

## SISCER Survival Analysis

### Problem Set 1 Answer Key

#### 1. (Life table analysis)

##### (a) Three approaches

- Approach 1

$t$	$n$	$d$	$w$	$q^r = d/n$	$p^r = 1 - q^r$	$\hat{S}^r = \prod p^r$
0-1	146	27	3	0.185	0.815	0.815
1-2	116	18	10	0.155	0.845	0.689
2-3	88	21	10	0.239	0.761	0.524
3-4	57	9	3	0.158	0.842	0.441
4-5	45	1	3	0.022	0.972	0.432
...						

- Approach 2

$t$	$n$	$d$	$w$	$q^l = d/(n - w)$	$p^l = 1 - q^l$	$\hat{S}^l = \prod p^l$
0-1	146	27	3	0.189	0.811	0.811
1-2	116	18	10	0.170	0.830	0.673
2-3	88	21	10	0.269	0.731	0.492
3-4	57	9	3	0.167	0.833	0.410
4-5	45	1	3	0.024	0.977	0.400
...						

- Approach 3

$t$	$n$	$d$	$w$	$q = d/(n - w/2)$	$p = 1 - q$	$\hat{S} = \prod p$
0-1	146	27	3	0.187	0.813	0.813
1-2	116	18	10	0.162	0.838	0.681
2-3	88	21	10	0.253	0.747	0.509
3-4	57	9	3	0.162	0.838	0.426
4-5	45	1	3	0.023	0.977	0.417
...						

##### (b) Use Greenwoods formula

$$se \left\{ \hat{S}(t) \right\} = \hat{S}(t) \left\{ \sum_{j=1}^t \frac{d_j}{(n_j - w_j/2)(n_j - d_j - w_j/2)} \right\}^{1/2}$$

and 95% confidence intervals for  $S(t)$  is  $\hat{S}(t) \pm 1.96 \times se[\hat{S}(t)]$ .

2. Exponential distributions.

- (a) Straightforward
- (b) Likelihood function

$$L(\theta) = \prod_{i=1}^n f(X_i; \theta)^{\Delta_i} S(X_i; \theta)^{1-\Delta_i}$$

and the MLE can be solved by

$$\hat{\theta} = \arg \max_{\theta \in \Theta} L(\theta) = \arg \max_{\theta \in \Theta} [\log L(\theta)] = \frac{\sum_i \Delta_i}{\sum_i X_i}.$$

It is the person-year analysis estimate. A variance estimate is

$$\hat{\text{var}}(\hat{\theta}) = \left[ -\frac{\partial^2 \log L(\hat{\theta})}{\partial \theta^2} \right]^{-1} = \frac{\hat{\theta}^2}{\sum_i \Delta_i}$$

Key assumptions are needed: (1) subjects are independent, (2) non-informative censoring.

3. Use computer to do this exercise

## **Supplement for Solution of Problem Set 1**

Question 1-(b)

95% CI calculation for Approach 3:

Years	$S(t)$	$V_j$	$\sum_{i=1}^j V_j$	$\left(\sum_{i=1}^j V_j\right)^{1/2}$	$S(t) \cdot \left(\sum_{i=1}^j V_j\right)^{1/2}$	95%CI	
						LB	UB
0-1	0.813	0.0016	0.0016	0.040	0.032	0.750	0.877
1-2	0.681	0.0017	0.0033	0.058	0.039	0.604	0.758
2-3	0.509	0.0041	0.0074	0.086	0.044	0.423	0.595
3-4	0.426	0.0035	0.0109	0.104	0.045	0.339	0.514
4-5	0.417	0.0005	0.0114	0.107	0.045	0.329	0.504
5-6	0.393	0.0017	0.0131	0.115	0.045	0.305	0.481
6-7	0.347	0.0052	0.0184	0.135	0.047	0.255	0.439
7-8	0.325	0.0042	0.0225	0.150	0.049	0.230	0.421
8-9	0.263	0.0224	0.0449	0.212	0.056	0.154	0.373
9-10	0.158	0.1333	0.1783	0.422	0.067	0.027	0.289

$$* V_j = \frac{d_j}{\left( n_j - \frac{w_j}{2} \right) \left( n_j - d_j - \frac{w_j}{2} \right)}$$



Three survival estimates are quite similar, however, the first approach produces higher estimates, while the second approach leads to lower estimates which is more conservative. The 95% CIs of the third one cover other two estimates, this approach is widely used in epidemiological studies with right censored data (e.g., in a longitudinal study, when failure event occurred in the interval between two follow-up visits).

### Question 2-(a)

$\because T \sim \exp(\theta)$

$\therefore$  the pdf:  $f(t; \theta) = \theta \cdot e^{-\theta t}$  ( $t > 0$ ) (see lecture -1, last two slides)

$\because \lambda(t) = \theta$

$$\text{and } \lambda(t) = \frac{f(t)}{S(t)}$$

$$\therefore \text{the survival function: } S(t) = \frac{f(t)}{\lambda(t)} = \frac{\theta \cdot e^{-\theta t}}{\theta} = e^{-\theta t}$$

$$\text{median of } T : S(t) = e^{-\theta t} = 0.5 \Rightarrow t = \frac{\log 0.5}{-\theta} = \frac{\log 2}{\theta}$$

### Question 2-(b)

(1) the likelihood function:

$$\begin{aligned} L(\theta) &= \prod_{i=1}^n f(X_i; \theta)^{\delta_i} S(X_i; \theta)^{1-\delta_i} \\ &= \prod_{i=1}^n \frac{f(X_i; \theta)^{\delta_i}}{S(X_i; \theta)^{\delta_i}} S(X_i; \theta) \\ &= \prod_{i=1}^n \lambda(X_i; \theta)^{\delta_i} S(X_i; \theta) \\ &= \prod_{i=1}^n \theta^{\delta_i} \cdot e^{-\theta X_i} \end{aligned}$$

(2) the log-likelihood:

$$\begin{aligned} l(\theta) &= \log \left[ \prod_{i=1}^n \theta^{\delta_i} \cdot e^{-\theta X_i} \right] \\ &= \sum_{i=1}^n [\log \theta^{\delta_i} + \log e^{-\theta X_i}] \\ &= \sum_{i=1}^n [\delta_i \log \theta - \theta X_i] \end{aligned}$$

(3) take first derivative:

$$\begin{aligned}\frac{\partial}{\partial \theta} l(\theta) &= \sum_{i=1}^n \left[ (\delta_i \log \theta)' - (\theta X_i)' \right] \\ &= \sum_{i=1}^n \left[ \delta_i \cdot \frac{1}{\theta} - X_i \right] \\ &= \frac{\sum_{i=1}^n \delta_i}{\theta} - \sum_{i=1}^n X_i\end{aligned}$$

(4) set  $\frac{\partial}{\partial \theta} l(\theta) = 0$ , solve for MLE of  $\theta$ :

$$\frac{\sum_{i=1}^n \delta_i}{\theta} - \sum_{i=1}^n X_i = 0 \Rightarrow \frac{\sum_{i=1}^n \delta_i}{\theta} = \sum_{i=1}^n X_i \Rightarrow \hat{\theta}_{MLE} = \frac{\sum_{i=1}^n \delta_i}{\sum_{i=1}^n X_i}$$

(5) Fisher information (see lecture 1, the last two slides):

$$\begin{aligned}I(\theta) &= E \left[ -\frac{\partial^2}{\partial \theta^2} \log L(\theta) \right] = E \left[ -\frac{\partial}{\partial \theta} \left( \frac{\sum_{i=1}^n \delta_i}{\theta} - \sum_{i=1}^n X_i \right) \right] \\ &= E \left[ - \left( \frac{\sum_{i=1}^n \delta_i}{\theta} \right)' - \left( \sum_{i=1}^n X_i \right)' \right] = E \left[ - \left( -\frac{\sum_{i=1}^n \delta_i}{\theta^2} \right) - 0 \right] = E \left[ \frac{\sum_{i=1}^n \delta_i}{\theta^2} \right] = \frac{\sum_{i=1}^n \delta_i}{\hat{\theta}_{MLE}^2}\end{aligned}$$

$$(6) \text{Var}(\hat{\theta}) = \frac{1}{I(\theta)} = \frac{\hat{\theta}_{MLE}^2}{\sum_{i=1}^n \delta_i}$$

Key assumption:

- (1) subjects are independent;
- (2) non-informative censoring (ie, failure time is independent of censoring time).

### **Question 2- (c).**

```
. gen z01=z1  
. replace z01=0 if z1==2  
. streg z01, d(exponential) nohr
```

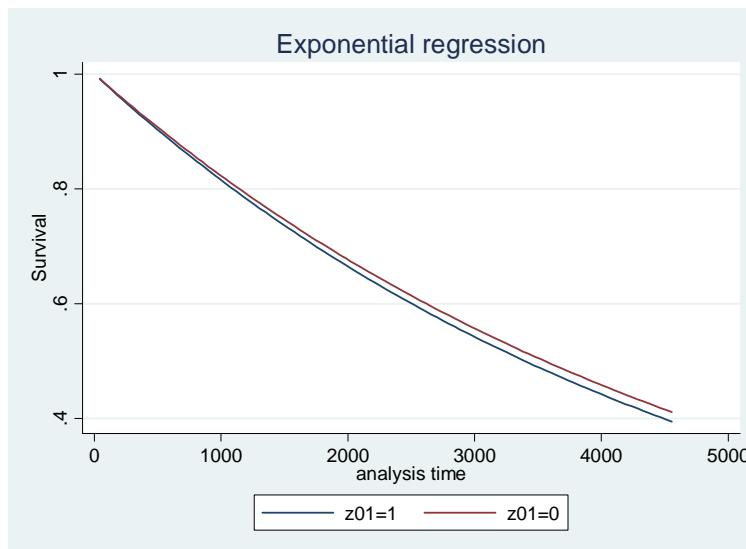
<u>_t</u>	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
z01	.0450511	.1790287	0.25	0.801	-.3058388 .3959409
_cons	-8.541941	.1290994	-66.17	0.000	-8.794971 -8.288911

```
. * hazard rate for two treatment groups  
. predictnl haz = predict(hazard), ci(haz_lb haz_ub)  
note: Confidence intervals calculated using Z critical values
```

```
. list _t haz haz_lb haz_ub z1 in 1/10
```

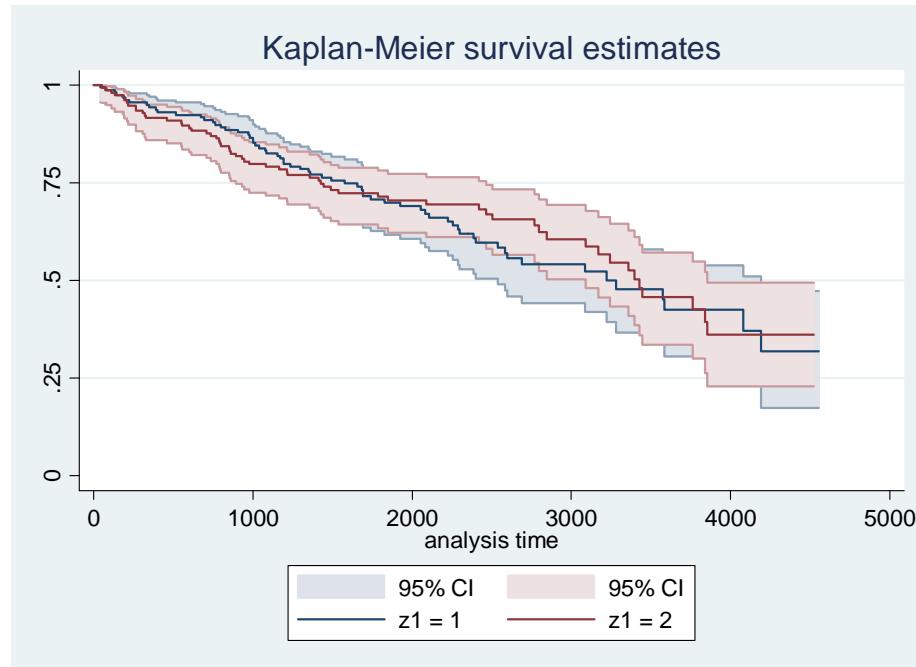
	_t	haz	haz_lb	haz_ub	z1
<hr/>					
1.	400	.0002041	.0001545	.0002537	1
2.	4500	.0002041	.0001545	.0002537	1
3.	1012	.0002041	.0001545	.0002537	1
4.	1925	.0002041	.0001545	.0002537	1
5.	1504	.0001951	.0001457	.0002445	2
<hr/>					
6.	2503	.0001951	.0001457	.0002445	2
7.	1832	.0001951	.0001457	.0002445	2
8.	2466	.0001951	.0001457	.0002445	2
9.	2400	.0002041	.0001545	.0002537	1
10.	51	.0001951	.0001457	.0002445	2

```
. * plot survival functions  
. stcurve, survival atl(z01=1) at2(z01=0)
```



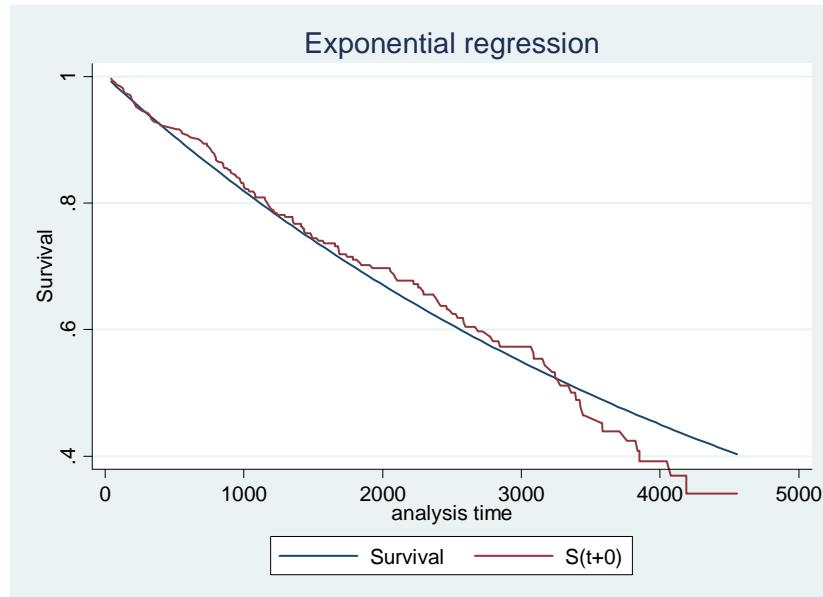
### Question 3-a

```
. sts graph, by (z1) gwood
```



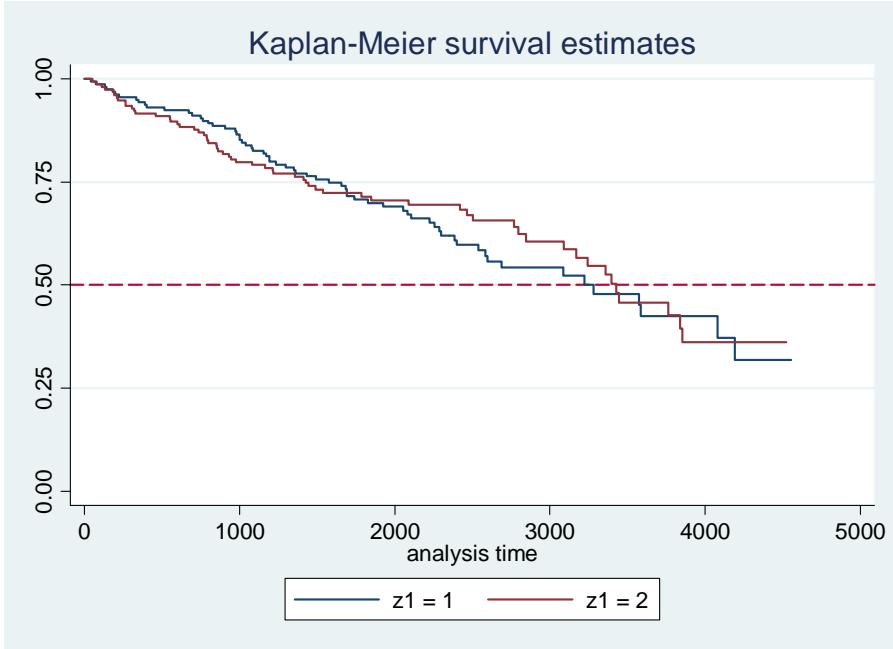
### Plot of exponential survival curve versus Kaplan Meier survival curve

```
* generate Kaplan Meier survival estimate using  
. sts gen st= s  
* plot exponential survival curve and Kaplan Meier survival curve in one graph  
. stcurve, survival addplot(line st _t, sort)
```



### Question 3-b

```
. sts graph, by(z1) yline(0.5, lpattern(dash))


```

```
. stci, by(z1)

      |      no. of
z1      |      subjects      50%      Std. Err.      [ 95% Conf. Interval]
-----+-----+-----+-----+-----+-----+
      1 |      158      3282      272.7255      2540      4191
      2 |      154      3428      174.1291      3090      3853
-----+-----+-----+-----+-----+
total |      312      3395      151.7444      3086      3839
```

```
. * estimate median time based on exponential regression model
. predict mt, median time
. list mt z1 in 1/10

+-----+
|      mt      z1 |
|-----+
1. | 3396.08      1 |
2. | 3396.08      1 |
3. | 3396.08      1 |
4. | 3396.08      1 |
5. | 3552.576     2 |
|-----+
6. | 3552.576     2 |
7. | 3552.576     2 |
8. | 3552.576     2 |
9. | 3396.08      1 |
10. | 3552.576     2 |
+-----+
```