

ORIGINAL ARTICLE

Optimal investment and consumption under partial information

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Abstract We present a unified approach for partial information optimal investment and consumption problems in a non-Markovian Itô process market. The stochastic local mean rate of return and the Wiener process cannot be observed by the agent, whereas the path-dependent volatility, the path-dependent interest rate and the asset prices can be observed. The main assumption is that the asset price volatility is a nonanticipative functional of the asset price trajectory. The utility functions are general and satisfy standard conditions. First, we show that the corresponding full information market is complete and in this setting we solve the problem using standard methods. Second, we transform the original partial information problem into a corresponding full information problem using filtering theory, and show that it follows that the market is observationally complete in the sense that any contingent claim adapted to the observable filtration is replicable. Using the solutions of the full information problem we then easily derive solutions to the original partial information problem.

Keywords Partial information · Utility maximization · Optimal investment and consumption · Stochastic control · Portfolio theory · Path-dependent volatility

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1 Introduction

The optimal investment and consumption problem is a fundamental object of study in financial mathematics. In this paper we consider a financial market living on a stochastic basis $(\Omega, \mathcal{F}, P, \underline{\mathcal{F}})$ carrying an *n*-dimensional asset price process *S* modeled as a non-Markovian Itô process. The (augmented) filtration generated by the asset prices $\underline{\mathcal{F}}^S$ formalizes the information set of an agent. The local mean rate of return process α and the driving Wiener process *W* are not assumed to be adapted to $\underline{\mathcal{F}}^S$ and the agent can therefore generally not observe these processes and the information is thus partial. The asset price volatility is, however, without loss of generality adapted to $\underline{\mathcal{F}}^S$. The main assumption is that the volatility of *S*, denoted by $\sigma(S)$, is a *regular* nonanticipative functional of the asset price trajectory. The market is shown to be *observationally complete*¹ in the sense that any contingent claim adapted to the observable filtration $\underline{\mathcal{F}}^S$ is replicable, and for this we need the assumption that $\sigma(S)$ is a regular nonanticipative functional.

For a fixed time interval and fixed initial wealth the agent faces the problem of optimizing expected utility of a continuous consumption stream and terminal wealth. The utility functions are general and satisfy standard conditions. This is a stochastic optimal control problem under partial information and the purpose of this paper is to characterize the agent's optimal consumption and optimal portfolio weights processes, as well as the resulting optimal wealth process. We also study logarithmic and power utility in two examples. We tackle the problem in two steps.

First, we solve the corresponding full information problem, which we obtain by assuming that the large filtration $\underline{\mathcal{F}}$ coincides with the observable filtration \mathcal{F}^{S} . Optimal investment and consumption problems with full information are typically studied using the dynamic programming approach or the martingale approach. In the dynamic programming approach solutions are typically presented in the form of non-linear infinite-dimensional partial differential equations called Hamilton-Jacobi-Bellman equations. This approach typically relies on the assumption that processes are Markovian. In the martingale approach optimization is performed ω by ω . Solutions are typically presented in the form of characterizations of the process dynamics of optimal wealth, optimal consumption and the optimal portfolio. This approach typically relies on market completeness. Under the assumption of full information we show that our market is complete, and since it is also non-Markovian we use the martingale approach. The optimal investment and consumption problem was first studied in Merton (1969, 1971), where the dynamic programming approach was used. The martingale approach was developed in Pliska (1986) and Karatzas et al. (1987). The literature on the optimal investment and consumption problem is considerable, see Chapter 3 in Karatzas and Shreve (1998) for an excellent survey.

Second, we transform the original partial information problem into a corresponding full information problem using the filtering theory separation principle, also known as the principle of separation of estimation and control. We express the originally

¹ The term *complete with respect to* $\underline{\mathcal{F}}^{S}$ is also used, see e.g. Hahn et al. (2007).

only partially observable dynamics of the asset prices in terms of fully observable dynamics of the same type as in the full information problem using a filter estimate of the unobservable process α and the *innovations process*, and show that it follows that the market is observationally complete. Using the solutions of the full information problem we then easily derive solutions to the original partial information problem.

Using the filtering theory separation principle in this type of two step approach is a well-known technique in the literature of optimal investment and consumption under partial information, see e.g. Lakner (1995), Putschögl and Sass (2008), Sass and Wunderlich (2010) and Björk et al. (2010).

One strand of the literature on partial information optimal investment and consumption focuses on the problem from a general perspective, i.e. with the aim of studying a financial market which is as general as possible while still obtaining interesting results. The present paper belongs to this strand. Much of the general theory of optimal investment and consumption under partial information was developed in Lakner (1995). That is where the notions of full and partial information were introduced. Using filtering theory and martingale methods the paper studies partial information optimal investment and partial information optimal consumption separately. Asset prices follow an Itô process with unobservable stochastic local mean rate of return and volatility of a similar type as in the present paper. The risk free interest rate is assumed to be zero. An explicit expression for the optimal consumption process is derived and the optimal portfolio process is implicitly characterized. Moreover, explicit solutions for logarithmic and power utility are obtained under more restrictive assumptions on the asset price dynamics. Putschögl and Sass (2008) study partial information optimal investment and consumption using Malliavin calculus and present explicit expressions for the optimal consumption process as well as the optimal portfolio process involving Malliavin derivatives. Asset prices follow an Itô process with unobservable stochastic local mean rate of return and constant volatility. The interest rate is constant. The special case of power utility is studied in an example. The use of Malliavin calculus and in particular the Clark-Ocone theorem in order to obtain explicit expressions for optimal quantities, such as the optimal portfolio, which generally can otherwise only be implicitly characterized, is also seen in Lakner (1998), where optimal investment for a similar asset price model is studied. Björk et al. (2010) study optimal investment in a setting similar to the one in the present paper. The Markovian special case and power and logarithmic utility are also studied in detail.

The other, much larger, strand of the literature studies optimal investment and consumption under partial information in more specific financial models. It should be noted that many of these papers also add to the general theory. There are two main types of partial information asset price models for which reasonably explicit expressions for the optimal quantities may be found using filtering theory. In the first one, the unobservable local mean rate of return is modeled as the solution of a linear SDE and the Kalman filter is used. In the second one, which is known as the *hidden Markov model* (HMM), the unobservable local mean rate of return is modeled as a function of the current state of a continuous-time Markov chain with a finite number of states. The corresponding filter is known as the Wonham filter or the HMM filter. In both of these models it is necessary to impose a restrictive structure on the asset price volatility in order to obtain explicit solutions.

The Kalman filter is used to study optimal investment and consumption under partial information in Dothan and Feldman (1986), Lakner (1998) and Putschögl and Sass (2008). The HMM filter is used in Elliott and Rishel (1994), Honda (2003), Sass and Haussmann (2004), Bäuerle and Rieder (2005) and Hahn et al. (2007).

Sass (2007) studies the problem under *convex constraints* on the portfolio strategy in an Itô process model where the volatility depends on the current state of the asset price return process. Sass and Wunderlich (2010) study a similar model under joint *budget constraints* and *shortfall risk constraints*.

Callegaro et al. (2006), Bäuerle and Rieder (2007) and Frey et al. (2012) study the problem in jump process models.

The main contribution of the present paper is a unified approach for partial information optimal investment and consumption problems in observationally complete non-Markovian Itô process markets with stochastic unobservable local mean rates of return, path-dependent volatility and path-dependent interest rates. We extend the general results of Björk et al. (2010) by giving the agent the possibility of continuous consumption. However, more important, we show that our market is (observationally) complete and we also clarify the fact that one must make restrictive assumptions regarding the asset price volatility for this to be the case. We extend the general results of Lakner (1995) in that our interest rate is path-dependent rather than zero. This extension is not trivial. In Lakner (1995) the filtration generated by the asset prices coincides with the filtration generated by the Wiener process under the risk-neutral measure. This is because asset prices have no drift under the risk-neutral measure, since the interest rate is zero. This implies that the standard martingale representation theorem can be directly used to prove that the market is complete. We extend the general results (not related to Malliavin calculus) of Putschögl and Sass (2008), mainly due to their assumption of constant volatility and constant interest rate. This extension is not trivial for reasons similar to the ones mentioned in relation to Lakner (1995).

For the corresponding full information problem we shed some light on the solutions for the special case of power utility. Specifically, we note that the price process of a certain financial derivative turns up as an ingredient in the solution for power utility. We also generalize the probability measure Q^0 studied in Björk et al. (2010).

The rest of this paper is structured as follows. In Sect. 2 we present general results which form the mathematical foundation of the paper. In Sect. 3 we introduce the financial market and formulate the problem which we study in the paper. In Sect. 4 we solve the problem in the special case of full information. By solving the problem we mean that we characterize the agent's optimal consumption and optimal portfolio weights processes, as well as the resulting optimal wealth process. The problem is solved for general utility functions satisfying standard conditions. We also study logarithmic and power utility as examples. In Sect. 5 we transform the original partial information problem into a corresponding full information problem. Using the solutions of the full information problem we derive solutions to the original partial information problem.

2 Two martingale representation results

In order to show that the financial market in this paper is observationally complete we need two martingale representation results. The results are not fundamentally new but they are usually not presented in this way. Much of the general theory directly related to these results was developed in the seminal paper of Fujisaki et al. (1972) and the references therein, and can be found in e.g. Kallianpur (1980) and Liptser and Shiryayev (2001). This section is to be considered an independent part of the paper.

Consider a stochastic basis $(\Omega, \mathcal{F}, P, \underline{\mathcal{F}})$ satisfying the usual conditions, where $\underline{\mathcal{F}} = \{\mathcal{F}_t\}_{0 \le t \le T}$ for some fixed terminal time *T*. The basis carries an *n*-dimensional Wiener process denoted by *W*. For any process ξ we use the notation that the augmented filtration generated by ξ is denoted by $\underline{\mathcal{F}}^{\xi}$. First we need two results from the theory of stochastic differential equations (Proposition 2.3).

Definition 2.1 Let (C_T, \mathcal{B}_T) be the measurable space of continuous functions x on [0, T] where, for all $t \in [0, T]$, \mathcal{B}_t is the σ -algebra $\sigma(x : x_s, 0 \le s \le t)$. The t-indexed $n \times n$ -dimensional functional $\beta_t : C_T \to \mathcal{R}^{n,n}$ is said to be nonanticipative if it is \mathcal{B}_t -measurable for all $t \in [0, T]$.

Definition 2.2 Let $\eta_t(x)$ be a nonanticipative functional. $\eta_t(x)$ is said to be regular if it satisfies the following Lipschitz and growth conditions, where $||\eta||^2 = \sum \eta_{ij}^2$,

$$\begin{aligned} ||\eta_t(x) - \eta_t(y)||^2 &\leq L_1 \int_0^t ||x_s - y_s||^2 dK_s + L_2 ||x_t - y_t||^2, \\ ||\eta_t(x)||^2 &\leq L_1 \int_0^t \left(1 + ||x_s||^2\right) dK_s + L_2 \left(1 + ||x_t||^2\right) \end{aligned}$$

for all $t \in [0, T]$ and continuous functions x and y on [0, T], where L_1 and L_2 are constants and K_t is a nondecreasing right-continuous function with $K_t \in [0, 1]$.

Proposition 2.3 Let $\eta_t(x)$ and $\beta_t(x)$ be regular nonanticipative functionals. Then the *SDE*

$$d\xi_t = \eta_t(\xi)dt + \beta_t(\xi)dW_t, \quad \xi_0 \in \mathcal{R}^n \tag{1}$$

has a unique strong solution ξ . Moreover, if $\beta_t(x)$ is invertible (for all t and x) then $\mathcal{F}^{\xi} = \mathcal{F}^{W}$.

Proofs of the two results in Proposition 2.3 can be found in e.g. Kallianpur (1980) and Liptser and Shiryayev (2001).²

² The proof of the first part of the proposition is found in Kallianpur (1980, Theorem 5.1.2) and in Liptser and Shiryayev (2001, Theorem 4.6). The proof of the second part is easiest found in Kallianpur (1980, Section 5.2) but the reasoning can also be found in the proof of Theorem 5.16 and Theorem 5.17 in Liptser and Shiryayev (2001). Regarding Definition 2.2: several different Lipschitz and growth conditions guaranteeing the existence of a unique strong solution are used in the literature (see e.g. Kallianpur 1980 or Liptser and Shiryayev 2001, or, for less general stochastic differential equations, Karatzas and Shreve 1991 or Øksendal 2003). Similar remarks apply to Definition 2.1.

In the rest of this section we will study an equation of the type

$$d\xi_t = \gamma_t dt + \beta_t(\xi) dW_t, \quad \xi_0 \in \mathcal{R}^n$$

where $\beta_t(x)$ is an invertible (for all *t* and *x*) regular nonanticipative functional and γ is an \mathcal{F} -progressively measurable process. We need the following regularity.

Assumption 2.4 $\int_0^T ||\beta_t^{-1}(\xi)\gamma_t||^2 dt < \infty$ a.s. and the local martingale $\tilde{L}_t \equiv e^{-\int_0^t (\beta(\xi)_s^{-1}\gamma_s)' dW_s - \frac{1}{2} \int_0^t ||\beta(\xi)_s^{-1}\gamma_s||^2 ds}$ is a martingale.³

We are now ready to present the first main result of this section.

Proposition 2.5 Let the process ξ satisfy the equation

$$d\xi_t = \gamma_t dt + \beta_t(\xi) dW_t, \quad \xi_0 \in \mathcal{R}^n \tag{2}$$

where $\beta_t(x)$ is an invertible (for all t and x) regular nonanticipative functional and γ is an $\underline{\mathcal{F}}^{\xi}$ -progressively measurable process.⁴ Let Y be any $\underline{\mathcal{F}}^{\xi}$ -martingale. Then there exists an $\underline{\mathcal{F}}^{\xi}$ -progressively measurable process a_t such that

$$Y_t = Y_0 + \int_0^t a_s dW_s \quad a.s.$$

for all t, where $\int_0^T ||a_s||^2 ds < \infty$ a.s.

Proof The proof is for the one-dimensional case and the extension is trivial. The Girsanov theorem implies that if we change the probability measure, on $\underline{\mathcal{F}}$, using the likelihood process \tilde{L} , and denote the resulting equivalent measure by \tilde{P} , then

$$d\tilde{W}_t = \beta_t(\xi)^{-1} \gamma_t dt + dW_t \tag{3}$$

is $\underline{\mathcal{F}}$ -Wiener under \tilde{P} (the second part of Assumption 2.4 implies that we can use the Girsanov theorem). This implies that

$$d\xi_t = \beta_t(\xi) d\tilde{W}_t. \tag{4}$$

From Proposition 2.3 and (4) it follows that $\underline{\mathcal{F}}^{\xi} = \underline{\mathcal{F}}^{\tilde{W}}$ and we can therefore use the standard martingale representation theorem (see e.g. Chapter 1 of Karatzas and Shreve 1998) to prove that any $\underline{\mathcal{F}}^{\xi}$ -martingale under \tilde{P} , denote it by \tilde{Y} , can be written as

$$\tilde{Y}_t = Y_0 + \int_0^t \tilde{a}_s d\,\tilde{W}_s$$

³ Here ' denotes transposition. However, ' will sometimes denote the derivative. What is meant should be clear from the context.

⁴ Note that we in this proposition restrict the process γ to be $\underline{\mathcal{F}}^{\xi}$ -progressively measurable, rather than just $\underline{\mathcal{F}}$ -progressively measurable.

 \tilde{P} -a.s., for some $\underline{\mathcal{F}}^{\tilde{W}}$ -progressively measurable process \tilde{a} (satisfying $\int_0^T ||\tilde{a}_s||^2 ds < \infty \tilde{P}$ -a.s.) and hence by (3)

$$\tilde{Y}_t = Y_0 + \int_0^t \tilde{a}_s \beta_s(\xi)^{-1} \gamma_s ds + \int_0^t \tilde{a}_s dW_s$$
(5)

 \tilde{P} -a.s. Now let *Y* be an $\underline{\mathcal{F}}^{\xi}$ -martingale under *P*. Abstract Bayes' theorem then implies that $\tilde{Y} \equiv \frac{Y}{\tilde{L}}$ is an $\underline{\mathcal{F}}^{\xi}$ -martingale under \tilde{P} and by (5) it follows that (use Itô's formula)

$$dY_t = d \left[\tilde{L}_t \tilde{Y} \right]_t$$

= $\left(\tilde{L}_t \tilde{a}_t - \tilde{L}_t \tilde{Y}_t \left(\beta_t \left(\xi \right)^{-1} \gamma_t \right) \right) dW_t$
= $\left(\tilde{L}_t \tilde{a}_t - Y_t \left(\beta_t \left(\xi \right)^{-1} \gamma_t \right) \right) dW_t.$

Hence *Y* is of the claimed form, with $a_t = \tilde{L}_t \tilde{a}_t - Y_t (\beta_t(\xi)^{-1} \gamma_t)$ (recall that *P* and \tilde{P} are equivalent).

Moreover, $\int_0^T ||a_s||^2 ds = \int_0^T ||\tilde{L}_s \tilde{a}_s - Y_s(\beta_s(\xi)^{-1}\gamma_s)||^2 ds < \infty$ a.s. follows from $\int_0^T ||\tilde{a}_s||^2 ds < \infty$ \tilde{P} -a.s., continuity of the trajectories of Y and \tilde{L} and the first part of Assumption 2.4.

In order to present the second main result of this section, Corollary 2.11, we need the following definition and result from filtering theory. The additional regularity of Assumption 2.7 is also needed.⁵

Definition 2.6 Denote the filter estimate process, with respect to the filtration $\underline{\mathcal{F}}^{\xi}$, for any process *Y* by \hat{Y} , where *Y* is assumed to be a measurable process satisfying $\int_0^T E\left[||Y_t||^2\right] dt < \infty$. Define the process \hat{Y} as the $\underline{\mathcal{F}}^{\xi}$ -progressively measurable modification of $E_{\underline{\mathcal{F}}^{\xi}}[Y_t], \forall t \in [0, T]$.

Assumption 2.7 $\int_0^T E\left[||\beta_t^{-1}(\xi)\gamma_t||^2\right] dt < \infty.$

Lemma 2.8 Let the process ξ satisfy the equation

$$d\xi_t = \gamma_t dt + \beta_t(\xi) dW_t, \quad \xi_0 \in \mathcal{R}^n \tag{6}$$

where $\beta_t(x)$ is an invertible (for all t and x) regular nonanticipative functional and γ is an $\underline{\mathcal{F}}$ -progressively measurable process. The process \overline{W} defined by

⁵ See e.g. Fujisaki et al. (1972), Kallianpur (1980, Chapter 8) or Liptser and Shiryayev (2001, Chapter 8) for similar results and definitions, and also justifications of the definitions. Recall (for a given filtration) that a progressively measurable process is adapted and measurable, that a process which is measurable and adapted has a progressively measurable modification, and that an adapted process with every sample path right-continuous or left-continuous is progressively measurable, see Karatzas and Shreve (1991, Chapter 1).

$$d\bar{W}_t = \beta_t^{-1}(\xi) \left(d\xi_t - \hat{\gamma}_t dt \right), \quad \bar{W}_0 = 0$$
(7)

is then $\underline{\mathcal{F}}^{\xi}$ -Wiener.

Remark 2.9 \overline{W} is referred to as an innovations process.

Lemma 2.8 is a slight modification of a well-known result in filtering theory. To see this note that if we rewrite (6) as

$$\beta_t(\xi)^{-1}d\xi_t = \beta_t(\xi)^{-1}\gamma_t dt + dW_t$$

then we are effectively in the setting of Fujisaki et al. (1972), where we find the result that the process \overline{W} defined by

$$d\bar{W}_t = \beta_t(\xi)^{-1}d\xi_t - \beta_t(\xi)^{-1}\gamma_t dt, \quad \bar{W}_0 = 0$$

is $\underline{\mathcal{F}}^{\xi}$ -Wiener (here we rely on Assumption 2.7). $\beta_t(\xi)^{-1}$ is adapted to $\underline{\mathcal{F}}^{\xi}$ and Lemma 2.8 follows directly. Now rewrite (7) as

$$d\xi_t = \hat{\gamma}_t dt + \beta_t(\xi) dW_t \tag{8}$$

and note that ξ in (8) is exactly the same process as ξ in (6). The only difference is that (8) is the $\underline{\mathcal{F}}^{\xi}$ -semimartingale representation of ξ while (6) is the $\underline{\mathcal{F}}$ -semimartingale representation of ξ . Using this, the following result follows directly from Proposition 2.5. First, however, we need regularity for (8) corresponding to Assumption 2.4.

Assumption 2.10 The local martingale (with respect to $\underline{\mathcal{F}}^{\xi}$) $e^{-\int_0^t (\beta(\xi)_s^{-1} \hat{\gamma}_s)' d\bar{W}_s - \frac{1}{2} \int_0^t ||\beta(\xi)_s^{-1} \hat{\gamma}_s||^2 ds}$ is an \mathcal{F}^{ξ} -martingale.

Regularity corresponding to the other part of Assumption 2.4 can be shown to follow from Assumption 2.7.

Corollary 2.11 Let the process ξ satisfy the equation

$$d\xi_t = \gamma_t dt + \beta_t(\xi) dW_t, \quad \xi_0 \in \mathcal{R}^n$$

where $\beta_t(x)$ is an invertible (for all t and x) regular nonanticipative functional and γ is an $\underline{\mathcal{F}}$ -progressively measurable process. Let Y be any $\underline{\mathcal{F}}^{\xi}$ -martingale. Then there exists an $\underline{\mathcal{F}}^{\xi}$ -progressively measurable process a_t such that

$$Y_t = Y_0 + \int_0^t a_s d\bar{W}_s \quad a.s.$$

for all t, where \overline{W} given by (7) is $\underline{\mathcal{F}}^{\xi}$ -Wiener and $\int_0^T ||a_s||^2 ds < \infty$ a.s.

3 The financial market and problem formulation

Consider an arbitrage free continuous time financial market living on a stochastic basis $(\Omega, \mathcal{F}, P, \underline{\mathcal{F}})$ satisfying the usual conditions, where *P* is the objective probability measure and $\underline{\mathcal{F}} = \{\mathcal{F}_t\}_{0 \le t \le T}$ for some fixed terminal time *T* which we interpret as the investment horizon of the agent. The market has the following components.

• *n* asset price processes S^i , for i = 1, ..., n, with dynamics given by

$$dS_t^i = \alpha_t^i S_t^i dt + \sigma_t^i (S) S_t^i dW_t^i, \quad S_0^i \in \mathcal{R}^{++}$$

which we describe in the more convenient vector form

$$dS_t = D(S_t)\alpha_t dt + D(S_t)\sigma_t(S)dW_t$$
(9)

where

- α is an \mathcal{F} -progressively measurable *n*-dimensional process satisfying $\int_0^T ||\alpha_t|| dt < \infty \text{ a.s.}$
- \tilde{W} is an *n*-dimensional \mathcal{F} -Wiener process
- $D(S_t)$ is the $n \times n$ diagonal matrix with the vector S_t as main diagonal
- $\sigma_t(\cdot)$ is an $n \times n$ -dimensional nonanticipative functional such that $D(x)\sigma_t(x)$ is regular and invertible (for all *t* and *x*) and $\int_0^T ||\sigma_t(S)||^2 dt < \infty$ a.s.
- A bank account process *B* which for the \mathcal{F}^{S} -progressively measurable process *r* (interpreted as the instantaneous short rate), satisfying $\int_{0}^{T} |r_t| dt < \infty$ a.s., evolves according to $dB_t = r_t B_t dt$, $B_0 = 1$.

We will now describe the setup of the optimal investment and consumption problem in more detail. This setup and also the optimization procedure of the full information version of our problem (in Sect. 4) is very much inspired from the one largely developed in Karatzas et al. (1987) and described in Karatzas and Shreve (1998), where the important difference is that we have adjusted their (full information) setup to our partial information setup.⁶

The market inhabits an agent with initial wealth $x_0 \ge 0$, a time-dependent consumption utility function $U_1(t, \cdot)$, and a (bequest) utility function $U_2(\cdot)$ for terminal wealth. The utility functions satisfy the standard conditions of Assumption 4.10. Based on the information generated by the asset prices, formalized as the filtration \mathcal{F}^S , the agent makes decisions about consumption and investments.

Definition 3.1 A consumption process is an $\underline{\mathcal{F}}^S$ -progressively measurable nonnegative process c_t satisfying $\int_0^T c_t dt < \infty$ a.s.

Definition 3.2 Let the amount of capital invested in each risky asset be given by the *n*-dimensional process π_t and in the bank account by the process π_t^0 , which are

⁶ Karatzas and Shreve (1998) also have a slightly different setup in other respects. For example, they analyze the portfolio process π_t rather than our portfolio weights process u_t , see Definition 3.2. The filtration of their market is moreover generated by the Wiener process driving the asset prices. The proofs of Karatzas and Shreve (1998) can easily be modified to our setup and are not included.

both $\underline{\mathcal{F}}^{S}$ -progressively measurable such that $\int_{0}^{T} |\pi_{t}^{0} + \pi_{t} \mathbf{1}||r_{t}|dt < \infty, \int_{0}^{T} |\pi_{t}(\alpha_{t} - r_{t}\mathbf{1})|dt < \infty \text{ and } \int_{0}^{T} ||\pi_{t}\sigma_{t}(S)||^{2}dt < \infty \text{ a.s.}^{7}$ For a given c_{t} the corresponding wealth process is

$$X_t \equiv x_0 + \int_0^t \left(\pi_s^0 + \mathbf{1}\pi_s\right) r_s ds + \int_0^t \pi_s \left(\alpha_s - r_s \mathbf{1}\right) ds$$
$$+ \int_0^t \pi_s \sigma_s \left(S\right) dW_s - \int_0^t c_s ds$$

with $x_0 \ge 0$. The corresponding portfolio weights process is $u_t \equiv \pi_t / X_t$. The portfolio weights process is said to be self-financing if⁸ $X_t = \pi_t^0 + \pi_t \mathbf{1} \ \forall t \in [0, T]$ and tame if the discounted wealth process $X_t B_t^{-1}$ is bounded from below by a real constant that does not depend on t (but possibly on the process π_t) a.s. Moreover, we denote the wealth process corresponding to any c_t and u_t by $X_t^{u,c}$.

The interpretation of a self-financing portfolio weights process is that the wealth of an agent with this portfolio weights process is equal to the cumulative gains earned from investment minus cumulative consumption plus initial wealth. The concept of a tame portfolio is introduced as these do not allow doubling strategies, see Karatzas and Shreve (1998, Chapter 1).

Definition 3.3 We call the pair (u_t, c_t) a consumption and portfolio weights process. For a fixed $x_0 \ge 0$, (u_t, c_t) is said to be admissible if u_t is self-financing and tame (given c_t) and if the corresponding wealth process satisfies $X_t^{u,c} \ge 0$ for all $t \in [0, T]$ a.s.

In the following all portfolio weights processes are assumed to be admissible. In the rest of the paper we will mainly study the optimal investment and consumption problem under partial information presented in Problem 1. The information is partial because the agent needs her consumption and portfolio weights process (u_t, c_t) to be adapted to $\underline{\mathcal{F}}^S$, while the local mean rate of return process α and the Wiener process W are not generally adapted to $\underline{\mathcal{F}}^S$.

Problem 1 Given initial wealth $x_0 \ge 0$ and utility functions $U_1(t, \cdot)$ and $U_2(\cdot)$ an agent wants to maximize the functional

$$E\left[\int_0^T U_1\left(t, c_t\right) dt + U_2\left(X_T^{u,c}\right)\right]$$

over the set of admissible consumption and portfolio weights processes (u_t, c_t) (which need to be adapted to $\underline{\mathcal{F}}^S$). Our task is to characterize the optimal consumption process, denoted by \bar{c}_t^* , the optimal portfolio weights process, denoted by \bar{u}_t^* , and the optimal wealth process, denoted by \bar{X}_t^* .

⁷ **1** is an *n*-dimensional vector with each element equal to 1.

⁸ Thus $(1 - u_t \mathbf{1})$ represents the weight in the bank account B_t .

4 The full information problem

In this section, we will study Problem 1 under the assumption that the agent has full information and we therefore let the large filtration coincide with the observable filtration, see Assumption 4.1. In Sect. 5 we will study the partial information problem we set out to study and consequently we will in that section drop Assumption 4.1 and show that the original partial information problem can be transformed into a full information problem of the type studied in this section.

Assumption 4.1 $\underline{\mathcal{F}} = \underline{\mathcal{F}}^{S}$.

Remark 4.2 We use the notation that a bar over a quantity indicates that this is a quantity in the original partial information setting, i.e. without Assumption 4.1, and that the same quantity without the bar is with this assumption. For example, \bar{c}_t^* denotes the optimal consumption process under partial information, i.e. without Assumption 4.1, whereas c_t^* denotes the optimal consumption process under full information, i.e. with Assumption 4.1.

In the rest of this section we will write $\underline{\mathcal{F}}$ and not $\underline{\mathcal{F}}^S$ even though they by assumption coincide. The reason is that we in Sect. 5 will study the filtration $\underline{\mathcal{F}}^S$ (without assuming that $\underline{\mathcal{F}} = \underline{\mathcal{F}}^S$) and we must make the distinction clear.

Consider the market price of risk process ϑ_t , which is an $\underline{\mathcal{F}}$ -progressively measurable modification such that $\vartheta_t = \sigma_t^{-1}(S)(\alpha_t - r_t \mathbf{1})$ a.s.

Assumption 4.3 The local martingales $L_t \equiv e^{-\int_0^t \vartheta_s' dW_s - \frac{1}{2}\int_0^t ||\vartheta_s||^2 ds}$ and $\tilde{L}_t \equiv e^{-\int_0^t (\sigma_s^{-1}(S)\alpha_s)' dW_s - \frac{1}{2}\int_0^t ||\sigma_s^{-1}(S)\alpha_s||^2 ds}$ are martingales. $\int_0^T ||\vartheta_t||^2 dt < \infty$ a.s. and $\int_0^T ||\sigma_t^{-1}(S)\alpha_t||^2 dt < \infty$ a.s.

We will now show that the full information market is complete and that a unique risk-neutral probability measure exits.

Definition 4.4 The market is said to be complete if every \mathcal{F}_T -measurable and integrable contingent claim ζ , with ζB_T^{-1} a.s. bounded from below, can be replicated by an admissible portfolio strategy. A contingent claim is said to be integrable if $E[\zeta B_T^{-1}L_T] < \infty$.¹⁰

Proposition 4.5

• There exists a unique equivalent risk-neutral probability measure Q, which is given by $\frac{dQ}{dP} = L_T$ on \mathcal{F}_T for

$$dL_t = -\vartheta_t' L_t dW_t, \quad L_0 = 1.$$
⁽¹⁰⁾

• The process W^Q defined by $dW_t^Q = \vartheta_t dt + dW_t$ is $\underline{\mathcal{F}}$ -Wiener under Q.

⁹ As noted in Björk et al. (2010), it is not necessarily the case that $\underline{\mathcal{F}}^S$ and $\underline{\mathcal{F}}^W$ coincide, see Tsirelson's counterexample in e.g. Rogers and Williams (1987).

¹⁰ For further details see e.g. Karatzas and Shreve (1998).

- The process \tilde{W} defined by $d\tilde{W}_t = \sigma_t^{-1}(S)\alpha_t dt + dW_t$ is $\underline{\mathcal{F}}$ -Wiener under the equivalent probability measure \tilde{Q} , given by $\frac{d\tilde{Q}}{dP} = \tilde{L}_T$ on \mathcal{F}_T . Moreover, S has no drift under \tilde{Q} and $\underline{\mathcal{F}} = \underline{\mathcal{F}}^{\tilde{W}}$.
- The market is complete.

Proof The second result and first part of the third result follow directly from the Girsanov theorem since L_t and \tilde{L}_t are a likelihood processes (see Assumption 4.3). We can therefore write the process S given by (9) as

$$dS_t = D(S_t)\alpha_t dt + D(S_t)\sigma_t(S)\left(-\vartheta_t dt + dW_t^Q\right)$$
(11)

which implies that $\vartheta_t = \sigma_t^{-1}(S)(\alpha_t - r_t \mathbf{1})$ gives us a risk-neutral measure. For any probability measure equivalent to *P* the corresponding likelihood process is an \mathcal{F} -martingale under *P* and by Proposition 2.5 (regularity is given in Assumption 4.3) we therefore know that any such likelihood process could be written on the same form as (10) giving another expression of the same type as (11). This implies that ϑ_t determines the risk-neutral measure uniquely, which is the first result. Simple calculations show that *S* has no drift under \tilde{Q} , and the last part of the third result follows from Proposition 2.3. The last part of the third result implies that we can use standard results about Wiener driven markets and the fourth result follows, see e.g. Karatzas and Shreve (1998, Chapter 1) (a slight modification of the proofs of these results is needed: instead of referring to the standard martingale representation theorem one must refer to a more general result such as Proposition 2.5).

Remark 4.6 The quadratic variation of *S* is without loss of generality adapted to $\underline{\mathcal{F}}^S$ and it therefore follows that the asset price volatility would be a functional of the trajectory of *S* even if we did not assume this. Generally, however, this volatility functional would not be regular or even nonanticipative, see e.g. Liptser and Shiryayev (2001, Chapter 4). If the volatility functional would not be nonanticipative and regular we would not be able to prove that the market is complete, as Propositions 2.3 and 2.5 rely on this.

The market has been shown to be complete under the assumption of full information and we can therefore use the standard approach for solving the optimization problem. We follow Karatzas and Shreve (1998), where proofs of results analogous to the results in the rest of this section can be found.

The next result, Lemma 4.9, can be interpreted as saying roughly the following: for any contingent claim maturing at T there is an admissible consumption and portfolio weights process such that the corresponding wealth will coincide with the contingent claim at T, assuming that the initial wealth is large enough. The following definition and regularity condition are needed.

Definition 4.7 The stochastic discount factor *M* is defined by $M_t = B_t^{-1}L_t$ for all $t \in [0, T]$.

Assumption 4.8
$$E\left[\int_0^T M_t dt + M_T\right] < \infty.$$

Lemma 4.9 Let initial wealth $x_0 \ge 0$ be given, let c_t be a consumption process and let ζ be some nonnegative \mathcal{F}_T -measurable random variable such that

$$E\left[\int_0^T M_t c_t dt + M_T \zeta\right] = x_0.$$

Then there exists an admissible consumption and portfolio weights process (u_t, c_t) such that final wealth $X_T^{u,c} = \zeta$.¹¹

Proposition 4.12 will reveal that optimal investment and consumption is directly related to the inverse of marginal utility and we must therefore introduce (generalized) inverse functions of the derivatives of the utility functions. This will be done in Assumption 4.10, where we will also specify the standard conditions that the utility functions satisfy.

Assumption 4.10 $U_2: \mathcal{R} \to [-\infty, \infty)$ and $U_1: [0, T] \times \mathcal{R} \to [-\infty, \infty)$.

 U_2 is concave, nondecreasing and upper semicontinuous and the half-line $dom(U_2) \equiv \{x \in \mathcal{R}; U_2(x) > -\infty\}$ is a nonempty subset of $[0, \infty)$. The derivative U'_2 is continuous, positive and strictly decreasing on the interior of $dom(U_2)$ and $U'_2(\infty) = 0$ (we use the notation $U'_2(\infty) = \lim_{x\to\infty} U'_2(x)$). U_1 satisfies the same conditions for each t.

 $\underline{c}_t \equiv \inf \{c \in \mathcal{R}; U_1(t, c) > -\infty\}$ is a continuous function of t with values in $[0, \infty)$.

 U_1 and U'_1 (the derivative is taken with respect to consumption) are continuous on $\{(t, c) \in [0, T] \times (0, \infty); c > \underline{c}_t\}$.

For fixed $t \in [0, T]$, the function $I_1(t, \cdot) : (0, \infty] \xrightarrow{onto} [\underline{c}_t, \infty)$ defined by

$$U_1'(t, I_1(t, y)) = \begin{cases} y, & y \in \left(0, \lim_{c \searrow c_t} U_1'(t, c)\right) \\ \lim_{c \searrow c_t} U_1'(t, c), & y \in \left[\lim_{c \searrow c_t} U_1'(t, c), \infty\right] \end{cases}$$

is strictly decreasing on $(0, \lim_{c \searrow \underline{c}_t} U'_1(t, c))$, equal to \underline{c}_t on $[\lim_{c \searrow \underline{c}_t} U'_1(t, c), \infty]$ and continuous on $(0, \infty]$.

For $\underline{x} \equiv \inf \{x \in \mathcal{R}; U_2(x) > -\infty\}$, the function $I_2 : (0, \infty] \to [\underline{x}, \infty)$ defined by

$$U_2'(I_2(y)) = \begin{cases} y, & y \in \left(0, \lim_{x \searrow \underline{x}} U_2'(x)\right) \\ \lim_{x \searrow \underline{x}} U_2'(x), & y \in \left[\lim_{x \searrow \underline{x}} U_2'(x), \infty\right] \end{cases}$$

satisfies the analogous conditions.

The following regularity condition is needed.

Assumption 4.11
$$\mathcal{X}(y) \equiv E\left[\int_0^T M_t I_1(t, yM_t) dt + M_T I_2(yM_T)\right] < \infty, \forall y \in (0, \infty).$$

¹¹ A slight modification is needed in the proof of Lemma 4.9 compared to the analogous result in Karatzas and Shreve (1998). Instead of referring to the standard martingale representation theorem one must refer to a more general result, such as Proposition 2.5.

The full information solution to Problem 1 for the general utility functions of Assumption 4.10 is presented in the following proposition.

Proposition 4.12 (Optimal investment and consumption under full information) *Let* $x_0 \in (\mathcal{X}(\infty), \infty)$. *The optimal wealth process*

$$X_t^* = E_{\mathcal{F}_t} \left[\int_t^T \frac{M_s}{M_t} I_1(s, \lambda M_s) \, ds + \frac{M_T}{M_t} I_2(\lambda M_T) \right]$$

where λ is determined by

$$E\left[\int_0^T M_t I_1\left(t, \lambda M_t\right) dt + M_T I_2\left(\lambda M_T\right)\right] = x_0.$$

The optimal consumption process

$$c_t^* = I_1\left(t, \lambda M_t\right).$$

The optimal portfolio weights process

$$u_t^* = \sigma_{X^*}(t)\sigma_t^{-1}(S)$$

where $\sigma_{X^*}(t)$ is the volatility of the optimal wealth process X_t^* .

In the rest of this section we will study the examples of power and logarithmic utility. Proposition 4.12 is used repeatedly.

4.1 Example 1: Power utility under full information

Let $U_1(t, c_t) = \frac{c_t^{\gamma}}{\gamma}$ and $U_2(X_T) = \frac{X_T^{\gamma}}{\gamma}$ with $\gamma < 1, \gamma \neq 0$. To simplify the calculations we introduce the notation $\beta = \frac{\gamma}{1-\gamma}$. Simple calculations show that $c_t^* = (\lambda M_t)^{\frac{1}{\gamma-1}}$ and $X_T^* = (\lambda M_T)^{\frac{1}{\gamma-1}}$, and that λ is given by

$$x_0 = E\left[\int_0^T M_t(\lambda M_t)^{\frac{1}{\gamma-1}} dt + M_T(\lambda M_T)^{\frac{1}{\gamma-1}}\right]$$
$$= \lambda^{\frac{1}{\gamma-1}} H_0$$

with H_0 defined through the last equality. This implies that $c_t^* = \frac{x_0}{H_0} M_t^{\frac{1}{\gamma-1}}$, $X_T^* = \frac{x_0}{H_0} M_T^{\frac{1}{\gamma-1}}$ and $\mathcal{X}(\infty) = 0$. We now introduce a process H, with H_0 as starting value, which we will use

We now introduce a process H, with H_0 as starting value, which we will use to characterize the solution c_t^* , X_t^* and u_t^* . The process H is a generalization of the process H studied in Björk et al. (2010). The process H of the present paper differs from the process H of Björk et al. (2010) because of the agent's possibility of consumption. **Definition 4.13** Define the process H by $H_t = E_{\mathcal{F}_t} \left[\int_t^T \left[\frac{M_s}{M_t} \right]^{-\beta} ds + \left[\frac{M_T}{M_t} \right]^{-\beta} \right]$ for all $t \in [0, T]$.

The optimal wealth process is then

$$X_{t}^{*} = \frac{x_{0}}{H_{0}} E_{\mathcal{F}_{t}} \left[\int_{t}^{T} \frac{M_{s}^{-\beta}}{M_{t}} ds + \frac{M_{T}^{-\beta}}{M_{t}} \right]$$
$$= x_{0} \frac{H_{t}}{H_{0}} M_{t}^{\frac{1}{\gamma - 1}}.$$

Lemma 4.14 The process H can be written in the following Itô process form

$$dH_t = \mu_H(t)H_t dt + \sigma_H(t)H_t dW_t$$

for some processes $\mu_H(t)$ and $\sigma_H(t)$.

Proof Use the definition of *H* and rewrite it as

$$H_{t} = E_{\mathcal{F}_{t}} \left[\int_{t}^{T} \left[\frac{M_{s}}{M_{t}} \right]^{-\beta} ds + \left[\frac{M_{T}}{M_{t}} \right]^{-\beta} \right]$$
$$= \left[E_{\mathcal{F}_{t}} \left[\int_{0}^{T} M_{s}^{-\beta} ds \right] - \int_{0}^{t} M_{s}^{-\beta} ds \right] M_{t}^{\beta} + M_{t}^{\beta} E_{\mathcal{F}_{t}} \left[M_{T}^{-\beta} \right].$$

The expected values in this expression are both martingales so they are by Proposition 2.5 driven by W. From the definition of M it follows that also M_t^β is driven by W. The result follows from Itô's formula and the obvious fact that H is strictly positive.

From the results above, the definition of M_t and Itô's formula it follows that

$$dX_{t}^{*} = d\left(x_{0}\frac{H_{t}}{H_{0}}M_{t}^{\frac{1}{\gamma-1}}\right) = d\left[x_{0}e^{\int_{0}^{t}(\ldots)ds + \int_{0}^{t}\left[\frac{-\vartheta_{x}'}{\gamma-1} + \sigma_{H}(s)\right]dW_{s}}\right]$$
$$= (\ldots)dt + \left[x_{0}e^{\int_{0}^{t}(\ldots)ds + \int_{0}^{t}\left[\frac{-\vartheta_{x}'}{\gamma-1} + \sigma_{H}(s)\right]dW_{s}}\right]\left[\frac{-\vartheta_{t}'}{\gamma-1} + \sigma_{H}(t)\right]dW_{t}$$
$$= (\ldots)dt + X_{t}^{*}\left[\frac{-\vartheta_{t}'}{\gamma-1} + \sigma_{H}(t)\right]dW_{t}.$$

This implies that $\sigma_{X^*}(t) = \frac{-\vartheta'_t}{\gamma-1} + \sigma_H(t)$, so that $u_t^* = \left[\frac{-\vartheta'_t}{\gamma-1} + \sigma_H(t)\right]\sigma_t^{-1}(S)$. The process *H* in Björk et al. (2010) is in that paper investigated by means of a probability measure Q^0 . In the following remark we perform a similar investigation of our process *H* under a probability measure Q^0 which is a generalization of the probability measure in Björk et al. (2010).

Remark 4.15 Consider the likelihood process (assume in this example that it is a martingale) $L_t^0 = e^{\int_0^t \beta \vartheta'_s dW_s - \frac{1}{2} \int_0^t \beta^2 ||\vartheta_s||^2 ds}$ and the measure Q^0 given by $\frac{dQ^0}{dP} = L_T^0$ on \mathcal{F}_T . Now use that $M_t^{-\beta} = B_t^{\beta} L_t^{-\beta} = B_t^{\beta} L_t^0 e^{\frac{1}{2} \int_0^t \frac{\beta}{1-\gamma} ||\vartheta_s||^2 ds}$ to see that the process *H* has the representation

$$H_t = E_{\mathcal{F}_t}^{Q^0} \left[\int_t^T \left[\frac{B_s}{B_t} \right]^\beta e^{\frac{1}{2} \int_t^s \frac{\beta}{1-\gamma} ||\vartheta_u||^2 du} ds + \left[\frac{B_T}{B_t} \right]^\beta e^{\frac{1}{2} \int_t^T \frac{\beta}{1-\gamma} ||\vartheta_u||^2 du} \right]$$

We now introduce a process Π which we use to characterize the solution $(c_t^*, X_t^*$ and $u_t^*)$ in a novel fashion.

Definition 4.16 Let Π be the price process of a derivative with dividend process $M_t^{-(\beta+1)}$ and derivative payoff $M_T^{-(\beta+1)}$ at time *T*, i.e. let

$$\Pi_t = E_{\mathcal{F}_t} \left[\int_t^T \frac{M_s}{M_t} M_s^{-(\beta+1)} ds + \frac{M_T}{M_t} M_T^{-(\beta+1)} \right].$$

In the following lemma we investigate the dynamics of Π , the connection between the volatilities of Π and H, and we also characterize the drift of the process H using the dynamics of the process Π .

Lemma 4.17 The process Π can be written in the following Itô process form

$$d\Pi_t = \mu_{\Pi}(t)\Pi dt + \sigma_{\Pi}(t)\Pi_t dW_t$$

for some process σ_{Π} and $\mu_{\Pi}(t) = r_t - \frac{M_t^{-(\beta+1)}}{\Pi_t} + \sigma_{\Pi}(t)\vartheta_t$. The volatility and the drift of the process H are characterized by $\sigma_H(t) = \sigma_{\Pi}(t) - (\beta + 1)\vartheta'_t$ and

$$\mu_H(t) = \mu_{\Pi}(t) - \frac{||\sigma_{\Pi}(t)||^2}{2} - (\beta + 1)\left(r_t + \frac{||\vartheta_t||^2}{2}\right) + \frac{||-(\beta + 1)\vartheta_t' + \sigma_{\Pi}(t)||^2}{2}.$$

Moreover $H_0 = \Pi_0$.

Proof By definition it directly follows that $H_t M_t^{-\beta-1} = \Pi_t$. Thus the dynamics of Π is clearly of the claimed form. Basic arbitrage theory implies that the return rate of a financial derivative with a dividend process is the risk free rate minus the dividend yield rate under the risk-neutral measure. This, together with Definition 4.16 and the Girsanov theorem gives the expression for μ_{Π} . The solutions of M_t and Π_t imply that

$$H_{t} = M_{t}^{\beta+1} \Pi_{t}$$

= $M_{0} \Pi_{0} e^{\int_{0}^{t} \left[\mu_{\Pi}(s) - \frac{1}{2} ||\sigma_{\Pi}(s)||^{2} + (\beta+1) \left(-r_{s} - \frac{1}{2} ||\vartheta_{s}||^{2} \right) \right] ds + \int_{0}^{t} \left[\sigma_{\Pi}(s) - (\beta+1)\vartheta_{s}' \right] dW_{s}}$

and Itô's formula then gives the second last result. The last result is trivial.

The solution for power utility is presented in the following proposition. The proof consists of basic manipulations of the results above.

Proposition 4.18 (Power utility under full information) Let $x_0 \in (0, \infty)$. The optimal wealth process

$$X_t^* = x_0 \frac{H_t}{H_0} M_t^{\frac{1}{\gamma - 1}} = x_0 \frac{\Pi_t}{\Pi_0}$$

The optimal consumption process

$$c_t^* = \frac{x_0}{H_0} M_t^{\frac{1}{\gamma - 1}} = \frac{x_0}{H_t} \frac{\Pi_t}{\Pi_0}.$$

The optimal portfolio weights process

$$u_t^* = \left[\frac{-\vartheta_t'}{\gamma - 1} + \sigma_H(t)\right] \sigma_t^{-1}(S) = \sigma_{\Pi}(t)\sigma_t^{-1}(S).$$

4.2 Example 2: Logarithmic utility under full information

Let $U_1(t, c_t) = ln(c_t)$ and $U_2(X_T) = ln(X_T)$. Simple calculations show that $c_t^* = (\lambda M_t)^{-1}, X_T^* = (\lambda M_T)^{-1}$, and that $\lambda = (T + 1)/x_0$. This implies that $c_t^* = \frac{x_0}{M_t(T+1)}, X_T^* = \frac{x_0}{M_T(T+1)}$ and $\mathcal{X}(\infty) = 0$. The optimal wealth process is consequently given by

$$\begin{aligned} X_t^* &= E_{\mathcal{F}_t} \left[\int_t^T \frac{M_s}{M_t} \frac{x_0}{M_s(T+1)} ds + \frac{M_T}{M_t} \frac{x_0}{M_T(T+1)} \right] \\ &= \frac{x_0}{M_t} \frac{T+1-t}{T+1}. \end{aligned}$$

This implies (use Itô's formula) that $\sigma_{X^*}(t) = \vartheta'_t$ and hence that the optimal portfolio weights process $u_t^* = \vartheta'_t \sigma_t^{-1}(S)$.

Remark 4.19 The results on optimal consumption for logarithmic utility in the present paper are contained in the more general results of Korn and Seifried (2013) who study optimal consumption under logarithmic utility in a general semimartingale setting. Their general setting allows for partial information and other concepts such as ambiguity, nonlinear wealth dynamics and trading constraints.

5 The partial information problem

In this section we will study the partial information optimization problem we originally set out to study. We therefore drop Assumption 4.1 so that $\mathcal{F}^S \subseteq \mathcal{F}$ where the inclusion

generally is strict. We allow some repetition and restate the problem, although omitting some details. The asset price dynamics are given by

$$dS_t = D(S_t)\alpha_t dt + D(S_t)\sigma_t(S)dW_t.$$
(12)

 $D(S_t)$ is the $n \times n$ diagonal matrix with the vector S_t as main diagonal and $D(S_t)\sigma_t(S)$ is an invertible regular nonanticipative functional. The process α and the Wiener process W are generally not adapted to the observable filtration $\underline{\mathcal{F}}^S$. The instantaneous short rate r is adapted to $\underline{\mathcal{F}}^S$. The task is to maximize

$$E\left[\int_0^T U_1(t,c_t)dt + U_2\left(X_T^{u,c}\right)\right]$$

over the set of admissible consumption and portfolio weights processes (u_t, c_t) (which need to be adapted to $\underline{\mathcal{F}}^S$), where $X_T^{u,c}$ is wealth at the terminal time T, and to characterize the resulting optimal consumption process \bar{c}_t^* , optimal portfolio weights process \bar{u}_t^* and optimal wealth process \bar{X}_t^* (recall that the bar indicates a quantity under partial information).

We solve this partial information problem by transforming it to a corresponding full information problem of the type studied in Sect. 4. The main tool for this transformation is a projection of the dynamics of the process *S* in (12) (which is only partially observable) to the observable filtration \mathcal{F}^S . This type of approach is standard in the literature of partial information optimal investment and consumption, see Sect. 1. We need the following regularity in order to use the filtering theory of Sect. 2.

Assumption 5.1 $\int_0^T E\left[||\sigma_t^{-1}(S)\alpha_t||^2\right] dt < \infty.$

Let $\hat{\alpha}$ denote the filter estimate process of the process α with respect to the observable filtration $\underline{\mathcal{F}}^{S}$ (see Definition 2.6 for details). Lemma 2.8 implies that the process \overline{W} defined by

$$d\bar{W}_t = (D(S_t)\sigma_t(S))^{-1} \left(dS_t - D(S_t)\hat{\alpha}_t dt \right), \quad \bar{W}_0 = 0$$
(13)

is $\underline{\mathcal{F}}^{S}$ -Wiener. Rewrite (13) as

$$dS_t = D(S_t)\hat{\alpha}_t dt + D(S_t)\sigma_t(S)d\bar{W}_t$$
(14)

and note that S in (14) is exactly the same process as S in (12).

Now consider the optimization problem stated in the beginning of this section, but instead of (12) describing the dynamics of the asset prices *S* let (14) do so. Moreover, consider the financial market as living on the filtration $(\Omega, \mathcal{F}, P, \underline{\mathcal{F}}^S)$, instead of $(\Omega, \mathcal{F}, P, \underline{\mathcal{F}})$. In other words the filtration that we now consider is the observable filtration $\underline{\mathcal{F}}^S$, and not the larger filtration $\underline{\mathcal{F}}$. It is then clear that we have transformed the partial information problem to a corresponding full information problem. We can therefore modify the relevant full information results of Sect. 4 to corresponding results

of the original partial information problem. The rest of this section is devoted to these corresponding results.¹²

The partial information setting unique equivalent risk-neutral measure \bar{Q} is given by $\frac{d\bar{Q}}{dP} = \bar{L}_T$ on \mathcal{F}_T^S for $d\bar{L}_t = -\bar{\vartheta}_t' \bar{L}_t d\bar{W}_t$, $\bar{L}_0 = 1$, where $\bar{\vartheta}_t = \sigma_t^{-1}(S)(\hat{\alpha}_t - r_t \mathbf{1})$ (defined in the same way as ϑ_t in Sect. 4). The partial information setting stochastic discount factor is $\bar{M}_t \equiv B_t^{-1} \bar{L}_t$. The necessary regularity conditions are collected in the following assumptions.

Assumption 5.2 The local martingale (with respect to $\underline{\mathcal{F}}^S$) \bar{L}_t is an $\underline{\mathcal{F}}^S$ -martingale. The same holds for $\tilde{\tilde{L}}_t \equiv e^{-\int_0^t (\sigma_s^{-1}(S)\hat{\alpha}_s)'d\bar{W}_s - \frac{1}{2}\int_0^t ||\sigma_s^{-1}(S)\hat{\alpha}_s||^2 ds}$. Moreover, $\int_0^T ||\bar{\vartheta}_t||^2 dt < \infty$ a.s.¹³

Assumption 5.3 $E\left[\int_0^T \bar{M}_t dt + \bar{M}_T\right] < \infty$ and $\bar{\mathcal{X}}(y) \equiv E\left[\int_0^T \bar{M}_t I_1(t, y\bar{M}_t) dt + \bar{M}_T I_2(y\bar{M}_T)\right] < \infty, \forall y \in (0, \infty).$

The results of Sect. 4 are based on the assumption that the stochastic basis $(\Omega, \mathcal{F}, P, \underline{\mathcal{F}})$ satisfies the usual conditions. We therefore need the following result.

Lemma 5.4 Let the measure \tilde{Q} be defined by $\frac{d\tilde{Q}}{dP} = \tilde{L}_T$ on \mathcal{F}_T^S so that the process $\tilde{\tilde{W}}$ defined by $d\tilde{\tilde{W}}_t = \sigma_t^{-1}(S)\hat{\alpha}_t dt + d\bar{W}_t$ is $\underline{\mathcal{F}}^S$ -Wiener under \tilde{Q} .

Then the observable filtration $\underline{\mathcal{F}}^{S} = \underline{\mathcal{F}}^{\overline{W}}$. Moreover, the stochastic basis $(\Omega, \mathcal{F}, P, \underline{\mathcal{F}}^{S})$ satisfies the usual conditions.

Proof The Girsanov theorem says that \tilde{W} is an $\underline{\mathcal{F}}^S$ -Wiener process under \tilde{Q} (use Assumption 5.2). Simple calculations show that *S* has no drift under this measure and the equality of the filtrations follows from Proposition 2.3 [see also (4) in the proof of Proposition 2.5]. The stochastic basis $(\Omega, \mathcal{F}, P, \underline{\mathcal{F}})$ satisfies the usual conditions by assumption and since $\underline{\mathcal{F}}^S$, by definition, is an augmented filtration we only need to check that $\underline{\mathcal{F}}^S$ is right-continuous. The augmented filtration generated by a Wiener process is right-continuous, see e.g. Karatzas and Shreve (1998).

The following result follows directly from the above and Proposition 4.5.

Corollary 5.5 The market is observationally complete in the sense that every \mathcal{F}_T^S -measurable contingent claim ζ , with ζB_T^{-1} a.s. bounded from below and $E[\zeta B_T^{-1}\bar{L}_T] < \infty$, can be replicated by an admissible portfolio strategy.

Remark 5.6 The approach described in the present paper does not rely on any particular utility function assumptions or on any particular optimization approach. Rather, it is a unified approach for studying partial information optimization problems in observationally complete non-Markovian Itô process markets with path-dependent volatility and path-dependent interest rates.

¹² Corollary 2.11 is a more formal version of this reasoning.

¹³ $\int_0^T ||\sigma_t^{-1}(S)\hat{\alpha}_t||^2 dt < \infty$ a.s. can be shown to follow from Assumption 5.1.

Obtaining a solution for Problem 1 is now just a matter of adding a bar to the relevant quantities in the full information solution (Proposition 4.12).

Theorem 5.7 (Optimal investment and consumption under partial information) Let $x_0 \in (\tilde{\mathcal{X}}(\infty), \infty)$. The optimal wealth process

$$\bar{X}_{t}^{*} = E_{\mathcal{F}_{t}^{S}} \left[\int_{t}^{T} \frac{\bar{M}_{s}}{\bar{M}_{t}} I_{1}\left(s, \bar{\lambda}\bar{M}_{s}\right) ds + \frac{\bar{M}_{T}}{\bar{M}_{t}} I_{2}\left(\bar{\lambda}\bar{M}_{T}\right) \right]$$

where $\overline{\lambda}$ is determined by

$$E\left[\int_0^T \bar{M}_t I_1\left(t, \bar{\lambda}\bar{M}_t\right) dt + \bar{M}_T I_2\left(\bar{\lambda}\bar{M}_T\right)\right] = x_0.$$

The optimal consumption process

$$\bar{c}_t^* = I_1\left(t, \bar{\lambda}\bar{M}_t\right)$$

The optimal portfolio weights process

$$\bar{u}_t^* = \sigma_{\bar{X}^*}(t)\sigma_t^{-1}(S)$$

where $\sigma_{\bar{X}^*}(t)$ is the volatility of the optimal wealth process \bar{X}^*_t .

The results in the examples of power and logarithmic utility under full information in Sect. 4 can also easily be transformed into corresponding partial information results by adding bars to the relevant quantities. We present the partial information logarithmic utility results. The power utility results are just as easily transformed.

Example 3: Logarithmic utility under partial information

The optimal wealth process $\bar{X}_t^* = \frac{x_0}{\bar{M}_t} \frac{T+1-t}{T+1}$, the optimal consumption process $\bar{c}_t^* = \frac{x_0}{\bar{M}_t(T+1)}$, and the optimal portfolio weights process $\bar{u}_t^* = \bar{\vartheta}_t' \sigma_t^{-1}(S)$.

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