Biostatistics 140.623 Third Term, 2017-2018

Laboratory Exercise 3 Answer Key

The times to "drug failure" (as determined by a treating psychiatrist) for 25 patients in a study comparing a new treatment for schizophrenia to a standard treatment as used in Self-Evaluation Problems Class 5 are:

Trt group	Times (weeks)
Standard	3, 5+, 6, 8, 8, 9, 13, 15+, 16, 16, 17, 18
New	4, 6, 9, 9, 10+, 11, 12, 13+, 14+, 16, 17, 18, 20

⁺ denotes a censored observation

The corresponding dataset can be found in the Stata file trt.dta

1. Check the listing below to confirm that the data set has been set up appropriately:

. list				
	trt	weeks	failure	id
1.	0	3	1	1
2.	0	5	0	2
3.	0	6	1	3
4.	0	8	1	4
5.	0	8	1	5
6.	0	9	1	6
7.	0	13	1	7
8.	0	15	0	8
9.	0	16	1	9
10.	0	16	1	10
11.	1	14	0	11
12.	0	17	1	12
13.	0	18	1	13
14.	1	4	1	14
15.	1	6	1	15
16.	1	9	1	16
17.	1	9	1	17
18.	1	10	0	18
19.	1	11	1	19
20.	1	12	1	20
21.	1	13	0	21
22.	1	16	1	22
23.	1	17	1	23
24.	1	18	1	24
25.	1	20	1	25

20

- 2. The next step is to define the data as "survival time data" using the "stset" command. Here the event is defined as drug failure.
- . stset weeks, failure(failure==1) id(id)

3. The **drug failure incidence rates** can be obtained in the new and standard treatment groups:

last observed exit t =

```
.stir trt
```

note: Exposed <-> trt==1 and Unexposed <-> trt==0

	trt Exposed	Unexposed	 Total		
Failure Time	10 159	10 134	20 293		
Incidence rate	.0628931	.0746269	.0682594		
	Point	estimate	[95% Conf.	Interval]	
Inc. rate diff.	01	17338	0722224	.0487549	
Inc. rate ratio	.84	127673	.3148131	2.256122	(exact)
Prev. frac. ex.	.15	72327	-1.256122	.6851869	(exact)
Prev. frac. pop	.08	353242	İ		
•	(midp)	Pr(k<=10) =		0.3529	(exact)
	(midp) 2*	Pr(k<=10) =		0.7058	(exact)

4. Next, we can obtain the **Kaplan-Meier estimates** of the survivor "drug-failure free" function for each treatment group: (Compare these to your solutions to **Self-Evaluation Problems** Class 5).

sts list if trt==0

Time	Beg. Total	Fail	Net Lost	Survivor Function	Std. Error	[95% Con	f. Int.]
3	12	1	0	0.9167	0.0798	0.5390	0.9878
5	11	0	1	0.9167	0.0798	0.5390	0.9878
6	10	1	0	0.8250	0.1128	0.4609	0.9533
8	9	2	0	0.6417	0.1441	0.3022	0.8483
9	7	1	0	0.5500	0.1499	0.2321	0.7829
13	6	1	0	0.4583	0.1503	0.1689	0.7102
15	5	0	1	0.4583	0.1503	0.1689	0.7102
16	4	2	0	0.2292	0.1370	0.0382	0.5143
17	2	1	0	0.1146	0.1061	0.0067	0.3917
18	1	1	0	0.0000	•	•	•

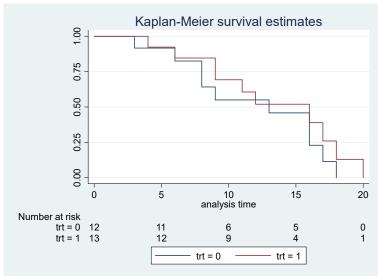
. sts list if trt==1

Time	Beg. Total	Fail	Net Lost	Survivor Function	Std. Error	[95% Cc	onf. Int.]
4	13	1	0	0.9231	0.0739	0.5664	0.9888
6	12	1	0	0.8462	0.1001	0.5122	0.9591
9	11	2	0	0.6923	0.1280	0.3734	0.8718
10	9	0	1	0.6923	0.1280	0.3734	0.8718
11	8	1	0	0.6058	0.1382	0.2943	0.8143
12	7	1	0	0.5192	0.1430	0.2246	0.7500
13	6	0	1	0.5192	0.1430	0.2246	0.7500
14	5	0	1	0.5192	0.1430	0.2246	0.7500
16	4	1	0	0.3894	0.1554	0.1152	0.6626
17	3	1	0	0.2596	0.1482	0.0454	0.5553
18	2	1	0	0.1298	0.1180	0.0076	0.4260
20	1	1	0	0.0000	•	•	•

5. The "survival curves" by treatment group can be plotted along with a risk table: Based upon the plot of the Kaplan-Meier curves for each treatment group, which treatment, if any, should be preferred?

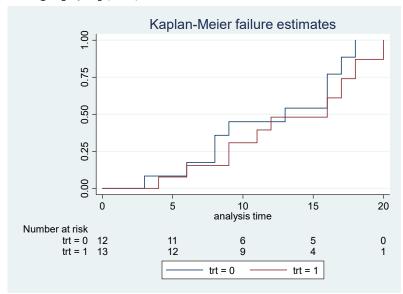
There is much overlap between the two curves and the sample sizes are very small. The curves do not appear to be distinguishable from each other.

.sts graph, by(trt) risktable



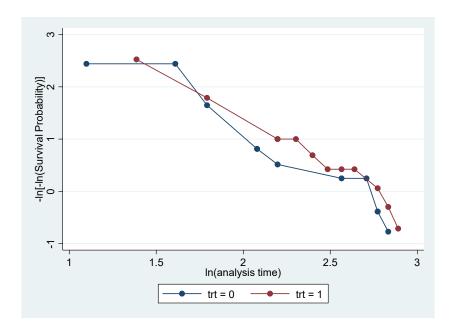
6. The cumulative hazard of drug failure by treatment group can be plotted using the "failure" option:

.sts graph, by(trt) failure risktable



7. What do you observe in the **complementary log log (CLL) plot** below?

```
.stphplot, by(trt)
```



We are hoping to observe a straight line relationship between the CLL function and log t. The sample sizes are very small, which makes the assessment difficult. However, a straight line relationship appears to hold for certain values of t. However, the curves cross during the early time period.

Also, the lines appear parallel for the two treatment groups; this would support the proportional hazards assumption that the ratio of hazards (comparing the two treatments) is constant (same) regardless of time t.

8. The **log-rank statistic** can be calculated to test the null hypothesis of no difference in overall survival by treatment.

```
failure _d: failure == 1
analysis time _t: weeks
id: id
```

Log-rank test for equality of survivor functions

trt	Events observed		Events expected
0	10 10	- 	8.27 11.73
Total	20		20.00
	<mark>chi2(1)</mark> Pr>chi2		0.76 0.3831

9. Set up the log-rank statistic to test whether overall drug failure differs between the two treatments. The log-rank test statistic can be computed by hand from the 2x2 tables based on each observed event time. In lab, you could split into groups, each group calculating the pieces needed from one 2x2 table.

where

$$\chi^{2}_{LR} = \frac{\left[\sum_{j} (a_{j} - E(a_{j}))\right]^{2}}{\sum_{j} Var(a_{j})} \text{ where } E(a_{j}) = d_{j} \cdot \frac{n_{ja}}{n_{j}} \text{ and } Var(a_{j}) = \frac{d_{j}(n_{j} - d_{j})n_{ja}n_{jb}}{n_{j}^{2}(n_{j} - 1)}$$

	Event	No Event	Total
Standard Trt	a_{j}		n _{ja}
New Trt	cj		n jb
Total	dj		nj

Observed Event Times:

Time=j = 3 weeks:

	Event	No Event	Total
Standard Trt	1	11	12
New Trt	0	13	13
Total	1	24	25

$$a_j=1$$
 $E(a_j)= (1)(12)/25$ $a_j-E(a_j)=13/25$ $E(a_j)=(1)(24)(12)(13)/[25^2(24)]=0.2496$

Time=4 weeks:

	Event	No Event	Total
Standard Trt	0	11	11
New Trt	1	12	13
Total	1	23	24

 $a_i = 0$ $E(a_i) = (1)(11)/24$ $a_i - E(a_i) = -11/24$ $Var(a_i) = (1)(23)(11)(13)/[24^2(23)] = 0.2483$

Time=6 weeks:

	Event	No Event	Total
Standard Trt	1	9	10
New Trt	1	11	12
Total	2	20	22

 $a_j = 1$ $E(a_j) = (2)(10)/22$ $a_j - E(a_j) = 2/22$ $Var(a_j) = (2)(20)(10)(12)/[22^2(21)] = 0.4723$

Time=8 weeks:

,	Event	No Event	Total
Standard Trt	2	7	9
New Trt	0	11	11
Total	2	18	20

 $a_i = 2$ $E(a_i) = (2)(9)/20$ $a_i - E(a_i) = 22/20$ $Var(a_i) = (2)(18)(9)(11)/[20^2(19)] = 0.4689$

Time=9 weeks:

	Event	No Event	Total
Standard Trt	1	6	7
New Trt	2	9	11
Total	3	15	18

 $a_j = 1$ $E(a_j) = (3)(7)/18$ $a_j - E(a_j) = 3/18$ $Var(a_j) = (3)(15)(7)(11)/[18^2(17)] = 0.6291$

Time=11 weeks:

1 Weeks				
	Event	No Event	Total	
Standard Trt	0	6	6	
New Trt	1	7	8	
Total	1	13	14	

 $a_j = 0$ $E(a_j) = (1)(6)/14$ $a_j - E(a_j) = -6/14$ $Var(a_j) = (1)(13)(6)(8)/[14^2(13)] = 0.2449$

Time=12 weeks:

	Event	No Event	Total
Standard Trt	0	6	6
New Trt	1	6	7
Total	1	12	13

 $a_i = 0$ $E(a_i) = (1)(6)/13$

 a_{j} - $E(a_{j}) = -6/13$

 $\overline{\text{Var}(a_i)}$ =(1)(12)(6)(7)/[13²(12)] = 0.2485

Time=13 weeks:

· · · · · · · · · · · · · · · · · · ·			
	Event	No Event	Total
Standard Trt	1	5	6
New Trt	0	6	6
Total	1	11	12

 $a_i = 1$ $E(a_i) = (1)(6)/12$

 $a_{j}-E(a_{j})=6/12$

 $\overline{\text{Var}(a_i)}$ =(1)(11)(6)(6)/[12²(11)] = 0.2500

Time=16 weeks:

	Event	No Event	Total
Standard Trt	2	2	4
New Trt	1	3	4
Total	3	5	8

 $a_i = 2$ $E(a_i) = (3)(4)/8$

 a_i -E(a_i) = 4/8 Var(a_i)=(3)(5)(4)(4)/[8²(7)] = 0.5357

Time=17 weeks:

	Event	No Event	Total
Standard Trt	1	1	2
New Trt	1	2	3
Total	2	3	5

 $a_i = 1$ $E(a_i) = (2)(2)/5$

 $\mathbf{a_{i}}\text{-}\mathbf{E}(\mathbf{a_{i}}) = 1/5$

 $Var(a_i)=(2)(3)(2)(3)/[5^2(4)]=0.3600$

Time=18 weeks:

	Event	No Event	Total
Standard Trt	1	0	1
New Trt	1	1	2
Total	2	1	3

 $a_i = 1$ $E(a_i) = (2)(1)/3$

 a_i - $E(a_i) = 1/3$

 $Var(\overline{a_j})=(2)(1)(1)(2)/[3^2(2)]=0.2222$

Time=20 weeks:

	Event	No Event	Total
Standard Trt	0	0	0
New Trt	1	0	1
Total	1	0	1

 $a_j = 0$ $E(a_j) = (1)(0)/1 = 0$

 $\mathbf{a}_{i}-\mathbf{E}(\mathbf{a}_{i})=\mathbf{0}$

 $\overline{\text{Var}(a_i)}$ =(1)(0)(0)(1)/[1²(0)] = undefined

Thus, we have:

$$\chi^{2}_{LR} = \frac{\left[\sum_{j} (a_{j} - E(a_{j}))\right]^{2}}{\sum_{j} Var(a_{j})}$$

$$= \frac{\left[(13/25) - (11/24) + (2/22) + (22/20) - (3/18) - (6/14) - (6/13) + (6/12) + (4/8) + (1/5) + (1/3)\right]^{2}}{(.2496 + .2483 + .4723 + .4689 + .6291 + .2449 + .2485 + .2500 + .5357 + .3600 + .2222)} = 0.7609$$

where p > 0.05 and we would fail to reject the null hypothesis of equal survival in both groups

(e.g. Ho: overall survival in the std treatment = overall survival in the new treatment).

Compare your calculation to that obtained by Stata above in step 8.