

LOG FILE  
April 17, 2016

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. do "C:\Prog_Stata\fedora160416.do"
. * 16 April 2016
. * EQ DCE Competition
. * Fedora Predictions
. * Benjamin M. Craig
. *
. * This code demonstrates my approach to modeling EQ-5D-5L valuation data
. * and how I produced the predictions for the DCE competition.
. *
. * To use this code, you will need access to STATA software and either
. * internet access or downloads of 2 comma-separated values (CSV) files:
. *     Prediction_file.csv
. *     Exploratory_Data.csv
. *
. * Specifically, this code estimates 1 model, Fedora, which includes a 20-
. * parameter regression (one for each health problem in the EQ-5D-5L) and
. * 2 ancillary parameters (i.e., 22 parameters).
. *
. * Most model assumes constant proportionality:
. *
. *     Value = Lifespan - Problems * Duration
. *
. * Fedora assumes:
. *
. *     Value = Lifespan^alpha - Problems * Duration^beta
. *
. * NOTE: if alpha and beta are 1, the model is constant proportional.
. *
. * The code is divided into 3 sections:
. *     (1) Import and recode responses into analytical data
. *     (2) Estimate Fedora using analytical data
. *     (3) Predict probabilities and estimate chi-square for Fedora
. *
. * For additional information, please email: benjamin.craig@iahpr.org
. * SECTION 1: Import and recode responses into analytical data
. local folder="http://iahpr.org/wordpress/wp-content/uploads/2016/03/"
. * Exploratory data
. import delimited using "`folder'/Exploratory_Data.csv", clear case(preserve)
(7 vars, 81,480 obs)
. save exploratory, replace
file exploratory.dta saved
. * Prediction file
. import delimited using "`folder'/Prediction_file.csv", clear case(preserve)
(5 vars, 3,200 obs)
. save predict, replace
file predict.dta saved
. * Recode the exploratory data
. use exploratory, clear
. * Switch A and B in each pair to be identical [left/right]
. gen switch=(H_A>H_B)
. replace choice=1-choice if switch==1
(40,931 real changes made)
. qui for X in any T H: gen D=X_A \ replace X_A=X_B if switch==1 \ /*
> */ replace X_B=D if switch==1 \ drop D
. * Calculate pair probabilities
. bysort p_id: egen C=mean(choice) /* pair probabilities */
. bysort p_id: gen N=_N                  /* pair sample sizes */
. bysort p_id: keep if _n==1
(79,920 observations deleted)
. drop r_id choice switch
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. save pairs, replace
file pairs.dta saved
. * Combine the blank prediction file with the pairs data
. use predict, clear
. gen switch=(H_A>H_B)
. qui for X in any T H: gen D=X_A \ replace X_A=X_B if switch==1 \ /*
> */ replace X_B=D if switch==1 \ drop D
. joinby p_id using pairs, unmatched(master)
. gen analysis=_merge==3
. drop switch _merge
. * Recode lifespans into years
. qui for X in any A B: gen X_time=substr(T_X,1,strpos(T_X," ")-1) \ /*
> */ destring X_time, replace force \ /*
> */ replace X_time=X_time/365.25 if strpos(T_X,"day")>0 \ /*
> */ replace X_time=X_time*7/365.25 if strpos(T_X,"week")>0 \ /*
> */ replace X_time=X_time/12 if strpos(T_X,"month")>0 \ /*
> */ replace X_time=0 if X_time==.
. * Separate the 5 domains of the EQ-5D-5L
. qui for X in any A B: tostring H_X, replace \ /*
> */ gen X_MO=substr(H_X,1,1) \ gen X_SC=substr(H_X,2,1) \ /*
> */ gen X_UA=substr(H_X,3,1) \ gen X_PD=substr(H_X,4,1) \ /*
> */ gen X_AD=substr(H_X,5,1) \ destring X_MO-X_AD, replace
. * Incorporate the 5 attributes using their additive differences in value
. gen dif_time=A_time-B_time
. qui for X in any MO SC UA PD AD: /*
> */ gen X1=(B_X>1)-(A_X>1) \ /*
> */ gen X2=(B_X>2)-(A_X>2) \ /*
> */ gen X3=(B_X>3)-(A_X>3) \ /*
> */ gen X4=(B_X>4)-(A_X>4)
. * Separate the disadvantages of A from the disadvantages of B
. qui for X in varlist MO1-AD4: gen A_X = X*(X>0)
. qui for X in varlist MO1-AD4: gen B_X = -X*(X<0)
. qui for X in any MO SC UA PD AD: /*
> */ gen a_X1=(A_X>1) \ /*
> */ gen a_X2=(A_X>2) \ /*
> */ gen a_X3=(A_X>3) \ /*
> */ gen a_X4=(A_X>4)
. qui for X in any MO SC UA PD AD: /*
> */ gen b_X1=(B_X>1) \ /*
> */ gen b_X2=(B_X>2) \ /*
> */ gen b_X3=(B_X>3) \ /*
> */ gen b_X4=(B_X>4)
. save analysis, replace
file analysis.dta saved
. * SECTION 2: Estimate Fedora using analytical data
. use analysis, clear
. * Constrain parameters of the 4 equations (A,B,a,b) to be equal
. qui for Y in numlist 1/20 \ /*
> */ X_A in varlist A_MO1-A_AD4 \ /*
> */ X_B in varlist B_MO1-B_AD4 : /*
> */ constraint Y [eq1a]X_A=[eq1b]X_B
. qui for Y in numlist 21/40 \ /*
> */ X_A in varlist a_MO1-a_AD4 \ /*
> */ X_B in varlist b_MO1-b_AD4 : /*
> */ constraint Y [eq2a]X_A=[eq2b]X_B
. qui for Y in numlist 41/60 \ /*
> */ X_A in varlist A_MO1-A_AD4 \ /*
> */ X_a in varlist a_MO1-a_AD4 : /*
> */ constraint Y [eq1a]X_A=[eq2a]X_a
. * Estimate Fedora with Constant Proportionality Assumption
. capture program drop wls_federal1
. program wls_federal1
1. args lnf `xb1` `xb2` `zb1` `zb2`

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2. tempvar P A_V B_V w
3.
. * Disadvantages of B
. qui gen double `A_V'=`xb1'*$ML_y4+($ML_y3-$ML_y4) if ($ML_y3>$ML_y4)
4. qui replace `A_V'=`xb1'*$ML_y3+`zb2'*($ML_y4-$ML_y3) if ($ML_y4>=$ML_y3)
5. qui replace `A_V'=`zb2'*$ML_y4 if ($ML_y3==0)
6. qui replace `A_V'=$ML_y3 if ($ML_y4==0)
7.
. * Disadvantages of A
. qui gen double `B_V'=`xb2'*$ML_y3+($ML_y4-$ML_y3) if ($ML_y4>$ML_y3)
8. qui replace `B_V'=`xb2'*$ML_y4+`zb1'*($ML_y3-$ML_y4) if ($ML_y3>=$ML_y4)
9. qui replace `B_V'=`zb1'*$ML_y3 if ($ML_y4==0)
10. qui replace `B_V'=$ML_y4 if ($ML_y3==0)
11.
. qui gen double `P'=`A_V`/(`A_V'+`B_V')
12.
. * Construct theoretical weights for Weighted Least Squares Estimation
. qui gen double `w'=(1/(`P'*(1-`P'))) /* Uncertainty Weight */
13. qui replace `w'=(4*($ML_y2^2)/(2*$ML_y2-1)) if (`P'<=0)+(`P'>=1)
/* Berkson Weight if w==. */
14. qui replace `lnf' =-((($ML_y1-`P')^2))*`w'
15. end
. mat m0=J(1,80,0.1)
. ml model lf wls_federal /*
> */ (eq1a:C= A_MO1-A_AD4 , nocons) /*
> */ (eq1b:N= B_MO1-B_AD4 , nocons) /*
> */ (eq2a:A_time= a_MO1-a_AD4 , nocons) /*
> */ (eq2b:B_time= b_MO1-b_AD4 , nocons) /*
> */ [fweight=N] if analysis==1, /*
> */ tech(nr) init(m0, copy) maximize missing /*
> */ norestore search(off) iterate(50) /*
> */ difficult constraint(1/60) collinear
Iteration 0: log likelihood = -7965.5838
Iteration 1: log likelihood = -5940.9886
Iteration 2: log likelihood = -5243.6017
Iteration 3: log likelihood = -5174.1328
Iteration 4: log likelihood = -5173.9628
Iteration 5: log likelihood = -5173.9628
. ml display

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	Number of obs	=	81,480
	Wald chi2(0)	=	.
	Prob > chi2	=	.

		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
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eq1a						
A_MO1		.0700658	.0037549	18.66	0.000	.0627064 .0774252
A_MO2		.0400127	.0042777	9.35	0.000	.0316286 .0483968
A_MO3		.1783307	.0086707	20.57	0.000	.1613364 .195325
A_MO4		.1269834	.0101185	12.55	0.000	.1071516 .1468153
A_SC1		.0243234	.0041759	5.82	0.000	.0161387 .0325081
A_SC2		.0447697	.0038815	11.53	0.000	.0371621 .0523773
A_SC3		.1746956	.0103272	16.92	0.000	.1544546 .1949366
A_SC4		.1875099	.0127832	14.67	0.000	.1624553 .2125646
A_UA1		.0727943	.0043238	16.84	0.000	.0643197 .0812689
A_UA2		.018828	.0059603	3.16	0.002	.007146 .0305101
A_UA3		.1223404	.0078527	15.58	0.000	.1069493 .1377315
A_UA4		.0560833	.0094131	5.96	0.000	.0376339 .0745327
A_PD1		.0608278	.0044962	13.53	0.000	.0520155 .0696401
A_PD2		.051347	.0052237	9.83	0.000	.0411088 .0615853
A_PD3		.5138365	.0106968	48.04	0.000	.4928712 .5348018
A_PD4		.1104702	.01351	8.18	0.000	.0839911 .1369493
A_AD1		.0904724	.0036101	25.06	0.000	.0833967 .0975481

A_AD2	.0686826	.0078742	8.72	0.000	.0532495	.0841158
A_AD3	.3122867	.010156	30.75	0.000	.2923813	.3321921
A_AD4	.0724193	.0062982	11.50	0.000	.0600751	.0847635
<hr/>						
eq1b						
B_MO1	.0700658	.0037549	18.66	0.000	.0627064	.0774252
B_MO2	.0400127	.0042777	9.35	0.000	.0316286	.0483968
B_MO3	.1783307	.0086707	20.57	0.000	.1613364	.195325
B_MO4	.1269834	.0101185	12.55	0.000	.1071516	.1468153
B_SC1	.0243234	.0041759	5.82	0.000	.0161387	.0325081
B_SC2	.0447697	.0038815	11.53	0.000	.0371621	.0523773
B_SC3	.1746956	.0103272	16.92	0.000	.1544546	.1949366
B_SC4	.1875099	.0127832	14.67	0.000	.1624553	.2125646
B_UA1	.0727943	.0043238	16.84	0.000	.0643197	.0812689
B_UA2	.018828	.0059603	3.16	0.002	.007146	.0305101
B_UA3	.1223404	.0078527	15.58	0.000	.1069493	.1377315
B_UA4	.0560833	.0094131	5.96	0.000	.0376339	.0745327
B_PD1	.0608278	.0044962	13.53	0.000	.0520155	.0696401
B_PD2	.051347	.0052237	9.83	0.000	.0411088	.0615853
B_PD3	.5138365	.0106968	48.04	0.000	.4928712	.5348018
B_PD4	.1104702	.01351	8.18	0.000	.0839911	.1369493
B_AD1	.0904724	.0036101	25.06	0.000	.0833967	.0975481
B_AD2	.0686826	.0078742	8.72	0.000	.0532495	.0841158
B_AD3	.3122867	.010156	30.75	0.000	.2923813	.3321921
B_AD4	.0724193	.0062982	11.50	0.000	.0600751	.0847635
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eq2a						
a_MO1	.0700658	.0037549	18.66	0.000	.0627064	.0774252
a_MO2	.0400127	.0042777	9.35	0.000	.0316286	.0483968
a_MO3	.1783307	.0086707	20.57	0.000	.1613364	.195325
a_MO4	.1269834	.0101185	12.55	0.000	.1071516	.1468153
a_SC1	.0243234	.0041759	5.82	0.000	.0161387	.0325081
a_SC2	.0447697	.0038815	11.53	0.000	.0371621	.0523773
a_SC3	.1746956	.0103272	16.92	0.000	.1544546	.1949366
a_SC4	.1875099	.0127832	14.67	0.000	.1624553	.2125646
a_UA1	.0727943	.0043238	16.84	0.000	.0643197	.0812689
a_UA2	.018828	.0059603	3.16	0.002	.007146	.0305101
a_UA3	.1223404	.0078527	15.58	0.000	.1069493	.1377315
a_UA4	.0560833	.0094131	5.96	0.000	.0376339	.0745327
a_PD1	.0608278	.0044962	13.53	0.000	.0520155	.0696401
a_PD2	.051347	.0052237	9.83	0.000	.0411088	.0615853
a_PD3	.5138365	.0106968	48.04	0.000	.4928712	.5348018
a_PD4	.1104702	.01351	8.18	0.000	.0839911	.1369493
a_AD1	.0904724	.0036101	25.06	0.000	.0833967	.0975481
a_AD2	.0686826	.0078742	8.72	0.000	.0532495	.0841158
a_AD3	.3122867	.010156	30.75	0.000	.2923813	.3321921
a_AD4	.0724193	.0062982	11.50	0.000	.0600751	.0847635
<hr/>						
eq2b						
b_MO1	.0700658	.0037549	18.66	0.000	.0627064	.0774252
b_MO2	.0400127	.0042777	9.35	0.000	.0316286	.0483968
b_MO3	.1783307	.0086707	20.57	0.000	.1613364	.195325
b_MO4	.1269834	.0101185	12.55	0.000	.1071516	.1468153
b_SC1	.0243234	.0041759	5.82	0.000	.0161387	.0325081
b_SC2	.0447697	.0038815	11.53	0.000	.0371621	.0523773
b_SC3	.1746956	.0103272	16.92	0.000	.1544546	.1949366
b_SC4	.1875099	.0127832	14.67	0.000	.1624553	.2125646
b_UA1	.0727943	.0043238	16.84	0.000	.0643197	.0812689
b_UA2	.018828	.0059603	3.16	0.002	.007146	.0305101
b_UA3	.1223404	.0078527	15.58	0.000	.1069493	.1377315
b_UA4	.0560833	.0094131	5.96	0.000	.0376339	.0745327
b_PD1	.0608278	.0044962	13.53	0.000	.0520155	.0696401
b_PD2	.051347	.0052237	9.83	0.000	.0411088	.0615853

	b_PD3	b_PD4	b_AD1	b_AD2	b_AD3	b_AD4	
	.5138365	.0106968	48.04	0.000	.4928712	.5348018	
	.1104702	.01351	8.18	0.000	.0839911	.1369493	
	.0904724	.0036101	25.06	0.000	.0833967	.0975481	
	.0686826	.0078742	8.72	0.000	.0532495	.0841158	
	.3122867	.010156	30.75	0.000	.2923813	.3321921	
	.0724193	.0062982	11.50	0.000	.0600751	.0847635	

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. mat b_fedora0=e(b)
. * NOTE: This is identical to the Bradley-Terry WLS model
. * in the Exploratory Analysis, except that it uses the theoretical weights.
. * Estimate Fedora with theta, alpha and beta
. capture program drop wls_federal1
. program wls_federal1
 1. args lnf xb1 xb2 zb1 zb2 alpha beta
 2. tempvar P A_V B_V T_A T_B D_A D_B w
 3. qui gen double `T_A'=`alpha'*ln($ML_y3) /* Alpha is the effect of lifespan */
 4. qui gen double `T_B'=`alpha'*ln($ML_y4)
 5. qui gen double `D_A'=`beta'*ln($ML_y3) /* Beta is the effect of duration */
 6. qui gen double `D_B'=`beta'*ln($ML_y4)
 7.
. * Disadvantages of B
. qui gen double `A_V'=`xb1'*exp(`D_B')+(exp(`T_A')-exp(`T_B')) if ($ML_y3>$ML_y4)
 8. qui replace `A_V'=`xb1'*exp(`D_A')+`zb2'* (exp(`D_B')-exp(`D_A')) if
($ML_y4>=$ML_y3)
 9. qui replace `A_V'=`zb2'*exp(`D_B') if ($ML_y3==0)
10. qui replace `A_V'=exp(`T_A') if ($ML_y4==0)
11.
. * Disadvantages of A
. qui gen double `B_V'=`xb2'*exp(`D_A')+(exp(`T_B')-exp(`T_A')) if ($ML_y4>$ML_y3)
12. qui replace `B_V'=`xb2'*exp(`D_B')+`zb1'* (exp(`D_A')-exp(`D_B')) if
($ML_y3>=$ML_y4)
13. qui replace `B_V'=`zb1'*exp(`D_A') if ($ML_y4==0)
14. qui replace `B_V'=exp(`T_B') if ($ML_y3==0)
15.
. qui gen double `P'=`A_V'/(`A_V'+`B_V')
16.
. * Construct theoretical weights for Weighted Least Squares Estimation
. qui gen double `w'=(1/(`P'*(1-`P'))) /* Uncertainty Weight */
17. qui replace `w'=(4*($ML_y2^2)/(2*$ML_y2-1)) if (`P'<=0)+(`P'>=1)
/* Berkson Weight if w==. */
18. qui replace `lnf' =-((($ML_y1-`P')^2))*`w'
19. end
. mat m1=b_fedora0,0.5,0.5
. ml model lf wls_federal1 /*
> */ (eq1a:C= A_M01-A_AD4 , nocons) /*
> */ (eq1b:N= B_M01-B_AD4 , nocons) /*
> */ (eq2a:A_time= a_M01-a_AD4 , nocons) /*
> */ (eq2b:B_time= b_M01-b_AD4 , nocons) /*
> */ /alpha /beta [fweight=N] if analysis==1, /*
> */ tech(nr) init(m1, copy) maximize missing /*
> */ nopreserve search(off) iterate(50) /*
> */ difficult constraint(1/60) collinear
Iteration 0: log likelihood = -5722.8212
Iteration 1: log likelihood = -5273.0728 (not concave)
Iteration 2: log likelihood = -3588.8411 (not concave)
Iteration 3: log likelihood = -3438.9318 (not concave)
Iteration 4: log likelihood = -3409.5025 (not concave)
Iteration 5: log likelihood = -3179.2126 (not concave)
Iteration 6: log likelihood = -2946.6595
Iteration 7: log likelihood = -2932.1761
Iteration 8: log likelihood = -2677.0851
Iteration 9: log likelihood = -2655.6007
Iteration 10: log likelihood = -2654.5209

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Iteration 11: log likelihood = -2654.5183  
 Iteration 12: log likelihood = -2654.5183  
 . ml display

							Number of obs	=	81,480
							Wald chi2(0)	=	.
							Prob > chi2	=	.
Log likelihood = -2654.5183									
							[95% Conf. Interval]		
		Coef.	Std. Err.	z	P> z				
eq1a									
	A_MO1	.0162336	.0011303	14.36	0.000	.0140183	.0184489		
	A_MO2	.0294416	.0023703	12.42	0.000	.024796	.0340873		
	A_MO3	.1309349	.0049108	26.66	0.000	.12131	.1405599		
	A_MO4	.0842535	.0049709	16.95	0.000	.0745107	.0939962		
	A_SC1	.0177395	.0014686	12.08	0.000	.0148611	.0206179		
	A_SC2	.0310703	.0020234	15.36	0.000	.0271044	.0350362		
	A_SC3	.1391488	.0057685	24.12	0.000	.1278428	.1504547		
	A_SC4	.0937537	.006057	15.48	0.000	.0818823	.1056252		
	A_UA1	.0140187	.0012422	11.29	0.000	.011584	.0164533		
	A_UA2	.0146191	.002391	6.11	0.000	.0099327	.0193054		
	A_UA3	.1136766	.004555	24.96	0.000	.104749	.1226042		
	A_UA4	.0368734	.0053118	6.94	0.000	.0264624	.0472845		
	A_PD1	.014749	.0015129	9.75	0.000	.0117838	.0177143		
	A_PD2	.0260796	.001882	13.86	0.000	.0223909	.0297683		
	A_PD3	.2686381	.0068625	39.15	0.000	.2551878	.2820883		
	A_PD4	.099022	.0065762	15.06	0.000	.086133	.1119111		
	A_AD1	.0357312	.0013896	25.71	0.000	.0330077	.0384547		
	A_AD2	.0536931	.0040137	13.38	0.000	.0458263	.0615599		
	A_AD3	.1484151	.0054766	27.10	0.000	.1376811	.1591491		
	A_AD4	.0468012	.0035243	13.28	0.000	.0398938	.0537086		
eq1b									
	B_MO1	.0162336	.0011303	14.36	0.000	.0140183	.0184489		
	B_MO2	.0294416	.0023703	12.42	0.000	.024796	.0340873		
	B_MO3	.1309349	.0049108	26.66	0.000	.12131	.1405599		
	B_MO4	.0842535	.0049709	16.95	0.000	.0745107	.0939962		
	B_SC1	.0177395	.0014686	12.08	0.000	.0148611	.0206179		
	B_SC2	.0310703	.0020234	15.36	0.000	.0271044	.0350362		
	B_SC3	.1391488	.0057685	24.12	0.000	.1278428	.1504547		
	B_SC4	.0937537	.006057	15.48	0.000	.0818823	.1056252		
	B_UA1	.0140187	.0012422	11.29	0.000	.011584	.0164533		
	B_UA2	.0146191	.002391	6.11	0.000	.0099327	.0193054		
	B_UA3	.1136766	.004555	24.96	0.000	.104749	.1226042		
	B_UA4	.0368734	.0053118	6.94	0.000	.0264624	.0472845		
	B_PD1	.014749	.0015129	9.75	0.000	.0117838	.0177143		
	B_PD2	.0260796	.001882	13.86	0.000	.0223909	.0297683		
	B_PD3	.2686381	.0068625	39.15	0.000	.2551878	.2820883		
	B_PD4	.099022	.0065762	15.06	0.000	.086133	.1119111		
	B_AD1	.0357312	.0013896	25.71	0.000	.0330077	.0384547		
	B_AD2	.0536931	.0040137	13.38	0.000	.0458263	.0615599		
	B_AD3	.1484151	.0054766	27.10	0.000	.1376811	.1591491		
	B_AD4	.0468012	.0035243	13.28	0.000	.0398938	.0537086		
eq2a									
	a_MO1	.0162336	.0011303	14.36	0.000	.0140183	.0184489		
	a_MO2	.0294416	.0023703	12.42	0.000	.024796	.0340873		
	a_MO3	.1309349	.0049108	26.66	0.000	.12131	.1405599		
	a_MO4	.0842535	.0049709	16.95	0.000	.0745107	.0939962		
	a_SC1	.0177395	.0014686	12.08	0.000	.0148611	.0206179		
	a_SC2	.0310703	.0020234	15.36	0.000	.0271044	.0350362		
	a_SC3	.1391488	.0057685	24.12	0.000	.1278428	.1504547		
	a_SC4	.0937537	.006057	15.48	0.000	.0818823	.1056252		
	a_UA1	.0140187	.0012422	11.29	0.000	.011584	.0164533		

a_UA2		.0146191	.002391	6.11	0.000	.0099327	.0193054
a_UA3		.1136766	.004555	24.96	0.000	.104749	.1226042
a_UA4		.0368734	.0053118	6.94	0.000	.0264624	.0472845
a_PD1		.014749	.0015129	9.75	0.000	.0117838	.0177143
a_PD2		.0260796	.001882	13.86	0.000	.0223909	.0297683
a_PD3		.2686381	.0068625	39.15	0.000	.2551878	.2820883
a_PD4		.099022	.0065762	15.06	0.000	.086133	.1119111
a_AD1		.0357312	.0013896	25.71	0.000	.0330077	.0384547
a_AD2		.0536931	.0040137	13.38	0.000	.0458263	.0615599
a_AD3		.1484151	.0054766	27.10	0.000	.1376811	.1591491
a_AD4		.0468012	.0035243	13.28	0.000	.0398938	.0537086
<hr/>							
eq2b							
b_MO1		.0162336	.0011303	14.36	0.000	.0140183	.0184489
b_MO2		.0294416	.0023703	12.42	0.000	.024796	.0340873
b_MO3		.1309349	.0049108	26.66	0.000	.12131	.1405599
b_MO4		.0842535	.0049709	16.95	0.000	.0745107	.0939962
b_SC1		.0177395	.0014686	12.08	0.000	.0148611	.0206179
b_SC2		.0310703	.0020234	15.36	0.000	.0271044	.0350362
b_SC3		.1391488	.0057685	24.12	0.000	.1278428	.1504547
b_SC4		.0937537	.006057	15.48	0.000	.0818823	.1056252
b_UA1		.0140187	.0012422	11.29	0.000	.011584	.0164533
b_UA2		.0146191	.002391	6.11	0.000	.0099327	.0193054
b_UA3		.1136766	.004555	24.96	0.000	.104749	.1226042
b_UA4		.0368734	.0053118	6.94	0.000	.0264624	.0472845
b_PD1		.014749	.0015129	9.75	0.000	.0117838	.0177143
b_PD2		.0260796	.001882	13.86	0.000	.0223909	.0297683
b_PD3		.2686381	.0068625	39.15	0.000	.2551878	.2820883
b_PD4		.099022	.0065762	15.06	0.000	.086133	.1119111
b_AD1		.0357312	.0013896	25.71	0.000	.0330077	.0384547
b_AD2		.0536931	.0040137	13.38	0.000	.0458263	.0615599
b_AD3		.1484151	.0054766	27.10	0.000	.1376811	.1591491
b_AD4		.0468012	.0035243	13.28	0.000	.0398938	.0537086
<hr/>							
alpha							
_cons		.2747722	.0066482	41.33	0.000	.261742	.2878024
<hr/>							
beta							
_cons		.1483939	.0064667	22.95	0.000	.1357194	.1610684
<hr/>							
. mat b_fedora1=e(b)							
. * Review the parameters of the two models							
. mat results=(b_fedora0[1,1..20],1,1',(b_fedora1[1,1..20],b_fedora1[1,81..82]))'							
. mat colnames results = CP Fedora							
. mat list results							
results[22,2]							
CP            Fedora							
eq1a:A_MO1		.07006582	.01623361				
eq1a:A_MO2		.04001268	.02944163				
eq1a:A_MO3		.17833074	.13093495				
eq1a:A_MO4		.12698344	.08425349				
eq1a:A_SC1		.02432339	.0177395				
eq1a:A_SC2		.0447697	.03107029				
eq1a:A_SC3		.17469562	.13914878				
eq1a:A_SC4		.18750992	.09375374				
eq1a:A_UA1		.07279431	.01401866				
eq1a:A_UA2		.01882804	.01461905				
eq1a:A_UA3		.12234042	.11367658				
eq1a:A_UA4		.0560833	.03687342				
eq1a:A_PD1		.06082777	.01474902				
eq1a:A_PD2		.05134702	.02607958				
eq1a:A_PD3		.51383655	.26863808				
eq1a:A_PD4		.11047017	.099022				

```

eq1a:A_AD1 .09047241 .03573116
eq1a:A_AD2 .06868265 .05369309
eq1a:A_AD3 .3122867 .14841512
eq1a:A_AD4 .07241929 .04680122
    r21      1 .27477218
    r22      1 .14839392
. * NOTE: The inclusion of 2 parameters (alpha and beta) relax the constant
. * proportionality assumption and reduce the chi square by over half (3166).
. * SECTION 3: Predict probabilities and estimate chi-square for Fedora
. use analysis, clear
. * Predict the probabilities for the Fedora model (theta, alpha, beta)
. gen A_f1=0
. qui for X in varlist A_MO1-A_AD4 \ Y in numlist 1/20: /*
> /* replace A_f1=A_f1+X*b_federal[1,Y]
. gen B_f1=0
. qui for X in varlist B_MO1-B_AD4 \ Y in numlist 1/20: /*
> /* replace B_f1=B_f1+X*b_federal[1,Y]
. gen a_f1=0
. qui for X in varlist a_MO1-a_AD4 \ Y in numlist 1/20: /*
> /* replace a_f1=a_f1+X*b_federal[1,Y]
. gen b_f1=0
. qui for X in varlist b_MO1-b_AD4 \ Y in numlist 1/20: /*
> /* replace b_f1=b_f1+X*b_federal[1,Y]
. qui gen double t_A=A_time^b_federal[1,81] /* Alpha is the effect of lifespan */
. qui gen double t_B=B_time^b_federal[1,81]
. qui gen double D_A=A_time^b_federal[1,82] /* Beta is the effect of duration */
. qui gen double D_B=B_time^b_federal[1,82]
. * Disadvantages of B
. qui gen double A_V=A_f1*D_B+(t_A-t_B) if (A_time>B_time)
. qui replace A_V=A_f1*D_A+b_f1*(D_B-D_A) if (B_time>=A_time)
. qui replace A_V=b_f1*D_B if (A_time==0)
. qui replace A_V=t_A if (B_time==0)
. * Disadvantages of A
. qui gen double B_V=B_f1*D_A+(t_B-t_A) if (B_time>A_time)
. qui replace B_V=B_f1*D_B+a_f1*(D_A-D_B) if (A_time>=B_time)
. qui replace B_V=a_f1*D_A if (B_time==0)
. qui replace B_V=t_B if (A_time==0)
. qui gen double P_f1=A_V/(A_V+B_V)
. replace P_f1=round(P_f1*1000)/1000
(3,200 real changes made)
. replace P_f1=0 if P_f1<0 /* If the prediction is less than 0 */
(0 real changes made)
. replace P_f1=1 if P_f1>1 /* If the prediction is greater than 1 */
(0 real changes made)
. rename P_f1 P_fedora
. save prediction, replace
file prediction.dta saved
. * Audit predictions and compute chi square
. use prediction, clear
. keep if analysis==1
(1,640 observations deleted)
. drop A_MO-b_AD4
. gen success=round(N*C,1)
. gen pv=0
. local i=1
. qui while `i'<=_N {
. replace pv=. if analysis==0
(0 real changes made)
. gen reject=(pv<0.01)*(analysis==1)
. tab reject
    reject |      Freq.      Percent        Cum.
-----+-----+
          0 |      1,495       95.83      95.83

```

```

      1 |       65      4.17     100.00
-----
      Total |    1,560     100.00
. twoway (scatter C P_fedora if reject==0) /*
> */      (scatter C P_fedora if reject==1), legend(off)
. * NOTE: It appears that 65 out of the 1560 predictions (4%; red) are rejected
. * at a p-value 0.01, which is acceptable for this competition. There is
. * certainly room for improvement; however, this would likely require
. * additional parameters beyond theta, alpha and beta.
. * Construct empirical weights for Weighted Least Squares Estimation
. gen double w=(1/(C*(1-C)))                      /* Uncertainty Weight */
(2 missing values generated)
. replace w=(4*(N^2)/(2*N-1)) if w==.             /* Berkson Weight if w==. */
(2 real changes made)
. gen fedora_sq=N*(C-P_fedora)^2*w
. tabstat fedora_sq, s(sum)
      variable |      sum
-----
      fedora_sq |   3567.711
-----
. * NOTE: The Bradley-Terry model (shown in the Exploratory Analysis code)
. * achieved a chi square of 8243, which was much lower than its closest
. * competitor, logit WLS (21521). By adding just 2 parameters, Fedora
. * reduces the chi square to 3568. Additional parameters may improve its
. * fit further; however, these predictions seem sufficient for my entry.
. * All are welcome to build from this code with proper citation.
. * Export Prediction File for Fedora
. use prediction, clear
. export delimited p_id T_A T_B H_A H_B P_fedora /*
> */ using "C:\data\crhe16\csv\Fedora_Predictions.csv", replace
file C:\data\crhe16\csv\Fedora_Predictions.csv saved

```