

EXERCISE 0. HYPOTHESIS TESTING IN A MNL MODEL FITTED BY MLE

(Not to be handed in, for discussion the week of Oct. 23)

This exercise estimates a multinomial logit model by maximum likelihood, and tests several hypotheses using Wald, LR, and LM tests. The application is discrete choice of travel mode for shopping, with bus and auto alternatives, or the alternative of staying at home. The data are artificially generated. The SST program ex0.cmd is in the class data directory, and can be run using the unix command `sst ex0`. Output from one run of this program, which describes the hypotheses and test procedures, is reproduced below. Using SST, or transcribing this program into TSP, carry out the tests requested below. This exercise may be helpful for practical understanding of the lectures on discrete response models.

SST Spool File: ex0.out
Wed Sep 29 07:05:06 1999

AN EXAMPLE OF MAXIMUM LIKELIHOOD ESTIMATION
AND HYPOTHESIS TESTING FOR A MULTINOMIAL LOGIT MODEL

GENERATE SOME MADE-UP DATA, SHOPPING CHOICE OF (1) BUS,
(2) AUTO, OR (3) HOME (NO TRIP)

```

range obs[1-400]
set u = nrnd
set tt2= 60 + 10*u + 5*nrnd           # travel time mode 2
set tt1= 90 + 5*u + 5*nrnd           # travel time mode 1
set tc2=2+0.04*tt2-.5*nrnd           # travel cost mode 2
set tc1=1.25+0.01*tt1-.5*nrnd        # travel cost mode 1
set u = log(-log(urnd))               # extreme value RV
set y1=-0.1*tt1-tc1+10+ log(-log(urnd)) - u
set y2=-0.1*tt2-tc2+10+ log(-log(urnd)) - u
set one=1
set z=0
set i=1*(y1>vmax(y2,0))+2*(y2>=vmax(y1,0))+3*(0>vmax(y1,y2))
label var[tt1] lab[travel time bus]
label var[tt2] lab[travel time auto]
label var[tc1] lab[travel cost bus]
label var[tc2] lab[travel cost auto]
label var[i] lab[choice] val[1 bus 2 auto 3 home]
cova var[tt1 tt2 tc1 tc2] cov

```

Variable: tt1 travel time bus

Mean	90.23568	Standard deviation	7.07601
Minimum	70.83353	Skewness	-1.56729e-002
Maximum	1.13171e+002	Kurtosis	2.96962
Valid observations	400		

Variable: tt2 travel time auto

```

Mean          60.29345   Standard deviation   11.58801
Minimum       19.05803   Skewness           -0.12238
Maximum       1.02718e+002 Kurtosis           3.41181
Valid observations    400

```

```
Variable:   tc1   travel cost bus
```

```

Mean          2.14241   Standard deviation   0.51380
Minimum       0.61633   Skewness           -0.14033
Maximum       3.72038   Kurtosis           2.98541
Valid observations    400

```

```
Variable:   tc2   travel cost auto
```

```

Mean          4.42621   Standard deviation   0.70304
Minimum       2.48229   Skewness           2.39518e-002
Maximum       6.53560   Kurtosis           2.80964
Valid observations    400

```

```
Correlation and Covariance matrix
```

```

          tt1      tt2      tc1
tt1  49.94471    59.03648    0.51180
tt2   0.72179    1.33946e+002    0.18747
tc1   0.14113    3.15654e-002    0.26333
tc2   0.52834    0.72451    -1.57086e-002

          tc2
          tt1  2.62176
          tt2  5.88772
          tc1 -5.66009e-003
          tc2  0.49303

```

```
freq var[i]
```

```
i choice
400 valid observations
```

```

          bus      auto      home
          1        2        3
-----
Count      57      143      200
Percent    14.25    35.75    50.00

```

```
THIS COMPLETES GENERATION OF MADE-UP DATA
```

```
ESTIMATION OF BASE MNL MODEL
```

```

mnl dep[i] covmat[cvmat] coef[beta] \
ivalt[tt: tt1 tt2 z tc: tc1 tc2 z d1: one z z d2: z one z]

```

```
***** MULTINOMIAL LOGIT *****
```

```
Dependent variable: i
```

```

Value      Label      Count      Percent
  1         bus        57         14.25

```

2	auto	143	35.75
3	home	200	50.00

ITERATION 1: OLD LLF = -4.39445e+002 STEP = 1.29693
 NEW LLF = -3.11731e+002 GRAD*DIREC = 2.30453e+002

ITERATION 2: OLD LLF = -3.11731e+002 STEP = 0.99416
 NEW LLF = -3.10431e+002 GRAD*DIREC = 2.61166

ITERATION 3: OLD LLF = -3.10431e+002 STEP = 1.00326
 NEW LLF = -3.10429e+002 GRAD*DIREC = 3.90115e-003

At convergence grad * dir = 1.48498e-009

Independent Variable	Estimated Coefficient	Standard Error	t-Statistic
tt	-8.57271e-002	1.50024e-002	-5.71422
tc	-1.25794	0.20111	-6.25483
d1	9.07420	1.28205	7.07790
d2	10.17796	0.99883	10.18987

auxiliary statistics	at convergence	initial
log likelihood	-310.43	-439.44
number of observations	400	
percent correctly predicted	64	

calc ll0=_llk

matrix cvmat

	[1]	[2]	[3]
[1]	2.25072e-004	-1.05793e-003	-1.82733e-002
[2]	-1.05793e-003	4.04472e-002	1.66843e-002
[3]	-1.82733e-002	1.66843e-002	1.64365
[4]	-8.68450e-003	-0.11046	1.00648

	[4]
[1]	-8.68450e-003
[2]	-0.11046
[3]	1.00648
[4]	0.99766

CONSIDER THE FOLLOWING HYPOTHESES

H1: beta_tt = 0.1*beta_tc

H2: beta_d1 = beta_d2

h3: beta_d1 = beta_d2 = 10

HYPHESIS H1 TESTS WHETHER TRAVEL TIME IS VALUED AT 10 CENTS/MINUTE OR \$6/HOUR. IT IS A ONE-DIMENSIONAL HYPOTHESIS.

HYPOTHESIS H2 TESTS WHETHER THE ALTERNATIVE-SPECIFIC EFFECTS FOR AUTO AND BUS ARE THE SAME. THIS WILL BE TRUE IF THESE REFLECT THE COMMON PAYOFF TO SHOPPING, BUT FALSE IF THERE IS ANY SYSTEMATIC DIFFERENCE IN THE ATTRACTIVENESS OF AUTO AND BUS OTHER THAN TRAVEL TIME AND COST DIFFERENCES.

HYPOTHESIS H3 TESTS WHETHER THE ALTERNATIVE-SPECIFIC EFFECTS FOR AUTO AND BUS ARE BOTH EQUAL TO 10. THIS IS A 2-DIMENSIONAL HYPOTHESIS

TESTING H1. FIRST, BY REPARAMETERIZATION, THIS CAN BE STATED AS A HYPOTHESIS THAT A PARAMETER IS ZERO. DEFINE A NEW VARIABLE, GENERALIZED COST $gc = tc + tt/10$, AND ESTIMATE THE MODEL BELOW. IF H1 IS TRUE, THEN WITH THIS REPARAMETERIZATION, THE COEFFICIENT ON tt SHOULD BE ZERO, AND A T-TEST ON THIS COEFFICIENT IS A TEST OF H1. FOR COMPARISON WITH OTHER TEST STATISTICS, THE SQUARE OF THIS T-STATISTIC IS ALSO COMPUTED.

```
set gc1 = tc1+tt1/10
set gc2 = tc2+tt2/10
mnl dep[i] ivalt[tt: tt1 tt2 z gc: gc1 gc2 z d1: one z z d2: z one z] \
coef[beta1] covmat[cvmat1]
```

***** MULTINOMIAL LOGIT *****
 Dependent variable: i

Value	Label	Count	Percent
1	bus	57	14.25
2	auto	143	35.75
3	home	200	50.00

```
ITERATION 1: OLD LLF = -4.39445e+002 STEP = 1.29693
NEW LLF = -3.11731e+002 GRAD*DIREC = 2.30453e+002

ITERATION 2: OLD LLF = -3.11731e+002 STEP = 0.99416
NEW LLF = -3.10431e+002 GRAD*DIREC = 2.61166

ITERATION 3: OLD LLF = -3.10431e+002 STEP = 1.00326
NEW LLF = -3.10429e+002 GRAD*DIREC = 3.90115e-003
```

At convergence grad * dir = 1.48498e-009

Independent Variable	Estimated Coefficient	Standard Error	t-Statistic
tt	4.00669e-002	2.90022e-002	1.38151
gc	-1.25794	0.20111	-6.25483
d1	9.07420	1.28205	7.07790
d2	10.17796	0.99883	10.18986

```
auxiliary statistics at convergence initial
log likelihood -310.43 -439.44
number of observations 400
percent correctly predicted 64
```

```
calc beta1[1]^2/cvmat1[1,1]
1.90857
```

SECOND, A WALD TEST CAN BE CALCULATED FROM THE BASE MODEL.

THE TEST STATISTIC IS ASYMP. CHI-SQUARED WITH 1 D.F. UNDER H1.

```
matrix A = {1; -0.1; 0; 0}
matrix A
      [ 1]
[ 1]  1.00000
[ 2] -0.10000
[ 3]  0.00000
[ 4]  0.00000

matrix beta'*A*inv(A'*cvmat*A)*A'*beta
      [ 1]
[ 1]  1.90857
```

THIRD, A LR TEST CAN BE CALCULATED BY RUNNING THE MODEL UNDER H1. THE TEST STATISTIC IS AGAIN ASYMP. CHI-SQUARED WITH 1 D.F.

```
mnl dep[i] ivalt[gc: gc1 gc2 z d1: one z z d2: z one z] \
      prob[pb2 pb3]
```

***** MULTINOMIAL LOGIT *****
 Dependent variable: i

Value	Label	Count	Percent
1	bus	57	14.25
2	auto	143	35.75
3	home	200	50.00

```
ITERATION 1: OLD LLF = -4.39445e+002 STEP = 1.29878
NEW LLF = -3.11608e+002 GRAD*DIREC = 2.30446e+002
```

```
ITERATION 2: OLD LLF = -3.11608e+002 STEP = 0.99555
NEW LLF = -3.11390e+002 GRAD*DIREC = 0.43751
```

At convergence grad * dir = 1.73179e-004

Independent Variable	Estimated Coefficient	Standard Error	t-Statistic
gc	-1.01759	9.78282e-002	-10.40186
d1	10.04989	1.08799	9.23710
d2	10.09161	0.99991	10.09254

auxiliary statistics	at convergence	initial
log likelihood	-311.39	-439.44
number of observations	400	
percent correctly predicted	63	

```
calc lla=_llk
calc lr = 2*(ll0 - lla)
calc lr
```

1.92212

FOURTH, A LM TEST CAN BE CALCULATED BY AUXILIARY REGRESSION:
CONSTRUCT THE SCORE FOR THE UNRESTRICTED MODEL AT THE
RESTRICTED ESTIMATES. REGRESS 1 ON THIS SCORE. THEN LM
IS THE SUM OF SQUARED FITTED VALUES.

```
set pb1 = 1 - pb2 - pb3    # fitted probabilities
set r1 = (i==1)-pb1      # generalized residuals
set r2 = (i==2)-pb2
set st = tt1*r1+tt2*r2   # score, tt
set sc = tc1*r1+tc2*r2   # score, tc
set sd1 = r1             # score, d1
set sd2 = r2             # score, d2
reg dep[(1)] ind[st sc sd1 sd2] pred[onehat]
```

***** ORDINARY LEAST SQUARES ESTIMATION *****

Dependent Variable: (1)

Independent Variable	Estimated Coefficient	Standard Error	t-Statistic
st	1.54258e-002	1.50540e-002	1.02469
sc	-0.22701	0.19346	-1.17338
sd1	-0.93538	1.27653	-0.73275
sd2	8.49096e-002	0.99540	8.53022e-002

Number of Observations 400
R-squared 1.00000
Corrected R-squared 0.00000
Sum of Squared Residuals 3.98165e+002
Standard Error of the Regression 1.00273
Durbin-Watson Statistic 9.42493e-003
Mean of Dependent Variable 1.00000

```
calc sum(onehat^2)
1.83473
```

THE HYPOTHESIS H1 IS TRUE, AND THE RESULTING TEST
STATISTICS ARE SIMILAR. TEST ANOTHER H1 HYPOTHESIS
THAT IS FALSE, SAY $\beta_{tt} = 0.05\beta_{tc}$. THE
STATISTICS IN THIS CASE ARE LIKELY TO ALL REJECT
THE NULL, BUT WILL BE MORE SPREAD OUT IN VALUES.
FOR THE EXERCISE, FORM THE HYPOTHESES H2 AND H3 AND
CALCULATE THE TEST STATISTICS FOR THEM USING THE
PROCEDURES GIVEN ABOVE, MODIFIED AS NEEDED.