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# Two-step actuarial valuations

Karim Barigou\* Daniël Linders† Fan Yang‡

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#### **Abstract**

We introduce the class of actuarial-consistent valuation methods for insurance liabilities which depend on both financial and actuarial risks, which imposes that all actuarial risks are priced via standard actuarial principles. We propose to extend standard actuarial principles by a new actuarial-consistent procedure, which we call "two-step actuarial valuations". In the coherent setting, we show that actuarial-consistent valuations are equivalent to two-step actuarial valuations. We also discuss the connection with "two-step market-consistent valuations" from Pelsser and Stadje (2014). In particular, we discuss how the dependence structure between actuarial and financial risks impacts both actuarial-consistent and market-consistent valuations.

**Keywords:** Fair valuation, two-step valuation, actuarial consistent, market consistent, Solvency II, incomplete market.

## 1 Introduction

Insurance liabilities, such as variable annuities, are complex combinations of different types of risks. Motivated by solvency regulations, the recent focus has been towards financial risks and the so-called market-consistent valuations. In this situation, the financial market is the main driver and actuarial risks only appear as the "second step". This paper goes against the tide and introduces the concept of *actuarial-consistent* valuations where actuarial risks are at the core of the valuation. We propose a two-step actuarial valuation that is first driven by actuarial information and is actuarial-consistent. Moreover, we show that actuarial-consistent valuations can always be expressed as the price of an appropriate hedging strategy.

Fundamental to insurance, actuarial valuation is typically based on a diversification argument which justifies applying the law of large numbers (LLN) among independent policyholders who face identical risks. For this reason, the actuarial valuation  $\rho(S)$  of a claim S is defined as the expectation under the real-world probability measure  $\mathbb P$  plus an additional risk margin to cover any undiversified and systematic risk, that is

$$\rho(S) = \mathbb{E}^{\mathbb{P}}[S] + \text{Risk margin.} \tag{1.1}$$

Financial valuation, on the other hand, is based on the no-arbitrage principal. Given the prices of the available traded assets, the value of a financial claim should be determined such that the market remains

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free of arbitrage if the claim is traded. Therefore, financial pricing is based on the idea of hedging and replication. It was shown in Delbaen and Schachermayer (2006) that no-arbitrage pricing implies that the prices of contingent claims can be expressed as expectations under a so-called risk-neutral measure  $\mathbb{Q}$ , that is

$$\rho(S) = \mathbb{E}^{\mathbb{Q}}[S]. \tag{1.2}$$

This approach dates back to the seminal paper of Black and Scholes (1973).

Insurance claims are nowadays non-trivial combinations of diversifiable and traded risks. It is therefore primordial to build a valuation framework which combines traditional actuarial and financial valuation. A simplifying approach in the actuarial literature is to assume independence between actuarial and financial risks such that the valuation can be split into a product of actuarial and financial valuations (see Fung et al., 2014; Da Fonseca and Zivevi, 2017; Ignatieva et al., 2016; Wüthrich, 2016 among others). However, the emergence of longevity-linked financial products and the pandemic situation showed us that future mortality cannot be assumed independent from evolution of the financial market. For this reason, different authors proposed general valuation approaches allowing for dependencies between actuarial and financial risks. For example, Pelsser and Stadje (2014) proposed a 'two-step market-consistent valuation' which extends standard actuarial principles by conditioning on the financial information. Dhaene et al. (2017) proposed a new framework for the fair valuation of insurance liabilities in a one-period setting; see also Dhaene (2020). The authors introduced the notion of a 'fair valuation', which they defined as a valuation which is both market-consistent (mark-to-market for any hedgeable part of a claim) and actuarial (mark-to-model for any claim that is independent of financial market evolution). This work was further extended in a multi-period discrete setting in Barigou and Dhaene (2019) and in continuous time in Delong et al. (2019). A 3-step valuation was introduced in Deelstra et al. (2020) for the valuation of claims which consists of traded, financial but also systematic risks. This approach was further generalized in Linders (2021).

The above-mentioned papers propose different valuation principles which have in common that they are all market-consistent valuations. Nowadays, several insurance regulation frameworks (e.g. Solvency II) require a market-consistent valuation of the insurance liabilities; see e.g. Möhr (2011). A market-consistent valuation assumes an investment in an appropriate replicating portfolio to offset the hedgeable part of the liability. The remaining part of the claim is managed by diversification and an appropriate capital buffer. However, Vedani et al. (2017) and Rae et al. (2018) raise some concerns about the market-consistent framework as this valuation follows market movements and can lead to excess volatility of the insurance liabilities. Moreover, market-consistency tends to be pro-cyclical and the use of a 1-year Value-at-Risk increases the risk of herd behaviour, hence reducing financial stability (Rae et al., 2018). In this direction, Le Courtois et al. (2021) replaced the market-consistent framework by a utility-consistent framework that accounts for the risk aversion of the market and the long-term nature of liabilities.

In this paper, we introduce the class of actuarial-consistent valuations for hybrid claims as an alternative for the market-consistent valuations. This new valuation framework is motivated by the requirement that any pure actuarial claim should be priced via a pure actuarial valuation and should not be managed using a risky investment. We will label this property of the valuation "actuarial-consistency". We also introduce the two-step actuarial valuations. Instead of first considering the hedgeable part of the claim, the two-step actuarial valuation will first price the actuarial part of the claim using an actuarial valuation. We show that every two-step actuarial valuation is actuarial-consistent. Moreover, in the coherent setting, we show the reciprocal: any actuarial-consistent valuation has a two-step actuarial representation.

<sup>&</sup>lt;sup>1</sup>This assumption is either made under the real-world measure  $\mathbb{P}$  or the risk-neutral measure  $\mathbb{Q}$ . We note that the independence under  $\mathbb{P}$  does not necessarily imply the independence under  $\mathbb{Q}$ ; see Dhaene et al. (2013). We also discuss this point in Lemma 4.2.

The two-step valuations are general in the sense that they do not impose linearity constraints on the actuarial and financial valuations. Therefore, they allow to account for the diversification of actuarial risks and/or the incompleteness of the financial market (e.g. non-linear financial pricing with bid-ask prices).

Hedge-based valuations were first introduced in Dhaene et al. (2017) to define the market-consistent valuations. We show that actuarial-consistent valuations can always be expressed as hedge-based valuations. The hedging strategy used in an actuarial-consistent valuation will only invest in the risk-free asset when valuating an actuarial claim. This is in contrast with the market-consistent hedge-based valuations, which may use the financial market to hedge actuarial claims.

The paper also provides a detailed comparison between two-step market and two-step actuarial valuations. We discuss how the dependence structure between actuarial and financial claims impacts both actuarial and market-consistent valuations. In the context of solvency regulations, we show how the two-step actuarial valuation can be decomposed into a best estimate (expected value) plus a risk margin to cover the uncertainty in the actuarial risks. The procedure will be illustrated with a portfolio of life insurance contracts with dependent financial and actuarial risks.

The rest of the paper is structured as follows. In Section 2, we describe financial and actuarial valuations. Section 3 discusses the notion of actuarial-consistency and introduces two-step actuarial valuations. In Section 4, we provide a detailed comparison between actuarial-consistent valuations and market-consistent valuations. Section 5 presents a detailed numerical application of the two-step actuarial valuation on a portfolio of equity-linked contracts. Section 6 concludes the paper.

## 2 Actuarial and financial valuations

All random variables introduced hereafter are defined on the filtered probability space  $(\Omega, \mathcal{F}, \mathbb{F}, \mathbb{P})$ . Equalities and inequalities between r.v.'s have to be understood in the  $\mathbb{P}$ -almost sure sense. The space  $L^2(\mathbb{P}, \mathbb{F})$  of square integrable random variables is denoted by  $\mathcal{C}$ . A contingent claim is a random liability of an insurance company that has to be paid at the deterministic future time T. Formally, a contingent claim is modeled by the random variable  $S \in \mathcal{C}$ . In what follows we are interested in the *valuation* of contingent claims.

**Definition 2.1 (Valuation)** A valuation is a mapping  $\Pi: \mathcal{C} \longrightarrow \mathbb{R}$  satisfying the following properties:

- Normalization:  $\Pi[0] = 0$ .
- Translation-invariance: For any  $S \in \mathcal{C}$  and  $a \in \mathbb{R}$ ,  $\Pi[S+a] = \Pi[S] + a$ .

A valuation  $\Pi$  attaches a real number to any claim, which we interpret as a 'value' of that claim. Later in the paper, several representation theorems can be derived under the additional assumption that the valuation is coherent. The concepts of linear and coherent valuations are therefore introduced in the next section.

#### 2.1 Linear and coherent valuations

We start by proceeding as in Buhlmann et al. (1992), Bühlmann (2000) and Wüthrich (2016) via the use of a linear valuation on the set of contingent claims  $S \in L^2(\mathbb{P}, \mathbb{F})$ .

**Definition 2.2 (Linear valuation)** A mapping  $\Pi: \mathcal{C} \longrightarrow \mathbb{R}$  is a linear valuation if for any  $S_1, S_2 \in \mathcal{C}$  and  $a, b \in \mathbb{R}$ , we have that

$$\Pi[aS_1 + bS_2] = a\Pi[S_1] + b\Pi[S_2]. \tag{2.1}$$

Since the linear case is too restrictive in an actuarial context, we also introduce the class of coherent valuations.

Coherent valuations can always be represented as upper expectations over a convex set of test probability measures  $\widetilde{\mathbb{P}}$ , different from the physical probability measure  $\mathbb{P}$ , such that the density function  $\varphi = \frac{d\widetilde{\mathbb{P}}}{d\mathbb{P}}$  is well-defined. The set  $\mathcal{P}$  is the set of all the density functions

$$\mathcal{P} = \{ \varphi \in L^2(\mathbb{P}, \mathbb{F}) \mid \varphi \ge 0, \mathbb{E}^{\mathbb{P}} [\varphi] = 1 \}.$$

**Definition 2.3 (Coherent valuation)** A mapping  $\Pi : \mathcal{C} \longrightarrow \mathbb{R}$  is a coherent valuation<sup>2</sup> if

$$\Pi[S] = \sup_{\varphi \in \mathcal{Q}} \mathbb{E}\left[\varphi S\right],$$

where Q is a unique, non-empty, closed convex subset of P.

A coherent valuation can then be understood as a worst-case expectation with respect to some class of probability measures. This can be motivated by the desire for robustness: the valuator does not only want to rely on a single measure  $\mathbb{P}$  for the occurrence of future events but prefers to test a set of plausible measures and value with the worst-case scenario. We also note that the set of linear valuations is a subset of the set of coherent valuations. One can show that the set of coherent valuations is positive homogeneous, translation-invariant, subadditive and monotone, see e.g. Artzner et al. (1999).

In the following subsections, we will consider two important types of claims, namely financial and actuarial claims, and their respective valuations.

#### 2.2 Financial and actuarial valuations

A claim  $S \in \mathcal{C}$  is a combination of different types of risks. In this paper, we assume risks can be divided in two groups: financial and actuarial risks. The financial risks are traded on a public exchange and market participants can buy and sell these financial risks at any quantity. The non-traded risks are referred to as actuarial risks. We note that a similar split was considered in Wüthrich (2016) between financial events (stocks, asset portfolio, inflation-protected bonds, etc) and insurance technical events (death, car accident, medical expenses, etc). Moreover, we remark that the issue of insurance-linked securities (e.g. longevity bonds) implies that a non-traded actuarial risk may become a traded financial risk.

## 2.2.1 Financial Valuation

We assume there is a financial market with  $n^{(1)}+1$  traded assets. The payoff of asset i at time t is denoted by  $Y_i(t)$ . We also refer to the traded assets as financial risks. We denote by  $\mathbf{Y}=\{(Y(t))\}_{0\leq t\leq T}$  the price process where  $Y(t)=(Y_0(t),Y_1(t),\ldots,Y_{n^{(1)}}(t))$  is the vector of the financial risks at time t. The payoff  $Y_0(t)$  denotes the payoff of the risk-free bank account at time t and we assume there is a

<sup>&</sup>lt;sup>2</sup>In the literature, it is also referred as a *coherent risk measure*.

constant and deterministic risk-free rate r.<sup>3</sup> The financial filtration is given by  $\mathbb{F}^{(1)} = \left\{ \mathcal{F}_t^{(1)} \right\}_{t \in [0,T]}$  with  $\mathcal{F}_T^{(1)} = \mathcal{F}^{(1)}$ . We assume  $\mathcal{F}^{(1)}$  is the  $\sigma$ -algebra generated by the financial risks, i.e.  $\mathcal{F}^{(1)} = \sigma\left(\boldsymbol{Y}\right)$ . The financial probability space is denoted by  $\left(\Omega, \mathcal{F}^{(1)}, \mathbb{F}^{(1)}, \mathbb{P}^{(1)}\right)$ .

A financial claim with maturity T is an  $\mathcal{F}_T^{(1)}$ -measurable random variable defined on the financial probability space. Otherwise stated, a financial claim only depends on the financial risks Y and its realization is completely known given the realization of the financial risks Y. The set of all financial claims is denoted by  $\mathcal{C}^{(1)}$ . We can always express a financial claim as a function of the financial risks. We have:

$$S^{(1)} = f(\mathbf{Y}), \text{ if } S^{(1)} \in \mathcal{C}^{(1)},$$
 (2.2)

for some function  $f^4$ .

The financial risks are traded on the financial market and all market participants can observe the prices at which each asset can be bought and sold. Note, however, that we do not assume that the price at which one can buy and sell at time t is equal. We assume that a financial valuation principle  $\pi^{(1)}$  is available to determine the price at which one can buy the traded payoffs:

$$\pi^{(1)}: \mathcal{C}^{(1)} \to \mathbb{R}.$$

The value  $\pi^{(1)}\left[S^{(1)}\right]$  has to be understood as the price one has to pay at time t=0 to receive the financial claim which pays the random amount  $S^{(1)}$  at maturity  $T^{5}$ .

The choice of the financial valuation principle  $\pi^{(1)}$  depends on the additional assumptions we impose on the financial market. Traditional financial pricing assumes that the market is complete and arbitrage-free such that the pricing rule is linear and unique (Black and Scholes, 1973). However, the existence of transaction costs and non-hedgeable payoffs lead to non-linear and non-unique valuations. In such situations, the market will decide which valuation principle is used and one has to use calibration to back out the valuation principle from the available traded assets. Below, we consider different possible choices for the financial valuation for different market situations.

1. The law of one price. Assume the market is arbitrage-free and that all financial assets are traded in the market and can be bought and sold at a unique price. One can prove that in this market setting, the no-arbitrage condition is equivalent with the existence of an equivalent martingale measure (EMM) Q. In this financial market where one can buy and sell any asset at a unique price, the financial valuation principle can be determined as follows:

$$\pi^{(1)}[S] = e^{-rT} \mathbb{E}^{\mathbb{Q}}[S].$$

The financial valuation is in this situation a linear valuation principle.

2. Imperfect market and bid-ask prices. In classical finance, markets are usually modelled as a counterparty for market participants. It is assumed that markets can accept any amount and direction of the trade (buy or sell) at the going market price. However, due to market imperfection, there is in practice a difference between the price the market is willing to buy (bid price) and the price the market is willing to sell (ask price). This difference, called the bid-ask spread, creates a two-price

<sup>&</sup>lt;sup>3</sup>For simplicity of presentation but our main results can be easily extended to stochastic interest rates.

<sup>&</sup>lt;sup>4</sup>We assume that all functions we encounter are Borel measurable.

 $<sup>^{5}</sup>$ This paper considers only the valuation at time 0. The dynamic setting in which one considers the time-t value (similar to Barigou et al. (2019)) is left for future research.

economy. In particular, the value  $\pi^{(1)}[S]$  which corresponds with the price required by the market to take over the financial claim S will typically be higher than the risk-neutral price. Indeed, the asymmetry in the market allows that market to take a more prudent approach when determining the price  $\pi^{(1)}[S]$ . Instead of using a single risk-neutral probability measure, a set of "stress-test measures" is selected from the set of martingale measures and the price is determined as the supremum of the expectations w.r.t. the stress-test measures:

$$\pi^{(1)}[S] = e^{-rT} \sup_{\mathbb{Q} \in \mathcal{Q}} \mathbb{E}^{\mathbb{Q}}[S].$$

For more details on conic finance, we refer to Madan and Cherny (2010) and Madan and Schoutens (2016). We remark that in this case, the valuation of financial claims is *coherent*, but *non-linear*.

In the remainder of the paper, we consider coherent financial valuations to account for bid-ask spread and market imperfections.

#### 2.2.2 Actuarial valuation

There are  $n^{(2)}$  non-traded risks (also called actuarial risks) which we denote by  $\boldsymbol{X} = \{(X(t))\}_{0 \leq t \leq T}$  the price process where  $X(t) = (X_0(t), X_1(t), \dots, X_{n^{(2)}}(t))$  is the vector of the actuarial risks at time t. The actuarial world is described by the probability space  $\left(\Omega, \mathcal{F}^{(2)}, \mathbb{F}^{(2)}, \mathbb{P}^{(2)}\right)$ . The actuarial filtration is generated by the actuarial risks, i.e.  $\mathbb{F}^{(2)} = \left\{\mathcal{F}^{(2)}_t\right\}_{t \in [0,T]}$  with  $\mathcal{F}^{(2)}_T = \mathcal{F}^{(2)}$ .

An *actuarial claim* is a  $\mathcal{F}_T^{(2)}$ -measurable random variable defined on the actuarial probability space. Equivalently stated, an actuarial claim only depends on X and its realization is completely known given the realization of the actuarial risks X. If we denote the set of actuarial claims by  $\mathcal{C}^{(2)}$ , we can write any claim  $S^{(2)} \in \mathcal{C}^{(2)}$  as a function of the actuarial risks:

$$S^{(2)} = f(\mathbf{X}), \text{ if } S^{(2)} \in \mathcal{C}^{(2)},$$

for some function f.

We assume that a valuation principle  $\pi^{(2)}$  is chosen to price actuarial claims. The actuarial valuation principle  $\pi^{(2)}$  is based on the idea of pooling and diversification. The value of a completely diversifiable portfolio has to correspond with its expectation under the physical measure  $\mathbb{P}$ . However, there is always an amount of residual actuarial risk present because one can never fully diversify away all the risk. Moreover, there are also systematic actuarial risks (e.g. longevity risk) which cannot be diversified away. Below, we briefly discuss the most important actuarial valuation principles (also called actuarial premium principles).

#### 1. Linear valuation:

$$\pi^{(2)}[S] = \mathbb{E}_{\tilde{\mathbb{p}}}[S].$$

The risk margin is modelled by an appropriate change of measure from  $\mathbb{P}$  to  $\widetilde{\mathbb{P}}$ . In terms of life tables, the change of measure can be interpreted as a switch from the second order life table (best-estimate survival or death probabilities) to a first order life table (survival or death probabilities that are chosen with a safety margin). For more details, see for instance Wüthrich (2016) and Norberg (2014).

## 2. Standard deviation principle:

$$\pi^{(2)}[S] = \mathbb{E}^{\mathbb{P}}[S] + \beta \sqrt{\operatorname{Var}^{\mathbb{P}}[S]}, \tag{2.3}$$

with  $\beta > 0$ .

In this case, the loading equals  $\beta$  times the standard deviation. It is well-known that  $\beta > 0$  is required in order to avoid getting ruin with probability 1 (see e.g. Kaas et al. (2008)). We note that the standard deviation principle is neither linear nor coherent.

#### 3. Coherent valuation:

$$\pi^{(2)}[S] = \rho[S],$$

where  $\rho$  is a coherent valuation. We remark that the coherent valuation can also be expressed as

$$\pi^{(2)}[S] = \sup_{\mathbb{P} \in \mathcal{Q}} \mathbb{E}^{\mathbb{P}}[S].$$

Therefore, model risk can be taken into account by considering a family of different distributions and the actuarial claim is valuated with the most conservative one.

## 2.3 Hybrid claims

Recall that a claim  $S \in \mathcal{C}$  is defined on the probability space  $(\Omega, \mathcal{F}, \mathbb{F}, \mathbb{P})$ , which is a combination of the financial and the actuarial probability space. The filtration  $\mathbb{F}$  is then defined as the minimal  $\sigma$ -algebra containing all events of  $\mathbb{F}^{(1)}$  and  $\mathbb{F}^{(2)}$ , i.e.  $\mathbb{F} = \sigma\left(\mathbb{F}^{(1)} \cup \mathbb{F}^{(2)}\right)$ . The probability measure  $\mathbb{P}$  is such that it is consistent with the financial and the actuarial probability measures:

$$\mathbb{P}[A] = \mathbb{P}^{(i)}[A], \text{ if } A \in \mathcal{F}^{(i)}.$$

Assume the valuation  $\Pi: \mathcal{C} \to \mathbb{R}$  is used to valuate claims in  $\mathcal{C}$ . A general claim in  $\mathcal{C}$  can contain both financial and actuarial risks and therefore the valuation of claims in  $\mathcal{C}$  cannot be solely based on the financial or the actuarial valuation principles. Moreover, the financial and actuarial worlds are dependent. Therefore, observing the values of financial (resp. actuarial) claims can provide information about the valuation of actuarial (resp. financial) claims.

Define by  $C^{(1,\perp)}$  the set of financial claims which are independent of the actuarial risks and by  $C^{(2,\perp)}$  the actuarial claims which are independent of the financial information:

$$\begin{split} S^{(1,\perp)} &\in \mathcal{C}^{(1,\perp)} \quad \text{if} \quad S^{(1,\perp)} &\in \mathcal{C}^{(1)} \text{ and } S^{(1,\perp)} \perp \boldsymbol{X}, \\ S^{(2,\perp)} &\in \mathcal{C}^{(2,\perp)} \quad \text{if} \quad S^{(2,\perp)} &\in \mathcal{C}^{(2)} \text{ and } S^{(2,\perp)} \perp \boldsymbol{Y}. \end{split}$$

We say that  $S^{(1,\perp)}$  is a pure financial claim, whereas  $S^{(2,\perp)}$  is called a pure actuarial claim. A pure financial claim does not contain any information about the actuarial risks and therefore the actuarial valuation  $\pi^{(2)}$  should not be used to valuate a pure financial claim. Hence, the valuation of a pure financial claim should only involve the financial valuation  $\pi^{(1)}$ . Similarly, the valuation of a pure actuarial claim should only be based on the actuarial valuation  $\pi^{(2)}$ . We require that the valuation principle  $\Pi$  is consistent with the financial valuation  $\pi^{(1)}$  and the actuarial principle  $\pi^{(2)}$  in that  $\Pi$  should correspond with financial valuation  $\pi^{(1)}$  for pure financial claims and with the actuarial valuation when considering pure actuarial claims.

**Definition 2.4 (Orthogonal-consistency)** A valuation  $\Pi: \mathcal{C} \to \mathbb{R}$  is said to be orthogonal consistent with the financial valuation  $\pi^{(1)}$  and the actuarial valuation  $\pi^{(2)}$  if we have that for i = 1, 2,

$$\Pi\left[S^{(i,\perp)}\right] = \pi^{(i)}\left[S^{(i,\perp)}\right], \text{ if } S^{(i,\perp)} \in \mathcal{C}^{(i,\perp)}.$$
(2.4)

We will show later that the two-step actuarial valuation introduced in this paper is orthogonal consistent with the financial and actuarial valuations.

Similar to Dhaene et al. (2017), a hybrid claim is a claim which depends on both the actuarial as well as the financial information, i.e. a hybrid claim S can be expressed as follows:

$$S$$
 is a hybrid claim  $\leftrightarrow S \in \mathcal{C} \setminus \left(\mathcal{C}^{(1)} \cup \mathcal{C}^{(2)}\right)$  .

Different valuation frameworks can be considered depending on how financial and actuarial valuations are merged together. In Section 3, we propose a two-step valuation which applies a financial valuation after conditioning on actuarial information. For this reason, we briefly introduce the concept of conditional valuations hereafter.

## 2.4 Conditional valuations

We define the notion of  $\mathcal{F}^{(i)}$ -conditional valuation which maps any claim S into an  $\mathcal{F}^{(i)}$ -measurable r.v, for i=1,2. This operator allows us to transform a claim S into a financial or actuarial claim.

**Definition 2.5** ( $\mathcal{F}^{(i)}$ -conditional valuation)  $A \mathcal{F}^{(i)}$ -conditional valuation is a mapping  $\Pi : \mathcal{C} \longrightarrow \mathcal{C}^{(i)}$  satisfying the following properties:

- Normalization:  $\Pi[0|\mathcal{F}^{(i)}] = 0$ .
- <u>Translation-invariance</u>: For any  $S \in \mathcal{C}$  and  $S^{(i)} \in \mathcal{C}^{(i)}$ , we have

$$\Pi\left[S + S^{(i)}|\mathcal{F}^{(i)}\right] = \Pi[S] + S^{(i)}.$$

• Positive homogeneity: For any  $S \in \mathcal{C}$  and any positive  $S^{(i)} \in \mathcal{C}^{(i)}$ , we have

$$\Pi \left[ S \times S^{(i)} | \mathcal{F}^{(i)} \right] = S^{(i)} \times \Pi \left[ S | \mathcal{F}^{(i)} \right]$$

**Remark 2.1** We remark that a  $\mathcal{F}^{(1)}$ -conditional valuation attaches the financial claim  $\Pi[S|\mathcal{F}^{(1)}]$  to any claim S. Since  $\mathbb{F}^{(1)}$  is the filtration generated by the price process Y, conditioning on  $\mathcal{F}_T^{(1)} = \mathcal{F}^{(1)}$  is equivalent to conditioning on the stochastic process Y(t) from time 0 to time T. Similarly, a  $\mathcal{F}^{(2)}$ -conditional valuation  $\Pi[S|\mathcal{F}^{(2)}]$  can be interpreted as conditioning on the actuarial risks X(t) from time 0 to time T.

## 3 Actuarial-consistency and two-step actuarial valuations

Given a financial valuation principle  $\pi^{(1)}$  and an actuarial valuation principle  $\pi^{(2)}$ , we search for general valuations  $\Pi$  that are consistent with both the financial valuation  $\pi^{(1)}$  and the actuarial valuation  $\pi^{(2)}$ , and study their properties.

#### 3.1 Actuarial-consistent valuations

This section starts with the concept of *actuarial-consistency* and the valuation of actuarial claims. Condition (2.4) for a valuation  $\Pi$  states that actuarial claims which are independent of the financial information, should be valued using the actuarial valuation principle  $\pi^{(2)}$ . A valuation  $\Pi$  is said actuarial-consistent if it ensures that all actuarial claims are priced with an actuarial valuation, even the ones that may be dependent to financial information. Before going to the valuation, we first introduce the notion of weak and strong actuarial consistency.

**Definition 3.1 (Strong actuarial-consistency)** A valuation  $\Pi$  is called strong actuarial-consistent (strong ACV) if for any actuarial claim  $S^{(2)} \in C^{(2)}$  the following holds:

$$\Pi \left[ S + S^{(2)} \right] = \Pi \left[ S \right] + \pi^{(2)} \left[ S^{(2)} \right].$$
 (3.1)

Strong actuarial-consistency implies linearity of the actuarial valuation principle (this issue was also discussed in Dhaene et al. (2017)). It would not be appropriate in an actuarial context, in which diversification benefits should be accounted for, to limit the actuarial valuation principles to linear valuations. Therefore, we define the notion of weak actuarial-consistency.

**Definition 3.2 (Weak actuarial-consistency)** A valuation  $\Pi$  is called weak actuarial-consistent (weak ACV) if for any actuarial claim  $S^{(2)} \in \mathcal{C}^{(2)}$  the following holds:

$$\Pi \left[ S^{(2)} \right] = \pi^{(2)} \left[ S^{(2)} \right].$$
 (3.2)

Weak actuarial-consistency only postulates that an actuarial valuation is applied for all actuarial claims. Note that weak actuarial-consistency is stronger than condition (2.4), which only states that independent actuarial claims are priced using the actuarial valuation. In Dhaene et al. (2017), the authors define a similar notion of actuarial-consistency, but the condition only holds for the claims which are independent of the financial filtration  $\mathbb{F}^{(1)}$ .

It is straightforward to show that strong ACV implies weak ACV. Following discussions from Pelsser and Stadje (2014), the following lemma proves that in case the actuarial valuation principle is linear and the valuation is coherent, weak ACV implies strong ACV.

**Lemma 3.1** Consider a coherent valuation  $\Pi$  with actuarial valuation  $\pi^{(2)}$  which is linear in the sense that

$$\pi^{(2)}\left[S^{(2)}\right] = \mathbb{E}\left[\varphi^{(2)}S^{(2)}\right],$$

for all actuarial claims  $S^{(2)}$ , with a positive  $\mathcal{F}^{(2)}$ -measurable density  $\varphi^{(2)}$ . Then, the following statements are equivalent:

- 1.  $\Pi$  is strong ACV.
- 2.  $\Pi$  is weak ACV.

**Proof.** The proof of  $(1) \to (2)$  is straightforward. In order to prove  $(2) \to (1)$ , we remark that since  $\Pi$  is coherent, we can write:

$$\Pi[S] = \sup_{\varphi \in \mathcal{Q}} \mathbb{E}\left[\varphi S\right]. \tag{3.3}$$

Weak actuarial consistency and linearity of the actuarial valuation principle  $\pi^{(2)}$  implies that for all actuarial claims  $S^{(2)} \in \mathcal{C}^{(2)}$ , we have that

$$\pi^{(2)} \left[ S^{(2)} \right] = \mathbb{E} \left[ \varphi^{(2)} S^{(2)} \right]$$

$$= \sup_{\varphi \in \mathcal{Q}} \mathbb{E} \left[ \varphi S^{(2)} \right]$$

$$= \sup_{\varphi \in \mathcal{Q}} \mathbb{E} \left[ \mathbb{E} \left[ \varphi S^{(2)} | \mathcal{F}^{(2)} \right] \right]$$

$$= \sup_{\varphi \in \mathcal{Q}} \mathbb{E} \left[ S^{(2)} \mathbb{E} \left[ \varphi | \mathcal{F}^{(2)} \right] \right]$$
(3.4)
$$(3.4)$$

where we used that  $S^{(2)}$  is  $\mathcal{F}^{(2)}$ -measurable. Since (3.4)=(3.5), it is sufficient to consider  $\varphi \in \mathcal{Q}$ :  $\mathbb{E}\left[\varphi|\mathcal{F}^{(2)}\right] = \varphi^{(2)}$ . Because  $\varphi^{(2)}$  is positive, we can then write  $\varphi = \varphi^{(2)}Z$  with  $Z \in \mathcal{Q}_{(2)} = \{\varphi \in L^2(\mathbb{P},\mathbb{F}) \mid \mathbb{E}\left[Z|\mathcal{F}^{(2)}\right] = 1\}$ . Thus, we have that:

$$\begin{split} \Pi\left[S+S^{(2)}\right] &= \sup_{Z\in\mathcal{Q}_{(2)}} \mathbb{E}\left[\varphi^{(2)}Z\left(S+S^{(2)}\right)\right] \\ &= \sup_{Z\in\mathcal{Q}_{(2)}} \left\{ \mathbb{E}\left[\varphi^{(2)}ZS\right] + \mathbb{E}\left[\varphi^{(2)}ZS^{(2)}\right] \right\} \\ &= \sup_{Z\in\mathcal{Q}_{(2)}} \left\{ \mathbb{E}\left[\varphi^{(2)}ZS\right] + \mathbb{E}\left[\mathbb{E}\left[\varphi^{(2)}ZS^{(2)}|\mathcal{F}^{(2)}\right]\right] \right\} \\ &= \sup_{Z\in\mathcal{Q}_{(2)}} \left\{ \mathbb{E}\left[\varphi^{(2)}ZS\right] + \mathbb{E}\left[\varphi^{(2)}S^{(2)}\mathbb{E}\left[Z|\mathcal{F}^{(2)}\right]\right] \right\} \\ &= \sup_{Z\in\mathcal{Q}_{(2)}} \left\{ \mathbb{E}\left[\varphi^{(2)}ZS\right] + \mathbb{E}\left[\varphi^{(2)}S^{(2)}\right] \right\} \\ &= \sup_{Z\in\mathcal{Q}_{(2)}} \left\{ \mathbb{E}\left[\varphi^{(2)}ZS\right] + \pi^{(2)}\left[S^{(2)}\right] \right\} \\ &= \sup_{Z\in\mathcal{Q}_{(2)}} \left\{ \mathbb{E}\left[\varphi^{(2)}ZS\right] \right\} + \pi^{(2)}\left[S^{(2)}\right] \\ &= \Pi[S] + \pi^{(2)}\left[S^{(2)}\right], \end{split}$$

which proves that  $\Pi$  is also strong ACV.

## 3.2 Actuarial-consistent hedgers

After having defined actuarial-consistent valuations, we will now introduce the corresponding class of actuarial-consistent hedgers. In Dhaene et al. (2017), the authors showed that any market-consistent valuation can be represented as the time-0 value of a market-consistent hedger. Similarly, in this section, we establish the relationship between actuarial-consistent valuations and their corresponding hedgers.

A trading strategy  $\nu$  is a real-valued vector  $(\nu_0, \nu_1, \dots, \nu_{n^{(1)}})$  where the component  $\nu_i$  denotes the number of units invested in asset i at time t=0 until maturity. We denote the set of all trading strategies by  $\Theta$ .

**Definition 3.3** A hedger  $\theta: \mathcal{C} \to \Theta$  is a function which maps a claim  $S \in \mathcal{C}$  into a trading strategy  $\theta_S$  and satisfies the following conditions

- 1.  $\theta$  is normalized:  $\theta_0 = (0, 0, \dots, 0)$ .
- 2.  $\theta$  is translation invariant:  $\theta_{S+a} = \theta_S + (a, 0, \dots, 0)$ , for any  $S \in \mathcal{C}$  and  $a \in \mathbb{R}$ .

The trading strategy  $\theta_S$  is called the hedge for the claim S. Now, we define the important class of actuarial-consistent hedgers.

**Definition 3.4** A hedger  $\theta_S$  for a claim  $S \in \mathcal{C}$  is said to be an actuarial-consistent hedger with actuarial valuation principle  $\pi^{(2)}$  if

$$\theta_{S^{(2)}} = \left(\pi^{(2)}[S^{(2)}], 0, 0, \dots, 0\right), \text{ for } S^{(2)} \in \mathcal{C}^{(2)}.$$

An actuarial consistent hedger will only allow investments in the risk-free bank account for actuarial claims. The higher potential returns in risky assets can be used to protect against the future losses from a claim. However, when an investment in risky assets is used for managing an actuarial claim, the insurer will be exposed to movements on the financial market. By considering actuarial-consistent hedgers, the insurer only adds risky investments to the portfolio if the claim he is trying to hedge contains financial risks.

**Example 1 (Actuarial-consistent hedgers)** Hereafter, we consider two examples of actuarial-consistent hedgers.

• Consider the hedger  $\theta$  such that

$$\theta_S = \begin{cases} \left(\pi^{(2)}[S], 0, 0, \dots, 0\right), \text{ for } S \in \mathcal{C}^{(2)} \\ \arg\min_{\boldsymbol{\mu} \in \Theta} \mathbb{E}^{\mathbb{P}}\left[ \left(S - \boldsymbol{\mu} \cdot \mathbf{Y}\right)^2 \right], \text{ for any } S \in \mathcal{C} \setminus \mathcal{C}^{(2)}. \end{cases}$$

Such hedger invests risk-free for actuarial claims and invests following quadratic hedging for all remaining claims. By construction, such hedger is actuarial-consistent.

• Assume that an insurer considers the two-step actuarial valuation for any claim S:

$$\Pi[S] = \pi^{(2)} \left[ \pi^{(1)} \left[ S | \mathcal{F}^{(2)} \right] \right].$$

If the insurer invests the whole value in risk-free asset, the following hedger is used:

$$\theta_S = (\Pi[S], 0, 0, \dots, 0), \text{ for any } S \in \mathcal{C}. \tag{3.6}$$

Such hedger is naturally an actuarial-consistent hedger.

In the following lemma, we show how one can decompose a hybrid claim into an actuarial and financial part and define an actuarial consistent hedger.

**Lemma 3.2** Consider a hybrid claim  $S \in C$  and a hedger  $\tilde{\theta}$ . We decompose the claim S into two parts as follows:

$$H_S^{(2)} = \mathbb{E}\left[S|\mathcal{F}^{(2)}\right] - \pi^{(2)}[S],$$
  
 $H_S^{(1)} = S - H_S^{(2)}.$ 

Note that  $H_S^{(2)} \in \mathcal{C}^{(2)}$ . Define the hedger  $\theta_S$  as follows:

$$\theta_S = \tilde{\theta}_{H_S^{(1)}}.$$

Then,  $\theta_S$  is an actuarial hedger.

**Proof.** Consider an actuarial claim  $S^{(2)} \in \mathcal{C}^{(2)}$ . Then  $S^{(2)}$  is  $\mathcal{F}^{(2)}$ —measurable and therefore  $H^{(2)}_{S^{(2)}} = S^{(2)} - \pi^{(2)} \left[S^{(2)}\right]$  and  $H^{(1)}_{S^{(2)}} = \pi^{(2)} \left[S^{(2)}\right] \in \mathbb{R}$ . Since a hedger  $\tilde{\theta}$  is translation invariant, we then find:

$$\theta_{S^{(2)}} = \tilde{\theta}_{H_{S^{(2)}}^{(1)}} = \left(\pi^{(2)} \left[S^{(2)}\right], 0, 0, \dots, 0\right),$$

which shows we have an actuarial consistent hedger.

Lemma 3.2 illustrates how one can derive an actuarial consistent hedger using a general hedger. The claim  $H_S^{(2)}$  can be interpreted as the actuarial part of the claim S.

Assume that we have a claim S and a hedger  $\theta$ . If we want to invest in the hedge  $\theta_S$ , we have to pay its time-0 value which is given by  $\pi^{(1)}[\theta_S \cdot Y]$ . In the next Lemma, we show that the resulting financial value is weak actuarial-consistent if the hedger is actuarial-consistent.

#### **Lemma 3.3** The following two statements are equivalent.

1. The valuation  $\Pi$  can be expressed as follows

$$\Pi[S] = \pi^{(1)} \left[ \theta_S \cdot Y \right], \tag{3.7}$$

where  $\theta$  is an actuarial consistent hedger.

2.  $\Pi$  is weak actuarial consistent.

**Proof.** Assume  $\theta$  is an actuarial consistent hedger and the valuation  $\Pi$  is given by (3.7). Then for any actuarial claim  $S^{(2)} \in \mathcal{C}^{(2)}$ , we have that  $\theta_{S^{(2)}} \cdot Y = \pi^{(2)} \left[ S^{(2)} Y_0 \right]$ . Taking into account  $Y_0 = 1$ , we find that  $\Pi$  is weak actuarial consistent. Assume now that  $\Pi$  is a weak actuarial-consistent valuation. Defining  $\theta$  as in (3.6) shows that  $\Pi[S] = \pi^{(1)} \left[ \theta_S \cdot Y \right]$ , where  $\theta$  is an actuarial-consistent hedger.

In order to determine a hedge-based value of S, one first splits this claim into a hedgeable claim, which (partially) replicates S, and a remaining claim. The value of the claim S is then defined as the sum of the financial value of the hedgeable claim and the actuarial value of the remaining claim, determined according to a pre-specified actuarial valuation.

**Definition 3.5** A valuation  $\Pi$  is called an actuarial hedge-based valuation with financial valuation  $\pi^{(1)}$  and actuarial valuation  $\pi^{(2)}$  if it can be expressed as follows:

$$\Pi[S] = \pi^{(1)} [\theta_S \cdot Y] + \pi^{(2)} [S - \theta_S \cdot Y],$$

where  $\theta_S$  is an actuarial-consistent hedger.

In the following theorem, it is proven that the class of weak actuarial-consistent valuations is equal to the class of actuarial hedge-based valuations.

**Theorem 3.1** A valuation  $\Pi$  is a weak actuarial-consistent valuation if, and only if, it is an actuarial hedge-based valuation.

**Proof.** Assume  $\Pi$  is an actuarial hedge-based valuation, i.e. we have that

$$\Pi[S] = \pi^{(1)} [\theta_S \cdot Y] + \pi^{(2)} [S - \theta_S \cdot Y],$$

where  $\theta_S$  is an actuarial hedger. Consider an actuarial claim  $S^{(2)} \in \mathcal{C}^{(2)}$ . Then

$$\theta_{S^{(2)}} = \left(\rho \left\lceil S^{(2)} \right\rceil, 0, 0, \dots, 0\right),\,$$

for some actuarial valuation  $\rho$ . Then it is straightforward to verify that

$$\Pi\left[S^{(2)}\right] = \pi^{(2)}\left[S^{(2)}\right],\,$$

which shows that an actuarial hedge-based valuation is a weak actuarial-consistent valuation.

Assume now that  $\Pi$  is a weak actuarial-consistent valuation. Then it follows from Lemma 3.3 that

$$\Pi[S] = \pi^{(1)} \left[ \theta_S \cdot Y \right],$$

where  $\theta_S = (\Pi[S], 0, \dots, 0)$ . Define the valuation  $\Pi'$  as follows:

$$\Pi'[S] = \pi^{(1)} [\theta_S \cdot Y] + \Pi [S - \theta_S \cdot Y].$$

Then  $\Pi'$  is an actuarial hedge-based valuation. Moreover, since  $\theta_S \cdot Y = \Pi[S]$ , we also find that  $\Pi' = \Pi$ , which ends the proof.

## 3.3 Two-step actuarial valuations

Hereafter, we introduce a class of actuarial-consistent valuations which we call *two-step actuarial valuations*. More specifically, in a first step we compute the financial value of S conditional on actuarial scenarios (the values of the actuarial assets X), i.e.  $\pi^{(1)}\left[S|\mathcal{F}^{(2)}\right]$ . Since this conditional payoff depends only on actuarial scenarios and is then  $\mathcal{F}^{(2)}$ -measurable, the quantity  $\pi^{(1)}\left[S|\mathcal{F}^{(2)}\right]$  should be valuated via a standard actuarial valuation  $\pi^{(2)}$ . This motivates the following definition.

**Definition 3.6 (Two-step actuarial valuation)** *The valuation*  $\Pi$  *is called a two-step actuarial valuation if it can be expressed as follows:* 

$$\Pi[S] = \pi^{(2)} \left[ \pi^{(1)} \left[ S | \mathcal{F}^{(2)} \right] \right],$$

where  $\pi^{(2)}$  is the actuarial valuation and  $\pi^{(1)}$  is the  $\mathcal{F}^{(2)}$ -conditional financial valuation.

Hence, the two-step actuarial valuation consists of applying the market-adjusted valuation  $\pi^{(1)}$  to the residual risk which remains after having conditioned on the future development of the actuarial risks, i.e. the filtration  $\mathbb{F}^{(2)}$ . It is straightforward to verify that the two-step actuarial valuation is orthogonal consistent.

In the following theorem, we show that any two-step actuarial valuation is weak actuarial-consistent. Moreover, if the valuation is coherent, the reciprocal is true: any weak actuarial-consistent valuation has a two-step actuarial valuation representation.

#### Theorem 3.2 (Characterization of weak actuarial-consistency)

The following properties hold:

- 1. Any two-step actuarial valuation  $\Pi$  is weak actuarial-consistent.
- 2. If  $\Pi$  is a coherent weak actuarial-consistent valuation, there exists an  $\mathcal{F}^{(2)}$ -conditional coherent valuation  $\pi^{(1)}$  and a coherent actuarial valuation  $\pi^{(2)}$  such that

$$\Pi[S] = \pi^{(2)} \left[ \pi^{(1)} \left[ S | \mathcal{F}^{(2)} \right] \right].$$
 (3.8)

**Proof.** (1) To prove that  $\Pi$  is weak actuarial-consistent, it is sufficient to notice that

$$\Pi \left[ S^{(2)} \right] = \pi^{(2)} \left[ \pi^{(1)} \left[ S^{(2)} | \mathcal{F}^{(2)} \right] \right]$$
$$= \pi^{(2)} \left[ S^{(2)} \right],$$

where we have used that  $S^{(2)}$  is  $\mathcal{F}^{(2)}$ -measurable. Moreover,  $\Pi$  is coherent since  $\pi^{(1)}$  and  $\pi^{(2)}$  are coherent.

(2)  $^{6}$  Because  $\Pi$  is coherent, we have that

$$\Pi[S] = \sup_{\varphi \in \mathcal{Q}} \mathbb{E}[\varphi S]$$

where Q is a unique, non-empty, closed convex subset of P. By weak market-consistency, for any actuarial claim  $S^{(2)}$ , the following also holds:

$$\Pi\left[S^{(2)}\right] = \sup_{\varphi^{(2)}} \mathbb{E}[\varphi^{(2)}S^{(2)}],$$

where the supremum is taken over a set of probability measures such that  $\varphi^{(2)}$  is  $\mathcal{F}^{(2)}$ -measurable. We can then write

$$\varphi = \varphi^{(2)} Z$$
,

with  $Z\in\mathcal{Q}_{(2)}:=\{Z\in L^2(\mathbb{P},\mathbb{F})\mid \mathbb{E}^{\mathbb{P}}\left[Z|\mathcal{F}^{(2)}\right]=1\}.$  Therefore, we have that

$$\begin{split} \Pi[S] &= \sup_{\varphi^{(2)}} \sup_{Z \in \mathcal{Q}_{(2)}} \mathbb{E}[\varphi^{(2)}ZS] \\ &= \sup_{\varphi^{(2)}} \sup_{Z \in \mathcal{Q}_{(2)}} \mathbb{E}\left[\varphi^{(2)}\mathbb{E}\left[ZS|\mathcal{F}^{(2)}\right]\right]. \end{split}$$

We remark that if we can prove that

$$\sup_{Z \in \mathcal{Q}_{(2)}} \mathbb{E}\left[\varphi^{(2)} \mathbb{E}\left[ZS|\mathcal{F}^{(2)}\right]\right] = \mathbb{E}\left[\varphi^{(2)} \sup_{Z \in \mathcal{Q}_{(2)}} \mathbb{E}\left[ZS|\mathcal{F}^{(2)}\right]\right],\tag{3.9}$$

then the proof is over since this relation implies that

$$\begin{split} \Pi[S] &= \sup_{\varphi^{(2)}} \mathbb{E} \left[ \varphi^{(2)} \sup_{Z \in \mathcal{Q}_{(2)}} \mathbb{E} \left[ ZS | \mathcal{F}^{(2)} \right] \right] \\ &= \pi^{(2)} \left[ \pi^{(1)} \left[ S | \mathcal{F}^{(2)} \right] \right]. \end{split}$$

<sup>&</sup>lt;sup>6</sup>We acknowledge that some arguments of the proof are similar to Theorem 3.10 in Pelsser and Stadje (2014)

To prove (3.9), we first observe that

$$\sup_{Z \in \mathcal{Q}_{(2)}} \mathbb{E}\left[\varphi^{(2)} \mathbb{E}\left[ZS|\mathcal{F}^{(2)}\right]\right] \leq \mathbb{E}\left[\varphi^{(2)} \sup_{Z \in \mathcal{Q}_{(2)}} \mathbb{E}\left[ZS|\mathcal{F}^{(2)}\right]\right].$$

Let us prove the other inequality. By definition of the supremum, there exists a sequence  $Z_n \in \mathcal{Q}_{(2)}$  with  $\mathbb{E}\left[Z_1S|\mathcal{F}^{(2)}\right] \leq \mathbb{E}\left[Z_2S|\mathcal{F}^{(2)}\right] \leq \cdots$  such that  $\lim_n \mathbb{E}\left[Z_nS|\mathcal{F}^{(2)}\right] = \sup_{Z \in \mathcal{Q}_{(2)}} \mathbb{E}\left[ZS|\mathcal{F}^{(2)}\right]$ . Thus, by the monotone convergence theorem,

$$\mathbb{E}\left[\varphi^{(2)} \sup_{Z \in \mathcal{Q}_{(2)}} \mathbb{E}\left[ZS|\mathcal{F}^{(2)}\right]\right] = \lim_{n} \mathbb{E}\left[\varphi^{(2)}\mathbb{E}\left[Z_{n}S|\mathcal{F}^{(2)}\right]\right]$$
$$\leq \sup_{Z \in \mathcal{Q}_{(2)}} \mathbb{E}\left[\varphi^{(2)}\mathbb{E}\left[ZS|\mathcal{F}^{(2x)}\right]\right],$$

which ends the proof.

Combining Theorems 3.1 and 3.2, we find the following representation theorem in the coherent case.

**Corollary 3.1** *If*  $\Pi$  *is a coherent valuation, the following statements are equivalent:* 

- 1. The valuation is weak actuarial-consistent.
- 2. The valuation is an actuarial hedge-based valuation.
- 3. The valuation is a two-step actuarial valuation.

We note that the coherent property is only necessary to show that any weak actuarial-consistent valuation has a two-step representation. In particular, any two-step actuarial valuation and actuarial hedge-based valuation are weak actuarial-consistent, even without the coherent property.

## 4 Fair valuations

Motivated by solvency regulations, the recent actuarial literature focused on market-consistent valuations which essentially require that financial risks are priced with a risk-neutral valuation. In this section, we provide a detailed comparison between actuarial-consistent and market-consistent valuations.

## 4.1 Market-consistency and two-step financial valuations

Hereafter, we only assume that a financial valuation  $\pi^{(1)}$  is used to value financial claims and search for valuations that are consistent with such requirement. In particular, we do not restrict ourselves to the linear risk-neutral valuation.

**Definition 4.1 (Strong market-consistency)** A valuation  $\Pi$  is called strong market-consistent (strong MCV) if for any financial claim  $S^{(1)}$  the following holds:

$$\Pi \left[ S + S^{(1)} \right] = \Pi \left[ S \right] + \pi^{(1)} \left[ S^{(1)} \right]. \tag{4.1}$$

In the literature, market-consistency is usually defined via a condition identical or similar to the condition (4.1) (see e.g. Pelsser and Stadje (2014), Dhaene et al. (2017) and Barigou et al. (2019)). However, strong market-consistency, similar to strong actuarial-consistency, implies linearity of the valuation. Therefore, strong MCV is too restrictive when the law of one price does not prevail.

**Definition 4.2 (Weak market-consistency)** A valuation  $\Pi$  is called weak market-consistent (weak MCV) if for any financial claim  $S^{(1)}$  the following holds:

$$\Pi \left[ S^{(1)} \right] = \pi^{(1)} \left[ S^{(1)} \right].$$
 (4.2)

Weak market-consistency only postulates that for financial claims, a financial valuation is applied. This weaker notion of market-consistency does not impose linearity of the financial valuation and allows for two-price economy and market incompleteness. We remark that Assa and Gospodinov (2018) also investigated these two types of market-consistency, that they called market-consistency of type I and type II.

**Definition 4.3 (Two-step financial valuation)** *The valuation*  $\Pi$  *is called a two-step financial valuation if it can be expressed as follows:* 

$$\Pi[S] = \pi^{(1)} \left[ \pi^{(2)} \left[ S | \mathcal{F}^{(1)} \right] \right],$$

where  $\pi^{(1)}$  is the financial valuation and  $\pi^{(2)}$  is the  $\mathcal{F}^{(1)}$ -conditional actuarial valuation.

Definition 4.3 can be seen as a generalization of Pelsser and Stadje (2014) where the linear risk-neutral operator is replaced by a general financial valuation  $\pi^{(1)}$ . Pelsser and Stadje (2014) showed that, under appropriate assumptions, strong market-consistent valuation is equivalent to two-step financial valuation where  $\pi^{(1)}$  is linear (see their Theorem 3.10). In the following theorem, this result is extended to the non-linear case. We note that the theorem is exactly Theorem 3.2 with the role of actuarial and financial valuations interchanged. The proof is therefore omitted.

## Theorem 4.1 (Characterization of weak market-consistency) The following properties hold:

- 1. Any two-step financial valuation  $\Pi$  is weak market-consistent.
- 2. If  $\Pi$  is a coherent weak market-consistent valuation, there exists an  $\mathcal{F}^{(1)}$ -conditional coherent valuation  $\pi^{(2)}$  and a coherent actuarial valuation  $\pi^{(1)}$  such that

$$\Pi[S] = \pi^{(1)} \left[ \pi^{(2)} \left[ S | \mathcal{F}^{(1)} \right] \right].$$
 (4.3)

## 4.2 Fair valuation: merging market- and actuarial-consistency

After having defined two broad classes of valuations: market-consistent and actuarial-consistent valuations, a natural question arises: Could we always define a *fair* valuation that is market-consistent *and* actuarial-consistent?

**Definition 4.4 (Fair valuation)** The valuation  $\Pi$  is fair if it is weak market-consistent and weak actuarial-consistent.

In general, it will not always be possible to define a fair valuation. Indeed, in a general probability space in which financial and actuarial risks are dependent, there is ambiguity on the valuation to be used: a market-consistent valuation calibrated to market prices *or* an actuarial-consistent valuation calibrated to historical actuarial data.

In the following lemma, we show that if the valuation is weak MCV and ACV and there exist a financial and actuarial claim that are equal a.s., then the financial and actuarial valuations should coincide. In particular, this lemma implies that for a given financial valuation  $\pi^{(1)}$  and actuarial valuation  $\pi^{(2)}$ , it is not always possible to define a fair valuation (i.e. a valuation that is weak MCV and ACV).

**Lemma 4.1** Assume that there exist a financial claim  $S^{(1)}$  and an actuarial claim  $S^{(2)}$  such that

$$S^{(1)} \stackrel{a.s.}{=} S^{(2)}. \tag{4.4}$$

If the valuation  $\Pi$  is weak MCV and weak ACV, then the following holds:

$$\pi^{(1)} \left[ S^{(1)} \right] = \pi^{(2)} \left[ S^{(1)} \right].$$

**Proof.** Since  $\Pi$  is weak MCV, we can write:

$$\Pi \left[ S^{(1)} \right] = \pi^{(1)} \left[ S^{(1)} \right]$$
$$= \pi^{(1)} \left[ S^{(2)} \right] \text{ by (4.4)}.$$

Moreover, weak ACV implies that

$$\begin{split} \Pi\left[S^{(2)}\right] &= \pi^{(2)}\left[S^{(2)}\right] \\ &= \pi^{(2)}\left[S^{(1)}\right] \text{ by (4.4)}. \end{split}$$

Because we identify claims which are equal a.s., we have that

$$\Pi\left[S^{(1)}\right] = \Pi\left[S^{(2)}\right],$$

which ends the proof.

To illustrate the previous lemma, we consider an example where financial and actuarial claims are comonotonic and show that a fair valuation cannot be properly defined.

**Example 2 (Comonotonic financial and actuarial claims)** Consider the financial claim  $S^{(1)}$  and the actuarial claim  $S^{(2)}$  which are given by

$$S^{(1)} = \max(Y, 100) = \begin{cases} 0, & \text{if } Y = 50, \\ 100, & \text{if } Y = 200. \end{cases} \text{ and } S^{(2)} = \begin{cases} 0, & \text{if } I = 0, \\ 100, & \text{if } I = 1, \end{cases}$$
 (4.5)

Hence,  $S^{(1)}$  is a call option with strike K=100 on the stock Y and  $S^{(2)}$  is a survival index on the random variable I which indicates if the policy holder survives (I=1) or dies (I=0) in the time interval. Assume moreover that the claims are comonotonic:  $\mathbb{P}[(Y,I)=(50,0)]=p$  and  $\mathbb{P}[(Y,I)=(50,0)]=p$ 

(200,1)] = 1 - p. Clearly, we have that  $S^{(1)} \stackrel{a.s.}{=} S^{(2)}$ . Let us note  $p = \mathbb{P}[I=1]$  and  $q = \mathbb{Q}[Y=200]$ . Then, any fair valuation  $\Pi$  should satisfy:

$$\Pi \left[ S^{(1)} \right] = 100q,$$

$$\Pi \left[ S^{(2)} \right] = 100 \left( p + \beta \sqrt{p(1-p)} \right),$$

if the financial valuation is the risk-neutral valuation and the actuarial valuation is the standard deviation principle. We then have two identical risks with two (possibly very different) values, creating inconsistency in the valuation mechanism.

With the emergence of the market for longevity derivatives, a valuator needs to make a choice between market-consistency and actuarial-consistency. For instance, consider a market with some traded longevity bonds and there is an issue of a new longevity product. One needs to decide to use either a market-consistent approach based on the traded longevity bonds in the market or an actuarial-consistent approach based on longevity trend assumptions.

In the following example, we illustrate this point and compare a market-consistent and an actuarial-consistent valuation in the presence of a longevity bond.

**Example 3 (Comparison between MCV and ACV)** (a) Consider a portfolio of pure endowments for  $l_x$  Belgian insureds of age x at time 0. The pure endowment guarantees a sum of 1 if the policyholder is still alive at maturity. The aggregate payoff can be written as

$$S = L_{x+T}$$

with  $L_{x+T}$  the number of policyholders who survive up to the maturity time T. Moreover, we assume that the financial market is composed of two assets: a risk-free asset  $Y^{(0)}(t) = e^{rt}$  and a longevity bond for which the payoff at maturity is  $Y^{(1)}(T) = \tilde{L}_{x+T}$ , the equivalent of  $L_{x+T}$  but for the Dutch population. First, we determine the expected value (called best-estimate) by a two-step actuarial valuation:

$$BE[S] = \mathbb{E}^{\mathbb{P}} \left[ \mathbb{E}^{\mathbb{Q}} \left[ e^{-rT} L_{x+T} | \mathcal{F}^{(2)} \right] \right]$$
$$= \mathbb{E}^{\mathbb{P}} \left[ e^{-rT} L_{x+T} \mathbb{E}^{\mathbb{Q}} \left[ 1 | \mathcal{F}^{(2)} \right] \right]$$
$$= e^{-rT} \mathbb{E}^{\mathbb{P}} \left[ L_{x+T} \right].$$

The actuarial-consistent valuation would suggest a full investment in the risk-free asset. Secondly, assuming that the Belgian population live slightly shorter than the Dutch population<sup>7</sup>:  $\mathbb{E}^{\mathbb{P}}\left[L_{x+T}|\tilde{L}_{x+T}\right] = \beta \tilde{L}_{x+T}$  with  $\beta < 1$ , we determine the best-estimate according to a two-step financial valuation:

$$BE^*[S] = \mathbb{E}^{\mathbb{Q}} \left[ e^{-rT} \mathbb{E}^{\mathbb{P}} \left[ L_{x+T} | \mathcal{F}^{(1)} \right] \right]$$
$$= \mathbb{E}^{\mathbb{Q}} \left[ e^{-rT} \beta \tilde{L}_{x+T} \right]$$
$$= \beta Y^{(1)}(0).$$

where  $Y^{(1)}(0)$  is the current price of the longevity bond. The market-consistent valuation would then suggest a full investment in the Dutch longevity bond.

<sup>&</sup>lt;sup>7</sup>For the reader interested in Dutch and Belgian mortality projections, we refer to Antonio et al. (2017)

(b) In order to better grasp the difference between the actuarial-consistent and market-consistent valuations, we introduce the following modelling assumptions.

Assume that the interest rate r=0, and the bivariate Belgian-Dutch population follows the distribution:  $(L_{x+T}, \tilde{L}_{x+T}) \sim \mathcal{N}(\boldsymbol{\mu}, \boldsymbol{\Sigma})$  with

$$\boldsymbol{\mu} = \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix}, \quad \boldsymbol{\Sigma} = \begin{pmatrix} \sigma_1^2 & \rho \sigma_1 \sigma_2 \\ \rho \sigma_1 \sigma_2 & \sigma_2^2 \end{pmatrix}.$$

Hence, both Belgian and Dutch populations are normal distributed with correlation  $\rho$ . Hereafter, we compare two-step actuarial and financial valuations with the following financial and actuarial valuations:

$$\pi^{(1)}\left[S\right] = \mathbb{E}^{\mathbb{Q}}\left[S\right],$$

$$\pi^{(2)}[S] = \mathbb{E}^{\mathbb{P}}\left[S\right] + \beta\sqrt{\operatorname{Var}^{\mathbb{P}}\left[S\right]}.$$

The two-step actuarial valuation of  $S = L_{x+T}$  is given by

$$\Pi^{(1)}[S] = \pi^{(2)} \left[ \mathbb{E}^{\mathbb{Q}} \left[ e^{-rT} L_{x+T} | L_{x+T} \right] \right] 
= \mathbb{E}^{\mathbb{P}} \left[ L_{x+T} \right] + \beta \sigma^{\mathbb{P}} \left[ L_{x+T} \right] 
= \mu_1 + \beta \sigma_1.$$
(4.6)

To determine the two-step financial valuation, we first notice that by standard results of normal distributions, we have

$$L_{x+T}|\tilde{L}_{x+T} = x \sim \mathcal{N}\left(\mu_1 + \rho \frac{\sigma_1}{\sigma_2} (x - \mu_2), (1 - \rho^2)\sigma_1^2\right).$$

Let us further assume that the distribution of  $\tilde{L}_{x+T}$  under  $\mathbb{Q}$  is

$$\tilde{L}_{x+T} \stackrel{\mathbb{Q}}{\sim} \mathcal{N}(\mu_2 - \sigma_2 \kappa, \sigma_2^2),$$

where  $\kappa > 0$  is the market price of risk for the longevity bond. Therefore, we find that

$$\Pi^{(2)}[S] = \mathbb{E}^{\mathbb{Q}} \left[ \pi^{(2)} \left[ L_{x+T} | \tilde{L}_{x+T} \right] \right] 
= \mathbb{E}^{\mathbb{Q}} \left[ \mu_1 + \rho \frac{\sigma_1}{\sigma_2} \left( \tilde{L}_{x+T} - \mu_2 \right) + \beta \sigma_1 \sqrt{1 - \rho^2} \right] 
= \mu_1 + \rho \frac{\sigma_1}{\sigma_2} \left( \mathbb{E}^{\mathbb{Q}} \left[ \tilde{L}_{x+T} \right] - \mu_2 \right) + \beta \sigma_1 \sqrt{1 - \rho^2} 
= \mu_1 - \rho \sigma_1 \kappa + \beta \sigma_1 \sqrt{1 - \rho^2}.$$
(4.7)

We can compare the two-step actuarial valuation (4.6) with the two-step financial valuation (4.7). Intuitively, the difference should reflect two aspects:

- 1. The dependence between Belgian and Dutch populations.
- 2. The risk premium on the Dutch longevity bond.

The results confirm the intuition: the difference is given by

$$\Pi^{(2)}[S] - \Pi^{(1)}[S] = \sigma_1 \left[ \beta \left( \sqrt{1 - \rho^2} - 1 \right) - \rho \kappa \right]. \tag{4.8}$$

We observe that the higher the correlation  $\rho$ , the higher the difference (this reflects the point 1.). Moreover, the absolute difference is an increasing function of the market price of risk  $\kappa$  (this reflects the point 2.).

If the valuator can choose between the risk-free investment or the Dutch longevity bond, he will go for the longevity bond if the benefits are higher than the costs, i.e. if the risk reduction of investing in the longevity bond is higher than the extra price he has to pay (given by Equation (4.8)). The prices at time 0 of both approaches and the residual losses at maturity are given in the table below:

$$\begin{array}{l} \underline{\textit{Price at time } 0} \\ \Pi^{(1)}[S] = \mu_1 + \beta \sigma_1 \\ \Pi^{(2)}[S] = \mu_1 - \rho \sigma_1 \kappa + \beta \sigma_1 \sqrt{1 - \rho^2} \end{array} \\ \begin{vmatrix} \underline{\textit{Residual loss at maturity}} \\ R_1 = L_{x+T} - \mu_1 - \beta \sigma_1 \sim \mathcal{N}(-\beta \sigma_1, \sigma_1^2) \\ R_2 = L_{x+T} - \left(\mu_1 + \rho \frac{\sigma_1}{\sigma_2} \left(\tilde{L}_{x+T} - \mu_2\right) + \beta \sigma_1 \sqrt{1 - \rho^2} \right) \\ \sim \mathcal{N}(-\beta \sigma_1 \sqrt{1 - \rho^2}, (1 - \rho^2) \sigma_1^2) \end{aligned}$$

From the table, we observe that the investment in the longevity bond leads to a decrease in the volatility of the residual loss but an increase in the expected loss. Notice that in case of comonotonic or countermonotonic risks (i.e.  $\rho = \pm 1$ ), the claim S can be completely hedged with the longevity bond and the residual loss  $R_2$  equals 0. The valuator will typically go for the longevity bond if the risk reduction (computed in terms of Value-at-Risk for simplicity) is higher than the extra price to pay:

$$\sigma_1 \left[ \beta \left( \sqrt{1 - \rho^2} - 1 \right) - \rho \kappa \right] < VaR_p[R_2] - VaR_p[R_1]$$
$$= \sigma_1 \left[ \Phi^{-1}(p) - \beta \right] \left( \sqrt{1 - \rho^2} - 1 \right).$$

On the other hand, if the longevity bond price is too high in comparison with the risk reduction, an actuarial-consistent valuation is then preferable.

**Remark 4.2** In this paper, we do not argue that one method is better than another; each one has pros and cons. While the second method allows to transfer the risk to the financial market, it comes also with a price: the liabilities become totally dependent on the longevity bond. In particular, an adverse shock on the Dutch population or a counter-party's default will have a direct effect on the assets backing the liabilities.

More generally, as pointed out by Vedani et al. (2017), market-consistent valuations are directly subject to market movements, and can lead to excess volatility, depending on the calibration sets chosen by the actuary. We also refer to Rae et al. (2018) for different concerns around the appropriateness of market-consistency to the insurance business.

In the next example, we consider the valuation of a hybrid claim via a two-step financial and actuarial valuation, and investigate the difference between the two-step operators.

**Example 4 (Two-step valuations for hybrid claims)** Consider an equity-linked contract for a life (x), which pays the call option  $(Y - K)_+$  in case the policyholder is alive at time T = 1 and 0 otherwise. We recall that the stock Y can go up to 200 or down to 50, the strike K = 100 and the policyholder survival is modelled by the indicator I. Therefore, the payoff of this contract is given by

$$S = (Y - K)_{+} \times I = \begin{cases} 100, & \text{if } Y = 200, I = 1, \\ 0, & \text{otherwise.} \end{cases}$$
 (4.9)

We assume that the financial valuation is the risk-neutral expectation and the actuarial valuation is the standard deviation principle:

$$\pi^{(1)}[S] = \mathbb{E}^{\mathbb{Q}}[S],$$
  
$$\pi^{(2)}[S] = \mathbb{E}^{\mathbb{P}}[S] + \beta \sqrt{Var^{\mathbb{P}}[S]}.$$

We consider the two-step valuations for the hybrid payoff (4.9):

1. Two-step financial valuation: The value of S is given by

$$\Pi^{(1)}[S] = \mathbb{E}^{\mathbb{Q}} \left[ \mathbb{E}^{\mathbb{P}} \left[ S | \mathcal{F}^{(1)} \right] + \beta \sqrt{Var^{\mathbb{P}} \left[ S | \mathcal{F}^{(1)} \right]} \right]$$
$$= \mathbb{E}^{\mathbb{Q}} \left[ (Y - K)_{+} \left( \mathbb{E}^{\mathbb{P}} \left[ I | Y \right] + \beta \sigma^{\mathbb{P}} \left[ I | Y \right] \right) \right].$$

If we note that

$$\mathbb{E}^{\mathbb{P}}\left[I|Y\right] = \left\{ \begin{array}{ll} \mathbb{P}[I=1|Y=50], & \textit{if } Y=50, \\ \mathbb{P}[I=1|Y=200], & \textit{if } Y=200, \end{array} \right.$$

then we find that the two-step financial value of S is

$$\Pi^{(1)}[S] = 100 \, q_Y \left( p_{I|Y=200} + \beta \sqrt{p_{I|Y=200} (1 - p_{I|Y=200})} \right), \tag{4.10}$$

where  $q_Y$  is the  $\mathbb{Q}$ -probability that Y goes up:  $q_Y = \mathbb{Q}[Y = 200]$  and  $p_{I|Y=200}$  is the  $\mathbb{P}$ -probability that the policyholder is alive given that the stock goes up:  $p_{I|Y=200} = \mathbb{P}[I = 1|Y = 200]$ .

2. Two-step actuarial valuation: The value of S is given by

$$\Pi^{(2)}[S] = \mathbb{E}^{\mathbb{P}}\left[\mathbb{E}^{\mathbb{Q}}\left[S|\mathcal{F}^{(2)}\right]\right] + \beta\sigma^{\mathbb{P}}\left[\mathbb{E}^{\mathbb{Q}}\left[S|\mathcal{F}^{(2)}\right]\right]$$
$$= \mathbb{E}^{\mathbb{P}}\left[I\mathbb{E}^{\mathbb{Q}}\left[(Y-K)_{+}|I|\right] + \beta\sigma^{\mathbb{P}}\left[I\mathbb{E}^{\mathbb{Q}}\left[(Y-K)_{+}|I|\right]\right].$$

Noting that

$$\mathbb{E}^{\mathbb{Q}}[(Y-K)_{+}|I] = \begin{cases} 100 \,\mathbb{Q}[Y=200|I=0], & \text{if } I=0, \\ 100 \,\mathbb{Q}[Y=200|I=1], & \text{if } I=1, \end{cases}$$

then we find that the two-step actuarial value of S is

$$\Pi^{(2)}[S] = 100 \, q_{Y|I=1} \left( p_I + \beta \sqrt{p_I (1 - p_I)} \right), \tag{4.11}$$

where  $p_I$  is the  $\mathbb{P}$ -probability that the policyholder is alive:  $p_I = \mathbb{P}[I=1]$  and  $q_{Y|I=1}$  is the  $\mathbb{Q}$ -probability that the stock goes up given that the policyholder is alive:  $q_{Y|I=1} = \mathbb{Q}[Y=200|I=1]$ .

If we compare the two-step financial and actuarial values (4.10) and (4.11), the structure is similar but the dependence between financial and actuarial risks is taken into account differently. In the first case, it is via the  $\mathbb{P}$ -probability of actuarial risks given financial scenarios, i.e.  $p_{I|Y=200}$  while in the second case, it is via the  $\mathbb{Q}$ -probability of financial risks given actuarial scenarios, i.e.  $q_{Y|I=1}$ . In case of

independence under  $\mathbb{P}$  and  $\mathbb{Q}$ , both valuations are equal. In case of dependence, both valuations (4.10) and (4.11) will in general be different as we illustrate below:

$$\Pi^{(2)}[S] - \Pi^{(1)}[S] = 100 \, q_{Y|I=1} \, p_I - 100 \, q_Y \, p_{I|Y=200}$$

$$+ 100 \, q_{Y|I=1} \, \beta \sqrt{p_I(1-p_I)} - 100 \, q_Y \, \beta \sqrt{p_{I|Y=200}(1-p_{I|Y=200})}.$$

Let us further assume that the difference between  $\mathbb{P}$  and  $\mathbb{Q}$  is given by a constant market price of risk  $\kappa$ :

$$\kappa = q_Y - p_Y,$$
  
=  $q_{Y|I=1} - p_{Y|I=1}.$ 

Therefore, by Bayes' Theorem, we find that

$$\Pi^{(2)}[S] - \Pi^{(1)}[S] = 100 \left( \frac{p_{I|Y=200}p_Y}{p_I} + \kappa \right) p_I - 100 (p_Y + \kappa) p_{I|Y=200} + 100 (p_{Y|I=1} + \kappa) \beta \sqrt{p_I(1 - p_I)} - 100 (p_Y + \kappa) \beta \sqrt{p_{I|Y=200}(1 - p_{I|Y=200})}.$$

After simplifications, we find that

$$\Pi^{(2)}[S] - \Pi^{(1)}[S] = 100\kappa \left( p_I - p_{I|Y=200} \right)$$

$$+ 100\kappa\beta \left( \sqrt{p_I(1-p_I)} - \sqrt{p_{I|Y=200}(1-p_{I|Y=200})} \right)$$

$$+ 100p_{Y|I=1}\beta \sqrt{p_I(1-p_I)} - 100p_Y\beta \sqrt{p_{I|Y=200}(1-p_{I|Y=200})}.$$

Similar to Example 3, we observe that the difference between the two-step valuations relies mainly on

- The risk premium  $\kappa$  which reflects the difference between the real-world measure  $\mathbb{P}$  and the risk-neutral measure  $\mathbb{Q}$ .
- The dependence between actuarial and financial risks (expressed as the difference between  $p_I$  and  $p_{I|Y=200}$  as well as the difference between  $p_Y$  and  $p_{Y|I=1}$ ).

We remark that in the literature, it is common to assume that financial and actuarial claims are independent (either under  $\mathbb{P}$  or  $\mathbb{Q}^8$ ). In that case, one can define a valuation that is MCV and ACV since the valuation is decoupled into two independent valuations, one for financial claims and one for actuarial claims. Even though extracting the  $\mathbb{Q}$ -dependence might be more complicated, different papers investigated the valuation under dependent financial and actuarial risks (see e.g. Liu et al. (2014), Deelstra et al. (2016), Zhao and Mamon (2018)).

In the following lemma, we show that if the conditional actuarial valuation of actuarial claims does not depend on the financial filtration, the two-step financial valuation is fair. Similar result holds for the two-step actuarial valuation.

**Lemma 4.2** Consider hybrid claims of the form  $S = S^{(1)} \times S^{(2)}$  where  $S^{(1)}$  is financial and  $S^{(2)}$  is actuarial. If one of the two following conditions holds:

<sup>&</sup>lt;sup>8</sup>Note that independence under  $\mathbb{P}$  does not necessarily imply independence under  $\mathbb{Q}$ , see Dhaene et al. (2013).

- 1.  $\Pi$  is a two-step financial valuation and  $\pi^{(2)} [S^{(2)} | \mathcal{F}^{(1)}] = \pi^{(2)} [S^{(2)}]$ .
- 2.  $\Pi$  is a two-step actuarial valuation and  $\pi^{(1)} [S^{(1)} | \mathcal{F}^{(2)}] = \pi^{(1)} [S^{(1)}]$ .

Then, we have that

$$\Pi\left[S^{(1)}\times S^{(2)}\right]=\pi^{(1)}\left[S^{(1)}\right]\times\pi^{(2)}\left[S^{(2)}\right].$$

*In particular, the valuation*  $\Pi$  *is fair:* 

$$\Pi \left[ S^{(1)} \right] = \pi^{(1)} \left[ S^{(1)} \right]$$

$$\Pi \left[ S^{(2)} \right] = \pi^{(2)} \left[ S^{(2)} \right].$$

**Proof.** The proof follows directly from the definition of the two-step valuations.

From Lemma 4.2, we observe that we can define a fair valuation under appropriate independence assumptions. The first condition will typically require independence under  $\mathbb{P}$  (e.g.  $\mathbb{E}\left[S^{(2)}|\mathcal{F}^{(1)}\right] = \mathbb{E}[S^{(2)}]$ ) and the second condition independence under  $\mathbb{Q}$  (e.g.  $\mathbb{E}^{\mathbb{Q}}\left[S^{(1)}|\mathcal{F}^{(2)}\right] = \mathbb{E}^{\mathbb{Q}}[S^{(1)}]$ ). Another possibility is to restrict the notion of actuarial-consistency to the risks which are independent of the financial market (see for instance Dhaene et al. (2017)).

**Example 4 (continued)** If we assume independence under  $\mathbb{P}$  in the two-step financial valuation or independence under  $\mathbb{Q}$  in the two-actuarial valuation, both valuations lead to a fair valuation. This is in line with Lemma 4.2.

## 5 Numerical illustration

Based on the two-step actuarial valuation introduced in the previous section, we show how the valuation can be decomposed into a best estimate and a risk margin. Moreover, we illustrate the valuation on a portfolio of equity-linked life insurance contracts with dependent financial and actuarial risks.

#### 5.1 Best estimate, risk margin and cost-of-capital valuation

#### 5.1.1 Best estimate

In Article 77 of the DIRECTIVE 2009/138/EC (European Commission (2009)), the best estimate is defined as the "the probability-weighted average of future cash-flows taking account of the time value of money" (expected present value of future cash-flows). Hence, the best estimate of an insurance liability can be interpreted as an appropriate estimation of the expected present value based on actual available information.

Based on our two-step actuarial valuation, we can define a broad notion of best estimate for a general claim S. Indeed, one can generate stochastic actuarial scenarios, determine the financial price in each scenario and then average over the different scenarios. This leads to the following definition.

**Definition 5.1** (Best estimate) For any claim  $S \in C$ , the best estimate is given by

$$BE[S] = \mathbb{E}^{\mathbb{P}}\left[\mathbb{E}^{\mathbb{Q}}\left[S|\mathcal{F}^{(2)}\right]\right].$$
 (5.1)

It turns out that the best estimate appears as a two-step actuarial valuation for which there is no distortion of the different measures, i.e.  $\pi^{(1)}[S] = \mathbb{E}^{\mathbb{Q}}[S]$  and  $\pi^{(2)}[S] = \mathbb{E}^{\mathbb{P}}[S]$ . In general, the expression (5.1) could be hardly tractable since we can possibly have an infinite number of actuarial scenarios. For practical purposes, we will often consider the *approximated* best estimate  $\widehat{BE}$  defined by

$$\widehat{BE}[S] = \sum_{i=1}^{n} \mathbb{P}[A_i] \,\mathbb{E}^{\mathbb{Q}}[S|A_i], \qquad (5.2)$$

for a finite number n of actuarial scenarios:  $A_1, A_2, ..., A_n \in \mathcal{F}^{(2)}$ .

The best estimate in Definition 5.1 appears as an average of risk-neutral valuations which are applied to the risk which remains after having conditioned on the actuarial filtration. Hereafter, we consider some special cases:

• For any actuarial risk  $S^{(2)}$ , we find that

$$BE[S^{(2)}] = \mathbb{E}^{\mathbb{P}} \left[ S^{(2)} \right].$$

• For any product claim S with independent actuarial and financial risks (under  $\mathbb{Q}$ ), we find that

$$BE[S] = \mathbb{E}^{\mathbb{P}} \left[ \mathbb{E}^{\mathbb{Q}} \left[ S^{(1)} \times S^{(2)} | \mathcal{F}^{(2)} \right] \right]$$

$$= \mathbb{E}^{\mathbb{P}} \left[ S^{(2)} \times \mathbb{E}^{\mathbb{Q}} \left[ S^{(1)} | \mathcal{F}^{(2)} \right] \right]$$

$$= \mathbb{E}^{\mathbb{P}} \left[ S^{(2)} \times \mathbb{E}^{\mathbb{Q}} \left[ S^{(1)} \right] \right]$$

$$= \mathbb{E}^{\mathbb{P}} \left[ S^{(2)} \right] \times \mathbb{E}^{\mathbb{Q}} \left[ S^{(1)} \right]. \tag{5.3}$$

Hence, the actuarial risk is priced via real-world expectation and the financial risk via risk-neutral expectation. In fact, for the Equation (5.3) to hold, it is sufficient that the financial claim  $S^{(1)}$  is independent from the actuarial filtration  $\mathcal{F}^{(2)}$ .

## 5.1.2 Risk margin

In order to motivate the risk margin, we recall that the best estimate is centered around the risk

$$\mathbb{E}^{\mathbb{Q}}\left[S|\mathcal{F}^{(2)}\right].$$

This risk represents the risk-neutral financial price of S conditional on the actuarial information. Looking at the tail of this (actuarial) risk will provide information on the actuarial scenarios which yield the worst financial price. Hence, applying an actuarial valuation on this conditional financial price allows to measure the impact of the actuarial uncertainty on the risk-neutral price. This motivates the following definition.

**Definition 5.2 (SCR for actuarial risk)** For any claim  $S \in \mathcal{C}$  and any actuarial valuation  $\pi^{(2)}$ , the SCR for actuarial risk is given by

$$SCR[S] = \pi^{(2)} \left[ \mathbb{E}^{\mathbb{Q}} \left[ S | \mathcal{F}^{(2)} \right] \right] - \mathbb{E}^{\mathbb{P}} \left[ \mathbb{E}^{\mathbb{Q}} \left[ S | \mathcal{F}^{(2)} \right] \right]. \tag{5.4}$$

It turns out that the SCR appears as a two-step actuarial valuation for which we deducted the best estimate. If the valuation is coherent, thanks to the representation theorem (see Theorem 2.3), the SCR for actuarial risk can be represented as

$$SCR\left[S\right] = \sup_{\tilde{\mathbb{D}}} \left\{ \mathbb{E}^{\tilde{\mathbb{P}}} \left[ \mathbb{E}^{\mathbb{Q}} \left[ S | \mathcal{F}^{(2)} \right] \right] - \mathbb{E}^{\mathbb{P}} \left[ \mathbb{E}^{\mathbb{Q}} \left[ S | \mathcal{F}^{(2)} \right] \right] \right\}$$

where the supremum is taken over a set of probability measures  $\tilde{\mathbb{P}}$  absolutely continuous to  $\mathbb{P}$ . Hence, the SCR for actuarial risk can be interpreted as a worst case scenario: we can consider a family of stressed actuarial models (e.g. different mortality dynamics) and define the SCR as the value under the worst-case model.

#### 5.1.3 Cost-of-capital valuation of insurance liabilities

In Solvency II, the fair value of insurance liabilities is defined as the sum of the best estimate and the risk margin in which the latter is defined as the cost of capital needed to cover the unhedgeable risks.

In the spirit of regulatory directives, we define a cost-of-capital valuation (CoC valuation) based on our two-step actuarial valuation. This CoC valuation is defined as the sum of the best estimate (expected present value) plus the risk margin (cost to cover unhedgeable risks) where the latter represents the cost of providing the SCR for actuarial risk.

**Definition 5.3 (Cost-of-capital valuation)** For any claim  $S \in \mathcal{C}$  and any actuarial valuation  $\pi^{(2)}$ , the cost-of-capital value of S is defined by

$$\rho[S] = BE[S] + iSCR[S] \tag{5.5}$$

with

$$\begin{split} BE[S] &= \mathbb{E}^{\mathbb{P}} \left[ \mathbb{E}^{\mathbb{Q}} \left[ S | \mathcal{F}^{(2)} \right] \right] \\ SCR\left[ S \right] &= \pi^{(2)} \left[ \mathbb{E}^{\mathbb{Q}} \left[ S | \mathcal{F}^{(2)} \right] \right] - \mathbb{E}^{\mathbb{P}} \left[ \mathbb{E}^{\mathbb{Q}} \left[ S | \mathcal{F}^{(2)} \right] \right], \end{split}$$

where i is the cost-of-capital rate and SCR stands for the SCR for actuarial risk.

## 5.2 Numerical application: Portfolio of GMMB contracts

In this subsection, we show how to determine the cost-of-capital value (5.5) for a portfolio of guaranteed minimum maturity benefit (GMMB) contracts underwritten at time 0 on  $l_x$  persons of age x. In particular, we detail the numerical procedure for the best estimate and the SCR for actuarial risk. Moreover, we compare the fair valuation with the setting of Brennan and Schwartz (1976) in which complete diversification of mortality is assumed.

The GMMB contract offers at maturity the greater of a minimum guarantee K and a stock value if the policyholder is still alive at that time. Let  $T_i$  be the remaining lifetime of insured  $i, i = 1, 2, \ldots, l_x$ , at contract initiation. The payoff per policy can be written as

$$S = \frac{L_{x+T}}{l_x} \times \max\left(Y^{(1)}(T), K\right),\tag{5.6}$$

with

$$L_{x+T} = \sum_{i=1}^{l_x} 1_{\{T_i > T\}}.$$

Here,  $L_{x+T}$  is the number of policyholders who survived up to time T and  $Y^{(1)}(T)$  is the value of the stock at time T.

We consider a continuous time setting for the stock and the force of mortality dynamics. Let us assume that the dynamics of the stock process and the population force of mortality are given by

$$dY^{(1)}(t) = Y^{(1)}(t) \left(\mu dt + \sigma dW_1(t)\right) \tag{5.7}$$

$$d\lambda(t) = c\lambda(t)dt + \xi dW_2(t), \tag{5.8}$$

with  $c, \xi, \mu$  and  $\sigma$  are positive constants, and  $W_1(t) = \rho W_2(t) + \sqrt{1-\rho^2}Z(t)$ . Here,  $W_2(t)$  and Z(t) are independent standard Brownian motions. The specification of a non-mean reverting Ornstein-Uhlenbeck (OU) process (5.8) for the mortality model allows negative mortality rates. However, Luciano and Vigna (2008) and Luciano et al. (2017) showed that the probability of negative mortality rates is negligible with calibrated parameters. The benefit of such specification is to allow tractability of mortality rates. Indeed, under Equation (5.8),  $\lambda(t)$  is a Gaussian process and  $\int_0^T \lambda(t) dv$  is normal distributed.

#### 5.2.1 Best-estimate computation

Since we want to determine the best estimate mortality, we assume that there is no risk premium in the actuarial market or, equivalently, that Equation (5.8) holds under  $\mathbb{P}$  and  $\mathbb{Q}$ . Therefore, the calibration of the mortality intensity is performed by estimating its dynamic under the real-world measure, and then using it under the risk-neutral measure. For the stock process, we define

$$dW_1^{\mathbb{Q}}(t) = \frac{\mu - r}{\sigma} dt + dW_1(t),$$

where  $\frac{\mu-r}{\sigma}$  represents the market price of equity risk. We can then write the dynamics under  $\mathbb Q$  as follows

$$dY^{(1)}(t) = Y^{(1)}(t) \left( rdt + \sigma dW_1^{\mathbb{Q}}(t) \right)$$
(5.9)

$$d\lambda(t) = c\lambda(t)dt + \xi dW_2^{\mathbb{Q}}(t). \tag{5.10}$$

The best estimate for the aggregate payoff (5.6) is given by

$$BE[S] = \mathbb{E}^{\mathbb{P}} \left[ \mathbb{E}^{\mathbb{Q}} \left[ e^{-rT} \frac{L_{x+T}}{l_x} \times \max \left( Y^{(1)}(T), K \right) | \mathcal{F}^{(2)} \right] \right].$$

Under the independence assumption between the force of mortality and the stock dynamics, one can easily show that the best estimate simplifies into

$$BE[S] = \mathbb{E}^{\mathbb{P}} \left[ \frac{L_{x+T}}{l_x} \right] \times \mathbb{E}^{\mathbb{Q}} \left[ e^{-rT} \max \left( Y^{(1)}(T), K \right) \right]$$
 (5.11)

$$= Tp_x \left[ Y^{(1)}(0)N(d_1) + Ke^{-rT} \left( 1 - N(d_2) \right) \right]$$
 (5.12)

$$= \underbrace{\mathbb{E}^{\mathbb{P}}\left[e^{-\int_{0}^{T}\lambda(v)dv}\right]}_{Tp_{T}}\left[Y^{(1)}(0)N(d_{1}) + Ke^{-rT}\left(1 - N(d_{2})\right)\right]$$
(5.13)

<sup>&</sup>lt;sup>9</sup>A similar approach is considered in Luciano et al. (2017)

with

$$d_1 = \frac{\ln\left(\frac{Y^{(1)}(0)}{K}\right) + \left(r + \frac{\sigma^2}{2}\right)T}{\sigma\sqrt{T}},$$
  
$$d_2 = d_1 - \sigma\sqrt{T}.$$

We remark that the survival probability  $_Tp_x$  can be obtained in closed-form (for details, see for instance Mamon (2004)):

$$_Tp_x = \mathbb{E}^{\mathbb{P}}\left[e^{-\int_0^T \lambda(v)dv}\right] = e^{A\lambda(0) + \frac{B}{2}},$$

with

$$A = \frac{1}{c} \left( 1 - e^{cT} \right)$$

$$B = \frac{\xi^2}{c^3} \left( cT + \frac{3}{2} - 2e^{cT} + \frac{1}{2}e^{2cT} \right).$$
(5.14)

Under the dependence assumption, we provide in the next proposition the approximated best estimate for the portfolio of GMMB contracts.

**Proposition 1** If we denote by  $_Tp_x^i$  (i=1,...,n) the survival rates for each actuarial scenario<sup>10</sup>, the approximated best estimate for the aggregate payoff of GMMB contracts:

$$\widehat{BE}[S] = \frac{1}{n} \sum_{i=1}^{n} \mathbb{E}^{\mathbb{Q}} \left[ e^{-rT} \frac{L_{x+T}}{l_x} \times \max \left( Y^{(1)}(T), K \right) \middle| L_{x+T} = l_x \,_T p_x^i \right]$$

is given by

$$\widehat{BE}[S] = \frac{1}{n} \sum_{i=1}^{n} T p_x^i \left( \widetilde{Y}^{(1)}(0) N(d_1) + e^{-rT} K \left( 1 - N(d_2) \right) \right), \tag{5.15}$$

with

$$\widetilde{Y}^{(1)}(0) = Y^{(1)}(0)e^{\frac{-\sigma\rho_0\sqrt{T}}{\sqrt{\frac{1}{2c}e^{2cT} - \frac{2}{c}e^{cT} + T + \frac{3}{2c}}} \left(\frac{c}{\xi} \ln T p_x^i + \frac{\lambda(0)}{\xi} \left(e^{cT} - 1\right)\right)} e^{-\frac{1}{2}\sigma^2 \rho_0^2 T},$$

$$\rho_0 = \frac{\rho\left(\frac{1}{c}e^{cT} - \frac{1}{c} - T\right)}{\sqrt{T\left(\frac{1}{2c}e^{2cT} - \frac{2}{c}e^{cT} + T + \frac{3}{2c}\right)}},$$

$$d_1 = \frac{\ln\left(\frac{\widetilde{Y}^{(1)}(0)}{K}\right) + \left(r + \frac{1}{2}\sigma^2(1 - \rho_0^2)\right)T}{\sigma\sqrt{\left(1 - \rho_0^2\right)T}},$$

$$d_2 = d_1 - \sigma\sqrt{\left(1 - \rho_0^2\right)T}.$$

**Proof.** The proof based on classical arguments of stochastic calculus can be found in Appendix A.

The approximated best estimate (5.15) appears as an average of Black-Scholes call option prices which are adjusted for the dependence between the population force of mortality and the stock price processes. In each call option, there is an adjustement of the current stock price  $Y^{(1)}(0)$  to  $\widetilde{S}^{(1)}(0)$ , taking into

<sup>&</sup>lt;sup>10</sup>We assume that the actuarial scenarios are generated by a Monte-Carlo sample of i.i.d. observations

account the realized survival rate  $_Tp_x^i$  in each actuarial scenario. It is also worth noticing that in case of independence ( $\rho = 0$ ), the approximated best estimate (5.15) converges to the best estimate (5.13).

To determine the best estimate (5.15), we only need to generate survival rates  $Tp_x^i$  (i = 1, ..., n) and plug them into the Black-Scholes option pricing formulas. Since the force of mortality dynamics is given by

$$d\lambda(t) = c\lambda(t)dt + \xi dW_2(t),$$

one can prove (for details, see Appendix A) that

$$\ln T p_x = -\int_0^T \lambda(s) ds \sim N\left(\frac{\lambda(0)}{c} \left(e^{cT} - 1\right), \frac{\xi^2}{c^3} \left(\frac{1}{2} e^{2cT} - 2e^{cT} + cT + \frac{3}{2}\right)\right).$$

We generate n=100000 mortality paths. The benchmark parameters for the stock and the force of mortality are given in Table 1. The mortality parameters follow from Luciano et al. (2017) while the financial parameters are based on Bernard and Kwak (2016). The mortality parameters correspond to UK male individuals who are aged 55 at time 0.

Parameter set for numerical analysis			
Force of mortality model: $c = 0.0750, \xi = 0.000597, \lambda(0) = 0.0087.$			
Financial model: $r = 0.02, T = 10, Y^{(1)}(0) = 1, K = 1, \sigma = 0.2.$			

Table 1: Parameter values used in the numerical illustration.

$\rho$	Best estimate	ho	Best estimate
-1.0	1.01132	0	1.00667
-0.9	1.01086	0.1	1.00618
-0.8	1.01041	0.2	1.00568
-0.7	1.00995	0.3	1.00517
-0.6	1.00950	0.4	1.00466
-0.5	1.00904	0.5	1.00414
-0.4	1.00858	0.6	1.00360
-0.3	1.00811	0.7	1.00307
-0.2	1.00764	0.8	1.00252
-0.1	1.00716	0.9	1.00196
		1.0	1.00141

Table 2: Best estimate for the GMMB contract using Equation (5.15).

Table 2 displays the best estimate per policy obtained using Equation (5.15) for a range of correlation coefficients:  $\rho \in [-1,1]$ . We observe that the best estimate slightly decreases with the increase of the correlation parameter. This can be justified by a compensation effect between the mortality and the stock dynamics:

• In case of positive dependence, high mortality scenarios (respectively low mortality scenarios) are linked with high stock values (respectively low stock values). In consequence, the expected value of the claim

$$S = \frac{L_{x+T}}{l_x} \times \max\left(Y^{(1)}(T), K\right)$$

will be reduced since high values of survivals  $L_{x+T}$  will be associated with low financial guarantees,  $\max (Y^{(1)}(T), K)$ , and vice-versa.

• On the other hand, in case of negative dependence, high survival rates will be linked with high financial guarantees, which implies a higher uncertainty and an increase of the best estimate.

## 5.2.2 Cost-of-capital value computation

The cost-of-capital value of the insurance liability S is then determined by

$$\rho[S] = BE[S] + iSCR[S],$$

where the cost-of-capital rate i is fixed at 6%.

Figure 1 represents the CoC value of S for a range of correlation coefficients:  $\rho \in [-1, 1]$ . Overall, we observe an increase of the CoC value of the GMMB contract under dependent mortality and equity risks. However, this effect is less pronounced for positive dependence. By comparison, the fair value of S under the assumption that mortality can be completely diversified (denoted by  $\rho^{\text{B-S}}$  for Brennan and Schwartz (1976)), is given by Equation (5.13):

$$\rho^{\text{B-S}}[S] = \mathbb{E}^{\mathbb{P}}\left[\frac{L_{x+T}}{l_x}\right] \times \mathbb{E}^{\mathbb{Q}}\left[e^{-rT} \max\left(Y^{(1)}(T), K\right)\right]$$

$$= Tp_x \left[Y^{(1)}(0)N(d_1) + Ke^{-rT}\left(1 - N(d_2)\right)\right]$$

$$= 1.0067.$$
(5.16)

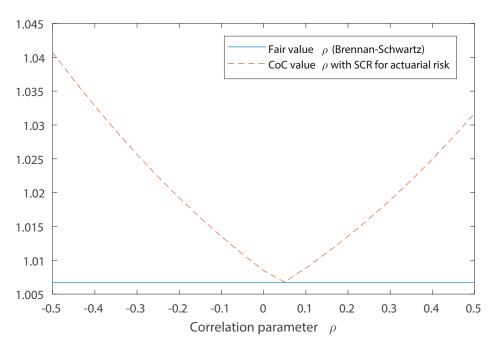


Figure 1: Comparison between the Cost-of-capital value for the GMMB contract under the two-step actuarial approach and the fair value of Brennan and Schwartz (1976).

From Figure 1, we clearly observe that this assumption underestimates the fair value of the contract since it does not take into account the actuarial uncertainty and the possible dependence with the financial market.

## 6 Concluding remarks

In this paper, we have proposed a general actuarial-consistent valuation for insurance liabilities based on a two-step actuarial valuation. Actuarial-consistency requires that traditional actuarial valuation based on diversification applies to all actuarial risks. We have shown that every two-step actuarial valuation is actuarial-consistent and in the coherent setting, any actuarial-consistent valuation has a two-step actuarial valuation representation. We also studied under which conditions it is feasible to define a valuation that is weak actuarial-consistent and market-consistent. In general, it is not possible and the valuator should decide whether the valuation is driven by current market prices or historical actuarial information.

Based on our two-step actuarial valuation, we have defined a cost-of-capital valuation in which the valuation is defined as the sum of a best estimate (expected value) and a risk margin (cost of providing the SCR for actuarial risks). The detailed numerical illustration has shown the important impact on risk management when relaxing the independence assumption between actuarial and financial risks. In an extended B-S financial market, we determined the cost-of-capital value of a GMMB contract under dependent financial and actuarial risks. It turns out that the dependence structure has an important impact on the fair valuation and the related SCR.

As pointed out by Liu et al. (2014), Solvency II Directive highly recommends the testing of capital adequacy requirements on the assumption of mutual dependence between financial markets and life insurance markets. In that respect, we believe that our two-step framework provides a plausible setting for the valuation of insurance liabilities with dependent financial and actuarial risks.

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## A Appendix: Proof of Proposition 1

**Proof.** We recall that the dynamics of the stock process and the population force of mortality under  $\mathbb{Q}$  are given by

$$dY^{(1)}(t) = Y^{(1)}(t) (rdt + \sigma dW_1(t))$$
(A.1)

$$d\lambda(t) = c\lambda(t)dt + \xi dW_2(t) \tag{A.2}$$

with  $c, \xi, \mu$  and  $\sigma_1$  are positive constants, and  $W_1(t) = \rho W_2(t) + \sqrt{1 - \rho^2} Z(t)$ . Here,  $W_2(t)$  and Z(t) are independent standard Brownian motions under  $\mathbb{Q}$ . From (A.2), we note that

$$d\left(e^{-ct}\lambda(t)\right) = -ce^{-ct}\lambda(t)dt + e^{-ct}d\lambda(t)$$
$$= \xi e^{-ct}dW_2(t).$$

Hence, the force of mortality is a Gaussian process:

$$\lambda(t) = \lambda(0)e^{ct} + \xi \int_0^t e^{-c(u-t)}dW_2(u).$$

Moreover, we find that

$$\begin{split} \int_0^T \lambda(s) ds &= \frac{\lambda(0)}{c} \left( e^{cT} - 1 \right) + \xi \int_0^T \int_0^s e^{-c(u-s)} dW_2(u) ds \\ &= \frac{\lambda(0)}{c} \left( e^{cT} - 1 \right) + \xi \int_0^T \int_u^T e^{-c(u-s)} ds dW_2(u) \\ &= \frac{\lambda(0)}{c} \left( e^{cT} - 1 \right) + \frac{\xi}{c} \int_0^T \left( e^{-c(u-T)} - 1 \right) dW_2(u) \\ &= \frac{\lambda(0)}{c} \left( e^{cT} - 1 \right) + \frac{\xi}{c} X_T, \end{split}$$

with

$$X_T = \int_0^T \left( e^{-c(u-T)} - 1 \right) dW_2(u) \sim N\left(0, \frac{1}{2c}e^{2cT} - \frac{2}{c}e^{cT} + T + \frac{3}{2c} \right).$$

We can also remark that

$$\mathbb{E}\left(W_1(T)X_T\right) = \mathbb{E}\left(\int_0^T dW_1(u) \int_0^T \left(e^{-c(u-T)} - 1\right) dW_2(u)\right)$$
$$= \rho\left(\frac{1}{c}e^{cT} - \frac{1}{c} - T\right),$$

which leads to

$$corr(W_1(T), X_T) = \frac{\rho(\frac{1}{c}e^{cT} - \frac{1}{c} - T)}{\sqrt{T(\frac{1}{2c}e^{2cT} - \frac{2}{c}e^{cT} + T + \frac{3}{2c})}} \equiv \rho_0.$$

We can then assume that

$$W_1(T) = \frac{\rho_0 \sqrt{T}}{\sqrt{\frac{1}{2c}e^{2cT} - \frac{2}{c}e^{cT} + T + \frac{3}{2c}}} X_T + \sqrt{T(1 - \rho_0^2)} Z,$$

where Z is a standard normal r.v. independent of  $X_T$ .

From

$$e^{-\int_0^T \lambda(s)ds} = {}_T p_x^i,$$

we find that

$$X_T = -\frac{c}{\xi} \ln T p_x^i - \frac{\lambda(0)}{\xi} \left( e^{cT} - 1 \right).$$

The stock price at time T can be written as

$$\begin{split} Y^{(1)}(T) &= Y^{(1)}(0)e^{(r-\frac{1}{2}\sigma^2)T + \sigma W_1(T)} \\ &= Y^{(1)}(0)e^{\frac{-\sigma\rho_0\sqrt{T}}{\sqrt{\frac{1}{2c}e^{2cT} - \frac{2}{c}e^{cT} + T + \frac{3}{2c}}} \left(\frac{c}{\xi} \ln T p_x^i + \frac{\lambda(0)}{\xi} (e^{cT} - 1)\right)} e^{(r - \frac{1}{2}\sigma^2)T + \sigma\sqrt{1 - \rho_0^2}\sqrt{T}Z} \\ &= \widetilde{S}^{(1)}(0)e^{\left(r - \frac{1}{2}\sigma^2\left(1 - \rho_0^2\right)\right)T + \sigma\sqrt{1 - \rho_0^2}\sqrt{T}Z}, \end{split}$$

with

$$\widetilde{S}^{(1)}(0) = Y^{(1)}(0)e^{\frac{-\sigma\rho_0\sqrt{T}}{\sqrt{\frac{1}{2c}e^{2cT} - \frac{2}{c}e^{cT} + T + \frac{3}{2c}}} \left(\frac{c}{\xi} \ln \tau p_x^i + \frac{\lambda(0)}{\xi} (e^{cT} - 1)\right)} e^{-\frac{1}{2}\sigma^2\rho_0^2T}.$$

Finally, we find that

$$\mathbb{E}^{\mathbb{Q}} \left[ e^{-rT} L_{x+T} \times \max \left( Y^{(1)}(T), K \right) \middle| e^{-\int_{0}^{T} \lambda(s) ds} = T p_{x}^{i} \right]$$

$$= l_{x} T p_{x}^{i} \mathbb{E}^{\mathbb{Q}} \left[ e^{-rT} K + e^{-rT} \max \left( Y^{(1)}(T) - K, 0 \right) \middle| e^{-\int_{0}^{T} \lambda(s) ds} = T p_{x}^{i} \right]$$

$$= l_{x} T p_{x}^{i} \left( \widetilde{S}^{(1)}(0) N(d_{1}) + e^{-rT} K \left( 1 - N(d_{2}) \right) \right),$$

which ends the proof.