Influence function calculation for Brier score for event time data

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IF Calculation

To describe the situation with competing risks (and also survival) we introduce a random variable $D \in \{1, 2\}$ which indicates the cause (i.e., type of the event) observed at time T such that D=1 means that the event of interest occurred, and D=2 that a competing risk occurred. As in the survival setting we let Q denote the joint probability measure of the uncensored data, $(T,D,X) \sim Q$, and P the joint probability measure of the right censored data $Z=(\tilde{T},\Delta,X) \sim P$ now with $\Delta=D1_{\{T\leqslant C\}}$ taking values in the set $\{0,1,2\}$. We are interested in the following definition of the time-dependent discrimination measure for cause 1. We can easily calculate the influence function for the Brier score, which can be written as:

$$\int \left\{ 1_{\{t \le \tau\}} - R(\tau \mid x) \right\}^2 Q(dt, dx)
= \int 1_{\{t \le \tau\}} - 21_{\{t \le \tau\}} R(\tau \mid x) + R(\tau \mid x)^2 Q(dt, dx)
= \int 1_{\{t \le \tau\}} (1 - 2R(\tau \mid x)) Q(dt, dx) + \int R(\tau \mid x)^2 Q(dx)
= \int \frac{1_{\{t \le \tau\}}}{G(t - |x)} (1 - 2R(\tau \mid x)) P(dt, 1, dx) + \int R(\tau \mid x)^2 P(dx)$$

We find

$$\begin{split} IC_{\text{Brier}}(\tilde{T}_{i}, \Delta_{i}, X_{i}; \tau) &= \partial_{\varepsilon} \int \frac{1_{\{t \leqslant \tau\}}}{G(t - |x)} (1 - 2R(\tau \mid x)) P_{\varepsilon}(dt, 1, dx) + \partial_{\varepsilon} \int R(\tau | x)^{2} P_{\varepsilon}(dx) \\ &= \int 1_{\{t \leqslant \tau\}} (1 - 2R(\tau \mid x)) \frac{d(\delta_{\{\tilde{T}_{i}, \Delta_{i}, X_{i}\}})(t, 1, x) + dP(t, 1, x) \left[f_{i}(t -, x) - 1\right]}{G(t - |x)} \\ &+ \int R(\tau | x)^{2} \left(\delta_{\{\tilde{T}_{i}, \Delta_{i}, X_{i}\}} - P\right) \\ &= 1_{\{\tilde{T}_{i} \leqslant \tau, \Delta_{i} = 1\}} (1 - 2R(\tau \mid X_{i})) \frac{1}{G(\tilde{T}_{i} - |X_{i})} \\ &+ \int 1_{\{t \leqslant \tau\}} (1 - 2R(\tau \mid x)) f_{i}(t -, x) \frac{dP(t, 1, x)}{G(t - |x)} + R(\tau | X_{i})^{2} \\ &- \int R(\tau | x)^{2} dP(x) - \int 1_{\{t \leqslant \tau\}} (1 - 2R(\tau \mid x)) \frac{dP(t, 1, x)}{G(t - |x)} \end{split}$$

The last term that is subtracted is the Brier score. Then using that

$$f_{i}(t,x) = \frac{\mathbb{1}_{\{\tilde{T}_{i} \leq t, \Delta_{i} = 0\}} \delta_{X_{i}}(x)}{G(\tilde{T}_{i}|X_{i})S(\tilde{T}_{i}|X_{i})} - \int_{0}^{\tilde{T}_{i} \wedge t} \frac{\delta_{X_{i}}(x)dP(s,0|x)}{G(s|X_{i})^{2}S(s|X_{i})^{2}}$$

we see that (plugging in f(t,x) instead of f(t-,x)!)

$$\begin{split} \int_{X} \int_{0}^{\tau} (1 - 2R(\tau|x)) f_{i}(t - , x) \frac{P(dt, 1, x)}{G(t - |x)} &= \int_{X} \int_{0}^{\tau} (1 - 2R(\tau|x)) \left[\frac{\mathbbm{1}_{\{\tilde{T}_{i} \leq t, \Delta_{i} = 0\}} \delta_{X_{i}}(x)}{G(\tilde{T}_{i}|X_{i}) S(\tilde{T}_{i}|X_{i})} - \int_{0}^{\tilde{T}_{i} \wedge t} \frac{\delta_{X_{i}}(x) P(ds, 0|x)}{G(s|X_{i})^{2} S(s|X_{i})^{2}} \right] \frac{P(dt, 1, x)}{G(t - |x)} \\ &= \int_{0}^{\tau} (1 - 2R(\tau|X_{i})) \left[\frac{\mathbbm{1}_{\{\tilde{T}_{i} \leq t, \Delta_{i} = 0\}}}{G(\tilde{T}_{i}|X_{i}) S(\tilde{T}_{i}|X_{i})} - \int_{0}^{\tilde{T}_{i} \wedge t} \frac{P(ds, 0|X_{i})}{G(s|X_{i})^{2} S(s|X_{i})^{2}} \right] \frac{P(dt, 1|X_{i})}{G(t - |X_{i})} \\ &= (i) - (ii) \end{split}$$

where

$$\begin{split} (i) &= \int_0^\tau (1 - 2R(\tau|X_i)) \frac{\mathbbm{1}_{\{\tilde{T}_i \leqslant t, \Delta_i = 0\}}}{G(\tilde{T}_i|X_i)S(\tilde{T}_i|X_i)} \frac{P(dt, 1|X_i)}{G(t - |X_i)} \\ &= (1 - 2R(\tau|X_i)) \frac{\mathbbm{1}_{\{\tilde{T}_i \leqslant \tau, \Delta_i = 0\}}}{G(\tilde{T}_i|X_i)S(\tilde{T}_i|X_i)} \int_{\tilde{T}_i}^\tau \frac{P(dt, 1|X_i)}{G(t - |X_i)} \\ &= (1 - 2R(\tau|X_i)) \frac{\mathbbm{1}_{\{\tilde{T}_i \leqslant \tau, \Delta_i = 0\}}}{G(\tilde{T}_i|X_i)S(\tilde{T}_i|X_i)} (F_1(\tau|X_i) - F_1(\tilde{T}_i|X_i)) \end{split}$$

Similarly, we have

$$(ii) = (1 - 2R(\tau|X_i)) \int_0^\tau \int_0^{\tilde{T}_i \wedge t} \frac{P(ds, 0|X_i)}{G(s|X_i)^2 S(s|X_i)^2} \frac{P(dt, 1|X_i)}{G(t - |X_i)}$$

If $\tilde{T}_i > \tau$, then this can be written as

$$\begin{split} (ii) &= (1 - 2R(\tau|X_i)) \int_0^\tau \int_s^\tau \frac{P(dt, 1|X_i)}{G(t - |X_i)} \frac{P(ds, 0|X_i)}{G(s|X_i)^2 S(s|X_i)^2} \\ &= (1 - 2R(\tau|X_i)) \int_0^\tau (F_1(\tau|X_i) - F_1(s|X_i)) \frac{P(ds, 0|X_i)}{G(s|X_i)^2 S(s|X_i)^2} \end{split}$$

On the other hand, if $\tilde{T}_i \leq \tau$, then

$$\begin{split} (ii) &= (1 - 2R(\tau|X_i)) \int_0^{\tilde{T}_i} \int_s^\tau \frac{P(dt, 1|X_i)}{G(t - |X_i)} \frac{P(ds, 0|X_i)}{G(s|X_i)^2 S(s|X_i)^2} \\ &= (1 - 2R(\tau|X_i)) \int_0^{\tilde{T}_i} (F_1(\tau|X_i) - F_1(s|X_i)) \frac{P(ds, 0|X_i)}{G(s|X_i)^2 S(s|X_i)^2} \end{split}$$

Thus

$$(ii) = (1 - 2R(\tau|X_i)) \int_0^{\tilde{T}_i \wedge \tau} (F_1(\tau|X_i) - F_1(s|X_i)) \frac{P(ds, 0|X_i)}{G(s|X_i)^2 S(s|X_i)^2}$$

Hence,

$$(i) - (ii) = (1 - 2R(\tau|X_i)) \left(\frac{I(\tilde{T}_i \leqslant \tau, \Delta_i = 0)}{G(\tilde{T}_i|X_i)S(\tilde{T}_i|X_i)} (F_1(\tau|X_i) - F_1(\tilde{T}_i|X_i)) - \int_0^{\tilde{T}_i \wedge \tau} \frac{(F_1(\tau|X_i) - F_1(s|X_i))}{G(s|X_i)^2 S(s|X_i)^2} P(ds, 0|X_i) \right)$$