UTILITY MAXIMIZATION IN MODELS WITH CONDITIONALLY INDEPENDENT INCREMENTS

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We consider the problem of maximizing expected utility from terminal wealth in models with stochastic factors. Using martingale methods and a conditioning argument, we determine the optimal strategy for power utility under the assumption that the increments of the asset price are independent conditionally on the factor process.

1. Introduction. A classical problem in Mathematical Finance is to maximize expected utility from terminal wealth in a securities market (cf. [20, 22] for an overview). This is often called the Merton problem, since it was first solved in a continuous-time setting by Merton [26, 27]. In particular, he explicitly determined the optimal strategy and the corresponding value function for power and exponential utility functions and asset prices modeled as geometric Brownian motions.

Since then, these results have been extended to other models of various kinds. For Lévy processes (cf. [3, 7, 8, 15]), the value function can still be determined explicitly, whereas the optimal strategy is determined by the root of a real-valued function. For some affine stochastic volatility models (cf. [19, 21, 23, 25]), the value function can also be computed in closed form by solving some ordinary differential equations, while the optimal strategy can again be characterized by the root of a real-valued function.

For more general Markovian models, one faces more involved partial (integro-)differential equations that typically do not lead to explicit solutions and require a substantially more complicated verification procedure to ensure the optimality of a given candidate strategy (cf., e.g., [35] for power and [31] for exponential utility). A notable exception is given by models where the stochastic volatility is independent of the other drivers of the asset price process. In this case, it has been shown that the optimal strategy is myopic, that is, only depends on the local dynamics of the asset price (cf., e.g., [11] for exponential and [4, 6, 24] for power utility). In particular, it can be computed without having to solve any differential equations.

In the present study, we establish that this generally holds for power utility, provided that the asset price has independent increments conditional on some arbitrary factor process. As in [11], the key idea is to condition on this process,
which essentially reduces the problem to studying processes with independent increments. This in turn can be done similarly as for Lévy processes in [15]. In the following, we make this idea precise. We first introduce our setup of processes with conditionally independent increments and prove that general Lévy-driven models fit into this framework if the stochastic factors are independent of the other sources of randomness. Subsequently, we then state and prove our main result in Section 3. Given conditionally independent increments of the asset price, it provides a pointwise characterization of the optimal strategy that closely resembles the well-known results for logarithmic utility (cf., e.g., [10]). Afterward, we present some examples. In particular, we show how the present results can be used to study whether the maximal expected utility that can be achieved in affine models is finite. For the proof of our main result we utilize that exponentials of processes with conditionally independent increments are martingales if and only if they are $\sigma$-martingales. A proof of this result is provided in the Appendix.

For stochastic background, notation and terminology we refer to the monograph of Jacod and Shiryaev [14]. In particular, for a semimartingale $X$, we denote by $L(X)$ the set of $X$-integrable predictable processes and by $\varphi \cdot X$ the stochastic integral of $\varphi \in L(X)$ with respect to $X$. Moreover, we write $\mathcal{E}(X)$ for the stochastic exponential of a semimartingale $X$. When dealing with stochastic processes, superscripts usually refer to coordinates of a vector rather than powers. By $I$ we denote the identity process, that is, $I_t = t$.

2. Setup. Our mathematical framework for a frictionless market model is as follows. Fix a terminal time $T \in \mathbb{R}_+$ and a filtered probability space $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \in [0,T]}, P)$. We consider traded securities whose price processes are expressed in terms of multiples of a numeraire security. More specifically, these securities are modeled by their discounted price process $S$, which is assumed to be a $(0, \infty)^d$-valued semimartingale. We consider an investor whose preferences are modeled by a power utility function $u(x) = x^{1-p}/(1-p)$ for some $p \in \mathbb{R}_+ \setminus \{0, 1\}$ and who tries to maximize expected utility from terminal wealth. Her initial endowment is denoted by $v \in (0, \infty)$. Trading strategies are described by $\mathbb{R}^d$-valued predictable stochastic processes $\varphi = (\varphi^1, \ldots, \varphi^d) \in L(S)$, where $\varphi^i_t$ denotes the number of shares of security $i$ in the investor’s portfolio at time $t$. A strategy $\varphi$ is called admissible if its discounted wealth process $V(\varphi) := v + \varphi \cdot S$ is nonnegative (no negative wealth allowed). An admissible strategy is called optimal, if it maximizes $\psi \mapsto E(u(V_T(\psi)))$ over all competing admissible strategies $\psi$.

We need the following very mild assumption. Since the asset price process is positive, it is equivalent to NFLVR by the fundamental theorem of asset pricing.

ASSUMPTION 2.1. There exists an equivalent local martingale measure, that is, a probability measure $Q \sim P$ such that the $S$ is a local $Q$-martingale.
Since the asset price process $S$ is positive, Assumption 2.1 and [14], I.2.27, imply that $S_{\infty} > 0$ as well. By [14], II.8.3, this means that there exists an $\mathbb{R}^d$-valued semimartingale $X$ such that $S^i = S^i_0 \exp(X^i)$ for $i = 1, \ldots, d$. We interpret $X$ as the returns that generate $S$ in a multiplicative way. To solve the utility maximization problem, we make the following crucial structural assumptions on the return process $X$.

\textbf{Assumption 2.2.}

1. The semimartingale characteristics $(B^X, C^X, \nu^X)$ (cf. [14]) of $X$ relative to some truncation function such as $h(x) = x 1_{\{|x| \leq 1\}}$ can be written as

$$B^X_t = \int_0^t b^X_s \, ds, \quad C^X_t = \int_0^t c^X_s \, ds, \quad \nu^X([0, t] \times G) = \int_0^t K^X_s(G) \, ds,$$

with predictable processes $b^X, c^X$ and a transition kernel $K^X$ from $(\Omega \times \mathbb{R}_+, \mathcal{E})$ into $(\mathbb{R}^d, \mathcal{B}^d)$. The triplet $(b^X, c^X, K^X)$ is called differential or local characteristics of $X$.

2. There is a process $y$ such that $X$ also is a semimartingale with local characteristics $(b^X, c^X, K^X)$ relative to the augmented filtration $\mathcal{G} := (\mathcal{G}_t)_{t \in [0, T]}$ given by

$$\mathcal{G}_t := \bigcap_{s > t} \sigma(\mathcal{F}_s \cup \sigma((y_r)_{r \in [0, T]})),$$

and such that $b^X_t, c^X_t$ and $K^X_t(G)$ are $\mathcal{G}_0$-measurable for fixed $t \in [0, T]$ and $G \in \mathcal{B}^d$. By [14], II.6.6, this means that $X$ has $\mathcal{G}_0$-conditionally independent increments, that is, it is a $\mathcal{G}_0$-PII.

\textbf{Remarks.}

1. In the present general framework, modelling the stock prices as ordinary exponentials $S^i = S^i_0 \exp(\tilde{X}^i)$, $i = 1, \ldots, d$ for some semimartingale $\tilde{X}$ leads to the same class of models (cf. [17], Propositions 2 and 3).

2. The first part of Assumption 2.2 essentially means that the asset price process has no fixed times of discontinuity. This condition is typically satisfied, for example, for diffusions, Lévy processes and affine processes.

3. The second part of Assumption 2.2 is the crucial one. It means that the local dynamics of the asset returns at time $t$ are determined by the evolution up to time $t$ of the process $y$, which can therefore be interpreted as a stochastic factor process.

In general, a semimartingale $X$ will not remain a semimartingale with respect to an enlarged filtration (cf., e.g., [30], Chapter VI, and the references therein). Even if the semimartingale property is preserved, the characteristics generally do not remain unchanged. Nevertheless, we now show that some fairly general models satisfy this property if the factor process $y$ is independent of the other sources of randomness in the model.
**Integrated Lévy models.** In this section, we assume that $X$ is modeled as

\[ X = y \cdot B \tag{2.1} \]

for an $\mathbb{R}^{d \times n}$-valued process $y \in L(B)$ and an independent $\mathbb{R}^n$-valued Lévy process $B$ with Lévy triplet $(b^B, c^B, K^B)$. Furthermore, we suppose that the underlying filtration $F$ is generated by $B$ and $y$ (or equivalently by $X$ and $y$ if $d = n$ and $y$ takes values in the invertible $\mathbb{R}^{d \times d}$-matrices). The following result shows that Assumption 2.2 is satisfied in this case.

**Lemma 2.3.** Relative to both $F$ and $G$, $X$ is a semimartingale with $\mathcal{G}_0$-measurable local characteristics $(b^X, c^X, K^X)$ given by

\[ b^X = yb^B + \int (h(yx) - yh(x)) K^B(dx), \quad c^X = yc^By^\top, \]

\[ K^X(G) = \int 1_G(yx) K^B(dx) \quad \forall G \in \mathcal{B}^d \setminus \{0\}. \]

In particular, Assumption 2.2 is satisfied.

**Proof.** Since $B$ is independent of $y$ and $F$ is generated by $y$ and $B$, it follows from [2], Theorem 15.5, that $B$ remains a Lévy process (and in particular a semimartingale), if its natural filtration is replaced with either $F$ or $G$. Since the distribution of $B$ does not depend on the underlying filtration, we know from the Lévy–Khintchine formula and [14], II.4.19, that $B$ admits the same local characteristics $(b^B, c^B, K^B)$ with respect to its natural filtration and both $F$ and $G$. Hence it follows from [14], III.6.30, III.6.19, that $y$ belongs to $L(B)$ and $X$ is a semimartingale with respect to $F$ and $G$. Its characteristics can be derived by applying [17], Proposition 2. The $\mathcal{G}_0$-measurability is obvious. \(\square\)

In particular, the prerequisites of Lemma 2.3 are satisfied for an $\mathbb{R}^{m}$-valued predictable process $\tilde{y}$ independent of a standard Brownian motion $B$, and a measurable function $f$ such that $(y_t)_{0 \leq t \leq T} = (f(t, (\tilde{y}_s)_{0 \leq s \leq t}))_{0 \leq t \leq T}$ belongs to $L(B)$. This exemplifies that the dynamics of the asset returns can depend on the whole history of the factor process.

**Time-changed Lévy models.** We now show that Assumption 2.2 also holds for time-changed Lévy models. For Brownian motion, stochastic integration and time changes lead to essentially the same models by the Dambis–Dubins–Schwarz theorem. For general Lévy processes with jumps, however, the two classes are quite different. More details concerning the theory of time changes can be found in [13], whereas their use in modeling is dealt with in [5, 17]. Here, we assume that the process $X$ is given by

\[ X = \int_0^\cdot \mu(y_s) \, ds + B_{f_0 y_s} \, ds \tag{2.2} \]
for a \((0, \infty)\)-valued predictable process \(y\) and a measurable mapping \(\mu : \mathbb{R} \to \mathbb{R}^d\) such that \(\int_0^T y_s \, ds < \infty\) and \(\int_0^T |\mu(y_s)| \, ds < \infty\), \(P\)-a.s., and an independent \(\mathbb{R}^d\)-valued Lévy process \(B\) with Lévy–Khintchine triplet \((b^B, c^B, K^B)\). Moreover, we suppose that the underlying filtration is generated by \(X\) and \(y\). We have the following analogue of Lemma 2.3.

**Lemma 2.4.** Relative to both \(\mathcal{F}\) and \(\mathcal{G}\), \(X\) is a semimartingale with \(\mathcal{G}_0\)-measurable local characteristics \((b^X, c^X, K^X)\) given by

\[
\begin{align*}
   b^X &= \mu(y) + b^By, \\
   c^X &= c^By, \\
   K^X(G) &= K^B(G)y
\end{align*}
\]

\(\forall G \in \mathcal{B}^d\). In particular, Assumption 2.2 is satisfied.

**Proof.** Relative to \(\mathcal{F}\), the assertion follows exactly as in the proof of [29], Proposition 4.3. For the corresponding statement relative to the augmented filtration \(\mathcal{G}\), let \(Y = \int_0^T y_s \, ds\) and \(U_r := \inf\{q \in \mathbb{R}_+ : Y_q \geq r\}\). Define the \(\sigma\)-fields

\[
\mathcal{H}_t := \bigcap_{s>t} \sigma(B_q)_{q \in [0,s]}, (U_r)_{r \in \mathbb{R}_+}.
\]

Since \(B\) is independent of \(y\) and hence \(Y\), it remains a Lévy process relative to the filtration \(\mathcal{H} := (\mathcal{H}_t)_{t \in \mathbb{R}_+}\). Its distribution does not depend on the underlying filtration, and hence we know from the Lévy–Khintchine formula and [14], II.4.19, that it is a semimartingale with local characteristics \((b^B, c^B, K^B)\) relative to \(\mathcal{H}\). By [17], Proposition 5, the time-changed process \((\tilde{B}_\theta)_{\theta \in [0,T]} := (B_{Y_\theta})_{\theta \in [0,T]}\) is a semimartingale on \([0, T]\) relative to the time-changed filtration \((\mathcal{H}_\theta)_{\theta \in [0,T]} := (\mathcal{H}_{Y_\theta})_{\theta \in [0,T]}\) with differential characteristics \((\tilde{b}, \tilde{c}, \tilde{K})\) given by

\[
\begin{align*}
   \tilde{b}_\theta &= b^B y_\theta, \\
   \tilde{c}_\theta &= c^B y_\theta, \\
   \tilde{K}_\theta(G) &= K^B(G)y_\theta
\end{align*}
\]

\(\forall G \in \mathcal{B}^d\). Furthermore, it follows as in the proof of [29], Proposition 4.3, that \(\mathcal{H}_t = \mathcal{G}_t\) for all \(t \in [0, T]\). The assertion now follows by applying [17], Propositions 2 and 3.

**Remarks.**

1. For the proof of Lemma 2.4 we had to assume that the given filtration is generated by the process \((y, X)\) or equivalently \((Y, X)\). In reality, though, the integrated volatility \(Y\) and the volatility \(y\) typically cannot be observed directly. Therefore the canonical filtration of the return process \(X\) would be a more natural choice. Fortunately, \(Y\) and \(y\) are typically adapted to the latter if \(B\) is an infinite activity process (cf., e.g., [34]).

2. A natural generalization of (2.2) is given by models of the form

\[
X = \int_0^\cdot \mu(y_s^{(1)}, \ldots, y_s^{(n)}) \, ds + \sum_{i=1}^n B^{(i)}_{Y^{(i)}}
\]
for $\mu : (0, \infty)^n \rightarrow \mathbb{R}^d$, strictly positive predictable processes $y^{(i)}, Y^{(i)} = \int_0^t y^{(i)}(s) \, ds$ and independent Lévy processes $B^{(i)}, i = 1, \ldots, n$. If one allows for the use of the even larger filtration generated by all $y^{(i)}, B^{(i)}_y, i = 1, \ldots, n$, the proof of Lemma 2.4 remains valid. If $Y^{(i)}$ is interpreted as business time in some market $i$, this class of models allows assets to be influenced by the changing activity in different markets.

3. Optimal portfolios. For asset prices with conditionally independent increments we can now characterize the solution to the Merton problem as follows.

**Theorem 3.1.** Suppose Assumptions 2.1, 2.2 hold, and assume there exists an $\mathbb{R}^d$-valued process $\pi \in L(X)$ such that the following conditions are satisfied up to a $dP \otimes dt$-null set on $\Omega \times [0, T]$:

1. $K^X(\{x \in \mathbb{R}^d : 1 + \pi^\top x \leq 0\}) = 0$.
2. $\int |x(1 + \pi^\top x)^{-p} - h(x)| K^X(dx) < \infty$.
3. For all $\eta \in \mathbb{R}^d$ such that $K^X(\{x \in \mathbb{R}^d : 1 + \eta^\top x < 0\}) = 0$, we have
   $$ (\eta^\top - \pi^\top) \left( b^X - pc^X \pi + \int \left( \frac{x}{(1 + \pi^\top x)^p} - h(x) \right) K^X(dx) \right) \leq 0. $$
4. $\int_0^T |\alpha_s| \, ds < \infty$, where
   $$ \alpha := (1 - p) \pi^\top b^X - \frac{p(1 - p)}{2} \pi^\top c^X \pi + \int ((1 + \pi^\top x)^{1-p} - 1 - (1 - p) \pi^\top h(x)) K^X(dx). $$

Then there exists a $\mathcal{G}_0$-measurable process $\tilde{\pi}$ satisfying conditions 1–4 such that the strategy $\varphi = (\varphi^1, \ldots, \varphi^d)$ defined as

$$ \varphi^i_t := \tilde{\pi}^i(t) \frac{v^i(\tilde{\pi} \cdot X)_t}{S^-_t}, \quad i = 1, \ldots, d, t \in [0, T], $$

is optimal with value process $V(\varphi) = v(\tilde{\pi} \cdot X)$. The corresponding maximal expected utility is given by

$$ E(u(V_T(\varphi))) = \frac{v^{1-p}}{1-p} E\left( \exp\left( \int_0^T \alpha_s \, ds \right) \right). $$

In particular, if $\pi$ is $\mathcal{G}_0$-measurable, it is possible to choose $\tilde{\pi} = \pi$.

**Proof.** Step 1: In view of conditions 1–4, the measurable selection theorem [32], Theorem 3, and [13], Proposition 1.1, show the existence of $\tilde{\pi}$, since $(b^X, c^X, K^X)$ are $\mathcal{G}_0$-measurable by Assumption 2.2. Hence we can assume without loss of generality that $\pi$ is $\mathcal{G}_0$-measurable, because we can otherwise pass to $\tilde{\pi}$ instead.
Step 2: Since \( \pi \) and hence \( \varphi \) is \( F \)-predictable by assumption and \( \mathcal{F}_t \subset \mathcal{G}_t \) for all \( t \in [0, T] \), it follows that \( \varphi \) is \( G \)-predictable as well. In view of Assumption 2.2, the local characteristics of \( X \) relative to \( F \) coincide with those relative to \( G \). Together with [14], III.6.30, this implies that we have \( \pi \in L(X) \) and hence \( \varphi \in L(S) \) w.r.t. \( G \), too.

Step 3: The wealth process associated to \( \varphi \) is given by
\[
V(\varphi) = v + \varphi \cdot S = v(1 + \mathcal{E}(\pi \cdot X)_- \cdot (\pi \cdot X)) = v\mathcal{E}(\pi \cdot X).
\]
Since condition 1 and [14], I.4.61, imply \( V(\varphi) > 0 \), the strategy \( \varphi \) is admissible.

Step 4: Let \( \psi \) be any admissible strategy. Together with Assumption 2.1 and [14], I.2.27, admissibility implies \( V(\psi) = 0 \) on the predictable set \( \{ V_-(\psi) = 0 \} \), because we can otherwise consider \( \tilde{\psi} := 1_{\{ V_-(\psi) > 0 \}} \psi \) without changing the wealth process. Consequently, we can write \( \psi^i = \eta^i V_-(\psi)/S^i \) for \( i = 1, \ldots, d \) and some \( \mathbb{R}^d \)-valued \( F \)-predictable process \( \eta \). The admissibility of \( \psi \) implies \( \eta_t^\top \Delta X_t \geq -1 \) which in turn yields
\[
K^X([x \in \mathbb{R}^d : 1 + \eta^\top_t x < 0]) = 0
\]
outside some \( dP \otimes dt \) null set. Moreover, it follows as above that \( \psi \in L(S) \) w.r.t. \( G \) as well. Since \( \int_0^T |\alpha_s| \, ds < \infty \) outside some \( P \)-null set by condition 4, the process
\[
L_t := \exp\left( \int_t^T \alpha_s \, ds \right) = L_0\mathcal{E}\left( \int_t^T \alpha_s \, ds \right)
\]
is indistinguishable from a \( \mathcal{G}_t \)-process of finite variation and hence a \( G \)-semimartingale because \( \pi \) and \( (b^X, c^X, K^X) \) are \( \mathcal{G}_0 \)-measurable. The local \( G \)-characteristics \( (b, c, K) \) of \( L/L_0 \) \( V(-p) V(\psi) \) can now be computed with [17], Propositions 2 and 3. In particular, we get
\[
K(G) = \int 1_G \left( \frac{L_0^{-1} V_-(\varphi)^{-p} V_-(\psi)}{(1 + \pi^\top x)^p} - 1 \right) K^X(dx)
\]
for all \( G \in \mathcal{B} \setminus \{0\} \), which combined with condition 2 yields
\[
\int_{\{|x| > 1\}} |x| K(dx) < \infty
\]
outside some \( dP \otimes dt \) null set. Moreover, insertion of the definition of \( \alpha \) leads to
\[
b = \int (h(x) - x) K(dx) + \frac{L_0}{L_0} V_-(\varphi)^{-p} V_-(\psi)(\eta^\top - \pi^\top)
\]
\[
\times \left( b^X - pc^X \pi + \int \frac{x}{(1 + \pi^\top x)^p} - h(x) K^X(dx) \right),
\]
and hence
\[
b + \int (x - h(x)) K(dx) \leq 0
\]
\(dP \otimes dt\)-almost everywhere on \(\Omega \times [0, T]\) by (3.2) and condition 3. In view of (3.3) and (3.5) the process \((L/L_0)V(\varphi)^{-p}V(\psi)\) is therefore a \(G\)-supermartingale by [18], Lemma A.2, and [16], Proposition 3.1.

**Step 5:** For \(\psi = \varphi\), (3.3), (3.4) and [18], Lemma A.2, show that \(Z := (L/L_0) \times (V(\varphi)/v)^{1-p}\) is a strictly positive \(G_0\)-martingale by [17], Proposition 3, \(\log(Z)\) is a \(G_0\)-PII, hence \(Z\) and in turn \((L/L_0)V(\varphi)^{1-p}\) are \(G\)-martingales by Lemma A.1.

**Step 6:** Now we are ready to show that \(\varphi\) is indeed optimal. Since \(u\) is concave, we have

\[
(3.6) \quad u(V_T(\psi)) \leq u(V_T(\varphi)) + u'(V_T(\varphi))(V_T(\psi) - V_T(\varphi))
\]

for any admissible \(\psi\). This implies

\[
E(u(V_T(\psi))|\mathcal{G}_0) \leq E(u(V_T(\varphi))|\mathcal{G}_0) + L_0E\left(\frac{L_T}{L_0}V_T(\psi)^{-p}V_T(\psi) - \frac{L_T}{L_0}V_T(\varphi)^{1-p}|\mathcal{G}_0\right)
\]

because \((L/L_0)V(\varphi)^{-p}V(\psi)\) is a \(G\)-supermartingale and \((L/L_0)V(\varphi)^{1-p}\) is a \(G\)-martingale, both starting at \(v^{1-p}\). Taking expectations, the optimality of \(\varphi\) follows. Likewise, the \(G\)-martingale property of \((L/L_0)V(\varphi)^{1-p}\) yields the maximal expected utility. \(\square\)

**Remarks.**

1. The first condition ensures that the wealth process of the optimal strategy is positive. It is satisfied automatically if the asset price process is continuous. In the presence of unbounded positive and negative jumps it rules out short selling and leverage. The second condition is only needed to formulate the crucial condition 3, which characterizes the optimal strategy. A sufficient condition for its validity is given by

\[
b^X - p c^X \pi + \int \left(\frac{x}{1 + \pi^T x} - h(x)\right)K^X(dx) = 0.
\]

While one does not have to require NFLVR if this stronger condition holds as well, it is less general than condition 3 in the presence of jumps (cf. [12] for a related discussion).

2. The fourth condition ensures that the maximal conditional expected utility is finite. By [14], III.6.30, it is automatically satisfied for \(\pi \in L(X)\) if \(X\) is continuous. Let us emphasize that the maximal unconditional expected utility does not necessarily have to be finite for \(p \in (0, 1)\). On the contrary, for \(p \in (1, \infty)\), the maximal expected utility is always finite because \(u(x) = x^{1-p}/(1 - p)\) is bounded from above in this case. Indeed, this can also be seen directly from Theorem 3.1, since conditions 1 and 3 combined with the Bernoulli inequality show that \(\alpha \geq 0\), respectively, \(\alpha \leq 0\) for \(p \in (0, 1)\), respectively, \(p \in (1, \infty)\).
3. Given the mild regularity condition 4, the optimal strategy at \( t \) is completely described by the local characteristics at \( t \), that is, it is myopic. This parallels well-known results for logarithmic utility (cf., e.g., [10]). It is important to note, however, that whereas the optimal strategy is myopic in the general semimartingale case for logarithmic utility, this only holds for power utility if the return process \( X \) has conditionally independent increments. Otherwise an additional nonmyopic term appears (see, e.g., [21, 23, 35]).

4. In the proof of Theorem 3.1 it is shown that the components of any admissible strategy \( \psi \) can be written as \( \psi^i = \eta^i V(\psi) - S^i_w \), where \( \eta^i \) represents the fraction of wealth invested into asset \( i \). This parametrization allows one to introduce convex constraints to the present setup by requiring these fractions to lie inside some nonempty convex set \( C \subset \mathbb{R}^d \). The most prominent example is given by the set \( C = [0, 1]^d \), which rules out short selling and leverage. If there exists a \( C \)-valued process \( \pi \) as in Theorem 3.1, it is optimal for the constrained problem as soon as condition 3 is satisfied for all \( \eta \in C \).

4. Examples. We now consider some concrete models where the results of the previous section can be applied. For ease of notation, we consider only a single risky asset (i.e., \( d = 1 \)), but the extension to multivariate versions of the corresponding models is straightforward.

**Generalized Black–Scholes models.** Let \( B \) be a standard Brownian motion, \( y \) an independent adapted càdlàg process and again denote by \( I \) the identity process \( I_t = t \). Consider measurable functions \( \mu : \mathbb{R} \to \mathbb{R} \) and \( \sigma : \mathbb{R} \to (0, \infty) \) such that \( \mu(y^-) \in L(I) \) and \( \sigma(y^-) \in L(B) \) and suppose the discounted stock price \( S \) is given by

\[
S = S_0 \mathbb{E}(\mu(y^-) \cdot I + \sigma(y^-) \cdot B).
\]

For \( X := \mu(y^-) \cdot I + \sigma(y^-) \cdot B \), [14], II.4.19, and [17], Proposition 3, yield \( b^X = \mu(y^-) \) as well as \( c^X = \sigma^2(y^-) \) and \( K^X = 0 \). In view of Lemma 2.3, Assumption 2.2 is satisfied. Define

\[
\pi := \frac{\mu(y^-)}{p\sigma^2(y^-)}.
\]

By Theorem 3.1 and the second remark succeeding it, the strategy \( \varphi := \pi v\mathcal{E}(\pi \cdot X)/S \) is optimal provided that \( \pi \in L(X) \). If \( y^- \) is \( E \)-valued for some \( E \subset \mathbb{R} \), this holds true, for example, if the mapping \( x \mapsto \mu(x)/\sigma^2(x) \) is bounded on compact subsets of \( E \).

**Remark 4.1.** This generalizes results of [6] by allowing for an arbitrary semimartingale factor process instead of a Lévy-driven Ornstein–Uhlenbeck (henceforth OU) process. Notice, however, that unlike [6] we only consider utility from terminal wealth and do not obtain a solution to more general consumption problems. Finiteness of the maximal expected utility is ensured in the case...
$p > 1$ in our setup, which complements the results of [6]. They consider the case $p \in (0, 1)$ and prove that the maximal expected utility is finite subject to suitable linear growth conditions on the coefficient functions $\mu(\cdot)$ and $\sigma(\cdot)$ and an exponential moment condition on the driver of the OU process.

Barndorff-Nielsen and Shephard [1]. If we set $\mu(x) := \kappa + \delta x$ for constants $\kappa, \delta \in \mathbb{R}$, let $\sigma(x) := \sqrt{x}$ and choose an OU process

$$
(4.1) \quad dy_t = -\lambda y_t + dZ_{\lambda t}, \quad y_0 \in (0, \infty),
$$

for a constant $\lambda > 0$ and some subordinator $Z$ in the generalized Black–Scholes model above, we obtain the model of Barndorff-Nielsen and Shephard [1]. Since $y_t \geq y_0 e^{-\lambda T} > 0$ in this case,

$$
\pi := \frac{\mu(y_-)}{p \sigma^2(y_-)} = \frac{\kappa}{py_-} + \frac{\delta}{p}
$$

is bounded and hence belongs to $L(X)$. Consequently, $\varphi_t = \pi V(\varphi)/S$ is optimal.

Remark 4.2. This recovers the optimal strategy obtained by [4]. Similarly to [6], Benth, Karlsen and Reikvam [4] consider the case $p \in (0, 1)$ and prove that the maximal expected utility is finite subject to an exponential moment condition on the Lévy measure $K^Z$ of $Z$. Our results complement this by ascertaining that the same strategy is always optimal (with not necessarily finite expected utility), as well as optimal with finite expected utility in the case $p > 1$.

Carr et al. [5]. In this section we turn to the time-changed Lévy models proposed by [5], that is, we let

$$
(4.2) \quad X_t = \mu t + B_{\int_0^t y_s \, ds}, \quad \mu \in \mathbb{R},
$$

for a Lévy process $B$ with Lévy–Khintchine triplet $(b^B, c^B, K^B)$ and an independent OU process $y$ given by (4.1). By Lemma 2.4, Assumption 2.2 holds. Hence we obtain the following corollary.

Corollary 4.3. Suppose $B$ has both positive and negative jumps and assume there exists a process $\pi$ such that the following conditions are satisfied:

1. $K^B((x \in \mathbb{R}^d : 1 + \pi x \leq 0)) = 0$.
2. $\int_0^T (\int |x(1 + \pi x)|^{-p} - h(x) K^B(dx)) \, dt < \infty$.
3. For all $\eta \in \mathbb{R}^d$ such that $K^B((x \in \mathbb{R}^d : 1 + \eta x < 0) = 0$, we have

$$
(\eta - \pi) \left( \frac{\mu}{y_-} + b^B \right) - pc^B \pi + \int \left( \frac{x}{(1 + \pi x)^p} - h(x) \right) K^B(dx) \leq 0.
$$

Then there exists a $\mathcal{F}_0$-measurable process $\bar{\pi} \in L(X)$ satisfying conditions 1–3 such that $\varphi = \bar{\pi} v \mathcal{E}(\pi \cdot X)/S_-$ is optimal.
PROOF. Since $B$ has both positive and negative jumps, the model satisfies Assumption 2.1 by [28], Lemma 4.42. Moreover, $\pi$ is bounded by condition 1. Hence it belongs to $L(X)$ and condition 2 implies that condition 4 of Theorem 3.1 is also satisfied. By Lemma 2.4, conditions 1–3 imply conditions 1–3 of Theorem 3.1. Consequently, the assertion immediately follows from Theorem 3.1. □

For $\mu = 0$ one recovers [19], Theorem 3.4, where the optimal fraction $\pi$ of wealth invested into stocks can be chosen to be deterministic. For $\mu \neq 0$, the optimal fraction depends on the current level of the activity process $y$. As for the generalized Black–Scholes models above, it is important to emphasize that the optimal strategy $\varphi$ is only ensured to lead to finite expected utility in the case $p > 1$. However, the results provided here allow us to complete the study of the case $p \in (0, 1)$ for $\mu = 0$ started in [19]. Using Corollary 4.3, we can now show that if there exists $\pi \in \mathbb{R}$ satisfying conditions 1–3, the exponential moment condition in [19], Theorem 3.4, is necessary and sufficient for the maximal expected utility to be finite. The key observation is that the random variable $\int_0^T \alpha_s ds$ from Theorem 3.1 turns out to be infinitely divisible for $\mu = 0$.

COROLLARY 4.4. Let $\mu = 0$ and suppose there exists $\pi \in \mathbb{R}$ satisfying the conditions of Corollary 4.3. Then the maximal expected utility corresponding to the optimal strategy $\varphi := \pi \nu \delta(\pi X) / S$ is always finite for $p > 1$, whereas for $p \in (0, 1)$ it is finite if and only if

\[
\int_0^T \int_1^\infty \exp\left(\frac{e^{-\lambda t} - 1}{\lambda} C z\right) K^Z(dz) dt < \infty,
\]

where

\[
C := (p - 1) b^B \pi + \frac{p(1 - p)}{2} c^B \pi^2 - \int ((1 + \pi x)^{1-p} - 1 - (1 - p) \pi h(x)) K^B(dx).
\]

If the maximal expected utility is finite, it is given by

\[
E(u(V_T(\varphi))) = \frac{v^{1-p}}{1-p} \exp\left(\int_0^T \left(\lambda b^Z \tilde{\alpha}(s) + \int (e^{\tilde{\alpha}(s)} - 1 - \tilde{\alpha}(s) h(z)) \lambda K^Z(dz) ds + \tilde{\alpha}(0) y_0\right) ds + \tilde{\alpha}(0) y_0\right) - 1)/\lambda.
\]
Proof. After inserting the characteristics of $X$, Theorem 3.1 shows that the maximal expected utility is given by

$$E(u(V_T(\varphi))) = \frac{u^{1-p}}{1-p} E\left(\exp\left(-C \int_0^T y_t \, dt\right)\right).$$

The process $(y, \int_0^T y_s \, ds)$ is an affine semimartingale by [17], Proposition 2, hence [17], Corollary 3.2, implies that the characteristic function of the random variable $\int_0^T y_s \, ds$ is given by

$$E\left(\exp\left(iu \int_0^T y_s \, ds\right)\right) = \exp\left(i bu + \int (e^{iux} - 1 - iuh(x)) K(dx)\right),$$

for all $u \in \mathbb{R}$, with

$$K(G) := \int_0^T \int_1^G \left(1 - e^{-\lambda t} - \frac{1 - e^{-\lambda t} - e^{-\lambda t} h(z)}{\lambda} K^Z(dz) \right) dt \quad \forall G \in \mathcal{B}$$

and

$$b := \lambda b^Z \left(\frac{e^{-\lambda T} - 1 + \lambda T}{\lambda^2} + y_0 \left(\frac{1 - e^{-\lambda T}}{\lambda}\right)\right) + \int_0^T \int \left(h \left(1 - \frac{e^{-\lambda t}}{\lambda} - \frac{1 - e^{-\lambda t}}{\lambda} h(z)\right) \lambda K^Z(dz) \right) dt.$$  

Since $K^Z$ is a Lévy measure, that is, satisfies $K^Z([0]) = 0$ and integrates $1 \wedge |x|^2$, one easily verifies that $b$ is finite and $K$ is a Lévy measure, too. By the Lévy–Khintchine formula, the distribution of $\int_0^T y_s \, ds$ is therefore infinitely divisible. Consequently (4.4) and [33], Theorems 7.10 and 25.17, yield that $E(u(V_T(\varphi)))$ is finite if and only if

$$\int_{\{|x| > 1\}} e^{-C x} K(dx) = \int_0^T \int_{\{(1 - e^{-\lambda t})z/\lambda > 1\}} \exp\left(\frac{e^{-\lambda t} - 1}{\lambda} C z\right) \lambda K^Z(dz) \, dt$$

is finite. Since $\lambda > 0$ and the Lévy measure $K^Z$ of the subordinator $Z$ is concentrated on $\mathbb{R}_+$ by [33], 21.5, the assertion follows. For $p > 1$, condition 3 of Corollary 4.3 and the Bernoulli inequality show that $C$ is positive. Consequently, (4.3) is always satisfied. □

Since the exponential moment condition in Corollary 4.4 depends on the time horizon, it is potentially only satisfied if $T$ is sufficiently small. This resembles the situation in the Heston model, where the maximal expected utility can be infinite for some parameters and sufficiently large $T$, if $p \in (0, 1)$ (cf. [19]). However, a qualitatively different phenomenon arises here. Whereas expected utility can only tend to infinity continuously in the Heston model, it can suddenly jump to infinity here. This means that the utility maximization problem is not stable with respect to the time horizon in this case.
EXAMPLE 4.5 (Sudden explosion of maximal expected utility). In the setup of Corollary 4.4 consider $p \in (0, 1)$, $K^B = 0$, $b^B \neq 0$, $c^B = 1$ and hence $C = (b^B)^2(p - 1)/2p < 0$. Define the Lévy measure

$$K^Z(dz) := \frac{1}{1_{(1, \infty)}(z)} \exp\left(\frac{C}{2\lambda} \frac{dz}{z^2}\right),$$

and let $b^Z = 0$ relative to the truncation function $h(z) = 0$ on $\mathbb{R}$. Setting $T_\infty := \log(2)/\lambda$, we obtain

$$\int_{1}^{\infty} \exp\left(\frac{e^{-\lambda t} - 1}{\lambda} Cz\right) K^Z(dz) \left\{ \begin{array}{ll} \leq 1, & \text{for } t \leq T_\infty, \\ = \infty, & \text{for } t > T_\infty. \end{array} \right.$$

Consequently, by Corollary 4.4, the maximal expected utility that can be obtained by trading on $[0, T]$ is finite for $T \leq T_\infty$ and satisfies

$$E(u(V_T(\varphi))) \leq \frac{v^{1-p}}{1-p} \exp(\log(2)/\lambda + |C/2\lambda|\gamma_0) < \infty.$$

Hence the maximal expected utility is actually bounded from above for $T \leq T_\infty$. For $T > T_\infty$, however, is is infinite by Corollary 4.4.

Since $u(V_T(\varphi)) = V_T(\varphi)^{1-p}/(1 - p)$ is an exponentially affine process for $\mu = 0$, the finiteness of the maximal expected utility is intimately linked to moment explosions of affine processes (cf. [9] and the references therein for more details).

APPENDIX: EXPONENTIAL MARTINGALES

In the proof of Theorem 3.1 we used that exponentials of processes with conditionally independent increments are martingales if and only if they are $\sigma$-martingales. In this appendix, we give a proof of this result.

**Lemma A.1.** Let $X$ be an $\mathbb{R}$-valued process with conditionally independent increments relative to some $\sigma$-field $\mathcal{H}$. If $X$ admits local characteristics $(b, c, K)$ with respect to some truncation function $h$, the following are equivalent:

1. $\exp(X)$ is a martingale on $[0, T]$.
2. $\exp(X)$ is a local martingale on $[0, T]$.
3. $\exp(X)$ is a $\sigma$-martingale on $[0, T]$.
4. Up to a $dP \otimes dt$-null set, we have $\int_{|x| > 1} e^x K(dx) < \infty$ and

$$b + \frac{c}{2} + \int (e^x - 1 - h(x)) K(dx) = 0.$$

**Proof.** The implications $1 \Rightarrow 2 \Rightarrow 3$ follow from [16], Lemma 3.1. Moreover, [16], Lemma 3.1, and [17], Proposition 3, yield $3 \Leftrightarrow 4$. Consequently, it remains to show $4 \Rightarrow 1$. 


By [16], Proposition 3.1, the $\sigma$-martingale $\exp(X)$ is a supermartingale. Therefore it suffices to show $E(\exp(X_T)) = 1$. In view of [2], Satz 44.3, a regular version $R(\omega, dx)$ of the conditional distribution of $X_T$ w.r.t. $\mathcal{H}$ exists. From [2], Section 44, and [14], II.6.6, we get

$$
\int e^{iux} R(\omega, dx)
= E(\exp(iuX_T)|\mathcal{H})(\omega)
= \exp\left(iuB_T(\omega) - \frac{1}{2}uC_T(\omega)u \right)
+ \int_{[0,T] \times \mathbb{R}^d} (e^{iux} - 1 - iuh(x)) v(\omega, dt, dx),
$$

where $B = b \cdot I$, $C = c \cdot I$ and $v = K \otimes I$ denote the semimartingale characteristics of $X$. By the Lévy–Khintchine formula [33], Theorem 8.1, $R(\omega, \cdot)$ is therefore a.s. infinitely divisible. Since any supermartingale is a special semimartingale by [13], Proposition 2.18, it follows from [16], Corollary 3.1, that $\exp(X^i)$ is a local martingale. Hence

$$
(A.2) \quad \int_{[0,T] \times \{x>1\}} e^x v(dt, dx) < \infty, \quad P\text{-a.s.}
$$

by [17], Proposition 3, and [16], Lemma 3.1. By [33], Theorems 7.10 and 25.17, (A.1) and (A.2) show that $\int e^x R(\omega, dx) = 1$, $P$-a.s. and hence

$$
E(\exp(X_T)) = \int \int e^x R(\omega, dx) P(d\omega) = 1.
$$

This proves the assertion. □

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