

## Handout: Model Building

How many predictors?

Guideline: You need at least 10 of each outcome for every predictor.

<b>y</b>					
		Frequency	Percent	Valid Percent	Cumulative Percent
Valid	.00	292	48.7	48.7	48.7
	1.00	308	51.3	51.3	100.0
Total		600	100.0	100.0	

For the HSB model we could use up to 29 variables according to this guideline, we are lucky because the data is balanced.

### Example 1

HSB

The variable  $X$  was calculated as the total of science writing, civic and math, therefore

$x = \text{math} + \text{science} + \text{writing} + \text{civic}$

The four variables are collinear and including them all in one model would make the parameter estimates undefined

### Example 2

HSB with the following predictors:

SES

school type

gender

$x$ , academic success

reading

writing

math

science

civics

Categorical Variables Codings				
		Frequency	Parameter coding	
			(1)	(2)
ses	1.00	139	1.000	.000
	2.00	299	.000	1.000
	3.00	162	.000	.000
gender	.00	327	.000	
	1.00	273	1.000	
school	.00	506	.000	
	1.00	94	1.000	

Omnibus Tests of Model Coefficients				
		Chi-square	df	Sig.
Step 1	Step	139.082	1	.000
	Block	139.082	1	.000
	Model	139.082	1	.000
Step 2	Step	28.161	1	.000
	Block	167.243	2	.000
	Model	167.243	2	.000
Step 3	Step	13.506	1	.000
	Block	180.749	3	.000
	Model	180.749	3	.000
Step 4	Step	12.806	2	.002
	Block	193.555	5	.000
	Model	193.555	5	.000

Variables in the Equation

		B	S.E.	Wald	df	Sig.	Exp(B)	95.0% C.I. for EXP(B)	
								Lower	Upper
Step 1	x	.027	.003	106.411	1	.000	1.028	1.022	1.033
	Constant	-7.055	.695	103.104	1	.000	.001		
Step 2	school(1)	1.421	.288	24.337	1	.000	4.143	2.355	7.287
	x	.027	.003	99.251	1	.000	1.027	1.022	1.033
Step 3	Constant	-7.160	.714	100.606	1	.000	.001		
	school(1)	1.449	.291	24.699	1	.000	4.257	2.404	7.537
	x	.040	.005	72.679	1	.000	1.041	1.031	1.051
	science	-.063	.018	12.879	1	.000	.939	.907	.972
	Constant	-7.308	.725	101.481	1	.000	.001		
Step 4	ses			12.506	2	.002			
	ses(1)	-.926	.289	10.256	1	.001	.396	.225	.698
	ses(2)	-.731	.238	9.413	1	.002	.481	.302	.768
	school(1)	1.374	.295	21.738	1	.000	3.952	2.218	7.043
	x	.039	.005	65.371	1	.000	1.039	1.030	1.049
	science	-.065	.018	13.502	1	.000	.937	.905	.970
Constant		-6.193	.786	62.014	1	.000	.002		

- a. Variable(s) entered on step 1: x.  
b. Variable(s) entered on step 2: school.  
c. Variable(s) entered on step 3: science.  
d. Variable(s) entered on step 4: ses.

**Model Summary**

Step	-2 Log likelihood	Cox & Snell R Square	Nagelkerke R Square
1	692.268 <sup>a</sup>	.207	.276
2	664.107 <sup>a</sup>	.243	.324
3	650.601 <sup>b</sup>	.260	.347
4	637.795 <sup>b</sup>	.276	.368

a. Estimation terminated at iteration number 4 because parameter estimates changed by less than .001.

b. Estimation terminated at iteration number 5 because parameter estimates changed by less than .001.

**Example 3**

HSB, compare models

$M_1$  with predictors school type,  $x$ , and science and

$M_2$  with predictors school type,  $x$ , science and SES.

Model	AIC
$M_1$	658.601
$M_2$	649.795

Therefore model  $M_2$  gives a better fit than model  $M_1$ , even when correcting for the number for parameters in the model.

**Example 4**

HSB for "best" model

**Classification Table<sup>a</sup>**

Observed		Predicted		Percentage Correct
		y		
		.00	1.00	
Step 1	y .00	208	84	71.2
	1.00	74	234	76.0
Overall Percentage				73.7

a. The cut value is .500

**Example 5**

HSB for "best" model

### Correlations

		y	Predicted probability
y	Pearson Correlation	1	.533**
	Sig. (2-tailed)		.000
	N	600	600
Predicted probability	Pearson Correlation	.533**	1
	Sig. (2-tailed)	.000	
	N	600	600

\*\* . Correlation is significant at the 0.01 level (2-tailed).

$$R^2=0.284.$$

### Model Summary

Step	-2 Log likelihood	Cox & Snell R Square	Nagelkerke R Square
1	639.691 <sup>a</sup>	.273	.365

a. Estimation terminated at iteration number 5 because parameter estimates changed by less than .001.