Duration Models Introduction to Single Spell Models

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• The hazard function gives the probability that a spell, denoted by the nonnegative random variable T with distribution g(t), will end at t, given that it has lasted until t:

$$h(t) = f(t|T > t) = \frac{g(t)}{1 - G(t)} \ge 0.$$

• Integrated hazard function (using G(0) = 0 to eliminate c):

$$H(t) = \int_0^t h(u)du = -\ln(1-G(t))|_0^t + c = -\ln(1-G(t)).$$

• Working backwards, we can derive g from h:

$$G(t) = 1 - e^{-\int_0^t h(u)du} = 1 - e^{-H(t)},$$

$$g(t) = h(t)[1 - G(t)] = h(t)e^{-H(t)},$$

$$H(t) = \int_0^t h(u) du.$$

 So the survival function, the probability that the spell lasts until t, i.e., T ≥ t, is

$$S(t) = 1 - G(t) = e^{-H(t)}$$
.

 The density and hazard function for T may have a number of qualities. If T has a nondefective duration density, then

$$\lim_{t\to\infty}\int_0^t h(u)du\to\infty\Longleftrightarrow S(\infty)=0$$

- Duration dependence arises when $\frac{\partial h(t)}{\partial t} \neq 0$.
- If $\frac{\partial h(t)}{\partial t} > 0$ (< 0), then we have positive (negative) duration dependence



 In constructing estimable models, we will often work with the conditional hazard function

$$h(t|x(t),\theta(t)),$$

where the regressor vector x(t) may include

- Entire past: $x_1(t) = \int_{\infty}^{t} k_1(z_1(u)) du$
- or future: $x_2(t) = \int_t^\infty k_2(z_2(u)) du$
- or both: $x_3(t) = \int_{-\infty}^{\infty} k_3(z_3(u), t) du$

of some variables.



Associated with the conditional hazard function is the conditional survival function

$$S(t|x(t), \theta(t)) = 1 - G(t|x(t), \theta(t)) = e^{-\int_0^t h(u|x(u), \theta(u))du}$$

and the conditional density of T

$$g(t|x(t),\theta(t)) = h(t|x(t),\theta(t)) \cdot [1 - G(t|x(t),\theta(t))]$$

= $h(t|x(t),\theta(t)) \cdot e^{-\int_0^t h(u|x(u),\theta(u))du}$.

- In these models we will assume
 - **1** $\theta(t)$ independent of x(t) and $\theta \sim \mu(\theta)$, $x \sim D(x)$
 - 2 No functional restrictions connecting the conditional distribution of $T|\theta, x$ and the marginal distribution of θ, x .



 A common specification of the conditional hazard is the proportional hazard specification:

$$h(t|x(t),\theta(t)) = \psi(t)\phi(x(t))\eta(\theta(t))$$

$$\ln h(t|x(t),\theta(t)) = \ln \psi(t) + \ln \phi(x(t)) + \ln \eta(\theta(t))$$

$$\psi(t) \ge 0, \quad \phi(x(t)) > 0, \quad \eta(\theta(t)) \ge 0 \quad \forall t.$$



Sampling Plans and Initial Condition Problems



- For interrupted spells, one of the following duration times may be observed:
 - time in state up to sampling date (T_b)
 - time in state after sampling date (T_a)
 - total time in completed spell observed at origin of sample $(T_c = T_a + T_b)$
- Duration of spells beginning after the origin date of the sample, denoted T_d, are not subject to initial condition problems.
- The intake rate, $k(-t_b)$, is the proportion of the population entering a spell at $-t_b$.



Assume

- a time homogenous environment, i.e. constant intake rate, $k(-t_b) = k, \forall b$
- a model without observed or unobserved explanatory variables.
- no right censoring, so $T_c = T_a + T_b$
- underlying distribution is nondefective
- $m = \int_0^\infty xg(x)dx < \infty$



• The proportion of the population experiencing a spell at t=0, the origin date of the sample, is

$$P_0 = \int_0^\infty k(-t_b)(1-G(t_b))dt_b = k \int_0^\infty (1-G(t_b))dt_b$$

$$= k \left[t_b(1-G(t_b))|_0^\infty - \int t_b d(1-G(t_b)) \right]$$

$$= k \int t_b g(t_b)dt_b = km,$$

where $1 - G(t_b)$ is the probability the spell lasts from $-t_b$ to 0 (or equivalently, from 0 to $-t_b$).



• So the density of a spell of length t_b interrupted at the beginning of the sample (t = 0) is

$$f(t_b) = rac{ ext{proportion surviving til } t = 0 ext{ from batch } t_b}{ ext{total surviving til } t = 0}$$
 $= rac{k(-t_b)(1-G(t_b))}{P_0} = rac{1-G(t_b)}{m}
eq g(t_b)$

• The probability that a spell lasts until t_c given that it has lasted from $-t_b$ to 0, is

$$g(t_c|t_b) = \frac{g(t_c)}{1 - G(t_b)}$$

• So the density of a spell that lasts for t_c is

$$f(t_c) = \int_0^{t_c} g(t_c|t_b) f(t_b) dt_b$$
$$= \int_0^{t_c} \frac{g(t_c)}{m} dt_b = \frac{g(t_c)t_c}{m}$$

Likewise, the density of a spell that lasts until t_a is

$$f(t_a) = \int_0^\infty g(t_a + t_b|t_b) f(t_b) dt_b$$

$$= \int_0^\infty \frac{g(t_a + t_b)}{m} dt_b$$

$$= \frac{1}{m} \int_{t_a}^\infty g(t_b) dt_b$$

$$= \frac{1 - G(t_a)}{m}$$

• So the functional form of $f(t_b) \approx f(t_a)$.



- Some useful results that follow from this model:
 - 1 If $g(t) = \theta e^{-t\theta}$, then $f(t_b) = \theta e^{-t_b\theta}$ and $f(t_a) = \theta e^{-t_a\theta}$. Proof:

$$g(t) = \theta e^{-t\theta} \to m = \frac{1}{\theta},$$
 $G(t) = 1 - e^{-t\theta} \to f(t_a) = \frac{1 - G(t)}{m} = \theta e^{-t\theta}$

2
$$E(T_a) = \frac{m}{2}(1 + \frac{\sigma^2}{m^2}).$$



Proof:

$$E(T_{a}) = \int t_{a}f(t_{a})dt_{a} = \int t_{a}\frac{1 - G(t_{a})}{m}dt_{a}$$

$$= \frac{1}{m} \left[\frac{1}{2}t_{a}^{2}(1 - G(t_{a}))|_{0}^{\infty} - \int \frac{1}{2}t_{a}^{2}d(1 - G(t_{a})) \right]$$

$$= \frac{1}{m} \int \frac{1}{2}t_{a}^{2}g(t_{a})dt_{a} = \frac{1}{2m}[var(t_{a}) + E^{2}(t_{a})]$$

$$= \frac{1}{2m}[\sigma^{2} + m^{2}]$$



- $E(T_b) = \frac{m}{2}(1 + \frac{\sigma^2}{m^2}).$
- **Proof**: See proof of Proposition 2.
- $E(T_c) = m(1 + \frac{\sigma^2}{m^2}).$
- Proof:

$$E(T_c) = \int \frac{t_c^2 g(t_c)}{m} dt_c = \frac{1}{m} (var(t_c) + E^2(t_c))$$

$$\rightarrow E(T_c) = 2E(T_a) = 2E(T_b), E(T_c) > m \text{ unless } \sigma^2 = 0$$



- $h'(t) > 0 \to E(T_a) = E(T_b) > m$.
- Proof: See Barlow and Proschan.
- $h'(t) < 0 \rightarrow E(T_a) = E(T_b) < m$.
- Proof: See Barlow and Proschan.



Pitfalls in Using Regression Methods to Analyze Duration Data



- **1** Density of duration in a spell (T) for an individual with fixed characteristics Z is f(t|Z).
- Assume
 - No time elapses between end of one spell and beginning of another,
 - No unobserved heterogeneity components,

 - 4 At origin, t = 0, of sample of length K, everyone begins a spell.
- \odot The expected length of spell in the population given Z is

$$E(T|Z) = \int_0^\infty tf(t|Z)dt = \frac{1}{\theta(Z)} = \beta Z.$$



1 The expected length of a spell in a sample frame of length k, however, is

$$\begin{split} E(T|Z,K) &= \int_0^K tf(t|Z)dt + K \int_K^\infty f(t|Z)dt \\ &= \int_0^K t\theta e^{-\theta t} dt + K \int_K^\infty \theta e^{-\theta t} dt \\ &= \left[-te^{-\theta t} \Big|_0^K + \int_0^K e^{-\theta t} dt \right] + K \left[\int_K^\infty \theta e^{-t\theta} dt \right] \\ &= \left[-Ke^{-\theta K} + \left(-\frac{1}{\theta} e^{-\theta t} \right) \Big|_0^K \right] + K \left[-e^{-\theta t} \Big|_K^\infty \right] \\ &= -Ke^{-\theta K} - \frac{1}{\theta} e^{-\theta K} + \frac{1}{\theta} + Ke^{-\theta K} \\ &= \beta Z (1 - e^{-\frac{K}{\beta Z}}) \neq \beta Z. \end{split}$$

- So OLS of T on Z will not estimate β . But as $K \to \infty$, the selection bias term $(\beta Z e^{-\frac{K}{\beta Z}})$ disappears.
- A widely used method to avoid this bias is to use only completed first spells.
- This results in another sort of selection bias,

$$E(T \mid Z, K, T < K) = \frac{\int_0^K tf(t|Z)dt}{\int_0^K f(t|Z)dt}$$
$$= \frac{-Ke^{-\theta K} - \frac{1}{\theta}e^{-\theta K} + \frac{1}{\theta}}{1 - e^{-\theta K}},$$

where recall that

$$\beta Z = \frac{1}{\theta}.$$

• As $K \to \infty$.

$$E(T \mid Z, K, T < K) = \beta Z.$$

