

READ CAREFULLY. Give an answer in the form of a number or numeric expression where possible. Show calculations for possible partial credit. Tables of the t distribution are provided. Use a critical value of $\alpha=0.05$ if not specified in a particular question.

1) 2 points each – Answer the following questions as true (T) or false (F) by **circling** the appropriate letter.

- T** **F** a) The equation $\hat{Y}_i = b_0 + b_1X_i$ would be a correct equation if the term “+ e_i ” was added to it.
- T** **F** b) For a simple linear regression the value of R^2 can never exceed 1.0.
- T** **F** c) Prior to doing regression analysis, the raw data values for Y_i should be examined to determine if they are normally distributed in order to meet the usual regression assumptions.
- T** **F** d) There are no *a priori* expectations or limitations on the distribution of the independent variables in a regression.
- T** **F** e) Hat values (i.e. h_{ii}) are NOT based on calculations of some change from deleted observations.
- T** **F** f) Logarithmic transformation of an Y_i variable in regression does not affect homogeneity of variance.
- T** **F** g) The sum of squares that a variable accounts for, or removes from the remaining error term when it enters the model, is called the extra sum of squares.
- T** **F** h) Sequential sums of squares (TYPE I) will always sum to a number that is equal to the model sum of squares.

All multiple choice questions have one correct response unless otherwise indicated.

Circle the letter corresponding to the best response.

- 2) 3 points – For an ordinary multiple regression, with $n = 17$ observations (Y_1, Y_2, \dots, Y_{17}) and 4 quantitative variables (X_1, X_2, X_3 and X_4) plus an intercept, the size of the $X'X$ matrix (X prime X) will be which of the following?
- a) 4 x 17.
- b) 4 x 4
- c) 17 x 17
- d) 5 x 5**
- 3) 3 points – Which of the following best describes a partial residual plot of Y and X_3 in a 4 variable multiple regression with X_1, X_2, X_3 and X_4 ?
- a) plot Y on X_3 after adjusting Y for X_1, X_2, X_3 and X_4
- b) plot Y on X_3 after adjusting both Y and X_3 for X_1, X_2, X_3 and X_4 .
- c) plot Y on X_3 after adjusting Y for X_1, X_2 and X_4
- d) plot Y on X_3 after adjusting both Y and X_3 for X_1, X_2 and X_4 .**

4) 3 points each – For the questions that follow, choose one diagnostic from this list (SAS names used).

Circle the letter corresponding to your answer.

a) Some diagnostic statistics require that the regression be fitted with an observation deleted in order to compare the result to the regression with the observation included. Which of the statistics below is calculated using a "deleted" observation?

- a) STUDENT **b) RSTUDENT** c) HATDIAG d) VIF

b) Which of the statistics below is NOT calculated from a "deleted" observation?

- a) COOK's D b) DFFITS c) DFBETAS **d) HATDIAG**

c) Some statistics are diagnostics for observations and some for variables. Which of the statistics below is a diagnostic for a variable?

- a) STUDENT b) RSTUDENT c) HATDIAG **d) VIF**

d) Some normally distributed statistics are "standardized" in order to obtain scores that approximate a Z distribution or t distribution. Which one of the following are "standardized"?

- a) R^2 **b) RSTUDENT** c) HATDIAG d) VIF

e) Which of the statistics below could be produced by "standardizing" the Y_i variable and the X_i variables before doing the regression ?

- a) RSTUDENT b) COOK's D c) HATDIAG **d) Standardized b_i**

f) Some diagnostic statistics are called "influence diagnostics". Which of the statistics below is NOT an influence diagnostic?

- a) COOK's D b) DFFITS c) DFBETAS **d) HATDIAG**

g) In regression analysis, how would the values of each of the TYPE II sum of squares compare to the values of the TYPE III sum of squares for the same variable?

- a) Type II larger b) Type II smaller **c) would be the same** d) cannot say

h) In regression analysis, how would the values of each of the TYPE I sum of squares compare to the values of the TYPE II sum of squares for the same variable?

- a) Type II larger b) Type II smaller c) would be the same **d) cannot say**

i) Which of the diagnostic tools below would NOT be used to examine the observations for potential outliers of Y_i values?

- a) STUDENT b) RSTUDENT **c) HATDIAG** d) a residual plot

j) Which of the statistics below would be the better statistic to use as a diagnostic for multicollinearity?

- a) STUDENT b) COOK's D c) R^2 **d) VIF**

Put your name on the output and RETURN THE OUTPUT WITH YOUR EXAM!

The remaining questions refer to the SAS output pages given separately. These pages contain output for a multiple regression fitted with PROC REG. t-tables are provided with the computer output. Questions requesting a “P value” may not always have a P value associated. If no P value is associated with the answer, state what statistic or graphic was used to answer the question.

A team of investigators are analyzing some crime of crime statistics from 47 states. They make some *a priori* decisions to streamline their multiple regression analysis.

- 1) they assume that there will be multicollinearity in the analysis if all variables are used.
- 2) they assume that a stepwise regression will eliminate the multicollinearity issues.
- 3) They plan to use the reduced set of variables from the stepwise regression to do their analysis.

The computer output provided contains (1) the SAS program, (2) the full model analysis with selected diagnostics, (3) the summary from the stepwise regression and (4) the reduced model analysis with selected diagnostics. **Unless otherwise stated the questions below will be about the reduced model analysis.**

9a) 3 points – The investigators believed that the full model would have multicollinearity? Based the original full model, does the evidence support this believe?

Circle one: Yes / No P value or statistic used = Condition No & VIF .

9b) 3 points – One of the key hypotheses by the investigators was that there would be a relationship between the wealth (i.e. the variable “Wealth”) of a state and crime. Based the original full model, does the evidence support this hypothesis?

Circle one: Yes / No P value or statistic used = 0.2533 .

9c) 3 points – The investigators used stepwise regression (forward selection with an option to remove non-significant variables) to reduce the model. Based on the summary of the stepwise regression, what was the first variable entered into the model and what proportion of the total variation did it accounted for?

Variable name Po1 Proportion of variance = 0.4728 .

9d) 3 points – The investigators used stepwise regression (forward selection with an option to remove non-significant variables) to reduce the model. Based on the stepwise regression summary, which variable was entered second in the model and what was its P value at the time it was entered?

Variable name Ineq P value or statistic used = 0.0016 .

The remaining questions should be answered with results of the “Reduced model multiple regression”.

9e) 4 points – The investigators used stepwise regression to eliminate the multicollinearity they believed would be present in the original model. Were they correct that multicollinearity, if present originally, was eliminated from the analysis?

Circle one: Yes / No P value or statistic used = Condition Number & VIF .

9f) 4 points – Of all the variables examined, which one, or ones, appear to reduce crime?

Variable names Prob Statistic used = negative slope .

9g) 4 points – Published studies indicate that the number of crimes is reduced as the mean years of schooling (Ed) increases. Does this contention appear to be correct?

Circle one: Yes / No P value or statistic used = positive slope .

9h) 3 points – Is the assumption of “normality” met?

Circle one: Yes / No P value or statistic used = 0.2903 .

9i) 3 points – Is the assumption of “homogeneity” met?

Circle one: Yes / No P value or statistic used = residual plot .

9j) 3 points – Which of the variables in the reduced model appears to be the most important to the analysis?

Variable name Po1 Statistic used = t value (& P value) .

9k) 3 points – What is the standard error of the intercept? 816.2794 .

9l) 3 points – The coefficient of the variable “Ineq” suggests that there are an additional 68.31 crimes for each additional percent of income inequality. Place a confidence interval on the effect of each percentage point of income inequality on crime.

$$P(\underline{38.90688} \leq \text{estimate} \leq \underline{97.71363}) = 0.95$$

9m) 3 points – The estimated, or predicted, rate of crime for the first state, labeled “a”, was 754.96. Place a confidence interval this estimate for this one, particular state.

$$P(\underline{309.01} \leq \text{estimate} \leq \underline{1200.92}) = 0.95$$

9n) 3 points – The 99.5% two-tailed t-value for 41 degrees of freedom is 2.97? The investigators have decided to use this value to define an “outlier”. Based on this value, do you think there are outliers in this analysis and if so, which observation?

Circle one: Yes / No Observation number = #29 or "C" (Rstudent) .

9o) 3 points – Are there any particularly unusual combinations of values of the X_i variables? If you find an observation that occupies an unusual value in the “X space” give the observation number.

Circle one: Yes / No Observation number = #18 or "r", et al. (hat value) .

$$h_{ii} = 2(p/n) = 2(6/47) = 0.2553$$

Office use only
Do not write in this box

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dm'log;clear;output;clear';
options ps=512 ls=120 nocenter nodate nonumber nolabel FORMCHAR="|----+|----+|-/\<>*" ;
TITLE1 'Crime rate statistics example';

ODS HTML style=minimal body='C:\Geaghan\Current\EXST7015\Fall2010\SAS\Crime
Rates\CrimeRateStats.HTML' ;

*****;
*** OzDASL - Australasian Data and Story Library ***;
*** http://www.statsci.org/data/index.html ***;
*****;
*** Criminologists are interested in the effect of punishment regimes on ***;
*** crime rates. This has been studied using aggregate data on 47 states ***;
*** of the USA for 1960. The data set contains the following columns: ***;
*****;

data crime; infile cards missover;
  input M So Ed Po1 Po2 LF MpF Pop NW U1 U2 Wealth Ineq Prob Time Crime ObsID $;
  label M = percentage of males aged 14-24 in total state population
        So = indicator variable for a southern state
        Ed = mean years of schooling of the population aged 25 years or over
        Po1 = per capita expenditure on police protection in 1960
        Po2 = per capita expenditure on police protection in 1959
        LF = labor force participation rate by urban males in the age-group 14-24
        M.F = number of males per 100 females
        Pop = state population in 1960 in hundred thousands
        NW = percentage of nonwhites in the population
        U1 = unemployment rate of urban males 14-24
        U2 = unemployment rate of urban males 35-39
        Wealth = wealth: median value of transferable assets or family income
        Ineq = income inequality: % of families earning below half the median income
        Prob = probability of imprisonment: ratio of commitment number to offenses
        Time = average time in months served by first offenders in state prisons
        Crime = crime rate: number of offenses per 100,000 population in 1960;
datalines; run;

data here

;

proc print data=crime;
  TITLE3 'Raw data list'; run;

proc reg data=crime LINEPRINTER;
  TITLE2 'Original Full model multiple regression';
  model Crime = M So Ed Po1 LF MpF Pop NW U1 Wealth Ineq Prob Time / clb vif collin;
run;

proc reg data=crime LINEPRINTER;
  TITLE2 'Stepwise regression';
  model Crime = M So Ed Po1 LF MpF Pop NW U1 Wealth Ineq Prob Time / selection=stepwise;
run;

proc reg data=crime LINEPRINTER; ID ObsID;
  TITLE2 'Reduced model multiple regression';
  model Crime = M Ed Po1 Ineq Prob / clb vif collin;
  Police:Test PO1 = 100;
  OUTPUT OUT=NEXT1 P=Predicted R=Resid cookd=cooksd dffits=dffits h=hatdiag
  STUDENT=student rstudent=rstudent lclm=lclm uclm=uclm lcl=lcl ucl=ucl;
run;

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proc print data=next1;
  TITLE3 'Listing of observation diagnostics';
  var ObsId crime Predicted Resid student rstudent cooks dffits lcl ucl lclm uclm; run;
options ps=512 ls=85;

proc univariate data=next1 normal plot; var Resid;
  TITLE3 'Residual analysis'; run;

options ls=120 ps=61;
proc plot data=next1; plot resid*predicted=obsid=ObsID / vref=0;
  TITLE3 'Residual plot'; run;

```

Crime rate statistics example
 Reduced model multiple regression
 Raw data list

Obs	M	So	Ed	Po1	Po2	LF	MpF	Pop	NW	U1	U2	Wealth	Ineq	Prob	Time	Crime	ID
1	15.1	1	9.1	5.8	5.6	0.510	95.0	33	30.1	0.108	4.1	3940	26.1	0.08460	26.2011	791	a
2	14.3	0	11.3	10.3	9.5	0.583	101.2	13	10.2	0.096	3.6	5570	19.4	0.02960	25.2999	1635	b
3	14.2	1	8.9	4.5	4.4	0.533	96.9	18	21.9	0.094	3.3	3180	25.0	0.08340	24.3006	578	c
4	13.6	0	12.1	14.9	14.1	0.577	99.4	157	8.0	0.102	3.9	6730	16.7	0.01580	29.9012	1969	d
5	14.1	0	12.1	10.9	10.1	0.591	98.5	18	3.0	0.091	2.0	5780	17.4	0.04140	21.2998	1234	e
6	12.1	0	11.0	11.8	11.5	0.547	96.4	25	4.4	0.084	2.9	6890	12.6	0.03420	20.9995	682	f
7	12.7	1	11.1	8.2	7.9	0.519	98.2	4	13.9	0.097	3.8	6200	16.8	0.04210	20.6993	963	g
8	13.1	1	10.9	11.5	10.9	0.542	96.9	50	17.9	0.079	3.5	4720	20.6	0.04010	24.5988	1555	h
9	15.7	1	9.0	6.5	6.2	0.553	95.5	39	28.6	0.081	2.8	4210	23.9	0.07170	29.4001	856	i
10	14.0	0	11.8	7.1	6.8	0.632	102.9	7	1.5	0.100	2.4	5260	17.4	0.04450	19.5994	705	j
11	12.4	0	10.5	12.1	11.6	0.580	96.6	101	10.6	0.077	3.5	6570	17.0	0.01620	41.6000	1674	k
12	13.4	0	10.8	7.5	7.1	0.595	97.2	47	5.9	0.083	3.1	5800	17.2	0.03120	34.2984	849	l
13	12.8	0	11.3	6.7	6.0	0.624	97.2	28	1.0	0.077	2.5	5070	20.6	0.04530	36.2993	511	m
14	13.5	0	11.7	6.2	6.1	0.595	98.6	22	4.6	0.077	2.7	5290	19.0	0.05320	21.5010	664	n
15	15.2	1	8.7	5.7	5.3	0.530	98.6	30	7.2	0.092	4.3	4050	26.4	0.06910	22.7008	798	o
16	14.2	1	8.8	8.1	7.7	0.497	95.6	33	32.1	0.116	4.7	4270	24.7	0.05210	26.0991	946	p
17	14.3	0	11.0	6.6	6.3	0.537	97.7	10	0.6	0.114	3.5	4870	16.6	0.07630	19.1002	539	q
18	13.5	1	10.4	12.3	11.5	0.537	97.8	31	17.0	0.089	3.4	6310	16.5	0.11980	18.1996	929	r
19	13.0	0	11.6	12.8	12.8	0.536	93.4	51	2.4	0.078	3.4	6270	13.5	0.01910	24.9008	750	w
20	12.5	0	10.8	11.3	10.5	0.567	98.5	78	9.4	0.130	5.8	6260	16.6	0.03480	26.4010	1225	t
21	12.6	0	10.8	7.4	6.7	0.602	98.4	34	1.2	0.102	3.3	5570	19.5	0.02280	37.5998	742	u
22	15.7	1	8.9	4.7	4.4	0.512	96.2	22	42.3	0.097	3.4	2880	27.6	0.08950	37.0994	439	v
23	13.2	0	9.6	8.7	8.3	0.564	95.3	43	9.2	0.083	3.2	5130	22.7	0.03070	25.1989	1216	w
24	13.1	0	11.6	7.8	7.3	0.574	103.8	7	3.6	0.142	4.2	5400	17.6	0.04160	17.6000	968	x
25	13.0	0	11.6	6.3	5.7	0.641	98.4	14	2.6	0.070	2.1	4860	19.6	0.06920	21.9003	523	y
26	13.1	0	12.1	16.0	14.3	0.631	107.1	3	7.7	0.102	4.1	6740	15.2	0.04170	22.1005	1993	z
27	13.5	0	10.9	6.9	7.1	0.540	96.5	6	0.4	0.080	2.2	5640	13.9	0.03610	28.4999	342	A
28	15.2	0	11.2	8.2	7.6	0.571	101.8	10	7.9	0.103	2.8	5370	21.5	0.03820	25.8006	1216	B
29	11.9	0	10.7	16.6	15.7	0.521	93.8	168	8.9	0.092	3.6	6370	15.4	0.02340	36.7009	1043	C
30	16.6	1	8.9	5.8	5.4	0.521	97.3	46	25.4	0.072	2.6	3960	23.7	0.07530	28.3011	696	D
31	14.0	0	9.3	5.5	5.4	0.535	104.5	6	2.0	0.135	4.0	4530	20.0	0.04200	21.7998	373	E
32	12.5	0	10.9	9.0	8.1	0.586	96.4	97	8.2	0.105	4.3	6170	16.3	0.04270	30.9014	754	F
33	14.7	1	10.4	6.3	6.4	0.560	97.2	23	9.5	0.076	2.4	4620	23.3	0.04950	25.5005	1072	G
34	12.6	0	11.8	9.7	9.7	0.542	99.0	18	2.1	0.102	3.5	5890	16.6	0.04080	21.6997	923	H
35	12.3	0	10.2	9.7	8.7	0.526	94.8	113	7.6	0.124	5.0	5720	15.8	0.02070	37.4011	653	I
36	15.0	0	10.0	10.9	9.8	0.531	96.4	9	2.4	0.087	3.8	5590	15.3	0.00690	44.0004	1272	J
37	17.7	1	8.7	5.8	5.6	0.638	97.4	24	34.9	0.076	2.8	3820	25.4	0.04520	31.6995	831	K
38	13.3	0	10.4	5.1	4.7	0.599	102.4	7	4.0	0.099	2.7	4250	22.5	0.05400	16.6999	566	L
39	14.9	1	8.8	6.1	5.4	0.515	95.3	36	16.5	0.086	3.5	3950	25.1	0.04710	27.3004	826	M
40	14.5	1	10.4	8.2	7.4	0.560	98.1	96	12.6	0.088	3.1	4880	22.8	0.03880	29.3004	1151	N
41	14.8	0	12.2	7.2	6.6	0.601	99.8	9	1.9	0.084	2.0	5900	14.4	0.02510	30.0001	880	O
42	14.1	0	10.9	5.6	5.4	0.523	96.8	4	0.2	0.107	3.7	4890	17.0	0.08890	12.1996	542	P
43	16.2	1	9.9	7.5	7.0	0.522	99.6	40	20.8	0.073	2.7	4960	22.4	0.05490	31.9989	823	Q
44	13.6	0	12.1	9.5	9.6	0.574	101.2	29	3.6	0.111	3.7	6220	16.2	0.02810	30.0001	1030	R
45	13.9	1	8.8	4.6	4.1	0.480	96.8	19	4.9	0.135	5.3	4570	24.9	0.05620	32.5996	455	W
46	12.6	0	10.4	10.6	9.7	0.599	98.9	40	2.4	0.078	2.5	5930	17.1	0.04660	16.6999	508	T
47	13.0	0	12.1	9.0	9.1	0.623	104.9	3	2.2	0.113	4.0	5880	16.0	0.05280	16.0997	849	U

Crime rate statistics example

Original Full model multiple regression

The REG Procedure

Model: MODEL1

Dependent Variable: Crime

Number of Observations Read 47
 Number of Observations Used 47

Analysis of Variance

Source	DF	Sum of Squares	Mean Square	F Value	Pr > F
Model	13	5303054	407927	8.53	<.0001
Error	33	1577874	47814		
Corrected Total	46	6880928			

Root MSE	218.66495	R-Square	0.7707
Dependent Mean	905.08511	Adj R-Sq	0.6804
Coeff Var	24.15960		

Parameter Estimates

Variable	DF	Parameter Estimate	Standard Error	t Value	Pr > t	Variance Inflation	95% Confidence Limits	
Intercept	1	-5627.80237	1599.42791	-3.52	0.0013	0	-8881.86292	-2373.74183
M	1	74.94071	42.90448	1.75	0.0900	2.79712	-12.34912	162.23053
So	1	73.95947	151.28993	0.49	0.6282	5.05179	-233.84220	381.76115
Ed	1	137.71236	60.48884	2.28	0.0294	4.40532	14.64689	260.77784
Pol	1	110.42746	24.05473	4.59	<.0001	4.91664	61.48773	159.36718
LF	1	-111.46341	1473.11420	-0.08	0.9401	3.40949	-3108.53678	2885.60996
MpF	1	10.83717	20.93543	0.52	0.6082	3.66139	-31.75628	53.43062
Pop	1	-0.88472	1.34660	-0.66	0.5157	2.52852	-3.62439	1.85495
NW	1	2.04891	6.48129	0.32	0.7539	4.27319	-11.13737	15.23519
U1	1	1286.10195	2751.25126	0.47	0.6432	2.36697	-4311.36083	6883.56473
Wealth	1	0.12429	0.10690	1.16	0.2533	10.23609	-0.09320	0.34178
Ineq	1	77.21985	23.53797	3.28	0.0024	8.48394	29.33150	125.10821
Prob	1	-4307.88573	2288.86666	-1.88	0.0687	2.60559	-8964.61997	348.84850
Time	1	0.51678	6.96375	0.07	0.9413	2.34314	-13.65108	14.68463

Crime rate statistics example

Original Full model multiple regression

Collinearity Diagnostics

Number	Eigenvalue	Condition Index	-----Proportion of Variation-----					
			Intercept	M	So	Ed	Po1	LF
1	11.97230	1.00000	0.00000274	0.00001966	0.00037927	0.00001659	0.00013480	0.00001009
2	1.08720	3.31844	0.00000190	4.939959E-7	0.05785	0.00004790	0.00103	0.00001402
3	0.54833	4.67271	0.00000703	0.00006592	0.00240	0.00006678	0.00059057	0.00003743
4	0.13003	9.59563	2.191751E-8	0.00005649	0.22952	0.00000118	0.00032869	0.00002344
5	0.10964	10.44964	0.00000936	0.00016458	0.23919	0.00002832	0.00005488	0.00001729
6	0.07871	12.33347	0.00001098	0.00040662	0.05325	0.00019317	0.15326	0.00003376
7	0.03444	18.64498	0.00000397	0.00055329	0.00511	0.00025550	0.00057362	0.00059071
8	0.02116	23.78679	0.00021365	0.00508	0.00721	0.00402	0.01442	0.00689
9	0.00986	34.84024	0.00012828	0.00014393	0.06111	0.02882	0.59482	0.00261
10	0.00360	57.66028	0.00032243	0.59331	0.00903	0.01213	0.01664	0.03073
11	0.00242	70.31930	0.00166	0.01128	0.13092	0.43510	0.10548	0.01455
12	0.00147	90.33782	0.00231	0.00031451	0.18404	0.42236	0.03216	0.60102
13	0.00066791	133.88445	0.34323	0.33530	0.00083729	0.09506	0.00368	0.07623
14	0.00017379	262.47111	0.65210	0.05330	0.01917	0.00190	0.07682	0.26725

Collinearity Diagnostics

Number	-----Proportion of Variation-----							
	MpF	Pop	NW	U1	Wealth	Ineq	Prob	Time
1	0.00000165	0.00076148	0.00048229	0.00009458	0.00002056	0.00003174	0.00041416	0.00018237
2	0.00000139	0.00710	0.03064	0.00012781	0.00011688	0.00002960	0.00268	0.00011798
3	0.00000606	0.27244	0.01215	0.00041612	0.00001536	0.00007109	0.01063	0.00050416
4	1.648241E-8	0.12955	0.44005	0.00010951	0.00001966	0.00000348	0.07333	0.01282
5	0.00000532	0.08401	0.09072	0.00008216	0.00017967	0.00024098	0.33274	0.03311
6	0.00000267	0.09793	0.01901	0.00430	0.00289	0.00661	0.00276	0.04557
7	0.00000726	0.00394	0.02592	0.28906	0.00026386	0.00060212	0.16283	0.14450
8	0.00022722	0.09489	0.00911	0.09684	0.00107	0.03636	0.25015	0.50016
9	0.00005078	0.09443	0.19498	0.00004779	0.06687	0.16076	0.03749	0.03117
10	0.00010347	0.05648	0.03947	0.00058946	0.04184	0.12505	0.00390	0.03714
11	0.00001750	0.00565	0.01255	0.03887	0.57409	0.10605	0.00010774	0.08951
12	0.00194	0.00807	0.05458	0.07918	0.06453	0.28457	0.00191	0.00279
13	0.06563	0.00273	0.04052	0.17276	0.24495	0.27835	0.10227	0.03265
14	0.93201	0.14202	0.02983	0.31752	0.00314	0.00126	0.01879	0.06978

Crime rate statistics example
Stepwise regression

All variables left in the model are significant at the 0.1500 level.

No other variable met the 0.1500 significance level for entry into the model.

Summary of Stepwise Selection

Step	Variable Entered	Variable Removed	Number Vars In	Partial R-Square	Model R-Square	C(p)	F Value	Pr > F
1	Pol		1	0.4728	0.4728	32.8690	40.36	<.0001
2	Ineq		2	0.1075	0.5803	19.3962	11.27	0.0016
3	Ed		3	0.0853	0.6656	9.1185	10.97	0.0019
4	M		4	0.0348	0.7004	6.1116	4.88	0.0327
5	Prob		5	0.0375	0.7379	2.7144	5.87	0.0199

Crime rate statistics example
Reduced model multiple regression

The REG Procedure

Model: MODEL1

Dependent Variable: Crime

Number of Observations Read 47
Number of Observations Used 47

Analysis of Variance

Source	DF	Sum of Squares	Mean Square	F Value	Pr > F
Model	5	5077637	1015527	23.09	<.0001
Error	41	1803290	43983		
Corrected Total	46	6880928			

Root MSE	209.72050	R-Square	0.7379
Dependent Mean	905.08511	Adj R-Sq	0.7060
Coeff Var	23.17136		

Parameter Estimates

Variable	DF	Parameter Estimate	Standard Error	t Value	Pr > t	Variance Inflation	95% Confidence Limits	
Intercept	1	-4064.57400	816.27940	-4.98	<.0001	0	-5713.08369	-2416.06431
M	1	79.68867	32.62004	2.44	0.0190	1.75773	13.81117	145.56618
Ed	1	160.15278	43.42207	3.69	0.0007	2.46788	72.46014	247.84543
Po1	1	121.22906	14.06269	8.62	<.0001	1.82675	92.82890	149.62923
Ineq	1	68.31025	14.55943	4.69	<.0001	3.52878	38.90688	97.71363
Prob	1	-3867.27087	1596.55188	-2.42	0.0199	1.37818	-7091.57279	-642.96894

Collinearity Diagnostics

Number	Eigenvalue	Condition Index	-----Proportion of Variation-----						
			Intercept	M	Ed	Po1	Ineq	Prob	
1	5.67805	1.00000	0.00004333	0.00013971	0.00013376	0.00161	0.00033768	0.00358	
2	0.23398	4.92614	0.00004420	0.00000242	0.00080182	0.09644	0.00187	0.30263	
3	0.06349	9.45722	0.00065379	0.00636	0.00012658	0.25901	0.04879	0.68712	
4	0.01934	17.13422	0.00327	0.00077196	0.10399	0.52747	0.18636	0.00113	
5	0.00409	37.26070	0.00062678	0.75305	0.16328	0.00367	0.46334	0.00041977	
6	0.00105	73.68864	0.99537	0.23967	0.73167	0.11179	0.29931	0.00512	

Crime rate statistics example
Reduced model multiple regression

The REG Procedure
Model: MODEL1

Test Police Results for Dependent Variable Crime

Source	DF	Mean Square	F Value	Pr > F
Numerator	1	100232	2.28	0.1388
Denominator	41	43983		

Crime rate statistics example
Reduced model multiple regression
Listing of observation diagnostics

Obs	ID	Crime	Predicted	Resid	student	rstudent	hatdiag	cooks	dffits	lcl	ucl	lclm	uclm
1	a	791	754.96	36.037	0.18201	0.17985	0.10864	0.00067	0.06279	309.01	1200.92	615.36	894.57
2	b	1635	1344.11	290.889	1.44403	1.46403	0.07739	0.02915	0.42401	904.49	1783.73	1226.29	1461.93
3	c	578	423.12	154.882	0.78792	0.78422	0.12148	0.01431	0.29162	-25.41	871.65	275.50	570.74
4	d	1969	1843.03	125.972	0.67140	0.66684	0.19960	0.01874	0.33301	1379.14	2306.92	1653.80	2032.25

Obs													
Obs	ID	Crime	Predicted	Resid	student	rstudent	hatdiag	cooks	dffits	lcl	ucl	lclm	uclm
5	e	1234	1346.78	-112.779	-0.56894	-0.56419	0.10660	0.00644	-0.19488	901.24	1792.32	1208.50	1485.06
6	f	682	820.29	-138.287	-0.70951	-0.70515	0.13630	0.01324	-0.28012	368.81	1271.77	663.92	976.65
7	g	963	704.05	258.953	1.26916	1.27896	0.05349	0.01517	0.30403	269.33	1138.76	606.09	802.00
8	h	1555	1371.26	183.735	0.92422	0.92254	0.10143	0.01607	0.30996	926.76	1815.77	1236.37	1506.16
9	i	856	771.25	84.755	0.42361	0.41933	0.08986	0.00295	0.13176	329.09	1213.41	644.28	898.21
10	j	705	818.11	-113.109	-0.56285	-0.55810	0.08180	0.00470	-0.16658	377.59	1258.63	696.97	939.25
11	k	1674	1170.66	503.338	2.53983	2.73284	0.10705	0.12888	0.94620	725.03	1616.29	1032.09	1309.23
12	l	849	696.40	152.604	0.75090	0.74684	0.06096	0.00610	0.19028	260.14	1132.65	591.82	800.97
13	m	511	809.40	-298.398	-1.51029	-1.53507	0.11245	0.04817	-0.54641	362.68	1256.12	667.37	951.43
14	n	664	728.79	-64.787	-0.32507	-0.32150	0.09692	0.00189	-0.10532	285.20	1172.38	596.93	860.64
15	o	798	767.19	30.809	0.15447	0.15262	0.09558	0.00042	0.04961	323.87	1210.51	636.25	898.13
16	p	946	944.09	1.913	0.00964	0.00952	0.10493	0.00000	0.00326	498.88	1389.29	806.89	1081.29
17	q	539	475.65	63.352	0.32492	0.32135	0.13565	0.00276	0.12731	24.29	927.00	319.65	631.64
18	r	929	831.73	97.266	0.65489	0.65027	0.49847	0.07105	0.64828	313.27	1350.20	532.71	1130.76
19	w	750	1229.21	-479.210	-2.39564	-2.55155	0.09024	0.09487	-0.80358	786.97	1671.45	1101.98	1356.44
20	t	1225	1030.44	194.562	0.95662	0.95560	0.05951	0.00965	0.24037	594.48	1466.40	927.12	1133.75
21	u	742	810.12	-68.125	-0.34396	-0.34023	0.10811	0.00239	-0.11846	364.28	1255.97	670.86	949.39
22	v	439	720.91	-281.909	-1.44914	-1.46949	0.13957	0.05677	-0.59184	268.78	1173.04	562.68	879.14
23	w	1216	1011.39	204.606	1.03504	1.03596	0.11153	0.02241	0.36704	564.86	1457.93	869.95	1152.84
24	x	968	824.10	143.904	0.70708	0.70270	0.05826	0.00516	0.17478	388.39	1259.80	721.86	926.33
25	y	523	664.17	-141.172	-0.72233	-0.71805	0.13156	0.01317	-0.27948	213.63	1114.71	510.55	817.80
26	z	1993	1733.92	259.080	1.39475	1.41153	0.21550	0.08906	0.73981	1266.97	2200.87	1537.30	1930.54
27	A	342	403.28	-61.277	-0.32101	-0.31747	0.17152	0.00356	-0.14445	-55.15	861.70	227.87	578.68
28	B	1216	1255.42	-39.420	-0.19999	-0.19764	0.11667	0.00088	-0.07183	807.86	1702.98	1110.75	1400.09
29	C	1043	1571.24	-528.242	-2.88542	-3.19253	0.23798	0.43335	-1.78410	1099.99	2042.49	1364.63	1777.86
30	D	696	714.50	-18.502	-0.09593	-0.09476	0.15421	0.00028	-0.04046	259.48	1169.53	548.18	880.82
31	E	373	411.03	-38.032	-0.19344	-0.19115	0.12112	0.00086	-0.07096	-37.42	859.49	263.63	558.43
32	F	754	716.59	37.406	0.18397	0.18178	0.06000	0.00036	0.04593	280.53	1152.65	612.84	820.34
33	G	1072	936.38	135.616	0.66903	0.66446	0.06579	0.00525	0.17632	499.14	1373.63	827.75	1045.02
34	H	923	981.40	-58.397	-0.28706	-0.28383	0.05909	0.00086	-0.07113	545.52	1417.27	878.44	1084.35
35	I	653	724.33	-71.326	-0.36664	-0.36273	0.13951	0.00363	-0.14606	272.21	1176.45	566.13	882.52
36	J	1272	1072.14	199.857	1.12073	1.12434	0.27698	0.08020	0.69590	593.53	1550.76	849.24	1295.05
37	K	831	1002.66	-171.661	-0.95836	-0.95739	0.27054	0.05677	-0.58305	525.25	1480.07	782.36	1222.96
38	L	566	607.30	-41.298	-0.20759	-0.20515	0.10016	0.00080	-0.06844	163.05	1051.54	473.26	741.34
39	M	826	804.07	21.928	0.10973	0.10840	0.09202	0.00020	0.03451	361.48	1246.67	675.59	932.55
40	N	1151	1158.00	-6.999	-0.03458	-0.03416	0.06865	0.00001	-0.00927	720.16	1595.84	1047.02	1268.97
41	O	880	828.13	51.869	0.27912	0.27596	0.21485	0.00355	0.14436	361.30	1294.96	631.81	1024.45
42	P	542	301.04	240.957	1.27344	1.28345	0.18597	0.06175	0.61345	-160.20	762.29	118.40	483.69
43	Q	823	1038.94	-215.942	-1.09121	-1.09382	0.10962	0.02443	-0.38379	592.79	1485.09	898.72	1179.17
44	R	1030	1106.67	-76.673	-0.38075	-0.37674	0.07802	0.00204	-0.10959	666.92	1546.42	988.37	1224.97
45	W	455	493.67	-38.674	-0.19702	-0.19470	0.12397	0.00092	-0.07324	44.65	942.70	344.55	642.80
46	T	508	878.02	-370.019	-1.82919	-1.88531	0.06965	0.04175	-0.51584	439.98	1316.06	766.24	989.79
47	U	849	889.05	-40.053	-0.19920	-0.19685	0.08081	0.00058	-0.05837	448.73	1329.37	768.65	1009.46

Crime rate statistics example
 Reduced model multiple regression
 Residual analysis

The UNIVARIATE Procedure

Variable: Resid

Moments

N	47	Sum Weights	47
Mean	0	Sum Observations	0
Std Deviation	197.994856	Variance	39201.9629
Skewness	-0.3946102	Kurtosis	0.98133542
Uncorrected SS	1803290.3	Corrected SS	1803290.3
Coeff Variation	.	Std Error Mean	28.880518

Basic Statistical Measures

Location		Variability	
Mean	0.00000	Std Deviation	197.99486
Median	-6.99881	Variance	39202
Mode	.	Range	1032
		Interquartile Range	220.57621

Tests for Location: Mu0=0

Test	-Statistic-	-----p Value-----	
Student's t	t 0	Pr > t	1.0000
Sign	M -0.5	Pr >= M	1.0000
Signed Rank	S 15	Pr >= S	0.8759

Tests for Normality

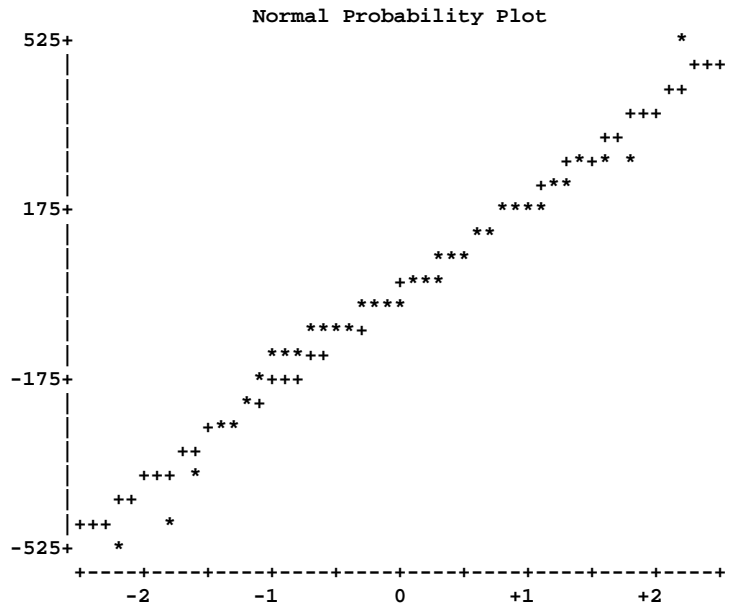
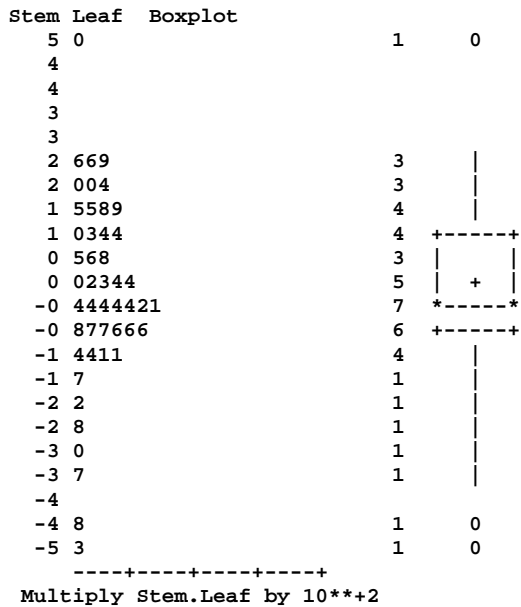
Test	--Statistic--	-----p Value-----	
Shapiro-Wilk	W 0.971049	Pr < W	0.2903
Kolmogorov-Smirnov	D 0.115245	Pr > D	0.1179
Cramer-von Mises	W-Sq 0.07433	Pr > W-Sq	0.2434
Anderson-Darling	A-Sq 0.494434	Pr > A-Sq	0.2139

Quantiles (Definition 5)

Quantile	Estimate
100% Max	503.33791
99%	503.33791
95%	259.08028
90%	240.95716
75% Q3	143.90368
50% Median	-6.99881
25% Q1	-76.67253
10%	-281.90902
5%	-370.01855
1%	-528.24222
0% Min	-528.24222

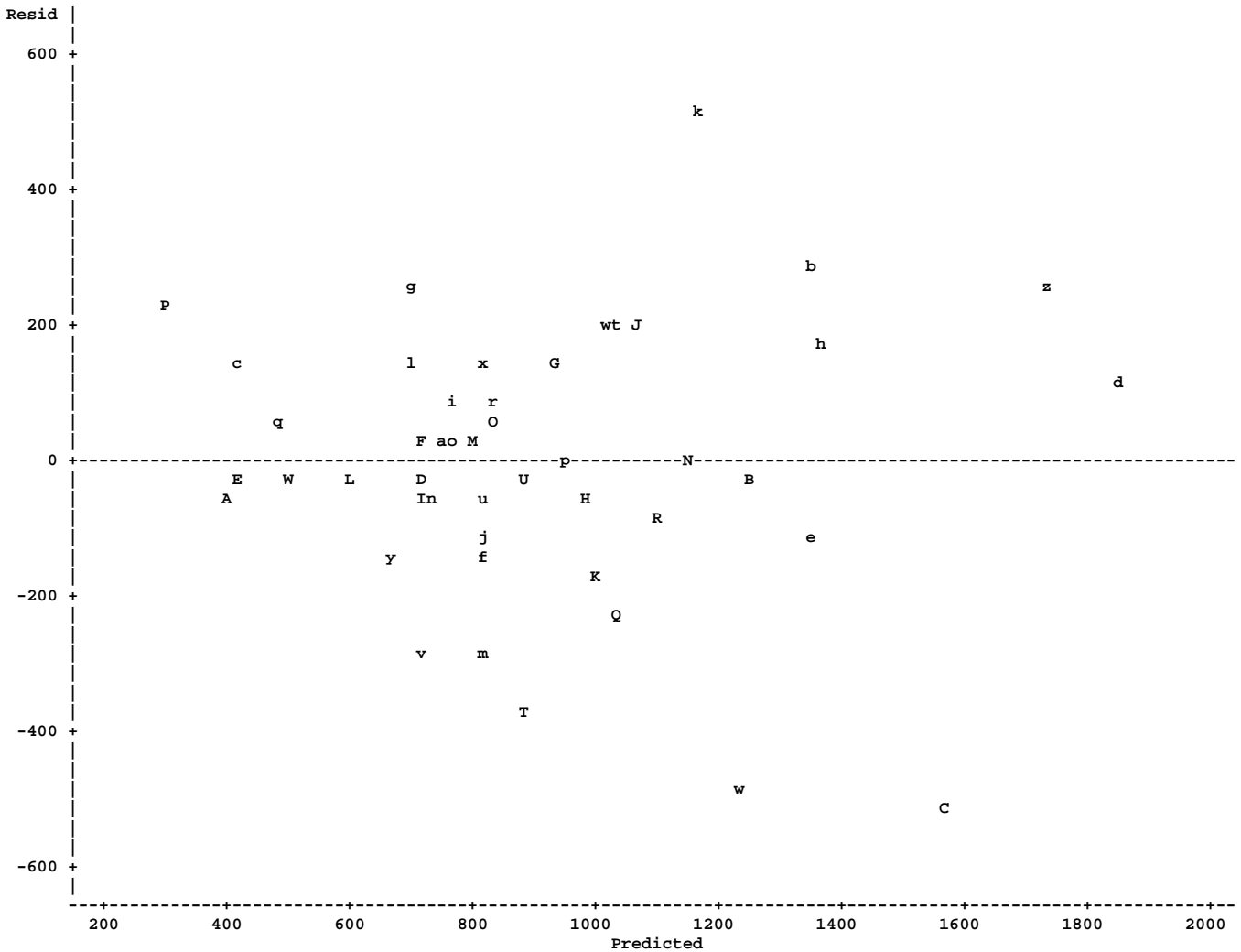
Extreme Observations

-----Lowest-----		-----Highest-----	
Value	Obs	Value	Obs
-528.242	29	240.957	42
-479.210	19	258.953	7
-370.019	46	259.080	26
-298.398	13	290.889	2
-281.909	22	503.338	11



Crime rate statistics example
Reduced model multiple regression
Residual plot

Plot of Resid*Predicted. Symbol is value of ObsID.



t - tables : Probability of a larger absolute value (two tailed test)											
<i>d.f.</i>	0.500	0.400	0.300	0.200	0.100	0.050	0.020	0.010	0.002	0.001	<i>d.f.</i>
1	1.000	1.376	1.963	3.078	6.314	12.706	31.821	63.656	318.289	636.578	1
2	0.816	1.061	1.386	1.886	2.920	4.303	6.965	9.925	22.328	31.600	2
3	0.765	0.978	1.250	1.638	2.353	3.182	4.541	5.841	10.214	12.924	3
4	0.741	0.941	1.190	1.533	2.132	2.776	3.747	4.604	7.173	8.610	4
5	0.727	0.920	1.156	1.476	2.015	2.571	3.365	4.032	5.894	6.869	5
6	0.718	0.906	1.134	1.440	1.943	2.447	3.143	3.707	5.208	5.959	6
7	0.711	0.896	1.119	1.415	1.895	2.365	2.998	3.499	4.785	5.408	7
8	0.706	0.889	1.108	1.397	1.860	2.306	2.896	3.355	4.501	5.041	8
9	0.703	0.883	1.100	1.383	1.833	2.262	2.821	3.250	4.297	4.781	9
10	0.700	0.879	1.093	1.372	1.812	2.228	2.764	3.169	4.144	4.587	10
11	0.697	0.876	1.088	1.363	1.796	2.201	2.718	3.106	4.025	4.437	11
12	0.695	0.873	1.083	1.356	1.782	2.179	2.681	3.055	3.930	4.318	12
13	0.694	0.870	1.079	1.350	1.771	2.160	2.650	3.012	3.852	4.221	13
14	0.692	0.868	1.076	1.345	1.761	2.145	2.624	2.977	3.787	4.140	14
15	0.691	0.866	1.074	1.341	1.753	2.131	2.602	2.947	3.733	4.073	15
16	0.690	0.865	1.071	1.337	1.746	2.120	2.583	2.921	3.686	4.015	16
17	0.689	0.863	1.069	1.333	1.740	2.110	2.567	2.898	3.646	3.965	17
18	0.688	0.862	1.067	1.330	1.734	2.101	2.552	2.878	3.610	3.922	18
19	0.688	0.861	1.066	1.328	1.729	2.093	2.539	2.861	3.579	3.883	19
20	0.687	0.860	1.064	1.325	1.725	2.086	2.528	2.845	3.552	3.850	20
21	0.686	0.859	1.063	1.323	1.721	2.080	2.518	2.831	3.527	3.819	21
22	0.686	0.858	1.061	1.321	1.717	2.074	2.508	2.819	3.505	3.792	22
23	0.685	0.858	1.060	1.319	1.714	2.069	2.500	2.807	3.485	3.768	23
24	0.685	0.857	1.059	1.318	1.711	2.064	2.492	2.797	3.467	3.745	24
25	0.684	0.856	1.058	1.316	1.708	2.060	2.485	2.787	3.450	3.725	25
26	0.684	0.856	1.058	1.315	1.706	2.056	2.479	2.779	3.435	3.707	26
27	0.684	0.855	1.057	1.314	1.703	2.052	2.473	2.771	3.421	3.689	27
28	0.683	0.855	1.056	1.313	1.701	2.048	2.467	2.763	3.408	3.674	28
29	0.683	0.854	1.055	1.311	1.699	2.045	2.462	2.756	3.396	3.660	29
30	0.683	0.854	1.055	1.310	1.697	2.042	2.457	2.750	3.385	3.646	30
32	0.682	0.853	1.054	1.309	1.694	2.037	2.449	2.738	3.365	3.622	32
34	0.682	0.852	1.052	1.307	1.691	2.032	2.441	2.728	3.348	3.601	34
36	0.681	0.852	1.052	1.306	1.688	2.028	2.434	2.719	3.333	3.582	36
38	0.681	0.851	1.051	1.304	1.686	2.024	2.429	2.712	3.319	3.566	38
40	0.681	0.851	1.050	1.303	1.684	2.021	2.423	2.704	3.307	3.551	40
45	0.680	0.850	1.049	1.301	1.679	2.014	2.412	2.690	3.281	3.520	45
50	0.679	0.849	1.047	1.299	1.676	2.009	2.403	2.678	3.261	3.496	50
75	0.678	0.846	1.044	1.293	1.665	1.992	2.377	2.643	3.202	3.425	75
100	0.677	0.845	1.042	1.290	1.660	1.984	2.364	2.626	3.174	3.390	100
∞	0.674	0.842	1.036	1.282	1.645	1.960	2.326	2.576	3.090	3.290	∞
<i>d.f.</i>	0.250	0.200	0.150	0.100	0.050	0.025	0.010	0.005	0.001	0.0005	<i>d.f.</i>

t - tables : Probability of a larger value (one tailed test)