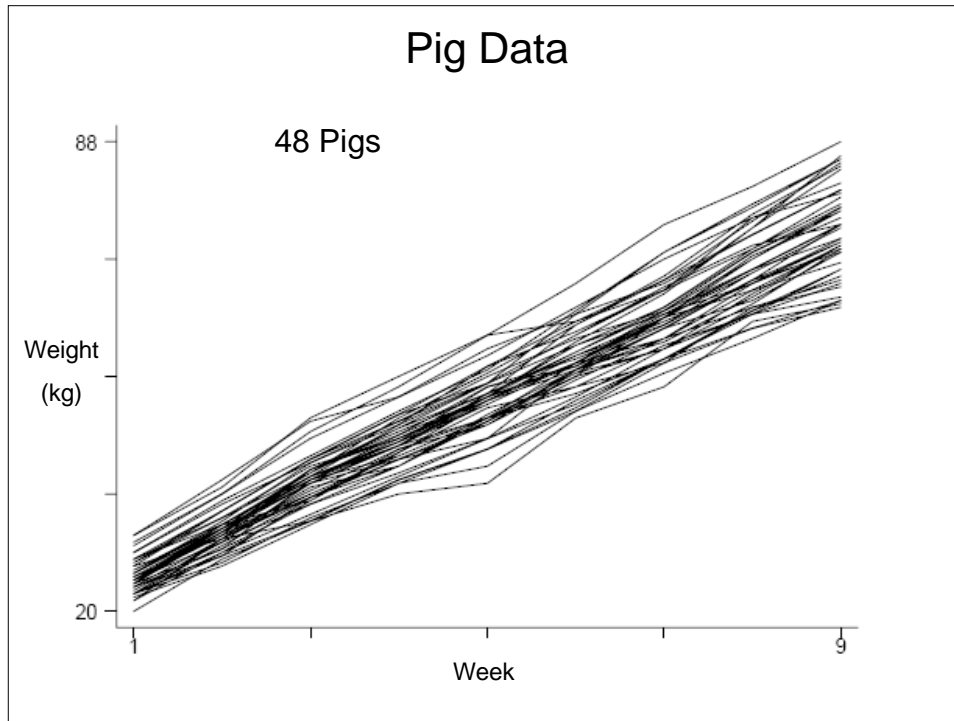


Lecture 3

Linear random intercept models

Example: Weight of Guinea Pigs

- **Body weights of 48 pigs in 9 successive weeks of follow-up (Table 3.1 DLZ)**
- **The response is measures at n different times, or under n different conditions. In the guinea pigs example the time of measurement is referred to as a "within-units" factor. For the pigs $n=9$**
- **Although the pigs example considers a single treatment factor, it is straightforward to extend the situation to one where the groups are formed as the results of a factorial design (for example, if the pigs were separated into males and female and then allocated to the diet groups)**



A) Linear model with random intercept

$$Y_{ij} = U_i + \beta_0 + \beta_1 t_j + \varepsilon_{ij}$$

$$U_i \sim N(0, \tau^2) \quad \text{Variance between}$$

$$\varepsilon_{ij} \sim N(0, \sigma^2) \quad \text{Variance within}$$

$$\rho = \frac{\tau^2}{\tau^2 + \sigma^2} \quad \text{Intraclass correlation coefficient}$$

Pigs data model 1 – OLS fit

. regress weight time

Source	SS	df	MS			
Model	111060.882	1	111060.882	Number of obs =	432	
Residual	8294.72677	430	19.2900622	F(1, 430) =	5757.41	
Total	119355.609	431	276.927167	Prob > F =	0.0000	
				R-squared =	0.9305	
				Adj R-squared =	0.9303	
				Root MSE =	4.392	

weight	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
time	6.209896	.0818409	75.88	0.000	6.049038	6.370754
_cons	19.35561	.4605447	42.03	0.000	18.45041	20.26081

OLS results

Pigs data model 1 – IND fit

. xtreg weight time, pa i(Id) corr(ind)

GEE population-averaged model		Number of obs =	432
Group variable:	Id	Number of groups =	48
Link:	identity	Obs per group: min =	9
Family:	Gaussian	avg =	9.0
Correlation:	independent	max =	9
Scale parameter:	19.20076	Wald chi2(1) =	5784.19
		Prob > chi2 =	0.0000
Pearson chi2(432):	8294.73	Deviance =	8294.73
Dispersion (Pearson):	19.20076	Dispersion =	19.20076

weight	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
time	6.209896	.0816513	76.05	0.000	6.049862	6.369929
_cons	19.35561	.4594773	42.13	0.000	18.45505	20.25617

Independence correlation model results

Example: Weight of Pigs

For this type of repeated measures study we recognize two sources of random variation

1. Between: There is heterogeneity between pigs, due for example to natural biological (genetic?) variation
2. Within: There is random variation in the measurement process for a particular unit at any given time. For example, on any given day a particular guinea pig may yield different weight measurements due to differences in scale (equipment) and/or small fluctuations in weight during a day

B) Marginal Model

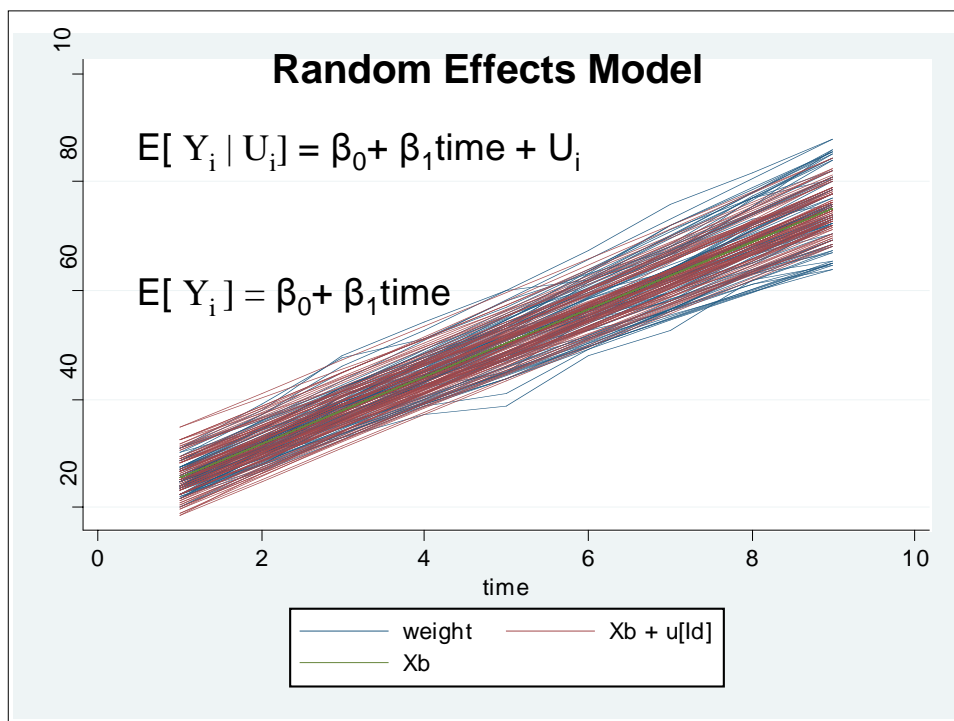
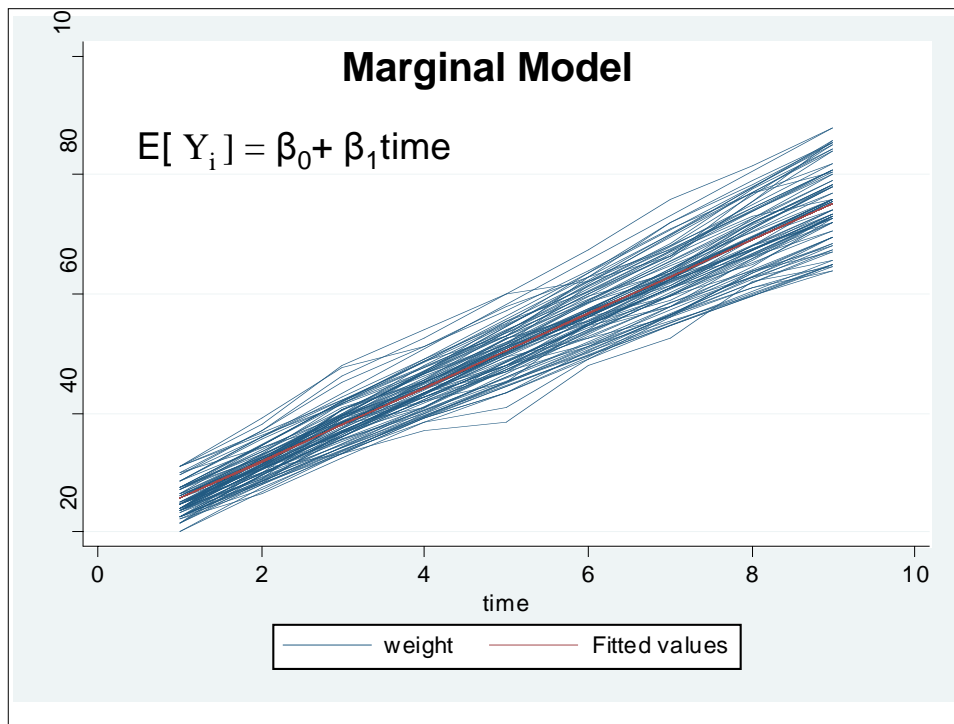
With a Uniform correlation structure

$$E[Y_{ij}] = \beta_0 + \beta_1 t_j$$

Model for the
mean

$$\text{cov}(Y_{ij}) = (\tau^2 + \sigma^2)[\rho \mathbf{1} \mathbf{1}' + (1 - \rho)I]$$

Model for the
covariance matrix



Models A and B are equivalent

$$E[Y_{ij} | U_i] = U_i + \beta_0 + \beta_1 t_j$$

$$E[Y_{ij}] = E[E[Y_{ij} | U_i]] = \beta_0 + \beta_1 t_j$$

$$\text{cov}(Y_{ij}) = \text{cov}[E[Y_{ij} | U_i]] + E[\text{cov}[Y_{ij} | U_i]]$$

$$\text{cov}[E[Y_{ij} | U_i]] = \text{cov}(1U_i) = \tau^2 11'$$

$$E[\text{cov}[Y_{ij} | U_i]] = E[\sigma^2 I] = \sigma^2 I$$

$$\text{cov}(Y_{ij}) = (\tau^2 + \sigma^2)[\rho 11' + (1 - \rho)I]$$

$$\rho = \frac{\tau^2}{\tau^2 + \sigma^2}$$

Pigs – Marginal model

```
xtreg weight time, pa i(Id) corr(exch)
```

Iteration 1: tolerance = 5.585e-15

```
GEE population-averaged model
Group variable:                Id
Link:                          identity
Family:                        Gaussian
Correlation:                   exchangeable
Scale parameter:               19.20076
Number of obs                  = 432
Number of groups               = 48
Obs per group: min             = 9
                             avg = 9.0
                             max = 9
Wald chi2(1)                   = 25337.48
Prob > chi2                    = 0.0000
```

weight	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
time	6.209896	.0390124	159.18	0.000	6.133433 6.286359
_cons	19.35561	.5974055	32.40	0.000	18.18472 20.52651

“Population Average”, Marginal Model with Exchangeable Correlation structure results

Pigs – RE model

```
xtreg weight time, re i(Id) mle
```

```
Random-effects ML regression      Number of obs   =    432
Group variable (i): Id           Number of groups =    48

Random effects u_i ~ Gaussian     Obs per group: min =    9
                                   avg   =    9.0
                                   max   =    9

Log likelihood = -1014.9268       LR chi2(1)      =  1624.57
                                   Prob > chi2      =    0.0000
```

weight	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
time	6.209896	.0390124	159.18	0.000	6.133433 6.286359
_cons	19.35561	.5974055	32.40	0.000	18.18472 20.52651
/sigma_u	3.84935	.4058114			3.130767 4.732863
/sigma_e	2.093625	.0755471			1.95067 2.247056
rho	.771714	.0393959			.6876303 .8413114

“Population Average”, Marginal Model with Exchangeable Correlation structure results

Pigs data model 1 – GEE fit

```
. xtgee weight time, i(Id) corr(exch)
```

```
Iteration 1: tolerance = 5.585e-15

GEE population-averaged model      Number of obs   =    432
Group variable:                    Id              Number of groups =    48
Link:                               identity           Obs per group: min =    9
Family:                             Gaussian         avg   =    9.0
Correlation:                        exchangeable     max   =    9

Scale parameter:                   19.20076       Wald chi2(1)    =  25337.48
                                   Prob > chi2      =    0.0000
```

weight	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
time	6.209896	.0390124	159.18	0.000	6.133433 6.286359
_cons	19.35561	.5974055	32.40	0.000	18.18472 20.52651

GEE fit – Marginal Model with Exchangeable Correlation structure results

Pigs data model 1 – GEE fit

```
. xtgee weight time, i(Id) corr(exch)  
. xtcorr
```

Estimated within-Id correlation matrix R:

	c9	c1	c2	c3	c4	c5	c6	c7	c8
r1	1.0000								
r2	0.7717	1.0000							
r3	0.7717	0.7717	1.0000						
r4	0.7717	0.7717	0.7717	1.0000					
r5	0.7717	0.7717	0.7717	0.7717	1.0000				
r6	0.7717	0.7717	0.7717	0.7717	0.7717	1.0000			
r7	0.7717	0.7717	0.7717	0.7717	0.7717	0.7717	1.0000		
r8	0.7717	0.7717	0.7717	0.7717	0.7717	0.7717	0.7717	1.0000	
r9	0.7717	0.7717	0.7717	0.7717	0.7717	0.7717	0.7717	0.7717	1.0000



GEE fit – Marginal Model with
Exchangeable Correlation structure results

One group polynomial growth curve model

- Similarly, if you want to fit a quadratic curve
 $E[Y_{ij} | U_i] = U_i + \beta_0 + \beta_1 t_j + \beta_2 t_j^2$

$$E(\mathbf{Y}_i) = \begin{pmatrix} 1 & t_1 & t_1^2 \\ 1 & t_2 & t_2^2 \\ \dots & & \\ 1 & t_n & t_n^2 \end{pmatrix} \begin{pmatrix} \beta_0 \\ \beta_1 \\ \beta_2 \end{pmatrix}$$

Pigs – RE model, quadratic trend

```
. gen timesq = time*time
. xtreg weight time timesq, re i(Id) mle
```

Random-effects ML regression
 Group variable (i): Id

Random effects u_i ~ Gaussian

Number of obs = 432
 Number of groups = 48

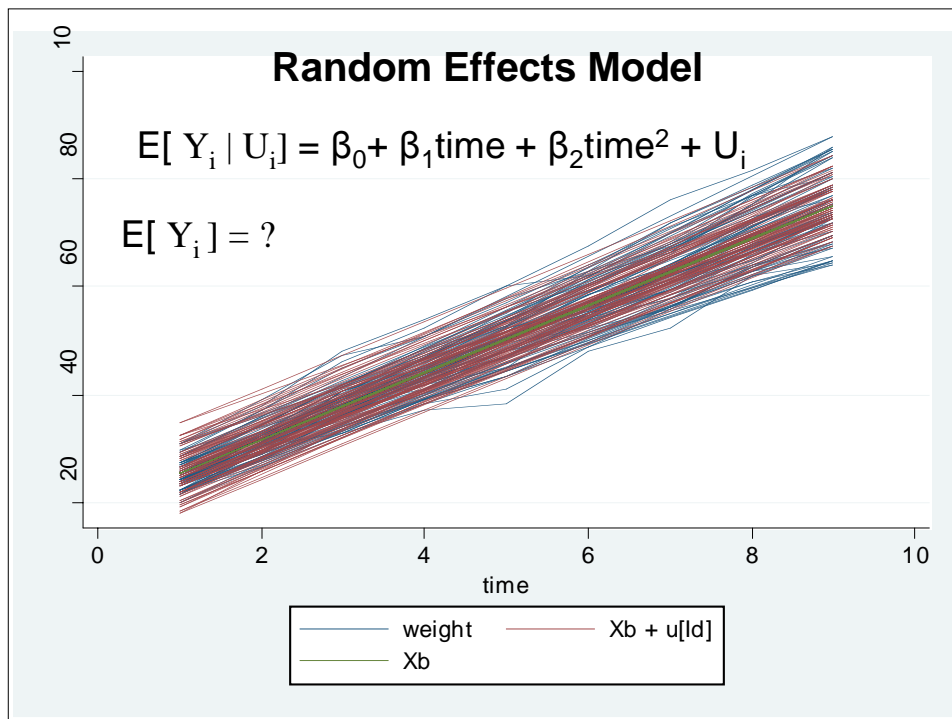
Obs per group: min = 9
 avg = 9.0
 max = 9

Log likelihood = -1014.5524

LR chi2(2) = 1625.32
 Prob > chi2 = 0.0000

weight	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
time	6.358818	.1763799	36.05	0.000	6.01312	6.704516
timesq	-.0148922	.017202	-0.87	0.387	-.0486075	.0188231
_cons	19.08259	.675483	28.25	0.000	17.75867	20.40651
/sigma_u	3.849473	.4057983			3.130909	4.732951
/sigma_e	2.091585	.0754733			1.948769	2.244866
rho	.7720686	.0393503			.6880712	.8415775

Exchangeable Correlation structure results



Pigs – Marg. model, quadratic trend

```
. xtgee weight time timesq, i(Id) corr(exch)
```

```
GEE population-averaged model
Group variable:          Id          Number of obs   =   432
Link:                   identity     Number of groups =    48
Family:                 Gaussian     Obs per group:  min =    9
Correlation:           exchangeable  avg             =   9.0
Scale parameter:      19.19317      max             =    9
Wald chi2(2)          = 25387.68
Prob > chi2           = 0.0000
```

weight	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
time	6.358818	.1763801	36.05	0.000	6.013119 6.704517
timesq	-.0148922	.017202	-0.87	0.387	-.0486076 .0188231
_cons	19.08259	.6754833	28.25	0.000	17.75867 20.40651

Exchangeable Correlation structure results

Pigs data model 1 – GEE fit

```
. xtcorr
```

Estimated within-Id correlation matrix R:

	c1	c2	c3	c4	c5	c6	c7	c8	c9
r1	1.0000								
r2	0.7721	1.0000							
r3	0.7721	0.7721	1.0000						
r4	0.7721	0.7721	0.7721	1.0000					
r5	0.7721	0.7721	0.7721	0.7721	1.0000				
r6	0.7721	0.7721	0.7721	0.7721	0.7721	1.0000			
r7	0.7721	0.7721	0.7721	0.7721	0.7721	0.7721	1.0000		
r8	0.7721	0.7721	0.7721	0.7721	0.7721	0.7721	0.7721	1.0000	
r9	0.7721	0.7721	0.7721	0.7721	0.7721	0.7721	0.7721	0.7721	1.0000

GEE fit – Marginal Model with
Exchangeable Correlation structure results

Pigs – Marginal model: AR(1)

```
xtgee weight time, i(Id) corr(AR1) t(time)
```

```
GEE population-averaged model
Group and time vars:      Id time      Number of obs   =   432
Link:                    identity      Number of groups =   48
Family:                  Gaussian      Obs per group: min =    9
Correlation:            AR(1)          avg             =   9.0
Scale parameter:        19.26754      max             =    9
Wald chi2(1)            = 6254.91
Prob > chi2              = 0.0000
```

weight	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
time	6.272089	.0793052	79.09	0.000	6.116654 6.427524
_cons	18.84218	.6745715	27.93	0.000	17.52004 20.16431

GEE-fit Marginal Model with AR1 Correlation structure

Pigs – RE model: AR(1)

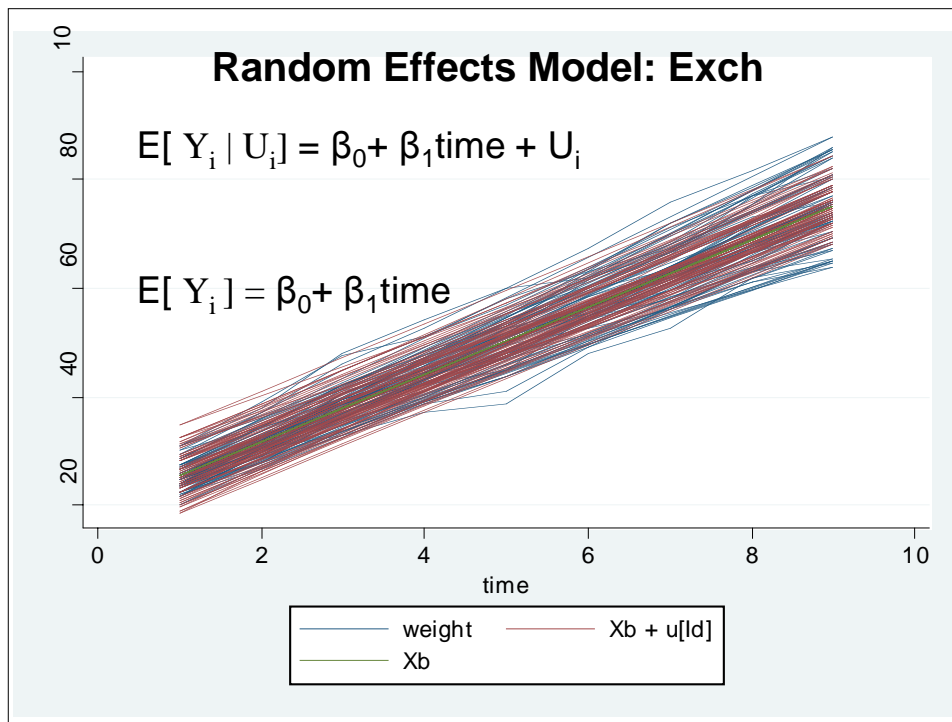
```
xtregar weight time
```

```
RE GLS regression with AR(1) disturbances
Group variable (i): Id      Number of obs   =   432
                             Number of groups  =   48
R-sq: within = 0.9851      Obs per group: min =    9
      between = 0.0000      avg             =   9.0
      overall  = 0.9305      max             =    9
Wald chi2(2)                = 12688.55
Prob > chi2                  = 0.0000
```

```
corr(u_i, Xb) = 0 (assumed)
```

weight	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
time	6.257651	.0555527	112.64	0.000	6.14877 6.366533	
_cons	19.00945	.6281622	30.26	0.000	17.77827 20.24062	
rho_ar	.73091237	(estimated autocorrelation coefficient)				
sigma_u	3.583343					
sigma_e	1.5590851					
rho_fov	.84082696	(fraction of variance due to u_i)				
theta	.60838037					

GEE-fit Marginal Model with AR1 Correlation structure



Important Points

- Modelling the correlation in longitudinal data is important to be able to obtain correct inferences on regression coefficients β
- There are correspondences between random effect and marginal models in the linear case because the interpretation of the regression coefficients is the same as that in standard linear regression