

Cox Proportional Hazards Model

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Introduction

“Survival analysis is a key technique in data-driven decision-making, which is now central to public interest because of COVID-19. Applying the correct technique for the specific question at hand is crucial for credible public health inferences. If you are interested in assessing how a risk factor or a potential treatment affects the progression of a disease—such as how long a patient takes to recover—then survival analysis techniques come into play. Survival analysis deeply respects the ultimate source of its data, often the disease experience or even the life and death of human patients. It seeks to exploit every last drop of information that this experience can render for saving lives—in particular, not only whether patients survived, but how long, and why. And it strives to do so with minimal assumptions, so that the data are truly driving the decision.”

—SAS Corporation

Cox Proportional Hazards Model

Formula for the Hazard

$h(t)$ the rate of occurrence.

$$h(t) = \lim_{\delta \rightarrow \infty} \frac{\text{probability of having an event before time } t + \delta}{\delta}$$

This definition per Johnson & Shih (2007).

$$h(t) = h_0(t)e^{\beta_1 x_1 + \beta_2 x_2 + \text{etc.}}$$

We don't directly estimate the hazard, but estimate the effect of covariates on the hazard.

The event (birth, death, program entry, program departure) is coded as 1, so we are estimating the association of the covariates with event occurrence.

Cox Proportional Hazards Model in Stata

Using a data set referenced frequently in Stata `help` and Stata YouTube Videos

```
. clear all  
  
. webuse drugtr // demonstration data set from Stata  
(Patient survival in drug trial)
```

Setup of Data

```
. stset // show st setup of data
-> stset studytime, failure(died)
Survival-time data settings
    Failure event: died!=0 & died<.
Observed time interval: (0, studytime]
Exit on or before: failure
```

```
48 total observations
0 exclusions
```

```
48 observations remaining, representing
31 failures in single-record/single-failure data
744 total analysis time at risk and under observation
    At risk from t = 0
    Earliest observed entry t = 0
    Last observed exit t = 39
```

```
. describe // show variables in data
Contains data from https://www.stata-press.com/data/r18/drugtr.dta
Observations: 48 Patient survival in drug trial
Variables: 8 3 Mar 2022 02:12
```

Variable name	Storage type	Display format	Value label	Variable label
studytime	byte	%8.0g		Months to death or end of exp.
died	byte	%8.0g		1 if patient died
drug	byte	%8.0g		Drug type (0=placebo)
age	byte	%8.0g		Patient's age at start of exp.
_st	byte	%8.0g		1 if record is to be used; 0 otherwise
_d	byte	%8.0g		1 if failure; 0 if censored
_t	byte	%10.0g		Analysis time when record ends
_t0	byte	%10.0g		Analysis time when record begins

```
Sorted by:
```

Kaplan-Meier Survivor Function (per Gabriela Ortiz, Stata)

$$S(t) = Pr(T > t)$$

```
. sts graph, scheme(michigan) // Kaplan-Meier Survivor Function
    Failure _d: died
    Analysis time _t: studytime

. graph export survival0.png, width(1000) replace
file
    /Users/agrogan/Desktop/GitHub/newstuff/categorical/survival-analysis-and-event-hi
> story/survival0.png saved as PNG format
```

Cox Proportional Hazards Model

```
. stcox age drug // run Cox Proportional Hazards Model
    Failure _d: died
    Analysis time _t: studytime
Iteration 0: Log likelihood = -99.911448
Iteration 1: Log likelihood = -83.551879
Iteration 2: Log likelihood = -83.324009
Iteration 3: Log likelihood = -83.323546
Refining estimates:
Iteration 0: Log likelihood = -83.323546
Cox regression with Breslow method for ties
No. of subjects = 48 Number of obs = 48
No. of failures = 31
Time at risk = 744
```

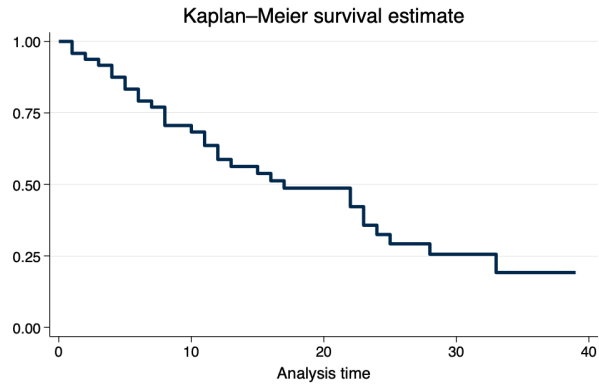


Figure 1: Kaplan-Meier Survivor Function

Log likelihood = -83.323546

LR chi2(2) = 33.18
Prob > chi2 = 0.0000

_t	Haz. ratio	Std. err.	z	P> z	[95% conf. interval]
age	1.120325	.0417711	3.05	0.002	1.041375 1.20526
drug	.1048772	.0477017	-4.96	0.000	.0430057 .2557622

Graph Survival Curves

```
. stcurve, survival scheme(michigan) // survival curve
note: function evaluated at overall means of covariates.

. graph export survival1.png, width(1000) replace
file
/Users/agrogan/Desktop/GitHub/newstuff/categorical/survival-analysis-and-event-hi
> story/survival1.png saved as PNG format
```

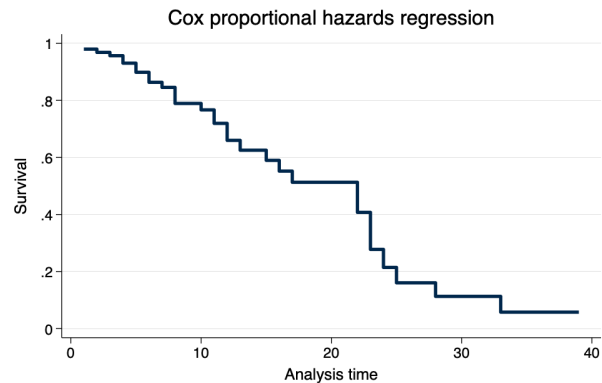


Figure 2: Survival Curve

```
. stcurve, survival at1(drug=0) at2(drug=1) scheme(michigan) // survival curve by group
note: function evaluated at specified values of selected covariates and overall means
of other covariates (if any).

. graph export survival2.png, width(1000) replace
file
/Users/agrogan/Desktop/GitHub/newstuff/categorical/survival-analysis-and-event-hi
```

```
> story/survival2.png saved as PNG format
```

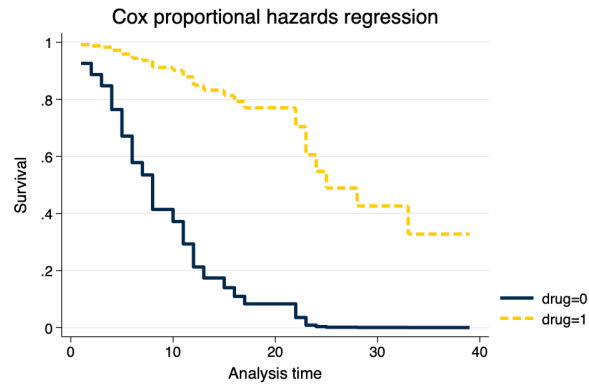


Figure 3: Survival Curve by Drug Group

Proportional Hazards Assumption

```
. estat phtest // formal test of PH assumption
```

Test of proportional-hazards assumption

Time function: Analysis time

	chi2	df	Prob>chi2
Global test	0.43	2	0.8064

```
. stphtplot, by(drug) scheme(michigan) // graphical test of PH assumption
```

Failure _d: died

Analysis time _t: studytime

```
. graph export ph.png, width(1000) replace
```

file

/Users/agrogan/Desktop/GitHub/newstuff/categorical/survival-analysis-and-event-hi

```
> story/ph.png saved as PNG format
```

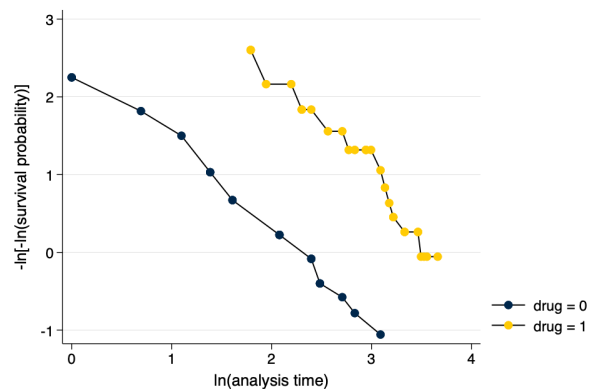


Figure 4: Graphical Assessment of Proportional Hazards Assumptions

References

Johnson, L. L., & Shih, J. H. (2007). CHAPTER 20 - An Introduction to Survival Analysis (J. I. Gallin & F. P. Ognibene, eds.). <https://doi.org/https://doi.org/10.1016/B978-012369440-9/50024-4>

Ragnar Frisch Centre for Economic Research (2020). Event History Analysis, Survival Analysis, Duration Analysis ,Transition Data Analysis, Hazard Rate Analysis. Oslo, Norway.