

Model 3

$$h(t, X) = h_0(t) \exp \{ \beta_1 R_x + \beta_2 \log WBC + \beta_3 (R_x \times \log WBC) \}$$

For model 3, the ^{estimated} hazard ratio reduces to $\exp \{ \hat{\beta}_1 + \hat{\beta}_3 \log WBC \}$

A 95% CI for $\exp[\hat{l}]$ is gotten by exponentiating the formula for the CI for \hat{l} .

$$\exp \{ \hat{l} \pm 1.96 \sqrt{\text{Var } \hat{l}} \}$$

The general formula is

$$\lambda = \beta_1 + \delta_1 W_1 + \dots + \delta_k W_k$$

where $X = (0, 1)$ exposure variable, $\beta_1 = \text{coeff of } X_1$

$\delta_j = \text{coeff of } X_1 \times W_j, j \in [k]$.

- Because the coefficients in the linear sum are estimated from the same dataset, the coefficients are correlated, so calculation of the variance must consider the (co)variances of the estimated coefficients.
- Computer can handle it.

VII. Adjusted Survival Curves Using the Cox PHL model

If there is no model to fit the survival data, use KM curves.

The Cox model allows for adjusted survival curves

$$h(t, X) = e^{\sum_{i=1}^p \beta_i X_i} \Rightarrow S(t, X) = [S_0(t)] e^{\sum_{i=1}^p \beta_i X_i}$$

$$\text{Estimate by } \hat{S}(t, X) = [\hat{S}_0(t)] e^{\sum_{i=1}^p \hat{\beta}_i X_i}$$

↑
found by computer

General formulas for adjusted survival curves comparing two groups.

$$\hat{S}(t, X_1) = [\hat{S}_0(t)]^{\exp[\hat{\beta}_1(1) + \sum_{i=2}^p \hat{\beta}_i X_i]}$$

$$\hat{S}(t, X_2) = [\hat{S}_0(t)]^{\exp[\hat{\beta}_1(0) + \sum_{i=2}^p \hat{\beta}_i X_i]}$$

Note that we use the mean to adjust for the other covariates.

You can average w/ all variables to get an overall survival curve.

VIII. The Meaning of the PH Assumption

The PH assumption requires that the HR is constant over time

$$\hat{h}(t, X^*) = c \hat{h}(t, X) \quad \forall t$$

$$HR = \frac{\hat{h}(t, X^*)}{\hat{h}(t, X)} = \frac{\hat{h}_0(t) \exp[\sum_{i=1}^p \hat{\beta}_i X_i^*]}{\hat{h}_0(t) \exp[\sum_{i=1}^p \hat{\beta}_i X_i]} = \exp[\hat{\beta} \cdot (X^* - X)]$$

Let $\hat{\theta} = \exp[\hat{\beta} \cdot (X^* - X)]$. $\hat{\theta}$ is constant

$$\Rightarrow c = \hat{\theta}$$

You need PH to be satisfied for Cox to be valid.

Ch 5 & 6 discuss non-proportional hazards.

The most sophisticated approach seems to be to use the extended Cox model.

IX. The Cox Likelihood

Likelihood:

- Typically based on outcome distribution
- Outcome distn not specified for Cox model (non-parametric)
- Instead we have a partial likelihood.

Scenario:

- Gary, Larry, Barry have lottery tickets
- Winning tickets chosen at times t_1, t_2, \dots
- Each person eventually chosen
- Can only be chosen once.

Q: What is the probability that the order chosen is Barry, Gary, Larry.

$$P = \frac{1}{3} \times \frac{1}{2} \times \frac{1}{1} = \frac{1}{6}$$

Modification:

Barry — 4 tickets

Gary — 1 ticket

Larry — 2 tickets

Q: P order is Barry, Larry, Gary

$$(\frac{4}{7})(\frac{1}{3})(\frac{2}{2}) = \frac{4}{21}$$

- Subject's number of tickets affects probability.
- Analogously, Cox model likelihood affected by patterns of covariates.

Individual hazards analogous to number of tickets

ID	TIME	STATUS	SMOKE
Barry	2	1	1
Garry	3	1	0
Harry	5	0 censored	0
Larry	8	1	1

ID	Barry	Garry	Harry	Larry
hazard	$h_0(t)e^{\beta_1}$	$h_0(t)e^0$	$h_0(t)e^0$	$h_0(t)e^{\beta_1}$

$$L = L_1 \times L_2 \times L_3$$

$$L_1 = \left[\frac{h_0(t)e^{\beta_1}}{h_0(t)e^{\beta_1} + h_0(t)e^0 + h_0(t)e^0 + h_0(t)e^{\beta_1}} \right]$$

$$L_2 = \left[\frac{h_0(t)e^0}{h_0(t)e^0 + h_0(t)e^0 + h_0(t)e^{\beta_1}} \right]$$

$$L_3 = \left[\frac{h_0(t)e^{\beta_1}}{h_0(t)e^{\beta_1}} \right]$$

Harry censored

* X. Using Age as the Time Scale

Time-on-study vs. age at follow up.

Possible choices for time 0:

- Study entry, birth - treatment time, etc.

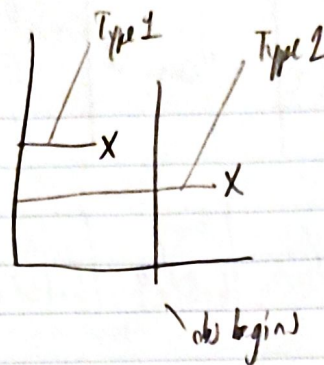
ex) Subject enters study at age 45. Their t_0 is 45 years, and their survival time is left-truncated at $t_0 = 45$.

Left-truncation

- Subject not observed before t_0
- If subject has event before t_0 , excluded from study
- If subject has event after t_0 , then included in study and assumed not at risk until t_0 .

Left truncation

Type 1	Type 2
• event before t_0	• event after t_0
• not included	• included
• effect of E underestimated	



Left censored means event did happen in past but we don't know when.

Time-on-study vs. age-at-follow-up.

If you use age-as-time-scale, then the risk set can increase as subjects are added.

Typically use time-on-study for clinical trials
(this applies to my case too!)

For left-truncated data, age as time scale better

I HAVE EVERYTHING I NEED NOW TO START