

Useful Stata Commands for Longitudinal Data Analysis

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Nuts and Bolts I

First some „Nuts and Bolts“ about data preparation with Stata.

Mathematical and Logical Expressions

+	add	~ [[]]	not	<	less than	ln()	natural log
-	subtract	&	and	<=	less than or equal	exp()	exponential
/	divide		or	>	greater than	sqrt()	square root
*	multiply			=	equal	abs()	absolute
^	power			~= [[]]	not equal		

RECODE

```
recode varname 1 3/5=7 //1 and 3 through 5 changed to 7

recode varname 2=1 .=. *=0 //2 changed to 1, all else is 0, . stays .
recode varname (2=1 yes) (nonmiss=0 no) //the same including labels, () needed

recode varname 5/max=max //5 through maximum changed to maximum (. stays .)
recode varname 1/2=1.5 2/3=2.5 //2 is changed to 1.5
recode varlist (2=1)(nonmiss=0) //you need () if recoding a varlist
```

Creating a Dummy

```
recode varname (2=1 yes) (nonmiss=0 no), into(dummy) //elegant solution I
generate dummy = varname==2 if varname<. //elegant solution II
tab varname, gen(dummy) //most simple but boring
```

Nuts and Bolts II

Comments

```
* ignore the complete line // ignore the rest excluding line break
/* ignore the text in between */ /// ignore the rest including line break
```

Be careful with missing values:

. == +∞, this might produce unwanted results. For instance, if you want to group a variable X, this is what you get

```
gen Xgrouped = X>2
```

	X	Xgrouped
1.	3	1
2.	2	0
3.	.	1
4.	1	0
5.	4	1

* better:

```
gen Xgrouped = X>2 if X<.
```

N.B.: . < .a < .b < ...

N.B.: X==. is true only if .
missing(X) is true for
all missing values

Data in wide-format: counting values in varlists

```
. egen numbl = anycount(var1-var3), v(1)
. egen numbmis = rowmiss(var1-var3)
. list var1 var2 var3 numbl numbmis
```

	var1	var2	var3	numbl	numbmis
1.	1	0	.	1	1
2.	1	0	0	1	0
3.	1	1	0	2	0
4.	1	1	1	3	0

Further example: number of valid episodes

```
egen nepi = rownonmiss(ts*)
```

Further example: max in "time finish"

```
egen maxage = rowmax(tf*)
```

Nuts and Bolts III

Missing Values

```
misstable summarize //gives overview of MV in the data
misstable patterns //MV patterns in the data
mvdecode _all, mv(-1) //-1 is set to . in all variables
mark nomiss //generates markervariable "nomiss"
markout nomiss Y X1 X2 X3 //0=somewhere missing, 1=nowhere missing
drop if nomiss == 0 //listwise deletion
```

Value-Label

```
label define geschlbl 1 "Man" 2 "Woman"
label value sex geschlbl
```

Display a Scalar

```
display 5*8
```

Regression Coefficients

```
regress, coeflegend //shows names of coefficients
display _b[bild] //displays a coefficient
```

Formating Output (permanent!)

```
set cformat %9.4f, permanently //format of coeff, S.E, C.I.
set pformat %5.3f, permanently //format of p-value
set showbaselevels on, permanently //display reference category
```

Nuts and Bolts IV

IF-Command

```
if expression {
  commands           //commands are executed if expression is true
}
```

GLOBAL Macros

```
* Directory where the data are stored
global pfa1 `"'I:\Daten\SOEP Analysen\Zufriedenheit\Fullsample\'"'
* Load data
cd $pfa1           //$pfa1 is expanded to "I:\Daten\..."
use Happiness, clear
```

Working with date functions

```
* Date information is transformed in "elapsed months since Jan. 1960"
gen birth = ym(birthy,birthm) //mdy(M,D,Y) if you have also days
gen birthc=birth
format birthc %tm           // %td if you have elapsed days
```

	id	birthy	birthm	birth	birthc
1.	1	1961	4	15	1961m4
2.	2	1963	11	46	1963m11

Note that Jan.1960 is month 0 here!!

Matching datasets: append and merge

A common task is to match information from different datasets

- **append**: Observations with information on the same variables are stored separately
- **merge**: Different variables are defined for the same observations, but stored separately

Consider the following SOEP example:

- We have the first two SOEP person data sets **ap.dta** and **bp.dta**
- The same 5 persons in each data set
- Variables: person id, year of wave, happiness (11-point scale 0-10, 10=very happy)

ap.dta			
	id	year	happy
1.	901	84	8
2.	1001	84	9
3.	1101	84	6
4.	1201	84	8
5.	1202	84	8

bp.dta			
	id	year	happy
1.	901	85	8
2.	1001	85	6
3.	1101	85	7
4.	1201	85	8
5.	1202	85	8

Matching datasets: append

	id	year	happy
1.	901	84	8
2.	1001	84	9
3.	1101	84	6
4.	1201	84	8
5.	1202	84	8
6.	901	85	8
7.	1001	85	6
8.	1101	85	7
9.	1201	85	8
10.	1202	85	8

append the rows of the second file
beyond the last row of the first:

```
use ap.dta
append using bp.dta
```

ap.dta is the master-file
bp.dta is the using-file

	id	year	happy
1.	901	84	8
2.	901	85	8
3.	1001	84	9
4.	1001	85	6
5.	1101	84	6
6.	1101	85	7
7.	1201	84	8
8.	1201	85	8
9.	1202	84	8
10.	1202	85	8

```
sort id year
```

Grouping observations of persons
together and ordering them by year
results in a

→ **panel dataset in long-format.**

Each row is called a

→ **“person-year”.**

Matching datasets: merge

Suppose that, for the persons in ap.dta, you need additional information on
variable `hhinc` which is stored in apequiv.dta. To match variables on identical
observations we can use `merge`.

ap.dta			
	id	year	happy
1.	901	84	8
2.	1001	84	9
3.	1101	84	6
4.	1201	84	8
5.	1202	84	8

apequiv.dta			
	id	year	hhinc
1.	901	84	9136.79
2.	1001	84	5773.51
3.	1101	84	10199.25
4.	1201	84	19776.77
5.	1202	84	19776.77

```
use ap.dta
merge 1:1 id using apequiv.dta
```

	id	year	happy	hhinc	_merge
1.	901	84	8	9136.79	3
2.	1001	84	9	5773.51	3
3.	1101	84	6	10199.25	3
4.	1201	84	8	19776.77	3
5.	1202	84	8	19776.77	3

STATA added a variable `_merge` which
equals 3 for all observations. This
indicates that all observations are part of
both files. If there were observations
which occur only in ap.dta (the master-
file), these would get value 1. Obs. which
occur only in apequiv.dta (the using-file),
would have `_merge==2`. (Naturally, obs.
of the first type would have missings on
`hhinc`, and obs. of the second type would
have missings on `happy`.)

Reshaping datasets from wide- to long-format

```
+-----+
| id ts1 tf1 st1 fail1 ts2 tf2 st2 fail2 ts3 tf3 st3 fail3 educ |
+-----+
| 1 19 22 1 1 22 26 2 1 26 29 1 0 9 |
+-----+
| 2 23 28 1 1 28 30 2 0 . . . . 13 |
+-----+
```

```
reshape long ts tf st fail, i(id) j(episode)
```

```
+-----+
| id episode ts tf st fail educ |
+-----+
| 1 1 19 22 1 1 9 |
| 1 2 22 26 2 1 9 |
| 1 3 26 29 1 0 9 |
+-----+
| 2 1 23 28 1 1 13 |
| 2 2 28 30 2 0 13 |
| 2 3 . . . . 13 |
+-----+
```

Here we have two persons, with 3 episodes each. In wide format all variables from the same episode need a common suffix. Here we simply numbered the episodes. The command for transforming in long format is `reshape long`. Then we list all episode-specific variables (without suffix). `i()` gives the person identifier variable and `j()` the new episode identifier variable created by Stata. All constant variables are copied to each episode.

How to repeat yourself without going mad: Loops

An extremely helpful technique to do tasks over and over again are loops. In Stata, there are (among others) `foreach`-loops and `forvalues`-loops. Both work in a similar way: They set a user-defined local macro to each element of a list of strings or list of numbers, and then execute the commands within the loop repeatedly, assuming that one element is true after the other.

```
foreach lname in list {
    commands referring to `lname`           //best for looping over strings
}                                           //or variable lists

forvalues lname = numlist {
    commands referring to `lname`           //best for looping over numbers
}
```

“`lname`” is the name of the local macro, “`list`” is any kind of list, “`numlist`” is a list of numbers (Examples: `1/10` or `0(10)100`).

The local can then be addressed by ``lname'` in the commands.

Loops

To append files ap.dta, bp.dta,..., wp.dta, one could type many appends. However, the following does the same much more efficiently:

```
use ap.dta
foreach wave in b c d e f g h i j k l m n o p q r s t u v w {
    append using `wave'p.dta
}
```

foreach also recognizes varlists:

```
foreach var of varlist ts1-ts10 {
    replace `var'=. if `var'==-3
}
```

forvalues loops over numlists:

```
forvalues k=1/10 {
    replace ts`k'=. if ts`k'==-3
}
```

Second counter:

"k" is the counter. Sometimes we need a second counter, derived from the first:

```
forvalues k=1/100 {
    local l=`k'+1
    ...
}
```

Finding the month from a date variable:

Imagine the month an event has happened is measured in months since January 1983 (months83). From this we want to create a new variable (month) telling us, in which month (January, ..., December) the event happened:

```
gen month = 0
forvalues j=1/12 {
    forvalues k=`j'(12)280 {
        quietly replace month = `j' if months83==`k'
    }
} //note that Jan.83 is 1 here!!
```

Loops Example: Converting EH Data to Panel Data

Note: Data are in process time (i.e. age). Therefore, we produce also panel data on an age scale (sequence data). Normally, panel data are in calendar time (i.e. years).

```
+-----+
|id ts1 tf1 st1 fail1 ts2 tf2 st2 fail2 ts3 tf3 st3 fail3 educ |
|-----|
| 1 19 22 1 1 22 26 2 1 26 29 1 0 9 |
|-----|
| 2 23 28 1 1 28 30 2 0 . . . . 13 |
|-----|
+-----+
```

```
egen maxage = rowmax(tf*) //generate the max value for the looping

forvalues j = 15/30 { //panels from age 15 to age 30
    generate s`j' = 0 if `j' < maxage //initializing with 0
    forvalues k = 1/3 {
        replace s`j' = st`k' if (`j' >= ts`k' & `j' < tf`k')
    }
}
```

```
+-----+
|id s15 s16 s17 s18 s19 s20 s21 s22 s23 s24 s25 s26 s27 s28 s29 s30 |
|-----|
| 1 0 0 0 0 1 1 1 2 2 2 2 1 1 1 . . |
|-----|
| 2 0 0 0 0 0 0 0 0 1 1 1 1 1 2 2 . |
|-----|
+-----+
```

Computations within panels (long-format)

- With panel data one often has to do computations within panels (groups)
- This is an example of a panel data set in long-format
 - Each record reports the observations on a person (id) in a specific year
 - This is termed “person-year”
 - A “panel” is defined as all person-years of a person

	id	year	X
1.	1	84	2
2.	1	85	4
3.	1	86	1
4.	1	87	6
5.	1	88	4
6.	2	84	3
7.	2	85	4

The basic idea

It is essential that one knows the following:

```
bysort bylist1 (bylist2): command
```

the by prefix; data are sorted according to bylist1 and bylist2 and computations are done for the groups defined by bylist1

_n system variable, contains the running number of the observation

_N system variable, contains the maximum number of observations

- This does the computations separately for each panel:

```
sort id  
by id: command
```

- bysort id: is a shortcut

- If the time ordering within the panels is important for the computations then use

```
sort id year  
by id: command
```

- bysort id (year): is a shortcut

Numbering person-years

Example: Numbering the person-years

```
gen recnr = _n //assigns a record ID
bysort id (year): gen pynr = _n //person-year ID (within person)
bysort id: gen pycount = _N //# of person-years (within person)
```

	id	year	X	recnr	pynr	pycount
1.	1	84	2	1	1	5
2.	1	85	4	2	2	5
3.	1	86	1	3	3	5
4.	1	87	6	4	4	5
5.	1	88	4	5	5	5
6.	2	84	3	6	1	2
7.	2	85	4	7	2	2

N.B.: If you now drop person-years, due to missing values (casewise deletion) pycount is no longer correct! Compute it anew.

Example: Statistics over persons

```
tabulate pycount if pynr==1 //distribution of person-years
```

Example: Identifying specific person-years

```
bysort id (year): gen first = 1 if _n==1 //first person-year
bysort id (year): gen last = 1 if _n==_N //last person-year
```

Using information from the year before

Explicit subscripting

It is possible to address specific values of a variable X (within a group) by using subscripts:

```
X[1] //X value of first person-year
X[_N] //X value of last person-year
X[_n-1] //X value of person-year before (X[0] is .)
```

```
bysort id (year): gen firstx = X[1] //firstx contains the first X-value
```

Example: Computing growth

```
bysort id (year): gen grx = (X - X[_n-1]) / X[_n-1]
```

	id	year	X	grx
1.	1	84	2	.
2.	1	85	4	1
3.	1	86	1	-.75
4.	1	87	6	5
5.	1	88	4	-.3333333
6.	2	84	3	.
7.	2	85	4	.3333333

Note:
Always think about
how your solution
behaves at the
first person-year!

Using the lag-operator

The Lag-Operator “L.” uses the observation in t-1. If this observation does not exist (due to a gap in the data) L.X returns a missing. X[_n-1] returns the value of the observation before, irrespective of any gaps.

```
bysort id (year): gen xn_1 = X[_n-1]
xtset id year
gen lx = L.X
```

	id	year	X	xn_1	lx
1.	1	84	2	.	.
2.	1	85	5	2	2
3.	1	87	3	5	.
4.	1	88	7	3	3

Finding statistics of X within persons

The egen command is very helpful (many more functions are available, see help egen):

```
bysort id (year): gen cumx = sum(X) //Summing up X
bysort id: egen maxx = max(X) //Maximum
bysort id: egen totx = total(X) //Sum
bysort id: egen meanx = mean(X) //Mean
bysort id: egen validx = count(X) //Number of nonmiss
bysort id: gen missx = _N-validx //Number of missings
```

	id	year	X	cumx	maxx	totx	meanx	validx	missx
1.	1	84	2	2	7	17	4.25	4	0
2.	1	85	5	7	7	17	4.25	4	0
3.	1	86	3	10	7	17	4.25	4	0
4.	1	87	7	17	7	17	4.25	4	0
5.	2	84	4	4	6	10	5	2	1
6.	2	85	.	4	6	10	5	2	1
7.	2	86	6	10	6	10	5	2	1

Variant: Finding statistics within person-episodes (spells)

Assume that we have a spell-indicator variable “spell”

```
bysort id: gen minX1 = min(X) if spell==1 //minimum within spelltype 1
bysort id spell: gen minX = min(X) //minimum within each spelltype
```

Deriving time-varying covariates I

In this context the **function sum(exp)** is very important (exp is a logical expression)

- exp can be 1 (true), 0 (false), or .
- sum(exp) returns a 0 in the first person-year also if exp==.

```
* marr is an indicator variable for the person-year of marriage
bysort id (year): gen married = sum(marr==1)          //married=1 after marriage
bysort id (year): gen ybefore = married[_n+1]-married //the year before marriage

* lf gives the activity status (0=out of lf, 1=employed, 2=unemployed)
bysort id (year): gen lfchg = sum(lf~=lf[_n-1] & _n~=1)    //# of changes in lf
```

	id	year	marr	lf	married	ybefore	lfchg
1.	1	84	-1	0	0	0	0
2.	1	85	-1	0	0	1	0
3.	1	86	1	1	1	0	1
4.	1	87	-1	1	1	0	1
5.	1	88	-1	1	1	.	1
6.	2	84	-1	0	0	0	0
7.	2	85	-1	0	0	1	0
8.	2	86	1	1	1	0	1
9.	2	87	-1	2	1	1	2
10.	2	88	1	1	2	.	3

Deriving time-varying covariates II

Identifying first and last occurrences of specific states. Here unemployment (lf==2)

```
* Identifying the first occurrence
bysort id (year): gen first = sum(lf==2)==1 & sum(lf[_n-1]==2)== 0

* Identifying the last occurrence
gsort id -year          //sorting in reverse time order
by id: gen last = sum(lf==2)==1 & sum(lf[_n-1]==2)==0    //do not sort again
sort id year
```

	id	year	lf	first	last
1.	1	84	0	0	0
2.	1	85	2	1	0
3.	1	86	1	0	0
4.	1	87	1	0	0
5.	1	88	2	0	1
6.	2	84	0	0	0
7.	2	85	0	0	0
8.	2	86	1	0	0
9.	2	87	2	1	1
10.	2	88	1	0	0

Copying time of first occurrence:

```
bysort id (first):    ///
gen yfirst = year[_N]
```

Missings / gaps in panels

When programming always be aware that there are certainly missings or even gaps (a whole person-year is missing) in the panels. These have the potential to wreck your analysis. Consider an example. We want to analyze the effect of being married on Y. We have a variable on civil status "fam" (0=single, 1=married, 2=divorce):

	id	year	fam
1.	1	84	0
2.	1	85	1
3.	1	86	1
4.	1	87	1
5.	1	88	2
6.	2	84	0
7.	2	85	1
8.	2	86	.
9.	2	87	1
10.	2	88	1
11.	2	89	1

How to deal with the missing? In this case it might make sense to impute 1 (see the example below, on how this could be done). Normally, however, one would drop the whole person-year (`drop if fam==.`) and create thereby a gap. This has to be taken into regard, when constructing time-varying covariates (see next slide).

Missings / gaps in panels

Example: Years since marriage

```
* This is the correct solution taking gaps into account
recode fam 2/max=. , into(marr)           //marriage indicator (spell)
bysort id: egen ymarr = min(year) if marr==1 //finding marriage year
gen yrsmarr = year - ymarr                //years since marriage
```

```
* This produces a wrong result
bysort id (year): gen yrsmarr1 = sum(marr[_n-1]) if marr==1
```

	id	year	fam	marr	ymarr	yrsmarr	yrsmarr1
1.	1	84	0	0	.	.	.
2.	1	85	1	1	85	0	0
3.	1	86	1	1	85	1	1
4.	1	87	1	1	85	2	2
5.	1	88	2
6.	2	84	0	0	.	.	.
7.	2	85	1	1	85	0	0
8.	2	87	1	1	85	2	1
9.	2	88	1	1	85	3	2
10.	2	89	1	1	85	4	3

Lessons for panel data preparation

- Make yourself comfortable with
 - merge and append
 - reshape
 - foreach and forvalues
 - by-Prefix
 - egen-functions
 - Explicit subscripting
- Always think about what happens with your solution
 - In the first person-year
 - If there are missings in the panel
 - If there are gaps in the panel
- List, list, and list
 - After each programming step try to understand what is going on by listing a few persons (complicated persons with missings, gaps, ...)
 - `list id year ... if id<4, sepby(id)`

Complex Examples

The following slides contain more complex examples

Filling up missings with the value from before, but only in between valid observations

```

* This is only an exercise, this kind of imputation usually makes no sense
gen X = inc
bysort id (year): gen first = sum(X<.)==1 & sum(X[_n-1]<.)== 0 //first valid inc
gsort id -year
by id: gen last = sum(X<.)==1 & sum(X[_n-1]<.)==0 //last valid inc
bysort id (year): gen spell = sum(first)-sum(last[_n-1]) //spell "being in panel"
* Filling in the value from before (this produces a cascade effect)
bysort id (year): replace X=X[_n-1] if X==. & spell==1
* Running # of missings encountered (this is only a little add on)
bysort id (year): gen nmiss=sum(missing(inc))

```

	id	year	inc	X	first	last	spell	nmiss
1.	1	84	1000	1000	1	0	1	0
2.	1	85	1100	1100	0	0	1	0
3.	1	86	.	1100	0	0	1	1
4.	1	87	.	1100	0	0	1	2
5.	1	88	1400	1400	0	1	1	2
6.	2	84	.	.	0	0	0	1
7.	2	85	2300	2300	1	0	1	1
8.	2	86	.	2300	0	0	1	2
9.	2	87	2400	2400	0	1	1	2
10.	2	88	.	.	0	0	0	3

Imputation of missings by linear interpolation

```

gen X = inc
bysort id (year): gen first = sum(X<.)==1 & sum(X[_n-1]<.)== 0 //first valid inc
gsort id -year
by id: gen last = sum(X<.)==1 & sum(X[_n-1]<.)==0 //last valid inc
bysort id (year): gen spell = sum(first)-sum(last[_n-1]) //indicator for spell "being in panel"

gen spellm = 1 if spell==1 & X==. //indicator for missing spell (MS)
bysort id spellm: gen lspellm=_N if spellm==1 //length of missing spell
bysort id spellm (year): gen nrspellm=_n if spellm==1 //numbering the person-years of MS

bysort id (year): gen incb = X[_n-1] if spellm==1 & spellm[_n-1]==. //last inc before MS
gsort id -year
by id: gen inca = X[_n-1] if spellm==1 & spellm[_n-1]==. //first inc after MS
bysort id (incb): replace incb = incb[1] if spellm==1 //filling up incb
bysort id (inca): replace inca = inca[1] if spellm==1 //filling up inca
sort id year
replace X = incb + nrspellm * ((inca-incb)/(lspellm+1)) if spellm==1 //imputing missing inc

```

	id	year	inc	X	spellm	lspellm	nrspellm	incb	inca
1.	1	84	1000	1000
2.	1	85	1100	1100
3.	1	86	.	1200	1	2	1	1100	1400
4.	1	87	.	1300	1	2	2	1100	1400
5.	1	88	1400	1400
6.	2	84
7.	2	85	2300	2300
8.	2	86	.	2350	1	1	1	2300	2400
9.	2	87	2400	2400
10.	2	88

Note:
Works only if
there is only
one MS per id

Creating a balanced panel

Sometimes one would like to “blow up” the dataset to a balanced one. In the following example the max person-years is 3. We create a new dataset, where every id has 3 observations.

```
* Starting with the real data (data.dta)
* Creating a list of the ids (idlist.dta)
bysort id: keep if _n==1
keep id
save idlist.dta

clear
set obs 3 //number of observations in balanced panel
gen time = _n
cross using idlist.dta //all pair wise combinations of time and id
merge 1:1 id time using data.dta //merge the real data
```

DATA.DTA			
	id	time	X
1.	1	1	4
2.	1	2	7
3.	1	3	2
4.	2	1	3
5.	2	3	5
6.	3	2	8

	id	time	X	_merge
1.	1	1	4	3
2.	1	2	7	3
3.	1	3	2	3
4.	2	1	3	3
5.	2	2	.	1
6.	2	3	5	3
7.	3	1	.	1
8.	3	2	8	3
9.	3	3	.	1

Converting EH Data to Panel Data (EH data.do)

EH Data (Marriage Episodes), Calendar Axis

id	birthy	ts1	tf1	end1	ts2	tf2	end2	inty
1	1971	1990	1993	1	1997	.	.	2000
2	1970	1993	1998	2	.	.	.	2000

EH Data, Age Axis

id	birthy	ts1	tf1	end1	ts2	tf2	end2	ts3	inty
1	1971	19	22	1	26	29	3	30	2000
2	1970	23	28	2	31	.	.	.	2000

end*:
1=divorce
2=death spouse
3=censoring

Panel Data, Age 17-30 (0=single, 1=married, 2=divorced, 3=widowed) These data could be used as an input for a sequence analysis!

id	s17	s18	s19	s20	s21	s22	s23	s24	s25	s26	s27	s28	s29	s30
1	0	0	1	1	1	2	2	2	2	1	1	1	1	.
2	0	0	0	0	0	0	1	1	1	1	1	3	3	3

Converting EH Data to Panel Data (EH data.do)

```
gen ageint = inty - birthy          //age at interview
egen nepi  = rownonmiss(ts*)        //number of valid marriage episodes

* Preset state 0 (single) over the whole sequence
forvalues j = 17/30 {
    generate s`j' = 0 if `j' <= ageint
}

if nepi>0 {                          //The rest is only for those who married at least once

    forvalues k=1/2 {
        replace tf`k' = inty if `k'==nepi & tf`k'==. //imputing inty for censored episodes
        replace end`k' = 3 if `k'==nepi & end`k'==. //flaging censored episodes with end==3
    }

    forvalues k=1/2 {                          //converting years to age
        replace ts`k' = ts`k' - birthy
        replace tf`k' = tf`k' - birthy
    }

    forvalues k=1/2 {                          //setting the endpoint of the sequence
        replace ts`k'=ageint+1 if `k'==nepi+1
    }
    gen ts3 = ageint+1 if nepi==2

    forvalues j=17/30 {
        forvalues k=1/2 {
            local l=`k'+1
            quietly replace s`j' = 1 if `j'>=ts`k' & `j'<=tf`k' //married
            quietly replace s`j' = 2 if `j'>=tf`k' & `j'< ts`l' & end`k'==1 //divorced
            quietly replace s`j' = 3 if `j'>=tf`k' & `j'< ts`l' & end`k'==2 //widowed
        }
    }
}
```