

**Final examination, 22 April 2015**

All aids are allowed, except a computer-like device (including tablets and smartphones) and personal assistance. The exam consists of 3 questions, which have equal weight (*10 points each*) and should all be answered; further detail about the points is given for specific parts of each question. The duration of the exam is 3 hours.

Generally, all statistical models used should be specified, and to such detail that it is clear which terms are present and in which form. Your answers should generally (unless specified otherwise) be based on the information provided. Nevertheless, if at some point you think it is necessary to carry out additional analysis in statistical software, explain carefully the purpose of your proposed analysis and how you would implement it in the statistical software.

**Question 1.**

A study explored how different compounds affected the concentration of a solution at which an individual first senses a specific odor; this is called the (odor) threshold concentration. The study involved a total of 16 different solutions as well as 8 individuals, termed judges. The threshold-determining procedure was time consuming, and therefore each judge could only be used for four solutions. The 16 solutions were formed by choosing either low or high concentrations of four compounds, in all possible combinations. For simplicity we denote these four compounds as A–D, each taking either the value 1 (low) or 2 (high). After a suitable normalization (of no interest here), the threshold concentrations were as shown in the data listing, for the 8 judges and the different compounds determining the solutions.

Judge	Solution compounds				Threshold conc.	Natural log conc.
	A	B	C	D		
1	1	1	1	1	8389	9.0347
	1	2	2	1	816	6.7044
	2	2	1	2	4	1.3863
	2	1	2	2	46	3.8286
2	2	1	1	1	4351	8.3782
	2	2	2	1	78	4.3567
	1	2	1	2	5941	8.6896
	1	1	2	2	27138	10.2087
3	2	2	1	1	6	1.7918
	2	1	2	1	262	5.5683
	1	1	1	2	1230	7.1148
	1	2	2	2	98	4.5850
4	1	2	1	1	375	5.9269
	1	1	2	1	33551	10.4208
	2	1	1	2	246	5.5053
	2	2	2	2	10	2.3026

(Note: the table continues on the next page)

(Note: the table is continued from the previous page)

Judge	Solution compounds				Threshold conc.	Natural log conc.
	A	B	C	D		
5	1	1	1	1	56034	10.9337
	1	2	2	1	25046	10.1285
	2	2	1	2	19	2.9444
	2	1	2	2	490	6.1944
6	2	1	1	1	2346	7.7605
	2	2	2	1	35	3.5553
	1	2	1	2	228	5.4293
	1	1	2	2	6842	8.8308
7	2	2	1	1	67	4.2047
	2	1	2	1	3081	8.0330
	1	1	1	2	50991	10.8394
	1	2	2	2	784	6.6644
8	1	2	1	1	40581	10.6111
	1	1	2	1	90293	11.4108
	2	1	1	2	19103	9.8576
	2	2	2	2	61	4.1109

Use the description of the study and the information contained in the Minitab listings on the next two pages (equivalent Stata listings are available upon request) to answer the following questions. Note that the outcome is the (natural) log threshold concentration.

- A) (3 points) Describe the experimental design in statistical terms, using standard descriptors such as factors, treatments, replication, blocks, balancedness, completeness, treatments, experimental units, hierarchical structure etc. Use your characterization of the design to motivate a statistical model for the data (possibly, but not necessarily, the model considered in the subsequent questions). Make sure to explain clearly the meaning of the different terms in your model.
- B) (4 points) Describe the statistical model used in the Minitab listings, and use the results to assess and interpret, by means of estimates and statistical tests, the different effects of the model. Draw conclusions about whether each of the compounds affects the threshold concentration for odor? — if yes, quantify such effect(s). Comment also on the importance of the judges in the experimental design and analysis.
- C) (1 point) The analysis was carried out on log scale. Do you think an analysis on original scale would have been likely to meet the model assumptions as well as on log scale? If not, explain how you would anticipate model assumptions on original scale to be violated, and why.
- D) (1 point) The experimental design includes some features that distinguish it from the standard designs discussed in the course. One way this became noticeable in the analysis was that inclusion of the interaction **A\*D** failed to produce any estimates for this term (both in Minitab and Stata; output not shown). Based on the listing of the experimental design, try to explain why this happened, and suggest at least one way to modify the model/analysis to enable estimation of the **A\*D** interaction.
- E) (1 point) For another run of the experiment involving three compounds and more judges (within reasonable numbers, say at most 20 judges), propose a “standard” design (among those covered

in the course) that could be used for the new experiment. Describe the proposed design and the advantages you see of using such a design.

Minitab listing for Question 1:

```
MTB > GLM 'lnconc' = judge a|b a|c b|c b|d c|d;
SUBC> Brief 2 ;
SUBC> Means a b c d a*b a*c b*c b*d c*d.
```

General Linear Model: lnconc versus judge, a, b, c, d

Factor	Type	Levels	Values
judge	fixed	8	1, 2, 3, 4, 5, 6, 7, 8
a	fixed	2	1, 2
b	fixed	2	1, 2
c	fixed	2	1, 2
d	fixed	2	1, 2

Analysis of Variance for lnconc, using Adjusted SS for Tests

Source	DF	Seq SS	Adj SS	Adj MS	F	P
judge	7	57.390	57.390	8.199	11.63	0.000
a	1	104.236	104.236	104.236	147.87	0.000
b	1	79.783	79.783	79.783	113.18	0.000
a*b	1	3.392	3.392	3.392	4.81	0.044
c	1	0.384	0.384	0.384	0.54	0.472
a*c	1	0.565	0.565	0.565	0.80	0.385
b*c	1	1.261	1.261	1.261	1.79	0.201
d	1	12.912	12.912	12.912	18.32	0.001
b*d	1	0.126	0.126	0.126	0.18	0.679
c*d	1	1.352	1.352	1.352	1.92	0.186
Error	15	10.574	10.574	0.705		
Total	31	271.975				

S = 0.839592 R-Sq = 96.11% R-Sq(adj) = 91.97%

Unusual Observations for lnconc

Obs	lnconc	Fit	SE Fit	Residual	St Resid
30	11.4108	12.6588	0.6120	-1.2479	-2.17 R

R denotes an observation with a large standardized residual.

Least Squares Means for lnconc

a	Mean	SE Mean
1	8.596	0.2099
2	4.986	0.2099

b	Mean	SE Mean
1	8.370	0.2099
2	5.212	0.2099

```
c
1 6.901 0.2099
2 6.681 0.2099
```

```
d
1 7.426 0.2099
2 6.156 0.2099
```

```
a*b
1 1 9.849 0.2968
1 2 7.342 0.2968
2 1 6.891 0.2968
2 2 3.082 0.2968
```

```
a*c
1 1 8.572 0.2968
1 2 8.619 0.2968
2 1 5.229 0.2968
2 2 4.744 0.2968
```

```
b*c
1 1 8.678 0.2968
1 2 8.062 0.2968
2 1 5.123 0.2968
2 2 5.301 0.2968
```

```
b*d
1 1 8.943 0.2968
1 2 7.797 0.2968
2 1 5.910 0.2968
2 2 4.514 0.2968
```

```
c*d
1 1 7.330 0.2968
1 2 6.471 0.2968
2 1 7.522 0.2968
2 2 5.841 0.2968
```

Stata do-file for Question 1: (Stata output available upon request)

```
anova lnconc judge a##b a##c b##c b##d c##d
margins a##b a##c b##c b##d c##d, asbalanced
```

**Question 2.**

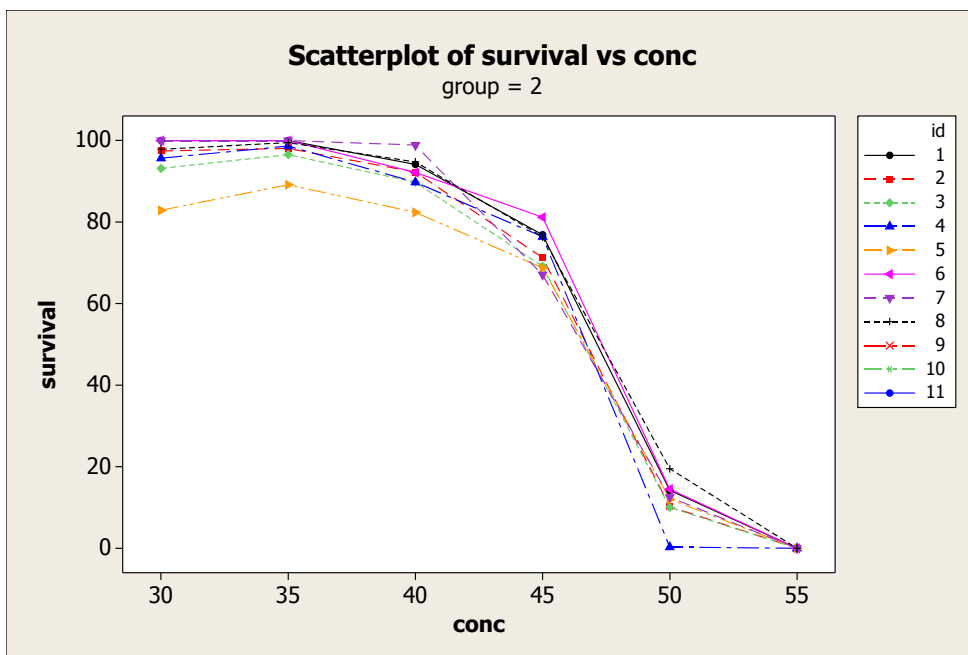
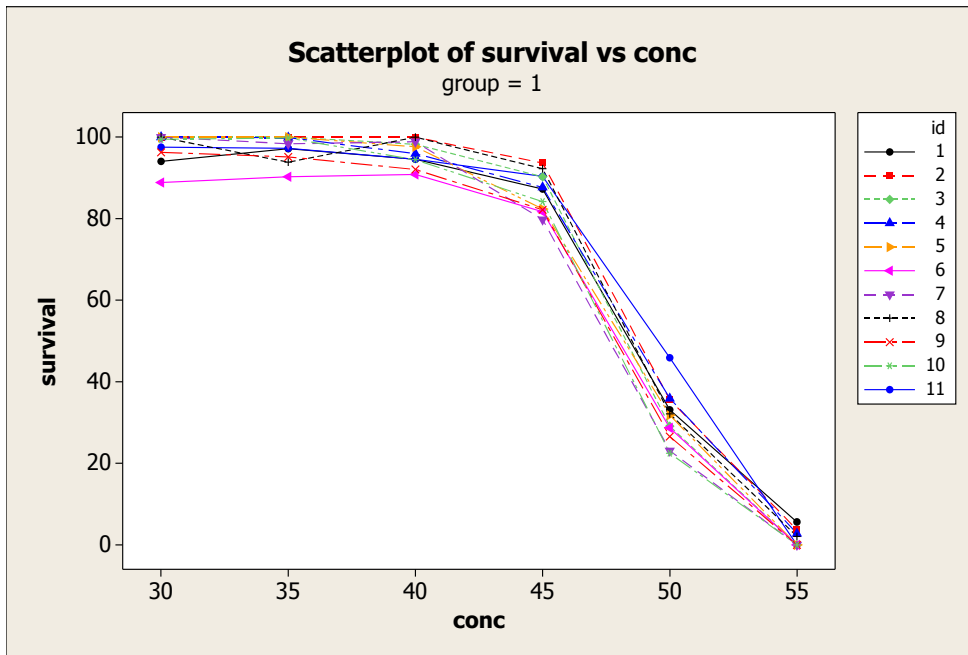
A study was carried out to detect differences in the percentage of red blood cell (RBC) survival at different ionic concentrations, among 3 different groups of mice, comprising 11, 8 and 11 animals, respectively. The table below shows the data obtained for some of the animals, and the (full) data are also displayed in 4 graphs on the next two pages.

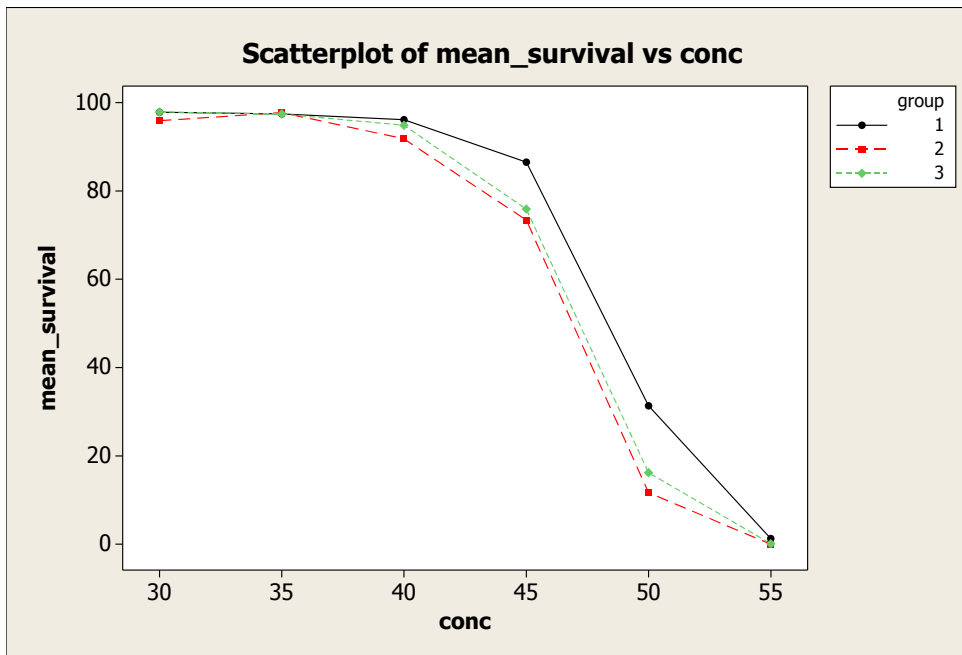
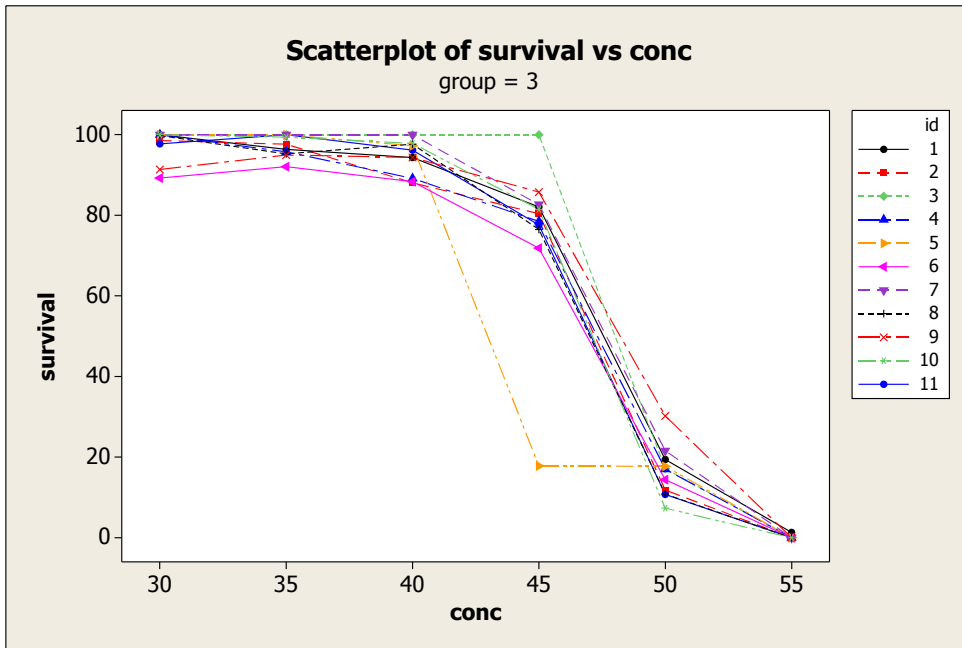
RBC survival (%)		Ionic concentration					
Group	Mouse	0.30	0.35	0.40	0.45	0.50	0.55
1	1	94.1	97.2	94.6	87.3	33.1	5.6
	2	100	100	100	93.7	35.6	3.7
	3	100	100	98.3	90.3	29.1	0
	4	100	99.9	96.0	87.7	35.9	2.8
	5	100	100	97.7	82.4	31.8	0
	...			...	...		
	11	97.6	97.3	94.6	90.4	45.9	0
2	1	100	100	94.2	76.9	14.2	0
	2	97.5	98.1	92.3	71.3	10.2	0
	3	93.2	96.6	89.8	69.2	10.0	0
	...		...	...			
	8	97.9	99.5	94.8	76.4	19.5	0
3	1	100	96.4	94.4	82.1	19.4	1.3
	2	98.6	97.7	88.1	80.4	11.8	0
	3	100	100	100	100	17.1	0
	4	100	95.8	89.2	78.5	17.0	0
	5	100	100	97.3	17.8	17.7	0
	...		...	...			
	11	97.8	100	96.2	77.6	10.7	0

- A) (3 points) Describe the data structure and experimental design; include any relevant diagrams to represent the data/design. Next, interpret the graphs; in particular, what do they tell us about the effects of groups, ionic concentrations and animals, on the outcome (RBC survival).
- B) (3 points) Describe the statistical analysis, and the underlying statistical model, carried out for the Minitab and Stata listings below; note that the Minitab and Stata listings do not contain exactly the same information. Draw conclusions from the analysis about the effects of interest in the study; **defer any discussion of model assumptions** to C). Make sure to quantify any difference(s) between groups, or describe any additional analyses you would want to carry out to assist with this part (ideally you should use the information provided as much as you can).
- C) (2 points) Use the graphical displays, in addition to the other information provided, to critically discuss the suitability of the model/analysis for these data. If you note some concerns with the model assumptions, try to assess how they might affect your conclusions from B). Include any suggestions for improvements of the statistical analysis you might have, and then give also details for how you would implement them in statistical software.
- D) (2 points) Suggest some (at least two) alternative ways to analyse the data that in your view would be helpful to explore the effects of groups and ionic concentrations (without including the analysis discussed above, or variants of it). Make sure to motivate clearly your proposed analyses, and explain how you would carry out the analysis, both in terms of the statistical

models used and the actual analysis in statistical software. Discuss, as well as the information provided allows you, whether there are concerns about the validity of the proposed analyses; ideally, you should not propose analyses that are not statistically valid.

Plots for Question 2, A):





Minitab listing and graphs for Question 2:

```
MTB > GLM 'survival' = group| conc id(group);
SUBC> Random 'id';
SUBC> Means group|conc;
```

General Linear Model: survival versus group, conc, id

Factor	Type	Levels	Values
group	fixed	3	1, 2, 3
conc	fixed	6	30, 35, 40, 45, 50, 55
id(group)	random	30	1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 1, 2, 3, 4, 5, 6, 7, 8, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11

Analysis of Variance for survival, using Adjusted SS for Tests

Source	DF	Seq SS	Adj SS	Adj MS	F	P
group	2	1388.9	1388.9	694.5	11.19	0.000
conc	5	279369.7	274381.7	54876.3	1409.14	0.000
group*conc	10	1833.4	1833.4	183.3	4.71	0.000
id(group)	27	1676.1	1676.1	62.1	1.59	0.044
Error	135	5257.3	5257.3	38.9		
Total	179	289525.4				

S = 6.24045 R-Sq = 98.18% R-Sq(adj) = 97.59%

Unusual Observations for survival

Obs	survival	Fit	SE Fit	Residual	St Resid
65	45.900	33.842	3.073	12.058	2.22 R
130	100.000	81.676	3.073	18.324	3.37 R
142	17.800	67.626	3.073	-49.826	-9.17 R
167	30.200	18.583	3.073	11.617	2.14 R

R denotes an observation with a large standardized residual.

... (some lines deleted)

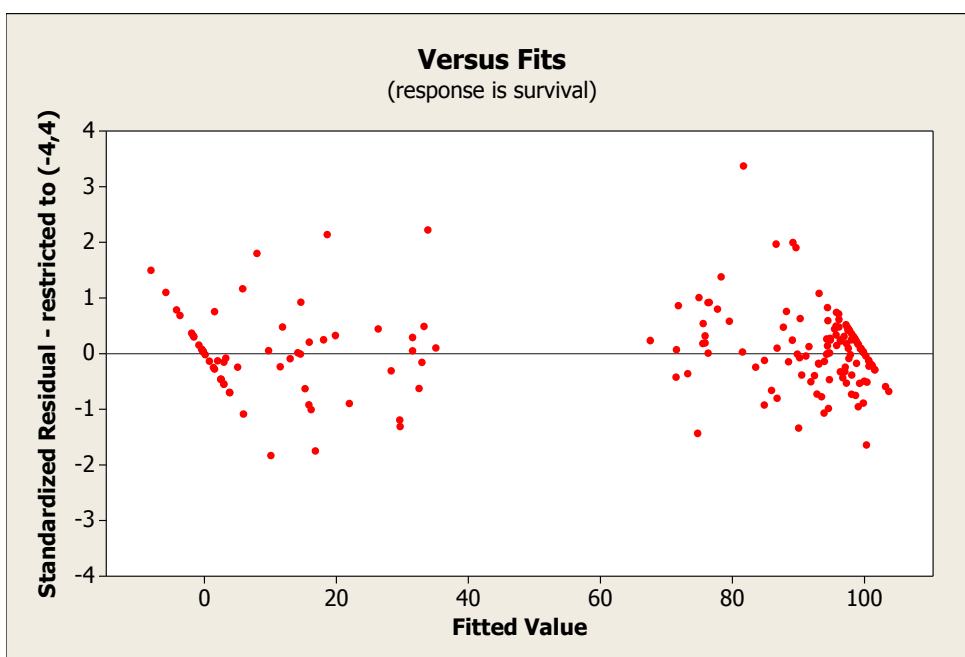
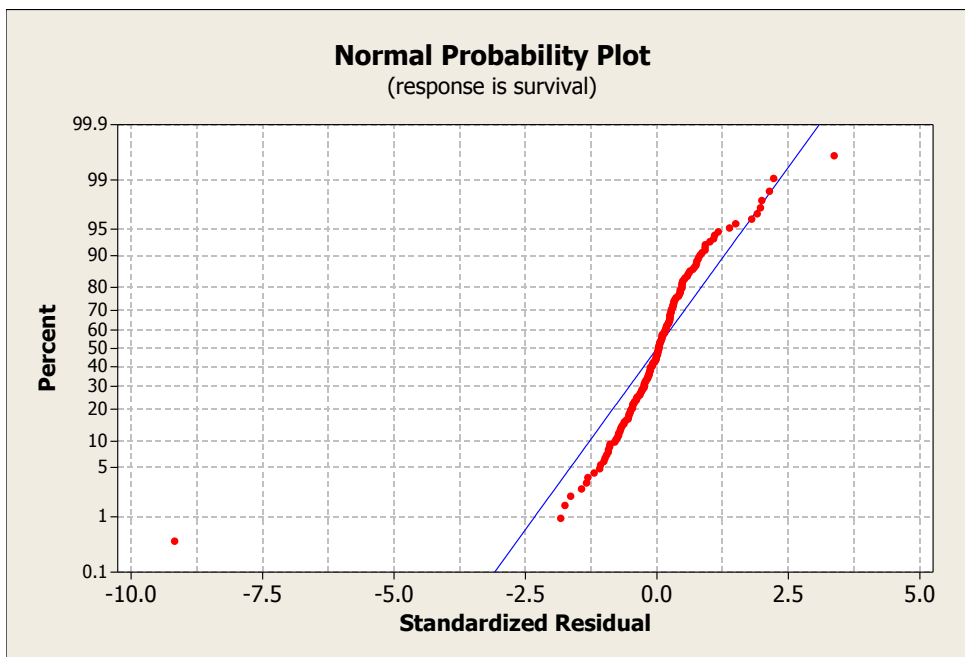
Variance Components, using Adjusted SS

Source	Estimated Value
id(group)	3.856
Error	38.943

Least Squares Means for survival

group	Mean
1	68.4333
2	61.7542
3	63.7318
conc	
30	97.2182
35	97.5481
40	94.2727
45	78.6163
50	19.7136
55	0.4697
group*conc	
1 30	97.8545
1 35	97.4545
1 40	96.1455
1 45	86.5455
1 50	31.3091
1 55	1.2909
2 30	95.9000
2 35	97.7625

2	40	91.8000
2	45	73.4125
2	50	11.6500
2	55	-0.0000
3	30	97.9000
3	35	97.4273
3	40	94.8727
3	45	75.8909
3	50	16.1818
3	55	0.1182



Stata do-file and listing for Question 2:

```
. generate uniqid=group*100+id
. mixed survival group##conc || uniqid:, reml
```

Performing EM optimization:

Performing gradient-based optimization:

Iteration 0: log restricted-likelihood = -553.42305

Iteration 1: log restricted-likelihood = -553.41878

Iteration 2: log restricted-likelihood = -553.41878

Computing standard errors:

```
Mixed-effects REML regression          Number of obs    =    180
Group variable: uniqid                 Number of groups  =     30

Obs per group: min =     6
                  avg =    6.0
                  max =     6
```

```
Wald chi2(17)    =   7243.22
Log restricted-likelihood = -553.41878  Prob > chi2     =    0.0000
```

survival	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
-----+-----						
group						
2	-1.954546	3.039843	-0.64	0.520	-7.912528 4.003436	
3	.0454552	2.789551	0.02	0.987	-5.421964 5.512874	
conc						
35	-.3999994	2.660937	-0.15	0.881	-5.615341 4.815342	
40	-1.709091	2.660937	-0.64	0.521	-6.924433 3.50625	
45	-11.30909	2.660937	-4.25	0.000	-16.52443 -6.09375	
50	-66.54545	2.660937	-25.01	0.000	-71.7608 -61.33011	
55	-96.56364	2.660937	-36.29	0.000	-101.779 -91.3483	
group#conc						
2 35	2.262499	4.10078	0.55	0.581	-5.774882 10.29988	
2 40	-2.390907	4.10078	-0.58	0.560	-10.42829 5.646474	
2 45	-11.17841	4.10078	-2.73	0.006	-19.21579 -3.141027	
2 50	-17.70454	4.10078	-4.32	0.000	-25.74193 -9.667164	
2 55	.6636369	4.10078	0.16	0.871	-7.373744 8.701018	
3 35	-.0727282	3.763134	-0.02	0.985	-7.448335 7.302878	
3 40	-1.318183	3.763134	-0.35	0.726	-8.693789 6.057424	
3 45	-10.7	3.763134	-2.84	0.004	-18.07561 -3.324394	
3 50	-15.17273	3.763134	-4.03	0.000	-22.54833 -7.797121	
3 55	-1.218182	3.763134	-0.32	0.746	-8.593789 6.157424	
_cons	97.85455	1.97251	49.61	0.000	93.9885 101.7206	

Random-effects Parameters	Estimate	Std. Err.	[95% Conf. Interval]
-----+-----			
uniqid: Identity			
var(_cons)	3.855533	2.924562	.8718006 17.05107
-----+-----			
var(Residual)	38.94323	4.740017	30.678 49.43527

```
LR test vs. linear regression: chibar2(01) = 2.70 Prob >= chibar2 = 0.0500
```

```
. testparm i.group
( 1) [survival]2.group = 0
( 2) [survival]3.group = 0

      chi2( 2) =    0.54
Prob > chi2 =    0.7648
```

```
. testparm i.conc
( 1) [survival]35.conc = 0
( 2) [survival]40.conc = 0
( 3) [survival]45.conc = 0
( 4) [survival]50.conc = 0
( 5) [survival]55.conc = 0

      chi2( 5) = 2454.64
Prob > chi2 =    0.0000
```

```
. testparm group#conc
( 1) [survival]2.group#35.conc = 0
( 2) [survival]2.group#40.conc = 0
( 3) [survival]2.group#45.conc = 0
( 4) [survival]2.group#50.conc = 0
( 5) [survival]2.group#55.conc = 0
( 6) [survival]3.group#35.conc = 0
( 7) [survival]3.group#40.conc = 0
( 8) [survival]3.group#45.conc = 0
( 9) [survival]3.group#50.conc = 0
(10) [survival]3.group#55.conc = 0

      chi2(10) =   47.08
Prob > chi2 =    0.0000
```

```
. predict stdres, rstandard
. predict fitted, fitted
. list if abs(stdres)>2
```

```
+-----+
| group  id  conc  survival  uniqid  stdres  fitted |
+-----+
65. |    1   11   50    45.9    111   2.186836  32.25315 |
130. |    3    3   45    100    303   3.517907  78.04667 |
142. |    3    5   45    17.8    305  -8.815203  72.81084 |
167. |    3    9   50    30.2    309   2.102931  17.07676 |
+-----+
```

```
. qnorm stdres
. scatter stdres fitted
```

Note: Stata graphs not included; use instead the Minitab graphs.

### Question 3.

A frequently used literature dataset (Johnson & Raven (1973), *Science* **179**, 893–895) gives the number of tortoise species on 30 Galapagos islands. The interest is in the relationship between the diversity of tortoises and several geographic variables also included in the data (species diversity on these islands was first studied by Charles Darwin in the 1830's). As our measure of diversity, we will consider the number of endemic species (**ES** in the table below). Some geographical information was missing for three of the 30 islands, therefore only the 27 islands with complete data are included here. The table below explains the different variables. Furthermore, the Minitab output on the following pages includes a listing of the full dataset.

Variable	Description
island	name of island
NS	total number of (tortoise) species on islands
ES	number of endemic species on island, i.e., occurring only on this island
area	area of the island (km <sup>2</sup> )
anear	area of the nearest island (km <sup>2</sup> )
dist	distance to the nearest island (km)
distSC	distance to Santa Cruz, the largest island, (km)
elev	highest elevation of the island (m)

Use the information contained in the Minitab listing and graphs displayed below to answer to the following questions. A Stata do-file which would give equivalent output is shown as well (without repeating the output; Stata output is available upon request).

- A) (*2 points*) Explain the statistical model on which the analysis is based, and interpret all the estimated parameters of the model (assuming that the model is valid and meaningful).
- B) (*4 points*) Discuss as fully as the provided information allows, whether the assumptions of the model seem to be met. For any issue that you may consider as a (potential) model violation, assess its impact on the analysis and give specific directions for how you would deal with it in the analysis.
- C) (*4 points*) Based on information provided, give a detailed outline of how you would develop a good (final) model for the number of endemic species (**ES**). Include in your outline any decisions you make about which predictors to include in the analysis leading to the final model; in particular, discuss whether the total number of species (**NS**) should be included as a predictor or not. Include also in your outline any analyses undertaken to examine in which form the predictors should be entered into the model.

Minitab listing for Question 3:

MTB > Correlation 'NS'-'elev'.

Correlations: NS, ES, area, anear, dist, distSC, elev

	NS	ES	area	anear	dist	distSC
ES	0.974 0.000					
area	0.614 0.001	0.618 0.001				
anear	0.034 0.865	0.095 0.639	0.185 0.355			
dist	-0.055 0.784	-0.049 0.810	-0.132 0.512	-0.114 0.571		
distSC	-0.243 0.222	-0.253 0.203	-0.133 0.509	0.061 0.763	0.599 0.001	
elev	0.725 0.000	0.778 0.000	0.751 0.000	0.562 0.002	-0.027 0.895	-0.047 0.817

Cell Contents: Pearson correlation  
P-Value

MTB > Name c9 "SRES1" c10 "TRES1" c11 "HI1" c12 "COOK1"  
MTB > Regress 'ES' 5 'area' 'anear' 'dist' 'distSC' 'elev';  
SUBC> SResiduals 'SRES1';  
SUBC> Tresiduals 'TRES1';  
SUBC> Hi 'HI1';  
SUBC> Cookd 'COOK1';  
SUBC> GFourpack;  
SUBC> RType 2;  
SUBC> Constant;  
SUBC> Brief 2.

Regression Analysis: ES versus area, anear, dist, distSC, elev

The regression equation is

$$ES = 9.54 - 0.00834 \text{ area} - 0.0174 \text{ anear} + 0.003 \text{ dist} - 0.0774 \text{ distSC} + 0.0807 \text{ elev}$$

Predictor	Coef	SE Coef	T	P
Constant	9.539	4.279	2.23	0.037
area	-0.008339	0.004677	-1.78	0.089
anear	-0.017427	0.003823	-4.56	0.000
dist	0.0027	0.2174	0.01	0.990
distSC	-0.07737	0.04528	-1.71	0.102
elev	0.08069	0.01144	7.05	0.000

S = 12.4693 R-Sq = 83.4% R-Sq(adj) = 79.4%

Analysis of Variance

Source	DF	SS	MS	F	P
Regression	5	16398.9	3279.8	21.09	0.000
Residual Error	21	3265.1	155.5		
Total	26	19664.1			

... (some lines deleted)

Unusual Observations

Obs	area	ES	Fit	SE Fit	Residual	St Resid
9	634	35.00	36.07	12.29	-1.07	-0.51 X
13	4669	89.00	95.11	12.28	-6.11	-2.81RX
16	60	37.00	60.31	6.33	-23.31	-2.17R
22	904	95.00	71.71	5.46	23.29	2.08R

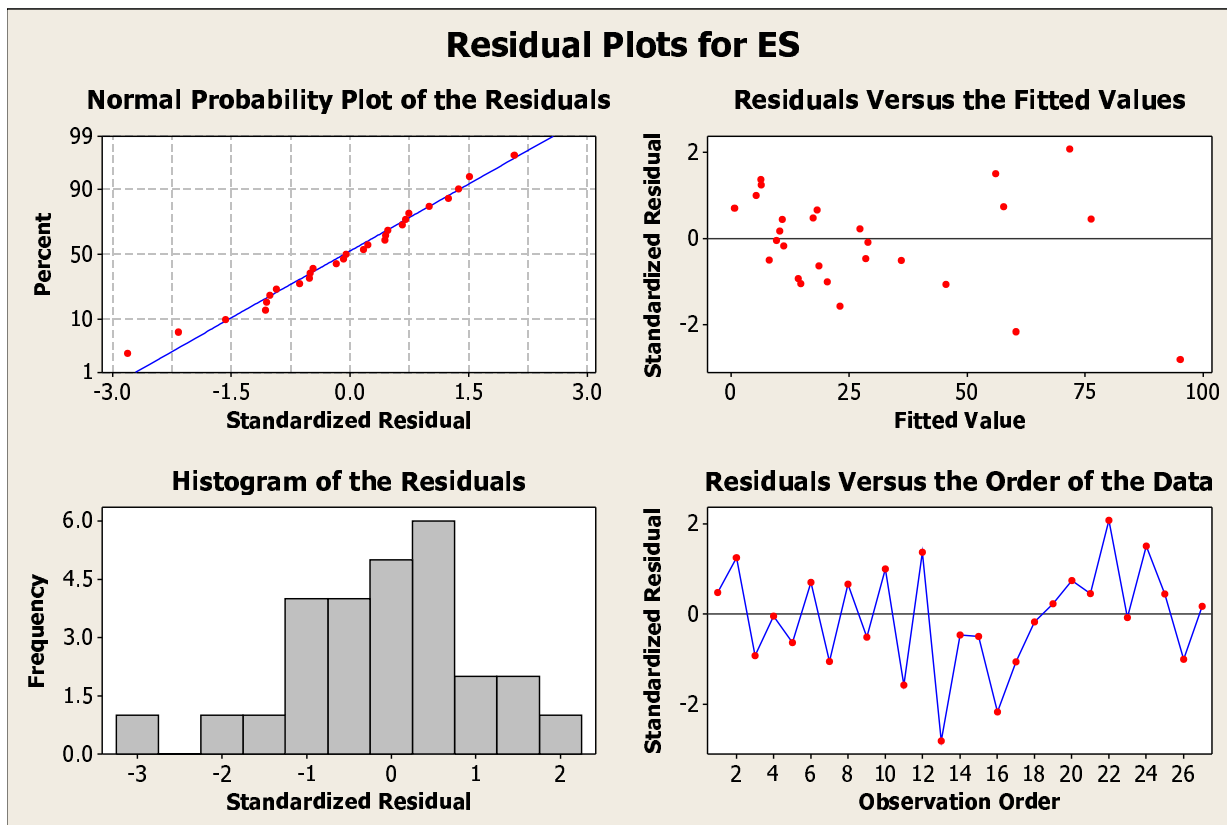
R denotes an observation with a large standardized residual.  
 X denotes an observation whose X value gives it large influence.

MTB > Print 'island'-'COOK1'.

Data Display

Row	island	NS	ES	area	anear	dist	distSC	elev	SRES1	TRES1	HI1	COOK1
1	Baltra	58	23	25.09	1.84	0.6	0.6	100	0.4774	0.4685	0.0904	0.003
2	Bartolome	31	21	1.24	572.33	0.6	26.3	109	1.2479	1.2657	0.1097	0.032
3	Caldwell	3	3	0.21	0.78	2.8	58.7	114	-0.9293	-0.9262	0.0679	0.010
4	Champion	25	9	0.10	0.18	1.9	47.4	46	-0.0489	-0.0477	0.0807	0.000
5	Daphne_Major	18	11	0.34	1.84	8.0	8.0	120	-0.6366	-0.6273	0.0858	0.006
6	Darwin	10	7	2.33	2.85	34.1	290.2	168	0.7026	0.6939	0.4775	0.075
7	Enderby	2	2	0.18	0.10	2.6	50.2	112	-1.0536	-1.0565	0.0660	0.013
8	Espanola	97	26	58.27	0.57	1.1	88.3	198	0.6603	0.6512	0.1006	0.008
9	Fernandina	93	35	634.49	4669.32	4.3	95.3	1494	-0.5135	-0.5043	0.9721	1.536
10	Gardner1	58	17	0.57	58.27	1.1	93.1	49	1.0029	1.0031	0.1207	0.023
11	Gardner2	5	4	0.78	0.21	4.6	62.2	227	-1.5730	-1.6344	0.0572	0.025
12	Genovesa	40	19	17.35	129.49	47.4	92.2	76	1.3745	1.4062	0.4481	0.255
13	Isabela	347	89	4669.32	634.49	0.7	28.1	1707	-2.8114	-3.4743	0.9696	42.043
14	Marchena	51	23	129.49	59.56	29.1	85.9	343	-0.4690	-0.4601	0.1054	0.004
15	Onslow	2	2	0.01	0.10	3.3	45.9	25	-0.5034	-0.4943	0.0828	0.003
16	Pinta	104	37	59.56	129.49	29.1	119.6	777	-2.1695	-2.4036	0.2576	0.272
17	Pinzon	108	33	17.95	0.03	10.7	10.7	458	-1.0650	-1.0686	0.1075	0.022
18	Las_Plazas	12	9	0.23	25.09	0.5	0.6	25	-0.1761	-0.1720	0.1101	0.000
19	Rabida	70	30	4.89	572.33	4.4	24.4	367	0.2269	0.2217	0.0636	0.000
20	San_Cristobal	280	65	551.62	0.57	45.2	66.5	716	0.7428	0.7346	0.3759	0.055
21	San_Salvador	237	81	572.33	4.89	0.2	19.8	906	0.4515	0.4428	0.2898	0.013
22	Santa_Cruz	444	95	903.82	0.52	0.6	0.0	864	2.0776	2.2747	0.1918	0.170
23	Santa_Fe	62	28	24.08	0.52	16.5	16.5	259	-0.0840	-0.0820	0.0955	0.000
24	Santa_Maria	285	73	170.92	0.10	2.6	49.2	640	1.5117	1.5628	0.1823	0.085
25	Seymour	44	16	1.84	25.09	0.6	9.6	30	0.4426	0.4339	0.1006	0.003
26	Tortuga	16	8	1.24	17.95	6.8	50.9	186	-1.0114	-1.0120	0.0481	0.008
27	Wolf	21	12	2.85	2.33	34.1	254.7	253	0.1703	0.1663	0.3417	0.002

*Minitab plots for Question 3:*



*Stata do-file for Question 3:* (Stata output available upon request)

```

pworth NS ES area anear dist distSC elev, sig
regress ES area anear dist distSC elev
predict fit, xb
predict stdres, rstandard
predict delres, rstudent
predict lev, hat
predict cook, cooks
scatter stdres fit
qnorm stdres
list NS-cook, table

```