The Impact of Fat Tails on Equilibrium Rates of Return and Term Premia

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Abstract: We investigate the impact of ignoring fat tails observed in the empirical distributions of macroeconomic time series on the equilibrium implications of the consumption-based asset-pricing model with habit formation. Fat tails in the empirical distributions of consumption growth rates are modeled as a dampened power law process that nevertheless guarantees finiteness of moments of all orders. This renders model-implied mean equilibrium rates of return and equity and term premia finite. Comparison with a benchmark Gaussian process reveals that accounting for fat tails lowers the model-implied mean risk-free rate by 20 percent, raises the mean equity premium by 80 percent and the term premium by 20 percent, bringing the model implications closer to their empirically observed counterparts.

Key phrases: asset-pricing model; habit formation; term premium; equity premium; fat tails; dampened power law

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1. INTRODUCTION

The empirical distributions of many economic and financial time series exhibit fat tails. This has been well documented in the literature. The presence of fat tails warrants the use of probability distributions that can accommodate the likelihood of large positive or negative shocks impacting the economy. Gaussian distributions do not admit this possibility. However, these distributions are well understood and analytically tractable. This explains their pervasive use across macroeconomics and finance.

Non-Gaussian fat-tailed distributions have been used to some extent, especially in models of asset pricing. In such studies, the underlying asset price is typically assumed to either follow a Gaussian distribution supplemented by features such as stochastic volatility and/or jumps (thus making the resulting distribution non-Gaussian), or simply assumed to follow a non-Gaussian distribution such as an $\alpha$-stable distribution. However, one reason why departures from a Gaussian distribution are not more common, despite its documented deficiencies, is the added complexity concomitant with such departures. Also, often in economic models, the use of many non-Gaussian distributions precludes the possibility of finding exact analytical solutions to equilibrium quantities of interest.

One may then ask what the cost of ignoring fat tails in economic model building would be. One answer is available from the options pricing literature where the assumption of normality is relaxed to accommodate the possibility of large movements in prices of underlying assets. In this vein, Hales (1997) finds that an options pricing model

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1 Studies documenting fat tails in macroeconomic series include, among others, Blanchard and Watson (1986), Balke and Fomby (1994), and Kiani and Bidarkota (2003).
with $\alpha$-stable distributions that capture fat tails due to McCulloch (1987) reduces pricing biases relative to the Gaussian Black-Scholes (1973) model for valuing foreign currency options. A second answer is available from the asset allocation literature. Here, Tokat et al. (2003) find that the optimal allocation of wealth between risk-free and risky assets could be up to 26 percent different when one accounts for fat tails in the empirical distributions of underlying data. In an asset-pricing context, Bidarkota and McCulloch (2003) find that accounting for fat tails in the dividends data generates an additional 13 percent of equilibrium equity returns in the standard consumption-based asset-pricing model.

In this study we seek to provide another answer to the question on the economic costs of ignoring fat tails in economic models. We provide an answer in the context of the popularly used consumption-based asset-pricing model of Lucas (1978), augmented with a habit-formation feature as in Abel (1999). We study two versions of the model – one in which exogenous consumption stochastically evolves as a non-Gaussian process exhibiting fat tails and the benchmark version in which consumption evolves as a Gaussian process. We calculate the model-implied equilibrium rates of return, the equity and the term premia in the two versions and compare them to evaluate the economic impact of modeling fat tails.

We model fat tails in this paper with the Dampened Power Law (DPL), recently utilized by Wu (2004) to examine the tail behavior of financial security returns and option prices. DPL nests $\alpha$-stable distributions but, unlike these, has the advantage that

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2 Outside the context of an explicit economic model, the danger inherent in ignoring fat tails in empirical distributions in economic analysis is illustrated in studies such as those on value-at-risk measures (Khindanova et al., 2001).
all moments are finite under strictly positive dampening. This renders model-implied
rates of return and equity and term premia finite under certain restrictions.

The paper is organized as follows. We set out the asset-pricing model in section 2.
In section 3, we specify the stochastic process that accounts for fat tails, discuss an
estimation method, and then specialize solutions to equilibrium quantities of interest
implied by the asset-pricing model to the postulated stochastic structure. In section 4, we
report maximum likelihood estimates of the model with data from the US, and calculate
the model-implied equilibrium rates of return and equity and term premia. In section 5,
we conclude with the main observations derived from our study.

2. THE ASSET-PRICING MODEL

In this section we provide a description of the asset-pricing model due to Abel
(1999) that forms the basis for our study. We specify preferences, define a canonical
asset, note the stochastic structure, outline the key steps for solving for equilibrium asset
prices, and define the rate of return on the canonical asset and the term premium. The
content of this section is largely derived from Abel (1999).

2.1 Preferences

The model economy is populated by a continuum of identical infinitely-lived
agents. It is a closed economy, producing a single completely perishable output. Thus,
consumption in every period must equal output.

A representative consumer maximizes expected lifetime utility given by:

\[ U_t = \mathbb{E}_t \left\{ \sum_{j=0}^{\infty} \frac{1}{(1 + \delta)^j} u(c_{t+j}, y_{t+j}) \right\} \]  (1)
where \[ u(c_t, v_t) = \frac{1}{1-r} \left( \frac{c_t}{v_t} \right)^{1-r}, \quad r > 0, \quad \delta > 0. \] (2)

Here, \( r \) is the coefficient of relative risk aversion (CRRA) and \( v_t \) is a benchmark level of consumption assumed exogenous to the individual consumer.

The benchmark level of consumption is assumed to depend on the aggregate per capita level of consumption \( C_t \) as follows:

\[ \nu_t = C_t^{h_0} C_{t-1}^{h_1} \left( G^t \right)^{h_2} \] (3)

where \( G \geq 1 \) and \( 0 \leq h_i \leq 1 \) for \( i = 0,1,2 \). Setting \( h_1 > 0 \) and \( h_0 = h_2 = 0 \) produces the ‘catching up with the Joneses’ utility specification of Abel (1990).

The intertemporal marginal rate of substitution (IMRS) between period \( t \) and period \( t+1 \) is given by:

\[ M_{t+1} = \frac{1}{1+\delta} \frac{u_c(c_{t+1}, v_{t+1})}{u_c(c_t, v_t)} \] (4)

Using equations (2) and (3) and recognizing that, in equilibrium,

\[ x_{t+1} = \frac{C_{t+1}}{C_t} = \frac{c_{t+1}}{c_t} \] (5)

the IMRS can be written as:

\[ M_{t+1} = \beta x_{t+1} x_t^{\theta} \] (6a)

where

\[ \beta = \frac{1}{1+\delta} G^{h_2(r-1)} > 0 \] (6b)

\[ A \equiv r(l-h_0) + h_0 > 0 \] (6c)

and

\[ \theta = h_1(r-1). \] (6d)

The equilibrium asset prices and returns depend on the IMRS.
In the model, there are six preference parameters - $\delta, r, h_0, h_1, h_2,$ and $G$ - and all six parameters determine the IMRS as is evident from the equations above. However, there are only three independent parameters - $\beta, A,$ and $\theta$ - that determine the IRMS as given in equation (6a).

2.2 The Canonical Asset

Abel (1999) introduces a canonical asset that includes fixed income securities of all maturities and equities as special cases. The canonical asset is an $n$-period asset, with the current period indexed by $t$ and terminal period by $t + n$. This asset pays $a_j y_{t+n-j}^\lambda$ in the period that is $j$ periods before the terminal period for $j = 0, \ldots, n-1$, where $y_{t+n-j} > 0$ is a random variable, $a_0 > 0$ is a constant, and $a_j \geq 0, j = 1, \ldots, n-1$ are constants. The parameter $\lambda$ takes the value zero for fixed income securities and one for equities.

Thus, the payoff for fixed-income securities in period $t + n - j$ is the known amount $a_j$. In the Lucas (1978) fruit-tree model, the dividend (per capita) on equity equals consumption per capita $C_t$. In terms of the canonical asset, this equity can be represented with $n = \infty$, $a_j = 1$ for all $j \geq 0$, and $y_t = C_t$.

Let $p_t(n, \lambda)$ denote the ex-payment price of the canonical $n$-period asset in period $t$. The dependence of this price on the sequence of constants $a_j, j = 0, \ldots, n-1$ and on the stochastic process for $y_t$ is suppressed for notational convenience. The gross rate of return on the canonical asset between period $t$ and period $t+1$ is given by:
\[
R_{t+1}(n, \lambda) = \frac{p_{t+1}(n-1, \lambda) + a_{n-1}y_{t+1}^\lambda}{p_t(n, \lambda)}, \quad \text{for } n \geq 1. \tag{7}
\]

2.3 The Stochastic Structure

The payoff growth rates \(z_{t+1} \equiv y_{t+1} / y_t\) and the consumption growth rates \(x_{t+1} \equiv C_{t+1} / C_t\) observable at the beginning of period \(t + 1\) are assumed throughout this paper to follow i.i.d. processes.

2.4 Asset Prices

Asset prices are postulated to be given by:

\[
p_t(n, \lambda) = \omega(n, \lambda)x_t^\theta y_t^\lambda
\]

where \(\omega(n, \lambda)\) is a function to be determined. The first order condition for utility maximization in this model is as follows:

\[
E_t\{R_{t+1}(n, \lambda)M_{t+1}\} = 1 \tag{9}
\]

Substituting equation (8) in equation (7) and the resulting expression for the gross rate of return into the first order condition above yields the following difference equation in \(\omega(n, \lambda)\) under the assumption that \(z_{t+1}\) and \(x_{t+1}\) follows i.i.d. processes:

\[
\omega(n, \lambda) = \kappa(\lambda)\omega(n-1, \lambda) + \beta a_{n-1}E_x\left\{x_t^{-A}z_{t+1}^\lambda\right\} \tag{10a}
\]

where

\[
\kappa(\lambda) = \beta E_x\left\{x_t^{-A}z_{t+1}^\lambda\right\}. \tag{10b}
\]

Throughout this paper, we assume as in Abel (1999) that \(0 < \kappa(\lambda) < 1\). This assumption guarantees that the difference equation (10a) converges as \(n\) grows.
The fact that the price of a zero-period asset has to be zero, i.e. that \( p_t(0, \lambda) = 0 \), provides the boundary condition for solving the difference equation (10a). Using this boundary condition and equation (8) in equation (10a) yields:

\[
\omega(1, \lambda) = \beta a_0 E \left\{ \frac{\lambda}{x_{t+1}^{\lambda} z_{t+1}^{\lambda}} \right\} > 0
\]  \hfill (11)

The solution to the difference equation with the above boundary condition, as can be easily verified, is given by:

\[
\omega(n, \lambda) = \frac{\omega(1, \lambda)}{a_0} \sum_{i=1}^{n} a_{i-1} [\kappa(\lambda)]^{n-i}.
\]  \hfill (12)

### 2.5 Expected Rate of Return on the Canonical Asset

Starting from equation (7), the expected rate of return on the one-period canonical asset can be shown to be:

\[
E[R_{t+1}(1, \lambda)] = \frac{E \left\{ \frac{\lambda}{z_{t+1}^{\lambda}} \right\}}{\beta E \left\{ \frac{\lambda}{x_{t+1}^{\lambda} z_{t+1}^{\lambda}} \right\}} = \frac{\theta + (1 - \Psi) a_{n-1} \omega(1, \lambda)}{\omega(n, \lambda)} E\{R_{t+1}(1, \lambda)\}
\]  \hfill (13)

The expected rate of return on an n-period canonical asset can then be shown to be:

\[
E[R_{t+1}(n, \lambda)] = \left[ \Psi + (1 - \Psi) a_{n-1} \frac{\omega(1, \lambda)}{\omega(n, \lambda)} \right] E\{R_{t+1}(1, \lambda)\}
\]  \hfill (14a)

where

\[
\Psi = \frac{E \left\{ \frac{\lambda}{x_{t+1}^{\lambda} z_{t+1}^{\lambda}} \right\}}{E \left\{ \frac{\lambda}{z_{t+1}^{\lambda}} \right\}} \frac{E \left\{ \frac{\lambda}{x_{t+1}^{\lambda} z_{t+1}^{\lambda}} \right\}}{E \left\{ \frac{\lambda}{x_{t+1}^{\lambda} z_{t+1}^{\lambda}} \right\}}
\]  \hfill (14b)

### 2.6 Term Premia

The term premium on an n-period asset is defined as:

\[
TP(n, \lambda) = \frac{E[R_{t+1}(n, \lambda)]}{E[R_{t+1}(1, \lambda)]} - 1
\]  \hfill (15)
Using equation (14a), this becomes:

\[ TP(n, \lambda) = (\Psi - 1)\Gamma(n, \lambda) \]  \hspace{1cm} (16a)

where

\[ \Gamma(n, \lambda) \equiv 1 - \frac{\alpha_{n-1}}{\alpha_0} \frac{\omega(1, \lambda)}{\omega(n, \lambda)} = 1 - \frac{a_{n-1}}{\sum_{i=1}^{n} a_{i-1}} \frac{1}{\sum_{i=1}^{n} \kappa(\lambda)^{n-i}} \]  \hspace{1cm} (16b)

The term \((\Psi - 1)\) is the term premium scale factor; it does not depend on the maturity \(n\). It can be easily seen from equation (14b) that when \(\theta = 0\), \(\Psi = 1\) and hence the term premium scale factor is zero. This happens when either utility is logarithmic \((r = 1)\) or when \(h_1 = 0\) from equation (6d)). For an \(n\)-period discount bond, \(a_1 = ... = a_{n-1} = 0\). Therefore, from equation (16b), \(\Gamma(n,0) = 1\) for \(n > 1\). Thus, the term premium is independent of the maturity \(n\) for pure discount bonds with more than one period to run.

3. EXOGENOUS DRIVING PROCESS AND EQUILIBRIUM RATES OF RETURN AND TERM PREMIA

In section 3.1 we define explicitly the dampened power law process for consumption growth rates, and in section 3.2 discuss estimation of the process. In section 3.3 we specialize the formulae for the term premium and the expected rates of return on the canonical asset to risk-free discount bonds and equity that pays consumption goods as in the Lucas (1978) fruit tree model. In section 3.4 we outline the benchmark Gaussian consumption process.

Economic and financial asset returns have been shown to possess distributions displaying power law decay in their tails, indicating that the tails are thicker than what one would find in the Gaussian case. These fat-tailed distributions are often said to have “power tails” that are inconsistent with the common Gaussian distributional assumption. However, most asset returns also converge to a Gaussian distribution when aggregated over time. This fact is inconsistent with the assumption of an $\alpha$-Stable distribution as a possible explanation for the observed power tails mentioned above, since time-aggregation of Stable distributions yields a Stable distribution. In a recent article, Wu (2004) focuses on reconciling these apparently contradicting observations by modeling asset returns with a Dampened Power Law (DPL).

The DPL model aims at reproducing the power tails observed in the finance and economics literatures, simultaneously allowing time aggregation to lead to Gaussian distributions (by permitting the Central Limit Theorem to hold). This is accomplished by the dampening of the tails of the probability distributions with an exponential function that nevertheless permits accurate modeling of the power tail distributions observed empirically. Dampening also guarantees the existence of finite moments of all orders. Without dampening, not all moments of the power tail distributions are finite and hence time aggregation would not necessarily lead to Gaussian behavior.

In the DPL setting, we assume that the consumption growth rates follow:

$$\ln(x_{t+1}) = \mu + \varepsilon_{t+1}, \quad (17a)$$

where $\varepsilon_t$ is a pure jump Lévy process following a Dampened Power Law (DPL), with its Lévy density – controlling the distribution – defined by:
\[ v(\varepsilon) = \begin{cases} \gamma_+ \exp\{-\beta_+ |\varepsilon|\} |\varepsilon|^{-\alpha-1}, & \varepsilon > 0 \\ \gamma_- \exp\{-\beta_- |\varepsilon|\} |\varepsilon|^{-\alpha-1}, & \varepsilon < 0 \end{cases} \]  

(17b)

where \( \alpha \in (0, 2] \) and \( \gamma_+, \gamma_-, \beta_+, \beta_- > 0 \). The \( \beta \) parameters control the amount of dampening, whereas the \( \gamma \) parameters determine the symmetry of the distribution. The \( \alpha \) parameter is identical to the \( \alpha \) parameter found in the traditional \( \alpha \)-Stable distributions, guiding the amount of leptokurtosis in the tails.

We adopt this DPL process for describing the exogenously evolving consumption growth rates in our paper because it allows simultaneously for fat tails and the existence of finite moments. The finance literature has recently produced a myriad of models that could be used for modeling fat tails, such as models of stochastic volatility with jumps developed in, among other studies, Bates (1996), Bakshi, Cao and Chen (1997), Duffie, Pan and Singleton (2000) or Pan (2002). However, these models have been only moderately successful at reconciling observed empirical facts with reasonable levels of jump magnitude and/or jump frequency, and have not in any way been proven superior to a general pure-jump Lévy process.

The main reason for incorporating jump features in an asset-pricing framework is the possibility this affords at replicating the levels of skewness and leptokurtosis observed in the data. However, this can also be achieved alternatively using an appropriate type of Lévy process, such as the Dampened Power Law process entertained here. Instead of being an amalgam of Brownian motions, Poisson jumps and stochastic volatility, the Dampened Power Law process embeds various features in one clean model and is thus intuitively appealing.
It needs to be emphasized that the primary purpose of this article is measuring the impact of fat tails on the equilibrium rates of return and equity and term premia, not necessarily in finding the best possible model that generates the observed leptokurtosis in the consumption data. Thus a comparison of the DPL process with stochastic volatility models with jumps is not undertaken for this reason.

3.2 Estimation Issues

Before estimating the parameters associated with the Lévy process, we first estimate the mean growth rate of consumption and use it to demean the series. Note that properties of logarithms imply that computing a simple arithmetic average of log differences of consumption is essentially the same as averaging the first and last observation, which can yield a noisy estimate of the mean growth rate. Hence, following Wu (2004), we regress log consumption levels on time t instead, estimating

\[ \ln C_t = a + bt + u_t \]  

(18)

From the discrete setting of equation (17a), the estimate for b is thus an estimate of the mean annualized growth rate of consumption \( \mu \). We use this estimate of \( \mu \) to detrend the consumption series, and model the log-difference of detrended consumption data as a pure-jump Levy DPL process.

The cumulant exponent of the Lévy DPL process given in section 3.1 above is derived in Wu (2004):

\[ k(s) = \Gamma(-\alpha)\gamma_+[(\beta_+ - s)^\alpha - \beta^{\alpha}_+] + \Gamma(-\alpha)\gamma_-[(\beta_- + s)^\alpha - \beta^{\alpha}_-] + sQ \]  

(19)

where

\[ Q = \gamma_+\beta_+^{\alpha-1}[\Gamma(-\alpha)\alpha + \Gamma(1-\alpha,\beta_+)] - \gamma_-\beta_-^{\alpha-1}[\Gamma(-\alpha)\alpha + \Gamma(1-\alpha,\beta_-)] \]  

(20)
Note that the expression for $Q$ depends on the choice of the truncation function selected during the application – needed for the derivation of the cumulant exponent – of the Lévy-Khinchine theorem to the DPL Lévy density. With $\varepsilon$ following the DPL process, the implicit truncation applied is the widely used and accepted truncation function $h(\varepsilon) = \varepsilon |\varepsilon| < 1$.

Since the cumulant exponent is defined as

$$k(s) = \frac{1}{t} \log E\left[\exp\{s\varepsilon_t\}\right]$$  \hspace{1cm} (21)

we can use the fact that $E[\exp\{s\varepsilon_t\}] = \exp\{t k(s)\}$ along with the expression for $k(s)$ from equations (19) and (20) in order to compute exponential moments of the DPL process. These exponential moments are directly utilized in computing the equilibrium expected riskless rate of return, the equity premium and the term premium in the empirical section of the paper.

The characteristic function of the DPL process for an interval of time $t$ is:

$$\Phi_{\varepsilon_t}(s) = E\left[\exp\{is\varepsilon_t\}\right] = tk(is)$$  \hspace{1cm} (22)

We can thus use the expression for $k(s)$ in order to recover the probability density function of the Lévy density by standard Fourier inversion transformation. Once the density function is retrieved, the parameters associated with the DPL can be estimated by maximum likelihood.

3.3 Equilibrium Rates of Return and Term Premia

In this section we specialize the formulae for the expected rates of return on the canonical asset and the term premium defined in sections 2.5 and 2.6 to risk-free discount
bonds and equity that pays consumption goods as in the Lucas (1978) fruit tree model.

For this kind of equity, the stochastic payoff equals consumption per capita so that \( y_t = C_t \), and hence the payoff growth rates equal consumption growth rates so that \( z_{t+1} = x_{t+1} \) in the notation introduced in section 2.3.

Using the above identities, starting from equation (13), the unconditional mean of the riskless rate on a discount bond for which \( \lambda = 0 \) can be shown to be:

\[
E \{ R_{t+1}(1,0) \} = \frac{E \{ x_t^{-\theta} \}}{\beta E \{ x_t^{\lambda-A} \}}
\]

(25)

Similarly, starting from equation (14a), the unconditional mean of equity returns for which \( \lambda = 1 \) can be shown to be:

\[
E \{ R_{t+1}(\infty,1) \} = [1 + \kappa(1)(\Psi - 1)]E \{ R_{t+1}(1,1) \}
\]

(26a)

where

\[
E \{ R_{t+1}(1,1) \} = \left[ \frac{E \{ x_{t+1}^{\lambda} \} E \{ x_{t+1}^{\lambda-A} \}}{E \{ x_{t+1}^{1-A} \}} \right] E \{ R_{t+1}(1,0) \},
\]

(26b)

and from equation (10b),

\[
\kappa(1) = \beta E \{ x_{t+1}^{\lambda+\theta-A} \}.
\]

(27)

From equation (14b), for discount bonds with \( \lambda = 0 \),

\[
\Psi = \frac{E \{ x_{t+1}^{\theta} \} E \{ x_{t+1}^{\lambda-A} \}}{E \{ x_{t+1}^{\theta-A} \}}
\]

(28)

As indicated in section 2.6, the term premium on discount bonds is just the term premium scale factor \((\Psi - 1)\).

Since the expressions for the mean risk-free rate, mean equity premium and term premium of equations (25)-(28) involve terms such as \( E \{ x^A \} \) or \( E \{ x^{\theta-A} \} \), we can compute these expressions using Wu’s (2004) cumulant exponent formula given in equations (19)-
Wu (2004) shows in his proposition 2 that with $\gamma_+, \gamma_-, \beta_+, \beta_- > 0$, the cumulant exponent is well defined only for $s \in (-\beta_-, \beta_+)$. Thus, we need to ensure that Wu’s proposition 2 is not violated when computing equilibrium mean rates of return and the term premium. Examining the expressions that need to be computed above, we can easily see that the following terms must lie in the interval $(-\beta_-, \beta_+)$: $\theta - A$, $1 + \theta - A$, $\theta$, $-A$, $1 + \theta$, $1 - A$, $1$, and $-\theta$. This imposes restrictions on the range of values that one can entertain for the preference parameters including the relative risk aversion coefficient used subsequently in calculating the model implied rates of return and the term premium.

3.4 The Log-Normal Case.

In the benchmark log-normal case, we assume that the consumption growth rates follow an i.i.d. Gaussian process:

$$\ln(x_{t+1}) = \mu + \epsilon_{t+1}, \quad \epsilon_{t+1} \sim \text{iid } \mathcal{N}(0, \sigma^2)$$

(29)

Given the moment generating function of the normal distribution, we can readily evaluate analytically all the expressions given in section 3.3 for the model-implied equilibrium quantities of interest when the consumption growth rates $x_{t+1}$ (which equal the payoff growth rates $z_{t+1}$) are assumed to be i.i.d. Gaussian. We present these formulae below.

From equation (25), the expected gross rate of return on a discount bond equals:

$$E \left\{ R_{t+1} (1, 0) \right\} = \beta^{-1} \exp \left\{ (A - \theta) \mu - \frac{1}{2} \left( A^2 - \theta^2 \right) \sigma^2 \right\}$$

(30)

From equation (26a), the expected gross rate of return on equity becomes:
\[
E[R_{t+1}(\infty,1)] = \beta^{-1}\left[1 + \kappa(1)\exp(\theta A \sigma^2) - \frac{1}{2} \exp\left(A \sigma^2\right)\exp\left((A - \theta)\mu - \frac{1}{2}(A^2 - \theta^2)\sigma^2\right)\right]
\]

(31)

where from equation (27),

\[
\kappa(1) = \beta \exp\left((1 + \theta - A)\mu + \frac{1}{2}(1 + \theta - A)^2\sigma^2\right).
\]

(32)

From equation (28),

\[
\Psi = \exp\left(\theta A \sigma^2\right)
\]

(33)

From equation (16a) and the discussion that follows, the term premium on riskless discount bonds becomes:

\[
TP(n,0) = \Psi - 1.
\]

(34)

4. EVALUATING MODEL-IMPLIED PREMIA AND EXPECTED RATES OF RETURN

In section 4.1 we discuss the consumption data series used and report summary statistics. In section 4.2 we report the maximum likelihood estimates of the DPL process and the benchmark Gaussian process for the data. In section 4.3 we compute the model-implied expected rates of return and the term premium and discuss the quantitative implications of modeling fat tails.

4.1 Characteristics of the Consumption Data

We use annual US real per capita consumption data on non-durables and services from Campbell and Cochrane (1999) spanning the period 1889-1997. Figure 1 plots the real consumption data and Table 1 presents summary statistics. The mean growth rate is
1.726 percent per annum. Kurtosis is measured to be in excess of, and statistically significantly greater than, 3 indicating fat tails in the empirical histogram. Normality is strongly rejected by the Jarque-Bera test (p-value is 4.43e-4).

### 4.2 Model Estimates for the Consumption Growth Rates

Table 2 presents empirical results on maximum likelihood estimates of equations (17) and (29). We estimate the most general unconstrained version of the DPL process presented in equations (17) and several restricted versions. The first row reports estimation results for the most general version. The second row reports estimation results for the symmetric dampening case where the dampening parameters $\beta_+ = \beta_-$. The third row reports estimates obtained by fitting a DPL process without dampening that is identical to fitting an $\alpha$-stable process to the consumption growth rates. The next row reports estimates obtained by fitting a symmetric $\alpha$-stable process. The last row reports results of fitting the benchmark Gaussian model. Incremental benefits of the most general version of the DPL process can be measured by the log-likelihood values in the last column.

The most general model yields estimates of 1.73 for $\alpha$, scaling parameters $\gamma_+ = 0.00029$ and $\gamma_- = 0.00034$, and dampening parameters $\beta_+ = 10$ and $\beta_- = 7$. The difference in scaling parameters $(\gamma_+ - \gamma_-)$ indicates a slight degree of negative skewness in the distribution of consumption growth rates. The dampening coefficients $\beta_+$ and $\beta_-$ are both large, and statistically significantly positive, thus guaranteeing the existence of finite moments of all orders (see Wu’s (2004) proposition 1). All other estimates are also statistically significant at the 0.05 level.
Estimates of the common parameters for the restricted nested models are generally similar to those for the unrestricted case reported above. In the symmetric dampening case the common dampening coefficient $\beta = 4$. Standard likelihood ratio (LR) test (not reported) would reject symmetric dampening in favor of the general model in equations (17) at the 0.10 significance level. Similarly an LR test would reject the Gaussian process in favor of the dampened power law process for consumption growth rates at the 0.10 significance level.

Armed with parameter estimates, rates of return and the equity and term premia implied by the model can now be computed alternatively under the DPL and Gaussian process for consumption growth rates. A comparison of these quantities would provide a quantitative assessment of the implications of modeling fat tails.

### 4.3 Implied Premia and Expected Rates of Return

As noted at the end of section 2.1 our asset-pricing model has six preference parameters. We need to select values for each of these parameters before we can compute equilibrium quantities of interest implied by our model. We follow Abel (1999) in imposing the following three restrictions on these parameters: $h_0 = 0$, $h_0 + h_1 + h_2 = 1$, and $G = 1 + \mu$. The first restriction $h_0 = 0$ implies that the benchmark level of consumption $v_t$ does not depend on the current period aggregate per capita level of consumption $C_t$ as evident from equation (3). The third restriction $G = 1 + \mu$ captures the intuitive notion that the growth rate of the benchmark level of consumption over time reflects the growth rate of the aggregate per capita level of consumption.
Following the model parameterization in Abel (1999), we choose $h_1 = 0.15$. We choose the rate of time preference $\delta$ to be 0.02. This leaves us with just one parameter to choose, namely, the coefficient of relative risk aversion $\gamma$. We report results for different values of $\gamma$ below. Following the discussion at the end of section 3.3, in order to ensure finiteness of exponential moments used in computing model-implied equilibrium rates of return and the term premium, the coefficient of relative risk aversion must be less than the estimate of $\beta_-$. Thus, using the estimate of $\beta_-$ for the unrestricted DPL model from section 4.2, $\gamma$ is constrained to be less than 7.

Table 3 presents the model-implied equilibrium expected risk-free rates, equity premium, and the term premium for the unrestricted DPL consumption growth rate process and the log-normal process. All rates of return and the term premium are expressed in percent per annum. As is evident from the table, the asset-pricing model is able to generate a low enough mean risk-free rate of under 3 percent per annum with relative risk aversion coefficient of about 6. The mean equity premium for this CRRA coefficient is 2.2 percent per annum, which is higher than what Mehra and Prescott (1985) are able to generate without habit formation but still much lower compared to the historical level of about 7 percent per annum.

The model generates a term premium on risk-free bonds of about 0.6 percent per annum with a CRRA coefficient of 6. Abel (1999) reports a term premium on long term US government-issued fixed-income securities of 170 basis points per year. Abel (1999) has greater success in replicating both the empirically observed mean risk-free rate and the equity and term premium with the asset-pricing model used here by incorporating leverage in the model. Our main objective in this paper is to evaluate the quantitative
importance of modeling fat tails on the model-implied equilibrium rates of return and the
term premium in a relatively simple framework. We therefore did not entertain the
possibility of leverage in the version of the asset-pricing model considered here.

Comparing the model-implied rates of return in the DPL and log-normal cases in
Table 3, we find that accounting for fat tails leads to a lower mean risk-free rate, and
higher mean equity and term premiums. More specifically, the DPL model is able to
lower the mean risk-free rate by as much as 20 percent, raise the mean equity premium
by 80 percent or more, and raise the term premium by 20 percent compared to the log-
normal case. Thus, accounting for fat tails produces a closer match of the quantitative
implications of the consumption-based asset-pricing model with the stylized facts
observed in the macroeconomic and financial data.

Figure 2 plots the model-implied equilibrium mean risk-free rate, the mean equity
and term premiums as a function of the coefficient of relative risk aversion in both the
DPL and the log-normal cases. As the graphs indicate, accounting for fat tails has greater
impact on the implied rates of return and the equity and term premiums at higher values
of the CRRA coefficient.

5. CONCLUSIONS

In this study we addressed the question: what are the costs of ignoring fat tails in
the empirical distributions of macroeconomic time series on the equilibrium implications
of macroeconomic models? We addressed this question within the context of the
consumption-based asset-pricing model, modified to incorporate habit formation as in
Abel (1999). We considered two versions of the model – one in which exogenous
consumption evolves as a stochastic dampened power law (DPL) process as in Wu
(2004) and the other benchmark version in which consumption follows a Gaussian
process. DPL nests $\alpha$-stable distributions but has the advantage that all moments are
finite under strictly positive dampening. This renders model-implied rates of return and
equity and term premia finite under certain restrictions.

We parameterized the two versions of the model with estimates derived from the
annual US monthly real per capita consumption data. Choosing suitable values for the
preference parameters of the model, our results show that accounting for fat tails
improves the ability of the asset-pricing model to replicate empirically observed mean
risk-free rate, equity and the term premia. Specifically, accounting for fat tails through a
DPL process generates 20 percent lower mean risk-free rate, 80 percent higher equity
premium, and 20 percent higher term premium compared to the log-normal case.
Table 1: Summary Statistics of Real Per Capita Consumption Data

<table>
<thead>
<tr>
<th></th>
<th>Mean</th>
<th>Variance</th>
<th>Skewness</th>
<th>Kurtosis</th>
<th>Normality test</th>
</tr>
</thead>
<tbody>
<tr>
<td>Real per capita consumption growth rates</td>
<td>1.726e-2</td>
<td>1.041e-3</td>
<td>-0.503</td>
<td>4.555</td>
<td>15.442</td>
</tr>
<tr>
<td></td>
<td>(3.105e-3)</td>
<td>(1.417e-4)</td>
<td>(0.984)</td>
<td>(4.847e-4)</td>
<td>(4.434e-4)</td>
</tr>
</tbody>
</table>

Notes to Table 1:

1. Numbers in parentheses in the first two columns are the standard errors for the mean and variance.

2. Numbers in parentheses in the third and fourth columns are the p-values for the null hypothesis of no skewness and no excess kurtosis, respectively.

3. The normality test gives the Jarque-Bera test statistic and the p-value in parentheses.
Table 2: Maximum Likelihood Model Estimates

<table>
<thead>
<tr>
<th></th>
<th>( \alpha )</th>
<th>( \gamma_+ )</th>
<th>( \gamma_- )</th>
<th>( \gamma_+ = \gamma_- )</th>
<th>( \beta_+ )</th>
<th>( \beta_- )</th>
<th>( \beta_+ = \beta_- )</th>
<th>( \sigma^2 )</th>
<th>( \log L )</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Dampened Power Law Process</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>( \ln(x_{t+1}) = \mu + \varepsilon_{t+1}, ) ( \varepsilon_{t+1} \sim \text{iid DPL}(\alpha, \gamma_+, \gamma_-, \beta_+, \beta_-) )</td>
<td>( 1.73 )</td>
<td>( 2.9e-4 )</td>
<td>( 3.4e-4 )</td>
<td>( 10.00 )</td>
<td>( 7.00 )</td>
<td></td>
<td></td>
<td>( 218.49 )</td>
<td></td>
</tr>
<tr>
<td></td>
<td>( (0.03) )</td>
<td>( (1.2e-4) )</td>
<td>( (1.2e-4) )</td>
<td>( (3.41) )</td>
<td>( (3.52) )</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>Gaussian Process</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>( \ln(x_{t+1}) = \mu + \varepsilon_{t+1}, ) ( \varepsilon_{t+1} \sim \text{iid N}(0, \sigma^2) )</td>
<td>( 2.9e-4 )</td>
<td>( 2.9e-4 )</td>
<td>( 2.9e-4 )</td>
<td>( 4.00 )</td>
<td></td>
<td></td>
<td></td>
<td>( 216.67 )</td>
<td></td>
</tr>
<tr>
<td></td>
<td>( (2.01e-7) )</td>
<td>( (2.01e-7) )</td>
<td>( (2.01e-7) )</td>
<td>( (0.03) )</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>Symmetric Dampening</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>( \alpha ) = \gamma_+ = \gamma_- = \beta_+ = \beta_- )</td>
<td>( 1.74 )</td>
<td>( 2.3e-4 )</td>
<td>( 2.6e-4 )</td>
<td>( 4.00 )</td>
<td></td>
<td></td>
<td></td>
<td>( 216.35 )</td>
<td></td>
</tr>
<tr>
<td></td>
<td>( (0.01) )</td>
<td>( (1.32e-6) )</td>
<td>( (1.32e-6) )</td>
<td>( (0.03) )</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>No Dampening</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>( \alpha ) = \gamma_+ = \gamma_- = \beta_+ = \beta_- )</td>
<td>( 1.74 )</td>
<td>( 2.1e-4 )</td>
<td>( 2.3e-4 )</td>
<td>( 2.4e-4 )</td>
<td>( 0 )</td>
<td>( 0 )</td>
<td>( (restricted) )</td>
<td>( (restricted) )</td>
<td>( 215.92 )</td>
</tr>
<tr>
<td></td>
<td>( (0.02) )</td>
<td>( (1.1e-4) )</td>
<td>( (1.1e-4) )</td>
<td>( (8.3e-5) )</td>
<td>( (restricted) )</td>
<td>( (restricted) )</td>
<td>( (restricted) )</td>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>No Dampening on Symmetric Stables</strong></td>
<td>( 1.74 )</td>
<td>( 2.6e-4 )</td>
<td>( 2.4e-4 )</td>
<td>( 0 )</td>
<td>( 0 )</td>
<td></td>
<td></td>
<td>( 215.92 )</td>
<td></td>
</tr>
<tr>
<td><strong>Gaussian</strong></td>
<td>( 2 )</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>( 1.11e-3 )</td>
<td>( 213.92 )</td>
</tr>
<tr>
<td></td>
<td>( (restricted) )</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>( (1.5e-4) )</td>
<td></td>
</tr>
</tbody>
</table>

Notes to Table 2:

1. Numbers in parentheses are the standard errors.
Table 3: Expected Rates of Return and Term Premia

<table>
<thead>
<tr>
<th></th>
<th>( r = 1.5 )</th>
<th>( r = 2 )</th>
<th>( r = 3 )</th>
<th>( r = 4 )</th>
<th>( r = 5 )</th>
<th>( r = 6 )</th>
<th>( r = 7 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>( E[R_{t+1}(1,0)] )</td>
<td>Log-Normal</td>
<td>4.550</td>
<td>4.464</td>
<td>4.208</td>
<td>3.839</td>
<td>3.359</td>
<td>2.770</td>
</tr>
<tr>
<td>% Difference</td>
<td>-0.2</td>
<td>-0.4</td>
<td>-1.1</td>
<td>-2.3</td>
<td>-4.6</td>
<td>-9.0</td>
<td>-19.1</td>
</tr>
<tr>
<td>Mean Equity Premium</td>
<td>DPL</td>
<td>0.390</td>
<td>0.540</td>
<td>0.871</td>
<td>1.247</td>
<td>1.671</td>
<td>2.155</td>
</tr>
<tr>
<td>( E[R_{t+1}(\infty,1)] - E[R_{t+1}(1,0)] )</td>
<td>Log-Normal</td>
<td>0.187</td>
<td>0.266</td>
<td>0.450</td>
<td>0.667</td>
<td>0.917</td>
<td>1.199</td>
</tr>
<tr>
<td>% Difference</td>
<td>108.4</td>
<td>102.6</td>
<td>93.5</td>
<td>86.9</td>
<td>82.3</td>
<td>79.7</td>
<td>80.4</td>
</tr>
<tr>
<td>Term Premium</td>
<td>DPL</td>
<td>0.014</td>
<td>0.036</td>
<td>0.110</td>
<td>0.222</td>
<td>0.376</td>
<td>0.577</td>
</tr>
<tr>
<td>( TP(n,0) )</td>
<td>Log-Normal</td>
<td>0.012</td>
<td>0.033</td>
<td>0.100</td>
<td>0.200</td>
<td>0.334</td>
<td>0.501</td>
</tr>
<tr>
<td>% Difference</td>
<td>8.4</td>
<td>8.8</td>
<td>9.8</td>
<td>11.1</td>
<td>12.8</td>
<td>15.3</td>
<td>20.0</td>
</tr>
</tbody>
</table>

Notes to Table 3:

1. All statistics are expressed in percent per annum.
2. % Difference is the difference in the relevant statistic between the DPL and the Log-Normal cases, relative to the value in the Log-Normal case.
Figure 1. Plots of Real Per Capita Consumption Data

Panel A
Figure 2. Expected Rates of Return and Term Premia

Panel A
Panel B

Expected Equity Premium

- Lognormal
- DPL

Percent per Annum

CRRA coefficient $\gamma$
REFERENCES


Hales, S.J., 1997. Valuation of foreign currency options with the paretian stable option pricing model, Ph.D. dissertation, Ohio State University, Department of Economics.


