

## Example: Remission Duration

```
. stcox sex, nohr efron
```

```
      failure _d: status  
analysis time _t: time
```

```
Cox regression -- Efron method for ties
```

```
No. of subjects =          42          Number of obs   =          42  
No. of failures =          30  
Time at risk   =          541  
  
Log likelihood = -92.885605          LR chi2(1)        =          0.60  
                                          Prob > chi2      =          0.4396
```

```
-----  
      _t |      Coef.   Std. Err.      z    P>|z|     [95% Conf. Interval]  
-----+-----  
      sex |  -.3101396   .4042745    -0.77   0.443   -1.102503   .4822238  
-----
```

```
. estat phtest
```

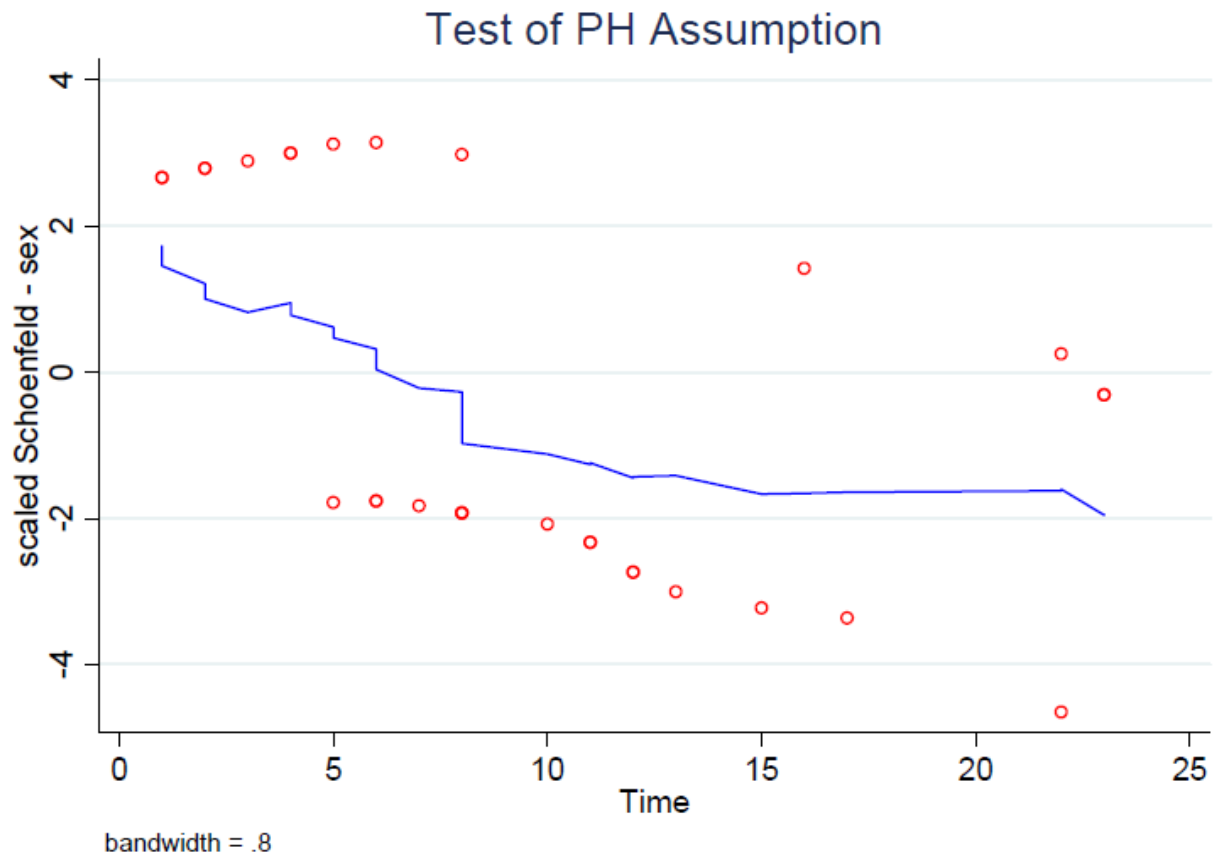
```
Test of proportional-hazards assumption
```

```
Time: Time
```

```
-----  
      |                      chi2      df      Prob>chi2  
-----+-----  
global test |                      10.90      1      0.0010  
-----
```

## Example: Remission Duration

```
. estat phtest, plot(sex)
```



## Example: Remission Duration

```
. stcox tx logwbc3 sex, nohr efron
```

```
      failure _d:  status  
analysis time _t:  time
```

```
Cox regression -- Efron method for ties
```

```
No. of subjects =          42          Number of obs =          42  
No. of failures =          30  
Time at risk   =          541  
  
Log likelihood = -69.590483          LR chi2(3)      =          47.19  
                                          Prob > chi2   =          0.0000
```

```
-----+-----  
      _t |      Coef.   Std. Err.      z    P>|z|     [95% Conf. Interval]  
-----+-----  
      tx | -1.503591   .4615127    -3.26  0.001   -2.408139   -.5990429  
logwbc3 |  1.681942   .3365836     5.00  0.000    1.022251    2.341634  
      sex |   .314678   .4545115     0.69  0.489   -.5761482    1.205504  
-----+-----
```

```
. estat phtest, detail
```

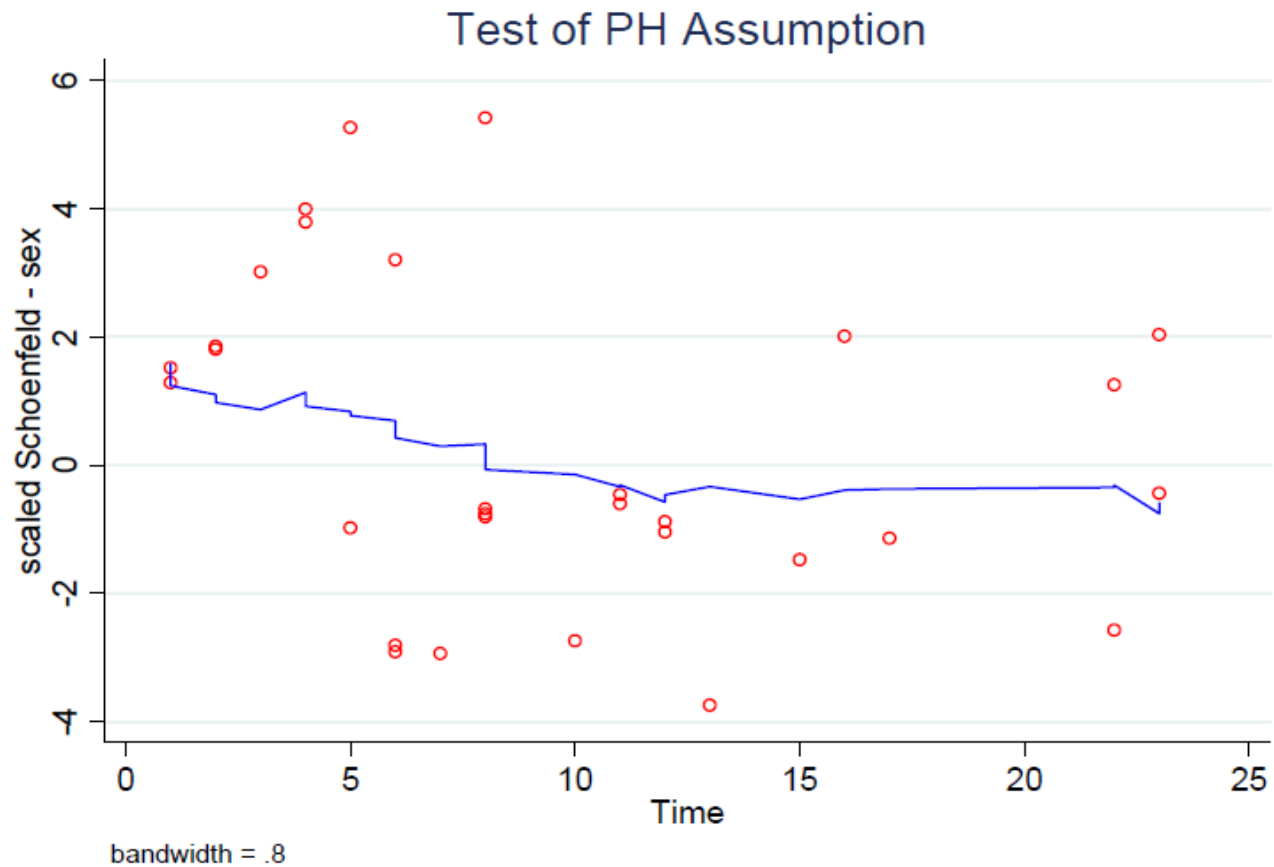
```
Test of proportional-hazards assumption
```

```
Time:  Time
```

```
-----+-----  
      |      rho      chi2      df      Prob>chi2  
-----+-----  
tx      |      0.05700      0.11      1      0.7425  
logwbc3 |      0.06864      0.21      1      0.6438  
sex     |     -0.29209      2.56      1      0.1094  
-----+-----  
global test |      2.82      3      0.4206  
-----+-----
```

## Example: Remission Duration

```
. estat phtest, plot(sex)
```



## Example: Remission Duration

Is the relationship with logwbc linear?

```
. gen lwbc2=logwbc3^2
```

```
. stcox group logwbc3 lwbc2, nohr
```

```
LR chi2(3)      =      44.50
Log likelihood  =     -71.73582          Prob > chi2      =      0.0000
```

_t	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
group	-1.366605	.4303963	-3.18	0.001	-2.210167	-.5230442
logwbc3	1.510339	.3221063	4.69	0.000	.8790224	2.141656
lwbc2	.2710913	.2558792	1.06	0.289	-.2304227	.7726052

## Model Checking Summary

### Cox PH Model

<u>Assumption</u>	<u>Technique</u>	<u>Stata</u>	<u>Remarks</u>
PH	-log-log plot	stphplot	graphical, cat. covars only
	obs v pred KM	stcoxkm	graphical, cat. covars only
	$H_0: \beta(t) = \beta$	estat phtest	test each cov., any kind of cov.

## What to do if PH fails?

### **Time dependent covariates:**

- Interaction between covariate and  $t$  or  $\log t$
- smoothed estimates of  $\beta(t)$
- See Biost 537 and other advanced courses

### **Stratification:**

- Discrete (grouped) covariates only
- Separate baseline hazard for each covariate class
- graphical analysis – no quantitative estimate of HR

## Stratified Cox Model

Remission duration study also includes information on gender

```
. tab sex
```

sex	Freq.	Percent	Cum.
female	22	52.38	52.38
male	20	47.62	100.00
Total	42	100.00	

```
. stcox group logwbc3 sex, nohr efron scaledsch(SCA*)
```

```
LR chi2(3) = 47.19
Log likelihood = -69.590483 Prob > chi2 = 0.0000
```

_t	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
group	-1.503591	.4615127	-3.26	0.001	-2.408139 - .5990429
logwbc3	1.681942	.3365836	5.00	0.000	1.022251 2.341634
sex	.314678	.4545115	0.69	0.489	-.5761482 1.205504

```
. estat phtest, detail log
```

Test of proportional-hazards assumption

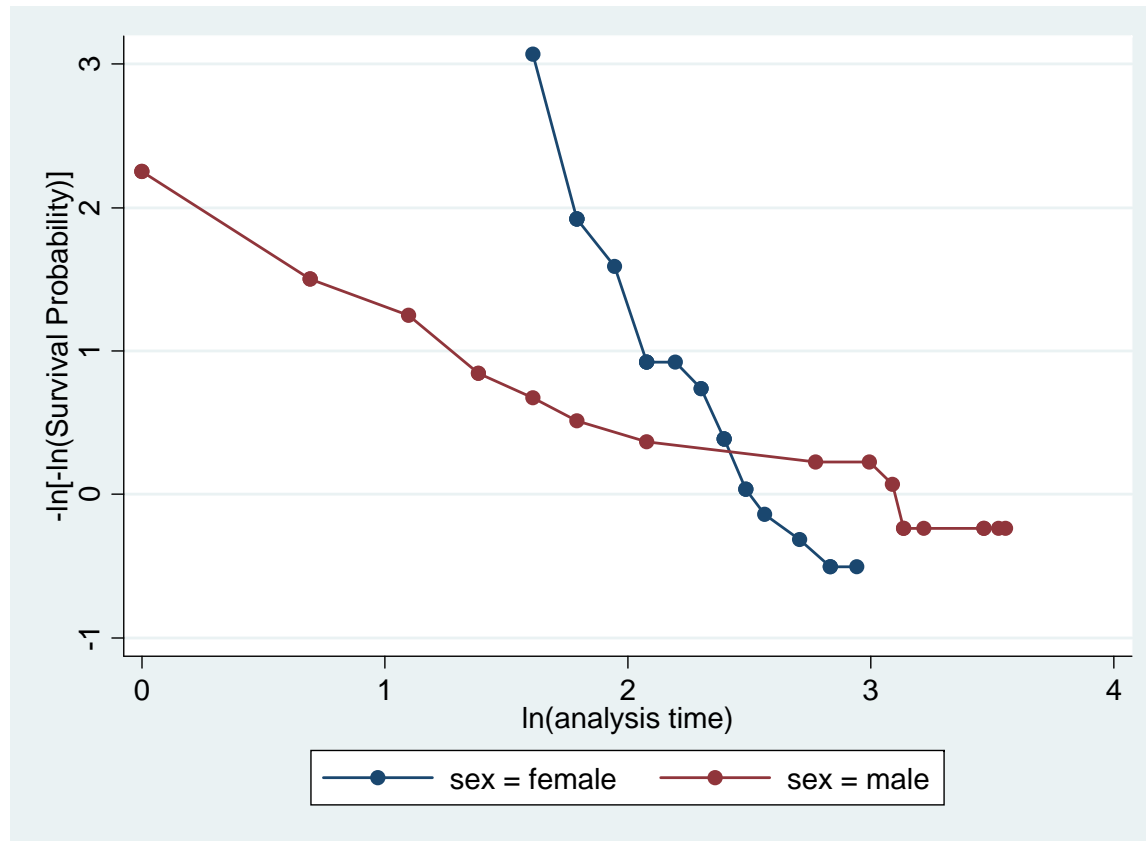
Time: Log(t)

	rho	chi2	df	Prob>chi2
group	0.13307	0.59	1	0.4431
logwbc3	0.08070	0.30	1	0.5867
sex	-0.36205	3.94	1	0.0472



## Stratified Cox Model

```
. stphtplot, by(sex)
```



## Stratified Cox Model

**Q:** Evidence is against PH assumption for sex (though less strong in multivariate model). If we don't accept PH assumption on *sex*, can we still make PH inference about *group* and *logwbc* even if *sex* does not satisfy the PH assumption?

**A:** Yes. In order to do this we can perform a “stratified” analysis. This is different from using dummy variables, and is different from using separate analyses by gender.

## Stratified Cox Model

### Idea:

We can use a model where, within each gender, we have the same PH model, but we allow men and women to have *different baseline hazards*:

$$\text{women : } h(t, X) = h_{0,0}(t) \exp(\beta_1 \cdot \text{group} + \beta_2 \cdot \text{logwbc3})$$

$$\text{men : } h(t, X) = h_{0,1}(t) \exp(\beta_1 \cdot \text{group} + \beta_2 \cdot \text{logwbc3})$$

Such a model is said to be “stratified on sex”. It is analogous to allowing interactions between sex and time, but we still have additive terms for the other covariates (no dependence on time).

## Stratified Cox Model

### **Proportional Hazards Model:**

$$\log[h(t, X)] = \log[h_0(t)] + \beta_1 \text{group} + \beta_2 \log \text{wbc3} + \beta_3 \text{sex}$$

### **Stratified Cox Model:**

$$\log[h(t, X)] = \log[h_{0,sex}(t)] + \beta_1 \text{group} + \beta_2 \log \text{wbc3}$$

**Q:** What's the interpretation of  $\beta_1$  in each model?

## Stratified Cox Model

### **Proportional Hazards:**

$$F : h(t, X) = h_0(t) \exp(\beta_1 \text{group} + \beta_2 \log\text{wbc3})$$

$$M : h(t, X) = h_0(t) \exp(\beta_1 \text{group} + \beta_2 \log\text{wbc3} + \beta_3)$$

### **Separate Models:**

$$F : h(t, X) = h_{0,1}(t) \exp(\beta_1^{(1)} \text{group} + \beta_2^{(1)} \log\text{wbc3})$$

$$M : h(t, X) = h_{0,2}(t) \exp(\beta_1^{(2)} \text{group} + \beta_2^{(2)} \log\text{wbc3})$$

### **Stratified Model #1:**

$$F : h(t, X) = h_{0,F}(t) \exp(\beta_1 \text{group} + \beta_2 \log\text{wbc3})$$

$$M : h(t, X) = h_{0,M}(t) \exp(\beta_1 \text{group} + \beta_2 \log\text{wbc3})$$

## Stratified Cox Model

### Stratified Model #2:

$$h(t, X) = h_{0,sex}(t) \exp(\beta_1 \text{group} + \beta_2 \log \text{wbc3} + \beta_3 \text{group} \times \text{sex})$$

$$F : h(t, X) = h_{0,F}(t) \exp(\beta_1 \text{group} + \beta_2 \log \text{wbc3})$$

$$M : h(t, X) = h_{0,M}(t) \exp(\beta_1 \text{group} + \beta_2 \log \text{wbc3} + \beta_3 \text{group})$$

## Example: Remission Duration

### Proportional hazards model

```
. stcox group logwbc3 sex, efron nohr
```

```
No. of subjects =          42                Number of obs =          42
No. of failures =          30
Time at risk    =          541
Log likelihood  = -69.590483                LR chi2(3)      =          47.19
                                                Prob > chi2    =          0.0000
```

```
-----+-----
      _t |          Coef.   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
      group | -1.503591   .4615127    -3.26   0.001   -2.408139   -.5990429
    logwbc3 |  1.681942   .3365836     5.00   0.000    1.022251    2.341634
         sex |  .314678   .4545115     0.69   0.489   -1.5761482   1.205504
-----+-----
```

## Example: Remission Duration

### Separate Models

#### Females

```
. stcox group logwbc3 if sex==0, nohr efron
```

```
Log likelihood = -33.090979          LR chi2(2) = 6.65
                                Prob > chi2 = 0.0361
```

_t	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
group	-.3112706	.5635539	-0.55	0.581	-1.415816	.7932747
logwbc3	1.206146	.5034893	2.40	0.017	.2193255	2.192967

#### Males

```
. stcox group logwbc3 if sex==1, nohr efron
```

```
Log likelihood = -20.760908          LR chi2(2) = 29.18
                                Prob > chi2 = 0.0000
```

_t	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
group	-1.977887	.739202	-2.68	0.007	-3.426697	-.5290782
logwbc3	1.742777	.5357723	3.25	0.001	.6926825	2.792871



## Stratified Cox Model – Remission Duration

### Stratified #1

```
. stcox group logwbc3, strata(sex) efron nohr bases(S) basech(H)
                                LR chi2(2) = 32.06
Log likelihood = -55.734815       Prob > chi2 = 0.0000
```

```
-----+-----
      _t |      Coef.   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
      group |  -0.9981037   .4735546   -2.11   0.035   -1.926254   -.0699538
      logwbc3 |  1.453654    .3440687    4.22   0.000    .7792919    2.128017
-----+-----
```

Stratified by sex

```
. estimates store model1
```

### Stratified #2

```
. gen txsex=group*sex
. stcox group logwbc3 txsex, strata(sex) nohr efron
                                LR chi2(3) = 35.28
Log likelihood = -54.126889       Prob > chi2 = 0.0000
```

```
-----+-----
      _t |      Coef.   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
      group |  -0.2865729   .5685327   -0.50   0.614   -1.400877    .8277308
      logwbc3 |  1.472627    .3517843    4.19   0.000    .7831426    2.162112
      txsex |  -1.642102    .9140899   -1.80   0.072   -3.433685    .1494813
-----+-----
```

Stratified by sex

```
. est stor model2
. lrtest model1 model2
```

```
Likelihood-ratio test                                LR chi2(1) = 3.22
(Assumption: model1 nested in model2)                 Prob > chi2 = 0.0729
```

## Time Varying Covariates

- Stratification allows us to make inferences about everything except the stratification variable. In effect, we have adjusted away the effect of sex.
- Suppose we wanted to make inferences on sex. Is it possible to model the time dependent HR?
- Yes! Include an interaction between sex and time.

Note: A different, but related, situation arises when we have a covariate for which we have measured changes over time e.g. sexual behavior, CD4 count

## Time Varying Covariates

Time-independent model:

$$\log h(t; X) = \log h_0(t) + \beta_1 \text{sex}$$

Time-dependent model:

$$\log h(t; X) = \log h_0(t) + \beta_1 \text{sex} + \beta_2 \text{sex} * t$$

(note: no main effect for t – it is part of  $\log h_0(t)$ )

What is the HR for sex?

$$\begin{aligned} \frac{h(t; \text{sex} = 1)}{h(t; \text{sex} = 0)} &= \frac{h_0(t) \exp(\beta_1 + \beta_2 t)}{h_0(t) \exp(0)} \\ &= \exp(\beta_1 + \beta_2 t) \end{aligned}$$

$\beta_2$  positive  $\Rightarrow$  HR increases over time

$\beta_2$  negative  $\Rightarrow$  HR decreases over time

## Stratified Cox Model – Remission Data

```
. stcox tx logwbc sex, nohr efron tvc(sex) texp(_t)
```

```
Cox regression -- Efron method for ties
```

```
No. of subjects =          42          Number of obs   =          42
No. of failures =          30
Time at risk    =          541
Log likelihood   = -67.641012          LR chi2(4)       =          51.09
                                          Prob > chi2     =          0.0000
```

```
-----+-----
      _t |          Coef.   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
main    |
      tx | -1.186389   .4828406    -2.46  0.014   -2.132739   -2.2400388
      logwbc |  1.535217   .3405832     4.51  0.000    .8676864    2.202748
      sex |  1.727716   .8966549     1.93  0.054   -.0296948    3.485128
-----+-----
tvc     |
      sex | -.1655309   .0904991    -1.83  0.067   -.3429058    .011844
-----+-----
```

```
Note: variables in tvc equation interacted with _t
```

## Stratified Cox Model – Remission Data

$$\text{Hr}_{\text{sex}}(t = 5) =$$

```
lincom [main]sex + 5*[tvc]sex, hr
```

_t	Haz. Ratio	Std. Err.	z	P> z	[95% Conf. Interval]	
(1)	2.459755	1.397205	1.58	0.113	.8079549	7.488532

$$\text{Hr}_{\text{sex}}(t = 25) =$$

```
lincom [main]sex + 25*[tvc]sex, hr
```

_t	Haz. Ratio	Std. Err.	z	P> z	[95% Conf. Interval]	
(1)	.0897653	.1417142	-1.53	0.127	.0040673	1.981143

### Overall LR test for sex:

```
stcox tx logwbc, nohr efron  
estimates store model0  
stcox tx logwbc sex, nohr efron tvc(sex)  
estimates store model1
```

```
lrtest model0 model1
```

```
Likelihood-ratio test  
(Assumption: model0 nested in model1)
```

```
LR chi2(2) = 4.37  
Prob > chi2 = 0.1122
```

## Survival Analysis and Sample Size

**Q:** What are the considerations for determining the sample size necessary when the study endpoint is a time-until-event?

### Planned Analysis:

- Assessment of percent surviving beyond  $t^*$ .
  - Need to know  $\alpha$ , power,  $S(t^*;X=0)$ ,  $S(t^*;X=1)$
  - Comparison of proportions (see STATA sampsi!)
- Assessment of survival function and/or hazard ratio.
  - Log-rank / Cox regression.

- Power depends on events only! 
$$L = \left( \frac{Z_{1-\alpha/2} + Z_{1-\beta}}{\log(HR)/2} \right)^2$$
  - $L$  is number of events (i.e. deaths)
  - $HR$  is relative hazard.
- N/arm based on expected event rate, duration of follow-up and  $HR$ .

## Sample Size Example

HPTN 039

Assume:

- 2 treatment arms with  $N$  subjects each
- $h_0 = 0.035/\text{yr}$  with  $HR = 0.5$
- $f = 1$  year follow-up
- $\alpha = 0.05$ , Power =  $1 - \beta = 0.90$

Compute:

- $L = [(1.96+1.28)/(-0.3466)]^2 = \mathbf{87.4}$
- $N = L/[h_0 * f * (1 + HR)] = 87.4/[(0.035)*(1)*(1.5)] = \mathbf{1665}$
- Inflate  $N$  further for expected lost to follow-up
- Compare to 1865/arm (sampsi) for  $p_0 = .035$  to  $p_1 = 0.0175$  at 1 year.
- Less censoring >> greater savings in sample size