

PSC 8185: Multilevel Modeling
Fitting Random Coefficient Binary Response Models in Stata

Consider the following two-level model random coefficient logit model. This is a Supreme Court decision making model, where level-1 units are justices' votes and level-2 units are court cases (justices' votes (3544) nested within 399 cases). For binary dependent variables, a Bernoulli sampling model is used, and I use a logit link here. For the logit link, first define $\text{Pr}(Y_{ij}=1) = p_{ij}$, which is the probability of a liberal vote for choice i in case j . Then define η_{ij} as the log-odds of p_{ij} : $\eta_{ij} = \log[p_{ij} / (1 - p_{ij})]$. This allows one to specify the log-odds as a linear function of the level-1 independent variables.

[Level-1 eq.] $\eta_{ij} = \beta_{0j} + \beta_{1j}PREF_{ij}$

[Level-2 eqs.] $\beta_{0j} = \gamma_{00} + \gamma_{01}SAL_j + \gamma_{02}COMP_j + \gamma_{03}ISSUEFAM_j + \gamma_{04}INFO_j +$
 $\gamma_{05}USAMICUS_j + \gamma_{06}USPARTY_j + \gamma_{07}STAT_j + u_{0j}$

$\beta_{1j} = \gamma_{10} + \gamma_{11}SAL_j + \gamma_{12}COMP_j + \gamma_{13}ISSUEFAM_j + \gamma_{14}INFO_j +$
 $\gamma_{15}USAMICUS_j + \gamma_{16}USPARTY_j + \gamma_{17}STAT_j + u_{1j}$

Reduced-form Representation of Model 2:

	$\eta_{ij} =$
<i>Effects on case outcome</i>	$\gamma_{00} + \gamma_{01}SAL_j + \gamma_{02}COMP_j + \gamma_{03}ISSUEFAM_j + \gamma_{04}INFO_j +$ $\gamma_{05}USAMICUS_j + \gamma_{06}USPARTY_j + \gamma_{07}STAT_j +$
<i>Avg. effect of preferences</i>	$\gamma_{10} * PREF_{ij} +$
<i>Cross-level interactions</i>	$\gamma_{11}SAL_j * PREF_{ij} + \gamma_{12}COMP_j * PREF_{ij} + \gamma_{13}ISSUEFAM_j * PREF_{ij} +$ $\gamma_{14}INFO_j * PREF_{ij} + \gamma_{15}USAMICUS_j * PREF_{ij} + \gamma_{16}USPARTY_j * PREF_{ij} +$ $\gamma_{17}STAT_j * PREF_{ij} +$
<i>Error components</i>	$u_{0j} + u_{1j} * PREF_{ij}$

Fitting the Two-Level Model using Full Maximum Likelihood (Adaptive Quadrature) in GLLAMM

```
set more off

*ADAPTIVE QUADRATURE MODELS

*Generate a constant; gllamm needs this
gen cons=1

*Specify random intercept and random coefficient
eq cons: cons
eq pref: lagpctlib_use

*Specify the level-2 variables that are a function of the random intercept
eq f1: salience_nyt_cen complex_3pt_cen lnissuecount_cen infoenv_01_cen
usamicus_cen usparty_cen statutory_cen

*Specify the level-2 variables that are a function of the random coefficient
eq f2: salience_nyt_cen complex_3pt_cen lnissuecount_cen infoenv_01_cen
usamicus_cen usparty_cen statutory_cen

*Model command; start with 4 quadrature points, then 8, then 12, etc.; use
*start values from the previous model (that's the "from(a)" command)
gllamm dir lagpctlib_use, i(caseid) l(logit) f(binom) nrf(2) eqs(cons pref)
geqs(f1 f2) adapt nip(4)

matrix a=e(b)

gllamm dir lagpctlib_use, i(caseid) l(logit) f(binom) nrf(2) eqs(cons pref)
geqs(f1 f2) adapt nip(8) from(a)

matrix a=e(b)

gllamm dir lagpctlib_use, i(caseid) l(logit) f(binom) nrf(2) eqs(cons pref)
geqs(f1 f2) adapt nip(12) from(a)

est store lag_adapt12

matrix a=e(b)

gllamm dir lagpctlib_use, i(caseid) l(logit) f(binom) nrf(2) eqs(cons pref)
geqs(f1 f2) adapt nip(16) from(a)

est store lag_adapt16

matrix a=e(b)

gllamm dir lagpctlib_use, i(caseid) l(logit) f(binom) nrf(2) eqs(cons pref)
geqs(f1 f2) adapt nip(20) from(a)

est store lag_adapt20

*Present results in a table form
est table lag_adapt12 lag_adapt16 lag_adapt20, se p varwidth(34) stats(N ll
aic)
```

GLLAMM OUTPUT

. *ADAPTIVE QUADRATURE MODELS

[Suppressing output from models using 4 and 8 quadrature points.]

```
. gllamm dir lagpctlib_use if lagpctlib_use~=., i(caseid) l(logit) f(binom)
nrf(2) eqs(cons pref) geqs(f1 f2) adapt nip (12) from(a)
```

Running adaptive quadrature

```
Iteration 0:   log likelihood = -1433.9683
Iteration 1:   log likelihood = -1433.9498
Iteration 2:   log likelihood = -1433.9498
```

Adaptive quadrature has converged, running Newton-Raphson

```
Iteration 0:   log likelihood = -1433.9498
Iteration 1:   log likelihood = -1433.9498   (backed up)
Iteration 2:   log likelihood = -1433.9496
Iteration 3:   log likelihood = -1433.9496
```

number of level 1 units = 3544

number of level 2 units = 399

Condition Number = 15.170465

gllamm model

log likelihood = -1433.9496

dir	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
lagpctlib_~e	4.469579	.3126479	14.30	0.000	3.8568	5.082357
_cons	-.3927239	.2204633	-1.78	0.075	-.824824	.0393763

Variances and covariances of random effects

***level 2 (caseid)

```
var(1): 15.246868 (2.1409271)
cov(2,1): .47004543 (1.589734) cor(2,1): .03597937
var(2): 11.194171 (2.0465321)
```

Regressions of latent variables on covariates

```
random effect 1 has 7 covariates:
saliency_nyt_cen: 1.3087985 (.53243653)
complex_3pt_cen: -.23641807 (.91949432)
lnissuecount_cen: .0830284 (.21363707)
infoenv_01_cen: -3.2768549 (2.4414709)
usamicus_cen: -1.2216683 (.55349651)
usparty_cen: -1.6855654 (.58330776)
statutory_cen: .20317709 (.45585559)
```

```

random effect 2 has 7 covariates:
saliency_nyt_cen: 1.4321357 (.63688317)
complex_3pt_cen: -.37590127 (1.0727686)
lnissuecount_cen: .52559336 (.25446341)
infoenv_01_cen: -.23535991 (2.9681077)
usamicus_cen: -.7769956 (.6769883)
usparty_cen: -1.3340209 (.70159345)
statutory_cen: -.6900348 (.54179413)

```

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. est store lag_adapt12
```

```
[Suppressing output from models using 16 and 20 quadrature points.]
```

```
.
. est table lag_adapt12 lag_adapt16 lag_adapt20, se p varwidth(34) stats(N ll aic)
```

		Variable	lag_ada~12	lag_ada~16	lag_ada~20

dir					
	lagpctlib_use		4.4695786	4.4671294	4.4646248
			.3126479	.31198263	.31178696
			0.0000	0.0000	0.0000
	_cons		-.39272387	-.38596767	-.38979855
			.22046332	.22660269	.22829931
			0.0749	0.0885	0.0877

cas1_1					
	cons		3.9047238	3.8868616	3.8851496
			.27414577	.26850628	.26867348
			0.0000	0.0000	0.0000

cas1_2					
	lagpctlib_use		3.3436029	3.3329991	3.3345159
			.30571974	.30443987	.30496215
			0.0000	0.0000	0.0000

cas1_2_1					
	_cons		.12037866	.12164403	.119327
			.40697491	.4042363	.40362999
			0.7674	0.7635	0.7675

f1					
	saliency_nyt_cen		1.3087985	1.3005355	1.3042134
			.53243653	.53845044	.54107965
			0.0140	0.0157	0.0159
	complex_3pt_cen		-.23641807	-.23030577	-.23172214
			.91949432	.93965606	.94469984
			0.7971	0.8064	0.8062
	lnissuecount_cen		.0830284	.08238374	.08341868
			.21363707	.22026835	.22284323
			0.6975	0.7084	0.7082
	infoenv_01_cen		-3.2768549	-3.2372372	-3.2487924
			2.4414709	2.4672407	2.4796085
			0.1795	0.1895	0.1901
	usamicus_cen		-1.2216683	-1.2033215	-1.2095659

		.55349651	.56258063	.56695643
		0.0273	0.0324	0.0329
	usparty_cen	-1.6855654	-1.6638382	-1.6713425
		.58330776	.59268672	.59686642
		0.0039	0.0050	0.0051
	statutory_cen	.20317709	.20579869	.20441877
		.45585559	.46632308	.47066661
		0.6558	0.6590	0.6641

f2				
	salience_nyt_cen	1.4321357	1.4270938	1.4280513
		.63688317	.63558729	.63648104
		0.0245	0.0247	0.0249
	complex_3pt_cen	-.37590127	-.37517886	-.3744169
		1.0727686	1.0706272	1.0717017
		0.7260	0.7260	0.7268
	lnissuecount_cen	.52559336	.52337804	.52371888
		.25446341	.25367009	.25423543
		0.0389	0.0391	0.0394
	infoenv_01_cen	-.23535991	-.24728057	-.24405967
		2.9681077	2.9606287	2.9653711
		0.9368	0.9334	0.9344
	usamicus_cen	-.7769956	-.77136087	-.77295672
		.6769883	.67531714	.67646903
		0.2511	0.2534	0.2532
	usparty_cen	-1.3340209	-1.3306189	-1.3306759
		.70159345	.7000968	.70095192
		0.0572	0.0574	0.0576
	statutory_cen	-.6900348	-.68659908	-.68781565
		.54179413	.54078971	.54172083
		0.2028	0.2042	0.2042

Statistics				
	N	3544	3544	3544
	ll	-1433.9496	-1434.0637	-1434.0918
	aic	2905.8993	2906.1274	2906.1837

legend: b/se/p