

From Stata to aML

Sara Ayllón
Department of Economics
University of Girona

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Introduction

- In this work, I show how to exploit Stata to run multilevel multiprocess models with aML.
- aML (applied Maximum Likelihood) has several advantages:
 - ▶ It easily handles three or more data levels (that can be mixed and matched)
 - ▶ It can estimate models with two or more different outcomes or data structures
 - ▶ It makes it easy to program complex errors structures
 - ▶ It is fast and free of charge
- aML is broader than `cmp` by Roodman (2011). It allows full simultaneity in systems of equations. It can account for random effects at different levels and offers a wider range of model types. Also, `gllamm` by Rabe-Hesketh includes some of aML features but for single-equation models.

Introduction

- But, aML is not a user-friendly programme:
 - ▶ It requires that the data set is prepared with a third party software package (as Stata)
 - ▶ It needs two different control files (to upload data and to run regressions)
 - ▶ It requires the researcher to typeset the starting values (tedious, time consuming)
 - ▶ It needs the starting values to be assigned in the same order as the regressor set (potential for user error)

Introduction

What can Stata do about it?

- EVERYTHING!
- One can create a single Stata do-file that:
 - ▶ Prepares the dataset
 - ▶ Writes the control files
 - ▶ Inputs the starting values
 - ▶ Calls aML using the `shell` command
 - ▶ Brings aML results back to Stata after estimation
 - ▶ Posts aML results so that they become 'official' within Stata for postestimation

Introduction

- Main advantages of joining Stata and aML:
 - ▶ One can easily change model specification
 - ▶ No need to manually open the Command Prompt window used by aML
 - ▶ No margin for typo errors as Stata automatically writes all the necessary files
 - ▶ It saves lots of time!
- There are other examples that connect Stata with software packages for running multilevel models (`runmlwin`, `runmplus`, `bugutils`, etc.)

Data preparation

- Use Stata `local` for writing a string that will contain the variables list (that way that can be easily modified):

```
loc level1 ""  
loc level1 "'level1' ivar1"
```

- Run single regressions in Stata and save the starting values using `estimates store` and `estout` (Jann, 2005)
- Use `outfile` and the local strings previously created to prepare the data set as required by aML
- Write the control file that uploads the data to aML with the `file open`, `file write` and `file close` commands (Gini and Pasquini, 2006)

Model specification and estimation

Model specification

- A bit more complicated... we can write the different parts that aML requires and save them as `.dta` files
- And, then, we simply append them to build the control file `.aml`

Model estimation

- Then one is ready to call aML within Stata. Three lines do all the job:

```
sh raw2aml -r data.r2a  
sh aml -r estimate.aml  
sh update -r estimate.out
```

And back to Stata...

- We can create an estimates table with the `mktab` utility in aML that can be imported with `insheet`.
- Once in Stata, we create a `bhat` matrix with the coefficients and a `vce` for variances and covariances using `mkmat`.
- We post them with `ereturn post` that way aML results become *official* within Stata.
- Now, we can use Stata postestimation commands.

A (naive) example...

Nlswork.dta (National Longitudinal Survey of Youth). Women aged 14 to 26 in 1968.

Let's imagine that we want to explain the probability of being married (*m*_{sp}) and that of belonging to a union. Moreover, we want to control for unobserved heterogeneity by integrating out a random effect in each equation and we wish to allow a free correlation between these individual-specific effects. We propose a bivariate RE probit:

$$M_{sp_{it}} = \beta_0 \text{Birthy}_i + \beta_1 \text{Race}_i + \beta_2 \text{Age}_{it} + \beta_3 \text{Lnwage}_{it} + \beta_4 \text{Union}_{it} + c_i + \epsilon_{it} \quad (1)$$

$$\text{Union}_{it} = \alpha_0 \text{Birthy}_i + \alpha_1 \text{Race}_i + \alpha_2 \text{Age}_{it} + \alpha_3 \text{Lnwage}_{it} + u_i + \lambda_{it} \quad (2)$$

We run the 'Stata_to_aML' do file and...

... in less than 2 minutes we have the results in Stata:

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
_cons_1	2.3125	.3274	7.06	0.000	1.670808	2.954192
birth_yr_1	-.031	.0063	-4.92	0.000	-.0433478	-.0186522
race_1	-.8534	.0383	-22.28	0.000	-.9284666	-.7783334
age_1	.0345	.0016	21.56	0.000	.0313641	.0376359
ln_wage_1	-.1564	.0331	-4.73	0.000	-.2212748	-.0915252
union_1	-.0891	.0344	-2.59	0.010	-.1565228	-.0216772
_cons_2	-2.3221	.3689	-6.29	0.000	-3.045131	-1.599069
birth_yr_2	-.0245	.0071	-3.45	0.001	-.0384157	-.0105843
race_2	.5751	.0417	13.79	0.000	.4933695	.6568305
age_2	-.0083	.0021	-3.95	0.000	-.0124159	-.0041841
ln_wage_2	.9753	.0357	27.32	0.000	.9053293	1.045271
eps_1	1.4076	.0221	63.69	0.000	1.364285	1.450915
eps_2	1.3215	.0271	48.76	0.000	1.268385	1.374615
rho	.0216	.0173	1.25	0.212	-.0123074	.0555074

This is an example of how well Stata can work with other software.

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