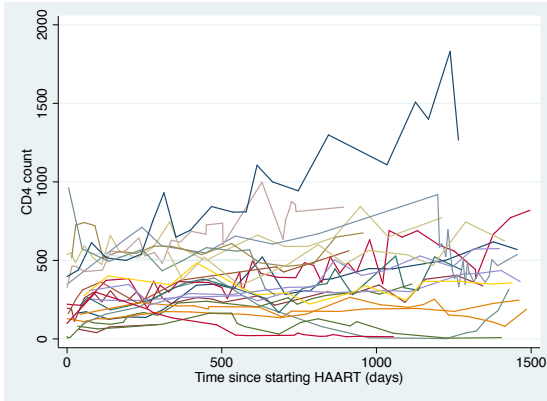


Longitudinal Data

Fall 2015



Chapter 7, Part II

Mixed, Random Effects, Random Coefficients, Multilevel, ...**Models**

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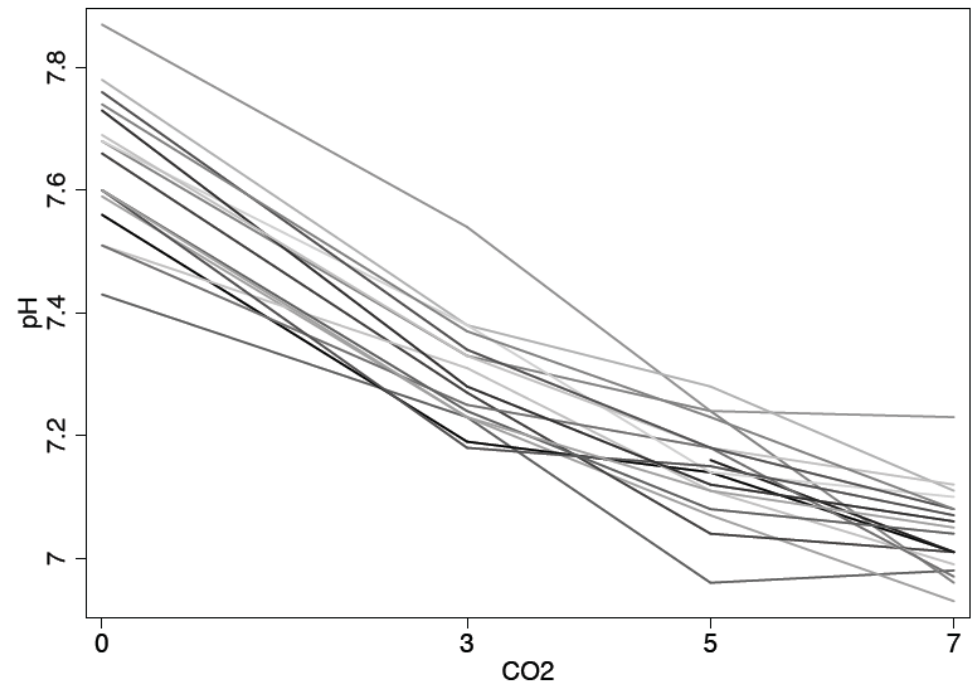
Hierarchical Models

pH vs. CO₂

- Functional ophthalmologic study of the impact of CO₂ levels on pH in eyes (Cohen, et al., 1992).
- 18 subjects, at two different visits, were fit with specialized goggles that regulated exposure to CO₂ (different levels for each eye) and subsequent measures were made of pH.
- Thus, the data consist of $18 \times 4 = 72$ nested measurements of eyes within visits within subjects.
- The outcome, pH, to be Y_{ijk} , for the i th person, j th visit, k th eye: $i=(1, \dots, m=18), j=(1, 2), k=(1, 2)$.

Data

	id	co2	ph	prph	visit	eye
1	1	0	7.432	76.4	1	1
2	1	5	6.956	59.4	1	2
3	1	3	7.232	80.3	2	1
4	1	7	6.982	74.2	2	2
5	2	0	7.594	67.8	1	1
6	2	7	6.929	61.1	1	2
7	2	3	7.230	73.4	2	1
8	2	5	7.073	65.9	2	2
9	3	3	7.329	87.6	1	1
10	3	7	6.963	55.8	1	2



Fitting the following Sequence of Models

- Model 1

$$Y_{ijk} = \beta_0 + \beta_1 X_{ijk} + \beta_2 (X_{ijk})^2 + e_{ijk}$$

- Model 2

$$Y_{ijk} = \beta_0 + \beta_{0i} + \beta_1 X_{ijk} + \beta_2 (X_{ijk})^2 + e_{ijk}$$

- Model 3

$$Y_{ijk} = \beta_0 + \beta_{0i} + \beta_{0ij} + \beta_1 X_{ijk} + \beta_2 (X_{ijk})^2 + e_{ijk}$$

- Model 4

$$Y_{ijk} = \beta_0 + \beta_{0i} + (\beta_1 + \beta_{1i})X_{ijk} + \beta_2 (X_{ijk})^2 + e_{ijk}$$

Model 1

$$Y_{ijk} = \beta_0 + \beta_1 X_{ijk} + \beta_2 (X_{ijk})^2 + e_{ijk}$$

- $E(e_{ijk})=0$, $cov(e_{ijk}, e_{ij'k'})=0$, $j \neq j'$ or $k \neq k'$.
- $Var(e_{ijk})= \sigma_e^2$.
- What does V_0 look like for this model?
- How does one interpret the parameters, β_0 , β_1 , and β_2 ?

Model 1

```
.mixed ph co2 co22, stddev reml
```

```
Mixed-effects REML regression                Number of obs      =           70

                                                Wald chi2(2)           =       463.58
Log restricted-likelihood =  58.499944        Prob > chi2             =       0.0000
```

```
-----+-----
      ph |      Coef.   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
      co2 |   -.1373144   .0137997    -9.95   0.000    - .1643613   - .1102674
     co22 |    .0073496   .0019106     3.85   0.000     .0036049   .0110942
    _cons |    7.646126   .0212018   360.64   0.000     7.604572   7.687681
-----+-----
```

```
-----+-----
Random-effects Parameters |   Estimate  Std. Err.      [95% Conf. Interval]
-----+-----
      sd(Residual) |   .0883141   .0074639     .0748325   .1042245
-----+-----
```

```
estat ic
```

```
-----+-----
      Model |      Obs   ll(null)   ll(model)      df         AIC         BIC
-----+-----
     model1 |       70         .   58.49994        4   -108.9999   -100.0059
-----+-----
```

Getting Measures of Association

```
. lincom 3*co2+9*co22      E(Y|CO2=3) - E(Y|CO2=0)
```

$$(1) \quad 3 \cdot [\text{ph}]_{\text{co2}} + 9 \cdot [\text{ph}]_{\text{co22}} = 0$$

ph	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
(1)	-.3457971	.0254904	-13.57	0.000	-.3957574	-.2958368

```
. lincom 5*co2+25*co22      E(Y|CO2=5) - E(Y|CO2=0)
```

$$(1) \quad 5 * [\text{ph}]_{\text{co2}} + 25 * [\text{ph}]_{\text{co22}} = 0$$

ph	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
(1)	-.502833	.0273284	-18.40	0.000	-.5563957 -.4492703

```
. lincom 7*co2+49*co22      E(Y|CO2=7)-E(Y|CO2=0)
```

$$(1) \quad 7 \cdot [\text{ph}]_{\text{co2}} + 49 \cdot [\text{ph}]_{\text{co22}} = 0$$

ph	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
(1)	-.6010725	.0286516	-20.98	0.000	-.6572287 -.5449163

Model 2

$$Y_{ijk} = \beta_0 + \beta_{0i} + \beta_1 X_{ijk} + \beta_2 (X_{ijk})^2 + e_{ijk}$$

- How does one interpret the parameters, β_0 , β_{0i} , and β_1 , β_2 ?
- $E(e_{ijk})=0$, $cov(e_{ijk}, e_{ij'k'})=0$, $j \neq j'$ or $k \neq k'$
- $Var(e_{ijk}) = \sigma^2_e$
- $Var(\beta_{0i}) = \sigma^2_{\beta_{0i}}$.
- $cov(e_{ijk}, \beta_{0i}) = 0$.
- What does V_0 look like for this model?

Model 2

```
.mixed ph co2 co22 || id:, stddev reml
```

Wald chi2(2) = 1346.68

Log restricted-likelihood = 76.647278

```
Prob > chi2      =    0.0000
```

ph	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
co2	-.1372775	.0080292	-17.10	0.000	-.1530143 -.1215406
co22	.0073546	.0011116	6.62	0.000	.0051759 .0095334
_cons	7.64567	.0209821	364.39	0.000	7.604546 7.686795

Random-effects Parameters		Estimate	Std. Err.	[95% Conf. Interval]	
id: Identity					
	sd(_cons)	.0715285	.0138821	.0488968	.1046354
	sd(Residual)	.0513741	.0051277	.0422459	.0624746

LR test vs. linear regression: $\chi^2(01) = 36.29$ Prob $\geq \chi^2 = 0.0000$

Residual intraclass correlation

Level	ICC	Std. Err.	[95% Conf. Interval]
id	.6596925	.1004796	.4463808 .8233416

Model	Obs	ll (null)	ll (model)	df	AIC	BIC
model2	70	.	76.64728	5	-143.2946	-132.0521

Model 3

$$Y_{ijk} = \beta_0 + \beta_{0i} + \beta_{0ij} + \beta_1 X_{ijk} + \beta_2 (X_{ijk})^2 + e_{ijk}$$

- How does one interpret the parameters, β_0 , β_{0i} , β_{0ij} and β_1 , β_2 ?
- $E(e_{ijk})=0$, $cov(e_{ijk}, e_{ij'k'})=0$, $j \neq j'$ or $k \neq k'$
- $Var(e_{ijk})= \sigma_e^2$
- $Var(\beta_{0i})= \sigma_{id}^2$
- $Var(\beta_{0ij})= \sigma_{visit}^2$
- $cov(e_{ijk}, \beta_{0i})= 0$, $cov(e_{ijk}, \beta_{0ij})= 0$, $cov(\beta_{0i}, \beta_{0ij})= 0$
- What does V_0 look like for this model?

Model 3

```
mixed ph co2 co22 || id: || visit:, stddev
```

		Observations per Group			
Group Variable	No. of Groups	Minimum	Average	Maximum	
id	18	2	3.9	4	
visit	35	2	2.0	2	

```
Log likelihood = 91.091123      Wald chi2(2)      = 1456.58
                                Prob > chi2      = 0.0000
```

ph	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
co2	-.1364223	.0078051	-17.48	0.000	-.1517199	-.1211246
co22	.0072612	.0010794	6.73	0.000	.0051455	.0093768
_cons	7.644394	.0203644	375.38	0.000	7.604481	7.684308

Random-effects Parameters		Estimate	Std. Err.	[95% Conf. Interval]	
id: Identity					
	sd(_cons)	.0684105	.0134426	.0465439	.1005501
visit: Identity					
	sd(_cons)	.0208853	.0144296	.005392	.0808969
	sd(Residual)	.0474441	.005692	.0375026	.0600211

```
LR test vs. linear regression:      chi2(2) = 38.01  Prob > chi2 = 0.0000
```

Model 3

estat icc

Residual intraclass correlation

Level	ICC	Std. Err.	[95% Conf. Interval]	
id	.6352526	.1100754	.407022	.8154659
visit id	.6944611	.0985583	.4776825	.8495975

```
. estat ic
```

Model	Obs	ll (null)	ll (model)	df	AIC	BIC
.	70	.	91.09112	6	-170.1822	-156.6913

Model 4

$$Y_{ijk} = \beta_0 + \beta_{0i} + (\beta_1 + \beta_{1i})X_{ijk} + \beta_2 (X_{ijk})^2 + e_{ijk}$$

- $E(e_{ijk})=0$, $cov(e_{ijk}, e_{ij'k'})=0$, $j \neq j'$ or $k \neq k'$
- $Var(e_{ijk}) = \sigma_e^2$
- $Var(\beta_{0i}) = \sigma_0^2$, $Var(\beta_{1i}) = \sigma_1^2$
- $cov(e_{ijk}, \beta_{0i}) = 0$, $cov(e_{ijk}, \beta_{1i}) = 0$, $cov(\beta_{0i}, \beta_{1i}) = \sigma_{01}$.
- What does variance-covariance of random effects look like?
- How does one interpret the parameters, β_0 , β_{0i} , β_{1i} , and β_1 , β_2 ?

Model 4

```
.mixed ph co2 co22 || id: co2, covariance(un)
```

```
Log likelihood = 94.57083      Wald chi2(2)      = 1109.29
                                Prob > chi2      = 0.0000
```

ph	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
co2	-.1372451	.0074464	-18.43	0.000	-.1518399	-.1226504
co22	.0073501	.0010082	7.29	0.000	.005374	.0093262
_cons	7.645654	.0249892	305.96	0.000	7.596676	7.694632

Random-effects Parameters		Estimate	Std. Err.	[95% Conf. Interval]	
id: Unstructured					
	var(co2)	.0000404	.0000312	8.88e-06	.0001837
	var(_cons)	.0088415	.0035064	.004064	.0192352
	cov(co2,_cons)	-.0005976	.0003306	-.0012456	.0000504
	var(Residual)	.002172	.0004247	.0014805	.0031864

```
LR test vs. linear regression:      chi2(3) = 44.97  Prob > chi2 = 0.0000
```

Model 4

```
. estat icc
```

Conditional intraclass correlation

Level	ICC	Std. Err.	[95% Conf. Interval]	
id	.8027868	.0711758	.6277709	.9076223

```
. estat ic
```

Akaike's information criterion and Bayesian information criterion

Model	Obs	ll(null)	ll(model)	df	AIC	BIC
.	70	.	94.57083	7	-175.1417	-159.4022

Comparing Models

Table 7.4: ESTIMATES(SE) AND AKAIKE'S INFORMATION CRITERION (AIC) OF FOUR HIERARCHICAL LINEAR MIXED EFFECTS MODELS FOR PH VERSUS $C0_2$

	Model 1		Model 2		Model 3		Model 4	
Parameter	Est	SE	Est	SE	Est	SE	Est	SE
β_0	7.6	.021	7.6	.021	7.6	.020	7.6	.025
β_1	-.14	.014	-.14	.008	-.14	.0078	-.14	.0074
β_2	.0073	.0019	.0074	.0011	.0073	.0011	.0074	.0010
$E(Y_{ijk} X_{ijk} = 5) - E(Y_{ijk} X_{ijk} = 0)$	-.50	.027	-.50	.016	-.50	.015	-.50	.016
$\sigma_{\beta_{0i}}$.072	.014	.068	.014	.094	.019
$\sigma_{\beta_{0ij}}$.021	.013		
$\sigma_{\beta_{1i}}$.0064	.0025
σ_e	.088	.0075	.051	.0051	.047	.0057	.047	.0046
<i>AIC</i>	-109		-143		-170		-175	

Model 4, Robust

```
.mixed ph co2 co22 || id: co2, covariance(un) stddev ro
```

```
Log pseudolikelihood = 94.57083      Wald chi2(2)      = 1260.49
                                Prob > chi2      = 0.0000
```

(Std. Err. adjusted for 18 clusters in id)

		Robust				
	ph	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
co2		-.1372451	.0080234	-17.11	0.000	-.1529707 -.1215196
co22		.0073501	.00105	7.00	0.000	.0052922 .0094081
_cons		7.645654	.0263427	290.24	0.000	7.594023 7.697285

Random-effects Parameters		Robust		
		Estimate	Std. Err.	[95% Conf. Interval]
id: Unstructured				
	sd(co2)	.0063553	.0014507	.004063 .009941
	sd(_cons)	.0940291	.0169891	.0659884 .1339852
	corr(co2,_cons)	-1	7.38e-09	-1 -.9999999
	sd(Residual)	.0466047	.0038592	.0396228 .0548169

Naïve vs Robust

Table 7.5: COMPARISONS OF NAIVE AND ROBUST INFERENCE FOR MODEL 4

	Est	SE Naive	SE Robust
β_0	7.6	.025	.026
β_1	-.14	.0074	.0080
β_2	.0074	.0010	.0010
$E(Y_{ijk} X_{ijk} = 5) - E(Y_{ijk} X_{ijk} = 0)$	-.50	.016	.017
$\sigma_{\beta_{0i}}$.094	.019	.017
$\sigma_{\beta_{1i}}$.0064	.0025	.0014
σ_e	.047	.0046	.0039