Plan for this lecture

- $\star\,$ Anything to say about the project?
- \star Recall some theory parts
 - partial likelihood function for a relative risk model
 - score test for a relative risk model
 - two-sample nonparametric log-rank test
- \star Cox regression with one binary covariate
- Score test for a Cox regression situation with one binary covariate
 two-sample log-rank test

Partial likelihood function for a relative risk model

★ Situation:

- *n* individuals
- individual *i* has covariate vector $x_i(t)$
- individual *i* has hazard rate and intensity process

$$\begin{aligned} \alpha(t|x_i(t)) &= \alpha_0(t)r(\beta, x_i(t)) \\ \lambda_i(t) &= Y_i(t)\alpha_0(t)r(\beta, x_i(t)) \end{aligned}$$

$$- N_i(t), Y_i(t), N_{\bullet}(t), Y_{\bullet}(t), \lambda_{\bullet}(t)$$

★ Partial likelihood

$$L(\beta) = \prod_{j} \left[\frac{r(\beta, x_{ij}(T_j))}{\sum_{l=1}^{n} Y_l(T_j) r(\beta, x_l(T_j))} \right]$$

★ For Cox regression models:

$$r(\beta, x_i(t)) = \exp\{\beta^T x_i(t)\}\$$

Score test for a relative risk model

★ Situation:

- *n* individuals
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$$\begin{aligned} \alpha(t|x_i(t)) &= \alpha_0(t)r(\beta, x_i(t)) \\ \lambda_i(t) &= Y_i(t)\alpha_0(t)r(\beta, x_i(t)) \end{aligned}$$

$$- N_i(t), Y_i(t), N_{\bullet}(t), Y_{\bullet}(t), \lambda_{\bullet}(t)$$

- * Score test: $H_0: \beta = \beta_0$
 - test statistic

$$\chi^2_{SC} = U(\beta_0)^T I(\beta_0)^{-1} U(\beta_0)$$

where χ^2_{SC} is approximately $\chi^2\text{-distributed}$ with q degrees of freedom when H_0 is true

$$- U(\beta) = \nabla \ell(\beta) - I(\beta) = -\nabla U(\beta) = -\nabla^2 \ell(\beta)$$

Two-sample nonparametric log-rank test

* Have two counting processes $N_1(t)$ and $N_2(t)$ with intensity processes

$$\lambda_1(t) = Y_1(t)\alpha_1(t)$$
$$\lambda_2(t) = Y_2(t)\alpha_2(t)$$

- ★ Log-rank test: H_0 : $\alpha_1(t) = \alpha_2(t), t \in [0, t_0]$
- test statistic

$$\chi^2_{LR} = \frac{Z_1(t_0)^2}{V_{11}(t_0)}$$

where χ^2_{LR} is approximately χ^2 distributed with 1 degree of freedom when ${\cal H}_0$ is true

$$- Z_1(t_0) = \int_0^{t_0} \frac{L(t)}{Y_1(t)} dN_1(t) - \int_0^{t_0} \frac{L(t)}{Y_2(t)} dN_2(t)$$

$$- V_{11}(t_0) = \int_0^{t_0} \frac{L^2(t)}{Y_1(t)Y_2(t)} dN_{\bullet}(t)$$

$$-L(t) = \frac{Y_1(t)Y_2(t)}{Y_{\bullet}(t)}$$