

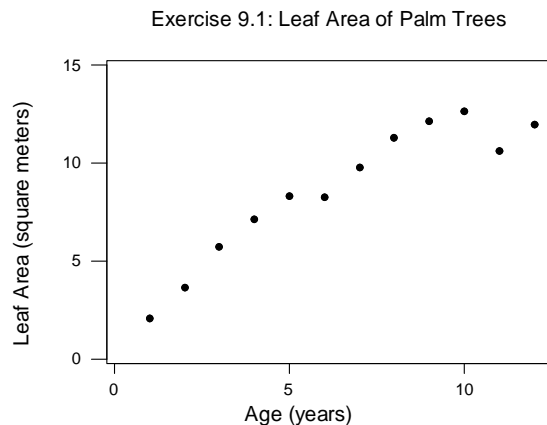
## CHAPTER 9

### A note on computing with SPSS (Version 11.5):

We use the SPSS software to fit the nonlinear regression models of Chapter 9. SPSS works like a spreadsheet program. We enter the data into the various columns of the spreadsheet and use the tabs: Analyze > Regression > Nonlinear. We write out the model equation and specify initial parameter values. We can save the fitted values and the residuals (also the derivatives of the objective function) into columns of the worksheet.

Several options for the iterative nonlinear estimation procedure are available. In the following examples we have used the Levenberg-Marquardt algorithm. Options for specifying the number of iterations and various convergence cutoffs are available. See the SPSS on-line help for further discussion and examples.

**9.1** A graph of the leaf area against the age of the palm tree is given below.



Note that there is not an abundance of data points to determine the model. The graph indicates that the relationship between leaf area and age is not linear; a quadratic component needs to be added to the model. The estimation results for the quadratic model  $y = \beta_0 + \beta_1 \text{Age} + \beta_2 \text{Age}^2 + \varepsilon$  (Minitab output) is shown below. The quadratic coefficient is clearly needed; the estimate of the coefficient for  $\text{Age}^{**2}$  is -0.09616, with a significant t-ratio of -4.95.

**Regression Analysis: Area (square meters) versus Age, Age\*\*2**

The regression equation is

$$\text{Area (square meters)} = -0.123 + 2.15 \text{ Age} - 0.0962 \text{ Age}^2$$

Predictor	Coef	SE Coef	T	P
Constant	-0.1234	0.7334	-0.17	0.870
Age	2.1496	0.2594	8.29	0.000
Age**2	-0.09616	0.01942	-4.95	0.001

S = 0.7096      R-Sq = 96.6%      R-Sq(adj) = 95.8%

Analysis of Variance

Source	DF	SS	MS	F	P
Regression	2	128.071	64.036	127.19	0.000
Residual Error	9	4.531	0.503		
Total	11	132.603			

Rasch/Sedlacek use the Gompertz model  $y = \mu + \varepsilon = \alpha \exp[-\beta \exp(-\gamma \text{Age})] + \varepsilon$  with parameters  $\alpha > 0, \beta > 0, \gamma > 0$ . Before fitting this model, we need to determine suitable starting values for the iterative nonlinear parameter estimation. The graph indicates that the saturation level for large values of Age is about 15. Hence a suitable starting value for  $\alpha$  is given by 15. For Age = 1, the response is about 2; for Age = 5, the response is roughly 7. The model equation implies  $-\beta \exp(-\gamma) = \ln(2/15)$  and  $-\beta \exp(-5\gamma) = \ln(7/15)$ . This implies  $\exp(4\gamma) = [\ln(2/15)]/[\ln(7/15)]$  and  $\gamma = \{\ln[\ln(2/15)]/\ln(7/15)\}/4 \approx 0.25$ . Finally,  $-\beta \exp(-\gamma) = \ln(2/15)$  and  $\beta = -\ln(2/15)\exp(\gamma) \approx 2.6$ . The starting values  $\alpha = 15, \beta = 2.6$  and  $\gamma = 0.25$  are used in the SPSS nonlinear regression routine. The (SPSS) outcome is given below:

Iteration	Residual SS	A	B	C
1	12.59000092	15.0000000	2.60000000	.250000000
1.1	15.64377972	11.3812687	2.34691685	.336739045
1.2	6.515778841	13.4641276	2.14122037	.271436482
2	6.515778841	13.4641276	2.14122037	.271436482
2.1	6.243186484	12.0109653	2.42204992	.359733910
3	6.243186484	12.0109653	2.42204992	.359733910
3.1	5.136619171	12.4921144	2.50012161	.359000316
4	5.136619171	12.4921144	2.50012161	.359000316
4.1	5.136518308	12.4937910	2.49764737	.358922047
5	5.136518308	12.4937910	2.49764737	.358922047
5.1	5.136518286	12.4936881	2.49773050	.358935226

Run stopped after 11 model evaluations and 5 derivative evaluations. Iterations have been stopped because the relative reduction between successive residual sums of squares is at most SSSCON = 1.000E-08

Nonlinear Regression Summary Statistics      Dependent Variable AREA

Source	DF	Sum of Squares	Mean Square
Regression	3	1023.43418	341.14473
Residual	9	5.13652	.57072
Uncorrected Total	12	1028.57070	
(Corrected Total)	11	132.60269	

R squared = 1 - Residual SS / Corrected SS = .96126

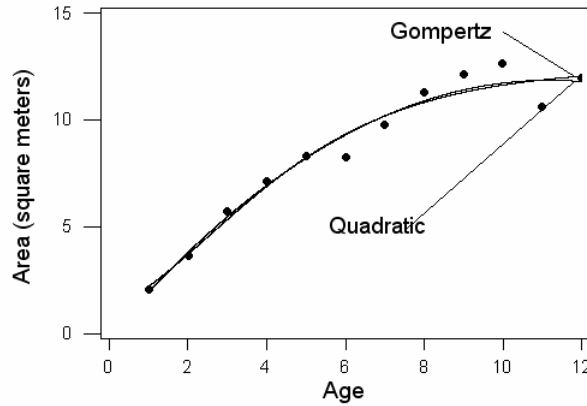
Parameter	Estimate	Asymptotic Std. Error	Asymptotic 95 % Confidence Interval	
			Lower	Upper
A ( $\alpha$ )	12.493688057	.683789772	10.946848127	14.040527986
B ( $\beta$ )	2.497730497	.440644079	1.500924338	3.494536656
C ( $\gamma$ )	.358935226	.066769083	.207893067	.509977385

Asymptotic Correlation Matrix of the Parameter Estimates

	A	B	C
A	1.0000	-.4983	-.8306
B	-.4983	1.0000	.8339
C	-.8306	.8339	1.0000

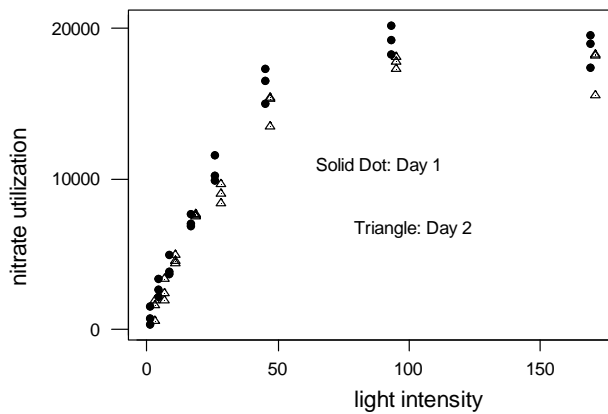
The estimate of  $\alpha$  is 12.5; the estimate of  $\beta$  is 2.5, and the estimate of  $\gamma$  is 0.36. All estimates are statistically significant. There is a fair amount of correlation, especially between the estimates of  $\gamma$  and  $\alpha$  (-0.83) and the estimates of  $\gamma$  and  $\beta$  (0.83). The coefficient of determination (0.961) is similar to the  $R^2$  from the quadratic regression. There is little difference between the fits of the quadratic regression (which is linear in the parameters) and the Gompertz model (which is nonlinear in the parameters). Both models lead to similar fitted curves. One difference is that the fitted values for the Gompertz model increase with age to an asymptotic value, whereas the quadratic curve starts to decrease with age after having reached a maximum. However, over the observed age range the two fitted models are virtually indistinguishable.

Exercise 9.1: Data and Fitted Models



**9.2** A scatter plot of nitrate utilization versus light intensity is shown below. We use solid circles for day 1 observations, and triangles for day 2 observations. Furthermore, we have added some jitter to the light intensity in order to emphasize the differences between the measurements of day 1 and day 2. The day 2 measurements are slightly lower, especially at increasing light intensity.

Exercise 9.2: Plot of nitrate utilization against light intensity



Michaelis-Menton model: Nitrate utilization reaches an asymptote of about 20,000 for large light intensity. Letting  $x$  go to infinity in the model equation

$$\frac{\beta_1 x}{\beta_2 + x} = \frac{\beta_1}{1 + (\beta_2 / x)} \approx 20,000$$

leads to the starting value  $\beta_1 \approx 20,000$ . Furthermore, the average nitrate utilization at light intensity 2.2 is 1075. Solving the model equation with  $\beta_1 = 20,000$  leads to the starting value  $\beta_2 = 38.7$ .

Using these starting values in the SPSS nonlinear regression routine results in the following estimation results:

Nonlinear Regression Summary Statistics      Dependent Variable NITRATE

Source	DF	Sum of Squares	Mean Square
Regression	2	6467226758.31	3233613379.15
Residual	46	96536195.6932	2098612.94985
Uncorrected Total	48	6563762954.00	
(Corrected Total)	47	2076766799.92	

R squared = 1 - Residual SS / Corrected SS = .95352

Parameter	Estimate	Asymptotic Std. Error	Asymptotic 95 % Confidence Interval	
			Lower	Upper
B1	23582.527043	889.35646658	21792.345325	25372.708760
B2	34.243774004	3.427314571	27.344947587	41.142600421

Asymptotic Correlation Matrix of the Parameter Estimates

	B1	B2
B1	1.0000	.8785
B2	.8785	1.0000

Exponential rise model: Nitrate utilization reaches an asymptote of about 20,000 for large light intensity. Letting  $x$  go to infinity in the equation for the exponential rise model leads to the starting value  $\beta_1 \approx 20,000$ . The average nitrate utilization at light intensity 2.2 is 1075. Solving the model equation with  $\beta_1 = 20,000$  leads to the

starting value  $\beta_2 = -\frac{1}{2.2} \ln \left[ 1 - \frac{1075}{20,000} \right] = 0.025$ . Using these starting values in the

SPSS nonlinear regression program results in the estimation results:

Nonlinear Regression Summary Statistics      Dependent Variable NITRATE

Source	DF	Sum of Squares	Mean Square
Regression	2	6504309173.87	3252154586.93
Residual	46	59453780.1310	1292473.48111
Uncorrected Total	48	6563762954.00	
(Corrected Total)	47	2076766799.92	

R squared = 1 - Residual SS / Corrected SS = .97137

Parameter	Estimate	Asymptotic Std. Error	Asymptotic 95 % Confidence Interval	
			Lower	Upper
B1	19014.305975	398.04663684	18213.079652	19815.532299
B2	.030021624	.001629334	.026741945	.033301303

Asymptotic Correlation Matrix of the Parameter Estimates

	B1	B2
B1	1.0000	-.7393
B2	-.7393	1.0000

Quadratic Michaelis-Menton model: Starting with  $\beta_1 = 20,000$  and  $\beta_2 = 38.7$  (from the earlier Michaelis-Menton model) and a small value for the parameter in the quadratic component ( $\beta_3 = 0.1$ ) leads to the following results:

Nonlinear Regression Summary Statistics      Dependent Variable NITRATE

Source	DF	Sum of Squares	Mean Square
Regression	3	6520540397.33	2173513465.78
Residual	45	43222556.6654	960501.25923
Uncorrected Total	48	6563762954.00	
(Corrected Total)	47	2076766799.92	

R squared = 1 - Residual SS / Corrected SS = .97919

Parameter	Estimate	Asymptotic Std. Error	Asymptotic 95 % Confidence Interval	
			Lower	Upper
B1	66769.034924	17585.504714	31350.010284	102188.05956
B2	137.82679758	43.735712594	49.738550634	225.91504453
B3	.011281055	.004496402	.002224837	.020337274

Asymptotic Correlation Matrix of the Parameter Estimates

	B1	B2	B3
B1	1.0000	.9964	.9941
B2	.9964	1.0000	.9856
B3	.9941	.9856	1.0000

Modified exponential rise model: Using  $\beta_1 = 20,000$  and  $\beta_2 = 0.025$  from the earlier exponential rise model and a small value for  $\beta_3 = 0.01$  leads to the following results:

Nonlinear Regression Summary Statistics      Dependent Variable NITRATE

Source	DF	Sum of Squares	Mean Square
Regression	3	6519117089.28	2173039029.76
Residual	45	44645864.7154	992130.32701
Uncorrected Total	48	6563762954.00	
(Corrected Total)	47	2076766799.92	

R squared = 1 - Residual SS / Corrected SS = .97850

Parameter	Estimate	Asymptotic Std. Error	Asymptotic 95 % Confidence Interval	
			Lower	Upper
B1	33551.454219	9502.1687711	14413.103896	52689.804543
B2	.018534079	.003572151	.011339397	.025728761
B3	.003221159	.001338559	.000525162	.005917155

Asymptotic Correlation Matrix of the Parameter Estimates

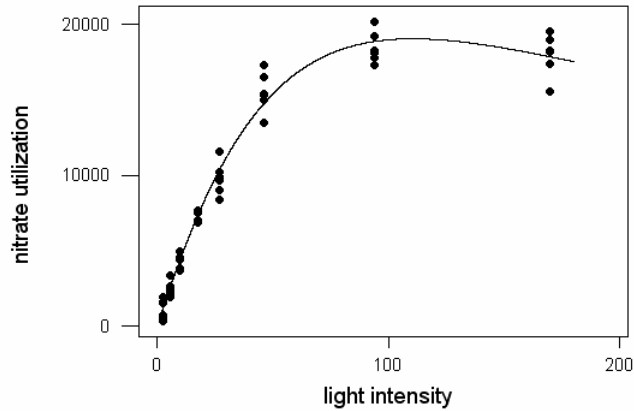
	B1	B2	B3
B1	1.0000	-.9898	.9948
B2	-.9898	1.0000	-.9741
B3	.9948	-.9741	1.0000

All four models lead to large  $R^2$ . The Michaelis-Menton and its quadratic extension lead to  $R^2$  of 0.954 and 0.979, respectively. Carrying out an F-test for the significance of the quadratic component in the Michaelis-Menton model leads to the F-statistic  $F = [96,536,195 - 43,222,556] / [43,222,556 / 45] = 55.5$ , which is highly significant. This shows that the quadratic extension represents a significant improvement.

Similarly, the exponential rise model and its extension lead to  $R^2$  of 0.971 and 0.979, respectively. The F-test for the significance of the extra component in the exponential rise model leads to the F-statistic  $F = [59,453,780 - 44,645,864] / [44,645,864 / 45] = 14.9$ , which is also highly significant.

The extensions are beneficial. The modified Michaelis-Menton and the modified exponential rise models perform similarly. In the following graph we show the fit of the quadratic Michaelis-Menton model; the fitted values of the modified exponential rise model are virtually indistinguishable.

Exercise 9.2: Fit of the quadratic Michaelis-Menten model



Standard Michaelis-Menten model with an indicator for the change of day: The final parameter estimates in the previous Michaelis-Menten model,  $\hat{\beta}_1 = 23,500$  and  $\hat{\beta}_2 = 34.2$ , are taken as the starting values in the iterative nonlinear estimation. Small values for the day indicator  $\alpha_1 = -1000, \alpha_2 = -1$  are used as the starting values for the two additional parameters. The estimation results are given below:

Nonlinear Regression Summary Statistics      Dependent Variable NITRATE

Source	DF	Sum of Squares	Mean Square
Regression	4	6477253424.57	1619313356.14
Residual	44	86509529.4274	1966125.66881
Uncorrected Total	48	6563762954.00	
(Corrected Total)	47	2076766799.92	

R squared = 1 - Residual SS / Corrected SS = .95834

Parameter	Estimate	Asymptotic Std. Error	Asymptotic 95 % Confidence Interval	
			Lower	Upper
B1	24743.334444	1241.1211323	22242.019158	27244.649730
B2	35.275400267	4.656586052	25.890667730	44.660132803
A1	-2328.743446	1720.3472191	-5795.875448	1138.3885567
A2	-2.172827290	6.626226364	-15.52710905	11.181454466



Asymptotic Correlation Matrix of the Parameter Estimates

	B1	B2	A1	A2
B1	1.0000	.8810	-.7214	-.6191
B2	.8810	1.0000	-.6356	-.7028
A1	-.7214	-.6356	1.0000	.8781
A2	-.6191	-.7028	.8781	1.0000

The F-statistic for testing the null hypothesis  $\alpha_1 = \alpha_2 = 0$  is  $F = [(96,536,195 - 86,509,529) / 2] / [86,509,529 / 44] = 2.55$ . The probability value from the F(2,44) distribution is  $P[F(2,44) \geq 2.55] = 1 - 0.91 = 0.09$ . Hence there is only weak evidence for including a day effect. The individual confidence intervals for  $\alpha_1$  and  $\alpha_2$  cover zero, which makes the individual interpretation of the two day-effect parameters difficult. These estimates are also quite correlated.

Quadratic Michaelis-Menton model with an indicator for the change of day: The final values from the earlier quadratic model  $\hat{\beta}_1 = 66,700, \hat{\beta}_2 = 138, \hat{\beta}_3 = 0.01$  and small values for the three parameters associated with the day indicators,  $\alpha_1 = -2000, \alpha_2 = -2, \alpha_3 = 0.001$ , are used as the starting values in the iterative nonlinear SPSS estimation. The estimation results are given below:

Run stopped after 10 model evaluations and 5 derivative evaluations. Iterations have been stopped because the relative reduction between successive residual sums of squares is at most SCON = 1.000E-08

Nonlinear Regression Summary Statistics      Dependent Variable NITRATE

Source	DF	Sum of Squares	Mean Square
Regression	6	6531740362.05	1088623393.67
Residual	42	32022591.9535	762442.66556
Uncorrected Total	48	6563762954.00	
(Corrected Total)	47	2076766799.92	

R squared = 1 - Residual SS / Corrected SS = .98458

Parameter	Estimate	Asymptotic Std. Error	Asymptotic 95 % Confidence Interval	
			Lower	Upper
B1	89797.916970	37540.345749	14038.432096	165557.40184
B2	186.61862445	89.984553967	5.022442558	368.21480635
A1	-38897.78690	39982.748874	-119586.2408	41790.667033
A2	-83.09078151	96.727346453	-278.2944695	112.11290653
B3	.016252421	.009207288	-.002328638	.034833481
A3	-.008449660	.009916211	-.028461384	.011562065

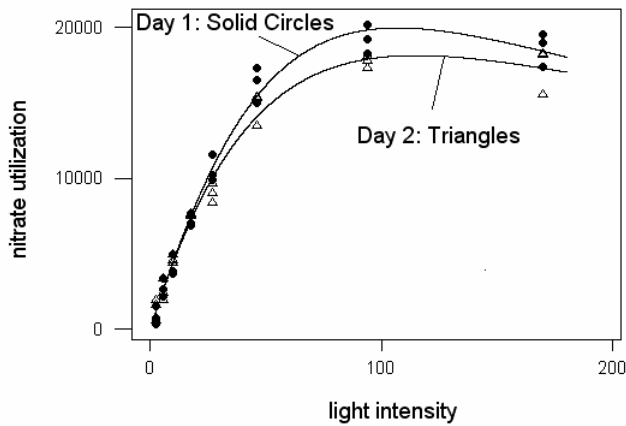
Asymptotic Correlation Matrix of the Parameter Estimates

	B1	B2	A1	A2	B3	A3
B1	1.0000	.9978	-.9389	-.9283	.9965	-.9252
B2	.9978	1.0000	-.9369	-.9303	.9913	-.9204
A1	-.9389	-.9369	1.0000	.9971	-.9356	.9953
A2	-.9283	-.9303	.9971	1.0000	-.9222	.9894
B3	.9965	.9913	-.9356	-.9222	1.0000	-.9285
A3	-.9252	-.9204	.9953	.9894	-.9285	1.0000

The F-statistic for testing the null hypothesis  $\alpha_1 = \alpha_2 = \alpha_3 = 0$  is given by  $F = [(43,222,556 - 32,022,591)/3]/[32,022,591/42] = 4.90$ . The probability value from the F(3,42) distribution is  $P[F(3,42) \geq 4.90] = 1 - 0.995 = 0.005$ , showing that the indicators for the day effect help explain the variation. Individually the three parameters are statistically insignificant and also highly correlated. This makes an individual interpretation of the estimates difficult.

The graph shown below compares the quadratic Michaelis-Menton model with and without the day indicator. The graph shows that the quadratic Michaelis-Menton model with a day indicator is capable of expressing the day differences.

Exercise 9.2: Quadratic Michaelis-Menton model with day indicator



### 9.3

Model 1: The logarithmic transformation of the first model leads to

$$\ln(y) = \ln(\beta_0) + \beta_1 \ln(x_1) + \beta_2 \ln(x_2) + \ln(\varepsilon)$$

A standard multiple linear regression of  $\ln(y)$  on  $\ln(x_1)$  and  $\ln(x_2)$  leads to the estimates of  $\alpha = \ln(\beta_0)$ ,  $\beta_1$ , and  $\beta_2$ . The estimate of  $\beta_0$  can be obtained

from  $\beta_0 = \exp(\alpha)$ . When carrying out the regression with the transformed variables we need to assume that the error  $\ln(\varepsilon)$  satisfies the standard regression assumptions.

Model 2: Taking the reciprocal of the response in the second model leads to

$$1/y = \beta_0 + \beta_1 x + \varepsilon$$

A simple linear regression of  $(1/y)$  on  $x_1$  leads to the estimates of  $\beta_0, \beta_1$ .

Model 3: The reciprocal of the response and a subsequent logarithmic transformation leads to the model

$$\ln[(1/y) - 1] = \beta_0 + \beta_1 x_1 + \ln(\varepsilon)$$

A simple linear regression of  $\ln[(1/y) - 1]$  on  $x_1$  leads to the estimates of  $\beta_0, \beta_1$ . We need to assume that the error  $\ln(\varepsilon)$  satisfies the standard regression assumptions.

#### 9.4 Search the literature.