# Summary Statistics Page 1

Sunday, December 6, 2020 4:39 PM

### **Continuous Variables**

. summ x1 x3 x4 x5 x7 x11 x13

Variable	0bs	Mean	Std. Dev.	Min	Max
x1	151	11.66225	3.257581	6	21
х3	151	493.5762	186.7685	120	985
x4	151	8.900662	3.526575	2	23
x5	151	48.09934	11.19866	29	90
x7	151	.1986755	.5035854	0	3
x11	151	.2980132	.7098798	0	4
x13	151	81.90728	3.988903	65	86

### **Categorical Variables**

. tab x2

x2	Freq.	Percent	Cum.
0	136	90.07	90.07
1	15	9.93	100.00
Total . tab x6	151	100.00	
<b>x</b> 6	Freq.	Percent	Cum.
0	44	29.14	29.14
1	107	70.86	100.00
Total . tab x8	151	100.00	
x8	Freq.	Percent	Cum.
1	11	7.28	7.28
2	63	41.72	49.01
3	54	35.76	84.77
4	10	6.62	91.39
5	13	8.61	100.00
Total	151	100.00	

Variable Name	Туре
x1	Continuous
x2	Categorical
x3	Continuous
x4	Continuous
x5	Continuous
x6	Categorical
x7	Continuous
x8	Categorical
<b>x</b> 9	Categorical
x10	Categorical
x11	Continuous
x12	Categorical
x13	Continuous
x14	Categorical

# Summary Statistics Page 2

Sunday, December 6, 2020 10:31 PM

# Categorical Variables Cont.

. tab x9

x9	Freq.	Percent	Cum.
0	66	43.71	43.71
1	85	56.29	100.00
Total . tab x10	151	100.00	
x10	Freq.	Percent	Cum.
0	64	42.38	42.38
1	87	57.62	100.00
Total . tab x12	151	100.00	
x12	Freq.	Percent	Cum.
1	55	36.42	36.42
2	52	34.44	70.86
3	28	18.54	89.40
4	8	5.30	94.70
5	8	5.30	100.00
Total . tab x14	151	100.00	
x14	Freq.	Percent	Cum.
0	75	49.67	49.67
1	76	50.33	100.00
Total	151	100.00	

# Logistic Regression w/ Odds Ratio Page 1

Sunday, December 6, 2020 7:44 PM

#### Model 1

. logit x2 x1 x3 x4 x7 x9 x10 x11 i.x12

note: 4.x12 != 0 predicts failure perfectly
4.x12 dropped and 8 obs not used

Iteration 0: log likelihood = -48.006155 log likelihood = -28.207472 Iteration 1: log likelihood = -9.5303157 Iteration 2: log likelihood = -5.6013261 Iteration 3: log likelihood = -4.161876 Iteration 4: log likelihood = -4.1183138 Iteration 5: log likelihood = -4.1157896 Iteration 6: Iteration 7: log likelihood = -4.1156154 Iteration 8: log likelihood = -4.1156142 Iteration 9: log likelihood = -4.1156142

Logistic regression Number of obs = 143  $LR \ chi2(10) = 87.78$  Prob > chi2 = 0.0000 Log likelihood = -4.1156142 Pseudo R2 = 0.9143

x2	Coef.	Std. Err.	z	P> z	[95% Conf.	Interval]
x1	.513863	.3922332	1.31	0.190	2548999	1.282626
x3	060554	.0313958	-1.93	0.054	1220885	.0009806
x4	-2.637786	1.689333	-1.56	0.118	-5.948819	.6732465
x7	-3.194342	45.99427	-0.07	0.945	-93.34146	86.95278
x9	-1.695295	3.299289	-0.51	0.607	-8.161782	4.771193
x10	3.244547	4.537504	0.72	0.475	-5.648798	12.13789
x11	-1.099486	11.69203	-0.09	0.925	-24.01545	21.81648
x12						
2	. 8463053	4.075242	0.21	0.835	-7.141022	8.833633
3	-1.140228	3.94459	-0.29	0.773	-8.871483	6.591027
4	0	(empty)				
5	9.037978	61.78213	0.15	0.884	-112.0528	130.1287
_cons	28.79344	19.34202	1.49	0.137	-9.116231	66.70311

Note: 60 failures and 2 successes completely determined.

. logit, or

note: 4.x12 != 0 predicts failure perfectly
4.x12 dropped and 8 obs not used

Logistic regression Number of obs = 143 LR chi2(10) = 87.78 Prob > chi2 = 0.0000 Log likelihood = -4.1156142 Pseudo R2 = 0.9143

x2	Odds Ratio	Std. Err.	z	P> z	[95% Conf.	Interval]
x1	1.671737	.6557105	1.31	0.190	.7749941	3.606096
x3	.941243	.029551	-1.93	0.054	.88507	1.000981
x4	.0715194	.1208202	-1.56	0.118	.0026089	1.960592
x7	. 0409935	1.885466	-0.07	0.945	2.90e-41	5.80e+37
x9	.1835451	.6055683	-0.51	0.607	.0002854	118.06
x10	25.6501	116.3874	0.72	0.475	.0035217	186818.4
x11	.3330421	3.89394	-0.09	0.925	3.72e-11	2.98e+09
x12						
2	2.331019	9.499464	0.21	0.835	.0007919	6861.165
3	.3197461	1.261267	-0.29	0.773	.0001403	728.5286
4	1	(empty)				
5	8416.738	520004	0.15	0.884	2.17e-49	3.27e+56
_cons	3.20e+12	6.18e+13	1.49	0.137	.0001099	9.31e+28

Note: \_cons estimates baseline odds.

Note:  $\overline{\text{60}}$  failures and 2 successes completely determined.

#### Model 2

. logit x2 x1 x3 x4 x7 x9 x10 x11

Iteration 0: log likelihood = -48.867438
Iteration 1: log likelihood = -29.001628
Iteration 2: log likelihood = -7.6920668
Iteration 3: log likelihood = -5.0095908
Iteration 4: log likelihood = -4.2762379
Iteration 5: log likelihood = -4.2413143
Iteration 7: log likelihood = -4.2413144

Logistic regression Number of obs = 151  $LR \ chi2(7) = 89.25$  Prob > chi2 = 0.0000 Log likelihood = -4.241314 Pseudo R2 = 0.9132

x2	Coef.	Std. Err.	z	P> z	[95% Conf.	Interval]
x1	.5918773	.3527541	1.68	0.093	099508	1.283263
х3	0586078	.0262244	-2.23	0.025	1100068	0072089
x4	-2.914245	1.48098	-1.97	0.049	-5.816912	0115773
x7	1.16314	6.854251	0.17	0.865	-12.27094	14.59722
x9	-1.265774	2.59441	-0.49	0.626	-6.350724	3.819176
x10	2.831577	3.649284	0.78	0.438	-4.320888	9.984042
x11	-1.166463	17.65575	-0.07	0.947	-35.77109	33.43817
_cons	29.1867	16.55568	1.76	0.078	-3.261837	61.63524

Note: 59 failures and 0 successes completely determined.

. logit, or

Logistic regression Number of obs = 151  $LR \ chi2(7) = 89.25$  Prob > chi2 = 0.0000 Log likelihood = -4.241314 Pseudo R2 = 0.9132

x2	Odds Ratio	Std. Err.	z	P> z	[95% Conf.	Interval]
x1	1.807378	.63756	1.68	0.093	.9052828	3.608393
x3	.9430765	.0247317	-2.23	0.025	.8958281	.9928171
x4	.054245	.0803357	-1.97	0.049	.0029768	.9884894
x7	3.199967	21.93337	0.17	0.865	4.69e-06	2185215
x9	.282021	.7316781	-0.49	0.626	.0017455	45.56666
x10	16.97221	61.9364	0.78	0.438	.0132881	21677.75
x11	.3114666	5.499176	-0.07	0.947	2.92e-16	3.33e+14
_cons	4.74e+12	7.84e+13	1.76	0.078	.038318	5.86e+26

Note: \_cons estimates baseline odds.

Note: 59 failures and 0 successes completely determined.

# Models Interpretations Page 1

Sunday, December 6, 2020 10:13 PM

#### Model #1

#### Coefficients

- x1 has a coefficient of 0.51
  - for every one minute that travel time increases, the driver is more likely to take the freeway instead of the rural road by 0.51
- x3 has a coefficient of -0.61
  - o for every one vehicle per hour increase, the driver is less likely to take the freeway to the rural road by -0.61
- x12.4 has no coefficient
  - this variable is collinear and therefore violates one of the assumptions and must be omitted for a valid model
- this model has a constant of 28.79
  - o given all other factors are zero, the driver is more likely to take the freeway over the rural road by 28.79

#### **Odds Ratios**

- x1 has an odds ratio of 1.67
  - for every one minute that the travel time increases, the driver is 1.67 times more likely to take the freeway over the rural road
- x3 has an odds ratio of 0.94
  - for every one vehicle per hour increase, the driver is 0.94 times more likely to take the freeway over the rural road. This ratio is less than one which means the chances of the driver taking the freeway is decreasing as the rate of cars per hour increases.
- x12.4 has an odds ratio of 1.00
  - o this variable is collinear and therefore, this variables is omitted from the final calculation
- this model has a constant of 3.20 \* 10<sup>12</sup>
  - If all other factors are ignored, the driver is 3.20 \* 10<sup>12</sup> times more likely to take the freeway over the rural road
  - o because this number is so high, not having any other information makes this model basically unusable

#### Model #2

#### Main Difference Relative to Model #1

This model excludes the x12 variable. By removing the x12 variable, variables x3 and x4 become more reliable to a realistic level with new P-Values under 0.05.

#### **Both Models**

- Neither model had more than two, if any variables under 0.05 which makes most of the data unreliable.
- If a logistic model only has x2 and one other variable, the model would work
- If all variables are included, the model breaks
- If any variable is added that isn't included in these two models, the model breaks
- by "breaks" I mean the model is too perfect and any data other than coefficients or odds ratios cannot be determined.

# Stata Code Page 1

Sunday, December 6, 2020

11:05 PM

```
import excel "C:\Users\Imm56\Documents\School\Poly 20 - 21\QMB 3200\Final Exam\Data_FinalTakeHome.xlsx",
sheet("Sheet1") firstrow
summ x1 x3 x4 x5 x7 x11 x13
summ x1 x3 x4 x5 x7 x11 x13
gen route1 = cond(x2 == 1, 1, 0)
gen route2 = cond(x2 == 2, 1, 0)
gen seatBeltYes = cond(x6 == 1, 1, 0)
gen seatBeltNo = cond(x6 == 0, 1, 0)
gen driverAge18 23 = cond(x8 == 1, 1, 0)
gen driverAge24_29 = cond(x8 == 2, 1, 0)
gen driverAge30_39 = cond(x8 == 3, 1, 0)
gen driverAge40 49 = \text{cond}(x8 == 4, 1, 0)
gen driverAge50_Up = cond(x8 == 5, 1, 0)
gen genderMale = cond(x9 == 1, 1, 0)
gen genderFemale = cond(x9 == 0, 1, 0)
gen marriedYes = cond(x10 == 0, 1, 0)
gen marriedNo = cond(x10 == 1, 1, 0)
gen annincomeLess_200k = cond(x12 == 1, 1, 0)
gen annIncome200k_299k = cond(x12 == 2, 1, 0)
gen annincome300k_399k = cond(x12 == 3, 1, 0)
gen annincome400k 499k = cond(x12 == 4, 1, 0)
gen annIncome500k_{p} = cond(x12 == 5, 1, 0)
gen carDomestic = cond(x14 == 1, 1, 0)
gen carForeign = cond(x14 == 0, 1, 0)
tab route1 route2 seatBeltYes seatBeltNo driverAge18 23 driverAge24 29 driverAge30 39 driverAge40 49 driverAge50
Up genderMale genderFemale marriedYes marriedNo annIncomeLess 200k annIncome200k 299k annIncome300k
399k annIncome400k 499k annIncome500k Up carDomestic carForeign
tab x12
tab x2
tab x6
tab x8
tab x9
tab x10
tab x14
logit route1 x1 x3 x4 x5 i.x6 x7 i.x8 i.x9 i.x10 x11 i.x12 x13 i.x14
logit route1 x1 x3 x4 x5 i.x6 x7 i.x8 i.x9 i.x10 x11 i.x12 x13 i.x14
logit route1 x1 x3 x4 x5 i.x6 x7 i.x8 i.x9 i.x10 x11 x13 i.x14
logit x2 x1 x3 x4 x5 i.x6 x7 i.x8 i.x9 i.x10 x11 i.x12 x13 i.x14
logit route1 x1 x3 x4 x5 i.x6 x7 i.x8 i.x9 i.x10 x11 i.x12 x13 i.x14
logit route2 x1 x3 x4 x5 i.x6 x7 i.x8 i.x9 i.x10 x11 i.x12 x13 i.x14
logit route1 x1 x3 x4 x5 x6 x7 i.x8 x9 x10 x11 i.x12 x13 x14
logit route1 x1 x3 x4 x5 x6 x7 i.x8 x9 x10 x11 i.x12 x13 x14
logit route1 x1 x3 x4 x5 x6 x7 i.x8 x9 x10 x11 x13 x14
logit route1 x1
logit route1 x1 x3
logit route1 x1 x3 x4
logit route1 x1 x3 x4 x5
```

# Stata Code Page 2

Sunday, December 6, 2020

11:08 PM

anova route1 x1 x3 x4 x5 x7 x11 x13 Ioneway route1 x1 x3 x4 x5 x7 x11 x13 logit route1 x1 x3 x4 x6 logit route1 x1 x3 x4 i.x6 logit route1 x1 x3 x4 logit route1 x1 x3 x4 x7 logit route1 x1 x3 x4 x7 i.x8 logit route1 x1 x3 x4 x7 x9 logit route1 x1 x3 x4 x7 x9 x10 logit route1 x1 x3 x4 x7 x9 x10 x11 logit route1 x1 x3 x4 x7 x9 x10 x11 x13 logit route1 x1 x3 x4 x7 x9 x10 x11 x14 logit route1 x1 x3 x4 x7 x9 x10 x11 logit, or logit route1 x8 logit route1 i.x8 logit route1 x3 x4 i.x8 replace x2 = x2-1logit x2 x1 x3 x4 x5 x7 i.x9 i.x10 x11 x12 logit x2 x3 x4 logit x2 x1 x3 x4 logit x2 x3 x4 x5 logit route1 x3 x4 logit route2 x1 x3 x4 x7 x9 x10 x11 logit route2 x3 x4 logit, or logit route1 x3 x4 logit, or logit x2 x1 logit x2 x3 logit x2 x4 logit x2 x5 logit x2 x6 logit x2 i.x6 logit x2 x7 logit x2 x8 logit x2 i.x8 logit x2 x9 logit x2 x10 logit x2 x11 logit x2 i.x12 tab x12 logit x2 x13 logit x2 x14 logit x2 x1 x3 x4 x6 x7 x8 x9 x10 x11 x12 x13 x14

logit route1 x1 x3 x4 x6 x7 x8 x9 x10 x11 x12 x13 x14 logit x2 x1 x3 x4 x6 x7 i.x8 x9 x10 x11 i.x12 x13 x14

logit x2 x1 x3 x4 x7 i.x9 i.x10 x11 x12

# Stata Code Page 3

11:09 PM

Sunday, December 6, 2020

logit x2 x1 x3 x4 x7 x9 i.x10 x11 x12 logit x2 x1 x3 x4 x7 x9 x10 x11 i.x12 logit, or logit x2 x3 x4 logit, or logit x2 x1 x3 x4 x5 x6 x7 i.x8 x9 x10 x11 i.x12 x13 x14 logit x2 x1 x3 x4 x6 x7 i.x8 x9 x10 x11 i.x12 x13 x14 logit x2 x1 x3 x4 x6 x7 x9 x10 x11 i.x12 x13 x14 logit x2 x1 x3 x4 x6 x7 x9 x10 x11 i.x12 x13 logit x2 x1 x3 x4 x6 x7 x9 x10 x11 i.x12 logit x2 x1 x3 x4 x7 x9 x10 x11 i.x12 logit x2 x1 x3 x4 x7 x9 x10 x11 anova x2 x1 x3 x4 x5 x6 x7 x8 x9 x10 x11 x12 x13 x14 logit x2 x1 x3 x4 x7 x9 x10 x11 logit x2 x3 x4 logit x2 x1 x3 x4 logit x2 x3 x4 i.x8 logit x2 x1 x3 x4 x7 x9 x10 x11 i.x12 logit, or logit x2 x3 x4 logit, or summ x1 x3 x4 x5 x7 x11 x13 tab x2 tab x6 tab x8 tab x9 tab x10 tab x12 tab x14 logit x2 x1 x3 x4 x7 x9 x10 x11 i.x12 logit, or logit route x1 x3 x4 x7 x9 x10 x11 i.x12 logit route2 x1 x3 x4 x7 x9 x10 x11 i.x12 logit, or logit x2 x1 x3 x4 x7 x9 x10 x11

log close clear

logit, or

My stata code is three pages because I was basically brute forcing the models until I found one that worked :( p.s. check out the times when all of these pages were made.