

# Optimal Consumption and Asset Allocation with Unknown

## Income Growth

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### Abstract

Recent empirical evidence supports the view that the income process has an individual-specific growth rate component (Baker (1997), Guvenen (2007b), and Huggett, Ventura, and Yaron (2007)). Moreover, the individual-specific growth component may be stochastic. Motivated by these empirical observations, I study an individual's optimal consumption-saving and portfolio choice problem when he does not observe his income growth. As in standard income fluctuation problems, the individual cannot fully insure himself against income shocks. In addition to the standard income-risk-induced precautionary saving demand, the individual also has learning-induced precautionary saving demand, which is greater when belief is more uncertain. With constant unobserved income growth, changes in belief are not predictable. However, with stationary stochastic income growth, belief is no longer a martingale. Mean reversion of belief reduces hedging demand on average and in turn mitigates the impact of estimation risk on consumption-saving and portfolio decisions.

*Keywords:* Incomplete markets, precautionary saving, learning, hedging, estimation risk.

*JEL classification:* E2, G11, G31

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## 1. Introduction

Intertemporal consumption-saving and portfolio allocation is a fundamental topic in modern economics. Almost all existing research on this topic assumes that the individual has complete information about the parameters of his income process such as the growth rate

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1 and volatility. The complete-information assumption may be a sensible starting point, if one  
2 believes that an individual's income process can be represented by a growth component com-  
3 mon to all individuals plus an idiosyncratic shock process (MaCurdy (1982) and Abowd and  
4 Card (1989)). Intuitively, if all individuals share the same income growth rate (for example,  
5 among individuals within the same education group), each individual can then estimate the  
6 "common" income growth by using panel data of income within his associated group.

7 However, Baker (1997), Guvenen (2007a), and Huggett, Ventura, and Yaron (2007) pro-  
8 vide convincing empirical evidence in support of a competing view that the income process  
9 has an important *individual-specific* growth component. If income growth is individual spe-  
10 cific, it then becomes much more difficult for each individual to estimate his own income  
11 growth. An individual enters the labor market with a prior belief about his future income  
12 growth and updates his belief over time based on realized incomes. Learning affects the in-  
13 dividual's consumption-saving and portfolio allocation decisions through two channels: the  
14 *expected* growth of income and precautionary saving demand induced by the individual's  
15 learning of his income growth.

16 Income volatility has two effects on consumption and saving. The first is the standard pre-  
17 cautionary saving demand against income fluctuations. Naturally, hedging demand against  
18 income risk is higher when the income stream is more volatile. More interestingly, unknown  
19 income growth also induces hedging demand, which is stochastic and depends on the agent's  
20 time-varying belief. For a given fixed spread between the two possible levels of income growth  
21 rates, estimation risk (induced by learning about his income growth) decreases with income  
22 volatility. This seemingly counter-intuitive result may be explained as follows. Past incomes  
23 from a more volatile income process provide less information about the unknown true income  
24 growth rate. Hence, the agent updates his belief less in response to unanticipated income in-  
25 novations. Therefore, estimation risk is smaller when the underlying income process is more  
26 volatile, *ceteris paribus*. The net impact of income volatility on hedging demand depends on  
27 the relative magnitude of these two opposing effects.

1       Recent work also provides strong empirical evidence consistent with the hypothesis that  
2 income growth is stochastic. Haider (2001) and Guvenen and Kuruscu (2008) document that  
3 cross-sectional dispersion in income growth has been rising since the 1970s. The stochas-  
4 tic feature of income growth further enriches the agent's learning problem, but makes his  
5 consumption-saving and portfolio decisions more complicated. The potential empirical im-  
6 portance of stochastic income growth on decision rules and utility costs calls for models  
7 incorporating *stochastic* income growth. Intuitively, when income growth is stochastic and  
8 unknown, belief change is locally predictable (due to the expected change in income growth)  
9 and hence belief is no longer a martingale as in the case of constant growth. For exam-  
10 ple, when the conditional probability of income growth being low is small, mean reversion  
11 (due to the stochastic transition of income growth from low to high) pulls the agent's be-  
12 lief upward in expectation. Intuitively, this mean reversion of belief makes shocks driving  
13 the change of belief no longer permanent, unlike in settings with unknown *constant* income  
14 growth. The stationary belief updating process in turn lowers the impact of estimation risk  
15 on consumption. Therefore, consumption responds less to change in belief.

16       This paper contributes to the literature on incomplete-markets consumption, saving,  
17 and portfolio choice with learning. Earlier papers that explore the role of partially observed  
18 income on consumption include Goodfriend (1992), Pischke (1995), and a collection of papers  
19 in Hansen and Sargent (1991). All these studies postulate that the agent's consumption is  
20 given by the certainty-equivalence-based permanent-income hypothesis (PIH) rule (Friedman  
21 (1957)), which precludes any possible effect of estimation risk on consumption. The most  
22 closely related papers are Guvenen (2007b) and Wang (2004). Guvenen (2007b) solves for the  
23 consumption rule numerically for agents with constant relative risk-averse utility. His work  
24 complements this one in terms of methodology and economic insights. Unlike Wang (2004),  
25 in the present paper, learning has implications not only on income volatility, but also on  
26 expected changes in income. More importantly, the conditional variance of belief updating  
27 is *stochastic*. As a result, learning induces *stochastic* belief-dependent precautionary saving

1 demand in this paper. Finally, unlike Guvenen (2007b) and Wang (2004), this paper also  
 2 studies the effect of estimation risk on hedging and portfolio allocation.

3 **2. Model Setup**

4 Consider a consumption-saving and portfolio allocation problem. An infinitely-lived agent  
 5 receives an exogenous perpetual stream of stochastic income. Let  $y(t)$  denote the *level* of  
 6 the agent’s time- $t$  labor income. Assume that the dynamics of  $\{y(t) : t \geq 0\}$  is given by

7 
$$dy(t) = (\alpha(t) - \kappa y(t)) dt + \sigma dZ(t), \tag{1}$$

8 where  $Z$  is a standard Brownian motion. The parameter  $\sigma$  measures the conditional volatility  
 9 of the income change over an incremental unit of time. The income growth parameter  
 10  $\{\alpha(t) : t \geq 0\}$  may change stochastically. The detailed specification for  $\alpha$  is deferred to the  
 11 next section. For convergence, assume  $r + \kappa > 0$ , i.e. income cannot grow too fast. When  
 12  $\kappa = 0$ , the income process (1) has a unit root (non-stationary). When  $\kappa > 0$ , the process  
 13 given in (1) is stationary and is known as an Ornstein-Uhlenbeck process.<sup>1</sup> In this case, the  
 14 parameter  $\kappa$  measures the degree of mean reversion. The discrete-time counterpart of (1)  
 15 when  $\kappa > 0$  is represented by the following first-order autoregressive (AR1) process:

16 
$$y(t + 1) = a_0(t) + a_1 y(t) + \hat{\sigma} \epsilon(t + 1), \tag{2}$$

17 where  $a_1 = e^{-\kappa}$ ,  $a_0(t) = \alpha(t) (1 - e^{-\kappa}) / \kappa$ ,  $\hat{\sigma} = \sigma \sqrt{(1 - e^{-2\kappa}) / (2\kappa)}$ , and  $\epsilon(t + 1)$  is a time-  
 18  $(t + 1)$  innovation drawn from the standard normal distribution. The above AR1 process  
 19 has been widely used to model income (Deaton (1992) and Attanasio (1999)). In the precau-  
 20 tionary saving literature, Caballero (1991) uses a discrete-time unit-root process ( $\kappa = 0$ ), a  
 21 special case of (2) to model labor income and derives a closed-form consumption rule. Wang  
 22 (2006) obtains the closed-form consumption rule and characterizes the stochastic precau-  
 23 tionary saving demand for a class of the income process known as “affine” models nesting  
 24 (1) as a special case.

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1 The agent can invest in both a risk-free asset (with a constant rate of return  $r$ ) and a  
 2 risky financial asset (e.g. the market portfolio). Investing in the risky asset offers the agent  
 3 both the opportunity to earn a higher expected return than the risk-free rate  $r$  and the  
 4 benefit of hedging labor-income related risk. The instantaneous return  $dR(t)$  of the market  
 5 portfolio over time increment  $dt$  is given by:

$$6 \quad dR(t) = (r + \zeta) dt + \nu dW(t), \quad (3)$$

7 where  $\zeta$  is the market risk premium,  $\nu$  is the volatility of the market return, and  $W$  is a  
 8 standard Brownian motion. Equation (3) specifies that the market return is independently  
 9 and identically distributed (iid). Let  $\rho$  be the (instantaneous) correlation between the labor  
 10 income process (1) and the return of the risky asset, i.e. the correlation between Brownian  
 11 motions  $Z$  and  $W$  is  $\rho$ . Let  $\eta = \zeta/\nu$  denote the Sharpe ratio of the market portfolio. Let  
 12  $\psi(t)$  denote the amount of wealth that the agent allocates to the market portfolio at time  
 13  $t$ , and hence  $x(t) - \psi(t)$  corresponds to time- $t$  wealth invested in the risk-free asset. The  
 14 agent's financial wealth dynamics is then given by

$$15 \quad dx(t) = (rx(t) + y(t) - c(t)) dt + \psi(t) (\zeta dt + \nu dW(t)) , \quad (4)$$

16 where the first term in (4) gives the sum of interest income  $rx$  (if all wealth is invested in  
 17 the risk-free asset) and labor income  $y$  minus consumption  $c$ . That is, the first term gives  
 18 the saving rate  $s = rx + y - c$  in standard self-insurance models if the agent can only invest  
 19 in the risk-free asset. The last term,  $\psi(t) (\zeta dt + \nu dW(t))$ , captures the “excess” return by  
 20 borrowing at the risk-free rate and investing in the risky asset.

21 The agent has a time-additive separable utility function given by

$$22 \quad U(c) = \mathbb{E} \left( \int_0^\infty e^{-\beta s} u(c(s)) ds \right) , \quad (5)$$

23 where  $\beta > 0$  is his subjective discount rate and  $\gamma > 0$  is the coefficient of absolute risk aversion  
 24 (CARA), i.e.  $u(c) = -e^{-\gamma c}/\gamma$ . It is well known that CARA utility gives much tractability in  
 25 deriving the consumption rule because it ignores the wealth effect. Merton (1971), Kimball

1 (1989), Caballero (1990, 1991), Svensson and Werner (1993), Davis and Willen (2000), and  
 2 Wang (2003, 2006) have all adopted CARA utility in analyzing the agent’s consumption-  
 3 saving decisions under incomplete markets with different income process specifications. The  
 4 agent chooses his consumption  $c$  and wealth allocation to the risky asset  $\psi$  to maximize his  
 5 utility given in (5) subject to his stochastic labor-income process (1), his wealth accumulation  
 6 process (4), and the transversality condition specified in the online appendix.

### 7 3. Model Analysis

8 In standard consumption-saving models, the agent knows both his income process and  
 9 the parameters governing his income process, such as the growth parameter  $\alpha$ . However,  
 10 much empirical evidence suggests that the agent’s income growth may be individual specific  
 11 and hence the agent does not necessarily know his income growth parameter. Learning  
 12 about income growth could potentially have a significant impact on the agent’s intertemporal  
 13 consumption-saving and portfolio allocation rules. Moreover, income growth  $\alpha$  may change  
 14 stochastically over time, further complicating the agent’s decision problem.

15 To understand the effects of learning and the stochastic feature of the income growth  
 16 on consumption-saving and portfolio allocation in an intuitive and pedagogical way, we  
 17 categorize our model into four special sub-models along two dimensions: whether the agent  
 18 knows the value of  $\{\alpha(t) : t \geq 0\}$ , and whether the agent’s income growth  $\{\alpha(t) : t \geq 0\}$   
 19 is stochastic. Table 1 summarizes the structure of the model development in this paper.

[Insert Table 1 here.]

20  
 21 Each special case will provide new insights on the effect of learning on precautionary saving.  
 22 I start with the models where the agent knows the value of his income growth parameter  $\alpha$ .

#### 23 3.1. Models I and III: Known (but possibly stochastic) income growth

24 First, I describe the dynamics for the income growth  $\{\alpha(t) : t \geq 0\}$ , and then analyze  
 25 the agent’s optimality. The agent’s income growth is often subject to both aggregate and

1 idiosyncratic risks. One parsimonious way to capture the stochastic nature of income growth  
 2 is to postulate that income growth  $\{\alpha(t) : t \geq 0\}$  varies stochastically over time between  $\alpha_1$   
 3 and  $\alpha_2 < \alpha_1$ , the two possible levels.<sup>2</sup> Let  $N(t)$  denote the regime for the agent's income  
 4 growth  $\alpha(t)$  at time  $t$ . That is, income growth  $\alpha(t)$  takes the value  $\alpha_{N(t)}$  for  $N(t) = 1, 2$ .  
 5 Fix a small time period  $\Delta t$ . If the time- $t$  growth rate is high, i.e.  $(\alpha(t) = \alpha_1)$ , the growth  
 6 rate remains high at time  $(t + \Delta t)$  with probability  $(1 - \lambda_1 \Delta t)$  and decreases to  $\alpha_2$  at time  
 7  $(t + \Delta t)$  with the remaining probability  $\lambda_1 \Delta t$ . Similarly, if the time- $t$  growth rate is low,  
 8 i.e.  $(\alpha(t) = \alpha_2)$ ,  $\lambda_2 \Delta t$  is the transition probability from low growth rate  $\alpha_2$  to  $\alpha_1$ , the high  
 9 value.<sup>3</sup> Now, I use dynamic programming to characterize the agent's optimality.

Wealth  $x$ , income  $y$ , and income growth are the three state variables. Let  $V(x, y, n)$  denote the value function when the income growth rate is  $\alpha_n$ , where  $n = 1, 2$ . When the current income growth is high  $(\alpha(t) = \alpha_1)$ , the agent's Hamilton-Jacobi-Bellman (HJB) equation is given by

$$\begin{aligned} \beta V(x, y, 1) = \max_{c, \psi} & u(c) + (rx + \psi\zeta + y - c) V_x(x, y, 1) + (\alpha_1 - \kappa y) V_y(x, y, 1) \\ & + \frac{\psi^2 \nu^2}{2} V_{xx}(x, y, 1) + \psi \rho \nu \sigma V_{xy}(x, y, 1) + \frac{\sigma^2}{2} V_{yy}(x, y, 1) \\ & + \lambda_1 (V(x, y, 2) - V(x, y, 1)). \end{aligned} \tag{6}$$

10 The left side of (6) is the “flow” value of the agent's value function. The right side of (6) is  
 11 the sum of his utility flow  $u(c)$  and the instantaneous expected changes in his value function.  
 12 Optimality of consumption and portfolio rules implies that the two sides of (6) are equated.  
 13 The  $V_x$  term describes the marginal increase of the agent's value function from saving. The  
 14  $V_y$  term captures the marginal increase of the agent's value function if income  $y$  increases by

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<sup>2</sup>Kimball (1989) assume that the “level” of income, rather than the growth rate (drift) of the income process, stochastically switches between two states in their study of the impact of precautionary saving on Ricardian (tax) equivalence. Indeed, the income process in Kimball (1989) is a special case of (1).

<sup>3</sup>For the continuous-time regime switching model, the implied stationary probabilities for high-income-growth and low-income-growth states are  $\lambda_2/(\lambda_1 + \lambda_2)$ , and  $\lambda_1/(\lambda_1 + \lambda_2)$ , respectively. See Hamilton (1989) for an early and important application of regime-switching models to economics and econometric analysis in discrete time.

1 a unit. The  $V_{xx}$ ,  $V_{xy}$ , and  $V_{yy}$  terms reflect the effects of stochastic return, income volatility  
 2 and their correlation on the agent's value function. The last term captures the effect of  
 3 stochastic transition of his income growth rate on the expected change in his value function.  
 4 Note that the value function changes discretely from  $V(x, y, 1)$  to  $V(x, y, 2)$  when the growth  
 5 rate changes.

6 The first-order condition (FOC) for consumption is  $u'(c) = V_x(x, y, 1)$ . That is, the  
 7 marginal utility of consumption  $u'(c)$  is equal to the marginal value of wealth  $V_x$ . The FOC  
 8 with respect to the portfolio rule  $\psi$  gives

$$\psi = -\frac{\zeta V_x(x, y, 1)}{\nu^2 V_{xx}(x, y, 1)} - \frac{\rho \sigma V_{xy}(x, y, 1)}{\nu V_{xx}(x, y, 1)}, \quad (7)$$

10 where the first term captures the standard risk-return tradeoff from investing in the market  
 11 portfolio, and the second term reflects the agent's motive to hedge against labor income  
 12 shocks. Next, we incorporate the effect of learning.

### 13 3.2. Models II and IV: Unknown (but possibly stochastic) income growth

14 If the agent does not know his income growth, his time- $t$  information set  $\mathcal{F}_t$  then only  
 15 contains the history of his past incomes  $\{y(s) : s \leq t\}$ , not the true (but possibly stochastic)  
 16 value of  $\alpha(t)$ . Let  $p(t)$  denote his time- $t$  belief that the growth rate is high (i.e.  $\alpha(t) = \alpha_1$ ),  
 17 in that  $p(t) = \text{Prob}(\alpha(t) = \alpha_1 | \mathcal{F}_t)$ . Let  $\mu$  denote the expected growth rate of the income  
 18 process. By definition, the expected growth rate  $\mu$  is a weighted average of the two possible  
 19 income growth rates, in that

$$\mu(t) = \mathbb{E}_t(\alpha(t)) = p(t)\alpha_1 + (1 - p(t))\alpha_2 = \alpha_2 + \delta p(t), \quad (8)$$

21 where

$$\delta = \alpha_1 - \alpha_2 \quad (9)$$

23 is the difference between the two possible values of  $\alpha$ . For a given small time period  $(t, t + \Delta t)$ ,  
 24 the change in income is  $(y(t + \Delta t) - y(t))$ . Out of this total change,  $(\mu(t) - \kappa y(t)) \Delta t$  is the ex-  
 25 pected change. The unanticipated change is given by  $(y(t + \Delta t) - y(t) - (\mu(t) - \kappa y(t)) \Delta t)$ .

1 Scaling by volatility  $\sigma\sqrt{\Delta t}$  and taking the limit as  $\Delta t \rightarrow 0$ , we may construct a “new”  
 2 Brownian motion process  $B$  as follows:

$$3 \quad dB(t) = (dy(t) - (\mu(t) - \kappa y(t)) dt) / \sigma, \quad (10)$$

4 which will serve as the innovations process for belief updating.

5 Re-writing the innovations process (10), we have the following innovations-based repre-  
 6 sentation of the income process (1):

$$7 \quad dy(t) = (\alpha_2 + \delta p(t) - \kappa y(t)) dt + \sigma dB(t), \quad (11)$$

8 where we use (8) for  $\mu(t)$ . Since the agent only observes his past income, the innovations  
 9 representation (11) will naturally be useful when we derive the agent’s optimal consumption  
 10 and portfolio rules. Using the results in Liptser and Shiriyayev (1977), we can write the belief  
 11 process as follows:

$$12 \quad dp(t) = (\lambda_2 - (\lambda_1 + \lambda_2) p(t)) dt + \sigma^{-1} \delta p(t) (1 - p(t)) dB(t), \quad (12)$$

13 where  $B$  is given in (10). Note that belief  $p$  and income  $y$  are perfectly correlated (one shock  
 14 model). We defer the economic interpretations of (12) to later sections when discussing  
 15 model intuition.

16 When the income growth rate  $\alpha$  is unknown, the optimization problem is not Markovian  
 17 with respect to the original information set  $\mathcal{F}_t$ , which only contains the history of income  
 18  $y$ . The belief updating process (12) and the innovations-representation process (11) for  
 19 income  $y$  jointly convey the same information as the agent’s original income process (1) and  
 20 his prior belief about his income growth do. We can transform the original non-Markovian  
 21 optimization problem into a Markovian one. That is, the agent maximizes his utility function  
 22 (5), subject to the innovations-based representation of his income process (11), his belief  
 23 updating process (12), his wealth accumulation equation (4), and the standard transversality  
 24 condition given in the online appendix.<sup>4</sup> Importantly, the agent’s learning about his income

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<sup>4</sup>Gennotte (1986) and Xia (2001) study the optimal asset allocation when the agent has incomplete information about his investment opportunities, such as the dividend growth rate or the expected stock return. See Detemple (1986), Wang (1993), and Veronesi (1999) for equilibrium asset pricing implications of learning.

1 growth implies that belief  $p$  is also a state variable in addition to wealth  $x$  and income  $y$ .

There are three state variables for the agent's optimization problem: wealth  $x$ , income  $y$ , and belief  $p$  that income growth  $\alpha$  is high. The HJB equation for the agent's value function  $J(x, y, p)$  is given as follows:

$$\begin{aligned}
 \beta J = \max_{c, \psi} & u(c) + (rx + \psi\zeta + y - c) J_x + \frac{\psi^2 \nu^2}{2} J_{xx} + (\alpha_2 + \delta p - \kappa y) J_y \\
 & + \psi \rho \nu \sigma J_{xy} + (\lambda_2 - (\lambda_1 + \lambda_2) p) J_p + \frac{\delta^2}{2\sigma^2} p^2 (1-p)^2 J_{pp} \\
 & + \psi \rho \nu \sigma^{-1} \delta p (1-p) J_{xp} + \delta p (1-p) J_{yp} + \frac{1}{2} \sigma^2 J_{yy}.
 \end{aligned} \tag{13}$$

2 The left side of (13) is the annuity (flow measure) of his value function. As for the HJB  
 3 equation (6) when the income growth  $\alpha$  is known, the right side includes standard terms,  
 4 such as  $J_x$ ,  $J_y$ ,  $J_{xx}$ ,  $J_{yy}$ , and  $J_{xy}$ . Unlike the HJB equation (6) for the case with known  
 5 income growth, the agent's learning has additional effects on decision making. For example,  
 6  $J_p$  and  $J_{pp}$  terms capture the effects of the agent's belief about income growth on his value  
 7 function. Since belief updating is solely driven by realized incomes, the agent's income  
 8 process is perfectly correlated with his belief updating, as reflected in the  $J_{yp}$  term in (13).  
 9 Finally, the  $J_{xp}$  term captures the agent's hedging demand induced by his estimation risk  
 10 (associated with belief updating).

#### 11 4. Model I: Known and constant growth parameter $\alpha$

12 I first solve the model and then discuss its economic implications.

13 **Model Solution.** First, consider the setting where  $\alpha(t)$  is *known* and is constant over  
 14 time. This is the standard consumption-saving and portfolio allocation problem for an agent  
 15 endowed with uninsurable stochastic labor income. The transition probability out of the  
 16 current income growth  $\alpha$  is zero. That is,  $\alpha(t) = \alpha$  for all  $t$ . The following proposition  
 17 summarizes the main results on consumption and portfolio rules for this setting, dubbed as  
 18 Model I:

**Proposition 1** *If the agent knows his constant income growth  $\alpha$ , his consumption  $c^*$  and wealth allocation to the risky asset  $\psi^*$  are given by*

$$c^*(t) = r(x(t) + g(y(t); \alpha)), \quad (14)$$

$$\psi^*(t) = \frac{\zeta}{\gamma r \nu^2} - \frac{\rho \sigma}{\nu} \frac{1}{r + \kappa}, \quad (15)$$

1 where the risk-adjusted certainty equivalent human wealth  $g(y; \alpha)$  is given by

$$2 \quad g(y; \alpha) = \frac{1}{r + \kappa} \left( y + \frac{\alpha - \rho \sigma \eta}{r} \right) - \frac{\gamma(1 - \rho^2)\sigma^2}{2(r + \kappa)^2} + \frac{\beta - r}{\gamma r^2} + \frac{\eta^2}{2\gamma r^2}. \quad (16)$$

3 **Model intuition and implications.** The agent's investment opportunity in the risky  
 4 asset has two effects. First, the risky asset offers a higher expected return and hence shall  
 5 raise the forward-looking agent's current consumption (Merton (1971)). This effect is cap-  
 6 tured by the first term in the agent's portfolio rule (15), and also by the constant positive  
 7 term  $\eta^2/(2\gamma r^2)$  in the agent's risk-adjusted certainty equivalent wealth  $g(y)$  given in (16).  
 8 Second, investing in the risky asset allows the agent to partially hedge against his labor in-  
 9 come risk (i.e. the second term in the portfolio allocation rule (15)). The agent has a higher  
 10 hedging demand, if the systematic volatility  $\rho\sigma$  is larger. Hedging changes the agent's labor  
 11 income growth from  $\alpha$  to  $(\alpha - \rho\sigma\eta)$ , and also reduces the non-diversifiable component of his  
 12 labor income volatility from  $\sigma$  to  $\sigma\sqrt{1 - \rho^2}$ . Since precautionary saving demand arises from  
 13 the agent's non-diversifiable idiosyncratic risk, hedging lowers the agent's precautionary sav-  
 14 ing demand. We measure the agent's precautionary saving demand as the amount by which  
 15 the certainty equivalent wealth  $g(y)$  is lower than the corresponding certainty equivalent  
 16 wealth under the PIH rule. It is immediate to see that the precautionary saving demand  
 17 only depends on idiosyncratic volatility  $\sigma\sqrt{1 - \rho^2}$  and is given by

$$18 \quad \pi(t) = \frac{\gamma(1 - \rho^2)\sigma^2}{2(r + \kappa)^2}. \quad (17)$$

19 The more persistent income shocks are (lower  $\kappa$ ), the greater the agent's precautionary saving  
 20 demand  $\pi$  is, which is consistent with our analysis on hedging demand  $\psi$ . If labor income is

1 perfectly correlated with the risky asset return, the agent can fully hedge his income risk. As  
 2 a result, his precautionary saving demand is zero (i.e. complete markets setting). Finally,  
 3 Model I nests Caballero (1991) and Wang (2006), settings where the agent cannot invest in  
 4 the risky asset ( $\psi(t) = 0$ ), as special cases.<sup>5</sup>

5 To sum up, Model I generates the following empirically testable predictions. First, pre-  
 6 cautionary saving demand is higher for the more persistent income process. Second, for  
 7 incomes more correlated with the market portfolio, hedging demand is higher, and hence  
 8 precautionary saving demand is lower.

9 Next, I analyze the case where income growth is constant but unknown.

## 10 5. Model II: Unknown and constant growth parameter $\alpha$

11 First, I analyze the agent’s Bayesian learning problem and then use dynamic program-  
 12 ming to solve for his decision rules. Finally, I highlight the model implications on learning-  
 13 induced precautionary saving.

14 **Model Solution.** When the agent does not know his income growth, he needs to use  
 15 his past realized incomes to estimate the likelihood that his income growth  $\alpha$  is high. Note  
 16 that Model II is a special case of the general learning model of Section 3.2. with  $\lambda_1 = \lambda_2 = 0$ .  
 17 We may write the belief updating process (12) as follows:

$$18 \quad dp(t) = \sigma^{-1} \delta p(t) (1 - p(t)) dB(t), \tag{18}$$

19 where  $B$  is the Brownian motion process under the *innovations* representation given in (10).  
 20 Note that the belief updating process (18) is a martingale. This is because the unknown

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<sup>5</sup>Wang (2006) extends the discrete-time CARA-Gaussian formulation of Caballero (1991) in a continuous-time setting to allow for conditionally heteroskedastic labor income process. The key advantage of introducing conditional heteroskedasticity of labor income process is that the agent’s marginal propensity to consume (MPC) out of labor income may be less than the MPC out of financial wealth, a desirable feature argued in Friedman (1957) , Hall (1978) , and Zeldes (1989). Using isoelastic-utility-based buffer-stock-type saving models, Deaton (1991), Carroll (1997), and Gourinchas and Parker (2002) also generate this desirable feature on the MPC.

1 growth rate  $\alpha$  is *constant*, and hence the change in the agent's belief must be unpredictable,  
 2 in that  $\mu(t) = \mathbb{E}_t(\alpha(t)) = \mathbb{E}_t(\mathbb{E}_s(\alpha(t))) = \mathbb{E}_t(\mu(s))$ , for any  $t < s$ .

3 The instantaneous volatility of belief updating is symmetric in  $p$  and  $(1 - p)$  because the  
 4 unobserved growth rate can only take two possible values:  $\alpha_1$  and  $\alpha_2$ . The greater the wedge  
 5  $\delta = \alpha_1 - \alpha_2 > 0$  is, the more volatile belief updating is. Moreover, a higher income volatility  
 6  $\sigma$  implies a less volatile belief updating. Intuitively, a higher realized value of income is more  
 7 informative about the unknown income growth if the income process is less volatile (lower  
 8  $\sigma$ ). The following proposition summarizes the results on consumption and portfolio rules  
 9 when the agent learns about his (constant) income growth.

**Proposition 2** *If income growth  $\alpha$  is constant but unknown to the agent, his consumption  $c^*$  and wealth allocation to the risky asset  $\psi^*$  are given by*

$$c^*(t) = r(x(t) + g(y(t); \alpha_2) + f(p(t))), \quad (19)$$

$$\psi^*(t) = \frac{\zeta}{\gamma r \nu^2} - \frac{\rho}{\nu} \frac{\sigma}{r + \kappa} - \xi(t), \quad (20)$$

10 where  $g(y; \alpha_2)$  is given by (16), learning-induced hedging demand  $\xi$  is

$$11 \quad \xi(t) = \frac{\rho}{\nu} \frac{\delta p(t) (1 - p(t)) f'(p(t))}{\sigma}, \quad (21)$$

and  $\{f(p) : 0 \leq p \leq 1\}$  solves the following non-linear ODE:

$$\begin{aligned} r f(p) = & \frac{\delta p}{r + \kappa} - \frac{\rho \eta}{\sigma} \delta p (1 - p) f'(p) + \frac{\delta^2}{2\sigma^2} p^2 (1 - p)^2 f''(p) \\ & - \gamma r (1 - \rho^2) \left[ \frac{1}{r + \kappa} \delta p (1 - p) f'(p) + \frac{\delta^2}{2\sigma^2} p^2 (1 - p)^2 f''(p) \right], \end{aligned} \quad (22)$$

12 subject to  $f(0) = 0$  and  $f(1) = \delta / (r(r + \kappa))$ .

13 **Model implications: Learning, precautionary saving, and hedging.** When the  
 14 agent learns about his income growth  $\{\alpha(t) : t \geq 0\}$ , his certainty equivalent wealth has  
 15 an additional term  $f(p)$ , which depends on  $p$ . If income growth is low with probability one

1 at all times (i.e.  $\alpha(t) = \alpha_2$  for all  $t$ ), we are back to Model I with constant known income  
 2 growth  $\alpha_2$  with  $f(0) = 0$ . Similarly, if income growth  $\alpha$  is always high (i.e.  $\alpha(t) = \alpha_1$  for all  
 3  $t$ ), we have  $f(1) = \delta/(r(r + \kappa))$ . Note that  $p = 0$  and  $p = 1$  are absorbing states here. For  
 4  $0 < p < 1$ , we need to solve for  $f(p)$  given in (22).

5 Learning has two effects: the expected growth rate of income and the precautionary  
 6 saving demand effect. In the following analysis, we separate these two effects on  $f(p)$ . Since  
 7 the *expected* growth rate of income exists for any agent, we first solve for  $f(p)$  for risk-  
 8 neutral agents ( $\gamma = 0$ ). Let  $\bar{f}(p)$  denote the certainty equivalent wealth satisfying (22) for  
 9 risk-neutral agents (i.e.  $\gamma = 0$ ). Intuitively,  $\bar{f}(p)$  captures the learning effect via the channel  
 10 of the *expected* growth rate of income. Let  $l(p) = \bar{f}(p) - f(p)$ , where  $f(p)$  solves (22) for  
 11 a given  $\gamma \geq 0$ . The wedge  $l(p)$  captures the second effect: learning-induced precautionary  
 12 saving demand.

13 Consider the special setting with  $\rho = 0$ , which includes the standard self-insurance (in-  
 14 come fluctuation) problem (with risk-free asset only) as a special case. When  $\rho = 0$ , the  
 15 solution for  $\bar{f}(p)$  is linear and is given by

$$16 \quad \bar{f}(p) = \frac{\delta p}{r(r + \kappa)}. \tag{23}$$

17 Figure 1 plots the “risk-adjusted” certainty equivalent wealth  $f(p)$  and the wedge  $l(p) =$   
 18  $\bar{f}(p) - f(p)$  for  $\gamma = 0, 1, 2$ . We use the following (annualized and continuously compounded)  
 19 parameters for the remainder of the paper unless otherwise noted. The interest rate  $r = 4\%$ ,  
 20 the dispersion of income growth  $\delta = \alpha_1 - \alpha_2 = 3\%$ , income volatility  $\sigma = 40\%$ , and the  
 21 degree of income mean reversion  $\kappa = 5\%$ .

[Insert Figure 1 here.]

22 The left panel of Figure 1 shows that  $f(p)$  is increasing in  $p$  for a given  $\gamma$ . Moreover,  
 23  $f(p)$  decreases with the coefficient of absolute risk aversion  $\gamma$ . The right panel plots the  
 24 learning-induced precautionary saving  $l(p) = \bar{f}(p) - f(p)$ . Note that  $l(0) = l(1) = 0$  because  
 25 there is no more uncertainty if  $p = 0, 1$  under constant but unknown growth (Model II).

1 Learning-induced precautionary saving  $l(p)$  is concave in  $p$ . Intuitively, when the agent is  
 2 more uncertain about his income growth (i.e. in the interior region of  $p$ ), learning induced  
 3 precautionary saving  $l(p)$  is higher, *ceteris paribus*. However, note that  $l(p)$  is not symmetric  
 4 around  $p = 1/2$ , and is rather skewed. This is due to the fact that  $f(p)$  is convex, (i.e.  
 5  $f'(1-p) > f'(p)$  for  $0 < p < 1/2$ ), and the fact that  $l(p)$  depends both on  $p(1-p)$  and  $f'(p)$ .  
 6 The nonlinear term in ODE (22) and the right panel of Figure 1 capture this asymmetry.

7 Now I consider the effect of learning when the agent can invest in the risky asset. As in  
 8 Model I and Merton (1971), the agent earns a higher expected return, (i.e. the first term  
 9  $\zeta/(\gamma r \nu^2)$  in (20)), hedges the systematic component of his labor income risk (the second  
 10 term in (20)), and also hedges the correlated component of his income growth risk (the  $\xi(t)$   
 11 term given in (21)). The hedging demand with respect to the labor income risk is the same  
 12 as in our Model I, Svensson and Werner (1993), and Davis and Willen (2000). The hedging  
 13 demand with respect to the estimation risk is however, stochastic, and depends on the time-  
 14 varying volatility  $\sigma^{-1}\delta p(1-p)$  of the belief updating process (18) and  $f'(p)$ , which measures  
 15 the sensitivity of  $f(p)$  with respect to belief  $p$ .

16 Income volatility  $\sigma$  has two *opposite* effects on the total hedging demand. On the one  
 17 hand, a higher income volatility  $\sigma$  increases the hedging demand of labor income risk. On  
 18 the other hand, incomes from a more volatile income process provide less precise information  
 19 about the unknown income growth  $\alpha$  given a fixed dispersion  $\delta = \alpha_1 - \alpha_2$ . Hence, the agent  
 20 updates his belief less in response to “unexpected” income news. Therefore, a higher income  
 21 volatility  $\sigma$  maps to a lower estimation risk and a lower hedging demand against estimation  
 22 risk, *ceteris paribus*.

[Insert Figure 2 here.]

23 The left panel of Figure 2 plots the certainty equivalent wealth  $f(p)$  for three levels of  
 24 income volatility  $\sigma = 0.1, 0.2, 0.3$  with correlation  $\rho = 0.5$  (the other parameters are the same  
 25 as those for Figure 1). Note that  $f(p)$  *increases* with income volatility  $\sigma$  for  $\rho > 0$ , whereas  
 26  $g(y; \alpha_2)$  decreases with  $\sigma$ . If  $\rho < 0$ , the opposite holds:  $f(p)$  increases with income volatility

$\sigma$  and is concave in  $p$ . The right panel of Figure 2 plots learning-induced hedging demand  $\xi(p)$ . Note that hedging demand  $\xi(p)$  decreases with income volatility  $\sigma$ . This seemingly counter-intuitive result is due to the assumption that income changes are less informative for a given  $\delta = \alpha_1 - \alpha_2$  (recall that the conditional volatility of income changes (in terms of levels) is constant in our model.) However, in reality, income volatility may increase with the income level. If so, then estimation risk will also increase with income volatility. In that case, the estimation-risk-induced precautionary saving and the standard income risk effect on precautionary saving may potentially move in the same direction, as shown by Guvenen (2007b). In that paper, the logarithmic income process is conditionally homoskedastic, which implies that the volatility of income increases with the level of income and hence estimation risk may increase with the level of income.

Empirically, an interesting and testable prediction is the effect of income risks on the precautionary saving demand induced by estimating income growth. We have highlighted a potential mechanism which makes the estimation risk lower when income is riskier. Again, we need to carefully control for the level effect of income growth estimation risk on precautionary saving demand. Having analyzed the impact of learning when income growth is constant, we now turn to the more general setting where income growth is stochastic.

## 6. Stochastic income growth

I first solve the model with stochastic income growth but without learning as a benchmark. Then, I solve the model with learning and interpret the economics of learning about stochastic income growth.

### Model III: Known and stochastic growth parameter $\alpha$ .

Recall that income growth is given by a regime-switching model. While stochastic, income growth is known to the agent. The agent's information set  $\mathcal{F}_t$  includes  $\{N(s) : s \leq t\}$ , where  $N(t) = 1, 2$  correspond to the high and the low income growth rates, respectively. The following proposition summarizes the main results of Model III.

1 **Proposition 3** *When the agent knows income growth  $\{\alpha(t) : t \geq 0\}$ , his portfolio allocation*  
 2 *is given by (15), and his consumption is given by*

$$3 \quad c^*(t) = r(x(t) + g(y(t); \alpha_2) + \phi_{N(t)}), \quad (24)$$

where  $g(y; \alpha_2)$  is given in (16) and  $\{\phi_1, \phi_2\}$  jointly solve

$$r\phi_1 = -\frac{\lambda_1}{\gamma r} (e^{-\gamma r(\phi_2 - \phi_1)} - 1) + \frac{\delta}{r + \kappa}, \quad (25)$$

$$r\phi_2 = -\frac{\lambda_2}{\gamma r} (e^{-\gamma r(\phi_1 - \phi_2)} - 1). \quad (26)$$

4 The portfolio rule is also given by (15), the same as in Model I. Equations (25) and (26)  
 5 jointly characterize the growth-dependent consumption profiles:  $\phi_1$  and  $\phi_2$ . Compared with  
 6 Model I, the *stochastic growth* feature of income induces additional precautionary saving  
 7 demand.

8 **Model IV: Unknown and stochastic growth parameter  $\alpha$ .**

Without observing his income growth  $\alpha$ , the agent uses his past incomes to estimate the likelihood that his income growth is high. Equation (12) gives the belief updating process. The intuition for the volatility specification in (12) is the same as the one for the belief process (18) in Model II, a special case of Model IV. See discussions on volatility in Section 5 for Model II. The intuition for the drift specification in (12) is richer than the one for Model II. Because the underlying unknown income growth  $\alpha$  is *stochastic*, the expected change in the agent's belief is no longer zero, unlike Model II in Section 5. Consider a small time period  $(t, t + \Delta t)$ . Suppose the current income growth is high (i.e.  $\alpha(t) = \alpha_1$ ). The conditional probability that income growth changes from  $\alpha_1$  to  $\alpha_2$  is  $\lambda_1 \Delta t$ . The size of this change is  $\alpha_2 - \alpha_1 = -\delta$ . Therefore, the expected change in income growth (conditional on  $\alpha(t) = \alpha_1$ ) is  $-\delta \lambda_1 \Delta t$ . The time- $t$  probability that  $\alpha(t) = \alpha_1$  is  $p(t) = \text{Prob}_t(\alpha(t) = \alpha_1)$ .

The unconditional expected change in income growth is thus given by

$$\mathbb{E}_t(\mu(t + \Delta t) - \mu(t)) = -p(t)\delta\lambda_1\Delta t + (1 - p(t))\delta\lambda_2\Delta t \quad (27)$$

$$= \delta[\lambda_2 - (\lambda_1 + \lambda_2)p(t)]\Delta t. \quad (28)$$

1 Using  $\mu(t) = \alpha_2 + \delta p(t)$ , we can show that the drift of  $p(t)$  is equal to  $(\lambda_2 - (\lambda_1 + \lambda_2)p(t))$ .  
 2 The above analysis on the drift and the analysis on volatility in the previous section jointly  
 3 provide an economically intuitive explanation for the Bayesian updating rule (12). When  
 4 the belief  $p(t)$  is larger than the (unconditional) long-run probability  $\lambda_2/(\lambda_1 + \lambda_2)$ , the belief  
 5  $p(t)$  is expected to move downward on average. This reflects the mean reversion property  
 6 of the belief process  $\{p(t) : t \geq 0\}$ . Using this belief updating rule, we solve the agent's  
 7 decision problem and summarize the results in the following proposition.

**Proposition 4** *When the agent does not know his stochastic income growth  $\alpha$ , his consumption  $c^*$  and wealth allocation  $\psi^*$  are given by (19) and (20), respectively, where  $g(y; \alpha_2)$  is given in (16), and  $\{f(p) : 0 \leq p \leq 1\}$  solves*

$$\begin{aligned} rf(p) = & \frac{\delta p}{r + \kappa} + \left[ (\lambda_2 - (\lambda_1 + \lambda_2)p) - \left( \frac{\rho\eta}{\sigma} + \gamma \frac{r(1 - \rho^2)}{r + \kappa} \right) \delta p(1 - p) \right] f'(p) \\ & + \frac{\delta^2}{2\sigma^2} p^2 (1 - p)^2 f''(p) - \frac{\gamma r (1 - \rho^2)}{2\sigma^2} \delta^2 p^2 (1 - p)^2 f'^2, \end{aligned} \quad (29)$$

*subject to the following boundary conditions:*

$$rf(0) = \lambda_2 f'(0), \quad (30)$$

$$rf(1) = \frac{\delta}{r + \kappa} - \lambda_1 f'(1). \quad (31)$$

## 8 Results, intuition, and implications.

9 Unlike Model II (with constant growth), the states  $p = 0$  and  $p = 1$  in Model IV are no  
 10 longer absorbing. The intuition is as follows. With stochastic growth, income growth may

1 become low with probability  $\lambda_1 \Delta t$  over time increment  $(t, t + \Delta t)$ , even if the agent knows for  
 2 sure his income growth is high at time  $t$ . Therefore, learning induces precautionary saving  
 3 demand at all times for any levels of belief  $p$ , provided that income growth is stochastic.

4 I now turn to the impact of income growth persistence  $(\lambda_1, \lambda_2)$  on consumption-saving  
 5 decisions. Recall that learning-induced precautionary saving  $l(p)$  is given by the difference  
 6 between the certainty equivalent wealth  $f(p)$  and  $\bar{f}(p)$ , where  $\bar{f}(p)$  solves (29) for  $\gamma = 0$ , i.e.  
 7  $l(p) = \bar{f}(p) - f(p)$ . The left and the right panels of Figure 3 plot the certainty equivalent  
 8 wealth  $f(p)$  and learning-induced precautionary saving  $l(p)$ , respectively. In addition to the  
 9 parameters we have used in Figures 1 and 2, we set market (portfolio) risk premium  $\zeta = 6\%$   
 10 and market (portfolio) return volatility  $\nu = 20\%$ .

[Insert Figure 3 here.]

11 On the left panel of Figure 3, we see that either a higher value of  $\lambda_1$  or of  $\lambda_2$  makes  $f(p)$   
 12 flatter. Of course, increasing  $\lambda_2$ , the transition probability from low income growth to high  
 13 income growth, on average makes income growth higher and hence  $f(p)$  larger. For example,  
 14 comparing the setting of  $(\lambda_1, \lambda_2) = (0, 0)$  with that of  $(\lambda_1, \lambda_2) = (0, 3\%)$ , we see that  $f(p)$  is  
 15 higher when  $\lambda_2 = 3\%$  than when  $\lambda_2 = 0$ . Moreover, with  $\lambda_1 = 0$ , the high-income-growth  
 16 state is absorbing, and  $f(1) = \delta / (r(r + \kappa))$  for both settings as we see from the figure.

17 The right panel of Figure 3 plots learning-induced precautionary saving demand  $l(p)$ . In-  
 18 tuitively, with stochastic growth, income growth is more transitory and hence precautionary  
 19 saving demand is lower when belief in the interior region, *ceteris paribus*. Comparing the set-  
 20 ting of  $(\lambda_1, \lambda_2) = (0, 0)$  with the setting of  $(\lambda_1, \lambda_2) = (3\%, 3\%)$ , we see that learning-induced  
 21 precautionary saving demand is higher with constant income growth (i.e.  $(\lambda_1, \lambda_2) = (0, 0)$ )  
 22 other than near the boundaries  $p = 0$  and  $p = 1$ .

23 Figure 4 plots learning-induced hedging demand  $\xi$  as a function of belief  $p$ . The left  
 24 panel analyzes the impact of transition intensities  $(\lambda_1, \lambda_2)$  on  $f(p)$ . As in Model II, learning-  
 25 induced hedging demand  $\xi$  is concave in  $p$ . More interestingly, hedging demand  $\xi$  decreases  
 26 with  $\lambda$ . Intuitively, the more transitive the income growth is, the lower learning-induced

1 hedging demand  $\xi$  is. The right panel of Figure 4 shows that learning-induced hedging  
 2 demand  $\xi$  decreases with income volatility  $\sigma$ . This result is consistent with the one for Model  
 3 II. Intuitively, the same realized changes in income from a less volatile stream are more  
 4 informative and reflect more changes in income growth. Hence, learning-induced hedging  
 5 demand is higher when income streams are less volatile.

[Insert Figure 4 here.]

6 When income growth is stochastic and unknown, the change in beliefs is locally pre-  
 7 dictable. Mean reverting beliefs imply that shocks are no longer permanent, unlike in settings  
 8 with unknown *constant* income growth. The stationarity of belief implies that the agent's  
 9 learning process is less volatile and the estimation risk is lower. Therefore, learning-induced  
 10 precautionary saving demand is lower and consumption responds less to the change in belief,  
 11 when unknown income growth is stochastic.

## 12 7. Conclusions

13 In this paper, I study the effect of learning about income growth on an individual's  
 14 consumption-saving and portfolio choice decisions when he cannot fully diversify his labor-  
 15 income risk. The individual uses the dynamic Bayesian rule to update his belief about his  
 16 income growth. Estimation risk naturally arises from his learning process. Importantly,  
 17 this estimation risk generates additional precautionary savings demand beyond the standard  
 18 income-risk-induced precautionary savings. By investing in the risky asset, the agent par-  
 19 tially hedges against both income risk and estimation risk. While higher income volatility  
 20 induces greater hedging demand against income shocks, higher income volatility (for a fixed  
 21 income growth wedge) also induces less volatile belief updating because realized income is  
 22 less informative about unknown income growth. Hence, higher income volatility implies  
 23 lower estimation risk, which in turn suggests a smaller learning-induced hedging demand.

24 When income growth is stochastic and unknown, the agent's learning about income  
 25 growth becomes even less volatile. Intuitively, the change in beliefs is locally predictable

1 due to the expected change in income growth and hence beliefs are no longer martingales.  
2 Mean reversion of beliefs makes shocks driving the change in beliefs more transitory, unlike  
3 in settings with unknown *constant* income growth. The stationary belief updating process  
4 in turn lowers the impact of estimation risk on consumption. Therefore, when beliefs are  
5 not extreme (i.e at corners  $p = 0, 1$ ), learning-induced precautionary saving demand is lower  
6 and consumption responds less to belief change, when unknown income growth is stochastic  
7 rather than constant.

8 The main objective of this paper is to study the effects of incomplete information about  
9 the income growth rate on his consumption and portfolio allocations, when the agent's  
10 income shocks are not insurable. In order to deliver this intuition in a simplest possible way,  
11 I have intentionally chosen the CARA utility for technical convenience. While analytically  
12 convenient, this utility specification ignores the wealth effect on consumption and portfolio  
13 allocation rules. The natural next step is to extend the analysis to settings with iso-elastic  
14 utility, which capture the wealth effect and hence allow for making quantitative assessments  
15 on the role of learning about income growth.

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Table 1:

Categorization of Models

income growth $\alpha$	known	unknown
constant	Model I	Model II
stochastic	Model III	Model IV

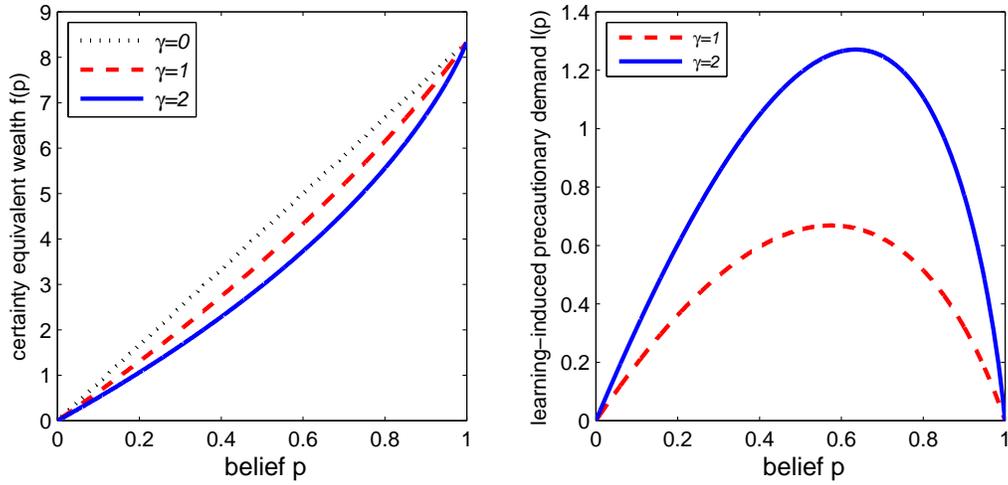


Figure 1: The certainty equivalent wealth  $f(p)$  and learning-induced precautionary saving  $l(p) = \bar{f}(p) - f(p)$  in Model II (with  $\rho = 0$ ): The effects of risk aversion.

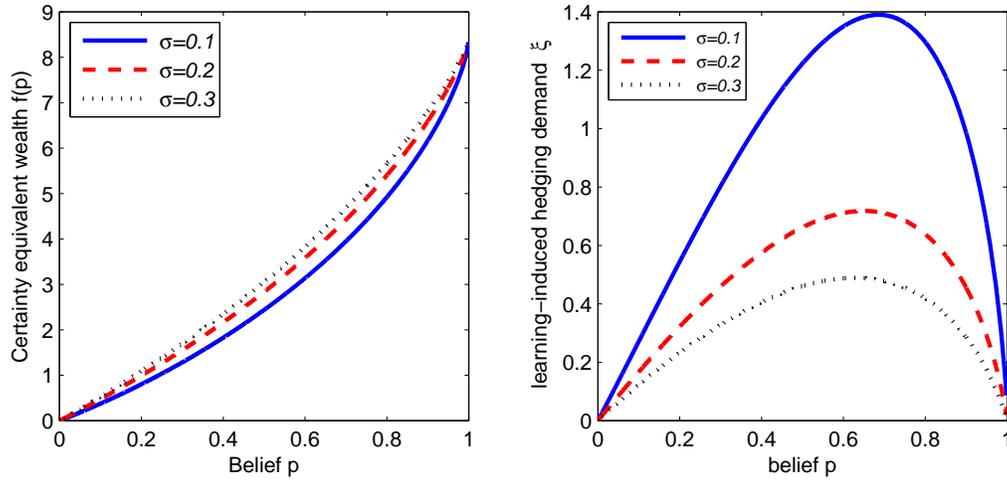


Figure 2: The certainty equivalent wealth  $f(p)$  and learning-induced hedging demand  $\xi(p)$  in Model II: The effects of income volatility  $\sigma$ . The correlation coefficient:  $\rho = 0.5$ . Other parameter values are the same as those for Figure 1.

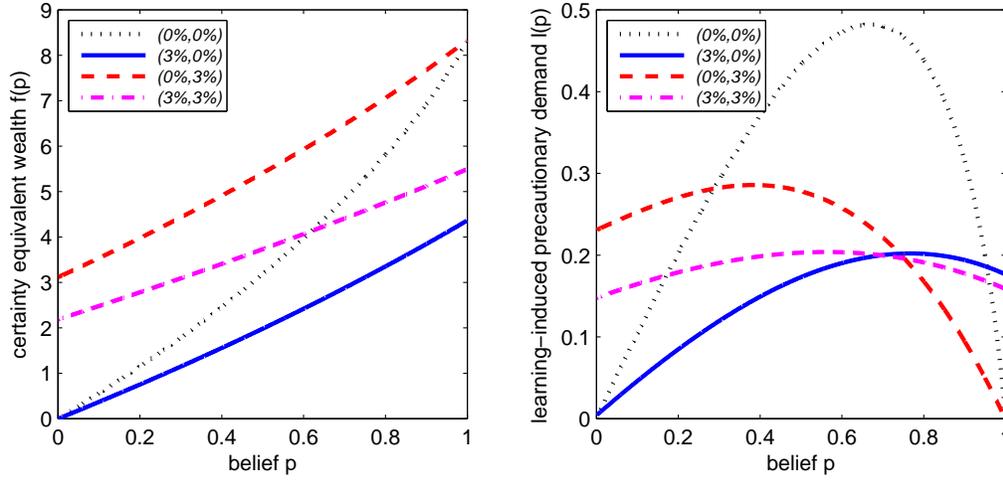


Figure 3: The certainty equivalent wealth  $f(p)$  and learning-induced precautionary demand  $l(p)$  in a stochastic income growth model with learning in Model IV: The effects of transition intensities  $(\lambda_1, \lambda_2)$ . Other parameter values are the same as those for Figure 2.

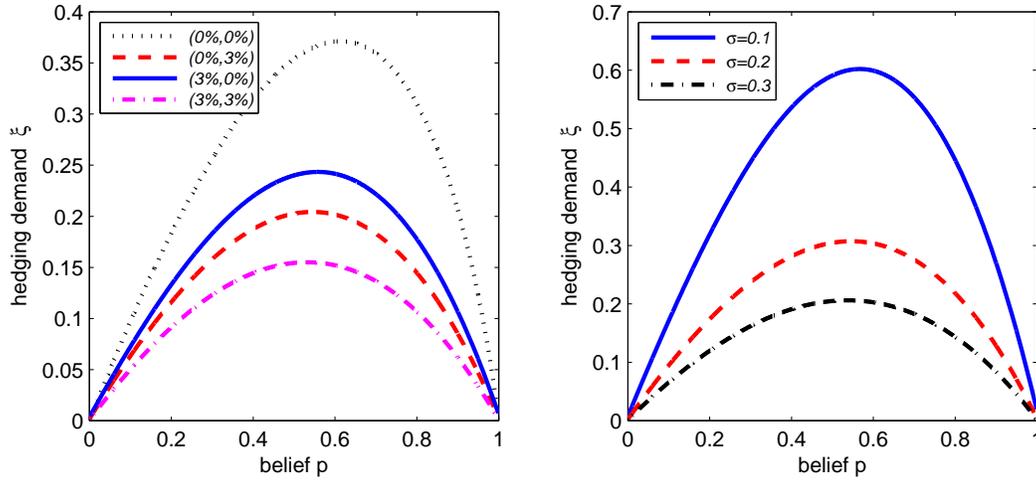


Figure 4: **Learning-induced hedging demand  $\xi(p)$  in a stochastic income growth model with learning (Model IV).** The left panel plots for various income growth transition parameters  $(\lambda_1, \lambda_2)$ . The right panel plots for three levels of  $\sigma$  for the setting with  $\lambda_1 = \lambda_2 = 3\%$ . Other parameter values are the same as those for Figure 3.