

# Reformatting Longitudinal Data As Survival Data

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## Introduction

Below you will find a simulated data set that might help you think about constructing a data set for survival analysis, or event history analysis.

These simulated data represent a common situation in which a *categorical status* (like a diagnosis of depression or PTSD) is observed at different time points.

An *event* is defined as a (relatively sudden) change from a status of 0 to a status of 1.

## Get The Data

```
. clear all  
  
. use "simulated-survival-data.dta", clear
```

In this example, we are going to think about how to take a data set of *statuses* and turn it into a data set of *events*.

## Look At The Data

Notice how this is a *wide* data set. Every individual has a *single row of data*, and information on **status** at each of the timepoints is contained in the *same row*.

```
. list
```

	id	status1	status2	status3
1.	1	0	0	0
2.	2	0	1	1
3.	3	0	0	1

## Notice How Individual Status Changes Over Time.

1. Status never changes for individual 1. The event never occurs for individual 1. Individual 1's event time is therefore *censored*.
2. Status changes at wave 2 for individual 2. The event therefore occurs for individual 2 at wave 2. Individual 2's event time is therefore *observed*.
3. Status changes at wave 3 for individual 3. The event is therefore conceptualized as occurring for individual 3 at wave 3. Individual 3's event time is therefore *observed*.

## How Do We Turn This Data Set Into A Data Set of Event Times?

First, we want to make sure that it is appropriate to conceptualize this data set of individuals as a data set for whom the event has not yet occurred.

Second, we want to create an *event time* out of these *status changes*.

Our code might look something as follows.

I am assuming in the code below that waves are 1 year apart, and you might want to adjust your code accordingly if waves are differentially spaced.

### Generate an Event Time

```
. * initialize to longest time

. * censored observations will have this value

. generate event_time = 3

. * change event time to 2 if status2 == 2
. * change event time to 1 if status1 == 1

. * notice that I am doing this in reverse order
. * to capture the earliest event time

. replace event_time = 2 if status2 == 1 // event time is 2 if status 2 is 1
(1 real change made)

. replace event_time = 1 if status1 == 1 // event time is 1 if status 1 is 1
(0 real changes made)
```

### Generate A Failure (Censoring) Indicator

```
. * failure becomes 1 for those
. * for whom event occurred at some timepoint

. generate failure = 0 // initialize

. replace failure = 1 if status1 == 1 | status2 == 1 | status3 == 1 // change failure to 1 if any
> status variable == 1
(2 real changes made)
```

You can see that our data now have an event time, and a censoring status.

```
. list, abbreviate(10) // list out the data
```

	id	status1	status2	status3	event_time	failure
1.	1	0	0	0	3	0
2.	2	0	1	1	2	1
3.	3	0	0	1	3	1

### stset the data

Inspection of the results from the `stset` command indicates that the data appears to have been `stset` correctly.

```

. stset event_time, failure(failure == 1)
      failure event:  failure == 1
obs. time interval:  (0, event_time]
exit on or before:  failure

```

---

3	total observations	
0	exclusions	

---

3	observations remaining, representing	
2	failures in single-record/single-failure data	
8	total analysis time at risk and under observation	
	at risk from t =	0
	earliest observed entry t =	0
	last observed exit t =	3

## Graph of Survival Function

```

. sts graph, scheme(michigan)
      failure _d:  failure == 1
      analysis time _t:  event_time

. graph export simulated-survival-data.png, width(1000) replace
(file simulated-survival-data.png written in PNG format)

```

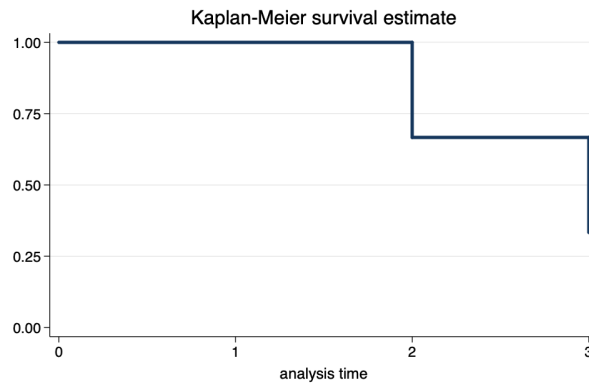


Figure 1: Kaplan-Meier Survivor Function