A Parsimonious Macroeconomic Model 
For Asset Pricing*

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Abstract

In this paper we study asset prices in a two-agent macroeconomic model with two key features: limited participation in the stock market and heterogeneity in the elasticity of intertemporal substitution in consumption (EIS). The model is consistent with some prominent features of asset prices that have been documented in the literature, such as a high equity premium; relatively smooth interest rates; procyclical variation in stock prices; and countercyclical variation in the equity premium, in its volatility, and in the Sharpe ratio. In this model, the risk-free asset market plays a central role by allowing the non-stockholders (who have low EIS) to smooth the fluctuations in their labor income. This process thus concentrates non-stockholders’ aggregate labor income risk among a small group of stockholders, who then demand a high premium for bearing the aggregate equity risk. Furthermore, this mechanism is consistent with the very small share of aggregate wealth held by non-stockholders in the US data, which has proved problematic for earlier models with limited participation. Finally, we show that this large wealth inequality is also important for the model’s ability to generate a countercyclical equity premium.

Keywords: Real business cycle model; the equity premium puzzle; limited stock market participation, elasticity of intertemporal substitution.

JEL classification: E32, E44, G12

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1 Introduction

Since the 1980s, a vast body of empirical research has documented some interesting and puzzling features of asset prices. For example, in a famous paper Mehra and Prescott (1985) have shown that the equity premium (i.e., the excess return on stocks over bonds) was hard to reconcile with a canonical consumption-based asset pricing model, and as it later turned out, with many of its extensions. A parallel literature in financial economics has found that the equity premium was predictable by a number of variables including the dividend yield, challenging the long-held view that stock returns follow a martingale (Campbell and Shiller (1988)). Other studies have documented that the expected equity premium, its volatility, and the ratio of the two—the conditional Sharpe ratio—move over time following a countercyclical business cycle pattern (Schwert (1989), and Chou, Engle, and Kane (1992)).

In this paper, we ask if these asset pricing phenomena can be explained in a parsimonious macroeconomic model with two key features: limited participation in the stock market and heterogeneity in the elasticity of intertemporal substitution in consumption (EIS). The limited nature of stock market participation and the concentration of stock wealth even among stockholders is well documented. For example, until the 1990s more than two-thirds of US households did not own any stocks at all, while the richest 1% held 48% of all stocks (Poterba and Samwick (1995), and Investment Company Institute (2002)). As for the heterogeneity in preferences, the empirical evidence that we review in Section 3 indicates that stockholders have a higher EIS than non-stockholders. The interaction of these two features is important, as will become clear below.

We choose the real business cycle model as the foundation that we build upon, to provide a contrast with the poor asset pricing implications of that framework that are well-known, which helps to highlight the role of the new features considered in this paper. Specifically, we study an economy with competitive markets and a neoclassical production technology where capital investments are subject to adjustment costs. There are two types of agents. The majority of households (first type) do not participate in the stock market where claims to the firm’s future dividend stream are traded. However, a risk-free bond is available to all households, so non-stockholders can also accumulate wealth and smooth consumption intertemporally. Finally, consistent with empirical evidence, non-stockholders are assumed to have a low EIS, whereas stockholders have a higher elasticity. To clarify the role played by different preference parameters, we employ Epstein-Zin preferences and disentangle risk aversion from the EIS. We find that heterogeneity in risk aversion plays no essential role, whereas heterogeneity in the EIS is essential, for the results of this paper.
We first examine a benchmark version of the model in which labor supply is inelastic. The calibrated model is consistent with some salient features of asset prices, such as a high equity premium with a plausible volatility, and a low average interest rate. Furthermore, the variability of the interest rate is very low in the US data, which has proved challenging to explain for some previous models that have otherwise successful implications. The standard deviation of the risk-free rate is about 4% – 6.5% in our model, which is still higher than in the US data, but quite low compared to these earlier studies. So, the present paper provides a step in the right direction as far as interest rate volatility is concerned.

Although there are now several papers that have made progress in explaining these unconditional moments in the context of production economies, some aspects of asset price dynamics have proved more difficult to generate. The present model is consistent with the procyclical variation in stock prices; the mean reversion in the equity premium; and the countercyclical variation in the expected equity premium, in its volatility, and in the conditional Sharpe ratio. The model also reproduces the long-horizon predictability of the equity premium, although the degree of predictability is quantitatively smaller than in the data.

This paper, as well as earlier models with limited participation, build on the empirical observation, first made by Mankiw and Zeldes (1991), that stockholders’ consumption growth is more volatile (and more highly correlated with returns) than that of non-stockholders. Therefore, a high equity premium can be consistent with the relatively smooth per capita consumption process in the US data, since stockholders only make up a small fraction of the population. Existing theoretical models differ precisely in the economic mechanisms they propose for generating this high volatility of stockholders’ consumption growth.

The mechanism in this paper differs from earlier studies (most notably, Saito (1995) and Basak and Cuoco (1998)) in some crucial ways. In particular, in these earlier models non-stockholders consume out of wealth, which they must invest in the bond market given the absence of any other investment opportunity. As a result, each period stockholders make interest payments to non-stockholders, which leverages the capital income of stockholders, thereby amplifying their consumption volatility. Although this is a potentially powerful mechanism, it only works quantitatively if these interest payments are substantial, which in turn requires non-stockholders to own a substantial fraction of aggregate wealth. But, in reality, non-stockholders own only one-tenth of aggregate wealth in the United States, and this counterfactual implication has been an important criticism raised against these models.

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One contribution of this paper is to propose a new economic mechanism, which avoids this counterfactual implication. Specifically, the mechanism results from the interaction of three factors. First, non-stockholders receive labor income every period, which is stochastic, and trade in the bond market for smoothing the fluctuations in their consumption. Second, because of their low EIS, non-stockholders have a stronger desire for consumption smoothing—and therefore need the bond market much more—than stockholders (who have a higher EIS and an additional asset for consumption smoothing purposes). However, and third, since the source of risk is aggregate, the bond market cannot eliminate this risk and merely reallocates it across agents. In equilibrium, stockholders make payments to non-stockholders in a countercyclical fashion, which serves to smooth the consumption of non-stockholders and amplifies the volatility of stockholders, who then demand a large premium for holding aggregate risk. As shown in Section 6.2, this mechanism is consistent with a very small wealth share of non-stockholders precisely because it is the cyclical nature of interest payments that is key, and not their average amount (which can very well be zero).

The same mechanism also explains why the equity premium is countercyclical. Essentially, because non-stockholders have very low wealth they become effectively more risk averse during recessions when their wealth falls even further (because with incomplete markets value functions have more curvature at low wealth levels.) This is not the case for stockholders who hold substantially more wealth. Consequently, during recessions, non-stockholders demand more consumption smoothing, which strengthens the mechanism described above—i.e., increased trade in the bond market, more volatile consumption growth for stockholders—generating a higher premium in recessions. In Section 5, we quantify the contribution of these channels to both the level and countercyclicality of the equity premium.

We also investigate the extent to which labor supply choice can be endogenized in this framework without compromising overall performance. We first find that Cobb-Douglas utility does not appear to be suitable for this task: it results in a deterioration of asset pricing results and generates labor hours much smoother than in the data. One reason for these results is that these preferences do not allow an independent calibration of the EIS and the Frisch labor supply elasticity parameters, which are both crucial for our analysis. This poor performance perhaps does not come as a surprise in light of the earlier results in the literature. For example, Lettau and Uhlig (2000), Boldrin et al. (2001), and Uhlig (2006) uncover various problems generated by endogenous labor supply in asset pricing models and identify certain labor market frictions that successfully overcome these difficulties. Incorporating the same frictions into the model with Cobb-Douglas utility could also work to improve its
performance, although this would be beyond the scope of the present paper. However, on a more positive note, we also consider the utility specification first introduced by Greenwood, Hercowitz, and Huffman (1988) and find that it performs better: it preserves the plausible asset pricing implications of the model with inelastic labor fairly well, and generates business cycle implications in the same ballpark as existing macroeconomic models. Although this version of the model also has some shortcomings that are discussed in the paper, it appears to provide a promising direction for endogenizing labor supply.

Finally, in a related framework, Danthine and Donaldson (2002) study asset prices in a two-agent model with an entrepreneur and a worker (where the worker lives hand-to-mouth and there is no labor supply choice). In this environment, labor contracts between the two agents act as “operational leverage” and affect asset prices in a way that is similar to limited participation. Storesletten, Telmer, and Yaron (2007) build a heterogeneous-agent life cycle model and show that persistent idiosyncratic shocks with countercyclical innovation variance could generate plausible unconditional moments. As noted earlier, one difference of this paper is our focus on the dynamics of asset prices, which is not studied in these papers.

2 The Model

Households. The economy is populated by two types of agents who live forever. The population is constant and is normalized to unity. Let \( \mu \in (0, 1) \) denote the measure of the second type of agents (who will be called “stockholders” later). Consumers are endowed with one unit of time every period, which they allocate between market work and leisure. We consider three different preference specifications in this paper that can be written as special cases of the following Epstein-Zin recursive utility function:

\[
U_t^i = \left(1 - \beta \right) u^i \left( c_t, 1 - l_t \right) + \beta \left( E_t \left( U_{t+1}^i \right)^{1-\alpha_i} \right)^{1-\alpha_i} \frac{1}{1-\beta} \]

for \( i = h, n \), where throughout the paper the superscripts \( h \) and \( n \) denote stockholders and non-stockholders respectively; \( c \) and \( l \) denote consumption and labor supply, respectively. For the parameterizations we consider below, the risk aversion parameter for static wealth gambles will be proportional to \( \alpha_i \), and the EIS will be \textit{inversely} proportional to \( \rho^i \), although the precise relationship will also depend on the choice of \( u \). As indicated by the superscripts, the two types are allowed to differ in their preference parameters.
It should be emphasized that the choice of recursive preferences is made for clarity: by disentangling risk aversion from the elasticity of intertemporal substitution, these preferences allow us to examine the impact of heterogeneity in the EIS on asset prices without generating corresponding differences in risk aversion that could confound the inference.

**The Firm.** There is an aggregate firm producing a single consumption good using capital \((K_t)\) and labor \((L_t)\) inputs according to a Cobb-Douglas technology: 
\[ Y_t = Z_t K_t^\theta L_t^{1-\theta}, \]
where \(\theta \in (0,1)\) is the factor share parameter. The technology level evolves according to:
\[ \log (Z_{t+1}) = \phi \log (Z_t) + \varepsilon_{t+1}, \quad \varepsilon \sim N (0, \sigma^2). \]

The firm’s managers maximize the value of the firm, which equals the value of the future dividend stream generated by the firm, \(\{D_{t+j}\}_{j=1}^{\infty}\), discounted by the marginal rate of substitution process of firm owners, \(\{\beta^j \Lambda_{t,t+j}\}_{j=1}^{\infty}\). Specifically, the firm’s problem is:

\[ P_t^s = \max_{\{I_{t+j}, L_{t+j}\}} Et \left[ \sum_{j=1}^{\infty} \beta^j \Lambda_{t,t+j} D_{t+j} \right] \]  

subject to the law of motion for capital, which features “adjustment costs” in investment:
\[ K_{t+1} = (1 - \delta) K_t + \Phi \left( \frac{I_t}{K_t} \right) K_t. \]  

\(P_t^s\) is the ex-dividend value of the firm, and we normalize the number of shares outstanding to unity (for convenience) so that \(P_t^s\) is also the stock price. The adjustment cost function \(\Phi (\cdot)\) is concave in investment, which captures the difficulty of quickly changing the level of capital installed in the firm.

Every period the firm sells one period bonds, at price \(P_t^f\), to finance part of its investment. The total supply of these bonds is constant over time and equals a fraction, \(\chi\), of the average capital stock owned by the firm (as in Jermann (1998), Danthine and Donaldson (2002)). As a result, the firm makes net interest payments in each period in the amount of \((1 - P_t^f) \chi K_t\) to bond owners.\(^2\) An equity share in this firm entitles its owner to the entire stream of future dividends.

\(^2\)The introduction of corporate debt into this framework allows us to model bonds as a positive net supply asset, which is more realistic. However, the Modigliani-Miller theorem holds in this framework in the sense that stockholders are able to fully undo the effect of leverage in their portfolio. Therefore, the existence of leverage has no effect on quantity allocations, which we have verified by solving the model without leverage.
dividends, which is given by the profits net of wages, investment, and interest payments:

$$D_t = Z_t K_t^{\theta} L_t^{1-\theta} - W_t L_t - I_t - (1 - P^f_t) \chi K.$$  

Financial Markets. There are two assets traded in this economy: the firm’s equity shares (stocks) as well as one-period bonds issued by the firm. The difference between the two groups of consumers is in their investment opportunity sets: the “non-stockholders” can freely trade the risk-free bond, but they are restricted from participating in the stock market. The “stockholders,” on the other hand, have access to both markets and hence are the sole capital owners in the economy. Finally, we impose portfolio constraints as a convenient way to prevent Ponzi schemes.

Individuals’ Dynamic Problem and the Equilibrium In a given period, the portfolio of each group can be expressed in terms of the beginning-of-period capital stock, $K$, the aggregate bond holdings of non-stockholders after production, $B$, and the technology level, $Z$. Let $\Upsilon$ denote the aggregate state vector $(K, B, Z)$. The dynamic programming problem of a stockholder can be expressed as follows:

$$V^h(\omega; \Upsilon) = \max_{c, l, b', s'} \left[ (1 - \beta) u(c, 1 - l) + \beta \left( E[V^h(\omega'; \Upsilon') \mid Z]^{1-\alpha} \right)^{\frac{1}{1-\alpha}} \right]$$

s.t.

$$c + P^f(\Upsilon) b' + P^a(\Upsilon) s' \leq \omega + W(K, Z) l$$

$$\omega' = b' + s' (P^a(\Upsilon') + D(\Upsilon'))$$

$$K' = \Gamma_K(\Upsilon)$$

$$B' = \Gamma_B(\Upsilon)$$

$$b' \geq B,$$

where $\omega$ denotes financial wealth; $b'$ and $s'$ are individual bond and stock holdings, respectively; the endogenous functions $\Gamma_K$ and $\Gamma_B$ denote the laws of motion for the wealth distribution which are determined in equilibrium; and $P^f$ is the equilibrium bond pricing function. The problem of a non-stockholder can be written as above with $s' \equiv 0$, and the superscript $h$ replaced with $n$. Finally, the stock return and the risk-free rate are defined as usual: $R^s = (P^a + D') / P^a - 1$, and $R^f = 1/P^f - 1$.

A stationary recursive competitive equilibrium for this economy is given by a pair of value functions, $V^i(\omega^i; \Upsilon)$, $i = h, n$; consumption, labor supply, and bond holding decision rules for each type of agent, $c^i(\omega^i; \Upsilon)$, $l^i(\omega^i; \Upsilon)$, and $b'^i(\omega^i; \Upsilon)$; a stockholding decision
rule for stockholders, \( s'(\omega^h; \Upsilon) \); stock and bond pricing functions, \( P_s(\Upsilon) \) and \( P_f(\Upsilon) \); a competitive wage function, \( W(K, Z) \); an investment function for the firm, \( I(\Upsilon) \); laws of motion for aggregate capital and the aggregate bond holdings of non-stockholders, \( \Gamma_K(\Upsilon) \), \( \Gamma_B(\Upsilon) \); and a marginal utility process \( \Lambda(\Upsilon) \), for the firm, such that:

1) Given the pricing functions and the laws of motion, the value function and decision rules of each agent solve that agent’s dynamic problem.

2) Given \( W(K, Z) \) and the equilibrium discount rate process obtained from \( \Lambda(\Upsilon) \), the investment function \( I(\Upsilon) \) and the labor choice of the firm, \( L(\Upsilon) \), are optimal.

3) All markets clear: we have (a) \( \mu b^h(\omega^h; \Upsilon) + (1 - \mu)b^n(\omega^n; \Upsilon) = \chi K/P_f(\Upsilon) \) (bond market); (b) \( \mu s'(\omega^h; \Upsilon) = 1 \) (stock market); and (c) \( L(\Upsilon) = \mu l^h(\omega^h; \Upsilon) + (1 - \mu) l^n(\omega^n; \Upsilon) \) (labor market), where \( \omega^i \) denotes the wealth of each type of agent in state \( \Upsilon \) in equilibrium.

4) Aggregate laws of motion are consistent with individual behavior: \( K' = (1 - \delta) K + \Phi(I(\Upsilon)/K) K \), and \( B' = (1 - \mu)b^n(\omega^n, \Upsilon) \).

5) There exists an invariant probability measure \( \mathbf{P} \) defined over the ergodic set of equilibrium distributions.

3 Quantitative Analysis

We use numerical methods to solve the model, since an analytical solution is not available. The details of the computational algorithm, the accuracy of the solution, and related issues are discussed in an appendix available from the author’s web site.

3.1 Baseline Parameterization

A model period corresponds to one month of calendar time to approximate the frequent trading in financial markets. Because asset pricing statistics are typically reported at annual frequencies and macroeconomic statistics are reported at quarterly frequencies, we aggregate financial variables and quantities to their respective reporting frequencies to calculate relevant statistics as explained below. Table 1 summarizes our baseline parameter choices.

The functional form for \( \Phi \) is specified as \( a_1 (I_t/K_t)^{1-1/\xi} + a_2 \), as in Jermann (1998), where \( a_1 \) and \( a_2 \) are constants chosen such that the steady state level of capital is invariant to \( \xi \). The curvature parameter \( \xi \) determines the severity of adjustment costs. As \( \xi \) approaches infinity, \( \Phi \) becomes linear, and investment is converted into capital one for one (frictionless economy limit). At the other extreme, as \( \xi \) approaches zero, \( \Phi \) becomes a constant function,
and the capital stock remains constant regardless of the investment level (exchange economy limit). We set $\xi = 0.40$, which is broadly consistent with the values reported in the empirical literature (see Christiano and Fisher (1998) for a survey of existing estimates). The calibration of the capital accumulation equation is completed by setting $\delta$ to 0.0066, implying a quarterly depreciation rate of 2%. As for the technology shock, we match the first order autocorrelation of 0.95 of the Solow residuals at quarterly frequencies by setting $\phi = 0.976$ at monthly frequency. We discretize the AR(1) process for $Z_t$ using a 15-state Markov process. The innovation standard deviation of the technology shock, $\sigma_{\varepsilon_t}$, is set later below.

Given the absence of idiosyncratic shocks in the present model, it does not seem realistic for borrowing constraints to bind frequently (for entire groups of population). Therefore, in the baseline case we calibrate these constraints to be quite loose—equal to 6 months of labor income for both types of agents—which never bind in our simulations.\footnote{In the web appendix we show that if constraints were tight enough to bind frequently, if anything this works to raise the equity premium.} As for the calibration of the leverage ratio, Masulis (1988, Table 1.3) reports that the leverage ratio (debt/book value) of US firms has varied between 13% and 44% from 1929 to 1986. With our calibration, the leverage ratio in the model is set to 20% of the average equity value and fluctuates between 11% and 32%. Moreover, this calibration also ensures that the firm is always able to pay its interest obligations so the corporate bond is default-free.

**Participation Rates.** Our model assumes a constant participation rate in the stock market, which seems to be a reasonable approximation for the period before the 1990s when the participation rate was either stable or increasing gradually (Poterba and Samwick (1995, Table 7)). In contrast, during the 1990s participation has increased substantially: from 1989 to 2002 the number of households who owned stocks increased by 74%, and by 2002 half of the US households had become stock owners (Investment Company Institute (2002)). Modeling the participation boom in this later period would require going beyond the stationary structure of our model, so instead, we exclude this later period (1992—) both when calibrating the participation rate and when comparing the model to the data. We set the participation rate in the model, $\mu$, to 20%, roughly corresponding to the average rate from 1962 to 1992 (a period during which participation data is available). Note that even during times when participation was higher, households in the top 20% have consistently owned more than 98% of stocks (Poterba and Samwick (1995, Table 9)).
Utility Functions. We consider three different specifications for the period utility function. First, to provide a simple and well-known benchmark, we begin with the case where labor supply is inelastic (i.e., leisure is not valued) and assume that the period utility function is of the standard power form: \( u(c, 1 - l) = c^{1-\rho} \). This is a useful benchmark that allows a direct comparison to the existing literature where inelastic labor supply is the most common assumption. In addition, this case allows us to illustrate the key mechanisms resulting from limited participation in their simplest form. To distinguish between different versions of the model, we will often refer to this case as the “CONS model.”

The remaining two specifications feature valued leisure for a full-blown quantitative analysis. The first one features a Cobb-Douglas function (hereafter, the “CD model”) commonly used in macroeconomic analysis: 
\[
\begin{align*}
  u(c, 1 - l) &= (c^\gamma (1 - l)^{1-\gamma})^{1-\rho}.
\end{align*}
\]
However, one restrictive property of this functional form is that \( \rho \) and \( \gamma \) jointly pin down the EIS, the fraction of time devoted to market work, and the Frisch labor supply elasticity. In other words, choosing the two parameters to match the first two empirical magnitudes automatically pins down the Frisch elasticity, which is a serious restriction given that we are interested in constructing a model that allows to study macro quantities and asset prices jointly. To overcome this difficulty we use a third utility function: 
\[
\begin{align*}
  u(c, 1 - l) &= \left(c - \psi \frac{c^{1+\gamma}}{1+\gamma}\right)^{1-\rho},
\end{align*}
\]
introduced by Greenwood et al. (1988, hereafter the “GHH model”). This specification has three distinct parameters that can be chosen to separately target the three parameters mentioned above. This feature will be useful in the analysis that follows.

Preference Parameters. There is a large body of empirical work documenting heterogeneity in the EIS across the population (see Guvenen (2006) for a more comprehensive review of the empirical evidence). These studies find that, by and large, non-stockholders (and the poor in general) have an elasticity of substitution that is very low—close to zero—while stockholders (and the wealthy in general) have an EIS that is higher. For example, Blundell et al. (1994) estimate that households in the top income quintile have an EIS that is three times that of households in the bottom quintile of the distribution. Similarly, Barsky et al. (1997) estimate the distribution of the EIS parameter in the population and find the average to be below 0.2, but find the highest percentiles to be exceeding unit elasticity.

One theoretical explanation for this observed heterogeneity is provided by Browning and Crossley (2000). They start with a model of choice where agents consume several goods with different income elasticities. Because the budget share of luxuries rises with wealth, the aggregate consumption bundle of wealthy individuals have more goods with high income
Table 1: Baseline Parameterization

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Value</th>
</tr>
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<tbody>
<tr>
<td>Parameters calibrated outside the model</td>
<td></td>
</tr>
<tr>
<td>$\beta^s$</td>
<td>Time discount rate</td>
</tr>
<tr>
<td>$eis^h$</td>
<td>EIS of stockholders</td>
</tr>
<tr>
<td>$eis^n$</td>
<td>EIS of non-stockholders</td>
</tr>
<tr>
<td>$\mu$</td>
<td>Participation rate</td>
</tr>
<tr>
<td>$\phi^*$</td>
<td>Persistence of aggregate shock</td>
</tr>
<tr>
<td>$\theta$</td>
<td>Capital share</td>
</tr>
<tr>
<td>$\xi$</td>
<td>Adjustment cost coefficient</td>
</tr>
<tr>
<td>$\delta^*$</td>
<td>Depreciation rate</td>
</tr>
<tr>
<td>$\bar{B}$</td>
<td>Borrowing limit</td>
</tr>
<tr>
<td>$\chi$</td>
<td>Leverage ratio</td>
</tr>
<tr>
<td>Parameters calibrated inside the model (to match targets)</td>
<td></td>
</tr>
<tr>
<td>$\sigma^*_r$</td>
<td>Standard deviation of shock (%)</td>
</tr>
<tr>
<td>$\alpha^h = \alpha^n$</td>
<td>Relative risk aversion</td>
</tr>
</tbody>
</table>

"**" indicates that the reported value refers to the implied quarterly value for a parameter that is calibrated to monthly frequency. $W$ is the average monthly wage rate in the economy. The last two parameters are chosen to (1) match the standard deviation of H-P filtered output in quarterly data (1.89%) and (2) generate an annual Sharpe ratio of 0.25. The standard deviation values refer to CONS/CD/GHH models, respectively.

elasticities than that of poor individuals. Browning and Crossley (2000) prove that this observation also implies that luxuries are easier to postpone than necessities, and consequently, that the EIS (with respect to total consumption) increases with wealth. Since stockholders are substantially wealthier than non-stockholders, this also implies heterogeneity in the EIS across these two groups as found in these studies.

To broadly capture the empirical evidence described above, we set the EIS of non-stockholders to 0.1 and assume an EIS that is three times higher (0.3) for stockholders (in all versions of the model). Finally, we set $\beta$ equal to 0.9966 (monthly) so as to match the US capital-output ratio of 2.5 in annual data.

With Cobb-Douglas preferences, there is only one additional parameter, $\gamma$, which is chosen to match the average time devoted to market activities (0.36 of discretionary time). We continue to keep the EIS values of both groups as above. However, as noted above, $\gamma$ and $\rho^i$ also determine the Frisch labor supply elasticity, which means that assuming heterogeneity in the EIS also implies unintended heterogeneity in the Frisch elasticity: 1.35 for stockholders and 0.69 for non-stockholders. Although such heterogeneity is difficult to justify with any empirical evidence we are aware of, there seems to be no practical way to get around this
problem with CD preferences. We will return back to this caveat later in the analysis.

The GHH specification provides more flexibility, with two additional parameters. The Frisch elasticity is now equal to $1/\gamma$ for both types of agents, which we set equal to 1. This value is consistent with the estimates reported in Kimball and Shapiro (2003). However, there is a fair degree of disagreement in the literature about the correct value of this parameter, so we also discuss below the effect of different values for $\gamma$ on the results. The average hours worked is given by: $L = \left(\frac{\bar{W}(\kappa)}{(1 + \gamma) \kappa}\right)^{1/\gamma}$, where $\bar{W}(\kappa)$ is the average wage rate in the economy whose dependence on $\kappa$ is made explicit. For a target value of $L = 0.36$, this equation is solved to obtain the required value of $\kappa$.

The existing empirical evidence on the risk aversion parameter is much less precise than one would like. Moreover, the limited evidence available pertains to the population average, whereas what will matter for asset prices in this framework is the risk aversion of stockholders, who constitute only a small fraction of the population, making those average figures even less relevant. Therefore, we calibrate the risk aversion of stockholders indirectly, i.e., by matching the model to some empirical targets.

Specifically, we first consider the CONS model. We choose the two parameters that are free at this point $(\alpha^h, \sigma_\varepsilon)$ to match two empirical targets: (1) the volatility of H-P filtered quarterly output (1.89%) and (2) an annual Sharpe ratio of 0.25. We then set the risk aversion of non-stockholders equal to the same value. Our target value for the Sharpe ratio is somewhat lower than the 0.32 figure we obtain in the century-long US data (see Table 2). This is because forcing the model to explain the full magnitude of the risk premium is likely to come at the expense of poor performance in other areas, such as macroeconomic behavior or asset price dynamics, which we are also interested in analyzing. The present choice is intended to balance these different considerations.

For practical considerations, we restrict our parameter search to integer values in the $\alpha^h$ direction (from 2 to 10) and considered 0.1 increments in the $\sigma_\varepsilon$ direction (from 0.05% to 2%). We minimize an equally weighted quadratic objective written in the percent deviation from each empirical target. The minimum is obtained for $\sigma_\varepsilon = 1.5\%$ (quarterly standard deviation) for the CONS model with $\alpha^h = 6$. For the CD and GHH models, we keep the risk aversion parameter at this value and choose $\sigma_\varepsilon$ in each case to match output volatility. The resulting values are $\sigma_\varepsilon = 1.5\%$ in the CD model and $\sigma_\varepsilon = 1.1\%$ in the GHH model.

These values of the innovation standard deviation are close to the values used by Boldrin et al. (2001), Danthine and Donaldson (2002), and Storesletten et al. (2007) in a context...
Table 2: Unconditional Moments of Asset Returns, Model with Inelastic Labor Supply

<table>
<thead>
<tr>
<th></th>
<th>US Data</th>
<th>CONS Model</th>
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<tbody>
<tr>
<td>$rra^b/rra^n$</td>
<td>6/6</td>
<td>6/6</td>
</tr>
<tr>
<td>$eis^b/eis^n$</td>
<td>0.3/0.1</td>
<td>0.3/0.3</td>
</tr>
</tbody>
</table>

A. Stock and Bond Returns

<p>| | | | | |</p>
<table>
<thead>
<tr>
<th></th>
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<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>$E(R^s - R^f)$</td>
<td>6.17</td>
<td>5.46</td>
<td>2.44</td>
<td>7.65</td>
</tr>
<tr>
<td>$\sigma(R^s - R^f)$</td>
<td>19.4</td>
<td>21.9</td>
<td>15.3</td>
<td>27.2</td>
</tr>
<tr>
<td>$\sigma(R^s)$</td>
<td>18.7</td>
<td>20.6</td>
<td>14.7</td>
<td>27.0</td>
</tr>
<tr>
<td>$\frac{E(R^s - R^f)}{\sigma(R^s - R^f)}$</td>
<td>0.32</td>
<td>0.25*</td>
<td>0.16</td>
<td>0.28</td>
</tr>
<tr>
<td>$E(R^f)$</td>
<td>1.94</td>
<td>1.31</td>
<td>3.20</td>
<td>0.24</td>
</tr>
<tr>
<td>$\sigma(R^f)$</td>
<td>5.44</td>
<td>6.65</td>
<td>4.55</td>
<td>8.52</td>
</tr>
</tbody>
</table>

B. Price-Dividend Ratio

<p>| | | | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>$E(P^s/D)$</td>
<td>22.1</td>
<td>27.2</td>
<td>25.9</td>
<td>29.5</td>
</tr>
<tr>
<td>$\sigma(\log(P^s/D))$</td>
<td>26.3</td>
<td>26.6</td>
<td>13.8</td>
<td>38.7</td>
</tr>
<tr>
<td>$\sigma(\Delta \log D)$</td>
<td>13.4</td>
<td>19.1</td>
<td>14.0</td>
<td>24.2</td>
</tr>
</tbody>
</table>

*The Sharpe ratio of 0.25 is one of the two empirical targets in our calibration. All statistics are reported in annualized percentages. Annual returns are calculated by summing up log monthly returns.

similar to ours. Nevertheless, these figures are quite high compared to the direct estimate of the volatility of Solow residuals for the post war period, which is about 0.7% (see, e.g., Cooley and Prescott (1995)). This suggests that it may be more appropriate to interpret the exogenous driving source in this class of models as encompassing more than just technology shocks (such as fiscal policy shocks, among others).

4 Model Results: Asset Prices

4.1 The Unconditional Moments of Asset Prices

We begin by discussing the unconditional moments of stock and bond returns, and then turn to the conditional moments in the next section. Table 2 displays the statistics from the simulated model along with their empirical counterparts computed from the historical US data covering the period 1890–1991. We first examine the inelastic labor supply case shown

4Boldrin et al. (2001) use permanent shocks with a standard deviation of 1.8 percent per quarter, Storesletten et al. (2007) also assume permanent shocks with a standard deviation of 3.3 percent per year. Danthine and Donaldson (2002) use a two-state Markov process with persistence of 0.97 and a standard deviation of 5.6 percent per quarter.

5The data are taken from Campbell (1999). The stock return and the risk-free rate are calculated from Standard and Poor’s 500 index and the six-month commercial paper rate (bought in January and rolled
in column 2. This case provides a useful benchmark, both because it is the most common case studied in the literature and because it allows us to explain the key mechanisms generated by limited participation without the added complexity of labor supply choice.

**The Equity Premium.** As shown in the second column of Table 2, in the calibrated model the target Sharpe ratio of 0.25 is attained with a moderately high risk aversion of 6. Clearly, a given Sharpe ratio can be generated by many different combinations of equity premium and volatility, so matching this target does not say anything about the numerator and the denominator. The corresponding equity premium is 5.45%, which is slightly lower than the historical figure of 6.2%. The volatility of the equity premium is 21.9% compared to 19.4% in the data. Therefore, the model generates an equity premium with mean and volatility that are in the right ballpark compared to the data.

**The Mechanism.** The high equity premium is generated by a general equilibrium mechanism that amplifies stockholders’ consumption growth volatility and does so in a procyclical fashion, causing them to demand a high equity premium. Specifically, the mechanism results from the interaction of three features of the model, which reinforce each other. First, limited participation creates an asymmetry in consumption smoothing opportunities: facing persistent (aggregate) labor income shocks, non-stockholders have to exclusively rely on the bond market, whereas stockholders have another margin—they can also adjust their equity holdings. Second, because of their low EIS, non-stockholders have a stronger desire for a smooth consumption process compared to stockholders. The combination of these two effects imply that non-stockholders need the bond market much more than stockholders. Third, and importantly, the bond market is not a very effective device for consumption smoothing in the face of aggregate risk, because it merely reallocates the risk rather than reducing it, as would be the case if shocks were idiosyncratic. As a result, non-stockholders’ desire for smooth consumption is satisfied via trade in the bond market, at the expense of higher volatility in stockholders’ consumption. Moreover, since these large fluctuations in stockholders’ consumption are procyclical, they are reluctant to own the shares of the aggregate firm that performs well in booms and poorly in recessions. Therefore, they demand a high equity premium. In Section 5, we quantify the role of this mechanism and contrast it with earlier models of limited participation, such as Saito (1995) and Basak and Cuoco (1998).

over in July), respectively. All returns are real and are obtained by deflating nominal returns with the consumption deflator series available in the same data set.
The Risk-Free Rate. Turning to the risk-free rate, the mean is 1.3%, which compares well to the low average interest rate of 1.9% in the data. It is important to note that the low risk-free rate is helped by the fact that the model abstracts from long-run growth and preferences are of the Epstein-Zin form. To see this, consider the following expression for the log risk-free rate, which holds as a fairly good approximation:\(^6\)

\[
r_f^t \approx -\ln \beta + \rho^h E_t \left( \Delta \ln c_{t+1}^h \right) + \kappa, \tag{3}
\]

where \(\kappa\) contains terms that involve the volatility of consumption and wealth, which turns out to be secondary for the present discussion.\(^7\)

With secular growth, the consumption growth term on the right-hand side would be non-zero—unlike in the present model—pushing the average risk-free rate up. For example, taking an annual growth rate of 1.5%, and setting \(\rho^h = 3.33\) as calibrated above, would imply \(r_f^t = 5.85\%\). As is well-known, this “risk-free rate puzzle” is even more severe with CRRA utility, because in this case it would be the risk aversion parameter that would appear in front of the second term, which is \(\alpha^h = 6\) in this case, implying \(r_f^t = 10.2\%\).

This discussion reiterates the well-known point that models with CRRA utility functions and long-run growth that match the equity premium typically imply a high average interest rate. Epstein-Zin preferences mitigate this problem if one assumes an EIS that is higher than the reciprocal of the risk aversion parameter, as is the case here.

Another well-documented feature of the interest rate—and as it turns out, a challenging one to explain—is its low volatility. The standard deviation is 5.44% in our historical sample, although different time periods (such as the post war sample) can yield values as low as 2% per year (see, e.g., Campbell 1999 for a discussion). The corresponding figure is 6.65% in the model (and further falls to 4.1% with endogenous labor supply below). Although this figure is higher than the empirical values, the low volatility of the interest rate has turned out to be quite difficult to generate, especially in production-based asset pricing models such as ours. For example, as we report in table 3, this volatility is 24.6% in Boldrin et al. (2001) and 10.6% in Danthine and Donaldson (2002); it is 11.5% in Jermann (1998) (not reported). Thus, the present model provides a step in the right direction. So, what explains the relatively low variability of interest rates in the model?

\(^6\)For an exact derivation of this expression, human wealth would need to be tradeable. Although this is not the case in the present model, the equation holds fairly well and provides a useful approximation.

\(^7\)Specifically, \(\kappa \equiv \frac{\theta - 1}{\theta} \sigma_\omega^2 - \frac{\theta}{2 \rho^2} \sigma_\nu^2\), where \(\theta = \frac{1 - \alpha^h}{1 - \rho^h}\), and the first volatility refers to the return on total wealth (including human wealth), whereas the second one is for consumption growth.
To understand the mechanism, consider the bond market diagram in Figure 1. The left panel depicts the case of a representative agent with a low EIS, which is a feature common to the models mentioned above. For example, both the endogenous and the external habit models imply a low EIS (despite differing in their risk aversion implications). With a low EIS, however, the interaction of the resulting inelastic (steep) bond demand curve with a bond supply that is *perfectly inelastic* at zero (because of the representative-agent assumption) means that even small shifts in the demand curve—due to labor income shocks and the consequent change in the demand for savings—generate large movements in the bond price, and hence, in the risk-free rate.

In the present model, the mechanism is different. First, for the following discussion it is convenient to label non-stockholders’ bond decision rule as the “bond demand” and the negative of stockholders’ bond decision rule as the “bond supply.” Now, notice that the majority—80%—of the population (non-stockholders) have a very low EIS as before, implying very inelastic bond demand (right panel). Turning to bond supply, the key difference here is that it is not inelastic at all. In fact, the stockholders’ supply curve is rather flat, both because of their higher EIS and because they have another asset—equity in the firm—that can act as a partial substitute for bond. As a result, a shift in the bond demand curve (resulting from fluctuations in non-stockholders’ labor income) of similar magnitude as before now results in smaller fluctuations in the interest rate, and the rest is reflected in the variability of trade volume.
The Price-Dividend Ratio. The average price-dividend (P/D) ratio in the CONS model is 27.2, which is about 20% higher than the average of 22.1 in the data. Its volatility is 26.6%, which compares fairly well with the empirical figure (26.2%). Finally, the volatility of dividend growth is 19.1%, which is too volatile compared to the 13.4% figure in the US data. This high volatility is due to the leverage in the capital structure and is one of the dimensions that the labor supply choice will help with.

4.1.1 The Role of Preference Heterogeneity

One advantage of Epstein-Zin utility is that it allows us to easily examine the impact of different preference parameters, as well as heterogeneity in these parameters, on asset prices. We conduct three experiments reported in the last three columns of Table 2. First (column 3), we keep all aspects of the baseline parameterization intact, but only increase non-stockholders’ EIS from 0.1 to 0.3, which eliminates all preference heterogeneity from the model. With this change, the equity premium falls significantly, from 5.5% to 2.44%, and the volatility of the premium falls from 21.9% to 15.3%. More importantly, the price of risk falls from 0.25 to 0.16. Moving down the column, the volatilities of all variables go down—by 30% to 50%. This makes some variables, such as the P/D ratio, too smooth compared to data, while bringing some others closer to their empirical counterparts, such as the interest rate and dividend growth volatilities. Overall, these results show that the EIS of non-stockholders has a major impact on asset prices, perhaps most importantly on the equity premium and the Sharpe ratio, which are key statistics that this model seeks to explain.

Second, an alternative way to eliminate preference heterogeneity is by reducing the EIS of stockholders from 0.3 to 0.1, which, as could be anticipated, has qualitatively the opposite effect (column 4). The equity premium now increases to 7.65%, but the volatility is also higher to 27.2%. As a result, the rise in the Sharpe ratio remains rather modest: it is 0.28, up from 0.25. Finally, other volatilities, that of the interest rate, P/D ratio, and dividend growth, are also significantly higher. These results show that the EIS for stockholders has a larger effect on volatilities, but a smaller one on the price of risk.\footnote{Stockholders’ EIS has a smaller effect on the price of risk than non-stockholders’ EIS precisely because the former group has access to better ways to smooth consumption. Consequently, when their EIS is lowered they respond by choosing a smoother consumption process, which keeps the rise in the Sharpe ratio small.}

Third, in the last column, we examine the effect of non-stockholders’ risk aversion, by doubling it to 12. Comparing column 5 to the baseline case in column 2 shows that this change has a minor effect, if at all, across the board. Surprisingly, doubling the risk aversion
of 80% of the population has very little impact on the unconditional moments of asset prices. Loosely speaking, this is due to the fact that non-stockholders’ only direct effect on asset prices is through the bond market, and their (precautionary) bond demand is largely determined by their EIS, but very little influenced by their risk aversion.\(^9\)

### 4.2 Asset Prices with Endogenous Labor Supply

In the previous section we have found that the benchmark model with inelastic labor supply generated plausible behavior for the unconditional moments of stock and short-term bond returns. Nevertheless, labor supply choice is central for any serious macroeconomic analysis. Therefore, we now relax the inelastic labor supply assumption and consider the two utility functions—Cobb-Douglas and GHH—described above.

**Results.** Table 3 reports the results (columns 2-4). To provide a comparison, the last two columns display the same set of statistics from two leading macro-asset pricing models proposed in the existing literature, namely Boldrin et al. (2001) and Danthine and Donaldson (2002). The first paper features an endogenous labor-leisure choice, whereas the latter paper has inelastic labor and is more comparable to the model analyzed in the previous section.

With Cobb-Douglas utility (CD model, column 2), the first point to observe is the rather large fall in the equity premium, which is now 2.65% (compared to 5.5% with inelastic labor supply), accompanied by a smaller fall in the volatility (which is 15.4% compared to 21.9% before), resulting in a fall in the Sharpe ratio from 0.25 to 0.17. Moving down the column, notice that both the risk-free rate and dividend growth are less volatile than before, and now are closer to their respective empirical values. The P/D ratio is also less volatile and is now too smooth compared to data. Overall, we view these results as a step backward compared to the model with inelastic labor supply.

We next turn to the GHH model in column 3, which delivers a more respectable equity premium level of 4.2%, with a volatility of 17.5%. The resulting Sharpe ratio is 0.24, slightly lower than in the model with inelastic labor supply. Furthermore, the volatility of the risk-free rate is 4.1%, which is lower than the model with inelastic labor supply (at 6.65%). As noted earlier, this low volatility is also an important improvement of this model over earlier production economy models. As for the price-dividend ratio, the volatility is about 70% that

\(^9\)Furthermore, in the web appendix we also report the results on asset price dynamics for \(\alpha^n = 12\) and find that they are very similar to the case with \(\alpha^n = 6\).
in the data, but higher than in the model with CD preferences. The volatility of dividend growth now falls to 11.2% annually. Overall, the model with GHH preferences display asset pricing behavior that is comparable to the model with inelastic labor supply.

In the next column, we explore the impact of lowering the Frisch elasticity closer to the values reported in the micro literature. We set $1/\gamma$ to 0.5 following Domeij and Floden (2006), who survey these micro studies and argue that part of the reason for the very low estimates is the bias arising from ignoring borrowing constraints. This change in calibration has a relatively modest effect on statistics: most notably, the equity premium falls slightly to 4% and the Sharpe ratio to 0.22. As we discuss in Section 6.1, the main drawback of this calibration is that it implies a labor hours process that is too smooth, which is also why macroeconomists typically use higher values similar to that in our baseline calibration.

It is useful to compare these results to earlier studies. In the working paper version, Boldrin et al. (2001) show that introducing a flexible labor supply choice reduces the equity premium substantially, from 4.45% to 0.15%, and the Sharpe ratio from 0.27 to 0.03.$^{10}$

---

**Table 3: Unconditional Moments of Asset Returns, Model with Elastic Labor Supply**

<table>
<thead>
<tr>
<th>Endog. leisure?</th>
<th>US Data</th>
<th>Model</th>
<th>BCF</th>
<th>D-D</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>A. Stock and Bond Returns</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$E(R_s - R_f)$</td>
<td>6.17</td>
<td>2.65</td>
<td>4.21</td>
<td>4.03</td>
</tr>
<tr>
<td>$\sigma(R_s - R_f)$</td>
<td>19.4</td>
<td>15.4</td>
<td>17.4</td>
<td>18.1</td>
</tr>
<tr>
<td>$\sigma(R_s)$</td>
<td>18.7</td>
<td>14.8</td>
<td>16.5</td>
<td>17.9</td>
</tr>
<tr>
<td>$E(R_s - R_f) / \sigma(R_s - R_f)$</td>
<td>0.32</td>
<td>0.17</td>
<td>0.24</td>
<td>0.22</td>
</tr>
<tr>
<td>$E(R_f)$</td>
<td>1.94</td>
<td>2.87</td>
<td>1.42</td>
<td>1.73</td>
</tr>
<tr>
<td>$\sigma(R_f)$</td>
<td>5.44</td>
<td>4.91</td>
<td>4.10</td>
<td>4.46</td>
</tr>
<tr>
<td><strong>B. Price-Dividend Ratio</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$E(P_s/D)$</td>
<td>22.1</td>
<td>25.7</td>
<td>24.7</td>
<td>25.9</td>
</tr>
<tr>
<td>$\sigma(log(P_s/D))$</td>
<td>26.3</td>
<td>13.6</td>
<td>17.8</td>
<td>19.1</td>
</tr>
<tr>
<td>$\sigma(\Delta log D)$</td>
<td>13.4</td>
<td>14.8</td>
<td>11.2</td>
<td>11.9</td>
</tr>
</tbody>
</table>

*Boldrin et al. (2001) report the time average of the conditional Sharpe ratio, $E(R_s-R_f) / \sigma(R_s)$, instead of the unconditional Sharpe ratio reported in the present paper. The statistics from Boldrin et al. (2001, BCF) refer to their benchmark model (called the "preferred two sector model") and has the best overall performance. The statistics from Danthine and Donaldson (2002, D-D) are from their Table 6, right column with smoothed dividends, which generates the best overall performance. A “—” indicates that the corresponding statistic has not been reported in that paper. In column 4, the Frisch elasticity is set to 0.5, and the $\sigma_D$ is recalibrated to 1.3% per quarter to match output volatility.

$^{10}$See Boldrin et al. (1999, table 2). The statement in the text is based on comparing the results in column 4 to column 8.
One goal of their paper, then, is to identify some prominent labor market frictions that overcome this difficulty associated with endogenous labor supply. Their baseline model features such a framework that matches the historical equity premium as well as a number of other unconditional moments (reported in column 4 of Table 3 below). In comparison, here flexible labor supply choice has a smaller negative impact on the Sharpe ratio, especially with GHH preferences. An important reason for this difference is that we consider preferences that are non-separable between consumption and leisure, whereas these authors study a separable specification that is linear in leisure. Another reason is that with GHH preferences, labor supply is entirely determined by the substitution effect in response to wage movements and is always procyclical. Therefore, they do not provide smoothing against procyclical income fluctuations.\footnote{More precisely, Boldrin et al. (2001) employ a specification where utility is linear in leisure and separable from consumption. Consequently, fluctuations in labor hours have no direct effect on utility, which makes it relatively costless to smooth consumption fluctuations by adjusting one’s labor supply. The only loss comes from the fact that to smooth fluctuations, labor supply would need to move in the opposite direction dictated by the substitution effect: rise in recessions when wages are low and fall in expansions when wages are high. With non-separable preferences, agents do care about fluctuations in leisure as well as how leisure comoves with consumption, which makes it more costly to adjust labor supply to suppress fluctuations in the marginal utility of consumption. Second, with GHH preferences there is no wealth effect on labor supply choice, so labor hours are strongly procyclical due to the substitution effect of wages over the business cycle, making it an even less effective tool for smoothing fluctuations in marginal utility. As a result, the price of risk does not fall as much in this framework when labor supply choice is endogenized.}

\section*{4.3 The Dynamics of Asset Prices}

\subsection*{4.3.1 Countercyclical variation in conditional moments}

We now examine the extent to which the limited participation model captures some salient aspects of asset price dynamics. As mentioned earlier, some of these features, such as the countercyclical price of risk, have been difficult to generate in some asset pricing models. The results reported below are from the GHH model, but we found these statistics to be remarkably robust across the different specifications analyzed in this paper.\footnote{The counterparts of the tables below for the CONS and CD models are reported in the web appendix.}

\subsection*{4.3.2 Mean Reversion and Predictability of Returns}

We begin with the price-dividend ratio, which is procyclical in the model, consistent with empirical evidence, with a correlation of 0.73 with output. The procyclicality follows from the
Table 4: The Autocorrelation Structure of Key Financial Variables

<table>
<thead>
<tr>
<th></th>
<th>Autocorrelation at lag j (years):</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>1</td>
</tr>
<tr>
<td>$r^s - r^f$</td>
<td></td>
</tr>
<tr>
<td>US Data</td>
<td>.03</td>
</tr>
<tr>
<td>Model</td>
<td>-.03</td>
</tr>
<tr>
<td>$\sum_{i=1}^{j} \rho[(r^s-r^f)<em>{t-i}, (r^s-r^f)</em>{t-i}]$</td>
<td></td>
</tr>
<tr>
<td>US Data</td>
<td>.03</td>
</tr>
<tr>
<td>Model</td>
<td>-.03</td>
</tr>
</tbody>
</table>

Notes: Statistics from the model are calculated from annualized values of each variable, except for the stock price, which is taken to be the value at the end of the year.

fact that when the persistent technology shock hits the economy, the stock price capitalizes all future productivity gains upon impact and thus increases substantially, while the initial response of dividends is muted due to the gradual rise of the capital stock after the shock (because of adjustment costs), making the ratio of the two variables procyclical.

A second well-known observation concerning the equity premium is that it tends to revert to its mean over time. The first row of Table 4 displays the autocorrelations of the equity premium from the US data. Notice that these autocorrelations are close to zero, which makes it difficult to measure them precisely. As a result, they are not uniformly negative. To circumvent this problem, an alternative statistic used in the literature aggregates consecutive autocorrelation coefficients, which reveals a stronger pattern of mean reversion (row 3). The second and fourth rows display the model counterparts of the two measures of mean reversion, which are consistent with the signs and rough magnitudes of these statistics in the data.

The finding of mean reversion is a clear departure from the martingale hypothesis of returns and is closely linked to the predictability of returns documented by Campbell and Shiller (1988), among others. To show this, we first regress log stock returns on the log price-dividend ratio using the historical US data (panel A of Table 5). The well-known pattern documented in the literature can be seen here: the coefficients are negative, indicating that a high price-dividend ratio forecasts lower returns in the future. Moreover, both the coefficients and the $R^2$ values are increasing with horizon. The model counterpart is reported in the last two columns. The coefficient estimates and the $R^2$s are consistent with empirical results: predictability is modest at the one-year horizon but increases steadily, reaching 35% at the seven year horizon. The coefficients also increase quickly first and then grow more slowly. Overall, the model generates significant predictability of stock returns.
Table 5: **Long-Horizon Regressions on Price-Dividend Ratio**

<table>
<thead>
<tr>
<th>Horizon (k)</th>
<th>US Data</th>
<th>Model</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Coefficient</td>
<td>$R^2$</td>
</tr>
<tr>
<td>A. Stock Returns</td>
<td>1</td>
<td>-.21</td>
</tr>
<tr>
<td></td>
<td>2</td>
<td>-.36</td>
</tr>
<tr>
<td></td>
<td>3</td>
<td>-.41</td>
</tr>
<tr>
<td></td>
<td>5</td>
<td>-.70</td>
</tr>
<tr>
<td></td>
<td>7</td>
<td>-.87</td>
</tr>
<tr>
<td>B. Excess Returns</td>
<td>1</td>
<td>-.22</td>
</tr>
<tr>
<td></td>
<td>2</td>
<td>-.39</td>
</tr>
<tr>
<td></td>
<td>3</td>
<td>-.47</td>
</tr>
<tr>
<td></td>
<td>5</td>
<td>-.77</td>
</tr>
<tr>
<td></td>
<td>7</td>
<td>-.94</td>
</tr>
</tbody>
</table>

Note: The table reports the coefficients and $R^2$ values of the regression of log stock returns (top panel) and log excess returns (bottom panel) on the log price-dividend ratio. Horizon (k) is in years.

Turning to excess returns in panel B (first four columns), the predictability in the US data is evident again. The last two columns report the same regressions using simulated data. The coefficients on the price-dividend ratio are negative and increase in absolute value with horizon consistent with the data. However, the amount of predictability is smaller than in the data: the $R^2$ of the regression is only 12% at the seven year horizon compared to 33% in the data. Thus, the model is qualitatively consistent with excess return predictability, but does not quantitatively capture the total magnitude observed in the US data.

### 4.3.3 Countercyclical Variation in the Price of Risk

We now examine the cyclical behavior of conditional moments. Figure 2 plots the typical simulated paths of the expected equity premium and Sharpe ratio against output over a period of 200 years.\(^{13}\) Consistent with the US data (Chou et al. (1992)), both the expected premium and the Sharpe ratio rise visibly in recessions and fall in expansions (correlation with output: $-0.55$ and $-0.63$, respectively). Moreover, the conditional volatility is also countercyclical with a correlation with output of $-0.42$ (to save space, the plot is not included in the figure). The conditional moments are also quite variable: the 95% confidence interval

\(^{13}\)Each series is rescaled by its standard deviation and shifted vertically to fit in the same graph. Thus, while the amplitudes are not informative, the comovement between the series is preserved.

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for the expected excess return extends from 1.3% to 4.8%, and the 95% confidence interval for its conditional volatility extends from 9.1% to 18.5%. The fact that expected returns are more variable than conditional volatility (as measured by their coefficient of variation) results in the countercyclicality of the Sharpe ratio mentioned above. With few exceptions,\textsuperscript{14} this countercyclicality of the market price of risk has been difficult to generate in consumption-based asset pricing models. In the next section, we explain the new mechanism in this model which generates this result.

5 Understanding the Sources of Equity Premium

The large and countercyclical equity premium arises from the interaction of (i) limited stock market participation; (ii) the low EIS of non-stockholders; and (iii) consumption smoothing in the face of persistent (aggregate) labor income shocks. What distinguishes the mechanism in this paper from earlier models of limited participation is (ii) and (iii), which makes the timing of trade in the bond market the driving force behind the high equity premium. In this section we quantify this mechanism. Because labor supply choice adds another level of complexity to the analysis, below we focus on the CONS model.

\textsuperscript{14}Most notably, Campbell and Cochrane (1999) and Bansal and Yaron (2004).
5.1 Why Is There a High Equity Premium?

The annual standard deviation of consumption growth in the CONS model is 3.61% for stockholders but only 1.48% for non-stockholders. To quantify the sources of stockholders’ consumption volatility, we first substitute the equilibrium conditions (3-a,b) as well as the expression for dividends into the budget constraint of a stockholder, and after some straightforward manipulation we obtain:

\[ c^h = \left( W + \frac{\theta Z K^g L^{1-\theta} - I}{\mu} \right) - \frac{(B - P^f B')}{\mu} \]

This expression provides a useful decomposition. Part of the variation in stockholders’ consumption comes from aggregate sources, that is, fluctuations in wage and capital income, which is denoted by \( A^h \). The scaling factor \( 1/\mu \) that appears in \( A^h \) reflects the fact that aggregate capital income risk is concentrated among stockholders, since they are the sole capital owners in this economy. However, this risk only has a modest contribution to the volatility of their consumption, as will become clear shortly. A second part of variability arises from trade with the non-stockholders in the bond market: the component denoted by \( a^h \) measures the net interest payments received by non-stockholders, which is made by stockholders indirectly through the firm they own. Using equation (4), stockholders’ consumption growth volatility can be approximated as:

\[ \text{var} (\Delta \log c^h) \approx \text{var}(\Delta \log A^h) + \text{var} \left( \frac{\Delta a^h}{A^h} \right) + 2\text{cov} \left( \Delta \log A^h, -\Delta a^h \right). \]  

Similarly, the consumption of non-stockholders can be written as: \( c^n = W + (B - P^f B') / (1 - \mu) \), and the variance of their consumption growth is obtained by replacing \( A^h \) and \( a^h \) in equation (5) with \( A^n \equiv W \) and \( a^n \equiv -a^h \mu / (1 - \mu) \), respectively.

The top panel of Table 6 shows that interest payments are very small on average, making up a small fraction of consumption: \( E(a^i) / E(c^i) \) is less than 1.1% for both stockholders and non-stockholders. However, interest payments are volatile, and more importantly, vary systematically with the business cycle. This is shown in the lower panel, which displays the fraction of consumption growth variance explained by each of the three terms in equation (5). For stockholders, only 19% of consumption growth variance is attributable to fluctuations in aggregate income, despite the fact that this component makes up nearly
Table 6: Decomposition of Consumption Volatility and Equity Premium

<table>
<thead>
<tr>
<th></th>
<th>Fraction of $E(c^i)$ explained by: $E(A^i)/E(c^i)$</th>
<th>$E(a^i)/E(c^i)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Stockholders</td>
<td>1.011</td>
<td>0.011</td>
</tr>
<tr>
<td>Non-stockholders</td>
<td>0.995</td>
<td>0.005</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th></th>
<th>Fraction of $\sigma^2(\Delta \log c^i)$ accounted for by: $\sigma^2(\Delta \log A^i)$</th>
<th>$\sigma^2(\Delta a^i)$</th>
<th>$2\sigma(\Delta \log A^i, -\Delta a^i)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Stockholders</td>
<td>$(3.61%)^2$</td>
<td>0.185</td>
<td>0.340</td>
</tr>
<tr>
<td>Non-stockholders</td>
<td>$(1.48%)^2$</td>
<td>3.13</td>
<td>0.61</td>
</tr>
</tbody>
</table>

\* $i=\text{h,n}$. See the text for the definitions of $a^i$ and $A^i$. 

all of their average consumption (row 1). Hence, the concentration of aggregate capital income risk, included in $A^h$, contributes only modestly to consumption fluctuations, and consequently to the equity premium. The main source of volatility for stockholders comes from the bond market: interest payments, $a^h$, account for the remaining three-quarters of variance ($0.34 + 0.475$). What is really crucial for this extra volatility is the timing of interest payments: $\text{corr}(\Delta \log A^h, \Delta a^h) = -0.953$, which means that the payments received by the non-stockholders increase exactly when aggregate income falls, i.e., in recessions. Therefore, consumption smoothing for non-stockholders comes at the expense of large fluctuations in stockholders’ consumption, so the covariance term in the third column accounts for nearly half of the total volatility. The flip side of this story is seen in the variance of non-stockholders: $\text{var}(\Delta \log c^n)$ would be more than three times higher, were it not for the consumption smoothing provided by stockholders through the bond market.

A Comparison to Earlier Models. The mechanism proposed in the present paper differs from earlier models with limited participation in two key dimensions. First, in Saito (1995) and Basak and Cuoco (1998) non-stockholders begin life with some financial wealth but do not have a labor income flow. The only way to sustain consumption is then by investing this wealth in the bond market. As a result, each period non-stockholders receive substantial interest payments from stockholders, which in turn leverages the capital income of the latter group and amplifies their consumption volatility. Although this is a potentially powerful mechanism, for it to work, interest payments needs to be substantial—as large as the consumption of non-stockholders—which in turn implies that non-stockholders must hold a large fraction of aggregate wealth. This counterfactual implication has been one of the main
criticisms raised against these models. In contrast, as noted above, the interest payments in
the present model are very small, and as we shall see below, the fraction of aggregate wealth
held by non-stockholders is also small and consistent with empirical evidence. Instead, the
amplification here is generated by the cyclical nature of these payments that serve to insure
non-stockholders against aggregate fluctuations.

A second difference is that preference heterogeneity is crucial in the present paper, which
is not the case in these earlier papers. For example, if preference heterogeneity is eliminated,
as in column 3 of Table 2, stockholders’ consumption becomes less volatile than that of non-
stockholders. (Specifically, the ratio $\sigma (\Delta \log c^h) / \sigma (\Delta \log c^n)$ falls by three fold, from 2.4 to
0.78.) This is because stockholders have access to two separate assets—one of which is the
capital stock in the firm—that they can use to smooth consumption effectively, and without
preference heterogeneity they choose to do exactly that. Therefore, preference heterogeneity
is a key component of the mechanism in this paper. This is clearly in contrast to the previous
papers mentioned above, where non-stockholders have no choice but to lend to stockholders
simply to survive, which amplifies the consumption volatility of the latter group regardless
of preference heterogeneity.

Empirical Evidence. Finally, the high volatility of stockholders’ consumption relative to
that of non-stockholders is well documented empirically. For example, Mankiw and Zeldes
(1991) find the ratio of standard deviations to be 1.6 using PSID data, although their con-
sumption measure consists of only food expenditures, which is likely to understate the true
volatility ratio for non-durables. Attanasio, Banks, and Tanner (2002) use expenditures on
non-durables and services from the UK Family Expenditure Survey and calculate stockhold-
ers’ volatility to be 1.5 to 2 times larger than that of non-stockholders. Another piece of
evidence comes from Ait-Sahalia, Parker and Yogo (2004) who document that the sales of
luxury goods (such as expensive French wine, charitable donations, rents on luxury condos
in Manhattan, etc.) display volatility that exceeds the standard deviation of aggregate non-
durables consumption by a factor of 5 to 10. They interpret this finding as indicative of
highly volatile expenditures by the very wealthy.

15These figures are likely to be downward biased because existing measures of stockholders’ consumption
are based on micro data sets that contain few “extremely rich” households, i.e., those in the top 1% of the
wealth distribution. At the same time, these households own nearly half of all stocks, and the top 0.5% own
37% of all stocks (Poterba and Samwick (1995)).
5.2 Why Is The Equity Premium Countercyclical?

An intuitive way of seeing why the equity premium is countercyclical in the model begins with observing that stockholders’ consumption growth (and, consequently, marginal utility) becomes more volatile during bad times: the (conditional) standard deviation is 32% higher when the economy is in a recession than when it is in a boom. As a result, they demand a higher (by 38%) premium for holding the equity risk during recessions.

The natural question that follows, of course, is why stockholders’ consumption volatility is countercyclical. To answer this question, two points should be noted. First, a well-known feature of the consumption-savings problem in the presence of uninsurable income risk is that the value function of individuals has more curvature at lower wealth levels and less curvature at higher wealth levels (e.g., Deaton (1992)). Second, as we further elaborate in Section 6.2, non-stockholders own substantially less wealth per capita than stockholders. To be precise, on average a non-stockholder’s wealth is equal to 36% of his annual income, whereas a stockholder’s wealth is 10.2 times his annual income. As a result, non-stockholders are on the part of their value functions that has high curvature and a relatively small change in their wealth holdings changes their attitudes toward risk and therefore significantly changes their precautionary savings demand. This is clearly not the case for stockholders: the lowest wealth these agents hold in our simulated sample is 5.8 times their average annual income (whereas it is −20% of average income for non-stockholders). As a result, stockholders’ demand for consumption smoothing changes little over the business cycle. As an illustration of this point, Figure 3 plots the normalized curvature of the value function (i.e., risk aversion with respect to wealth gambles), $-\omega V_{\omega}^{\alpha}/V_{\omega}^{\alpha}$, for each type of agent as a function of realized output using simulated data from the CONS model. We normalize the average value to 1 for both groups so that the vertical axis can be interpreted as the percentage change in risk attitudes over the business cycle. The effect described above can be clearly seen here: non-stockholders’ risk aversion increases significantly in recessions (and consequently, their demand for insurance), whereas it barely moves for stockholders. Combining this countercyclical change in non-stockholders’ demand for consumption smoothing with the mechanism described in the previous section implies that there will be more trade in the risk-free asset

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16 A recession is defined as the states of the world when output is one standard deviation or more below the mean: $Y < E(Y) - \sigma(Y)$. An expansion is defined analogously: $Y > E(Y) + \sigma(Y)$. If the definition is changed to simply above or below the mean, we get: $E[\sigma_t(\Delta c_t^{h+1})|Y < E(Y)] / E[\sigma_t(\Delta c_t^{h+1})|Y > E(Y)] = 1.18$.

17 Notice that risk aversion does not only depend on the curvature of the utility function $\alpha$, but also depends on the market structure. In particular, risk aversion is constant and equal to $\alpha$ under complete markets, whereas with incomplete markets it is amplified for a given $\alpha$ and varies with an individual’s resources.
market in recessions. This is exactly what happens: \( \rho (|a_t^h|, Y_t) = -0.55 \), which shows that the trade volume as defined above (i.e., the average size of interest payments) increases during recessions. Notice also that, despite the 32% rise in stockholders’ consumption growth volatility mentioned above, non-stockholders’ consumption volatility barely changes (goes down by 4%) as a result of this increased trading. As a result, aggregate consumption growth volatility is only 6% higher in recessions and therefore masks the large rise in the volatility of stockholders who make up only a small fraction of the population.

Finally, the stock price is obtained as the discounted sum of future dividends using stockholders’ consumption growth as the discount factor (from (1)). Thus, the countercyclical variation in the latter also generates stock prices—and hence returns—that are more volatile during recessions. Since both the Sharpe ratio and the volatility is countercyclical, the product of the two—the conditional equity premium—is also countercyclical.

## 6 Model Results: Macroeconomic Quantities

In this section we examine the implications of the model for macroeconomic quantities along two dimensions. We first analyze the performance of the model for business cycle statistics and then turn to the cross-sectional implications.
Table 7: Business Cycle Statistics in the US Data and in the Model

<table>
<thead>
<tr>
<th>Leisure preferences</th>
<th>Data Volatilities</th>
<th>Model Volatilities</th>
<th>BCF linear</th>
<th>D-D none</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>CD</td>
<td>GHH Baseline</td>
<td>GHH low Frisch</td>
<td></td>
</tr>
<tr>
<td>$\sigma (Y)$</td>
<td>1.89</td>
<td>1.97</td>
<td>1.95</td>
<td>1.96</td>
</tr>
<tr>
<td>$\sigma (C) / \sigma (Y)$</td>
<td>0.40</td>
<td>0.92</td>
<td>0.78</td>
<td>0.75</td>
</tr>
<tr>
<td>$\sigma (I) / \sigma (Y)$</td>
<td>2.39</td>
<td>1.38</td>
<td>1.76</td>
<td>1.86</td>
</tr>
<tr>
<td>$\sigma (L) / \sigma (Y)$</td>
<td>0.80</td>
<td>0.07</td>
<td>0.50</td>
<td>0.32</td>
</tr>
</tbody>
</table>

Cross-correlation with Output

<table>
<thead>
<tr>
<th></th>
<th>Data</th>
<th>CD</th>
<th>GHH Baseline</th>
<th>GHH low Frisch</th>
<th>BCF linear</th>
<th>D-D none</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\rho (Y, C)$</td>
<td>0.76</td>
<td>0.99</td>
<td>0.99</td>
<td>0.99</td>
<td>0.95</td>
<td>0.96</td>
</tr>
<tr>
<td>$\rho (Y, I)$</td>
<td>0.96</td>
<td>0.99</td>
<td>0.94</td>
<td>0.97</td>
<td>0.97</td>
<td>0.93</td>
</tr>
<tr>
<td>$\rho (Y, L)$</td>
<td>0.78</td>
<td>0.96</td>
<td>0.98</td>
<td>0.99</td>
<td>0.86</td>
<td>—</td>
</tr>
</tbody>
</table>

*The statistics for the US data are taken from Boldrin et al. (2001, Table 1). The model statistics are computed after simulated data have been aggregated to quarterly frequency, logged, and then HP filtered with a bandwidth parameter of 1600.

6.1 Business-Cycle Implications

Table 7 displays standard business cycle statistics from the US data as well as their counterparts from the CD model (column 2) and GHH model (columns 3-4). As before, the last two columns report the corresponding figures from Boldrin et al. (2001) and Danthine and Donaldson (2002) for comparison.

In column 2, the first row reports the volatility of output, which was one of our two calibration targets and is matched fairly closely. The volatility of consumption (normalized by the volatility of output) is about 0.4 in the US data, but is overstated at 0.92 in the model. In contrast, investment is too smooth, with a volatility of 1.38 in the model compared to 2.39 in the US data. Although these discrepancies are bothersome, the most glaring shortcoming of the CD model is seen in the behavior of labor hours, which is an order of magnitude too smooth compared to data (0.07 versus 0.80). However, the behavior of the aggregate labor supply masks significant heterogeneity between the two groups’ behavior. In particular, labor hours are quite volatile but countercyclical for stockholders ($\sigma (l^h) / \sigma (Y) = 0.77$; correlation with output: $-0.96$), whereas it is procyclical but smoother for non-stockholders because of their low Frisch labor supply elasticity ($\sigma (l^n) / \sigma (Y) = 0.24$; correlation: $+0.97$). The behavior of aggregate hours largely mirrors that of non-stockholders (i.e., procyclical and smooth), because they make up a substantial majority of the population. However, aggregate hours
are even smoother than that of non-stockholders, exactly because the labor supply of the two groups moves in opposite directions over the business cycle, partly canceling each other. At the same time, the procyclicality of aggregate hours is an encouraging finding, in light of earlier studies that have consistently found countercyclical movements in representative agent models with capital adjustment costs (for discussions, see Boldrin et al. (2001) and Uhlig (2006)). Overall, however, we conclude from these results that Cobb-Douglas utility is not a promising direction to pursue in this framework for endogenizing labor supply.

Turning to the GHH model, the volatility of consumption is lower, whereas the volatility of investment is higher compared to the CD model, and the model is now in the same ballpark as the other two production-based asset pricing models reported in the table. The more significant improvement is in the volatility of labor hours, where the CD model failed most dramatically: now, the volatility of hours is much higher than before (equal to half the volatility of output) and is closer to the data. This improvement is due to the flexibility afforded by GHH preferences for calibrating the Frisch elasticity independently of the EIS. Aggregate hours (as well as each group’s labor hours) are also procyclical, which is due to the lack of wealth effects with GHH preferences. Finally, column 3 reports the statistics from the GHH model when Frisch elasticity is lowered to 0.5. As seen here, the major impact of this change is a large fall in the volatility of labor hours, which is a major reason why macroeconomic studies typically use higher values as in our baseline calibration. To sum up, the macroeconomic implications of GHH preferences are tolerable—i.e., comparable to earlier work—and are significantly better than the CD model.

The procyclicality of aggregate hours is due to the cross-sectional heterogeneity generated by limited participation. To see why this helps, first note that adjustment costs generate a volatile and procyclical stock price (as in the data). In a representative agent model, such as the one studied by Boldrin et al (2001), this implies a large positive wealth effect during expansions, which overcomes the substitution effect from a higher wage rate and causes labor hours to drop during expansions. In contrast, in the present model, 80% of the population are non-stockholders and therefore do not experience the described wealth effect. Instead, they mainly respond to the substitution effect of higher wages and increase their labor hours. Although stockholders behave in a similar fashion to the representative agent described above, they only make up a small fraction of the population. Overall, this result suggests that the countercyclicality of labor hours in earlier studies arises from the interaction of adjustment costs with the representative agent assumption. To increase the volatility of aggregate hours in our framework, one would need to increase the Frisch elasticity of non-stockholders and/or assume a smaller fraction of households to be stockholders. Unfortunately, a higher Frisch elasticity also means a higher EIS for non-stockholders, which contradicts the main empirical evidence this paper builds upon.
6.2 Distributional Implications

Given the central role played by heterogeneity for the asset pricing results of this paper, an important question to ask is whether or not the implied cross-sectional behavior is broadly consistent with empirical evidence. We first look at the wealth distribution implied by the GHH model (Figure 4), which reveals substantial inequality between the two groups: on average stockholders own 92% of aggregate wealth, whereas non-stockholders own the remaining 8%. (These shares are fairly robust across specifications: stockholders’ share is 86% in the CD model and 89% in the CONS model.) These figures compare fairly well to the US data, where in the 1980s and 1990s stockholders owned an average of 82% of all net worth (fungible wealth, including housing) and more than 90% of financial assets (see Poterba and Samwick (1995) and Guvenen (2006)). As noted above, this implication for wealth inequality is an important difference between the present model and earlier models with limited participation, which required substantial wealth holdings for non-stockholders.

This discussion, however, raises a question of its own: Is the amount of trade in the risk-free asset implied by the model quantitatively plausible, or do we need a substantial amount of trade to sustain the mechanism that generates the equity premium? One way to answer this question is by looking at the trade volume, which we define as the (absolute) change in non-stockholders’ bond holdings during a given year compared to the level of their bond holdings: $E(\Delta B)/E(B)$. This measure is less than 1.1% per year across all versions of the model reported in Tables 2 and 3. This modest figure perhaps does not
come as a surprise, given the gradual shifts in the wealth shares shown earlier in Figure 4. As an alternative measure, we can compare the size of non-stockholders’ per capita transactions to their wage income: \( E \left( \left| \frac{B_t - P_t B_t^f}{1-\mu} \right| \right) / E(W) \) is less than 1.9% annually across all simulations.\(^{19}\) For a non-stockholding household with an annual income of $50,000, this upper bound implies a net change, for example in their savings account, of $950 during the year. These relatively small payments entail large changes in stockholders’ consumption, because they are synchronized across all non-stockholders (i.e., they result from aggregate shocks) and the effect is concentrated among a small number of stockholders.\(^{20}\)

### 7 Extensions and Discussions

**Rising Participation in the Stock Market.** Given the various implications of limited participation that the preceding analysis has highlighted, a natural question to ask is: what are the implications of the rising participation in the stock market observed since the 1990s? A complete answer to this question would require solving for the transition path during this period, which will add substantially to the already high computational demands of the model. However, some insights can be gained by comparing outcomes across stationary equilibria with different participation rates, keeping in mind the well-known caveats associated with drawing inference from such a comparison for transitions.

We solve the GHH model with \( \mu \) set equal to 30%, corresponding to the high-participation economy.\(^{21}\) In this case, the equity premium is lower—3.8% compared to 4.2% before—since aggregate risk is now shared among a larger group of households. It is also less volatile, which mitigates the fall in the Sharpe ratio: 0.22 versus 0.24 before. It is important to note that the 3.8% figure is the ex ante equity premium, i.e., what investors expect to receive looking forward. Therefore, this lower ex ante premium in the new steady state does not necessarily imply that the realized premium during the transition is low. To the contrary, because equity becomes a less risky asset with higher participation, its price rises (27.8 compared to 24.7 in the first steady state) which is likely to generate a higher realized equity premium along

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\(^{19}\)To construct the measure, we first annualize the total transactions by adding up \( B_t - P_t^f B_t \) over 12 consecutive months and then take the absolute value.

\(^{20}\)To see this, note that \( E \left( \left| \frac{B_t - P_t^f B_t^f}{\mu} \right| \right) / E(Cs) = 4.9\% \).

\(^{21}\)Although the fraction of households who own any positive amount of equity has been around 50% in the United States, since 2000, many of the new participants hold very small amounts of equity. The 30% figure is chosen to roughly correspond to the set of households who own 99% of all equity outstanding (including indirect holdings through defined contribution plans; see, e.g., Poterba (2000)).
the transition due to capital gains. This would be consistent with the experience of the US economy after the 1990’s (with a booming stock market and high realized premium). However, these results should be viewed as suggestive about the potential impact of rising participation on asset prices, because it is not clear how long such a transition would take given that the new participants are entering the stock market with a substantially lower wealth level than existing stockholders.

**Endogenizing the Heterogeneity in the EIS.** We have assumed so far that the heterogeneity in the EIS is an exogenously given characteristic of stockholders and non-stockholders. We now show how this heterogeneity can be generated endogenously. To this end, assume that both agents have identical expected utility functions that feature “benchmark consumption levels”: \( u^i = \left( c^i - aC \right)^{1-\rho} / (1 - \rho) \), where \( C \) is the aggregate consumption in a given period, which is taken to be exogenous by individuals. Because these preferences are non-homothetic, the EIS of an individual endogenously rises with his consumption and therefore with his wealth. But note that wealth inequality in this framework is mainly due to limited participation and is quite robust to changes in the curvature of the utility of both agents (see Guvenen (2006) for a detailed analysis of this point). Therefore, with these new preferences stockholders continue to be much wealthier and consequently consume more than non-stockholders, endogenously generating the same kind of heterogeneity in the EIS assumed exogenously in the main framework above.

We solve the model with these preferences and set \( \rho = 3 \) and \( a = 0.6 \). When aggregate consumption is normalized to 1, the per capita consumption in this model is 1.55 for stockholders and 0.89 for non-stockholders, which generates \( eis^h \approx \frac{\rho^h}{(\rho^h - aC)} = 0.21 \) and \( eis^n \approx 0.11 \) (and stockholders own 91% of aggregate wealth as conjectured above). We set the stockholders’ EIS to a lower value than in the baseline calibration so as not to generate a risk aversion that is too low: with this utility function, it is approximately equal to \( 1/eis^h \approx 4.77 \). The resulting equity premium is 5.1% and the Sharpe ratio is 0.23. The volatility of the interest rate is 7.1%. One notable difference from before is seen in the dynamics of asset prices where the countercyclicality of the Sharpe ratio and equity premium become stronger (correlation with output \(-0.82\) and \(-0.90\), respectively), which is due to the fact that now the risk aversion of both agents changes over the business cycle in a countercyclical fashion.\(^{23}\)

\(^{22}\)In addition: \( E(R^f) = 1.65\% \), \( \sigma(R^f) = 6.41\% \), \( \sigma(\log(P^s/D)) = 23.5\% \), \( \sigma(\Delta \log(D)) = 13.1\% \)

\(^{23}\)Other statistics are as follows: \( E(R^f) = 1.45\% \), \( E(\log(P^s/D)) = 26.8 \), \( \sigma(\log(P^s/D)) = 24.8 \), \( \sigma(\log(P^s/D)) = 27.5 \). The results on asset price dynamics for the parameterizations in this section are reported in the web appendix.
8 Conclusions

We have studied the implications of a two-agent model with limited stock market participation and heterogeneity in the EIS for asset prices and macroeconomic quantities. Our findings suggest that the cross-sectional heterogeneity generated by the interaction of these two features could be important for several macro-asset pricing phenomena.

The model highlights a new mechanism for the equity premium that results from the limited consumption smoothing opportunities available to the majority of the population, who are non-stockholders. These households turn to the bond market, which effectively transfers their (aggregate) labor income risk and concentrates it among a small group of stockholders. As a result, these households demand to be compensated in the form of a high equity premium. In addition to generating broadly plausible behavior for asset prices, this mechanism is also consistent with the very low wealth holdings of non-stockholders in the US data, which has been problematic for previous models of limited participation. This significant skewness in wealth holdings, in turn, combines with the basic mechanism above to provide a novel explanation for the countercyclicality of the equity premium: essentially, because non-stockholders are very poor and poorly insured, they become effectively more risk averse during recessions when their wealth falls even further. Therefore, stockholders provide more insurance to these agents and bear more risk during these periods. This leads them to demand a higher ex ante premium to hold the equity risk. Finally, another implication of this framework is the fairly smooth interest rate process generated by the model despite the very low EIS of the majority of households (non-stockholders). This result is due to the higher EIS of the stockholders who are on the other side of the bond market.

We have also investigated the extent to which the labor supply decision can be endogenized in this framework. Cobb-Douglas utility gave rise to some counterfactual implications and therefore did not seem to offer a promising direction. In contrast, GHH preferences did not cause deterioration in asset pricing implications compared to the inelastic labor case and generated reasonable macroeconomic behavior.

A limitation of the present model as a macroeconomic framework is that it abstracts from long-run growth. Essentially, we assume that the economy is stationary in levels to circumvent the difficulties associated with obtaining a stationary wealth distribution with preference heterogeneity and long-run growth. In a recent paper, Chan and Kogan (2002) are able to generate a stationary wealth distribution by introducing “benchmark consumption levels” into preferences. Although incorporating a similar idea into this framework would be
a valuable extension, this would also introduce issues beyond the scope of our analysis. This remains an important area for extension of this model.

We hope that our results would also encourage further research on the reasons behind limited participation, which is not addressed in this paper. Furthermore, given the central role played by limited participation in this model, another important research avenue is to investigate the consequences of the recent trends in participation observed in most countries for asset prices as well as for wealth inequality and welfare.

References


