Analytic Strategies: Simultaneous, Hierarchical, and Stepwise Regression

This discussion borrows heavily from <u>Applied Multiple Regression/Correlation Analysis</u> for the Behavioral Sciences, by Jacob and Patricia Cohen (1975 edition).

The simultaneous model. In the simultaneous model, all K IVs are treated simultaneously and on an equal footing. Such a research strategy is clearly most appropriate when we have no logical or theoretical basis for considering any variable to be prior to any other, either in terms of a hypothetical causal structure of the data or in terms of its relevance to the research goals.

The hierarchical model. An alternative strategy to the simultaneous model is one in which the K IVs are entered cumulatively according to some specified hierarchy which is dictated in advance by the purpose and logic of the research. The hierarchical model calls for a determination of R^2 and the partial coefficients of each variable at the point at which it is added to the equation. Note that with the addition of the ith IV, the MRC analysis *at that stage* is simultaneous in i variables.

Perhaps the most straightforward use of the hierarchical model is when IVs can be ordered with regard to their temporally or logically determined priority. Thus, for example, when two IVs are sex (male or female) and an attitudinal variable, sex must be considered the causally prior variable since it antedates attitude.

The hierarchical MRC analysis may proceed by entering the IVs in the specified order and determining R² after each additions. Suppose, for example, when we regress Y on X1 R²_{Y1} = .10. Then, when we regress Y on X1 and X2, R²_{Y12} = .15. We can say that, after controlling for X1, X2 accounts for 5% of the variance in Y, i.e. $sr^2_{2\cdot 1} = R^2_{Y12} - R^2_{Y1} = .15 - .10 = .05$.

Note that the R^2 contribution of any variable depends upon what else is in the equation.

Hierarchical analysis of the variables typically adds to the researcher's understanding of the phenomena being studied, since it requires thoughtful input by the researcher in determining the order of entry of IVs, and yields successive tests of the validity of the hypotheses which determine that order.

Stepwise regression. *Forward* stepwise regression programs are designed to select from a group of IVs the one variable at each stage which has the largest sr^2 , and hence makes the largest contribution to R^2 . (This will also be the variable that has the largest T value.) Such programs typically stop admitting IVs into the equation when no IV makes a contribution which is statistically significant at a level specified by the user. Thus, the stepwise procedure defines an a posteriori order based solely on the relative uniqueness of the variables in the sample at hand.

Backwards stepwise regression procedures work in the opposite order. The dependent variable is regressed on all K independent variables. If any variables are statistically insignificant, the one making the smallest contribution is dropped (i.e. the variable with the smallest sr^2 , which will also be the variable with the smallest T value). Then the K - 1 remaining variables are

regressed on Y, and again the one making the smallest contribution is dropped. The procedure continues until all remaining variables are statistically significant.

Note that forwards and backwards regression need not produce the same final model.

Problems with stepwise regression.

✓ When an investigator has a large pool of potential IVs and very little theory to guide selection among them, stepwise regression is a sore temptation. If the computer selects the variables, the investigator is relieved of the responsibility of making decisions about their logical or causal priority or relevance before the analysis. However, this atheoritical approach tends not to be viewed kindly. Most social scientists believe that more orderly advances are made when researchers armed with theories provide a priori hierarchical ordering which reflects causal hypotheses rather than when computers order IVs post and ad hoc for a given sample.

✓ Another serious problem arises when a relatively large number of IVs is used. Since the significance test of an IV's contribution to R^2 proceeds in ignorance of the large number of other such tests being performed at the same time for the other competing IVs, there can be very serious capitalization on chance. The result is that neither the statistical significance tests for each variable nor the overall tests on the multiple R^2 at each step are valid. (For example, if you have 20 IVs, by chance alone one of them is likely to be significant at the .05 level.)

✓ A related problem is that the ad hoc order produced from a set of IVs in one sample is likely not to be found in other samples from the same population. When there are trivial differences between variables, the computer will dutifully choose the largest for addition at each step. In other samples and, more important, in the population, such differences may well be reversed. When the competing IVs are substantially correlated with each other, the problem is likely to be compounded, since the losers in the competition may not make a sufficiently large unique contribution to be entered at any subsequent step before the problem is terminated by "nonsignificance."

 \checkmark Sometimes with a large number of IVs, variables that were entered into the equation early no longer have nontrivial (i.e. significant) relationships after other variables have been added. However, many programs (e.g. SPSS) allow for the removal of such variables.

✓ Although it is not a common phenomenon, it is possible that neither of two variables alone would reach the criterion for acceptance into the equation, yet if both were entered they would make a useful contribution to R^2 . This can occur when *suppression* is present, e.g.



Conditions under which concerns are minimized

 \checkmark The research goal is entirely or primarily predictive (technological), and not at all, or only secondarily, explanatory (scientific). (Still, even research which is only concerned with prediction will likely do better with a theoretically guided model.)

 \checkmark N is very large, and the original K (that is, before forward stepwise selection) is not too great; a K/N ratio of at least 1 to 40 is prudent.

 \checkmark Particularly if the results are to be substantively interpreted, a cross-validation of the stepwise analysis in a new sample should be undertaken, and only those conclusions that hold for both samples should be drawn. Alternatively, the original sample may be randomly divided in half, and the two half-samples treated in this manner.

 \checkmark Stepwise and hierarchical regression can be combined. An investigator may be clear that some groups of variables are logically, causally, or structurally prior to others, and yet have no basis of ordering variables within such groups. Hence, the researcher can decide what order the groups of variables should be entered in, and then let the computer decide within each group what the sequencing should be. This type of analysis is likely to be primarily hierarchical (between classes of IVs) and only incidentally stepwise (within classes).

Stepwise regression example

In this section, I will show how stepwise regression could be used with the <u>Education</u>, <u>Occupation and</u> <u>Earnings</u> example from Sewell and Hauser (1975).

As you look through the handout, make sure you can confirm the different claims that are made. In particular, for each model, check whether the predictions for the next variable to be entered or removed are correct; also check to see whether the R^2 is as predicted.

NOTE!!! SPSS's old style of formatting output is better for purposes of my presentation, ergo I am continuing to use it. Later on I give SPSS's current printout.

```
10-Dec-91 SPSS RELEASE 4.1 FOR IBM OS/MVS
00:49:38
            UNIVERSITY OF NOTRE DAME IBM 370/3081 U OS/MVS
   1 0 Set width = 125 .

    Title 'Stepwise Regression Example' .
    Adapted from Education, Occupation, and Earnings:

   4 0 * Achievement in the Early Career, by William H. Sewell
   5
     0 * and Robert M. Hauser. 1975, Academic Press.
   6 0 * See especially pages 72 and 79.
   7 0 MATRIX DATA VARIABLES=ROWTYPE_ V M X I Q U W Y.
There are 2,951,800 bytes of memory available.
The largest contiguous area has 2,944,880 bytes.
MATRIX DATA has already allocated 440 bytes.
More memory will be allocated to store the data to be read.
  20 0 END DATA.
Preceding task required .04 seconds CPU time; .59 seconds elapsed.
  21 0 VARIABLE LABELS
         V "Father's educational attainment"
  22 0
 23 0
24 0
          M "Mother's educational attainment"
X "Status of F's Occ when son grad from HS"
           I "Parent's average income, 1957-1960"
  25 0
  26 0
           Q "Son's score on Henmon-Nelson IQ test"
           U "Son's educational attainment"
  27 0
  28 0
           W "Son's 1964 Occupation (Duncan SEI)"
           Y "Son's average earnings, 1965-1967" .
  29 0
  30 0
 31 0 Subtitle "Stepwise forward selection" .
32 0 REGRESSION /MATRIX IN(*)
                     /DESCRIPTIVES DEF
  33 0
                     /VARIABLES V M X Q U W
  34
     0
                     /CRITERIA=POUT(.051)
  35 0
                    /STATISTICS DEF CHANGE OUTS ZPP TOL
  36 0
                     /DEPENDENT W
  37 0
                     /METHOD FORWARD V M X O U.
  38 0
  39
     0
```

There are 2,954,104 bytes of memory available. The largest contiguous area has 2,947,640 bytes.

Forward selection

10-Dec-91 Stepwise Regression Example 00:49:43 Stepwise forward selection * * * * MULTIPLE REGRESSION * * * * Listwise Deletion of Missing Data Mean Std Dev Label V 10.200 3.008 Father's educational attainment 10.410 2.848 Mother's educational attainment М 33.110 22.410 Status of F's Occ when son grad from HS 99.860 14.570 Son's score on Henmon-Nelson IQ test х Q U 13.220 1.676 Son's educational attainment 42.110 23.340 Son's 1964 Occupation (Duncan SEI) W N of Cases = 2069Correlation: Х U М W V Q .425 .316 .260 .291 .254 .211 .184 .263 V 1.000 .524 1.000 .524 .524 .211 М .288 1.000 .262 Х .425 .288 Q .254 .211 .184 1.000 .448 .377 .316 .260 .291 .448 1.000 U .621

.262

.211

W

.263

Note that U has the biggest correlation with W (.621). Ergo, U will be the first variable entered into the equation. R^2 will equal .621² = .3856.

.377

.621

1.000

Forward selection (Continued)

10-Dec-91 Stepwise Regression Example 00:49:44 Stepwise forward selection
* * * * MULTIPLE REGRESSION * * * *
Equation Number 1 Dependent Variable W Son's 1964 Occupation (Duncan SE
Block Number 1. Method: Forward Criterion PIN .0500 V M X Q U
Variable(s) Entered on Step Number 1 U Son's educational attainment
Multiple R .62100 R Square .38564 Adjusted R Square .38534 F Change 1297.48233 Standard Error 18.29856 Signif F Change .0000
Analysis of Variance DF Sum of Squares Mean Square Regression 1 434445.63509 434445.63509 Residual 2067 692108.94571 334.83742 F = 1297.48233 Signif F = .0000
Variables in the Equation
Variable B SE B Beta Correl Part Cor Partial Tolerance VIF T Sig T U 8.648055 .240086 .621000 .621000 .621000 1.000000 1.000 36.021 .0000 .0000 (Constant) -72.217286 3.199336 -22.573 .0000
Variables not in the Equation
VariableBeta InPartialToleranceVIFMin TolerTSig TV.074170.089779.9001441.111.9001444.097.0000M.053132.065455.9324001.073.9324002.982.0029X.088809.108401.9153191.093.9153194.956.0000Q.123599.140980.7992961.251.7992966.473.0000
(1) As "Variables not in the equation" shows, Q would have the biggest T value if it were added to the equation, which means that adding Q would produce the greatest increase in R^2 . Ergo, Q gets entered next.
(2) sr_Q is not reported but is easily computed. Using the information from the "Beta in" and "Tolerance" columns in the "Variables not in the equation" section, we get $sr_Q = b'_Q * sqrt(Tol_Q) = .123599 * sqrt(.799296) = .1105,$ $sr_Q^2 = .1105^2 = .01221$ so adding Q will cause R ² to increase from its current value of .38564
to .39785.

(3) Note that, in a bivariate regression, b' (BETA), the bivariate correlation (CORR), the semipartial correlation (PART CORR), and the partial correlation (PARTIAL) are all identical.

Forward selection (continued)

10-Dec-91 00:49:44	Stepwise R Stepwise f	egression orward se	Example								
			* * * *	MULT	IPI	LERE	GRESS	SION *	* * *		
Equation Num	nber 1 D	ependent	Variable	W So	n's 19	964 Occupa	ation (Du	ncan SE			
Variable(s)	Entered on	Step Num	ber 2	Q	Sor	n's score	on Henmoi	n-Nelson IQ	test		
Multiple R R Square Adjusted R S Standard Err	Square . For 18.	63075 39785 39727 12019	R So F Ch Sign	uare Chan ange if F Chan	ge 41 ge	.01221 .89505 .0000					
Analysis of Regression Residual F = 682.	Variance DF 2066 52386	Sum 4 6 Signif	of Square 48201.5079 78353.0728 F = .0000	s Me 6 224 4 - Variabl	an Squ 100.75 328.34 es in	tare 398 127 the Equat	ion				
Variable U Q (Constant)	7.87694 .19799 -81.79498	B 0 .26 6 .03 5 3.49	SE B 5925 .5 0590 .1 6676	Beta C 65628 .6 23599 .3	orrel 21000 77000	Part Cor .505690 .110501	Partial .545976 .140980	Tolerance .799296 .799296	VIF 1.251 1.251	T 29.621 6.473 -23.392	Sig T .0000 .0000 .0000
		Variab	les not in	the Equa	tion -						
Variable V M X	Beta In .059783 .041095 .081889	Partial .072449 .050830 .100764	Tolerance .884329 .921223 .911720	VI 1.13 1.08 1.09	F Mir 1 . 6 . 7 .	Toler 755588 770639 754271	T 3.301 2.313 4.602	Sig T .0010 .0208 .0000			

- (1) As "Variables not in the equation" shows, X would have the largest T value if added to the equation, i.e. it would produce the greatest increase in R^2 . Ergo, X gets entered next.
- (2) sr_x is not reported but is easily computed. Using the information from the "Beta in" and "Tolerance" columns in the "Variables not in the equation" section, we get $sr_x = b'_x * sqrt(Tol_x) = .081889 * sqrt(.911720) = .07182,$ $sr_0^2 = .07182^2 = .00612$ so adding X will cause R² to increase from its current value of .39785 to .40397.

Forward selection (continued)

10-Dec-91 00:49:44	Stepwise Reg Stepwise for	ression Exampl ward selection	2							
		* * *	* M U	LTIP	LE RE	GRESS	SION *	* * *		
Equation Nur	mber 1 Depo	endent Variabl	e W	Son's 1	964 Occupa	tion (Dur	ican SE			
Variable(s)	Entered on S	tep Number 3.	. x	St	atus of F'	s Occ whe	en son grad i	from HS		
Multiple R R Square Adjusted R S Standard Ern	.63 .40 Square .40 ror 18.03	558 397 R 310 F 233 S	Square Change ignif F	Change 2 Change	.00611 1.18189 .0000					
Analysis of Regression Residual F = 466	Variance DF 3 2065 .52140	Sum of Squ 455089.1 671465.4 Signif F = .0	ares 1434 5646 000	Mean Sq 151696.3 325.1 iables in	uare 7145 6487 the Equat	ion				
Variable U Q X (Constant)	B 7.579367 .189193 .085288 -79.805965	SE B .272420 .030501 .018531 3.506456	Beta .544260 .118104 .081889	Correl .621000 .377000 .262000	Part Cor .472683 .105381 .078191	Partial .522162 .135244 .100764	Tolerance .754271 .796153 .911720	VIF 1.326 1.256 1.097	T 27.822 6.203 4.602 -22.760	Sig T .0000 .0000 .0000 .0000
Variable V M	Beta In Pa: .034126 .0 .023996 .0	Variables not rtial Toleran 38740 .7681 29068 .8746	in the 2 20 79	Equation VIF Mi 1.302 1.143	n Toler .735815 .740276	T 1.761 1.321	Sig T .0783 .1866			

End Block Number 1 PIN = .050 Limits reached.

As "Variables not in the equation" shows, V would have the greatest T value (and hence produce the greatest increase in \mathbb{R}^2) if it were added to the equation. HOWEVER, the T value for V is not significant at the .05 level, so we do not want to include it. Ergo, STOP NOW!

Backwards Selection

10-Dec-91 Stepwise Regression Example 00:49:44 Stepwise forward selection
Preceding task required .14 seconds CPU time; 3.16 seconds elapsed.
40 0 Subtitle "Stepwise backward selection".
41 0 REGRESSION /MATRIX IN(*)
42 0 /VARIABLES V M X Q U W
43 0 /CRITERIA=POUT(.051)
44 0 /STATISTICS DEF CHANGE OUTS ZPP TOL
45 0 / DEPENDENT W
46 0 /METHOD ENTER V M X Q U
47 0 /METHOD BACKWARDS .
48 0

There are 2,954,096 bytes of memory available. The largest contiguous area has 2,947,640 bytes.

> 1956 bytes of memory required for REGRESSION procedure. O more bytes may be needed for Residuals plots.

Backwards selection (continued)

10-Dec-91 00:49:47	Stepwise Regre Stepwise backy	ession Examp ward selecti	le on							
		* *	** M U	LTIP	LE RE	GRESS	ION *	* * *		
Listwise Dei	letion of Miss	ing Data								
Equation Num Block Number	nber 1 Deper r 1. Method:	ndent Variab Enter	le W V	Son's 1 M	964 Occupa X	tion (Dun Q	can SE U			
Variable(s)	Entered on Ste	ep Number 1 2 3 4 5	U M X Q V	So Mo St So Fa	n's educat ther's edu atus of F' n's score ther's edu	ional att cational s Occ whe on Henmon cational	ainment attainment n son grad -Nelson IQ attainment	from HS test		
Multiple R R Square Adjusted R S Standard Err	.6363 .4049 Square .4039 For 18.0258	37 97 :: 53 35	R Square C F Change Signif F C	Change 28 Change	.40497 0.81132 .0000					
Analysis of Regression Residual F = 280	Variance DF 2063 .81132 S:	Sum of Sq 456221. 670332. ignif F = .	uares 67830 90250 0000	Mean Sq 91244.3 324.9	uare 3566 3112					
			Vari	ables in	the Equat	ion				
Variable J M X 2 V (Constant)	B 7.490510 .102339 .071806 .181830 .220009 -80.759043	SE B .276501 .165107 .019918 .030746 .166816 3.560080	Beta .537879 .012488 .068945 .113508 .028354	Correl .621000 .211000 .262000 .377000 .263000	Part Cor .460081 .010527 .061227 .100438 .022399	Partial .512245 .013645 .079124 .129116 .029025	Tolerance .731644 .710609 .788648 .782980 .624039	VIF 1.367 1.407 1.268 1.277 1.602	T 27.090 .620 3.605 5.914 1.319 -22.685	Sig T .0000 .5354 .0003 .0000 .1874 .0000

End Block Number 1 All requested variables entered.

- (1) We begin by estimating the model in which all independent variables are included. Note that some are not statistically significant. M has the smallest semipartial correlation, which means that dropping it would produce the smallest decrease in R². So, the next step is to remove M from the equation.
- (2) Let H = the set of all variables currently in the equation. Let G_k = the set of all variables in the equation except the variable that is to be dropped. Then,

 $R_{YGm}^2 = R_{YH}^2 - sr_m^2 = .40497 - .010527^2 = .40486.$ Ergo, after dropping M, R² will equal .40486.

Backwards selection (continued)

10-Dec-91 00:49:47	Stepwise Regre	ession Exampl ard selection	le on							
		* * *	** M U	LTIPL	ERE	GRESS	SION *	* * *		
Equation Nur Block Number	mber 1 Depen r 2. Method:	dent Variabl Backward	le W Criteri	Son's 19 on POUT	64 Occupa .0510	tion (Dur	ncan SE			
Variable(s)	Removed on Ste	ep Number 6.	М	Mot	her's edu	cational	attainment			
Multiple R R Square Adjusted R S Standard Err Analysis of Regression Residual F = 351	.6362 .4048 Square .4037 ror 18.0231 Variance DF 4 2064 .02283 Si	9 6 F 1 F 6 S Sum of Squ 456096.8 670457.7 gnif F = .0	R Square C F Change Signif F C Mares 34110 73970 0000	hange - hange Mean Squ 114024.21 324.83	.00011 .38420 .5354 are 028 418					
		-	Vari	ables in	the Fault	ion				
Variable U X Q V (Constant)	B 7.503414 .072599 .182813 .264791 -80.445490	SE B .275675 .019874 .030700 .150336 3.523431	Beta .538806 .069706 .114121 .034126	Correl .621000 .262000 .377000 .263000	Part Cor .462185 .062031 .101116 .029909	Partial .513934 .080149 .129961 .038740	Tolerance .735815 .791912 .785070 .768120	VIF 1.359 1.263 1.274 1.302	T 27.218 3.653 5.955 1.761 -22.832	Sig T .0000 .0003 .0000 .0783 .0000
	V	Variables not	t in the E	quation -						
Variable M	Beta In Part .012488 .013	ial Tolerar 645 .7106	nce 509 1	VIF Min .407 .	Toler 624039	T .620	Sig T .5354			

- (1) As the "variables in the equation" shows, V has the smallest semipartial correlation, which means that dropping it would produce the smallest decrease in R^2 . Since the effect of V is not statistically significant, remove it next.
- (2) Let H = the set of all variables currently in the equation. Let G_k = the set of all variables in the equation except the variable that is to be dropped. Then, $R_{YGv}^2 = R_{YH}^2 - sr_v^2 = .40486 - .029909^2 = .40397.$

Ergo, after dropping V, R^2 will equal .40397.

Backwards selection (continued)

10-Dec-91 00:49:47	Stepwise Reg Stepwise bac	ression Exampl kward selectio	e n							
		* * *	* M U	LTIPI	LE RE	GRESS	ION *	* * *		
Equation Nur	mber 1 Dep	endent Variabl	e W	Son's 19	964 Occupa	tion (Dun	can SE			
Variable(s)	Removed on S	tep Number 7.	. v	Fat	cher's edu	cational	attainment			
Multiple R R Square Adjusted R S Standard Err	.63 .40 Square .40 ror 18.03	558 397 R 310 F 233 S	Square (Change ignif F (Change Change	00089 3.10228 .0783					
Analysis of Regression Residual F = 466	Variance DF 3 2065 .52140	Sum of Squ 455089.1 671465.4 Signif F = .0	ares 1434 6646 000	Mean Sq 151696.3 325.1(the Equat	ion				
Variable U X Q (Constant)	B 7.579367 .085288 .189193 -79.805965	SE B .272420 .018531 .030501 3.506456	Beta .544260 .081889 .118104	Correl .621000 .262000 .377000	Part Cor .472683 .078191 .105381	Partial .522162 .100764 .135244	Tolerance .754271 .911720 .796153	VIF 1.326 1.097 1.256	T 27.822 4.602 6.203 -22.760	Sig T .0000 .0000 .0000 .0000
		Variables not	in the H	Equation ·						
Variable V M	Beta In Pa .034126 .0 .023996 .0	rtial Toleran 38740 .7681 29068 .8746	ce 20 1 79 1	VIF Min L.302 L.143	n Toler 735815 740276	T 1.761 1.321	Sig T .0783 .1866			

End Block Number 2 POUT = .051 Limits reached.

As the "Variables in the equation" shows, X has the smallest semipartial correlation; HOWEVER, the effect of X is significant at the .05 level, so we do not want to exclude it. Ergo, STOP NOW!

MODERN SPSS OUTPUT. SPSS has changed its formatting in recent years. The new style is prettier but makes it harder to do the sort of step by step discussion given above. Here is what modern output looks like. You should have little trouble seeing how the information in both sets of printouts is the same.

Stepwise Regression Example Adapted from Education, Occupation, and Earnings: * Achievement in the Early Career, by William H. Sewell * and Robert M. Hauser. 1975, Academic Press. * See especially pages 72 and 79. MATRIX DATA VARIABLES=ROWTYPE_ V M X I Q U W Y. BEGIN DATA. MEAN10.20010.41033.1106.44399.86013.22042.110STDDEV3.0082.84822.4103.14114.5701.67623.340 7.538 1.676 23.340 2.589 2069 2069 Ν 2069 2069 2069 2069 2069 2069 CORR 1.000 CORR 0.524 1.000 1.000 CORR 0.425 0.288 CORR 0.320 0.238 0.457 1.000 CORR 0.254 0.211 0.184 0.184 1.000 0.448 0.279 CORR 0.316 0.260 0.291 1.000 0.621 1.000 CORR 0.263 0.211 0.262 0.237 0.377 CORR 0.093 0.083 0.102 0.184 0.162 0.203 0.220 1.000 END DATA. VARIABLE LABELS V "Father's educational attainment" M "Mother's educational attainment" X "Status of F's Occ when son grad from HS" I "Parent's average income, 1957-1960" Q "Son's score on Henmon-Nelson IQ test" U "Son's educational attainment" W "Son's 1964 Occupation (Duncan SEI)" Y "Son's average earnings, 1965-1967" . "Stepwise forward selection" . Subtitle Stepwise Regression Example Stepwise forward selection REGRESSION /MATRIX IN(*) /DESCRIPTIVES DEF /VARIABLES V M X Q U W /CRITERIA=POUT(.051) /STATISTICS DEF CHANGE OUTS ZPP CI SES TOL /DEPENDENT W /METHOD FORWARD V M X Q U.

Regression

Descriptive Statistics

	Mean	Std. Deviation	Ν
V Father's educational attainment	10.200000	3.0080000	2069
M Mother's educational attainment	10.410000	2.8480000	2069
X Status of F's Occ when son grad from HS	33.110000	22.4100000	2069
Q Son's score on Henmon-Nelson IQ test	99.860000	14.5700000	2069
U Son's educational attainment	13.220000	1.6760000	2069
W Son's 1964 Occupation (Duncan SEI)	42.110000	23.3400000	2069

Correlations

		V Father's educational attainment	M Mother's educational attainment	X Status of F's Occ when son grad from HS	Q Son's score on Henmon-Nel son IQ test	U Son's educational attainment	W Son's 1964 Occupation (Duncan SEI)
Pearson Correlation	V Father's educational attainment	1.000	.524	.425	.254	.316	.263
	M Mother's educational attainment	.524	1.000	.288	.211	.260	.211
	X Status of F's Occ when son grad from HS	.425	.288	1.000	.184	.291	.262
	Q Son's score on Henmon-Nelson IQ test	.254	.211	.184	1.000	.448	.377
	U Son's educational attainment	.316	.260	.291	.448	1.000	.621
	W Son's 1964 Occupation (Duncan SEI)	.263	.211	.262	.377	.621	1.000

Variables Entered/Removed[®]

Model	Variables Entered	Variables Removed	Method
1	U Son's educational attainment		Forward (Criterion: Probability-of-F-to-enter <= .050)
2	Q Son's score on Henmon-Nelson IQ test		Forward (Criterion: Probability-of-F-to-enter <= .050)
3	X Status of F's Occ when son grad from HS		Forward (Criterion: Probability-of-F-to-enter <= .050)

a. Dependent Variable: W Son's 1964 Occupation (Duncan SEI)

Model Summary

					Change Statistics					
			Adjusted	Std. Error of	R Square					
Model	R	R Square	R Square	the Estimate	Change	F Change	df1	df2	Sig. F Change	
1	.621 ^a	.386	.385	18.2985633	.386	1297.482	1	2067	.000	
2	.631 ^b	.398	.397	18.1201897	.012	41.895	1	2066	.000	
3	.636 ^c	.404	.403	18.0323286	.006	21.182	1	2065	.000	

a. Predictors: (Constant), U Son's educational attainment

b. Predictors: (Constant), U Son's educational attainment, Q Son's score on Henmon-Nelson IQ test

c. Predictors: (Constant), U Son's educational attainment, Q Son's score on Henmon-Nelson IQ test, X Status of F's Occ when son grad from HS

		Sum of				
Model		Squares	df	Mean Square	F	Sig.
1	Regression	434445.635	1	434445.635	1297.482	.000 ^a
	Residual	692108.946	2067	334.837		
	Total	1126554.581	2068			
2	Regression	448201.508	2	224100.754	682.524	.000 ^b
	Residual	678353.073	2066	328.341		
	Total	1126554.581	2068			
3	Regression	455089.114	3	151696.371	466.521	.000 ^c
	Residual	671465.466	2065	325.165		
	Total	1126554.581	2068			

ANOVAd

a. Predictors: (Constant), U Son's educational attainment

b. Predictors: (Constant), U Son's educational attainment, Q Son's score on Henmon-Nelson IQ test

c. Predictors: (Constant), U Son's educational attainment, Q Son's score on Henmon-Nelson IQ test, X Status of F's Occ when son grad from HS

d. Dependent Variable: W Son's 1964 Occupation (Duncan SEI)

						Coe	ficients	1						
		Unstan Coef	dardized ficients	Stan Coe	dardized			95% Confidence	ce Interval for B	Cor	Correlations		Collinearity Statistics	
Model		В	Std. Error	Beta	Std. Error	t	Sig.	Lower Bound	Upper Bound	Zero-order	Partial	Part	Tolerance	VIF
1	(Constant)	-72.217	3.199			-22.573	.000	-78.492	-65.943					
	U Son's educational attainment	8.648	.240	.621	.017	36.021	.000	8.177	9.119	.621	.621	.621	1.000	1.000
2	(Constant)	-81.795	3.497			-23.392	.000	-88.652	-74.938					
	U Son's educational attainment	7.877	.266	.566	.019	29.621	.000	7.355	8.398	.621	.546	.506	.799	1.251
	Q Son's score on Henmon-Nelson IQ test	.198	.031	.124	.019	6.473	.000	.138	.258	.377	.141	.111	.799	1.251
3	(Constant)	-79.806	3.506			-22.760	.000	-86.683	-72.929					
	U Son's educational attainment	7.579	.272	.544	.020	27.822	.000	7.045	8.114	.621	.522	.473	.754	1.326
	Q Son's score on Henmon-Nelson IQ test	.189	.031	.118	.019	6.203	.000	.129	.249	.377	.135	.105	.796	1.256
	X Status of F's Occ when son grad from HS	.085	.019	.082	.018	4.602	.000	.049	.122	.262	.101	.078	.912	1.097

a. Dependent Variable: W Son's 1964 Occupation (Duncan SEI)

Excluded Variables^d

						Collin	Collinearity Sta	
Model		Beta In	t	Sig.	Partial Correlation	Tolerance	VIF	Minimum Tolerance
1	V Father's educational attainment	.074 ^a	4.097	.000	.090	.900	1.111	.900
	M Mother's educational attainment	.053 ^a	2.982	.003	.065	.932	1.073	.932
	X Status of F's Occ when son grad from HS	.089 ^a	4.956	.000	.108	.915	1.093	.915
	Q Son's score on Henmon-Nelson IQ test	.124 ^a	6.473	.000	.141	.799	1.251	.799
2	V Father's educational attainment	.060 ^b	3.301	.001	.072	.884	1.131	.756
	M Mother's educational attainment	.041 ^b	2.313	.021	.051	.921	1.086	.771
	X Status of F's Occ when son grad from HS	.082 ^b	4.602	.000	.101	.912	1.097	.754
3	V Father's educational attainment	.034 [°]	1.761	.078	.039	.768	1.302	.736
	M Mother's educational attainment	.024 ^c	1.321	.187	.029	.875	1.143	.740

a. Predictors in the Model: (Constant), U Son's educational attainment

b. Predictors in the Model: (Constant), U Son's educational attainment, Q Son's score on Henmon-Nelson IQ test

c. Predictors in the Model: (Constant), U Son's educational attainment, Q Son's score on Henmon-Nelson IQ test, X Status of F's Occ when son grad from HS

d. Dependent Variable: W Son's 1964 Occupation (Duncan SEI)

Subtitle "Stepwise backward selection".
Stepwise Regression Example
Stepwise backward selection
REGRESSION /MATRIX IN(*)
/DESCRIPTIVES DEF
/VARIABLES V M X Q U W
/CRITERIA=POUT(.051)

/CRITERIA=POUT(.051)
/STATISTICS DEF CHANGE OUTS ZPP CI SES TOL
/DEPENDENT W
/METHOD ENTER V M X Q U
/METHOD BACKWARDS .

Regression

Descriptive Statistics

	Mean	Std. Deviation	N
V Father's educational attainment	10.200000	3.0080000	2069
M Mother's educational attainment	10.410000	2.8480000	2069
X Status of F's Occ when son grad from HS	33.110000	22.4100000	2069
Q Son's score on Henmon-Nelson IQ test	99.860000	14.5700000	2069
U Son's educational attainment	13.220000	1.6760000	2069
W Son's 1964 Occupation (Duncan SEI)	42.110000	23.3400000	2069

			Correlations				
		V Father's educational attainment	M Mother's educational attainment	X Status of F's Occ when son grad from HS	Q Son's score on Henmon-Nel son IQ test	U Son's educational attainment	W Son's 1964 Occupation (Duncan SEI)
Pearson Correlation	V Father's educational attainment	1.000	.524	.425	.254	.316	.263
	M Mother's educational attainment	.524	1.000	.288	.211	.260	.211
	X Status of F's Occ when son grad from HS	.425	.288	1.000	.184	.291	.262
	Q Son's score on Henmon-Nelson IQ test	.254	.211	.184	1.000	.448	.377
	U Son's educational attainment	.316	.260	.291	.448	1.000	.621
	W Son's 1964 Occupation (Duncan SEI)	.263	.211	.262	.377	.621	1.000

Variables Entered/Removed[®]

Model	Variables Entered	Variables Removed	Method
1	U Son's educational attainment, M Mother's educational attainment, X Status of F's Occ when son grad from HS, Q Son's score on Henmon-Nelson IQ test, V Father's educational attainment		Enter
2		M Mother's educational attainment	Backward (criterion: Probability of F-to-remove >= .051).
3		V Father's educational attainment	Backward (criterion: Probability of F-to-remove >= .051).

a. All requested variables entered.

b. Dependent Variable: W Son's 1964 Occupation (Duncan SEI)

Model Summary

					Change Statistics					
Madal	Б	D Caucro	Adjusted	Std. Error of	R Square	E Change	461	45	Sig. E.Changa	
woder	ĸ	R Square	R Square	the Estimate	Change	r Change	un	uiz	Sig. F Change	
1	.636 ^a	.405	.404	18.0258459	.405	280.811	5	2063	.000	
2	.636 ^b	.405	.404	18.0231567	.000	.384	1	2063	.535	
3	.636 ^c	.404	.403	18.0323286	001	3.102	1	2064	.078	

a. Predictors: (Constant), U Son's educational attainment, M Mother's educational attainment, X Status of F's Occ when son grad from HS, Q Son's score on Henmon-Nelson IQ test, V Father's educational attainment

b. Predictors: (Constant), U Son's educational attainment, X Status of F's Occ when son grad from HS, Q Son's score on Henmon-Nelson IQ test, V Father's educational attainment

c. Predictors: (Constant), U Son's educational attainment, X Status of F's Occ when son grad from HS, Q Son's score on Henmon-Nelson IQ test

ANOVAd

Model		Sum of Squares	df	Mean Square	F	Sia.
1	Regression	456221 678	5	01244 336	280 811	000a
1 ·	rtegression	430221.070	5	31244.330	200.011	.000
	Residual	670332.903	2063	324.931		
	Total	1126554.581	2068			
2	Regression	456096.841	4	114024.210	351.023	.000 ^b
	Residual	670457.740	2064	324.834		
	Total	1126554.581	2068			
3	Regression	455089.114	3	151696.371	466.521	.000 ^c
	Residual	671465.466	2065	325.165		
	Total	1126554.581	2068			

a. Predictors: (Constant), U Son's educational attainment, M Mother's educational attainment, X Status of F's Occ when son grad from HS, Q Son's score on Henmon-Nelson IQ test, V Father's educational attainment

b. Predictors: (Constant), U Son's educational attainment, X Status of F's Occ when son grad from HS, Q Son's score on Henmon-Nelson IQ test, V Father's educational attainment

c. Predictors: (Constant), U Son's educational attainment, X Status of F's Occ when son grad from HS, Q Son's score on Henmon-Nelson IQ test

d. Dependent Variable: W Son's 1964 Occupation (Duncan SEI)

Coefficients^a

		Unstandardized Standardized				05% Confidence Interval for R		Correlations			Collinearity			
Model		B	Std Error	Beta	Std Error	l .	Sig	Jower Bound	Lipper Bound	Zero-order	Partial	Part	Tolerance	
1	(Constant)	-80 759	3 560	Dela	Stu. LIIU	-22 685	000	-87 741	-73 777	Zero-order	Failiai	Fait	TOIETATICE	VII
	V Father's educational attainment	.220	.167	.028	.021	1.319	.187	107	.547	.263	.029	.022	.624	1.602
	M Mother's educational attainment	.102	.165	.012	.020	.620	.535	221	.426	.211	.014	.011	.711	1.407
	X Status of F's Occ when son grad from HS	.072	.020	.069	.019	3.605	.000	.033	.111	.262	.079	.061	.789	1.268
	Q Son's score on Henmon-Nelson IQ test	.182	.031	.114	.019	5.914	.000	.122	.242	.377	.129	.100	.783	1.277
	U Son's educational attainment	7.491	.277	.538	.020	27.090	.000	6.948	8.033	.621	.512	.460	.732	1.367
2	(Constant)	-80.445	3.523			-22.832	.000	-87.355	-73.536					
	V Father's educational attainment	.265	.150	.034	.019	1.761	.078	030	.560	.263	.039	.030	.768	1.302
	X Status of F's Occ when son grad from HS	.073	.020	.070	.019	3.653	.000	.034	.112	.262	.080	.062	.792	1.263
	Q Son's score on Henmon-Nelson IQ test	.183	.031	.114	.019	5.955	.000	.123	.243	.377	.130	.101	.785	1.274
	U Son's educational attainment	7.503	.276	.539	.020	27.218	.000	6.963	8.044	.621	.514	.462	.736	1.359
3	(Constant)	-79.806	3.506			-22.760	.000	-86.683	-72.929					
	X Status of F's Occ when son grad from HS	.085	.019	.082	.018	4.602	.000	.049	.122	.262	.101	.078	.912	1.097
	Q Son's score on Henmon-Nelson IQ test	.189	.031	.118	.019	6.203	.000	.129	.249	.377	.135	.105	.796	1.256
	U Son's educational attainment	7.579	.272	.544	.020	27.822	.000	7.045	8.114	.621	.522	.473	.754	1.326

a. Dependent Variable: W Son's 1964 Occupation (Duncan SEI)

Excluded Variables^C

						Collinearity Statistics		
Model		Beta In	t	Sig.	Partial Correlation	Tolerance	VIF	Minimum Tolerance
2	M Mother's educational attainment	.012 ^a	.620	.535	.014	.711	1.407	.624
3	M Mother's educational attainment	.024 ^b	1.321	.187	.029	.875	1.143	.740
	V Father's educational attainment	.034 ^b	1.761	.078	.039	.768	1.302	.736

a. Predictors in the Model: (Constant), U Son's educational attainment, X Status of F's Occ when son grad from HS, Q Son's score on Henmon-Nelson IQ test, V Father's educational attainment

b. Predictors in the Model: (Constant), U Son's educational attainment, X Status of F's Occ when son grad from HS, Q Son's score on Henmon-Nelson IQ test

c. Dependent Variable: W Son's 1964 Occupation (Duncan SEI)