# Just tired of endless loops! or parallel: Stata module for parallel computing

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# Agenda

Motivation

What is and how does it work

Benchmarks

 $\label{eq:Syntax} \mbox{Syntax and Usage}$ 

- ▶ Both computation speeds and size of data are ever increasing
- Often our work is easily broken down into independent chunks
- But, implementing parallel computing even for these "embarrassingly parallel" problems is not easy, most of this due to lack of (user-friendly) statistical computing tools.
- parallel aims to make a contribution to these issues.

What is and how does it work

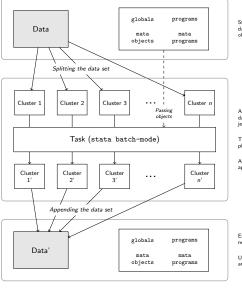
Benchmarks

Syntax and Usage

- NB: StataMP is available but only parallelizes some internal commands.
   Therefore a user-provided solution is needed.
- ▶ Inspired by the R package "snow" (several other examples exists: StataMP, HTCondor, Matlab's Parallel Toolbox, etc.)
- Is designed to be used in multiprocessor machines (e.g. simultaneous multi-threading, multiple cores, multiple sockets).
- It implements parallel computing methods through an OS's shell scripting (using Stata in batch mode).
- On Windows we have a compiled plugging allowing
  - Functionality when the parent Stata is in batch-mode
  - Seamless user experience by launching the child programs in a hidden desktop (otherwise GUI for each steals focus)
- Depending on the task, can reach near to (or over) linear speedups proportional to the number of physical cores of the computer.
- ▶ Thus having a quad-core computer can lead to a 400% speedup.

## What is and how does it work

#### How does it work?



Starting (current) stata instance loaded with data plus user defined globals, programs, mata objects and mata programs

A new stata instance (batch-mode) for every data-clusters. Programs, globals and mata objects/programs are passed to them.

The same algorithm (task) is simultaneously applied over the data-clusters

After every instance stops, the data-clusters are appended into one.

Ending (resulting) stata instance loaded with the new data.

User defined globals, programs, mata objects and mata programs remind unchanged.

## What is and how does it work

How does it work?

- ▶ Method is *split-apply-combine* like MapReduce
- ▶ Straightforward usage when there is observation or group-level work
- ▶ If each iteration needs the entire dataset, then can be reformulated as
  - 1. Save real dataset to file
  - 2. Generate in-memory data as the list of tasks. For example,
    - List of seeds for each bootstrap resampling
    - List of parameter values for simulations
  - Child process reads its list of tasks, loads the real dataset, and does it's own looping over tasks
- If the list of tasks is not initially definable then the "nodata" alternative mechanism allows for more flexibility.

## What is and how does it work

Parallel's backend

#### When the user enters

```
parallel: gen n = _N
```

## parallel takes the command and writes something like this

```
cap clear all
                                                                  cap clear all
  cd ~
                                                                  cd ~
1 set seed 34815
                                                                1 set seed 98327
  set memory 16777216b
                                                                  set memory 16777216b
  cap set maxvar 5000
                                                                  cap set maxvar 5000
  cap set matsize 400
                                                                  cap set matsize 400
2 local pll_instance 1
                                                                2 local pll_instance 2
  local pll_id efcql2tspr
                                                                  local pll_id efcql2tspr
  capture {
                                                                  capture {
                                                                  noisily {
  noisily {
3 use __pllefcql2tsprdataset if _efcql2tsprcut == 1
                                                                3 use __pllefcql2tsprdataset if _efcql2tsprcut == 2
  gen n = N
                                                                  gen n = N
4 save __pllefcgl2tsprdta1, replace
                                                                4 save __pllefcgl2tsprdta2, replace
  local result = _rc
                                                                  local result = _rc
  cd ~
                                                                  cd ~
5 mata: write_diagnosis(st_local("result"),
                                                                5 mata: write_diagnosis(st_local("result"),
  >"__pllefcql2tsprfinito1")
                                                                  >"__pllefcql2tsprfinito2")
```

What is and how does it work

# Benchmarks

Syntax and Usage

Simple example: Serial replace

## Serial fashion

do mydofile.do

## Parallel fashion

parallel do mydofile.do

## Table: Serial replacing using a loop on a Linux Server (16 clusters)

	100,000	1,000,000	10,000,000
CPU	1.43	16.94	144.68
Total	0.34	3.20	12.49
Setup	0.00	0.00	0.00
Compute	0.32	3.07	11.54
Finish	0.02	0.12	0.95
Ratio (compute) Ratio (total)	4.50 4.22 (26%)	5.51 5.30 (30%)	12.53 11.58 (72%)

Tested on a Intel Xeon X470 (hexadeca-core) machine

#### Monte Carlo simulation (Windows Machine)

#### Serial fashion

do myexperiment.do

#### Parallel fashion

parallel do myexperiment.do, nodata

```
Figure: myexperiment.do
local num_of_intervals = 50
if length("'pll.id'") == 0 (
    local start = 1
    local end = 'num_of_intervals'
    local ntot = floor('num_of_intervals'/$PLL_CLUSTERS)
    local start = ('pll_instance' - 1)*'ntot' + 1
    local end = ('pll_instance') + 'ntot'
    if 'pll_instance' == $FLLCLUSTERS local end = 10
local reps 10000
forval i='start'/'end' (
    qui use census2, clear
    gen true_y = age
    gen z_factor = region
    sun z.factor, meanonly
    scalar znu = r(mean)
         cen v1 = .
         gen y2 = .
         local c = 'i'
         set seed 'c'
         simulate c=r(c) nul=r(nul) se_nul = r(se_nul) ///
                 mu2=r(mu2) se_mu2 = r(me_mu2), ///
                  saving(cc'i', replace) nodots reps('reps'): ///
ncsimull, c('c')
```

#### Table: Monte Carlo Experiment on a Windows Machine (4 clusters)

	2	4
CPU	111.49	114.13
Total	58.02	37.48
Setup	0.00	0.00
Compute	58.02	37.48
Finish	0.00	0.00
Ratio (compute) Ratio (total)	1.92 1.92 (96%)	3.04 3.04 (76%)

Tested on a Intel i3 2120 (dual-core) machine

Monte Carlo simulation (Unix Machine)

## Serial fashion

do myexperiment.do

## Parallel fashion

 ${\tt parallel} \ {\tt do} \ {\tt myexperiment.do}, \ {\tt nodata}$ 

## Table: Monte Carlo Experiment on a Linux Server (16 clusters)

	2	4	8	16
CPU	164.79	164.04	162.84	163.89
Total	69.85	34.28	19.00	10.78
Setup	0.00	0.00	0.00	0.00
Compute	69.85	34.28	19.00	10.78
Finish	0.00	0.00	0.00	0.00
Ratio (compute) Ratio (total)	2.36 2.36 (118%)	4.78 4.78 (120%)	8.57 8.57 (107%)	15.21 15.21 (95%)

Tested on a Intel Xeon X470 (hexadeca-core) machine

Reshaping Administrative Data

## Serial fashion

```
reshape wide tipsolic rutemp opta derecho ngiros, /// i(id) j(time)
```

## Parallel fashion

```
parallel, by(id) :reshape wide tipsolic rutemp opta derecho ngiros, /// i(id) j(time)
```

## Table: Reshaping wide a large database on a Linux Server (8 clusters)

	100,000	1,000,000	5,000,000
CPU	5.51	72.70	392.97
Total	2.33	17.46	86.44
Setup	0.00	0.00	0.00
Compute	1.83	12.42	57.93
Finish	0.50	5.04	28.51
Ratio (compute) Ratio (total)	3.01 2.37 (29%)	5.85 4.16 (52%)	6.78 4.55 (57%)

Tested on a Intel Xeon X470 (hexadeca-core) machine

What is and how does it work

Benchmarks

Syntax and Usage

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Recommendations on its usage

#### parallel suits ...

- ▶ Monte-Carlo simulation.
- Extensive nested control flow (loops, while, ifs, etc.).
- Bootstrapping/Jackknife.
- Simulations in general.

#### parallel doesn't suit ...

- (already) fast commands.
- ► Regressions, ARIMA, etc.
- Linear Algebra.
- Whatever StataMP does better.
- (Currently) Tasks that already take up all of RAM.

- ▶ Brings parallel computing to many more commands than StataMP
- its major strengths/advantages are in simulation models and non-vectorized operations such as control-flow statements.
- Depending on the proportion of the algorithm that can be de-serialized, it is possible to reach near to linear scale speedups.
- parallel can be incorporated into other user commands as an optional speedup.
- Caveat: Has not been tested on Stata 15.
- Contribute, find help, and report bugs at http://github.com/gvegayon/parallel

# Thank you very much!

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