PORTFOLIO REBALANCING AND ASSET PRICING
WITH HETEROGENEOUS INATTENTION (*)

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(*) I am especially indebted to Matthias Kredler for his guidance. I thank Henrique Basso, Adrian Buss, James Costain, Juanjo Dolado, Andrés Erosa, Daria Finocchiaro, Robert Kirby, Hanno Lustig, Iacopo Morchio, Salvador Ortigueira, Alessandro Peri, Josep Pijoan-Mas, Carlos Ramirez, Rafael Repullo, José-Víctor Ríos-Rull, Manuel Santos, Pedro Sant’Anna, Hernán Seoane, Marco Serena, Nawid Siassi, Ctirad Slavík, Xiaojun Song, Nikolas Tsikas, two anonymous referees and the participants of several Conferences, Workshops and Seminars for their helpful criticism, suggestions and insights. This paper was previously circulated under the title «Asset Pricing with Heterogeneous Inattention». The views expressed in this paper are those of the author and do not necessarily represent the views of Banco de España or the Eurosystem.

Documentos de Trabajo. N.º 1633
2016
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ISSN: 1579-8666 (on line)
Abstract

Can households’ inattention to the stock market quantitatively account for the inertia in portfolio rebalancing? I address this question by introducing an observation cost into a production economy with heterogeneous agents. In this environment inattention changes endogenously over time and across agents. I find that inattention explains the inertia in portfolio rebalancing and its heterogeneity across households. Inattention also rationalises the limited stock market participation observed in the data, and improves the asset pricing performance of the model. Finally, I present a novel testable implication linking the effects of inattention on portfolio choices and asset prices to households’ funding liquidity.

Keywords: observation cost, limited stock market participation, equity premium.

JEL classification: G11, G12.
Resumen

¿Puede la desatención de los hogares a la bolsa de valores explicar cuantitativamente la inercia en el rebalanceo de las carteras? Para abordar esta pregunta, este documento introduce un coste de observación en una economía de producción con agentes heterogéneos. En este entorno, la desatención cambia de forma endógena en el tiempo y entre los agentes. En el análisis cuantitativo, la desatención explica la inercia en el rebalanceo de la cartera y su heterogeneidad entre los hogares. La desatención también racionaliza la limitada participación en el mercado de valores observada en los datos y mejora el rendimiento de los precios de los activos del modelo. Por último, se presenta una novedosa implicación comprobable que vincula los efectos de la desatención en las elecciones de cartera y los precios de los activos a la liquidez de financiación de los hogares.

Palabras clave: coste de observación, participación limitada en el mercado de valores, prima de riesgo.

Códigos JEL: G11, G12.
1 Introduction

Can households’ limited attention to the stock market quantitatively account for the inertia in households’ portfolio rebalancing? I address this question by introducing an observation cost into a production economy with heterogeneous agents, idiosyncratic labor income risk, and borrowing constraints. In this environment inattention changes endogenously over time and across agents. I discipline the quantitative analysis by calibrating the observation cost to match the duration of inattention of the median household observed in the data.

I find that inattention accounts for half of the inertia in portfolio rebalancing and its heterogeneity across stockholders. In the model, as it is in the data, wealthy stockholders invest much more actively than poor stockholders. Inattention also reconciles the amount of limited stock market participation observed in the data with a low per-period participation cost. In addition, I show that inattention improves the asset pricing performance of the model. Importantly, I highlight a novel testable implication that links households’ inattention to households’ funding liquidity: inattention matters quantitatively on the dynamics of portfolio rebalancing and asset prices only if borrowing constraints are tight enough.

This paper studies the role of households’ inattention by relaxing the assumption that agents are always aware of the state of the economy. Despite standard models postulate that households continuously collect information on the stock market and derive optimal consumption/savings plans, in the data we observe a different pattern. For example, Ameriks et al. (2003) show that households plan infrequently, and wealthy agents plan more often than poor ones. Alvarez et al. (2012) use data from two Italian surveys and find that the median household pays attention to the stock market every
3 months. Furthermore, there is a large heterogeneity in inattention across households: 24% of households observe their financial portfolios less often than twice per year, while 20% of them do it more often than once per week. Finally, Rossi (2010), Da et al. (2011), and Sichermann et al. (2016) find that the allocation of attention comoves with stock returns.¹

This evidence has motivated a new strand of the literature, which studies the implications of households’ limited attention on portfolio choices and asset prices. Nevertheless, the results are still inconclusive. Gabaix and Laibson (2002) and Rossi (2010) show that models with inattention can account for the lumpiness in portfolio adjustments and the dynamics of asset prices. Conversely, Chen (2006) and Finocchiaro (2011) find that inattention has negligible effects on portfolio choices and the level of the equity premium.²

In this paper I evaluate whether the observed duration of households’ inattention can quantitatively account for the inertia in households’ portfolio rebalancing. I build on Reis (2006) and develop a tractable theory of endogenous inattention with heterogeneity across agents. I propose a model that introduces an observation cost into the environment of Krusell and Smith (1997, 1998). Namely, I consider a production economy with incomplete markets, idiosyncratic labor income risk, and heterogeneous agents, who incur in an observation cost whenever they collect information on the aggregate states of the economy and formulate a new plan for financial investment. This feature generates a novel trade-off: attentive households take better decisions, but also bear higher costs. As a result, households decide to plan at infrequent dates and stay inattentive meanwhile. Inattentive agents do not gather

¹Few other papers show that investors’ attention affects stock prices and portfolio choices, e.g. Barber and Odean (2008), Brunnermeier and Nagel (2008), Della Vigna and Pollet (2009), Hirshleifer et al. (2009), and Mondria et al. (2010).

²Lynch (1996), Duffie (2010), and Chien et al. (2011, 2012) study the implications of exogenous infrequent portfolio rebalancing on asset prices. In the Supplementary Appendix, I study how the results of the model change when inattention is considered as either an endogenous variable or an exogenous one.
new information on the aggregate states of the economy and their financial portfolios change by inertia following the realizations of stock and bond returns.

When I bring the model to the data, I discipline the role of inattention by calibrating the observation cost to match the duration of inattention of the median household estimated by Alvarez et al. (2012). I also discipline the interaction of inattention with households’ funding liquidity by calibrating the tightness of the borrowing constraints to match the share of households with negative wealth. These choices imply that the aim of the paper is not to use inattention to match portfolio rebalancing and asset prices. Rather, the model can be used to address the following question: once the observation cost and the tightness of the borrowing constraints are pinned down by the data, how much of the dynamics of portfolio rebalancing and asset prices can be accounted for only by households’ inattention?

Looking at the results of the model, I find that the duration of inattention depends negatively on households’ wealth - in line with the evidence of Ameriks et al. (2003) - because observation costs are relatively higher for poor agents. The cyclicality of inattention depends on the marginal gain and the marginal cost of being attentive and actively investing in the stock market. Both forces are countercyclical, but they asymmetrically affect different agents. Poor households plan in expansions because they cannot afford the observation cost in bad times. Instead, wealthy agents plan in recession to benefit of the higher expected return to equity. Overall the level of inattention is countercyclical.

Second, households’ portfolio rebalancing displays substantial inertia. As long as inattentive agents do not invest actively, their financial positions follow passively the realizations of the aggregate shock. On average, households actively offset around 73% of the passive variations in the risky share. Hence,
inertia drives 27% of the changes in the financial portfolios. Since in the data inertia characterizes 50% of the variations in the risky share, as documented in Calvet et al. (2009), the model can account for 54% of the observed inertia in portfolio rebalancing. In the model, the inertia is entirely determined by the observation cost. Indeed, when the observation cost is set to zero, households always actively adjust their financial positions.

The model generates a large heterogeneity in the degree of portfolio rebalancing across households. Wealthy agents are attentive often enough to actively rebalance their portfolios period-by-period. Instead, poor stockholders are very inattentive and offset just 43% of the passive variation in their portfolios. This result is consistent with the evidence of Calvet et al. (2009), who find that although on average portfolio rebalancing is rather inertial, wealthy households invest very actively.

Third, inattention provides a rationale to the limited stock market participation. The model can account for the share of market participants observed in the data with a low per-period participation cost. Without inattention, matching the same share of stockholders requires a participation cost which is four times larger. Inattention is a barrier to financial investment because households anticipate that during the periods of inattention they end up investing sub-optimally.

Fourth, inattention improves the asset pricing performance of the model. On the one hand, inattention raises the volatility of stock returns, by boosting the movements in the marginal productivity of capital. As inattentive agents cannot immediately adjust their portfolios to the realizations of the aggregate shock, individual financial investment alternates between periods of inaction and periods of sharp adjustments. As a result, aggregate investment becomes more volatile and less correlated with the realizations of the aggregate productivity shock. These two channels raise the volatility of the
marginal productivity of capital. On the other hand, inattention induces countercyclical variations in the equity premium. This result is usually obtained through consumption habits or long run risk. Instead, here it is just the by-product of the observation cost, that concentrates the aggregate risk on a small measure of agents. As in Chien et al. (2012), at each point of time there are few attentive investors that actively trade stocks and bear the whole aggregate risk of the economy, commanding a higher return rate on equity.

Finally, I provide a novel testable implication that links households’ inattention to households’ funding liquidity. I find that inattention matters quantitatively on the dynamics of portfolio rebalancing and asset prices only if borrowing constraints are tight enough. When borrowing constraints are loose, households can borrow and smooth away any investment mistake made during inattention. Moreover, all households participate in the stock market with buy-and-hold positions, as in Chen (2006). In this case, households dilute the observation cost by trading very infrequently, portfolio rebalancing is passive, and inattention does not affect asset prices.

2 Related Literature

This paper studies households’ inattention to the stock market. In the literature, inattention is usually achieved either by making agents gathering information and planning financial investment at discrete dates (e.g., Duffie and Sun, 1990; Lynch, 1996; Gabaix and Laibson, 2002; Chen, 2006; Reis, 2006; Rossi, 2010; Finocchiaro, 2011; Chien et al., 2011, 2012), or through learning with capacity constraints (e.g., Sims, 2003; Peng, 2005; Huan and
Liu, 2007).³ I follow the first strand of the literature because of my emphasis on the effects of inattention on agents’ portfolio choices. Indeed, I study a heterogeneous agent economy, in which the individual portfolio choice is not trivially determined. This feature avoids having a representative agent which in equilibrium holds the portfolio of the market. Models featuring learning with capacity constraint can be extended to the case of heterogeneous agents and idiosyncratic shocks only by neglecting the existence of higher-order beliefs, as discussed in Porapakkarm and Young (2008).⁴ Yet, Angeletos and La’O (2009) show that higher-order beliefs do play a crucial role in the dissemination of information across agents.

My paper differs from the literature on inattention on two main dimensions. First, I discipline the role of inattention by calibrating the observation cost to match the actual duration of inattention for the median household. In this way, I can evaluate whether the observed level of inattention can quantitatively account for the heterogeneous dynamics in portfolio rebalancing across households. Second, I highlight a novel testable mechanism that links inattention - and its quantitative effects on portfolio choices and asset prices - to the tightness of households’ borrowing constraints.

3 The Model

In the economy there is a representative firm that uses capital and labor to produce a consumption good. On the other side, there is a unit measure of ex-ante identical agents. Households are ex-post heterogeneous because they bear an uninsurable idiosyncratic labor income risk. Moreover, they incur

³Inattention is also closely tied to the concepts of information acquisition, e.g. Grossman and Stiglitz (1980) and Peress (2004), and uncertainty, see Veronesi (1999) and Andrei and Hasler (2015).

⁴When agents have imperfect common knowledge and differ in their information set, they need to forecast other agents’ forecast, and so on so forth. In this case, equilibrium prices do not depend only on the infinite-dimensional distribution of agents across wealth, but also on the infinite-dimensional distribution of beliefs.
in a monetary observation cost whenever they collect information on the aggregate states of the economy and take the optimal decisions on portfolio choices. Households can invest in three assets: a risk-free bond, risky capital, and a transaction account that yields no interest payment. The transaction account is liquid: inattentive households finance consumption expenditure only using the transaction account.

### 3.1 The Firm

The production sector of the economy consists of a representative firm, which produces a consumption good $Y_t \in \mathbf{Y} \subset \mathbb{R}_+$ using a Cobb-Douglas production function

$$Y_t = z_t N_t^{1-\eta} K_t^{\eta}$$

where $\eta \in (0,1)$ denotes the capital income share. The variable $z_t \in \mathbf{Z} \subset \mathbb{R}_+$ follows a stationary Markov process with transition probabilities $\Gamma_z(z',z) = \Pr(z_{t+1} = z' | z_t = z)$. The firm hires $N_t \in \mathbf{N} \subset \mathbb{R}_+$ workers at the wage $w_t$, and rents from the households the stock of physical capital $K_t \in \mathbf{K} \subset \mathbb{R}_+$ at the interest rate $r_t^s$. Physical capital depreciates at a rate $\delta \in (0,1)$ after production. The firm chooses capital and labor to equate the marginal productivities to prices, as follows

$$r_t^s = \eta z_t N_t^{1-\eta} K_t^{\eta-1} - \delta$$

$$w_t = (1 - \eta) z_t N_t^{-\eta} K_t^\eta.$$
3.2 Households

The economy is populated by a measure one of ex-ante identical households. They are infinitely lived, discount the future at the rate $\beta \in (0, 1)$, and maximize lifetime utility

$$E_0 \sum_{t=0}^{\infty} \beta^t U(c_t)$$

where $c_t \in C \subset \mathbb{R}^+$ denotes consumption at time $t$. I consider a CRRA utility function $U(c) = \frac{c^{1-\theta}}{1-\theta}$, where $\theta$ denotes the risk aversion of the households.

3.2.1 Idiosyncratic Shocks

Households bear an idiosyncratic labor income risk which consists of two components. First, households are hit by a shock $e_t \in E \subset \{0, 1\}$, which determines their employment status. A household has a job when $e_t = 1$ and is unemployed when $e_t = 0$. I assume that $e_t$ follows a stationary Markov process with transition probabilities

$$\Gamma_e(z, z', e, e') = \Pr(e_{t+1} = e'|e_t = e, z_t = z, z_{t+1} = z').$$

Although the shock is idiosyncratic and washes out in the aggregate, its transition probabilities depend on the aggregate productivity shock. In this way, both the idiosyncratic uncertainty and the unemployment rate of the economy rise in recessions.\(^5\) Second, when a household is employed, it faces a further shock $\xi_t \in \Xi \subset \mathbb{R}^+$, which determines the efficiency units of hours worked. This shock is orthogonal to the aggregate productivity shock and follows a stationary Markov process with transitional probabilities

$$\Gamma_\xi(\xi, \xi') = \Pr(\xi_{t+1} = \xi'|\xi_t = \xi).$$

\(^5\)Mankiw (1986) and Krueger and Lustig (2010) show that a countercyclical idiosyncratic uncertainty raises the price of risk. Storesletten et al. (2007) find that in the data labor income risk peaks in recessions.
When a household is unemployed, it receives a constant unemployment benefit $\bar{w} > 0$, which is financed through a lump sum tax $\tau$ applied to employed agents. Households’ labor income $l_t$ is then

$$l_t = w_t \xi_t e_t + \bar{w} (1 - e_t) - \tau e_t.$$ (7)

As in Pijoan-Mas (2007), I consider two sources of idiosyncratic uncertainty just for quantitative reasons. In the calibration of the model, the employment shock disciplines the correlation between the individual labor income risk and the aggregate shock, whereas I use the shock to the efficiency units of labor to match the cross-sectional distribution of labor income observed in the data.

### 3.2.2 Market Arrangements

Households can allocate their overall wealth $\omega_t \in \Omega \subseteq \mathbb{R}$ between consumption and savings. Households can save in three different ways. First, households own the capital of the economy. Each household holds $s_t \in S \equiv [s, \infty]$ units of capital, which are either rented to the firm or traded among households. Capital is risky and yields the rate $r_s^t$, as defined in (2). Second, households can also invest in a one-period non-contingent bond $b_t \in B \equiv [b, \infty]$, which is in zero net supply. The bond yields a risk-free rate $r_b^t$. Households face exogenous borrowing constraints for both assets and cannot go shorter than $s$ in risky equity and $b$ in risk-free bonds. When these values equal zero, households cannot take short positions at all. Third, as in Abel et al. (2007, 2013), households can also save in a transaction account $a_t \in A \equiv [0, \infty]$. The transaction account yields no interest payment.\(^6\)

In addition, households incur in a fixed per-period participation cost $\phi$ whenever investing in the stock market, that is, whenever $s_t \neq 0$. This cost

\(^6\)In equilibrium households save in the transaction account only if the observation cost is positive. The only rationale of this account is to provide liquid funds to inattentive households for financing their consumption.
may prevent households from investing in the stocks. In the quantitative analysis, I use this cost to match the observed amount of limited stock market participation, and evaluate to what extent inattention can reconcile a large share of non-participants with a low per-period participation cost.

In this framework, markets are incomplete because households cannot trade claims which are contingent on the realizations of the idiosyncratic shock. As long as the labor income risk cannot be fully insured, households are ex post heterogeneous in wealth, consumption, and portfolio choices.

3.2.3 Observation Cost

Households incur in a monetary observation cost proportional to their labor income $\chi l_t$ whenever they acquire information on the aggregate states of the economy and define their optimal choices on stocks and bonds. This cost is a reduced form for the financial and time opportunity expenditures bore by households to figure out the optimal composition of the financial portfolio.

The observation cost induces the households to plan infrequently and stay inattentive meanwhile. Planning dates are defined as dates $d_i \in D \subset \mathbb{N}$ such that $d_{i+1} \geq d_i$ for any $i$. At a planning date $d_i$, households pay the cost $\chi l_{d_i}$, collect the information on the aggregate states of the economy, and decide the next planning date $d_{i+1}$. Moreover, households decide the stream of consumption throughout the period of inattention $[c(e_i, \xi_i), c(e_{d_{i+1}-1}, \xi_{d_{i+1}-1})]$, and the investment in the transaction account $a_{d_{i+1}}$, risky capital $s_{d_{i+1}}$, and risk-free bonds $b_{d_{i+1}}$. Importantly, the stream of consumption throughout the period of inattention is set conditional on the realizations of the idiosyncratic shocks $e_t$ and $\xi_t$. Instead, at non planning dates, households are inattentive and follow the pre-determined plan for consumption set in the previous planning date. I assume that inattentive households finance consumption using the transaction account and the stream of labor income. Throughout
inattention, interest payments $r_i^a$ and $r_i^b$ are reinvested in equity and bonds, respectively.

In the model, attentive households observe all the states of the economy, while inattentive households have a limited amount of information. I assume that inattentive households do not observe the aggregate states of the economy, although they are always fully aware of the realizations of the idiosyncratic shocks $e_t$ and $\xi_t$. In the benchmark economy, I assume that throughout inattention only the choice of consumption - and not the choices on the composition of the financial portfolio - depends on the realizations of the idiosyncratic shocks. This assumption is consistent with the empirical evidence of Alvarez et al. (2012), who find that just 6% of the households adjust their portfolio more often than they observe it.\footnote{The Supplementary Appendix relaxes this assumption, by considering a version of the model in which during inattention also the choices of bonds and stocks depend on the realizations of the idiosyncratic shocks.}

I further characterize the conditions governing inattention in the model. To maintain the existence of credit imperfections, I postulate that inattention breaks out exogenously whenever households hit the borrowing constraints. In such a case, an unmodeled financial intermediary calls the attention of the households, which pay the observation cost and become attentive. Moreover, I assume that households become attentive when their consumption plan cannot be financed anymore by the transaction account and the labor income.

These assumptions affect the realized duration of inattention. A household that at time $d_i$ decides not to observe the states of the economy until $d_{i+1}$ ceases to be inattentive at the realized new planning date $\lambda(d_{i+1})$, which is the minimum between the desired new planning date $d_{i+1}$ and the periods in which either the household hits the borrowing constraint, $\{j \in [d_i, d_{i+1}) : b_{j+1} < b \text{ or } s_{j+1} < s\}$, or consumption cannot be financed anymore by the liquid funds, $\{j \in [d_i, d_{i+1}) : c_j > a_j + l_j\}$. 

\[\]
3.2.4 Value Function

To define the aggregate states of the households’ problem, I introduce the distribution of the agents $\gamma_t$ - defined over households’ idiosyncratic states, the decisions of inattention, the portfolio choices, and the consumption path $\{\omega_t, e_t, \xi_t, d_t, a_t, b_t, c_t\}$ - which characterizes the probability measure on the $\sigma$-algebra generated by the Borel set $J \equiv \Omega \times E \times \Xi \times D \times A \times S \times B \times C$.

Roughly speaking, $\gamma_t$ keeps track of all the heterogeneity among agents. In this environment $\gamma_t$ is an aggregate state. Indeed, Krusell and Smith (1997, 1998) show that prices depend on the entire distribution of agents across their idiosyncratic states. The distribution $\gamma_t$ evolves over time following a law of motion

$$
\gamma_{t+1} = H(\gamma_t, z_t, z_{t+1}).
$$

The operator $H(\cdot)$ pins down the changes in the measure $\gamma_t$ taking as given an initial value and the realizations of the aggregate shock $z_t$.

The structure of the problem takes also into account how the information is revealed to the agents. The state variables of this economy $x_t \equiv \{\omega_t, e_t, \xi_t; z_t, \gamma_t\}$ are random variable defined on a filtered probability space $(X, F, P)$. $X$ denotes the set including all the possible realizations of $x_t$, $F$ is the filtration $\{F_t, t \geq 0\}$ consisting of the $\sigma$-algebra that controls how the information on the states of the economy is disclosed to the agents, and $P$ is the probability measure defined on $F$. Hereafter, I define the expectation of a variable $v_t$ conditional on the information set at time $k$ as $E_k [v_t] = \int v_t dP(F_k) = \int v(x_t) dP(F_k)$. The state vector $P (v_t | F_k) = P (v_t | x_k)$ is a sufficient statistics for the probability of any variable $v_t$ because of the Markov structure of $x_t$.

The presence of observation costs and inattentive agents implies some measurability constraints on the expectations of the households. Namely,
a planning date \( d_i \) defines a new filtration \( \mathcal{F}_s \) such that \( \mathcal{F}_s = F_{d_i} \) for \( s \in [d_i, \lambda(d_{i+1})] \). Hence, any decision made throughout the duration of inattention is conditional on the information at time \( d_i \), because households do not update their information on the aggregate states of the economy until the new planning date \( \lambda(d_{i+1}) \). Taking into account this measurability constraint, I write agents’ recursive problem as

\[
V(\omega_t, e_t, \xi_t; z_t, \gamma_t) = \max_{d_i \in \mathcal{E}(\omega_t, e_t, \xi_t, \gamma_t)} \mathbb{E}_t \left[ \sum_{j=t}^{\lambda(d)} \beta^{j-t} U(c(e_j, \xi_j)) \right] + \beta^{\lambda(d) - t} V(\omega_{\lambda(d)}, e_{\lambda(d)}, \xi_{\lambda(d)}; z_{\lambda(d)}, \gamma_{\lambda(d)})
\]

s.t. \( \omega_t = c(e_t, \xi_t) + a_{t+1} + s_{t+1} + b_{t+1} + \phi I_{\{s_{t+1} \neq 0\}} \) (10)

\[
\omega_{\lambda(d)} = s_{t+1} \prod_{j=t+1}^{\lambda(d)} \left( 1 + r^a_j (z_j, \gamma_j) \right) + b_{t+1} \prod_{j=t+1}^{\lambda(d)} \left( 1 + r^b_j (z_j, \gamma_j) \right) + a_{t+1} + \ldots + \sum_{j=t+1}^{\lambda(d)} l_j (z_j, \gamma_j) - \sum_{j=t+1}^{\lambda(d)-1} c(e_j, \xi_j) - \sum_{j=t+2}^{\lambda(d)} \phi I_{\{s_j \neq 0\}} - \chi l_{\lambda(d)} (z_{\lambda(d)}, \gamma_{\lambda(d)})
\]

(11)

\[
\gamma_{\lambda(d)} = H(\gamma_t, z_{\lambda(t)}) \quad (12)
\]

(13)

\[
\lambda(d) = \min_{j \in \{t+1, d\}} \left\{ d_i, b_{t+1} \prod_{k=t+1}^{j} \left[ 1 + r^a_k (z_k, \gamma_k) \right] < b, \quad s_{t+1} \prod_{k=t+1}^{j} \left[ 1 + r^a_k (z_k, \gamma_k) \right] < b, \quad \sum_{k=t+1}^{j} c(e_k, \xi_k) > a_{t+1} + \sum_{k=t+1}^{j} l_k (z_k, \gamma_k) \right\}
\]

(14)

Equation (10) denotes the budget constraint of the agents, who use their wealth to consume, invest in the two assets, save in the transaction account, and pay the participation cost in case they own stocks. Equation (11) shows the evolution over time of wealth, which depends on the consumption stream and the returns to investment throughout inattention. At the realized new planning date \( \lambda(d) \) agents incur in the observation cost \( \chi l_{\lambda(d)} \). Equation (12)
defines the law of motion of the distribution of agents $\gamma_t$ conditional on the history of aggregate shocks $z^{\lambda(d)}$. Finally, Equation (13) denotes the borrowing constraints faced by the households, whereas Equation (14) describes the new realized planning date $\lambda(d)$.

### 3.3 Equilibrium

#### 3.3.1 Definition of Equilibrium.

A competitive equilibrium for this economy is a value function $V$, a set of policy functions $\{g^c, g^b, g^s, g^a, g^d\}$, a set of prices $\{r^b, r^s, w\}$, and a law of motion $H(\cdot)$ for the measure of agents $\gamma$ such that

- Given the prices $\{r^b, r^s, w\}$, the law of motion $H(\cdot)$, the exogenous transition matrices $\{\Gamma^z, \Gamma^e, \Gamma^\xi\}$, the value function $V$, and the set of policy functions $\{g^c, g^b, g^s, g^a, g^d\}$ solve the household’s problem;
- The bonds market clears, $\int g^bd\gamma = 0$;
- The capital market clears, $\int g^sd\gamma = K'$;
- The labor market clears, $\int e\xi d\gamma = N$;
- The unemployment benefit is financed by a lump sum tax on employed households, $\int \bar{w}(1-e)d\gamma = \int \tau ed\gamma$;
- The law of motion $H(\cdot)$ is generated by the optimal decisions $\{g^c, g^b, g^s, g^a, g^d\}$, the transition matrices $\{\Gamma^z, \Gamma^e, \Gamma^\xi\}$, and the history of aggregate shocks $z$.

#### 3.3.2 First-Order Conditions.

Gabaix and Laibson (2002) consider an environment in which agents are exogenously inattentive for a fixed number of periods. In their model, the Euler equation for consumption holds only for the mass of attentive agents because inattentive households are off their equilibrium condition. Instead,
here the Euler equations of both attentive and inattentive agents hold in equilibrium. At a planning date $t$ the Euler equation is a standard stochastic inter-temporal condition that reads

$$\mathbb{E}_t \left[ M_{\lambda(d),t} \prod_{k=t+1}^{\lambda(d)} \left( r^s_k (z_k, \gamma_k) - r^b_k (z_k, \gamma_k) \right) \right] = 0 \quad (15)$$

where $M_{\lambda(d),t} = \beta^{\lambda(d)-t} \frac{U'(c_{\lambda(d)})}{U'(c_t)}$ denotes households' stochastic discount factor. This condition posits that the optimal share of stocks in the portfolio equalizes the compounded expected discounted returns from stocks and bonds throughout the period of inattention.

The Euler equation of an inattentive agent between time $v$ and $q$, with $t < v < q < \lambda$ is deterministic and equals

$$M_{q,v} \prod_{k=v+1}^{q} \left( r^s_k (z_k, \gamma_k) - r^b_k (z_k, \gamma_k) \right) = 0. \quad (16)$$

Inattentive agents do not gather any new information on the states of the economy, and therefore they behave as if there were no uncertainty. Agents get back to the stochastic inter-temporal conditions as soon as they reach a new planning date and update their information set. As agents alternate between attention and inattention, they also shift from stochastic to deterministic Euler equations.\(^8\)

### 3.4 Inattention in the Model and in the Data

In the model inattentive households do not observe the aggregate states of the economy, whereas they are always fully aware of the realizations of the idiosyncratic shocks $e_t$ and $\xi_t$. The model represents a tractable extension to the case of heterogeneous agents of the inattentiveness proposed by Reis

---

\(^8\)In either case, the Euler equations are not satisfied with equality for borrowing constrained agents.
(2006). Although in Reis (2006) inattentive agents do not receive any flow of information, I relax this condition by allowing inattentive households to observe at least their idiosyncratic sources of uncertainty. From this point of view, this model bridges the gap between the inattentiveness of Reis (2006) and the rational inattention of Sims (2003). In Sims (2003), households choose how to allocate their limited capacity of information acquisition, by deciding the noise up to which they observe all the relevant variables of the economy. My model can be considered a limiting case of Sims (2003), in which households decide to allocate their entire capacity to observe the idiosyncratic shocks, up to the point that the noise around the idiosyncratic shocks disappears, while the noise on the aggregate states goes to infinity.

The definition of inattention of the model slightly differs from the definition of inattention of the survey studied by Alvarez et al. (2012). In their data, inattention refers to the frequency with which households observe their financial portfolio. Instead, the model considers a broader definition of inattention, by focusing on the frequency of observation of the aggregate states of the economy.

Furthermore, in the model households always adjust their portfolios upon paying the observation cost. Although this is not always the case in the data, the correlation between observing and adjusting the financial portfolio is very high and equals 0.45. Alvarez et al. (2012) show that such a correlation can be rationalized with the presence of both observation costs and portfolio transaction costs. Since I study an economy in which households do not face the additional friction of the transaction cost, my model implies a lower bound on the effects of inattention on portfolio inertia and asset prices.\footnote{I abstract from the portfolio transaction costs for purely computational reasons. The introduction of portfolio transaction costs requires the addition of yet another state variable, which would limit substantially the computational tractability of the model. For instance, following the choices I made in the calibration exercise, this further state variable would inflate the grid points from 5,184,000 up to 311,040,000.}

\footnotetext{9}
4 Calibration

The calibration strategy follows Krusell and Smith (1997, 1998) and Pijoan-Mas (2007). Some parameters are calibrated to match salient facts of the U.S. economy, while others (e.g., the risk aversion of the household) are set to values estimated in the literature. Throughout the quantitative analysis, I set one period of the model to correspond to one month. Nevertheless, I report the asset pricing statistics aggregated at the annual frequency to be consistent with the literature.

First of all, I calibrate the aggregate shock to match the volatility of output growth. The idiosyncratic labor income risk is defined to target the cross-sectional distribution of labor income, and its correlation with the aggregate unemployment rate. It is important to have a realistic variation in labor income because the choice of inattention, and consequently the effect of the observation cost on portfolio rebalancing and asset prices, depends on the budget of the households. The observation cost is defined to replicate the duration of inattention of the median household, while the participation cost is set to match the share of stockholders observed in the data. Finally, I calibrate three parameters that capture the amount of wealth in the economy: the time discount factor $\beta$ and the borrowing constraint on stocks $s$ and bonds $b$. The discount factor is set to match the U.S. annual capital to output ratio of 2.5, which yields a value of $\beta = 0.9951$. The calibration of the borrowing constraint is very important because I show that the quantitative implications of inattention depends crucially on how tighten borrowing constraints are. First, I equalize the level of the constraint on stocks and bonds, that is, $s = b$. Second, I pin down the level of both constraints by matching the fraction of households with negative wealth, which in the data is around 10%, as shown in Diaz-Gimenez et al. (2011). This choice implies a value of $s = b = -5.96,$
which is equivalent to around three times the monthly income of the median household.

The parameters set to values estimated in the literature are the capital share of the production function $\eta$, the capital depreciation rate $\delta$, and the risk aversion of the household $\theta$. I choose a capital share $\eta = 0.40$, as suggested by Cooley and Prescott (1995). The depreciation rate equals $\delta = 0.0066$ to match a 2% quarterly depreciation. The risk aversion of the household is $\theta = 5$, which gives an inter-temporal elasticity of substitution of 0.2.

### 4.1 Aggregate Productivity Shock

I assume that the aggregate productivity shock $z_t$ follows a two-state first-order Markov chain, with values $z_g$ and $z_b$ denoting the realizations in good and bad times, respectively. The two parameters of the transition function are calibrated targeting a duration of 2.5 quarters for both states. The values $z_g$ and $z_b$ are instead defined to match the standard deviation of the Hodrick-Prescott filtered quarterly aggregate output, which is 1.89% in the data. These values are model dependent, and vary with the specification of the environment.

### 4.2 Idiosyncratic Labor Income Shock

**Employment Status.** The employment shock $e_t$ follows a two-state first-order Markov chain, which requires the calibration of ten parameters that define four transition matrices two by two. I consider the ten targets of Krusell and Smith (1997, 1998). I first define four conditions that create a one-to-one mapping between the state of the aggregate shock and the level of unemployment: regardless of the previous realizations of the shock, the good productivity shock $z_g$ comes always with an unemployment rate $u_g$, and the
bad productivity shock $z_b$ with an unemployment rate $u_b$. In this way, the realization of the aggregate shock pins down the unemployment rate of the economy. The four conditions are

\[
1 - u_g = u_g \Gamma_e (z_g, z_g, 0, 1) + (1 - u_g) \Gamma_e (z_g, z_g, 1, 1) \\
1 - u_g = u_b \Gamma_e (z_b, z_g, 0, 1) + (1 - u_b) \Gamma_e (z_b, z_g, 1, 1) \\
1 - u_b = u_g \Gamma_e (z_g, z_b, 0, 1) + (1 - u_g) \Gamma_e (z_g, z_b, 1, 1) \\
1 - u_b = u_b \Gamma_e (z_b, z_b, 0, 1) + (1 - u_b) \Gamma_e (z_b, z_b, 1, 1).
\]

The levels of the unemployment rate in good time and bad time are defined to match the actual average and standard deviation of the unemployment rate. Using data from the Bureau of Labor Statistics from 1948 to 2012, I obtain that the two moments equal 5.67% and 1.68%, respectively. Under the assumption that the unemployment rate fluctuates symmetrically around its mean, I find $u_g = 0.0402$ and $u_b = 0.0732$. Two further conditions come by matching the expected duration of unemployment in good times (6 months) and bad times (10 months). Finally, I set both the job finding probability when moving from the good state to the bad one and the probability of losing the job in the transition from the bad state to the good one to zero.

**Unemployment Benefit.** I set the monthly unemployment benefit $\bar{w}$ to be 5% of the average monthly labor earning. Although different values of the benefit affect the lower end of the wealth distribution, they have no sizable effect on the dynamics of portfolio rebalancing and asset prices.

**Efficiency Units of Hour.** The shock to the efficiency unit of hour $\xi_t$ follows a three-state first-order Markov chain. The values of the shock and the transition function are calibrated to match three facts on the cross-sectional dispersion of labor earnings across households: the share of labor earnings...
held by the top 20\%, the share of labor earnings held by the bottom 40\%, and the Gini coefficient of labor earnings. Table 1 reports the calibrated values and the transition function of the shock $\xi_t$, while Table 2 compares the three statistics on the distribution of labor earnings in the data and in the model.

Table 1: Parameters of the shock to the efficiency units of hour

<table>
<thead>
<tr>
<th>$\xi_1$</th>
<th>$\xi_2$</th>
<th>$\xi_3$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\xi_1$</td>
<td>6</td>
<td></td>
</tr>
<tr>
<td>$\xi_2$</td>
<td>2</td>
<td></td>
</tr>
<tr>
<td>$\xi_3$</td>
<td>1</td>
<td></td>
</tr>
</tbody>
</table>

| $\Gamma_\xi (\xi_1, \cdot)$ | 0.9850 | 0.0025 | 0.0050 |
| $\Gamma_\xi (\cdot, \xi_2)$ | 0.0100 | 0.9850 | 0.0100 |
| $\Gamma_\xi (\cdot, \xi_3)$ | 0.0050 | 0.0125 | 0.9850 |

Note: The efficiency unit of hours $\xi_t$ follows a first-order Markov chain with transition function $\Gamma_\xi$.

Table 2: The distribution of labor earnings

<table>
<thead>
<tr>
<th></th>
<th>Target</th>
<th>Model</th>
<th>Data</th>
</tr>
</thead>
<tbody>
<tr>
<td>Share earnings top 20%</td>
<td>62.1%</td>
<td>63.5%</td>
<td></td>
</tr>
<tr>
<td>Share earnings bottom 40%</td>
<td>4.4%</td>
<td>4.2%</td>
<td></td>
</tr>
<tr>
<td>Gini index</td>
<td>0.57</td>
<td>0.64</td>
<td></td>
</tr>
</tbody>
</table>

Note: the data is from Díaz-Giménez et al. (2011).

4.3 Participation Cost

I calibrate the fixed per-period participation cost $\phi$ to match the amount of limited stock market participation observed in the data. Favilukis (2013)
reports that in the U.S. in 2007 the share of stockholders is 59.4%. The model matches this moment with a participation cost that equals $\phi = 0.019$, which amounts to 0.8% of households’ monthly labor earnings. For example, if the average household earns an income of around $3,000 per month, the cost equals $24.

### 4.4 Observation Cost

I discipline the amount of inattention risk in the model by calibrating the observation cost to match the duration of inattention of the median household. Alvarez et al. (2012) estimate that the median household observes its portfolio every 3 months. The model matches this moment with an observation cost that equals $\chi = 0.029$, which amounts to 2.9% of households’ monthly labor earnings. For example, if the average household earns an income of around $3,000 per month, the cost equals $87.

### 4.5 Computation of the Model

The computation of heterogeneous agent models with aggregate uncertainty are cumbersome because the distribution $\gamma_t$, a state of the problem, is an infinite-dimensional object. I approximate $\gamma_t$ using a finite set of moments of the distribution of aggregate capital $K_t$ - as in Krusell and Smith (1997, 1998), Pijoan-Mas (2007) and Gomes and Michaelides (2008) - and the number of inattentive agents in the economy in every period $\zeta_t$. On the one hand, the approximation using a finite set of moments of aggregate capital $K_t$ can be interpreted as if the agents of the economy were bounded rational, ignoring higher-order moments of $\gamma_t$. Nevertheless, this class of models generates almost linear economies, in which it is sufficient to consider just the first moment of the distribution of capital to have a perfect fit for the approximation.
On the other hand, inattention adds a further term $\zeta_t$, which signals active investors about the degree of the informational frictions in the economy. This condition adds a further law of motion upon which to converge. The presence of inattention implies one further complication. The decision of the agents on the duration of inattention requires the evaluation of their value function over a wide range of different time horizons. I report the details of the computational algorithm in the Supplementary Appendix.

5 Results

I compare the results of the benchmark model with three alternative calibrations. In the first one, the observation cost is zero and there is no inattention. In the second one, the observation cost is more severe and amounts to $\chi = 0.058$. Finally, I consider an economy in which agents are more risk averse, with $\theta = 8$. I calibrate each version of the model to match the volatility of aggregate output growth, the cross-sectional distribution of labor earnings, the amount of limited stock market participation, the level of aggregate wealth, and the fraction of households with negative wealth. Results are computed from a simulated path of 3,000 agents over 10,000 periods.

5.1 Inattention

The observation cost is calibrated to a 3 months duration of inattention for the median household. It turns out that such a cost prevents a large fraction of agents from gathering information on the stock market. Table III shows that in the model, in any given month, the average fraction of inattentive agents in the economy equals 44%. Furthermore, Figure 1 shows that there is a negative correlation between wealth and inattention, in line with the empirical evidence of Ameriks et al. (2003) and Alvarez et al. (2012).
Consistently with the data, the model generates a sizable heterogeneity in the duration of inattention across households. For example, the wealthiest 20% of households observe the aggregate states of the economy every period. Instead, poor agents cannot afford the observation cost and end up being more inattentive. In the model, the poorest 20% of households stay inattentive for 9 months on average. These results point out that in the model inattention behaves both as a time-dependent and a state-dependent rule. Indeed, at each point of time households set a time-dependent rule, deciding how long to stay inattentive. Yet, when a household becomes wealthier, it opts for shorter periods of inattention. Thus, inattention looks as if it were conditional on wealth.¹⁰

Figure 1: Optimal Choice of Inattention

![Figure 1: Optimal Choice of Inattention](image)

Note: the figure plots the policy function of inattention $g^d$ as a function of wealth $\omega$. The idiosyncratic shocks are set to $e_t = 1$ and $\xi_t = 2$. The aggregate shock is $z_t = z_g$ and the aggregate capital equals its mean.

The dynamics of inattention over the business cycle depend on two forces. On the one hand, the countercyclical equity premium induces households to plan in recessions because it is the moment in which the cost of inattention

¹⁰Reis (2006) labels this property of inattention as “recursive time-contingency”. See Alvarez et al. (2012) and Abel et al. (2007, 2013) for further characterizations of the dynamics of inattention over time.
Table 3: Inattention

<table>
<thead>
<tr>
<th>Inattention</th>
<th>$\chi = 0.029$</th>
<th>$\chi = 0$</th>
<th>$\chi = 0.058$</th>
<th>$\theta = 8$</th>
<th>Data</th>
</tr>
</thead>
</table>

A. Duration of inattention (months)

<table>
<thead>
<tr>
<th></th>
<th>Median</th>
<th>Median - good times</th>
<th>Median - bad times</th>
<th>75th percentile - good times</th>
<th>75th percentile - bad times</th>
<th>25th percentile - good times</th>
<th>25th percentile - bad times</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>3.0</td>
<td>0</td>
<td>3.6</td>
<td>4.0</td>
<td>3.0</td>
<td>3.6</td>
<td>7.0</td>
</tr>
<tr>
<td></td>
<td>2.8</td>
<td>0</td>
<td>3.2</td>
<td>3.7</td>
<td>-</td>
<td>3.7</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td>3.2</td>
<td>0</td>
<td>3.9</td>
<td>4.4</td>
<td>-</td>
<td>4.4</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td>1.0</td>
<td>0</td>
<td>1.1</td>
<td>1.2</td>
<td>-</td>
<td>1.2</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td>0.6</td>
<td>0</td>
<td>0.9</td>
<td>0.8</td>
<td>-</td>
<td>0.8</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td>6.1</td>
<td>0</td>
<td>6.5</td>
<td>6.7</td>
<td>-</td>
<td>6.7</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td>6.7</td>
<td>0</td>
<td>7.0</td>
<td>7.2</td>
<td>-</td>
<td>7.2</td>
<td>-</td>
</tr>
</tbody>
</table>

B. Fraction of inattentive agents

<table>
<thead>
<tr>
<th></th>
<th>Median</th>
<th>Median - good times</th>
<th>Median - bad times</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>0.44</td>
<td>0</td>
<td>0.48</td>
</tr>
<tr>
<td></td>
<td>0.41</td>
<td>0</td>
<td>0.44</td>
</tr>
<tr>
<td></td>
<td>0.50</td>
<td>0</td>
<td>0.54</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Note: the variable $\chi$ defines the observation cost and $\theta$ is the risk aversion of agents, which equals 5 in the benchmark model. Good times denote the periods in which the aggregate productivity shock is $z_t = z_g$ and bad times denote the periods in which the aggregate productivity shock is $z_t = z_b$. The fraction of inattentive agents are reported in percentage values. Data is from Alvarez et al. (2012).

in terms of foregone financial returns is highest. On the other hand, the severity of the observation cost fluctuates following households’ wealth. In recessions, households are poorer and cannot afford the observation cost. The results show that wealthy agents tend to be attentive in bad times, to profit from the higher equity premium. For example, in the model the agents at the 75-th percentile of the wealth distribution are on average inattentive for
1 month in good times and 0.6 months in bad times. Instead, the direct cost of inattention affects relatively more poor agents, which prefer to plan in expansions. The agents at the 25-th percentile of the wealth distribution are on average inattentive for 6.1 months in good times and 6.7 months in bad times. Overall, inattention is countercyclical: both the duration of inattention for the median agent and the fraction of inattentive agents in the economy rise in recession. The countercyclicality of inattention is consistent with the empirical evidence of Sichermann et al. (2016), who report that 401(k) retirement account logins fall in bearish markets.

Increasing the size of the observation cost to $\chi = 0.058$ extends the duration of inattention for the median agent up to 3.6 months. Also a risk aversion of $\theta = 8$ does increase the duration of inattention, which goes up to 4 months. This last result is in line with the evidence provided by Alvarez et al. (2012), who show that more risk averse investors observe their portfolio less frequently. This outcome is the net result of two counteracting forces. Households with a higher risk aversion change their portfolio towards risk-free bonds, decreasing the need of observing the stock market. At the same time, more risk averse agents have a stronger desire for consumption smoothing, which induces them to keep track of their investments more frequently. In the model, the first channel offsets the second one, implying a longer duration of inattention for more risk averse agents.

### 5.2 Stock Market Participation

Favilukis (2013) reports that in 2007 just 59.4% of the households participate in the stock market. I show that inattention can reconcile this amount of limited stock market participation with a low per-period participation cost. Although the model is calibrated to match exactly the participation rate
observed in the data, the level of the participation cost required to get the right number of stockholders is endogenous. The amount of this cost is then informative of the extent in which inattention can rationalize the limited stock market participation.

Table 4 shows that, in the benchmark economy with a positive observation cost, the model matches the observed share of stockholders with a participation cost that amounts to 0.8% of households’ average monthly income. If the average household earns an income of around $3,000 per month, the cost of participating in the market for one entire year equals $288. When I abstract from the observation cost, the model requires a participation cost which is four times larger: the model matches the same share of participants with a participation cost that amounts to 3.0% of households’ average monthly income. Following the previous example, in this case the cost of participating in the market for one entire year rises up to $1,080.

Table 4: Participation to the stock market

<table>
<thead>
<tr>
<th>Variable</th>
<th>$\chi = 0.029$</th>
<th>$\chi = 0$</th>
<th>$\chi = 0.058$</th>
<th>$\theta = 8$</th>
<th>Data</th>
</tr>
</thead>
<tbody>
<tr>
<td>Fraction of Shareholders</td>
<td>59.4%</td>
<td>59.4%</td>
<td>59.4%</td>
<td>59.4%</td>
<td>59.4%</td>
</tr>
<tr>
<td>Per-Period Participation Cost</td>
<td>0.8%</td>
<td>3.0%</td>
<td>0.6%</td>
<td>0.7%</td>
<td>-</td>
</tr>
<tr>
<td>(in terms of avg. monthly labor income)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$\frac{\sigma(\Delta \log c_S)}{\sigma(\Delta \log c_{NS})}$</td>
<td>0.86</td>
<td>0.39</td>
<td>0.91</td>
<td>0.98</td>
<td>1.60</td>
</tr>
<tr>
<td>Gini Index of Wealth</td>
<td>0.66</td>
<td>0.42</td>
<td>0.69</td>
<td>0.71</td>
<td>0.82</td>
</tr>
</tbody>
</table>

Note: the variable $\chi$ defines the observation cost and $\theta$ is the risk aversion of agents, which equals 5 in the benchmark model. All the economies are calibrated to match the share of stock market participants observed in the data. The ratio $\frac{\sigma(\Delta \log c_S)}{\sigma(\Delta \log c_{NS})}$ compares the standard deviation of the consumption growth of stockholders $\sigma(\Delta \log c_S)$ with the standard deviation of consumption growth of non-stockholders $\sigma(\Delta \log c_{NS})$. Data is from Favilukis (2013), Mankiw and Zeldes (1991) and Díaz-Giménez et al. (2011).
In the literature limited stock market participation is either assumed exogenously, as in Saito (1996), Basak and Cuoco (1998), and Guvenen (2009), or it is derived endogenously through the presence of trading costs, as in Gomes and Michaelides (2008). The results of Table 4 show that in my economy the limited stock market participation is determined by the *interaction* between the observation cost and the participation cost. For any given value of the participation cost, the presence of the observation cost further reduces the number of households that decide to hold equity. In the model, inattention is a barrier to financial investment because households anticipate that, during their inattention periods, they cannot actively manage their portfolios and end up investing sub-optimally. Hence, the households that opt for long periods of inattention decide not to participate in the stock market.

**Figure 2: Optimal Portfolio Choices**

Note: the figure plots the policy functions of investment in risky assets $g^a$ (continuous line) and risk free bonds $g^b$ (dashed line) as a function of wealth $\omega$. The idiosyncratic shocks are set to $e_t = 1$ and $\xi_t = 2$. The aggregate shock is $z_t = z_g$ and the aggregate capital equals its mean.

The model also successfully predicts that stockholders are on average wealthier than non-stockholders. Figure 2 shows that stockholders tend to be the wealthiest agents of the economy. For example, the poorest 8% of the
households do not hold any risky capital because they are inattentive very often.

The model fails in reproducing the higher consumption growth volatility of stockholders with respect to non-stockholders. In the model, stockholders turn out to be wealthy agents that are able to self-insure their consumption stream, experiencing thereby a lower volatility than non-stockholders.

5.3 Portfolio Rebalancing

In this Section, I study to what extent stockholders manage to actively offset the passive variations in their portfolios. To understand the implications of inattention on portfolio rebalancing, I take the simulated data of my model and replicate the regression that Calvet et al. (2009) run on a panel of Swedish households. In this way I quantify whether - and to what extent - households’ inattention can account for the dynamics of portfolio rebalancing observed in the data.

Let me first define the risky share $\alpha_{i,t}$ of the household $i$ at time $t$ as

$$
\alpha_{i,t} = \frac{s_{i,t+1}}{s_{i,t+1} + b_{i,t+1}},
$$

that is, the ratio of risky capital over the sum of risky capital and risk free bonds. I decompose the variations over time in the risky share $\alpha_{i,t+1} - \alpha_{i,t}$ in two components: the passive change $P_{i,t+1}$ and the active change $A_{i,t+1}$.

Consider a stockholder that at time $t$ invests in stocks $s_{i,t+1}$ and bonds $b_{i,t+1}$. In the next period, her positions amount to $(1 + r_{t+1}^s) s_{i,t+1}$ stocks and $(1 + r_{t+1}^b) b_{i,t+1}$ bonds. If the stockholder does not adjust the portfolio, the new risky share equals

$$
\alpha_{i,t+1}^{P} = \frac{(1 + r_{t+1}^s) s_{i,t+1}}{(1 + r_{t+1}^s) s_{i,t+1} + (1 + r_{t+1}^b) b_{i,t+1}} = \frac{(1 + r_{t+1}^s) \alpha_{i,t+1}}{(r_{t+1}^s - r_{t+1}^b) \alpha_{i,t} + (1 + r_{t+1}^b)}.
$$
I refer to the variable \( \alpha_{i,t+1}^P \) as the passive risky share, that is, the risky share that is expected in the case stockholders do not rebalance at all their portfolios.

I define the passive change \( P_{i,t+1} \) as the change in the risky share for a stockholder that does not adjust her financial portfolio

\[
P_{i,t+1} = \alpha_{i,t+1}^P - \alpha_{i,t}.
\]

Then, I define the active change \( A_{i,t+1} \) as the residual change in the risky share that is not accounted for by the passive risky share \( \alpha_{i,t+1}^P \), that is

\[
A_{i,t+1} = \alpha_{i,t+1} - \alpha_{i,t+1}^P.
\]

\( A_{i,t+1} \) does not capture any mechanic change in the risky share and therefore quantifies the amount of active rebalancing of a stockholder. This measure is defined such that the overall change in the risky share is the sum of the passive and active change

\[
\alpha_{i,t+1} - \alpha_{i,t} = A_{i,t+1} + P_{i,t+1}.
\]

Following Calvet et al. (2009), I study the dynamics of active and passive portfolio rebalancing across households by estimating the panel regression

\[
A_{i,t+1} = \text{constant} + \psi \times P_{i,t+1} + \mu \times \alpha_{i,t} + \epsilon_{i,t+1}.
\]

The coefficient \( \psi \) defines the amount of passive change which is offset by the active rebalancing of stockholders. Instead, the coefficient \( \mu \) captures the dependence of the adjustments in the financial portfolio on the previous share invested in stocks. A fully passive stockholder would have both \( \psi = 0 \) and \( \mu = 0 \). I estimate the regressions at the yearly frequency, over a sample of
about 800 periods. I consider only stockholders that maintain a position in stocks for at least two consecutive years. The regression is estimated over a panel of households over a total of around 1,200,000 stockholder-period observations. To be consistent with Calvet et al. (2009), I compute the annual risky share and related variables from the simulated data taken from the last month of each year.

Table 5.3 reports the estimates of the parameter $\psi$ of the regression above in four different cases: I consider all the years of my simulated data (Panel A), I consider only the expansionary periods (Panel B), I consider all the years focusing only on stockholders in the 75th percentile of the wealth distribution (Panel C), I consider all the years focusing only on stockholders in the 25th percentile of the wealth distribution (Panel D).

Panel A shows that in the model stockholders actively offset around 73% of the passive change in their portfolio share. This result implies that inertia accounts for the remaining 27% of the movements in the financial portfolios. Since Calvet et al. (2009) find that inertia characterizes 50% of the changes in the risky share of Swedish households, the model is able to account for 54% of the inertia in portfolio rebalancing observed in the data. In the model, the inertia in portfolio rebalancing is entirely driven by inattention. Indeed, when I shut down the observation cost, households always actively manage their financial positions.

Panel B shows that the amount of active rebalancing increases during expansions, going up to 77%. This result is consistent with the evidence of Calvet et al. (2009), who find that the amount of active rebalancing decreased in Sweden from 2000 to 2002, a period of bearish stock market. In the model, stockholders are on average more attentive during expansions, and therefore the average amount of active rebalancing rises in good times.
Table 5: Panel Regression of Active and Passive Portfolio Rebalancing

<table>
<thead>
<tr>
<th></th>
<th>$\chi = 0.029$</th>
<th>$\chi = 0$</th>
<th>$\chi = 0.058$</th>
<th>$\theta = 8$</th>
<th>Data</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>$\chi = 0$</td>
<td>$\chi = 0.058$</td>
<td>$\theta = 8$</td>
<td></td>
</tr>
<tr>
<td>Active Change</td>
<td></td>
<td>$-0.730$</td>
<td>$-0.997$</td>
<td>$-0.703$</td>
<td>$-0.712$</td>
</tr>
<tr>
<td>$\hat{\psi}$ (Passive Change)</td>
<td></td>
<td>$-0.774$</td>
<td>$-0.998$</td>
<td>$-0.756$</td>
<td>$-0.766$</td>
</tr>
<tr>
<td>Adjusted $R^2$</td>
<td>0.70</td>
<td>0.76</td>
<td>0.72</td>
<td>0.71</td>
<td>0.12</td>
</tr>
<tr>
<td>C. All Years - Stockholders in the 75th Percentile of the Wealth Distribution</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$\hat{\psi}$ (Passive Change)</td>
<td></td>
<td>$-0.876$</td>
<td>$-0.999$</td>
<td>$-0.868$</td>
<td>$-0.860$</td>
</tr>
<tr>
<td>Adjusted $R^2$</td>
<td>0.73</td>
<td>0.77</td>
<td>0.71</td>
<td>0.73</td>
<td></td>
</tr>
<tr>
<td>D. All Years - Stockholders in the 25th Percentile of the Wealth Distribution</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$\hat{\psi}$ (Passive Change)</td>
<td></td>
<td>$-0.425$</td>
<td>$-0.996$</td>
<td>$-0.419$</td>
<td>$-0.406$</td>
</tr>
<tr>
<td>Adjusted $R^2$</td>
<td>0.74</td>
<td>0.80</td>
<td>0.79</td>
<td>0.79</td>
<td></td>
</tr>
</tbody>
</table>

Note: the table reports the yearly panel regressions of the active change $A_{i,t+1}$ in portfolio rebalancing on a constant, the passive change $P_{i,t+1}$ in portfolio rebalancing and the share of the portfolio invested in stocks in the previous year $\alpha_{i,t}$. I consider households that participate in the stock market over two consecutive years ($t$ and $t + 1$). Passive Change $\hat{\psi}$ is the estimated amount of passive change which is actively offset by the households by adjusting their portfolios. The variable $\chi$ defines the observation cost and $\theta$ is the risk aversion of agents, which equals 5 in the benchmark model. Data is from Calvet et al. (2009).
Panel C and D compare the amount of active rebalancing across wealthy and poor stockholders. Panel C shows that wealthy households offset 88% of the passive change in the risky share. This value is twice as large as the amount of rebalancing estimated across poor households, which equals 43%. Again, wealthy stockholders can afford to be attentive often enough to have a very active management of their portfolios. Instead, poor households end up being inattentive, and their portfolios follow by inertia the realizations of stock and bond returns. This result highlights that the model is able to capture the heterogeneity in the inertia of portfolio rebalancing across households.

5.4 Asset Pricing Moments

5.4.1 Stock and Bond Returns

Panel A of Table 6 reports the results of the model on the level and the dynamics of stock returns, bond returns, and the equity premium. First, I discuss the standard deviations because the observation cost almost doubles the volatility of stock returns. In the benchmark model the standard deviation of returns is 11.5%, which accounts for around 60% of the volatility observed in the data, that is 19.3%. Nonetheless, without inattention the standard deviation would be just 6.3%.

The observation cost boosts the volatility of returns because it alters the dynamics of the marginal productivity of capital. Since inattentive agents cannot immediately adjust their portfolios to the realizations of the aggregate shock, individual financial investment alternates between periods of inaction and periods of sharp adjustments. The effects of inattention on the marginal productivity of capital can be understood - and quantified - by looking at how the observation cost changes the behavior of aggregate investment. In the model, inattention alters the dynamics of aggregate investment in two
ways. First, inattention raises the volatility of aggregate investment. In the economy with the observation costs, the standard deviation of investment at the quarterly frequency equals 2.5%. Although this value is lower than the empirical counterpart of 4.5%, when I abstract from inattention the volatility of investment shrinks even more, down to 1.4%. Second, inattention reduces the correlation between investment and output. In an economy without the observation costs, investment moves one-to-one with output: the correlation equals 0.99. With inattention, the correlation drops to 0.93, closer to the value of 0.95 observed in the data. The increase in the volatility of investment and the reduction of the correlation between investment and output raise the fluctuations of both the marginal productivity of capital and the stock returns. Absent these changes in investment, inattention would not affect the volatility of stock returns.

As far as the volatility of the risk-free rate is concerned, I find a standard deviation of 3.7%, which is lower than its empirical counterpart, that equals 5.4%. Note that standard models usually deliver excessively volatile risk-free rates. For example, Jermann (1998) and Boldrin et al. (2001) report a standard deviation between 10% and 20%. The mechanism that prevents volatility to surge is similar to the one explored by Guvenen (2009). Poor agents have a strong desire to smooth consumption, and their high demand of precautionary savings offsets any large movements in bond returns. Although in Guvenen (2009) the strong desire for consumption smoothing is achieved by a low elasticity of inter-temporal substitution, here it is the observation cost that forces the agents to insure against the risk of infrequent planning.

When looking at the level of the equity premium reported in Panel B of Table 6, I find that the model generates a wedge between stock returns and bond yields which equals 2.6%. In the economy without the observation cost, the equity premium is just 0.8%. Hence, in the model households’ inattention
### Table 6: Asset pricing moments

<table>
<thead>
<tr>
<th>Variable</th>
<th>Moment</th>
<th>$\chi = 0.029$</th>
<th>$\chi = 0$</th>
<th>$\chi = 0.058$</th>
<th>$\theta = 8$</th>
<th>Data</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>$\theta$</td>
<td>0.029</td>
<td>0.058</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>$\theta$</td>
<td>0.029</td>
<td>0.058</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>A. Stock and bond returns</td>
<td>Mean</td>
<td>4.47</td>
<td>1.91</td>
<td>4.81</td>
<td>5.08</td>
<td>8.11</td>
</tr>
<tr>
<td>Stock return</td>
<td>Std. dev.</td>
<td>11.51</td>
<td>6.31</td>
<td>11.97</td>
<td>12.23</td>
<td>19.30</td>
</tr>
<tr>
<td>Risk-free return</td>
<td>Mean</td>
<td>1.83</td>
<td>1.10</td>
<td>1.84</td>
<td>1.98</td>
<td>1.94</td>
</tr>
<tr>
<td></td>
<td>Std. dev.</td>
<td>3.65</td>
<td>2.62</td>
<td>4.00</td>
<td>4.39</td>
<td>5.44</td>
</tr>
<tr>
<td>B. Equity premium</td>
<td>Mean</td>
<td>2.64</td>
<td>0.81</td>
<td>2.97</td>
<td>3.10</td>
<td>6.17</td>
</tr>
<tr>
<td>Equity premium</td>
<td>Std. dev.</td>
<td>11.35</td>
<td>6.27</td>
<td>11.81</td>
<td>12.02</td>
<td>19.49</td>
</tr>
<tr>
<td>Sharpe ratio</td>
<td>Mean</td>
<td>0.23</td>
<td>0.13</td>
<td>0.25</td>
<td>0.26</td>
<td>0.32</td>
</tr>
<tr>
<td>C. Cyclical dynamics</td>
<td>Std. dev. - good times</td>
<td>11.33</td>
<td>6.29</td>
<td>8.45</td>
<td>12.07</td>
<td>-</td>
</tr>
<tr>
<td>Stock returns</td>
<td>Std. dev. - bad times</td>
<td>11.70</td>
<td>6.34</td>
<td>8.69</td>
<td>12.44</td>
<td>-</td>
</tr>
<tr>
<td>Equity premium</td>
<td>Mean - good times</td>
<td>2.48</td>
<td>0.80</td>
<td>2.75</td>
<td>2.83</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td>Mean - bad times</td>
<td>2.97</td>
<td>0.82</td>
<td>3.22</td>
<td>3.40</td>
<td>-</td>
</tr>
</tbody>
</table>

Note: the variable $\chi$ defines the observation cost and $\theta$ is the risk-aversion of agents, which equals 5 in the benchmark model. All statistics are computed in expectation and reported in annualized percentage values. Annual returns are defined as the sum of log monthly returns. The equity premium is $r_{t+1}^P = \mathbb{E}[r_{t+1}^S - r_{t+1}^B]$. The Sharpe ratio is defined as the ratio between the equity premium and its standard deviation. Good times denote the periods in which the aggregate productivity shock is $z_t = z_g$ and bad times denote the periods in which the aggregate productivity shock is $z_t = z_b$. Data is from Campbell (1999) and Guvenen (2009).

to the stock market accounts for around 30% of the observed level of the equity premium.

The observation cost raises the equity premium through two channels. First, as in Guvenen (2009), the limited participation in the stock market concentrates the entire aggregate risk of the economy on a small measure of stockholders, who accordingly demand a higher compensation for holding
equity. Second, inattention exacerbates the curvature of the value function of the agents. Figures 3 - 4 show that inattention raises the implied risk aversion of the households. Although in an economy without observation costs households’ value function is rather flat, with inattention the value function becomes both more concave and more responsive to aggregate conditions. In this way, inattention amplifies the risk associated to holding stocks, especially in bad times.

Overall the model falls shorter in accounting for asset prices when compared to other papers that in the literature study the role of inattention. For instance Chien et al. (2011, 2012) consider a model with inattentive agents which delivers asset prices moments much closer to the data. Although these papers follow different calibrations strategies, the main difference with my paper lies in the modeling of inattention. Chien et al. (2011, 2012) consider an economy with an exogenous measure of agents which trade at exogenously fixed intermittent dates. In the Supplementary Appendix, I study a version of my model with exogenous inattention, in which the duration of inattention is calibrated to match exactly the cross-sectional duration of inattention obtained in the economy with observation costs. I find that moving from an endogenous to an exogenous inattention leads to an overstatement of both the inertia in portfolio rebalancing and the level of the equity premium. Intuitively, when inattention is endogenous, households can choose optimally when to observe the aggregate states of the economy. As a result, households have yet another choice for smoothing their consumption stream, which leads to a more frequent rebalancing of their portfolio and to a lower price of risk.
Figure 3: Slope of the Value Function - Attentive Economy

Note: the figure plots the slope of agents' value function as a function of wealth $\omega$ in an economy no observation cost, i.e. $\chi = 0$. Bad times (continuous line) and good times (dashed line) denote periods in which the aggregate productivity shock is $z_t = z_b$ and $z_t = z_g$. The idiosyncratic shocks are set to $\epsilon_t = 1$ and $\xi_t = 2$.

Figure 4: Slope of the Value Function - Inattentive Economy

Note: the figure plots the slope of agents' value function as a function of wealth $\omega$ in an economy with observation costs, i.e. $\chi = 0.029$. Bad times (continuous line) and good times (dashed line) denote periods in which the aggregate productivity shock is $z_t = z_b$ and $z_t = z_g$. The idiosyncratic shocks are set to $\epsilon_t = 1$ and $\xi_t = 2$. 
5.4.2 Cyclical Dynamics

Inattention generates countercyclical variations in both the stock returns volatility and the equity premium, as shown in Panel C of Table 6. Since the observation cost bites more strongly in recessions, there are very few active investors in the economy, which implies that the investment in physical capital is low and very responsive to the decision of the marginal attentive stockholder. Instead, when the observation cost goes to zero the volatility becomes rather acyclical.

Inattention leads to an equity premium which is countercyclical and displays large variations over the cycle, a result which is usually obtained through consumption habits (Campbell and Cochrane, 1999) or long-run risk (Bansal and Yaron, 2004). In the model, the equity premium equals 2.48% in good times and 2.97% in bad times. This result is in line with the empirical evidence on a positive risk-return trade-off. Again, the cyclicality disappears when I shut down the observation costs.

5.5 The Role of Borrowing Constraints

Chen (2006) considers a Lucas-tree economy where heterogeneous agents face an observation cost, finding that any household owns stock, the portfolio rebalancing is mostly passive, and the equity premium is zero. Instead, in my model there is a vast heterogeneity in the degree of portfolio rebalancing across households, the equity premium is 2.64%, and the observation cost reconciles the amount of limited stock market participation observed in the data with a low participation cost. What is the main feature of the model that allows households’ inattention to matter quantitatively on the dynamics of both portfolio rebalancing and asset prices? In this Section, I show a novel testable mechanism that links households’ inattention to households’ funding
liquidity. I find that the tightness of the borrowing constraints shapes the implications of households’ inattention: inattention affects the dynamics of portfolio rebalancing and asset prices only if borrowing constraints are tight enough.

In what follows, I compare three economies which differ only for the level of the borrowing constraints. The first one is the benchmark model, where the borrowing constraints equal minus three times the monthly income of the median household. In the second case, I consider an economy in which agents cannot borrow at all, while in the last case the constraints are loose and equal minus six times the monthly income of the median household. In this way, I can identify how the effects of households’ inattention depend on credit market frictions. Furthermore, I keep constant the participation cost across the three alternative economies to disentangle the contribution of the borrowing constraints - and their interaction with inattention - on the stock market participation decisions of the households.

Panel A of Table 7 reports three moments from the three economies: the fraction of stockholders, the amount of passive rebalancing that households offset by actively changing their financial portfolio, and the level of the equity premium.

Panel A shows that when I consider an environment with very tight borrowing constraints, the fraction of stockholders falls down dramatically to 50.7%. Portfolio rebalancing becomes more active: now the households can offset 78.9% of the passive change in the risky share. This increase in the active management of financial portfolios is mainly due to a composition effect. The drop in the participation rate implies that the few stockholders of the economy are very wealthy and can afford to incur in the observation cost very often. Hence, stockholders manage on average more actively their portfolios.

When I consider loose borrowing constraints, both the share of stock-
holders and the amount of inertia in portfolio rebalancing rise substantially. In this economy, the households decide to participate in the stock market and they dilute the observation cost by trading very infrequently. Anyway, the loose borrowing constraints allow agents to borrow sufficiently to smooth away any eventual mistake made throughout inattention. As a result, inattention does not affect the price of risk and the equity premium is zero, as in Chen (2006).

Table 7: The role of borrowing constraints

<table>
<thead>
<tr>
<th>Variable</th>
<th>Benchmark</th>
<th>Tight Constraints</th>
<th>Loose Constraints</th>
</tr>
</thead>
<tbody>
<tr>
<td>A. Inattentive economy - $\chi = 0.029$</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>% Stockholders</td>
<td>59.4</td>
<td>50.7</td>
<td>77.1</td>
</tr>
<tr>
<td>$\hat{\psi}$ (Passive Change)</td>
<td>$-0.730$</td>
<td>$-0.789$</td>
<td>$-0.186$</td>
</tr>
<tr>
<td>Equity premium</td>
<td>2.64</td>
<td>6.31</td>
<td>0.36</td>
</tr>
<tr>
<td>B. Attentive economy - $\chi = 0$</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>% Stockholders</td>
<td>89.8</td>
<td>88.2</td>
<td>91.5</td>
</tr>
<tr>
<td>$\hat{\psi}$ (Passive Change)</td>
<td>$-0.998$</td>
<td>$-0.989$</td>
<td>$-0.996$</td>
</tr>
<tr>
<td>Equity premium</td>
<td>0.81</td>
<td>5.01</td>
<td>0.04</td>
</tr>
</tbody>
</table>

Note: The variable $\chi$ defines the observation cost. In the “Benchmark” model, borrowing constraints on bonds and stocks equal around minus three times the monthly income of the median households, that is, $b = s = -5.96$. The “Tight Constraints” model does not allow short sales in neither stocks nor bonds, that is, $s = b = 0$. In the “Loose Constraints” model borrowing constraints on bonds and stocks equal around minus three times the monthly income of the median households, that is, $b = s = -11.92$. Passive Change $\hat{\psi}$ is the estimated amount of passive change which is actively offset by the households by adjusting their portfolios. The parameter is estimated in a panel regression of the active change in portfolio rebalancing on the passive change in portfolio rebalancing

$$ ActiveChange_{i,t+1} = \text{Constant} + \hat{\psi} \times \text{PassiveChange}_{i,t+1} + \mu \times \text{RiskyShare}_{i,t} + \epsilon_{i,t+1}. $$

Importantly, these differences across economies are not driven only by the changes in the tightness of the borrowing constraints, because most of
the action hinges on the interaction between borrowing constraints and the observation cost. To show this mechanism, I replicate the exercise using the three economies above without the observation cost. Panel B of Table 7 reports the results. In this case, the differences in the tightness of the borrowing constraint per se can alter the equity premium, but have no effect whatsoever on the dynamics of portfolio rebalancing.

6 Conclusion

This paper studies whether households’ inattention to the stock market explains the inertia in households’ portfolio rebalancing. To answer this question, I introduce an observation cost into an otherwise standard production economy with heterogeneous agents, idiosyncratic labor income risk, and borrowing constraints. In this model inattention changes endogenously over time and across agents. To discipline the quantitative analysis, I calibrate the observation cost to match the duration of inattention of the median household estimated by Alvarez et al. (2012).

The quantitative results show that inattention accounts for half of the inertia in portfolio rebalancing, and explains its heterogeneity across households. In the model, as it is in the data, wealthy households invest much more actively than poor households. Moreover, inattention can rationalize the limited stock market participation observed in the data, and improves the asset pricing performance of the model. Finally, I highlight a novel testable implication that links households’ inattention to households’ funding liquidity: inattention matters quantitatively on the dynamics of portfolio rebalancing and asset prices only if borrowing constraints are tight enough.
References


Supplementary Appendix
A.1 Computation of the Model

This Section describes the computational algorithm I used to numerically solve the model. The algorithm is an extension to the case of inattention of the standard heterogeneous agent model with aggregate uncertainty and two assets, implemented by Krusell and Smith (1998), Pijoan-Mas (2007) and Gomes and Michaelides (2008).

The numerical computation of heterogeneous agent model with aggregate uncertainty and two assets is very cumbersome. The reason is twofold. First, one of the endogenous aggregate state of the problem is the distribution of the agents over their idiosyncratic states \( \gamma_t \), which is an infinite-dimensional object. As noted by Krusell and Smith (1997), agents need to know the entire distribution \( \gamma_t \) in order to generate rational expectations on prices. To circumvent this insurmountable curse of dimensionality, the state space has to be somehow reduced. I approximate the entire distribution \( \gamma_t \) by a set of moments \( m < \infty \) of the stock of aggregate capital \( K_t \), as in Krusell and Smith (1997), and the number of inattentive agents in the economy \( \zeta_t \). On the one hand, the approximation with a finite set of moments of \( K \) can be interpreted as if the agents of the economy were bounded rational, ignoring higher-order moments of \( \gamma_t \). As in previous studies, I find that \( m = 1 \) is enough to have an almost perfect approximation of \( \gamma_t \). That is, the mean of aggregate capital \( \bar{K}_t \) is a sufficient statistics that capture virtually all the information that agents need to forecast future prices. On the other hand, the variable \( \zeta_t \) signals agents about the degree of informational frictions in the economy. When every agent is attentive, the model shrinks down to the standard Krusell and Smith (1998). Instead, when there is a (non-negligible) measure of inattentive agent, which is the case at the core of my analysis, the model departures from the standard setting. Since the presence of the observation costs pins down different equilibria, and therefore different paths of futures prices, agents are required to be aware of the extent of the informational frictions in the economy whenever taking their optimal choices on consumption and savings.
Second, when extending the basic Krusell and Smith (1997) algorithm to the case of an economy with more than one asset, the market for bonds does not clear at all dates and states: the total bondholdings implied by the model is almost a random walk. Since total bondholdings experience large movements over time, it is not always possible to achieve the clearing of the market. I therefore follow the modified algorithm of Krusell and Smith (1998), where agents perceive the bond return as a state of the economy. The equilibrium bond return is then the one in which the bond return perceived by the agents and the one implied by the optimal decisions of the agents coincide.

The presence of the observation cost adds a further complication. Agents have to decide their optimal duration of inattention. This step requires the derivation of the household’s maximization procedure not just in one case (i.e., today vs. the future), but in a much wider set of alternatives. In the model households can decide whether to be attentive today and tomorrow, whether to be attentive today and inattentive for the following period, or to be attentive today and inattentive for the following two periods, and so on and so forth. Accordingly, I define a grid over all the potential durations of inattention that agents can choose, solve the model over each grid point, and eventually take the maximum among the different value functions to derive the optimal choice of inattention.

The computation of the model requires the convergence upon six forecasting rules which predict the future mean of the stock of aggregate capital, the future price of the bond, and the future number of inattentive agents for both the aggregate shocks \( z_b \) and \( z_g \). The procedure yields a set of twenty-two different parameters upon which to converge. This algorithm is very time-consuming and makes at the moment computationally infeasible any extension of the model that inflates either the mechanisms or the number of states.

In what follows, I first describe the computational algorithm in Section A.1.1. Then, I discuss the problem of the household given the forecasting rule on future prices in Section A.1.2. Finally, Section A.1.3 concentrates on the derivation of the equilibrium forecasting rules. I also show that the substitution of the entire distribution \( \gamma_t \) with the first moment of aggregate capital \( \bar{K}_t \) and the number of inattentive agents \( \zeta_t \) yields an almost perfect approximation.
A.1.1 Algorithm

The algorithm works around nine main steps, as follows:

1. Guess the set of moments $m_t$ of aggregate capital $K_t$ upon which to approximate the distribution of agents $\gamma_t$;

2. Guess the functional forms for the forecasting rule of the set of moments $m_t$, the number of inattentive agents in the economy $\zeta_t$, and the bonds’ risk-free return $r^b_t$;

3. Guess the parameters of the forecasting rules;

4. Solve the household’s problem;

5. Simulate the economy:

   (a) Set an initial distribution of agents over their idiosyncratic states $\omega$, $e$ and $\xi$;

   (b) Find the interest rate $r^{bs}$ that clears the market for bonds. Accordingly, guess an initial condition $r^{b,0}$, solve the household’s problem in which agents perceive the bond return $r^{b,0}$ as a state, and obtain the policy functions $g^c(\omega, e, \xi; z, m, \zeta, r^{b,0})$, $g^b(\omega, e, \xi; z, m, \zeta, r^{b,0})$, $g^a(\omega, e, \xi; z, m, \zeta, r^{b,0})$, $g^s(\omega, e, \xi; z, m, \zeta, r^{b,0})$ and $g^d(\omega, e, \xi; z, m, \zeta, r^{b,0})$. Use the policy functions on bondholdings $g^b$ to check whether the market clears, that is, whether the total amount of bond equals zero. If there is an excess of bond supply, then change the initial condition to $r^{b,1} < r^{b,0}$. If there is an excess of bond demand, then change the initial condition to $r^{b,1} > r^{b,0}$. Iterate until the convergence on the interest rate $r^{bs}$ that clears the market.

   (c) Derive next period distribution of agents over their idiosyncratic states $\omega$, $e$ and $\xi$ using the policy functions $g^c(\omega, e, \xi; z, m, \zeta, r^{bs})$, $g^b(\omega, e, \xi; z, m, \zeta, r^{bs})$, $g^a(\omega, e, \xi; z, m, \zeta, r^{bs})$, $g^s(\omega, e, \xi; z, m, \zeta, r^{bs})$, $g^d(\omega, e, \xi; z, m, \zeta, r^{bs})$. 
\[ g^s(\omega, e, \xi; z, m, \zeta, r^{b,0}) \text{ and } g^d(\omega, e, \xi; z, m, \zeta, r^{b}) \] and the law of motions for the shocks \(z, e\) and \(\xi\).

(d) Simulate the economy for a large number of periods \(T\) over a large measure of agents \(N\). Drop out the first observations which are likely to be influenced by the initial conditions.

6. Use the simulated series to estimate the forecasting rules on \(m_t, \zeta_t\) and \(r_t^b\) implied by the optimal decisions of the agents;

7. Check whether the coefficients of the forecasting rules implied by the optimal decisions of the agents coincide with the one guessed in step (3). If they coincide, go to step (8). Otherwise, go back to step (3);

8. Check whether the functional forms of the forecasting rule as chosen in step (2) give a good fit of the approximation of the state space of the problem. If this is the case, go to step (9). Otherwise, go back to step (2);

9. Check whether the set of moments \(m_k\) of aggregate capital \(K\) yields a good approximation of the distribution of agents \(\gamma\). If this is the case, the model is solved. Otherwise, go back to step (1).

A.1.2 Household’s Problem

I solve the household’s problem using value function iteration techniques. I discretize the state space of the problem as follows. First, I guess that the first moment of aggregate capital and the number of inattention agents are sufficient statistics describing the evolution of the distribution of agents \(\gamma_t\). Later on, I evaluate the accuracy of my conjecture. Then, I follow Pijoan-Mas (2007) by stacking all the shocks, both the idiosyncratic and the aggregate ones, in a single vector \(\epsilon_t\), which has 8 points: four points - one for unemployed agents and three different level for employed agents - for each aggregate shock \(z_t\). For the wealth \(\omega_t\) I use a grid of 60 points.
on a logarithmic scale. Instead, for the possible durations of inattention $d_t$, I use a grid of 30 points: the first 25 points are equidistant and goes from no inattention at all, 1 month of inattention until 2 years of inattention. The following 5 grid points are equidistant on a quarterly basis. In this respect, the assumption made in the model on when inattention breaks out exogenously are very helpful in the definition of the grid. Indeed, agents will not choose too long durations of inattention because they take into account the probability of being called attentive because either they hit the borrowing constraints or they run out of liquid funds. For example, in the benchmark model the largest point of the grid yields a duration of inattention of 3 years. Yet, this choice is hardly picked up by households in the simulations done to solve the model. Without the two assumptions on the exogenous ending of inattention, then some households could theoretically be inattentive forever, which would require a wider grid for the choice variable $d_t$. Then, for the grids of the first moment of aggregate capital $\bar{K}_t$ and the number of inattentive agents $\zeta_t$ I use 6 points since the value function does not display a lot of curvature along these dimensions. To sum up, any value functions is computed over a total of 518,400 different grid points. Furthermore, I need to take into account that the solution method requires the households to perceive the bond return as a state of the economy. I use a grid for $r^b_t$ formed by 10 points, which yields a total of 5,184,000 grid points. Decisions rules off the grid are evaluated using a cubic spline interpolation around along the values of wealth $\omega_t$ and a bilinear interpolation around the remaining endogenous state variables. Finally, the solution of the model is simulated on a set of 3,000 agents over $T=10,000$ time periods. In any evaluation of the simulated series, the first 1,000 observations are dropped out.

The household’s problem used in step (4) of the algorithm modifies the standard structure presented in the text to allow for the approximation of the measure of agents $\mu_t$ with the first moment of aggregate capital $\bar{K}_t$ and the number of inattentive agents $\zeta_t$. Then, I postulate three forecasting rules $(R_1 (\cdot), R_2 (\cdot), R_3 (\cdot))$ for aggregate capital $K_t$, the number of inattentive
agents $\zeta_t$ and the return of the bond $r^b_t$, respectively. The household’s problem reads

$$V(\omega_t, e_t, \xi_t; K_t, \zeta_t) = \max_{d, [c(e_t, \xi_t), c(e_{\lambda(d-1)}, \xi_{\lambda(d)-1})], a_{t+1}, s_{t+1}, b_{t+1}} \mathbb{E}_t \left[ \sum_{j=t}^{\lambda(d)} \beta^{j-t} U(c_j) + \ldots + \beta^{\lambda(d)-t} V(\omega_{\lambda(d)}, e_{\lambda(d)}, \xi_{\lambda(d)}; K_{\lambda(d)}, \zeta_{\lambda(d)}) \right]$$

s.t.  
$$\omega_t = c(e_t, \xi_t) + a_{t+1} + s_{t+1} + b_{t+1} + \phi I_{\{s_{t+1} \neq 0\}}$$

$$\omega_{\lambda(d)} = s_{t+1} \prod_{j=t+1}^{\lambda(d)} \left( 1 + r^s_j (K_j, \zeta_j) \right) + b_{t+1} \prod_{j=t+1}^{\lambda(d)} \left( 1 + r^b_j (K_j, \zeta_j) \right) + \ldots + a_{t+1} + \sum_{j=t+1}^{\lambda(d)} l_j (K_j, \zeta_j) - \sum_{j=t+1}^{\lambda(d)-1} c(e_j, \xi_j) - \chi l_{\lambda(d)} (K_{\lambda(d)}, \zeta_{\lambda(d)}) - \sum_{j=t+2}^{\lambda(d)-1} \phi I_{\{s_j \neq 0\}}$$

$$K_{\lambda(d)} = R_1(K_t, \zeta_t, [z_t, z_{\lambda(d)}])$$

$$\zeta_{\lambda(d)} = R_2(K_t, \zeta_t, [z_t, z_{\lambda(d)}])$$

$$r^b_{\lambda(d)} = R_3(K_t, \zeta_t, [z_t, z_{\lambda(d)}])$$

$$s_{t+1} \geq s, \quad b_{t+1} \geq b, \quad a_{t+1} \geq 0, \quad c(e_t, \xi_t) \geq 0$$

$$\lambda(d) = \min_{j \in [t+1, d]} \left\{ d, b_{t+1} \prod_{k=t+1}^j (1 + r^b_k) < b, \quad s_{t+1} \prod_{k=t+1}^j (1 + r^s_k) < s, \quad \sum_{k=t+1}^j c(e_k, \xi_k) > a_{t+1} + \sum_{k=t+1}^j l_k \right\}$$
Instead, in step (5b) of the problem the households perceive the return of the bond \( r_t^b \) as a state of the economy, as follows

\[
V \left( \omega_t, e_t, \xi_t; K_t, \zeta_t, r_t^b \right) = \max_{d, \left[ c(e_t, \xi_t), c(e_{(d)-1}, \xi_{(d)-1}) \right]} \mathbb{E}_t \left[ \sum_{j=t}^{\lambda(d)} \beta^{j-t} U (c_j) + \ldots + \beta^{\lambda(d)-t} V \left( \omega_{\lambda(d)}, e_{\lambda(d)}, \xi_{\lambda(d)}; K_{\lambda(d)}, \zeta_{\lambda(d)}; r_{\lambda(d)}^b \right) \right]
\]

s.t. \( \omega_t = c(e_t, \xi_t) + a_{t+1} + s_{t+1} + b_{t+1} + \phi_{\{s_{t+1} \neq 0\}} \)

\[
\omega_{\lambda(d)} = s_{t+1} \prod_{j=t+1}^{\lambda(d)} \left( 1 + r_j^s (K_j, \zeta_j, r_j^b) \right) + b_{t+1} \prod_{j=t+1}^{\lambda(d)} \left( 1 + r_j^b (K_j, \zeta_j, r_j^b) \right) + \ldots + a_{t+1} + \sum_{j=t+1}^{\lambda(d)} l_j (K_j, \zeta_j, r_j^b) - \sum_{j=t+1}^{\lambda(d)-1} c(e_j, \xi_j) - \chi l_{\lambda(d)} (K_{\lambda(d)}, \zeta_{\lambda(d)}; r_{\lambda(d)}^b) - \sum_{j=t+2}^{\lambda(d)-1} \phi_{\{s_j \neq 0\}}
\]

\( K_{\lambda(d)} = R_1 \left( K_t, \zeta_t, [z_t, z_{\lambda(d)}] \right) \)

\( \zeta_{\lambda(d)} = R_2 \left( K_t, \zeta_t, [z_t, z_{\lambda(d)}] \right) \)

\( r_{\lambda(d)}^b = R_3 \left( K_t, \zeta_t, [z_t, z_{\lambda(d)}] \right) \)

\( s_{t+1} \geq \xi, \quad b_{t+1} \geq \bar{b}, \quad a_{t+1} \geq 0, \quad c(e_t, \xi_t) \geq 0 \)

\( \lambda(d) = \min_{j \in [t+1, d]} \left\{ d, b_{t+1} \prod_{k=t+1}^{j} (1 + r_k^b) < \bar{b}, s_{t+1} \prod_{k=t+1}^{j} (1 + r_k^s) < \xi, \ldots \right\} \)

\( \ldots \sum_{k=t+1}^{j} c(e_k, \xi_k) > a_{t+1} + \sum_{k=t+1}^{j} l_k \right\} \)

I use this problem to simulate the economy given the return to the bond \( r_t^b \) as a perceived state for the households. I follow Gomes and Michaelides (2008) by aggregating agents’ bond demands and determining the bond return that clears the market through linear interpolation. This value is then used to recover the implied optimal decisions of the agents, which are then aggregated to form the aggregate variables that become that state variables in the following time period.
Furthermore, I have to define the parametric form that let the choices of consumption throughout inattention depend on the realizations of the idiosyncratic shocks $e_t$ and $\xi_t$. Namely, I posit that $c(e_t, \xi_t) = \rho_1 * (1 - e_t) + \rho_2 e_t \xi_t$. In this way, $\rho_1$ captures the absolute amount that an unemployed household consumes, whereas $\rho_2$ determines how consumption changes as a function of the shock to the efficiency unit of hours that employed households experience.

A.1.3 Equilibrium Forecasting Rules

I follow Krusell and Smith (1997, 1998) by defining log-linear functional forms for the forecasting rules of the mean of aggregate stock capital $\bar{K}_t$, the number of inattentive agents $\zeta_t$ and the bond return $r^b_t$. Namely, I use the following law of motions:

$$\log \bar{K} = \alpha_0(z) + \alpha_1(z) \log \bar{K} + \alpha_2(z) \log \zeta$$
$$\log \zeta = \beta_0(z) + \beta_1(z) \log \bar{K} + \beta_2(z) \log \zeta$$
$$r^b = \gamma_0(z) + \gamma_1(z) \log \bar{K} + \gamma_2(z) \log \zeta + \gamma_3(z) (\log \bar{K})^2 + \gamma_4(z) (\log \zeta)^2$$

The parameters of the functional forms depend on the aggregate shock $z_t$. Indeed, there is a set of three forecasting rule for each of the two realizations of the aggregate shock $z$, resulting in a total of six forecasting rules and twenty-two parameters, upon which to find convergence.

I find the equilibrium forecasting rules as follows. First, I guess a set of initial conditions $\{\alpha_0^0(z), \alpha_1^0(z), \alpha_2^0(z), \beta_0^0(z), \beta_1^0(z), \gamma_0^0(z), \gamma_1^0(z), \gamma_2^0(z), \gamma_3^0(z), \gamma_4^0(z)\}$. Then, given such rules I solve the household’s problem. I take the simulated series to then re-estimate the forecasting rules, which yields a new set of implied parameters $\{\alpha_0^1(z), \alpha_1^1(z), \alpha_2^1(z), \beta_0^1(z), \beta_1^1(z), \beta_2^1(z), \gamma_0^1(z), \gamma_1^1(z), \gamma_2^1(z), \gamma_3^1(z), \gamma_4^1(z)\}$. If the two sets coincide (up to a numerical wedge), then these values correspond to the equilibrium forecasting rules. Otherwise, I use the latter set of coefficients as a new initial guess.
For the benchmark specification of the model, I find the following equilibrium forecasting rules for \( z_t = z_g \)

\[
\log \bar{K} = 0.130 + 0.976 \log \bar{K} - 0.251 \log \zeta \quad \text{with } R^2 = 0.993482
\]

\[
\log \zeta = -0.206 + 0.039 \log \bar{K} + 0.881 \log \zeta \quad \text{with } R^2 = 0.995397
\]

\[
r^b = 1.042 - 0.073 \log \bar{K} + 0.012 \log \zeta + \\
+ 0.010 (\log \bar{K})^2 + 0.005 (\log \zeta)^2 \quad \text{with } R^2 = 0.998526
\]

and the following equilibrium forecasting rules for \( z_t = z_b \)

\[
\log \bar{K} = 0.089 + 0.981 \log \bar{K} - 0.240 \log \zeta \quad \text{with } R^2 = 0.994102
\]

\[
\log \zeta = -0.214 + 0.043 \log \bar{K} + 0.840 \log \zeta \quad \text{with } R^2 = 0.997249
\]

\[
r^b = 1.036 - 0.073 \log \bar{K} + 0.024 \log \zeta + \\
+ 0.008 (\log \bar{K})^2 + 0.008 (\log \zeta)^2 \quad \text{with } R^2 = 0.998388
\]

Note that the \( R^2 \) are all above 0.99. This result points out that approximating the distribution of agents \( \gamma_t \) with the first moment of aggregate capital \( \bar{K}_t \) and the number of inattentive agents \( \zeta_t \) implies basically no discharge of relevant information that agents can use to forecast future prices.

### A.2 Further Results

#### A.2.1 An Alternative Model of Inattention

How do the results of the model change if inattentive agents can change their portfolio choices? I address this question by comparing the benchmark economy, in which the consumption of inattentive households depends on the realization of the idiosyncratic shocks, to a counterfactual economy, in which both consumption and the holdings of bonds and stocks of inattentive
households depend on the realization of the idiosyncratic shocks. More precisely, I consider an environment which is exactly similar to the benchmark economy, with the only difference that throughout inattention the portfolio choices partially change as a function of the realizations of the idiosyncratic shocks. In this new version of the model, although inattentive households are not aware of the amount of their financial portfolio, they blindly decide to partially change it.\footnote{I assume that the liquid transaction account clears the movements in stocks and bonds throughout inattention, i.e., an inattentive household uses funds from the transaction account to increase its holdings of bonds and stocks, and vice versa.}

In particular, I let the policy function of the portfolio choices throughout inattention to be defined as follows: for any \( t \in \mathbb{N} : d_i < t < d_{i+1} \), \( s(\epsilon_t, \xi_t) = [\rho_s^1 (1 - \epsilon_t) + \rho_s^2 \epsilon_t \xi_t] \rho_s^\omega \omega_d \) and \( b(\epsilon_t, \xi_t) = [\rho_b^1 (1 - \epsilon_t) + \rho_b^2 \epsilon_t \xi_t] \rho_b^\omega \omega_d \). This specification posits that the optimal portfolio choices move throughout inattention following the realizations of the idiosyncratic shocks \( \epsilon_t \) and \( \xi_t \), and also depend on the value of households’ wealth at the last planning date \( \omega_d \). This last condition follows the idea that households set their choices looking at the level of their wealth, and the best prediction they can have of the evolution of their wealth during inattention is given by the observation of their wealth they made at the last planning date.

Table A.1 compares some implications of the benchmark model with those of the counterfactual economy. The Table shows that the counterfactual economy leads to a lower amount of rebalancing inertia, since households can now actively offset around 84\% of the passive change in the risky share of their portfolio. Also the equity premium shrinks, from 2.64\% to 1.83\%. In addition, the model requires a much higher participation cost to match the observed amount of limited stock market participation.

Although I acknowledge that my modeling of inattention leads to a larger portfolio inertia and equity premium than the counterfactual economy considered in this Section, the underlying assumptions of the benchmark model are more consistent with the data. Indeed, Alvarez et al. (2012) find that in the data only 6\% of households adjust their portfolio more often than they...
observe it (i.e., almost no investor blindly changes the composition of its financial portfolio). Thus, a model of inattention should be able to account for this fact, by reproducing a pattern in which inattentive households do not modify their portfolios.

Table A.1: An Alternative Model of Inattention

<table>
<thead>
<tr>
<th>Variable</th>
<th>Benchmark</th>
<th>Counterfactual Economy</th>
</tr>
</thead>
<tbody>
<tr>
<td>Per-Period Participation Cost (in terms of avg. monthly labor income)</td>
<td>0.8%</td>
<td>1.5%</td>
</tr>
<tr>
<td>( \hat{\psi} ) (Passive Change)</td>
<td>(-0.730)</td>
<td>(-0.844)</td>
</tr>
<tr>
<td>Equity premium</td>
<td>2.64</td>
<td>1.83</td>
</tr>
</tbody>
</table>

Note: In the “Benchmark” model, the consumption choices of inattentive households change as a function of the realizations of the idiosyncratic shocks. In the “Counterfactual Economy”, I consider an environment in which the choices of consumption, bonds, and stocks of inattentive households change as a function of the realizations of the idiosyncratic shocks.

A.2.2 The Role of the Endogeneity of Inattention

How do the results of the model change if inattention is exogenous? I address this question by comparing the benchmark economy, in which households face an observation cost and inattention is endogenously determined, with a counterfactual environment, in which inattention is exogenous. More precisely, I consider an environment which is exactly similar to the benchmark economy, with the only difference that inattention is exogenous. I discipline the comparison by calibrating the amount of exogenous inattention to match exactly the cross-sectional distribution of the duration of inattention generated by the benchmark economy. In this way, the only difference between the two models is given by whether inattention is considered as either an exogenous variable or an endogenous one.
Table A.2 compares some implications of the benchmark economy with those recovered from the model with exogenous inattention. The Table shows that an exogenous inattention requires a lower participation cost to match the observed amount of limited stock market participation. Importantly, an exogenous inattention raises both the lumpiness of portfolio rebalancing - the active rebalancing goes from 73% down to 61% - and also the equity premium, which goes from 2.64% to 3.20%.

Intuitively, when inattention is endogenous, households can choose optimally when to observe the aggregate states of the economy. As a result, households have yet another choice for smoothing their consumption stream, which leads to a more frequent rebalancing of their portfolio, and to a lower price of risk.

These results suggest that the endogeneity of inattention could partially explain why the benchmark economy falls shorter in accounting for asset prices when compared to other models that consider inattention, such as Chien et al. (2011, 2012). From this point of view, the endogeneity of inattention poses further quantitative challenges to any model that aims at matching both the level and the dynamics of asset prices.

Table A.2: Endogenous vs. Exogenous Inattention

<table>
<thead>
<tr>
<th>Variable</th>
<th>Benchmark</th>
<th>Exogenous Inattention</th>
</tr>
</thead>
<tbody>
<tr>
<td>Per-Period Participation Cost (in terms of avg. monthly labor income)</td>
<td>0.8%</td>
<td>0.5%</td>
</tr>
<tr>
<td>$\psi$ (Passive Change)</td>
<td>−0.730</td>
<td>−0.611</td>
</tr>
<tr>
<td>Equity premium</td>
<td>2.64</td>
<td>3.20</td>
</tr>
</tbody>
</table>

Note: In the “Benchmark” model, inattention is endogenous because households face an observation cost. In the “Exogenous Inattention” model I consider an environment which is completely similar to the “Benchmark” model with the only difference that inattention is exogenous. I calibrate the amount of exogenous inattention to match exactly the cross-sectional distribution of the duration of inattention generated by the “Benchmark” model.
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