

Competing-risks regression

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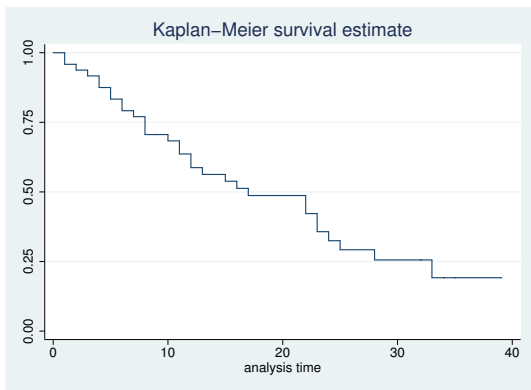


1. Overview
2. Competing-risks events
3. Cumulative incidence functions
4. The Cox regression approach
5. The Fine and Gray approach (`stcrreg`)
6. Concluding remarks

- Competing risks are events that prevent an event of interest from occurring, rather than just prevent you from seeing it happen (censoring)
- When you have competing events, you want to focus on the cumulative incidence function rather than the survival function
- You can use Cox regression with competing risks, but you have to either
 1. Modify your interpretations
 2. Go through a lot of work to assess covariate effects
- Competing-risks regression by the method of Fine and Gray (1999) is a useful alternative
- Implemented in the `stcrreg` command, new to Stata 11

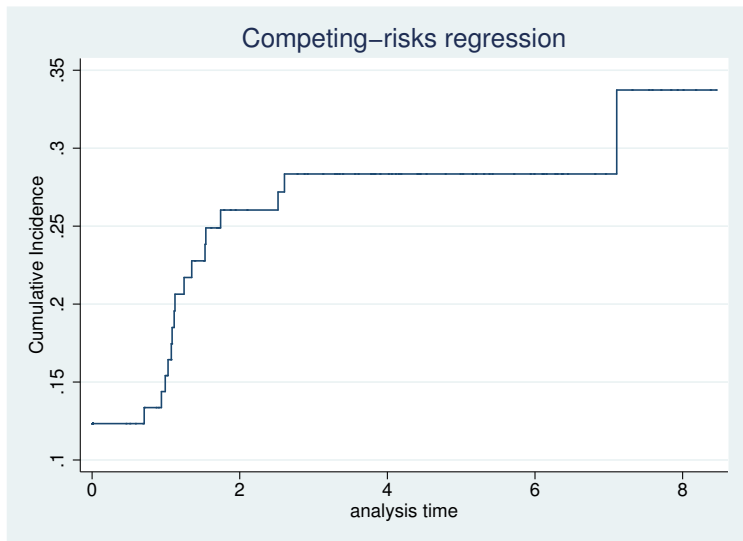
- A competing-risk event is an event that impedes what a researcher actually “wants” to see
- For example, if a researcher is interested in recurrence of breast cancer, a heart attack would be competing event
- This is not the same as censoring. With censoring, the event of interest still happens at a later time. You just don't get to see when
- What confuses the matter is that, for some purposes, it is okay to think of competing events as censorings

```
. sysuse cancer, clear  
(Patient Survival in Drug Trial)  
. stset studytime died  
  (output omitted)  
. sts graph
```



- If the event of interest is recurrent breast cancer, a Kaplan-Meier curve answers the question "What is the probability of no breast cancer before 5 months?"
- In a competing-risks setting, a Kaplan-Meier curve is inadequate for two reasons
- First, it fails to acknowledge that breast cancer may never occur, i.e. the probability of breast cancer after time zero is not equal to 1.0

- Second, with competing risks it is better to reverse the temporal ordering of the question
- It makes better sense to ask “What is the probability of breast cancer within 5 months?” than to ask “What is the probability that nothing happens for the first five months, but when something does happen I want it to be breast cancer and not a heart attack?”
- As such, you should use the cumulative incidence function (CIF) instead
- A CIF is just the probability that the event of interest occurs before a given time



Example

- Geskus (2000) analyzed data from 324 HIV-positive men from the Amsterdam Cohort Studies on HIV infection and AIDS.
- During the course of infection, the syncytium inducing (SI) phenotype appeared in some
- Appearance of SI typically means bad news
- Time to SI appearance in the absence of AIDS is the analysis of interest
- In this context, AIDS with no SI acts as the competing event
- Covariate of interest is variable `ccr5` indicating a mutation in the C-C chemokine receptor 5 gene

```
. webuse hiv_si, clear
(HIV and SI as competing risks)
. describe
Contains data from http://www.stata-press.com/data/r11/hiv_si.dta
  obs:          324                HIV and SI as competing risks
  vars:          4                 3 Apr 2009 13:40
  size:          5,184 (99.9% of memory free)  (_dta has notes)
```

variable name	storage type	display format	value label	variable label
patnr	int	%8.0g		ID
time	float	%9.0g		Years from HIV infection
status	byte	%10.0g	stat	1 = AIDS, 2 = SI, 0 = event-free
ccr5	byte	%9.0g	ccr5	1 if WM (deletion in C-C chemokine receptor 5 gene)

Sorted by:

- The research question is: “How does the `ccr5` mutation affect the incidence of SI in the presence of AIDS as a competing risk?”
- Standard Cox regression does not answer that question
- That does not mean it is completely useless, however
- First, you can modify your interpretation if you wish
- Second, you can use more Cox regression to answer our research question, but it is a lot of work. But it is possible

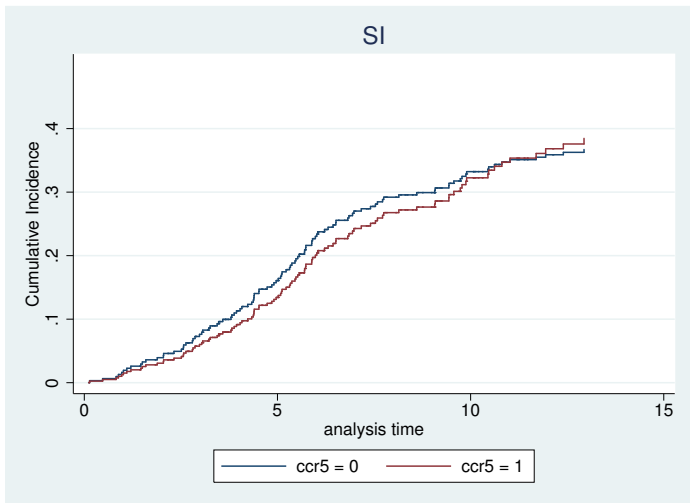
```

. stset time, failure(status == 2)           // SI is the event of interest
  (output omitted)
. stcox ccr5
  (output omitted)
Cox regression -- no ties
No. of subjects =           324                Number of obs   =           324
No. of failures =           107                LR chi2(1)       =           1.19
Time at risk    = 2261.959996                Prob > chi2      =           0.2748
Log likelihood  = -549.73443

```

_t	Haz. Ratio	Std. Err.	z	P> z	[95% Conf. Interval]
ccr5	.7755334	.1846031	-1.07	0.286	.4863914 1.23656

- The previous results may be interpreted as those for a world where the competing event does not occur
- A full CIF analysis with Cox regression would require the following additional steps
 1. Predict the hazard contributions from the previous Cox regression
 2. Fit a Cox regression with AIDS instead as the event of interest
 3. Predict the hazard contributions from that model fit
 4. Calculate the estimated CIF manually for both `ccr5==0` and `ccr5==1`; see page 209 of [ST] for details
 5. Plot the results
- Moral: Assessing covariate effects on the CIF using Cox regression is a lot of work



- An easier way to do CIF covariate analysis is with competing risks regression, according to the model of Fine and Gray (1999)
- They posit a model for the hazard of the subdistribution for the failure event of interest, known as the *subhazard*
- Put simply, they model the CIF directly knowing full well it is not a proper distribution function
- Covariates affect the subhazard proportionally, similar to Cox regression
- You do this in Stata 11 using `stcrreg`. `stcurve` after `stcrreg` will plot comparative CIFs for you

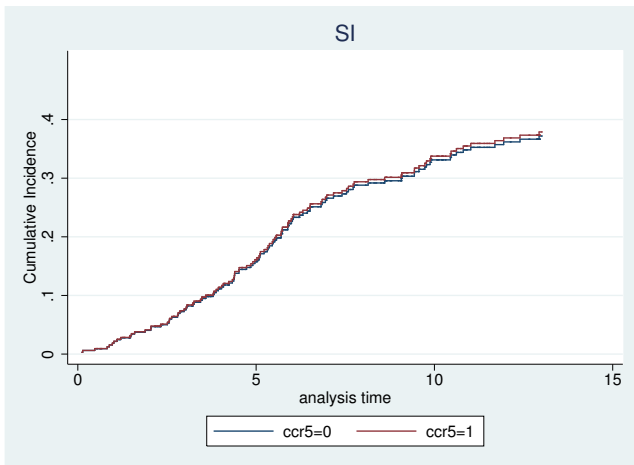
```

. stset time, failure(status == 2)           // SI is the event of interest
  (output omitted)
. stcrreg ccr5, compete(status == 1)       // AIDS is the competing event
  (output omitted)
Competing-risks regression
Failure event : status == 2
Competing event: status == 1
Log pseudolikelihood = -579.06241
No. of obs           = 324
No. of subjects      = 324
No. failed           = 107
No. competing        = 113
No. censored         = 104
Wald chi2(1)         = 0.01
Prob > chi2          = 0.9172

```

_t	SHR	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
ccr5	1.023865	.2324119	0.10	0.917	.6561827	1.597574


```
. stcurve, cif at1(ccr5=0) at2(ccr5=1) title(SI) range(0 13) yscale(range(0 0.5))
```



- Competing risks are events that prevent an event of interest from occurring
- If you have competing risks, you want to look at CIFs instead of survival functions
- CIF analysis with Cox regression is possible, but difficult
- `stcrreg` followed by `stcurve` is the easier way to go
- Keep in mind, however, that easier does not mean correct. There are model assumptions to be made for either of the two approaches

- Fine, J. and R. Gray. 1999. A proportional hazards model for the subdistribution of a competing risk. *Journal of the American Statistical Association*. **94**: 496–509.
- Geskus, R. B. 2000. On the inclusion of prevalent cases in HIV/AIDS natural history studies through a marker-based estimate of time since seroconversion. *Statistics in Medicine*. **19**: 1753–1769.