Actuarial-consistency and two-step actuarial valuations: a new paradigm to insurance valuation

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ABSTRACT
This paper introduces new valuation schemes called actuarial-consistent valuations 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 case valuations are coherent, 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, A. & Stadje, M. (2014). Time-consistent and market-consistent evaluations. Mathematical Finance 24(1), 25–65. In particular, we discuss how the dependence structure between actuarial and financial risks impacts both actuarial-consistent and market-consistent valuations.

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, an actuarial premium principle (called actuarial valuation in this paper) is typically based on a diversification argument which justifies applying the law of large numbers (LLN) among independent policyholders who face identical risks (see Denuit et al. 2006). Consistent with insurance regulation, the actuarial valuation \( \rho(S) \) of a discounted claim \( S \) is represented as the expectation under the real-world probability measure \( P \) plus an additional risk margin to cover any undiversified and systematic risk, that is

\[
\rho(S) = E_P[S] + \text{Risk margin. (1)}
\]

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

$$\rho(S) = E^Q [S].$$

(2)

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

Insurance claims are nowadays non-trivial combinations of diversifiable and undiversifiable insurance risk, and traded financial risks. It is therefore primordial to build a valuation framework which combines traditional actuarial and financial valuation. Over the last two decades, several researchers have worked on the interplay between financial and actuarial valuation. Embrechts (2000) offers a detailed comparison of insurance and finance pricing mechanisms. Møller (2002) addresses the aspects of the interplay between finance and insurance by combining traditional actuarial and financial pricing principles. A simplifying approach in the actuarial literature is to assume independence between actuarial and financial risks1 such that the valuation can be split into a product of actuarial and financial valuations (see Fung et al. 2014, Ignatieva et al. 2016, Wüthrich 2016, Da Fonseca & Ziveyi 2017 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 (Sharif et al. 2020, Harjoto et al. 2021). For this reason, different authors proposed general valuation approaches allowing for dependencies between actuarial and financial risks. For instance, valuation under dependent mortality and interest risks was investigated in Liu et al. (2014), Deelstra et al. (2016), and Zhao & Mamon (2018). Moreover, Pelsser & 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 & 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).

11. Market-consistent valuation and its shortfalls

The above-mentioned papers propose different valuation principles which have in common that they are all market-consistent valuations. The Solvency II insurance regulation directive introduced a prospective and risk-based supervisory approach on January 1, 2016. Pillar 1 of this directive requires 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, concerns about the appropriateness of the market-consistent valuation for long-term insurance business were raised:

- Le Courtois et al. (2021) pointed out market-consistency can lead to a substantial misvaluation of the effective wealth of an insurance company that deals with long-term commitments. This approach also induces high instability and excess volatility in the balance sheet indicators of an insurance company (Vedani et al. 2017, Rae et al. 2018).

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1 This assumption is either made under the real-world measure $P$ or the risk-neutral measure $Q$. We note that the independence under $P$ does not necessarily imply the independence under $Q$; see Dhaene et al. (2013). We also discuss this point in Lemma 3.3.
• Plantin et al. (2008) found that the damage done by marking-to-market (that is market-consistency) is greatest when claims are illiquid and long term, which is precisely the case of balance sheet of insurance companies.
• Market-consistency tends to be pro-cyclical and the use of a 1-year Value-at-Risk increases the risk of herd behavior, hence reducing financial stability. Market-consistent valuation tends to minimise the value of liabilities when markets are bullish and over-estimate them in times of crisis (Rae et al. 2018).

1.2. Alternative valuations to market-consistency
The second pillar of the Solvency II directive allows the insurer to proceed to an alternative assessment of the company’s overall solvency needs using different recognition and valuation bases. In this purpose, 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. The framework is not anymore market-consistent but market-implied and utility-consistent. The authors found that this alternative valuation provides less volatility than the traditional market-consistent approach. Muermann (2003) investigates the valuation of catastrophe derivatives and proposes an actuarial-consistent valuation approach that is consistent with existing insurance premiums to exclude arbitrage opportunities.

1.3. Why actuarial-consistency is an important alternative?
In the spirit of Muermann (2003) and Le Courtois et al. (2021), 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 actuarial claim (such as a pure endowment) should be priced via an actuarial valuation and should not be managed using a risky investment. We will label this property of the valuation ‘actuarial-consistency’. Such property plays an important role in life insurance where the development of the longevity market and the longevity-linked securities is growing in the recent years (Blake & Cairns 2020). Indeed, in several countries, the valuation of the longevity risk is often required to follow regulatory life tables that impose minimum loadings for actuarial claims. Under a market-consistent setting, the presence of undervalued longevity-linked securities might create a potential undervaluation of life insurance business. Therefore, we believe it is relevant, for certain insurance claims, to investigate valuation mechanisms that are not market-consistent and would encourage risk-free investments rather than longevity transfers to the capital market.

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, if the valuation is coherent, we show the reciprocal: any actuarial-consistent valuation has a two-step actuarial representation. 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.

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2 As pointed out in Blake & Cairns (2020), the Prudential Regulatory Authority (regulatory authority for insurance companies in the UK) expressed concerns that too much longevity risk is transferred offshore so that if the offshore reinsurance firm failed, UK pensioners might not get their pensions.
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 we can define a best-estimate based on the two-step actuarial valuation. This 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 3.3, we provide a detailed comparison between actuarial-consistent valuations and market-consistent valuations. Section 4 presents a best estimate valuation based on the two-step actuarial valuation and a detailed numerical application of the two-step actuarial valuation on a portfolio of equity-linked contracts. Section 5 concludes the paper.

2. Actuarial and financial valuations

All random variables introduced hereafter are defined on the probability space \((\Omega, \mathcal{F}, \mathbb{P})\). Equalities and inequalities between r.v.’s have to be understood in the \(\mathbb{P}\)-almost sure sense. The space of bounded random variables is denoted by \(L^{\infty}(\Omega, \mathcal{F}, \mathbb{P})\) or \(L^{\infty}(\mathcal{F})\) for short. In this paper, we consider a one-period setting. A contingent claim is a random liability of an insurance company that has to be paid at the deterministic future time \(T\). Formally, a discounted claim is modeled by the random variable \(S \in L^{\infty}(\mathcal{F})\). In what follows we are interested in the valuation of discounted claims.

Suppose a \(\sigma\)-algebra \(G \subset \mathcal{F}\) is the information available to the agent. We define a \(G\)-conditional valuation as follows.

**Definition 2.1 (G-conditional valuation):** A \(G\)-conditional valuation is a mapping \(\Pi[\cdot|G] : L^{\infty}(\mathcal{F}) \rightarrow L^{\infty}(G)\) satisfying the following properties:

- **Normalization:** \(\Pi[0|G] = 0\).
- **Translation-invariance:** For any \(S \in L^{\infty}(\mathcal{F})\) and \(\lambda \in L^{\infty}(G)\), we have
  \[
  \Pi[S + \lambda|G] = \Pi[S|G] + \lambda.
  \]
- **Convexity:** For any \(S_1, S_2 \in L^{\infty}(\mathcal{F})\) and \(\lambda \in L^{\infty}(G)\) with \(0 \leq \lambda \leq 1\), we have
  \[
  \Pi[\lambda S_1 + (1 - \lambda)S_2|G] \leq \lambda \Pi[S_1|G] + (1 - \lambda)\Pi[S_2|G].
  \]
- **Positive homogeneity:** For any \(S \in L^{\infty}(\mathcal{F})\) and any positive \(\lambda \in L^{\infty}(G)\), we have
  \[
  \Pi[\lambda S|G] = \lambda \Pi[S|G].
  \]

Note that if \(G\) is chosen to be the trivial \(\sigma\)-algebra, then \(\Pi[\cdot|G]\) is a real number and we simply write the valuation \(\Pi[\cdot]\) without conditioning on \(G\).

Apart from the properties in Definition 2.1, other properties that a valuation may have are

- **Monotonicity:** for any \(S, R \in L^{\infty}(\mathcal{F})\), \(\Pi(S) \leq \Pi(R)\) if \(S \leq R\).
- **Subadditivity:** for any \(S, R \in L^{\infty}(\mathcal{F})\), \(\Pi(S + R) \leq \Pi(S) + \Pi(R)\).
- **Fatou property:** for \(S, S_1, S_2, \ldots \in L^{\infty}(\mathcal{F})\), \(\sup_{n \in \mathbb{N}} \|S_n\|_{\infty} < \infty\), where \(\| \cdot \|_{\infty} = \text{ess sup}(\| \cdot \|)\), and \(S_n \overset{a.s.}{\rightarrow} S\), then
  \[
  \lim \inf_{n \rightarrow \infty} \Pi(S_n) \geq \Pi[S].
  \]

According to Artzner et al. (1999), coherent valuations (also known as coherent risk measures) are defined as follows.
Definition 2.2 (Coherent valuation): A coherent valuation is a valuation which is translation-invariant, positive homogeneous, monotone and subadditive.

A few dual representations of coherent valuations are available in the literature. Here we present the most popular version, which is defined for bounded random variables on a general space.

Proposition 2.1 (Delbaen & Biagini 2000): A coherent valuation with Fatou property \( \Pi : L^{\infty}(\mathcal{F}) \to \mathbb{R} \) has the following representation for any \( S \in L^{\infty}(\mathcal{F}) \)

\[
\Pi[S] = \sup_{Q \in \mathcal{M}} \mathbb{E}^Q[S],
\]

where \( \mathcal{M} \) is a collection of probability measures absolutely continuous w.r.t. \( \mathbb{P} \).

From the above dual representation, coherent valuations can 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.

Moreover, we also introduce linear valuations in the following sense.

Definition 2.3 (Linear valuation): A valuation \( \Pi : L^{\infty}(\mathcal{F}) \to \mathbb{R} \) is linear, if there exists a finitely additive measure \( Q \) absolutely continuous w.r.t. \( \mathbb{P} \), \( Q(\Omega) = 1 \), such that for any \( S \in L^{\infty}(\mathcal{F}) \)

\[
\Pi[S] = \mathbb{E}^Q[S].
\]

In this paper, we assume risks can be divided in two groups: financial and actuarial risks. Then a hybrid discounted claim \( S \in L^{\infty}(\mathcal{F}) \) is a combination of both the actuarial as well as the financial 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.

2.1. Financial valuation

We assume there is a financial market with \( n^{(1)} + 1 \) traded assets and denote by \( Y = (Y_0, Y_1, \ldots, Y_{n^{(1)}}) \) the vector of prices of assets at time \( T \), where \( Y_0 \) is the risk-free asset assumed to be constant and equal to 1. We also refer to the traded assets as financial risks. The \( \sigma \)-algebra generated by the financial risks \( Y \) is denoted by \( \mathcal{F}^{(1)} \subset \mathcal{F} \).

A financial claim is an \( \mathcal{F}^{(1)} \)-measurable random variable. 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 discounted financial claims is denoted by \( L^{\infty}(\mathcal{F}^{(1)}) \). We can always express a discounted financial claim as a function of the financial risks. We have:

\[
S^{(1)} = f(Y), \quad \text{if } S^{(1)} \in L^{\infty}(\mathcal{F}^{(1)}),
\]

for some function \( f^3 \).

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. Moreover, we assume that the price at which one

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3 We assume that all functions we encounter are Borel measurable.
can buy and sell is equal and that a financial valuation principle \( \pi^{(1)} \) is available to determine the price of the traded payoffs:

\[
\pi^{(1)} : L^\infty(\mathcal{F}^{(1)}) \to \mathbb{R}.
\]

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 & 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, frictionless and that all financial assets are discretely 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) \( Q \) (Dalang et al. 1990). 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:

\[
\pi^{(1)}[S] = E_Q[S].
\]

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

(2) **Imperfect market and bid-ask prices.** In classical finance, markets are usually modeled 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 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:

\[
\pi^{(1)}[S] = \sup_{Q \in \mathcal{Q}} E_Q[S],
\]

where \( \mathcal{Q} \) is a convex set of probability measures absolutely continuous with respect to the original probability. For more details on conic finance, we refer to Madan & Cherny (2010) and Madan & Schoutens (2016).

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

### 2.2. Actuarial valuation

Suppose there are \( n^{(2)} \) non-traded risks (also called actuarial risks) which we denote by \( X = (X_1, \ldots, X_{n^{(2)}}) \) the vector of prices of the actuarial risks at time \( T \). The \( \sigma \)-algebra generated by \( X \) is denoted by \( \mathcal{F}^{(2)} \subset \mathcal{F} \).

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

$$S^{(2)} = f(X), \quad \text{if } S^{(2)} \in L^\infty(\mathcal{F}^{(2)}),$$

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. A completely diversifiable portfolio can be valued with its expectation under the physical measure $\mathbb{P}$. However, there is always an amount of residual actuarial risk present because one cannot aggregate an infinite amount of policies. Moreover, there are also systematic actuarial risks (e.g. longevity risk) which cannot be diversified away. Indeed, aggregating a large number of systematic risks will not result in the desired risk reduction. For this reason, actuarial valuations include a risk margin to cover any non-diversifiable risk. Below, we briefly discuss the most important actuarial valuation principles (also called actuarial premium principles).

1) **Linear principle:**

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

The risk margin is modeled by an appropriate change of measure from $\mathbb{P}$ to $\tilde{\mathbb{P}}$. 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 Norberg (2014) and Wüthrich (2016).

2) **Standard deviation principle:**

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

with $\beta \geq 0$.

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 recall that the coherent valuation can also be represented as a supremum of a set of measures (see Proposition 2.1). 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 $S \in L^\infty(\mathcal{F})$ is defined on the probability space $(\Omega, \mathcal{F}, \mathbb{P})$. The $\sigma$-algebra $\mathcal{F}$ contains $\mathcal{F}^{(1)}$ and $\mathcal{F}^{(2)}$. That is $(\mathcal{F}^{(1)} \cup \mathcal{F}^{(2)}) \subset \mathcal{F}$.

Assume the valuation $\Pi : L^\infty(\mathcal{F}) \rightarrow \mathbb{R}$ is used to valuate discounted claims in $L^\infty(\mathcal{F})$. A general claim in $L^\infty(\mathcal{F})$ can contain both financial and actuarial risks and therefore the valuation of claims in $L^\infty(\mathcal{F})$ cannot be solely based on the financial or the actuarial valuation principles. Moreover, the financial and actuarial risks are dependent. Therefore, observing the values of financial (resp. actuarial) claims can provide information about the valuation of actuarial (resp. financial) claims.
Define by \( L^\infty(F^{(1,\perp)}) \) the set of financial claims which are independent of the actuarial risks and by \( L^\infty(F^{(2,\perp)}) \) the actuarial claims which are independent of the financial information:

\[
S^{(1,\perp)} \in L^\infty(F^{(1,\perp)}) \quad \text{if} \quad S^{(1,\perp)} \in L^\infty(F^{(1)}) \quad \text{and} \quad S^{(1,\perp)} \perp X,
\]
\[
S^{(2,\perp)} \in L^\infty(F^{(2,\perp)}) \quad \text{if} \quad S^{(2,\perp)} \in L^\infty(F^{(2)}) \quad \text{and} \quad S^{(2,\perp)} \perp Y.
\]

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 \( \Pi : L^\infty(F) \to \mathbb{R} \) is said to be orthogonal consistent with the financial valuation \( \pi^{(1)} \) and the actuarial valuation \( \pi^{(2)} \) if we have that for \( i = 1, 2 \),

\[
\Pi\left[S^{(i,\perp)}\right] = \pi^{(i)}\left[S^{(i,\perp)}\right], \quad \text{if} \quad S^{(i,\perp)} \in L^\infty(F^{(i,\perp)}).
\]

(5)

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

In this paper, a hybrid claim is a general \( F \)-measurable claim, which may depend on actuarial and/or financial information i.e. a hybrid claim \( S \) can be expressed as follows:

\[
S \text{ is a hybrid claim} \leftrightarrow S \in L^\infty(F).
\]

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.

### 3. Actuarial-consistent 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.

This section starts with introducing the concept of *actuarial-consistency*. Condition (5) 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 actuarial-consistent if all actuarial claims are priced with an actuarial valuation, even the ones that may be dependent to financial information.

**Definition 3.1 (Actuarial-consistency):** A valuation \( \Pi \) is called actuarial consistent (ACV) with an actuarial valuation \( \pi^{(2)} \) if for any actuarial claim \( S^{(2)} \in L^\infty(F^{(2)}) \) the following holds:

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

(6)

Actuarial consistency postulates that an actuarial valuation is applied for all actuarial claims. Note that actuarial consistency is stronger than condition (5), 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 $\mathcal{F}^{(1)}$.

In this section, we define two new classes of actuarial-consistent valuations. The first class are the actuarial hedge-based valuations and the second class are the two-step actuarial valuations.

3.1. Actuarial hedge-based valuations

Since we consider a one-period setting, we value a discounted claim and determine a corresponding risk management strategy.

We start with introducing the 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 in a one-period setting. This result was further generalized in Barigou & Dhaene (2019) and Chen et al. (2021). 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 $(\nu_0, \nu_1, \ldots, \nu_n)$ where the component $\nu_i$ denotes the number of units invested in asset $i$ at time $t = 0$ until maturity with the asset 0 is the risk-free asset with constant interest rate $r$. We denote the set of all these static trading strategies by $\Theta_1$.

**Definition 3.2:** A hedger $\theta : L^\infty(\mathcal{F}) \to \Theta$ is a function which maps a claim $S \in L^\infty(\mathcal{F})$ into a trading strategy $\theta_S$ and satisfies the following conditions

1. $\theta$ is normalized: $\theta_0 = (0, 0, \ldots, 0)$.
2. $\theta$ is translation invariant: $\theta_{S+a} = \theta_S + (a, 0, \ldots, 0)$, for any $S \in L^\infty(\mathcal{F})$ and $a \in \mathbb{R}$.

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

**Definition 3.3:** A hedger is said to be an actuarial-consistent hedger if there exists a valuation $\pi$ such that

$$\theta_{S^{(2)}} = \left( \pi [S^{(2)}], 0, 0, \ldots, 0 \right), \quad \text{for any } S^{(2)} \in L^\infty(\mathcal{F}^{(2)}).$$

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

1. Consider the hedger $\theta$ such that

$$\theta_S = \begin{cases} (\pi^{(2)}[S], 0, 0, \ldots, 0), & \text{for } S \in L^\infty(\mathcal{F}^{(2)}) \\ \arg \min_{\mu \in \Theta} \mathbb{E}^P [(S - \mu \cdot Y)^2], & \text{for } S \in L^\infty(\mathcal{F}) \setminus L^\infty(\mathcal{F}^{(2)}), \end{cases}$$

where $Y$ is the financial risks defined in Subsection 2.1 and $\Theta$ is the set of all hedging strategies, that is the set of all $(n^{(1)} + 1)$- dimensional real-valued vectors. Such hedger invests risk-free for actuarial claims and invests following quadratic hedging for all remaining claims. By
construction, such hedger is actuarial consistent. For details on risk-minimizing strategies for insurance processes, we refer to Møller (2001) and Delong et al. (2019).

(2) Assume that an insurer considers the two-step actuarial valuation for any claim $S$:

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

where $\pi [\cdot \mid \mathcal{F}^{(2)}] : L^\infty (\mathcal{F}) \rightarrow L^\infty (\mathcal{F}^{(2)})$ is an $\mathcal{F}^{(2)}$-conditional valuation and $\pi^{(2)}$ is an actuarial valuation.

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

$$
\theta_S = (\Pi [S], 0, 0, \ldots, 0), \quad \text{for any } S \in L^\infty (\mathcal{F}).
$$

(7) 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 3.1:** Consider a hybrid claim $S \in L^\infty (\mathcal{F})$ and a hedger $\tilde{\theta}$. We decompose the claim $S$ into two parts as follows:

$$
H^{(2)}_S = \mathbb{E} \left[ S \mid \mathcal{F}^{(2)} \right] - \mathbb{E} [S],
$$

$$
H^{(1)}_S = S - H^{(2)}_S.
$$

Note that $H^{(2)}_S \in L^\infty (\mathcal{F}^{(2)})$. Define the hedger $\theta_S$ as follows:

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

Then, $\theta_S$ is an actuarial hedger.

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

$$
\theta_{S^{(2)}} = \tilde{\theta}_{H^{(2)}_S} = \left( \mathbb{E} \left[ S^{(2)} \right], 0, 0, \ldots, 0 \right),
$$

which shows we have an actuarial-consistent hedger. ■

Lemma 3.1 illustrates how one can derive an actuarial-consistent hedger using a general hedger. The claim $H^{(2)}_S$ 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 actuarial consistent if the hedger is actuarial consistent.

**Lemma 3.2:** The following two statements are equivalent.

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

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

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

2. $\Pi$ is actuarial consistent.
Proof: Assume \( \theta \) is an actuarial-consistent hedger and the valuation \( \Pi \) is given by (8). Then for any actuarial claim \( S^{(2)} \in L^\infty(\mathcal{F}^{(2)}) \), we have that \( \theta_{S^{(2)}} \cdot Y = \pi^{(2)}[S^{(2)}Y_0] \). Taking into account \( Y_0 = 1 \), we find that \( \Pi \) is actuarial consistent. Assume now that \( \Pi \) is an actuarial-consistent valuation. Defining \( \theta \) as in (7) shows that \( \Pi[S] = \pi^{(1)}[\theta \cdot Y] \), 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 3.4:** A valuation \( \Pi \) is called an actuarial hedge-based valuation with financial valuation \( \pi^{(1)} \) and actuarial-consistent valuation \( \pi \) if it can be expressed as follows:

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

where \( \theta \) is an actuarial-consistent hedger.

In the following theorem, we prove that the class of actuarial-consistent valuations is equal to the class of actuarial hedge-based valuations.

**Theorem 3.1:** A valuation \( \Pi \) is an 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

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

where \( \theta \) is an actuarial hedger. Consider an actuarial claim \( S^{(2)} \in L^\infty(\mathcal{F}^{(2)}) \). Then

\[
\theta_{S^{(2)}} = \left( \pi^{(2)}\left[ S^{(2)} \right], 0, 0, \ldots, 0 \right),
\]

for some actuarial valuation \( \pi^{(2)} \). Then it is straightforward to verify that

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

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

Assume now that \( \Pi \) is an actuarial-consistent valuation. Then it follows from Lemma 3.2 that

\[
\Pi[S] = \pi^{(1)}[\theta \cdot Y],
\]

where \( \theta = (\Pi[S], 0, \ldots, 0) \). Define the valuation \( \Pi' \) as follows:

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

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

We remark that all the results in this subsection are satisfied for general valuations that are only normalized and translation-invariant. Positive homogeneity and convexity properties, which are assumed for a valuation in Definition 2.1, are not necessary for the equivalence result of Theorem 3.1 to hold.
3.2. 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. $\pi [S | \mathcal{F}^{(2)}]$ . Since this conditional payoff depends only on actuarial scenarios and is then $\mathcal{F}^{(2)}$-measurable, in the second step the quantity $\pi [S | \mathcal{F}^{(2)}]$ should be valuated via a standard actuarial valuation $\pi^{(2)}$ . By conditioning on actuarial scenarios, the two-step actuarial valuations possess appealing properties. In particular, we show in Theorem 3.2 that any two-step actuarial valuation is actuarial-consistent and provide a characterization result.

**Definition 3.5 (Two-step actuarial valuation)**: The valuation $\Pi_1$ is called a two-step actuarial valuation if there exists an $\mathcal{F}^{(2)}$-conditional valuation $\pi^{(2)}$ : $L^{\infty}(\mathcal{F}) \rightarrow L^{\infty}(\mathcal{F}^{(2)})$ such that

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

where $\pi^{(2)} : L^{\infty}(\mathcal{F}^{(2)}) \rightarrow \mathbb{R}$ is an actuarial valuation.

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

**Example 3.2**: As an example of a two-step actuarial valuation, we note that Møller (2002) proposed a modified version of the standard deviation principle:

$$\Pi [S] = \mathbb{E}^Q [S] + a \left( \mathbb{V}a^{r} \left[ \mathbb{E}^Q \left[ S | \mathcal{F}^{(2)} \right] \right] \right)^{1/2},$$

which corresponds to applying the traditional standard deviation principle to the no-arbitrage price of $S$ conditional on the actuarial filtration (see Equation (5.5) in Møller (2002)).

In the following theorem, we show that any two-step actuarial valuation is actuarial-consistent. Moreover, if the valuation is coherent, we provide a characterization of the two-step actuarial valuation.

**Theorem 3.2 (Characterization of actuarial consistency)**: The following properties hold:

1. Any two-step actuarial valuation $\Pi$ is actuarial-consistent.
2. If $\Pi$ is a coherent and actuarial-consistent valuation with a linear actuarial valuation $\pi^{(2)} : L^{\infty}(\mathcal{F}^{(2)}) \rightarrow \mathbb{R}$, then there exists a $\mathcal{F}^{(2)}$-conditional coherent valuation $\pi^{(2)} [\cdot | \mathcal{F}^{(2)}] : L^{\infty}(\mathcal{F}) \rightarrow L^{\infty}(\mathcal{F}^{(2)})$ such that for any $S \in L^{\infty}(\mathcal{F})$,

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

where

$$\pi [S | \mathcal{F}^{(2)}] = \text{ess sup}_{Z \in \mathcal{R}} \mathbb{E}[ZS | \mathcal{F}^{(2)}],$$

with $\mathcal{R} = \{ \xi \in L^1(\mathcal{F}) : \mathbb{E}[\xi | \mathcal{F}^{(2)}] = 1 \}$.

**Proof**: (1) To prove that $\Pi$ is actuarial consistent, it is sufficient to notice that for any $S^{(2)} \in L^{\infty}(\mathcal{F}^{(2)})$,

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

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

where we have used that $S^{(2)}$ is $\mathcal{F}^{(2)}$-measurable.
(2) Because \( \Pi \) is coherent, we have that
\[
\Pi[S] = \sup_{\xi \in \mathcal{M}} \mathbb{E}[\xi S].
\]

Since \( \pi^{(2)} \) is linear, by Definition 2.3 there exists a probability measure \( Q \) absolutely continuous w.r.t. \( \mathbb{P} \) such that for any \( S^{(2)} \in L^\infty(\mathcal{F}^{(2)}) \)
\[
\pi^{(2)}[S^{(2)}] = \mathbb{E}^Q[S^{(2)}],
\]
which means equivalently there exists a density \( \varphi^{(2)} \in \mathcal{M}^{(2)}=[\xi \in L^1(\mathcal{F}^{(2)}): \mathbb{E}[\xi]=1] \) such that
\[
\pi^{(2)}[S^{(2)}] = \mathbb{E}^Q[S^{(2)}] = \mathbb{E}[\varphi^{(2)}S^{(2)}].
\]
By the same arguments in the proof of Proposition 3.3 of Pelsser & Stadje (2014), we can decompose \( \xi \) as \( \xi = \varphi^{(2)}Z \) with \( Z \in \mathcal{R} = [\xi \in L^1(\mathcal{F}): \mathbb{E}[\xi | \mathcal{F}^{(2)}]=1]. \) Thus
\[
\Pi[S] = \sup_{\xi \in \mathcal{M}} \mathbb{E}[\xi S] \\
= \sup_{Z \in \mathcal{R}} \mathbb{E}[\varphi^{(2)}ZS] \\
= \sup_{Z \in \mathcal{R}} \mathbb{E}[\varphi^{(2)}\mathbb{E}[ZS | \mathcal{F}^{(2)}]].
\]

By the same arguments in the proof of Theorem 3.10 of Pelsser & Stadje (2014), we can decompose
\[
\Pi[S | \mathcal{F}^{(2)}] = \text{ess sup}_{Z \in \mathcal{R}} \mathbb{E}[ZS | \mathcal{F}^{(2)}].
\]

Now we define \( \pi [S | \mathcal{F}^{(2)}] = \text{ess sup}_{Z \in \mathcal{R}} \mathbb{E}[ZS | \mathcal{F}^{(2)}] \). It is straightforward to verify that \( \pi [S | \mathcal{F}^{(2)}] \) is a \( \mathcal{F}^{(2)} \)-conditional valuation. All we are left to show is
\[
\sup_{Z \in \mathcal{R}} \mathbb{E}[\varphi^{(2)}\mathbb{E}[ZS | \mathcal{F}^{(2)}]] = \mathbb{E}[\varphi^{(2)}\text{ess sup}_{Z \in \mathcal{R}} \mathbb{E}[ZS | \mathcal{F}^{(2)}]].
\]
This can be proved using the same arguments in the proof of Theorem 3.10 of Pelsser & Stadje (2014).

Combining Theorems 3.1 and 3.2, we find the following representation corollary.

**Corollary 3.1:** The following holds:

1. Any two-step actuarial and any actuarial hedge-based valuation is actuarial-consistent.
2. Any actuarial-consistent valuation is an actuarial hedge-based valuation.

### 3.3. Comparisons of two-step actuarial valuation and two-step financial valuation

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.

**Definition 3.6 (Strong market-consistency):** A valuation \( \Pi \) is called strong market-consistent (strong MCV) if for any financial claim \( S^{(1)} \) the following holds:
\[
\Pi[S + S^{(1)}] = \Pi[S] + \pi^{(1)}[S^{(1)}].
\]
In the literature, market-consistency is usually defined via a condition identical or similar to the condition (11) (see e.g. Pelsser & Stadje (2014), Dhaene et al. (2017) and Barigou et al. (2019)). When the financial valuation $\pi^{(1)}$ is linear, Proposition 3.3 of Pelsser & Stadje (2014) shows that the strong market-consistency is equivalent to the following weak market-consistency.

**Definition 3.7 (Weak market-consistency):** A valuation $\Pi$ is called weak market-consistent (weak MCV) if for any financial claim $S^{(1)}$ the following holds:

$$\Pi \left[ S^{(1)} \right] = \pi^{(1)} \left[ S^{(1)} \right]. \quad (12)$$

We remark that Assa & Gospodinov (2018) also investigated these two types of market-consistency, that they called market-consistency of type I and type II.

Following the definition of the two-step market valuation in Pelsser & Stadje (2014), we define the class of two-step financial valuations.

**Definition 3.8 (Two-step financial valuation):** The valuation $\Pi$ is called a two-step financial valuation if there exists an $\mathcal{F}^{(1)}$-conditional valuation $\pi \left[ \cdot \mid \mathcal{F}^{(1)} \right] : L^\infty (\mathcal{F}) \rightarrow L^\infty (\mathcal{F}^{(1)})$ such that

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

where $\pi^{(1)} : L^\infty (\mathcal{F}^{(1)}) \rightarrow \mathbb{R}$ is a financial valuation.

After having defined two broad classes of two-step 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 3.9 (Fair valuation):** A two-step valuation $\Pi$ is fair if it is weak market consistent and 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 when financial risks and actuarial risks are independent, then the two-step financial valuation and the two-step actuarial valuation coincide. We emphasize that the following lemma holds true when the second step valuation is linear. For example, if $\Pi$ is a two-step actuarial valuation, then the second step $\pi^{(2)}$ can be a linear principle, a standard deviation principle or a coherent valuation; if $\Pi$ is a two-step financial valuation, then the second step $\pi^{(1)}$ can be the non-arbitrage pricing valuation given the financial market is arbitrage free and complete.

**Lemma 3.3:** Assume that $\mathcal{F}^{(1)}$ and $\mathcal{F}^{(2)}$ are independent and for any $S \in L^\infty (\mathcal{F})$ it can be expressed as $S = S^{(1)} S^{(2)}$ with $S^{(1)} \in L^\infty (\mathcal{F}^{(1)})$ and $S^{(2)} \in L^\infty (\mathcal{F}^{(2)})$. If a coherent two-step valuation $\Pi : L^\infty (\mathcal{F}) \rightarrow \mathbb{R}$ with a linear second step valuation is a fair valuation, then

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

where $\pi^{(1)} : L^\infty (\mathcal{F}^{(1)}) \rightarrow \mathbb{R}$ and $\pi^{(2)} : L^\infty (\mathcal{F}^{(2)}) \rightarrow \mathbb{R}$. 

...
Proof: Suppose \( \Pi \) is a fair and coherent two-step actuarial valuation with a linear second step actuarial valuation, by Theorem 3.2, there exists a \( F^2 \)-conditional valuation \( \pi [ \cdot \mid F^2 ] : \mathcal{L}^\infty (F) \rightarrow L^\infty (F^2) \) such that

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

where in the second step we used the positive homogeneity of \( \pi [ \cdot \mid F^2 ] \) and the last step is due to the independence of \( F^1 \) and \( F^2 \). Because \( \Pi \) is also weak market consistent, we have

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

which leads to \( \Pi[S] = \pi^1 [S^{(1)}] \pi^2 [S^{(2)}] \). Similar result follows if \( \Pi \) is a fair and coherent two-step financial valuation with a linear second step financial valuation by applying Theorem 3.10 of Pelsser & Stadje (2014).

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. In particular, we compare the following two valuations: for any \( S \in \mathcal{L}^\infty (F) \), we define

\[
\Pi^{(1)}[S] = \pi^2 \left( \mathbb{E}^Q \left( S \mid F^2 \right) \right), \quad (13)
\]

\[
\Pi^{(2)}[S] = \mathbb{E}^Q \left( \pi^2 \left( S \mid F^1 \right) \right). \quad (14)
\]

Here for simplicity, we abuse the notation a little. In (13), the conditional valuation \( \mathbb{E}^Q [S \mid F^2] \) is in fact \( \mathbb{E}^P [ZS \mid F^2] \) for some \( Z \in \{ \xi \in L_+^1 (F) : \mathbb{E}[\xi \mid F^2] = 1 \} \). That is \( Q \) in (13) is an absolutely continuous probability measure with respect to \( P \) conditional on \( F^2 \). In (14), \( Q \) is an absolutely continuous probability measure with respect to \( P \) in the usual sense; that is \( \mathbb{E}^Q [S^{(1)}] = \mathbb{E}^P [ZS^{(1)}] \) for some \( Z \in \{ \xi \in L_+^1 (F^1) : \mathbb{E}[\xi] = 1 \} \). \( \pi^2 \) in (13) is the usual actuarial valuation defined on \( L^\infty (F^2) \) while \( \pi^2 [S \mid F^1] \) in (14) is a conditional valuation defined on \( L^\infty (F) \) sharing similar structure with \( \pi^2 \). For example, if \( \pi^2 [S^{(2)}] = \mathbb{E}^P [S^{(2)}] \), then \( \pi^2 [S \mid F^1] = \mathbb{E}^P [S \mid F^1] \).

Example 3.3 (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 \( Y^{(0)} = 1 \) and a
longevity bond for which the payoff at maturity is \( Y^{(1)} = \tilde{L}_{x+T} \), the equivalent of \( L_{x+T} \) but for the Dutch population. First, we determine the value by the two-step actuarial valuation:

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

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\(^4\) : \( \mathbb{E}^P [L_{x+T} | \tilde{L}_{x+T}] = \beta \tilde{L}_{x+T} \) with \( \beta < 1 \), we determine the value according to the two-step financial valuation:

\[
\Pi^{(2)}[S] = \mathbb{E}^Q \left[ \mathbb{E}^P \left[ L_{x+T} | \mathcal{F}^{(1)} \right] \right] \\
= \mathbb{E}^Q \left[ \beta \tilde{L}_{x+T} \right] \\
= \beta Y^{(1)}(0),
\]

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.

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

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

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

Hence, both Belgian and Dutch populations are normal distributed with correlation \( \rho \). Hereafter, we compare the two-step actuarial and financial valuations given by

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

where \( \pi^{(2)} \) is the standard deviation principle (4) and \( \pi^{(2)}[\cdot | \mathcal{F}^{(1)}] \) is the conditional standard deviation principle:

\[
\Pi[S | \mathcal{F}^{(1)}] = \mathbb{E}^P \left[ S | \mathcal{F}^{(1)} \right] + \beta \sqrt{\text{Var}\left[ S | \mathcal{F}^{(1)} \right]}.
\]

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

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

\(^4\) For the reader interested in Dutch and Belgian mortality projections, we refer to Antonio et al. (2017)
To determine the two-step financial valuation, we first notice that by standard results of normal distributions, we have

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

Let us further assume that the distribution of \( \tilde{L}_{x+T} \) under \( \mathbb{Q} \) is \( \tilde{L}_{x+T} \mathbb{Q} \sim N(\mu_2 - \sigma_2 \kappa, \sigma_2^2) \), where \( \kappa > 0 \) is the market price of risk for the longevity bond. Therefore, we find that

\[
\Pi^{(2)}[S] = \mathbb{E}^{\mathbb{Q}} \left[ \pi^{(2)} \left( L_{x+T} | \tilde{L}_{x+T} \right) \right] = \mathbb{E}^{\mathbb{Q}} \left[ \mu_1 + \rho \frac{\sigma_1}{\sigma_2} (\tilde{L}_{x+T} - \mu_2) + \beta \sigma_1 \sqrt{1 - \rho^2} \right] = \mu_1 + \rho \frac{\sigma_1}{\sigma_2} \mathbb{E}^{\mathbb{Q}} \left[ \tilde{L}_{x+T} \right] - \mu_2 + \beta \sigma_1 \sqrt{1 - \rho^2} = \mu_1 - \rho \sigma_1 \kappa + \beta \sigma_1 \sqrt{1 - \rho^2}. \tag{18}
\]

We can compare the two-step actuarial valuation (17) with the two-step financial valuation (18). 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

\[
\Pi^{(2)}[S] - \Pi^{(1)}[S] = \sigma_1 \left[ \beta \left( \sqrt{1 - \rho^2} - 1 \right) - \rho \kappa \right]. \tag{19}
\]

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 (19)). The prices at time 0 of both approaches and the residual losses at maturity are given in the table below:

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

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:

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

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.

\[\text{Remark 3.1:}\] 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.

\[\text{Example 3.4 (Two-step valuations for hybrid claims):}\] 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. Suppose that the stock \(Y\) can go up to 200 or down to 50, the strike \(K = 100\) and the policyholder survival is modeled by the indicator \(I\). Therefore, the payoff of this contract is given by

\[
S = (Y - K)_+ \times I = \begin{cases} 100, & \text{if } Y = 200, I = 1, \\ 0, & \text{otherwise.} \end{cases} \tag{20}
\]

Similar to Example 3.3, we consider the two-step financial and actuarial valuations given by Equations (15) and (16) for the hybrid payoff (20):

1. \textit{Two-step financial valuation:} The value of \(S\) is given by

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

If we note that

\[
\mathbb{E}^P \left[ I | Y \right] = \begin{cases} \mathbb{P}[I = 1 | Y = 50], & \text{if } Y = 50, \\ \mathbb{P}[I = 1 | Y = 200], & \text{if } Y = 200, \end{cases}
\]

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

\[
\Pi^{(1)}[S] = 100 q_Y \left( p_{I|Y=200} + \beta \sqrt{p_{I|Y=200}} (1 - p_{I|Y=200}) \right), \tag{21}
\]

where \(q_Y\) is the \(Q\)-probability that \(Y\) goes up: \(q_Y = Q[Y = 200]\) and \(p_{I|Y=200}\) is the \(P\)-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

\[
\Pi^{(2)} [S] = \mathbb{E}^{\mathbb{P}} \left[ \mathbb{E}^{\mathbb{Q}} \left[ S \mid \mathcal{F}^{(2)} \right] \right] + \beta \sigma^{\mathbb{P}} \left[ \mathbb{E}^{\mathbb{Q}} \left[ S \mid \mathcal{F}^{(2)} \right] \right]
\]

\[
= \mathbb{E}^{\mathbb{P}} \left[ \mathbb{E}^{\mathbb{Q}} \left[ (Y - K)_{+} \mid I \right] \right] + \beta \sigma^{\mathbb{P}} \mathbb{E}^{\mathbb{Q}} \left[ (Y - K)_{+} \mid I \right].
\]

Noting that

\[
\mathbb{E}^{\mathbb{Q}} \left[ (Y - K)_{+} \mid I \right] = \begin{cases} 100 \text{ if } I = 0, \\ 100 \text{ if } I = 1,
\end{cases}
\]

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

\[
\Pi^{(2)} [S] = 100 q_{Y \mid I=1} (p_I + \beta \sqrt{p_I(1 - p_I)}),
\]

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

If we compare the two-step financial and actuarial values (21) and (22), 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. \( p_I \mid Y=200 \) while in the second case, it is via the \( \mathbb{Q} \)-probability of financial risks given actuarial scenarios, i.e. \( q_{Y \mid I=1} \). In case of independence under \( \mathbb{P} \) and \( \mathbb{Q} \), both valuations are equal. In case of dependence, both valuations (21) and (22) will in general be different as we illustrate below:

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

\[
+ 100 q_{Y \mid I=1} \beta \sqrt{p_I(1 - p_I)} - 100 q_Y \beta \sqrt{p_I \mid Y=200(1 - p_I \mid Y=200)}.
\]

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

\[
\kappa = q_Y - p_Y,
\]

\[
= q_{Y \mid I=1} - p_{Y \mid I=1}.
\]

Therefore, by Bayes' Theorem, we find that

\[
\Pi^{(2)} [S] - \Pi^{(1)} [S] = 100 \left( p_I \mid Y=200 \frac{p_Y}{p_I} + \kappa \right) p_I - 100 \left( p_Y + \kappa \right) p_I \mid Y=200
\]

\[
+ 100 \left( p_{Y \mid I=1} + \kappa \right) \beta \sqrt{p_I(1 - p_I)}
\]

\[
- 100 \left( p_Y + \kappa \right) \beta \sqrt{p_I \mid Y=200(1 - p_I \mid Y=200)}.
\]

After simplifications, we find that

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

\[
+ 100 \kappa \beta \left( \sqrt{p_I(1 - p_I)} - \sqrt{p_I \mid Y=200(1 - p_I \mid Y=200)} \right)
\]

\[
+ 100 p_{Y \mid I=1} \beta \sqrt{p_I(1 - p_I)} - 100 p_Y \beta \sqrt{p_I \mid Y=200(1 - p_I \mid Y=200)}.
\]

Similar to Example 3.3, 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 \( P \) and the risk-neutral measure \( Q \).

• 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 \( p_Y | l = 1 \)).

We remark that in the literature, it is common to assume that financial and actuarial claims are independent (either under \( P \) or \( Q \)). 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 \( Q \)-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 & Mamon 2018).

Example 3.5 (continued): If we assume independence under \( P \) in the two-step financial valuation or independence under \( Q \) in the two-step actuarial valuation, both valuations lead to a fair valuation. This is in line with Lemma 3.3.

4. Numerical illustration: equity-linked contracts

Based on the two-step actuarial valuation introduced in Section 3, we show how we can define a best estimate valuation. Moreover, we illustrate such valuation on a portfolio of equity-linked life insurance contracts with dependent financial and actuarial risks in Section 4.2.

4.1. 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 then 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 4.1 (Best estimate):** For any claim \( S \in L^\infty(\mathcal{F}) \), the best estimate is given by

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

(23)

where \( Q \) is an absolutely continuous probability measure with respect to \( P \) conditional on \( \mathcal{F}^{(2)} \).

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. \( \pi^{(1)}[S] = \mathbb{E}^Q[S] \) and \( \pi^{(2)}[S] = \mathbb{E}^P[S] \). In general, the expression (23) 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 \( \overline{BE} \) defined by

\[
\overline{BE}[S] = \sum_{i=1}^{n} \mathbb{P}[A_i] \mathbb{E}^Q[S | A_i],
\]

(24)

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

---

\(^5\) Note that independence under \( P \) does not necessarily imply independence under \( Q \), see Dhaene et al. (2013).
The best estimate in Equation (23) 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
  \[ \text{BE}[S^{(2)}] = \mathbb{E}^P \left[ S^{(2)} \right]. \]

- For any product claim $S$ with independent actuarial and financial risks (under $\mathbb{Q}$), we find that
  \[
  \text{BE}[S] = \mathbb{E}^P \left[ \mathbb{E}^Q \left[ S^{(1)} \times S^{(2)} | \mathcal{F}^{(2)} \right] \right]
  = \mathbb{E}^P \left[ S^{(2)} \times \mathbb{E}^Q \left[ S^{(1)} | \mathcal{F}^{(2)} \right] \right]
  = \mathbb{E}^P \left[ S^{(2)} \right] \times \mathbb{E}^Q \left[ S^{(1)} \right].
  \]

Hence, the actuarial risk is priced via real-world expectation and the financial risk via risk-neutral expectation. In fact, for Equation (25) to hold, it is sufficient that the financial claim $S^{(1)}$ is independent from the actuarial filtration $\mathcal{F}^{(2)}$.

### 4.2. Numerical application: portfolio of GMMB contracts

In this subsection, we show how to determine the best-estimate for a portfolio of guaranteed minimum maturity benefit (GMMB) contracts underwritten at time 0 on $l_x$ persons of age $x$. 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

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

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

\[
\begin{align*}
    dY^{(1)}(t) &= Y^{(1)}(t) \left( \mu \ dt + \sigma \ dW_1(t) \right), \\
    d\lambda(t) &= c\lambda(t) \ dt + \xi \ dW_2(t),
\end{align*}
\]

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 (28) for the mortality model allows negative mortality rates. However, Luciano & 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 (28), $\lambda(t)$ is a Gaussian process and $\int_0^T \lambda(t) \, dv$ is normal distributed.

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 (28) 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.\(^6\) For the stock process, we define

$$d W_1^Q(t) = \frac{\mu - r}{\sigma} \, dt + d W_1(t),$$

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

$$d Y^{(1)}(t) = Y^{(1)}(t) \left( r \, dt + \sigma \, d W_1^Q(t) \right)$$

$$d \lambda(t) = c \lambda(t) \, dt + \xi \, d W_2^Q(t).$$

The best estimate for the aggregate payoff (26) is given by

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

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

$$BE[S] = \mathbb{E}^\mathbb{P} \left[ \mathbb{E}^\mathbb{Q} \left[ e^{-r T} \max \left( Y^{(1)}(T), K \right) \right] \right]$$

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

$$= \mathbb{E}^\mathbb{P} \left[ e^{-\int_0^T \lambda(v) \, dv} \right] \left[ Y^{(1)}(0) N(d_1) + Ke^{-r T} (1 - N(d_2)) \right]$$

with

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

$$d_2 = d_1 - \sigma \sqrt{T}.$$
Proposition 4.1: If we denote by \( T_{pi}^x \) \((i = 1, \ldots, n)\) the survival rates for each actuarial scenario\(^7\), the approximated best estimate for the aggregate payoff of GMMB contracts:

\[
\overline{BE}[S] = \frac{1}{n} \sum_{i=1}^{n} \mathbb{E}^Q \left[ e^{-rT} \frac{L_{x+T}}{l_x} \times \max \left( Y^{(1)}(T), K \right) \right] \text{l}_{x+T} = l_x T_{pi}^x
\]

is given by

\[
\overline{BE}[S] = \frac{1}{n} \sum_{i=1}^{n} T_{pi}^x \left( \tilde{Y}^{(1)}(0)N(d_1) + e^{-rT} K(1 - N(d_2)) \right), \tag{35}
\]

with

\[
\tilde{Y}^{(1)}(0) = Y^{(1)}(0) e^{\frac{-\sigma \rho_0 \sigma T}{2} \left( \frac{1}{2} e^{2cT} - \frac{1}{2} e^{T} + T + \frac{3}{2} \right)} \left( \frac{1}{\sqrt{2\pi}} \sigma^2 \rho_0 \right),
\]

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

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

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

**Proof:** The proof based on classical arguments of stochastic calculus can be found in Appendix A. \(\blacksquare\)

The approximated best estimate (35) 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 adjustment of the current stock price \(Y^{(1)}(0)\) to \(\tilde{Y}^{(1)}(0)\), taking into account the realized survival rate \(T_{pi}^x\) in each actuarial scenario. It is also worth noticing that in case of independence \((\rho = 0)\), the approximated best estimate (35) converges to the best estimate (33).

To determine the best estimate (35), we only need to generate survival rates \(T_{pi}^x\) \((i = 1, \ldots, n)\) 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

\[
\ln T_{pi}^x = -\int_0^T \lambda(s) ds \sim N \left( \frac{\lambda(0)}{c} \left( e^{ct} - 1 \right), \frac{\xi^2}{c^3} \left( \frac{1}{2} e^{2ct} - 2 e^{ct} + cT + \frac{3}{2} \right) \right).
\]

We generate \(n = 100,000\) 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 & Kwak (2016). The mortality parameters correspond to UK male individuals who are aged 55 at time 0.

Table 2 displays the best estimate per policy obtained using Equation (35) for a range of correlation coefficients: \(\rho \in [-1, 1]\). We observe that the best estimate slightly decreases with the increase of the

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\(^7\) We assume that the actuarial scenarios are generated by a Monte-Carlo sample of i.i.d. observations.
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. 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 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.

As pointed out by Liu et al. (2014), Solvency II Directive strongly 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.

Acknowledgments

Karim Barigou acknowledges the financial support of the Research Foundation – Flanders (FWO) (PhD funding) and the Joint Research Initiative on ‘Mortality Modeling and Surveillance’ funded by AXA Research Fund (postdoc funding). The authors would also like to thank Jan Dhaene from KU Leuven for useful discussions and helpful comments.
on this manuscript. Finally, we sincerely thank the editor and anonymous referees for their pertinent remarks that significantly improved the quality of the manuscript.

Disclosure statement

No potential conflict of interest was reported by the author(s).

Funding

This work was supported by AXA Research Fund [JRI : ‘Mortality Modeling and Surveillance’] and Research Foundation Flanders [PhD grant].

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### Appendix. Proof of Proposition 4.1

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

$$
\begin{align*}
\text{d}Y^{(1)}(t) &= Y^{(1)}(t) (r \, \text{d}t + \sigma \, \text{d}W^1(t)) \\
\text{d}\lambda(t) &= c \lambda(t) \, \text{d}t + \xi \, \text{d}W^2(t)
\end{align*}
$$

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 (A2), we note that

$$
\begin{align*}
\text{d} (e^{-\mu t} \lambda(t)) &= -ce^{-\mu t}\lambda(t) \, \text{d}t + e^{-\mu t} \text{d}\lambda(t) \\
&= \xi e^{-\mu t} \, \text{d}W^2(t).
\end{align*}
$$
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
\[
\int_0^T \lambda(s) \, ds = \frac{\lambda(0)}{c} \left( e^T - 1 \right) + \xi \int_0^T \int_0^s e^{-c(u-s)} \, dW_2(u) \, ds
\]
\[ = \frac{\lambda(0)}{c} \left( e^T - 1 \right) + \xi \int_0^T \int_u^T e^{-c(u-s)} \, ds \, dW_2(u)
\]
\[ = \frac{\lambda(0)}{c} \left( e^T - 1 \right) + \xi \int_0^T \left( e^{-c(u-T)} - 1 \right) \, dW_2(u)
\]
\[ = \frac{\lambda(0)}{c} \left( e^T - 1 \right) + \frac{\xi}{c} X_T,
\]

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
\[ E(W_1(T)X_T) = E \left( \int_0^T \, dW_1(u) \int_0^T \left( e^{-c(u-T)} - 1 \right) \, dW_2(u) \right)
\]
\[ = \rho \left( \frac{1}{c} e^T - \frac{1}{c} - T \right),
\]
which leads to
\[ \text{corr}(W_1(T), X_T) = \frac{\rho \left( \frac{1}{c} e^T - \frac{1}{c} - T \right)}{\sqrt{T \left( \frac{1}{2c} e^{2cT} - \frac{2}{c} e^{cT} + T + \frac{3}{2c} \right)}} = \rho_0. \]

We can then assume that
\[ W_1(T) = \frac{\rho_0 \sqrt{T}}{\sqrt{\frac{1}{2c} e^{2cT} - \frac{2}{c} e^{cT} + T + \frac{3}{2c}}} X_T + \sqrt{T \left( 1 - \rho_0^2 \right)} Z,
\]

where \( Z \) is a standard normal r.v. independent of \( X_T \).

From
\[ e^{-\int_0^T \lambda(s) \, ds} = T P^j_x,
\]
we find that
\[ X_T = -\frac{c}{\xi} \ln T P^j_x = -\frac{\lambda(0)}{\xi} \left( e^T - 1 \right).
\]

The stock price at time \( T \) can be written as
\[
Y^{(1)}(T) = Y^{(1)}(0)e^{(r-\frac{1}{2}\sigma^2)T} + \sigma W_1(T)
\]
\[ = Y^{(1)}(0)e^{\sqrt{\frac{\sigma^2}{2\pi}} \frac{\sigma \sqrt{T}}{1 - r\sigma^2 + \frac{1}{4}(e^T - 1)} \left( \frac{1}{2} \ln T P^j_x + \frac{1}{2}(e^T - 1) \right) e^{(r-\frac{1}{2}\sigma^2)T} + \sigma \sqrt{1 - r\sigma^2 + \frac{1}{4}(e^T - 1)} \sqrt{T} \, \sigma \sqrt{T} \, \sigma \sqrt{T},
\]

with
\[
\tilde{S}^{(1)}(0) = Y^{(1)}(0)e^{\sqrt{\frac{\sigma^2}{2\pi}} \frac{\sigma \sqrt{T}}{1 - r\sigma^2 + \frac{1}{4}(e^T - 1)} \left( \frac{1}{2} \ln T P^j_x + \frac{1}{2}(e^T - 1) \right) e^{(r-\frac{1}{2}\sigma^2)T}}.
\]

Finally, we find that
\[
E^{\mathbb{Q}} \left[ e^{-rT} L_{x+T} \times \max \left( (Y^{(1)}(T, K) e^{-\int_0^T \lambda(s) \, ds} = T P^j_x \right) \right]
\]
\[ = l_x T P^j_x E^{\mathbb{Q}} \left[ e^{-rT} K + e^{-rT} \max \left( (Y^{(1)}(T) - K, 0) e^{-\int_0^T \lambda(s) \, ds} = T P^j_x \right) \right]
\]
\[ = l_x T P^j_x \left( \tilde{S}^{(1)}(0)N(d_1) + e^{-rT} K (1 - N(d_2)) \right),
\]

which ends the proof.