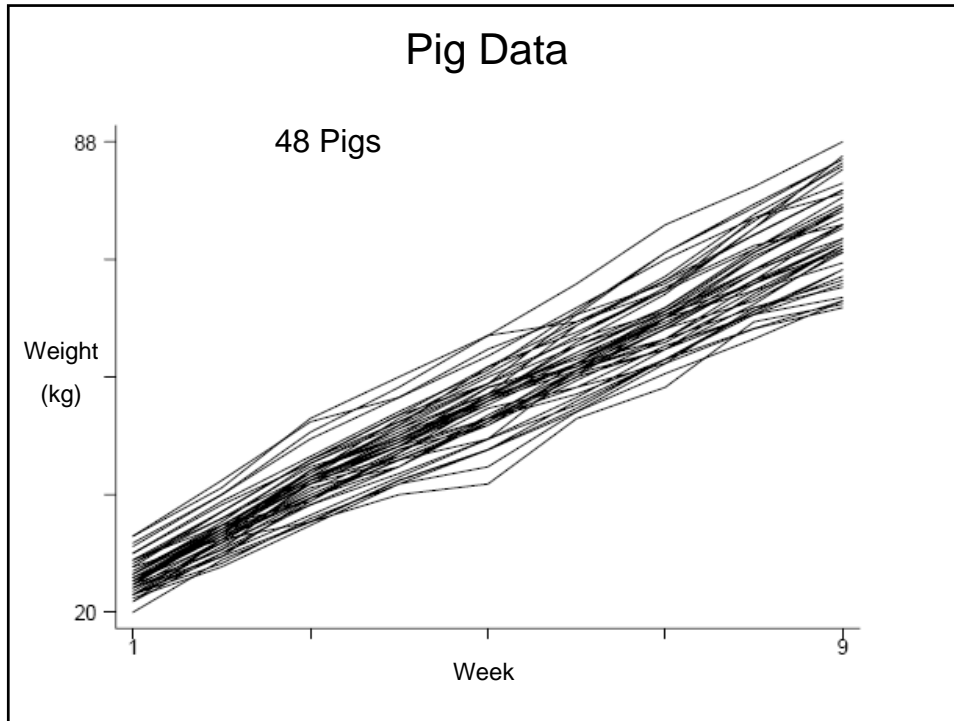


## Lecture 3

### 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 measured 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)**



## 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

### 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

### A) Linear model with random intercept

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

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

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

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

Intraclass correlation coefficient, i.e. correlation within measurements from pig i

$$\text{corr}(Y_{ij}, Y_{ik}) = \frac{\text{Cov}(Y_{ij}, Y_{ik})}{\sqrt{\text{Var}(Y_{ij})} \sqrt{\text{Var}(Y_{ik})}}$$

$$\text{corr}(Y_{ij}, Y_{ik}) = \frac{\tau^2}{\sqrt{\tau^2 + \sigma^2} \sqrt{\tau^2 + \sigma^2}}$$

## 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

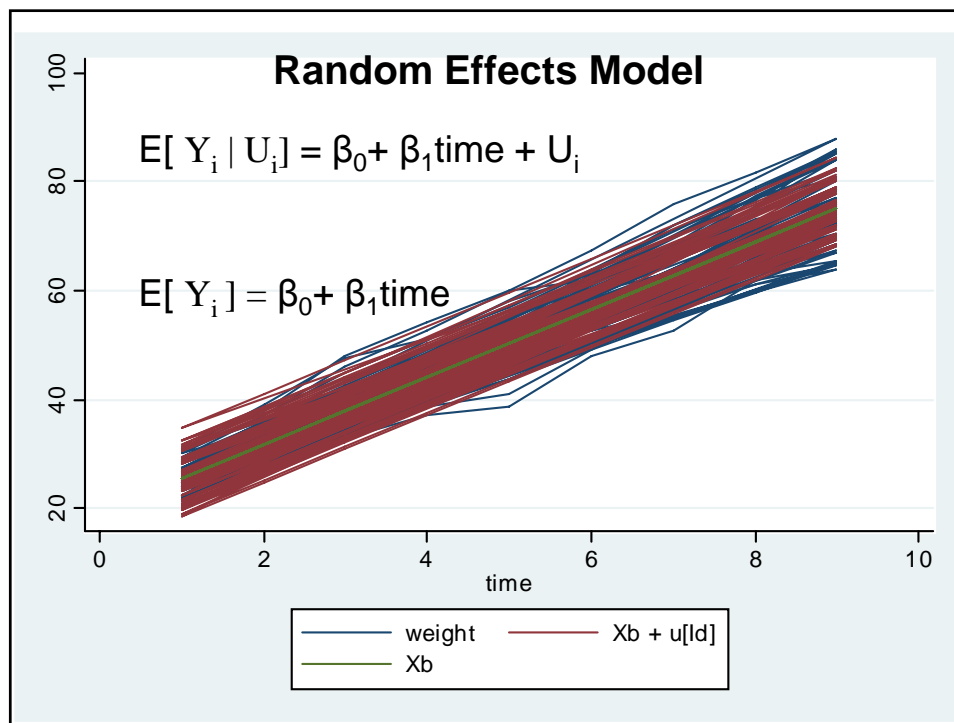
Log likelihood = -1014.9268           LR chi2(1)      =  1624.57
                                       Prob > chi2      =   0.0000
```

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

Linear model with a random intercept -  
“conditional model”

## Interpretation of results:

- Time effect: Among pigs with similar genetic variation (random effect), weight increases by 6.2 kg per week (95% CI: 6.1 to 6.3)
- Estimate of heterogeneity across pigs:  $\sigma_u^2 = 3.8^2 = 14.4$
- Estimate of variation in weights within a pig over time:  $\sigma_e^2 = 2.1^2 = 4.4$
- Fraction of total variability attributable to heterogeneity across pigs: 0.77
- This is also a measure of intraclass correlation, within pig correlation....



## B) Marginal Model With a Uniform or Exchangeable correlation structure

$$E[Y_{ij}] = \beta_0 + \beta_1 t_j \quad \text{Model for the mean}$$

$$\text{corr}(Y_{ij}, Y_{ik}) = \rho$$

$$\text{Var}(Y_{ij}) = \sigma_*^2 = (\tau^2 + \sigma^2) \quad \text{Model for the covariance matrix}$$

$$\text{Cov}(Y_{ij}, Y_{ik}) = \rho(\tau^2 + \sigma^2)$$

## Pigs – Marginal model

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

Iteration 1: tolerance = 5.585e-15

GEE population-averaged model		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

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

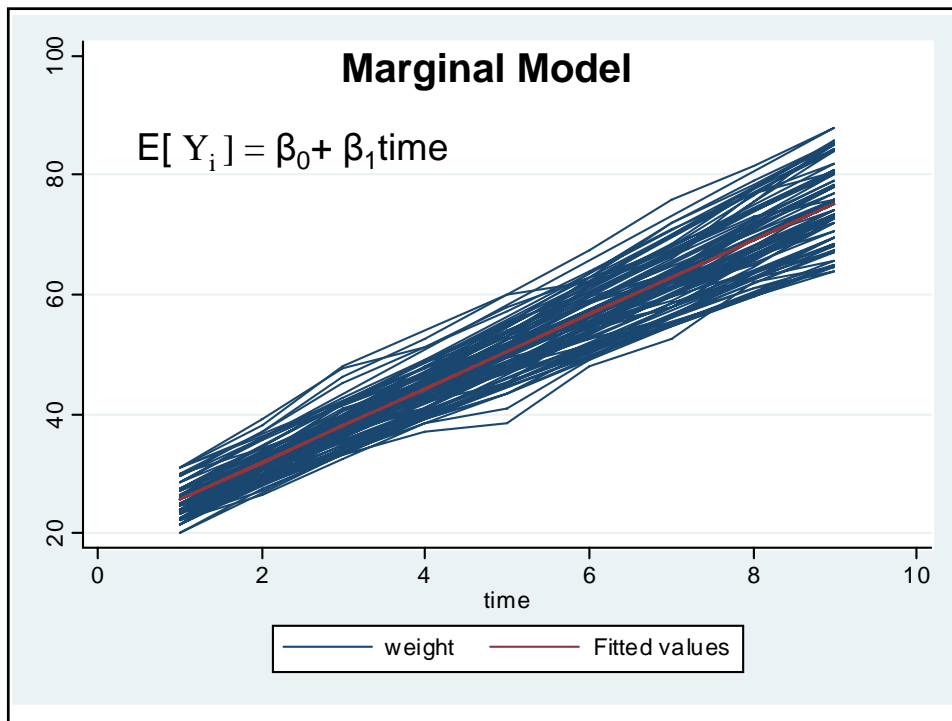
# Pigs data model 1 – GEE fit

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

Estimated within-Id correlation matrix R:

	c1	c2	c3	c4	c5	c6	c7	c8	c9
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



### Models A and B are equivalent

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

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

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

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

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

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

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

### One group polynomial growth curve model

- Similarly, if you want to fit a quadratic curve

$$E[Y_{ij} | U_i] = U_i + \beta_0 + \beta_1 t_j + \beta_2 t_j^2$$

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



## Pigs – Marg. model, quadratic trend

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

```
GEE population-averaged model
Group variable:          Id
Link:                   identity
Family:                 Gaussian
Correlation:            exchangeable
Scale parameter:       19.19317

Number of obs   = 432
Number of groups = 48
Obs per group: min = 9
                avg = 9.0
                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 – 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

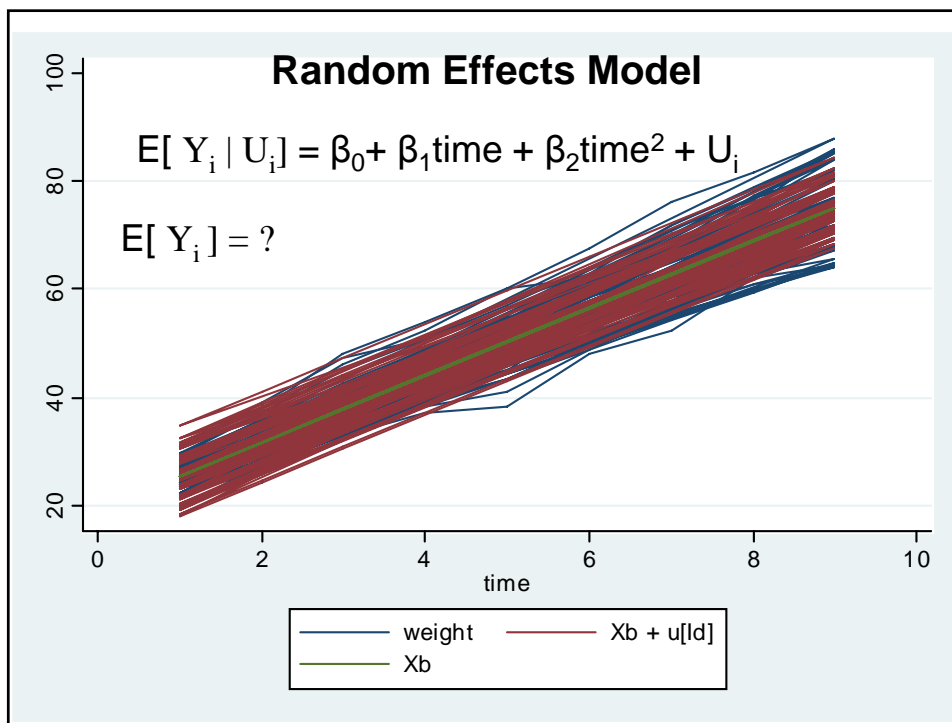
Log likelihood = -1014.5524

Number of obs = 432  
 Number of groups = 48  
 Obs per group: min = 9  
 avg = 9.0  
 max = 9

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 – Marginal model: AR(1)

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

```
GEE population-averaged model
Group and time vars:      Id time
Link:                     identity
Family:                   Gaussian
Correlation:              AR(1)
Scale parameter:         19.26754

Number of obs      =    432
Number of groups  =    48
Obs per group: min =     9
                  avg =    9.0
                  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
R-sq:  within = 0.9851
        between = 0.0000
        overall = 0.9305

Number of obs      =    432
Number of groups  =    48
Obs per group: min =     9
                  avg =    9.0
                  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
```

Random effects model with AR1 Correlation structure

## **Important Points**

- Modeling the correlation in longitudinal data is important to be able to obtain correct inferences on regression coefficients  $\beta$
- 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