

## Cox PH Model

Estimation of survival function

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## Cox PH model

- Recall the Cox's proportional hazard model  

$$h(t | \mathbf{Z}) = h_0(t) \exp(\mathbf{Z}^T \boldsymbol{\beta}).$$
- The coefficients  $\boldsymbol{\beta}$  quantifies the effects of covariates on relative risk.
- The coefficients  $\boldsymbol{\beta}$  can be estimated by maximizing the partial likelihood function.
- How to estimate the baseline?

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## Baseline survival

- In practice, we are also interested in estimating the survival probability for a new patient with a given set of covariates  $S(t | \mathbf{Z})$ .
- Note that in the Cox's model
 
$$\begin{aligned} h(t | \mathbf{Z}) &= h_0(t) \exp(\mathbf{Z}^T \boldsymbol{\beta}) \\ \Rightarrow H(t | \mathbf{Z}) &= H_0(t) \exp(\mathbf{Z}^T \boldsymbol{\beta}) \\ \Rightarrow \exp[-H(t | \mathbf{Z})] &= \exp[-H_0(t) \exp(\mathbf{Z}^T \boldsymbol{\beta})] \\ \Rightarrow S(t | \mathbf{Z}) &= [S_0(t)]^{\exp(\mathbf{Z}^T \boldsymbol{\beta})} \end{aligned}$$
- After  $\boldsymbol{\beta}$  being estimated, we need to estimate the baseline survival  $S_0(t)$ .

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### Recall Kaplan-Meier estimator

- Recall the relationships between  $h(t)$ ,  $H(t)$ , and  $S(t)$  for discrete r.v.s
  - $- h(t_i) = P(T = t_i \mid T \geq t_i)$
  - $- H(t) = \sum_{t_i \leq t} h(t_i)$
  - $- S(t) = \prod_{t_i \leq t} (1 - h(t_i))$

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### Recall Kaplan-Meier estimator

- In the univariate case, individuals all have the same hazard rate  $h(t_i)$  at  $t_i$ .
- For a subject alive just prior to  $t_i$ ,  

$$P(\text{Failure at } t_i \mid \text{alive prior to } t_i) = h(t_i)$$
- Therefore
  - $\frac{d_i}{Y_i} \approx h(t_i) \text{ or } d_i \approx E(d_i \mid R_i) = Y_i h(t_i)$
  - $- \hat{h}(t_i) = \frac{d_i}{Y_i}$
  - $- \hat{H}(t) = \sum_{t_i \leq t} \frac{d_i}{Y_i}$
  - $- \hat{S}(t) = \prod_{t_i \leq t} \left(1 - \frac{d_i}{Y_i}\right)$

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### Cox PH model

- Different hazard for different individuals
  - Let  $d_i$  be the total number of events at  $t_i$
- $$d_i \approx E(d_i \mid R_i) = \sum_{j \in R(t_i)} E(I(\text{subject } j \text{ has event at } t_i) \mid R_i)$$
- $$= \sum_{j \in R(t_i)} P(\text{subject } j \text{ has event at } t_i \mid R_i) = \sum_{j \in R(t_i)} h_0(t_i) \exp(Z_j^T \beta)$$
- $$= h_0(t_i) \sum_{j \in R(t_i)} \exp(Z_j^T \beta)$$
- Therefore estimate  $h_0(t_i)$  by
 
$$\hat{h}_0(t_i) = \frac{d_i}{\sum_{j \in R(t_i)} \exp(Z_j^T \beta)}$$

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### Baseline estimation

- Therefore

$$\hat{H}_0(t) = \sum_{t_i \leq t} \hat{h}_0(t_i) = \sum_{t_i \leq t} \frac{d_i}{\sum_{j \in R(t_i)} \exp(Z_j^T \beta)}$$

- The baseline survival function can be estimated by

$$\hat{S}_0(t) = \prod_{t_i \leq t} (1 - \hat{h}_0(t_i)) = \prod_{t_i \leq t} \left( 1 - \frac{d_i}{\sum_{j \in R(t_i)} \exp(Z_j^T \beta)} \right)$$

- This is referred as Product-Limit method in SAS.

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### Survival estimation

- Therefore the survival function for patients with covariates  $\mathbf{Z}$  can be estimated by

$$\hat{S}(t | \mathbf{Z}) = [\hat{S}_0(t)]^{\exp(\mathbf{Z}^T \beta)}$$

- $\hat{S}_0(t)$  is the estimated baseline survival for individuals with  $\mathbf{Z} = 0$ .
- Sometimes the baseline is meaningless that no individual can have  $\mathbf{Z} = 0$ .
- For example, if  $Z = \text{age}$ , all patients are over 30 years old in the study.
- But can construct covariates so that  $\mathbf{Z} = 0$  is representative by **centering the covariates**.
- Let  $\mathbf{Z}^* = \mathbf{Z} - \bar{\mathbf{Z}}$ , then  $\mathbf{Z}^* = 0$  corresponding to the baseline being "average" individuals.
- A good reason to center the covariates in the analysis.

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### SAS Example

```
data covin;
input stage age id;
cards;
2 50 1
3 40 2
;
```

- Create a data set to estimate survival functions for two potential patients.
- We will estimate the survival functions
 
$$S_1(t | \text{Stage} = 2, \text{age} = 50)$$

$$S_2(t | \text{Stage} = 3, \text{age} = 40)$$

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## SAS Example

- BASELINE Statement in PROC PHREG
  - OUT=SAS-data-set
  - COVARIATES=SAS-data-set
  - TIMELIST=list (eg. timelist=5,20 to 50 by 10)
  - keyword=name (specifies the statistics to be included in the OUT= data set)
  - METHOD(NELSON, PL)

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## SAS Example

```
proc phreg data=larynx plots(overlay)=survival;
class stage;
model stime*censor(0)=stage age/risklimits
/*computes confidence intervals for hazard ratios*/
baseline out=out covariates=covin survival=_all_
/method=PL rowid=id;
hazardratio 'H1'age / units=10 cl=both;
hazardratio 'H2'stage / cl=both;
contrast "age" age 10/estimate=exp;
contrast "Stage 1 vs Stage 2" stage 1 -1 0
/estimate=exp;
contrast "Stage 2 vs Stage 3" stage 0 1 -1
/estimate=exp;
run;
```

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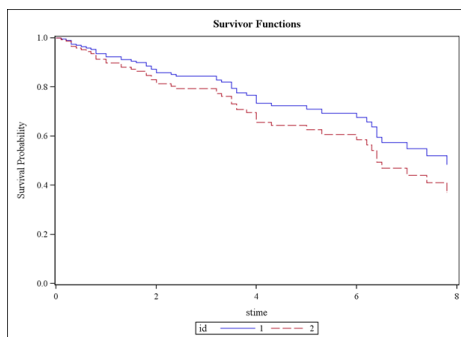
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