit Exceeded in PD

Common Threshold Model (based on Transformed swage)  
 Tobit Model - The dependent variable is: gswage  
 The data set is: martins2

Grad:	8.7110	LogL:	-4927.4116	Size:	0.500
Grad:	11.2640	LogL:	-4114.9475	Size:	1.000
Grad:	0.3296	LogL:	-4037.0659	Size:	1.000
Grad:	0.6883	LogL:	-4027.2845	Size:	1.000
Grad:	0.2088	LogL:	-4025.1593	Size:	1.000
Grad:	0.0829	LogL:	-4024.6870	Size:	1.000
Grad:	0.0183	LogL:	-4024.5574	Size:	1.000
Grad:	0.0071	LogL:	-4024.5110	Size:	1.000
Grad:	0.0043	LogL:	-4024.4923	Size:	1.000
Grad:	0.0035	LogL:	-4024.4844	Size:	1.000
Grad:	0.0028	LogL:	-4024.4810	Size:	1.000
Grad:	0.0020	LogL:	-4024.4796	Size:	1.000
Grad:	0.0013	LogL:	-4024.4790	Size:	1.000
Grad:	0.0009	LogL:	-4024.4788	Size:	1.000
Grad:	0.0006	LogL:	-4024.4787	Size:	1.000
Grad:	0.0004	LogL:	-4024.4786	Size:	1.000
Grad:	0.0002	LogL:	-4024.4786	Size:	1.000
Grad:	0.0002	LogL:	-4024.4786	Size:	1.000
Grad:	0.0001	LogL:	-4024.4786	Size:	1.000
Grad:	0.0001	LogL:	-4024.4786	Size:	1.000

Regressor	Coefficient	Std. Error	t-stat	Prob> t
Con	-3.59134	0.37410	-9.59998	0.00000
edu	0.51670	0.01613	32.03960	0.00000
pexp	0.10019	0.02821	3.55156	0.00039
pexp2	-0.00188	0.00056	-3.35441	0.00081
pexpchd	-0.01187	0.00818	-1.45046	0.14707
pexpch2	0.00017	0.00024	0.74149	0.45847
Sigma	2.66724	0.03791	70.34793	0.00000

-----

---Geometrically Scaled Log-Wages---

Iteration Limit Exceeded in PD  
 Iteration Limit Exceeded in PD  
 Iteration Limit Exceeded in PD

Common Threshold Model (based on Transformed lgwage)

Tobit Model - The dependent variable is: lgwage

The data set is: martins2

Grad:	8.9935	LogL:	-4914.8767	Size:	0.500
Grad:	14.4850	LogL:	-4071.6009	Size:	0.500
Grad:	6.4257	LogL:	-2518.4323	Size:	1.000
Grad:	0.0692	LogL:	-2392.5370	Size:	1.000
Grad:	0.0504	LogL:	-2354.7076	Size:	1.000
Grad:	0.0375	LogL:	-2346.8652	Size:	1.000
Grad:	0.0491	LogL:	-2344.5035	Size:	1.000
Grad:	0.0314	LogL:	-2343.7956	Size:	1.000
Grad:	0.0176	LogL:	-2343.6064	Size:	1.000
Grad:	0.0073	LogL:	-2343.5607	Size:	1.000
Grad:	0.0056	LogL:	-2343.5509	Size:	1.000
Grad:	0.0021	LogL:	-2343.5505	Size:	1.000
Grad:	0.0038	LogL:	-2343.5462	Size:	0.500
Grad:	0.0006	LogL:	-2343.5460	Size:	1.000
Grad:	0.0012	LogL:	-2343.5457	Size:	0.500
Grad:	0.0002	LogL:	-2343.5456	Size:	1.000
Grad:	0.0004	LogL:	-2343.5456	Size:	0.500
Grad:	0.0001	LogL:	-2343.5456	Size:	1.000

Regressor	Coefficient	Std. Error	t-stat	Prob> t
Con	-0.97134	0.11959	-8.12244	0.00000
edu	0.14008	0.00611	22.93052	0.00000
pexp	0.03328	0.00885	3.76006	0.00017
pexp2	-0.00067	0.00017	-3.82255	0.00014
pexpchd	-0.00464	0.00247	-1.88020	0.06021
pexpch2	0.00008	0.00007	1.12471	0.26083
Sigma	0.80155	0.01984	40.39400	0.00000

## Code:

```

1 % Computer Take-Home 2: Question 2
2 % Metrics III
3 %Oscar Martinez
4
5 %Diary
6 diary Q1_Output_Oscar_Martinez.txt
7

```



```
8 %Introduction
9 fprintf('_____\\n')
10 ;
11 fprintf('Oscar Martinez \\t Take-Home 2: Question 2 \\t Metrics III\\n');
12 fprintf('_____\\n')
13 ;
14 clear
15 clc
16 load martins;
17 % shorten length of pexpchd2;
18 pexpch2=pexpchd2;
19 save martins1;
20 clear;
21 % Now identify common threshold and transform lwage
22 load martins1;
23 nemploy=(employ==0);
24 nobs=size(employ,1);
25
26 % create vector of maximums ;
27 mins=ones(nobs,1).*max(lwage);
28 lwage2=lwage.*employ+mins.*nemploy;
29 % generate lambda;
30 lambda=min(lwage2)
31 % create transformed dependent variable as lwage2;
32 % lwage2 equals lwage minus the minimum of non-zero ;
33 % values when employ=1 and 0 otherwise;
34 lwage2=(lwage2-lambda).*employ;
35 wage2=exp(lwage2);
36 clear lambda nobs ans;
37 save martins2;
38 clear;
39
40 %define text strings for tobit models;
41 data='martins2';
42 dep2='lwage2';
43 ind='edu      pexp      pexp2      pexpchd pexpch2 ';
44 bin='employ';
45
46 % Use HL to generate startingvalues;
47 [s1b,s2b]=tobit_hl(data,dep2,bin,ind);
48 theta=tobit2a(data,dep2,bin,ind,s2b);
49 theta(7)=exp(theta(7));
50 fprintf('Common Threshold Model (based on Transformed lwage)');
```

```

51 [theta, vc, stderr]=tobit3(data,dep2,bin,ind,theta);
52
53 %95% Confidence Interval
54 BCo=theta(2);
55 BErr=stderr(2);
56 LBCI=BCo-1.96*BErr; %Lower CI
57 HBCI=BCo+1.96*BErr; %Upper CI
58 fprintf('\n');
59 fprintf('\n-----Part A----- \n');
60 fprintf('---Confidence Interval---\n');
61 fprintf('(%2.4f,%2.4f)',LBCI, HBCI);
62
63 %Part B
64 fprintf('\n');
65 fprintf('\n-----Part B----- \n');
66 fprintf('---Variance-Covariance Matrix---\n');
67 vc
68 fprintf('      Alpha      Beta      Delta      Gamma      Theta      Lambda
        Sigma \n');
69
70 % get stats needed for marginal effect
71 load martins
72 lwage(lwage==0)=NaN;
73 M=[child, pexp, edu, lwage];
74 fprintf('\n---Median Values---');
75 MED=nanmedian(M);
76 MWAGE=exp(MED(1,4)); MPEXP=MED(1,2); MCHD=MED(1,1); MEDU=MED(1,3);
77 MEDout=[MED MWAGE]
78 fprintf('      child      pexp      edu      lwage      wage\n');
79
80
81 fprintf('\n---Marginal Effect---');
82 %Marginal Effect
83 M1=theta(3)+2*theta(4)*MPEXP+theta(5)*MCHD+2*theta(6)*MPEXP*MCHD; %Marginal
    Effect w/o lwage
84 MargEff=M1*MWAGE
85
86 %Variances
87 fprintf('\n---Std Err---');
88 VB=vc(2,2); VD=vc(3,3); VG=vc(4,4); VT=vc(5,5); VL=vc(6,6); CDG=vc(3,4);
    CDT=vc(3,5);
89 CDL=vc(3,6); CGT=vc(4,5); CGL=vc(4,6); CTL=vc(5,6);
90
91 %Could've combined them all into one equation but this is easier to error
92 %correct

```

```

93  VARR1=(VD+4*VG*MPEXP^2+MCHD^2*VT+4*MPEXP^2*MCHD^2*VL);
94  VARR2=2*2*MPEXP*CDG+2*MCHD*CDT+2*2*MPEXP*MCHD*CDL;
95  VARR3=2*2*MPEXP*MCHD*CGT+2*2*MPEXP*2*MPEXP*MCHD*CGL;
96  VARR4=2*MCHD*2*MPEXP*MCHD*CTL;
97  VSUM=VARR1+VARR2+VARR3+VARR4;
98  VMARG=MWAGE^2*VSUM;
99  MargSD=sqrt(VMARG)
100
101  %95% Confidence Interval
102  LBCI2=MargEff-1.96*MargSD;    %Lower CI
103  HBCI2=MargEff+1.96*MargSD;    %Upper CI
104  fprintf('——Confidence Interval——\n');
105  fprintf(' (%2.4f,%2.4f)',LBCI2, HBCI2);
106
107
108  fprintf('\n');
109  fprintf('\n————Part C———— \n');
110
111  %Backup Relevant Files
112  theta1=theta;
113  vc1=vc;
114
115  %define text strings for tobit models;
116  data='martins2';
117  dep2='wage2';
118  ind='edu      pexp      pexp2      pexpchd pexpch2 ';
119  bin='employ';
120
121  % Use HL to generate starting values;
122  [s1b,s2b]=tobit_hl2(data,dep2,bin,ind);
123  theta=tobit5a(data,dep2,bin,ind,s2b);
124  theta(7)=exp(theta(7));
125  %fprintf('Common Threshold Model (based on Transformed wage)\n');
126  [theta, vc, stderr]=tobit4(data,dep2,bin,ind,theta);
127
128  %CI
129  BCo=theta(2);
130  BErr=stderr(2);
131  LBCI=BCo-1.96*BErr;    %Lower CI
132  HBCI=BCo+1.96*BErr;    %Upper CI
133  fprintf('——Confidence Interval——\n');
134  fprintf(' (%2.4f,%2.4f)',LBCI, HBCI);
135
136  %Part D
137  fprintf('\n');

```

```
138 fprintf('\n————Part D———— \n');
139
140 %Backup Relevant Files
141 theta2=theta;
142 vc2=vc;
143
144 %Create Geometric scaled wages
145 load martins2;
146 gmwage=geomean(wage2)
147 gswage=(1/gmwage)*wage2;
148 lgwage=log(gswage);
149 save martins3;
150
151 %define text strings for tobit models;
152 data='martins3';
153 dep2='gswage';
154 ind='edu      pexp      pexp2      pexpchd pexpch2 ';
155 bin='employ';
156
157 fprintf('\n ———Geometrically Scaled Wages—— \n');
158
159 % Use HL to generate startingvalues;
160 [s1b,s2b]=tobit_hl2(data,dep2,bin,ind);
161 theta=tobitPD(data,dep2,bin,ind,s2b);
162 theta=tobitPD(data,dep2,bin,ind,theta);
163 theta=tobitPD(data,dep2,bin,ind,theta);
164 theta(7)=exp(theta(7));
165 fprintf('\n Common Threshold Model (based on Transformed swage)\n');
166 [theta, vc, stderr]=tobit4(data,dep2,bin,ind,theta);
167
168 %define text strings for tobit models;
169 data='martins3';
170 dep2='lgwage';
171 ind='edu      pexp      pexp2      pexpchd pexpch2 ';
172 bin='employ';
173
174 fprintf('\n ——— \n');
175
176 fprintf('\n ———Geometrically Scaled Log-Wages—— \n');
177
178 % Use HL to generate startingvalues;
179 [s1b,s2b]=tobit_hl2(data,dep2,bin,ind);
180 theta=tobitPD(data,dep2,bin,ind,s2b);
181 theta=tobitPD(data,dep2,bin,ind,theta);
182 theta=tobitPD(data,dep2,bin,ind,theta);
```

```
183 theta(7)=exp(theta(7));  
184 fprintf('\n Common Threshold Model (based on Transformed lgwage)\n');  
185 [theta, vc, stderr]=tobit4(data,dep2,bin,ind,theta);  
186  
187 %closing output  
188 diary off
```