

Biost 518: Applied Biostatistics II
 #### Emerson, Winter 2008

Homework #3 Key
 #### Annotated Stata Log File
 #### February 9, 2008

The following output was used to generate the numbers that I wanted to present
 #### in tables, as well as the plots I wanted to present as figures in the paper.
 #### I note that Stata does not present its output in a form suitable for presentation.
 #### Numbers need to be rounded to an interesting number of significant digits, and
 #### the columns and rows need to facilitate comparison of relevant measures.
 #### I used Excel to manipulate this output into the form I wanted, then copied the
 #### resulting tables to the MS-Word document.

Comments edited into the log file produced by Stata are
 #### on the lines that start with the four '#' signs and are
 #### printed in italics.

The Stata commands are put in **bold face**.

Stata output is displayed in regular typeface in blue.

Data was read in and processed according to the leukemia.do file.

#####
 #### Problem 1: Analysis of induction of complete remission by treatment and sex
 #########
 #####

Problem 1a: Descriptive statistics

```
. tabulate male cr, row
| Key          |
| frequency    |
| row percentage |
|              cr |
| male |       0      1     Total |
| 0 |     14      51     65 |
|   | 21.54    78.46 100.00 |
| 1 |     27      38     65 |
|   | 41.54    58.46 100.00 |
| Total |     41      89    130 |
|       | 31.54    68.46 100.00 |
```

```
. bysort tx: tabstat cr, stat(n mean) by(male) col(stat)
-> tx = 0
```

Summary for variables: cr
 by categories of: male

male	N	mean
0	30	.7
1	35	.4857143
Total	65	.5846154

```
-> tx = 1
```

Summary for variables: cr
 by categories of: male

male	N	mean
0	35	.8571429
1	30	.7
Total	65	.7846154

Problem 1b: Comparing the incidence of complete remission by treatment group
 #### I choose to focus in the difference in probability of CR

```
. cs cr tx
```

	tx		Total
	Exposed	Unexposed	
Cases	51	38	89
Noncases	14	27	41
Total	65	65	130
Risk	.7846154	.5846154	.6846154
	Point estimate		[95% Conf. Interval]
Risk difference	.2	.0439899	.3560101
Risk ratio	1.342105	1.054389	1.708332
Attr. frac. ex.	.254902	.0515836	.4146336
Attr. frac. pop	.1460674		
	chi2(1) = 6.02		Pr>chi2 = 0.0141

Problem 1c: Assessing the possibility of confounding by sex
 #### Association between sex and treatment can be assessed from part 1a
 #### Now looking at an association between CR and sex in the control group

```
. cs cr male if tx==0
```

	male		Total
	Exposed	Unexposed	

Cases	17	21		38
Noncases	18	9		27
Total	35	30		65
Risk	.4857143	.7		.5846154
	Point estimate		[95% Conf. Interval]	
Risk difference	-.2142857		-.4473239 .0187525	
Risk ratio	.6938776		.4588232 1.04935	
Prev. frac. ex.	.3061224		-.0493498 .5411768	
Prev. frac. pop	.1648352			
	chi2(1) =	3.05	Pr>chi2 =	0.0805

Problem 1d: Comparing the incidence of complete remission by treatment group
after adjustment for sex using the Mantel-Haenszel statistic

. cs cr tx, by(male) or				
male	OR	[95% Conf. Interval]	M-H Weight	
0	2.571429	.7807859 8.404801	1.615385	(Cornfield)
1	2.470588	.8981522 6.780976	2.353846	(Cornfield)
Crude	2.588346	1.20652 5.545979		
M-H combined	2.511628	1.144075 5.513866		

Test of homogeneity (M-H) chi2(1) = 0.002 Pr>chi2 = 0.9609

Test that combined OR = 1:
 Mantel-Haenszel chi2(1) = 5.33
 Pr>chi2 = 0.0210

Problem 1e: Comparing the incidence of complete remission by treatment group
after adjustment for sex using logistic regression. I use the robust SE to
ensure that any lack of model fit does not lead to anti-conservative inference
(I could not be sure that the lack of model fit would necessarily have lead
to anti-conservative inference—it could have been conservative.)

. logistic cr tx male, robust					
Logistic regression					
					Number of obs = 130
					Wald chi2(2) = 10.41
					Prob > chi2 = 0.0055
					Pseudo R2 = 0.0713
Log pseudolikelihood = -75.252997					
Robust					
cr	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]
tx	2.511585	1.011183	2.29	0.022	1.140901 5.529019
male	.3981549	.1603001	-2.29	0.022	.1808639 .8765004

```
#### Problem 1g: Assessing the possibility of an interaction between treatment and sex
#### This could have been assessed using the output for part 1d. Here I use logistic regression
#### I do not use the robust standard error estimates in this "saturated" model in which
#### each of the four treatment-sex groups is estimated independently (there are four
#### parameters in the model)
```

```

. g txm= tx * male

. logistic cr tx male txm
Logistic regression                                         Number of obs = 130
                                                               LR chi2(3) = 11.57
                                                               Prob > chi2 = 0.0090
Log likelihood = -75.251793                                Pseudo R2 = 0.0714

```

	cr	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]
tx	2.571429	1.610099	1.51	0.131	.7536936	8.77312
male	.4047619	.211528	-1.73	0.084	.1453326	1.127292
txm	.9607843	.7835976	-0.05	0.961	.1942683	4.75171

```
#####
# ### Problem 2: Analysis of time to death by treatment and sex
#####
```

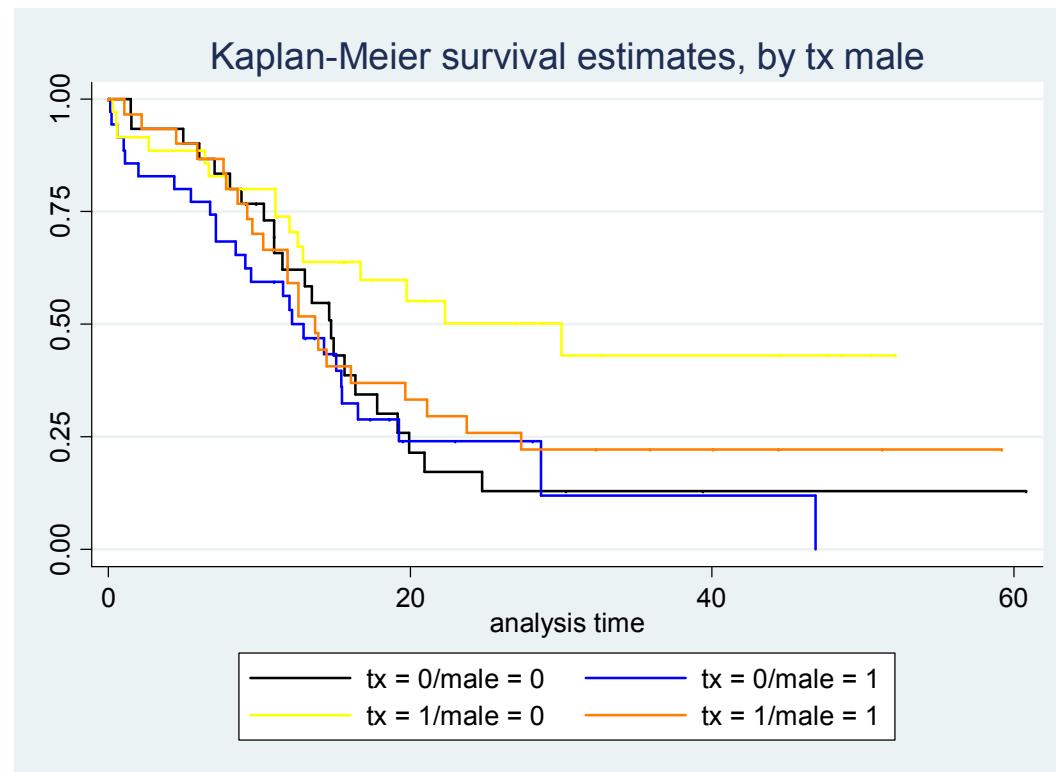
```
#### Problem 2a: Descriptive statistics  
#### I prefer to measure time to death in months, rather than days
```

```
. replace ttodeath= ttodeath/3  
(130 real changes made)
```

```
. stset ttodeath death  
    failure event: death != 0 & death < .  
obs. time interval: (0, ttodeath]  
exit on or before: failure
```

130 total obs.
0 exclusions
130 obs. remaining, representing
87 failures in single record/single failure data
64669 total analysis time at risk, at risk from t = 0
earliest observed entry t = 0
last observed exit t = 1848

```
. sts graph , by(tx male) col(black blue yellow orange)
    failure _d: death
    analysis time _t: ttodeath
```



```
. sts list, by(tx male) at(12 24)
    failure _d: death
    analysis time _t: ttodeath
        Beg.          Survivor      Std.
        Time       Total      Fail   Function     Error   [95% Conf. Int.]

```

	Time	Total	Fail	Beg.	Survivor	Std.
				Function	Error	[95% Conf. Int.]
tx=0 male=0	12	18	11	0.6206	0.0907	0.4192 0.7697
	24	5	11	0.1721	0.0767	0.0555 0.3427
tx=0 male=1	12	18	16	0.5317	0.0859	0.3526 0.6820
	24	4	8	0.2406	0.0817	0.1030 0.4095

<code>tx=1 male=0</code>						
12	23	9	0.7385	0.0751	0.5570	0.8546
24	11	6	0.5017	0.0968	0.3033	0.6712
<code>tx=1 male=1</code>						
12	17	12	0.5911	0.0913	0.3925	0.7441
24	8	9	0.2586	0.0835	0.1154	0.4287

Note: Survivor function is calculated over full data and evaluated at indicated times; it is not calculated from aggregates shown at left.

Problem 2b: Analysis of a treatment effect using proportional hazards regression.
I use robust SE to relax assumptions about proportional hazards

<code>. stcox tx, robust</code>						
failure _d: death						
analysis time _t: ttodeath						
Iteration 0: log pseudolikelihood = -368.044						
Iteration 1: log pseudolikelihood = -365.41537						
Iteration 2: log pseudolikelihood = -365.41523						
Refining estimates:						
Iteration 0: log pseudolikelihood = -365.41523						
 Cox regression -- Breslow method for ties						
No. of subjects	=	130	Number of obs	=	130	
No. of failures	=	87				
Time at risk	=	2127.269739				
			Wald chi2(1)	=	5.29	
<u>Log pseudolikelihood</u>	=	-365.41523	Prob > chi2	=	0.0214	
		Robust				
	t Haz. Ratio	Std. Err.	z	P> z	[95% Conf. Interval]	
	tx .60737	.1316259	-2.30	0.021	.3971784	.9287977

Problem 2c: Assessing the possibility of confounding by sex
Relationships between sex and treatment were examined in problem 1a. Here I
consider the existence of an association between time to death and sex independent
of treatment by only using the control group.

<code>. stcox male if tx==0, robust</code>						
failure _d: death						
analysis time _t: ttodeath						
Iteration 0: log pseudolikelihood = -166.9936						

```

Iteration 1: log pseudolikelihood = -166.84092
Iteration 2: log pseudolikelihood = -166.84092
Refining estimates:
Iteration 0: log pseudolikelihood = -166.84092
Cox regression -- Breslow method for ties
No. of subjects      =          65           Number of obs     =          65
No. of failures       =          49
Time at risk          = 899.9342122
Wald chi2(1)        =         0.32
Log pseudolikelihood = -166.84092
Prob > chi2          = 0.5702
| Robust
t | Haz. Ratio Std. Err.      z   P>|z| [95% Conf. Interval]
male | 1.171496 .3265651    0.57  0.570   .6783604  2.023116

```

Problem 2d: Analysis of a treatment effect after adjustment for sex
 #### using proportional hazards regression.
 #### I use robust SE to relax assumptions about proportional hazards

```

. stcox tx male, robust
failure _d: death
analysis time _t: ttodeath
Iteration 0: log pseudolikelihood = -368.044
Iteration 1: log pseudolikelihood = -364.12782
Iteration 2: log pseudolikelihood = -364.12768
Refining estimates:
Iteration 0: log pseudolikelihood = -364.12768
Cox regression -- Breslow method for ties
No. of subjects      =          130           Number of obs     =          130
No. of failures       =          87
Time at risk          = 2127.269739
Wald chi2(2)        =         7.60
Log pseudolikelihood = -364.12768
Prob > chi2          = 0.0223
| Robust
t | Haz. Ratio Std. Err.      z   P>|z| [95% Conf. Interval]
tx | .6182218 .1335386   -2.23  0.026   .4048378  .9440773
male | 1.413015 .2972737    1.64  0.100   .9355541  2.134149

```

Problem 2f: Investigating the possibility of an interaction between treatment and sex.
 #### I use robust SE to relax assumptions about proportional hazards

```
. stcox tx male txm, robust
      failure _d: death
      analysis time _t: ttodeath
Iteration 0:  log pseudolikelihood = -368.044
Iteration 1:  log pseudolikelihood = -363.73989
Iteration 2:  log pseudolikelihood = -363.65662
Iteration 3:  log pseudolikelihood = -363.65657
Refining estimates:
Iteration 0:  log pseudolikelihood = -363.65657
Cox regression -- Breslow method for ties
No. of subjects      =          130                      Number of obs     =          130
No. of failures       =           87
Time at risk          =    2127.269739
Log pseudolikelihood = -363.65657
                                         Wald chi2(3)      =        7.45
                                         Prob > chi2     =     0.0588
                                         Robust
   t | Haz. Ratio  Std. Err.      z      P>|z|  [95% Conf. Interval]
   tx | .4898034  .1594606    -2.19    0.028  .2587643  .9271271
   male | 1.174754  .3115312     0.61    0.544  .6985833  1.975494
   txm | 1.525956  .6617335     0.97    0.330  .6522543  3.569991
```