

Directions: All students are to answer both 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. Diggle, Liang & Zeger (1994) and Venables and Ripley 1999, p.209 give an example of repeated measurements on the 'size' – defined as $\log(\text{height}) + 2 \log(\text{diameter})$ – of 79 Sitka spruce trees, 54 of which were grown in ozone-enriched chambers and 25 of which were controls. The trees were measured on eight different days in 1989 (see the SAS analysis in the attached *Appendix*). This analysis uses the classical linear mixed models analysis. Use the output to thoroughly analyze these repeated-measures data.
 - (a) For each part of the analysis (i.e., for each of the PROC MIXEDs), listing the necessary assumptions paying particular attention to the assumed form of the variance-covariance matrix.
 - (b) Identify and justify which one of the three covariance structures used in the PROC MIXEDs is most appropriate for these data. 'Justify' here means report the relevant test statistic(s) and results. Clearly specify how many variance components are estimated for each of the three runs of PROC MIXED.
 - (c) Using the appropriate analysis, comment on whether you feel the profiles for the two treatment curves are the same (some people call the curves "parallel"), reporting the relevant test statistic and p-value. Give your final conclusion.
2. On page 18 of Davidian & Giltinan, *Nonlinear Models for Repeated Measurement Data*, the authors present the data of Kwan et al (1976). Here, we fit the mixed nonlinear model of the form

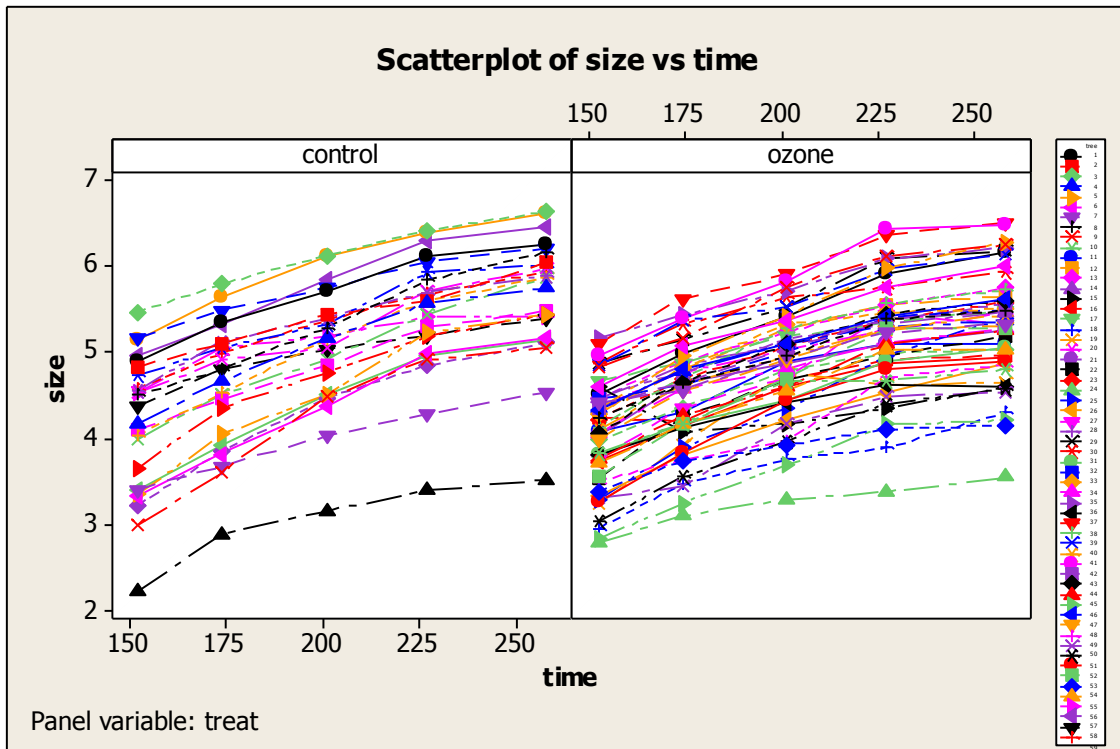
$$E(Y) = \beta_1 e^{-\beta_2 x} + \beta_3 e^{-\beta_4 x}$$

In this expression, Y = plasma concentration of the drug 'cefamandole' and x = time (in minutes post dosing). For this study, a dose of 15 mg (per kg body weight) of the drug was administered by ten-minute intravenous infusion to six healthy volunteers. The data are analyzed in SAS in the *Appendix*. The program first graphs the data then runs four runs of PROC NLMIXED. To understand the model function, the above four-parameter exponential (SE4) model function is graphed in the Appendix (p.9) using the values of $\beta_1 = 2.7733$, $\beta_2 = 2.8139$, $\beta_3 = 0.7870$, $\beta_4 = 0.4195$; also graphed are the two components, $[\beta_1 e^{-\beta_2 x}]$ and $[\beta_3 e^{-\beta_4 x}]$. For this model function, note that the 'upper asymptote' is $\beta_1 + \beta_3$; also β_2 is the initial slope parameter; and β_4 is the final slope parameter.

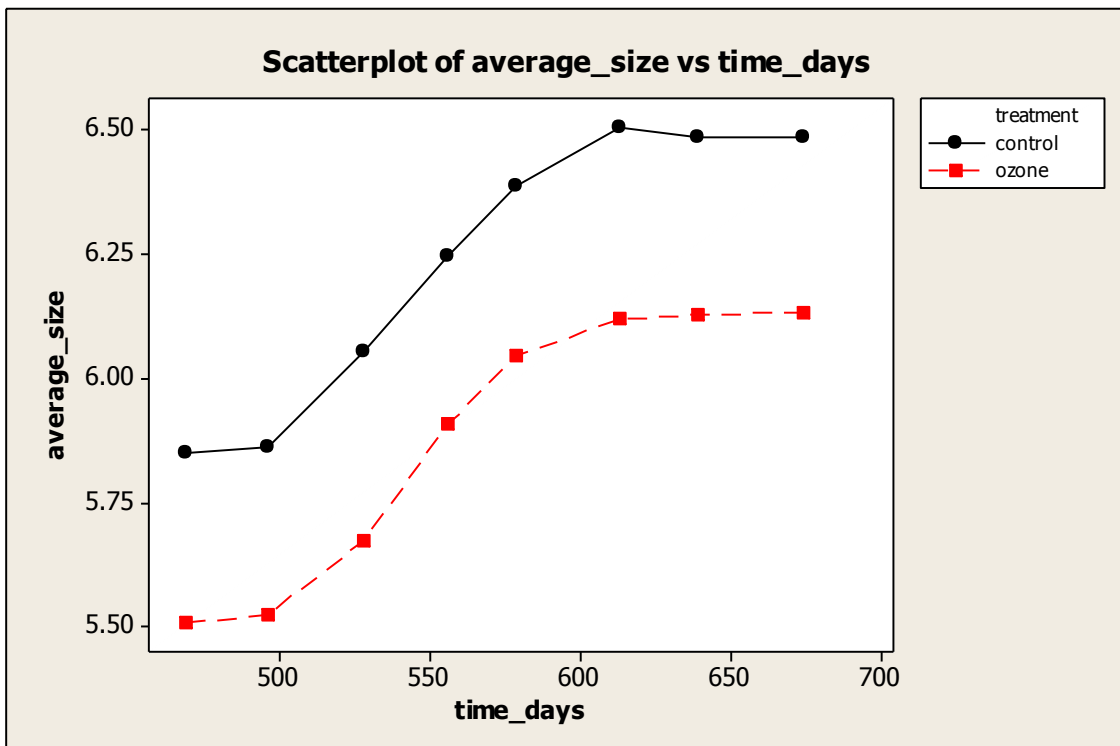
- (a) Listing all necessary assumptions, explain what is being done in each of the four runs of NLMIXED, commenting on which model is being fit and identifying the underlying assumptions. Be specific, and in each case write out the assumed covariance matrix for the model parameters. Also, identify which models are special cases of others (i.e., which are nested), and identify which they are nested in.
- (b) Choose the NLMIXED model/analysis which **best** describes these data, and give specific reasons for why you chose the model you did, and why you rejected each of the others. Give test statistics, degrees of freedom and p-values to justify your claims.
- (c) The parameters β_2 and β_4 are important since they address the rate of decrease (elimination) of the expected concentration function. Contrasting the estimated standard errors (SE's) of these two parameters for the first NLMIXED with those for each of the second through fourth NLMIXED's, why are these SEs lower for the latter three NLMIXED's than for the first NLMIXED?

Homework 13 Appendix

Graphs for First Exercise



Graphs of means for both treatments (average size versus time)



First Exercise – MIXED (CS) Approach

```
proc mixed data=sitka89;  
  title 'MIXED approach with CS error structure';  
  class treat tree time;  
  model size=treat time treat*time;  
  repeated / type=cs sub=tree(treat);  
run;
```

MIXED approach with CS error structure

The Mixed Procedure Model Information

Data Set	WORK.SITKA89
Dependent Variable	size
Covariance Structure	Compound Symmetry
Subject Effect	tree(treat)
Estimation Method	REML
Residual Variance Method	Profile
Fixed Effects SE Method	Model-Based
Degrees of Freedom Method	Between-Within

Class Level Information

Class	Levels	Values
treat	2	control ozone
tree	79	1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79
time	8	469 496 528 556 579 613 639 674

Dimensions

Covariance Parameters	2
Columns in X	27
Columns in Z	0
Subjects	79
Max Obs Per Subject	8

Number of Observations

Number of Observations Read	632
Number of Observations Used	632
Number of Observations Not Used	0

The Mixed Procedure

Iteration History

Iteration	Evaluations	-2 Res Log Like	Criterion
0	1	1260.72452373	
1	1	-614.89217445	0.00000000

Convergence criteria met.

Covariance Parameter Estimates

Cov Parm	Subject	Estimate
CS	tree(treat)	0.4033
Residual		0.009478

Fit Statistics	
-2 Res Log Likelihood	-614.9
AIC (smaller is better)	-610.9
AICC (smaller is better)	-610.9
BIC (smaller is better)	-606.2

Null Model Likelihood Ratio Test		
DF	Chi-Square	Pr > ChiSq
1	1875.62	<.0001

Type 3 Tests of Fixed Effects				
Effect	Num DF	Den DF	F Value	Pr > F
treat	1	77	5.30	0.0241
time	7	539	545.99	<.0001
treat*time	7	539	0.66	0.7071

First Exercise – MIXED (AR1) Approach

```
proc mixed data=sitka89;
  title 'MIXED approach with AR(1) error structure';
  class treat tree time;
  model size=treat time treat*time;
  repeated / type=ar(1) sub=tree(treat);
run;
```

MIXED approach with AR(1) error structure		
The Mixed Procedure		
Model Information		
Data Set	WORK.SITKA89	
Dependent Variable	size	
Covariance Structure	Autoregressive	
Subject Effect	tree(treat)	
Estimation Method	REML	
Residual Variance Method	Profile	
Fixed Effects SE Method	Model-Based	
Degrees of Freedom Method	Between-Within	

Class Level Information		
Class	Levels	Values
treat	2	control ozone
tree	79	1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79
time	8	469 496 528 556 579 613 639 674

Dimensions			
Covariance Parameters		2	
Columns in X		27	
Columns in Z		0	
Subjects		79	
Max Obs Per Subject		8	
Number of Observations			
Number of Observations Read		632	
Number of Observations Used		632	
Number of Observations Not Used		0	
Iteration History			
Iteration	Evaluations	-2 Res Log Like	Criterion
0	1	1260.72452373	
1	2	-892.46138132	0.00114345
2	1	-893.74611704	0.00004058
3	1	-893.78818880	0.00000005
4	1	-893.78824278	0.00000000
Convergence criteria met.			
Covariance Parameter Estimates			
Cov Parm	Subject	Estimate	
AR(1)	tree(treat)	0.9911	
Residual		0.4272	
Fit Statistics			
-2 Res Log Likelihood		-893.8	
AIC (smaller is better)		-889.8	
AICC (smaller is better)		-889.8	
BIC (smaller is better)		-885.0	
Null Model Likelihood Ratio Test			
DF	Chi-Square	Pr > ChiSq	
1	2154.51	<.0001	
Type 3 Tests of Fixed Effects			
	Num	Den	
Effect	DF	DF	F Value Pr > F
treat	1	77	5.13 0.0263
time	7	539	135.39 <.0001
treat*time	7	539	2.03 0.0500

First Exercise – MIXED (UN) Approach

```
proc mixed data=sitka89;
  title 'MIXED approach with Unstructured error structure';
  class treat tree time;
  model size=treat time treat*time;
  repeated / type=un sub=tree(treat);
run;
```

MIXED approach with Unstructured error structure

The Mixed Procedure

Model Information			
Data Set	WORK.SITKA89		
Dependent Variable	size		
Covariance Structure	Unstructured		
Subject Effect	tree(treat)		
Estimation Method	REML		
Residual Variance Method	None		
Fixed Effects SE Method	Model-Based		
Degrees of Freedom Method	Between-Within		
Class Level Information			
Class	Levels	Values	
treat	2	control ozone	
tree	79	1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79	
time	8	469 496 528 556 579 613 639 674	
Dimensions			
Covariance Parameters	36		
Columns in X	27		
Columns in Z	0		
Subjects	79		
Max Obs Per Subject	8		
Number of Observations			
Number of Observations Read	632		
Number of Observations Used	632		
Number of Observations Not Used	0		
Iteration History			
Iteration	Evaluations	-2 Res Log Like	Criterion
0	1	1260.72452373	
1	1	-1246.66153301	0.00000000
Convergence criteria met.			
Covariance Parameter Estimates			
Cov Parm	Subject	Estimate	
UN(1,1)	tree(treat)	0.4486	
UN(2,1)	tree(treat)	0.4455	
UN(2,2)	tree(treat)	0.4425	
UN(3,1)	tree(treat)	0.4189	
UN(3,2)	tree(treat)	0.4166	
UN(3,3)	tree(treat)	0.4012	
UN(4,1)	tree(treat)	0.4080	
UN(4,2)	tree(treat)	0.4060	
UN(4,3)	tree(treat)	0.3890	
UN(4,4)	tree(treat)	0.3877	
UN(5,1)	tree(treat)	0.4223	
UN(5,2)	tree(treat)	0.4201	
UN(5,3)	tree(treat)	0.4019	
UN(5,4)	tree(treat)	0.4004	
UN(5,5)	tree(treat)	0.4229	
UN(6,1)	tree(treat)	0.4119	

UN(6,2)	tree(treat)	0.4099
UN(6,3)	tree(treat)	0.3944
UN(6,4)	tree(treat)	0.3930
UN(6,5)	tree(treat)	0.4115
UN(6,6)	tree(treat)	0.4061
UN(7,1)	tree(treat)	0.3986
UN(7,2)	tree(treat)	0.3967
UN(7,3)	tree(treat)	0.3807
UN(7,4)	tree(treat)	0.3790
UN(7,5)	tree(treat)	0.3950
UN(7,6)	tree(treat)	0.3909
UN(7,7)	tree(treat)	0.3851
UN(8,1)	tree(treat)	0.4085
UN(8,2)	tree(treat)	0.4065
UN(8,3)	tree(treat)	0.3918
UN(8,4)	tree(treat)	0.3909
UN(8,5)	tree(treat)	0.4077
UN(8,6)	tree(treat)	0.4030
UN(8,7)	tree(treat)	0.3936
UN(8,8)	tree(treat)	0.4079

Fit Statistics	
-2 Res Log Likelihood	-1246.7
AIC (smaller is better)	-1174.7
AICC (smaller is better)	-1170.1
BIC (smaller is better)	-1089.4

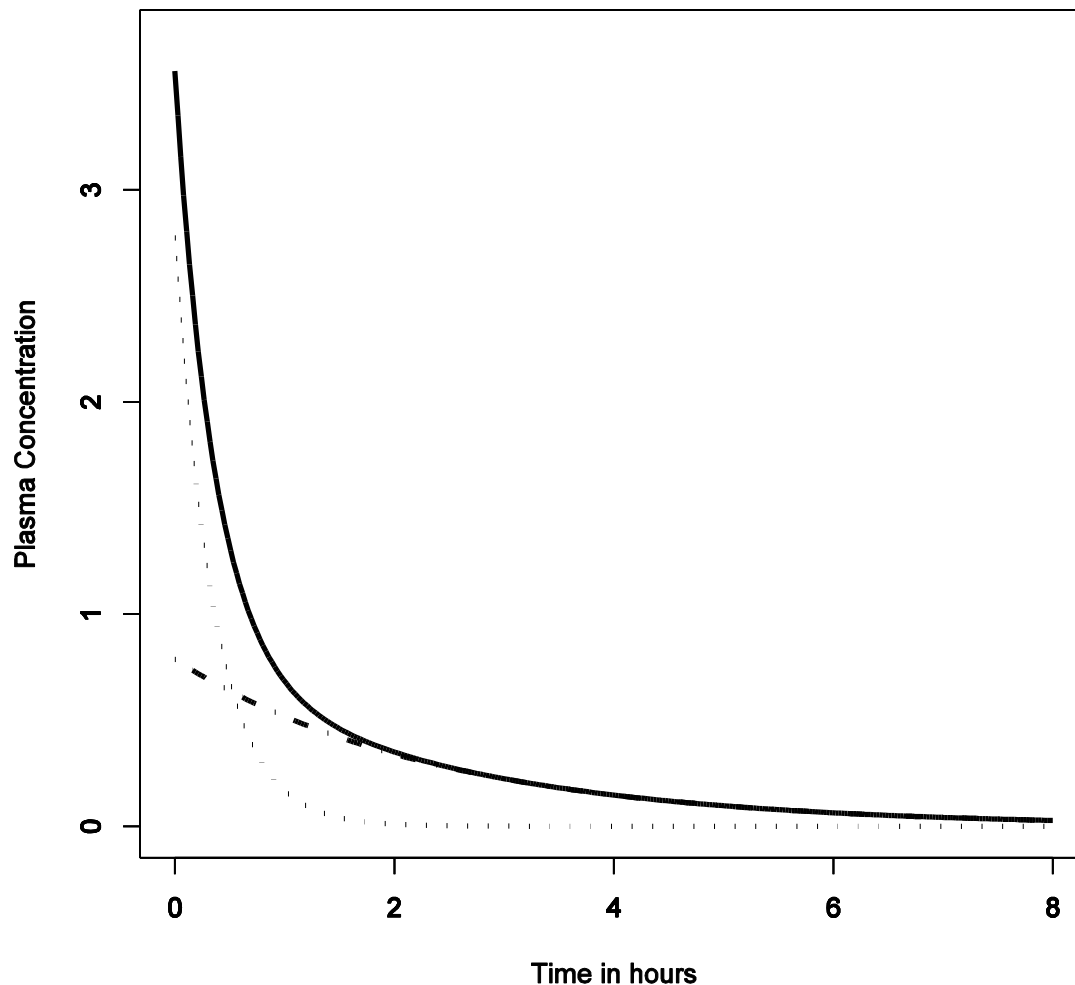
Null Model Likelihood Ratio Test		
DF	Chi-Square	Pr > ChiSq
35	2507.39	<.0001

Type 3 Tests of Fixed Effects				
	Num	Den		
Effect	DF	DF	F Value	Pr > F
treat	1	77	5.30	0.0241
time	7	77	141.48	<.0001
treat*time	7	77	2.15	0.0485

The fitted variance-covariance for these data follows

0.4486							
0.4455	0.4425	<i>This matrix is symmetric</i>					
0.4189	0.4166	0.4012					
0.4080	0.4060	0.3890	0.3877				
0.4223	0.4201	0.4019	0.4004	0.4229			
0.4119	0.4099	0.3944	0.3930	0.4115	0.4061		
0.3986	0.3967	0.3807	0.3790	0.3950	0.3909	0.3851	
0.4085	0.4065	0.3918	0.3909	0.4077	0.4030	0.3936	0.4079

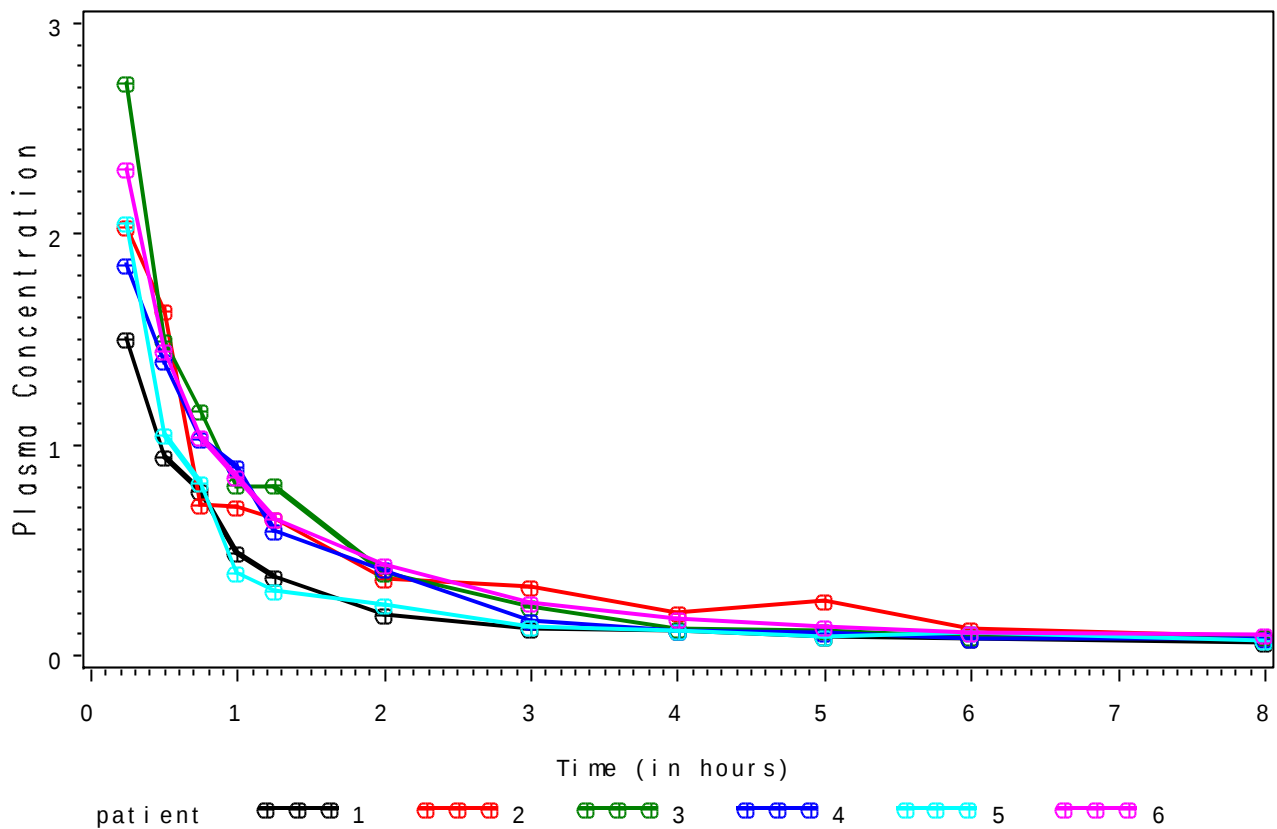
Second Exercise Plot of Theoretical Model Function (upper-most curve)



Second Exercise Plot of Data for 6 Patients

```
data one;
  do patient=1 to 6;
    do time=.25 to 1.25 by .25,2,3,4,5,6,8;
      input y @@; plasma=y/100; drop y; output;
    end; end; cards;
150 94 78 48 37 19 12 11 8 7 5 203 163 71 70 64 36 32 20 25 12 8
272 149 116 80 80 39 22 12 11 8 8 185 139 102 89 59 40 16 11 10 7 7
205 104 81 39 30 23 13 11 8 10 6 231 144 103 84 64 42 24 17 13 10 9
;
axis1 label=(a=90 h=1.2 'Plasma Concentration');
axis2 label=(h=1.2 'Time (in hours)');
symbol1 v=+ i=join; symbol2 v=x i=join; symbol3 v=* i=join;
symbol4 v=dot i=join; symbol5 v=circle i=join; symbol6 v=$ i=join;
proc gplot;
  title 'Kwan*s Plasma Concentration data - 6 patients';
  plot plasma*time=patient / vaxis=axis1 haxis=axis2;
run;
```

Kwan*s Plasma Concentration data — 6 patients



Second Exercise – First NLMIXED

```
proc nlmixed;
  parms b1=3.5 b2=3.6 b3=.5 b4=.5 s2e=.1;
  rhs=b1*exp(-b2*time)+b3*exp(-b4*time);
  model plasma~normal(rhs,s2e);
run;
```

The NLMIXED Procedure	
Specifications	
Data Set	WORK.ONE
Dependent Variable	plasma
Distribution for Dependent Variable	Normal
Optimization Technique	Dual Quasi-Newton
Integration Method	None
Dimensions	
Observations Used	66
Observations Not Used	0
Total Observations	66
Parameters	5

Fit Statistics								
-2 Log Likelihood				-47.3				
AIC (smaller is better)				-37.3				
AICC (smaller is better)				-36.3				
BIC (smaller is better)				-26.3				
Parameter Estimates								
Parameter	Estimate	Standard Error	DF	t Value	Pr > t	Alpha	Lower	Upper
b1	2.7734	0.2485	66	11.16	<.0001	0.05	2.2773	3.2695
b2	2.4263	0.5997	66	4.05	0.0001	0.05	1.2289	3.6236
b3	0.6067	0.3010	66	2.02	0.0479	0.05	0.005707	1.2078
b4	0.3356	0.1542	66	2.18	0.0331	0.05	0.02778	0.6434
s2e	0.02860	0.004979	66	5.74	<.0001	0.05	0.01866	0.03854

Second Exercise – Second NL MIXED

```
proc nlmixed;
  parms b1=3.5 b2=3.6 b3=.5 b4=.5 s2e=.1 s2b1=.1 s2b3=.1 cb13=0;
  rhs=(b1+u1)*exp(-b2*time)+(b3+u3)*exp(-b4*time);
  model plasma~normal(rhs,s2e);
  random u1 u3~normal([0,0],[s2b1,cb13,s2b3]) subject=patient;
run;
```

The NL MIXED Procedure Specifications	
Data Set	WORK.ONE
Dependent Variable	plasma
Distribution for Dependent Variable	Normal
Random Effects	u1 u3
Distribution for Random Effects	Normal
Subject Variable	patient
Optimization Technique	Dual Quasi-Newton
Integration Method	Adaptive Gaussian Quadrature
Dimensions	
Observations Used	66
Observations Not Used	0
Total Observations	66
Subjects	6
Max Obs Per Subject	11
Parameters	8
Quadrature Points	1
Fit Statistics	
-2 Log Likelihood	-112.2
AIC (smaller is better)	-96.2
AICC (smaller is better)	-93.7
BIC (smaller is better)	-97.9
Parameter Estimates	
Standard	

Parameter	Estimate	Error	DF	t Value	Pr > t	Alpha	Lower	Upper
b1	2.7733	0.2761	4	10.04	0.0006	0.05	2.0066	3.5400
b2	2.8139	0.3600	4	7.82	0.0014	0.05	1.8144	3.8135
b3	0.7870	0.1731	4	4.55	0.0104	0.05	0.3064	1.2676
b4	0.4195	0.07116	4	5.90	0.0041	0.05	0.2220	0.6171
s2e	0.006739	0.001298	4	5.19	0.0066	0.05	0.003134	0.01034
s2b1	0.3346	0.2338	4	1.43	0.2256	0.05	-0.3145	0.9837
s2b3	0.04391	0.03076	4	1.43	0.2266	0.05	-0.04148	0.1293
cb13	-0.00179	0.05782	4	-0.03	0.9767	0.05	-0.1623	0.1587

Second Exercise – Third NLMIXED

```
proc nlmixed;
  parms b1=3.5 b2=3.6 b3=.5 b4=.5 s2e=.1 s2b1=.1;
  rhs=(b1+u1)*exp(-b2*time)+b3*exp(-b4*time);
  model plasma~normal(rhs,s2e);
  random u1~normal(0,s2b1) subject=patient;
run;
```

The NLMIXED Procedure								
Specifications								
Data Set			WORK.ONE					
Dependent Variable			plasma					
Distribution for Dependent Variable			Normal					
Random Effects			u1					
Distribution for Random Effects			Normal					
Subject Variable			patient					
Optimization Technique			Dual Quasi-Newton					
Integration Method			Adaptive Gaussian Quadrature					
Dimensions								
Observations Used			66					
Observations Not Used			0					
Total Observations			66					
Subjects			6					
Max Obs Per Subject			11					
Parameters			6					
Quadrature Points			1					
Fit Statistics								
-2 Log Likelihood			-92.9					
AIC (smaller is better)			-80.9					
AICC (smaller is better)			-79.4					
BIC (smaller is better)			-82.1					
Parameter Estimates								
Parameter	Estimate	Standard Error	DF	t Value	Pr > t	Alpha	Lower	Upper
b1	2.7617	0.2816	5	9.81	0.0002	0.05	2.0377	3.4856
b2	2.1108	0.2473	5	8.53	0.0004	0.05	1.4750	2.7467
b3	0.4522	0.1426	5	3.17	0.0248	0.05	0.08557	0.8189
b4	0.2617	0.08520	5	3.07	0.0277	0.05	0.04271	0.4807
s2e	0.01098	0.002005	5	5.48	0.0028	0.05	0.005828	0.01614

s2b1	0.3680	0.2330	5	1.58	0.1750	0.05	-0.2309	0.9669
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Second Exercise – Fourth NLMIXED

```
proc nlmixed;
  parms b1=3.5 b2=3.6 b3=.5 b4=.5 s2e=.1 s2b1=.1 s2b3=.1;
  rhs=(b1+u1)*exp(-b2*time)+(b3+u3)*exp(-b4*time);
  model plasma~normal(rhs,s2e);
  random u1 u3~normal([0,0],[s2b1,0,s2b3]) subject=patient;
run;
```

The NLMIXED Procedure								
Specifications								
Data Set			WORK.ONE					
Dependent Variable			plasma					
Distribution for Dependent Variable			Normal					
Random Effects			u1 u3					
Distribution for Random Effects			Normal					
Subject Variable			patient					
Optimization Technique			Dual Quasi-Newton					
Integration Method			Adaptive Gaussian Quadrature					
Dimensions								
Observations Used			66					
Observations Not Used			0					
Total Observations			66					
Subjects			6					
Max Obs Per Subject			11					
Parameters			7					
Quadrature Points			1					
Fit Statistics								
-2 Log Likelihood			-112.2					
AIC (smaller is better)			-98.2					
AICC (smaller is better)			-96.3					
BIC (smaller is better)			-99.7					
Parameter Estimates								
	Standard							
Parameter	Estimate	Error	DF	t Value	Pr > t	Alpha	Lower	Upper
b1	2.7736	0.2756	4	10.06	0.0005	0.05	2.0085	3.5388
b2	2.8132	0.3594	4	7.83	0.0014	0.05	1.8155	3.8110
b3	0.7865	0.1724	4	4.56	0.0103	0.05	0.3080	1.2651
b4	0.4192	0.07058	4	5.94	0.0040	0.05	0.2233	0.6152
s2e	0.006741	0.001297	4	5.20	0.0065	0.05	0.003139	0.01034
s2b1	0.3334	0.2302	4	1.45	0.2211	0.05	-0.3058	0.9726
s2b3	0.04374	0.03016	4	1.45	0.2206	0.05	-0.04000	0.1275