

Directions: Answer each of the four exercises below showing all relevant work. Conclusions and justifications are to be given in clear detailed English. Please type up your solutions or write very neatly.

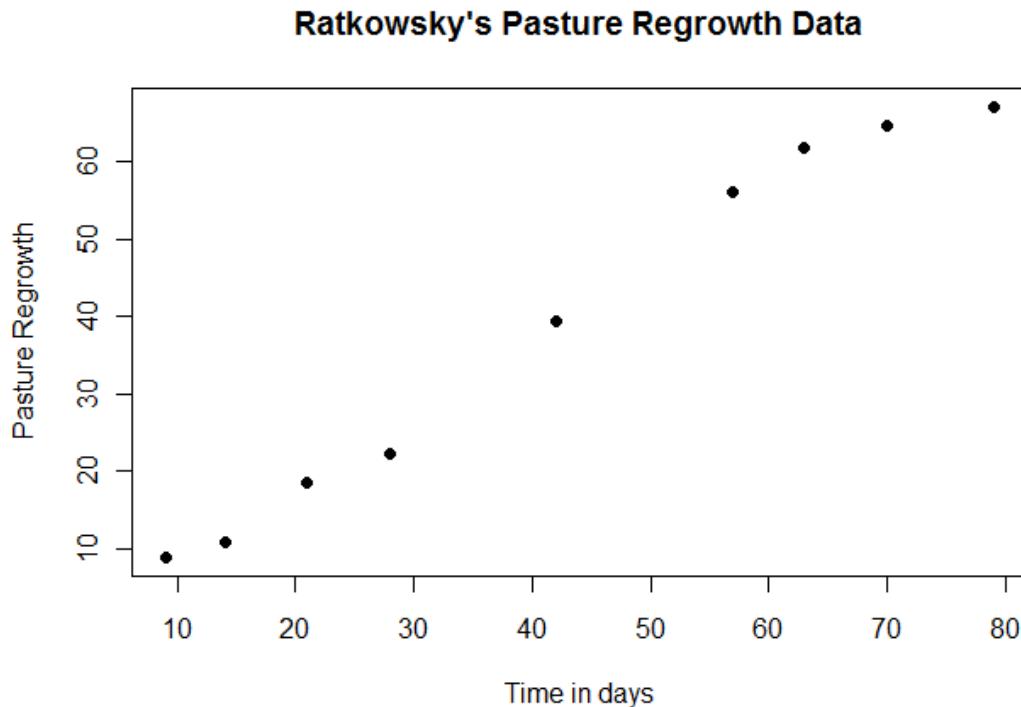
1. Huet, Bouvier, et al (*Statistical Tools for Nonlinear Regression*, p.2) use the *Pasture Regrowth* dataset from Ratkowsky (*Nonlinear Regression*, p.88) to fit a certain four-parameter sigmoidal growth model. In the dataset,  $Y$  = pasture regrowth since last grazing, and  $X$  = time, and for our present purposes, let's assume that the data are independent measurements. The nonlinear model function that these authors used to fit the data is somewhat complicated – and coming up with starting values for the model parameters is not easy and comes only come after we understand the roles these parameters play. These data are given, plotted, and analyzed in the Appendix.
  - (a) List all the needed assumptions for the given `proc nlin` analysis. Give an example of conditions where – in the context of this situation – the above required independent-measurements assumption would not be met.
  - (b) After examining SAS Program B (`proc nlin`), write down the assumed 4-parameter model function that the researchers fit to the data; see the right-hand side of the `model` statement.
  - (c) Assuming that  $\theta_4$  is positive and using algebra and one ‘limit’, clearly give the roles of the  $\theta_1$  and  $\theta_2$  parameters. (Hint: Which parameters – or functions of parameters – are the upper and lower asymptotes for this model?) Upon examining the graph of the data below, what are your “eyeball estimates” of these two parameter values?
  - (d) To obtain NLIN starting values for  $\theta_3$  and  $\theta_4$ , we use the following approach: write down the expression with ‘y’ on the left-hand side and the above assumed nonlinear model function on the right-hand side (with no error term for now), substitute in our eyeball estimates for the upper and lower asymptotes and solve so that the new right-hand expression is a linear model in  $\log(x)$ . Next, relate what you have found above to the simple linear regression (`proc reg`) performed in SAS Program A, and use SAS Output A to report the starting values for  $\theta_3$  and  $\theta_4$ . Verify that these starting values (or approximations to these) are used in SAS Program B.
  - (e) Using SAS Output B (`proc nlin`), report the estimate of  $\sigma^2$  here.
  - (f) Using SAS Output B (`proc nlin`), do a two-sided Wald test that  $\theta_4 = 3$  using  $\alpha = 1\%$ . Redo this two-sided Wald test using  $\alpha = 5\%$ . Clearly report your test statistics, p-values, and conclusions in both cases.
  - (g) Repeat both tests done in part (f) but using Likelihood Ratio tests instead.
  - (h) In examining the listing of the residuals in Output C and the Residual Plot, it is apparent that one of the residuals (at  $x = 21$ ) may be ‘large’. If the `proc nlin` were to be rerun with this potential outlier removed, would the estimate of the lower asymptote increase or decrease?
2. In *Nonlinear Regression Analysis and its Applications* (1988, p.269), Bates and Watts report data from Treloar (1974) regarding the “velocity” of an enzymatic reaction. The number of counts per minute of radioactive product from the reaction was measured as a function of substrate concentration (in ppm), and from these counts the initial rate, or “velocity,” of the reaction was calculated (in counts/min<sup>2</sup>). The experiment was conducted with the enzyme treated with puromycin (variable ‘treat’ = “yes”) and again with the enzyme untreated (‘treat’ = “no”). The velocity is assumed to depend on the substrate concentration according to the usual Michaelis-Menton ( $MM_2$ ) equation. In the word of the authors, it has been hypothesized that the “ultimate velocity parameter” ( $\theta_1$ ) should be affected by introduction of the Puromycin, but not necessarily the “half-velocity parameter” ( $\theta_2$ ). Here,  $Y$  = velocity and  $X$  = concentration.

- (a) Clearly list all the needed assumptions for the given `proc nlin` analyses (see the Appendix) in the context of this situation/exercise.
- (b) Write down the model function which is fit in SAS Program A. Clearly indicate the roles of  $\theta_3$  and  $\theta_4$  used in this model function.
- (c) Based on the output, give estimates for the  $MM_2$  model parameters (upper asymptote and  $LD_{50}$ ) for both the treated and untreated curves. Report these estimates in a  $2 \times 2$  table.
- (d) Test both hypotheses indicated by the authors' claims (above) using one-at-a-time Wald hypothesis tests. In both cases – and using the model in SAS Program A – report the hypotheses, test statistics, degrees of freedom, p-values, and conclusions.
- (e) Using the full-and-reduced (likelihood-based) F-test, test whether the half-velocity parameters are equal, reporting the calculated test statistic, degrees of freedom, p-value, and your conclusion.
- (f) Using the full-and-reduced (likelihood-based) F-test – and assuming the half-velocity parameters are indeed the same – test whether the ultimate velocity parameters are equal, reporting the calculated test statistic, degrees of freedom, p-value, and your conclusion.
- (g) Finally, compare the model function in Program/Output A with that in Program/Output D – are they equivalent? Why/why not? In what way are the approaches different? In which situation(s) is Program/Output A preferred, and in which situation(s) is Program/Output D preferred? Be clear in your explanation.
3. In “Calibration and assay development using the four-parameter logistic curve” (*Chem. Intell. Lab. Systems*, 1993, p.97), O’Connell *et al* fit the four-parameter log-logistic (LL4) model function to their radioimmunoassay (RIA) data. The data are analyzed in SAS in the Appendix using one run of `proc nlin` and then three runs of `proc nlmixed`.
- (a) Clearly list all the needed assumptions for the run of `proc nlin` and the three runs of `proc nlmixed` in the context of this exercise/situation.
- (b) Looking at the residual plot of the `proc nlin` fit in Output A, comment on whether all necessary assumptions appear to be met. (Note that the `proc nlmixed` fit in Output C fits this same homoskedastic or constant variance nonlinear model as in this `proc nlin`).
- (c) In Output B, researchers are trying to obtain a good model for the model variance for these data. Explain what is being done in the `proc nlmixed` run in Output D: what is the ‘model’ and what are the roles of the new parameter(s). Perform a likelihood-based test of whether the extra parameter ( $\rho = \rho$ ) in the variance is needed, writing out your hypotheses, test statistic, degrees of freedom, p-value, and clear conclusion. Does this conclusion seem sensible considering the residual plot on p.3?
- (d) It turns out that the `proc nlmixed` in Output E involves another – potentially more appropriate but also more complicated – way of modeling the variance for these data. Comparing Outputs C and E, perform a likelihood-based hypothesis test testing for homoskedasticity, again writing out your hypotheses, test statistic, degrees of freedom, p-value, and clear conclusion.
- (e) Compare the parameter estimates for the model parameters (the  $\theta$ s) and – more importantly – the associated standard errors for Outputs C and E. What has changed and how?
4. Simulated data are graphed and examined in the Appendix starting on p.18 using two runs of `proc nlin`.
- (a) Clearly list all the needed assumptions for the first `proc nlin` analysis.

- (b) Write down the model function being fit in Output A. What is the precise relevance/function of the parameter named 'phi' (denoted  $\phi$ )?
- (c) Write down the OLS estimate of 'phi'. Also, perform a two-sided WALD test that the true value of 'phi' is equal to  $-0.40$  (using  $\alpha = 5\%$ ), and clearly indicate the 95 % Wald Confidence Interval (WCI) for this parameter. For the test, give the hypotheses, test statistic, p-value, and degrees of freedom.
- (d) Repeat part (c) using the likelihood test. Remember to give the hypotheses, test statistic, p-value, and degrees of freedom.
- (e) It turns out that the 95% likelihood-based confidence interval for 'phi' for these data is very different from the reported 95% WCI. In your opinion, is this true CI shifted to the right or to the left of the WCI? Clearly and succinctly explain your answer, making reference to the plot of the data and fitted curve.

### Homework 3 Appendix

#### Exercise 1 Graph



#### Exercise 1 SAS Program/Output A

```
data one;
  do x=9,14,21,28,42,57,63,70,79;
    input y @@;
    y=y/100;
    output;
  end;
datalines;
893 1080 1859 2233 3935 5611 6173 6462 6708
;
data two;
  set one;
  ny=log(-(log((70-y)/65)));
  nx=log(x);
proc reg data=two;
  model ny=nx;
run;
```

```
The REG Procedure
Model: MODEL1
Dependent Variable: ny
```

Number of Observations Read	9
Number of Observations Used	9

Analysis of Variance					
Source	DF	Sum of Squares	Mean Square	F Value	Pr > F
Model	1	16.87291	16.87291	550.90	<.0001
Error	7	0.21439	0.03063		
Corrected Total	8	17.08730			
Root MSE		0.17501	R-Square	0.9875	
Dependent Mean		-0.53872	Adj R-Sq	0.9857	
Coeff Var		-32.48571			
Parameter Estimates					
Variable	DF	Parameter Estimate	Standard Error	t Value	Pr >  t
Intercept	1	-7.18999	0.28932	-24.85	<.0001
nx	1	1.88511	0.08032	23.47	<.0001

### Exercise 1 SAS Program/Output B

```
proc nlin data=one;
parms th1=70 th2=65 th3=-7 th4=2;
model y=th1-th2*exp(-exp(th3)*(x**th4));
output out=three r=resids p=preds;
run;
```

The NLIN Procedure					
Dependent Variable y					
Method: Gauss-Newton					
Iterative Phase					
Iter	th1	th2	th3	th4	Sum of Squares
0	70.0000	65.0000	-7.0000	2.0000	832.1
1	68.9082	64.3067	-6.1978	1.5712	276.5
2	63.5647	52.7166	-9.3162	2.4354	85.3555
3	69.8974	61.6943	-9.0989	2.3440	9.7022
4	69.9211	61.6539	-9.2072	2.3777	8.3768
5	69.9575	61.6846	-9.2082	2.3776	8.3759
6	69.9552	61.6815	-9.2089	2.3778	8.3759
7	69.9552	61.6815	-9.2089	2.3778	8.3759

NOTE: Convergence criterion met.

Estimation Summary					
Method		Gauss-Newton			
Iterations		7			
Observations Read		9			
Observations Used		9			
Observations Missing		0			

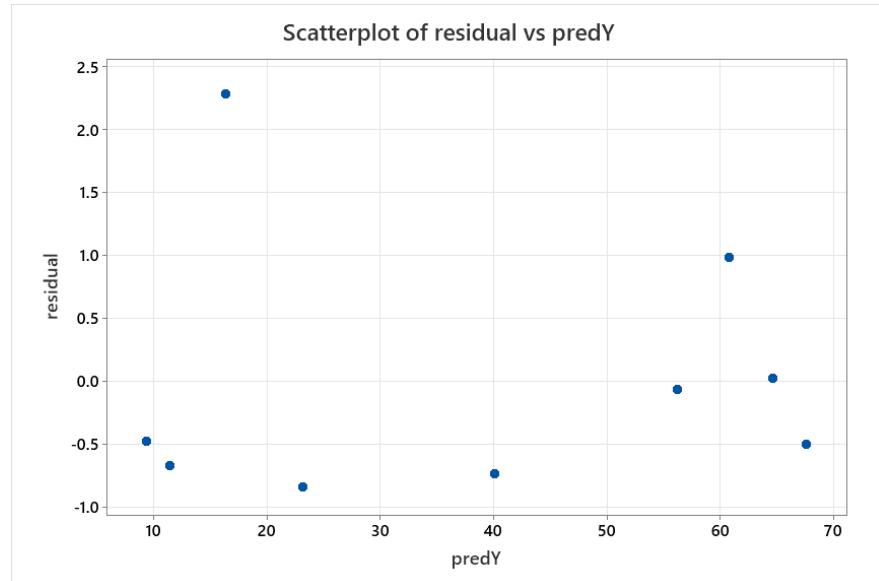
Source	DF	Sum of Squares	Mean Square	F Value	Approx Pr > F
Model	1	16.87291	16.87291	550.90	<.0001
Error	7	0.21439	0.03063		
Corrected Total	8	17.08730			

Model	3	4639.7	1546.6	923.22	<.0001
Error	5	8.3759	1.6752		
Corrected Total	8	4648.1			
<b>Approx</b>					
Parameter	Estimate	Std Error	Approximate	95% Confidence Limits	
th1	69.9552	2.3620	63.8835	76.0269	
th2	61.6815	3.1927	53.4744	69.8885	
th3	-9.2089	0.8173	-11.3098	-7.1080	
th4	2.3778	0.2210	1.8098	2.9459	
<b>Approximate Correlation Matrix</b>					
	th1	th2	th3	th4	
th1	1.0000000	0.9251613	0.7095438	-0.7658736	
th2	0.9251613	1.0000000	0.8615146	-0.8906628	
th3	0.7095438	0.8615146	1.0000000	-0.9955752	
th4	-0.7658736	-0.8906628	-0.9955752	1.0000000	

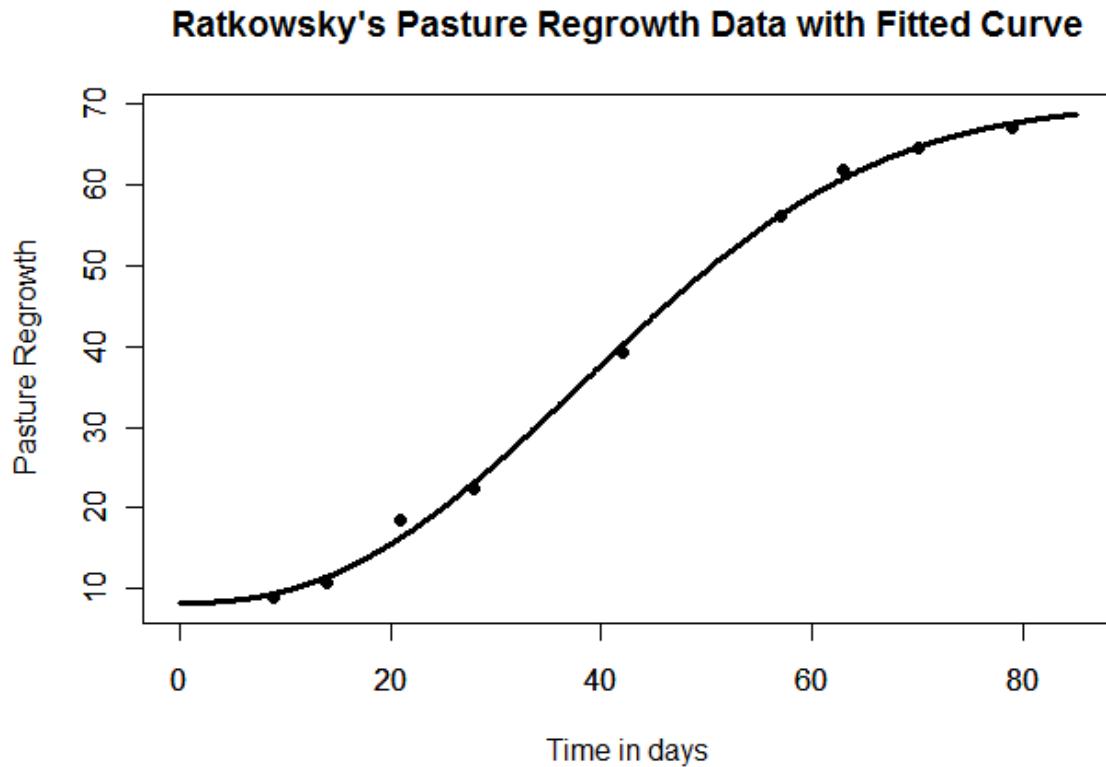
### Exercise 1 SAS Program and Output C

proc print noobs; run;	x	y	preds	resids
	9	8.93	9.4107	-0.48069
	14	10.80	11.4693	-0.66931
	21	18.59	16.3057	2.28432
	28	22.33	23.1737	-0.84374
	42	39.35	40.0846	-0.73458
	57	56.11	56.1766	-0.06655
	63	61.73	60.7442	0.98581
	70	64.62	64.5949	0.02506
	79	67.08	67.5803	-0.50032

### Exercise 1 SAS Residual Plot



## Exercise 1 Fitted Nonlinear Model and Data



## Exercise 1 SAS Program/Output D

```
proc nlin data=one;
  parms th1=70 th2=55 th3=-12;
  th4=3;
  model y=th1-th2*exp(-exp(th3)*(x**th4));
run;
```

The NLIN Procedure  
Dependent Variable y  
Method: Gauss-Newton

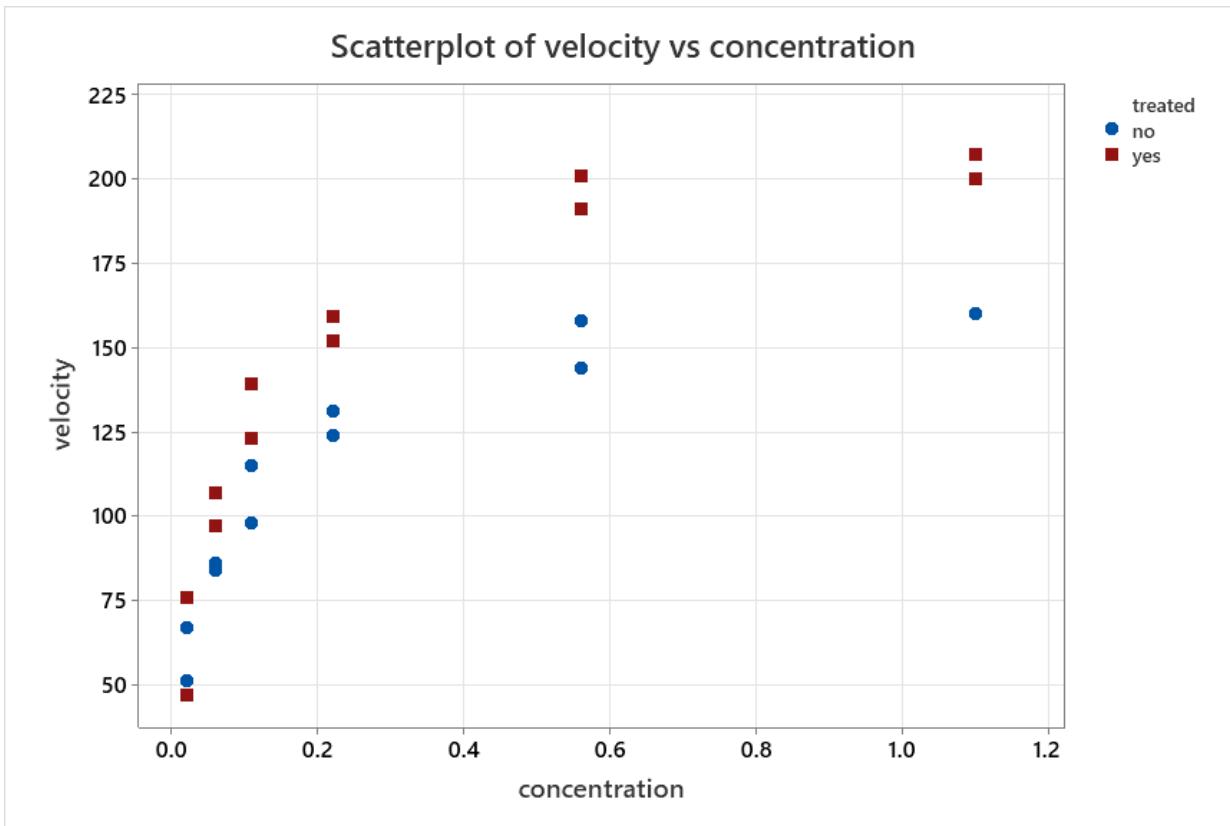
Iter	Iterative Phase			Sum of Squares
	th1	th2	th3	
0	70.0000	55.0000	-12.0000	112.9
1	64.8344	53.7243	-11.5162	28.9029
2	66.7202	56.0710	-11.5451	19.2408
3	66.7573	56.0959	-11.5467	19.2384
4	66.7590	56.0969	-11.5469	19.2384
5	66.7592	56.0969	-11.5469	19.2384

NOTE: Convergence criterion met.

Estimation Summary

Method						Gauss - Newton	
Iterations						5	
Observations Read						9	
Observations Used						9	
Observations Missing						0	
Source		DF	Sum of Squares	Mean Square	F Value	Pr > F	Approx
Model		2	4628.8	2314.4	721.81	<.0001	
Error		6	19.2384	3.2064			
Corrected Total		8	4648.1				
Approx							
Parameter		Estimate	Std Error	Approximate 95% Confidence Limits			
th1		66.7592	1.4695	63.1633	70.3550		
th2		56.0969	1.6005	52.1806	60.0133		
th3		-11.5469	0.0999	-11.7912	-11.3026		
Approximate Correlation Matrix							
th1		th1	th2	th3			
th1		1.0000000	0.7565905	-0.7648290			
th2		0.7565905	1.0000000	-0.3861020			
th3		-0.7648290	-0.3861020	1.0000000			

## Exercise 2 Graph and Listing of Data



conc	treat	dummy	dumyes	dumno	velocity
0.02	no	0	0	1	67
0.02	no	0	0	1	51
0.06	no	0	0	1	84
0.06	no	0	0	1	86
0.11	no	0	0	1	98
0.11	no	0	0	1	115
0.22	no	0	0	1	131
0.22	no	0	0	1	124
0.56	no	0	0	1	144
0.56	no	0	0	1	158
1.10	no	0	0	1	160
0.02	yes	1	1	0	76
0.02	yes	1	1	0	47
0.06	yes	1	1	0	97
0.06	yes	1	1	0	107
0.11	yes	1	1	0	123
0.11	yes	1	1	0	139
0.22	yes	1	1	0	159
0.22	yes	1	1	0	152
0.56	yes	1	1	0	191
0.56	yes	1	1	0	201
1.10	yes	1	1	0	207
1.10	yes	1	1	0	200

## Exercise 2 SAS Program and Output A

```
proc nlin data=one;
parms th1=150 th2=0.10 th3=0 th4=0;
model velocity=((th1+th3*dummy)*conc)/(th2+th4*dummy+conc);
run;
```

The NLIN Procedure  
 Dependent Variable velocity  
 Method: Gauss-Newton

Iter	Iterative Phase				Sum of Squares
	th1	th2	th3	th4	
0	150.0	0.1000	0	0	45433.4
1	158.5	0.0239	53.5635	0.0151	8794.1
2	155.1	0.0376	53.7329	0.0182	2379.9
3	158.9	0.0453	53.0719	0.0175	2066.1
4	160.0	0.0473	52.5673	0.0167	2055.3
5	160.2	0.0476	52.4332	0.0165	2055.1
6	160.3	0.0477	52.4085	0.0164	2055.1
7	160.3	0.0477	52.4044	0.0164	2055.1
8	160.3	0.0477	52.4038	0.0164	2055.1

NOTE: Convergence criterion met.

#### Estimation Summary

Method	Gauss-Newton
Iterations	8
Observations Read	23
Observations Used	23
Observations Missing	0

NOTE: An intercept was not specified for this model.

Source	DF	Sum of Squares	Mean Square	F Value	Approx Pr > F
Model	4	417562	104390	965.14	<.0001
Error	19	2055.1	108.2		
Uncorrected Total	23	419617			

#### Approx

Parameter	Estimate	Std Error	Approximate	95% Confidence Limits
th1	160.3	6.8960	145.8	174.7
th2	0.0477	0.00828	0.0304	0.0650
th3	52.4038	9.5510	32.4135	72.3942
th4	0.0164	0.0114	-0.00751	0.0403

#### Approximate Correlation Matrix

	th1	th2	th3	th4
th1	1.0000000	0.7768268	-0.7220184	-0.5628691
th2	0.7768268	1.0000000	-0.5608833	-0.7245748
th3	-0.7220184	-0.5608833	1.0000000	0.7712219
th4	-0.5628691	-0.7245748	0.7712219	1.0000000

### Exercise 2 SAS Program and Output B

```
proc nlin data=one;
  parms th1=150 th2=0.10 th3=0;
  th4=0;
  model velocity=((th1+th3*dummy)*conc)/(th2+th4+conc);
run;
```

The NLIN Procedure  
Dependent Variable velocity  
Method: Gauss-Newton

#### Iterative Phase

Iter	th1	th2	th3	Sum of Squares
0	150.0	0.1000	0	45433.4
1	161.6	0.0321	47.9824	9612.9
2	162.3	0.0482	41.2622	2595.2
3	165.7	0.0561	41.6938	2251.4
4	166.5	0.0577	41.9696	2241.1

5	166.6	0.0579	42.0189	2240.9
6	166.6	0.0580	42.0251	2240.9
7	166.6	0.0580	42.0259	2240.9
8	166.6	0.0580	42.0260	2240.9

NOTE: Convergence criterion met.

#### Estimation Summary

Method	Gauss-Newton
Iterations	8
Observations Read	23
Observations Used	23
Observations Missing	0

NOTE: An intercept was not specified for this model.

Source	DF	Sum of Squares	Mean Square	F Value	Approx Pr > F
Model	3	417376	139125	1241.70	<.0001
Error	20	2240.9	112.0		
Uncorrected Total	23	419617			

Parameter	Estimate	Std Error	Approximate	95% Confidence Limits
th1	166.6	5.8074	154.5	178.7
th2	0.0580	0.00591	0.0456	0.0703
th3	42.0260	6.2721	28.9426	55.1093

#### Approximate Correlation Matrix

	th1	th2	th3
th1	1.0000000	0.6112817	-0.5405580
th2	0.6112817	1.0000000	0.0644066
th3	-0.5405580	0.0644066	1.0000000

### Exercise 2 SAS Program and Output C

```
proc nlin data=one;
  parms th1=150 th2=0.10;
  th3=0;
  model velocity=((th1+th3*dummy)*conc) / (th2+conc);
run;
```

The NLIN Procedure  
 Dependent Variable velocity  
 Method: Gauss-Newton

#### Iterative Phase

Iter	th1	th2	Sum of Squares
0	150.0	0.1000	45433.4
1	190.7	0.0398	11454.1
2	187.5	0.0536	7424.5
3	190.1	0.0591	7280.7

4	190.7	0.0602	7276.6
5	190.8	0.0604	7276.5
6	190.8	0.0604	7276.5
7	190.8	0.0604	7276.5
NOTE: Convergence criterion met.			
Source	DF	Sum of Squares	Mean Square
Model	2	412340	206170
Error	21	7276.5	346.5
Uncorrected Total	23	419617	
Parameter	Estimate	Std Error	Approximate 95% Confidence Limits
th1	190.8	8.7646	172.6 209.0
th2	0.0604	0.0108	0.0380 0.0828
Approximate Correlation Matrix			
th1	1.0000000	0.7757154	
th2	0.7757154	1.0000000	

Exercise 2 SAS Program and Output D (after introducing TWO dummy variables in the dataset for puromycin YES called 'dumyes' and for puromycin NO called 'dumno')

```

proc nlin data=one;
parms th1yes=150 th1no=150 th2yes=0.10 th2no=0.10;
th1=th1yes*dumyes+th1no*dumno;
th2=th2yes*dumyes+th2no*dumno;
model velocity=(th1*conc) / (th2+conc);
run;

```

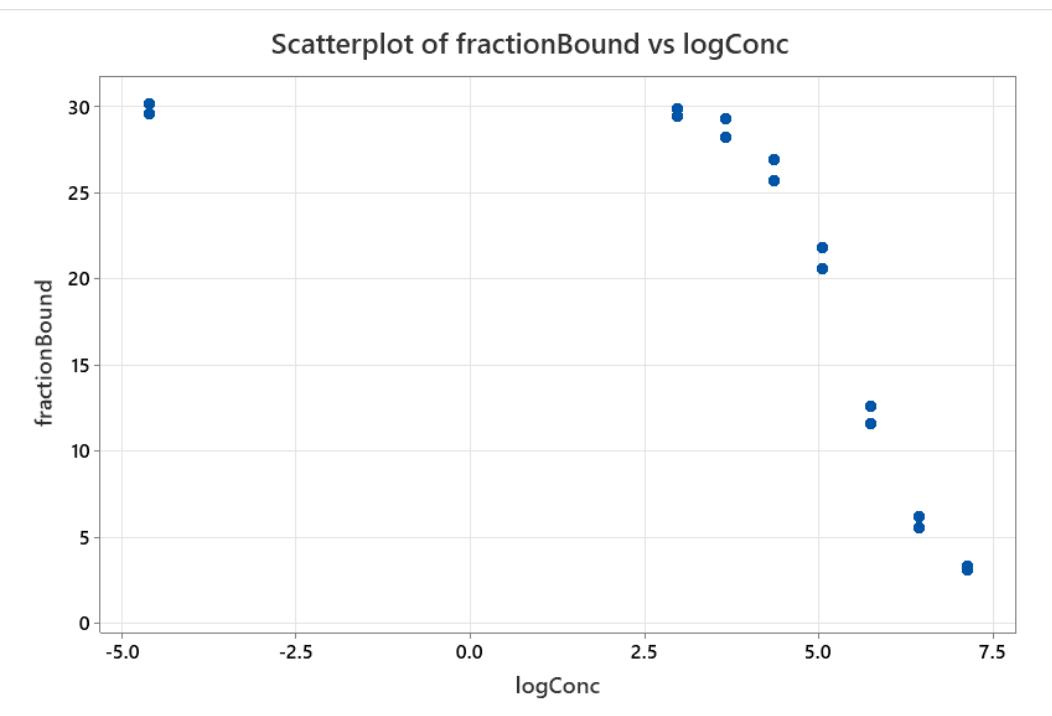
The NLIN Procedure  
 Dependent Variable velocity  
 Method: Gauss-Newton

Iter	Iterative Phase				Sum of Squares
	th1yes	th1no	th2yes	th2no	
0	150.0	150.0	0.1000	0.1000	45433.4
1	212.0	158.5	0.0390	0.0239	8794.1
2	208.8	155.1	0.0558	0.0376	2379.9
3	212.0	158.9	0.0628	0.0453	2066.1
4	212.6	160.0	0.0640	0.0473	2055.3
5	212.7	160.2	0.0641	0.0476	2055.1
6	212.7	160.3	0.0641	0.0477	2055.1
7	212.7	160.3	0.0641	0.0477	2055.1
8	212.7	160.3	0.0641	0.0477	2055.1

NOTE: Convergence criterion met.

Estimation Summary					
Method	Gauss-Newton				
Iterations	8				
Observations Read	23				
Observations Used	23				
Observations Missing	0				
NOTE: An intercept was not specified for this model.					
Source	DF	Sum of Squares	Mean Square	F Value	Approx Pr > F
Model	4	417562	104390	965.14	<.0001
Error	19	2055.1	108.2		
Uncorrected Total	23	419617			
Parameter	Estimate	Std Error	Approximate 95% Confidence Limits		
th1yes	212.7	6.6081	198.9	226.5	
th1no	160.3	6.8960	145.8	174.7	
th2yes	0.0641	0.00788	0.0476	0.0806	
th2no	0.0477	0.00828	0.0304	0.0650	
Approximate Correlation Matrix					
	th1yes	th1no	th2yes	th2no	
th1yes	1.0000000	0.0000000	0.7650837	0.0000000	
th1no	0.0000000	1.0000000	0.0000000	0.7768268	
th2yes	0.7650837	0.0000000	1.0000000	0.0000000	
th2no	0.0000000	0.7768268	0.0000000	1.0000000	

### Exercise 3 Graph and Data Listing



conc	fraction_bound	log_conc
0.0	30.16	-4.60517
0.0	29.58	-4.60517
19.4	29.87	2.96579
19.4	29.43	2.96579
38.8	28.19	3.65868
38.8	29.33	3.65868
77.5	26.96	4.35041
77.5	25.72	4.35041
155.0	21.82	5.04349
155.0	20.59	5.04349
310.0	12.62	5.73660
310.0	11.57	5.73660
620.0	5.56	6.42974
620.0	6.17	6.42974
1240.0	3.33	7.12287
1240.0	3.07	7.12287

### Exercise 3 SAS Program/Output A (NLIN)

```

proc nlin data=one;
  parms th1=30 th2=0 th3=300 th4=2;
  if conc=0 then do;
    mean=th1;
  end;
  else do;
    t=(conc/th3)**th4;
    mean=th2+(th1-th2)/(1+t);
  end;
  model fraction_bound=mean;
  output out=two r=residual p=predicted_value;
run;
proc print noobs;
  var residual predicted_value;
run;

```

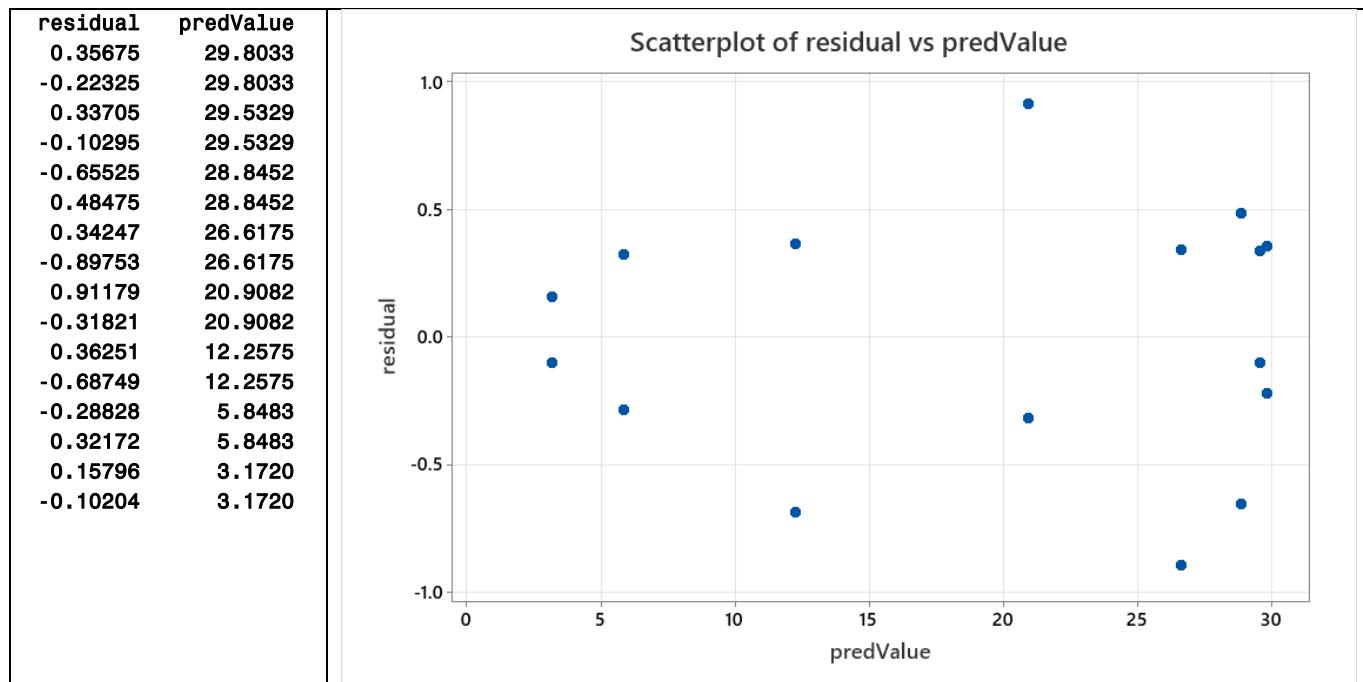
The NLIN Procedure  
Dependent Variable fraction\_bound

Source	DF	Sum of Squares		Mean Square	F Value	Pr > F	Approx
		Model	Error				
Model	3	1704.2		568.1	1869.08	<.0001	
Error	12	3.6471		0.3039			
Corrected Total	15	1707.8					

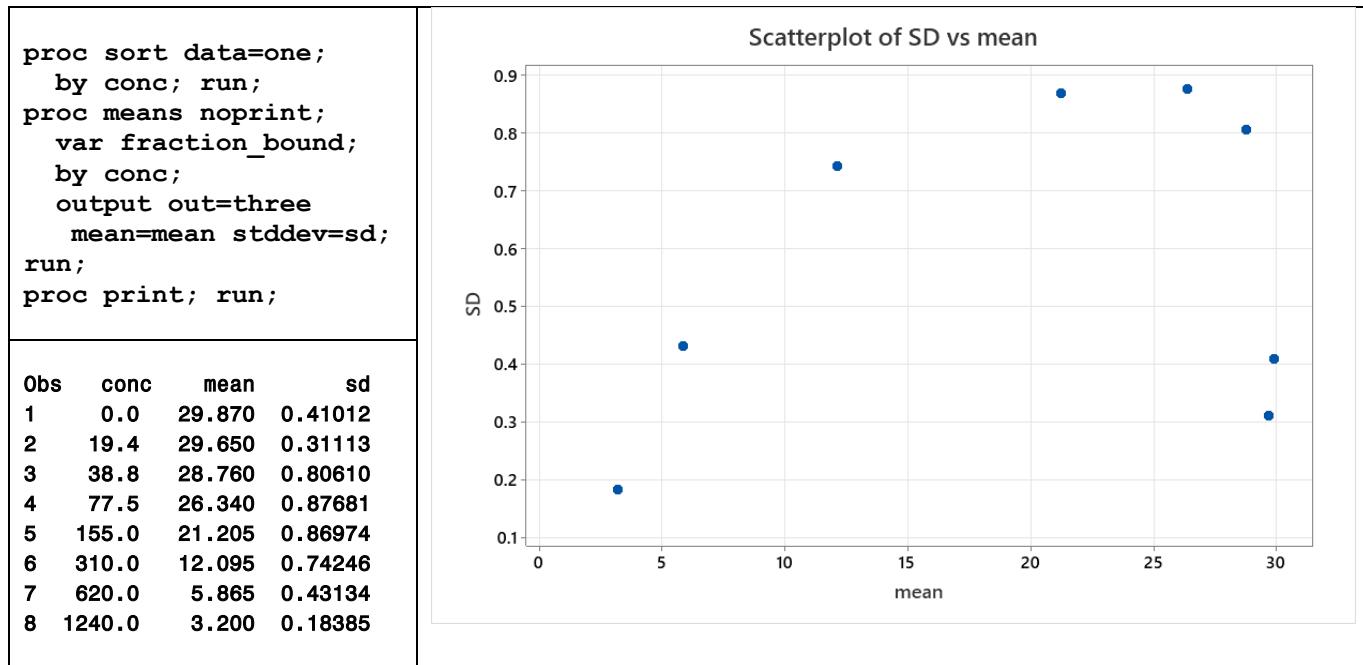
Parameter	Estimate	Std Error	Approx		
			Approximate	95% Confidence	Limits
th1	29.8033	0.2658	29.2242	30.3823	
th2	1.9929	0.5667	0.7582	3.2276	

th3	232.4	8.6272	213.6	251.2
th4	1.8619	0.1122	1.6174	2.1065

### Exercise 3 NLIN Residuals and Residual Plot



### Exercise 3 SAS Program/Output B



### Exercise 3 SAS Program and Output C

```

proc nlmixed data=one;
  parms th1=30 th2=0 th3=300 th4=2 sigma=1;
  if conc=0 then do;
    mean=th1;
  end;
  else do;
    t=(conc/th3)**th4; den=1+t;
    mean=th2+(th1-th2)/den;
  end;
  var=sigma*sigma;
  model fraction_bound ~ normal(mean,var);
run;

```

The NL MIXED Procedure

Fit Statistics

-2 Log Likelihood	21.7
AIC (smaller is better)	31.7
AICC (smaller is better)	37.7
BIC (smaller is better)	35.6

Parameter Estimates

Parameter	Standard		DF	t Value	Pr >  t	Alpha	Lower	Upper
	Estimate	Error						
th1	29.8033	0.2329	16	127.97	<.0001	0.05	29.3095	30.2970
th2	1.9929	0.4927	16	4.04	0.0009	0.05	0.9484	3.0374
th3	232.44	7.4175	16	31.34	<.0001	0.05	216.72	248.17
th4	1.8619	0.09942	16	18.73	<.0001	0.05	1.6512	2.0727
sigma	0.4774	0.08440	16	5.66	<.0001	0.05	0.2985	0.6564

### Exercise 3 SAS Program and Output D

```

proc nlmixed data=one;
  parms th1=30 th2=0 th3=300 th4=2 sigma=0.01 rho=0;
  if conc=0 then do;
    mean=th1;
  end;
  else do;
    t=(conc/th3)**th4; den=1+t;
    mean=th2+(th1-th2)/den;
  end;
  var=sigma*sigma*(mean)**rho;
  model fraction_bound ~ normal(mean,var);
run;

```

The NL MIXED Procedure

Fit Statistics

-2 Log Likelihood	19.4
-------------------	------

AIC (smaller is better)	31.4							
AICC (smaller is better)	40.7							
BIC (smaller is better)	36.0							
Parameter Estimates								
Standard								
Parameter	Estimate	Error	DF	t Value	Pr >  t	Alpha	Lower	Upper
th1	29.7840	0.2777	16	107.24	<.0001	0.05	29.1952	30.3728
th2	2.0618	0.2811	16	7.33	<.0001	0.05	1.4659	2.6577
th3	231.74	5.6829	16	40.78	<.0001	0.05	219.69	243.79
th4	1.8784	0.08705	16	21.58	<.0001	0.05	1.6939	2.0629
sigma	0.1223	0.09460	16	1.29	0.2144	0.05	-0.07822	0.3229
rho	0.9410	0.5501	16	1.71	0.1065	0.05	-0.2252	2.1071

### Exercise 3 SAS Program and Output E

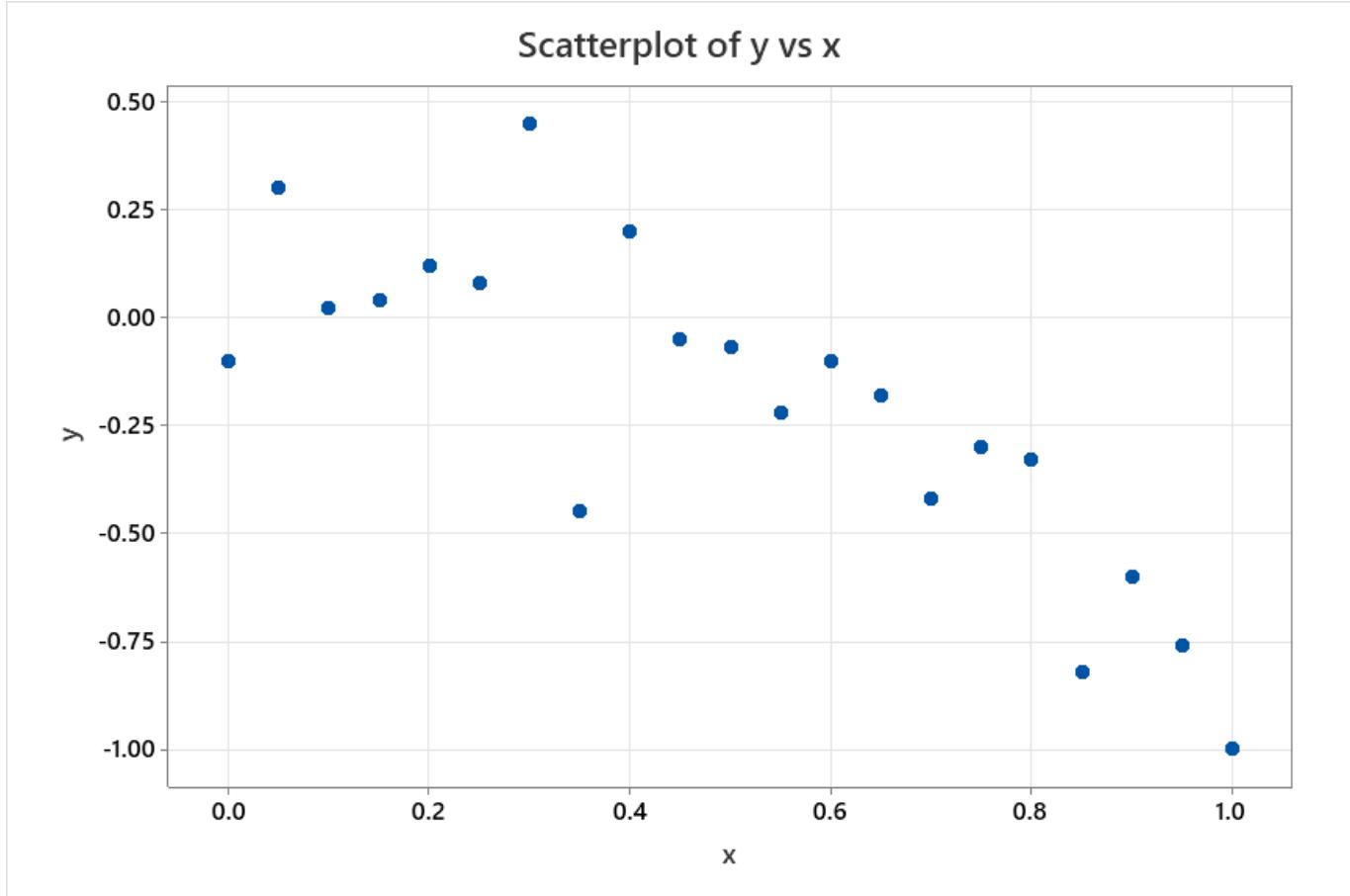
```

proc nlmixed data=one;
parms th1=30 th2=2 th3=250 th4=2 sigma=0.01 rho1=5 rho2=4;
if conc=0 then do;
  mean=th1;
end;
else do;
  t=(conc/th3)**th4; den=1+t;
  mean=th2+(th1-th2)/den;
end;
var=.00001*sigma*sigma*((mean)**rho1)*((1.1*th1-mean)**rho2);
model fraction_bound ~ normal(mean,var);
run;

```

The NL MIXED Procedure								
Fit Statistics								
-2 Log Likelihood								15.4
AIC (smaller is better)								29.4
AICC (smaller is better)								43.4
BIC (smaller is better)								34.8
Parameter Estimates								
Standard								
Parameter	Estimate	Error	DF	t Value	Pr >  t	Alpha	Lower	Upper
th1	29.8710	0.1771	16	168.64	<.0001	0.05	29.4955	30.2464
th2	1.9814	0.2295	16	8.63	<.0001	0.05	1.4948	2.4680
th3	230.04	6.7797	16	33.93	<.0001	0.05	215.67	244.42
th4	1.8332	0.08383	16	21.87	<.0001	0.05	1.6555	2.0109
sigma	0.2476	0.6171	16	0.40	0.6936	0.05	-1.0606	1.5557
rho1	2.9106	1.0545	16	2.76	0.0139	0.05	0.6751	5.1461
rho2	2.0161	0.9914	16	2.03	0.0589	0.05	-0.08554	4.1176

#### Exercise 4 Scatterplot of Data



#### Exercise 4 SAS Program/Output A

```
proc nlin;
  parms b0=0 b2=-0.5 phi=0.25;
  model y=b0-2*phi*b2*x+b2*x*x;
run;
```

The NLIN Procedure  
Dependent Variable y  
Method: Gauss-Newton

Iterative Phase				Sum of Squares
Iter	b0	b2	phi	
0	0	-0.5000	0.2500	2.3100
1	0.0424	-1.5529	0.0508	1.9069
2	0.0424	-1.5529	0.1859	0.6451

NOTE: Convergence criterion met.

Estimation Summary

Method						Gauss-Newton	
Iterations						2	
Observations Read						21	
Observations Used						21	
Observations Missing						0	
Source		DF	Sum of Squares	Mean Square	F Value	Pr > F	Approx
Model		2	2.1702	1.0851	30.28	<.0001	
Error		18	0.6451	0.0358			
Corrected Total		20	2.8153				
						Approx	
Parameter		Estimate	Std Error	Approximate 95% Confidence Limits			
b0		0.0424	0.1130	-0.1950		0.2798	
b2		-1.5529	0.5056	-2.6150		-0.4907	
phi		0.1859	0.1113	-0.0480		0.4197	
Approximate Correlation Matrix							
		b0	b2	phi			
b0		1.0000000	0.7084130	-0.8891903			
b2		0.7084130	1.0000000	-0.9188135			
phi		-0.8891903	-0.9188135	1.0000000			

#### Exercise 4 SAS Program/Output B

```
proc nlin;
  parms b0=0 b2=-0.5;
  phi=-0.40;
  model y=b0-2*phi*b2*x+b2*x*x;
run;
```

The NLIN Procedure  
 Dependent Variable y  
 Method: Gauss-Newton

Iterative Phase			Sum of Squares
Iter	b0	b2	
0	0	-0.5000	1.4278
1	0.2189	-0.5642	0.7852

NOTE: Convergence criterion met.

Estimation Summary	
Method	Gauss-Newton
Iterations	1
Observations Read	21
Observations Used	21
Observations Missing	0

Source	DF	Sum of Squares	Mean Square	F Value	Approx Pr > F
Model	1	2.0301	2.0301	49.13	<.0001
Error	19	0.7852	0.0413		
Corrected Total	20	2.8153			

Parameter	Estimate	Std Error	Approximate Confidence Limits		
			b0	b2	
b0	0.2189	0.0744	0.0633	0.3746	
b2	-0.5642	0.0805	-0.7327	-0.3957	

Approximate Correlation Matrix			
	b0	b2	
b0	1.0000000	-0.8026837	
b2	-0.8026837	1.0000000	