

Running MLwiN from within Stata: the `runmlwin` command

e-Stat meeting
University of Bristol
7th April 2011

George Leckie and Chris Charlton
Centre for Multilevel Modelling
University of Bristol

INTRODUCTION

Existing multilevel modelling commands in Stata

- Stata provide the `xtmixed`, `xtmelogit` and `xtmepoisson` commands to fit multilevel models
 - Limited range of models can be specified
 - Computationally quite slow to fit models
- Sophia Rabe-Hesketh and Anders Skrondal provide the `gllamm` command
 - Wide range of models can be specified
 - Computationally slow to fit models
- Other user-written multilevel modelling commands include: `hlm`, `realcomimpute`, `runmplus`, `sabre`, `winbugs`

Multilevel modelling in MLwiN

1. Estimation of multilevel models for continuous, binary, **ordered categorical**, **unordered categorical** and count data
2. Fast estimation via classical and **Bayesian** methods
3. Estimation of multilevel models for cross-classified and **multiple membership** non-hierarchical data structures
4. Estimation of multilevel **multivariate response models**, **multilevel spatial models**, **multilevel measurement error models**, **multilevel multiple imputation models** and **multilevel factor models**

RAUDENBUSH (1993) CROSS- CLASSIFIED MODELLING EXAMPLE

Scottish neighbourhood study on child educational attainment

- Scottish neighbourhood study on child educational attainment
- 2310 students nested within 17 schools and 524 neighbourhoods
- First analysed by Garner and Raudenbush (1991)
- Re-analysed by Rabe-Hesketh and Skrondal (2008), Raudenbush (1993), Raudenbush and Bryk (2002) and others

Journal of Educational Statistics
Winter 1993, Vol. 18, No. 4, pp. 321–349

A Crossed Random Effects Model for Unbalanced Data With Applications in Cross-Sectional and Longitudinal Research

Stephen W. Raudenbush
Michigan State University

Key words: hierarchical models, maximum likelihood, covariance components

Hierarchical linear models have found widespread application when the data have a nested structure—for example, when students are nested within classrooms (a two-level nested structure) or students are nested within classrooms and classrooms are nested within schools (a three-level nested structure). Often, however, the data will have a more complex nested structure. In Example 1, students are nested within both neighborhoods and schools; however, a school can draw students from multiple neighborhoods, and a neighborhood can send students to multiple schools. In Example 2, children are nested within classrooms during the first year of the study; however, each child finds himself or herself with a new teacher and a new set of classmates

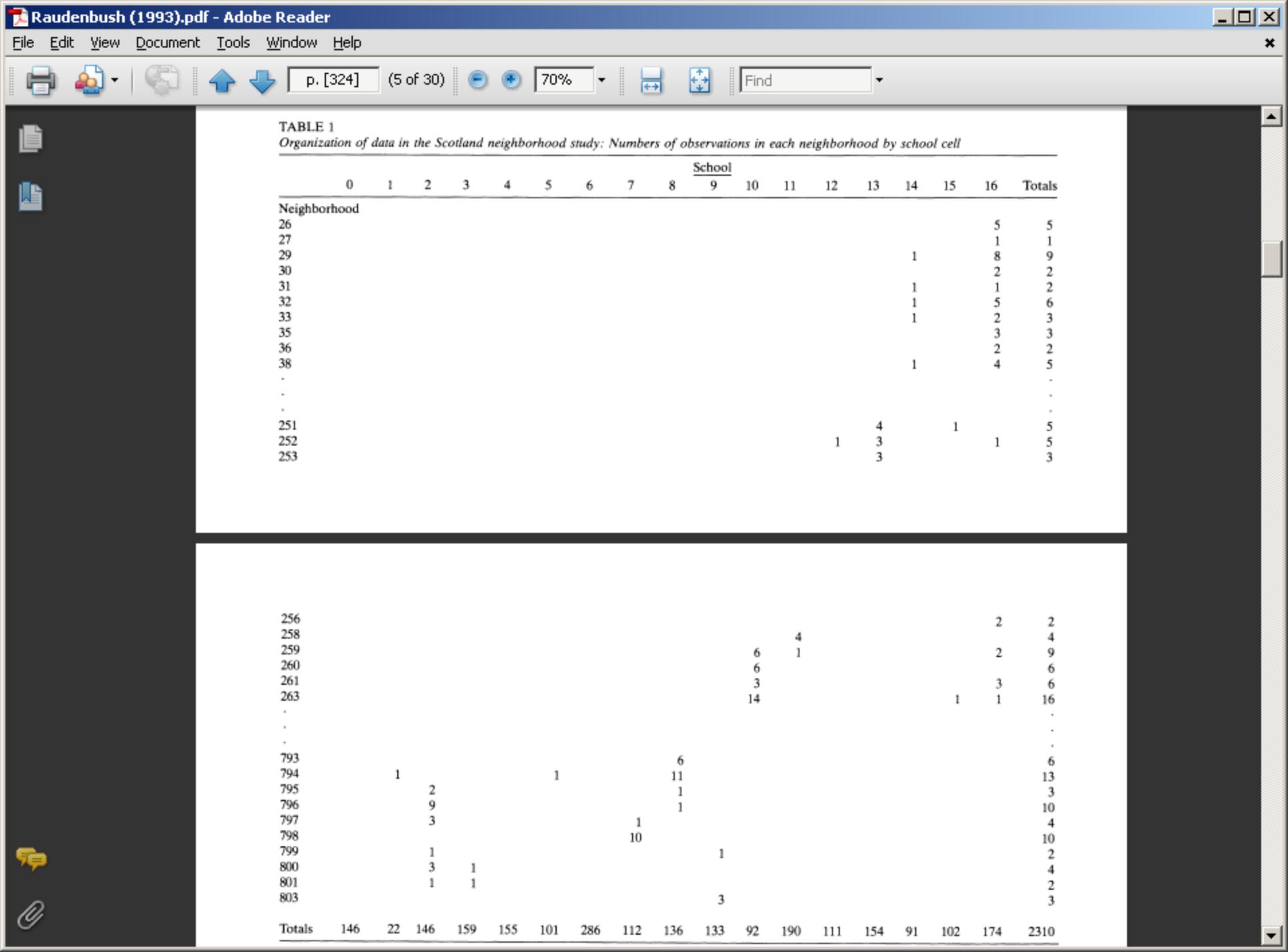
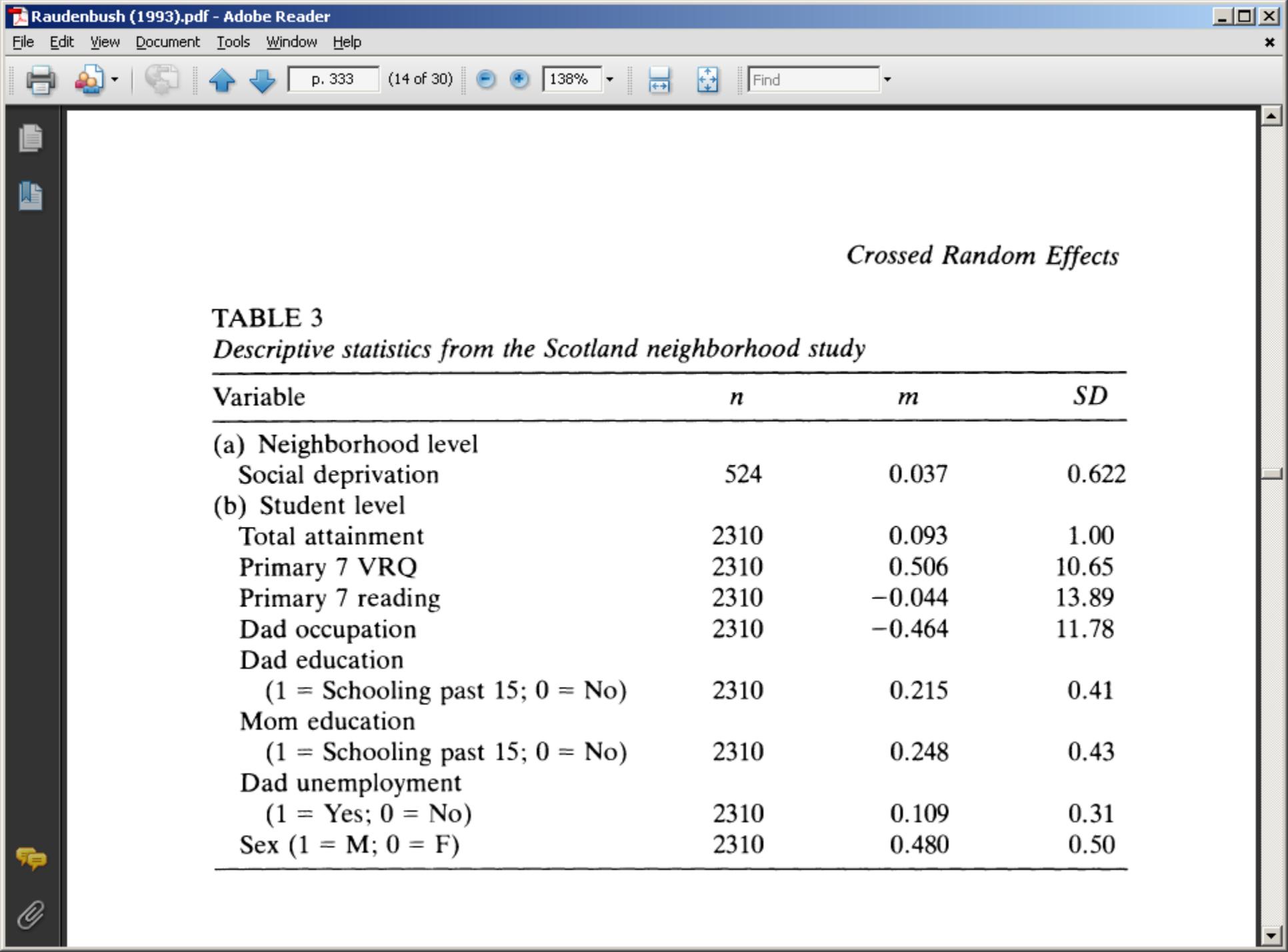


TABLE 1

Organization of data in the Scotland neighborhood study: Numbers of observations in each neighborhood by school cell

Neighborhood	School																Totals	
	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15		16
26																	5	5
27																	1	1
29															1		8	9
30																	2	2
31															1		1	2
32															1		5	6
33															1		2	3
35																	3	3
36																	2	2
38															1		4	5
.																		.
.																		.
251														4		1		5
252													1	3			1	5
253														3				3
256																	2	2
258												4						4
259											6	1					2	9
260											6							6
261											3						3	6
263											14					1	1	16
.																		.
.																		.
793										6								6
794		1				1				11								13
795			2							1								3
796			9							1								10
797			3						1									4
798								10										10
799			1								1							2
800			3		1													4
801			1		1													2
803										3								3
Totals	146	22	146	159	155	101	286	112	136	133	92	190	111	154	91	102	174	2310



Crossed Random Effects

TABLE 3
Descriptive statistics from the Scotland neighborhood study

Variable	<i>n</i>	<i>m</i>	<i>SD</i>
(a) Neighborhood level			
Social deprivation	524	0.037	0.622
(b) Student level			
Total attainment	2310	0.093	1.00
Primary 7 VRQ	2310	0.506	10.65
Primary 7 reading	2310	-0.044	13.89
Dad occupation	2310	-0.464	11.78
Dad education			
(1 = Schooling past 15; 0 = No)	2310	0.215	0.41
Mom education			
(1 = Schooling past 15; 0 = No)	2310	0.248	0.43
Dad unemployment			
(1 = Yes; 0 = No)	2310	0.109	0.31
Sex (1 = M; 0 = F)	2310	0.480	0.50

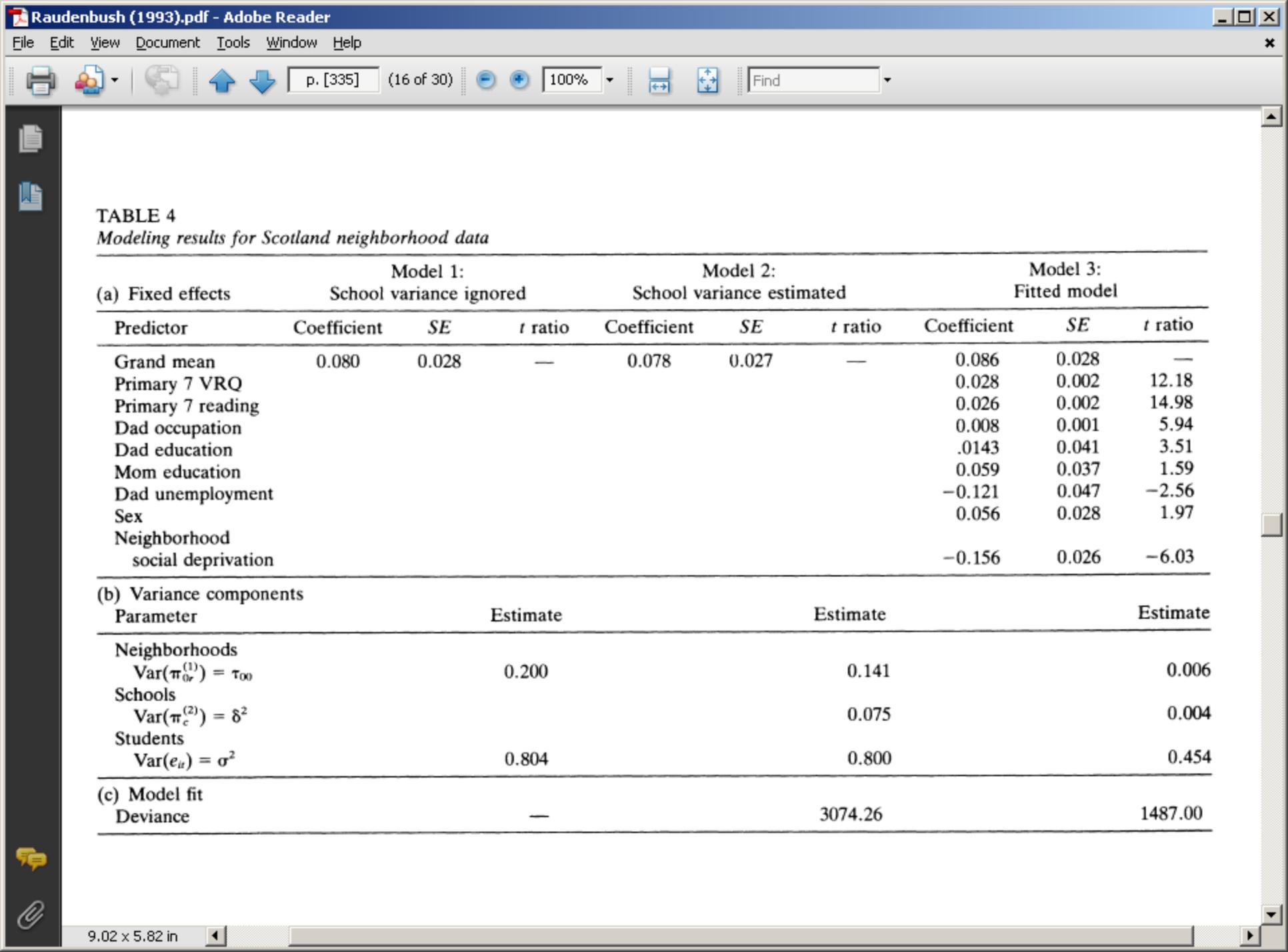


TABLE 4

Modeling results for Scotland neighborhood data

(a) Fixed effects	Model 1: School variance ignored			Model 2: School variance estimated			Model 3: Fitted model		
	Coefficient	SE	t ratio	Coefficient	SE	t ratio	Coefficient	SE	t ratio
Grand mean	0.080	0.028	—	0.078	0.027	—	0.086	0.028	—
Primary 7 VRQ							0.028	0.002	12.18
Primary 7 reading							0.026	0.002	14.98
Dad occupation							0.008	0.001	5.94
Dad education							.0143	0.041	3.51
Mom education							0.059	0.037	1.59
Dad unemployment							-0.121	0.047	-2.56
Sex							0.056	0.028	1.97
Neighborhood social deprivation							-0.156	0.026	-6.03
(b) Variance components									
Parameter	Estimate			Estimate			Estimate		
Neighborhoods $\text{Var}(\pi_{0r}^{(1)}) = \tau_{00}$	0.200			0.141			0.006		
Schools $\text{Var}(\pi_c^{(2)}) = \delta^2$				0.075			0.004		
Students $\text{Var}(e_{it}) = \sigma^2$	0.804			0.800			0.454		
(c) Model fit									
Deviance	—			3074.26			1487.00		



Review

Command _rc

Variables

Name	Label	Type	Format
------	-------	------	--------

STATA (R)
11.2
Statistics/Data Analysis
MP - Parallel Edition

Copyright 2009 StataCorp LP
StataCorp
4905 Lakeway Drive
College Station, Texas 77845 USA
800-STATA-PC <http://www.stata.com>
979-696-4600 stata@stata.com
979-696-4601 (fax)

2-user 2-core Stata network perpetual license:

Serial number: 50110514919

Licensed to: Centre for Multilevel Modelling
University of Bristol

Notes:

1. (/m# option or -set memory-) 500.00 MB allocated to data
2. (/v# option or -set maxvar-) 5000 maximum variables

running C:\Program Files (x86)\Stata11\sysprofile.do ...

running C:\Users\gl9158\profile.do ...

.

Command



Review

Command _rc

Variables

Name	Label	Type	Format
------	-------	------	--------

STATA (R)
11.2
Statistics/Data Analysis
MP - Parallel Edition

Copyright 2009 StataCorp LP
StataCorp
4905 Lakeway Drive
College Station, Texas 77845 USA
800-STATA-PC <http://www.stata.com>
979-696-4600 stata@stata.com
979-696-4601 (fax)

2-user 2-core stata network perpetual license:

Serial number: 50110514919

Licensed to: Centre for Multilevel Modelling
University of Bristol

Notes:

1. (/m# option or -set memory-) 500.00 MB allocated to data
2. (/v# option or -set maxvar-) 5000 maximum variables

running C:\Program Files (x86)\Stata11\sysprofile.do ...

running C:\Users\gl9158\profile.do ...

.

Command

use "http://www.stata-press.com/data/mlmus2/neighborhood.dta", clear



Review

▲	Command	_rc
1	use "http://www.stata-press.co..."	

Variables

Name	Label	Type	Format
neighid		int	%8.0g
schid		byte	%8.0g
attain		float	%9.0g
p7vrq		float	%9.0g
p7read		float	%9.0g
dadocc		float	%9.0g
dadunemp		byte	%8.0g
daded		byte	%8.0g
momed		byte	%8.0g
male		byte	%8.0g
deprive		float	%9.0g
dummy		byte	%8.0g

STATA (R)
Statistics/Data Analysis

11.2

Copyright 2009 StataCorp LP

StataCorp

4905 Lakeway Drive

College Station, Texas 77845 USA

800-STATA-PC

979-696-4600

979-696-4601 (fax)

<http://www.stata.com>stata@stata.com

2-user 2-core stata network perpetual license:

Serial number: 50110514919

Licensed to: Centre for Multilevel Modelling

University of Bristol

Notes:

1. (/m# option or -set memory-) 500.00 MB allocated to data
2. (/v# option or -set maxvar-) 5000 maximum variables

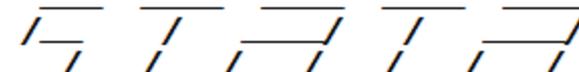
running C:\Program Files (x86)\Stata11\sysprofile.do ...

running C:\Users\gl9158\profile.do ...

. use "http://www.stata-press.com/data/mlmus2/neighborhood.dta", clear

Command



 (R)
11.2
Statistics/Data Analysis

MP - Parallel Edition

Copyright 2009 StataCorp LP

StataCorp

4905 Lakeway Drive

College Station, Texas 77845 USA

800-STATA-PC

<http://www.stata.com>

979-696-4600

stata@stata.com

979-696-4601 (fax)

2-user 2-core Stata network perpetual license:

Serial number: 50110514919

Licensed to: Centre for Multilevel Modelling
University of Bristol

Notes:

1. (/m# option or -set memory-) 500.00 MB allocated to data
2. (/v# option or -set maxvar-) 5000 maximum variables

running C:\Program Files (x86)\Stata11\sysprofile.do ...

running C:\Users\gl9158\profile.do ...

. use "<http://www.stata-press.com/data/mlmus2/neighborhood.dta>", clear

.

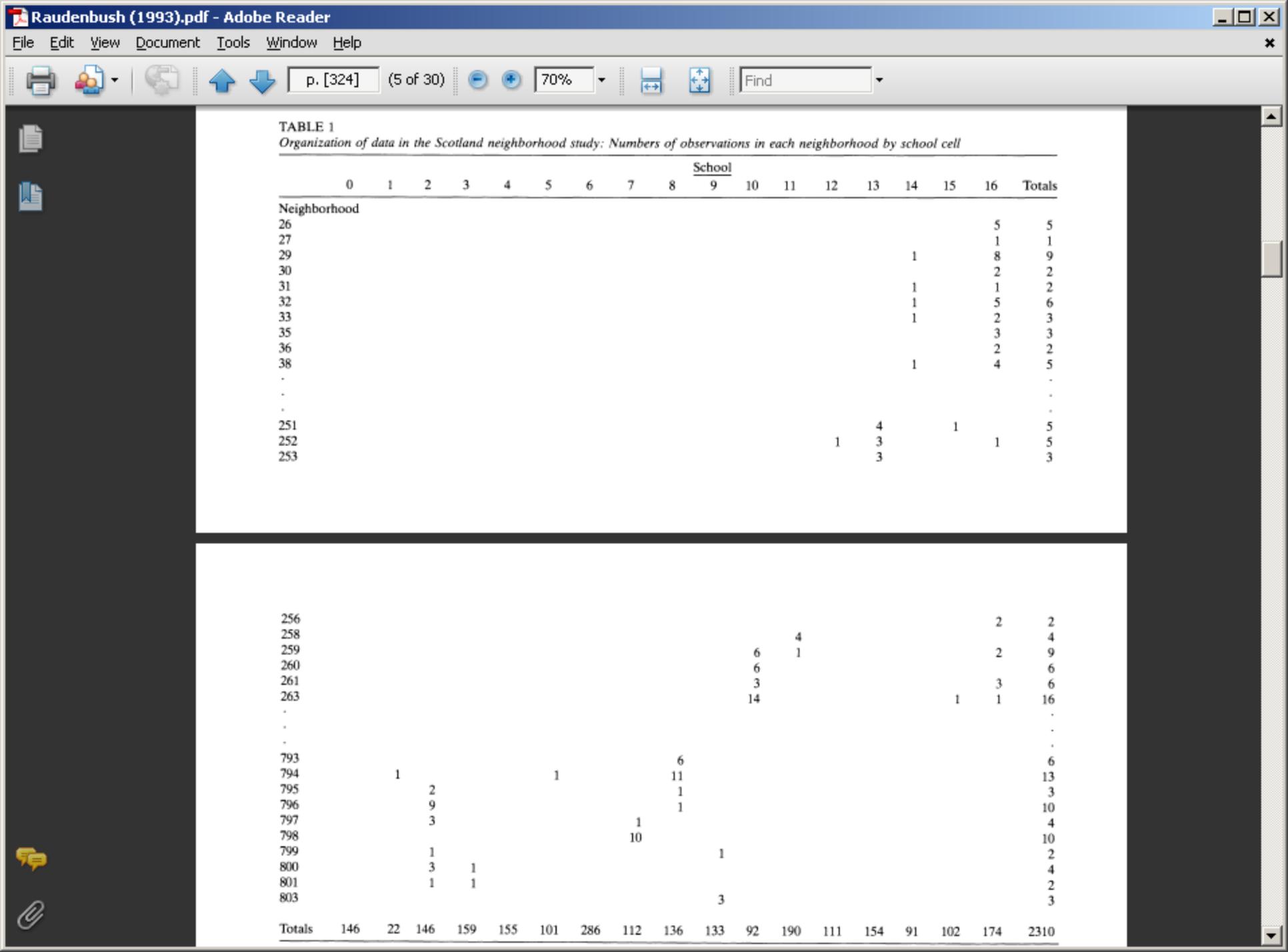


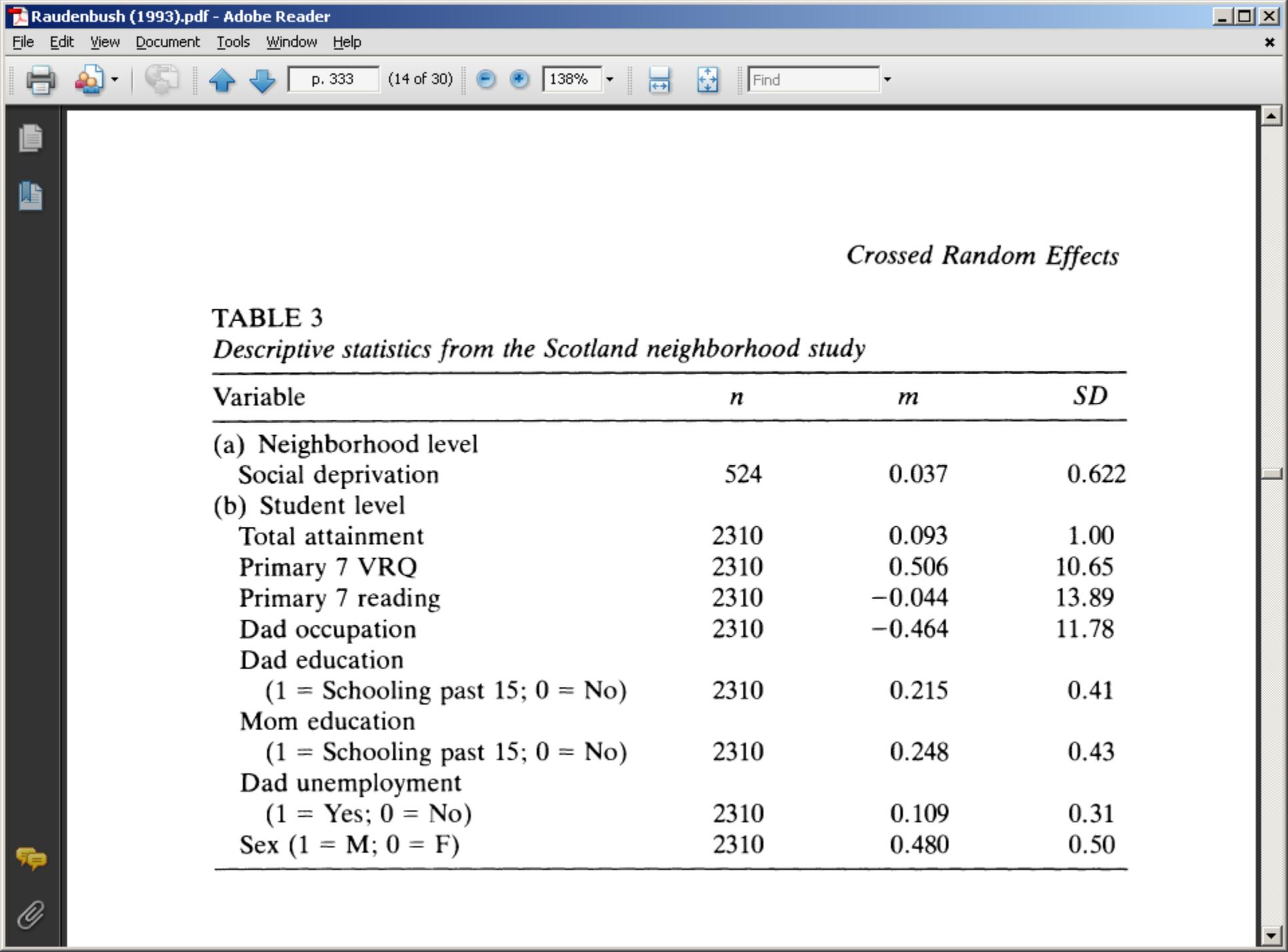
TABLE 1

Organization of data in the Scotland neighborhood study: Numbers of observations in each neighborhood by school cell

Neighborhood	School																Totals	
	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15		16
26																	5	5
27																	1	1
29															1		8	9
30																	2	2
31															1		1	2
32															1		5	6
33															1		2	3
35																	3	3
36																	2	2
38															1		4	5
.																		.
.																		.
251														4		1		5
252													1	3			1	5
253														3				3
256																	2	2
258												4						4
259											6	1					2	9
260											6							6
261											3						3	6
263											14					1	1	16
.																		.
.																		.
793										6								6
794		1				1				11								13
795			2							1								3
796			9							1								10
797			3						1									4
798								10										10
799			1								1							2
800			3		1													4
801			1		1													2
803										3								3
Totals	146	22	146	159	155	101	286	112	136	133	92	190	111	154	91	102	174	2310

```
. table neighid schid if inrange(neighid,26,38) | inrange(neighid,251,263) | inrange(neighid
> ,793,800)
```

neighid	schid								
	2	8	10	15	16	17	18	19	20
26									5
27									1
29							1		8
30									2
31							1		1
32							1		5
33							2		2
35									3
36									2
37									1
38							1		4
251						4		1	
252					1	3			1
253						3			
256									2
258				5					
259				6	1		2		
260				7					
261				4			3		
262				5		1	1		
263				14		1	1		
793		1	7						
794	1	1	12						
795	1		1						
796	9								
797	4	1							
798	9	1							
799	1		1						



Crossed Random Effects

TABLE 3
Descriptive statistics from the Scotland neighborhood study

Variable	<i>n</i>	<i>m</i>	<i>SD</i>
(a) Neighborhood level			
Social deprivation	524	0.037	0.622
(b) Student level			
Total attainment	2310	0.093	1.00
Primary 7 VRQ	2310	0.506	10.65
Primary 7 reading	2310	-0.044	13.89
Dad occupation	2310	-0.464	11.78
Dad education			
(1 = Schooling past 15; 0 = No)	2310	0.215	0.41
Mom education			
(1 = Schooling past 15; 0 = No)	2310	0.248	0.43
Dad unemployment			
(1 = Yes; 0 = No)	2310	0.109	0.31
Sex (1 = M; 0 = F)	2310	0.480	0.50

IGLS ESTIMATION

Model 1 fitted using IGLS

$$\text{attain}_i = \beta_0 + u_{\text{neighid}(i)}^{(2)} + e_i$$

$$u_j^{(2)} \sim N(0, \sigma_{u^{(2)}}^2), \quad e_i \sim N(0, \sigma_e^2)$$

```
. runmlwin attain cons, ///  
    level2(neighid: cons) ///  
    level1(studentid: cons)
```



```
. runmlwin attain cons, level2(neighid: cons) level1(studentid: cons) nopause
```

```
MLwin 2.23 multilevel model          Number of obs      =      2310
Normal response model
Estimation algorithm: IGLS
```

Level Variable	No. of Groups	Observations per Group		
		Minimum	Average	Maximum
neighid	524	1	4.4	16

```
Run time (seconds) =      1.59
Number of iterations =      3
Log likelihood      = -3207.9849
Deviance           =  6415.9697
```

attain	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
cons	.0820249	.0284355	2.88	0.004	.0262923	.1377574

Random-effects Parameters	Estimate	Std. Err.	[95% Conf. Interval]	
Level 2:				
var(cons)	.2015358	.0255229	.1515119	.2515597
Level 1:				
var(cons)	.8043717	.0265334	.7523673	.8563761

Model 2 fitted using IGLS

$$attain_i = \beta_0 + u_{schid(i)}^{(3)} + u_{neighid(i)}^{(2)} + e_i$$

$$u_j^{(3)} \sim N(0, \sigma_{u^{(3)}}^2), \quad u_j^{(2)} \sim N(0, \sigma_{u^{(2)}}^2), \quad e_i \sim N(0, \sigma_e^2)$$

```
. tabulate schid, gen(s)
. forvalues i = 1/16 {
.     constraint define `i' [RP3]var(s`i') = [RP3]var(s17)
. }
. runmlwin attain cons, ///
    level3(cons: s1-s17, diagonal) ///
    level2(neighid: cons) ///
    level1(studentid: cons) constraints(1/16)
```



```
. quietly tabulate schid, generate(s)

. forvalues s = 1/16 {
2.
. constraint define `s' [RP3]var(s`s') = [RP3]var(s17)
3.
. }

. runmlwin attain cons, level3(cons: s1-s17, diagonal) level2(neighid: cons) level1(studenti
> d: cons) constraints(1/16) nopause
```

```
( 1) [RP3]var(s1) - [RP3]var(s17) = 0
( 2) [RP3]var(s2) - [RP3]var(s17) = 0
( 3) [RP3]var(s3) - [RP3]var(s17) = 0
( 4) [RP3]var(s4) - [RP3]var(s17) = 0
( 5) [RP3]var(s5) - [RP3]var(s17) = 0
( 6) [RP3]var(s6) - [RP3]var(s17) = 0
( 7) [RP3]var(s7) - [RP3]var(s17) = 0
( 8) [RP3]var(s8) - [RP3]var(s17) = 0
( 9) [RP3]var(s9) - [RP3]var(s17) = 0
(10) [RP3]var(s10) - [RP3]var(s17) = 0
(11) [RP3]var(s11) - [RP3]var(s17) = 0
(12) [RP3]var(s12) - [RP3]var(s17) = 0
(13) [RP3]var(s13) - [RP3]var(s17) = 0
(14) [RP3]var(s14) - [RP3]var(s17) = 0
(15) [RP3]var(s15) - [RP3]var(s17) = 0
(16) [RP3]var(s16) - [RP3]var(s17) = 0
```

Warning: variance matrix is nonsymmetric or highly singular

```
MLwiN 2.23 multilevel model          Number of obs      =      2310
Normal response model
Estimation algorithm: IGLS
```

Level Variable	No. of Groups	Observations per Group		
		Minimum	Average	Maximum

Model 3 fitted using IGLS

$$\begin{aligned} \text{attain}_i = & \beta_0 + \beta_1 \text{p7vrq}_i + \beta_2 \text{p7read}_i + \beta_3 \text{dadocc}_i + \beta_4 \text{daded}_i \\ & + \beta_5 \text{momed}_i + \beta_6 \text{dadunemp}_i + \beta_7 \text{male}_i + \beta_8 \text{deprive}_i \\ & + u_{\text{schid}(i)}^{(3)} + u_{\text{neighid}(i)}^{(2)} + e_i \end{aligned}$$

$$u_j^{(3)} \sim N(0, \sigma_{u^{(3)}}^2), \quad u_j^{(2)} \sim N(0, \sigma_{u^{(2)}}^2), \quad e_i \sim N(0, \sigma_e^2)$$

```
. runmlwin attain cons p7vrq p7read dadocc daded ///  
  momed dadunemp male deprive, ///  
  level3(cons: s1-s17, diagonal) ///  
  level2(neighid: cons) ///  
  level1(studentid: cons) constraints(1/16)
```

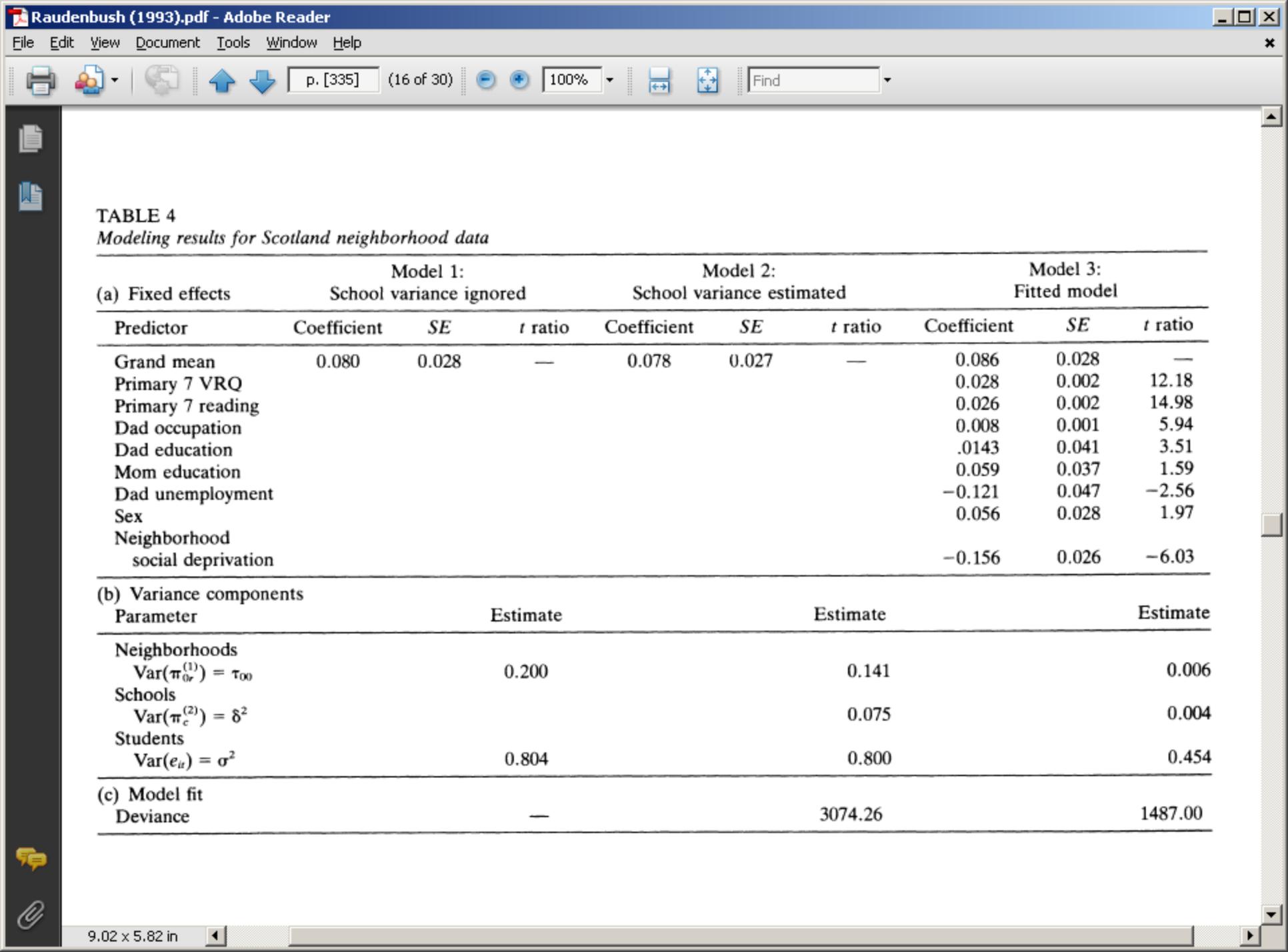



TABLE 4

Modeling results for Scotland neighborhood data

(a) Fixed effects	Model 1: School variance ignored			Model 2: School variance estimated			Model 3: Fitted model		
	Coefficient	SE	t ratio	Coefficient	SE	t ratio	Coefficient	SE	t ratio
Grand mean	0.080	0.028	—	0.078	0.027	—	0.086	0.028	—
Primary 7 VRQ							0.028	0.002	12.18
Primary 7 reading							0.026	0.002	14.98
Dad occupation							0.008	0.001	5.94
Dad education							.0143	0.041	3.51
Mom education							0.059	0.037	1.59
Dad unemployment							-0.121	0.047	-2.56
Sex							0.056	0.028	1.97
Neighborhood social deprivation							-0.156	0.026	-6.03
(b) Variance components									
Parameter	Estimate			Estimate			Estimate		
Neighborhoods $\text{Var}(\pi_{0r}^{(1)}) = \tau_{00}$	0.200			0.141			0.006		
Schools $\text{Var}(\pi_c^{(2)}) = \delta^2$				0.075			0.004		
Students $\text{Var}(e_{it}) = \sigma^2$	0.804			0.800			0.454		
(c) Model fit									
Deviance	—			3074.26			1487.00		



```
. estimates table model1a model2a model3a, stats(deviance) keep("FP1:" "RP2:" "RP3:var(s1)"
> "RP3:var(s2)" "RP3:var(s17)" "RP1:") b(%9.3f) style(online)
```

Variable	model1a	model2a	model3a
FP1			
cons	0.082	0.075	0.030
p7vrq			0.028
p7read			0.026
dadocc			0.008
daded			0.144
momed			0.059
dadunemp			-0.121
female			0.056
deprive			-0.156
RP2			
var(cons)	0.202	0.141	0.004
RP3			
var(s1)		0.075	0.004
var(s2)		0.075	0.004
var(s17)		0.075	0.004
RP1			
var(cons)	0.804	0.799	0.456
Statistics			
deviance	6415.970	6356.711	4769.335

.
.

MCMC ESTIMATION

Model 1 fitted using MCMC

$$\text{attain}_i = \beta_0 + u_{\text{neighid}(i)}^{(2)} + e_i$$

$$u_j^{(2)} \sim N(0, \sigma_{u^{(2)}}^2), \quad e_i \sim N(0, \sigma_e^2)$$

```
. runmlwin attain cons, ///  
    level2(neighid: cons) ///  
    level1(studentid: cons) ///  
mcmc(on) initsprevious
```


Model 2 fitted using MCMC

$$\text{attain}_i = \beta_0 + u_{\text{schid}(i)}^{(3)} + u_{\text{neighid}(i)}^{(2)} + e_i$$

$$u_j^{(3)} \sim N(0, \sigma_{u^{(3)}}^2), \quad u_j^{(2)} \sim N(0, \sigma_{u^{(2)}}^2), \quad e_i \sim N(0, \sigma_e^2)$$

```
. matrix b = (0, .075, .15, .8)

. runmlwin attain cons, ///

    level3(schid: cons) ///

    level2(neighid: cons) ///

    level1(studentid: cons) ///

mcmc(cc) initsb(b)
```


Model 3 fitted using MCMC

$$\begin{aligned} \text{attain}_i = & \beta_0 + \beta_1 \text{p7vrq}_i + \beta_2 \text{p7read}_i + \beta_3 \text{dadocc}_i + \beta_4 \text{daded}_i \\ & + \beta_5 \text{momed}_i + \beta_6 \text{dadunemp}_i + \beta_7 \text{male}_i + \beta_8 \text{deprive}_i \\ & + u_{\text{schid}(i)}^{(3)} + u_{\text{neighid}(i)}^{(2)} + e_i \end{aligned}$$

$$u_j^{(3)} \sim N(0, \sigma_{u^{(3)}}^2), \quad u_j^{(2)} \sim N(0, \sigma_{u^{(2)}}^2), \quad e_i \sim N(0, \sigma_e^2)$$

```
. matrix b = (0,0,0,0,0,0,0,0,0,0,1,1,1)
. runmlwin attain cons p7vrq p7read dadocc daded ///
  momed dadunemp male deprive, ///
  level3(schid: cons) ///
  level2(neighid: cons) ///
  level1(studentid: cons) mcmc(cc) initsb(b)
```



```
. runmlwin attain cons p7vrq p7read dadocc daded momed dadunemp male deprive, level3(schid:
> cons) level2(neighid: cons) level1(studentid: cons) mcmc(cc) initsb(b) nopause
```

```
MLwiN 2.23 multilevel model          Number of obs      =      2310
Normal response model
Estimation algorithm: MCMC
```

Level Variable	No. of Groups	Observations per Group		
		Minimum	Average	Maximum
schid	17	22	135.9	286
neighid	524	1	4.4	16

```
Burnin          =      500
Chain           =      5000
Run time (seconds) =      11.4
Deviance (dbar) =      4744.81
Deviance (thetabar) =      4704.18
Effective no. of pars (pd) =      40.63
Bayesian DIC    =      4785.44
```

attain	Mean	Std. Dev.	z	ESS	[95% Cred. Interval]	
cons	.0908623	.028849	3.15	1642	.0341885	.1490847
p7vrq	.0275759	.0022769	12.11	4555	.0232042	.0320144
p7read	.0262263	.0017901	14.65	4683	.0226469	.0297636
dadocc	.0080741	.0013762	5.87	4678	.0053839	.0107383
daded	.1427545	.0411568	3.47	5455	.0616242	.2235425
momed	.060508	.0379714	1.59	4707	-.0130428	.1339119
dadunemp	-.1224706	.0468447	-2.61	4503	-.2136019	-.0290561
male	-.0556571	.0281836	-1.97	4665	-.1103975	-.0002866
deprive	-.1562529	.0260999	-5.99	3698	-.2078412	-.1057263

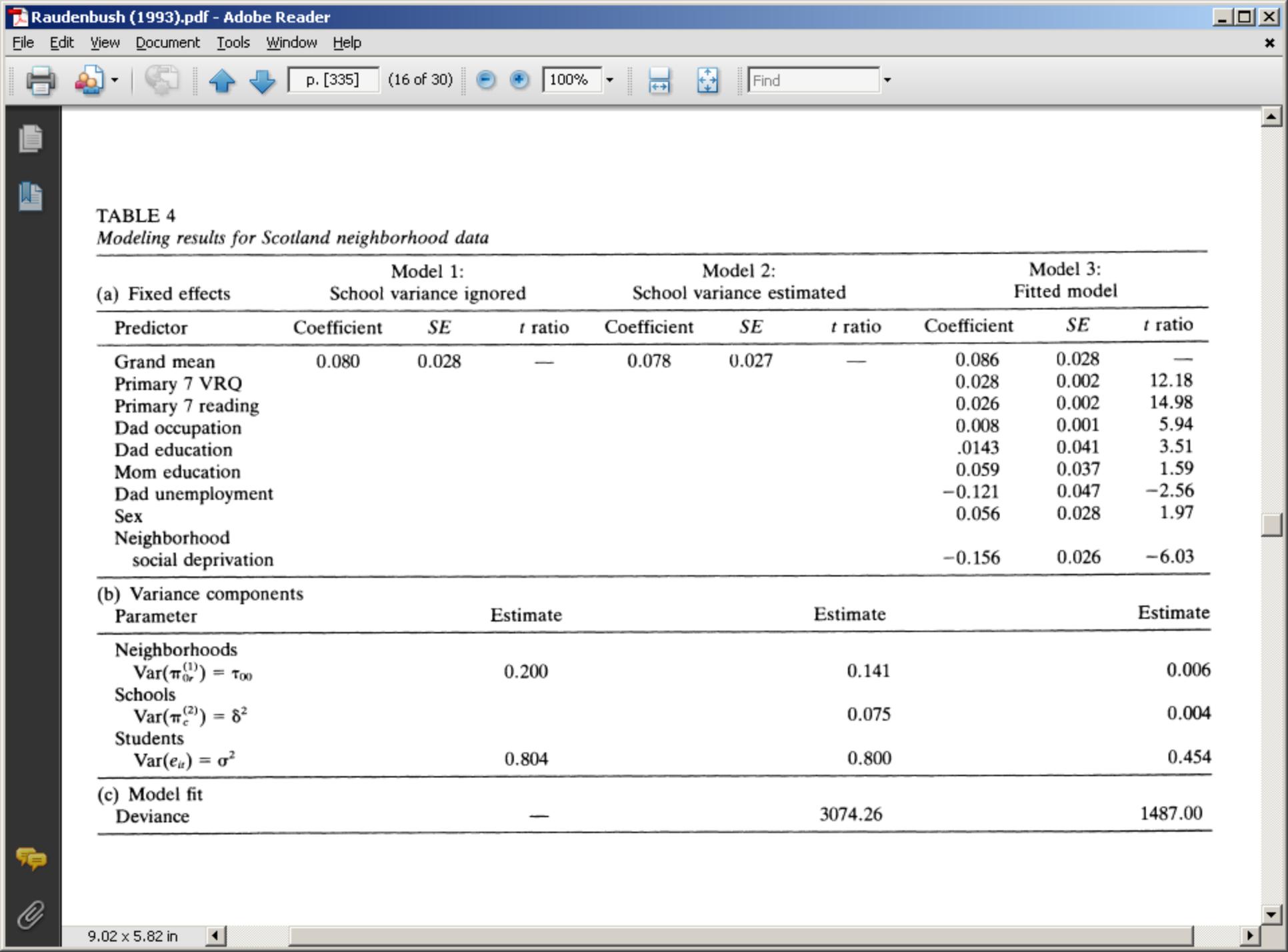


TABLE 4

Modeling results for Scotland neighborhood data

(a) Fixed effects	Model 1: School variance ignored			Model 2: School variance estimated			Model 3: Fitted model		
	Coefficient	SE	t ratio	Coefficient	SE	t ratio	Coefficient	SE	t ratio
Grand mean	0.080	0.028	—	0.078	0.027	—	0.086	0.028	—
Primary 7 VRQ							0.028	0.002	12.18
Primary 7 reading							0.026	0.002	14.98
Dad occupation							0.008	0.001	5.94
Dad education							.0143	0.041	3.51
Mom education							0.059	0.037	1.59
Dad unemployment							-0.121	0.047	-2.56
Sex							0.056	0.028	1.97
Neighborhood social deprivation							-0.156	0.026	-6.03
(b) Variance components									
Parameter	Estimate			Estimate			Estimate		
Neighborhoods $\text{Var}(\pi_{0r}^{(1)}) = \tau_{00}$	0.200			0.141			0.006		
Schools $\text{Var}(\pi_c^{(2)}) = \delta^2$				0.075			0.004		
Students $\text{Var}(e_{it}) = \sigma^2$	0.804			0.800			0.454		
(c) Model fit									
Deviance	—			3074.26			1487.00		



```
. estimates table model1b model2b model3b, stats(deviance) keep("FP1:" "RP2:" "RP3:" "RP1:")
> b(%9.3f) style(online)
```

Variable	model1b	model2b	model3b
FP1			
cons	0.083	0.096	0.035
p7vrq			0.028
p7read			0.026
dadocc			0.008
daded			0.143
momed			0.061
dadunemp			-0.122
female			0.056
deprive			-0.156
RP2			
var(cons)	0.203	0.142	0.005
RP3			
var(cons)		0.100	0.006
RP1			
var(cons)	0.806	0.800	0.457
Statistics			
deviance			

```
.
.
.
.
```

WORK EFFICIENTLY



Bristol_e-Stat.do

▼ ×

```
52
53 * Open the Raudenbush (1993) Scottish neighbourhood study data
54 use "http://www.stata-press.com/data/mlmus2/neighborhood.dta", clear
55
56 * Replicate Table 1 of Raudenbush (1993). The table shows that the data are
57 * cross-classified. Students are said to be nested within the cross-
58 * classification of schools by neighbourhoods. To account for both school
59 * and neighbourhood effects in our models of student attainment, we will
60 * need to fit cross-classified multilevel model using runmlwin.
61 table neighid schid if inrange(neighid,26,38) | inrange(neighid,251,263) ///
62 | inrange(neighid,793,800)
63
64 * Replicate Table 3 of Raudenbush (1993). The table gives summary statistics.
65 egen pickone = tag(neighid)
66
67 tabstat deprive if pickone==1, stat(mean sd) format(%4.3f) columns(stats)
68
69 tabstat attain p7vrq p7read dadocc daded momed dadunemp male, ///
70 stat(mean sd) format(%4.3f) columns(stats)
71
72
73 * Generate a unique student identifier variable which will be the level 1
74 * unit identifier variable in the runmlwin command
75 gen studentid = _n
76
77 * Generate a variable cons to act as the constant or intercept variable in
78 * the runmlwin models
79 gen cons = 1
80
81 * Sort the data by students within neighbourhoods (otherwise runmlwin will
82 * complain when we try to fit the following model)
83 sort neighid studentid
84
85 * Fit a two-level (students within neighbourhoods) variances components
86 * model to attain. This model is referred to as model 1 in Table 4 of
87 * Raudenbush (1993). Note, you will need to click the "Resume Macro" button
```

RESOURCES TO HELP YOU LEARN
RUNMLWIN

help runmlwin

Title

runmlwin - Run the MLWIN multilevel modelling software from within Stata

Syntax

```
runmlwin responses_and_fixed_part, random_part [discrete(discrete_options)] [mcmc(mcmc_options)]
[general_options]
```

where the syntax of *responses_and_fixed_part* is one of the following

for univariate continuous, binary, proportion and count response models

```
depvar indepvars [if] [in]
```

for univariate ordered and unordered categorical response models

```
depvar indepvars1 [(indepvars2, contrast(numlist)) ... ] [if] [in]
```

where *indepvars1* are those independent variables which appear with separate coefficients in each of every log-odds contrast, while *indepvars2* are those independent variables which appear with common coefficients for those log-odds contrasts specified in **contrast**(*numlist*). Contrasts can be thought of as the separate "subequations" or "arms" of a multinomial response model. These contrasts are indexed 1,2,... up to the total number of contrasts included in the model. The total number of contrasts will be one less than the number of response categories.

for multivariate response models

```
(depvar1 indepvars1, equation(numlist))
(depvar2 indepvars2, equation(numlist))
[(depvar3 indepvars3, equation(numlist))]
[... ]
[if] [in]
```

where **equation**(*numlist*) specifies equation numbers. Equation numbers are indexed 1,2,... up to the total number of equations (i.e. response variables) included in the model.

and the syntax of *random_part* is

```
[ ... ] [level2(levelvar: [varlist] [, random_part_options])]
level1(levelvar: [varlist] [, random_part_options])
```

Examples

IMPORTANT. The following examples will only work on your computer once you have installed MLwin and once you have told **runmlwin** what the mlwin.exe file address is. See *Remarks on installation instructions* above for more information.

(a) Continuous response models

Two-level models

Setup

- use <http://www.bristol.ac.uk/cmm/media/runmlwin/tutorial>, clear

Two-level random-intercept model, analogous to xtreg (fitted using IGLS)

(See page 28 of the MLwin User Manual)

(You will need to click the "Resume macro" button twice in MLwin to fit the model.)

- `runmlwin normexam cons standlrt, level2(school: cons) level1(student: cons)`

Two-level random-intercept and random-slope (coefficient) model (fitted using IGLS)

(See page 59 of the MLwin User Manual)

- `runmlwin normexam cons standlrt, level2 (school: cons standlrt) level1 (student: cons)`

Refit the model suppressing the two pauses in MLwin (fitted using IGLS)

(See page 59 of the MLwin User Manual)

- `runmlwin normexam cons standlrt, level2 (school: cons standlrt) level1 (student: cons) nopause`

Refit the model, where this time we additionally calculate the level 2 residuals (fitted using IGLS)

(See page 59 of the MLwin User Manual)

- `runmlwin normexam cons standlrt, level2 (school: cons standlrt, residuals(u)) level1 (student: cons)`

Two-level random-intercept and random-slope (coefficient) model with a complex level 1 variance function (fitted using IGLS)

(See page 99 of the MLwin User Manual)

- `matrix A = (1,1,0,0,0,1)`
- `runmlwin normexam cons standlrt girl, level2(school: cons standlrt) level1(student: cons standlrt girl, elements(A))`

Two-level random-intercept and random-slope (coefficient) model using MCMC (where we first fit the model using IGLS to obtain initial values for the MCMC chains)

(See page 71 of the MLwin MCMC Manual)

- `runmlwin normexam cons standlrt, level2 (school: cons standlrt) level1 (student: cons)`
- `runmlwin normexam cons standlrt, level2 (school: cons standlrt) level1 (student: cons) mcmc(on) initsprevious`

Multivariate response models

Bristol University | Centre for Multilevel Modelling | runmlwin: Running MLwiN from within Stata - Mozilla Firefox

File Edit View History Bookmarks Title Tools Help

Bristol University | Centre for Multilevel Mod... +

http://www.bristol.ac.uk/cmm/software/runmlwin/ W Wiktionary (en)

skip to content university home | study | research | global | contacting people | a-z index | news | help search search

University of BRISTOL Centre for Multilevel Modelling

Centre for Multilevel Modelling home Contacts News MLwiN Online course

SOFTWARE

- MLwiN
- Realcom
- MLPowSim
- runmlwin**

- Presentations
- Examples
- Citations
- User Forum

CMM software support

University home > Centre for Multilevel Modelling... > Software > runmlwin

runmlwin: Running MLwiN from within Stata

runmlwin is a Stata command which allows Stata users to run the powerful MLwiN multilevel modelling software from within Stata.

The multilevel models fitted by **runmlwin** are often considerably faster than those fitted by the Stata's **xtmixed**, **xtmelogit** and **xtmepoisson** commands. The range of models which can be fitted by **runmlwin** is also much wider than those commands. **runmlwin** also allows fast estimation on large data sets for many of the more complex multilevel models available through the user written **gllamm** command.

MLwiN has the following features:

1. Estimation of multilevel models for continuous, binary, count, ordered categorical and unordered categorical data
2. Fast estimation via classical and Bayesian methods
3. Estimation of multilevel models for cross-classified and multiple membership nonhierarchical data structures
4. Estimation of multilevel multivariate response models, multilevel spatial models, multilevel measurement error models and multilevel multiple imputation models

These details with a screen shot are available on our **runmlwin** [leaflet](#) (pdf, 0.1mb)

Presentations

We have provided a range of presentations showcasing **runmlwin**. These presentations provide a quick overview of how the command works and the range of models which can be fitted. [More >>](#)

Download



SOFTWARE University home > Centre for Multilevel Modelling... > Software > runmlwin > Presentations

MLwiN
Realcom
MLPowSim
runmlwin
→ Presentations
→ Examples
→ Citations
→ User Forum
CMM software support

Presentations using runmlwin

- UK Stata Users' Group, 17th Meeting (16th September 2011)
 - [Slides](#) (PDF, 2.0mb)
 - [Stata do-file](#) (do, 0.1mb) to replicate all analyses presented in the slides.
- University of Bristol, Mplus/MLwiN User Group (MUGS) meeting (14th June 2011)
 - [Slides](#) (PDF, 2.3mb)
 - [Stata do-file](#) (do, 0.1mb) to replicate all analyses presented in the slides.
- Modern Modeling Methods (M3) Conference, University of Connecticut (26th May 2011)
 - [Slides](#) (PDF, 3.2mb)
 - [Stata do-file](#) (do, 0.1mb) to replicate all analyses presented in the slides.
- 2011 American Sociological Association Spring Methodology Conference, Tilburg University (20th May 2011)
 - [Slides](#) (PDF, 2.0mb)
 - [Stata do-file](#) (do, 0.1mb) to replicate all analyses presented in the slides.
- University of Bristol, e-Stat meeting (7th April 2011)
 - [Slides](#) (PDF, 1.7mb)
 - [Stata do-file](#) (do, 0.1mb) to replicate all analyses presented in the slides.
- 8th International Amsterdam Multilevel Conference (17th March 2011)



skip to content

[university home](#) | [study](#) | [research](#) | [global](#) | [contacting people](#) | [a-z index](#) | [news](#) | [help](#) 

Centre for Multilevel Modelling

[Centre for Multilevel Modelling home](#) | [Contacts](#) | [News](#) | [MLwiN](#) | [Online course](#)

SOFTWARE

MLwiN

Realcom

MLPowSim

runmlwin

→ Presentations

→ **Examples**

→ Citations

→ User Forum

CMM software support

[University home](#) > [Centre for Multilevel Modelling...](#) > [Software](#) > [runmlwin](#) > [Examples](#) Examples using runmlwin

MLwiN User Manual

These do-files and log files replicate the analyses reported in the [MLwiN User Manual](#) (PDF, 4.6 mb) Rasbash, J., Steele, F., Browne, W.J. and Goldstein, H. (2009) Centre for Multilevel Modelling, University of Bristol.

Note that we have not created do-files for Chapters 1, 8 or 19 of the manual as no models are fitted in those chapters. We have also not yet attempted to replicate the analysis in Chapter 17.

- 1 - Introducing Multilevel Models
- 2 - Introduction to Multilevel Modelling ([do](#) | [log](#))
- 3 - Residuals ([do](#) | [log](#))
- 4 - Random Intercept and Random Slope Models ([do](#) | [log](#))
- 5 - Graphical Procedures for Exploring the Model ([do](#) | [log](#))
- 6 - Contextual Effects ([do](#) | [log](#))
- 7 - Modelling the Variance as a Function of Explanatory Variables ([do](#) | [log](#))
- 8 - Getting Started with your Data
- 9 - Logistic Models for Binary and Binomial Responses ([do](#) | [log](#))
- 10 - Multinomial Logistic Models for Unordered Categorical Responses ([do](#) | [log](#))
- 11 - Fitting an Ordered Category Response Model ([do](#) | [log](#))
- 12 - Modelling Count Data ([do](#) | [log](#))
- 13 - Fitting Models to Repeated Measures Data ([do](#) | [log](#))
- 14 - Multivariate Response Models ([do](#) | [log](#))

- Rasbash, J., Charlton, C., Browne, W.J., Healy, M. and Cameron, B. 2009. MLwiN Version 2.1. Centre for Multilevel Modelling, University of Bristol.

For models fitted using MCMC estimation, we ask that you additionally cite:

- Browne, W.J. 2009. MCMC Estimation in MLwiN, v2.13. Centre for Multilevel Modelling, University of Bristol.

Papers using runmlwin

Please let George Leckie (g.leckie@bristol.ac.uk) know of any further publications using **runmlwin** including forthcoming papers, books, PhD theses, etc.

- Cheung, C., Goodman, D., Leckie, G. and Jenkins, J. (2011) [Understanding Contextual Effects on Externalizing Behaviors in Children in Out-of-home Care: Influence of Workers and Foster Families](#). *Children and Youth Services Review*, 33, 2050-2060.
- Chung, H. and Beretvas, S.N. (2011) [The Impact of ignoring multiple membership data structures in multilevel models](#). *British Journal of Mathematical and Statistical Psychology*. *Forthcoming*.
- Leckie, G. and Baird, J.-A. (2011) [Rater effects on essay scoring: A multilevel analysis of severity drift, central tendency and rater experience](#). *Journal of Educational Measurement*. *Forthcoming*.
- Leckie, G., Pillinger, R., Jones, K. and Goldstein, H. (2011) [Multilevel modelling of social segregation](#). *Journal of Educational and Behavioral Statistics*. *Forthcoming*.
- Paternoster, L., Howe, L. D., Tilling, K., Weedon, M. N., Freathy, R. M., Frayling, T. M., Kemp, J. P., Davey Smith, G., Timpson, N. J. Ring, S. M., Evans, D. M. and Lawlor, D. A. (2011) [Adult height variants affect birth length and growth rate in children](#). *Human Molecular Genetics*. *Forthcoming*

Books discussing runmlwin

- Snijders, T. and Bosker, R. (2011) [Multilevel Analysis: An Introduction to Basic and Advanced Multilevel Modeling](#), Second Edition. *Sage*. *Forthcoming*.



runmlwin user forum

Forum rules

NEWTOPIC*

Search this forum...

Search

20 topics • Page 1 of 1

ANNOUNCEMENTS

REPLIES

VIEWS

LAST POST

**Do-files to replicate entire MLwiN User & MCMC Manuals**

by GeorgeLeckie » Mon Apr 18, 2011 5:30 pm

0

123

by GeorgeLeckie

Mon Apr 18, 2011 5:30 pm

**Welcome to the runmlwin discussion forum**

by GeorgeLeckie » Fri Apr 01, 2011 4:06 pm

0

130

by GeorgeLeckie

Fri Apr 01, 2011 4:06 pm

TOPICS

REPLIES

VIEWS

LAST POST

**MVs & error message 'line too long'**

by julia1633 » Mon Aug 15, 2011 3:17 pm

1 2

17

125

by julia1633

Sun Aug 28, 2011 12:05 am

**runmlwin in Batch mode - gui causing error?**

by ash » Sat Aug 27, 2011 6:43 am

2

20

by ash

Sat Aug 27, 2011 9:07 pm

**Bug in residuals(u, savechains("u.dta", replace)) ?**

by ash » Mon Aug 01, 2011 7:06 pm

4

76

by GeorgeLeckie

Wed Aug 03, 2011 6:25 pm

**Predictions via the runmlwin interface: a clarification**

by ewancarr » Tue Jul 26, 2011 6:49 pm

6

84

by GeorgeLeckie

Wed Jul 27, 2011 7:04 pm

**highly correlated multivariate dependents -> numerical error**

by ash » Sat Jul 23, 2011 10:48 am

1

67

by GeorgeLeckie

Mon Jul 25, 2011 3:49 pm

**Input dataset contains double precision data...**

by ewancarr » Wed Jul 13, 2011 3:55 pm

2

81

by ewancarr

Wed Jul 13, 2011 9:16 pm

**Modelling Count Data (example do-file) - mismatch error**

by leap » Tue Jul 12, 2011 10:18 am

1

48

by ChrisCharlton

Wed Jul 13, 2011 3:32 pm

**Error code: r(-1073740777);**

by pd65 » Mon Jul 04, 2011 11:01 am

7

207

by GeorgeLeckie

Thu Jul 07, 2011 3:15 pm

**MCMC estimation**

by janna » Fri Apr 08, 2011 9:20 am

5

293

by ChrisCharlton

Thu Jul 09, 2011 10:09 am

Citing `runmlwin`

- If you use `runmlwin` in your work, please cite `runmlwin`
- Leckie, G. and Charlton, C. (2011) *runmlwin: Stata module for fitting multilevel models in the MLwiN software package*. Centre for Multilevel Modelling, University of Bristol.
- We can then add you to the list of papers using `runmlwin` on our website
- <http://www.bristol.ac.uk/cmm/software/runmlwin/citations>