Title

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Intro 7 — Models for panel data

Description Remarks and examples Also see

Description

This introduction covers the command `cmxtmixlogit`. This is the only one of the `cm` estimation commands that explicitly models panel data. Other `cm` estimation commands, however, can be used with panel data when run with an appropriate variance estimator, that is, `vce(cluster` *idvar*`)`, `vce(bootstrap, cluster(`*idvar*`))`, or `vce(jackknife, cluster(`*idvar*`))`.

`cmxtmixlogit` fits a mixed logit model to panel choice data. `cmxtmixlogit` models a sequence of choices rather than a single choice, as commands for cross-sectional data do. As with `cmmixlogit`, random coefficients can be fit to model the correlation of choices across alternatives, and the property of independence of irrelevant alternatives (IIA) is not assumed. See *Overview of CM commands for discrete choices* in [CM] **Intro 5**, and see [CM] **Intro 8** if you are not familiar with this assumption.

Remarks and examples

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Remarks are presented under the following headings:

Data layout for panel choice data

In panel choice data, decision makers make multiple choices at different times. The data layout is similar to that for cross-sectional data, the difference being that there are repeated cases for each decision maker.

Here is an example of panel choice data. These fictitious data represent individuals' choices of transportation mode at multiple times. We list the data for the first two people:

```
. use https://www.stata-press.com/data/r17/transport
(Transportation choice data)
. list if id <= 2, sepby(t)
```

	id	t	alt	choice	trcost	trtime	age	income	parttime
1.	1	1	Car	1	4.14	0.13	3.0	3	Full-time
2.	1	1	Public	0	4.74	0.42	3.0	3	Full-time
3.	1	1	Bicycle	0	2.76	0.36	3.0	3	Full-time
4.	1	1	Walk	0	0.92	0.13	3.0	3	Full-time
5.	1	2	Car	1	8.00	0.14	3.2	5	Full-time
6.	1	2	Public	0	3.14	0.12	3.2	5	Full-time
7.	1	2	Bicycle	0	2.56	0.18	3.2	5	Full-time
8.	1	2	Walk	0	0.64	0.39	3.2	5	Full-time
9.	1	3	Car	1	1.76	0.18	3.4	5	Part-time
10.	1	3	Public	0	2.25	0.50	3.4	5	Part-time
11.	1	3	Bicycle	0	0.92	1.05	3.4	5	Part-time
12.	1	3	Walk	0	0.58	0.59	3.4	5	Part-time
13.	2	1	Car	0	4.36	0.23	3.0	2	Full-time
14.	2	1	Public	0	4.43	0.43	3.0	2	Full-time
15.	2	1	Bicycle	0	1.25	1.23	3.0	2	Full-time
16.	2	1	Walk	1	0.89	0.12	3.0	2	Full-time
17.	2	2	Car	0	7.14	0.23	3.2	3	Part-time
18.	2	2	Public	1	1.54	0.12	3.2	3	Part-time
19.	2	2	Bicycle	0	2.75	0.95	3.2	3	Part-time
20.	2	2	Walk	0	0.53	1.64	3.2	3	Part-time
21.	2	3	Car	0	6.69	0.17	3.4	2	Full-time
22.	2	3	Public	1	1.32	0.34	3.4	2	Full-time
23.	2	3	Bicycle	0	0.60	0.49	3.4	2	Full-time
24.	2	3	Walk	0	0.68	0.63	3.4	2	Full-time

Individuals (identified by the variable `id`) at each of three time points (time variable `t`) could choose between four modes of transportation (alternatives variable `alt`) with the one chosen alternative indicated by the binary variable `choice`. The first person chose to use a car at all three time points. The second person walked at time = 1 and took public transportation at the other two times.

Cost of travel (`trcost`, measured in $) and travel time (`trtime`, measured in hours) are alternative-specific variables. Variables `age` (measured in decades), `income` (annual income measured in $10,000), and `parttime` (indicating a part-time or full-time job) are case specific.

Before we can fit the model, we must `cmset` the data. For panel data, `cmset` requires three variables: first, the variable identifying individuals (`id`), second, the time variable (`t`), and third, the alternatives variable (`alt`). (`cmxtmixlogit`, like `cmmixlogit`, can fit models without explicitly identified alternatives. In this case, there is no alternatives variable, and the option `noalternatives` is specified.)

```
. cmset id t alt
note: case identifier _caseid generated from id and t.
note: panel by alternatives identifier _panelaltid generated from id and alt.

                    Panel data: Panels id and time t
              Case ID variable: _caseid
         Alternatives variable: alt
Panel by alternatives variable: _panelaltid (strongly balanced)
                 Time variable: t, 1 to 3
                         Delta: 1 unit

Note: Data have been xtset.
```

The notes displayed by `cmset` say it has created two new variables: `_caseid` and `_panelaltid`. Let's list their values for the first two individuals.

```
. list id t alt _caseid _panelaltid if id <= 2, sepby(alt) abbr(11)
```

	id	t	alt	_caseid	_panelaltid
1.	1	1	Car	1	1
2.	1	2	Car	2	1
3.	1	3	Car	3	1
4.	1	1	Public	1	2
5.	1	2	Public	2	2
6.	1	3	Public	3	2
7.	1	1	Bicycle	1	3
8.	1	2	Bicycle	2	3
9.	1	3	Bicycle	3	3
10.	1	1	Walk	1	4
11.	1	2	Walk	2	4
12.	1	3	Walk	3	4
13.	2	1	Car	4	5
14.	2	2	Car	5	5
15.	2	3	Car	6	5
16.	2	1	Public	4	6
17.	2	2	Public	5	6
18.	2	3	Public	6	6
19.	2	1	Bicycle	4	7
20.	2	2	Bicycle	5	7
21.	2	3	Bicycle	6	7
22.	2	1	Walk	4	8
23.	2	2	Walk	5	8
24.	2	3	Walk	6	8

`_caseid` is a variable that identifies cases. For choice model data, remember that a case is a single statistical observation but consists of multiple Stata observations. Each distinct value of panel ID × time represents a single statistical observation, that is, a case. The values of `_caseid` correspond to the distinct values of panel ID × time, in this example the values of `id` × `t`.

`_panelaltid` is a variable that uniquely identifies the distinct values of panel ID × alternative. We will explain why this variable is needed when we show an example with time-series operators. But you can skip over the explanation. These new variables make `cmxtmixlogit` work as you would expect. You need not be concerned about them, just leave them in your dataset.

A cmxtmixlogit model

Continuing with the previous example, we wish to model the effect of travel cost (`trcost`), travel time (`trtime`), income, and age on the choice of transportation mode.

We assume that all individuals have the same preferences with respect to travel cost but that preferences with respect to travel time are heterogeneous, and we model these heterogeneous preferences with a random coefficient for `trtime` by specifying the option `random(trtime)`.

The dependent variable is `choice`, the binary variable indicating which alternative was chosen. The variable `trcost` is included following the dependent variable; placing it in this position means that it should have a fixed coefficient. Specifying `casevars(age income)` includes the case-specific variables `age` and `income` in the model with fixed coefficients.

```
. cmxtmixlogit choice trcost, random(trtime) casevars(age income)
Fitting fixed parameter model:
Fitting full model:
Iteration 0:   log simulated-likelihood =  -1025.707  (not concave)
Iteration 1:   log simulated-likelihood = -1014.2513
Iteration 2:   log simulated-likelihood = -1006.2212
Iteration 3:   log simulated-likelihood = -1005.9904
Iteration 4:   log simulated-likelihood = -1005.9899
Iteration 5:   log simulated-likelihood = -1005.9899
```

```
Mixed logit choice model                        Number of obs        =      6,000
                                                Number of cases      =      1,500
Panel variable: id                              Number of panels     =        500
Time variable: t                                Cases per panel: min =          3
                                                                 avg =        3.0
                                                                 max =          3
Alternatives variable: alt                      Alts per case:   min =          4
                                                                 avg =        4.0
                                                                 max =          4
Integration sequence:      Hammersley
Integration points:               594           Wald chi2(8)         =     432.68
Log simulated-likelihood = -1005.9899           Prob > chi2          =     0.0000
```

choice	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
alt						
trcost	-.8388216	.0438587	-19.13	0.000	-.9247829	-.7528602
trtime	-1.508756	.2641554	-5.71	0.000	-2.026492	-.9910212
/Normal						
sd(trtime)	1.945596	.2594145			1.498161	2.526661
Car	(base alternative)					
Public						
age	.1538915	.0672638	2.29	0.022	.0220569	.2857261
income	-.3815444	.0347459	-10.98	0.000	-.4496451	-.3134437
_cons	-.5756547	.3515763	-1.64	0.102	-1.264732	.1134222
Bicycle						
age	.20638	.0847655	2.43	0.015	.0402426	.3725174
income	-.5225054	.0463235	-11.28	0.000	-.6132978	-.4317131
_cons	-1.137393	.4461318	-2.55	0.011	-2.011795	-.2629909
Walk						
age	.3097417	.1069941	2.89	0.004	.1000372	.5194463
income	-.9016697	.0686042	-13.14	0.000	-1.036132	-.7672078
_cons	-.4183279	.5607111	-0.75	0.456	-1.517302	.6806458

The coefficients for `trcost` and `trtime` are negative, indicating that as cost and travel time increase, the probability of selecting a method of travel decreases. In the `Public`, `Bicycle`, and `Walk` sections of the output, we see coefficients for the case-specific variables. These are each interpreted relative to the base alternative `Car`. We can use `margins` to more easily interpret the results of this model; see [CM] **Intro 1** and [CM] **margins**.

Because we did not specify a distribution in the `random()` option, we got the default distribution for the random coefficient, which is the normal distribution. Other options for the distribution are available. If we had multiple variables in the `random()` option, we could optionally specify `corrmetric()` to pick the form of the correlation modeled. See [CM] **cmxtmixlogit** for more information on options for random coefficients.

Time-series operators

When you cmset panel data with specified alternatives, your data are automatically xtset. You can type xtset to see the settings:

```
. xtset
Panel variable: _panelaltid (strongly balanced)
 Time variable: t, 1 to 3
         Delta: 1 unit
```

_panelaltid becomes the "panel" identifier for viewing the data as xt data. This is why cmxtmixlogit needs this variable. It is created so you can use Stata's time-series operators (see [U] **11.4.3.6 Using factor variables with time-series operators**) with cmxtmixlogit. For instance, if you want to include lags of alternative-specific variables in your model, the lags must be specific to the alternative, and Stata's time-series lag operator needs to know how to do this.

To illustrate, we add a lag trtime to our earlier model. We also specify correlated for the random coefficients of trtime and its lag so that the distributions of the random coefficients are correlated. Note that because of the additional complexity of this model, it is computationally intensive and may take a few minutes to fit.

```
. cmxtmixlogit choice, random(trtime L.trtime, correlated) casevars(age income)
Fitting fixed parameter model:
Fitting full model:
Iteration 0:   log simulated-likelihood = -726.49438  (not concave)
Iteration 1:   log simulated-likelihood = -725.73356
Iteration 2:   log simulated-likelihood = -724.30029
Iteration 3:   log simulated-likelihood = -720.40177
Iteration 4:   log simulated-likelihood = -720.28639
Iteration 5:   log simulated-likelihood = -720.07741
Iteration 6:   log simulated-likelihood = -720.07434
Iteration 7:   log simulated-likelihood = -720.07411
Iteration 8:   log simulated-likelihood = -720.07411
Refining estimates:
Iteration 0:   log simulated-likelihood = -720.07411
Iteration 1:   log simulated-likelihood = -720.07411
```

```
Mixed logit choice model                        Number of obs        =      4,000
                                                Number of cases      =      1,000
Panel variable: id                              Number of panels     =        500
Time variable: t                                Cases per panel: min =          2
                                                                 avg =        2.0
                                                                 max =          2
Alternatives variable: alt                      Alts per case:   min =          4
                                                                 avg =        4.0
                                                                 max =          4
Integration sequence:      Hammersley
Integration points:               625           Wald chi2(8)         =      82.87
Log simulated-likelihood = -720.07411           Prob > chi2          =     0.0000
```

choice	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
alt						
trtime						
--.	-1.02391	.3884411	-2.64	0.008	-1.785241	-.2625792
L1.	-.7797073	.3843897	-2.03	0.043	-1.533097	-.0263174
/Normal						
sd(trtime)	.8594882	.6233604			.2074386	3.56115
corr(trtime, L.trtime)	.4457922	.6071271	0.73	0.463	-.7639532	.9614328
sd(L.trtime)	1.576005	.4241405			.9299912	2.670768
Car	(base alternative)					
Public						
age	.0848749	.0715193	1.19	0.235	-.0553005	.2250502
income	-.208774	.0336985	-6.20	0.000	-.2748219	-.1427261
_cons	.1079519	.3923718	0.28	0.783	-.6610826	.8769865
Bicycle						
age	.2542854	.1066569	2.38	0.017	.0452418	.4633291
income	-.3155109	.0531635	-5.93	0.000	-.4197094	-.2113123
_cons	-.462521	.5845974	-0.79	0.429	-1.608311	.6832688
Walk						
age	.5830396	.1878859	3.10	0.002	.21479	.9512892
income	-.8183397	.1207108	-6.78	0.000	-1.054929	-.5817508
_cons	.0269189	.9301377	0.03	0.977	-1.796118	1.849955

Including the lag of `trtime` in this model may not have made much conceptual sense, but we did so for the purpose of showing how to use time-series operators with `cmxtmixlogit`.

Using other cm estimation commands with panel data

`cm` estimation commands for cross-sectional data can also be used with panel data. The estimates from these commands have a population-averaged interpretation when used with panel data. The `cmsettings` tell these cross-sectional `cm` commands that the data are panel data. In this case, by default, the `cm` commands report cluster–robust standard errors that account for the within-panel correlation.

Here is what we get if we run a cmclogit model on our previous panel choice data.

```
. cmclogit choice trcost trtime, casevars(age income)
note: data were cmset as panel data, and the default vcetype for panel data is
      vce(cluster id); see cmclogit.

Iteration 0:   log pseudolikelihood = -1197.9902
Iteration 1:   log pseudolikelihood = -1035.4817
Iteration 2:   log pseudolikelihood = -1027.6346
Iteration 3:   log pseudolikelihood = -1027.6227
Iteration 4:   log pseudolikelihood = -1027.6227

Conditional logit choice model                  Number of obs      =      6,000
Case ID variable: _caseid                       Number of cases    =       1500

Alternatives variable: alt                      Alts per case: min =          4
                                                               avg =        4.0
                                                               max =          4

                                                   Wald chi2(8)    =     335.13
Log pseudolikelihood = -1027.6227                  Prob > chi2     =     0.0000

                                  (Std. err. adjusted for 500 clusters in id)
```

choice	Coefficient	Robust std. err.	z	P>|z|	[95% conf.	interval]
alt						
trcost	-.7667673	.0464592	-16.50	0.000	-.8578258	-.6757089
trtime	-.6572159	.1700226	-3.87	0.000	-.990454	-.3239778
Car	(base alternative)					
Public						
age	.1588594	.0784292	2.03	0.043	.0051409	.3125779
income	-.3479798	.0405743	-8.58	0.000	-.4275039	-.2684557
_cons	-.8253419	.3651235	-2.26	0.024	-1.540971	-.109713
Bicycle						
age	.2025874	.0867382	2.34	0.020	.0325835	.3725912
income	-.4538989	.0436598	-10.40	0.000	-.5394705	-.3683273
_cons	-1.505446	.4571108	-3.29	0.001	-2.401367	-.6095252
Walk						
age	.307546	.1077107	2.86	0.004	.0964369	.518655
income	-.7648748	.0616934	-12.40	0.000	-.8857917	-.6439579
_cons	-.959179	.5054328	-1.90	0.058	-1.949809	.0314511

By default, cmclogit used the variance estimator given by vce(cluster id). If you wish to change the variance estimator, simply set the vce() option to what you want.

Also see

[CM] **Intro 1** — Interpretation of choice models

[CM] **Intro 2** — Data layout

[CM] **Intro 3** — Descriptive statistics

[CM] **Intro 4** — Estimation commands

[CM] **cmclogit** — Conditional logit (McFadden's) choice model

[CM] **cmxtmixlogit** — Panel-data mixed logit choice model