The States vs. the states:

On the Welfare Cost of Business Cycles in the U.S.

Stéphane Pallage

Michel A. Robe\*

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#### Abstract

Extant estimates of the welfare cost of business cycles suggest that this cost is quite low and might well be minuscule. Those estimates are based on consumption data for the United States as a whole. The volatility of aggregate consumption, however, is much stronger at the state level. We argue that, because interstate risk sharing is imperfect, much information about actual consumption volatility is lost by averaging consumption figures across all 50 U.S. states. Using state-level consumption data, we show that the welfare cost of macroeconomic volatility is in fact very substantial. Surprisingly, in many states, the welfare gain from eliminating business cycles can exceed the gain from increasing the long-term growth rate by 1% forever. Our results have implications for several key issues in economics and finance.

**Keywords:** Business cycles, Consumption volatility, Growth, Welfare, Regional data

JEL classification: E32; E60

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### 1 Introduction

In an influential paper that has shaped the debate on the importance of business cycles, Robert E. Lucas Jr. evaluated the welfare cost of business cycles in the United States at a minuscule 0.05% of permanent consumption (Lucas, 1987). In contrast, Lucas estimated that the welfare cost of being denied an additional percentage point of consumption growth in perpetuity was very large – several hundred times larger than the cost of consumption volatility. Lucas performed his calculations within a representative agent model calibrated to match key moments of the aggregate per capita U.S. consumption series. The model was very simple, but the results were very important. They suggested a shift, both at the research and the policy levels, from trying to remove business cycles to promoting growth.

Many studies have challenged Lucas's results. Some obtain estimates of the welfare cost of aggregate consumption volatility that can be more than an order of magnitude higher than the Lucas estimates. However, all studies that focus on what constitutes a reasonable model to compute this cost conclude, like Lucas, that it is far below the welfare benefit of an additional percentage point of growth in perpetuity. In Pallage & Robe (forthcoming), we identify a series of developing countries in which this conclusion may be overturned. Using several models and reasonable parameterizations, we find that the welfare gain from removing aggregate consumption volatility can in fact exceed the welfare benefit of an extra 1% of growth forever. Those countries, however, are all characterized by relatively low growth and by high consumption volatility or high shock persistence.

The present paper provides the first evidence that, even in the U.S. and abstracting from any negative impact of volatility on growth [e.g., Barlevy (2001); Epaulard & Pommeret (2002)], the welfare cost of macroeconomic volatility could well approach or exceed the welfare gain from increasing the long-term growth rate by 1%. Extant studies of the welfare cost of business cycles rely on consumption data for the United States as a whole. We argue that, since interstate risk sharing is imperfect,<sup>2</sup> considerable information about actual consumption volatility is likely lost by

<sup>&</sup>lt;sup>1</sup>Imrohoroğlu (1989) takes into account the inability of agents to fully insure against idiosyncratic employment shocks. She reports business-cycle costs only three to five times higher than does Lucas (1987). Atkeson & Phelan (1994), Krusell & Smith (1999), Storesletten, Telmer, & Yaron (2001) and Beaudry & Pages (2001) also find small or moderate costs of aggregate fluctuations despite assuming that markets are incomplete. Obstfeld (1994) and Dolmas (1998) use representative-agent models, but question Lucas's use of isoelastic preferences and his choice of stochastic process governing consumption. Both authors find that the welfare cost of consumption volatility can be significant but is lower than the gain from higher growth. For reasonable preference parameters, the magnitude of the cost remains small at business cycle frequencies when habit formation is explicitly considered [Otrok (2001a, 2001b)].

<sup>&</sup>lt;sup>2</sup>See, e.g., Asdrubali, Sorensen, & Yosha (1996); Hess & Shin (1998); Athanasoulis & van Wincoop (2001); del

aggregating consumption figures across all 50 U.S. states. Using annual state-level consumption data similar to Ostergaard, Sorensen, & Yosha (2002) and del Negro (2002), we document that per capita private consumption indeed fluctuates much more than the corresponding U.S.—wide figures would suggest. Consequently, the welfare cost of business cycles should be much larger than suggested by previous estimates. Our results support this intuition. Using several models, we show that the costs of consumption volatility are sizeable in all 50 individual states. What is more important, we identify many states where the welfare cost of business cycles approaches or exceeds the gain from increased growth.

Of course, business cycles are defined as the common component of economic fluctuations, not the idiosyncratic variation. State data are more volatile than U.S. data; county or city data, more volatile yet; and household data, more volatile still. As we disaggregate, we necessarily get bigger benefits to risk reduction. Why, then, should states provide the relevant level of aggregation?

One reason is that, because policies that affect macroeconomic volatility often are chosen by state authorities (e.g., fiscal policy, financial intermediation laws) or have consequences that vary by state (e.g., monetary policy), it is in fact appropriate to measure the cost of aggregate fluctuations at the state level – as opposed to some other level of aggregation.

The second, more important, reason is that, because the usual assumption of complete markets is not an accurate description of how consumption risk is shared, it is very important to model asymmetric economic shocks. Under complete markets, looking at the loss of a representative U.S. agent would be sufficient: because individual consumption is proportional to aggregate consumption, an aggregate loss due to higher volatility would translate in the same loss for all agents. However, with incomplete markets, this is not so. If much of the variance is borne by residents in some areas, and if these agents cannot effectively insure their risk, then the aggregate U.S. volatility is a not a good predictor of the loss incurred by at least some agents. The problem with business cycles is that the decrease in income is, indeed, borne much more heavily by the regions whose industries suffer the most. This problem is compounded by the fact that, under concave utility functions, agents are much worse off facing a large loss with small probability than a smaller loss with high probability. Combined with incomplete markets, the welfare loss ex-ante of macroeconomic volatility is thus much larger than it would be in models with a representative U.S. agent.

It should be clear that the consumption data we use already reflect all the risk sharing that actually takes place. Undoubtedly, additional risk sharing could be achieved, either across states within the U.S. or with foreign countries, and many authors have tried to measure the welfare

Negro (2002); and references therein.

gains from such opportunities.<sup>3</sup> Our goal is different. Individuals live in one of 50 U.S. states. Those states experience aggregate shocks. We want to measure the welfare cost of the resulting consumption volatility.

Our conclusion is that business cycles truly matter to the residents of many states. We find that (i) business cycles create very large costs in the United States, regardless of the model economy used to compute these costs, and (ii) eliminating business cycles might yield welfare benefits comparable to those from having an extra 1% growth forever. To wit, in nine states, we show that the welfare costs of business cycles can exceed the gains from increased growth even when the costs are the lower bound estimates that we compute with the original Lucas (1987) endowment economy. Our findings therefore have important implications for some key issues in economics, such as fiscal federalism and balanced-budget mandates. They also help reconcile the results of quantitative models of the cost of fluctuations with evidence from survey data that economic agents strongly dislike macroeconomic volatility [diTella, MacCulloch, & Oswald (2001); Wolfers (2002)]. Finally, because both the equity-premium and home-bias puzzles are directly linked to consumption volatility, our conclusion that aggregate volatility is very costly raises new questions for these puzzles.

The remainder of the paper is organized as follows. Section 2 discusses modelling. Section 3 outlines the methods used to compute the welfare cost of economic fluctuations and the welfare benefit from higher growth in the various model economies. Section 4 calibrates these models. Section 5 summarizes the results. Section 6 concludes.

### 2 Model environments

Our focus in this paper is on measuring the costs of consumption volatility. Whatever the reason for that volatility (be it high state-level output volatility, lack of risk sharing through financial markets, imperfections or lack of redistributive federal fiscal policies, etc.), consumption patterns are what economic agents value at the end of the day. Hence, our approach is to measure the costs of business cycles by modelling those agents' consumption streams as "given" and to compute their expected utilities from simulated streams whose moments match those of actual consumption data.

Throughout the paper, we will work within a simple model in which an infinitely-lived representative agent is provided with sequences of consumption generated from a stochastic process calibrated to match key elements of actual U.S. data. By assuming that all idiosyncratic shocks

<sup>&</sup>lt;sup>3</sup>See, e.g., van Wincoop (1994, 1999); Hess & Shin (2000); Crucini & Hess (2000); Athanasoulis & van Wincoop (2001); Athanasoulis & Shiller (2001); Davis, Nalewaik, & Willen (2001); and references cited in those papers.

can be perfectly insured and assuming away possible costs related to distributional issues or to the deleterious effect of aggregate volatility on growth, we ensure that our estimates of the welfare cost of business cycles are conservative.

Our "benchmark" economy will be the Lucas (1987) model in which the representative consumer has isoelastic preferences over consumption:

$$U = E \sum_{t=0}^{\infty} \beta^t \frac{c_t^{1-\gamma} - 1}{1-\gamma} \tag{1}$$

where  $\beta$  is the agent's discount factor,  $\gamma$  is his coefficient of constant relative risk aversion, and  $c_t$  denotes his consumption at time t. In this benchmark economy, consumption streams are generated by the process:

$$\ln c_t = \alpha + t \ln(1+g) - \frac{1}{2}\sigma_z^2 + z_t \quad \text{with } z_t \sim N(0, \sigma_z^2)$$
 (2)

where g is the mean growth rate of real per capita consumption.

Process (2) is trend-stationary, i.e., mean consumption is assumed to follow a deterministic trend. Obstfeld (1994) and Dolmas (1998) consider alternative processes for which the trend component of consumption is itself stochastic:

$$\psi_t = (1 - a)(1 + g) + a\psi_{t-1} + \epsilon_t \quad \text{with } \epsilon_t \rightsquigarrow N(0, \sigma_\epsilon^2)$$
(3)

where  $\psi_t = c_t/c_{t-1}$  is the growth factor and g is the mean growth rate of real per capita consumption. This AR process is the same as that considered in Dolmas (1998) and, when a = 0, is similar to the martingale representation proposed by Obstfeld (1994).

In our second model economy, we will retain the CRRA utility specification (1), but depart from the benchmark by considering process (3) instead of process (2). This change will help identify the importance of the consumption process when computing welfare cost estimates. That is, it will give an idea of the welfare effect of removing all consumption volatility, including the volatility brought about by changes in the growth rate of consumption.

When assessing these welfare effects, two factors are at work. Understanding the agent's attitude towards risk, on the one hand, is key in gauging the welfare cost of volatility. The agent's willingness to substitute over time, on the other hand, weighs heavily on the welfare cost of changes in consumption growth. The CRRA utility posited in models 1 and 2 does not allow us to disentangle these two effects. In a third experiment, we will therefore consider an alternative specification of preferences that allows a decoupling of risk aversion and intertemporal substitution. Specifically,

we will use the utility specification known as the Epstein & Zin (1989) recursion:

$$U_{t} = \left(c_{t}^{1-\theta} + \beta \left[E_{t}(U_{t+1}^{1-\gamma})\right]^{\frac{1-\theta}{1-\theta}}\right)^{\frac{1}{1-\theta}} \tag{4}$$

where  $\frac{1}{\theta}$  is the elasticity of intertemporal substitution. Specification (4) reduces to (1) when  $\theta = \gamma$ .

# 3 Computing the welfare cost of aggregate fluctuations

As is common, we define the cost of business cycles as the percentage consumption increase at all dates and in all states,  $\lambda$ , that would render the representative agent indifferent, given his preferences, between a world of uncertainty [with consumption following (2) or (3)] and one of certainty [i.e., with the same laws of motion but  $\sigma_z^2 = 0$  or  $\sigma_\epsilon^2 = 0$ , respectively]. The welfare benefit of an extra 1% of growth,  $\eta$ , is then measured as the across—the—board percentage consumption increase that would be needed for the same agent to give up a one percentage point increase in the mean growth rate [i.e., g + 1%] if volatility were kept constant.

In the benchmark economy (1)-(2), the welfare cost of consumption volatility,  $\lambda$ , and that of being deprived an extra 1% of growth,  $\eta$ , both have closed-form solutions [Lucas (1987), Obstfeld (1994)]:

$$\lambda = e^{\frac{\gamma \sigma_z^2}{2}} - 1 = \frac{\gamma \sigma_z^2}{2} \tag{5}$$

$$\eta = \left(\frac{1 - \beta e^{g(1-\gamma)}}{1 - \beta e^{(g+1\%)(1-\gamma)}}\right)^{\frac{1}{1-\gamma}} - 1 \tag{6}$$

In contrast, unless shocks to the consumption growth rate have no persistence in process (3)(i.e., a=0), the welfare cost of business cycles in economies 2 and 3 does not have a closed form solution. Note, though, that our second model economy is a subcase of the third when  $\gamma=\theta$ . But in the third economy, we can make the agent's utility recursion more tractable computationally by combining properties of the utility specification (4) and of the stochastic process for consumption (3). Precisely, the Epstein & Zin (1989) utility function is linearly homogeneous and, when consumption growth follows a stationary AR(1) process, can be rewritten as the following Bellman equation:

$$V(c,\psi) = cW(\psi) = c\left\{1 + \beta (E[\psi'W(\psi')|\psi]^{1-\gamma})^{\frac{1-\theta}{1-\gamma}}\right\}^{\frac{1}{1-\theta}}$$

where  $\psi'$  denotes next period's growth factor and follows process (3). We approximate this law of motion by a finite-state Markov chain, using a method proposed by Tauchen (1986) and adapting

code from Ljungqvist & Sargent (2000). After solving for the value function W(.) by iteration, we compute the cost  $\lambda$  as:

$$\lambda = \frac{W_d}{\sum_{\psi} \pi(\psi) W(\psi)} - 1$$

where  $W_d$  is the lifetime utility from deterministic consumption growth and, in the stochastic case, each state  $\psi$  is weighed by the unconditional probability of being at that state,  $\pi(\psi)$ . This averaging ensures that our cost estimate is independent of the initial state.<sup>4</sup>

The computation of the welfare benefit of an additional 1% of yearly consumption growth in economies 2 and 3 is carried out in a similar fashion. The new process for consumption growth can be written as:

$$\psi' = (1-a)(1+q+0.01) + a\psi + \epsilon$$
 with  $\epsilon \rightsquigarrow N(0, \sigma_{\epsilon}^2)$ 

After approximating this modified process by a finite-state Markov chain, we solve for the corresponding value function  $W_g$  and the unconditional probability distribution  $\pi_g(\psi)$ . The welfare benefit of the additional percentage point of growth,  $\eta$ , is then simply:

$$\eta = \frac{\sum_{\psi} \pi_g(\psi) W_g(\psi)}{\sum_{\psi} \pi(\psi) W(\psi)} - 1$$

## 4 Calibration

In order to quantify the welfare cost brought about by macroeconomic fluctuations in every U.S. state, we must parameterize each model economy, solve it numerically, and carry out robustness checks. We focus on the 50 states, for which private consumption data can be constructed, and exclude entities such as Puerto Rico or the District of Columbia.

For each of the 50 U.S. states and for the United States as a whole, we estimate laws of motion (2) and (3) using real per capita private consumption data over the 1960–1995 period. Quarterly state consumption figures are not available, so we rely on annual data. State-level consumption series are constructed from proprietary retail sales data originally published by Sales and Marketing Management (S&MM), using procedures described in del Negro (2002). Those retail sales are only

<sup>&</sup>lt;sup>4</sup>This method is similar in spirit to that used by Dolmas (1998) and, with the same calibration [based on quarterly U.S. consumption data] yields welfare—cost estimates extremely close to those reported by that author. For countries where a = 0 in process (3), a closed—form approximation exists [Obstfeld (1994)] and is consistent with our cost estimates.

a proxy for private consumption but, as Ostergaard et al. (2002) point out, they are the best data available at the state level. We focus on total private consumption, and consider non-durable consumption in robustness tests. (i) For each state and the United States as a whole, we calculate total private consumption for a given year by multiplying the relevant retail sales by the ratio of total U.S. private consumption (Bureau of Economic Analysis, BEA 2002) to overall U.S. retail sales (S&MM) for that year. This re-scaling presents the advantage of adjusting our consumption estimates for the consumption of services not included in the original retail sales series. (ii) For non-durables, we do not re-scale the S&MM non-durable retail sales series in a similar fashion (i.e., by the U.S. ratio of non-durable retail sales to non-durable private consumption). The reason is that, if the sales-to-consumption ratio is not the same for all states, then re-scaling may introduce some extraneous noise into the constructed consumption series. Because the goods-to-services ratio is larger for non-durables than for total consumption, such extra noise would have a larger impact on non-durable consumption estimates.<sup>5</sup> We deflate all these consumption series using the U.S. CPL.<sup>6</sup>

For process (2), we parameterize  $\sigma_z^2$  to the variance of the residuals from regressing log real per capita private consumption on time. In Table 1, we provide the 50+1 regression estimates of  $\sigma_z^2$  and of the mean growth rate g. By computing  $\sigma_z^2$  under the assumption that the shocks in (2) are i.i.d., we ensure the direct comparability of our welfare cost figures with similar calculations in other papers on the cost of business cycles. By abstracting from any impact that volatility in the growth component could have on the representative agent's welfare, parameterization (2) guarantees that cost computations in our first model economy focus purely on the welfare cost of cyclical fluctuations and, hence, yield conservative cost estimates.

With process (3), we calibrate the model for each state and for the United States to match moments of the real per capita private consumption growth series. We obtain the mean growth rate g, the persistence parameter a and the residual variance  $\sigma_{\epsilon}^2$  from a standard AR(1) fit. When the slope coefficient, a, is not statistically significant, we re-estimate the other parameters by regressing the consumption growth rates on a constant. For the United States, California, Illinois and New York as well as Missouri, Rhode Island and Tennessee, an AR(1) process provides the best

<sup>&</sup>lt;sup>5</sup>Ostergaard et al. (2002) use the same proxy for non-durable consumption. Note that S&MM non-durable retail sales, summed up across all states, are a very close substitute for U.S. private consumption expenditures on non-durable goods (to the exclusion of services) reported by the Bureau of Economic Analysis. Between 1960-95, the mean ratio of the two series is 1.02 and the correlation between them is 0.99.

<sup>&</sup>lt;sup>6</sup>Our conclusions are qualitatively robust to using state CPI figures instead. Hence, we only report figures computed with the U.S. CPI.

fit. For Georgia, Montana, and Nebraska, an AR(1) fit cannot be rejected at the 10% significance level. For all the other states, the persistence parameter a is not significantly different from 0. In Table 2, we report the regression estimates.

To calibrate the preference parameters in (1) and (4), we rely on previous estimates. For the United States, the discount factor  $\beta$  typically is set between 0.95 and 0.97 for yearly data. We therefore choose 0.96 as a base value for our computations. Neither the coefficient of relative risk aversion  $\gamma$  nor the elasticity of intertemporal substitution  $\frac{1}{\theta}$  has an accepted standard value. We use the values  $\gamma \in \{1.5, 2, 2.5, 5, 10\}$ , which are within that parameter's recognized range [Mehra & Prescott (1985)]. In the third experiment, we take values for  $\theta \in \{1.5, 2, 2.5, 5\}$ , which is in line with extant papers [e.g., Obstfeld (1994), Dolmas (1998)].

### 5 Results

This section summarizes our main results. First, the estimate of the welfare losses due to macroeconomic fluctuations is much larger when the stochastic processes describing per capita consumption are estimated at the level of individual U.S. states than when the estimates are obtained using aggregate data for the United States as a whole. Depending on the model economy, the welfare cost for the average state is between three and five times the U.S. estimate. In contrast, the benefit from higher growth varies little from state to state and is comparable with previous estimates. Second, our estimates of the welfare cost of consumption volatility are high in absolute terms. On average over the 50 U.S. states, our lower–bound estimate (obtained in the first model economy) already exceeds 0.5% of permanent consumption at the moderate risk–aversion level of  $\gamma = 2.5$ . The median cost figures in the other two model economies are one or two orders of magnitude higher. Third, for reasonable parameter values, the benefit from shutting off all consumption volatility approaches or even exceeds in many states the welfare benefit of an extra 1% of growth in perpetuity – even in the benchmark model.

#### 5.1 First economy: the benchmark

Table 1 gives, for the 50 U.S. states and the United States as a whole, the estimates of consumption volatility,  $\sigma_z$ , and average consumption growth rate, g. The table also reports estimates of the welfare cost of business cycles,  $\lambda$ , and of the welfare gain from increasing the mean growth rate by 1% forever,  $\eta$ . Depending on the posited level of risk aversion,  $\lambda$  ranges from 0.1% ( $\gamma = 1.5$ ) to 0.7% ( $\gamma = 10$ ) of permanent consumption for the United States as a whole. Computations at the

state level, however, yield much larger figures: the median (average) state estimate ranges from 0.3% (0.35%) to 1.9% (2.4%) of the representative agent's permanent consumption.

Those figures might seem pale in comparison with the welfare gain from permanently higher growth, which ranges from about 3.6% ( $\gamma = 10$ ) to over 20% ( $\gamma = 1.5$ ). They are, however, almost six times larger than extant estimates [e.g., Obstfeld (1994)]. More importantly, there are nine states whose residents, if sufficiently risk averse, would strictly prefer to see business cycles eliminated to receiving an extra 1% of growth in perpetuity. These states are Alaska, Connecticut, Hawaii, Mississippi, New Hampshire, Rhode Island, Vermont and Wyoming, as well as Massachusetts. With the exception of the latter, those states have some of the smallest populations in the U.S. Still, this result matters because (i) this is the first such finding; (ii) the result was obtained within a model environment that, by construction, yields conservative volatility cost estimates; and (iii) there are many more states where, at the same risk aversion level, the cost of business cycles does not exceed but nevertheless approaches the gain from higher growth.

To put our figures in perspective, it is worth comparing our cost estimates for the United States as a whole to those obtained in previous studies. The welfare gains from increasing growth reported in Table 1 are directly comparable to existing figures. Our consumption volatility (and, hence, cost) estimates, in contrast, are higher than previously reported. First, they are significantly higher than the numbers in Lucas (1987). This discrepancy comes mostly from the fact that the original analysis assumes process (2) but  $\sigma_z^2$  is calibrated to the variance of the cyclical component of Hodrick-Prescott filtered logarithms of real per capita private consumption. Our volatility figure for the time regression, 3.76\%, is also approximately 1.4 times the corresponding figure for total consumption in Table 1.A of Obstfeld (1994), 2.66%. The difference in time periods used in the two studies (1950-90 vs. 1960-95) explains only a small part of this large discrepancy. The main reason for the difference is that the figures for population and for inflation-adjusted private consumption expenditures reported in the 1991 Economic Report of the President, on which the Obstfeld (1994) study relies, yield a much less volatile per capita consumption series than the corresponding data reported in more recent issues of the same Report (our 1960-1995 data are directly comparable to those reported in the 1995 through 2002 Reports). As a result, we find that the welfare cost of business cycles for the United States as a whole is almost twice that previously estimated. At a risk aversion level  $\gamma = 2$ , for example, column 4 in Table 1 shows that the representative U.S. consumer would be willing to part with 0.14% of consumption forever to eliminate all consumption

<sup>&</sup>lt;sup>7</sup>Such an estimate is by nature smaller than the corresponding estimate obtained by fitting a time trend – see Dolmas (1998) for a similar point, and Lucas (2003) for updated cost figures similar to ours.

fluctuations around the long-term time trend. The comparable figure in Obstfeld (1994) is 0.07%. In contrast, mean growth rates are robust to the data revisions, and we find welfare gains from increasing growth by 1% forever that are similar to those reported in Obstfeld (1994).

#### 5.2 Second and third economies

Table 2 provides, for each state and the United States, the parameters estimated by fitting the autoregressive process (3) to the private consumption data, plus a matrix that gives the welfare cost of consumption volatility  $\lambda$  (left panel) and the gain  $\eta$  from an additional 1% growth forever (right panel) for various combinations of the representative agent's preference parameters,  $\gamma$  and  $\theta$ . The diagonal elements of each matrix ( $\gamma = \theta$ ) are the cost estimates in the CRRA case, i.e., in economy 2.

The median cost figures in these two model economies are one or two orders of magnitude higher than those reported in Table 1. More importantly, at the risk aversion level  $\gamma=10$ , the benefit from shutting off all consumption volatility exceeds the welfare benefit of an extra 1% of growth in perpetuity in 28 states – regardless of the elasticity from intertemporal substitution.<sup>8</sup> Figure 1 provides a geographic survey of these 28 states. They are the nine states identified in Table 1 (Alaska, Connecticut, Hawaii, Massachusetts, Mississippi, New Hampshire, Rhode Island, Vermont and Wyoming), plus Arizona, Colorado, Delaware, Georgia, Idaho, Illinois, Iowa, Maine, Montana, North and South Dakotas, Nebraska, Nevada, New York, Oklahoma, Oregon, Tennessee, Utah and West Virginia. While many of these states have small populations or land mass, together they accounted for 34.1% of the population of the United States in 1995.<sup>9</sup>

<sup>&</sup>lt;sup>8</sup>Using state-level consumption data, Crucini & Hess (2000) argue that better intranational risk sharing can generate sizeable welfare gains. Their result is derived under CRRA preferences and the assumption that consumption follows a random walk, which might seem to make it comparable with our findings for economy 2. However, there are key differences. First, the result depends on a particular specification of the permanent income process that, as the authors point out, could be mis-specified. The main difference, though, is that we compute the welfare gains from removing aggregate consumption volatility at the state level – not just the part of consumption volatility that is not accounted for by fluctuations in U.S. wide aggregate consumption. For that reason, their results are not directly comparable to ours.

<sup>&</sup>lt;sup>9</sup>Excluding the three states (Georgia, Montana, Nebraska) for which an AR(1) fit is only significant at the 10% level still leaves a ratio of 30.4%. Even omitting Massachusetts from the group (because the cost of aggregate fluctuations in that state is merely very close to the benefit of higher growth) would leave us with 28.1% of the U.S. population.

#### 5.3 Discussion

The analysis thus far has focused on private consumption figures, including purchases of durable goods. It is well know, however, that such purchases are quite volatile. We therefore re-ran the analysis using non-durable retail sales. Though volatility cost estimates are somewhat smaller, our main qualitative findings are robust and the thrust of our results does not change. This robustness is consistent with previous studies that suggest that estimates of the welfare cost of business cycles are not drastically reduced by excluding durables from the consumption series.<sup>10</sup>

We rely on data from 1960 to 1995 to show that consumption volatility is high and very costly. A natural question is whether volatility has changed during the sample period. Blanchard & Simon (2001), in particular, argue that U.S. output and consumption volatility have both fallen significantly between 1950 and 2000. We therefore re-ran the analysis with data from 1969 to 1995.<sup>11</sup> The volatility figures are indeed smaller, but not by much. For the United States as a whole, for example, the estimate of  $\sigma_z$  in process (2) falls from 3.76% to 3.22%. The persistence and residual volatility estimates for process (3), likewise, are little changed: shock persistence is a bit weaker (a = 0.4057, vs. 0.4266 in Table 2) but shocks are a bit stronger ( $\sigma_{\epsilon} = 2.26\%$ , vs. 2.07% in Table 2). Comparable results obtain at the state level. Unsurprisingly, then, we find that our main findings are robust to the period change.<sup>12</sup>

Overall, we find that using a U.S. national average to compute the welfare cost of business cycles yields excessively low cost estimates, because the latter are obtained by averaging out a large amount of state—level consumption risk that in fact was not shared. Still, one might question the interpretation of these state-level volatilities by observing that moving is a very important part of interstate risk sharing [e.g., Blanchard & Katz (1992)]. Our analysis admittedly abstracts from that possibility. To the extent that moving costs are non-negligible, however, it is not clear that the cost of business cycles would be significantly lower if such a possibility were integrated into the

<sup>&</sup>lt;sup>10</sup>See, e.g., Obstfeld (1994). Note that, in line with Asdrubali et al. (1996), our robustness analysis proxies non-durable consumption by retail sales of non-durable goods. The latter cannot capture components of consumption such as the service flow from the housing stock. To the extent that consumption of these additional items is much smoother, our robustness checks may still overstate the welfare cost of consumption volatility.

<sup>&</sup>lt;sup>11</sup>This alternative sample preserves enough degrees of freedom for AR(1) and AR(2) regressions, yet covers the very period during which interstate risk sharing should have improved following the growth of U.S. financial markets, increased opportunities for borrowing and lending, and the development of the mortgage-backed securities industry.

<sup>&</sup>lt;sup>12</sup>One possible explanation for not finding much volatility reduction is that neither the initial analysis nor the robustness analysis covers the 1950s, when volatility was particularly high [see, e.g., Figures 9 and 10 in Blanchard & Simon (2001)].

analysis.

# 6 Conclusion and suggestions for further work

Overall, our findings suggest that business cycles truly matter to the residents of many U.S. states. Our results highlight several possible venues for further research. For example, there is broad evidence that almost all industries are somewhat geographically concentrated [e.g., Ellison & Glaeser (1997)]. It will therefore be interesting to ascertain whether differences in the cost of fluctuations across U.S. states are due primarily to different diversification levels of their local economies, to differences in the extent of risk sharing between various states, to some other factor, or to differences in the binding nature of balanced budget requirements between various states.

Indeed, the paper also points to the importance of fiscal policy as a smoothing mechanism. Just as monetary policy, being decided at the federal level, should appropriately look at aggregate figures for the United States as a whole, it suffers from the downside of not being able to differentiate between the economic circumstances of the various states. This "one size fits all" approach may therefore be less effective at smoothing out economic fluctuations than state—level fiscal policies may be. Yet many states operate under balanced—budget fiscal policy rules that can amplify business cycles and bring about instability [e.g., King, Plosser, & Rebelo (1988); Schmitt-Grohe & Uribe (1997)]. This last observation in turn raises the question of the extent to which fiscal policy decisions should be left at the state level vs. how much of a responsibility the federal government should have. Relatedly, it could be interesting to examine in the light of our results some potential costs of the "3% maximum yearly deficit" rule adopted by European Union members that have joined the single—currency euro zone.

Our findings also have important implications for some of the key puzzles in financial economics. First, some recent studies have documented that investors exhibit not only international but also intranational home bias [e.g., Coval & Moskowitz (1999); Huberman (2001); Hess & Shin (2001)]. Such investment patterns are all the more perplexing if costs associated with consumption volatility are high. Second, the equity premium puzzle is directly linked to an apparent lack of consumption volatility. That is, the volatility of U.S. residents' consumption streams has long been thought to be too low to explain the level of equity returns observed in the United States over the last 50 years. There may in fact be less of a puzzle after all because, for the residents of most U.S. states, the volatility of those very consumption streams is in fact much higher than measures for the average U.S. consumer would suggest.

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Table 1: Benchmark economy

State	Regr	ession	Welfar	re cost c	f busines	s cycle	λ (%)	Welfare	e benefit c	of extra 10	Z growth	n (%)
State	g (%)	$\sigma_z$ (%)	$\gamma = 1.5$	$\gamma=2$	$\gamma = 2.5$	$\gamma = 5$	$\gamma = 10$	$\gamma = 1.5$	$\gamma=2$	$\gamma = 2.5$	$\gamma = 5$	$\gamma = 10$
								·				
AK	3.35	15.15	1.74	2.32	2.91	5.90	12.15	17.53	12.89	10.15	4.77	2.15
AL	2.36	7.17	0.39	0.51	0.64	1.29	2.60	19.32	14.96	12.17	6.18	2.95
AR	$\frac{2.13}{1.71}$	$6.11 \\ 6.85$	$0.28 \\ 0.35$	$0.37 \\ 0.47$	$0.47 \\ 0.59$	$0.94 \\ 1.18$	$\frac{1.88}{2.37}$	19.78 $20.68$	$15.53 \\ 16.68$	$12.75 \\ 13.96$	$6.62 \\ 7.58$	$3.21 \\ 3.81$
$egin{array}{l} AZ \ CA \end{array}$	1.71	$\frac{0.83}{4.83}$	0.33 $0.18$	$0.47 \\ 0.23$	0.39 $0.29$	0.58	$\frac{2.37}{1.17}$	20.66 $21.66$	18.02	15.90 $15.41$	8.86	$\frac{3.61}{4.65}$
CO	1.86	4.42	0.15	0.20	0.23 $0.24$	0.38	0.98	20.36	16.02 $16.26$	13.52	7.22	$\frac{4.03}{3.58}$
$^{ m CT}$	2.14	8.53	0.55	0.73	0.91	1.84	3.71	19.76	15.50	12.73	6.60	3.20
DE	1.98	6.76	0.34	0.46	0.57	1.15	2.31	20.09	15.92	13.15	6.93	3.40
$\operatorname{FL}$	1.97	6.06	0.28	0.37	0.46	0.92	1.85	20.10	15.94	13.17	6.94	3.41
GA	2.35	6.74	0.34	0.45	0.57	1.14	2.29	19.34	14.98	12.20	6.19	2.96
HI	3.18	8.79	0.58	0.78	0.97	1.95	3.94	17.82	13.20	10.45	4.97	2.26
IA	1.47	6.84	0.35	0.47	0.59	1.18	2.37	21.23	17.42	14.76	8.27	4.25
ID	1.20	6.90	0.36	0.48	0.60	1.20	2.41	21.89	18.34	15.77	9.20	4.88
IL	1.32	5.93	0.26	0.35	0.44	0.88	1.77	21.59	17.92	15.31	8.76	4.59
IN	1.68	6.14	0.28	0.38	0.47	0.95	1.90	20.75	16.78	14.07	7.67	3.87
KS	1.75	5.26	0.21	0.28	$0.35 \\ 0.32$	0.69	1.39	20.58	16.55	13.83	7.47	3.74
KY LA	$2.41 \\ 2.41$	$5.09 \\ 4.64$	$0.19 \\ 0.16$	$0.26 \\ 0.22$	$0.32 \\ 0.27$	$0.65 \\ 0.54$	$1.30 \\ 1.08$	19.21 $19.21$	14.83 $14.83$	$12.05 \\ 12.05$	$6.08 \\ 6.08$	$\frac{2.89}{2.89}$
MA	$\frac{2.41}{2.07}$	8.22	$0.10 \\ 0.51$	$0.22 \\ 0.68$	0.27 $0.85$	1.70	$\frac{1.08}{3.44}$	19.21	14.63 $15.67$	12.03 $12.90$	6.73	$\frac{2.69}{3.28}$
MD	$\frac{2.07}{2.13}$	5.37	$0.31 \\ 0.22$	0.00	0.36	0.72	1.45	19.78	15.52	12.75	6.61	3.20 $3.21$
ME	$\frac{2.15}{2.57}$	6.52	0.32	0.43	0.53	1.07	2.15	18.91	14.47	11.68	5.82	2.74
MI	1.75	4.72	0.17	0.22	0.28	0.56	1.12	20.59	16.57	13.84	7.48	3.75
MN	2.08	4.24	0.13	0.18	0.22	0.45	0.90	19.89	15.67	12.90	6.73	3.28
MO	1.70	4.73	0.17	0.22	0.28	0.56	1.12	20.69	16.70	13.98	7.60	3.82
MS	2.28	9.51	0.68	0.91	1.14	2.29	4.62	19.48	15.16	12.37	6.33	3.04
MT	1.59	5.40	0.22	0.29	0.37	0.73	1.47	20.96	17.05	14.36	7.92	4.03
NC	2.46	5.41	0.22	0.29	0.37	0.74	1.48	19.12	14.72	11.94	6.01	2.85
ND	2.11	4.94	0.18	0.24	0.31	0.61	1.23	19.82	15.58	12.80	6.66	3.23
NE	1.51	6.65	0.33	0.44	0.55	1.11	2.24	21.15	17.31	14.64	8.16	4.18
NH	2.91	7.64	0.44	0.58	0.73	1.47	2.96	18.29	13.75	10.97	5.32	2.45
NJ	1.97	5.46	0.22	0.30	0.37	0.75	1.50	20.12	15.96	13.20	6.96	3.42
NM NV	$1.95 \\ 1.54$	$\frac{4.88}{7.67}$	$0.18 \\ 0.44$	$0.24 \\ 0.59$	$0.30 \\ 0.74$	$0.60 \\ 1.48$	$\frac{1.20}{2.98}$	$20.14 \\ 21.07$	$15.99 \\ 17.20$	$13.23 \\ 14.52$	$6.99 \\ 8.06$	$\frac{3.44}{4.12}$
NY	1.34 $1.24$	6.05	$0.44 \\ 0.27$	$0.39 \\ 0.37$	$0.74 \\ 0.46$	0.92	$\frac{2.96}{1.85}$	$\frac{21.07}{21.79}$	18.20	14.52 $15.62$	9.05	4.12
OH	1.74	4.99	0.19	0.25	0.40	0.62	1.25	20.61	16.59	13.86	7.50	3.76
OK	1.85	7.47	0.42	0.56	0.70	1.41	2.83	20.38	16.29	13.55	7.24	3.60
OR	1.79	5.10	0.20	0.26	0.33	0.65	1.31	20.50	16.46	13.73	7.39	3.69
PA	1.90	5.11	0.20	0.26	0.33	0.65	1.31	20.26	16.14	13.39	7.12	3.52
RI	1.78	9.09	0.62	0.83	1.04	2.09	4.22	20.53	16.48	13.75	7.41	3.70
SC	2.81	6.57	0.32	0.43	0.54	1.08	2.18	18.46	13.94	11.16	5.46	2.53
$^{\mathrm{SD}}$	1.93	7.07	0.38	0.50	0.63	1.26	2.53	20.19	16.05	13.29	7.04	3.47
TN	2.34	7.21	0.39	0.52	0.65	1.31	2.63	19.34	14.99	12.21	6.20	2.96
TX	1.96	5.86	0.26	0.34	0.43	0.86	1.73	20.14	15.98	13.22	6.98	3.43
UT	1.33	5.84	0.26	0.34	0.43	0.86	1.72	21.57	17.90	15.28	8.74	4.57
VA	2.69	3.74	0.11	0.14	0.18	0.35	0.70	18.68	14.19	11.41	5.63	2.63
VT	2.14	9.43	0.67	0.89	1.12	2.25	4.55	19.76	15.51	12.73	6.60	3.20
WA WI	$1.67 \\ 1.85$	$3.16 \\ 6.50$	$0.08 \\ 0.32$	$0.10 \\ 0.42$	$0.13 \\ 0.53$	$0.25 \\ 1.06$	$0.50 \\ 2.13$	$20.76 \\ 20.36$	$16.79 \\ 16.27$	$14.08 \\ 13.52$	$7.68 \\ 7.22$	$\frac{3.87}{3.58}$
WV	$\frac{1.83}{2.21}$	$\frac{6.50}{7.37}$	$0.32 \\ 0.41$	$0.42 \\ 0.55$	$0.53 \\ 0.68$	1.00 $1.37$	$\frac{2.13}{2.76}$	19.62	15.27 $15.33$	13.52 $12.55$	6.46	3.38 $3.12$
WY	1.49	12.00	1.09	1.45	1.82	3.66	7.46	21.19	17.36	14.69	8.21	4.22
US	1.84	3.76	0.11	0.14	0.18	0.35	0.71	20.38	16.30	13.55	7.25	3.60
Mean	2.00	6.52	0.35	0.47	0.59	1.18	2.38	20.10	15.97	13.23	7.03	3.48
Median	1.96	6.13	0.28	0.38	0.47	0.94	1.89	20.14	15.99	13.23	6.99	3.44
Min	1.20	3.16	0.08	0.10	0.13	0.25	0.50	17.53	12.89	10.15	4.77	2.15
Max	3.35	15.15	1.74	2.32	2.91	5.90	12.15	21.89	18.34	15.77	9.20	4.88

Note: The table compares the welfare cost of aggregate fluctuations (left panel) and the welfare gain from a permanent extra 1% consumption growth (right panel), in economy 1 [CRRA preferences and consumption growth following process (2)] for different values of the risk aversion parameter  $\gamma$ . For each state and the U.S. as a whole, the parameter estimates for process (2) are obtained from a standard linear regression using state-level consumption data from 1960 to 1995.

Table 2: Second and third model economies

A 7.7					Welfa	re cost		ess cycle	$, \lambda (\%)$	Welfar	e gain of		% growth	$\eta$ (%)
AK	0.0343	$egin{matrix} a \ 0 \end{bmatrix}$	$\sigma_{\epsilon}$ $0.0732$	$\theta$	1.5	2	$_{2.5}^{\gamma}$	5	10	1.5	2	$_{2.5}^{\gamma}$	5	10
				1.5	6.72	9.14	11.66	25.95	66.38	17.57	17.83	18.11	19.61	23.54
				2	5.17	7.04	8.99	20.16	53.20	13.17	13.45	13.75	15.46	20.55
				2.5	4.19	5.71	7.30	16.44	44.46	10.49	10.76	11.05	12.73	18.23
				5	2.10	2.87	3.67	8.40	24.42	5.07	5.25	5.44	6.64	11.63
$_{ m AL}$	0.0275	$egin{array}{c} a \ 0 \end{array}$	$\sigma_{\epsilon}$ $0.0407$	θ	1.5	2	$_{2.5}^{\gamma}$	5	10	1.5	2	$_{2.5}^{\gamma}$	5	10
				$\frac{\theta}{1.5}$	2.15	2.89	3.64	7.51	16.06	18.29	18.38	18.47	18.92	19.90
				2	1.72	2.30	2.90	6.00	12.91	14.00	14.09	14.19	14.71	15.86
				2.5	1.42	1.91	2.40	4.98	10.78	11.30	11.40	11.49	12	13.16
				5	0.75	1.01	1.27	2.65	5.81	5.63	5.70	5.76	6.12	6.99
AR	0.0254	$egin{matrix} a \ 0 \end{bmatrix}$	$\sigma_{\epsilon}$ $0.0344$	0	1.5	2	$^{\gamma}_{2.5}$	5	10	1.5	2	$_{2.5}^{\gamma}$	5	10
				heta 1.5	1.57	2.10	2.64	5.39	11.31	18.62	18.69	18.75	19.09	19.79
				2	1.26	1.69	2.12	4.36	9.18	14.39	14.46	14.53	14.92	15.74
				2.5	1.05	1.41	1.78	3.65	7.71	11.69	11.76	11.83	12.21	13.04
				5	0.57	0.76	0.96	1.98	4.22	5.91	5.96	6.01	6.28	6.91
AZ	0.0191	$egin{matrix} a \ 0 \end{bmatrix}$	$\sigma_{\epsilon}$ $0.0526$	0	1.5	2	$^{\gamma}_{2.5}$	5	10	1.5	2	$^{\gamma}_{2.5}$	5	10
				heta 1.5	4.02	5.43	6.86	14.58	33.25	20.29	20.47	20.65	21.61	23.83
				2	3.38	$\frac{3.43}{4.57}$	5.79	12.42	29.00	16.41	16.63	16.86	18.10	23.83 $21.21$
				2.5	2.92	3.95	5.00	10.80	25.73	13.76	13.99	14.23	15.56	19.12
				5	1.71	2.32	2.95	6.49	16.51	7.50	7.69	7.89	9.06	12.81
CA	0.0119	$a \ 0.3148$	$\sigma_{\epsilon}$ $0.0290$		1.5	2	$^{\gamma}_{2.5}$	5	10	1.5	2	$^{\gamma}_{2.5}$	5	10
				$\frac{\theta}{1.5}$	1.86	2.77	3.70	8.57	19.62	21.81	21.95	22.08	22.76	24.27
				2	1.65	$\frac{2.11}{2.46}$	3.29	7.67	17.90	18.45	18.63	18.81	19.77	24.21 $22.01$
				2.5	1.48	2.21	2.96	6.93	16.46	15.98	16.18	16.38	17.47	20.14
				5	0.99	1.46	1.95	4.65	11.73	9.50	9.69	9.89	11.00	14.20
СО	g 0.0000	a	$\sigma_{\epsilon}$			0	$\gamma$	-	10	1 5	0	$\gamma$	-	1.0
	0.0202	0	0.0457	$\theta$	1.5	2	2.5	5	10	1.5	2	2.5	5	10
				1.5	2.96	3.99	5.03	10.51	23.07	19.91	20.04	20.17	20.85	22.37
				2	2.47	3.33	4.20	8.83	19.68	15.94	16.10	16.26	17.11	19.11
				$\frac{2.5}{5}$	$\frac{2.12}{1.22}$	$\frac{2.85}{1.64}$	$\frac{3.60}{2.08}$	$7.61 \\ 4.45$	$17.15 \\ 10.38$	$13.26 \\ 7.10$	$13.42 \\ 7.23$	$13.59 \\ 7.36$	$14.48 \\ 8.10$	$16.67 \\ 10.12$
CT	g	a	$\sigma_\epsilon$	9	1.22	1.04	$\frac{2.08}{\gamma}$	4.40	10.36	7.10	1.23	$\gamma$	0.10	10.12
01	0.0207	0	0.0563	θ	1.5	2	2.5	5	10	1.5	2	$2.5^{\prime}$	5	10
				1.5	4.54	6.14	7.78	16.66	38.74	20.00	20.20	20.40	21.48	24.03
				2	3.78	5.12	6.49	14.05	33.57	16.04	16.28	16.53	17.91	21.49
				2.5	3.23	4.38	5.57	12.13	29.65	13.36	13.61	13.88	15.35	19.44
DE	_	_	=	5	1.85	2.52	3.21	7.15	18.83	7.18	7.38	7.59	8.86	13.19
DE	0.0200	$egin{matrix} a \ 0 \end{bmatrix}$	$\sigma_{\epsilon}$ $0.0526$	θ	1.5	2	$_{2.5}^{\gamma}$	5	10	1.5	2	$\gamma \ 2.5$	5	10
				1.5	3.98	5.37	6.79	14.41	32.80	20.09	20.27	20.44	21.38	23.55
				2	3.33	4.49	5.69	12.19	28.38	16.16	16.37	16.59	17.79	20.79
				2.5	2.85	3.86	4.89	10.55	25.03	13.49	13.71	13.94	15.22	18.61
T)*				5	1.65	2.24	2.85	6.25	15.74	7.28	7.46	7.65	8.75	12.21
$\operatorname{FL}$	0.0206	$egin{array}{c} a \ 0 \end{array}$	$\sigma_{\epsilon}$ $0.0412$	$\theta$	1.5	2	$_{2.5}^{\gamma}$	5	10	1.5	2	$_{2.5}^{\gamma}$	5	10
				$\frac{\theta}{1.5}$	2.39	3.21	4.04	8.37	18.02	19.75	19.86	19.96	20.50	21.67
				2	1.99	2.67	3.36	7.00	15.24	15.74	15.87	15.99	16.66	18.17
				$2.\overline{5}$	1.70	2.28	2.88	6.01	13.18	13.06	13.19	13.32	14.01	15.62
				5	0.97	1.31	1.65	3.48	7.81	6.94	7.04	7.14	7.70	9.11

Note: See end of Table 2.

Table 2 cont.

					Welfa	re cost	of busi	ness cycle	e, $\lambda$ (%)	Welfar	e gain of	extra 1	% growth	ι, η (%)
GA	0.0269	$a \ 0.2658$	$\frac{\sigma_{\epsilon}}{0.0349}$	$\theta$	1.5	2	$_{2.5}^{\gamma}$	5	10	1.5	2	$\gamma \ 2.5$	5	10
				1.5	2.08	3.05	4.04	9.23	21.13	18.39	18.51	18.63	19.26	20.65
				2	1.66	2.43	3.21	7.36	17.02	14.11	14.25	14.38	15.10	16.79
				2.5	1.38	2.01	2.66	6.10	14.20	11.42	11.55	11.68	12.39	14.12
***				5	0.74	1.06	1.40	3.19	7.55	5.71	5.80	5.89	6.41	7.74
HI	0.0319	$egin{array}{c} a \ 0 \end{array}$	$\sigma_{\epsilon} \ 0.0550$	θ	1.5	2	$_{2.5}^{\gamma}$	5	10	1.5	2	$\gamma \ 2.5$	5	10
				1.5	3.79	5.11	6.46	13.70	31.04	17.66	17.81	17.96	18.76	20.61
				2	2.95	3.98	5.03	10.71	24.55	13.28	13.44	13.60	14.50	16.68
				2.5	2.41	3.25	4.11	8.77	20.27	10.60	10.75	10.91	11.78	13.99
- 4				5	1.23	1.66	2.10	4.51	10.67	5.15	5.25	5.35	5.95	7.64
IA	0.0198	$egin{array}{c} a \ 0 \end{array}$	$\sigma_{\epsilon}$ $0.0461$	θ	1.5	2	$_{2.5}^{\gamma}$	5	10	1.5	2	$_{2.5}^{\gamma}$	5	10
				$\frac{\theta}{1.5}$	3.04	4.09	5.15	10.78	23.73	20.00	20.13	20.26	20.97	22.53
				2	2.54	3.42	4.32	9.09	20.32	16.04	16.21	16.37	17.26	19.34
				$2.\overline{5}$	2.18	2.94	3.71	7.85	17.75	13.37	13.54	13.71	14.64	16.93
				5	1.26	1.70	2.15	4.61	10.83	7.19	7.32	7.46	8.24	10.38
ID	0.0194	$egin{matrix} a \ 0 \end{bmatrix}$	$\sigma_{\epsilon} \ 0.0540$	0	1.5	2	$^{\gamma}_{2.5}$	5	10	1.5	2	$_{2.5}^{\gamma}$	5	10
				heta 1.5	4.23	5.71	7.22	15.39	35.36	20.24	20.43	20.62	21.63	23.98
				2	3.55	4.80	6.08	13.08	30.84	16.35	16.58	16.82	18.12	21.44
				$2.5^{-2}$	3.05	4.13	5.25	11.37	27.37	13.69	13.93	14.18	15.58	19.39
				5	1.78	2.42	3.08	6.81	17.59	7.44	7.64	7.85	9.08	13.14
IL	$g\\0.0159$	$a \ 0.3456$	$\sigma_{\epsilon}$ $0.0342$	0	1.5	2	$^{\gamma}_{2.5}$	5	10	1.5	2	$^{\gamma}_{2.5}$	5	10
				$\frac{\theta}{1.5}$	2.61	3.94	5.31	12.65	30.38	20.87	21.06	21.24	22.23	24.51
				2	$\frac{2.01}{2.24}$	3.39	4.57	10.98	27.12	17.17	17.41	17.65	18.97	22.31
				$2.5^{-2}$	1.96	2.97	4.00	9.68	24.51	14.57	14.82	15.08	16.54	20.49
				5	1.22	1.82	2.45	6.01	16.58	8.20	8.42	8.65	10.01	14.59
IN	$g\\0.0211$	$egin{matrix} a \ 0 \end{matrix}$	$\sigma_{\epsilon} \ 0.0402$	0	1.5	2	$^{\gamma}_{2.5}$	5	10	1.5	2	$^{\gamma}_{2.5}$	5	10
				$\frac{\theta}{1.5}$	2.26	3.03	3.81	7.89	16.92	19.62	19.72	19.82	20.32	21.42
				2	1.87	$\frac{3.03}{2.51}$	3.17	6.57	14.23	15.58	15.72	15.82	16.44	17.83
				$2.5^{-2}$	1.60	2.14	2.70	5.62	12.26	12.90	13.01	13.13	13.77	15.25
				5	0.91	1.22	1.54	3.23	7.19	6.81	6.90	7.00	7.51	8.77
KS	0.0206	$egin{matrix} a \ 0 \end{bmatrix}$	$\sigma_{\epsilon}$ $0.0387$		1.5	2	$_{2.5}^{\gamma}$	5	10	1.5	2	$_{2.5}^{\gamma}$	5	10
				$\theta$	2.10	2.82	3.55	7.32	15.63	19.71	19.80	19.89	20.36	21.38
				$\frac{1.5}{2}$	$\frac{2.10}{1.75}$	$\frac{2.82}{2.35}$	$\frac{3.33}{2.95}$	6.12	13.03 $13.17$	15.69	15.80 $15.80$	15.89	16.49	17.78
				$2.5^{-2}$	1.49	2.01	2.53	5.25	11.37	13.01	13.12	13.23	13.83	15.20
				5	0.85	1.15	1.45	3.03	6.69	6.90	6.99	7.08	7.56	8.73
KY	$g\\0.0271$	$egin{array}{c} a \ 0 \end{array}$	$\sigma_{\epsilon}$ $0.0337$	_	1.5	2	$^{\gamma}_{2.5}$	5	10	1.5	2	$\gamma \ 2.5$	5	10
				$\frac{\theta}{1.5}$	1 47	1 07	9.40	5.06	10.58	18.28	18.34	10 40	19 70	19.35
				$\frac{1.5}{2}$	$\frac{1.47}{1.17}$	$\frac{1.97}{1.57}$	$\frac{2.48}{1.98}$	$\frac{5.06}{4.04}$	8.49	$18.28 \\ 13.98$	18.34 $14.05$	$18.40 \\ 14.12$	$18.70 \\ 14.46$	19.35 $15.21$
				$2.5^{2}$	0.97	1.31	1.64	3.36	7.08	11.29	11.36	11.42	11.76	12.5
				5	0.52	0.69	0.87	1.79	3.80	5.62	5.67	5.71	5.95	6.49
LA	$g\\0.0254$	$egin{matrix} a \ 0 \end{bmatrix}$	$\sigma_{\epsilon}$ $0.0315$		1.5	2	$\gamma \ 2.5$	5	10	1.5	2	$^{\gamma}_{2.5}$	5	10
				$\theta$	1.01	1 75	0.00	4 40	0.01	10.00	10.00	10.71	10.00	10 50
				1.5	1.31	1.75	2.20	4.48	9.31	18.60	18.66	18.71	18.98	19.56
				$\frac{2}{2.5}$	$\frac{1.05}{0.88}$	$\frac{1.41}{1.18}$	$1.77 \\ 1.48$	$\frac{3.61}{3.02}$	$7.55 \\ 6.33$	$14.36 \\ 11.66$	$14.42 \\ 11.72$	14.48 $11.78$	$14.80 \\ 12.09$	$15.47 \\ 12.77$
				∠.5 5	$0.88 \\ 0.47$	0.64	0.80	$\frac{3.02}{1.64}$	3.46	5.89	5.93	5.97	6.20	6.70
					0.11	5.01	5.00	1.01	5.10	0.00	5.00	5.01	0.20	0.10

Table 2 cont.

$\eta$ (%)	% growth		e gain of	Welfare	, λ (%)	ness cycle		re cost	Welfa					7 A A
10	5	$_{2.5}^{\gamma}$	2	1.5	10	5	$_{2.5}^{\gamma}$	2	1.5	θ	$\sigma_{\epsilon}$ $0.0447$	$egin{array}{c} a \ 0 \end{array}$	0.0178	MA
22.92	21.40	20.71	20.58	20.45	22.70	10.35	4.96	3.93	2.92	1.5				
19.91	17.83	16.95	16.78	16.62	19.73	8.86	4.21	3.34	2.48	2				
17.60	15.27	14.32	14.15	13.98	17.44	7.73	3.66	2.90	2.15	$2.5^{-2}$				
11.10	8.81	7.98	7.83	7.69	11.01	4.68	2.19	1.73	1.28	5				
10	5	$\gamma \ 2.5$	2	1.5	10	5	$_{2.5}^{\gamma}$	2	1.5		$\sigma_{\epsilon}$ 0.0291	$a \\ 0$	$g\\0.0215$	MD
10	3	2.0	2	1.5	10	3	2.0	2	1.0	$\theta$	0.0291	U	0.0215	
20.28	19.76	19.50	19.45	19.40	8.24	3.98	1.96	1.56	1.17	1.5				
16.38	15.74	15.44	15.38	15.32	6.85	3.29	1.62	1.29	0.96	2				
13.72	13.05	12.75	12.69	12.63	5.85	2.81	1.37	1.10	0.82	2.5				
7.46	6.93	6.70	6.65	6.61	3.34	1.59	0.78	0.62	0.46	5	<u>σ</u>			ME
10	5	$\gamma \ 2.5$	2	1.5	10	5	$_{2.5}^{\gamma}$	2	1.5	θ	$\sigma_{\epsilon}$ $0.0557$	$egin{array}{c} a \ 0 \end{array}$	0.0238	10112
22.94	20.64	19.65	19.47	19.28	36.00	15.62	7.32	5.78	4.28	1.5				
19.85	16.80	15.59	15.37	15.16	30.28	12.86	5.98	4.72	3.49	$^{2}$				
17.49	14.14	12.90	12.67	12.46	26.14	10.92	5.05	3.98	2.94	2.5				
10.93	7.80	6.80	6.63	6.47	15.48	6.14	2.80	2.20	1.62	5				
10	5	$\gamma \ 2.5$	2	1.5	10	5	$_{2.5}^{\gamma}$	2	1.5		$\sigma_{\epsilon}$ $0.0408$	$egin{array}{c} a \ 0 \end{array}$	0.0207	MI
01.60	00.46	10.02	10.02	10.79	1769	0.90	2.06	0.15	0.24	$\theta$				
21.60 $18.07$	$20.46 \\ 16.60$	19.93 $15.95$	19.83 $15.83$	$19.72 \\ 15.71$	17.63 $14.88$	$8.20 \\ 6.85$	$\frac{3.96}{3.29}$	$\frac{3.15}{2.62}$	$\frac{2.34}{1.95}$	1.5				
15.52	13.95	13.93 $13.28$	13.03 $13.15$	13.71 $13.02$	12.87	5.88	$\frac{3.29}{2.82}$	2.02 $2.23$	1.66	2.5				
9.02	7.65	7.11	7.01	6.91	7.60	3.39	1.61	1.28	0.95	5				
10	5	$\gamma \ 2.5$	2	1.5	10	5	$\gamma \ 2.5$	2	1.5	ŭ	$\sigma_{\epsilon}$ 0.0366	$a \\ 0$	$g\\0.0212$	MN
										$\theta$				
20.99	20.11	19.70	19.62	19.54	13.70	6.47	3.15	2.5	1.87	1.5				
17.28	16.17	15.67	15.57	15.48	11.47	5.38	2.61	2.07	1.55	2				
14.66	13.50	12.98	12.89	12.79	9.85	4.59	2.22	1.77	1.32	2.5				
8.25	7.29	6.88	6.80	6.73	5.71	2.62	1.26	1.00	0.74	5	_			МО
10	5	$_{2.5}^{\gamma}$	2	1.5	10	5	$_{2.5}^{\gamma}$	2	1.5	$\theta$	$\sigma_{\epsilon}$ 0.0378	$egin{array}{c} a \ 0 \end{array}$	0.0212	WIO
21.12	20.18	19.74	19.65	19.56	14.66	6.90	3.35	2.66	1.99	1.5				
17.44	16.25	15.72	15.61	15.51	12.29	5.73	2.78	2.21	1.64	2				
14.84	13.58	13.03	12.93	12.82	10.56	4.90	2.37	1.88	1.40	$2.5^{-}$				
8.40	7.35	6.92	6.84	6.76	6.15	2.80	1.34	1.07	0.79	5				
1.0	-	$\gamma$		1 5	1.0	_	$\gamma$	0	1 5		$\sigma_{\epsilon}$	a	<i>g</i>	MS
10	5	2.5	2	1.5	10	5	2.5	2	1.5	$\theta$	0.0397	0.2816	0.0300	
20.91	19.02	18.20	18.04	17.89	29.14	12.29	5.27	3.96	2.68	1.5				
17.08	14.80	13.87	13.70	13.54	23.15	9.64	4.12	3.10	2.10	$^{2}$				
14.41	12.08	11.18	11.01	10.85	19.12	7.89	3.37	2.54	1.72	2.5				
7.96	6.16	5.53	5.42	5.32	9.92	4.00	1.72	1.30	0.90	5				
10	5	$_{2.5}^{\gamma}$	2	1.5	10	5	$_{2.5}^{\gamma}$	2	1.5	^	$\sigma_{\epsilon}$ $0.0553$	a - 0.2800	0.0194	МТ
22.48	21.10	20.47	20.35	20.23	22.79	11.18	6.06	5.08	4.12	heta 1.5				
19.27	17.43	16.63	16.48	$\frac{20.23}{16.33}$	22.79 19.60	9.48	5.11	$\frac{3.08}{4.28}$	$\frac{4.12}{3.46}$	$\frac{1.5}{2}$				
16.86	14.83	13.98	13.82	13.67	17.19	8.23	4.41	$\frac{4.28}{3.69}$	2.99	$\frac{2}{2.5}$				
10.32	8.41	7.69	7.56	7.43	10.67	4.95	2.62	2.19	1.77	2.5 5				
-0.02	U.11	$\gamma$			-5.51	1.00	$\gamma$	0		9	$\sigma_\epsilon$	a	g	NC
10	5	2.5	2	1.5	10	5	2.5	2	1.5	$\theta$	0.0348	0	0.0273	
19.39	18.70	18.38	18.31	18.25	11.30	5.39	2.63	2.10	1.56	1.5				
15.25	14.46	14.09	14.02	13.95	9.07	4.30	2.10	1.67	1.25	2				
12.54	11.75	11.40	11.33	11.26	7.55	3.57	1.74	1.39	1.03	2.5				
6.52	5.95	5.70	5.65	5.60	4.05	1.90	0.92	0.73	0.55	5				

Table 2 cont.

$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	5 20.32 21. 16.42 18. 13.75 15. 7.48 9. 5 21.75 25. 18.26 23. 15.72 21. 9.18 16. 5 19.85 21. 15.81 18. 13.11 15. 6.96 8. 5 20.65 21. 16.86 18. 14.23 15. 7.89 9.	20.32 16.42 13.75 7.48 5 21.75 18.26 15.72 9.18 5 19.85 15.81 13.11 6.96
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	16.42 18. 13.75 15. 7.48 9. 5 21.75 25. 18.26 23. 15.72 21. 9.18 16. 5 19.85 21. 15.81 18. 13.11 15. 6.96 8. 5 20.65 21. 16.86 18. 14.23 15.	16.42 13.75 7.48 5 21.75 18.26 15.72 9.18 5 19.85 15.81 13.11 6.96
NE   $\frac{1}{2}$   $\frac{1}{2}$	16.42 18. 13.75 15. 7.48 9. 5 21.75 25. 18.26 23. 15.72 21. 9.18 16. 5 19.85 21. 15.81 18. 13.11 15. 6.96 8. 5 20.65 21. 16.86 18. 14.23 15.	16.42 13.75 7.48 5 21.75 18.26 15.72 9.18 5 19.85 15.81 13.11 6.96
NE $\begin{array}{c c c c c c c c c c c c c c c c c c c $	13.75 15. 7.48 9. 5 21.75 25. 18.26 23. 15.72 21. 9.18 16. 5 19.85 21. 15.81 18. 13.11 15. 6.96 8. 5 20.65 21. 16.86 18. 14.23 15.	13.75 7.48 5 21.75 18.26 15.72 9.18 5 19.85 15.81 13.11 6.96
NE $\begin{array}{c c c c c c c c c c c c c c c c c c c $	7.48 9.  5 21.75 25. 18.26 23. 15.72 21. 9.18 16.  5 19.85 21. 15.81 18. 13.11 15. 6.96 8.  5 20.65 21. 16.86 18. 14.23 15.	5 21.75 18.26 15.72 9.18 5 19.85 15.81 13.11 6.96
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	21.75 25. 18.26 23. 15.72 21. 9.18 16. 5 19.85 21. 15.81 18. 13.11 15. 6.96 8. 5 20.65 21. 16.86 18. 14.23 15.	21.75 18.26 15.72 9.18 5 19.85 15.81 13.11 6.96
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	18.26 23. 15.72 21. 9.18 16. 5 19.85 21. 15.81 18. 13.11 15. 6.96 8. 5 20.65 21. 16.86 18. 14.23 15.	18.26 15.72 9.18 5 19.85 15.81 13.11 6.96
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	18.26 23. 15.72 21. 9.18 16. 5 19.85 21. 15.81 18. 13.11 15. 6.96 8. 5 20.65 21. 16.86 18. 14.23 15.	18.26 15.72 9.18 5 19.85 15.81 13.11 6.96
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	15.72 21. 9.18 16. 5 19.85 21. 15.81 18. 13.11 15. 6.96 8. 5 20.65 21. 16.86 18. 14.23 15.	15.72 9.18 5 19.85 15.81 13.11 6.96
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	9.18 16. 5 19.85 21. 15.81 18. 13.11 15. 6.96 8. 5 20.65 21. 16.86 18. 14.23 15.	9.18 5 19.85 15.81 13.11 6.96
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	5 19.85 21. 15.81 18. 13.11 15. 6.96 8. 5 20.65 21. 16.86 18. 14.23 15.	5 19.85 15.81 13.11 6.96
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	19.85 21. 15.81 18. 13.11 15. 6.96 8. 5 20.65 21. 16.86 18. 14.23 15.	19.85 15.81 13.11 6.96
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	15.81 18. 13.11 15. 6.96 8. 5 20.65 21. 16.86 18. 14.23 15.	15.81 13.11 6.96
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	13.11 15. 6.96 8. 5 20.65 21. 16.86 18. 14.23 15.	13.11 6.96 5
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	6.96 8.  5  20.65 21. 16.86 18. 14.23 15.	6.96 5
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	5 20.65 21. 16.86 18. 14.23 15.	5
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	20.65 21. 16.86 18. 14.23 15.	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	16.86 18. 14.23 15.	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	14.23 15.	20.65
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$		
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	7.89 9.	14.23
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$		7.89
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	5	5
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	00.00 01	20.20
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	20.28 21. 16.37 17.	
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	13.71 15.	
NV $g$ $a$ $\sigma_{\epsilon}$ $\gamma$	7.45 8.	
heta	5	5
$1.5  5.84  7.92  10.07  22.01  53.85 \qquad 20.73  21.00  21.27$	22.74 26.	22.74
2  4.95  6.73  8.58  19.05  48.87  16.97  17.31  17.66	19.65   25.	
2.5 4.30 5.85 7.47 16.79 44.97 14.35 14.71 15.08	17.29 24.	
$5  2.56  3.50  4.50  10.49  33.53 \qquad 8.00  8.31  8.64$ NY $q  a  \sigma_\epsilon \qquad \qquad \gamma \qquad \qquad \gamma$	10.77 21.	10.77
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	5	5
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	23.66 26.	23.66
2 1.90 3.13 4.39 11.36 29.55 18.86 19.15 19.45	21.11 25.	
2.5 1.74 2.84 3.98 10.39 28.13 16.44 16.77 17.11	19.05 24.	
$5  1.27  1.99  2.74  7.24  23.52 \qquad 9.97  10.29  10.63$	12.79 21.	
OH $g$ $a$ $\sigma_{\epsilon}$ $\gamma$ $\gamma$	-	_
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	5	5
1.5  1.65  2.22  2.78  5.70  11.99  19.62  19.70  19.77	20.13 20.	20.13
2 1.37 1.84 2.31 4.75 10.06 15.59 15.67 15.76	16.20 17.	
2.5  1.17  1.57  1.97  4.07  8.65  12.90  12.99  13.08	13.53 14.	
5  0.67  0.90  1.13  2.33  5.03  6.82  6.89  6.95	7.32 8.	
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	5	5
heta		
$1.5  3.17  4.27  5.38  11.29  24.97 \qquad 19.89  20.04  20.17$	20.91 22.	
2 2.64 3.56 4.50 9.49 21.31 15.92 16.08 16.26	17.17 19.	
2.5  2.26  3.05  3.85  8.17  18.59  13.24  13.41  13.59	4 4 2 2 4 7	14.55
5 1.30 1.75 2.22 4.77 11.28 7.08 7.22 7.36	14.55 16. 8.16 10.	

Table 2 cont.

					Welf	are cost		ess cycle	, λ (%)	Welfar	e gain of		% growtl	h, η (%)
OR	0.0207	$a \\ 0$	$\sigma_{\epsilon}$ $0.0471$	θ	1.5	2	$_{2.5}^{\gamma}$	5	10	1.5	2	$_{2.5}^{\gamma}$	5	10
				1.5	3.14	4.23	5.34	11.18	24.72	19.82	19.96	20.09	20.82	22.43
				2	2.61	3.52	4.45	9.37	21.04	15.82	15.99	16.16	17.06	19.19
				2.5	2.23	3.01	3.80	8.06	18.30	13.14	13.31	13.49	14.43	16.76
PA	a	a	σ	5	1.28	1.72	2.18	4.68	11.04	7.00	7.14	7.28	8.05	10.21
IA	0.0200	0	$\sigma_{\epsilon}$ $0.0347$	θ	1.5	2	$_{2.5}^{\gamma}$	5	10	1.5	2	$^{\gamma}_{2.5}$	5	10
				1.5	1.70	2.27	2.85	5.85	12.32	19.79	19.86	19.93	20.31	21.12
				2	1.41	1.90	2.38	4.90	10.39	15.79	15.88	15.97	16.43	17.44
				$\frac{2.5}{5}$	$\frac{1.21}{0.70}$	$\frac{1.62}{0.93}$	$\frac{2.04}{1.18}$	$\frac{4.21}{2.44}$	$8.98 \\ 5.28$	$13.11 \\ 6.98$	$13.20 \\ 7.05$	$13.29 \\ 7.13$	$13.77 \\ 7.51$	$14.84 \\ 8.42$
RI	g	a	$\sigma_\epsilon$	3	0.70	0.93	$\gamma$	2.44	0.28	0.98	7.00	$\gamma$ .13	7.31	8.42
101	0.0153	0.3942	0.0529	$\theta$	1.5	2	$2.5^{\prime}$	5	10	1.5	2	$2.\overline{5}$	5	10
				1.5	7.16	11.28	15.68	43.18	151.06	21.66	22.22	22.82	26.33	38.10
				2	6.24	9.85	13.77	39.79	186.32	18.21	18.97	19.80	25.30	56.74
				2.5	5.52	8.74	12.27	37.06	305.77	15.70	16.54	17.48	24.40	118.12
$\mathbf{SC}$	g	a	$\sigma_\epsilon$	5	3.56	5.60	$7.94$ $\gamma$	28.70	-	9.22	10.03	$10.99$ $\gamma$	21.18	-
	0.0306	0	0.0376	θ	1.5	2	$2.5^{\prime}$	5	10	1.5	2	2.5	5	10
				1.5	1.77	2.37	2.98	6.11	12.91	17.66	17.73	17.80	18.15	18.92
				2	1.38	1.85	2.33	4.79	10.16	13.28	13.36	13.43	13.82	14.68
				2.5	1.13	1.52	1.91	3.93	8.36	10.61	10.68	10.75	11.13	11.96
$_{ m SD}$	a	a	σ	5	0.58	0.78	0.98	2.03	4.35	5.15	5.20	5.25	5.50	6.09
БD	0.0237	0	$\sigma_{\epsilon}$ $0.0540$	θ	1.5	2	$_{2.5}^{\gamma}$	5	10	1.5	2	$^{\gamma}_{2.5}$	5	10
				1.5	4.02	5.42	6.86	14.57	33.25	19.28	19.45	19.62	20.54	22.67
				2	3.28	4.43	5.61	12	27.92	15.15	15.36	15.56	16.68	19.48
				2.5	2.76	3.73	4.73	10.18	24.05	12.46	12.66	12.87	14.02	17.06
TN	g	a	<b>a</b>	5	1.52	2.07	2.62	5.73	14.16	6.47	6.62	6.78	7.70	10.49
111	0.0286	0.4060	$\sigma_{\epsilon}$ 0.0371	θ	1.5	2	$_{2.5}^{\gamma}$	5	10	1.5	2	$_{2.5}^{\gamma}$	5	10
				1.5	2.97	4.60	6.28	15.47	38.83	18.18	18.38	18.59	19.69	22.36
				2	2.35	3.62	4.94	12.20	31.34	13.87	14.09	14.32	15.60	18.99
				2.5	1.94	2.97	4.05	10.02	26.19	11.17	11.39	11.61	12.88	16.48
TX	a	a	σ	5	1.05	1.56	2.09	5.09	13.93	5.54	5.68	5.84	6.74	9.79
1 //	0.0197	0	0.0238	θ	1.5	2	$_{2.5}^{\gamma}$	5	10	1.5	2	$_{2.5}^{\gamma}$	5	10
				1.5	0.79	1.06	1.33	2.69	5.51	19.75	19.79	19.82	19.99	20.35
				2	0.66	0.89	1.11	2.25	4.63	15.75	15.79	15.83	16.04	16.48
				2.5	0.57	0.76	0.95	1.93	3.98	13.07	13.11	13.16	13.37	13.83
UT	_	_		5	0.33	0.44	0.55	1.12	2.32	6.95	6.99	7.02	7.19	7.56
01	0.0174	$a \\ 0$	$\sigma_{\epsilon}$ $0.0456$	θ	1.5	2	$\gamma \ 2.5$	5	10	1.5	2	$\gamma \ 2.5$	5	10
				1.5	3.06	4.12	5.19	10.86	23.92	20.56	20.70	20.84	21.56	23.18
				2	2.60	3.51	4.43	9.33	20.90	16.77	16.94	17.12	18.06	20.29
				2.5	2.26	3.05	3.85	8.17	18.56	14.14	14.32	14.51	15.52	18.03
3.7.A				5	1.35	1.83	2.32	4.99	11.87	7.82	7.98	8.14	9.04	11.58
VA	0.0286	$a \\ 0$	$\sigma_{\epsilon}$ $0.0347$	θ	1.5	2	$^{\gamma}_{2.5}$	5	10	1.5	2	$_{2.5}^{\gamma}$	5	10
				$\frac{\theta}{1.5}$	1.53	2.05	2.58	5.28	11.06	18.00	18.06	18.12	18.44	19.10
				2	1.21	1.63	2.04	4.18	8.80	13.67	13.74	13.81	14.16	14.91
				$2.\overline{5}$	1.00	1.34	1.68	3.46	7.29	10.98	11.05	11.11	11.45	12.20
				5	0.52	0.70	0.88	1.81	3.86	5.41	5.45	5.50	5.73	6.27

Table 2 cont.

					Welfa	are cost		ss cycle,	$\lambda$ (%)	Welfar	e gain of		% growth	n, η (%)
VT	0.0205	$egin{matrix} a \ 0 \end{bmatrix}$	$\sigma_{\epsilon} \ 0.0587$	0	1.5	2	$^{\gamma}_{2.5}$	5	10	1.5	2	$_{2.5}^{\gamma}$	5	10
				$\frac{\theta}{1.5}$	4.97	6.72	8.53	18.39	43.48	20.10	20.32	20.54	21.74	24.62
				2	4.15	5.62	7.14	15.57	38.00	16.16	16.43	16.71	18.26	22.39
				2.5	3.55	4.82	6.13	13.48	33.82	13.49	13.77	14.06	15.72	20.54
				5	2.04	2.78	3.55	8.01	22.05	7.28	7.51	7.75	9.21	14.62
WA	g	a	$\sigma_\epsilon$				$\gamma$					$\gamma$		
	0.0181	0	0.0289	$\theta$	1.5	2	2.5	5	10	1.5	2	2.5	5	10
				1.5	1.20	1.61	2.02	4.11	8.52	20.17	20.23	20.28	20.55	21.11
				2	1.02	1.36	1.71	3.48	7.26	16.27	16.34	16.41	16.74	17.46
				2.5	0.88	1.18	1.48	3.02	6.32	13.62	13.69	13.75	14.11	14.87
X X 7 T				5	0.52	0.70	0.87	1.79	3.80	7.39	7.45	7.50	7.80	8.45
WI	0.0221	$egin{array}{c} a \ 0 \end{array}$	$\sigma_{\epsilon}$ 0.0409	θ	1.5	2	$_{2.5}^{\gamma}$	5	10	1.5	2	$_{2.5}^{\gamma}$	5	10
				1.5	2.31	3.10	3.90	8.07	17.35	19.41	19.51	19.61	20.12	21.24
				2	1.90	$\frac{3.10}{2.55}$	3.22	6.68	14.49	15.32	15.44	15.56	16.18	17.58
				2.5	1.61	2.17	2.73	5.69	12.42	12.63	12.75	12.87	13.50	14.98
				5	0.90	1.22	1.53	3.22	7.18	6.61	6.70	6.79	7.29	8.52
WV	g	a	$\sigma_\epsilon$				$\gamma$					$\gamma$		
	0.0246	0	0.0459	$\theta$	1.5	2	2.5	5	10	1.5	2	2.5	5	10
				1.5	2.84	3.82	4.81	10.04	21.97	18.95	19.07	19.19	19.81	21.20
				2	2.30	3.10	3.91	8.18	18.10	14.76	14.90	15.04	15.78	17.50
				2.5	1.93	2.60	3.28	6.90	15.38	12.06	12.20	12.34	13.09	14.88
				5	1.05	1.42	1.79	3.80	8.68	6.18	6.28	6.38	6.95	8.43
WY	0.0202	$egin{matrix} a \ 0 \end{bmatrix}$	$\sigma_{\epsilon}$ 0.0784	$\theta$	1.5	2	$_{2.5}^{\gamma}$	5	10	1.5	2	$_{2.5}^{\gamma}$	5	10
				1.5	9.21	12.61	16.10	37.51	_	20.70	21.12	21.55	24.04	_
				2	7.73	10.63	13.71	32.60	_	16.91	17.44	18.01	21.5	_
				$2.5^{-2}$	6.65	9.17	11.87	28.86	_	14.27	14.84	15.45	19.46	_
				5	3.88	5.39	7.04	18.43	=	7.92	8.41	8.96	13.21	=
USA	g	a	$\sigma_\epsilon$				$\gamma$					$\gamma$		
	0.0199	0.4266	0.0207	$\theta$	1.5	2	2.5	5	10	1.5	2	2.5	5	10
				1.5	1.05	1.63	2.21	5.21	11.71	19.73	19.81	19.89	20.29	21.15
				2	0.88	1.35	1.83	4.32	9.78	15.73	15.82	15.91	16.41	17.49
				2.5	0.75	1.15	1.56	3.68	8.37	13.05	13.14	13.24	13.75	14.89
				5	0.45	0.67	0.89	2.05	4.71	6.93	7.01	7.08	7.48	8.42
Median	g	a	$\sigma_\epsilon$			_	$\gamma$	_	4.0	<u>.</u> –	_	$\gamma$		4.5
	0.0207	0.0000	0.0409	Δ	1.5	2	2.5	5	10	1.5	2	2.5	5	10
				$\theta$	0.64	9 07	4.09	10.49	99.70	10.70	10.70	10.06	20.41	01 66
				1.5	$\frac{2.64}{2.17}$	$\frac{3.87}{3.11}$	4.93	10.43	22.70	$19.72 \\ 15.70$	19.79	19.86	20.41	21.66
				$\frac{2}{2.5}$	$\frac{2.17}{1.83}$	$\frac{3.11}{2.72}$	$\frac{4.16}{3.49}$	$8.84 \\ 7.64$	$19.60 \\ 17.15$	$15.70 \\ 13.02$	$15.79 \\ 13.12$	15.87 $13.19$	$16.55 \\ 13.89$	$18.10 \\ 15.56$
				2.5 5	1.05	$\frac{2.72}{1.51}$	$\frac{3.49}{1.99}$	4.39	10.10	6.91	6.99	7.05	7.60	9.06
				J	1.00	1.01	1.00	1.00	10.10	0.51	0.00	1.00	1.00	0.00

Note: The table compares the welfare cost of aggregate fluctuations (left panel) and the welfare gain from a permanent extra 1% consumption growth (right panel), in economies 2 [CRRA preferences and consumption growth following process (3)] and 3 [Epstein-Zin preferences and same law of motion for consumption growth] for different values of the preference parameters  $\gamma$  and  $\theta$ . For each state and the U.S. as a whole, the **parameter estimates** for process (3) are obtained from a standard AR(1) fit using state-level consumption data from 1960 to 1995. When the slope coefficient a is not statistically significant (at least at the 10% level), the other parameters are re-estimated by regressing the consumption growth rate on a constant. Cost estimates related to economy 2 appear as diagonal elements of the matrices ( $\gamma = \theta$ ). No numbers are reported when the cost estimates explode.

Figure 1: Eliminating business cycles vs. Promoting growth

Note: The figure identifies states where the welfare cost of business cycle,  $\lambda$ , exceeds the welfare gain of an extra 1% growth forever,  $\eta$  for values of the representative agent's risk aversion ( $\gamma=10$ , all model economies) and elasticity of intertemporal substitution ( $\theta<5$ , economy 3) that fall within accepted ranges. States are shown in black when  $\lambda>\eta$  regardless of the model economy. Dark-gray states are those where  $\lambda>\eta$  except in the benchmark Lucas economy [CRRA preferences and consumption following process (2)]. Massachusetts is shaded in very dark gray, because  $\lambda>\eta$  in the benchmark model and the cost of aggregate fluctuations there is extremely close to that of higher growth in economy 3 (i.e.  $\lambda\simeq\eta$ .)

