Equilibrium Portfolio Strategies in the Presence of Sentiment Risk and Excess Volatility

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ABSTRACT

Our objective is to identify the trading strategy that would allow an investor to take advantage of “excessive” stock price volatility and “sentiment” fluctuations. We construct a general-equilibrium “difference-of-opinion” model of sentiment in which there are two classes of agents, one of which is overconfident about a public signal, while still being intertemporal optimizers. The overconfident investors overreact to the signal and change their expectations too often, sometimes being excessively optimistic, sometimes being excessively pessimistic. They thereby introduce an additional risk factor, cause stock prices to be excessively volatile, and induce rational investors to be conservative in their bond and equity investment. Moreover, the rational investor’s portfolio strategy is based not just on a current price divergence but also on their prediction about future sentiment behavior and the speed of convergence of prices.
The excess-volatility puzzle – identified by Shiller (1981) and LeRoy and Porter (1981) – points to a form of market inefficiency. But, so far, the investment strategy that would serve to exploit that form of inefficiency and would cause those responsible for excess volatility to part with their wealth has not been identified. Suppose that a financial market is deemed to be affected by fluctuations in market “sentiment” so that sentiment volatility causes prices to be more volatile than what would be justified by dividend volatility alone.\(^1\) Suppose further that a Bayesian, intertemporally optimizing investor trades in that market. We would like to know what investment policy this person will undertake in equilibrium.

To address this question, we build a general-equilibrium model of a financial market in which a subpopulation of investors trades on “sentiment” and generates excess volatility. In our model, some investors are overconfident in the sense that they give too much credence to a public information signal. One way to capture that behavioral feature has recently been proposed by Scheinkman and Xiong (2003). In their model of a “tree” economy, a stream of dividends is paid. Some aspect of the stochastic process of dividends is not observable by anyone. All investors are risk neutral, are constrained from short selling, and receive information in the form of the current dividend and some public signals. Agents have different beliefs about the correlation between innovations in the signal and innovations in the unobserved variables: “overconfident” agents are people who steadfastly believe that this correlation is a positive number when, in fact, it is equal to zero. This causes them to give too much weight to the signals.

Here, we consider a similar setting except that all investors are risk averse and are allowed to sell short. One class of agents knows the true correlation. They form “proper” beliefs by using Bayes’ formula. A second class of investors are overconfident about the public signal. As a result, they change their beliefs too often about economic prospects: when they receive a signal, they overreact to it, which then generates excessive stock price movements. We say that volatility is “excessive” when, for the given utility functions of agents, the level of volatility is larger than it would be if all agents knew the correlation to be equal to zero, and we refer to the fluctuations in the probability beliefs of overconfident agents relative to agents with the proper beliefs as fluctuations in “sentiment.” The non-zero correlation number to which overconfident investors adhere, is the single parameter in our model that causes it to differ from a traditional rational-expectations general equilibrium model.

In their contest with investors who process information rationally under the proper beliefs, we want the traders who are overconfident to be full-fledged intertemporal optimizers nonetheless. It is well-known that complete irrationality in the manner of positive “feedback traders” à la De Long et al. (1990b) can amplify the volatility of stock prices and that the additional volatility creates “noise-trader risk” for rational arbitrageurs, thereby creating a limit to arbitrage. However, feed-
back traders or traders acting randomly may not be the best representation of irrational behavior because they are “sitting ducks” for rational investors. Furthermore, it is not clear where the consumption units they are losing are coming from. For this reason, we prefer to model a general equilibrium economy where the overconfident traders are intertemporal optimizers with fully specified budget constraints, even if they have overconfident beliefs. In this way, welfare analysis and the analysis of gains and losses of the two groups of traders remain meaningful.\(^2\)

We identify three distinct aspects of the portfolio strategy of the investors who process information properly.\(^3\) First, these investors may not agree today with the market about its current estimate of the growth rate of dividends: there is a difference of opinion. When the investors with the proper beliefs are more optimistic than the market, they increase their investment in equity, while decreasing their investment in bonds (because equity and bonds are positively correlated). Second, all investors, whether they agree today or not, know that they will revise their forecasts of growth, with consequent future changes in the rate of interest. Third, if there is a difference of opinion today, investors with the proper beliefs are aware that the overconfident investors will revise their probability beliefs differently from the way their own will be revised: the difference of opinion drives sentiment, which is stochastic, and sentiment risk carries a risk premium. The second and third effects cause investors with the proper beliefs to hedge; they hold fewer shares of equity than would be optimal in a market without excess volatility. Thus, our analysis illustrates how “risk arbitrage” must be based not just on a current price divergence but also on a model of overconfident behavior, so that a protection can be put in place.

The model we develop can be viewed as an equilibrium model of investor sentiment, in the sense of Barberis et al. (1998). Several studies have supported the hypothesis that agents active in the financial markets exhibit aspects of behavior that deviate from rationality.\(^4\) This was done typically on the basis of some natural experiments, for instance, spin offs, share repurchases, initial public offerings, reactions to news, etc. In order to sort out which behavioral aspects actually exist in the marketplace, it would be important to conduct tests using data on asset prices. For that purpose, one must deduce theoretically the behavior of asset prices and portfolio choices that will prevail in the financial markets as a result of a particular deviation from rationality. That was the intent of three classic papers in this strand of the literature, which can be called “behavioral equilibrium theory”: Barberis et al. (1998), Daniel et al. (1998), and Hong and Stein (1999). The first two of these papers feature a single group of agents who are non-Bayesian.\(^5\) The model of Hong and Stein (1999), like ours, features two groups of agents with heterogeneous beliefs who, in contrast to the agents in our model, are not intertemporal optimizers.

The heterogeneity of beliefs between agents needs to be regenerated; otherwise Bayes’ law causes it to die out.\(^6\) There exist basically two ways of doing that: either agents receive different informa-
tion or they receive the same information but process it differently. The first, more sophisticated approach is to let agents receive private signals, as in the vast “Noisy-Rational Expectations” literature originating from the work of Grossman and Stiglitz (1980), Hellwig (1980) and Wang (1993), in which agents learn also from price, a channel that is not present in our model. In most renditions of noisy-rational expectations equilibria, the market includes noise traders who behave randomly, a feature we would like to avoid. The second approach is the one utilized in “difference-of-opinion” models such as Harris and Raviv (1993), Kandel and Pearson (1995), and Cecchetti et al. (2000); see Morris (1995) for a discussion of this approach. Agents disagree about the basic model they believe in, or about some fixed model parameter, and they do not learn from each others’ behavior: they “agree to disagree”. Under this approach, agents are viewed as being non-Bayesian, but, one can say equally well, as do Biais and Bossaerts (1998), that agents remain Bayesian and place infinite trust on some aspect of their prior. The second approach is more easily tractable while capturing some of the same phenomena as the first one. It is the approach we adopt here, as it has been in the recent, related work of David (2008). In our model, the non-zero correlation number to which overconfident investors adhere is the mechanism by which heterogeneity of beliefs is regenerated.

On the technical side, we adopt the exponential linear-quadratic framework, introduced in the work of Constantinides (1992) on the term structure, in the work of Kim and Omberg (1996) on dynamic portfolio choice, and now extensively used in term-structure and volatility modeling (see Cheng and Scaillet (2007)). It is a very flexible functional setting in that it can handle any finite number of state variables. In our model, asset prices are weighted averages of exponential linear-quadratic functions of state variables, with time varying weights. Ours is, to our knowledge, the first general-equilibrium application of that mathematical framework.

The balance of this paper covers the following material. In Section I, we describe agents’ beliefs and the way beliefs evolve over time. In Section II, we determine the equilibrium allocation of consumption and the equilibrium pricing kernel. In Section III, we produce the explicit solution for all securities prices. In Section IV, we derive and study the diffusion matrix and all moments of securities prices; we verify, of course, that overconfident investors bring about excessive volatility. In Section V, we identify the main factors driving the portfolio strategy of the rational trader and we show how it can be implemented via trading in stocks and bonds, as functions of the values of the fundamental state variables of the economy. In Section VI, we explain the link that exists between the portfolio strategy and the trader’s ability to predict returns in the long run. In Section VII, we derive the speed of impoverishment of the overconfident traders and demonstrate that excess volatility may be a long-lived phenomenon. Section VIII contains the conclusion. Throughout, we highlight our main results in propositions while technical results are presented in lemmas, with their mathematical derivations provided in the appendix.
I. Beliefs and Information Structure

We wish to develop a dynamic general-equilibrium model where investors have heterogeneous expectations about some aspect of the economy. In the first subsection below, we specify an endowment economy populated by two groups of investors, Group A and Group B, who harbor different expectations about the process driving aggregate dividends. In the second subsection, we show that the dynamics of expectations that we have specified can be viewed as the result of a learning process of the kind proposed by Scheinkman and Xiong (2003).

A. Beliefs and Their Dynamics

We consider an economy with an aggregate dividend-flow diffusion process \( \{ \delta_t \} \), which we regard as the “fundamental” variable.

**Assumption 1:** Under the beliefs of Group B, the process for aggregate dividends is driven by the following pair of stochastic differential equations, which defines a Markovian system in two state variables, \( \{ \delta_t, \hat{f}^B_t \} \):

\[
\frac{d\delta_t}{\delta_t} = \hat{f}^B_t dt + \sigma_\delta dW^B_{\delta,t},
\]

\[
d\hat{f}^B_t = -\zeta (\hat{f}^B_t - \bar{f}) dt + \frac{\gamma^B}{\sigma_\delta} dW^B_{\delta,t},
\]

where \( W^B_{\delta} \) is a one-dimensional process that is Brownian under the probability measure that reflects the expectations of Group B.

From Equation (1), we see that \( \hat{f}^B_t \) is the growth rate of \( \delta \) conditionally expected by Group B and Equation (2) postulates that this expected growth rate follows a mean reverting Ornstein-Uhlenbeck process, with the mean-reverting parameter given by \( \zeta > 0 \). The coefficients \( \gamma^B \) and \( \sigma_\delta \) are assumed to be positive and constant. An interpretation of \( \gamma^B \) is given in the next subsection. We can think of the pair of equations (1)–(2) as the “fundamental system”.

Our economy is a heterogeneous-expectations economy, where the belief of Group A about the expected growth rate of aggregate dividends differs from that of Group B.

**Assumption 2:** Group A believes that the expected growth rate of aggregate dividends is equal to:

\[
\hat{f}^A_t = \hat{f}^B_t - \bar{g}_t.
\]
Hence $\hat{g}_t$ is the “difference of opinion”. The dynamics of the difference of opinion are specified to follow an Ornstein-Uhlenbeck process that mean reverts ($\psi > 0$) to zero:

$$
\begin{align*}
    d\hat{g}_t &= -\psi \hat{g}_t dt + \sigma_{\hat{g},\delta} dW^B_{\delta,t} + \sigma_{\hat{g},s} dW^B_{s,t}, \\
    \sigma_{\hat{g},\delta} &\geq 0; \quad \sigma_{\hat{g},s} \leq 0.
\end{align*}
$$

where $W^B_s$ is a second one-dimensional process, independent of $W_\delta$, that is Brownian under the probability measure that reflects the expectations of Group B.

In the equation above, the volatilities $\sigma_{\hat{g},\delta}$ and $\sigma_{\hat{g},s}$ are constant with the signs postulated. An interpretation for them and a justification for the sign assumptions are given in the next subsection.

Because of the difference of opinion about the conditionally expected growth rate of aggregate dividends, the two groups have different probability beliefs:

**Assumption 3:** Group A differs from Group B in its beliefs about the aggregate dividends process. Group A’s probability beliefs at time $t$ are represented by a change of measure $\eta$, where $\{\eta_t\}$ is a strictly positive martingale process. For any event $e_u$ belonging to the $\sigma$-algebra of time $u$, we have:

$$
E^A_t [1_{e_u}] = E^B_t [\frac{\eta_u}{\eta_t} 1_{e_u}].
$$

In the eyes of Group B, $\eta$ captures the way in which Group A’s probability beliefs differ from theirs. We call $\eta$ the “sentiment” variable. Girsanov’s theorem tells us how the difference of opinion gets encoded into $\eta$ to generate different probability beliefs. When Group B is currently comparatively pessimistic ($\hat{g}_t < 0$), Group A views positive innovations in $\delta$ (or $s$) as more probable than Group B does. This, from Girsanov’s theorem, implies positive innovations in the change of measure $\eta$ for those states of nature in which $\delta$ (or $s$) has positive innovations:

**Lemma 1:** (Girsanov)

$$
\frac{d\eta_t}{\eta_t} = -\hat{g}_t \frac{1}{\sigma_\delta} dW^B_{\delta,t}.
$$

We can think of the pair of equations (4)–(6) as the “sentiment system”.

The joint dynamics of the four state variables, $\{\delta, \eta, \hat{f}^B, \hat{g}\}$, in the eyes of Group B are completely characterized by the Markovian system (1), (2), (4), and (6). Three of the four state variables, namely $\delta, \eta$ and $\hat{f}^B$, are always perfectly correlated, but $\delta$ and $\hat{f}^B$ are always positively correlated with each other, while the diffusion vector of $\eta$ has the sign opposite to the sign of $\hat{g}$. Of these four state variables, two will have a direct, immediate effect on the economy. They are the fundamental, $\delta$, and the sentiment, $\eta$. The fundamental moves independently of sentiment but sentiment is correlated with the fundamental for reasons that will be made clear shortly. The
other two state variables, \( \hat{f}^B \) and \( \hat{g} \), have only an indirect effect in that they solely act on the first two: \( \hat{f}^B \) is the current estimate of the drift of \( \delta \) by Population \( B \), and the difference of opinion, \( \hat{g} \), determines the diffusion of sentiment, \( \eta \). Functionally speaking, the “fundamental system” (1)–(2) and the “sentiment system” (4)–(6) are unrelated to each other but they are correlated. Instantaneously, the four state variables, \( \{ \delta, \hat{f}^B, \eta, \hat{g} \} \), which are driven by only two Brownians, \( W^B_\delta \) and \( W^B_s \), have the following \( 4 \times 2 \) diffusion matrix:

\[
\begin{bmatrix}
\delta \sigma_\delta > 0 & 0 \\
\frac{\gamma_B}{\sigma_\delta} > 0 & 0 \\
-\eta \frac{\sigma_{\delta}}{\sigma_s} & 0 \\
\sigma_{\hat{g}, \delta} \geq 0 & \sigma_{\hat{g}, s} \leq 0
\end{bmatrix}.
\] (7)

Notice a property of the diffusion matrix of state variables: although the difference of opinion is affected by both the \( W_\delta \) shock and the \( W_s \) shock, the \( W_s \) shock has an impact on only the difference-of-opinion state variable, \( \hat{g} \).

In summary, there are two distinct effects of heterogeneous beliefs and their dynamics.\(^8\) One determines the volatility of sentiment, the other the volatility of volatility of sentiment so that, in essence, our model is a model of stochastic volatility in a state variable. The difference of opinion \( \hat{g} \) scales the diffusion of \( \eta \), which implies that \( \eta \) has a diffusion that can take large positive or negative values. Sentiment or heterogeneity of beliefs, \( \eta \), is volatile. This first effect is cumulative over time, or long term. Second, the difference of opinion, \( \hat{g} \), itself is stochastic. Even when the two groups of investors happen to agree today about the growth rate of aggregate dividends (\( \hat{g} = 0 \)), all investors still know that they will revise their future estimates of the growth rate, and that they will do so in different ways, so that they will not agree tomorrow. This second effect (the effect of the dynamics of heterogeneous beliefs) is instantaneous or short-term.

To understand the long-run effects of a random shock pair \( dW^B_t = \{ dW^B_{\delta,t}, dW^B_{s,t} \} \) occurring at date \( t \), we shall have occasion in Section VI to use Malliavin calculus, which is a form of calculus that tells how shocks occurring today affect the values of variables in the distant future.

B. An Information-Processing Interpretation for the Formation of Beliefs

In the previous subsection, the dynamics for the beliefs of the two agents are specified exogenously. In this subsection, we show that these dynamics can be viewed as being the result of a learning process similar to that proposed by Scheinkman and Xiong (2003).

We start by re-specifying the process for aggregate dividends \( \delta \) under the effective measure.
Assumption 1': The stochastic process for $\delta$ is:

$$\frac{d\delta_t}{\delta_t} = f_t dt + \sigma_\delta dZ^\delta_t,$$

where $Z^\delta$ is a Brownian under the effective probability measure, which governs empirical realizations of the process. The conditional expected growth rate of aggregate dividends, $f_t$, behaves according to:

$$df_t = -\zeta (f_t - \overline{f}) dt + \sigma_f dZ^f_t; \quad \zeta > 0,$$

where $Z^f$ is a Brownian under the effective probability measure.

We assume that the conditional expected growth rate of dividends, $f$, is not observed by any agent, and thus, must be estimated by them.

Assumption 2': All investors estimate, or filter out, the current value of $f$ and its future behavior using the observation of the current dividend, $\delta$, and the observation of a public signal, $s$, which has the following process:

$$ds_t = \sigma_s dZ^s_t,$$

where $Z^s$ is a third Brownian under the objective probability measure as well.

All three Brownians, $\{Z^\delta, Z^f, Z^s\}$, are uncorrelated with each other (under the objective probability measure and any measure equivalent to it) so that, instantaneously, innovations $dZ^s$ in the signal convey no information about innovations $dZ^f$ in the unobserved variable. That is, the signal is pure noise.

Finally, we assume that agents in Group A are overconfident about the signal while agents in Group B are not.

Assumption 3': Agents in Group A perform their filtering under the overconfident belief that the signal, $s$, has positive correlation $\phi \in [0, 1]$ with $f$ when, in fact, it has zero correlation. The “model” they have in mind is:

$$ds_t = \sigma_s \phi dZ^f_t + \sigma_s \sqrt{1 - \phi^2} dZ^s_t.$$

Group B, on the other hand, believes properly that the true correlation is zero.

Now, because of the assumed nonzero correlation $\phi$ in the eyes of Group A, the signal provides the agents of that group with short-run, albeit false, information about the current shock to the dividend growth rate.

This single number $\phi$, which we call the “overconfidence coefficient”, parameterizes the degree of excess confidence placed in the signal by Group A. It could be argued that overconfident agents
should gradually detect from their empirically-realized consumption losses that their beliefs are overconfident. We do not address this issue, but in Section VII we show that the consumption losses of the overconfident group are “so gradual” that detection may be difficult.

From filtering theory (see Lipster and Shiryaev (2001, Theorem 12.7, page 36)), the conditional expected values, \( \hat{f}^A \) and \( \hat{f}^B \), of \( f \) according to individuals of Group \( A (\phi \neq 0) \) and Group \( B \) obey the following stochastic differential equations:\(^{10}\)

\[
d\hat{f}^A_t = -\zeta \left( \hat{f}^A_t - \bar{f} \right) dt + \frac{\gamma^A}{\sigma^2} \left( \frac{d\delta}{\delta} - \hat{f}^A_t \right) dt + \frac{\phi \sigma_f}{\sigma_s} ds, \tag{12}
\]

\[
d\hat{f}^B_t = -\zeta \left( \hat{f}^B_t - \bar{f} \right) dt + \frac{\gamma^B}{\sigma^2} \left( \frac{d\delta}{\delta} - \hat{f}^B_t \right) dt. \tag{13}
\]

The numbers \( \gamma^A \) and \( \gamma^B \) are the steady-state variances of \( f \) as estimated by Group \( A \) and \( B \) respectively.\(^{11}\) These variances would normally be deterministic functions of time. But for simplicity we assume, as did Scheinkman and Xiong (2003), that there has been a sufficiently long period of learning for people of both groups to converge to their level of variance, irrespective of their prior.

In our model, no agent knows the true state of the economy. Hence, the objective measure is not defined on either agent’s \( \sigma \)-algebra and we can ignore it for the purpose of calculating the equilibrium. In the previous subsection, we wrote the above stochastic differential equations directly in terms of processes that are Brownian motions under subjective probability measures. We have used \( B \)'s probability measure as the reference measure. Under this measure, the process for the signal is:

\[
ds_t = \sigma_s dW^B_{s,t}. \tag{14}
\]

Based on Assumptions 1’–3’, the following lemma gives the relation between the specification of the model in the previous subsection and this subsection.

**Lemma 2:** In Equation (4) the mean-reversion parameter for the difference-of-opinion process, \( \psi \), and the volatilities, \( \sigma_{g,\delta} \) and \( \sigma_{g,s} \), are given by:

\[
\psi = \zeta + \frac{\gamma^A}{\sigma^2} > 0, \tag{15}
\]

\[
\sigma_{g,\delta} = \frac{\gamma^B - \gamma^A}{\sigma^2} \geq 0, \tag{16}
\]

\[
\sigma_{g,s} = -\phi \sigma_f \leq 0. \tag{17}
\]

From the above lemma, we see that the mean-reversion parameter, \( \psi \), is positive. The diffusion coefficients have the property that \( \sigma_{g,\delta} \geq 0 \) and \( \sigma_{g,s} \leq 0 \). Moreover, both coefficients are equal to
zero when the overconfidence coefficient $\phi$ takes the value zero so that $\hat{g}$ has zero diffusion and the rank of the diffusion matrix drops from 2 to 1. This is because, in that case, the signal shock is bogus and ignored by all.

II. Equilibrium Allocation of Consumption and Pricing

We are interested in the interaction between two groups, who harbor different and evolving expectations. Differences in risk aversion and differences in the rate of impatience are not our main focus. So, we restrict our analysis to the setting in which

**Assumption 4:** Both groups have power utility with the same risk aversion, $1 - \alpha$, and rate of impatience, $\rho$.

A. The Individual’s Optimization Problem

Assuming a complete financial market, we can use the martingale, “static” formulation (as done in Cox and Huang (1989) and Karatzas et al. (1987)). Then, the problem of Group $B$ is to maximize the expected utility from lifetime consumption:

$$\sup_c E^B \int_0^\infty e^{-\rho t} \frac{1}{\alpha} (c^B_t)^\alpha \ dt; \quad \alpha < 1,$$

subject to the lifetime budget constraint:

$$E^B \int_0^\infty \xi_t^B c_t^B dt = \bar{\theta}^B E^B \int_0^\infty \xi_t^B \delta_t dt,$$

where $\xi^B$ is the change of measure from Group $B$’s probability measure to the risk neutralized measure (which we determine in the next subsection) and $\bar{\theta}^B$ is the share of equity with which $B$ is initially endowed. The first-order condition for consumption equates marginal utility to $\lambda^B \xi^B_t$, where $\lambda^B$ is the Lagrange multiplier of the budget constraint (19):

$$e^{-\rho t} (c_t^B)^{\alpha-1} = \lambda^B \xi^B_t.$$ (20)

Group $A$ holds an initial share $\bar{\theta}^A = 1 - \bar{\theta}^B$ of the equity and faces an analogous optimization problem. The only difference is that Group $A$ uses a probability measure that is different from that of Group $B$. Under $B$’s probability measure, the problem of $A$ can be stated as follows:

$$\sup_c E^B \int_0^\infty \eta_t e^{-\rho t} \frac{1}{\alpha} (c_t^A)^\alpha \ dt,$$ (21)
subject to the lifetime budget constraint:

\[ E^B \int_0^\infty \xi_t^B c_t^A dt = \bar{\theta}^A E^B \int_0^\infty \xi_t^B \delta_t dt. \]  

(22)

The first-order condition for consumption in this case is

\[ \eta_t e^{-\rho t} (c_t^A)^{\alpha - 1} = \lambda^A \xi_t^B, \]  

(23)

where \( \lambda^A \) is the Lagrange multiplier of the budget constraint (22).

**B. Equilibrium Pricing Measure**

An equilibrium is a price system and a pair of consumption-portfolio processes such that: (i) investors choose optimally their consumption-portfolio strategies, given their perceived price processes; (ii) the perceived security price processes are consistent across investors; and (iii) commodity and securities markets clear.

The aggregate resource constraint, from (20) and (23), is

\[ \left( \frac{\lambda^A \xi_t^B e^{\rho t}}{\eta_t} \right)^{\frac{1}{\alpha - 1}} + \left( \frac{\lambda^B \xi_t^B e^{\rho t}}{\eta_t} \right)^{\frac{1}{\alpha - 1}} = \delta_t. \]  

(24)

Solving this equation:

\[ \xi_t^B(\delta_t, \eta_t) = e^{-\rho t} \left[ \left( \frac{\eta_t}{\lambda^A} \right)^{\frac{1}{1 - \alpha}} + \left( \frac{1}{\lambda^B} \right)^{\frac{1}{1 - \alpha}} \right]^{1 - \alpha} \delta_t^{\alpha - 1}, \]  

(25)

and, therefore:

\[ c_t^A = \omega(\eta_t) \delta_t, \]  

(26)

\[ c_t^B = (1 - \omega(\eta_t)) \delta_t, \]  

(27)

where:

\[ \omega(\eta_t) \triangleq \frac{\left( \frac{\eta_t}{\lambda^A} \right)^{\frac{1}{1 - \alpha}}}{\left( \frac{\eta_t}{\lambda^A} \right)^{\frac{1}{1 - \alpha}} + \left( \frac{1}{\lambda^B} \right)^{\frac{1}{1 - \alpha}}}, \]  

(28)

is the share of consumption of Group A. The consumption-sharing rule is linear in \( \delta \) because both groups have the same risk aversion. But its slope is stochastic because the share of consumption allocated to each group, \( \omega(\eta_t) \), is driven by sentiment, \( \eta \).

The equilibrium value of \( \xi^B \) – the martingale pricing density under B’s probability measure – depends on \( \eta \), the probability density of A relative to B, that is, sentiment. In addition to reflecting
the abundance or scarcity of goods, as is usual in the absence of state preference or heterogeneous beliefs, the state prices also incorporate a power (or Hölder) average of the probability beliefs of the two groups (given by the term that is in square brackets in Equation (25)). As \( \eta \) fluctuates, average probability belief or “sentiment” fluctuates with it. In writing his/her budget constraint based on \( \xi^B \), Agent B anticipates A’s beliefs. This reflects “higher-order beliefs.”

Notice in Equation (25) that the functional forms of \( \xi^B \) with respect to \( \delta \) and with respect to \( \eta \) are very different from each other. This is because fundamental risk and sentiment risk have very different economic effects on utility and marginal utility. From Equation (25), one can show that the first derivative of \( \xi^B (\delta, \eta) \) with respect to \( \delta \) is negative while the second derivative is positive. The second derivative of the function \( \xi^B (\delta, \eta) \) with respect to \( \eta \) has the same sign as \( \alpha \). The cross derivative of the function \( \xi^B (\delta, \eta) \) is unambiguously negative. These derivatives have the following economic interpretation.

The fundamental, \( \delta \), has the customary aggregate effect on both groups: when output increases, marginal utility decreases. Thus, an increase in future expected output decreases the expected value of discount factors. Furthermore, marginal utility is convex with respect to \( \delta \). Jensen’s inequality implies that an increase in fundamental risk increases the expected value of discount factors, which is the familiar precautionary-saving motive.\(^{16}\)

In contrast to the fundamental, which is an aggregate shock, sentiment acts as a wedge between the two groups.\(^{17}\) Because the second derivative of \( \xi^B (\delta, \eta) \) with respect to \( \eta \) has the same sign as \( \alpha \), if \( \alpha < 0 \) (risk aversion greater than 1), discount factors are concave with respect to \( \eta \) so that an increase in sentiment risk (the variance of \( \eta \)), by Jensen’s inequality, reduces the expected values of all the future stochastic discount factors written with respect to \( B \)’s measure.


Because there are two Brownians that agents care about, \( W^B_\delta \) and \( W^B_s \), three securities that are linearly independent are required to complete financial markets and implement the equilibrium. The choice of securities is arbitrary. We assume that there is a riskless, instantaneous bank deposit with a rate of interest \( r \). The second security, equity or total wealth, pays the aggregate dividend \( \delta \) perpetually. The third security we introduce is a perpetual bond paying continuously and indefinitely a coupon equal to one unit of consumption per unit of time. We shall denote the price at date \( t \) of the bond as \( P_t \) and the price of equity as \( F_t \).
The equilibrium price of a fixed-income bond, which we denote by $P$, can be obtained directly from the pricing measure (25):

$$P \left( \hat{f}^B, \eta, \hat{g}, t \right) = \int_t^\infty \mathbb{E}^{B, \hat{f}^B, \hat{g} \mid \xi_t^B} \left[ \frac{\xi_{tu}^B}{\xi_t^B} \right] du. \quad (29)$$

Similarly, the equilibrium price of long-lived equity is the discounted sum of all future dividends. This is also the total wealth of the economy, which we denote by $F$:

$$F \left( \delta, \hat{f}^B, \eta, \hat{g}, t \right) = \int_t^\infty \mathbb{E}^{B, \eta, \hat{f}^B, \hat{g} \mid \xi_t^B} \left[ \frac{\xi_{tu}^B}{\xi_t^B} \delta_{tu} \right] du = \delta \int_t^\infty \mathbb{E}^{B, \eta, \hat{f}^B, \hat{g} \mid \xi_t^B} \left[ \frac{\xi_{tu}^B}{\xi_t^B} \delta_t \right] du. \quad (30)$$

We shall also have occasion to refer to the single-payoff versions of these securities, in which case we shall add a superscript $T$ for the maturity date of the payoff. So, for instance, the price today of a claim paying a single dividend $\delta_T$ is:

$$F_T \left( \delta, \hat{f}^B, \eta, \hat{g}, t \right) = \mathbb{E}^{B, \delta, \hat{f}^B, \hat{g} \mid \xi_t^B} \left[ \frac{\xi_{Tu}^B}{\xi_t^B} \delta_{Tu} \right]. \quad (31)$$

### III. Transform Analysis and Other Technical Issues

From Equation (25) for the martingale pricing density under $B$’s probability measure and Equations (29) and (30) for the prices of the bond and equity, we see that the joint conditional distribution of $\eta_u$ and $\delta_u$, given $\delta_t, \eta_t, \hat{f}^B_t, \hat{g}_t$ at $t$ will be needed to characterize the prices of distant-maturity claims, and hence, portfolio policies. That joint distribution is not easy to obtain but, remarkably, its characteristic or moment generating function or Fourier transform $E^{B, \hat{f}^B, \hat{g} \mid \xi_t^B} \left[ \left( \frac{\delta_u}{\delta} \right)^\varepsilon \left( \frac{\eta_u}{\eta} \right)^\chi \right]; u \geq t; \varepsilon, \chi \in \mathbb{R}$ (or $\mathbb{C}$) can be obtained in closed form.

**Proposition 1:** 1. The moment-generating-function for the joint distribution of $\delta$ and $\eta$ at maturity $u$ under the measure of Group $B$ is given by

$$E^{B, \hat{f}^B, \hat{g} \mid \xi_t^B} \left[ \left( \frac{\delta_u}{\delta} \right)^\varepsilon \left( \frac{\eta_u}{\eta} \right)^\chi \right] = H_f \left( \hat{f}^B, t; u; \varepsilon \right) \times H_g \left( \hat{g}, t; u; \varepsilon, \chi \right), \quad (32)$$

where

$$H_f \left( \hat{f}^B, t; u; \varepsilon \right) = \exp \left\{ \varepsilon \left[ \hat{f} (u - t) + \frac{1}{\zeta} \left( \hat{f}^B - \hat{f} \right) \left[ 1 - e^{-\zeta (u - t)} \right]\right] + \frac{1}{2} \varepsilon (\varepsilon - 1) \sigma_B^2 (u - t) \right. + \frac{\varepsilon^2 \gamma^B}{2 \zeta^2} \left[ 1 - e^{-\zeta (u - t)} \right]^2 + \frac{\varepsilon^2 \sigma^2}{4 \zeta^3} \left[ 2 \zeta (u - t) - 3 + 4 e^{-\zeta (u - t)} - e^{-2 \zeta (u - t)} \right], \quad (33)$$

$$H_g \left( \hat{g}, t; u; \varepsilon, \chi \right) = \exp \left\{ A_1 (\chi; u - t) + \varepsilon^2 A_2 (\chi; u - t) + \varepsilon \hat{g} B (\chi; u - t) + \hat{g}^2 C (\chi; u - t) \right\}. \quad (34)$$
and where the functions $A_1$, $A_2$, $B$ and $C$ are given in the proof. Moreover, the moment-generating-function is finite and real for $0 \leq \chi \leq 1$ and $\varepsilon \in \mathbb{R}$.

2. $\frac{\partial}{\partial \hat{f} B} H_f \left( \hat{f}^B, u, t; \varepsilon \right) = H_f \left( \hat{f}^B, u, t; \xi \right) \frac{\varepsilon}{\xi} \left[ 1 - e^{-\zeta(u-t)} \right]$ and has the sign of $\varepsilon$.

3. For $0 \leq \chi \leq 1$, in a neighborhood of $\hat{g} = 0$, the derivative $\frac{\partial H_g}{\partial g}$ is nonnegative if $\varepsilon < 0$ ($\varepsilon B(\chi; u - t) \geq 0$) and nonincreasing ($C(\chi; u - t) \leq 0$).

Observe, from Equations (32), (33) and (34), that the moment-generating-function for the joint distribution of $\delta$ and $\eta$ takes the form of a product of a function $H_f$, which is linear exponential in $\hat{f}^B$, with a function $H_g$, which is quadratic exponential in $\hat{g}$ with all coefficients being functions only of time and being available in closed form. The formula is a generalization of Heston (1993) and Kim and Omberg (1996) (recently extended by Liu (2007)) and perhaps others. Our economy’s dynamic system belongs to the category of exponential affine-quadratic models. Cheng and Scaillet (2007) in a term-structure context, have recently clarified the manner in which exponential affine-quadratic models can be reformulated as exponential affine models, for which solutions are well-known.

The inverse transformation to obtain securities prices can be performed in general by means of the Fast Fourier Transform (see the appendix). For the sake of speed, precision and simplicity, we conduct the calculation of these prices for the special case in which risk aversion is an integer, $(1 - \alpha) \in \mathbb{N}$, which excludes the cases of risk aversion smaller than 1 (that is, $\alpha > 0$). While in general the measure (25) is a weighted power mean of the two terms corresponding to the two groups, in the special case of risk aversion being an integer, the pricing measure can be written in an alternative way by expanding the bracket into an exact finite sum by virtue of the binomial formula:

$$\left[ \left( \frac{\eta u}{\chi^A} \right)^{1-\alpha} + \left( \frac{1}{\lambda^B} \right)^{1-\alpha} \right]^{1-\alpha} = \frac{1}{\chi^B} \sum_{j=0}^{1-\alpha} C_{1-\alpha}^j \left( \frac{\eta u \lambda^B}{\chi^A} \right)^{1-\alpha}$$

$$= \frac{1}{\chi^B} \sum_{j=0}^{1-\alpha} C_{1-\alpha}^j \left[ \frac{\omega(\eta u)}{1 - \omega(\eta u)} \right]^j,$$

where $C_{1-\alpha}^j$ denotes the binomial coefficient $\binom{1-\alpha}{j}$. Therefore, using the moment generating function obtained in (32), the price of the bond in (29) is obtained by integrating the following over future times $u$:

$$\mathbb{E}_B^{\eta, f^B, \hat{g}} \left[ \frac{f^B(\delta^g, \eta u)}{\xi^B_t(\delta, \eta)} \right] = e^{-\rho(u-t)} H_f \left( \hat{f}^B, t, u; \alpha - 1 \right) \times [1 - \omega(\eta)]^{1-\alpha}$$

$$\times \sum_{j=0}^{1-\alpha} C_{1-\alpha}^j \left[ \frac{\omega(\eta)}{1 - \omega(\eta)} \right]^j H_g \left( \hat{g}, t, u; \alpha - 1, \frac{j}{1 - \alpha} \right),$$

15
and the price of equity in (30) is obtained from:

$$\mathbb{E}^B_{\eta, \hat{f}^B, \tilde{g}} \left[ \frac{\xi^B(\delta, \eta)}{\xi^B(\delta, u)} \frac{\delta_u}{\delta} \right] = e^{-\rho(u-t)} H_f \left( \hat{f}^B, t, u; \alpha \right) \times [1 - \omega(\eta)]^{1-\alpha} \times \sum_{j=0}^{1-\alpha} C_j \left[ \frac{\omega(\eta)}{1 - \omega(\eta)} \right]^j H_g \left( \hat{g}, t, u; \alpha, \frac{j}{1-\alpha} \right).$$

(37)

Asset prices in our framework are weighted sums of exponential-quadratic functions. The time varying weights, $\omega(\eta)$, capture the role of the fluctuating distribution of consumption in the population, itself arising from sentiment fluctuations, $\eta$.

### IV. Excess Volatility

In this section, we verify that overconfident beliefs induce higher volatility of asset returns than would be the case if all investors had proper beliefs. To do so, we study the diffusion matrix of stocks and bonds, which will also be needed when making up portfolios.

The diffusion vector of the stock, $F$, and the bond, $P$, is its sensitivity or “exposure” to the shocks in the fundamental and in the signal, $dW^B_\delta$ and $dW^B_s$, respectively. It can be obtained from the gradient of the price function postmultiplied by the diffusion matrix of state variables (Equation (7)). Each security’s price exposure, therefore, has four components corresponding to the four elements of the gradient vector: $\left( \frac{\partial F}{\partial \delta}, \frac{\partial F}{\partial \hat{f}^B}, \frac{\partial F}{\partial \eta}, \frac{\partial F}{\partial \tilde{g}} \right)$ for the stock, and $\left( \frac{\partial P}{\partial \delta}, \frac{\partial P}{\partial \hat{f}^B}, \frac{\partial P}{\partial \eta}, \frac{\partial P}{\partial \tilde{g}} \right)$ for the bond, all these derivatives being known in closed form. The diffusion vectors are:

$$\begin{bmatrix} \text{diffF} \\ \text{diffP} \end{bmatrix} = \begin{bmatrix} \frac{\partial F}{\partial \delta} & \frac{\partial F}{\partial \hat{f}^B} & \frac{\partial F}{\partial \eta} & \frac{\partial F}{\partial \tilde{g}} \\ \frac{\partial P}{\partial \delta} & \frac{\partial P}{\partial \hat{f}^B} & \frac{\partial P}{\partial \eta} & \frac{\partial P}{\partial \tilde{g}} \end{bmatrix} \begin{bmatrix} \delta \sigma_\delta \\ \frac{\gamma^B}{\sigma_\delta} \\ \frac{\eta}{\sigma_\delta} \\ \frac{\gamma^A - \gamma^A}{\sigma_\delta} \end{bmatrix}.$$  

(38)

Given the emphasis of this article, however, it will be useful to separate the diffusion components that arise from the movements of the difference of opinion, $\hat{g}$. We define the diffusion arising from the first three state variables as:

$$\begin{bmatrix} \text{diff_3F} \\ \text{diff_3P} \end{bmatrix} = \begin{bmatrix} \delta \sigma_\delta \\ \frac{\gamma^B}{\sigma_\delta} \\ \frac{\eta}{\sigma_\delta} \end{bmatrix}.$$  

(39)

Clearly, the total exposures are:

$$\begin{bmatrix} \text{diffF} \\ \text{diffP} \end{bmatrix} = \begin{bmatrix} \text{diff}_3F \\ \text{diff}_3P \end{bmatrix} + \begin{bmatrix} \frac{\partial F}{\partial \tilde{g}} \\ \frac{\partial P}{\partial \tilde{g}} \end{bmatrix} \begin{bmatrix} \frac{\gamma^B - \gamma^A}{\sigma_\delta} \\ -\phi \sigma_f \end{bmatrix}.$$  

(40)
The derivatives \( \frac{\partial F}{\partial \hat{g}} \) and \( \frac{\partial P}{\partial \hat{g}} \) capture the exposures to changes in the difference of opinion. By definition, the exposures to \( W_s \) shocks arise only from these derivatives. Item 3 of Lemma 1 above implies that, in a neighborhood of \( \hat{g} = 0 \),

\[
\left. \frac{\partial F}{\partial \hat{g}} \right|_{\hat{g}=0} \geq 0, \quad \text{and} \quad \left. \frac{\partial P}{\partial \hat{g}} \right|_{\hat{g}=0} \geq 0,
\]

when risk aversion is greater than 1, because \( \varepsilon \) is set at \( \alpha \) and at \( \alpha - 1 \) respectively. It also implies that these derivatives are nonincreasing in \( \hat{g} \).

A. Illustration

In order to illustrate the effect of expectations and their dynamics on securities prices, we specify numerical values for the parameters of the model. Even though our objective is not to match the magnitude of particular moments in the data, we would like to work with parameter values that are reasonable. The parameter values that we specify are based on the estimation undertaken in Brennan and Xia (2001) for a model similar to ours. We limit ourselves to the case in which risk aversion is greater than one (\( \alpha < 0 \)). The particular values chosen for all the parameters are listed in Table I.

[Table 1 about here.]

As we analyze securities’ returns and portfolio strategies, we display results for the following four cases:

Case 1. Where all agents have the proper beliefs (\( \phi = 0 \)) and are in agreement about the growth rate of aggregate dividends (\( \hat{g} = 0 \), or zero difference of opinion). This corresponds to the setting in a standard model where all agents have homogeneous beliefs and identical priors.

Case 2. Where all agents have the proper beliefs (\( \phi = 0 \)) but disagree about the growth rate (\( \hat{g} \neq 0 \)). This corresponds to the setting where agents have the same beliefs but different priors.

Case 3. Where one group of agents is overconfident (\( \phi = 0.95 \)) but currently in agreement with the other group about the growth rate (\( \hat{g} = 0 \)). Because Group A is overconfident in the signal, even though they currently agree with Group B, they will disagree in the future.

Case 4. Where one group of agents is overconfident (\( \phi = 0.95 \)) and disagrees today about the growth rate (\( \hat{g} \neq 0 \)). In this case, agents of the two groups have different beliefs and these beliefs fluctuate randomly over time.
Our results are displayed in figures. Each plot in the figures has on the x-axis either current difference of opinion, $\hat{g}$, or the relative share of aggregate consumption of the overconfident Groups $A$, $\omega$. Each plot has typically three curves on it, with the dotted line representing the case where all agents have the proper beliefs ($\phi = 0$), the dashed line representing the case where Group $A$ is overconfident ($\phi = 0.95$) and the continuous line representing the construct “$\text{diff}_3$” containing three terms only, agents $A$ being, however, overconfident.

Note that when $\phi = 0$, everyone has proper beliefs and so they ignore the signal. This implies that effectively there is only one shock in the economy. Consequently, one risky security is sufficient to complete the financial market; that is, the equity and the bond are redundant relative to each other.

[Figure 1 about here.]

Figure 1 portrays the relative diffusion vectors for stocks and bonds with respect to the $W_\delta$ shock. We draw three inferences from this figure. First, we see that the dotted curves, for the case of proper beliefs, and the solid line, for the case of overconfident beliefs but ignoring the derivative with respect to $\hat{g}$, are very close to each other in all the plots:

\[
\begin{bmatrix}
\frac{\text{diff}_F}{F} \\
\frac{\text{diff}_P}{P}
\end{bmatrix}
\approx
\begin{bmatrix}
\frac{\text{diff}_3F}{F} \\
\frac{\text{diff}_3P}{P}
\end{bmatrix}
\quad \phi \neq 0.
\]

(42)

Analyzing this small difference, one would find that the derivatives $\left(\frac{1}{F} \frac{\partial F}{\partial \delta}, \frac{1}{P} \frac{\partial P}{\partial \delta}\right)$ for the stock and $\left(\frac{1}{P} \frac{\partial P}{\partial \delta}, \frac{1}{P} \frac{\partial P}{\partial \eta}\right)$ for the bond hardly change with $\phi$ and that the difference arises mostly from a slight dependence on $\phi$ of the derivatives with respect to sentiment $\frac{1}{F} \frac{\partial F}{\partial \eta}$ and $\frac{1}{P} \frac{\partial P}{\partial \eta}$. Second, comparing the dashed line for the case of overconfidence to the other two in any plot, we conclude that the changes in the diffusion vector of assets arising from overconfidence are almost entirely the result of the $\hat{g}$ component. To understand the effect of $\hat{g}$, recall from Lemma 1, that the intercept and slope of the two partial derivatives $\frac{\partial F}{\partial g}$ and $\frac{\partial P}{\partial g}$ have respectively the signs of the $\varepsilon B$ ($\geq 0$) and $C$ ($\leq 0$) terms of the moment generating function. Furthermore, as we have indicated earlier in Lemma 2, overconfidence has a marked effect on the diffusion of the difference of opinion $\hat{g}$, with the signs indicated in Equation (7). Because $(\gamma^B - \gamma^A)/\sigma_\delta \geq 0$, the traders know that a positive $dW_\delta^B$ shock will increase the difference of opinion (see also Equation (4)) and because $-\phi \sigma_f \leq 0$, a positive $dW_\delta^B$ shock will decrease it.

Thirdly, we see from Figure 1 that equity is mostly positively exposed to the realized innovation in the fundamental, $dW_\delta^B$, whereas the bond is negatively exposed. The reason for this difference between the equity and the bond is that a positive shock to the fundamental affects equity in
two ways: it changes the immediate payoff upward but also the valuation operator downward; the positive effect dominates, at least in a neighborhood of $\hat{g} = 0$. For bonds, an innovation in the fundamental has only a valuation effect, which is negative ($\text{diff}P < 0$).

The signal innovation, $dW^B_s$, is similar in its effects but it has no immediate payoff implication, so that both the equity and the bond have a negative exposure to it, which, however, is not displayed in the figure.

B. Volatilities and Correlation

The volatilities of assets are obtained from the diffusion vectors described in the previous subsection. The effect of the overconfidence of Group A is generally to increase the volatility of asset prices. This occurs because of the greatly increased volatility of the state variable $\hat{g}$ representing the difference of opinion, which also increases the volatility of sentiment, $\eta$.

Figure 2 plots the volatilities of the rate of return on equity and the bond, and the correlation between them. In the first column of the figure, these three quantities are plotted against difference of opinion, $\hat{g}$, for the two cases of overconfident beliefs ($\phi = 0.95$, dashed line) and proper beliefs ($\phi = 0$, dotted line). In the second column of the figure, these three quantities are plotted for the same two cases but now against the relative weight of Group A, $\omega$.

From the first two plots in the left column of Figure 2, we see that overconfident investors create “noise” that increases the volatility of both risky assets—the stock and the bond. The last row of the plots in Figure 2 shows that the correlation between stock and bond returns increases or decreases with the presence of overconfident investors, as well as with the difference of opinion between the two investor groups, depending on whether it is equal to plus or minus 1 in the absence of overconfidence. In a neighborhood of $\hat{g} = 0$, when risk aversion is greater than 1, it increases since the prices of the equity and the bond move in the same direction when expectations of future growth fluctuate.

An exception to the increase in volatility arising from overconfidence is to be seen in the right-hand column, middle row. When the consumption share of overconfident investors, $\omega$, is sufficiently low, it is possible for the volatility of bonds to be reduced by overconfidence, because of the opposite effects of $\eta$ and $\hat{g}$ on the volatility of the bond price.

In summary, overconfident investors create “noise”, which tends to increase the volatility of both stock and bond returns and also the correlation between them. The volatilities and correlation mostly increase with an increase in the relative weight of Group A in the population. We are going
to see now that overconfident investors also chase away the investors with proper beliefs from the bond and equity markets.

V. Equity-and-Bond Portfolio Strategy of Group B Analyzed According to Motive

We now study the fluctuations of the wealth of Group B and deduce from it the main features of the portfolio strategy of the investors with the proper beliefs.

The wealth of agents of Group B can be determined by applying the same approach that we used to find the equity and bond prices (see the appendix). To do so, we interpret the wealth of agents of Group B as the price of a “security” whose flow payoff at future times $u$ is the consumption $(27)$ of these investors:

$$
F^B_{\delta, \tilde{f}^B, \eta, \tilde{g}, t} = \int_t^\infty \mathbb{E}^B_{\delta, \eta, \tilde{f}^B, \tilde{g}} \left[ \mathbb{S}^B u \right] du
$$

$$
= \delta \int_t^\infty e^{-\rho(u-t)} H_f \left( \tilde{f}^B, t, u; \alpha \right) \times [1 - \omega(\eta)]^{1-\alpha}
$$

$$
\times \sum_{j=0}^{-\alpha} C^j_{-\alpha} \left[ \frac{\omega(\eta)}{1 - \omega(\eta)} \right]^j H_g \left( \tilde{g}, t, u; \alpha, \frac{j}{1-\alpha} \right) du.
$$

Following Cox and Huang (1989), the portfolio choice of Group B in terms of equities and bonds can be calculated from Group B’s demand for exposure to $(W^B_{\delta}, W^B_{\eta})$ shocks, which are themselves obtained by multiplying the gradient vector of $B$’s wealth with respect to the four state variables by the diffusion matrix of the four state variables given in Equation (7). If the investors had available elementary securities on the shocks, the exposures would indicate the desired amounts of holdings. If, however, they have access to an equity share and a bond, the investor needs to use these to synthesize the desired exposures. He/she solves for $\theta^T$, a $1 \times 2$ vector, the following system of two equations:

$$
F^B \begin{bmatrix} \frac{\partial F^B}{\partial \tilde{f}^B} & \frac{\partial F^B}{\partial \eta} & \frac{\partial F^B}{\partial \tilde{g}} \end{bmatrix} =
\begin{bmatrix}
\frac{\sigma_{\delta}}{\sigma_{\delta}} & 0 \\
0 & 0 \\
-\eta \frac{\tilde{g}}{\sigma_{\delta}} & 0 \\
\gamma B - \gamma A & -\phi \sigma_f
\end{bmatrix}
$$

$$
= \theta^T \begin{bmatrix} F & 0 \\
0 & 0 \\
0 & -\eta \frac{\tilde{g}}{\sigma_{\delta}} \\
\frac{\gamma B - \gamma A}{\sigma_{\delta}} & -\phi \sigma_f
\end{bmatrix},
$$

(44)
where the left hand-side contains the investor’s target exposures and the right-hand side the exposures of the available securities, which we have analyzed in Section IV. We now study the terms of the left-hand side of the equation.

A. Target Exposures

Defining:

\[
\text{diff}_3 F_B \triangleq \left[ \begin{array}{c}
F_B \\
\frac{\partial F_B}{\partial \beta} \\
\frac{\partial F_B}{\partial \eta} \\
\frac{\partial F_B}{\partial \beta} \\
-\eta \frac{\partial F_B}{\partial \beta}
\end{array} \right],
\]

we can write:

\[
\text{diff} F_B = \text{diff}_3 F_B + \frac{\partial F_B}{\partial \beta} \left[ \begin{array}{c}
\gamma_B - \gamma_A \\
\frac{\gamma_B - \gamma_A}{\sigma_d} \\
-\phi_f 
\end{array} \right].
\]

(46)

Figure 3 about here.

In Figure 3, the components of the exposure strategy as fractions of B’s wealth are drawn against the current difference of opinion, \( \hat{g} \), and against the current weight of the overconfident group A, \( \omega \), the exact same format being used as in Figure 1 for the exposures of equity and the bond.

Three conclusions emerge from the comparison of the three curves in both plots of Figure 3. First, the dotted and the solid line are close to each other:

\[
\text{diff}_3 F_B \bigg|_{\phi=0} \approx \text{diff}_3 F_B \bigg|_{\phi \neq 0}.
\]

(47)

Second, the dashed line is vastly different from the other two. As was the case for equity and for the same reasons, the only components of exposure demands that are markedly affected by the presence of overconfidence are the \( \hat{g} \) components. As has been well explained by Merton (1973), a state variable has an impact on Group B’s wealth for two possible economic reasons: (i) it can change the prospect for the immediate return on a given portfolio held by Group B, and (ii) it can affect the investment opportunities in the future that Group B will face when rebalancing their portfolio. The effects of Type (ii) can be explained by means of the concept of “favorable or unfavorable shift in the investment opportunity set” introduced by Merton (1973), where “a favorable shift” is defined as a change in a state variable such that, for given immediately anticipated returns, consumption rises for a given level of wealth.\(^{30}\) Lemma 1 implies that, in a neighborhood of \( \hat{g} = 0 \), when risk aversion is greater than 1, the derivative \( \partial F_B / \partial \beta \) is nonnegative: an increase in the difference of opinion is an unfavorable shift for Group B (as it is for everyone). Because \( \frac{\gamma_B - \gamma_A}{\sigma_d} \geq 0 \),
the traders with the proper beliefs know that a positive \( dW^B \) shock will increase the difference of opinion (see also Equation (4)). To offset this, they construct their portfolio to have a positive exposure to \( dW^B \). \(^{31}\)

Finally, a comparison of the gap between the dotted and the dashed curves in Figure 3 and in the equity plot of Figure 1 reveals that this gap is smaller in Figure 3. This is the effect of a general result, which is strongly supported by the numerical experiments we have conducted: To each maturity \( T \), Group B desires to have an exposure to difference-of-opinion risk per unit of wealth that is smaller than that contained in one consumption-unit worth of equity:

\[
\frac{1}{F^B,T} \left. \frac{\partial F^B,T}{\partial g^T} \right|_{\hat{g}=0} < \frac{1}{F^T} \left. \frac{\partial F^T}{\partial g^T} \right|_{\hat{g}=0}.
\] (48)

### B. Portfolio Choice

We discuss the cases \( \phi = 0 \) and \( \phi \neq 0 \) separately because, as mentioned above, if everyone has proper beliefs (\( \phi = 0 \)) then of the two risky securities one is redundant. That is, the portfolio equation (44) reduces to:

\[
\begin{bmatrix}
F^B \\
\frac{\partial F^B}{\partial f^B} \\
\frac{\partial F^B}{\partial \eta} \\
\frac{\partial F^B}{\partial g^T}
\end{bmatrix}
= \theta^T_{\phi=0}
\begin{bmatrix}
\frac{\sigma_\delta}{\sigma_\delta} 0 \\
\frac{\gamma_B}{\sigma_\delta} \eta 0 \\
-\eta \frac{\bar{g}}{\sigma_\delta} 0 \\
0 0
\end{bmatrix}
\begin{bmatrix}
F \\
\frac{\partial F}{\partial f^B} \\
\frac{\partial F}{\partial \eta} \\
\frac{\partial F}{\partial g^T}
\end{bmatrix}
\begin{bmatrix}
\frac{\sigma_\delta}{\sigma_\delta} 0 \\
\frac{\gamma_B}{\sigma_\delta} \eta 0 \\
-\eta \frac{\bar{g}}{\sigma_\delta} 0 \\
0 0
\end{bmatrix},
\]

so that the portfolio is indeterminate.

When, in addition to \( \phi = 0 \), there is zero difference of opinion (\( \hat{g} = 0 \)), then both agents are identical. While the portfolio choice remains, of course, indeterminate, a useful "symmetric" benchmark might be a situation in which Group B holds no bonds and holds equity in proportion to its share of wealth: \( \left. \frac{F^B}{F^T} \right|_{\phi=0} = 0 \). \(^{32}\)

Turning now to the case \( \phi \neq 0 \), a few algebraic manipulations\(^{33}\) indicate that the system of equations (44) is equivalent to:

\[
\begin{bmatrix}
F^B \\
\frac{\partial F^B}{\partial f^B} \\
\frac{\partial F^B}{\partial \eta} \\
\frac{\partial F^B}{\partial g^T}
\end{bmatrix}
= \theta^T
\begin{bmatrix}
\frac{\sigma_\delta}{\sigma_\delta} 0 \\
\frac{\gamma_B}{\sigma_\delta} \eta 0 \\
-\eta \frac{\bar{g}}{\sigma_\delta} 0 \\
0 0 1
\end{bmatrix}
\begin{bmatrix}
F \\
\frac{\partial F}{\partial f^B} \\
\frac{\partial F}{\partial \eta} \\
\frac{\partial F}{\partial g^T}
\end{bmatrix}
\begin{bmatrix}
\frac{\sigma_\delta}{\sigma_\delta} 0 \\
\frac{\gamma_B}{\sigma_\delta} \eta 0 \\
-\eta \frac{\bar{g}}{\sigma_\delta} 0 \\
0 0 1
\end{bmatrix},
\]

which can be written:

\[
\begin{bmatrix}
\text{diff}_3 F^B \\
\frac{\partial F^B}{\partial g^T}
\end{bmatrix}
= \theta^T
\begin{bmatrix}
\text{diff}_3 F \\
\frac{\partial F}{\partial g^T}
\end{bmatrix}.
\] (49)
Our principal result follows from this equation:

**Proposition 2:** For as long as \( \phi \neq 0 \), the portfolio choice is independent of the specific value of \( \phi \) except through the value functions \( F, P \) and \( F^B \), and the solution is:

\[
\theta_F = \begin{vmatrix}
\text{diff}_3 \frac{\partial F^B}{\partial g}
\text{diff}_3 \frac{\partial \text{diff}_3 F^B}{\partial g}
\text{diff}_3 \frac{\partial F}{\partial g}
\end{vmatrix}; \quad \theta_P = \begin{vmatrix}
\text{diff}_3 \frac{\partial \text{diff}_3 F^B}{\partial g}
\text{diff}_3 \frac{\partial \text{diff}_3 F}{\partial g}
\text{diff}_3 \frac{\partial \text{diff}_3 F}{\partial g}
\end{vmatrix},
\]

where \( \theta_F \) is the number of units of equity demanded and \( \theta_P \) the number of units of the bond.

We have already noticed a property of the diffusion matrix of state variables: although the difference of opinion is affected by both the output shock and the signal shock, the signal shock has an impact on only the difference-of-opinion state variable, \( \hat{g} \). For that reason, it is possible to redefine the shocks in such a way that the output shock no longer has an impact on the difference of opinion, while the redefined signal shock incorporates both effects on the difference of opinion.\(^{34}\) Then, the value of the parameter cannot matter because it affects equally the exposures of Group \( B \)'s wealth and the exposures of equity and the bond to the redefined signal shock. If \( \phi \) is increased, that causes both the prices of the securities and the wealth of \( B \) to fluctuate more by the same factor. As a result, there is no need to change the number of units of the two securities that are held.

Observe also that Proposition 2 is stated for a given value of the state variable \( \hat{g} \), the behavior of which is vastly affected by a change in \( \phi \). So, what we want to say is that the effect of the overconfidence parameter \( \phi \) is mainly through \( \hat{g} \), the difference of opinion. The time path of portfolio choices would be very much affected by a change in the value of \( \phi \). But, once the effect of \( \phi \) is accounted for in \( \hat{g} \), the remaining effects of \( \phi \) are small.

[Figure 4 about here.]

Figure 4 contains a lot of information about the solution (50). The dashed lines represent the portfolio demands relative to \( B \)'s wealth (portfolio shares), that is, \( \left( \frac{\text{diff}_3 F^F}{\text{diff}_3 F}, \frac{\text{diff}_3 F^P}{\text{diff}_3 F} \right) \) for the case \( \phi = 0.95 \). The solid line represents the limit of (50) when \( \phi \to 0 \). Evidently, these two lines in all four plots are very close to each other. So, even though the proposition rigorously says that the portfolio demand does not depend on the specific value of \( \phi \) “except through the value functions \( F, P, \) and \( F^B \),” we can practically ignore the caveat when examining the fractional composition of the portfolio. The demand for securities depends negligibly on the degree of overconfidence (although equilibrium prices and consumption allocations do depend on it), for as long as it is not equal
to zero. The only thing that matters for $B$’s portfolio demand is whether there exist people in the market who are somewhat overconfident. This conclusion is obviously in line with our earlier remarks about Equations (42) and (47). The modicum of variation in portfolio demand in relation to the value of $\phi$ is mostly due to the sensitivity $\frac{1}{F^T} \frac{\partial F^B}{\partial \eta}$ to sentiment risk. In the bottom plots of Figure 4, the demands are drawn against the consumption share $\omega$ of overconfident investors in the population. Evidently, even a limitingly small presence of overconfident investors in the market causes Group $B$ to follow the portfolio strategy that takes overconfidence into account.

When $\tilde{g} = 0$, the demands for equity and for bonds are:

$$\theta_F = \frac{F^B}{F} \bigg|_{\tilde{g}=0} \left[ \sigma_\delta + \frac{1}{F^T} \frac{\partial F^B}{\partial \eta} \frac{g^B}{\sigma_s} + \frac{1}{F} \frac{\partial F^B}{\partial \eta} \frac{g^B}{\sigma_s} \right] \bigg|_{\tilde{g}=0}; \theta_P = \frac{F^B}{P} \bigg|_{\tilde{g}=0} \left[ \sigma_\delta + \frac{1}{F^T} \frac{\partial F^B}{\partial \eta} \frac{g^B}{\sigma_s} + \frac{1}{F} \frac{\partial F^B}{\partial \eta} \frac{g^B}{\sigma_s} \right] \bigg|_{\tilde{g}=0}.$$  

(51)

Because, for given maturity $T$, $\frac{1}{F^{T,T}} \frac{\partial F^{B,T}}{\partial \eta} = \frac{1}{F^T} \frac{\partial F^T}{\partial \eta} > 0$, (Lemma 1) and since

$$0 < \frac{1}{F^{T,T}} \frac{\partial F^{B,T}}{\partial \eta} \bigg|_{\tilde{g}=0} < \frac{1}{F^T} \frac{\partial F^T}{\partial \eta} \bigg|_{\tilde{g}=0}$$  

(Lemma 1 and Equation (48)), it follows that similar equalities and inequalities hold also but approximately for the integrated prices and thus we have the following result: When $\tilde{g} = 0$, Group $B$ holds a negative amount of bonds and fewer units of equity than the number corresponding to their share of wealth:

$$\theta_F < \frac{F^B}{P} \quad \text{and} \quad \theta_P < 0.$$  

(52)

The reason Group $B$ holds smaller positions than those in the “symmetric benchmark” is that their share of wealth invested in equity and bonds would contain too much difference-of-opinion risk. Risk averse investors $B$ with the proper beliefs are deterred by the presence of the overconfident traders, whose difference of opinion is a source of risk in their eyes. Hence, investors with the proper beliefs prefer to take refuge in the riskless short-term asset, unless they are very optimistic about future growth. Thus, the short-term deposit is the only safe haven from sentiment risk. These results imply that the presence of overconfident investors not only distorts the stock and bond markets, but also scares away rational investors. In the words of De Long et al. (1990a), “noise traders create their own space.”

That property is, of course, illustrated in Figure 4 (top left plot) as are a number of additional features of Group $B$’s demand. The demand schedule for equity is upward sloping as a function of $\hat{g}$. The reason for this is that even though Group $B$ are driven away by the presence of the overconfident traders, they overcome their fear when they are very optimistic about future growth.
At the same time, $B$’s demand schedule for bonds is downward sloping (top right plot). When $\hat{g}$ increases, Group $B$ is increasingly optimistic about the future growth rate of output. Investing in equity, however, also exposes them to future changes in discount rates and in the beliefs of others. Group $B$ uses bonds to hedge these other risks.

VI. On Long-Run Predictability

There exists a logical link between the phenomenon of excessive volatility and the long-run predictability of stock returns. Campbell and Shiller (1988a,b) and Cochrane (2001, page 394 ff), have pointed out that the dividend-price ratio would be constant over time if dividends were unpredictable (specifically, if they followed a geometric Brownian walk) and expected returns were constant. Because the dividend-price ratio is changing, its changes must be predicting either future changes in dividends or future changes in expected returns. This statement is true in any economic model, unless there are violations of the transversality conditions. Empirically, the dividend-price ratio hardly predicts subsequent dividends. It must, therefore, predict returns. But, if it predicts returns, it can serve as valuable information for a rational person trading in the market. We now show how that aspect is embedded in our model and in the equilibrium portfolio strategy that we have just described. As we saw, that strategy was crucially driven by the derivatives of the price functions with respect to the difference of opinion: $\frac{\partial F^B}{\partial g}$, $\frac{\partial F}{\partial g}$, and $\frac{\partial P}{\partial g}$. In essence, we want to produce an interpretation of these derivatives in terms of anticipated returns.

In performing that task, it will be convenient to use Malliavin derivatives. Just as the standard derivative measures the local change of a function to a small change in an underlying variable, the Malliavin derivative measures the change in a path-dependent function implied by a small change in the initial value of the underlying Brownian motions. It provides the “impulse-response function” following a shock in the initial values. In the context of our model, Malliavin calculus allows for a very clean and insightful interpretation of the results, and in particular, allows us to distinguish between instantaneous effects and long-term effects.

Denoting by $D^B_{\delta,t}X_u$ the response at time $u$ of a process $X$ to a unit $dW^B_\delta$ shock having occurred at time $t$ with $u \geq t$, and by $D^B_tX_u = \{D^B_{\delta,t}X_u, D^B_{s,t}X_u\}$ the row vector of responses to the two shocks, $\{dW^B_{\delta,t}, dW^B_{s,t}\}$, we compute below the Malliavin derivatives of the four state variables. Observe that $D^B_tX_t$ is another notation for the diffusion vector of the process $X$ at time $t$. 25
Lemma 3: The Malliavin derivatives for the four state variables are:

\[
\mathcal{D}_t^B \hat{F}_u^B = e^{-\zeta(t-u)} \begin{bmatrix} \frac{\gamma^B}{\sigma_\delta} & 0 \end{bmatrix},
\]

\[
\mathcal{D}_t^B \hat{g}_u = e^{-\psi(u-t)} \begin{bmatrix} \frac{\gamma^B - \gamma^A}{\sigma_\delta} & -\phi \sigma_f \end{bmatrix},
\]

\[
\frac{\mathcal{D}_t^B \delta_T}{\delta_T} = \left[ \sigma_\delta \ 0 \right] + \frac{1}{\zeta} \left( 1 - e^{-\zeta(T-t)} \right) \begin{bmatrix} \frac{\gamma^B}{\sigma_\delta} \ 0 \end{bmatrix},
\]

\[
\frac{\mathcal{D}_t^B \eta_T}{\eta_T} = \left[ \frac{-\hat{g}_u}{\sigma_\delta} \ 0 \right] - \int_t^T e^{-\psi(u-t)} \left( \frac{\hat{g}_u}{\sigma_\delta} du + \frac{dW_{\delta,u}^B}{\sigma_\delta} \right) \begin{bmatrix} \frac{\gamma^B - \gamma^A}{\sigma_\delta} & -\phi \sigma_f \end{bmatrix}.
\]

Observe from Equations (53) and (54) that the responses in \( \hat{F}_u^B \) and \( \hat{g}_u \) follow deterministic paths. From Equations (55) and (56) we see that the perturbations in the fundamental \( \delta \) and the sentiment \( \eta \) accumulate the perturbations in \( \hat{F}_u^B \) and \( \hat{g}_u \); shocks occurring today have a declining effect on future values of the fundamental and the sentiment.\(^{37}\) Given Equations (25), (30) and (35), one can, for instance, calculate the Malliavin derivatives of the discounted price \( F_t^T \) of a single dividend to be paid at time \( T > t \):\(^{38}\)

\[
\mathcal{D}_t^B F_t^T = E_t^B \left[ \frac{\xi^B}{\xi_t^B} \left( (\alpha - 1) \frac{\mathcal{D}_t \delta_T}{\delta_T} + \omega(\eta_T) \frac{\mathcal{D}_t \eta_T}{\eta_T} - (\alpha - 1) \frac{\mathcal{D}_t \delta_T}{\delta_T} - \omega(\eta_t) \frac{\mathcal{D}_t \eta_t}{\eta_t} + \frac{\mathcal{D}_t \delta_T}{\delta_T} \right) \right] \]

\[
= F_t^B \frac{\mathcal{D}_t \delta_t}{\delta_t} + E_t^B \left[ \frac{\xi^B}{\xi_t^B} \delta_T \left[ \omega(\eta_T) - \omega(\eta_t) \right] \frac{\mathcal{D}_t \eta_t}{\eta_t} + \alpha \left\{ \int_t^T \left( \mathcal{D}_t \hat{F}_u^T \right) du \right\} \right]
\]

\[
+ \omega(\eta_T) \left\{ - \int_t^T \left( \mathcal{D}_t \hat{g}_u \right) \left[ \frac{\hat{g}_u}{\sigma_\delta} du + \frac{dW_{\delta,u}^B}{\sigma_\delta} \right] \right\}. \]

Because \( \mathcal{D}_t^B F_t^T \) is the diffusion vector of equity, there is a direct association between the four terms of the Malliavin derivative and the four partial derivatives:

\[
\frac{\partial F_t^T}{\partial \delta} = \frac{F_t^T}{\delta_t}, \quad (58)
\]

\[
\frac{\partial F_t^T}{\partial \hat{F}_u^B} = \alpha F_t^T \frac{1}{\zeta} \left[ 1 - e^{-\zeta(T-t)} \right], \quad (59)
\]

\[
\frac{\partial F_t^T}{\partial \eta} = E_t^B \left[ \delta_T \frac{\xi^B}{\xi_t^B} \omega(\eta_T) \right] - \xi_t^B F_t^T \omega(\eta_t), \quad (60)
\]

\[
\frac{\partial F_t^T}{\partial \hat{g}_u} = -E_t^B \left\{ \omega(\eta_T) \delta_T \frac{\xi^B}{\xi_t^B} \int_t^T e^{-\psi(u-t)} \left[ \frac{\hat{g}_u}{\sigma_\delta} du + \frac{dW_{\delta,u}^B}{\sigma_\delta} \right] \right\}. \quad (61)
\]

Equation (61) in particular, and similar ones written for the bond and the wealth of \( B \), provide us with the promised interpretation of the derivative \( \frac{\partial F_t^T}{\partial \hat{g}_u} \) with respect to the difference of opinion.
it captures the covariation over the entire investment horizon between future output and future changes in the difference of opinion weighted by the future weight of the overconfident population. We have thus identified the relevant statistics that the asset manager of Group B needs to have in mind over the entire future. The explicit solutions for \( F, P \) and \( F^B \) that we have exhibited allow one to forecast the terms in the curly brackets on the right-hand side of (61).

We would like to show now how sentiment risk contributes to anticipated returns in general, irrespective of the specific menu of securities available in the market.

A. Instantaneous Pricing of Risk

By construction (see Cox and Huang (1989)), the instantaneous market price of risk (or Sharpe ratio) is equal to minus the diffusion of the pricing measure. It is the instantaneous response of the stochastic discount factor to shocks occurring today. Knowing the pricing measure (25), Itô’s lemma gives directly the following result (which we state without proof):

**Lemma 4:** In equilibrium, the market prices of risk in the eyes of Group B are:

\[
\frac{\mathcal{D}_t^B \xi_t^B}{\xi_t^B} = - (\kappa_t^B) + (\alpha - 1) \frac{\mathcal{D}_t^B \delta_t}{\delta_t} + \omega (\eta_t) \frac{\mathcal{D}_t^B \eta_t}{\eta_t}
\]

\[
= - \begin{bmatrix} (1 - \alpha) \sigma_\delta & 0 \end{bmatrix} - \hat{g} \omega (\eta) \begin{bmatrix} \frac{1}{\sigma_\delta} & 0 \end{bmatrix}.
\]

(62)

(63)

From Equation (63), we see that the prices of risk \( \kappa_t^B \) contain an instantaneous premium for the output shock \( W_\delta \) but no instantaneous premium for the signal shock \( W_s \). If there is no difference of opinion (\( \hat{g} = 0 \)), the prices of risk \( \kappa_t^B \) include only the traditional reward for fundamental risk \( (1 - \alpha) \sigma_\delta \). As soon as there is a difference of opinion, investors realize that “sentiment” will fluctuate randomly in response to output shocks. Hence, they start charging a premium also for the risk arising from the vagaries of others. The premium is proportional to the product of the difference of opinion \( \hat{g} \) with the relative weight \( \omega \) of the overconfident population.

It is noteworthy that, once the current values of the state variables \( \eta, \hat{f}^B, \hat{g} \) (describing the current population and its expectations) are given, the instantaneous return and risk reward on immediate-maturity instruments do not depend on the degree of overconfidence, \( \phi \), of Group A. The overconfidence coefficient, \( \phi \), affects only the future dynamics of the state variables. For that reason, it has an impact on returns but only for assets maturing beyond the immediate date, a topic to which we turn now.
B. Long-Run Pricing of Risk

The multiperiod rate of return of an asset that delivers a unit payoff at time $T$ in a given state and that was bought at time $t$ ($t < T$) in a given state is $\xi_t / \xi_T$. However, that would be the relevant long-run return if one were to buy that asset at $t$ and hold it until $T$. If there exists a financial market that allows repricing and retrading tomorrow of assets bought today, we show now that this is the not the concept of long-run return that is relevant for portfolio choice. As will be apparent in Equation (67) below, the long-run excess return that is relevant is the long-run response $D_t^B \xi_T^B$ of the stochastic discount factor to shocks occurring today. From Equation (25), we have:

$$
\frac{D_t^B \xi_T^B}{\xi_T^B} = (\alpha - 1) \frac{D_t^B \delta_T}{\delta_T} + \omega(\eta_T) \frac{D_t^B \eta_T}{\eta_T}.
$$

(64)

Notice again the crucial role of the relative weight of the overconfident group in the perturbation of the pricing kernel.

The long-run return has two components: the first arising from the fluctuations in output and the second arising from the vagaries of the overconfident population. The term, $\omega(\eta_T) \frac{D_t^B \eta_T}{\eta_T}$, which would not be present in a market without overconfident investors ($\omega = 0$), is a predictable component and it modifies the long-run behavior of returns. Long-run returns are very much affected by the current values of state variables, a situation commonly referred to in the Finance literature as “return predictability”. In this model, however, return predictability is not as simple as has been commonly envisaged in the empirical Finance literature. For instance, there exists sometimes a positive relation between expected return on equity and the dividend yield and sometimes a negative one. This matter is analyzed by Berrada (2006).

Using (56) to expand $\frac{D_t^B \eta_T}{\eta_T}$ in (64), one can write:

$$
\frac{D_t^B \xi_T^B}{\xi_T^B} = \frac{D_t^B \xi_T^B}{\xi_T^B} + (\alpha - 1) \left\{ \int_t^T \left( D_{t, u}^B \frac{\hat{f}_{u, B}}{\hat{f}_u} \right) du \right\} + \left[ \omega(\eta_T) - \omega(\eta_t) \right] \frac{D_t^B \eta_T}{\eta_T} \\
+ \omega(\eta_T) \left\{ - \int_t^T \left( D_{t, u}^B g_{u} \left[ \frac{\hat{W}_{u, B}^B}{\sigma_{\delta}^2} dW_{\epsilon, u}^B + \frac{dW_{\delta, u}^B}{\sigma_{\delta}} \right] \right) \right\}. \quad (65)
$$

Thus, the long-run price of risk is equal to the short-run price of risk, plus the impact $D_t^B \hat{f}_{u}^B$ of future changes in beliefs of Group $B$, plus the short-run impact of the sentiment weighted by the future change in the weight of the overconfident population, and the impact $D_t^B \hat{g}_{u}$ of future changes in the difference of opinion. Because future consumption is a function of the state price for the future maturity, these four components drive Group $B$’s portfolio strategy.
C. Portfolio Strategy in Terms of Short-Run and Long-Run Returns

Let $F_{t}^{B,T}$ denote the price at date $t$ today of a single-maturity unit of consumption to be consumed at date $T > t$:

$$F_{t}^{B,T} \triangleq E_{t}^{B} \left[ \frac{\xi_{t}^{B} \xi_{T}^{B}}{\xi_{t}^{B}} e_{t}^{B} \right] = (\lambda^{B} e_{t}^{B}) \left[ \frac{1}{\xi_{t}^{B}} \right] = \left( \frac{1}{\xi_{t}^{B}} \right)^{\frac{1}{\alpha - 1}} + 1.$$ \hfill (66)

The Malliavin derivatives of the price can be written:

$$D_{t}^{B} F_{t}^{B,T} = -\frac{1}{1 - \alpha} F_{t}^{B,T} \frac{D_{t}^{B} \xi_{t}^{B}}{\xi_{t}^{B}} + \frac{\alpha}{\alpha - 1} E_{t}^{B} \left[ c_{t}^{B} \xi_{T}^{B} \left( \frac{D_{t}^{B} \xi_{T}^{B}}{\xi_{t}^{B}} - \frac{D_{t}^{B} \xi_{t}^{B}}{\xi_{t}^{B}} \right) \right],$$ \hfill (67)

$$\frac{D_{t}^{B} F_{t}^{B,T}}{F_{t}^{B,T}} = \frac{(\kappa^{B})^{T}}{1 - \alpha} + \frac{\alpha}{\alpha - 1} \frac{1}{F_{t}^{B,T}} E_{t}^{B} \left[ c_{t}^{B} \xi_{T}^{B} \left( \frac{D_{t}^{B} \xi_{T}^{B}}{\xi_{t}^{B}} - \frac{D_{t}^{B} \xi_{t}^{B}}{\xi_{t}^{B}} \right) \right].$$ \hfill (68)

**PROPOSITION 3:** Group B’s portfolio reflects two motivations:

1. The first term of the equation is a myopic portfolio seeking to reap immediate excess return per unit of risk,

2. The second term is an intertemporal hedge which incorporates the prospect of longer-run returns. For each future consumption date $T$, the second term can be further split into:

   (a) A hedge against future shocks to $\tilde{f}_{u}^{B}$, $u \in [t, T]$:

   $$\frac{\alpha}{\alpha - 1} \frac{1}{F_{t}^{B,T}} E_{t}^{B} \left[ c_{t}^{B} \xi_{T}^{B} \left\{ \int_{t}^{T} \left( D_{t}^{B} \tilde{f}_{u}^{B} \right) du \right\} \right];$$

   (b) A myopic exposure to the immediate movement in the sentiment $\eta$, which is also a hedge against future movements in the equilibrium distribution of consumption:

   $$\frac{\alpha}{\alpha - 1} \frac{1}{F_{t}^{B,T}} E_{t}^{B} \left[ c_{t}^{B} \xi_{T}^{B} \left[ \omega(\eta_{T}) - \omega(\eta_{t}) \right] \frac{D_{t}^{B} \eta_{t}}{\eta_{t}} \right];$$

   (c) A hedge against the product of A’s future consumption share $\omega(\eta_{T})$ with the future shocks to $\tilde{g}_{u}$, $u \in [t, T]$:

   $$\frac{\alpha}{\alpha - 1} \frac{1}{F_{t}^{B,T}} E_{t}^{B} \left[ c_{t}^{B} \xi_{T}^{B} \left( \omega(\eta_{T}) \right) \left\{ - \int_{t}^{T} \left( D_{t}^{B} \tilde{g}_{u} \right) \left( \frac{\tilde{g}_{u}}{\sigma_{\delta}^{2}} du + \frac{dW_{\delta,u}^{B}}{\sigma_{\delta}} \right) \right\} \right].$$

As Equations (65) and (67) reveal, the more “strategic” exploitation of the long-run predictability created by overconfident investors is imbedded in the intertemporal hedge. Group B knows that
its share of consumption will fluctuate, that it will revise its expectations of growth, that the other group also will and that it will do so in a manner different from theirs. So, these are the three considerations that are incorporated in when they choose the hedging component of their portfolio.

In contrast to the intertemporal hedging portfolio in Merton (1971), which is expressed in terms of the partial derivatives of the investor’s value function, the three expressions given in Item 2 of Proposition 3 show explicitly how Group B’s expectations of future growth, sentiment, and difference of opinion influence the hedging component of the portfolio. For each of these portfolio components, securities-market implementation requires the simultaneous use of stocks and bonds.

**VII. Epilogue: Survival**

We now return to the question we asked originally concerning the potential for gains that the excessive volatility creates for the investors with the proper beliefs who follow the portfolio strategy that we described in the previous section. By asking whether rational risk arbitrageurs can take advantage of overconfident investors, we simultaneously ask whether investors with the proper beliefs eliminate overconfident investors from the economy very quickly, or whether overconfident investors can survive for some time. The main goal of this exercise is to demonstrate that the phenomena we have analyzed in this article do not go away quickly, contra Alchian (1950) and Friedman (1953).

The survival of irrational traders is an issue that was raised by De Long et al. (1990a, 1991) in a partial-equilibrium setting, in which traders did not affect prices. The survival of excessively optimistic or pessimistic agents, in an economy in which one category of agents knows the true probability distribution, is the focus of recent papers by Kogan et al. (2006) and Yan (2006). Here, however, we consider a different kind of overconfident agents, who change their mind too frequently, being sometimes too optimistic and at other times too pessimistic about the growth rate of aggregate dividends, as compared to investors with the proper beliefs, not as compared to the truth. Kogan et al. (2006) considers agents who consume only at some terminal horizon date so that their saving rate is not optimized and the growth of the economy is not an important factor in the analysis. Yan (2006), in contrast, considers agents who consume intertemporally and make optimal savings decisions. In an economy with a specification that is very close to ours, Berrada (2004) has discussed the issue of survival by means of simulations.

Most previous studies, with the exception of Berrada (2004), have defined “survival” in terms of the “irrational” agents’ asymptotic share of wealth as the horizon goes to infinity. In general, however, wealth is a sufficient summary statistic neither of an agent’s welfare nor of his or her influence on asset prices. Under von Neuman-Morgenstern, time-additive utility, the share of consumption is such a summary statistic. Therefore, we measure the survival of overconfident
agents in the economy by studying the evolution of the share of the total dividend that will be consumed by them. The probability distribution of this share is computed under the objective, or true, probability measure rather than under the measure of either Group A or B.

Group A’s ratio of consumption to aggregate dividends is given by Equation (28). To compute the distribution of this ratio under the true or effective probability measure, we need the conditional distribution of $\eta_\omega$, given $\eta_t, f_t, \hat{f}_t^A$, and $\hat{f}_t^B$ at $t$. As in Section III, we first obtain its characteristic function (see Lemma 5 in the appendix). Then, the expected value and the probability distribution of the share of Group A’s consumption in (28) can be obtained by means of Fourier inversion (as shown in Equation (VIII) of the appendix).

[Figure 5 about here.]

We study the case where Groups A and B start out with the same estimate of the future growth rate, that is, $\hat{f}_t^A = \hat{f}_t^B$ and also consume the same share of aggregate dividend, $\omega_0 = 1/2$. Recall also that we are assuming that both categories of agents have the same, time additive, isoelastic utility function. In the left plot of Figure 5, we plot the probability density function of Group A’s share of consumption after the passage of different numbers of years. We see from the left plot that as time passes, the density moves to the left, and thus, Group A’s share of consumption is decreasing. To understand the rate at which this share is decreasing, in the right plot we plot the expected value of this share, $E_{P_0}[\omega_u]$, against time measured in years. This plot in the right plot considers three cases: one, where Group A has the proper beliefs ($\phi = 0$), the second where Group A is overconfident with $\phi = 0.50$, and the third where it is even more confident with $\phi = 0.95$.

Both plots confirm that ultimately, overconfident agents become extinct in that their share of consumption vanishes. This is simply the result of the fact that $\omega$ is a monotonic function (28) of a positive martingale $\eta$. As is the case for any positive martingale, the probability mass (or the mode of the distribution) accumulates towards zero. Equation (6) implies that

$$
\eta_t = \exp \left[ -\frac{1}{2} \int_0^t \hat{\eta}_u^2 \frac{1}{\sigma_\delta^2} du - \int_0^t \frac{1}{\sigma_\delta} dW^{f}_{\delta,u} \right],
$$

so that this long-run decrease takes place at the rate $\hat{\eta}^2$. But, the more interesting observation is that, in contrast to what is typically assumed in models of rational asset pricing, overconfident agents do not lose out right away. For instance, the right-hand side plot in Figure 5 shows that after 100 years the overconfident agents’ expected share of consumption of the aggregate dividends is still at 20% compared to the initial share of 50%. Recall that in Figure 2 we showed that the relation between the volatility of equity returns and the consumption share of Group A is close to being linear. The fact that the overconfident group is not eliminated from the population instantly implies that the phenomenon of excess volatility will also not be eliminated quickly.
VIII. Conclusion

In a capital market characterized by excessive volatility, we have analyzed the return behavior that would prevail in equilibrium and the trading strategy that would allow a rational investor with the proper beliefs to take advantage of the excess volatility generated by the presence of overconfident investors. To achieve our goal, we have constructed a general-equilibrium “difference-of-opinion” model in which stock prices are excessively volatile, using the device proposed by Scheinkman and Xiong (2003). In our model, there are two groups of agents, and one group (overconfident) believes that the magnitude of the correlation between the innovations in the signal and innovations in some unobserved variable (the expected growth rate of dividends) is larger than it actually is. Consequently, when a signal is received, this group of agents adjusts their beliefs too much and overreacts to it, which then generates excessive stock price movements. The excess movement was regarded as a “sentiment” factor.

For given beliefs, however, both classes of agents are rational in their decision making, in the sense that both are intertemporal optimizers. In this way, the overconfident investors are not sitting ducks. We show that investors with the proper beliefs have to engage in a fairly intricate investment strategy to triumph over the overconfident ones. And their victory can be achieved only in the fairly long run.

We believe that our undertaking brings two benefits. First, given that we have worked out in careful detail the optimal portfolio strategies to be followed by rational investors with the proper beliefs, our model should be of practical use to hedge funds who play the price-convergence game and in so doing expose themselves to “market sentiment” risk. They often have at their disposal perfect-market pricing models that allow them to spot pricing anomalies. But, that is not sufficient information to put into place a “risk arbitrage” strategy, including the optimal timing of trades into the strategy, of trades out of the strategy, plus the accompanying hedges. For that purpose, hedge funds also need a model of the equilibrium stochastic process which describes how sentiment will drive price spreads. We provide one such model.

Second, the model parsimoniously combines the technical virtues of continuous-time, rational-expectations equilibrium asset pricing models (including the use of the martingale approach) with a single, well-defined, almost axiomatic deviation from the case where all agents have the proper beliefs. In this way, it has allowed us to analyze the equilibrium consequences of that specific deviation. We hope that this model, or similar models obtained by this method, can become workhorses in the development of behavioral equilibrium theory.
Appendix

Proof for Lemma 2: Substituting Equations (1) and (14) into (12) and (13) gives the expressions in Equations (15), (16), and (17). The sign for each expression can then be established using the expressions for $\gamma_A$ and $\gamma_B$ in Footnote 11 and by showing that $\gamma_A$ is decreasing in $\phi$, while $\gamma_B$ is equal to $\gamma_A$ evaluated at $\phi = 0$.

Proof for Proposition 1: To prove this proposition, we need to determine:

$$H\left(\delta, \eta, \hat{f}^B, \hat{g}, t, u; \varepsilon, \chi\right) = \mathbb{E}^B_{\delta, \eta, \hat{f}^B, \hat{g}}\left[(\delta_u)^\varepsilon (\eta_u)^\chi\right]. \tag{A1}$$

This function satisfies the linear PDE:

$$0 \equiv \mathcal{L}H\left(\delta, \eta, \hat{f}^B, \hat{g}, t, u; \varepsilon, \chi\right) + \frac{\partial H}{\partial t}\left(\delta, \eta, \hat{f}^B, \hat{g}, t, u; \varepsilon, \chi\right), \tag{A2}$$

with the initial condition $H\left(\delta, \eta, \hat{f}^B, \hat{g}, t, t; \varepsilon, \chi\right) = \delta^\varepsilon \eta^\chi$, and where $\mathcal{L}$ is the differential generator of $\left(\delta_t, \eta_t, \hat{f}^B_t, \hat{g}_t\right)$ under the probability measure of Group $B$.

Spelling out (A2) we have:

$$0 = \frac{\partial H}{\partial \delta} \hat{f}^B - \frac{\partial H}{\partial f^B} \zeta\left(\hat{f}^B - \hat{f}\right) - \frac{\partial H}{\partial \hat{g}} g\left(\zeta + \frac{\gamma^A}{\sigma^2}\right) + \frac{1}{2} \frac{\partial^2 H}{\partial \delta^2} \left(\delta \sigma^2\right)^2 + \frac{1}{2} \frac{\partial^2 H}{\partial \eta^2} \left(\eta \sigma^2\right)^2 + \frac{1}{2} \frac{\partial^2 H}{\partial \hat{g}^2} \left(\frac{\gamma^B - \gamma^A}{\sigma^2}\right)^2 + \phi \sigma^2 \left(\phi \sigma^2\right)^2$$

$$+ \frac{1}{2} \frac{\partial^2 H}{\partial \hat{f}^2} \left(\frac{\gamma^B - \gamma^A}{\sigma^2}\right)^2 - \frac{\partial^2 H}{\partial \delta \partial \eta} \delta \eta \hat{g} + \frac{\partial^2 H}{\partial \delta \partial \hat{g}} \delta \hat{g} + \frac{\partial^2 H}{\partial \delta \partial \hat{f}^B} \delta \gamma^B$$

$$+ \frac{\partial^2 H}{\partial \eta \partial \hat{g}} \eta \hat{g} \left(\frac{\gamma^B - \gamma^A}{\sigma^2}\right) - \frac{\partial^2 H}{\partial \eta \partial \hat{f}^B} \eta \hat{g} \left(\frac{\gamma^B - \gamma^A}{\sigma^2}\right) + \frac{\partial^2 H}{\partial \hat{g} \partial \hat{f}^B} \left(\frac{\gamma^B - \gamma^A}{\sigma^2}\right) \gamma^B + \frac{\partial H}{\partial t}.$$

The solution of this PDE is

$$H\left(\delta, \eta, \hat{f}^B, \hat{g}, t, u; \varepsilon, \chi\right) = \delta^\varepsilon \eta^\chi \times H_f\left(\hat{f}^B, t, u; \varepsilon\right) \times H_g\left(\hat{g}, t, u; \varepsilon, \chi\right), \tag{A4}$$

where $H_f(\hat{f}^B, t, u; \varepsilon)$ and $H_g(\hat{g}, t, u; \varepsilon, \chi)$ are defined in (33) and (34). Substituting (A4) into the PDE and simplifying, we find that the functions of time, $A_1, A_2, B$ and $C$, that are present in (34)
need to solve the following ODEs:

\[
C'(u-t) = aC^2(u-t) - 2bC(u-t) + c, \quad C(0) = 0, \quad (A5)
\]

\[
B'(u-t) = B(u-t) [aC(u-t) - b] + k + le^{-\zeta(u-t)}
+ 2 \left[ m + ne^{-\zeta(u-t)} \right] C(u-t), \quad B(0) = 0, \quad (A6)
\]

\[
A'_1(u-t) = \frac{a}{2} C(u-t), \quad A_1(0) = 0, \quad (A7)
\]

\[
A'_2(u-t) = B(u-t) \left[ m + ne^{-\zeta(u-t)} + \frac{a}{4} B(u-t) \right], \quad A_2(0) = 0, \quad (A8)
\]

where

\[
a = 2 \left[ \left( \frac{\gamma^B - \gamma^A}{\sigma_\delta} \right)^2 + (\phi \sigma_f)^2 \right], \quad (A9)
\]

\[
b = \zeta + \frac{\gamma^A}{\sigma_\delta^2} + \chi \left( \frac{\gamma^B - \gamma^A}{\sigma_\delta^2} \right), \quad (A10)
\]

\[
c = \frac{1}{2} \chi (\chi - 1) \frac{1}{\sigma^2}, \quad (A11)
\]

and

\[
k = -\chi \left[ 1 + \frac{\gamma^B}{\zeta} \frac{1}{\sigma_\delta^2} \right], \quad (A12)
\]

\[
l = \chi \frac{\gamma^B}{\zeta} \frac{1}{\sigma_\delta^2}, \quad (A13)
\]

\[
m = \gamma^B - \gamma^A + \frac{\gamma^B}{\zeta} \left( \frac{\gamma^B - \gamma^A}{\sigma_\delta^2} \right), \quad (A14)
\]

\[
n = -\frac{\gamma^B}{\zeta} \left( \frac{\gamma^B - \gamma^A}{\sigma_\delta^2} \right), \quad (A15)
\]

Of the ODEs (A5)–(A8), all are first-degree linear with constant coefficients, except (A5), which is a Riccati (i.e., quadratic) equation. Radon’s lemma (see, e.g., Cheng and Scaillet (2007)) says that one solution is of the form: \( C(u-t) = \frac{y(u-t)}{x(u-t)} \), where \( y \) and \( x \) satisfy a system of two linear ODEs with constant coefficients. We can, therefore, obtain the solution. Denoting

\[
q = \sqrt{b^2 - ac}, \quad (A16)
\]
and

\[
\begin{align*}
  v_1 &= 0, & \vartheta_1 &= \frac{2cn + k (b + q)}{q}, & (A17) \\
  v_2 &= 2q, & \vartheta_2 &= \frac{2cm + k (b - q)}{q}, & (A18) \\
  v_3 &= \zeta, & \vartheta_3 &= \frac{2cn + l (b + q)}{q - \zeta}, & (A19) \\
  v_4 &= 2q + \zeta, & \vartheta_4 &= \frac{2cn + l (b - q)}{q + \zeta}, & (A20) \\
  v_5 &= q, & \vartheta_5 &= -(\vartheta_1 + \vartheta_2 + \vartheta_3 + \vartheta_4), & (A21)
\end{align*}
\]

we obtain

\[
\begin{align*}
  C (u - t) &= \frac{c \left( 1 - e^{-2q(u-t)} \right)}{q + b + (q - b) e^{-2q(u-t)}}, & (A22) \\
  B (u - t) &= \frac{\sum_{i=1}^{5} \vartheta_i e^{-v_i(u-t)}}{q + b + (q - b) e^{-2q(u-t)}}, & (A23) \\
  A_1 (u - t) &= \frac{a}{2} \int_{t}^{u} C (\tau - t) \, d\tau & (A24) \\
  &= \frac{1}{2} \left[ (b - q) (u - t) + \ln (2q) - \ln \left( q + b + (q - b) e^{-2q(u-t)} \right) \right], \\
  A_2 (u - t) &= \int_{t}^{u} B (\tau - t) \left[ m + ne^{-\zeta(\tau-t)} + \frac{a}{4} B (\tau - t) \right] \, d\tau \\
  &= \sum_{j=1}^{5} \vartheta_j \left[ mD_1 (v_j; u - t) + nD_1 (v_j + \zeta; u - t) \right] + \frac{a}{4} \sum_{i,j=1}^{5} \vartheta_i \vartheta_j D_2 (v_i + v_j; u - t)
\end{align*}
\]

where, denoting by \( \mathcal{H} \) the standard hypergeometric function,

\[
D_1 (p; u - t) = \int_{t}^{u} \frac{e^{-p(\tau-t)}}{q + b + (q - b) e^{-2q(u-t)}} \, d\tau & (A26) \\
= \begin{cases} \\
  \frac{2q(u-t)-\ln(2q)+\ln(q+b+(q-b)e^{-2q(u-t)})}{2q(q+b)} & p = 0, \\
  \frac{1}{p(q+b)} \left[ \mathcal{H} \left( 1, \frac{p}{2q}, 1 + \frac{p}{2q}, \frac{-q-b}{q+b} e^{-2q(u-t)} \right) - e^{-p(u-t)} \mathcal{H} \left( 1, \frac{p}{2q}, 1 + \frac{p}{2q}, \frac{-q-b}{q+b} e^{-2q(u-t)} \right) \right], & p > 0,
\end{cases}
\]

and

\[
D_2 (p; u - t) = \int_{t}^{u} \frac{e^{-p(\tau-t)}}{[q + b + (q - b) e^{-2q(u-t)}]^2} \, d\tau \\
= \frac{1}{2q(q+b)} \left[ \frac{1}{2q} - \frac{e^{-p(u-t)}}{q + b + (q - b) e^{-2q(u-t)}} + (2q - p) D_1 (p; u - t) \right]. & (A27)
\]
To show that the function $H_g(\hat{g}, t, u; \varepsilon, \chi)$ is well defined for $\chi \in [0, 1]$ and $u \geq t$, first note that the radicand in the expression (A16) for $q$ may be written as a quadratic trinomial of $\chi$:

$$b^2 - ac = q_2\chi^2 + q_1\chi + q_0,$$

where

$$q_2 = -\frac{\phi^2 \sigma_f^2}{\sigma_\delta^2},$$

$$q_1 = 2\frac{\phi^2 \sigma_f^2}{\sigma_\delta^2},$$

$$q_0 = \zeta^2 + (1 - \phi^2) \frac{\sigma_f^2}{\sigma_\delta^2}. $$

Because $\phi \in [0, 1]$ and $\zeta > 0$, it is immediate that $q_2 \leq 0$, $q_0 > \zeta^2$, and

$$q_2 + q_1 + q_0 = \zeta^2 + \frac{\sigma_f^2}{\sigma_\delta^2} > \zeta^2. $$

So, when $\chi \in [0, 1]$, $q = \sqrt{b^2 - ac}$ is real. Moreover, $q > \zeta$ so that $\{q_i\}_{i=1}^5$ are finite.

Taking into account that $a \geq 0$, and for $\chi \in [0, 1]$, $c \leq 0$ and

$$b = \chi \sqrt{\zeta^2 + \frac{\sigma_f^2}{\sigma_\delta^2} + (1 - \chi) \sqrt{\zeta^2 + (1 - \phi^2) \frac{\sigma_f^2}{\sigma_\delta^2}}},$$

we obtain that $q - b \geq 0$ and

$$q + b + (q - b) e^{-2q(u-t)} \geq q + b > 0.$$ 

Therefore, when $\chi \in [0, 1]$ and $u \geq t$, the functions $C(u - t)$ and $B(u - t)$ are well defined and bounded; the integrals $A_1(u - t)$ and $A_2(u - t)$ are convergent, and thus, their closed-form expressions (A25) and (A8) are also well defined.

To derive the properties of $H_g$, note that

$$\frac{1}{H_g} \frac{\partial H_g}{\partial \hat{g}}(\hat{g}, t, u; \varepsilon, \chi) = \varepsilon B(\chi; u-t) + 2\hat{g} C(\chi; u-t). $$

Because $\eta_u$ is a martingale under the measure of Group $B$, for $0 \leq \chi \leq 1$, it must be the case by Jensen’s inequality that $\mathbb{E}_B^B \left[ \left( \frac{\eta_u}{\eta} \right)^\chi \right] \leq 1$ for any value of $\hat{g}$. Because from (34)

$$\mathbb{E}_B^B \left[ \left( \frac{\eta_u}{\eta} \right)^\chi \right] = H_g(\hat{g}, t, u; 0, \chi) = \exp \left\{ A_1(\chi; u - t) + \hat{g}^2 C(\chi; u - t) \right\},$$

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this implies $C(\chi; u - t) \leq 0$.

We prove that $B(\chi; u - t)$ is non-positive by contradiction. Assume the function $B(u - t)$ takes positive values. Then because this function is smooth and $B(0) = 0$, there must exist $u_1 \geq t$ such that $B(u_1 - t) = 0$ and $B'(u_1 - t) > 0$. However, it can be shown that the RHS in (A6) is nonpositive at $u_1$. Actually, because $B(u_1 - t) = 0$ the first term in (A6) is equal to zero. Then from (A12) and (A13), $k, k + l \leq 0$ and hence $k + le^{-\zeta(u_1 - t)} \leq 0$. Finally, $m + ne^{-\zeta(u_1 - t)} \geq 0$ because from (A14) and (A15)

\[ m + n = \gamma^B - \gamma^A \geq 0, \]  
\[ m \equiv \frac{a}{4\zeta} \geq 0, \]  

where the last identity can be verified by substituting the expressions for $\gamma^A$ and $\gamma^B$ into (A9) and (A14). Therefore, by contradiction, we have shown that $B(u - t) \leq 0$.

**Proof for Lemma 3:** The solution of Equation (2) is:

\[ \hat{f}_u = \hat{f}_t + (\tau - \hat{f}_t) \times \left[ 1 - e^{-\zeta(u-t)} \right] + \int_t^u \frac{\gamma^B}{\sigma_\delta} dW_{\delta,v}^B, \]

Hence, it follows that:

\[ D_B f_{u} = e^{-\zeta(u-t)} \begin{bmatrix} \gamma^B & 0 \end{bmatrix}. \]  
\[ \text{(A39)} \]

Similarly, the solution of Equation (4) is:

\[ \hat{g}_u = \hat{g}_t \times e^{-\psi(u-t)} + \int_t^u \sigma_{\bar{g}_u} dW_{\delta,v}^B + \sigma_{\bar{g}_u} dW_{s,v}^B, \]

and therefore, it follows that:

\[ D_B g_u = e^{-\psi(u-t)} \begin{bmatrix} \sigma_{\bar{g}_u} & \sigma_{\bar{g}_u} \end{bmatrix}. \]  
\[ \text{(A40)} \]

Because the solution of Equation (1) is:

\[ \delta_T = \delta_t \exp \left[ \int_t^T \left( \frac{\hat{f}_u^B}{2} - \frac{1}{2} \sigma_\delta^2 \right) du + \sigma_\delta \int_t^T dW_{\delta,u}^B \right], \]

it follows that:

\[ D_B \delta_T = \delta_t \times \left\{ \begin{bmatrix} \sigma_\delta & 0 \end{bmatrix} + \int_t^T \left( D_B \hat{f}_u \right) du \right\}, \]
\[ \frac{D_B \delta_T}{\delta_T} = D_B \delta_t + \int_t^T \left( D_B \hat{f}_u \right) du, \]
\[ D_B \delta_T = \delta_t \times \left\{ \begin{bmatrix} \sigma_\delta & 0 \end{bmatrix} + \frac{1}{\zeta} \left[ 1 - e^{-\zeta(T-t)} \right] \begin{bmatrix} \gamma^B & 0 \end{bmatrix} \right\}. \]  
\[ \text{(A41)} \]
Similarly, Equation (6) for \( \eta \) implies:

\[
\eta_t = \exp \left[ -\frac{1}{2} \int_0^t \frac{\hat{g}^2_{\eta}}{\sigma_\delta^2} du - \int_0^t \hat{g}_u \frac{1}{\sigma_\delta} dW^B_{\delta,u} \right],
\]  
(A42)

so that:

\[
D_t^B \eta_t = \eta_T \times \left\{ \left[ -\frac{\hat{g}}{\sigma_\delta} \ 0 \right] - \int_t^T (D_t^B \hat{g}_u) \left[ \frac{1}{\sigma_\delta} du + \frac{dW^B_{\delta,u}}{\sigma_\delta} \right] \right\},
\] 
\[
D_t^B \eta_t = \frac{D_t^B \eta_t}{\eta_t} - \int_t^T (D_t^B \hat{g}_u) \left[ \frac{1}{\sigma_\delta} du + \frac{dW^B_{\delta,u}}{\sigma_\delta} \right],
\] 
(A43)

**Lemma 5:** The moment generating function of \( \eta \) under the effective probability measure is

\[
\mathbb{E}_{\tilde{g}^A, \tilde{g}^B} \left[ \eta_t^\chi \right] = H_P (\tilde{g}^A, \tilde{g}^B, t, u; \chi),
\]  
(A44)

where \( \tilde{g}^A \triangleq \tilde{f}^A - f \), \( \tilde{g}^B \triangleq \tilde{f}^B - f \), and

\[
H_P (\tilde{g}^A, \tilde{g}^B, t, u; \chi) = \exp \left\{ A_P (\chi; u - t) + C^A (\chi; u - t) (\tilde{g}^A)^2 + C^B (\chi; u - t) (\tilde{g}^B)^2 \right\},
\] 
(A45)

for certain functions of time \( A_P, C^A, C^B \), and \( C^{AB} \) that are given in the proof.

**Proof of Lemma 5:** We wish to compute the following expected value:

\[
\tilde{H} (\eta, \tilde{g}^A, \tilde{g}^B, t, u; \chi) = \mathbb{E}_{\eta, \tilde{g}^A, \tilde{g}^B} [\eta_t^\chi].
\]  
(A46)

Under the objective probability measure, the processes \( \eta, \tilde{g}^A \) and \( \tilde{g}^B \) obey the following stochastic differential equations:

\[
d\tilde{g}^A_t = -\tilde{g}^A_t \left( \zeta + \frac{\gamma^A}{\sigma_\delta} \right) dt + \frac{\gamma^A}{\sigma_\delta} dZ^\delta_t + \phi^A f dZ^f_t - \sigma^A f dZ^f_t,
\] 
(A47)

\[
d\tilde{g}^B_t = -\tilde{g}^B_t \left( \zeta + \frac{\gamma^B}{\sigma_\delta} \right) dt + \frac{\gamma^B}{\sigma_\delta} dZ^\delta_t - \sigma^B f dZ^f_t,
\] 
(A48)

\[
\frac{d\eta_t}{\eta_t} = \left( \tilde{g}^B_t - \tilde{g}^A_t \right) \frac{1}{\sigma^2} dt - \left( \tilde{g}^B_t - \tilde{g}^A_t \right) \frac{1}{\sigma_\delta} dZ^\delta_t.
\] 
(A49)

The function \( \tilde{H} (\eta, \tilde{g}^A, \tilde{g}^B, t, u; \chi) \) satisfies the linear PDE:

\[
0 \equiv \mathcal{L} \tilde{H} (\eta, \tilde{g}^A, \tilde{g}^B, t, u; \chi) + \frac{\partial \tilde{H}}{\partial t} (\eta, \tilde{g}^A, \tilde{g}^B, t, u; \chi),
\]  
(A50)
with the initial condition $H (\eta, \hat{g}^A, \hat{g}^B, t, 0; \chi) = \eta^\chi$, where $\mathcal{L}$ is the differential generator of $(\eta_t, \hat{g}^A_t, \hat{g}^B_t)$ under the objective probability measure. Spelling out (A50) using (A47) and (A48), we have:

$$
0 = \frac{\partial \tilde{H}}{\partial \eta} \eta (\hat{g}^B_t - \hat{g}^A_t) \hat{g}^A_t \frac{1}{\sigma^2} - \frac{\partial \tilde{H}}{\partial g^A} \left( \zeta + \frac{\gamma^A}{\sigma^2} \right) \hat{g}^A - \frac{\partial \tilde{H}}{\partial g^B} \left( \zeta + \frac{\gamma^B}{\sigma^2} \right) \hat{g}^B
+ \frac{1}{2} \frac{\partial^2 \tilde{H}}{\partial \eta^2} \left[ \eta (\hat{g}^B_t - \hat{g}^A_t) \right]^2 \frac{1}{\sigma^2} + \frac{1}{2} \frac{\partial^2 \tilde{H}}{\partial (g^A)^2} \left( \frac{(\gamma^A)^2}{\sigma^2} + (\phi \sigma f)^2 + \sigma^2_f \right)
+ \frac{1}{2} \frac{\partial^2 \tilde{H}}{\partial (\hat{g}^B)^2} \left( \frac{(\gamma^B)^2}{\sigma^2} + \sigma^2_f \right) - \frac{\partial^2 \tilde{H}}{\partial \eta \partial g^B} \eta (\hat{g}^B_t - \hat{g}^A_t) \frac{\gamma^A}{\sigma^2} - \frac{\partial^2 \tilde{H}}{\partial g^A \partial \hat{g}^B} \left( \frac{\gamma^A \gamma^B}{\sigma^2} + \sigma^2_f \right) + \frac{\partial \tilde{H}}{\partial t}. \quad (A51)
$$

The appropriate solution of this PDE is

$$
\tilde{H} (\eta, \hat{g}^A, \hat{g}^B, t, u; \chi) = \eta^\chi \times H_P (\hat{g}^A, \hat{g}^B, t, u; \chi), \quad (A52)
$$

where $H_P (\hat{g}^A, \hat{g}^B, t, u; \chi)$ is defined in (A45) and

$$
A_P (u - t) = \int_t^u (\tau - t) \left[ C^A \left( \frac{(\gamma^A)^2}{\sigma^2} + \sigma^2_f \right) + C^B \left( \frac{(\gamma^B)^2}{\sigma^2} + \sigma^2_f \right) + 2C^{AB} \left( \frac{\gamma^A \gamma^B}{\sigma^2} + \sigma^2_f \right) \right] d\tau. \quad (A53)
$$

By Radon’s lemma, the functions of time $C^A, C^{AB}$ and $C^B$ are obtained as elements of the matrix $Z = \begin{pmatrix} C^A & C^{AB} \\ C^{AB} & C^B \end{pmatrix}$, itself defined as follows. Let matrices $X (u - t)$ and $Y (u - t)$ be the unique solution of the linear Cauchy problem

$$
\begin{align*}
\begin{cases}
\dot{X} = Q^{11} X + Q^{12} Y, & X (0) = I, \\
\dot{Y} = Q^{21} X + Q^{22} Y, & Y (0) = 0,
\end{cases}
\end{align*} \quad (A54)
$$

where $I$ is the $2 \times 2$ identity matrix. Then $Z (u - t) = Y (u - t) [X (u - t)]^{-1}$. The coefficients in (A54) are given by:

$$
Q^{21} = \begin{pmatrix}
\frac{1}{2} \chi (\chi - 1) \frac{1}{\sigma^2} & -\frac{1}{2} \chi^2 \frac{1}{\sigma^2} \\
-\frac{1}{2} \chi^2 \frac{1}{\sigma^2} & \frac{1}{2} \chi (\chi + 1) \frac{1}{\sigma^2}
\end{pmatrix}, \quad (A55)
$$

$$
Q^{11} = - (Q^{22})^\top = \begin{pmatrix}
\zeta + (1 - \chi) \frac{\gamma^A}{\sigma^2} & \chi \frac{\gamma^A}{\sigma^2} \\
-\chi \frac{\gamma^B}{\sigma^2} & \zeta + (1 + \chi) \frac{\gamma^B}{\sigma^2}
\end{pmatrix}, \quad (A56)
$$

$$
Q^{12} = \begin{pmatrix}
-2 \left( \frac{(\gamma^A)^2}{\sigma^2} + (\phi \sigma f)^2 + \sigma^2_f \right) & -2 \left( \frac{\gamma^A \gamma^B}{\sigma^2} + \sigma^2_f \right) \\
2 \left( \frac{\gamma^A \gamma^B}{\sigma^2} + \sigma^2_f \right) & -2 \left( \frac{(\gamma^B)^2}{\sigma^2} + \sigma^2_f \right)
\end{pmatrix}, \quad (A57)
$$

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Determining wealth and prices of equity and bond: Knowing the characteristic function (32), the single-maturity claims prices \( E_{\eta, \tilde{f}^B, \tilde{g}} \left[ \frac{E marketed}{} \right] \), \( E_{\eta, \tilde{f}^B, \tilde{g}} \left[ \frac{E marketed}{} \right] \) and \( E_{\eta, \tilde{f}^B, \tilde{g}} \left[ \frac{E marketed}{} \right] \) can be obtained by one of two methods. One is applicable only if \( 1 - \alpha \in \mathbb{N} \). Then the bracket \( \left[ \frac{1}{\lambda^\alpha} + \left( \frac{1}{\lambda^\alpha} \right)^{-1} \right]^{1-\alpha} \) can be expanded into an exact finite series by virtue of the binomial formula as in Equation (35) of the text. The overall calculation in this case is greatly simplified. It leads to the prices of single-maturity claims (36), (37) and (43).\(^1\) The second method is general in that it applies for any value of risk aversion. This method is the inverse Fourier transform, for which the formulae are given below and which can be computed by means of the Fast Fourier Transform:

\[
\begin{align*}
\mathbb{E}^B_{\eta, \tilde{f}^B, \tilde{g}} \left[ \frac{E marketed}{} \right] &= e^{-\rho(u-t)} H_f \left( \tilde{f}^B, t, u; \alpha - 1 \right) \\
&\times \int_0^\infty \left( \frac{1 - \omega(\eta)}{1 - \omega(\tilde{\eta})} \right)^{1-\alpha} \left[ \frac{1}{2\pi} \int_{-\infty}^{+\infty} \left( \frac{\tilde{\eta}}{\eta} \right)^{-i\chi} H_g \left( \tilde{g}, t, u; \alpha - 1, i\chi \right) d\chi \right] \frac{d\tilde{\eta}}{\tilde{\eta}}, \\
\mathbb{E}^B_{\eta, \tilde{f}^B, \tilde{g}} \left[ \frac{E marketed}{} \right] &= e^{-\rho(u-t)} H_f \left( \tilde{f}^B, t, u; \alpha \right) \\
&\times \int_0^\infty \left( \frac{1 - \omega(\eta)}{1 - \omega(\tilde{\eta})} \right)^{1-\alpha} \left[ \frac{1}{2\pi} \int_{-\infty}^{+\infty} \left( \frac{\tilde{\eta}}{\eta} \right)^{-i\chi} H_g \left( \tilde{g}, t, u; \alpha, i\chi \right) d\chi \right] \frac{d\tilde{\eta}}{\tilde{\eta}}, \\
\mathbb{E}^B_{\eta, \tilde{f}^B, \tilde{g}} \left[ \frac{E marketed}{} \right] &= e^{-\rho(u-t)} H_f \left( \tilde{f}^B, t, u; \alpha \right) \\
&\times \int_0^\infty \left( \frac{1 - \omega(\eta)}{1 - \omega(\tilde{\eta})} \right)^{-\alpha} \left[ \frac{1}{2\pi} \int_{-\infty}^{+\infty} \left( \frac{\tilde{\eta}}{\eta} \right)^{-i\chi} H_g \left( \tilde{g}, t, u; \alpha, i\chi \right) d\chi \right] \frac{d\tilde{\eta}}{\tilde{\eta}}.
\end{align*}
\]

Similarly, the share of consumption of Group A under the objective probability measure is:

\[
\mathbb{E}^P_{\eta, \tilde{f}^A, \tilde{g}^B} \left[ \omega(\eta_u) \right] = \int_0^\infty \omega(\tilde{\eta}) \left[ \frac{1}{2\pi} \int_{-\infty}^{+\infty} \left( \frac{\tilde{\eta}}{\eta} \right)^{-i\chi} H_P \left( \tilde{g}^A, \tilde{g}^B, t, u; i\chi \right) d\chi \right] \frac{d\tilde{\eta}}{\tilde{\eta}}.
\]

Finally, we would like to consider only economies where the prices of perpetual claims are finite. Because the prices of perpetual claims involve time integrals of (A58) and (A59) (or (36) and (37)) and these integrals have an infinite upper bound, we need to check for convergence. The conditions for convergence of prices are derived in the lemma below

**Lemma 6**: The growth conditions for the price of perpetual equity, \( F \), and the price of the perpetual bond, \( P \), to be well defined are, respectively:

\[
G(\alpha) < \rho \quad \text{and} \quad G(\alpha - 1) < \rho,
\]

(A61)
where

\[ G(\varepsilon) = \varepsilon \mathcal{I} + \frac{1}{2} \varepsilon(\varepsilon - 1)\sigma_{o}^2 + \frac{\varepsilon^2 \sigma_f^2}{2\zeta^2}. \]  
(A62)

**Proof for Lemma 6**: Because for \( u \geq t \) and \( \chi \in [0, 1] \), \( \left| \frac{b-q}{q+b}e^{-q(u-t)} \right| < 1 \), we can take Taylor series in (A23):

\[
B(u - t) = \frac{\sum_{i=1}^{5} \partial_{i}e^{\psi_{i}(u-t)}}{(q + b)\left(1 - \frac{b-q}{q+b}e^{-2q(u-t)}\right)} = \sum_{i=1}^{5} \frac{\partial_{i}e^{\psi_{i}(u-t)}}{(q + b)} \sum_{j=0}^{\infty} \left[ \frac{b-q}{q+b}e^{-2q(u-t)} \right]^j,
\]

\[
= \sum_{j=0}^{\infty} \left[ h_{j}^{I}e^{-jq(u-t)} + h_{j}^{II}e^{-(jq+\zeta)(u-t)} \right],
\]

(A63)

where \( h_{j}^{I} \) and \( h_{j}^{II} \) are certain functions of \( \chi \) (and independent of time) such that the series in (A63) is uniformly convergent.

Define

\[
\Pi \left( \tilde{f}^B, \tilde{g}, t, u; \varepsilon, \chi \right) = e^{-\rho(u-t)}H_f \left( \tilde{f}^B, t, u; \varepsilon \right) H_g \left( \tilde{g}, t, u; \varepsilon, \chi \right).
\]

(A64)

Because in (34) \( A_1, A_2, C \leq 0 \), we have \( \Pi \leq \overline{\Pi} \), where

\[
\overline{\Pi} \left( \tilde{f}^B, \tilde{g}, t, u; \varepsilon, \chi \right) = e^{-\rho(u-t)}H_f \left( \tilde{f}^B, t, u; \varepsilon \right) \exp \left[ \varepsilon \tilde{g}B(\chi; u - t) \right].
\]

(A65)

Substituting (A63) and (33), we find that the function \( \Pi \) may be written as:

\[
\Pi \left( \tilde{f}^B, \tilde{g}, t, u; \varepsilon, \chi \right) = \pi_0 \left( \tilde{f}^B, \tilde{g}; \varepsilon, \chi \right) \exp \left\{ [G(\varepsilon) - \rho](u - t) + \pi_1(\varepsilon)e^{-2\zeta(u-t)} \right\}
\]

\[
+ \sum_{j=0}^{\infty} \left[ \pi_j^{I}(\tilde{g}; \varepsilon, \chi)e^{-(j+1)q(u-t)} + \pi_j^{II} \left( \tilde{f}^B, \tilde{g}; \varepsilon, \chi \right)e^{-(jq+\zeta)(u-t)} \right],
\]

(A66)

where the series converges uniformly.

Following the line of argument in Brennan and Xia (2001, Theorem 6)\(^{42}\), one can easily show that, when \( \chi \in [0, 1] \), the integrals

\[
\int_{t}^{\infty} \Pi \left( \tilde{f}^B, \tilde{g}, t, u; \varepsilon, \chi \right) du
\]

(A67)

and

\[
\int_{t}^{\infty} \Pi \left( \tilde{f}^B, \tilde{g}, t, u; \varepsilon, 0 \right) du = \int_{t}^{\infty} e^{-\rho(u-t)}H_f \left( \tilde{f}^B, t, u; \varepsilon \right) du
\]

(A68)

are finite if and only if \( G(\varepsilon) - \rho < 0 \).
Assume first that \(1 - \alpha \in \mathbb{N}\). Then (30) and (29) imply that the prices are well defined as long as the integral \(\int_{t}^{\infty} \Pi \left( \tilde{f}^{B}, \tilde{g}, t, u; \varepsilon, \chi \right) du\) is convergent for every \(\chi \in J = \left\{ \frac{j}{1 - \alpha} \right\}_{j=0}^{1 - \alpha}\) and when \(\varepsilon = \alpha\) for the stock and \(\varepsilon = \alpha - 1\). Therefore, convergence of integral (A68) is a necessary condition (because \(\chi = 0 \in J\)) and convergence of integral (A67) is a sufficient condition (because \(\Pi \leq \bar{\Pi}\)) for the prices to be well defined. It only remains to notice that, for each price, these conditions are identical and are given by (A61).

Now, let \(\alpha\) be any real number such that \(1 - \alpha > 0\). Observe that when \(y > 0\), the function \(f(x) = (1 + y^{1/x})^x\) increases with \(x > 0\):

\[
f'(x) = \frac{1}{x} \left(1 + y^{1/x}\right)^{x-1} \left[x \left(1 + y^{1/x}\right) \ln \left(1 + y^{1/x}\right) - y^{1/x} \ln y\right] > \frac{1}{x} \left(1 + y^{1/x}\right)^{x-1} \left[xy^{1/x} \ln \left(y^{1/x}\right) - y^{1/x} \ln y\right] = 0. \tag{A69}
\]

Therefore,

\[
\frac{1}{\lambda^B} < \left[ \left( \frac{\eta u}{\lambda^A} \right)^{\frac{1}{1 - \alpha}} + \left( \frac{1}{\lambda^B} \right)^{\frac{1}{1 - \alpha}} \right]^{1 - \alpha} \leq \left[ \left( \frac{\eta u}{\lambda^A} \right)^{\frac{1}{1 - \alpha}} + \left( \frac{1}{\lambda^B} \right)^{\frac{1}{1 - \alpha}} \right]^{1 - [\alpha]}. \tag{A70}
\]

Replacing the bracket \(\left[ \left( \frac{\eta u}{\lambda^A} \right)^{\frac{1}{1 - \alpha}} + \left( \frac{1}{\lambda^B} \right)^{\frac{1}{1 - \alpha}} \right]^{1 - \alpha}\) with its lower and upper bounds (A70) in the expressions (29) and (30) for securities market prices and applying the above results for the case of \(1 - \alpha \in \mathbb{N}\), we obtain necessary (when the lower bound is substituted; \(\chi = 0\)) and sufficient (when the upper bound is substituted; \(\chi = \frac{j}{1 - \alpha}\), \(j = 0, ..., 1 - [\alpha]\)) conditions for the prices to be well defined. The fact that, for each price, the necessary and sufficient conditions are identical and are given by (A61) completes the proof.
REFERENCES


Footnotes

1 Whether financial-market volatility is actually excessive has been debated. The literature on the equity-premium puzzle has developed a number of models, such as habit-formation models (see Constantinides (1990); Abel (1990); Campbell and Cochrane (1999)), in which the effective discount rate is strongly time varying even though the consumption stream remains very smooth. Using models of that kind, Menzly et al. (2004) have recently calibrated a model of the U.S. stock market in which the volatility of stock returns is larger than the one observed in the data. In another line of investigation, Bansal and Yaron (2004) and Hansen et al. (2005) find that allowing for a small long-run predictable component in dividend growth rates and for very high elasticity of intertemporal substitution can generate several observed asset-pricing phenomenon, including volatility of the market return. Spiegel (1998) shows that excess volatility can be explained also in an overlapping-generations model with small shocks to the supply of the multiple risky assets. Bhamra and Uppal (2008) show that differences in risk aversions can also lead to large volatility.

2 Models of feedback trading do not discuss the budget constraint of the feedback traders and, therefore, leave unclear the origin of the gains that the rational arbitrageurs would make. And, even when noise traders pursue an explicit objective, as is done in De Long et al. (1990a), one must be careful not to confuse “noise risk” with some output risk induced by the noise risk, as has been pointed out recently by Loewenstein and Willard (2006). Restricting our analysis to a pure exchange general equilibrium economy allows us to maintain a clean distinction between output risk and noise risk.

3 The question being answered in our paper is the same as the one that was raised by Williams (1977) and Ziegler (2000) in a simpler setting in which the expected growth rate of dividends is constant (although unobserved) and in which there are fewer securities. In these two papers, the investor whose strategy one is studying is assumed to be of negligible weight in the market, in contrast to our setting.

4 For a survey of the behavioral view of asset pricing, see Shefrin (2005).

5 The agents in Barberis et al. (1998) update their beliefs using Bayes’ formula but they do so on the basis of the wrong prior category of models, which they refuse to update.

6 For a comprehensive study of the influence of heterogeneous beliefs on asset prices, see Basak (2005) and Jouini and Napp (2007).

7 One class of models shows that, with sufficiently deviant priors, Bayesian, rational learning alone can serve to develop theoretical models with volatility that matches the data, by assuming that investors do not know the true stochastic process of dividends (Barsky and De Long (1993)). As investors do not know the expected growth rate of dividends, prices are revised when they receive information about it. These price revisions go beyond the change in the current dividend because the current dividend contains information about future dividends. A similar argument has been made by Bullard and Duffy (2001), Buraschi and Jiltsov (2006), David and Veronesi (2002),
Detemple and Murthy (1994), Duffie et al. (2002), Gallmeyer (2000), Timmermann (1993, 1996), and Veronesi (1999). Brennan and Xia (2001) calibrate a model in which a single type of investors populate the financial market and learn about the expected growth rate of dividends and, separately, about the expected growth rate of consumption. In that model, as in ours, the expected growth rate of dividends is unobservable and needs to be filtered out, which then contributes positively to the volatility of the stock price. They find that they can match several moments of stock returns.

8We emphasize that two state variables, $\hat{g}$ and $\eta$, are needed in our Markovian formulation to capture the dynamics of heterogeneous beliefs and their effects, unlike what happens in, e.g., Barberis et al. (1998). As is noted below, one captures an instantaneous effect and the other a lasting effect.

9Note that the process specified for the signal has only a diffusion term but no drift term. We choose this specification because it makes it easier to interpret the results. For the case where the signal has also a drift term, we refer the reader to earlier versions of this paper.

10Observe once again that output $\delta$ serves as a signal, which causes an update of the growth rate of output, just as the signal $s$ does.

11The steady-state variances of $f$ as estimated by Group A and Group B are, respectively:

$$
\gamma^A \triangleq \sigma^2_\delta \left( \sqrt{\zeta^2 + (1 - \phi^2) \frac{\sigma^2_f}{\sigma^2_\delta}} - \zeta \right), \quad \gamma^B \triangleq \sigma^2_\delta \left( \sqrt{\zeta^2 + \frac{1}{\sigma^2_f} \frac{1}{\sigma^2_\delta}} - \zeta \right).
$$

As has been pointed out by Scheinkman and Xiong (2003), $\gamma^A$ decreases as $\phi$ rises, which is the reason for which Group A is called “overconfident”. $\gamma^A$ starts at the value $\gamma^B$ when $\phi = 0$ and would reach $\gamma^A = 0$ when $\phi \to 1$. The signal can lead Group A ultimately to complete (and foolish) unconditional certainty.

12David (2008) says that the fluctuating difference of measure $\eta$ between the two groups makes the market “effectively incomplete”. That is a matter of semantics. Analytically, the equilibrium can be obtained by complete-market methods. It would probably be more descriptive of the analytical structure that is reflected in Equation (21) below, to say that the fluctuating $\eta$ causes the utility function of agents $A$ to become “effectively state dependent” (i.e. non von Neuman-Morgenstern) relative to the probability measure of Group B. See Riedel (2001).

13We could also have defined $\xi^A$, the density that makes prices martingales under $A$’s probability measure. For any event $e$: $E^A [\xi^A 1_e] = E^B [\eta \xi^A 1_e] = E^B [\xi^B 1_e]$, which implies that $\xi^B = \eta \xi^A$. The martingale pricing density would then be defined relative to each agent’s probability measure. But the risk neutral measure is the same in the end.

14For arbitrary (but von Neuman Morgenstern) utility, $\omega$ would be defined as the ratio of Group A’s absolute risk tolerance over the sum of absolute risk tolerances of Group A and Group B.
See Lintner (1969) and Basak (2005). In the isoelastic case, this ratio reduces to the share of consumption.

15 Along any sample path of the economy, $\omega(\eta)$ is monotonically increasing with $\eta$. Thus, we can use $\omega$ as a representation of $\eta$. As state variable, $\eta$ is equivalent to the consumption shares of the two subpopulations.

16 For arbitrary (but von Neuman Morgenstern) utility, the curvature of $\xi_B$ with respect to the fundamental $\delta$ is equal to total absolute risk aversion multiplied by total absolute prudence. See Basak (2005).

17 For arbitrary (but von Neuman Morgenstern) utility, Basak (2005) shows that the curvature of the $\xi_B$ with respect to $\eta$ is given by a combination of the risk aversions and the prudences of the two groups. One can verify on his formula that the knife-edge case of zero curvature is the case in which both groups have log utility. A special case of his result is obtained here for isoelastic utility.

18 The “growth conditions” required to guarantee that the time integrals in the case of perpetual securities converge (and that the interchange of the integration and expectation operators in (29) and (30) is allowed) are provided in Lemma 6 in the appendix. Similar to the way it has been done in Brennan and Xia (2001), one can show that the condition for the equity price to converge requires that the long-run growth rate of aggregate dividends be less than the long-run risk free rate. It is worth noting that this condition depends only on the anticipated behavior of dividends, $\mathbb{E}_B^B[(\delta g)^{\varepsilon}]$, and not on the anticipated behavior of sentiment, $\mathbb{E}_B^B[(\eta g)^{\chi}]$, which implies that it is independent of the heterogeneity of agents in the economy.

19 That would be done in our context simply by introducing an additional state variable: $Z \triangleq \hat{g}^2$.

20 Yan (2006) uses a similar approach. Recall that $1 - \alpha > 0$. The parameter $\chi$ in the function $H_g$ takes values ranging from 0 to 1.

21 To obtain the bond price, the parameter $\varepsilon$ in the $H_g$ function is set at $\alpha - 1$, which is negative unambiguously.

22 To obtain the stock price, the parameter $\varepsilon$ in the $H_g$ function is set at $\alpha$, which is negative when risk aversion is greater than 1.

23 The range of parameter values that can be considered is restricted by the need to satisfy the growth conditions in Lemma 6, so that the prices of perpetual assets (equity and consol bond) are well defined. This limits, in particular, the range of values for the discount rate, or for risk aversion, that can be considered. Because of this constraint, the risk aversion we consider is somewhat too low to account for the equity premium by itself. The presence of overconfident traders, however, suffices to bring the equity premium up to realistic levels.
We have not set the volatility of the signal, $\sigma_s$, at any particular number because that is immaterial and what matters is only the signal’s correlation with the expected rate of growth of the fundamental.

When we vary the parameter $\phi$, we adjust the ratio $\frac{\lambda_B \eta}{\lambda_A}$ in such a way that the time-0 lifetime budget constraints of the two groups still hold, with unchanged time-0 endowments of securities.

Prior to integrating the present values of payoffs over future times, these logarithmic derivatives would be exactly independent of $\phi$. The value of $\phi$ affects only the relative weighting of the future payoffs.

This is in part the result of the unambiguous negative effect of an increase in $\hat{f}^B$ on the price of equity for the case where $\alpha < 0$. However, a fundamental shock $dW^B_\delta$ has an effect also on the other state variables.

The values produced by the model for the volatility of bond returns (and interest rates) are regrettably too high to fit real-world data. With a risk aversion smaller than 1, David (2008) was able to match the volatility of interest rates much better. Alternatively, if one wanted to match interest-rate volatility, one could introduce habit formation.

Again, Formula (43) applies only when risk aversion is an integer greater than zero, therefore, at least equal to 1. Hence it applies only for $\alpha < 0$. The parameter $\varepsilon$ of the characteristic function is set at $\alpha$ and the parameter $\chi$ takes values ranging from 0 to $-\frac{\alpha}{1-\alpha} > 0$. The latter is a positive rational number smaller than 1.

For instance, because $\varepsilon$, in the calculation of $F^{B,T}$ (for a single-maturity claim) is set at $\alpha < 0$, Lemma 1 implies that:

$$\frac{1}{F^{B,T}} \frac{\partial F^{B,T}}{\partial \hat{f}^B} = \frac{1}{F^T} \frac{\partial F^T}{\partial \hat{f}^B} < 0,$$

and therefore:

$$\frac{1}{F^B} \frac{\partial F^B}{\partial \hat{f}^B} < 0; \frac{1}{F} \frac{\partial F}{\partial \hat{f}^B} < 0.$$

An increase in their estimate of growth is a favorable shift for Group B (as it is for everyone): their wealth decreases, their consumption increases.

Because $-\phi \sigma_f \leq 0$, the opposite is true for the signal shock.

Another possibility would be to let the equilibrium portfolio be the limit as $\phi \to 0$ (from above or from below) of the portfolio applicable when $\phi \neq 0$ (see below Equation (50)). That limit would just be equal to the value of (50) calculated at $\phi = 0$. It will soon be apparent in which way that portfolio would differ from the “symmetric” benchmark.
These algebraic manipulations are equivalent to postmultiplying both sides of Equation (44) by the matrix:

$$\begin{bmatrix}
\gamma B - \gamma A \\
\sigma \phi \sigma_f
\end{bmatrix}.$$

The redefined vector of shocks is:

$$\begin{bmatrix}
\frac{1}{\gamma} \\
\frac{1}{\sigma \phi \sigma_f}
\end{bmatrix} \begin{bmatrix}
dW_B \\
dW_s
\end{bmatrix}$$

the 2 × 2 matrix above being the inverse of the matrix in Footnote 33. The two shocks are no longer independent of each other but that is not important for a portfolio-composition equation that applies to wealth exposures to both shocks.

For recent work studying long run risk and return, see Alvarez and Jermann (2005) and Hansen and Scheinkman (2005).

An introduction to Malliavin calculus with applications to problems in finance can be found in Detemple and Zapatero (1991, Appendix A) and Detemple et al. (2003, Appendix D). For additional details on Malliavin calculus, see Ocone and Karatzas (1991), Nualart (1995), and Øksendal (1997).

The time-integral in the right-hand side of (56) contains terms that interact the effect of current shocks \(D_t \widehat{g}_{u}\) with future shocks \(\frac{dW^B_{\delta, u}}{\sigma}\), because the current shocks have an impact on the diffusion \(\widehat{g}_{u}\) that is applied to future shocks to get the diffusion of \(\eta_u\).

Here we follow closely Detemple et al. (2003), but using our four state variables, whereas they consider shocks to the pricing kernel, the rate of interest and the prices of risk.

The risk-neutral measures for Groups A and B differ only in the market prices of risk:

$$\begin{bmatrix}
\kappa^A \\
\kappa^B
\end{bmatrix} = \begin{bmatrix}
(1 - \alpha) \sigma \delta \\
\frac{1}{\sigma \delta}
\end{bmatrix} \begin{bmatrix}
\widehat{g}_{u} [1 - \omega (\eta)] \\
- \phi \sigma_f
\end{bmatrix}.$$

To see that \(H_f\), which is defined in (33), is the moment generating function for \(\delta_u / \delta\) under Group B’s measure, one can verify that \(\delta^\varepsilon H_f(\widehat{f}^B, t; u, \varepsilon)\) solves the PDE in (A4) but with all terms that are the partial derivatives of \(H\) with respect to either \(\eta\) or \(\widehat{g}\) dropped.

The \(\chi\) argument belongs to \([0, 1]\) allowing us to apply Lemma 1 to conclude that \(H_{\delta}\) is well defined.

To show that for integral (A67), one should additionally note that, because \(q, \zeta > 0\) and the series in (A66) is uniformly convergent, it tends to zero when \(u \to \infty\).
Table I
Choice of parameter values and benchmark values of the state variables

This table lists the parameter values used for all the figures in the paper. These values are similar to the estimation results reported in Brennan and Xia (2001). The table also indicates the benchmark values of state variables, which are the reference values taken by all state variables except for the particular one being varied in a given graph.

<table>
<thead>
<tr>
<th>Name</th>
<th>Symbol</th>
<th>Value</th>
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</thead>
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<tr>
<td><strong>Parameters for aggregate endowment and the signal</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Long-term average growth rate of aggregate endowment</td>
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</tr>
<tr>
<td>Volatility of expected growth rate of endowment</td>
<td>( \sigma_f )</td>
<td>0.03</td>
</tr>
<tr>
<td>Volatility of aggregate endowment</td>
<td>( \sigma_\delta )</td>
<td>0.13</td>
</tr>
<tr>
<td>Mean reversion parameter</td>
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</tr>
<tr>
<td><strong>Parameters for the agents</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Agent A’s correlation between signal and mean growth rate</td>
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</tr>
<tr>
<td>Agent B’s correlation between signal and mean growth rate</td>
<td>—</td>
<td>0</td>
</tr>
<tr>
<td>Agent A’s initial share of aggregate endowment</td>
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</tr>
<tr>
<td>Time-preference parameter for both agents</td>
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</tr>
<tr>
<td>Relative risk aversion for both agents</td>
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</tr>
<tr>
<td><strong>Benchmark values of the state variables</strong></td>
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</tr>
<tr>
<td>The level of aggregate dividends</td>
<td>( \delta )</td>
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</tr>
<tr>
<td>The change from B’s measure to A’s measure</td>
<td>( \eta )</td>
<td>1</td>
</tr>
<tr>
<td>The population average belief about expected rate of growth</td>
<td>( \hat{f}^B )</td>
<td>( \hat{f} )</td>
</tr>
<tr>
<td>The difference in opinions: ( \hat{f}^B - \hat{f}^A )</td>
<td>( \hat{g} )</td>
<td>-0.03</td>
</tr>
</tbody>
</table>
Figure 1
Diffusion of equity and bond
The figure has four plots. The top two show the diffusion of equity and bond rates of return with respect to the output shock as functions of the current difference of opinion, $\hat{g}$, for a value of sentiment such that $\omega = 0.5$. The bottom two plots show the same as functions of $\omega$ for a value of the difference of opinion $\hat{g} = -0.03$. Each plot in the figure has three curves on it, with the dotted line representing the case where all agents have the proper beliefs ($\phi = 0$), the dashed line representing the case where Group $A$ is overconfident ($\phi = 0.95$) and the solid line representing the construct “diff$_3$” containing three terms only, Group $A$ being, however, overconfident. All other parameter values used in this figure are given in Table I.
Figure 2
Volatilities of equity and bond returns and their correlation
The first column of this figure plots against dispersion in beliefs, $\hat{g}$, the volatilities of equity and bond returns and their correlation, assuming that the two groups of investors have equal weight, $\omega = 0.5$. In the second column of this figure, the same quantities are plotted but now against the relative weight of Group A in the population, $\omega$, assuming that $\hat{g} = -0.03$. There are two curves in each plot: the dotted curve is for the case of proper beliefs ($\phi = 0$) and the dashed curve is for overconfident beliefs ($\phi = 0.95$). All other parameter values used in this figure are given in Table I.
Figure 3
Diffusion of wealth
The figure has two plots. The one on the left shows the diffusion of the wealth of Group $B$ with respect to output shocks as a function of the current difference of opinion, $\hat{g}$, for a value of sentiment such that $\omega = 0.5$. The right-hand plot shows the same as a function of $\omega$, for a value of the difference of opinion $\hat{g} = -0.03$. Each plot in the figure has three curves on it, with the dotted line representing the case where all agents have the proper beliefs ($\phi = 0$), the dashed line representing the case where Group $A$ is overconfident ($\phi = 0.95$), and the continuous line representing the construct “diff$_3$” containing three terms only, Group $A$ being, however, overconfident. All other parameter values used in this figure are given in Table I.
Figure 4

Portfolio choice

The figure has four plots. The top two show Group B’s demand for equity and bond as a fraction of their wealth against the current difference of opinion $\hat{g}$ for a value of sentiment such that $\omega = 0.5$. The bottom two plots show the same against $\omega$ for a value of the difference of opinion $\hat{g} = -0.03$. Each plot in the figure has two curves on it, with the dashed line representing the case where Group A is overconfident (with $\phi = 0.95$) and the solid line representing the limit of the latter when $\phi \to 0$. All other parameter values used in this figure are given in Table I.
Figure 5
Survival of the overconfident Group A
The plot on the left gives the probability density function (pdf) of Group A’s share of consumption, $\omega_u$, after the passage of $u$ years. The plot on the right shows the expected value under the objective measure of Group A’s consumption share, $E^P_0 \omega_u$, as a function of time measured in years, with current time assumed to be 0 and future time denoted by $u$ on the $x$-axis. In the second plot, the dotted line represents the case where $\phi = 0$ and all agents have the correct beliefs, while the solid line represents the case where $\phi = 0.50$ and the dashed line represents the case where $\phi = 0.95$ implying that Group A is overconfident. The parameter values used here are given in Table I. In particular, the two groups of investors have equal initial weights, $\omega_0 = 1/2$. 