

We have already seen, e.g., the simple joint life status $(x : \bar{n})$ along with the corresponding future life time r.v.'s $T(x : \bar{n})$ and $K(x : \bar{n})$. We shall generalize the notion to the case when $T(\bar{n})$ is not a degenerate r.v. We shall thus have $T(x : y)$.

Definition 0.1 (Future lifetime of the joint life status)

The r.v. $T(x : y)$ having an (absolutely) continuous c.d.f. is given by

$$T(x : y) := \min(T(x), T(y)) = \begin{cases} T(x), & T(x) \leq T(y) \\ T(y), & T(x) > T(y) \end{cases},$$

with

$${}_t p_{x:y} = \mathbf{P}[T(x) \geq t, T(y) \geq t], \quad t \geq 0,$$

and

$${}_t q_{x:y} = {}_t q_x + {}_t q_y - \mathbf{P}[T(x) < t, T(y) < t], \quad t \geq 0.$$

Note.

For independent $T(x)$ and $T(y)$, we certainly have that

$${}_t p_{x;y} = {}_t p_x \cdot {}_t p_y = 1 - {}_t q_x - {}_t q_y + {}_t q_x \cdot {}_t q_y, \quad t \geq 0,$$

and

$${}_t q_{x;y} = {}_t q_x + {}_t q_y - {}_t q_x \cdot {}_t q_y, \quad t \geq 0.$$

Question.

Can we calculate the p.d.f. of $T(x : y)$? Remember that

$${}_t p_{x;y} \mu(x : y + t) = \frac{d}{dt} {}_t q_{x;y} = -\frac{d}{dt} {}_t p_{x;y} \stackrel{\text{ind}}{=} -\frac{d}{dt} {}_t p_x \cdot {}_t p_y.$$

Proposition 0.1

Let $T(x)$ and $T(y)$ be independent future lifetimes of (x) and (y) . Then

$${}_t p_{x;y} \mu(x : y + t) = {}_t p_x \cdot {}_t p_y (\mu(x + t) + \mu(y + t)), \quad t > 0.$$

Proof.

We have that

$$\begin{aligned}\frac{d}{dt}({}_t p_x \cdot {}_t p_y) &= {}_t p'_x {}_t p_y + {}_t p_x \cdot {}_t p'_y \\ &= -{}_t p_x \mu(x+t) {}_t p_y - {}_t p_y \mu(y+t) {}_t p_x \\ &= -{}_t p_x \cdot {}_t p_y (\mu(x+t) + \mu(y+t)),\end{aligned}$$

which completes the proof. □

Note.

It readily follows that

$$\mu((x : y) + t) \stackrel{ind}{=} \mu(x+t) + \mu(y+t), \quad t > 0.$$

Proposition 0.2

Let $T(x)$ and $T(y)$ be the future lifetimes of (x) and (y) . Then

$${}_t p_{x:y} \mu((x : y) + t) = \int_t^\infty f_{T(x), T(y)}(t, v) dv + \int_t^\infty f_{T(x), T(y)}(u, t) du.$$

Leibniz integral rule

Under certain continuity assumptions, we have that

$$\begin{aligned} \frac{d}{dt} \int_{\alpha(t)}^{\beta(t)} \psi(t, v) dv &= \frac{d}{dt} \varphi(\alpha(t), \beta(t), t) \\ &= \int_{\alpha(t)}^{\beta(t)} \frac{d}{dt} \psi(t, v) dv + \frac{d\beta(t)}{dt} \psi(t, \beta(t)) - \frac{d\alpha(t)}{dt} \psi(t, \alpha(t)) \end{aligned}$$

Proof of Proposition 0.2.

We are interested in the derivative

$$f_{T(x:y)}(t) = -\frac{d}{dt}\bar{F}_{T(x:y)}(t) = -\frac{d}{dt}\mathbf{P}[T(x) \geq t, T(y) \geq t], \quad t \geq 0.$$

Let

$$\psi(t, v) = \int_t^\infty f_{T(x), T(y)}(u, v) du,$$

we are then interested in the derivative of

$$\bar{F}_{T(x), T(y)}(t, t) = \int_t^\infty \int_t^\infty f_{T(x), T(y)}(u, v) dudv = \int_t^\infty \psi(t, v) dv.$$

So we can use Leibniz rule with $\beta(t) = \infty$ and $\alpha(t) = t$, and hence,

$$\frac{d\beta(t)}{dt}\psi(t, \beta(t)) = 0, \quad \frac{d\alpha(t)}{dt}\psi(t, \alpha(t)) = \int_t^\infty f_{T(x), T(y)}(u, t) du.$$

Also, using the Leibniz rule again or just due to the FTC,

Proof.

$$\frac{d}{dt}\psi(t, v) = \frac{d}{dt} \int_t^\infty f_{T(x), T(y)}(u, v) du = -f_{T(x), T(y)}(t, v)$$

Hence

$$\begin{aligned} & \frac{d}{dt} \bar{F}_{T(x), T(y)}(t, t) \\ &= - \left(\int_t^\infty f_{T(x), T(y)}(t, v) dv + \int_t^\infty f_{T(x), T(y)}(u, t) du \right), \end{aligned}$$

that completes the proof. □

Note.

In the case of independent future lifetimes, we have that

$$\int_t^\infty f_{T(x), T(y)}(t, v) dv = f_{T(x)}(t) \int_t^\infty f_{T(y)}(v) dv = f_{T(x)}(t) \bar{F}_{T(y)}(t)$$

and also that

$$\int_t^\infty f_{T(x), T(y)}(u, t) du = f_{T(y)}(t) \int_t^\infty f_{T(x)}(u) du = f_{T(y)}(t) \bar{F}_{T(x)}(t),$$

which gives

$${}_t p_x \mu(x+t) {}_t p_y + {}_t p_y \mu(y+t) {}_t p_x = {}_t p_x {}_t p_y (\mu(x+t) + \mu(y+t)),$$

i.e., the p.d.f. of $T(x : y)$ under independence.

Question.

What happens with $K(x)$ and $K(y)$?

Proposition 0.3

The p.n.f. of $K(x : y)$ is

$$\mathbf{P}[K(x : y) = k] = {}_k p_{x:y} \cdot q_{x+k:y+k} = {}_k q_{x:y}, \quad k = 0, 1, \dots$$

Proof.

We have already proven that for any life status (u), the p.m.f. of $K(u)$ is

$$\mathbf{P}[K(u) = k] = {}_k p_u \cdot q_{u+k}, \quad k = 0, 1, \dots$$

Thus, for $(u) = (x : y)$, we have that

$$\mathbf{P}[K(x : y) = k] = {}_k p_{x:y} \cdot q_{(x:y)+k}, \quad k = 0, 1, \dots$$

Moreover, in the case of the joint life status

$$q_{(x:y)+k} = q_{x+k:y+k},$$

which completes the proof. □

Note.

Under independence,

$$\mathbf{P}[K(x : y) = k] = {}_k p_x \cdot {}_k p_y (1 - p_{x+k} p_{y+k}).$$

In other words,

$$\mathbf{P}[K(x : y) = k] = {}_k p_x \cdot {}_k p_y (q_{x+k} + q_{y+k} - q_{x+k} q_{y+k}).$$

Proposition 0.4

The probability that the joint life status will not survive k time units is

$$\mathbf{P}[K(x : y) \leq k] = {}_{k+1} q_{x:y}, \quad k = 0, 1, \dots$$

Note.

The joint life status can be generalized to take into account more than two lives. Then we shall have $T(x_1 : x_2 : \dots : x_n)$.

Note.

We have already seen the last survivor life status $(\overline{x : y})$. In the case of two lives we conveniently have that

$$T(x : y) + T(\overline{x : y}) = T(x) + T(y).$$

Definition 0.2 (Future lifetime of the last survivor life status)

The r.v. $T(\overline{x:y})$ having an (absolutely) continuous c.d.f. is given by

$$T(\overline{x:y}) := \max(T(x), T(y)) = \begin{cases} T(x), & T(x) > T(y) \\ T(y), & T(x) \leq T(y) \end{cases},$$

with

$${}_tq_{\overline{x:y}} = \mathbf{P}[T(x) < t, T(y) < t], \quad t \geq 0,$$

and

$${}_tp_{\overline{x:y}} = {}_tp_x + {}_tp_y - \mathbf{P}[T(x) \geq t, T(y) \geq t], \quad t \geq 0.$$

These can be further specialized under independence of $T(x)$ and $T(y)$.

Proposition 0.5

Let $(x : y)$ and $\overline{x : y}$ be the joint and last survivor life statuses, respectively. Then

$${}_tq_{x:y} + {}_tq_{\overline{x:y}} = {}_tq_x + {}_tq_y, \quad t \geq 0.$$

Proof.

Let $\mathcal{A} := \{T(x) < t\}$ and $\mathcal{B} := \{T(y) < t\}$, $t \geq 0$. Then

$$\mathcal{A} \cap \mathcal{B} = \{T(\overline{x : y}) < t\}$$

and

$$\mathcal{A} \cup \mathcal{B} = \{T(x : y) < t\}.$$

Then employ

$$\mathbf{P}\{\mathcal{A} \cup \mathcal{B}\} = \mathbf{P}\{\mathcal{A}\} + \mathbf{P}\{\mathcal{B}\} - \mathbf{P}\{\mathcal{A} \cap \mathcal{B}\}$$

to complete the proof. □

Note.

We can use the proposition to obtain, say, ${}_tq_{\overline{x:y}}$. Namely,

$$\begin{aligned}{}_tq_{\overline{x:y}} &= {}_tq_x + {}_tq_y - {}_tq_{x:y} \\ &= {}_tq_x + {}_tq_y - ({}_tq_x + {}_tq_y - \mathbf{P}[T(x) < t, T(y) < t]) \\ &= \mathbf{P}[T(x) < t, T(y) < t], \quad t \geq 0.\end{aligned}$$

Note.

As ${}_tq_u = 1 - {}_tp_u$ for any $t \geq 0$ and any life status (u), we have that

$${}_tp_{x:y} + {}_tp_{\overline{x:y}} = {}_tp_x + {}_tp_y, \quad t \geq 0.$$

Hence

$${}_tp_{\overline{x:y}} = {}_tp_x + {}_tp_y - \mathbf{P}[T(x) \geq t, T(y) \geq t], \quad t \geq 0.$$

Proposition 0.6

The force of mortality of the last survivor life status is

$$\mu(\overline{x:y} + t) = \frac{{}_t p_x \mu(x + t) + {}_t p_y \mu(y + t) - {}_t p_{x:y} \mu((x : y) + t)}{{}_t p_{\overline{x:y}}},$$

where $t > 0$.

Proof.

Because of Proposition 0.5, it holds that

$${}_t p_x \mu(x + t) + {}_t p_y \mu(y + t) = {}_t p_{x:y} \mu((x : y) + t) + {}_t p_{\overline{x:y}} \mu(\overline{x:y} + t),$$

for $t > 0$. Dividing by ${}_t p_{\overline{x:y}} \neq 0$ throughout completes the proof. □

Corollary 0.1

Under independence we have that

$$\mu(\overline{(x : y)} + t) = \frac{{}_t p_x \cdot {}_t q_y \mu(x + t) + {}_t p_y \cdot {}_t q_x \mu(y + t)}{{}_t p_x \cdot {}_t q_y + {}_t p_y \cdot {}_t q_x + {}_t p_x \cdot {}_t p_y},$$

where $t > 0$.

Proof.

The following completes the proof

$$\begin{aligned} & \mu(\overline{(x : y)} + t) \\ = & \frac{{}_t p_x \mu(x + t) + {}_t p_y \mu(y + t) - {}_t p_x {}_t p_y (\mu(x + t) + \mu(y + t))}{{}_t p_{\overline{x:y}}} \\ = & \frac{{}_t p_x \mu(x + t) + {}_t p_y \mu(y + t) - {}_t p_x {}_t p_y \mu(x + t) - {}_t p_x {}_t p_y \mu(y + t)}{{}_t p_x + {}_t p_y - {}_t p_x \cdot {}_t p_y} \\ = & \frac{{}_t p_x (1 - {}_t p_y) \mu(x + t) + {}_t p_y (1 - {}_t p_x) \mu(y + t)}{{}_t q_x \cdot {}_t p_y + {}_t q_y \cdot {}_t p_x + {}_t p_x \cdot {}_t p_y}. \end{aligned}$$

Note.

The relation we had for c.d.f.'s and p.d.f.'s of $T(x : y)$ and $T(\overline{x} : \overline{y})$ does not hold in the case of the force of mortality, i.e.,

$$\mu((x : y) + t) + \mu(\overline{x} : \overline{y} + t) \neq \mu(x + t) + \mu(y + t).$$

Proposition 0.7

The p.m.f. of $K(\overline{x} : \overline{y})$ is

$$\mathbf{P}[K(\overline{x} : \overline{y}) = k] = {}_k p_x \cdot q_{x+k} + {}_k p_y \cdot q_{y+k} - {}_k p_{x:y} q_{(x+k:y+k)},$$

for $k = 0, 1, \dots$

Proof.

Note that Proposition 0.5 does not assume any continuity restrictions. Thus

$${}_k q_x + {}_k q_y = {}_k q_{x:y} + {}_k q_{\overline{x} : \overline{y}}, \quad k = 0, 1, \dots$$

Proof.

As the above is true for any $k = 0, 1, \dots$, we also have that

$${}_{k+1}q_x + {}_{k+1}q_y = {}_{k+1}q_{x:y} + {}_{k+1}\overline{q_{x:y}}.$$

Then, observing that, for any (u) ,

$${}_{k+1}q_u - {}_kq_u = {}_kp_u \cdot q_{u+k},$$

we have that

$$\mathbf{P}[K(\overline{x:y}) = k] = {}_kp_x \cdot q_{x+k} + {}_kp_y \cdot q_{y+k} - {}_kp_{x:y}q_{(x:y)+k},$$

which along with

$$q_{(x:y)+k} = q_{x+k:y+k}$$

completes the proof. □

Note.

Under independence and for $k = 0, 1, \dots$, we have that

$$\begin{aligned} & \mathbf{P}[K(\overline{x} : \overline{y}) = k] \\ &= {}_k p_x q_{x+k} + {}_k p_y q_{y+k} - {}_k p_x \cdot {}_k p_y (q_{x+k} + q_{y+k} - q_{x+k} q_{y+k}) \\ &= {}_k p_x (1 - {}_k p_y) q_{x+k} + {}_k p_y (1 - {}_k p_x) q_{y+k} + {}_k p_x \cdot {}_k p_y q_{x+k} q_{y+k} \\ &= {}_k q_y \cdot {}_k p_x q_{x+k} + {}_k q_x \cdot {}_k p_y q_{y+k} + {}_k p_x \cdot {}_k p_y q_{x+k} q_{y+k}. \end{aligned}$$

Question.

Can we find treat the life status $(\overset{1}{x} : y)$ in a similar manner?

Definition 0.3 (Future lifetime of an ordered life status)

The r.v. $T(x^1 : y)$ having an (absolutely) continuous c.d.f. is given by

$$T(x^1 : y) := \begin{cases} T(x), & T(x) \leq T(y) \\ \infty, & T(x) > T(y) \end{cases},$$

with, for $t \geq 0$,

$${}_t q_{x:y}^1 = \mathbf{P}[T(x) < t, T(x) \leq T(y)] = \int_0^t \int_u^\infty f_{T(x), T(y)}(u, s) ds du.$$

Proposition 0.8

The p.d.f. of the above r.v. is

$$f_{T(x^1 : y)}(t) = \int_t^\infty f_{T(x), T(y)}(t, s) ds, \quad t \geq 0.$$

Proof.

Let

$$\psi(u) := \int_u^\infty f_{T(x), T(y)}(u, s) ds,$$

and take the derivative

$$\frac{d}{dt} \int_0^t \psi(u) du = \psi(t) = f_{T(x:y)}^1(t).$$

This completes the proof. □

Corollary 0.2

Under independence,

$${}_t q_{x:y}^1 = \int_0^t {}_s p_y \cdot {}_s p_x \mu(x+s) ds, \quad t \geq 0.$$

Proof.

Indeed, for non-negative t , we have that

$$\begin{aligned} {}_tq_{1;x:y} &= \int_0^t \int_u^\infty f_{T(x),T(y)}(u, s) ds du \\ &= \int_0^t \left(\int_u^\infty f_{T(y)|T(x)}(s|u) ds \right) f_{T(x)}(u) du \\ &= \int_0^t \mathbf{P}[(T(y) > u | T(x) = u)] {}_u p_x \mu(x + u) du \\ &\stackrel{\text{ind}}{=} \int_0^t {}_u p_y \cdot {}_u p_x \mu(x + u) du, \end{aligned}$$

as required. □

Note.

The p.d.f. is easily obtained as:

$$f_{T(x:y)}^1(t) = {}_t p_y \cdot {}_t p_x \mu(x + t), \quad t \geq 0.$$

Corollary 0.3

The c.d.f. of the r.v. $T(x : \bar{n})$ is

$${}_t q_{x:\bar{n}}^1 = \begin{cases} {}_t q_x, & t \leq n \\ {}_n q_x, & t > n \end{cases} \quad \text{and} \quad {}_t p_{x:\bar{n}}^1 = \begin{cases} {}_t p_x, & t \leq n \\ {}_n p_x, & t > n \end{cases} .$$

Proof.

Note that, say, for $t > n$,

$$\begin{aligned} {}_t q_{x:\bar{n}}^1 &= \int_0^n {}_s p_{\bar{n}} \cdot {}_s p_x \mu(x+s) ds + \int_n^t {}_s p_{\bar{n}} \cdot {}_s p_x \mu(x+s) ds \\ &= \int_0^n {}_s p_x \mu(x+s) ds = {}_n q_x. \end{aligned}$$

The other part follows in the same fashion, and this completes the proof. □

Question.

Is the probability ${}_tq_{x:\overline{2}|y}$ the same as ${}_tq_{x:y}^1$?

Definition 0.4 (Future lifetime of an ordered life status)

The r.v. $T(x : \overline{2}|y)$ having an (absolutely) continuous c.d.f. is given by

$$T(x : \overline{2}|y) := \begin{cases} T(y), & T(x) < T(y) \\ \infty, & T(x) \geq T(y) \end{cases},$$

with

$${}_tq_{x:\overline{2}|y} = \mathbf{P}[T(x) \leq T(y) \leq t] = \int_0^t \int_u^t f_{T(x), T(y)}(u, s) ds du, \quad t \geq 0.$$

Proposition 0.9

The p.d.f. of $T(x : \overset{2}{y})$ is

$$f_{T(x:\overset{2}{y})}(t) = \int_0^t f_{T(x),T(y)}(u,t) du, \quad t \geq 0.$$

Proof.

Let

$$\psi(u,t) := \int_u^t f_{T(x),T(y)}(u,s) ds.$$

Then we need to find the derivative

$$\frac{d}{dt} t q_{x:\overset{2}{y}} = \frac{d}{dt} \int_0^t \psi(u,t) du = \psi(t,t) + \int_0^t \frac{\partial}{\partial t} \psi(u,t) du,$$

which results in

$$\frac{d}{dt} t q_{x:\overset{2}{y}} = \int_0^t f_{T(x),T(y)}(u,t) du.$$

Corollary 0.4

Under independence, we have that

$${}_nq_{x:\overline{2}} = \int_0^n {}_sp_y \cdot {}_sp_x \mu(x+s) ds = {}_np_y \int_0^n {}_sp_x \mu(x+s) ds.,$$

for $n \geq 0$. Also,

$${}_np_{x:\overline{2}} \mu((x:\overline{2}) + n) = {}_np_y \mu(y+n) \cdot {}_nq_x,$$

for $n \geq 0$.

Proof.

For the density, the result follows immediately. Further,

Proof.

Under independence,

$$\begin{aligned} {}^nq_{x:\overline{2}|y} &= \int_0^n \int_s^n f_{T(x), T(y)}(s, t) dt ds \\ &= \int_0^n \int_s^n f_{T(y)|T(x)}(t|s) f_{T(x)}(s) dt ds \\ &= \int_0^n \mathbf{P}[s < T(y) \leq n | T(x) = s] f_{T(x)}(s) ds \\ &= \int_0^n \mathbf{P}[s < T(y) \leq n | T(x) = s] {}_s p_x \mu(x + s) ds \\ &\stackrel{ind}{=} \int_0^n \mathbf{P}[s < T(y) \leq n] {}_s p_x \mu(x + s) ds \\ &= \int_0^n ({}_s p_y - {}_n p_y) {}_s p_x \mu(x + s) ds, \end{aligned}$$

as required. □

Note.

It readily follows that

$$T(x^1 : y) \leq T(x : y^2).$$

Hence

$$\mathbf{P}[T(x^1 : y) < t] = {}_t q_{x:y}^1 \geq {}_t q_{x:y}^2 = \mathbf{P}[T(x : y^2) < t], \quad t \geq 0,$$

and also

$${}_t p_{x:y}^1 \leq {}_t p_{x:y}^2, \quad t \geq 0.$$

Proposition 0.10

The full expectancy of life is

$$\overset{\circ}{e}_{x:y} = \int_0^{\infty} {}_t p_{x:y} dt$$

for the joint life status, and

$$\overset{\circ}{e}_{\overline{x:y}} = \int_0^{\infty} {}_t p_{\overline{x:y}} dt$$

for the last survivor life status. Also, it holds that

$$\overset{\circ}{e}_{\overline{x:y}} + \overset{\circ}{e}_{x:y} = \overset{\circ}{e}_x + \overset{\circ}{e}_y.$$

Proof.

Recalling the formula for $\overset{\circ}{e}_u$ along with the fact that $T_x + T_y = T_{x:y} + T_{\overline{x:y}}$ completes the proof. □

Proposition 0.11

We have that

$$\mathbf{Var}[T(x : y)] = 2 \int_0^{\infty} t {}_t p_{x:y} dt - (\overset{\circ}{e}_{x:y})^2,$$

and

$$\mathbf{Var}[T(\overline{x : y})] = 2 \int_0^{\infty} t {}_t p_{\overline{x:y}} dt - (\overset{\circ}{e}_{\overline{x:y}})^2.$$

Also

$$\mathbf{Cov}[T(x : y), T(\overline{x : y})] = \mathbf{Cov}[T(x), T(y)] + (\overset{\circ}{e}_x - \overset{\circ}{e}_{x:y})(\overset{\circ}{e}_y - \overset{\circ}{e}_{x:y}).$$

Proof.

The formulas for the variances are straight forward. As to the covariance:

Proof.

$$\begin{aligned} & \mathbf{Cov}[T(x : y), T(\overline{x : y})] \\ &= \mathbf{E}[T(x : y)T(\overline{x : y})] - \mathbf{E}[T(x : y)]\mathbf{E}[T(\overline{x : y})] \\ &= \mathbf{E}[T(x)T(y)] - \mathbf{E}[T(x : y)]\mathbf{E}[T(\overline{x : y})] \\ &= \mathbf{E}[T(x)T(y)] \\ &\quad - \mathbf{E}[T(x : y)](\mathbf{E}[T(x)] + \mathbf{E}[T(y)] - \mathbf{E}[T(x : y)]) . \end{aligned}$$

Further, adding and subtracting $\mathbf{E}[T(x)]\mathbf{E}[T(y)]$, we obtain

$$\begin{aligned} & \mathbf{Cov}[T(x : y), T(\overline{x : y})] \\ &= \mathbf{Cov}[T(x), T(y)] \\ &\quad + (\mathbf{E}[T(x)] - \mathbf{E}[T(x : y)])(\mathbf{E}[T(y)] - \mathbf{E}[T(x : y)]) , \end{aligned}$$

which completes the proof. □

Corollary 0.5

If $T(x)$ and $T(y)$ are uncorrelated, then

$$\mathbf{Cov}[T(x : y), T(\overline{x : y})] = (\overset{\circ}{e}_x - \overset{\circ}{e}_{x:y})(\overset{\circ}{e}_y - \overset{\circ}{e}_{x:y}).$$