Asset pricing with costly short sales *

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Abstract

We study a dynamic general equilibrium model with costly-to-short stocks and heterogeneous beliefs. The closed-form solution to the model shows that costly short sales drive a wedge between the valuation of assets that promise identical cash flows but are subject to different trading arrangements. Specifically, we show that the price of an asset is given by the risk-adjusted present value of future cash flows which include both dividends and an endogenous lending yield. This formula implies that returns satisfy a modified CAPM and sheds light on recent findings about the explanatory power of lending fees in the cross-section of returns. In particular, we show that once returns are appropriately adjusted for lending fees, stocks with low and high shorting costs offer similar risk-return tradeoffs.

Keywords: Shorting fees; Securities lending; Heterogeneous beliefs; Dynamic equilibrium.

JEL Codes: D51, D52, G11, G12.

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1 Introduction

Securities lending and borrowing is a critical function that makes financial markets more efficient through improved liquidity and price discovery. In the U.S. alone, short selling accounts for more than a quarter of trading volume in the stock market and the value of securities on loans has recently surpassed $1.4trillion (Gensler 2021). Historically, the primary suppliers of shares to loan have mostly been investment firms, pension funds, insurance companies, passive funds, and exchange traded funds (ETFs). However, the practice of securities lending is now extending to non institutional investors because it can be a significant revenue source that often offsets management fees and transaction costs.\footnote{See, e.g., Table 3 in iShares Report on Securities Lending. In addition, a number of important broker-dealers have recently started lending programs that allow retail customers to earn incremental income. See, for example, the Fully Paid Lending program of Fidelity and the similar programs put in place by TD Ameritrade/Charles Schwab, BNY Mellon, and E-Trade among others.}

For example, Kashyap, Kovrijnykh, Li and Pavlova (2020) report that securities lending contributed 5% of the total revenues of both BlackRock and State Street in 2017 while the data provider DataLend (2022) reports that the global revenues of security lenders have been growing steadily in the last decade to reach a level in excess of $9billion in 2021.

As a last indication of the current importance of securities lending, we note that the U.S. Securities and Exchange Commission Rule 10c–1, which is currently under review, will soon create a new reporting and disclosure regime for all participants in the securities lending market (Gensler 2021).

Despite the crucial importance of short selling and the extensive literature on the effects of short sales constraints on asset returns,\footnote{See among others Seneca (1967), Miller (1977), Harrison and Kreps (1978), Figlewski (1981), Diether, Malloy and Scherbina (2002), Jones and Lamont (2002), Mitchell, Pulvino and Stafford (2002) Scheinkman and Xiong (2003), Ofek, Richardson and Whitelaw (2004), and Atmaz and Basak (2019) for key contributions, and either Reed (2013) or Jiang, Habib and Hasan (2020) for comprehensive surveys.} there are only few studies that explicitly analyze the role of securities lending in the price formation process.\footnote{The short list of such studies includes Duffie (1996), D’Avolio (2002), Duffie, Gärleanu and Pedersen (2002), Cohen, Diether and Malloy (2007), Drechsler and Dreschler (2018) (henceforth DD) and, more recently, Nutz and Scheinkman (2020), Atmaz, Basak and Ruan (2021), Gärleanu, Panageas and Zheng (2021), and Chen, Kaniel and Opp (2022) which we briefly review below.} In particular, there is currently no commonly accepted theoretical model for the joint endogenous determination of asset returns and lending fees. We contribute to bridging this gap by developing a
tractable dynamic general equilibrium model of asset prices and lending fees with a focus on the return-augmenting effect of securities lending.

Specifically, we consider a continuous-time Lucas economy populated by two groups of investors who have logarithmic utility and heterogeneous dogmatic beliefs about the growth rate of the economy. The financial market includes a riskless asset, and two long-lived risky assets that each represent a claim to a constant fraction of the aggregate dividend. The first risky asset (asset 1) can be shorted at a cost that is to be determined in equilibrium, while the second risky asset (asset 2) cannot be shorted. The assumption that the two risky assets have proportional dividends implies that they are Siamese twins (see e.g., Froot and Dabora (1999) and De Jong, Rosenthal and Van Dijk (2009)). This assumption serves two purposes. First, it allows us to easily account for the fact that the shares of a given stock are often not all available for lending because of variety of reasons that include the existence of different share classes (Mei, Scheinkman and Xiong 2009), dual or cross-listings as in the case of ADRs (Blau, Van Ness and Warr 2012), or the occurrence of a partial public offering (Lamont and Thaler 2003). Second, and perhaps more importantly, this assumption allows us to study the effect of costly short sales on ex-ante identical assets within a single general equilibrium model where all markets clear rather than across different models.

An investor who wants to short a stock must first borrow the required shares from another investor who holds a long position. In line with the empirical evidence (D’Avolio 2002, Baklanova, Copeland and McCaughrin 2015, Gensler 2021, Chen et al. 2022) we assume that the securities lending market is intermediated by a finite number of lending agents. In our model, investors wanting to go short over the next instant are randomly matched with one of the lending agents. Each lending agent sets a shorting fee to maximize the flow of shorting revenues taking as given the aggregate short demand schedule of the investors who are matched with her. Once the terms of the short transactions are set, each lending agent borrows the required shares from the custodian bank that holds securities on behalf of long investors, lends them over an infinitesimal time

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4The assumption of a constant disagreement simplifies the solution of the model by fixing the identity of the optimist and pessimist, and thereby limiting the number of state variables. Moving to a richer environment with a stochastic disagreement leads to a more cumbersome, yet fully explicit, characterization of the equilibrium but does not affect the underlying economics. We develop such an extension in Appendix B.
interval, and transfers back the securities and the induced shorting fees to the custodian bank who in turn redistributes them on a value weighted basis to long investors.\(^5\) This approach is, to the best of our knowledge, new to the literature and allows us to easily integrate costly short sales into an otherwise standard dynamic asset pricing model. Our modelling implies that the shortable asset 1 entitles its owners to a convenience yield that can be endogenously determined in equilibrium by matching the aggregate flows of lending fees paid and received by investors.

If asset 1 could be shorted at no cost, then the short sale constraint on asset 2 would have no impact and, as a result, pricing would be linear in the sense that both assets would offer the same price-dividend ratio. By contrast, our analysis shows that costly short sales drive a wedge between the valuation of the two assets and thereby result in nonlinear pricing. In particular, the value of asset 1 in our model represents a fraction of the market portfolio that is strictly greater than its share of dividends and which varies endogenously across times and states to reflect the impact of the short selling frictions at play in the model. This contribution formalizes, within a general equilibrium model, the intuition in Cochrane (2002) and Cherkes, Jones and Spatt (2013) according to which the valuation of a shortable asset includes not only the present value of its future dividends but also the present value of future lending revenues. Perhaps of greater interest, this nonlinearity provides a rational explanation for the mispricing observed in certain famous equity carve-outs (Lamont and Thaler 2003), such as the partial spin-off of Palm by 3Com that we use to quantitatively illustrate the implications of the model.

Our theoretical framework provides a backdrop to the recent empirical findings of Beneish, Lee and Nichols (2015) and DD who document that stocks with high lending fees exhibit low average excess returns that cannot be explained by standard factor models such as the three- and four-factor models of Fama and French (1993) and Carhart (1997). In particular, DD argue that these negative excess returns are compensation for the systematic risk borne by the small fraction of investors who account for most of the shorting activity. They refer to this finding as the shorting premium and construct a

\(^5\)For simplicity we assume throughout that the lending agents are benevolent and have full bargaining power over investors. However, the model is easily extended to the case where the lending agents (and possibly also the custodian bank) retain a fraction of lending revenues and/or where investors have some bargaining power in their interactions with the lending agents. In either extension the structure of the equilibrium and the qualitative properties of the model remain unchanged.
portfolio risk factor labeled CME (for cheap-minus-expensive to short) that earns the corresponding abnormal return. Our model offers an alternative explanation for these findings that is aligned with the literature (e.g., Fama and French 2010) that questions the possibility to generate abnormal returns by stock picking, and argues that the estimated alphas may result from return-augmenting activities like securities lending. Specifically, our framework implies that stock returns satisfy a modified CAPM that includes an explicit downward adjustment for lending fees and, therefore, predicts that lending fees should not have any significant impact on the cross-section of returns provided that an appropriate correction is applied before running the estimation.

To empirically test this prediction, we use a sample similar to that of DD in which monthly stock returns from CRSP and accounting information from Compustat are matched with shorting cost and utilization measures provided by Markit Securities Finance. We sort firms into decile portfolios according to their average shorting costs and estimate on these portfolios the standard three- and four-factor models of Fama and French (1993) and Carhart (1997). As in DD we find that the unadjusted stock returns in our sample exhibit a significant shorting premium. Indeed, the CME portfolio obtained by going long in stocks of decile 1 and short in stocks of decile 10 produces strongly significant abnormal returns of 64bps/month relative to the three-factor model and 56bps/month relative to the four-factor model.

We then use the shorting costs and utilization measure data provided by Markit Securities Finance to adjust returns for the lending revenues and shorting costs as indicated by our model and repeat the estimation of the two factor models on these adjusted returns. The results are fully aligned with the predictions of our theoretical model. In particular, we find that, after adjustment, the estimated alpha for decile 10 relative to the four-factor model jumps up from a statistically significant value of −37bps/month to a non-significant value of −15bps/month. Furthermore, this negative abnormal return almost exactly offsets the positive abnormal return on the first decile and, as a result, the CME portfolio no longer exhibits any abnormal return relative to the adjusted four-factor model. Similar results hold relative to the three-factor model and these results continue to hold relative to either model even if we only use the 50% of decile 10 stocks that are most expensive to short in our construction of the CME portfolio.
Related literature

To capture the fact that locating shares to borrow may be a time consuming process, Duffie et al. (2002) develop a model with a single stock in which search costs and bargaining over fees generate a deterministic price process that includes the present value of the lending fees that accrue to holders of long positions. Vayanos and Weill (2008) extend the search model of Duffie et al. (2002) to include two assets with different lending fees and show that the resulting equilibrium can help understand phenomena such as the different pricing of on-the-run and off-the-run Treasury bonds. By contrast, we study an otherwise frictionless stochastic general equilibrium model where positive loan fees arise due to the presence of an intermediated securities lending market.

Our paper is naturally related to the large body of theoretical literature that studies the impact of shorting constraints in trading models where agents have heterogenous beliefs. Earlier contributions in this literature, including the seminal papers of Miller (1977) and Harrison and Kreps (1978), feature discrete-time partial equilibrium models with risk-neutral investors in which the combination of heterogenous beliefs with the impossibility of short selling gives rise to speculative episodes where the asset price exceeds its fundamental value to the most optimistic investor. Scheinkman and Xiong (2003) extend the original setting of Harrison and Kreps (1978) to continuous-time and use it to study the occurrence and properties of bubbles. Detemple and Murthy (1997) and, later, Gallmeyer and Hollifield (2008) study a similar problem but in a dynamic general equilibrium setting with risk-averse investors and show that the imposition of a short sale ban may result in either a price increase or a price decrease relative to a frictionless model. More recently, Nutz and Scheinkman (2020) study a continuous-time version of the model of Harrison and Kreps (1978) in which agents can short the asset subject to exogenous quadratic costs but these costs are dissipated and thus do not accrue to holders of long positions. Our paper advances this literature by proposing a tractable way to model the endogenous determination of securities lending fees in a dynamic general equilibrium setting where lending fees are rebated to holders of long positions.

Our focus and contributions markedly differ from Atmaz et al. (2021) who develop a CARA-Normal model with heterogeneous beliefs, two independent risky assets and a
riskless asset with exogenous return. In their model, stock prices and shorting costs are linear functions of two Gaussian processes that represent dividends and the stochastic disagreement among agents. Therefore, prices and shorting costs can be negative and, perhaps of greater concern, the model does not rule out situations where some agents that are a priori assumed to be long only end up holding short positions without paying the corresponding cost. By contrast, prices and shorting fees are nonnegative in our framework and, rather than independent assets, we consider Siamese twin stocks which allows us to elicit the premium associated with the possibility of shorting an asset and facilitate a clear identification in our empirical exercise.

Perhaps closest to us is the contemporaneous paper by Gărleanu et al. (2021) who also propose a continuous-time general equilibrium model that features logarithmic agents with heterogenous beliefs and costly-to-short stocks. The main difference with our paper resides in the modelling of the shorting friction. Specifically, Gărleanu et al. (2021) take the shorting cost as an \textit{exogenously} given function of short interest that they interpret as a cost of matching between competitive brokers and dealers. They show that this modelling produces multiple equilibria and use this feature to analyze situations, such as the recent GameStop episode, where fears among short sellers lead to self-fulfilling run-type behavior. By contrast, we construct a securities lending market operated by a single intermediary, in light of the existing market conditions in the U.S., and show that our modelling produces a unique equilibrium that is not nested among the multiple equilibria of Gărleanu et al. (2021) because in our model the \textit{endogenous} shorting cost cannot be expressed as a deterministic function of the \textit{endogenous} short interest. More recently, Chen, Kaniel and Opp (2022) introduce asymmetric information in a partial equilibrium version of our model to evaluate the implications of non-competitive lending fees. They quantify price \textit{wedges} due to the incremental impact that lenders assign to stocks due to the fee income, similar to the effect we capture \textit{endogenously} by using assets 1 and 2 in our model. Their empirical results show a significant impact of fees on stock valuations, particularly for small and micro-cap stocks, similar to our empirical exercise where we find that shorting costs and lending yields are negatively related to firm size.

Our work is also related to the broad literature on rational models of limits to arbitrage. See Gromb and Vayanos (2010) for a survey. We highlight Basak and Croitoru
who study a dynamic general equilibrium model with a risky asset and a derivative in zero net supply to show that mispricing can arise between two securities that carry the same risk, if agents are subject to portfolio constraints that prevent them from exploiting the induced arbitrage opportunity. Banerjee and Graveline (2014) obtain similar conclusions in a static CARA-Normal model with quasi-redundant assets and costly short sales. By contrast, we study the implications of costly shorting in a dynamic setting where all risky assets are in positive supply so that both expected returns and volatilities are endogenously determined.

Other contributions to the study of lending fees include Duffie (1996), Krishnamurthy (2002), and Blocher, Reed and Van Wesep (2013). More recently, Nezafat and Schroder (2022) study the role of private information in the equity lending market in a static rational expectations model with endogenous loan fees. There is also a growing literature on strategic short selling that studies the role of short selling in the transmission of information about firm fundamentals. For example, Goldstein and Guembel (2008) show that this channel can lead to negative spillovers, Goldstein, Ozdenoren and Yuan (2013) show that it may distort investment decisions, and Brunnermeier and Oehmke (2014) show that it may lead to situations where short sellers can force a vulnerable institution to liquidate assets at fire-sale prices. In the same vein, Brunnermeier and Pedersen (2005) and Carlin, Lobo and Viswanathan (2007) develop predatory trading models where short sellers exploit undercapitalized arbitrageurs.

In addition to DD and Beneish et al. (2015), our paper directly relates to the empirical literature that investigates the effect of short selling frictions. See Reed (2013) for a survey of this extensive body of work. Consistent with our arguments, Blocher and Whaley (2015) show that ETF managers tend to tilt their portfolios toward stocks with higher lending fees as a way of enhancing their returns, Prado (2015) suggests that institutional investors buy shares in response to increases in lending fees, and Johnson and Weitzner (2019) report that fund managers in their sample overweight assets with high lending fees. Our paper is also consistent with D’Avolio (2002) who documents that shorting costs are positively related to the dispersion of beliefs and with the findings of Nagel (2005), Blocher et al. (2013), and Prado, Saffi and Sturgess (2016) regarding the effects of supply and demand shocks in the securities lending market.
The remainder of the paper is organized as follows. The model is presented in Section 2. Section 3 provides a detailed account of the equilibrium construction. Section 4 discusses the endogenous determination of shorting costs and their properties in the one- and two-risky assets cases. Section 5 presents our empirical exercise and Section 6 concludes. The proofs of all results are provided in Appendix A and an extension of our benchmark model to the case of stochastic disagreement is found in Appendix B.

2 The model

2.1 Fundamental uncertainty

We consider a continuous-time economy on an infinite horizon. There is a single non-storable good available for consumption at every date $t \geq 0$ and we assume that its supply $e_t$ evolves according to

$$\frac{de_t}{e_t} = \mu dt + \sigma dZ_t,$$

for some exogenously given constants $\mu$ and $\sigma > 0$, where the process $(Z_t)_{t \geq 0}$ is a Brownian motion under some reference probability.

2.2 Agents

The economy is populated by two agents that we index by $a \in \{o, p\}$. Agents observe aggregate consumption as well as the prices of traded assets, but do not observe the increments of the Brownian motion and disagree about their perception of the dynamics of the aggregate consumption process. Specifically, we assume that from the point of view of agent $a$, this process evolves according to

$$\frac{de_t}{e_t} = \mu^{(a)} dt + \sigma dZ^{(a)}_t$$

(2)
for some constant $\mu^{(a)}$, where the process $(Z_t^{(a)})_{t \geq 0}$ is a Brownian motion under the subjective probability of agent $a$. We denote by

$$\Delta = \frac{\delta}{\sigma} \equiv \frac{\mu^{(o)} - \mu^{(p)}}{\sigma}$$

(3)

the disagreement per unit of volatility and assume that $\Delta \geq 0$ so that agent $o$ can be interpreted as being an optimist and agent $p$ as being a pessimist. The assumption of a constant disagreement simplifies the solution of the model by fixing the identity of the optimist and pessimist, and thereby limiting the number of state variables. Moving to a richer model with a stochastic disagreement process leads to a more cumbersome, yet fully explicit, characterization of the equilibrium but does not affect the underlying economics, see Appendix B for such an extension.

Finally, we assume that conditional on their beliefs, the two agents have homogenous logarithmic preferences given by

$$E_t^{(a)} \left[ \int_0^\infty e^{-\rho t} \log c_t dt \right]$$

(4)

for some constant discount rate $\rho > 0$, where $E_t^{(a)}[\cdot]$ denotes an expectation under the agent’s subjective probability measure conditional on the observation of the paths of dividends and market prices, up to date $t \geq 0$.

As is well-known, this specification implies that agents have marginal propensity to consume equal to $\rho$ and choose their portfolio to optimize an instantaneous quadratic criterion (see (18) below). In the context of our model, this myopic behavior also implies that–up to a slight reinterpretation and the addition of a linear term in the drift of the endogenous state variable–all the asset pricing results we derive below remain unchanged if instead of two agents with constant beliefs we consider a steady state population of agents subject to idiosyncratic shocks that shift their perceived growth rate back and forth between a low and a high value.
2.3 Traded assets

The financial market consists of three long-lived assets: A locally riskless asset in zero net supply and two risky securities in positive supply of one unit each. The price of the riskless asset evolves according to

$$dS_{0t} = r_t S_{0t} dt$$  \hspace{1cm} (5)$$

for some interest rate process $r_t$ that is to be determined in equilibrium. On the other hand, we assume that risky asset $i \in \{1, 2\}$ is a claim to a fraction $\eta_i \geq 0$ of aggregate consumption, and that its price evolves according to

$$dS_{it} + \eta_i e_t dt = r_t S_{it} dt + S_{it} \sigma_{it} \left(dZ_{t}^{(a)} + \theta_{it}^{(a)} dt\right),$$ \hspace{1cm} (6)$$

where the volatility coefficient $\sigma_{it}$ and the perceived market prices of risk

$$\theta_{it}^{(a)} = \Delta + \theta_{it}^{(p)}$$ \hspace{1cm} (7)$$

are to be determined endogenously in equilibrium. To ensure that the market portfolio $M_t \equiv S_{1t} + S_{2t}$ is a claim to the whole aggregate consumption, we naturally assume that the fractions paid by the risky assets are such that $\eta_1 + \eta_2 = 1$.

The risky assets in our model have proportional dividends and thus are Siamese twins (see e.g., Froot and Dabora (1999) and De Jong et al. (2009)). This assumption serves two purposes. First, it allows us to easily account for the fact that the shares of a given stock are often not all available for lending because of a variety of reasons that include the existence of different share classes (Mei et al. 2009), dual or cross-listings as in the case of ADRs (Blau et al. 2012), or the occurrence of an equity carve-out (Lamont and Thaler 2003). Second, and perhaps more importantly, this assumption allows us to study the effect of costly short sales on ex-ante identical assets within a single general equilibrium model where all markets clear rather than across different models.
2.4 Shorting frictions

Our point of departure from existing equilibrium models with heterogenous beliefs is that the risky assets are subject to different trading arrangements. Specifically, we assume that shares of asset 2 cannot be shorted whereas shares of asset 1 can be shorted by incurring a flow cost per dollar of short as long as the position is maintained.

To sell short, one must first borrow the required shares. In line with the evidence reported by D’Avolio (2002), Baklanova et al. (2015), Gensler (2021), and Chen et al. (2022) among others, we assume that the lending market is intermediated by a finite number $n$ of ex-ante identical lending agents and that securities are held for investors by a custodian bank. At time $t \geq 0$ an investor who wishes to short over the next instant is randomly matched to one of the lending agents. Each lending agent $i$ sets a shorting fee $\Phi_i$ to maximize her flow of shorting revenues taking as given the aggregate short demand schedule of the investors who are matched with her at time $t \geq 0$. Once the terms of the short transactions are set, each lending agent borrows the required shares from the custodian bank, lends them over an infinitesimal time interval to the investors matched with her, and transfers the induced shorting fees to the custodian bank who in turn redistributes them on a value weighted basis to holders of long positions. See Figure 1 for an illustration of this mechanism in a model with a single intermediary.

We assume that a given investor can only be matched with a single lending agent at each point in time. As a result, each lending agent enjoys some degree of market power over the group of investors who are matched with her at a given point in time. To simplify the presentation, we focus on the case where intermediaries are benevolent and enjoy full bargaining power over investors in the determination of the shorting fee. However, the model is easily extended to the case where intermediaries—and possibly also the custodian bank—retain a fraction of lending fees and do not enjoy full bargaining over investors. See Remark 1 below for a discussion of the latter extension.

The fact that shorting fees eventually accrue to investors who are long induces a form of interaction between investors. For this interaction to remain competitive, investors have to take as given not only the costs incurred when taking short positions but also the fees that they may receive when they hold shares of asset 1. We model this feature
Figure 1: The shorting mechanism. In this example with a single lending agent, the short interest in asset 1 amounts to \( \nu = \sum_{i} \nu_i \geq 0 \) shares while the aggregate long position sums up to \( \sum_{k} n_k = 1 + \nu \) shares. As a result, the equilibrium lending yield is \( \Gamma_t = \frac{\nu}{1 + \Phi_t} \).

by assuming that agents take as given the flow cost of shorting as well as the flow rate of lending fees \( \Gamma_t \) that each dollar invested in asset 1 generates, and determine this rate endogenously in equilibrium by requiring that the flow of lending fees received by long agents equals the flow of costs paid by short agents. To make a clear distinction between the flows paid by short agents and those received or anticipated by long agents, we refer to \( \Phi_t \) as the shorting cost charged by lending agent \( i \) and to \( \Gamma_t \) as the lending yield.

In our model, investors of type \( o \) are more optimistic than investors of type \( p \) and both have logarithmic utility. Therefore, we know that investors of type \( p \) are short whenever investors of type \( o \) are and, since all investors cannot be short simultaneously, it follows that shorting activity can only come from the pessimists in equilibrium. Furthermore, the assumption of logarithmic utility implies that the short demand schedule of any investor is proportional to her wealth. Combining these observations shows that, up to a multiplicative factor, all intermediaries face the same optimization problem in equilibrium and it follows that they will all select the same shorting fee \( \Phi_t \equiv \Phi_t \) at every point in
time. In particular, we may assume without loss of generality that the lending market is intermediated by a single agent and thus set \( n \equiv 1 \) from now on.

### 2.5 Definition of equilibrium

Combining the above shorting mechanism with the usual self-financing condition shows that the wealth of agent \( a \) evolves according to

\[
dW_t^{(a)} = \left( r_t W_t^{(a)} - c_t + \Lambda(\pi_t; \Phi_t, \Gamma_t) \right) dt + \sum_i \pi_{it} \sigma_i \left( dZ_t^{(a)} + \theta_{it}^{(a)} dt \right),
\]

where \( c_t \geq 0 \) represents her consumption rate, \( \pi_t \in \mathbb{R} \times \mathbb{R}_+ \) denote the amounts she invests in the risky assets, and the nonlinear term

\[
\Lambda(m; \Phi_t, \Gamma_t) \equiv m^+ \Gamma_t - m^- \Phi_t
\]

captures the flow rate of lending fees that she pays or receives.

As is standard, we require agents to maintain strictly positive wealth at all times. Therefore, the optimization problem solved by agent \( a \) is

\[
\sup_{(c, \pi)} E^{(a)} \left[ \int_0^\infty e^{-\rho t} \log c_t dt \right] \text{ subject to } (8), (9), \text{ and } W_t^{(a)} > 0.
\]

Whenever they exist, we denote by \((c_t^{(a)}(\Phi, \Gamma), \pi_t^{(a)}(\Phi, \Gamma))\) the optimal consumption and optimal portfolio of agent \( a \), taking as given the traded asset prices and the pair of flow rates \( (\Phi, \Gamma) \) that characterize the short market.

**Definition 1.** An equilibrium is a price process \((S_{0t}, S_{1t}, S_{2t})\), a shorting cost \( \Phi_t \), and a lending yield \( \Gamma_t \) such that

\[
\Phi_t \in \arg\max_x \left\{ \sum_a \pi_{1t}^{(a)}(x, \Gamma)^- x \right\},
\]

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and all markets clear:

\[ \sum_{a} c_t^{(a)}(\Phi, \Gamma) = e_t, \quad \text{(Consumption),} \quad (12) \]

\[ \sum_{a} \pi_t^{(a)}(\Phi, \Gamma) = S_{it}, \quad \text{(Risky asset } i \in \{1, 2\}), \quad (13) \]

\[ \sum_{a} \Lambda \left( \pi_t^{(a)}(\Phi, \Gamma); \Phi_t, \Gamma_t \right) = 0, \quad \text{(Lending market),} \quad (14) \]

where the mapping \( \Lambda \) is defined in (9).

The above definition is similar to the classical definition of an equilibrium by Radner (1972) but includes two additional conditions to accommodate the presence of costly short sales in a dynamic general equilibrium setting.

The first condition (11) is an optimality condition that results from our modelling of the security lending market and which requires that the shorting cost maximizes shorting revenues taking as given market prices, the lending yield anticipated by agents, and their short demand schedules \( (a, x) \mapsto \pi_t^{(a)}(x, \Gamma)^- \). The second condition (14) requires that the flow of fees received by holders of long positions must equal the flow of lending fees paid by short agents. We treat this condition as a market clearing condition as it matches the flows exchanged between agents, but one could equally view this requirement as a rational expectations condition which ensures that the lending fees anticipated by agents are indeed realized along the equilibrium path.

### 3 Equilibrium

In this section, we sequentially build an equilibrium for our economy with shorting costs. To facilitate the analysis, we focus throughout on the characterization of an equilibrium in which the asset volatilities are strictly positive at all times.

#### 3.1 Individual optimality

The absence of arbitrage requires the net Sharpe ratios offered by the different assets to be such that it is not possible to generate a locally riskless return by combining assets.
In our model, this constraint can be intuitively expressed as

$$\max \left\{ \theta_{2t}^{(a)}, \theta_{1t}^{(a)} + \gamma_t \right\} \leq \theta_{1t}^{(a)} + \phi_t,$$

(15)

where $\gamma_t \equiv \Gamma_t / \sigma_{1t}$ and $\phi_t \equiv \Phi_t / \sigma_{1t}$ denote the lending yield and the shorting cost per unit of volatility. The interpretation of this inequality is clear: The left hand side is the largest expected excess return that can be generated from the point of view of agent $a$ by going long in either of the risky assets, while the right hand side is the expected excess return generated by going short in asset 1. In addition to this no-arbitrage requirement, in equilibrium we must have that

$$\theta_{2t}^{(a)} = \theta_{1t}^{(a)} + \gamma_t,$$

(16)

for otherwise one of the assets would dominate the other and markets would not clear. This equality shows that, in our model, an agent wanting to go long is indifferent between the risky assets once fees are taken into account and will lead to some indeterminacy in the characterization of optimal portfolios: see Proposition 1 below. Importantly, given (16) the no-arbitrage condition (15) boils down to $\gamma_t \leq \phi_t$ which simply requires that borrowing asset 1 to hold it does not generate riskless profits.

The assumption of logarithmic utility implies that, under the above conditions, the optimal consumption rate of agent $a$ is given by

$$c_t^{(a)} = \rho W_t^{(a)}$$

(17)

and that the fractions of her wealth $p_{it}^{(a)} = \pi_{it}^{(a)} / W_t^{(a)}$ that she optimally invests in the risky assets solve the mean-variance problem

$$\max_{p \in \mathbb{R} \times \mathbb{R}^+} \left\{ \Lambda (p_1; \Phi_t, \Gamma_t) + \sum_i p_i \sigma_{it} \theta_{it}^{(a)} - \frac{1}{2} (p_1 \sigma_{1t} + p_2 \sigma_{2t})^2 \right\}.$$ (18)

The following proposition derives the solution to this problem and summarizes the optimal trading behavior of agents, taking as given market prices and the rates ($\Phi, \Gamma$) that characterize the short market.
Proposition 1. Assume that condition (16) holds and let $\phi_t \geq \gamma_t$. Then the optimal portfolio of agent $a$ satisfies

$$
\begin{align*}
    p_{1t}^{(a)}(\Phi, \Gamma)\sigma_{1t} &= \mathbb{1}_{\{\theta_{1t}^{(a)} \geq 0\}} x_t + \mathbb{1}_{\{\theta_{1t}^{(a)} + \phi_t \leq 0\}} \left( \theta_{1t}^{(a)} + \phi_t \right), \\
    p_{2t}^{(a)}(\Phi, \Gamma)\sigma_{2t} &= \mathbb{1}_{\{\theta_{2t}^{(a)} \geq 0\}} \left( \theta_{2t}^{(a)} - x_t \right),
\end{align*}
$$

where $x_t$ is an arbitrary process such that $0 \leq x_t \leq \theta_{2t}^{(a)}$.

The optimal trading strategy in Proposition 1 admits an intuitive interpretation. If the net Sharpe ratio $\theta_{2t}^{(a)} = \gamma_t + \theta_{1t}^{(a)}$ that agent $a$ associates with long positions in the risky assets is positive, then agent $a$ naturally goes long at the optimum but there is a degree of freedom in the determination of her optimal portfolio because any $p \in \mathbb{R}^2_+$ that delivers the efficient risk exposure

$$
\sum_i p_i \sigma_{it} = \arg\max_{x \in \mathbb{R}} \left\{ x \theta_{2t}^{(a)} - \frac{1}{2} x^2 \right\} = \theta_{2t}^{(a)}
$$

is optimal. On the other hand, if $\theta_{2t}^{(a)} \leq 0$ then the agent clearly does not want to go long in either risky asset. Whether she goes short in asset 1 depends on the sign of the Sharpe ratio $-(\theta_{1t}^{(a)} + \phi_t)$ that she associates with a short position in asset 1. If this quantity is positive, then she shorts asset 1 to achieve the efficient risk exposure

$$
\sum_a p_a \sigma_{1t} = \arg\max_{x \in \mathbb{R}} \left\{ x \left( \theta_{1t}^{(a)} + \phi_t \right) - \frac{1}{2} x^2 \right\} = \theta_{1t}^{(a)} + \phi_t,
$$

and otherwise she invests only in the riskless asset.

3.2 Equilibrium shorting cost

Proposition 1 and the discussion preceding it show that the total flow of lending fees induced by a shorting cost process $\Phi_t$ is well-defined only if $\phi_t = \Phi_t / \sigma_{1t} \geq \gamma_t$ in which case it is explicitly given by

$$
\sum_a \pi_{1t}^{(a)}(\Phi, \Gamma)^{-} \Phi_t = \sum_a \phi_t \left( \theta_{1t}^{(a)} + \phi_t \right)^- W_{t}^{(a)}. \tag{22}
$$
In our model, agent o is more optimistic than agent p and both have logarithmic utility. Therefore, we know that agent p is short whenever agent o is short and, because agents cannot be short simultaneously in equilibrium, it follows that any shorting activity must come from the pessimist alone. In particular, we must have

\[ \theta_{2t}^{(o)} = \theta_{1t}^{(o)} + \gamma_t > 0, \]  

so that the optimist is long at all times. In combination with (16), this implies that the Sharpe ratios perceived by the optimist are such that

\[ \forall y \geq \gamma_t : \left( \theta_{1t}^{(o)} + y \right)^- = \left( \theta_{2t}^{(o)} - \gamma_t + y \right)^- = 0. \]  

As a result, the sum in (22) reduces to the contribution of the pessimist and it follows that the equilibrium shorting cost satisfies

\[ \phi_t \in \arg\max_{y \geq \gamma_t} \left\{ y \left( \theta_{1t}^{(p)} + y \right)^- \right\} = \max \left\{ \gamma_t, -\frac{1}{2} \theta_{1t}^{(p)} \right\} + 1 \{ \gamma_{1t} \geq 0 \} \mathbb{R}^+. \]  

The above expression shows that when \( \theta_{1t}^{(p)} \geq 0 \), the shorting cost is undetermined because in such states neither agent wants to go short irrespective of the cost set by the lending agent. To facilitate the presentation, we from now select the smallest element of the above set of maximizers, i.e., we let

\[ \phi_t = \max \left\{ \gamma_t, -\frac{1}{2} \theta_{1t}^{(p)} \right\}. \]  

This selection is without loss of generality and simply amounts to setting the flow cost to zero on the endogenous set of states \( L \equiv \{(\omega, t) : \gamma_t = 0\} \), where the shorting market is inactive in equilibrium.

Substituting the optimal portfolios of Proposition 1 into the market clearing conditions shows that the equilibrium lending yield and shorting cost are related by

\[ \gamma_t \sigma_{1t} S_{1t} = (\phi_t - \gamma_t) \left( \theta_{1t}^{(p)} + \phi_t \right)^- W_t^{(p)}. \]  

17
Since the diffusion of asset 1 and the wealth of the pessimist are both strictly positive in equilibrium, this identity combined with (16) and (26) implies that (see the Appendix A for a detailed argument)

\[ \mathcal{L} = \left\{ (\omega, t) : \theta_{2t}^{(p)} \geq 0 \right\}, \tag{28} \]

and substituting back into (26) shows that

\[ \phi_t = -\frac{1}{2} \theta_{1t}^{(p)} \mathbb{1}_{\{S\}}, \tag{29} \]

where

\[ S \equiv (\Omega \times \mathbb{R}_+) \setminus \mathcal{L} = \left\{ (\omega, t) : \theta_{2t}^{(p)} < 0 \right\} \tag{30} \]

denotes the set of states where the shorting market is active. These expressions provide key information about the equilibrium properties of the shorting market. Specifically, the equivalent set identities (28) and (30) show that the market is active in equilibrium exactly in states where \( \theta_{2t}^{(p)} = \theta_{1t}^{(p)} + \gamma_t < 0 \) so that the pessimist perceives a strictly negative net Sharpe ratio on long positions in either risky asset. Equation (29) shows that, in those states, the lending agent sets the flow cost \( \phi_t \) to one half of the Sharpe ratio \( -\theta_{1t}^{(p)} \) that the pessimist could have obtained absent frictions. In response, the pessimist scales down her short demand by a half relative to the frictionless case and, as a result, the flow of lending fees received by the optimist, i.e.

\[ \left( \pi_{1t}^{(p)} \right)^{-} \Phi_t = \frac{1}{4} \left( \theta_{1t}^{(p)} \right)^2 W_t^{(p)} \mathbb{1}_{\{S\}}, \tag{31} \]

amounts to a fourth of her optimal frictionless excess return.

**Remark 1** (Nash bargaining). Since the lending agent upholds only the interests of asset owners, this fraction constitute an upper bound on the share that alternative price setting mechanisms may attribute to the optimist. In particular, if the cost was determined by Nash bargaining between the lending agent and the pessimist(s) then (29) would be replaced by \( -\frac{b}{2} \theta_{1t}^{(p)} \mathbb{1}_{\{S\}} \), where \( b \in [0, 1] \) represents the bargaining power of the lending
agent. In response, the pessimist would now scale down her optimal short demand by a factor $1 - \frac{b}{2}$ and, as a result, the optimist would capture a fraction $\frac{b}{2}(1 - \frac{b}{2})$ of her optimal frictionless excess return. We focus throughout the paper on the polar case where $b = 1$ because it leads to simpler algebraic expressions for equilibrium outcomes, but the structure of the equilibrium and the qualitative properties of the model, including the characterization of the trading regions, remain the same when $b < 1$.

3.3 State variables and trading regions

Combining the equilibrium restriction (16) with the expression of the equilibrium shorting cost in (25) shows that the set

$$\left\{(\omega, t) : \theta_{2t}^{(p)} < 0 \text{ and } \theta_{1t}^{(p)} + \phi_t \geq 0 \right\} \neq \emptyset. \quad (32)$$

Therefore, it follows from Proposition 1 that the pessimist is strictly long in either or both risky assets in the interior of the set $\mathcal{L}$ where $\theta_{2t}^{(p)} > 0$, is fully invested in the riskless asset on the boundary where

$$\partial \mathcal{L} \equiv \left\{(\omega, t) : \theta_{2t}^{(p)} = 0 \right\}, \quad (33)$$

and is strictly short in asset 1 on the set $\mathcal{S}$ where $\theta_{2t}^{(p)} < 0$. As we now show, the scale invariance of logarithmic preferences allows us to characterize these sets and the resulting pricing of risk/time in terms of a single endogenous state variable

$$s_t \equiv \frac{c_t^{(o)}}{e_t} \in [0, 1] \quad (34)$$

that tracks the consumption share of the optimist. To construct the equilibrium evolution of this state variable, we take as reference the subjective probability of the optimist. This choice is without loss in generality.
Since the marginal propensity to consume of both agents is $\rho$, it follows from market clearing that the price of the market portfolio is

$$M_t = \sum_i S_{it} = \sum_a W^{(a)}_t = \frac{1}{\rho} \sum_a c^{(a)}_t = \frac{e_t}{\rho},$$  \hspace{1cm} (35)$$

and that the endogenous state variable can be expressed as

$$s_t = \frac{W^{(o)}_t}{M_t} = 1 - \frac{W^{(p)}_t}{M_t}. \hspace{1cm} (36)$$

On the other hand, combining (16), (23), and (27) with the results of Proposition 1 shows that the wealth of the agents evolve according to

$$\frac{dW^{(o)}_t}{W^{(o)}_t} = (r_t - \rho) \, dt + \theta^{(o)}_{2t} \left( dZ^{(o)}_t + \theta^{(o)}_{2t} \, dt \right),$$  \hspace{1cm} (37)$$

and

$$\frac{dW^{(p)}_t}{W^{(p)}_t} = (r_t - \rho) \, dt + \mathbb{1}_{\{L\}} \left( \theta^{(o)}_{2t} - \Delta \right) \left( dZ^{(o)}_t + \theta^{(o)}_{2t} \, dt \right)$$

$$+ \mathbb{1}_{\{S\}} \left( \theta^{(o)}_{2t} - \Delta - \gamma_t + \phi_t \right) \left( dZ^{(o)}_t + \left( \theta^{(o)}_{2t} - \gamma_t + \phi_t \right) \, dt \right), \hspace{1cm} (38)$$

subject to (29) and

$$\gamma_t \sigma_t S_{1t} = \mathbb{1}_{\{S\}} (1 - s_t) \left( \theta^{(o)}_{2t} - \Delta - \gamma_t + \phi_t \right) M_t. \hspace{1cm} (40)$$

Next, applying Itô’s lemma to the second equality in (36) and matching terms, pins down the equilibrium interest rate $r_t$ and the equilibrium market price of risk $\theta^{(o)}_{2t}$ as functions of the state variable and the lending yield:

$$\theta^{(o)}_{2t} = \theta^*(s_t) - \mathbb{1}_{\{S\}} \frac{(1 - s_t) \left( s_t \Delta - \sigma - \gamma_t \right)}{1 + s_t}.$$  \hspace{1cm} (41a)
and

\[ r_t = r^*(s_t) + \mathbb{1}_{\{s\}} \frac{s_t(1-s_t) \left( \sigma + \gamma_t + \Delta \right) (s_t \Delta - \sigma - \gamma_t)}{(1 + s_t)^2}, \]  

(41b)

where

\[ \theta^*(s_t) \equiv \sigma + (1 - s_t) \Delta, \]  

(42)

\[ r^*(s_t) \equiv \rho - \sigma^2 + \mu(o) s_t + \mu(p)(1 - s_t), \]  

(43)

denote the market price of risk and the interest rate that would prevail in an otherwise identical economy where either or both of the risky assets can be freely shorted (see e.g., Detemple and Murthy (1997)).

On the shorting region, we have from (29) that \( \phi_t = -\frac{1}{2} \theta^{(p)}_t > \gamma_t \) and combining this with (16) and (41a) shows that over that region

\[ \sigma + \gamma_t < s_t \Delta. \]  

(44)

Therefore, it follows from (41) that the presence of costly short selling increases the interest rate and decreases the market price of risk relative to the frictionless case. To understand this result, observe that costly short sales trigger a reduction in the short demand of the pessimist which in turn implies that the optimist’s equilibrium portfolio does not have to be as leveraged as in the frictionless case and, thus, explains the changes in the interest rate and market price of risk.

Substituting (30) into (41a) shows that the net Sharpe ratio perceived by the pessimist on long risky asset positions satisfies

\[ \theta^{(p)}_{2t} = \mathbb{1}_{\{\theta^{(p)}_{2t} \geq 0\}} \left( \sigma - s_t \Delta \right) + \mathbb{1}_{\{\theta^{(p)}_{2t} < 0\}} \frac{\gamma_t (1 - s_t) + 2 (\sigma - s_t \Delta)}{1 + s_t}. \]  

(45)

This implies that \( \mathcal{L} \) of (28) is contained in the set of states where the consumption share of the optimist lies below the threshold

\[ s^* = s^*(\Delta) \equiv \min \left\{ 1, \frac{\sigma}{\Delta} \right\} \]  

(46)
Figure 2: Equilibrium trading regions. The figure illustrates the shape of the equilibrium trading regions and allows us to determine the equilibrium configuration that occurs for each level of disagreement among agents.

and, because the second term on the right hand side of (45) is nonnegative at $s_t = s^*$, we conclude that the trading regions are explicitly given by

$$
\mathcal{L} = \left\{ (\omega, t) : \theta_{2t}^{(p)} \geq 0 \right\} = \left\{ (\omega, t) : 0 < s_t \leq s^* \right\}, \quad (47a)
$$

$$
\mathcal{S} = (\Omega \times \mathbb{R}_+) \setminus \mathcal{L} = \left\{ (\omega, t) : \theta_{2t}^{(p)} < 0 \right\} = \left\{ (\omega, t) : s^* < s_t < 1 \right\}. \quad (47b)
$$

These expressions show that shorting occurs only in states where the optimists represent a large enough share of the economy. This is intuitive. Indeed, when optimists are very few, prices mostly reflect the opinion of the pessimists and shorting is not necessary. On the contrary, when a large fraction of agents are optimists, equilibrium prices reflect more closely the opinion of the optimists and shorting becomes necessary for the pessimists to express their perception of the risky assets as being overpriced.

As illustrated in Figure 2, the characterization of the trading regions in (47) shows that two mutually exclusive types of equilibria may arise in our model. The first type occurs if
the disagreement among agents is so small that $\Delta \leq \sigma$. In that case, both agents are long in the risky asset throughout the state space, and the existence of the shorting market is irrelevant so that the equilibrium is the same as in an otherwise identical frictionless economy with heterogeneous beliefs (see Detemple and Murthy (1997)). The second type of equilibrium occurs when the disagreement among agents is such that $\Delta > \sigma$. In that case, the equilibrium includes two non empty trading regions: up to the locus of points $s_t = \sigma / \Delta$, both agents are long and the shorting market is inactive, while strictly above that locus agent 2 holds a short position in asset 1 and the shorting costs she incurs generate a strictly positive flow of lending revenues for the optimist.

Applying Itô’s lemma to both sides of the first equality in (36) and matching terms finally shows that the state variable evolves according to

$$\frac{ds_t}{s_t(1 - s_t)} = m(s_t, \gamma_t) \, dt + v(s_t, \gamma_t) \, dZ_t^{(o)},$$

(48a)

with the functions defined by

$$v(s_t, \gamma_t) = v(s_t, \gamma_t; \Delta) \equiv \Delta - \mathbb{1}_{\{s_t > s^*(\Delta)\}} \frac{s_t \Delta - (\gamma_t + \sigma)}{1 + s_t},$$

(48b)

$$m(s_t, \gamma_t) = m(s_t, \gamma_t; \Delta) \equiv (1 - s_t) \, v(s_t, \gamma_t)^2 + \mathbb{1}_{\{s_t > s^*(\Delta)\}} \frac{(s_t \Delta - (\gamma_t + \sigma))(\gamma_t + \Delta) s_t - \sigma}{(1 + s_t)^2}.$$

(48c)

Importantly, the drift and the diffusion of the endogenous state variable are equal to zero at both $s_t = 0$ and $s_t = 1$. This implies that 0 and 1 are absorbing boundaries for the consumption share process of the optimist, and will allow us to easily derive boundary conditions for equilibrium prices in the next section.

### 3.4 Price representation

Having characterized the instantaneous pricing of risk and time, we now turn to the pricing of long lived assets. To this end, let

$$\xi_{t,u}^{(o)} = e^{\rho u - \rho u} \frac{C_t^{(o)}}{C_u^{(o)}} = e^{-\rho(u-t)} \frac{s_t \gamma_t}{s_u \gamma_u},$$

(49)
denote the normalized marginal utility of the reference agent.

**Proposition 2.** In equilibrium

\[
S_{1t} = E_t^{(o)} \left[ \int_t^\infty \xi_{t,u}^{(o)} (e_{1u} + S_{1u} \Gamma_u) \, du \right], \tag{50}
\]

\[
S_{2t} = E_t^{(o)} \left[ \int_t^\infty \xi_{t,u}^{(o)} e_{2u} \, du \right], \tag{51}
\]

where \( e_{it} = \eta_i e_i \) denotes the dividend rate of asset \( i = \{1, 2\} \).

The above proposition shows that there are no rational bubbles in our model. Indeed, the equilibrium prices of the two assets are given by the risk-adjusted present value of the cash flows that they deliver to holders of long positions. The novelty is that, in our model, the cash flows of risky asset 1 include an endogenous component \( S_{1t} \Gamma_t \) that accounts for the lending fees generated by each share of the asset along the equilibrium path.

This endogenous cash flow component is strictly positive over a set of positive measure if and only if the disagreement \( \Delta > \sigma \) so that the shorting region is non empty. In that case, the equilibrium price-dividend ratio of asset 2

\[
PD_{2t} \equiv \frac{S_{2t}}{e_{2t}} = E_t^{(o)} \left[ \int_t^\infty e^{-\rho(u-t)} \left( \frac{s_t}{s_u} \right) \, du \right] \tag{52}
\]

is strictly lower than that of asset 1

\[
PD_{1t} \equiv \frac{S_{1t}}{e_{1t}} = E_t^{(o)} \left[ \int_t^\infty e^{-\rho(u-t)} \frac{s_t}{s_u} \left( 1 + PD_{1u} \Gamma_u \right) \, du \right], \tag{53}
\]

and the premium

\[
PD_{1t} - PD_{2t} = E_t^{(o)} \left[ \int_t^\infty e^{-\rho(u-t)} \frac{s_t}{s_u} PD_{1u} \Gamma_u du \right] = \frac{1}{\eta_1} \left( \frac{1}{\rho} - PD_{2t} \right) > 0 \tag{54}
\]

gives the risk-adjusted present value of the stream of holding benefits that accrue to owners of asset 1 in the form of lending fees.

The above inequality implies that, in the presence of costly short sales, the equilibrium pricing rule is nonlinear as the risky assets have different price-dividend ratios despite the fact that they are Siamese twins. Specifically, since \( PD_{2t} < 1/\rho \) from (54), we have
that the share of asset 2 in the market portfolio

\[ \frac{S_{2t}}{M_t} = \rho \eta_2 PD_{2t} < \eta_2 \]  

is strictly lower than the share of aggregate dividends that it pays out, while the share of asset 1 in the market portfolio

\[ \frac{S_{1t}}{M_t} = \rho \eta_1 PD_{1t} = 1 - \rho \eta_2 PD_{2t} > \eta_1 \]  

exceeds its share of dividends. This nonlinearity is entirely driven by the presence of costly short sales and provides a rational explanation for the apparent mispricing observed in the period following certain corporate restructurings. For example, Lamont and Thaler (2003) report that after the spin-off by 3Com of 5% of its subsidiary Palm, the extrapolation of the value of the traded Palm shares resulted in an implied valuation that exceeded the market capitalization of the subsidiary 3Com.

As suggested by Cherkes et al. (2013), the key to understand this phenomenon is the observation that at the time of this apparent mispricing, the costs associated with shorting Palm were very high because only the 5% of freely traded Palm shares could be lent to investors wanting to establish a short position. In our model, the \( \eta_1 = 5\% \) of freely traded Palm shares are akin to asset 1 so that their price in (50) should include a sizable lending fee component in (54), while the remaining Palm shares held by 3Com are part of asset 2 whose equilibrium price in (51) only reflects the present value of future dividends. We quantitatively illustrate this feature of the model in Section 4.2.

**Remark 2.** The strict inequality (54) holds irrespective of whether the shorting market is currently active or not. Indeed, we show in Appendix A that the equilibrium evolution of \( s_t \) on the long region implies

\[ P^{(o)} \left[ \sup_{u \geq t} s_u \in S \right| 0 < s_t \leq s^* \right] = 1, \]

so that the optimist can be certain that, starting from any point in \( L \), her consumption share will eventually enter the open region \( S \) where the trading of the pessimist generates strictly positive lending fees.
4 Equilibrium prices and lending yield

To complete the construction of the equilibrium, it now remains to compute the lending yield and the risky asset prices. To facilitate the presentation, we start with the simpler case of a single risky asset where the solution is in closed-form before turning to the more challenging case of two risky assets. We then calibrate the model to briefly discuss the 3Com/Palm spin-off puzzle.

4.1 One risky asset

When the weight \( \eta_1 = 1 \), the single risky asset \( S_{1t} = M_t \) is the market portfolio and its volatility equals that of the aggregate dividend. Substituting these quantities into (29) and (40) and using (41) shows that in equilibrium

\[
\Phi_t = \mathbb{1}_{\{s_t>s^*\}} \frac{(\delta + \Gamma_t)s_t - \sigma^2}{1+s_t},
\]

(58)

\[
\Gamma_t = \mathbb{1}_{\{s_t>s^*\}} \frac{(1-s_t)(s_t\delta - \Gamma_t - \sigma^2)((\delta + \Gamma_t)s_t - \sigma^2)}{\sigma^2(1+s_t)^2},
\]

(59)

where the constant

\[
\delta \equiv \sigma \Delta = \mu^{(o)} - \mu^{(p)} \geq 0
\]

(60)

denotes the unscaled difference in beliefs between the two agents. Solving this system delivers the following result.

**Proposition 3.** With a single risky asset, the equilibrium shorting cost and the equilibrium lending yield are given by

\[
\Phi_t = \mathbb{1}_{\{s_t>s^*\}} \frac{s_t(1-s_t)\delta - 2\sigma^2 + \sqrt{s_t^2(1-s_t)^2\delta^2 + 4\sigma^4s_t}}{2(1-s_t)},
\]

(61)

and

\[
\Gamma_t = \mathbb{1}_{\{s_t>s^*\}} \frac{-s_t((1-s_t)^2\delta + 4\sigma^2) + (1+s_t)\sqrt{s_t^2(1-s_t)^2\delta^2 + 4\sigma^4s_t}}{2s_t(1-s_t)},
\]

(62)
and both are increasing and convex in $\delta$.

The positive relation between the shorting cost and the difference in beliefs is intuitive. Indeed, an increase in $\delta$ implies that agent $p$ becomes more pessimistic than agent $o$ in relative terms and thus triggers an upward shift in her short demand schedule which in turn leads to an increase of the shorting cost. To understand the comparative statics of the lending yield, note that due to market clearing we have

$$\Gamma_t = \frac{\Phi_t \Upsilon_t}{1 + \Upsilon_t},$$

where the utilization ratio $\Upsilon_t = \pi_{1t}^{(p)} / S_{1t}$ tracks the fraction of shortable shares that are on loan. This measure of short interest is affected by changes in $\delta$ both directly through the perceived risk premia and indirectly through the equilibrium shorting cost. However, combining Proposition 1 and (29) reveals that in equilibrium we have

$$\Upsilon_t = \left(1 - s_t \sigma\right) \left(\theta_{1t}^{(p)} + \phi_t\right)^- = \left(1 - s_t \sigma^2\right) \Phi_t,$$

which implies that the comparative statics of $\Upsilon_t$ and thus of $\Gamma_t$ are the same as those of the equilibrium shorting cost $\Phi_t$. In particular, since the shorting cost is increasing in $\delta$, this identity shows that, in equilibrium, there exists a positive relation between short interest and the divergence in beliefs. This implication of the model is consistent with extensive empirical evidence. In particular, it is well documented that there exists a positive relation between short interest and the dispersion of analysts’ earning forecasts taken as a proxy for heterogenous beliefs (see e.g., D’Avolio (2002), Duffie et al. (2002)), and Table 1 below confirms that the same relation holds in the sample that we use for our empirical application in Section 5.

To illustrate the magnitude of the shorting cost and its dependence on the wealth distribution, Figure 3 plots $\Phi_t$ and $\Gamma_t$ as function of the consumption share of the optimist in a model with $\sigma = 10\%$ and $\delta = 5\%$. The figure shows that the shorting cost starts from zero at the lower end of the shorting region, increases until it reaches a maximum
Figure 3: Equilibrium shorting cost and lending yield. The solid and dash-dotted lines represent the equilibrium shorting cost and lending yield as functions of the consumption share of the optimist in a single asset model with $\sigma = 0.1$ and $\delta = 0.05$. The dashed/dotted lines represent the impact of a 10% increase/decrease in the divergence of beliefs $\delta$.

and tappers off to a limit that is given by

$$
\lim_{s_t \to 1} \Phi_t = (1 - s^*) \frac{\delta}{2} = (\Delta - \sigma) \frac{\sigma}{2}
$$

as a result of (61). To understand this limit, recall that as $s_t \to 1$ the model converges to one where only the optimist is present. As a result, the market price of risk perceived by the pessimist must converge to its frictionless counterpart $\theta^*(1) - \Delta = \Delta(s^* - 1)$ and the expression for the limiting cost now follows from (29).

The bottom curves of the figure show that the lending yield is a bell-shaped function of $s_t$ that starts from zero at the lower end of the shorting region and comes back to zero as the wealth share of the optimist approaches one. The apparent discrepancy between the limiting behavior of the shorting cost and the lending yield as $s_t \to 1$ can be traced back to the economic nature of these objects. Indeed, $\Phi_t$ represents a price that can
be meaningfully understood in the limit as the cost for a short position of infinitesimal size (see e.g., Davis (1998), Hugonnier and Kramkov (2004), Hugonnier, Kramkov and Schachermayer (2005)) whereas \( \Gamma_t \) is a flow rate that can be strictly positive only in states where the pessimist holds a non infinitesimal fraction of aggregate wealth.

As usual with logarithmic preferences, the price \( S_{1t} = e_t / \rho \) of the single risky asset is unaffected by the presence of frictions. However, it is important to recall that, in our model, this price comprises two parts. Indeed, it follows from Proposition 2 that

\[
S_{1t} = E_t^{(o)} \left[ \int_t^\infty \xi^{(o)}_{u,t} e_u du \right] + E_t^{(o)} \left[ \int_t^\infty \xi^{(o)}_{u,t} S_{1u} \Gamma_u du \right],
\]

where the first term

\[
E_t^{(o)} \left[ \int_t^\infty \xi^{(o)}_{u,t} e_u du \right] = E_t^{(o)} \left[ \int_t^\infty e^{-\rho(u-t)} \left( \frac{s_t}{S_u} \right) du \right]
\]

(67)

gives the (risk-adjusted) present value of futures dividends, i.e., the fundamental value of the asset, and the second captures the present value of the flows of lending fees associated with ownership of the asset. As we discuss below in the two risky assets case, the fact that the lending yield is a deterministic function of the endogenous state variable implies that both components can be computed from the solution to a boundary value problem for a nonlinear differential equation, see (72) and (80).

### 4.2 Two risky assets

Consider now the model with two risky assets. Equation (54) shows that in order to compute the equilibrium asset prices it is sufficient to compute the price-dividend ratio of asset 2 or, equivalently, its market share

\[
w_t = \frac{S_{2t}}{M_t} = \rho E_t^{(o)} \left[ \int_t^\infty \xi^{(o)}_{u,t} e_{2u} du \right] = \rho \eta_2 E_t^{(o)} \left[ \int_t^\infty e^{-\rho(u-t)} \left( \frac{s_t}{S_u} \right) du \right].
\]

(68)

This expression makes it clear that \( w_t \) and thus the asset prices

\[
(S_{1t}, S_{2t}) = (1 - w_t, w_t) M_t
\]

(69)

29
depend on an expectation over the future path of the endogenous state variable. On the other hand, since

\[ \sigma_t S_{1t} = ((1 - w_t) \sigma - \text{diff}_t(w)) M_t, \]  

(70)

it follows from (48) that the drift and diffusion of \( s_t \) on the shorting region depend on \( w_t \) and its diffusion coefficient

\[ \text{diff}_t(w) = \frac{1}{d} \langle w_t, Z_t^{(d)} \rangle \]  

(71)

through the lending market clearing condition (40). Therefore, the triple \((s_t, w_t, \text{diff}_t(w))\) is the solution to a Forward Backward Stochastic Differential Equation over an infinite horizon (FBSDE, see Ma and Yong (1999) for a thorough introduction).

Since the evolution of the process \( s_t \) is fully determined by \((s_t, w_t, \text{diff}_t(w))\), it is natural to look for Markovian equilibria in which \( w_t = w(s_t) \) for some sufficiently regular bounded function such that

\[ w(0) = w(1) = \eta_2, \]  

(72)

where the equalities follow from the fact that the endogenous state variable is absorbed at the endpoints of the unit interval. Furthermore, Itô’s lemma and (70) show that for such a solution we have

\[ \text{diff}_t(w) = s_t (1 - s_t) v(s_t, \gamma_t) w'(s_t) \]  

(73)

and therefore

\[ \sigma_t S_{1t} = ((1 - w(s_t)) \sigma - s_t (1 - s_t) v(s_t, \gamma_t) w'(s_t)) M_t, \]  

(74)
where the function \( v(s_t, \gamma_t) \) is defined by (48b). Substituting into the short market clearing condition (40), gives a quadratic equation

\[
\frac{\gamma_t \sigma}{1 - s_t} = \frac{\gamma_t (\gamma_t + \Delta + \sigma) w'(s_t)}{(1 - w(s_t))(1 + s_t)} + \frac{(s_t \Delta - \gamma_t - \sigma)((\gamma_t + \Delta)s_t - \sigma)}{(1 - w(s_t))(1 + s_t)^2}
\] (75)

that implicitly determines the lending fee

\[
\gamma_t = \gamma(s_t, w(s_t), w'(s_t))
\] (76)

as a function of \( s_t, w(s_t), \) and \( w'(s_t) \) for all \( s_t > s^* \). Substituting this function into (48) then shows that the endogenous state variable evolves according to the autonomous stochastic differential equation defined by

\[
d s_t = m[w](s_t)dt + v[w](s_t)dZ_t^{(o)},
\] (77)

with the deterministic functions

\[
(m[w](s), v[w](s)) \equiv s(1 - s)(m, v)(s, \gamma(s, w(s), w'(s))).
\] (78)

This implies that \( s_t \) is a Markov diffusion and, since the process

\[
e^{-\rho t} \frac{w(s_t)}{s_t} + \rho \eta_2 \int_0^t e^{-\rho u} \frac{du}{s_u} = \rho \eta_2 E_t^{(o)} \left[ \int_0^\infty e^{-\rho u} \frac{du}{s_u} \right]
\] (79)

is by construction a martingale, it follows that the market weight is a piecewise twice continuously differentiable solution to

\[
\rho \left( \frac{w(s)}{s} \right) = \frac{w(s)}{s} \left( \frac{w(s)}{s} \right) + \frac{1}{2} \left( \frac{w(s)}{s} \right)^2 + \frac{\rho \eta_2}{s},
\] (80)

subject to the boundary condition (72).

This nonlinear boundary value problem is too complex to admit an explicit solution. We therefore resort to numerical methods to illustrate the quantitative implications of the model. As a first step, we start by observing that on the long region \([0, s^*]\) the differential
equation simplifies to
\[ p w(s) = \rho \eta_2 + \frac{1}{2} s^2 (1 - s)^2 \Delta^2 w''(s). \] (81)

A direct calculation shows that, for any given \( \varepsilon \in (0, \eta_2) \), the unique solution to this equation with \( w(0) = \eta_2 \) and \( w(s^*) = \varepsilon \) is explicitly given by
\[ w(s; \varepsilon) = \eta_2 + (\varepsilon - \eta_2) \left( \frac{s}{s^*} \right)^{\frac{1}{2} + \frac{1}{2} \sqrt{1 + \frac{2\rho \Delta}{\eta_2}}} \left( \frac{1 - s}{1 - s^*} \right)^{\frac{1}{2} - \frac{1}{2} \sqrt{1 + \frac{2\rho \Delta}{\eta_2}}}. \] (82)

Relying on this solution over the long region, we now combine a traditional shooting approach with a collocation method (see, e.g., Miranda and Fackler (2004) and Dangl and Wirl (2004)) to construct a global solution as follows: For each value of \( \varepsilon \) we implement a Chebyshev collocation to numerically solve equation (80) on the shorting interval \( [s^*, 1] \) subject to the initial conditions
\[ 0 = w(s^*) - w(s^*; \varepsilon) = w'(s^*) - w'(s^*; \varepsilon), \] (83)
and then numerically vary the value of the free constant until the solution satisfies the terminal boundary condition \( w(1) = \eta_2 \) required by (72).

To illustrate the quantitative implications of the model in the two asset case, we fix the underlying parameters \( (\sigma, \delta, \rho) = (10, 5, 1)\% \) and set \( e_t = 1 \) so that the equilibrium value of the market portfolio is normalized to 100. In the left panel of Figure 4, we plot the equilibrium price of asset 1
\[ S_{1t} = M_t - S_{2t} = (1 - w(s_t)) M_t \] (84)
and the present value of its future dividends, i.e., its fundamental value,
\[ f_1(s_t) M_t \equiv E^{(o)}_t \left[ \int_t^\infty \xi_{u,t}^{(o)} e_{1u} du \right] = \frac{\eta_1 w(s_t) M_t}{1 - \eta_1} \] (85)
when \( \eta_1 = 50\% \), so that half of the asset supply is available for shorting. In a frictionless environment, these functions would be constant and equal to \( \eta_1 M_t = 50 \) because, absent
shorting costs, asset 1 only entitles its owner to a constant share of dividends. As shown by the figure, this is no longer the case in the presence of a shorting friction. Indeed, since \( s^* = \sigma^2 / \delta = 0.2 < 1 \), we have \( S \neq \emptyset \) and it follows that asset 1 entitles its owners to strictly more than its share of dividends. Since the value of the market is fixed, this implies that the equilibrium price of asset 2, given by the risk-adjusted present value of its dividends, must account for less than \( 1 - \eta_1 = 50\% \) of the market, and it follows that the price of asset 1 must strictly exceed its frictionless value \( \eta_1 M_t = 50 \), which in turn must exceed the fundamental value of the asset.

The difference between the market value of the asset and the risk-adjusted present value of its dividends, that is

\[
\ell(s_t)M_t \equiv E_t^{(\sigma)} \left[ \int_t^\infty \ell^{(\sigma)}_{u,t} S_{1u} \Gamma_u du \right] = \left( 1 - \frac{w(s_t)}{1 - \eta_1} \right) M_t, \tag{86}
\]

represents the risk-adjusted present value of the lending fees associated with ownership of asset 1. The figure shows that this difference is bell-shaped as function of the state variable and can amount to as much as 10\% of the market portfolio when half of the asset supply can be shorted. To illustrate the impact of the supply parameter \( \eta_1 \) on the lending component, we plot in the right panel the relative contribution

\[
\frac{\ell(s_t)}{1 - w(s_t)} = \frac{1 - \eta_1 - w(s_t)}{(1 - \eta_1)(1 - w(s_t))} \tag{87}
\]

of this component to the price of asset 1 for different values of \( \eta_1 \) ranging from 1\% to 100\%. As shown by the figure, this contribution is also single-peaked as a function of the state variable and monotone decreasing in \( \eta_1 \). The latter property is intuitive. Indeed, as \( \eta_1 \) decreases, the dividend component of the asset cash flows naturally decreases but the lending fees component remains essentially unchanged because the demand for short positions is not directly affected by the supply parameter \( \eta_1 \), and it follows that lending fees must account for a larger share of the equilibrium asset price.

As discussed in Section 3.4, the nonlinearity of the equilibrium pricing rule can help explain apparent mispricing episodes such as the partial spin-off of Palm by 3Com. To illustrate this point, we identify asset 1 with the 5\% of shortable Palm shares and asset 2
Figure 4: Decomposition of the equilibrium price. The left panel plots the price of the asset 1 (solid line) and the risk-adjusted present value of its dividends (dashed line) when half of the supply is available for shorting. The right panel plots the present value of lending fees as a fraction of the asset price for different values of $\eta_1$. In both panels the parameters of the model are set to $\sigma = 10\%$, $\delta = 5\%$, $\rho = 1\%$, $e_t = 1$, and the hatched region indicates the interval over which no shorting activity takes place in equilibrium.

with the remaining shares held by 3Com. From the figure, we read that the equilibrium price of the block of shortable Palm shares evaluated at the point $s^*$ is given by $(1 - w(s^*))M_t = 5.87$ and includes $15.5\%$ of lending fees. Extrapolating this price to the remaining Palm shares values asset 2 at $(0.95/0.05)(1 - w(s^*))M_t = 111.41$ which exceeds the value $M_t = 100$ of the conglomerate and represents a premium of

$$
\left( \frac{0.95}{0.05} \right) \left( \frac{1 - w(s^*)}{w(s^*)} \right) - 1 = 18.35\% \quad (88)
$$

relative to the equilibrium price $w(s^*)M_t = 94.14$ of asset 2. Note that these figures are conservative because they are evaluated at the point $s^*$ that signals the entry into the shorting region. If instead we used as reference the point $\text{argmax} (1 - w(s)) \approx 0.62$, then the price of the block of shortable Palm shares would include $32\%$ of lending fees and the relative premium on asset 2 would increase to $46.84\%$. 

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**Figure 5:** Equilibrium with two risky assets. This figure plots the PD ratio and the volatility of the shortable asset (1st row), the shorting cost and the lending yield per unit of volatility (2nd row), and the shorting cost and the lending yield (3rd row) as functions of the consumption share of the optimist for different values of $\eta_1$ in a model with $\sigma = 10\%$, $\delta = 5\%$, $\rho = 1\%$ and $e_t = 1$. In each panel the hatched region indicates the interval of states over which no shorting activity takes place in equilibrium.
Turning to the lending market, Figure 5 plots the price-dividend ratio of the shortable asset PD_{1t} and its volatility \( \sigma_{1t} \), as well as the shorting cost and lending fees both per unit of volatility \((\phi_t, \gamma_t)\) and unscaled \((\Phi_t, \Gamma_t)\). The middle panels show that, when expressed in units of risk, the shorting cost and the lending yield are decreasing in the dividend share \(\eta_t\) and otherwise behave similarly as in the one asset case of Section 4.1 which here corresponds to the dash-dotted lines. The former feature can be understood as follows: As \(\eta_t\) decreases, asset 1 becomes more scarce so the share of total lending fees that each share entitles to increases. This tends to push the price up and the market price of risk down which in turn implies that the intermediary can charge a higher cost.

The bottom panels show that these intuitive properties no longer hold when the shorting cost and the lending yield are expressed as flow rates per dollar of short. This change can be traced back to the oscillatory behavior of the asset volatility in the top right panel, which in turn is implied by the behavior of the PD ratio in the top left panel of the figure. Indeed, since the PD ratio is hump-shaped and the diffusion of the endogenous state variable \(s_t(1 - s_t)\bar{\pi}[w](s_t) \geq 0\) vanishes at the endpoints of the state space, we have that the excess volatility

\[
\sigma_{1t} - \sigma = \frac{s_t(1 - s_t)\bar{\pi}[w](s_t)(-w'(s_t))}{1 - w(s_t)}
\]

is positive (negative) over the interval where the PD ratio \((1 - w(s_t))/(\eta_t \rho)\) is increasing (decreasing) and equal to zero at \(s_t = 0, s_t = 1\), and at the point where the PD ratio reaches its maximum. The top right and bottom panels of the figure show that the amplitude of the volatility oscillation is decreasing in the dividend share \(\eta_t\) and gets gradually transferred to the shorting costs and the lending yield as the shortable asset becomes scarce.
5 Empirical application

It follows from (6), (14), and (16) that, in equilibrium, the expected excess returns on the two risky assets can be expressed as

\[
\frac{1}{dt} E_t^{(o)} \left[ \frac{dS_{it} + \eta_i e_{it} dt}{S_{it}} \right] - r_t = \sigma_i b_{i2t}^{(o)} - 1_{\{i=1\}} \Gamma_t \\
= \sigma_i b_{i2t}^{(o)} - 1_{\{i=1\}} \left( \frac{\Upsilon_t}{1 + \Upsilon_t} \right) \Phi_t,
\]

where \( \Phi_t \) is the shorting cost, \( \Gamma_t \) is the lending yield, and \( \Upsilon_t \) is the fraction of the available inventory that is on loan. This shows that, within our simple framework, the two assets offer the same risk-return tradeoff once the lending revenues of asset 1 are taken into account, and the same logic suggests that in a more general model including multiple assets and sources of risk, risky securities with different shorting costs should offer equivalent risk-adjusted investment opportunities provided that the lending revenues that they generate are correctly controlled for.

This observation prompts us to revisit some of the findings of DD who document that stocks with higher shorting costs exhibit significantly lower returns that cannot be explained by standard risk factors and argue that these negative returns are a compensation for the systematic risk borne by the small fraction of investors who account for most of the shorting activity. If upheld in the data, the above prediction of our model would provide an alternative explanation: The returns of stocks with different shorting costs in DD and Beneish et al. (2015) only appear to be different because the lending revenues that they generate are not properly accounted for. In particular, the adjusted returns of a CME portfolio that is long in cheap-to-short stocks and short in expensive-to-short stocks should be explained by traditional factor models such as the three factors model of Fama and French (1993) and the four factors model of Carhart (1997).

To test these predictions we assume that the beliefs of the optimist coincide with those of the econometrician and construct a sample spanning the decade 2004–14 in which the monthly stock returns of U.S. equities from CRSP are matched with accounting information from Compustat, analyst forecasts from IBES, and the time series of shorting costs and utilization rates provided by Markit Securities Finance at the individual stock
level. Following DD, we then sort stocks into deciles at the end of each month based on their volume-weighted shorting cost over the previous 30 days. Each decile contains, on average, 270 stocks. The summary statistics of Table 1 confirm that the characteristics of our sample are very similar to those of the sample used in DD, Table 2. Specifically, shorting costs are low for most stocks and their distribution is highly skewed, with the average varying between 0.56bps/month and 4.98bps/month for deciles 1 through 9 and jumping to 57.79bps/month in decile 10. Measures of volatility and realized returns also closely track the characteristics of the sample used by DD. One exception is decile 10 for which the average excess return is positive in our sample and negative in DD, although we note that median returns for decile 10 are negative.

Consistent with previous studies, our sample also displays a negative relation between shorting costs and firm size (Saffi and Sigurdsson 2011), a positive relation between shorting costs and the dispersion of analyst forecasts (D’Avolio 2002), and a positive relation between shorting costs and the portfolio level utilization rate

$$Util_t^{[n]} = \frac{1}{270} \sum_{i=1}^{270} Util_{it}$$

(92)

that gives the average over each decile of the stock level utilization rates $Util_{it}$ computed by MSF as the ratio of the value of shares on loan to the total value of lendable shares (Beneish et al. 2015). Table 1 also reports the lending yield of each decile portfolio defined as the average

$$Yield_t^{[n]} = \frac{1}{270} \sum_{i=1}^{270} Yield_{it} = \frac{1}{270} \sum_{i=1}^{270} \left( \frac{Util_{it}}{1 + Util_{it}} \right) Cost_{it}$$

(93)

of the lending yields of the stocks in the portfolio computed from the shorting costs and the utilization rate as in the right hand side of (91). This variable is essentially aligned with the shorting cost but of much lower magnitude since our model assumes that lending revenues are shared by all owners on a value-weighted basis.
Table 1: Summary statistics for decile portfolios. This table reports mean (median) quantities associated with ten portfolios of stocks sorted on their shorting cost. The data set covers the period from January 2004 to January 2014 and is filtered to exclude stocks with a price below $5 per share. The resulting sample contains on average 270 stocks per decile.

<table>
<thead>
<tr>
<th>Decile n</th>
<th>Cost\textsuperscript{[n]} (bps/month)</th>
<th>Yield\textsuperscript{[n]} (bps/month)</th>
<th>Util\textsuperscript{[n]} (%)</th>
<th>Vol\textsuperscript{[n]} (%/month)</th>
<th>Disp\textsuperscript{[n]} (%)</th>
<th>MCap\textsuperscript{[n]} ($billion)</th>
<th>r\textsubscript{t} - r\textsubscript{t} (%/month)</th>
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</thead>
<tbody>
<tr>
<td>1 (Cheap)</td>
<td>0.56</td>
<td>0.05</td>
<td>9.05</td>
<td>2.18</td>
<td>8.86</td>
<td>13.20</td>
<td>0.82</td>
</tr>
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<td></td>
<td>0.52</td>
<td>0.02</td>
<td>4.82</td>
<td>1.80</td>
<td>2.53</td>
<td>3.59</td>
<td>0.63</td>
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<td>2.31</td>
<td>9.59</td>
<td>5.87</td>
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<td>1.80</td>
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<td>2.58</td>
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<td>0.29</td>
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<td>8.09</td>
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<td>3.26</td>
<td>7.14</td>
<td>0.21</td>
<td>-0.45</td>
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</table>

Cost\textsuperscript{[n]}: MSF reports the value-weighted average shorting cost for each security over the past 1, 3, 7, and 30 days where the weight assigned to a loan fee is the dollar value of the outstanding balance of the loan divided by the total dollar value of outstanding balances for that time period. Like DD we analyze trading strategies that are rebalanced monthly and therefore use the 30-day value-weighted average fee as our measure of a stock’s shorting cost and then average across stocks within each decile. If an observation is missing the 30-day value-weighted average fee, we simply drop it from the sample.

Yield\textsuperscript{[n]}: The monthly lending yield of decile \( n \) computed according to (93) using the shorting costs and utilization rates of the stocks in each portfolio.

Util\textsuperscript{[n]}: Utilization rate computed by MSF as the ratio of the value of assets on loan (from Beneficial Owners) to the total value of lendable assets and then averaged within each decile. MSF sources this data from several custodians and prime brokers. See Beneish et al. (2015) and Ramachandran and Tayal (2021) for a detailed description of the MSF data set.

Vol\textsuperscript{[n]}: Monthly volatility of the portfolio excess return measured as the sum of squared daily returns over a month as in (French et al. 1987, Schwert 1989).

Disp\textsuperscript{[n]}: Dispersion of analysts’ earnings-per-share forecast scaled by the absolute value of the average outstanding forecast as in (Diether et al. 2002).

\( r\textsubscript{t} - r\textsubscript{t} \): Monthly excess returns on decile-\( n \) computed using equal weights and the time series of the risk free rate \( r\textsubscript{t} \) provided in the data library of Kenneth French.
We then use the time series of monthly returns of the ten decile portfolios to estimate an empirical asset pricing model of the form

\[ r_t^{[n]} - r_t = \alpha^{[n]} + \sum_{k=1}^{K} \beta_k^{[n]} f_{kt} + \varepsilon_t^{[n]}, \]  

(94)

where \( f_t \in \mathbb{R}^K \) is a vector of risk factors and \( \varepsilon_t^{[n]} \) is an error term. Following DD and most of the empirical literature, we take as benchmarks the three-factor model of Fama and French (1993) with \( f_t = (\text{mkt}_t, \text{smb}_t, \text{hml}_t) \) and its four-factor extension by Carhart (1997) that also includes the momentum factor. The alphas estimated from these benchmark specifications along with their statistical significance are reported in the *benchmark* columns of Table 2. Consistent with Beneish et al. (2015) and DD, the results of this benchmark exhibit a strong shorting premium relative to either factor model. In particular, the unadjusted returns of the CME portfolio generate a strongly significant unexplained excess return of 64bps/month relative to the three-factor model and of 56bps/month relative to the four-factor model, and as in DD, these results become even more pronounced if we only use the most expensive-to-short half of decile 10 in the construction of the CME portfolio.

To show that this shorting premium can be explained within our framework, we estimate the adjusted factor models defined by

\[ r_t^{[n]} - r_t + \text{Yield}_t^{[n]} = \alpha^{[n]} + \sum_{k=1}^{K} \beta_k^{[n]} f_{kt} + \varepsilon_t^{[n]}, \]  

(95)

where \( f_t \in \mathbb{R}^K \) is the same vector of observable risk factors as before, \( \varepsilon_t^{[n]} \) is an error term, and the adjustment on the left hand side represents the average lending yield of the stocks in decile \( n \) computed according to (93). As can be seen from the *adjusted* columns of Table 2, the adjustment has little impact on the returns of deciles 1 to 9 in which shorting costs are minute. However, there is a sizable effect from adding the lending yield to returns of decile 10 as the estimated alpha jumps up from \(-49\text{bps}\) to \(-27\text{bps/month}\) relative to the three-factor model and from \(-37\text{bps}\) to \(-15\text{bps/month}\) relative to the four-factor model. What is perhaps more revealing is that the estimated alpha becomes statistically insignificant relative to both factor models.

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Table 2: Estimated alphas for decile and CME portfolios. The CME portfolios are constructed by going long in the stocks of decile 1 and short in the stocks of either decile 10 or decile 10b which corresponds to the more expensive-to-short half of decile 10. The adjusted returns of the CME portfolios are defined by (96) whereas the adjusted returns of the portfolio labeled 10 (Short) correspond to the net returns on a short position in decile 10. The t-statistics are computed using Newey-West standard errors with 12 lags and the superscripts *, **, and *** indicate statistical significance at the 10%, 5% and 1% levels.

| Factor model          | Decile n | $\alpha^{[n]}$ (%) | $|t|$-stat. | $\bar{\alpha}^{[n]}$ (%) | $|t|$-stat. |
|-----------------------|----------|-------------------|----------|--------------------------|----------|
| **Benchmark**         |          |                   |          |                          |          |
| Fama and French (1993)| 1 (Cheap)| 0.16              | (1.55)  | 0.16                     | (1.56)   |
|                       | 2        | 0.29***           | (3.65)  | 0.29***                  | (3.66)   |
|                       | 3        | 0.35***           | (5.04)  | 0.35***                  | (5.05)   |
|                       | 4        | 0.28***           | (3.43)  | 0.28***                  | (3.44)   |
|                       | 5        | 0.23              | (1.47)  | 0.23                     | (1.48)   |
|                       | 6        | 0.25**            | (2.06)  | 0.25**                   | (2.07)   |
|                       | 7        | 0.59***           | (2.92)  | 0.59***                  | (2.93)   |
|                       | 8        | 0.52*             | (1.98)  | 0.52**                   | (1.99)   |
|                       | 9        | 0.38              | (1.38)  | 0.39                     | (1.42)   |
|                       | 10 (Expensive) | −0.49*          | (1.88)  | −0.09                    | (1.05)   |
|                       | 10 (Short)| 0.49*            | (1.88)  | −0.09                    | (0.35)   |
| CME                   | [1–10]   | 0.64***           | (3.03)  | 0.06                     | (0.29)   |
|                       | [1–10b]  | 1.24***           | (4.29)  | 0.22                     | (0.78)   |
| **Adjusted**          |          |                   |          |                          |          |
|                       |          |                   |          |                          |          |
| Carhart (1997)        | 1 (Cheap)| 0.19**            | (2.60)  | 0.19**                   | (2.61)   |
|                       | 2        | 0.31***           | (3.70)  | 0.32***                  | (3.71)   |
|                       | 3        | 0.37***           | (6.42)  | 0.37***                  | (6.43)   |
|                       | 4        | 0.31***           | (4.47)  | 0.31***                  | (4.48)   |
|                       | 5        | 0.27**            | (2.19)  | 0.27**                   | (2.20)   |
|                       | 6        | 0.28**            | (2.53)  | 0.28**                   | (2.54)   |
|                       | 7        | 0.65***           | (4.53)  | 0.65***                  | (4.54)   |
|                       | 8        | 0.60***           | (3.61)  | 0.60***                  | (3.63)   |
|                       | 9        | 0.45**            | (2.24)  | 0.46**                   | (2.30)   |
|                       | 10 (Expensive) | −0.37**         | (2.10)  | −0.15                    | (0.88)   |
|                       | 10 (Short)| 0.37**           | (2.10)  | −0.21                    | (1.29)   |
| CME                   | [1–10]   | 0.56***           | (3.01)  | −0.03                    | (0.14)   |
|                       | [1–10b]  | 1.17***           | (4.48)  | 0.15                     | (0.59)   |
To remain consistent with the model, we cannot simply compute the adjusted returns on the CME portfolios by subtracting the adjusted return of decile 10 from that of decile 1, as this would impose a counterfactual identity between the shorting cost and the lending yield of decile 10. Instead, we define the adjusted CME returns as

\[
\text{CME}_{t}^{[1-m]} \equiv \left( r_{1}^{[t]} + \text{Yield}_{1}^{[t]} \right) + \left( -r_{m}^{[t]} - \text{Cost}_{m}^{[t]} \right),
\]

where the first term is the adjusted return on a long position in decile 1 and the second is the net return on a short position in decile \( m \in \{10, 10b\} \). As can be seen from the highlighted cells of Table 2, the results of the corresponding adjusted regressions strongly support the predictions of our model. Indeed, the estimated alphas for the long/short portfolio, which were positive and strongly significant in the benchmark case, are now very close to zero and statistically insignificant for both CME portfolios.

The impact of the shorting friction that our model captures is readily observed in the row labeled 10 (Short) which reports estimation results for the net returns of a short position in decile 10. The entry in the benchmark column displays the exact opposite to the entry for decile 10, whereas the entry in the adjusted column accounts for the shorting costs. As our results demonstrate, this operation not only changes the sign of the estimated alpha but also its statistical significance.

6 Conclusion

We study a dynamic general equilibrium model with costly short sales and heterogeneous beliefs. The closed-form solution to the model reveals how costly short sales drive a wedge between the valuation of assets that promise identical cash flows but are subject to different trading arrangements. In our model, the price of an asset is given by the risk-adjusted present value of the cash flows it promises but these cash flows include both dividends and an endogenous lending yield. This pricing formula implies that, after adjusting for lending revenues, asset returns satisfy a standard intertemporal capital asset pricing model and allows us to shed light on recent findings about the explanatory power of shorting costs in the cross-section of stock returns. In particular, we show empirically
that once returns are appropriately adjusted for lending fees, stocks with low and high shorting costs offer equivalent risk-return tradeoffs.

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A Proofs

Proof of Proposition 1. The solution follows from a direct application of the Karush, Kuhn, and Tucker conditions to (18) subject to (15) and (16).

Proof of Proposition 2. Let \( \xi_t = \xi_t^{(o)} \) with

\[
- \frac{d\xi_t}{\xi_t} = r_t dt + \theta_t^{(o)} dZ_t^{(o)}
\]

(97)
denote the marginal utility of the optimist. By construction, we have that

\[
N_{it} = \xi_t S_{it} + \int_0^t \xi_u (e_u + \mathbb{1}_{\{i=1\}} S_{1u} \Gamma_u) \, du
\]

(98)

are local martingales under \( P^{(o)} \) and it follows from Lemma 1 below that these processes are martingales over any finite horizon. In particular, we have that

\[
\xi_t^{(o)} S_{it} = E_t^{(o)} \left[ \xi_T S_{iT} + \int_t^T \xi_u (e_u + \mathbb{1}_{\{i=1\}} \Gamma_u S_{1u}) \, du \right]
\]

(99)

for all finite \( T < \infty \) and therefore

\[
\xi_t^{(o)} S_{it} = \lim_{T \to \infty} E_t^{(o)} [\xi_T S_{iT}] + E_t^{(o)} \left[ \int_t^\infty \xi_u (e_u + \mathbb{1}_{\{i=1\}} \Gamma_u S_{1u}) \, du \right]
\]

(100)

by monotone convergence, since the terms below the integral are all nonnegative. To complete the proof, it remains to show that the limit is zero. Let \( \lambda_t = 1/s_t - 1 \). As shown in the proof of Lemma 1 below, we have that

\[
\xi_T S_{iT} \leq \xi_T M_T = e^{-\rho T} M_0 \left( \frac{s_0}{s_T} \right) = e^{-\rho T} M_0 \left( \frac{1 + \lambda_T}{1 + \lambda_0} \right) \leq e^{-\rho T} M_0 \left( \frac{1 + \Lambda_T}{1 + \lambda_0} \right)
\]

(101)

for some \( P^{(o)} \)-martingale with initial value \( \lambda_0 \) and therefore

\[
\lim_{T \to \infty} E_t^{(o)} [\xi_T S_{iT}] \leq \lim_{T \to \infty} \frac{e^{-\rho T} M_0}{1 + \lambda_0} \left( 1 + E_t^{(o)} [\Lambda_T] \right) = \lim_{T \to \infty} e^{-\rho T} M_0 = 0,
\]

(102)

where the last equality uses the fact that \( \rho > 0 \). Since \( \xi_T S_{iT} \geq 0 \), this in turn implies that the limit is zero and the proof is complete.

Lemma 1. The process \( N_{it} \) is a \( P^{(o)} \)-martingale on \([0, T]\) for any \( T < \infty \).

Proof. By construction, we have that

\[
0 \leq N_{it} \leq N_t \equiv N_{1t} + N_{2t} = \xi_t M_t + \int_0^t \xi_u (e_u + S_{1u} \Gamma_u) \, du,
\]

(103)
and it is thus sufficient to show that the process $P_t$ is a martingale under $P^{(o)}$ over the finite time interval $[0, T]$. Since $S_{it} \sigma_{it} \geq 0$ we have that

$$S_{it} \sigma_{it} \leq \sum_{j=1}^{2} S_{jt} \sigma_{jt} = M_t \sigma.$$  \hspace{1cm} (104)

On the other hand, using (59) and the fact that $\gamma_t \leq \phi_t$ shows that we have

$$\gamma_t \leq \phi_t = 1_{\{s_t > s^*\}} \frac{s_t \Delta - (\gamma_t + \sigma)}{1 + s_t} \leq \Delta.$$  \hspace{1cm} (105)

Combining this inequality with the definition of $\xi_t$, we deduce that there are strictly positive constants such that

$$|N_t| \leq \xi_t M_t + \int_0^t \xi_u M_u \left( \rho + \gamma_u \sigma \right) du \leq \xi_t M_t + \int_0^t \xi_u M_u \left( \rho + \Delta \sigma \right) du \leq C + C' \sup_{u \in [0, T]} \lambda_u$$  \hspace{1cm} (106)

for all $t \in [0, T]$, where $\lambda_t \equiv 1/s_t - 1$. Using Itô’s lemma and the dynamics of the consumption share process in (48) shows that

$$d\lambda_t = \lambda_t \left( g(s_t, \gamma_t) dZ^{(o)}_t - f(s_t, \gamma_t) dt \right)$$  \hspace{1cm} (108)

for some functions $f, g : [0, 1] \times [0, \phi] \rightarrow \mathbb{R}$ such that $f(s, \gamma) \geq 0$ and $|g(s, \gamma)| \leq \Delta$. Therefore, Novikov’s condition implies that

$$\Lambda_t \equiv e^{\int_0^t f(s_u, \gamma_u) du} \lambda_t = \lambda_0 \exp \left( - \int_0^t g(s_u, \gamma_u) dZ^{(o)}_u - \frac{1}{2} \int_0^t |g(s_u, \gamma_u)|^2 du \right)$$  \hspace{1cm} (109)

is a $P^{(o)}$-martingale on any finite time interval and it thus follows from Doob’s maximal inequality that for any $q > 1$ we have:

$$E^{(o)} \left[ \sup_{u \in [0, T]} \Lambda^q_u \right] \leq E^{(o)} \left[ \sup_{u \in [0, T]} \Lambda^q_u \right] \leq \frac{q}{q - q} E^{(o)} \left[ \Lambda^q_T \right].$$  \hspace{1cm} (110)

Now, since $|g(s, \gamma)| \leq \Delta$ and the function $x^q$ is convex for any $q > 1$, it follows from the mean comparison results of Hajek (1985, e.g., Theorem 1.3) that

$$E^{(o)} \left[ \Lambda^q_T \right] \leq \Lambda^0_T E^{(o)} \left[ e^{q \Delta Z^{(o)}_T - \frac{q}{2} \Delta^2 T} \right] = e^{\frac{1}{2} q(q-1) \Delta^2 T} \Lambda^q_0.$$  \hspace{1cm} (111)

This implies that the right hand side of (107) is $P^{(o)}$-integrable and the required result finally follows from the dominated convergence theorem.  

\[\text{■}\]
Proof of equation (28). If \( \theta_{2t}^{(p)} \geq 0 \) then it follows from \( \gamma_t \leq \phi_t \) and (16) that we have
\[
\theta_{1t}^{(p)} + \phi_t \geq \theta_{1t}^{(p)} + \gamma_t = \theta_{2t}^{(p)} \geq 0,
\]
and therefore \( \gamma_t = 0 \) due to (27). To establish the converse implication, assume towards a contradiction that we have the lending yield \( \gamma_t = 0 \) but \( \theta_{2t}^{(p)} < 0 \). Then it follows from (16), (26) and (27) that we have
\[
0 = \phi_t \left( \theta_{1t}^{(p)} + \phi_t \right) W_t^{(p)} - \frac{1}{4} \left( \theta_{2t}^{(p)} \right)^2 W_t^{(p)}
\]
and therefore \( \theta_{2t}^{(p)} \geq 0 \), since the wealth of the pessimist is strictly positive.

Proof of equation (32). If \( \theta_{2t}^{(p)} < 0 \) then
\[
\theta_{1t}^{(p)} + \phi_t = \theta_{1t}^{(p)} + \max \left\{ \gamma_t, -\frac{1}{2} \theta_{1t}^{(p)} \right\} = \max \left\{ \theta_{1t}^{(p)} + \gamma_t, \frac{1}{2} \theta_{1t}^{(p)} \right\} \geq \max \left\{ \theta_{2t}^{(p)}, \frac{1}{2} (\theta_{2t}^{(p)} - \gamma_t) \right\} < 0,
\]
where the first equality follows from (26) and the third follows from (16).

Proof of Proposition 3. For \( s_t > s^* \) we have that (59) is equivalent to \( g_t(\gamma) = 0 \) with the quadratic function defined by
\[
g_t(\gamma) \equiv (1 - s_t)(s_t \Delta - \gamma - \sigma)((\gamma + \Delta)s_t - \sigma) - \gamma \sigma (1 + s_t)^2.
\]
Since
\[
g_t(0) = (1 - s_t)(s_t \Delta - \sigma)^2 > 0,
\]
\[
g_t'(0) = -\sigma (1 + s_t)^2 - (1 - s_t)^2 (s_t \Delta - \sigma) < 0,
\]
\[
g_t''(\gamma) = -s_t(1 - s_t) < 0,
\]
and
\[
\lim_{\gamma \to \infty} g_t(\gamma) = -\infty,
\]
it is clear that (59) admits a unique strictly positive solution. A direct calculation shows that this solution is given by (62) and substituting into (58) gives (61). The comparative statics follow by (61), (63), and (64) by differentiation. We omit the details.

Proof of equation (57). First observe that
\[
P_t^{(o)} \left[ \left\{ \sup_{u \geq t} s_u \in S \right\} \right] = P_t^{(o)} [\{ \tau^* < \infty \}],
\]
\(49\)
where the stopping time
\[ \tau^* \equiv \inf \{ u \geq t : s_t \geq s^* \} \] (122)
denotes the first time at or after \( t \geq 0 \) that the Itô process \( s_t \) finds itself in the shorting region.

To obtain the required probability, we will compute
\[ g_t \equiv E^{(o)}_t \left[ e^{-\lambda \tau^*} \right] = E^{(o)}_t \left[ e^{-\lambda \tau^*} 1_{\{\tau^* < \infty\}} \right], \] (123)
and then let \( \lambda \downarrow 0 \). On the time interval \([t, \tau^*]\), we have from (48) that the consumption share of the optimist evolves according to the autonomous SDE
\[ ds_t = s_t (1 - s_t) \Delta \left( dZ^{(o)}_t + (1 - s_t) \Delta dt \right). \] (124)
Therefore, it follows from well-known results (see, e.g., Karatzas and Shreve (1988, Chapter 5.7.A)) that \( g_t = g(s_t) \), where the function \( g : [0, 1] \to \mathbb{R} \) is the unique bounded function such that
\[ \lambda g(s) = s(1 - s)^2 \Delta^2 \left( g'(s) + \frac{1}{2} sg''(s) \right), \quad 0 \leq s \leq s^*, \] (125)
\[ g(s) = 1, \quad s^* \leq s \leq 1. \] (126)
Solving this differential equation gives
\[ g(s_t) = 1_{\{s_t > s^*\}} + 1_{\{s_t \leq s^*\}} \left\{ s^* \left( \frac{1 - s_t}{1 - s^*} \right) \right\}^{\frac{1}{2} \frac{1}{2} \sqrt{1 + \frac{2 \lambda s}{s_t}}}, \] (127)
and the desired result now follows from the dominated convergence theorem by letting the constant \( \lambda \downarrow 0 \) in the definition of \( g_t \).

\[ \blacksquare \]

**B Stochastic disagreement**

In this appendix, we discuss the construction of an equilibrium in an extension of the model where the divergence in beliefs is stochastic and time-varying.

Assume that the economy is populated by two agents indexed by \( a \in \{1, 2\} \) who have different perceptions of the evolution of the aggregate dividend process. Specifically, assume that in the eyes of agent \( a \)
\[ \frac{de_t}{e_t} = \mu_t^{(a)} dt + \sigma dZ_t^{(a)}, \] (128)
for some agent-specific Brownian motions $Z^{(a)}$ and growth rate process $\mu^{(a)}_t$ such that the scaled divergence in beliefs

$$
\Delta_t \equiv \frac{1}{\sigma} \left( \mu^{(1)}_t - \mu^{(2)}_t \right)
$$

is adapted to the filtration generated by the observation of the aggregate dividend process. As a typical example, one could consider an Ornstein-Uhlenbeck process for the disagreement, i.e., a process of the form

$$
d\Delta_t = -\lambda \Delta_t dt + dZ^{(o)}_t = -(1 + \lambda) \Delta_t dt + dZ^{(p)}_t,
$$

for some strictly positive constant $\lambda$, but the exact specification of the divergence process is unimportant for the arguments of this appendix. All the other building blocks of the model, i.e., the agents’ preferences, the assets they trade, and the shorting mechanism remain the same as in the benchmark model of Section 2.

If the disagreement never changes sign then this model is essentially equivalent to the benchmark model of Section 2 with the identification $[o, p] = [1, 2]$ if the disagreement is always positive and $[o, p] = [2, 1]$ in the opposite case. Now assume that the disagreement is not signed. In this case, the identity of the optimist is a stochastic process that changes back and forth between $o_t = 1$ when the disagreement is positive and $o_t = 2$ when it is negative. As a result, the equilibrium can be constructed by analogy with that of the benchmark model by observing that the consumption share of agent 1 evolves like the consumption share of the optimist in the benchmark model at times where the disagreement is nonnegative, and as the consumption share of the pessimist at times where it is negative. For brevity we only outline the main steps.

Let $s_t \in [0, 1]$ denote the consumption share of agent 1 which we will use as an endogenous state variable. Proceeding along the lines of Sections 3.2 and 3.3 shows that the equilibrium shorting cost and lending yield satisfy

$$
-\phi_t = 1_{\{S^{(1)}\}} \frac{1}{2} \theta^{(1)}_{2t} + 1_{\{S^{(2)}\}} \frac{1}{2} \theta^{(2)}_{2t}
$$

and

$$
\gamma_t \sigma_1 (1 - w_t) = (\phi_t - \gamma_t) \phi_t \left( 1_{\{S^{(1)}\}} s_t + 1_{\{S^{(2)}\}} (1 - s_t) \right),
$$

where

$$
S^{(2)} = \left\{(\omega, t) : \theta^{(2)}_{2t} < 0 \leq \theta^{(1)}_{2t} \right\} = \left\{(\omega, t) : s_t > s^* \left( \Delta_t^+ \right) \right\}
$$

gives the region of the state space over which agent 2 is short in asset 1 and agent 1 holds long positions in both risky assets, and

$$
S^{(1)} = \left\{(\omega, t) : \theta^{(1)}_{2t} < 0 \leq \theta^{(2)}_{2t} \right\} = \left\{(\omega, t) : 1 - s_t > s^* \left( \Delta_t^- \right) \right\}
$$

(133)

(134)
gives the region over which agent 1 is short in asset 1 and agent 2 holds long positions in both risky assets. This, in turn, implies that the shorting market is endogenously inactive over the region given by

\[ L \equiv (\Omega \times \mathbb{R}_+) \cup \{ (\omega, t) : \min_a \theta^{(a)}_{2t} \geq 0 \} \]

\[ = \{ (\omega, t) : 1 - s^* (\Delta_t^-) \leq s_t \leq s^* (\Delta_t^+) \} \]

and substituting these expressions into (41) and (48) shows that the equilibrium interest rate, the equilibrium market price of risk perceived by agent 1, and the equilibrium evolution of her consumption share are explicitly given by

\[ \theta^{(1)}_{2t} = \theta^*(s_t) - \mathbb{1}_{\{s_t > s^*(\Delta_t^+)\}} \frac{(1 - s_t) (s_t \Delta_t^+ - \sigma - \gamma_t)}{1 + s_t} \]

\[ - \mathbb{1}_{\{1 - s_t > s^*(\Delta_t^-)\}} \frac{s_t ((1 - s_t) \Delta_t^- - \sigma - \gamma_t)}{2 - s_t} \]

\[ r_t = r^*(s_t) + \mathbb{1}_{\{s_t > s^*(\Delta_t^+)\}} \frac{s_t (1 - s_t) (\Delta_t^+ + \sigma + \gamma_t) (s_t \Delta_t^+ - \sigma - \gamma_t)}{(1 + s_t)^2} \]

\[ + \mathbb{1}_{\{1 - s_t > s^*(\Delta_t^-)\}} \frac{s_t (1 - s_t) (\Delta_t^- + \sigma + \gamma_t) ((1 - s_t) \Delta_t^- - \sigma - \gamma_t)}{(2 - s_t)^2} \]

and

\[ \frac{ds_t}{s_t (1 - s_t)} = m(s_t, \gamma_t; \Delta_t^+) dt + v(s_t, \gamma_t; \Delta_t^+) dZ_t^{(1)} \]

\[ - m(1 - s_t, \gamma_t; \Delta_t^-) dt - v(1 - s_t, \gamma_t; \Delta_t^-) dZ_t^{(1)} \]

where the functions \( m(s, \gamma; \Delta) \) and \( v(s, \gamma; \Delta) \) are defined as in (48b) and (48c). See Figure 6 for an illustration of the equilibrium trading regions.

To complete the construction of the equilibrium, it now remains to solve for the equilibrium lending yield \( \gamma_t \) and to compute the asset prices. In the one asset case, the derivation follows the same steps as in Section 4.1. In particular, we find that the equilibrium shorting cost and lending yield are given by

\[ (\Phi_t, \Gamma_t) = (\Phi, \Gamma) (s_t, \Delta_t^+) + (\Phi, \Gamma) (1 - s_t, \Delta_t^-) \]

with the functions \( \Phi(s, \Delta) \) and \( \Gamma(s, \Delta) \) implicitly defined by the right hand sides of (61) and (62). The comparative statics are very similar to those of the benchmark model with a constant disagreement. In particular, the flow rates \( (\Gamma_t, \Phi_t) \) and the equilibrium utilization ratio

\[ \Upsilon_t = \frac{1 - s_t}{\sigma^2} \Phi (s_t, \Delta_t^+) + \frac{s_t}{\sigma^2} \Phi (1 - s_t, \Delta_t^-) \]

are all convex and \( u \)-shaped in the disagreement \( \Delta_t \).
Figure 6: Trading regions with a stochastic disagreement. The figure illustrates the shape of the equilibrium trading regions and allows us to determine the configuration that occurs for each level of disagreement among agents.

The representation of equilibrium prices—or of the fundamental value in the one asset case—is slightly more complex than in the benchmark model because each agent successively participates on both sides of the shorting market. Proposition 1 and the above characterization of the equilibrium trading regions imply that the normalized marginal utility of agent 1 evolves according to

\[ -d\xi^{(1)}(t)/\xi^{(1)}(t) = r_t dt + \left( \theta^{(1)}(t) + 1 \{ S^{(2)}(t) \} \gamma_t + 1 \{ S^{(1)}(t) \} \phi_t \right) dZ^{(1)}_t. \] (145)

Under appropriate integrability assumptions on the disagreement process \( \Delta_t \), this expression can be combined with arguments similar to those of the proof of Proposition 2 to show that the equilibrium prices satisfy

\[ S_{2t} = E^{(1)}_t \left[ \int_t^\infty \xi^{(1)}_{t,u} e_{2u} du \right], \] (146)

and

\[ S_{1t} = E^{(1)}_t \left[ \int_t^\infty \xi^{(1)}_{t,u} \left( \epsilon_{1u} + 1 \{ s_u > s^*(\Delta_u^+) \} S_{1u} \Gamma_u + 1 \{ 1 - s_u > s^*(\Delta_u^+) \} S_{1u} \Phi_u \right) du \right]. \] (147)
To understand this expression, note that from the point of view of agent 1 the cash flows that are relevant to the equilibrium valuation of asset 1 depend on which side of the shorting market the agent is. On the set $L$, the only relevant cash flow is the dividend $e_{1t}$ since the shorting market is inactive. On the set $S^{(2)}$, the agent is long in asset 1 so that the relevant cash flows are the dividend and the lending yield $S_{1t}\Gamma_t$ associated with each share of the asset, and finally on $S^{(1)}$, the agent is short so that the relevant cash flows are now given by the dividend and the shorting cost $S_{1t}\Phi_t$ required to maintain a short position. Importantly, if the disagreement process is always positive then the latter region is empty and we recover (30).

To derive a differential equation for the equilibrium price of asset 1, we assume that the scaled disagreement follows an autonomous diffusion process

$$d\Delta_t = \mu(\Delta_t)dt + \Sigma(\Delta_t)dZ_t^{(1)},$$

with values in some set $D \subset \mathbb{R}$ and then proceed as in Section 4.2 albeit with an additional state variable. Specifically, we look for an equilibrium in which

$$S_{1t} = w(s_t, \Delta_t) M_t$$

for some sufficiently regular function $w : [0, 1] \times D \rightarrow [0, 1]$ such that

$$w(0, \Delta) = w(1, \Delta) = \eta_1, \quad \forall \Delta \in D.$$

Itô’s lemma and (70) show that, in such a Markovian equilibrium, the diffusion coefficient of asset 1 satisfies

$$\frac{\text{diff}_t(S_1)}{M_t} = \sigma w(s_t, \Delta) + w'_\Delta(s_t, \Delta_t)\Sigma(\Delta_t)$$

$$+ w'_s(s_t, \Delta_t) \left( v(s_t, \gamma_t, \Delta_t^+) - v(1 - s_t, \gamma_t, \Delta_t^-) \right).$$

Substituting into (131) and (132) then gives a linear-quadratic system that implicitly determines the shorting cost and the lending fee as functions

$$(\Phi[w](s_t, \Delta_t), \Gamma[w](s_t, \Delta_t))$$

of $s_t$, $\Delta_t$, $w(s_t, \Delta_t)$, and the derivatives $(w'_s, w'_\Delta)(s_t, \Delta_t)$. Taking these functions as given, it follows from (141) that the endogenous state variable evolves according to

$$ds_t = \bar{v}[w](s_t, \Delta_t)dt + \bar{v}[w](s_t, \Delta_t)dZ_t^{(1)}$$
for some explicit drift and diffusion functions \( (\bar{m}, \bar{v})[w](s, \Delta) \). This, in turn, implies that the pair \((s_t, \Delta_t)\) forms a Markov process and, since

\[
e^{-\rho t} \frac{w(s_t, \Delta_t)}{s_t} + \int_0^t e^{-\rho u} \left( \rho \eta_1 \mathbb{1}_{\{s_u > s^*(\Delta_u^+)\}} w(s_u, \Delta_u) \Gamma[w](s_u, \Delta_u) + \mathbb{1}_{\{1-s_u > s^*(\Delta_u^-)\}} w(s_u, \Delta_u) \Phi[w](s_u, \Delta_u) \right) \frac{du}{s_u}
\]

is a martingale as a result of (147), we deduce that the function \( u \equiv \frac{w}{s} \) is a piecewise twice continuously differentiable solution to

\[
(\rho - \beta[w](s, \Delta)) u = \frac{\rho \eta_1}{s} + \mu(\Delta) u'_\Delta + \frac{1}{2} \Sigma(\Delta)^2 u''_{\Delta \Delta}
\]

\[
+ \bar{m}[w](s, \Delta) u'_s + \bar{v}[w](s, \Delta) \Sigma(\Delta) u''_{s \Delta} + \frac{1}{2} \bar{v}[w](s, \Delta)^2 u''_{s s \Delta},
\]

subject to the boundary condition (150), where

\[
\beta[w](s, \Delta) \equiv \mathbb{1}_{\{s > s^*(\Delta^+)\}} \Gamma[w](s, \Delta) + \mathbb{1}_{\{1-s > s^*(\Delta^-)\}} \Phi[w](s, \Delta)
\]

denotes the additional cash flow per dollar of asset value in (147) as a function of the state variables \( s, \Delta \) taking as given \( w(\cdot) \) and \( w'(\cdot) \).

A numerical solution to this nonlinear boundary value problem can in principle be constructed using the same collocation approach as in the constant disagreement case of Section 4.2, albeit in two dimensions and subject to the caveat that the differential equation no longer admits an explicit solution on the long region in general. We leave the challenges of this implementation for future research.