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A General Surplus Decomposition Principle in Life Insurance

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Joint work with Marcus C. Christiansen

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- Overarching decomposition principle is missing

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Goal: Find adapted processes D_1, \ldots, D_m that start at zero with

$$R(t)-R(0)=D_1(t)+\cdots+D_m(t)$$

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The random vector $D(t) = (D_1(t), \dots, D_m(t))$ defined by

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Transition to a sequence of partitions with vanishing step lengths

• $(\mathcal{T}_n)_n$ sequence of partitions on [0, t] with $\lim_{n \to \infty} \max_l |t_l^n - t_{l-1}^n| = 0$

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The random vector $D(t) = (D_1(t), \dots, D_m(t))$ that satisfies

$$D_i(t) = \lim_{n \to \infty} D_i^n(t)$$

is called ISU (infinitesimal sequential updating) decomposition of $R(t) = \varrho(X^t)$ with respect to $(\mathcal{T}_n)_n$.

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- Insurance cash flow B with

$$\mathrm{d}B(t) = \sum_{j} I_j(t-) \,\mathrm{d}B_j(t) + \sum_{jk:j \neq k} b_{jk}(t) \,\mathrm{d}N_{jk}(t)$$

Individual surplus (cf. Norberg, 1999)

$$R(t) = -\int_{[0,t]} rac{1}{\kappa(s)} \mathrm{d}B(s) - \sum_j rac{1}{\kappa(t)} I_j(t) V_j^*(t),$$

where

$$\mathbb{E}^*\left[\int_t^T \frac{\kappa^*(t)}{\kappa^*(s)} \mathrm{d}B(s) \middle| Z(t) = j\right] \text{ and } \mathrm{d}\kappa(t) = \kappa(t-) \mathrm{d}\Phi(t), \ \kappa(0) = 1.$$

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Proposition: It holds

$$R(t) = -H((\Phi^*, \Lambda^*) + (\Phi - \Phi^*, N - \Lambda^*)^t),$$

where for any valuation basis $(\overline{\Phi}, \overline{\Lambda})$ the mapping H is defined by

$$\begin{split} H((\overline{\Phi},\overline{\Lambda})) &\coloneqq \sum_{j} \int_{[0,T]} \frac{1}{\overline{\kappa}(s)} \overline{p}_{aj}(0,s-) \mathrm{d}B_{j}(s) \\ &+ \sum_{j,k:j \neq k} \int_{(0,T]} \frac{1}{\overline{\kappa}(s)} \overline{p}_{aj}(0,s-) b_{jk}(s) d\overline{\Lambda}_{jk}(s) \end{split}$$

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In case 2), we obtain the ISU decomposition

$$\begin{split} D_u(t) &= -\sum_{jk:j \neq k} \int_{(0,t]} \frac{1}{\kappa(s)} I_j(s-) R_{jk}^*(s) \mathrm{d}(N_{jk} - \Lambda_{jk})(s), \\ D_j(t) &= \int_{(0,t]} \frac{1}{\kappa(s-)} I_j(s-) \Big(V_j^*(s-) \mathrm{d}(\widetilde{\Phi} - \Phi^*)(s) - \sum_{k:k \neq j} R_{jk}^*(s) \mathrm{d}(\Lambda_{jk} - \Lambda_{jk}^*)(s) \Big), \end{split}$$

where $\widetilde{\Phi}(t) = \Phi(t) - [\Phi, \Phi]^c(t) - \sum_{0 < s \le t} (1 + \Delta \Phi(s))^{-1} (\Delta \Phi(s))^2$.

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Splitting the financial risk into an unsystematic and a systematic part, one can replicate the surplus formula of Asmussen & Steffensen (2020).

The mean portfolio (revaluation) surplus is given by (cf. Norberg, 1999)

$$\overline{R}(t) = \mathbb{E}\bigg[-\int_{[0,t]} rac{1}{\kappa(s)} \mathrm{d}B(s) - \sum_j rac{1}{\kappa(t)} I_j(t) V_j^*(t) \bigg| \Phi, \Lambda\bigg],$$

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ISU decomposition yields again traditional decomposition formulas by Ramlau-Hansen (1991) and Norberg (1999).

Conclusion

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THANK YOU FOR YOUR ATTENTION!

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Alternative decomposition principles - One-at-a-time

The random vector $D(t) = (D_1(t), \dots, D_m(t), \overline{D}(t))$ defined by

$$D_1(t) = \sum_{t_l,t_{l+1}\in\mathcal{T}} \Big(U(t_{l+1},t_l,\ldots,t_l) - U(t_l,t_l,\ldots,t_l) \Big),$$

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is called the OAT (one-at-a-time) decomposition of R(t) - R(0) w.r.t. T. (cf. Biewen, 2014)

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Theorem: If the ISU decomposition is independent of update order, then ISU (for each update order), IOAT and averaged ISU yield the same decomposition.