

# Biostatistics (MATH11230)

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# Proportional hazards model

## General context

- ↪ We end the course discussing regression models for time to event/survival data.
- ↪ There are various ways for modelling the dependency between the event time and the factors that might affect it.
- ↪ The two most common approaches are known as the **proportional hazards model** and the **accelerated failure time model**.
- ↪ Due to time constraints, here we will only cover the first class of methods: proportional hazards models.

# Proportional hazards model

## Comparing two groups

- ↪ Let us start with the simplest example: suppose that two groups of (male) patients receive either a standard treatment or a new treatment for prostate cancer.
- ↪ The primary goal is to identify whether patients in the two treatment groups have a different survival experience.
- ↪ Let  $h_S(t)$  and  $h_N(t)$  be the hazards of death at time  $t$  for patients in the standard and new treatment groups, respectively.
- ↪ A convenient model is to assume that the hazard at time  $t$  for a patient in the new treatment group is proportional to the hazard at the same time in the standard treatment.
- ↪ This proportional hazards model can be expressed as

$$h_N(t) = \psi h_S(t), \quad t \geq 0, \quad (1)$$

where  $\psi$  is a constant.

# Proportional hazards model

## Comparing two groups

- ↪ An implication of the proportional hazards model is that the corresponding true survival functions for individuals in the new and in the standard treatments do not cross.
- ↪ To see why, recall that by assumption  $h_N(t) = \psi h_S(t)$ . Integrating both sides of this expression, multiplying by (-1) and exponentiating gives

$$\exp \left\{ - \int_0^t h_N(u) du \right\} = \exp \left\{ - \int_0^t \psi h_S(u) du \right\}.$$

- ↪ Further note that

$$\exp \left\{ - \int_0^t \psi h_S(u) du \right\} = \left[ \exp \left\{ - \int_0^t h_S(u) du \right\} \right]^\psi,$$

and remember that

$$S(t) = \exp \left\{ - \int_0^t h(u) du \right\}.$$

# Proportional hazards model

## Comparing two groups

↪ Therefore, if  $S_N(t)$  and  $S_S(t)$  are the survival functions for the two groups, then

$$S_N(t) = [S_S(t)]^\psi.$$

- ↪ The parameter  $\psi$  is called the hazard ratio: it is the ratio of the hazard of death at any time for an individual in the standard treatment.
- ↪ If  $0 < \psi < 1$ , the hazard of death at time  $t$  is smaller for an individual on the new drug relative to an individual on the standard treatment.
- ↪ The new treatment is then an improvement on the standard treatment.
- ↪ On the other hand, if  $\psi > 1$ , the hazard of death at time  $t$  is greater for an individual in the new drug, and the standard treatment is superior.

# Proportional hazards model

## Comparing two groups

- ↪ An alternative way of expressing the model in (1) leads to a model that can be more easily generalised.
- ↪ Suppose that survival times are available on  $n$  individuals and denote the hazard function for the  $i$ th of these by  $h_i(t)$ , for  $i = 1, \dots, n$ .
- ↪ Also, let  $h_0(t)$  be the hazard function for an individual in the standard treatment group.
- ↪ The hazard function for an individual on the new treatment is then  $\psi h_0(t)$ .
- ↪ The relative hazard cannot be negative and so it convenient to set  $\psi = \exp(\beta)$ .
- ↪ Any value of  $\beta \in (-\infty, +\infty)$  will lead to a positive value of  $\psi$ .

# Proportional hazards model

## Comparing two groups

- ↪ Now let  $X$  be an indicator variable, which takes the value zero if an individual is on the standard treatment and one if the individual is on the new drug.
- ↪ The hazard function for individual  $i$  is then written as

$$\begin{aligned} h_i(t) &= h(t \mid X = x_i) = h_0(t)e^{\beta x_i} \\ &= \begin{cases} h_0(t), & \text{if } x_i = 0 \\ h_0(t)e^{\beta}, & \text{if } x_i = 1 \end{cases} \end{aligned}$$

- ↪ This is the proportional hazards model for the comparison of two treatment groups.
- ↪ Here  $\psi = e^{\beta}$  represents the ratio of the hazard of death for an individual on the new treatment relative to one on the standard treatment at any time  $t$ .

# Proportional hazards model

## General model

- ↪ The model we have considered just now can be generalised to the situation where the hazard of death (or, more generally, the hazard of the event of interest occurring) at a particular time point depends on the values of  $x_1, \dots, x_p$  of  $p$  explanatory variables/risk factors/covariates  $X_1, \dots, X_p$ .
- ↪ The values of these variables will be assumed to have been recorded at the time origin of the study.
- ↪ We will not cover them but there are extensions of the model to cover the situation where the values of one or more explanatory variables change over time.
- ↪ Let  $\mathbf{x}_i = (x_{i1}, \dots, x_{ip})'$  be the explanatory variables for individual  $i$ .



# Proportional hazards model

## General model

↪ The proportional hazards model is given by

$$h_i(t) = h(t | \mathbf{x}_i) = h_0(t) \exp(x_{i1}\beta_1 + x_{i2}\beta_2 + \dots + x_{ip}\beta_p) = h_0(t) \exp(\mathbf{x}_i' \boldsymbol{\beta}), \quad (2)$$

where  $\boldsymbol{\beta} = (\beta_1, \dots, \beta_p)'$ .

↪ Note that there is no intercept  $\beta_0$  in model (2).

↪ Obviously,

$$h(t | \mathbf{x}_i = \mathbf{0}) = h_0(t),$$

and, therefore,  $h_0(t)$  is often called the baseline hazard function.

↪ It can be interpreted as the hazard function for a standard subject, which is a subject with  $\mathbf{x} = \mathbf{0}$ .

# Proportional hazards model

## General model

- ↪ Any parametric hazard function can be used for  $h_0(t)$  and as we will see next week,  $h_0(t)$  can be left completely unspecified without sacrificing the ability to estimate  $\beta$ , by the use of Cox's semiparametric proportional hazards model.
- ↪ The hazard function for the  $i$ th subject always has the same general shape as  $h_0(t)$ , but can be, say, doubled or halved, depending in the individual's risk factors.
- ↪ The term  $\exp(x_i'\beta)$  is in many cases the function of primary interest as it describes the (relative) effects of the risk factors.
- ↪ Note that the model separates clearly the effect of time from the effect of the risk factors.

# Proportional hazards model

## General model

↪ The proportional hazards model can also be written in terms of the survival function:

$$\begin{aligned} S_i(t) = S(t \mid \mathbf{x}_i) &= \exp \left( - \int_0^t h_i(u) du \right) \\ &= \exp \left( - \int_0^t h_0(u) \exp(x_i' \beta) du \right) \\ &= \left[ \exp \left( - \int_0^t h_0(u) du \right) \right]^{\exp(x_i' \beta)} \\ &= S_0(t)^{\exp(x_i' \beta)}, \end{aligned}$$

where  $S_0(t)$  is the baseline survival function.

# Proportional hazards model

## General model

↪ On the log-hazard scale this model can be written as

$$\log h_i(t) = \log h_0(t) + x_i' \beta.$$

↪ The coefficient  $\beta_j$  associated with the risk factor  $X_j$  represents the change in the log hazard when  $X_j$  is increased by one unit, all the other risk factors being constant, that is,

$$\beta_j = \log h(t \mid X_1, X_2, \dots, X_j + 1, \dots, X_p) - \log h(t \mid X_1, X_2, \dots, X_j, \dots, X_p),$$

which is equivalent to the log of the ratio of the hazards at time  $t$ .

↪ Equivalently,

$$\frac{h(t \mid X_1, X_2, \dots, X_j + 1, \dots, X_p)}{h(t \mid X_1, X_2, \dots, X_j, \dots, X_p)} = e^{\beta_j}.$$

↪ Therefore,  $e^{\beta_j}$  represents the ratio of the hazards for a one unit change in  $X_j$  at any time  $t$  while keeping all the other risk factors constant.

# Proportional hazards model

## General model

- ↪ The main assumption of this model is the **proportional hazards assumption**, that is, the fact that the ratio of the hazard functions for two subjects with covariates  $\mathbf{x}_i$  and  $\mathbf{x}_l$  is constant over time, that is,

$$\frac{h(t \mid \mathbf{x}_i)}{h(t \mid \mathbf{x}_l)} = \frac{\exp(\mathbf{x}_i' \boldsymbol{\beta})}{\exp(\mathbf{x}_l' \boldsymbol{\beta})} = \exp\{(\mathbf{x}_i - \mathbf{x}_l)' \boldsymbol{\beta}\}.$$

- ↪ In other words, all the time dependency is captured by the baseline hazard function which is common to all observations.
- ↪ The proportional hazards model also further assumes that the relationship between the risk factors and the log hazard function is linear.

# Proportional hazards model

Toy example (from Klein and Moeschberger, 2003, p248)

↪ Let us consider for the sake of simplicity/illustration that the only risk factor that needs to be considered is race.

↪ Let us consider that race is a three level factor, whose levels are: black, white, hispanic.

↪ Define

$$X_1 = \begin{cases} 1, & \text{if the subject is black,} \\ 0, & \text{otherwise,} \end{cases} \quad X_2 = \begin{cases} 1, & \text{if the subject is white,} \\ 0, & \text{otherwise.} \end{cases}$$

↪ Hispanic is therefore the reference category.

# Proportional hazards model

Toy example (from Klein and Moeschberger, 2003, p248)

↪ The proportional hazards model is

$$h(t | X_1, X_2) = h_0(t) \exp(X_1 \beta_1 + X_2 \beta_2)$$
$$= \begin{cases} h_0(t), & \text{if } X_1 = X_2 = 0 \text{ (hispanic subject),} \\ h_0(t) \exp(\beta_1), & \text{if } X_1 = 1, X_2 = 0 \text{ (black subject),} \\ h_0(t) \exp(\beta_2), & \text{if } X_1 = 0, X_2 = 1 \text{ (white subject).} \end{cases}$$

- ↪ The risk of the event occurring among black subjects relative to the risk of the event occurring among hispanic subjects is  $e^{\beta_1}$ .
- ↪ The risk of the event occurring among white subjects relative to the risk of the event occurring among hispanic subjects is  $e^{\beta_2}$ .
- ↪ The risk of the event occurring among black subjects relative to the risk of the event occurring among white subjects is  $e^{\beta_1 - \beta_2}$ .