

Multilevel models for analyzing people's daily moving behaviour

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OUTLINE

- **Data**

- hierarchical data structure

- **Statistical model**

- multilevel generalized linear model

- **Application**

- A survey on daily moving behaviours of the people residing in the area of Pisa

HIERARCHICAL DATA STRUCTURE

LEVEL	UNIT	SUBSCRIPT	RANGE	TOTAL
1	Individual	$j = 1, \dots, n_i$	$1 \leq n_i \leq 5$	$N=806$
2	Family	$i = 1, \dots, m_k$	$1 \leq m_k \leq 76$	$M=373$
3	Area	$k = 1, \dots, K$		$K=35$

$$N = \sum_{k=1}^K \sum_{i=1}^{m_k} n_i = 806$$

$$M = \sum_{k=1}^K m_k = 373$$

FEATURES OF THE DATA

- **Dependence**

- Correlation between individual observations in the same level.

- **Ignoring hierarchical data structure**

- Misleading statistical inference

- **Traditional statistical models should be adapted**

- Using random variables and relative assumptions to describe this correlation

MULTILEVEL GENERALIZED LINEAR MODEL

- **Probability distribution**

$$Y \sim \text{Exponential Family}$$

- **Conditional expectation**

$$E[Y | \mathbf{x}, \mathbf{z}, \mathbf{u}] = \mu$$

- **Linear predictor**

$$\eta = \sum_{v=1}^V x_v \beta_v + \sum_{l=2}^L \sum_{r=1}^{R_l} u_r^{(l)} z_r^{(l)}$$

- **Link function**

$$g(\mu) = \eta$$

2-LEVEL RANDOM INTERCEPT MODEL NORMAL

$$L = 2 \quad R_l = 1 \quad V = 2 \quad j = 1, \dots, n_i \quad i = 1, \dots, m_k$$

Distribution $Y_{ji} \sim N(\mu_{ji}, \sigma^2)$

Expectation $E(Y_{ji} | x_{ji}, z_{ji}, u_{1i}^{(2)}) = \mu_{ji} \quad -\infty < \mu_{ji} < +\infty$

Linear predictor $\eta_{ji} = x_{1ji}\beta_{1i} + x_{2ji}\beta_{2i} + u_{1i}^{(2)}z_{1ji}^{(2)}$

Link $\mu_{ji} = \eta_{ji}$

$$x_{1ji} = 1 \quad z_{1ji}^{(2)} = 1$$

3-LEVEL RANDOM INTERCEPT MODEL NORMAL

$$L = 3 \quad R_l = 1 \quad V = 2 \quad j = 1, \dots, n_i \quad i = 1, \dots, m_k \quad k = 1, \dots, K$$

Distribution $Y_{jik} \sim N(\mu_{jik}, \sigma^2)$

Expectation $E(Y_{jik} \mid x_{jik}, z_{jik}, u_{1i}^{(2)}, u_{1k}^{(3)}) = \mu_{jik} \quad -\infty < \mu_{jik} < +\infty$

Linear predictor $\eta_{jik} = x_{1jik} \beta_{1ik} + x_{2jik} \beta_2 + u_{1i}^{(2)} z_{1jik}^{(2)} + u_{1k}^{(3)} z_{1jik}^{(3)}$

Link $\mu_{jik} = \eta_{jik}$

$$x_{1jik} = 1 \quad z_{1jik}^{(2)} = 1 \quad z_{1jik}^{(3)} = 1$$

2-LEVEL RANDOM INTERCEPT MODEL POISSON

$$L = 2 \quad R_l = 1 \quad V = 2 \quad j = 1, \dots, n_i \quad i = 1, \dots, m_k$$

Distribution $Y_{ji} \sim P(\mu_{ji})$

Expectation $E(Y_{ji} | x_{ji}, z_{ji}, u_{1i}^{(2)}) = \mu_{ji} \quad 0 < \mu_{ji} < +\infty$

Linear predictor $\eta_{ji} = x_{1ji} \beta_{1i} + x_{2ji} \beta_2 + u_{1i}^{(2)} z_{1ji}^{(2)}$

Link $\log(\mu_{ji}) = \eta_{ji}$

$$x_{1ji} = 1 \quad z_{1ji}^{(2)} = 1$$

RANDOM EFFECTS DISTRIBUTION

$$u_r^{(l)} \sim N(0, \sigma_{lr}^2) \quad l = 1, \dots, L \quad r = 1, \dots, R_l$$

- **Covariance between two random effects at the same level**

$$COV(u_r^{(l)}, u_{r'}^{(l')}) = \sigma_{lrr'}^2 \quad \text{with } l = l' \text{ and } r \neq r'$$

- **Covariance between two random effects at different level**

$$COV(u_r^{(l)}, u_{r'}^{(l')}) = 0 \quad \text{with } l \neq l' \text{ and } r \neq r'$$

MAXIMUM LIKELIHOOD ESTIMATORS

- **Define the likelihood for the data**

$$L(Y|w) = \int f(Y|u, w) p(u|w) du$$

- w denote a vector containing all parameters to be estimated
 - $f()$ is the probability distribution of the outcome
 - $p()$ is the probability distribution of the random effects
- **Maximize the likelihood respect to \mathcal{W} in order to make inferences about \mathcal{W}**
 - **Optimization Algorithms**
 - **Differences between linear and non linear multilevel model**

TESTING HYPOTHESIS

- **Fixed effects**

$$H_0 : \beta_v = 0$$

- Wald test

$$W = \frac{\hat{\beta}_v}{\hat{\sigma}_{\hat{\beta}_v}}$$

- **Random effects**

$$H_0 : \sigma_{lr}^2 = 0$$

- Likelihood Ratio test $LRT = 2 \times \text{abs}(L1-L0)$

where L1 and L0 are the log-likelihoods evaluated in a model with and without the variance component σ_{lr}^2 , that is, the random effect $u_r^{(l)}$.

STATISTICAL SOFTWARE

- **Several packages are available**
 - STATA (command GLLAMM and XT)
 - SPLUS or R (command LME and NLME)
 - SAS (command PROC MIXED)
 - MLwiN
 - HLM

- **Differences in computational algorithms**
 - Newton-Raphson
 - Expectation-Maximization or Fisher Scoring
 - Iterative Generalized Least Squares

MODEL 1

```
. xi: gllamm distance gender i.agec, f(gaussian) l(identity) i(family)
```

number of level 1 units = 806

number of level 2 units = 373

log likelihood = -4516.0369 (some STATA output is omitted)

distance	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]		
gender	-18.55207	4.294402	-4.32	0.000	-26.96894	-10.13519	
_Iagec_2	23.31702	9.492722	2.46	0.014	4.711627	41.92241	0-14 15-30
_Iagec_3	27.22336	8.56821	3.18	0.001	10.42997	44.01674	31-44
_Iagec_4	20.49728	8.890619	2.31	0.021	3.071982	37.92257	45-59
_Iagec_5	2.963706	10.46916	0.28	0.777	-17.55547	23.48288	60-74
_Iagec_6	-5.604856	29.3878	-0.19	0.849	-63.20389	51.99418	>74
_cons	24.56164	8.354864	2.94	0.003	8.186403	40.93687	

Variance at level 1

3227.0578 (209.92252)

Variances and covariances of random effects

level 2 (family)

var(1): 1562.0955 (307.4231)

INTRA-FAMILY CORRELATION = $1562 / (3227 + 1562) = 0.32$

MODEL 2

```
. xi: gllamm distance gender i.agec, f(gaussian) l(identity) i(family area)
```

```
number of level 1 units = 806
```

```
number of level 2 units = 373
```

```
number of level 3 units = 35
```

```
log likelihood = -4513.9259 (some STATA output is omitted)
```

distance	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]		
gender	-18.57917	4.264296	-4.36	0.000	-26.93704	-10.22131	
_Iagec_2	22.05173	9.455582	2.33	0.020	3.519131	40.58433	0-14
_Iagec_3	26.95951	8.51035	3.17	0.002	10.27953	43.63948	15-30
_Iagec_4	20.09742	8.85395	2.27	0.023	2.743992	37.45084	31-44
_Iagec_5	2.625714	10.39971	0.25	0.801	-17.75734	23.00877	45-59
_Iagec_6	-6.245908	29.18454	-0.21	0.831	-63.44656	50.95474	60-74
_cons	24.89028	8.625134	2.89	0.004	7.985325	41.79523	>74

Variance at level 1

```
3180.402 (209.47161)
```

Variances and covariances of random effects

level 2 (family)

```
var(1): 1479.8355 (300.42951)
```

```
scalar LRT = 2*abs(-4513.9259 - -4516.0369)
```

```
.di chiprob(1, LRT)
```

0.03990366

level 3 (area)

```
var(1): 135.86588 (95.478454)
```

MODEL 1 (adaptive quadrature)

. xi: gllamm distance gender i.agec, f(gaussian) l(identity) i(family) adapt

number of level 1 units = 806

number of level 2 units = 373

log likelihood = -4508.9392 (some STATA output is omitted)

distance	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]		
gender	-18.65784	3.999824	-4.66	0.000	-26.49735	-10.81833	
_Iagec_2	22.63511	9.034982	2.51	0.012	4.926874	40.34335	0-14
_Iagec_3	26.48049	7.971745	3.32	0.001	10.85615	42.10482	15-30
_Iagec_4	20.19226	8.470851	2.38	0.017	3.589694	36.79482	31-44
_Iagec_5	2.842191	10.06534	0.28	0.778	-16.88551	22.56989	45-59
_Iagec_6	-5.707835	27.53451	-0.21	0.836	-59.67449	48.25882	60-74
_cons	24.60227	8.010341	3.07	0.002	8.902292	40.30225	>74

Variance at level 1

2691.102 (209.34499)

Variances and covariances of random effects

level 2 (family)

var(1): 2233.4213 (359.34996)

INTRA-FAMILY CORRELATION = $2233 / (2691 + 2233) = 0.45$

MODEL 2 (adaptive quadrature)

```
. xi: gllamm distance gender i.agec, f(gaussian) l(identity) i(family area) adapt
```

```
number of level 1 units = 806
```

```
number of level 2 units = 373
```

```
number of level 3 units = 35
```

```
log likelihood = -4507.8158 (some STATA output is omitted)
```

distance	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]		
gender	-18.64834	3.997244	-4.67	0.000	-26.48279	-10.81388	
_Iagec_2	21.86722	9.037881	2.42	0.016	4.153303	39.58114	0-14
_Iagec_3	26.34994	7.970704	3.31	0.001	10.72765	41.97223	15-30
_Iagec_4	19.98335	8.472274	2.36	0.018	3.378	36.5887	31-44
_Iagec_5	2.683085	10.04687	0.27	0.789	-17.00843	22.3746	45-59
_Iagec_6	-6.018096	27.51967	-0.22	0.827	-59.95566	47.91947	60-74
_cons	24.92912	8.274806	3.01	0.003	8.710801	41.14744	>74

Variance at level 1

```
2692.1052 (209.67197)
```

Variances and covariances of random effects

level 2 (family)

```
var(1): 2125.0703 (359.25033)
```

```
scalar LRT = 2*abs(-4508.9392 - -4507.8158)
```

```
.di chiprob(1, LRT)
```

```
.13389047
```

level 3 (area)

```
var(1):103.8059(92.49525)
```


MODEL 3

```
. xi: gllamm ntrip gender i.agec, f(poisson) l(log) i(family)
```

```
number of level 1 units = 806
```

```
number of level 2 units = 373
```

```
log likelihood = -1628.4103
```

```
(some STATA output is omitted)
```

ntrip	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]		
gender	-.0464929	.0359264	-1.29	0.196	-.1169073	.0239215	
_Iagec_2	.2817658	.0825903	3.41	0.001	.1198919	.4436398	0-14
_Iagec_3	.2987986	.0781225	3.82	0.000	.1456814	.4519158	15-30
_Iagec_4	.294807	.0778875	3.79	0.000	.1421503	.4474638	31-44
_Iagec_5	.3061211	.088164	3.47	0.001	.1333227	.4789194	45-59
_Iagec_6	-.0988595	.2861549	-0.35	0.730	-.6597128	.4619938	60-74
_cons	1.144427	.073767	15.51	0.000	.9998468	1.289008	>74

Variances and covariances of random effects

```
level 2 (family)
```

```
var(1): .03422704 (.00978456)
```

```
. display 2*abs(-1628.4103 - -1638.08°) °(-1638.08 from poisson regression)
```

```
19.3392
```

```
. display chiprob(1, 19.3392)
```

```
0.00001094
```

MODEL 3 (adaptive quadrature)

```
. xi: gllamm ntrip gender i.agec, f(poisson) l(log) i(family) adapt
```

```
number of level 1 units = 806
```

```
number of level 2 units = 373
```

```
log likelihood = -1628.4102
```

ntrip	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]		
gender	-.0464929	.0359265	-1.29	0.196	-.1169076	.0239218	
_Iagec_2	.2817775	.0825924	3.41	0.001	.1198994	.4436555	0-14
_Iagec_3	.2988007	.0781228	3.82	0.000	.1456829	.4519185	15-30
_Iagec_4	.2948099	.0778881	3.79	0.000	.142152	.4474678	31-44
_Iagec_5	.3061284	.088165	3.47	0.001	.1333281	.4789287	45-59
_Iagec_6	-.0988579	.2861577	-0.35	0.730	-.6597167	.4620009	60-74
_cons	1.144424	.0737682	15.51	0.000	.9998409	1.289007	>74

Variances and covariances of random effects

```
level 2 (family)
```

```
var(1): .03422967 (.00978768)
```

```
. display 2*abs(-1628.4102 - -1638.08°) °(-1638.08 from poisson regression)
```

19.3392

```
. display chiprob(1, 19.3392)
```

0.00001094

SUMMARY

- **Close relation between features of the data and statistical techniques**
 - Multilevel Statistical models take into account a hierarchical data structure
- **Define a multilevel generalized linear model and obtained MLEs using the STATA command GLLAMM**
 - Normal and Poisson probability distributions as special case in the exponential family

POTENTIAL FUTURE RESEARCH

- **Testing variance components and confidence intervals**
 - Score test
- **Assessing the adequacy of multilevel models**
 - Assumptions of the random variables
- **Multi-stage sampling design**

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