The Fiscal Roots of Inflation

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Abstract

Unexpected inflation devalues nominal government bonds. It must therefore correspond to a decline in expected future surpluses, or a rise in their discount rates, so that the real value of debt equals the present value of surpluses. I measure each component using a vector autoregression, via responses to inflation, recession, surplus and discount rate shocks. Discount rates, rather than deficits, account for most inflation variation, and discount rates explain why large deficits often do not cause inflation. Also, long-term debt is important to understand fiscal-monetary interactions. Smooth inflation slowly devalues outstanding long-term bonds.

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1 Introduction

This paper measures the fiscal roots of inflation. Higher inflation devalues nominal government debt. Higher inflation must therefore correspond to lower surplus/GDP ratios, lower GDP growth, or higher discount rates for government debt. I develop a set of linearized identities that express these connections. I measure these components in impulse-response functions of a simple vector autoregression (VAR).

I find that shocks to inflation largely correspond to shocks to discount rates. For example, consider 2008-2009. There was a big recession, and a sharp fall in inflation which raised the real value of nominal debt. Yet deficits exploded. How can this be? Well, perhaps people expected higher future primary surpluses to pay back the cumulated deficits and more, making nominal debt more valuable in real terms. Aside from its implausibility, I do not find this pattern in the data. But nominal and real interest rates on government debt fell sharply. Perhaps this lower discount rate for government surpluses increased their present value, justifying a lower than expected price level? I find that this is the case: the change in expected returns is large and persistent enough quantitatively to account for inflation shocks in the data.

I also examine shocks to surpluses and shocks to discount rates. These shocks come with essentially no inflation. Shocks to surpluses are highly correlated with shocks to discount rates, so the surplus and discount rate terms of the present value formula largely offset. Viewed in ex-post terms, persistent deficits come at the same time as low returns. Low returns bring back the value of debt, without needing repayment via later surpluses, or devaluation via an initial inflation. The strong correlation between discount rates and deficits provide fiscal roots of the otherwise-puzzling absence of inflation.

The two observations are not contradictory. There are multiple sources of variation in the data. Not all business cycles are alike. When we isolate a shock to inflation, we see events in which discount rates and deficits do not offset. When we isolate a shock to discount rates or deficits, we see a different slice of data, in which they do offset and there is not much inflation or deflation. If you sip a drink on the counter at Starbucks’ and find it bitter, the cup is likely to contain only coffee. But the average cup on the counter that contains coffee also contains sugar, and is not bitter.

I also find an important role for long-term debt. Simple models focus on one-period debt, and only a price-level jump can devalue such debt. With long-term debt, a
slow inflation can devalue long-term bonds when they come due. Expectation of such future inflation lowers nominal bond prices, restoring present value balance in place of a price-level jump. This mechanism, more plausible than one-time price-level jumps, is evident in the data, with expected future inflation accounting for large fractions of changes in the present value of debt.

I interpret the results through the lens of the fiscal theory of monetary policy: models with interest rate targets, fiscal theory of the price level, and potentially sticky prices, as described in Cochrane (2020a), Cochrane (2020b). (More literature below.) In this interpretation, changes in expected surpluses and discount rates cause unexpected inflation. In this interpretation, we study the fiscal roots rather than the fiscal consequences of inflation. This paper establishes a set of facts that will be useful for constructing such models, as atheoretical VARs guided the construction of conventional monetary models. The fact that discount rates account for much inflation variation is key to making a fiscal-theory analysis reasonable, and to allow a fiscal-theory model to account for events such as 2009, or the converse rise in inflation in low-deficit booms. A fiscal theory model must include time-varying discount rates with the right sign and cyclical pattern. My causal language below refers to this interpretation.

But the identities whose terms I measure hold in almost all macroeconomic models used to quantitatively address inflation, and therefore form a useful set of stylized facts for monetary and fiscal interaction in a wide set of models. (The calculations assume that the present value is finite – loosely that $r > g$. I presume this case without much comment, and I leave interpretation of the facts in $r < g$ models for another day.) The computations of this paper are deliberately “measurement without theory,” in the classic sense articulated by Koopmans (1947). I do not estimate any structural parameters, identify any structural shocks, or test one model vs. another. A “shock” only means a movement in a variable that is not forecast by the VAR, without structural interpretation.

In particular, standard new-Keynesian / DSGE models posit an opposite causal- ity of monetary-fiscal policy coordination. An interest-rate policy and an equilibrium-selection policy by the central bank determines inflation. Fiscal policy reacts “passively,” raising or lowering surpluses to validate inflation-induced changes in the value of government debt. Absent identifying assumptions, the fiscal theory of monetary policy causal story is observationally equivalent to this passive-fiscal causal story. Therefore, the
results of this paper can also be interpreted as measures of the fiscal adjustments to inflation that a standard new-Keynesian model must envision. The fact that discount rates do much of the adjusting, rather than the ex-post lump-sum taxes alluded to in many theoretical footnotes, changes the fiscal underpinnings of such models substantially.

Since the analysis is based on identities that hold in these and many other models, the empirical results do nothing to establish one or another causal story. But which element in an identity moves – whether surpluses or discount rates account for variation in the value of government debt – is still an interesting measurement, that bears on the construction of any theory.

More narrowly, this paper addresses a common attempt at armchair refutation of fiscal theory: We have huge debt and deficits, and no inflation. Debt and deficits increase in recessions, where inflation declines. The theory must be wrong. No. First, a low real interest rate is a low discount rate. A low real interest rate can account for a large value of nominal debt, and thereby low inflation. It does so in the estimates. Second, the government debt valuation equation holds equally in conventional monetary theories, and in the standard new-Keynesian and monetarist models in particular. If there is a puzzle in the fiscal foundations of inflation, it applies equally to conventional theories, and does not reject fiscal theory in favor of those other theories. As a paper on pure facts, I do not offer here theory or evidence on why interest rates are so low, or why they decline amid the deficits of a recession. Given that interest rates behave as they do, inflation makes fiscal sense.

2 Literature

The technique in this paper is adapted from asset pricing. The general approach to linearizing the valuation identity follows Campbell and Shiller (1988). Appendix C relates impulse-response calculations to asset price variance decompositions. The summary of this literature in Cochrane (2011b) and the treatment of identities in Cochrane (2007) are obvious precursors to this work. The uniting theme in the former is that asset price and return variation corresponds largely to variation in discount rates.

The analysis of government finances, how debt is paid off, grown out of, or inflated away, is a huge literature. Hall and Sargent (1997), Hall and Sargent (2011) are the most
important recent precursors. Hall and Sargent focus on the market value of debt, as I do, not the face value reported by the Treasury, and consequent proper accounting for interest costs.

Cochrane (1998) constructs a linearized present value equation similar to that used here, and uses it to decompose the value of government debt. Cochrane (2019) improves on that calculation, using the value identity (2). Both papers find that variation in expected primary surpluses is an important determinant of the value of debt.

The main methodological novelty is that this paper uses the innovation identities, (3) and (5) below, to focus on inflation, and eliminate the value of debt from the identities, paralleling VAR-based return decompositions from asset pricing such as Campbell and Ammer (1993). As we get different results by focusing on returns rather than prices, I find a greater role for discount rates here by focusing on innovations and inflation rather than the level of the value of debt.

The fiscal theory of monetary policy is the latest step in a long literature on the fiscal theory of the price level, starting with Leeper (1991), that integrates fiscal theory with sticky-price models and interest rate targets. Sims (2011), Cochrane (2017) are immediate antecedents. Cochrane (2020a) works out such a model with the S-shaped surplus processes I find here, calculates inflation decompositions and response functions in the model, and reviews the literature. It is not quite the theory paper corresponding to this work. I do not here identify the structural monetary and fiscal policy shocks studied there. To keep the theoretical points transparent, I do not there extend the model with all the embellishments necessary to be estimated, tested, and match dynamics in the data.

Much of the fiscal theory literature has pursued various theoretical controversies. A big point of this paper is to begin productively and quantitatively to use fiscal theory to understand US data. Fiscal theory will in the end be judged by its usefulness, as all past theories have been judged.
3 Identities

To develop the identities linking inflation to surpluses, growth, and discount rates, start with a linearized version of the government debt flow identity,

\[ \rho v_{t+1} = v_t + r_{t+1} - \pi_{t+1} - g_{t+1} - s_{t+1}. \]  

The quantity \( v_t \) is the log of the ratio of the market value of debt, henceforth just “debt,” to GDP. The log debt to GDP ratio at the end of period \( t + 1 \), \( v_{t+1} \), is equal to its value at the end of period \( t \), \( v_t \), increased by the log nominal return on the portfolio of government bonds \( r^n_{t+1} \), less inflation \( \pi_{t+1} \), less log GDP growth \( g_{t+1} \), and less the scaled real primary surplus to GDP ratio \( s_{t+1} \). The parameter \( \rho \) is a constant of linearization, \( \rho = e^{r-g} \), which I take to be \( \rho = 1 \) in the numerical results. I derive this identity in Appendix A.

All variables in (1) are logs, except the surplus. I Taylor expand the level of the surplus, to allow the surplus to be negative. As a result the surplus is scaled to generate percentage units. The variable \( s_t \) is \( \rho \) times the ratio of primary surplus to GDP scaled by the debt to GDP ratio at the linearization point. With \( \rho = 1 \), \( s_t \) can also represent the real primary surplus divided by the previous period’s debt. Either definition leads to the same linearization. In the data, I impute the surplus from the other terms of (1), so its definition only matters when one wishes to assess an independent data source on surpluses. For brevity, I refer to \( s_t \) simply as the “surplus.” With \( \rho < 1 \) there is also a constant in the linearization, or the variables are deviations from steady state.

Iterating forward, we have a present value identity,

\[ v_t = \sum_{j=1}^{\infty} \rho^{j-1} s_{t+j} + \sum_{j=1}^{\infty} \rho^{j-1} g_{t+j} - \sum_{j=1}^{\infty} \rho^{j-1} (r^n_{t+j} - \pi_{t+j}). \]  

Taking expected values, the debt to GDP ratio is the present value of future surplus to GDP ratios, discounted at the ex-post real return, and adjusted for growth. (Higher GDP growth, with the same surplus to GDP ratio, gives rise to greater surpluses.)

Taking time \( t + 1 \) innovations \( \Delta E_{t+1} \equiv E_{t+1} - E_t \) and rearranging, we have an
unexpected inflation identity,

\[
\Delta E_{t+1} \pi_{t+1} - \Delta E_{t+1} r^n_{t+1} = - \sum_{j=0}^{\infty} \rho^j \Delta E_{t+1} s_{t+1+j} - \\
- \sum_{j=0}^{\infty} \rho^j \Delta E_{t+1} g_{t+1+j} + \sum_{j=1}^{\infty} \rho^j \Delta E_{t+1} \left( r^n_{t+1+j} - \pi_{t+1+j} \right).
\] (3)

A decline in the present value of surpluses, coming either from a decline in surplus to GDP ratios, a decline in GDP growth, or a rise in discount rates, must correspond to a lower real value of the debt. This reduction can come about by unexpected inflation, or by a decline in nominal long-term bond prices. I use time \( t + 1 \) to denote unexpected events, and time 1 as the date of a shock in the impulse-response functions.

The second term on the left hand side of (3) is a key point of the analysis. For example, when there is a negative innovation to the present value of surpluses on the right hand side of (3), a decline in nominal long-term bond prices and consequent negative return \( \Delta E_{t+1} r^n_{t+1} \) can lower the real value of debt, in place of unexpected inflation \( \Delta E_{t+1} \pi_{t+1} \). In this way, long-term debt can buffer fiscal shocks.

What determines the long-term bond return \( r^n_{t+1} \)? Appendix B linearizes the return of the government bond portfolio around a geometric maturity structure, in which the face value of maturity \( j \) debt declines at rate \( \omega^j \),

\[
\Delta E_{t+1} r^n_{t+1} = - \sum_{j=1}^{\infty} \omega^j \Delta E_{t+1} r^n_{t+1+j} = - \sum_{j=1}^{\infty} \omega^j \Delta E_{t+1} \left( \left( r^n_{t+1+j} - \pi_{t+1+j} \right) + \pi_{t+1+j} \right). \] (4)

Lower nominal bond prices, and a lower ex-post bond return, mechanically correspond to higher bond expected nominal returns, which in turn are composed of real returns and inflation.

We can then eliminate the bond return in (3)-(4) to focus on inflation and fiscal affairs alone,

\[
\sum_{j=0}^{\infty} \omega^j \Delta E_{t+1} \pi_{t+1+j} = - \sum_{j=0}^{\infty} \rho^j \Delta E_{t+1} s_{t+1+j} - \sum_{j=0}^{\infty} \rho^j \Delta E_{t+1} g_{t+1+j} + \\
+ \sum_{j=1}^{\infty} (\rho^j - \omega^j) \Delta E_{t+1} \left( r^n_{t+1+j} - \pi_{t+1+j} \right). \] (5)
I focus on this decomposition.

I assume here that the expected terminal condition vanishes, $E_{t+1} \lim_{T \to \infty} \rho^T v_{t+T} = 0$, and the sums converge. Equality $r = g$ and $\rho = 1$ are also allowable, since I take deviations from the mean and the variables are stationary.

While there is an enormous literature on the $r < g$ possibility, most recently Blanchard (2019), I don’t have anything novel to say here about necessary or sufficient conditions in economic models for the sums to converge or the limiting discounted debt/GDP ratio to vanish, or about the astonishing implications for public finance if debt really does not need to be repaid. The sums converge and the terminal condition vanishes in the VAR estimates, and I interpret the calculations in light of economic models that have the same features. How to interpret the data on inflation and the value of debt, through the lens of models with $r > g$, a truly infinite value of government debt, or debt that never needs to be repaid, I leave as questions for another day.

### 3.1 Mechanisms

The identities highlight several interesting mechanisms which we can look for in the data.

Consider the simple case with constant expected returns $E_t r^n_{t+1} = E_t \pi_{t+1}$. With one-period debt, $\omega = 0$, there is only one term on the left-hand side of (5), $\Delta E_{t+1} \pi_{t+1}$. Shocks to the present value of surpluses must be soaked up by a price-level jump, which absent default is the only way to devalue outstanding debt. With long-term debt, $\omega > 0$, however, a shock to the present value of surpluses can result in a drawn out period of inflation, which slowly devalues outstanding long-term bonds. In the identity (3), the term $r^n_{t+1}$ marks the future inflation to market, as future inflation in (4) lowers that return.

The latter is a more realistic vision of the US economy, where we see drawn-out inflation accompanying fiscal problems as in the 1970s, not one-time price-level jumps. In fact, equation (5) allows the entire effect of the fiscal shock to show up in expected future inflation with no movement in current inflation $\Delta E_{t+1} \pi_{t+1} = 0$. This is how continuous-time models that disallow price-level jumps work (Cochrane (2017)). This case provides an important counterexample to the usual intuition gained from one-period models, in which fiscal shocks only give rise to a one-period inflation surprise. We can productively look for fiscal roots of drawn-out inflation.

In both the fiscal theory of monetary policy and in standard new-Keynesian the-
ories, monetary policy controls the path of nominal interest rates and thereby controls the path of expected inflation. For example, in the simplest frictionless model, 
\[ i_t = E_t \pi_{t+1} \] so the interest rate target sets expected inflation directly. Thus monetary policy determines whether fiscal shocks result in smooth inflation in (5), and whether there is a large bond-return term in the one-period accounting (3).

As the maturity structure of government debt lengthens, \( \omega \) increases, and the discount rate terms in (5) get smaller. When \( \omega = \rho \), almost a perpetuity, the discount rate term drops out. Intuitively, a government that funds itself with near-perpetuities can pay off its current debt while ignoring real interest rate variation, just as a household that takes out a fixed rate mortgage is immune from rate variation.

If surpluses and growth are constant as well as discount rates in (5), with long term debt a rise in expected future inflation results in a decline in current inflation. This is an important mechanism for monetary policy to temporarily reduce inflation. The rise in interest rates raises long-term inflation, but with no change in fiscal policy, that rise must lower near-term inflation.

With one-period debt, expected inflation may continue to be high after an initial inflation shock, but this fact has no impact on one-period unexpected inflation or this fiscal accounting. With \( \omega = 0 \), \( \Delta E_1 \pi_j \) for \( j > 1 \) is irrelevant (though interesting) in (5). With long-term debt, the weighted sum of changes in expected inflation substitutes for inflation at time 1, but only the \( \omega \)-weighted sum. Additional persistence in inflation, though interesting for matching data, has no fiscal consequence or consequence for understanding unexpected inflation.

With time-varying expected returns, interesting additional dynamics can emerge. A higher nominal interest rate results in lower nominal bond prices and less inflation today. But with sticky prices, the higher nominal interest rate raises the real interest rate and discount rate. This is an inflationary force, which offsets the direct deflationary force of the higher nominal rate. If a deficit shock comes with lower real interest rates and expected bond returns, the latter raise the value of debt and offset the inflationary effect of the surplus shock. We will these effects in the data.
4 Data and VAR estimates

I use data on the market value of government debt held by the public and the nominal rate of return of the government debt portfolio from Hall, Payne, and Sargent (2018). I use standard BEA data for GDP and total consumption. I use the GDP deflator to measure inflation. I use CRSP data for the three-month Treasury rate. I use the 10-year constant maturity government bond yield from 1953 on and the yield on long-term United States bonds before that date to measure a long-term bond yield.

I measure the debt to GDP and surplus to GDP ratios by the ratios of debt and surplus to consumption, times the average consumption to GDP ratio. Debt to GDP ratios are often used to compare countries, but in our time-series application they introduce cyclical variation in GDP. We want only a detrending divisor, and an indicator of the economy’s long-run level of tax revenue and spending. Potential GDP has a severe look-ahead bias. Consumption is a decent stochastic trend for GDP.

I infer the primary surplus from the flow identities. This calculation measures how much money the government actually borrows. NIPA surplus data, though broadly similar, does not obey the flow identity.

I infer the surplus for the VAR from the linearized identity (1), at an annual frequency. By doing so, the data obey the identity exactly. Therefore VAR estimates of the decompositions add up exactly with no approximation error. The approximation errors are much smaller than sampling errors, so this choice just produces clearer tables.

To measure the accuracy of the linear approximation, I also infer the monthly real primary surplus from the exact nonlinear flow identity, Appendix equation (11). I then carry the surplus to the end of the year using the government bond return. This procedure produces an annual series for which the nonlinear flow identity (11) continues to hold in annual data.

I approximate around $r = g$ or $\rho = 1$. The variables are all stationary, impulse-responses and expected values converge, so downweighting higher order terms by, say, $0.99^j$ vs. 1.0 makes little difference to the results. Since the value of the debt $v_t$ is stationary, $\lim_{T \to \infty} E_t v_{t+T} = 0$ without $\rho$ weighting. The parameter $\rho$ is only the arbitrary point about which one takes a Taylor expansion of the one-period flow relation. It need not be determined by a long run average $r - g$ in the economy. With $\rho = 1$, the same linearization applies to the surplus to value ratio, which is a bit more accurate. One can also view
the unweighted $\rho = 1$ identities as $r \to g$ limits.

Figure 1: Surplus. “Linear” is inferred from the linearized flow identity, and is the definition used in VAR analysis. “$s_v$” is the exact ratio of the primary surplus to the previous year’s market value of the debt. “$s_y$” is the exact ratio of surplus to consumption, scaled by the average consumption to GDP ratio and the average value of debt. Vertical shading denotes NBER recessions.

Figure 1 presents the surplus and compares three measures. The “Linear, $s_t$” line imputes the surplus from the linearized flow identity (1) directly at the one-year horizon, which is the measure I use in the following analysis. The “$s_v$” and “$s_y/e^v$” lines both infer the surplus from the exact nonlinear flow identity, Appendix equation (11). The “$s_v$” line presents the ratio of the exact surplus to the previous year’s value of the debt. The “$s_y/e^v$” line presents the exact surplus to GDP ratio – actually, the ratio of surplus to consumption, times the average consumption to GDP ratio – scaled by the average value to GDP ratio $e^{E(v_t)}$.

The first piece of news in Figure 1 is that there are primary surpluses. One’s impression of endless deficits comes from the deficit including interest payments on the debt. NIPA measures (not shown) also show regular positive primary surpluses. Steady
primary surpluses from 1947 to 1975 helped to pay off WWII debt. The year 1975 started an era of large primary deficits, interrupted by the strong surpluses of the late 1990s. Postwar primary surpluses also have a clear cyclical pattern. The primary surplus correlates very well with the unemployment rate (not shown), a natural result of procyclical tax revenues, automatic (e.g. unemployment insurance) and discretionary countercyclical spending.

The three measures in Figure 1 are close. The graph is a measure of the accuracy of the linearized identity (1). The linearized identity is a slightly closer approximation to the surplus to value ratio $sv$. The difference is largest when the value of debt is far from its mean, both in WWII and in the 1970s.

I use a postwar data sample 1947-2018 for the main VAR analysis, as is conventional in empirical macroeconomics. One may well suspect that financing that war, and expectations and reality of paying off war debt, follow a different pattern than fiscal-monetary policy in the subsequent decades of largely cyclical deficits. Appendix G includes results from 1930-2018, including the great depression and WWII. The results are quite different, in ways traceable to a few influential data points. That analysis suggests that using full sample results to characterize the post WWII regime is not a good idea.

### 4.1 Vector autoregression

Table 1 presents OLS estimates of the VAR coefficients. Each column is a separate regression. I orthogonalize shocks later, so the order of variables has no significance. The VAR includes the central variables for the inflation identity – nominal return on the government bond portfolio $r^n$, consumption growth rate $g$, inflation $\pi$, surplus $s$ and value $v$. I include the three-month interest rate $i$ and the 10 year bond yield $y$ as they are important forecasting variables for growth, inflation, and long-term bond returns.

It is important to include the value of debt $v_t$ in the VAR, even if we are calculating terms of the innovation identity (3) that does not reference that variable. When we deduce from the present value identity (2) expressions $v_t = E_t(\cdot)$, we must include $v_t$ in the information set that takes the expectation. The surplus typically follows an s-shaped process, in which deficits today are followed by surpluses in the future. The process will not be properly recovered by VARs that do not include the value of debt. (See Cochrane (2020b), Cochrane (2020a) for discussion and examples.)
Table 1: OLS VAR estimate. Sample 1947-2018. One (two) stars means the estimate is one (two) Monte Carlo standard errors away from zero.

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<th></th>
<th>$r_{t+1}$</th>
<th>$g_{t+1}$</th>
<th>$\pi_{t+1}$</th>
<th>$s_{t+1}$</th>
<th>$v_{t+1}$</th>
<th>$i_{t+1}$</th>
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<td>$r_{t}$</td>
<td>-0.17**</td>
<td>-0.02</td>
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<td>0.20*</td>
<td>0.16*</td>
<td>1.37**</td>
<td>-2.00**</td>
<td>0.28</td>
<td>0.06</td>
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<tr>
<td>$\pi_{t}$</td>
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<td>-0.14*</td>
<td>0.53**</td>
<td>-0.25</td>
<td>-0.29</td>
<td>0.09</td>
<td>0.04</td>
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<tr>
<td>$s_{t}$</td>
<td>0.12**</td>
<td>0.03</td>
<td>-0.03*</td>
<td>0.35**</td>
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<td>-0.04</td>
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<td>-0.02**</td>
<td>0.04*</td>
<td>0.98**</td>
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<td>$i_{t}$</td>
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<td>-0.40*</td>
<td>0.29*</td>
<td>0.50</td>
<td>-0.72</td>
<td>0.73**</td>
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<td>$y_{t}$</td>
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<td>0.54**</td>
<td>-0.17</td>
<td>-0.04</td>
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<td>0.11</td>
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<td>Corr $\varepsilon, \varepsilon_{\pi}$</td>
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<td>-0.24</td>
<td>1.00</td>
<td>-0.14</td>
<td>-0.11</td>
<td>0.21</td>
<td>0.31</td>
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<tr>
<td>$R^2$</td>
<td>0.71*</td>
<td>0.17*</td>
<td>0.73*</td>
<td>0.48</td>
<td>0.97*</td>
<td>0.82*</td>
<td>0.90*</td>
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<td>$100 \times \text{std}(x)$</td>
<td>4.08</td>
<td>1.68</td>
<td>2.16</td>
<td>6.61</td>
<td>37.00</td>
<td>2.96</td>
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I use a single lag. Adding the last variable, the long-term rate, already introduces slight wiggles in the impulse-response function indicative of overfitting. More lags are insignificant forecasters, and add additional wiggles without much changing results. The results depend on long-run forecasts, which are controlled by the most persistent combination of variables. Fast-moving variables that improve short-term forecasts have little effect on long-term forecasts.

I compute standard errors from a Monte Carlo, described in the Appendix. The stars in Table 1 represent one or two standard errors above zero. Since we aren’t testing anything, stars are just a visual way to show standard errors without another table.

In the first column, the long-term bond yield $y_t$ forecasts the government bond portfolio return $r_{t+1}^n$ (1.93). The negative coefficient on the three-month rate $i_t$ means that the long-short spread also forecasts those returns. Since the $y_t$ and $i_t$ coefficients are not repeated in forecasting inflation and growth, the long rate and long-short spread forecast real, growth-adjusted, and excess returns on government bonds, as we expect from the long literature in which yield spreads forecast bond risk premia (Fama and Bliss (1987), Campbell and Shiller (1991), Cochrane and Piazzesi (2005)). The long rate $y_t$ is thus an important state variable for measuring expected bond returns, the relevant discount rate for our present value computations. (It’s common to use the yield spread $y_t - i_t$ as a forecasting variable. However, when forecasting inflation, we also want to include...
the level of the short rate \( i_t \). Forecasts using short rate \( i_t \) and yield spread \( y_t - i_t \) are the same as the forecasts using short rate \( i_t \) and yield \( y_t \), though the coefficients are not the same.)

Growth \( g_t \) is only very slightly persistent (0.20). The term spread \( y_t - i_t \) also predicts economic growth, and reinforcing the importance of the interest rates as state variables.

Inflation \( \pi_t \) is moderately persistent (0.53). The interest rate and growth help a bit to predict inflation, but not much else does. We will see inflation responses that mostly look like AR(1) decay.

The surplus is somewhat persistent (0.35). Growth \( g_t \) predicts higher surpluses, an important and realistic feedback mechanism. Inflation forecasts deficits (-0.25), so we expect that to some extent inflation may be related to subsequent deficits. Debt also forecasts surpluses (0.04), which is important to the following dynamics. Deficits raise debt, and then larger debts lead to surpluses which slowly pay off some of the debt accumulated from the deficits. This response does not imply passive fiscal policy, as discussed in more detail below.

The value of the debt is very persistent (0.98). It thus becomes the most important state variable for long-run calculations. A larger surplus \( s_t \) forecasts lower debt, \( v_{t+1} \), \((-0.24)\), as one expects. The long-run yield \( y_t \) forecasts a rise in the value of debt \( v_{t+1} \), as we expect given its effect on the expected return \( r_{t+1}^{\pi} \).

The short rate \( i_t \) and long yield \( y_t \) are also persistent (0.73, 0.46) and the long yield is forecast by the interest rate, again reflecting standard yield curve dynamics.

For calculations reported below, I use the standard notation

\[
x_{t+1} = Ax_t + \varepsilon_{t+1}
\]

(6) to denote this VAR.

5 Responses and decomposition estimates

I start by examining the fiscal roots of a simple inflation shock, an unexpected movement in inflation \( \Delta E_1 \pi_1 = \varepsilon_1^\pi = 1 \). I allow all other variables to move contemporaneously to the inflation shock. In either reading of causality, we want to measure simultaneous movements of inflation and other variables, for example that a shock to current and future
surpluses caused inflation, or that a shock to inflation caused changes in fiscal policy within the period. To measure how much other variables typically move conditional on seeing an inflation shock, I fill in shocks to the other variables by running regressions of their shocks on the inflation shock. For each variable \( z \), I run

\[
\varepsilon_{z,t+1} = b_{z,\pi} \varepsilon_{\pi,t+1} + \eta_{z,t+1}.
\]

Then I start the VAR (6) at

\[
\varepsilon_1 = \begin{bmatrix} b_{r,\pi} & b_{g,\pi} & \varepsilon_{\pi,1} = 1 & b_{s,\pi} & \ldots \end{bmatrix}'.
\]

This procedure is equivalent to the usual orthogonalization of the shock covariance matrix, but it is more transparent and it generalizes more easily later. I denote the VAR innovations as the change in expectations at time 1, i.e. \( \Delta E_1 \), and thus the response of variable \( x, j \) periods in the future is \( \Delta E_{1,x,j} \).

Figure 2 plots responses to this inflation shock. The “Inflation” rows of Table 2 present the terms of the inflation and bond return decompositions for these responses. (I discuss the remaining rows of Table 2 later.) Figure 2 also presents some of the main terms in the decomposition identities, (3), (4), (5).

In any interpretation, these responses and calculations answer the question, “if we see an unexpected 1% inflation, how should we revise our forecasts of other variables?” In a fiscal-theoretic interpretation, they answer “what changes in expectations caused the 1% inflation?” As shown in Appendix C, the inflation decompositions care also decompositions of the variance of unexpected inflation: They answer the question, “What fraction of the variance of unexpected inflation is due to each component?”

Table 3 presents Monte Carlo quantiles of the sampling distributions of the terms of the inflation decompositions in Table 2. Figure 9, below, plots quantiles of the impulse-response functions. A reader hungry to see that sampling variation may wish to peek. Sampling variation merits a longer discussion, however, which I postpone until we see the message in the point estimates.

In Figure 2, the inflation shock is moderately persistent, largely following the AR(1) dynamics we noticed in the VAR coefficients. As result, the weighted sum \( \sum_{j=0}^{\infty} \omega^j \Delta E_{1,\pi,1+j} = 1.59\% \), greater than the 1% initial shock.
Figure 2: Responses to a 1% inflation shock.

The inflation shock coincides with a deficit $s_1$, which builds with a hump shape. That hump shape largely represents the -0.25 coefficient by which inflation forecasts surpluses. One might think these persistent deficits account for inflation. But surpluses eventually rise to pay back almost all of the incurred debt. The sum of all surplus responses is $-0.06\%$, essentially zero.

The line marked $r^n - \pi$ plots the response of the real discount rate, $\Delta E_1(r^n_{1+j} - \pi_{1+j})$. These points are plotted at the time of the ex-post return, $1 + j$, so they are the expected return one period earlier, at time $j$. The line starts at time 2, where the terms of the discount-rate sums in the inflation decompositions start, and representing the time-1 expected return. After two periods, this discount rate rises and stays persistently positive. The weighted sum of discount rate terms is 1.04% while the unweighted sum is 1.00% (really 1.004%). The weight $\omega = 0.69$, chosen to make the identity (4) hold exactly for this response function, so weighting by $1$ vs. $1 - \omega^j$ makes little difference in the face of this persistent response.

Weighted or unweighted, the discount rate terms account for 1% inflation. A higher discount rate lowers the value of government debt, an inflationary force.
\[
\sum_{j=0}^{\infty} \omega^j \Delta E_{1 \pi_1+j} = -\sum_{j=0}^{\infty} \Delta E_{1 s_1+j} - \sum_{j=0}^{\infty} \Delta E_{1 g_1+j} + \sum_{j=1}^{\infty} (1 - \omega^j) \Delta E_{1}(r^n_{1+j} - \pi_{1+j})
\]

<table>
<thead>
<tr>
<th>Event</th>
<th>Inflation</th>
<th>Recession</th>
<th>Surplus</th>
<th>Disc. Rate</th>
<th>Surplus, no i</th>
</tr>
</thead>
<tbody>
<tr>
<td>(\pi)</td>
<td>1.59</td>
<td>-2.36</td>
<td>-0.10</td>
<td>-0.18</td>
<td>0.38</td>
</tr>
<tr>
<td>(s)</td>
<td>-( -0.06)</td>
<td>-( -1.15)</td>
<td>-( -0.66)</td>
<td>-( -0.54)</td>
<td>-( -0.52)</td>
</tr>
<tr>
<td>(g)</td>
<td>-( -0.49)</td>
<td>-( -1.46)</td>
<td>-( -0.34)</td>
<td>-( -0.28)</td>
<td>-( -0.48)</td>
</tr>
<tr>
<td>(r^n - \pi)</td>
<td>+( 1.04)</td>
<td>+( -4.96)</td>
<td>+( -1.10)</td>
<td>+( -1.00)</td>
<td>+( -0.62)</td>
</tr>
</tbody>
</table>

\[
\Delta E_1 \pi_1 - \Delta E_1 r^n_1 = -\sum_{j=0}^{\infty} \Delta E_1 s_1+j - \sum_{j=0}^{\infty} \Delta E_1 g_1+j + \sum_{j=1}^{\infty} \Delta E_1(r^n_{1+j} - \pi_{1+j})
\]

<table>
<thead>
<tr>
<th>Event</th>
<th>Inflation</th>
<th>Recession</th>
<th>Surplus</th>
<th>Disc. Rate</th>
<th>Surplus, no i</th>
</tr>
</thead>
<tbody>
<tr>
<td>(\pi)</td>
<td>1.00</td>
<td>-1.00</td>
<td>0.02</td>
<td>-0.03</td>
<td>0.36</td>
</tr>
<tr>
<td>(s)</td>
<td>-( -0.56)</td>
<td>-( 1.19)</td>
<td>-( -0.27)</td>
<td>-( -0.28)</td>
<td>-( -0.03)</td>
</tr>
<tr>
<td>(g)</td>
<td>-( -0.49)</td>
<td>-( -1.15)</td>
<td>-( -0.66)</td>
<td>-( -0.54)</td>
<td>-( -0.52)</td>
</tr>
<tr>
<td>(r^n - \pi)</td>
<td>+( 1.00)</td>
<td>+( -4.79)</td>
<td>+( -1.25)</td>
<td>+( -1.13)</td>
<td>+( -0.67)</td>
</tr>
</tbody>
</table>

\[
\Delta E_1 r^n_1 = -\sum_{j=1}^{\infty} \omega^j \Delta E_1(r^n_{1+j} - \pi_{1+j}) - \sum_{j=1}^{\infty} \omega^j \Delta E_1 \pi_1+j
\]

<table>
<thead>
<tr>
<th>Event</th>
<th>Inflation</th>
<th>Recession</th>
<th>Surplus</th>
<th>Disc. Rate</th>
<th>Surplus, no i</th>
</tr>
</thead>
<tbody>
<tr>
<td>(r^n - \pi)</td>
<td>-0.56</td>
<td>1.19</td>
<td>0.27</td>
<td>0.28</td>
<td>0.03</td>
</tr>
<tr>
<td>(\pi)</td>
<td>-( -0.03)</td>
<td>-( 0.17)</td>
<td>-( -0.15)</td>
<td>-( -0.13)</td>
<td>-( -0.05)</td>
</tr>
<tr>
<td></td>
<td>-( -0.59)</td>
<td>-( -1.36)</td>
<td>-( -0.12)</td>
<td>-( -0.15)</td>
<td>-( -0.02)</td>
</tr>
</tbody>
</table>

Table 2: Terms of the inflation and bond return identities. The inflation shock is a 1 percent unexpected rise in inflation. The recession shock is a 1 percent unexpected decline in inflation and growth. The surplus shock is a 1 percent unexpected decline in the sum of current and future surpluses. The discount rate shock is a 1 percent unexpected decline the sum of current and future expected returns. The Surplus, no i shock holds the interest rate constant for two years after a surplus shock. Sample 1947-2018

Inflation also is also correlated with a persistent decline in economic growth \(g\). The stagflationary episodes of the 1970s drive this result. The growth decline contributes 0.49% to the inflation decompositions.

Overall, then,

1. A 1% shock to inflation corresponds to a roughly 1.5% decline in the present value of surpluses. A rise in discount rate contributes about 1%, and a decline in growth
\[ \sum_{j=0}^{\infty} \omega^j \Delta E_1 \pi_{1+j} = - \sum_{j=0}^{\infty} \Delta E_1 s_{1+j} - \sum_{j=0}^{\infty} \Delta E_1 g_{1+j} + \sum_{j=1}^{\infty} (1 - \omega^j) \Delta E_1 (r_{1+j}^n - \pi_{1+j}) \]

<table>
<thead>
<tr>
<th>Event</th>
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<th>Value 2</th>
<th>Value 3</th>
<th>Value 4</th>
</tr>
</thead>
<tbody>
<tr>
<td>Inflation 25%</td>
<td>1.38</td>
<td>-(-0.69)</td>
<td>-(-0.72)</td>
<td>+(0.16)</td>
</tr>
<tr>
<td>Inflation 75%</td>
<td>1.64</td>
<td>-(-0.23)</td>
<td>-(-0.22)</td>
<td>+(1.46)</td>
</tr>
<tr>
<td>Recession 25%</td>
<td>-2.41</td>
<td>-(-1.28)</td>
<td>-(-1.45)</td>
<td>+(4.84)</td>
</tr>
<tr>
<td>Recession 75%</td>
<td>-2.05</td>
<td>-(-0.49)</td>
<td>-(-0.57)</td>
<td>+(2.43)</td>
</tr>
<tr>
<td>Surplus 25%</td>
<td>-0.11</td>
<td>-(-0.78)</td>
<td>-(-0.39)</td>
<td>-(1.11)</td>
</tr>
<tr>
<td>Surplus 75%</td>
<td>0.02</td>
<td>-(-0.61)</td>
<td>-(-0.22)</td>
<td>-(0.98)</td>
</tr>
<tr>
<td>Disc. Rate 25%</td>
<td>-0.26</td>
<td>-(-0.63)</td>
<td>-(-0.34)</td>
<td>-(1.00)</td>
</tr>
<tr>
<td>Disc. Rate 75%</td>
<td>-0.13</td>
<td>-(-0.46)</td>
<td>-(-0.18)</td>
<td>-(1.00)</td>
</tr>
<tr>
<td>Surplus, no i 25%</td>
<td>0.21</td>
<td>-(-0.78)</td>
<td>-(-0.48)</td>
<td>-(0.76)</td>
</tr>
<tr>
<td>Surplus, no i 75%</td>
<td>0.45</td>
<td>-(-0.52)</td>
<td>-(-0.22)</td>
<td>-(0.50)</td>
</tr>
</tbody>
</table>

\[ \Delta E_1 \pi_1 - \Delta E_1 r_{1}^n = - \sum_{j=0}^{\infty} \Delta E_1 s_{1+j} - \sum_{j=0}^{\infty} \Delta E_1 g_{1+j} + \sum_{j=1}^{\infty} \Delta E_1 (r_{1+j}^n - \pi_{1+j}) \]

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<th>Value 3</th>
<th>Value 4</th>
</tr>
</thead>
<tbody>
<tr>
<td>Inflation 25%</td>
<td>1.00</td>
<td>-(-0.71)</td>
<td>-(-0.69)</td>
<td>-(-0.72)</td>
</tr>
<tr>
<td>Inflation 75%</td>
<td>1.00</td>
<td>-(-0.39)</td>
<td>-(-0.23)</td>
<td>-(-0.22)</td>
</tr>
<tr>
<td>Recession 25%</td>
<td>-1.00</td>
<td>-(-0.96)</td>
<td>-(-1.28)</td>
<td>-(-1.45)</td>
</tr>
<tr>
<td>Recession 75%</td>
<td>-1.00</td>
<td>-(-1.40)</td>
<td>-(-0.49)</td>
<td>-(-0.57)</td>
</tr>
<tr>
<td>Surplus 25%</td>
<td>0.00</td>
<td>-(-0.21)</td>
<td>-(-0.78)</td>
<td>-(-0.39)</td>
</tr>
<tr>
<td>Surplus 75%</td>
<td>0.09</td>
<td>-(-0.34)</td>
<td>-(-0.61)</td>
<td>-(-0.22)</td>
</tr>
<tr>
<td>Disc. Rate 25%</td>
<td>-0.07</td>
<td>-(-0.25)</td>
<td>-(-0.63)</td>
<td>-(-0.34)</td>
</tr>
<tr>
<td>Disc. Rate 75%</td>
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<td>-(-0.42)</td>
<td>-(-0.46)</td>
<td>-(-0.18)</td>
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<tr>
<td>Surplus, no i 25%</td>
<td>0.18</td>
<td>-(-0.08)</td>
<td>-(-0.78)</td>
<td>-(-0.48)</td>
</tr>
<tr>
<td>Surplus, no i 75%</td>
<td>0.38</td>
<td>-(-0.07)</td>
<td>-(-0.52)</td>
<td>-(-0.22)</td>
</tr>
</tbody>
</table>

\[ \Delta E_1 r_{1}^n = - \sum_{j=1}^{\infty} \omega^j \Delta E_1 (r_{1+j}^n - \pi_{1+j}) - \sum_{j=1}^{\infty} \omega^j \Delta E_1 \pi_{1+j} \]

<table>
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<th>Value 4</th>
</tr>
</thead>
<tbody>
<tr>
<td>Inflation 25%</td>
<td>-0.71</td>
<td>-(-0.12)</td>
<td>-(-0.38)</td>
<td></td>
</tr>
<tr>
<td>Inflation 75%</td>
<td>-0.39</td>
<td>-(-0.19)</td>
<td>-(-0.64)</td>
<td></td>
</tr>
<tr>
<td>Recession 25%</td>
<td>0.96</td>
<td>-(-0.17)</td>
<td>-(-1.41)</td>
<td></td>
</tr>
<tr>
<td>Recession 75%</td>
<td>1.40</td>
<td>-(-0.28)</td>
<td>-(-1.05)</td>
<td></td>
</tr>
<tr>
<td>Surplus 25%</td>
<td>0.21</td>
<td>-(-0.24)</td>
<td>-(-0.13)</td>
<td></td>
</tr>
<tr>
<td>Surplus 75%</td>
<td>0.34</td>
<td>-(-0.12)</td>
<td>-(-0.05)</td>
<td></td>
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<tr>
<td>Disc. Rate 25%</td>
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<td>-(-0.24)</td>
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<tr>
<td>Disc. Rate 75%</td>
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<td>-(-0.11)</td>
<td></td>
</tr>
<tr>
<td>Surplus, no i 25%</td>
<td>-0.08</td>
<td>-(-0.18)</td>
<td>-(-0.00)</td>
<td></td>
</tr>
<tr>
<td>Surplus, no i 75%</td>
<td>0.07</td>
<td>-(-0.00)</td>
<td>-(-0.10)</td>
<td></td>
</tr>
</tbody>
</table>

Table 3: Monte Carlo quantiles of the inflation and bond return identities.
accounts for about 0.5% of that decline. Changes in the surplus/GDP ratio account for nearly nothing. The additional 0.5% fiscal shock corresponds to a persistent rise in expected inflation, which slowly devalues outstanding long-term bonds, and produces a 1.5% overall rise in inflation weighted by the maturity structure of debt.

This is an important finding for matching the fiscal theory to data, or for understanding the fiscal side of standard passive-fiscal models. Thinking in both contexts has focused on the presence or absence of surpluses, or surplus to GDP ratios – lump sum taxes in many discussions – not discount rate or growth effects. Thinking in both contexts has considered one-period unexpected inflation, to devalue one-period bonds, not a rise in expected inflation that slowly devalues outstanding long-term bonds.

Turn to Table 2 for a more systematic view of the inflation decompositions, and to see the role of one-period bond returns $\Delta E_1 r^n_1$. The top row of the top panel presents the just-discussed overall decomposition (5) of current and expected future inflation in terms of surplus, growth and discount rate shocks. The second and third panels express the inflation decomposition in one-period terms, using the bond return $r^n_1$. The sum of surpluses and sum of growth rate terms are the same in this second panel as in the top panel, but I repeat them so one can see the terms of each identity more clearly. In the first row of the second panel, the 1% inflation shock corresponds to a roughly 1.56% overall fiscal shock. That shock comes similarly from a tiny 0.06% decline in surpluses, a 1.004% rise in discount rate and 0.49% reduction in growth. Here, the extra 0.56% fiscal shock is absorbed by a 0.56% decline in the value of government debt, $r^n_1$.

Turning to the last panel, we see that -0.56% return on government debt comes almost entirely from expected inflation (0.59%) not a higher real discount rate (0.03%).

Discount rates matter in the inflation decompositions of the top two panels but not in this return decomposition because the former have weights that emphasize long-term movements ($1$ and $1 - \omega^j$), while the $\omega^j$ weights of the bottom panel emphasizes a short-run movement in discount rate. With these weights, the early discount rate declines shown in Figure 2 match nearly exactly the subsequent persistent rise.

Comparing the two analyses, you see how the government bond return essentially marks to market the expected future inflation of the top panel. Here, the roughly 1.5% fiscal shock is absorbed 1% by inflation, and 0.5% by a decline in long-term bond prices. The last panel ties the two decompositions together, showing that the decline in long-
term bond prices reflects higher expected future inflation.

In sum,

- The 1.5% fiscal shock that comes with 1% unexpected inflation is buffered by an 0.5% decline in bond prices, which corresponds to 0.5% additional expected future inflation. The additional expected inflation slowly devalues long-term bonds as they come due, a loss in value marked to market in the fall in bond prices.

Figure 3: Responses to 1% inflation shock

Figure 3 adds detail to the bond pricing responses. The interest rate \( i \), bond yield \( y \), and expected return \( r^n \) all move together and persistently. Again, the graph plots the return \( r^n \), the expected return is one period earlier. The sawtooth pattern in \( r^n \) at time 3 comes from a slightly negative eigenvalue of the VAR, which is far below statistical significance. The return shock \( r^n_1 \) moves down sharply as expected subsequent returns rise. Bond prices decline when yields rise. This is the picture of a “parallel shift” in the yield curve, with no sizable change in risk premiums.

The rise in real discount rates stems from the apparent disconnect between nominal returns and inflation. Inflation is initially above nominal rates, giving a few periods
of lower real rates. When inflation declines below the persistent nominal rates, implied 
real interest rates rise on the right hand side of the graphs.

![Figure 4: Responses to 1% inflation shock](image)

Figure 4 plots the response of surplus and value of debt to the unexpected inflation 
shock. The debt to GDP ratio $v_1$ declines on impact, reflecting the offsetting forces of 
deficits, inflation, bond returns, and growth in (1). The deficit ($s_1 = -0.58\%$) and lower 
growth ($g_1 = -0.33\%$) raise the value of debt to GDP ratio. But inflation ($\pi_1 = 1\%$) and a 
negative bond return ($r_n^1 = -0.56\%$) combine to reduce it. Together these forces produce 
the impact response $v_1 = -0.65\%$. The long string of deficits and rise in expected real 
returns then raises the value of debt. But eventually surpluses rise and pay down the 
debt.

The s-shaped surplus response is a crucial lesson. It means that early debts are 
repaid, at least in part, by following surpluses. The surplus does not follow an AR(1)-like 
process. Mechanically, this pattern is a result of the VAR coefficient of surplus on lagged 
debt, and the persistence of debt. Thus, the finding is econometrically robust; it does not 
rely on a tenuous measurement of high-order surplus autocorrelations.

However, this analysis illustrates the vital importance of including debt in the VAR.
Without debt in the VAR, the surplus is positively autocorrelated throughout, and sur-
pluses never rise to pay off deficits. If we specify a theoretical model with AR(1) surplus,
we miss the crucial fact that governments do promise, and people do expect, subsequent
surpluses to pay off debts at least in part.

The 0.04 VAR coefficient of surpluses on debt and the s-shaped surplus response
do not mean that the estimates measure a passive fiscal policy. The active vs. passive fis-
cal question concerns how surpluses respond to changes in the value of debt induced by
multiple-equilibrium inflation. We cannot measure off-equilibrium responses from data
drawn from equilibrium. Suppose, for example, that surpluses are completely exoge-
nous. Suppose that when a government borrows money (negative surplus) it commits
to future positive surpluses to completely repay bondholders, an s-shaped pattern with
\[
\sum \rho^j s_{t+j} = 0,
\]
but the schedule of those surpluses is fixed and independent of inflation. That’s active fiscal policy. Yet we observe deficits, which run up debts, and then surpluses
which seem to “respond” to those debts and to pay them off. (For more on this point see
Cochrane (2020a) and Cochrane (2020b). Leeper and Li (2017) also show that regressions
of surplus on debt do not establish passive fiscal policy.)

5.1 Disclaimers

I use the words “shock,” and “response,” which have become conventional in the VAR
literature, and compactly describe the calculations for those familiar with VARs. The cal-
culations do not imply or require a causal structure, nor do they make any pretense to
measure structural shocks. A “shock” here is only an “innovation,” a movement in a vari-
able not forecast by the VAR. A “response” is a change in VAR expectations of a future
variable coincident with such a movement.

In fact, my fiscal theory interpretation offers a reverse causal story: News about
future surpluses and discount rates causes inflation to move. That news in turn reflects
news about future productivity, fiscal and monetary policy and other truly exogenous or
structural disturbances. Many VAR exercises do attempt to find an “exogenous” move-
ment in a variable by careful construction of shocks, or they attempt to measure struc-
tural shocks, and they attempt to measure responses as effects of causal shocks. I do
not.

I do not assume that people use only the VAR information set to form expecta-
tions. Since we start with an identity (1) that holds ex-post, or under agents' information sets, the identity holds using any coarser information set that includes the value of debt. The model $v_t = E(x_{t+1} | \Omega_t)$ implies $v_t = E(x_{t+1} | I_t \subset \Omega_t)$, so long as $v_t \in I_t$. But “unexpected” here means relative to the VAR information set. People may see a lot more. A decomposition using larger information sets, survey forecasts, or people's full information sets, may be different. The VAR forecasts are correct on average, but they integrate out other variables which agents may see.

Likewise, one is tempted to explore stochastic volatility, time-varying parameters, stochastically changing regimes, subsample variation, or other nonlinearities. Such additions may help to predict the variables, and they may change answers, but they do not invalidate the linear VAR. The linear VAR recovers the Wold representation. There always is a Wold linear representation of a covariance-stationary time series, even if the true process is governed by such nonlinear processes. Nonlinearities are fundamentally the same question as additional variables.

Why not present fancier specifications? The answer is, sample size and a bit of peeking. I tried to add some extra variables and lags, but explorations showed them to be economically insignificant and visibly to add overfitting noise. A quick look at the standard errors in Table 3, Figure 9, and discussed in Section 6 puts a quick damper on any desire to chop up the sample.

The value of debt in particular moves very slowly in this sample, declining from 1945 to 1974, and then rising again with a bit of interruption in the late 1990s. Interest rates and inflation also move slowly. The long run dynamics of the VAR are driven by the autocorrelation of the persistent state variables, the debt in particular, and how those slow-moving state variables forecast the other variables one period ahead. When a forecasting variable crosses its mean once every two decades, you just can’t get too fancy with subsample variation or nonlinearities and time-varying parameters.

And, most of all, when the long-run forecasts of interest are driven by slow-moving state variables and their one-step ahead forecast of other variables, adding stochastic volatility and other nonlinearities which may help (a bit) to forecast one-step ahead, does not change the long-run forecasts.

In particular, the surplus of Figure 1 seems to ask for different regimes, interestingly correlated with different inflation and growth regimes. But keep in mind that the
most important regression coefficient here is the coefficient of surplus on debt. Debt was large from 1945-1965 with positive surpluses, low in the low-surplus 1970s, and large again in the high-surplus late 1990s. The crucial finding that larger debts drive larger surpluses (on average, if not lately) is driven by this low-frequency variation.

The correlation of the level of inflation with deficits is also appetizing, but the analysis here focuses on unexpected inflation. That is a higher, business-cycle, frequency phenomenon. It links unexpected inflation to revisions in long-run forecasts.

Why use annual data? A practical reason is that the surplus and deficit data have strong seasonals, which otherwise need adjustment. Seasonally adjusting, which uses ex-post information, then forecasting, then finding implied long-run forecasts is a delicate business. Most of all, the point here is to measure long-run responses. With quarterly seasonally adjusted data one would surely want to include four lags. Raising an annual VAR to many powers to calculate long-run forecasts is already fraught. Raising a quarterly four-lag VAR to four-times higher powers is even more fraught. Higher frequency data does not always help to make long-run forecasts. Minute by minute data rather than daily or even yearly data does not improve estimates of the speed of climate change.

Sometimes simpler is not just easier, it’s better, more robust, and clearer too.

### 5.2 Recession, or aggregate demand shocks

We can use the same procedure to understand the fiscal underpinnings of other shocks. For any interesting $\varepsilon_1$, we can compute impulse-response functions, and thereby the terms of the inflation decompositions. I show in Appendix C that we can consider these calculations as a decomposition of the covariance of unexpected inflation with the shock $\varepsilon_1$, rather the decomposition of the variance of unexpected inflation.

I start with a shock that moves inflation and growth in the same direction. The inflation shock in Figure 2 is stagflationary, in that growth falls when inflation rises. Unexpected inflation is, in this sample, negatively correlated with unexpected consumption (and also GDP) growth. The stagflationary episodes in the 1970s outweigh the simple Phillips curve episodes.

However, it is interesting to examine the response to disinflations which come in recessions, and inflations that come in expansions, following a conventional Phillips
curve. Such events are common, as in the recession following the 2008 financial crisis. But they pose a fiscal puzzle. In such a recession, deficits soar, yet inflation declines. How is this possible? As I outlined in the introduction, future surpluses or lower discount rates could give that deflationary force, needed whether fiscal policy is active or passive. Can we see these effects in the data, and which one is it?

To answer that question, I simply specify $\varepsilon_1^\pi = -1$, $\varepsilon_1^g = -1$. The model is linear, so the sign doesn’t matter, but the story is clearer for a recession. To give it a name, I call this a “recession shock” in the Tables. We could also call it an “aggregate demand” shock, because output and inflation move in the same direction, as opposed to “aggregate supply” shocks which move output and inflation in opposite directions. These are just memorable labels, with no pretense to identify structural shocks of any model.

Again, we want shocks to other variables to have whatever value they have, on average, conditional on the inflation and output shock. To initialize the other shocks of the VAR, then, I run a multiple regression

$$
\varepsilon_{t+1}^z = b_{z,\pi} \varepsilon_{t+1}^\pi + b_{z,g} \varepsilon_{t+1}^g + \eta_{t+1}^z
$$

for each variable $z$. I fill in the other shocks at time 1 from their predicted variables given $\varepsilon_1^\pi = -1$ and $\varepsilon_1^g = -1$, i.e. I start the VAR at

$$
\varepsilon_1 = - \left[ b_{\pi,\pi} + b_{\pi,g} \quad \varepsilon_1^\pi = 1 \quad \varepsilon_1^g = 1 \quad b_{s,\pi} + b_{s,g} \quad \ldots \right]'.
$$

Figure 5 presents responses to this shock, and Table 2 collects the inflation decomposition elements in the “Recession” rows.

Both inflation $\pi$ and growth $g$ responses start at -1%, by construction. Inflation is once again persistent, with a $\omega$-weighted sum of current and expected future inflation equal to -2.36%. Consumption growth $g$ returns rapidly, but does not much overshoot zero, so the level of consumption does not recover much at all. Consumption is roughly a random walk in response to this shock. The nominal interest rate $i$ falls in the recession, and recovers a bit more slowly than inflation. Long-term bond yields $y$ also fall, but not as much as the short-term rate, for about 4 years. We see here the standard upward-sloping yield curve of a recession. The expected bond return follows the long-term yield. The persistent fall in expected return corresponds to a large positive ex-post bond return...
Figure 5: Responses to a recession or aggregate demand shock, $\varepsilon_1^\pi = \varepsilon_1^q = -1$. 
The recession includes a large deficit $s$, which continues for three years. In short, we see a standard picture of a recession similar to 2008-2009.

The large deficits in recessions puzzle a simplistic interpretation of the fiscal theory. Why do we not see inflation at times with such large deficits? Surpluses subsequently turn positive, paying down some of the debt. But the total surplus is still -1.15. Left to their own devices, surpluses would produce a 1.15% inflation during the recession. A potential story that disinflation results from future surpluses more than matching today’s deficits is wrong. Growth also adds an inflationary force. The decline in consumption is essentially permanent, and would lead on its own to another 1.46% inflation.

Discount rates are the central story for disinflation in recessions. After one period, expected real returns $r - g$ decline persistently, accounting for 4.96% cumulative deflation.

In terms of the unexpected inflation accounting in the second and third panels of Table 2, again surpluses and growth provide a total $1.15\% + 1.46\% = 2.61\%$ fiscal loosening, an inflationary force. The unweighted sum of future discount rates provides a 4.79% deflationary force, for an overall fiscal shock of 2.19% deflation. Of that, 1% results in unexpected deflation and 1.19% is soaked up by lower long-term bond prices. In the bottom panel, that 1.19% overwhelmingly represents lower expected inflation, essentially marking it to market for a one-period accounting.

In sum, rounding the numbers,

- **Disinflation in a recession is driven by a lower discount rate, reflected in lower interest rates and bond yields.** For each 1% disinflation and growth shock, the expected return on bonds falls so much that the present value of debt rises by nearly 5%. This discount rate shock overcomes a 1.1% inflationary shock coming from persistent deficits, and 1.5% inflationary shock coming from lower growth. The overall fiscal shock is 1.6%, with the extra 0.6% spread to future disinflation and soaked up by long-term bond prices.

The opposite conclusions hold of inflationary shocks in a boom. Discount rate variation gives us a fiscal Phillips curve, accounting for the otherwise puzzling correlation of deficits with disinflation and surpluses with inflation.

The relative magnitudes of the inflation and growth shocks that I used to define a “recession” or “aggregate demand shock” are (obviously) arbitrary. Growth fell about
twice as much as inflation in 2008, but inflation fell a bit more than growth in 1982. Other recessions have been stagflationary.

To produce a better number one must write a model and find an identification in the data to separate “supply” or “stagflationary” Phillips-curve shift shocks from “demand” or “movement along the Phillips curve” shocks, and one must thereby define precisely just what kind of events we seek to evaluate. Rather than belabor the point with such a calculation, or fill the paper with multiple graphs, I choose a simple and transparent value of 1% less growth and 1% less inflation. The calculations report correctly “How do expectations of other variables change if we observe that inflation and growth both decline by 1%?” The only quibble is whether some other combination of inflation and growth shocks might be more interesting.

## 5.3 Surplus and discount rate shocks

We have studied what happens to surpluses and to discount rates given that we see unexpected inflation. What happens to inflation if we see changes in surpluses or discount rates? These are not the same questions. An inflation shock may come, on average, with a discount rate shock, but a discount rate shock may not come on average with inflation.

I calculate here how the variables in the VAR react to an unexpected change in current and expected future primary surpluses including growth, \( \Delta E_1 \sum_{j=0}^{\infty} (s_{t+j} + g_{t+j}) = 1 \), and other shocks to the VAR take their average values given this innovation. I call this a “surplus shock.” The results are almost the same with or without the growth term in the shock definition. Then I calculate how the variables in the VAR react to an unexpected change in discount rates, \( \Delta E_1 \sum_{j=1}^{\infty} (1 - \omega^j)(r_{t+1}^{n_t} - \pi_{t+1}) = 1 \), again letting all other variables take their average values given this innovation. I call this a “discount rate shock.” These shocks take a step in the direction of monetary and fiscal policy shocks, as studied in Cochrane (2020a), but have many orthogonalization and identification steps to go before they can take on that mantle. For now, they represent the effects of (or, more carefully, the correlates of) fiscal and interest rate changes, no matter how the latter are brought about.
The response of the sum of future surpluses and growth to a shock $\varepsilon_1$ is

$$\Delta E_1 \sum_{j=0}^{\infty} (s_{1+j} + g_{t+j}) = (a_s + a_g)' (I - A)^{-1} \varepsilon_1.$$ 

To calculate how VAR shocks respond to a surplus shock, I run for each variable $z$ a regression

$$\varepsilon_{z, t+1} = b_z \left[(a_s + a_g)' (I - A)^{-1} \varepsilon_{t+1} \right] + \eta_{t+1}^z \tag{7}$$

where $a_z$ pulls variable $z$ from the VAR, $a_z' x_t = z_t$. Then, I start the surplus-shock response function at

$$\varepsilon_1 = - \left[ b_{r^n} \ b_g \ b_{\pi} \ ... \right]' .$$

I plot a negative surplus shock, i.e. a deficit shock, as that sign tells an easier story.

Similarly, to calculate responses to a discount-rate shock, I run

$$\varepsilon_{z, t+1} = b_z \left\{ (a_r - a_{\pi})' \left[ A(I - A)^{-1} - \omega A(I - \omega A)^{-1} \right] \varepsilon_{t+1} \right\} + \eta_{t+1}^z .$$

I start the discount-rate response function with the negative of these regression coefficients as well, capturing the response to a discount rate decline.

Figure 6 presents the responses to the deficit shock, and Figure 7 presents the responses to the discount rate shock. Table 2 collects relevant contributions to the inflation decompositions.

The sum of surplus and growth responses to the deficit shock are $-0.66 - 0.34 = -1.00$ by construction. Surpluses still have an s-shaped pattern, but the initial deficits are not matched by subsequent surpluses.

This decline in surpluses and growth has essentially no effect on inflation. Starting in year 2, inflation declines – the “wrong” direction given deficits and lower growth – by less then a tenth of a percent, and the overall weighted sum of inflation declines by a tenth of a percent. Why is there no inflation? Because discount rates also decline, with a weighted sum of 1.10%, almost exactly matching the surplus decline. The lower panel of Figure 6 adds insight. We see a sharp and persistent decline in the interest rate, long-term bond yield, and expected bond return, along with deficits and the growth decline.

This figure captures the event of a widening deficit, accompanied by a decline in growth and interest rates, i. e. a recession. These deficits are on average not directly
Figure 6: Responses to a surplus and growth shock, $\Delta E_1 \sum_{j=0}^{\infty} (s_{1+j} + g_{1+j}) = -1$. 
Figure 7: Responses to a discount-rate shock \( \Delta E_1 \sum_{j=1}^{\infty} (1 - \omega^j) \left( r_{1+j}^n - \pi_{1+j} \right) = 1. \)
repaid by subsequent surpluses or growth. Instead, real interest rates decline persistently
in the recession and its aftermath. This decline in real returns essentially pays for the
deficits. Ex-post, a low real return brings the value of debt back rather than larger taxes
or lower spending. There is, on average, very little inflation or deflation. The opposite
sign occurs for positive shocks.

The response to the discount rate shock in Figure 7 is, surprisingly, almost exactly
the same. The weighted discount rate response ($\sum (1 - \omega^j)$) is -1.00 here by construction.
This discount rate decline should be deflationary, and it is – but the disinflation peaks at
-0.1% and the weighted sum is only -0.18%. A sharp growth and surplus decline accompanies
this discount rate decline, with a pattern almost exactly the same as we found
from the growth and surplus shock. In the bottom panel, the expected return decline
comes with a decline in interest rates and bond yields, as we would expect.

Clearly, the surplus + growth shock and the expected return shock have isolated
essentially the same events – recessions in which growth falls, deficits rise, interest rates
fall, and, on average in this sample, inflation doesn’t move much, and the converse pattern of expansions. The correlation of the surplus+growth and discount rate shocks is 0.96. To continue the coffee story from the introduction, if you sample cups of coffee at
Starbucks that have a lot of sugar in them, they are likely to have a lot of coffee as well,
and if you sample cups with a lot of coffee, they are likely to have a lot of sugar in them.

The responses to a one-period surplus shock, $\Delta E_1 s_1 = 1$, a pure growth shock
$\Delta E_1 g_1 = 1$ and a one-period discount rate shock $\Delta E_1 r^2_n = 1$ are all quite similar as well.

The fiscal roots of the absence of inflation, in the end, characterize these business-
cycle movements in the data. This sort of event, apparently common in the data, is not
much studied in macroeconomic public finance. We study deficits in recessions, primary
surpluses in booms (at least before the most recent one), and wonder whether surpluses
pay off the deficits, or whether state-contingent default via inflation plays a part. We do
not often tell a story that the higher deficits in recessions are resolved by lower interest
rates, bringing back the debt/GDP ratio, and vice versa – a temporary $r < g$ effect – without surpluses, and that this can be a part of a responsible public policy that does not rely
on unexpected inflation to devalue debt.

In sum,

- Surplus and discount rate shocks paint the same picture: Deficits are mostly not
repaid by subsequent growth or surpluses, but do not produce inflation. Instead, deficits come with periods of extended low expected returns. Discount rate declines come with offsetting deficits and do not produce much deflation. Discount rate and deficit shocks move together. Discount rate variation explains why deficits, not repaid by future surpluses, do not result in inflation.

5.4 A surplus shock without accommodation

The fact that interest rates move in opposition to the surplus shock is obviously key to the noninflationary result. What if there is a surplus shock and the Federal Reserve does not accommodate the shock, or its economic correlates, with the prominent interest decline seen in Figure 6? To answer this question, I modify the surplus+growth shock so that the short-term interest rate remains constant for two years. I now run

\[
\varepsilon_{t+1}^z = b_{z,s} \left[ (a_s + a_g)' (I - A)^{-1} \varepsilon_{t+1} \right] + b_{z,i0} \varepsilon_{t+1}^i + b_{z,i1} \left( a_i' A \varepsilon_{t+1} \right) + \eta_{t+1}^z.
\]

The last term before the error is the expected interest rate one year forward. Then, I initialize the VAR at

\[
\varepsilon_1 = \left[ b_{r,s}, b_{g,s}, b_{\pi,s}, \ldots \right]'.
\]

Figure 8 presents the responses, and Table 2 collects the terms of the identities. Starting in the bottom panel of Figure 8, verify that the interest rate now stays constant for two years, by construction. This behavior contrasts with the strong interest rate decline in the bottom panel of Figure 6. Except for the one-period expected return decline in year two, the long-term bond yields and expected returns follow the interest rate. All decline eventually.

Turning to the upper panel, the sum of surplus (-0.52) and growth (-0.48) shocks remains -1.00% by construction. Deficits are initially much larger than 0.52%, but much of this immediate deficit is repaid by higher long-term surpluses, so in the end the fiscal shock is split equally between surpluses and growth. The discount rate term is now reduced to 0.62% - 0.67%, however, so the surplus shock now produces 0.36% immediate and 0.38% weighted sum inflation.

In sum, without the interest rate response, the fiscal shock does result in unexpected inflation. We see here a parallel of the theoretical analysis that central bank ac-
Figure 8: Responses to a surplus and growth shock with no interest rate movement for two years, $\Delta E_1 \sum_{j=0}^{\infty} (s_{1+j} + g_{1+j}) = 1$, $\Delta E_1 i_1 = 0$, $\Delta E_1 i_2 = 0$. 

\[ \pi, \Sigma = 0.38 \]
\[ g, \Sigma = -0.48 \]
\[ r^n - \pi, \Sigma = -0.67, \Sigma 1 - \omega = -0.62 \]
\[ s, \Sigma = -0.52 \]
commodation of shocks, via the interest rate target, smooth forward and thereby reduce unexpected inflation, even though the bank cannot control fiscal policy.

(Holding the interest rate constant for one period produces a similar though slightly weaker result. Interest rates drift down after the one period, so the discount rate effect is slightly stronger and inflation slightly less.)

Will the real recession please stand up? How do we have by one calculation recessions with disinflations, and by another recessions with no change in inflation? Alas, our macroeconomy is not a one-factor model, with all time-series moving in lockstep. Different (true, structural) shocks dominate different events. The recessions of the 1970s featured stagflation, those since 1990 did not. All recessions are not the same. Sometimes inflation falls, sometimes it doesn't. I have examined five, hopefully interesting, slices of the full covariance matrix of shocks. They are different.

6 Standard errors

I have delayed a discussion of standard errors because there is nothing important to test. Identities are identities. If $x = y + z$ and $x$ moves, $y$ or $z$ must move, and all we can do is to measure which one moves. Standard errors only serve to give us a sense of how accurate the measurement is. In addition, unlike the case in asset pricing, no important economic hypothesis rests on whether one of surpluses or discount rates do not move. Asset pricing finds the hypothesis that expected returns are constant over time interesting to test.

I run a Monte Carlo to evaluate sampling distributions. Appendix D gives details. Most of the interesting statistics – variance decompositions, impulse-response functions, $(I - A)^{-1}$, etc. – are nonlinear functions of the underlying data, and the near-unit root in value $v_t$ also induces non-normal distributions. For these reasons, I largely characterize the sampling distribution by the interquartile range – the 25% and 75% points of the sampling distribution.

Table 3 collects the sampling quantiles for the variance decompositions of Table 2. Figure 9 presents the main components of the impulse-response function relevant to the inflation variance decomposition. The bands are 25% and 75% points of the sampling distribution, the dashed line is the median, and the solid line is the estimate.

Start with the “Inflation” shock in Table 3. In the second panel, inflation quantiles
Figure 9: Distribution of the responses to an inflation shock. The bands are 25% and 75% points of the sampling distribution, the dashed line is the median, and the solid line is the estimate.

are 1.00 because the shock is defined as a 1% movement in inflation in every sample. The 1.59% weighted sum of inflation has 1.38% to 1.64% quantiles in the top panel. The -0.06% sum of future surpluses has quartiles -0.69% to 0.23%. The -0.49% sum of growth rates has quantiles -0.72% to -0.22%. The 1.04% (weighted) and 1.00% (unweighted) discount rate term has quantiles 0.16% to 1.46% and 0.16% to 1.55%. That discount rates matter is a pretty solid conclusion, but deficits may contribute more to unexpected inflation than the point estimate suggests.

There are several sources of this rather large sampling variation. First, the shocks are large. As shown in Table 1, the surplus innovation has a 4.75 percentage point standard deviation, and value 6.55 percentage points, compared to 1.12 percentage points for inflation. Our friend $\sigma/\sqrt{T}$ starts off badly.

Second, the shocks are imperfectly correlated. This matters, because in each case I find movements in other variables contemporaneous with the shock of interest by run-
ning a regression of the other shocks on the shock of interest. The sampling uncertainty of this orthogonalization adds to that of the VAR coefficient estimates. We see a correspondingly wide band around the initial surplus and growth responses in Figure 9. There is hope in this observation, however. Higher frequency data can better identify shock correlations, at the cost that one must model the strong seasonal in primary surpluses. Moreover, other shock identifications may have better measured correlations.

Third, we measure sums of future surpluses and discount rates. The value of the debt \( v_t \) is the main long-run state variable, and uncertainty about its evolution adds to the uncertainty about the sum of surpluses. The coefficient of value \( v_t \) on its own lag is 0.98 in Table 1, so small variations in that value lead to large variation in \( (I - A)^{-1} \) sums. Appendix E shows that the last two sources of variation contribute about equally.

Table 3 also presents 25% and 75% quantiles for the recession, surplus and discount rate shocks of Table 2. The -1.15% total surplus response to a recession shock has quantiles -1.28% to 0.49%, spanning zero, while the -4.96% and -4.79% discount rate response has quantiles from -4.84% to -2.43% and -4.84% to -2.35%. The conclusion that discount rate variation is a central part of the story for understanding aggregate-demand inflation is fairly solid. The small inflation and offsetting surplus and discount rate responses to surplus and discount rate shocks are similarly measured.

It would be nice if the elements of the identities were more precisely measured. But there is nothing one can do within the framework of this VAR to improve on them, so it’s worth examining point estimates while awaiting more data or other approaches such as model-based estimates that impose prior structure. The rather large sampling variation should, however, discourage one from the inevitable temptation to split up the sample or add complexity to the specification.

### 7 Concluding comments

This analysis evidently just scratches the surface. One can apply these decompositions to any VAR, or to the impulse-responses of any theoretical model. Such calculations beckon.

In particular, it is interesting to apply the inflation decompositions to model predictions or empirical estimates of well-identified monetary and fiscal policy shocks. Sup-
pose, for example, that monetary policy follows $i_t = \phi_{i,\pi} \pi_t + \phi_{i,x} x_t + u_{i,t}$, and fiscal policy follows $s_t = \phi_{s,\pi} \pi_t + \phi_{s,x} x_t + u_{s,t}$, with persistent disturbances $u_{i,t}$ and $u_{s,t}$, and in particular an s-shaped moving average of $u_{s,t}$, reflecting partial repayment promises. With such a specification, it is interesting to compute responses and inflation decompositions to orthogonal shocks to monetary and fiscal policy disturbances $u_{i,t}$ and $u_{s,t}$. The Federal Reserve cannot directly control fiscal policy, so fiscal theory of monetary policy models suggest that it is interesting to shock monetary policy $u_{i,t}$, while holding fiscal policy in some sense fixed. Yet we likely want to allow the systematic part of fiscal policy to respond to economic events, reflecting the rise in tax revenues with income and inflation, and the automatic stabilizers and predictable stimulus spending in recessions, as modeled by the $\phi_{s,\pi} \pi_t + \phi_{s,x} x_t$ part of the above fiscal policy rule. That, I think, is the kind of “what happens if we raise interest rates?” question a Federal Reserve official might have in mind. Cochrane (2020a) presents such calculations from a simple fiscal theory of monetary policy model.

By contrast, New-Keynesian models imply a fiscal policy shock $u_{s,t}$ that moves passively with the monetary policy shock to $u_{i,t}$. Responses to such correlated disturbances are interesting too, but a different question. Here, we might focus on the neglected fiscal side of new-Keynesian models to examine what the “passive” fiscal policy is.

It may also be interesting to know what happens without the systematic fiscal response, which setting $\phi_{s,\pi} = \phi_{s,x} = 0$ can answer. Such a calculation illuminates the economic operation of monetary policy, not just as a leader of a customary fiscal response. The contrast tells us how important monetary-fiscal policy coordination is. Likewise, it is interesting to know what is the response to fiscal shocks, changes in $u_{s,t}$, assuming the central bank follows its customary rule represented by $\phi$ terms and not holding interest rates constant as in my last Figure.

Alas, making such calculations in data require one to solve the formidable identification problems of estimating the $\phi$ coefficients, given that the right hand variables react to the disturbances. The state of the art for identifying monetary policy disturbances and measuring the reaction function goes well beyond the simple recursive and long-run strategies available in the atheoretical annual VAR here, to include highly detailed identification assumptions, high frequency data, narrative approaches, and other devices, and
the literature still does not offer a robustly successful result on which one can build (Ramey (2016), Cochrane (2011a)). And even this voluminous literature has not started to think about how we can identify monetary policy shocks that are orthogonal to fiscal policy shocks. In the data, monetary and fiscal authorities are likely to respond to the same underlying shocks, as we found a strong correlation between interest rate or discount rate shocks and surplus shocks here. Teasing out monetary policy shocks that are orthogonal to fiscal policy shocks, as well as all the other desired orthogonality, requires some thought. I attempted monetary and fiscal policy shocks by recursive identification in this data, but one-year interest-rate, inflation, and growth shocks are all highly correlated. Assuming all of that correlation flows from interest rates to inflation and growth results in positive effects of interest rates on inflation and growth. Assuming all correlation reflects rule-like responses of interest rates to inflation and growth eliminates the unexpected inflation response we wish to measure. Obviously, reality lies in between.

Additional measurements beckon. Quarterly or monthly data are attractive, offering potentially better measurement of correlations and shock orthogonalization but requiring us to model the strong seasonality in surpluses, and not to let seasonal adjustment, which uses ex-post data, influence forecasts. Debt data go back centuries, allowing and requiring us to think what is the same and different across different periods of history. Inflation through wars and under the gold standard may well have different fiscal foundations than in the postwar environment. Appendix G finds quite different behavior in 1930-1947, though that sample is dominated by a few influential data points and does not offer by itself enough evidence to measure a different regime. A narrative counterpart, especially for big episodes such as the 1970s and 1980s, awaits. Different countries under different monetary and exchange rate regimes and different fiscal constraints will behave differently. A parallel investigation of exchange rates beckons, following Jiang (2019a), Jiang (2019b). One could define shocks in many additional interesting ways. The treatment of debt can be refined in many ways. In particular, the maturity structure is not geometric, and varies over time, and active management of the maturity structure is in theory an interesting policy for stabilizing inflation, or creating it (Cochrane (2001)).

I omitted analysis of the remaining shocks in the VAR. A shock to any other variable, orthogonal to the inflation shock, can move all of the other terms of the inflation identities. Such movements must offset: In (5), if a shock does not move the in-
flation term, but does move the sum of future surpluses, then it must also move the sum of growth rates or discount rates. These additional effects are large. The variation in $\Delta E_1 \sum_{j=0}^\infty s_1^j$ when other shocks move is large; the corresponding movement in the discount rate term is also large, and the two movements are negatively correlated. We get a hint of that behavior in the surplus+growth and discount rate shock responses. I do not pursue this question because it is much more interesting if one can give some structural or economic interpretation to the shocks to other variables.

Perhaps most of all, linking these theory-free characterizations to explicit fiscal theory of monetary policy models such as Cochrane (2020a), or at least to explicit models of discount rates and long-term debt management, is an obviously important step. However, such models need to be elaborated to the point that they can match data, which requires considerable complication of model elements, as has been the case in the empirical new-Keynesian DSGE literature, and to surmount difficult identification and estimation challenges of all their structural parameters. It's important, but not easy.
References


Online Appendix to “The Fiscal Roots of Inflation”

A Derivation of the linearized identities

In this appendix I derive the linearized identities (1), (2), and (3),

\[ v_{t+1} = v_t + r_n^{t+1} - \pi_{t+1} - g_{t+1} - s_{t+1}, \]  

and

\[ \Delta E_{t+1} \pi_{t+1} - \Delta E_{t+1} \left( r_n^{t+1} - g_{t+1} \right) 
= - \sum_{j=0}^{\infty} \Delta E_{t+1} s_{t+1+j} + \sum_{j=1}^{\infty} \Delta E_{t+1} \left( r_n^{t+1+j} - \pi_{t+1+j} - g_{t+1+j} \right), \]  

I also define the variables more carefully.

The symbols are as follows:

\[ V_t = M_t + \sum_{j=0}^{\infty} Q_t^{(t+j+1)} B_t^{(t+1+j)} \]

is the nominal end-of-period market value of debt, where \( M_t \) is non-interest-bearing money, \( B_t^{(t+j)} \) is zero-coupon nominal debt outstanding at the end of period \( t \) and due at the beginning of period \( t + j \), and \( Q_t^{(t+j)} \) is the time \( t \) price of that bond, with \( Q_t^{(t)} = 1 \). Taking logs,

\[ v_t \equiv \log \left( \frac{V_t}{Y_t P_t} \right) \]

is log market value of the debt divided by GDP, where \( P_t \) is the price level and \( Y_t \) is real GDP or another stationarity-inducing divisor such as consumption, potential GDP, etc. I use consumption times the average GDP to consumption ratio in the empirical work, but I will call \( Y \) and ratios to \( Y \) “GDP” for brevity. The quantity

\[ R_t^{n+1} \equiv \frac{M_t + \sum_{j=1}^{\infty} Q_t^{(t+j)} B_t^{(t+1+j)}}{M_t + \sum_{j=1}^{\infty} Q_t^{(t+j)} B_t^{(t+1+j)}} \]
is the nominal return on the portfolio of government debt, i.e. how the change in prices from the end of \( t \) to the beginning of \( t + 1 \) affects the value of debt held between periods. The quantity

\[
r_{t+1}^n \equiv \log(R_{t+1}^n)
\]

is the log nominal return on that portfolio.

\[
\pi_t \equiv \log \left( \frac{P_t}{P_{t-1}} \right), \quad g_t \equiv \log \left( \frac{Y_t}{Y_{t-1}} \right)
\]

are log inflation and GDP growth rate.

We can accommodate explicit default, so the formulas can also apply to countries that borrow in foreign currency such as the members of the Euro. An explicit default is a reduction in the nominal quantity of debt between periods. The \( B_t^{(t+j)} \) in the numerator of (10) represents the post-default number of bonds outstanding, i.e. at the beginning of period \( t + 1 \), while the \( B_t^{(t+j)} \) in the denominator represents the pre-default number of bonds outstanding, i.e. at the end of period \( t \). A partial default then shows up as a low return. To handle default one would, of course, add notation distinguishing the pre- and post- default quantity of debt in the definition of return.

We start with the nonlinear flow identity,

\[
M_t + \sum_{j=1}^{\infty} Q_{t+1}^{(t+j)} B_{t}^{(t+j)} = P_{t+1} s_{t+1} + M_{t+1} + \sum_{j=1}^{\infty} Q_{t+1}^{(t+1+j)} B_{t+1}^{(t+1+j)}.
\]

(11)

Here, \( s_{t+1} \) denotes the real primary (not including interest payments) surplus or deficit. At the beginning of period \( t + 1 \), money \( M_t \) and bonds \( B_t^{(t+1+j)} \) are outstanding. Money \( M_{t+1} \) at the end of period \( t \) and beginning of period \( t + 1 \) then equals money \( M_t \), money printed up to redeem bonds \( B_t^{(t+1)} \), less money soaked up by a primary surplus \( P_{t+1} s_{t+1} \), or conversely printed to finance a primary deficit, and less money soaked up by net new bond sales, or printed to finance long-term bond purchases, \( \sum_{j=1}^{\infty} Q_{t+1}^{t+1+j} (B_{t+1}^{(t+1+j)} - B_t^{(t+1+j)}) \).

Using the definition of return, (11) becomes

\[
\left( M_t + \sum_{j=1}^{\infty} Q_{t}^{(t+j)} B_{t}^{(t+j)} \right) R_{t+1}^n = P_{t+1} s_{t+1} + \left( M_{t+1} + \sum_{j=1}^{\infty} Q_{t+1}^{(t+1+j)} B_{t+1}^{(t+1+j)} \right).
\]
or,
\[ V_t R^n_{t+1} = P_{t+1} s_{t+1} + V_{t+1}. \]

The nominal value of government debt is increased by the nominal rate of return, and decreased by primary surpluses. This seems easy. The algebra all comes from properly defining the return on the portfolio of government debt.

Expressing the result as ratios to GDP, we have a flow identity
\[
\frac{V_t}{P_t Y_t} \times \frac{R^n_{t+1}}{G_{t+1}} \frac{P_t}{P_{t+1}} = \frac{s_{t+1}}{Y_{t+1}} + \frac{V_{t+1}}{P_{t+1} Y_{t+1}},
\]
where \( G_{t+1} \equiv Y_{t+1}/Y_t \).

We can iterate this flow identity (12) forward to express the nonlinear government debt valuation identity as
\[
\frac{V_t}{P_t Y_t} = \sum_{j=1}^{\infty} \prod_{k=1}^{j} \frac{1}{R^n_{t+k}} \left( \frac{P_t}{P_{t+1}} \right) \frac{s_{t+j}}{Y_{t+j}},
\]
where \( \Pi_{t+1} \equiv P_{t+1}/P_t \). The market value of government debt at the end of period \( t \), as a fraction of GDP, equals the present value of primary surplus to GDP ratios, discounted at the government debt rate of return less the GDP growth rate. I assume here that the right hand side converges. Otherwise, keep the limiting debt term or iterate a finite number of periods. Given that the value of debt is finite, the sums converge iff the terminal condition converges. And, in the end we only need convergence in expectation. Roughly speaking, it is sufficient that debt to GDP is bounded and the proper nonlinear expected version of \( r > g \) holds.

The nonlinear present value identities (12) and (13) are cumbersome. I linearize the flow equation (12) and then iterate forward to obtain a linearized version of (13). Taking logs of (12),
\[
v_t + r^n_{t+1} - \pi_{t+1} - g_{t+1} = \log \left( \frac{s_{t+1}}{Y_{t+1}} + \frac{V_{t+1}}{P_{t+1} Y_{t+1}} \right)
\]
(14)

I linearize this equation in the level of the surplus, not its log as one conventionally does in asset pricing, since the surplus is often negative. To linearize in terms of the
surplus/GDP ratio, Taylor expand the last term,

\[ v_t + r^n_{t+1} - \pi_{t+1} - g_{t+1} = \log(e^v + sy) + \frac{e^v}{e^v + sy} (v_{t+1} - v) + \frac{1}{e^v + sy} (sy_{t+1} - sy) \]

where

\[ sy_{t+1} \equiv \frac{sph_{t+1}}{Y_{t+1}} \tag{15} \]

denotes the surplus to GDP ratio, and variables without subscripts denote a steady state of (14). With \( r \equiv r^n - \pi \),

\[ r - g = \log \frac{e^v + sy}{e^v}. \]

Then,

\[ v_t + r^n_{t+1} - \pi_{t+1} - g_{t+1} = \left[ \log(e^v + sy) - \frac{e^v}{e^v + sy} \left( v + \frac{sy}{e^v} \right) \right] + \frac{e^v}{e^v + sy} v_{t+1} + \frac{e^v}{e^v + sy} \frac{sy_{t+1}}{e^v} \]

\[ v_t + r^n_{t+1} - \pi_{t+1} - g_{t+1} = \left[ v + r - g - \frac{e^v}{e^v + sy} \left( v + \frac{e^v + sy}{e^v} - 1 \right) \right] + \rho v_{t+1} + \rho \frac{sy_{t+1}}{e^v} \]

\[ v_t + r^n_{t+1} - \pi_{t+1} - g_{t+1} = [r - g + (1 - \rho) (v - 1)] + \rho v_{t+1} + \rho \frac{sy_{t+1}}{e^v} \tag{16} \]

where

\[ \rho \equiv e^{-(r-g)}. \tag{17} \]

Suppressing the small constant, and thus interpreting variables as deviations from means, the linearized flow identity is

\[ v_t + r^n_{t+1} - \pi_{t+1} - g_{t+1} = \rho \frac{sy_{t+1}}{e^v} + \rho v_{t+1}. \tag{18} \]

Iterating forward, the present value identity is

\[ v_t = \sum_{j=1}^{T} \rho^{j-1} \left[ \rho \frac{sy_{t+j}}{e^v} - (r^n_{t+j} - \pi_{t+j} - g_{t+j}) \right] + \rho^T v_T. \tag{19} \]

If we linearize around \( r - g = 0 \), then the constant in (18) is zero \( (sy = 0) \), and we obtain the linearized flow and present value identities (8) and (9), with the symbol \( s_t \) representing \( sy_t / e^v \). There is nothing wrong with expanding about \( r = g \). The point of expansion need not be the sample mean.
To approximate in terms of the surplus to value ratio, write (14) as

\[ v_t + r^n_{t+1} - \pi_{t+1} - g_{t+1} = \log \left( \frac{V_t}{P_t Y_t} \frac{sp_{t+1}}{Y_{t+1}} + \frac{V_{t+1}}{P_{t+1} Y_{t+1}} \right) \]

\[ r^n_{t+1} - \pi_{t+1} - g_{t+1} = \log \left( \frac{V_t}{P_t Y_t} \frac{sp_{t+1}}{Y_{t+1}} + \frac{V_{t+1}}{P_{t+1} Y_{t+1}} \right) \]

\[ r^n_{t+1} - \pi_{t+1} - g_{t+1} = \log (sv_{t+1} + e^{v_{t+1} - v_t}) \]

At a steady state

\[ r - g = \log (1 + sv) \]

(20)

\[ e^{r-g} = 1 + sv. \]

Taylor expanding around a steady state,

\[ r^n_{t+1} - \pi_{t+1} - g_{t+1} = \log (1 + sv) + \frac{1}{(1 + sv)} (sv_{t+1} - sv + v_{t+1} - v_t) \]

\[ v_t + (1 + sv) \left( r^n_{t+1} - \pi_{t+1} - g_{t+1} \right) = [(1 + sv) \log (1 + sv) - sv] + sv_{t+1} + v_{t+1} \] (21)

The linearized flow identity (8) follows, with the symbol \( s_t \) representing the surplus to value ratio \( s_t = sv_t \), if we suppress the constant, using deviations from means in the analysis, or if we use \( r = g \) or \( sv = 0 \), as a point of expansion.

The linearizations in terms of the surplus to value ratio \( sv_t \) are more accurate. The units of the flow identities (8), (18) are rates of return. Dividing the surplus by the previous period’s value gives a better approximation to the growth in value, when the value of debt is far from the steady state.

A constant ratio of surplus to market value of debt for any price level path leads to a passive fiscal policy: An unexpected deflation raises the real value of debt. If surpluses always rise in response, they validate the lower price level. Thus, although on the equilibrium path one can describe dynamics via either linearization, if one wants to think about how fiscal-theory equilibria are formed, it is better to describe a surplus that does not react to price level changes, so only one value \( v_t \) emerges, as is the case in (19). For such purposes, the surplus to GDP definition is appropriate, as well as adopting a linearization point \( r > g \) and \( \rho < 1 \). It’s also better to use the nonlinear versions of the identities
for determinacy issues. The analysis of this paper is about what happens in equilibrium, and does not require an active-fiscal assumption, so the difference is irrelevant here.

I infer the surplus from the linearized flow identity (8) so which concept the surplus corresponds to makes no difference to the analysis. The difference is only the accuracy of approximation, how close the surplus recovered from the linearized flow identity corresponds to a surplus recovered from the nonlinear exact identity (14).

B Linearizing the bond return formula

Here I derive the linearized identity

\[ r^n_{t+1} \approx \omega q_{t+1} - q_t, \]

which leads to (4),

\[ \Delta E_{t+1} r^n_{t+1} = - \sum_{j=1}^{\infty} \omega^j \Delta E_{t+1} \left[ (r^n_{t+1+j} - \pi_{t+1+j}) + \pi_{t+1+j} \right]. \]

I also derive expectations-hypothesis bond-pricing equations.

\[ E_t r^n_{t+1} = i_t \]
\[ \omega E_t q_{t+1} - q_t = i_t. \]

These equations are used in the sticky-price model Cochrane (2020a).

Denote the maturity structure by

\[ \omega_{j,t} \equiv \frac{B^{(t+j)}_t}{B^{(t+1)}_t} \]

and \( B_t \equiv B^{(t+1)}_t \). Then the end of period \( t \) nominal market value of debt is

\[ \sum_{j=1}^{\infty} B^{(t+j)}_t Q^{(t+j)}_t = B_t \sum_{j=1}^{\infty} \omega_{j,t} Q^{(t+j)}_t. \]

(I ignore money to keep the formulas simple.) Define the price of the government debt
portfolio

\[ Q_t = \sum_{j=1}^{\infty} \omega_{j,t} Q_t^{(t+j)}. \]

The return on the government debt portfolio is then

\[ R_{t+1}^n = \frac{\sum_{j=1}^{\infty} B_t^{(t+j)} Q_t^{(t+j)} - \sum_{j=1}^{\infty} \omega_{j,t} Q_t^{(t+j)} Q_t^{(t+1)} + \sum_{j=1}^{\infty} \omega_{j,t} Q_t^{(t+j)} Q_t^{(t+1)}}{Q_t}. \] (22)

I loglinearize around a geometric maturity structure, \( B_t^{(t+j)} = B_t \omega_t^{j-1} \), or equivalently \( \omega_{j,t} = \omega_t^{j-1} \). I use variables with no subscripts to denote the linearization points, and tildes to denote deviations from those points.

When we linearize, we move bond prices holding the maturity structure at its steady-state, geometric value, and then we move the maturity structure while holding bond prices at their steady-state value. As a result, changes in maturity structure have no first-order effect on the linearized bond return. At the steady state \( Q_t^{(t+j)} = 1/(1+i)^j \),

\[ R_{t+1}^n = \frac{\sum_{j=1}^{\infty} \omega_{j,t} (1+i)^{j-1} Q_t^{(t+j)} Q_t^{(t+1)}}{\sum_{j=1}^{\infty} \omega_{j,t} (1+i)^{j}} = (1+i) \]

independently of \( \{w_{j,t}\} \). Intuitively, at the steady state bond prices, all bonds give the same return, so all portfolios of bonds give the same return. Moreover, maturity structure is a time-\( t \) variable in the definition of return \( R_{t+1}^n \). The return from \( t \) to \( t+1 \) is not affected by the time \( t+1 \) maturity structure. (Changes in maturity structure might affect returns if there is price pressure in bond markets. These are formulas for measurement, however, and such effects would show up as changes in measured prices coincident with changes in quantities.)

Maturity structure only has a second-order interaction effect on the bond portfolio return. For example, a longer maturity structure at \( t \) raises the bond portfolio return at \( t+1 \) if there is also a level shock, raising long-maturity bond returns at \( t+1 \). A longer maturity structure at \( t \) it raises the expected return if the yield curve at \( t \) is also temporarily upward sloping. But a linear VAR and a linear decomposition do not include interaction effects.

To be clear, I measure the bond portfolio return \( r_{t+1}^n \) directly, and exactly, and this measure includes all variation in maturity structure. The linearization only affects the
decomposition of the bond portfolio return to future inflation and future expected returns. A second-order approximation would effectively use a different $\omega$ in the decomposition formula for different dates, as well as estimate a VAR with parameters that depend on the maturity structure or interaction terms. But variation in the geometric maturity structure parameter $\omega$ makes little difference to the results. And the sample is too short to add more variables, interaction terms, or time-varying parameters.

The term of the linearization with steady-state bond prices and changing maturity thus adds nothing. The linearization only includes a linearization with steady-state, geometric maturity structure and changing bond prices. Linearizing (22) then, we have

$$r_{t+1} = \log (1 + \omega e^{q+1}) - q_t \approx \log \left( \frac{1 + \omega Q}{Q} \right) + \frac{\omega Q}{1 + \omega Q} \tilde{q}_{t+1} - \tilde{q}_t$$

(23)

where as usual variables without subscripts are steady state values and tildes are deviations from steady state. In a steady state,

$$Q = \sum_{j=1}^{\infty} \omega^{j-1} \frac{1}{(1+i)^j} = \left( \frac{1}{1+i} \right) \left( \frac{1}{1 - \frac{\omega}{1+i}} \right) = \frac{1}{1+i - \omega}.$$  \hspace{1cm} (24)

The limits are $\omega = 0$ for one-period bonds, which gives $Q = 1/(1+i)$, and $\omega = 1$ for perpetuities, which gives $Q = 1/i$. The terms of the approximation (23) are then

$$\frac{1 + \omega Q}{Q} = 1 + i$$

$$\frac{\omega Q}{1 + \omega Q} = \frac{\omega}{1 + i}$$

so we can write (23) as

$$r_{t+1} \approx i + \frac{\omega Q}{1 + i} \tilde{q}_{t+1} - \tilde{q}_t.$$

since $i < 0.05$ and $\omega \approx 0.7$, I further approximate to

$$r_{t+1} \approx i + \omega \tilde{q}_{t+1} - \tilde{q}_t.$$  \hspace{1cm} (25)

I find the value of $\omega$ that best fits the return identity, rather than measure the maturity structure directly, so the difference between $\omega$ and $\omega/(1+i)$ makes no practical difference.

To derive the bond return identity (4), iterate (25) forward to express the bond
price in terms of future returns,

\[
\tilde{q}_t = - \sum_{j=1}^{\infty} \omega^j \tilde{r}^n_{t+j}.
\]

Take innovations, move the first term to the left hand side, and divide by \(\omega\),

\[
\Delta E_{t+1} \tilde{r}^n_{t+1} = - \sum_{j=1}^{\infty} \omega^j \Delta E_{t+1} \tilde{r}^n_{t+1+j}.
\]  

(26)

Then add and subtract inflation to get (4),

\[
\Delta E_{t+1} \tilde{r}^n_{t+1} = - \sum_{j=1}^{\infty} \omega^j \Delta E_{t+1} \left(\tilde{r}^n_{t+1+j} - \tilde{\pi}_{t+1+j} + \tilde{\pi}_{t+1+j}\right).
\]  

(27)

The expectations hypothesis states that expected returns on bonds of all maturities are the same,

\[
E_t r^n_{t+1} = i_t
\]

\[
i + \omega E_t \tilde{q}_{t+1} - \tilde{q}_t = i_t
\]

\[
\omega E_t \tilde{q}_{t+1} - \tilde{q}_t = \tilde{i}_t
\]

In the text, all variables are deviations from steady state, so I drop the tilde notation.

### C A variance decomposition

I use the elements of the impulse response function and their sums to calculate the terms of the unexpected inflation identity (3). We can interpret this calculation as an decomposition of the variance of unexpected inflation. Multiply both sides of (3) by \(\Delta E_{t+1} \pi_{t+1}\) and take expectations, giving

\[
var \left(\Delta E_{t+1} \pi_{t+1}\right) = cov \left[\Delta E_{t+1} \pi_{t+1}, \Delta E_{t+1} \left(\tilde{r}^n_{t+1} - g_{t+1}\right)\right]
\]

\[
= - \sum_{j=0}^{\infty} cov \left[\Delta E_{t+1} \pi_{t+1}, \Delta E_{t+1} \tilde{s}_{t+1+j}\right] + \sum_{j=1}^{\infty} cov \left[\Delta E_{t+1} \pi_{t+1}, \Delta E_{t+1} \left(\tilde{r}^n_{t+1+j} - \pi_{t+1+j} - g_{t+1+j}\right)\right].
\]  

(28)
Unexpected inflation may only vary to the extent that it covaries with current bond returns, or if it forecasts surpluses or real discount rates.

Dividing by \( \text{var } (\Delta E_{t+1}\pi_{t+1}) \), we can express each term as a fraction of the variance of unexpected inflation coming from that term. This decomposition adds up to 100%, within the accuracy of approximation, but it is not an orthogonal decomposition, nor are all the elements necessarily positive. Each term is also a regression coefficient of future long-run variables on unexpected inflation.

The two approaches give exactly the same result – the terms of (28) are exactly the terms of the impulse-response function, to an inflation shock orthogonalized last, i.e. a shock that moves all variables at time 1 including \( \Delta E_{1}\pi_{1} \).

To see this fact, write the VAR in standard notation

\[
x_{t+1} = Ax_{t} + \varepsilon_{t+1}
\]  

so

\[
\Delta E_{t+1} \sum_{j=1}^{\infty} x_{t+j} = (I - A)^{-1}\varepsilon_{t+1}.
\]

Let \( a \) denote vectors which pull out each variable, i.e.

\[
\pi_{t} = a'_{\pi} x_{t}, \ s_{t} = a'_{s} x_{t},
\]

etc. Then the present value identity (3) reads and may be calculated as

\[
a'_{\pi} \varepsilon_{t+1} - (a_{r} - a_{g})' \varepsilon_{t+1} = -a'_{s}(I - A)^{-1}\varepsilon_{t+1} + a'_{rg}(I - A)^{-1}A\varepsilon_{t+1}
\]

where

\[
a_{rg} \equiv a_{r} - a_{\pi} - a_{g}.
\]

We can calculate the variance decomposition (28) by

\[
a'_{\pi} \Omega a_{\pi} - (a_{r} - a_{g})' \Omega a_{\pi} = -a'_{s}(I - A)^{-1}\Omega a_{\pi} + a'_{rg}(I - A)^{-1}A\Omega a_{\pi}
\]
where $\Omega = \text{cov}(\varepsilon_{t+1}, \varepsilon'_{t+1})$, and then divide by $a'_n \Omega a_\pi$ to express the result as a fraction,

$$1 - (a_r n - a_g)' \frac{\Omega a_\pi}{a'_n \Omega a_\pi} = -a'_s (I - A)^{-1} \frac{\Omega a_\pi}{a'_n \Omega a_\pi} + a'_r g (I - A)^{-1} A \frac{\Omega a_\pi}{a'_n \Omega a_\pi}. \quad (32)$$

To show that this variance decomposition is the same as the elements and sum of elements of the impulse-response function to an inflation shock, orthogonalized last, note that the regression coefficient of any other shock $\varepsilon^z$ on the inflation shock is

$$b_{\varepsilon^z, \varepsilon^\pi} = \frac{\text{cov}(\varepsilon^z_{t+1}, \varepsilon^\pi_{t+1})}{\text{var}(\varepsilon^\pi_{t+1})} = \frac{a'_s \Omega a_\pi}{a'_n \Omega a_\pi},$$

so the VAR shock, consisting of a unit movement in inflation $\varepsilon^\pi_1 = 1$ and movements $\varepsilon^z_1 = b_{\varepsilon^z, \varepsilon^\pi}$ in each of the other variables is given by

$$\varepsilon_1 = \frac{\Omega a_\pi}{a'_n \Omega a_\pi}.$$

We recognize in (32) the responses and sums of responses to this shock. Dividing (28) by the variance of unexpected inflation, or examining the terms of (32), we recognize that each term is also the coefficient in a single regression of each quantity on unexpected inflation.

In an analogous way, we can interpret the responses to other shocks as a decomposition of the covariance of unexpected inflation with that shock, based on

$$\text{cov} \left( \Delta E_{t+1} \pi_{t+1} \varepsilon_{t+1} \right) - \text{cov} \left[ \varepsilon_{t+1}, \Delta E_{t+1} \left( r^n_{t+1} - g_{t+1} \right) \right] = -\sum_{j=0}^{\infty} \text{cov} \left[ \varepsilon_{t+1}, \Delta E_{t+1} s_{t+1+j} \right] + \sum_{j=1}^{\infty} \text{cov} \left[ \varepsilon_{t+1}, \Delta E_{t+1} \left( r^n_{t+1+j} - \pi_{t+1+j} - g_{t+1+j} \right) \right].$$

This variance decomposition is similar in style to the decomposition of return variance in Campbell and Ammer (1993). To avoid covariance terms, however, it follows the philosophy of the price/dividend variance decomposition in Cochrane (1992), extended to a multivariate context. With $x = y + z$, I explore $\text{var}(x) = \text{cov}(x, y) + \text{cov}(x, z)$ rather than $\text{var}(x) = \text{var}(y) + \text{var}(z) + 2\text{cov}(y, z)$. 


D Monte Carlo details

To evaluate sampling distributions I run a simple Monte Carlo. I start with the estimated VAR. I find the covariance matrix of the residuals $\varepsilon_{t+1}$. The identity (1) implies

$$\varepsilon_{s,t+1} = \varepsilon_{r,t+1} - \varepsilon_{g,t+1} - \varepsilon_{\pi,t+1} - \varepsilon_{v,t+1}. \tag{33}$$

Since I infer the surplus data $s_t$ from (1), the data obey this identity and the covariance matrix of residuals is singular. Thus I simulate iid shocks from the covariance matrix of all shocks except the surplus, and then I infer the surplus shock from the identity (33).

I initialize the VAR at the first data point, thereby generating the conditional sampling distribution. I simulate forward 50,000 artificial data samples using the estimated VAR parameters. I re-estimate the VAR and I calculate impulse responses and inflation decompositions in each artificial sample. I tabulate the sampling distribution of these quantities and report quantiles.

In a very few artificial samples, the VAR estimate has eigenvalues greater than or equal to one, so $(I-A)^{-1}$ cannot be computed. I omit these 38 out of 50,000 samples. As a result the reported quantiles are slightly smaller than actual quantiles. Avoiding these infinities and beyond is one reason that I report quantiles rather than standard errors. More generally, the distribution of statistics is not normal.

It is also not always possible to find $\omega \in [0, 1]$ to satisfy the return identity, so many Monte Carlo draws use a best fit value of $\omega$ in which the return identity does not hold. Weights have little effect on the results however, so this fact seems to have little effect. Since this is what I would have done in sample had I not been able to find an $\omega \in [0, 1]$ that satisfied the return identity, this fact just fills out the correct sampling distribution.

I run the Monte Carlo using sample estimates, and in particular the estimated 0.98 coefficient of debt on lagged debt. Near unit roots are biased down, and one might wish also to run a Monte Carlo with a bias-corrected estimate with eigenvalues closer to one. That procedure would likely lead to somewhat larger sampling distributions.

Between the conditional Monte Carlo – starting at the first data point – the problem of draws with $A$ eigenvalues greater than one, near-unit roots, and non-normal error distributions, one could likely find sampling experiments that produce even larger distributions. Generating data from models with stochastic volatility, time-varying means,
sample breaks, Markov-switching, and so forth may do even more.

But remember, I am not testing anything, so the point is simply to give a sense of the sampling error of the measurements. My main conclusion is that the sampling distribution of the response functions and decompositions, though narrow enough that the qualitative results are reasonably reliable, is still pretty wide already, steering me away from model complications. Sampling exercises that produce even wider distributions would only emphasize that point.

### E Sources of sampling variation

Table 4 includes the regression of other shocks on inflation shock that starts off the main inflation decomposition, and thus determines the instantaneous response in Figures 2 and 9. The table also includes the correlation matrix of the shocks.

To measure the relative contribution of the shock correlation and the long-run response function given the shock identification as sources of variation, Table 5 includes two other sampling calculations. The “no b” columns resample data using the original regression of shocks $\varepsilon_{t+1}$ on inflation shocks $\varepsilon_{t+1}^\pi$, the top row of Table 4, in each sample. The VAR coefficients still vary across samples, but the identification of the inflation shock does not. The “no A” columns likewise keep constant the VAR regression coefficients, but
Table 5: Decomposition of unexpected inflation variance – distribution quantiles. No b holds the initial response constant across trials. No A holds the VAR regression coefficients constant across trials

<table>
<thead>
<tr>
<th>Component</th>
<th>Fraction Estimate</th>
<th>No b 25%</th>
<th>No b 75%</th>
<th>No b 25%</th>
<th>No b 75%</th>
<th>No A 25%</th>
<th>No A 75%</th>
</tr>
</thead>
<tbody>
<tr>
<td>Inflation $\pi_1$</td>
<td>1.00</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Bond return $(r^n_1 - g_1)$</td>
<td>-0.23</td>
<td>-0.45</td>
<td>0.00</td>
<td>-0.23</td>
<td>-0.23</td>
<td>-0.45</td>
<td>0.00</td>
</tr>
<tr>
<td>Future $\Sigma s$</td>
<td>-0.06</td>
<td>-0.69</td>
<td>0.23</td>
<td>-0.60</td>
<td>0.14</td>
<td>-0.69</td>
<td>0.23</td>
</tr>
<tr>
<td>Future $\Sigma r - g$</td>
<td>1.17</td>
<td>0.42</td>
<td>1.57</td>
<td>0.63</td>
<td>1.37</td>
<td>0.42</td>
<td>1.57</td>
</tr>
</tbody>
</table>

reestimate the shock regression in each sample. Turning off either source of sampling variation reduces that variation, but not as much as you might think. Sampling variation is still large in either case, and variances add, not standard deviations. Moreover the sampling variation associated with shock orthogonalization – the “no A” exercise – does not go away no matter how small the shocks. Both left and right hand sides of the shock on shock regressions get smaller at the same rate.

**F 1980-2018 subsample results**

This section presents results using the 1980-2018 subsample. Much monetary macroeconomics isolates this period as having a different set of correlations that the earlier 1970s inflation, 1960s under Bretton woods, etc. Breaking the sample also allows us to see if the results are stable across subsamples.

Table 6 presents OLS VAR regression coefficients, parallel to Table 1. Table 7 compiles inflation decompositions, parallel to Table 2. Figures 10, 11 and 12 plot responses to inflation shocks, paralleling Figures 2, 3, and 4.

The broad pattern of Figure 10 is similar to the full postwar sample. There are some differences. The surplus and growth shocks are now positively correlated with the inflation shock, seen in the period 1 responses. There is less need to isolate a separate growth+inflation shock in this period, dominated by “aggregate demand” rather than “stagflation” episodes.

However, the surplus and growth responses turn negative after one period, as they are in the full sample. Higher inflation strongly forecasts a lower surplus, -1.55 in Table
Table 6: OLS VAR estimate. Sample 1980-2018. One (two) stars means the estimate is one (two) Monte Carlo standard errors away from zero.

<table>
<thead>
<tr>
<th></th>
<th>$r_{t+1}$</th>
<th>$g_{t+1}$</th>
<th>$\pi_{t+1}$</th>
<th>$s_{t+1}$</th>
<th>$v_{t+1}$</th>
<th>$i_{t+1}$</th>
<th>$y_{t+1}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$r_t^n$</td>
<td>-0.22</td>
<td>0.05</td>
<td>-0.10**</td>
<td>-0.25*</td>
<td>0.08</td>
<td>-0.04*</td>
<td>0.07</td>
</tr>
<tr>
<td>$g_t$</td>
<td>-0.11</td>
<td>0.13</td>
<td>0.06</td>
<td>0.76</td>
<td>-1.06</td>
<td>0.20*</td>
<td>-0.04</td>
</tr>
<tr>
<td>$\pi_t$</td>
<td>-0.04</td>
<td>-0.57**</td>
<td>0.67**</td>
<td>-1.55</td>
<td>1.41</td>
<td>-0.08</td>
<td>-0.12</td>
</tr>
<tr>
<td>$s_t$</td>
<td>0.10*</td>
<td>0.07*</td>
<td>-0.02</td>
<td>0.38**</td>
<td>-0.34*</td>
<td>-0.02</td>
<td>-0.04*</td>
</tr>
<tr>
<td>$v_t$</td>
<td>-0.00</td>
<td>0.00</td>
<td>-0.00</td>
<td>0.05</td>
<td>0.95**</td>
<td>0.01</td>
<td>0.00</td>
</tr>
<tr>
<td>$i_t$</td>
<td>-0.12</td>
<td>-0.27*</td>
<td>0.20*</td>
<td>1.14*</td>
<td>-1.19</td>
<td>0.61**</td>
<td>0.31*</td>
</tr>
<tr>
<td>$y_t$</td>
<td>1.61**</td>
<td>0.67**</td>
<td>-0.08</td>
<td>0.05</td>
<td>0.98</td>
<td>0.32*</td>
<td>0.57**</td>
</tr>
</tbody>
</table>

$100 \times std(\varepsilon_{t+1})$ | 2.44 | 1.10 | 0.51 | 5.17 | 7.00 | 1.15 | 0.93 |

$\text{Corr } \varepsilon, \varepsilon$ | -0.40 | 0.14 | 1.00 | 0.11 | -0.31 | 0.27 | 0.42 |

$R^2$ | 0.73* | 0.54* | 0.88* | 0.50* | 0.94* | 0.85* | 0.89* |

$100 \times std(x)$ | 4.74 | 1.63 | 1.48 | 7.30 | 28.88 | 2.97 | 2.84 |

rather than -0.25 in Table 1, and similarly higher inflation forecasts lower growth -0.57 rather than -0.14. The overall responses are then similar to the full period.

Surpluses then recover and turn positive as before. The sum of the surplus response remains small, 0.19 rather than -0.06.

Figure 11 explores the long-run surplus response, and you can see the same dynamics playing out. Inflation forecasts a rise in debt (1.41 in Table 6), and the period of deficits also raises debt (-0.34). But the rise in debt leads to a rise in surpluses, which slowly pay down much of that debt.

The expected return also rises in figure 10, and accounts for all the inflation and more in this subsample as it does in the main estimate.

Figure 12 shows the interest rate response in more detail. The wiggly response, which I pointed out in the postwar sample and is a result of slight overfitting there, is even more pronounced here. However, wiggles aside, the basic picture is similar. Interest rates and the expected bond return rise together, and almost permanently in response to the inflation shock. They do not rise as much as inflation, giving a few periods of negative expected returns, but their rise is so much more persistent than that of inflation that we see a very long period of high expected returns on the right side of the graph. As in the full sample, the much greater persistence of yield-curve changes than of inflation generates the long-term discount rate rise which accounts for most of the inflation shock.
\[\sum_{j=0}^{\infty} \omega^j \Delta E_1 \pi_{1+j} = -\sum_{j=0}^{\infty} \Delta E_1 s_{1+j} - \sum_{j=0}^{\infty} \Delta E_1 g_{1+j} + \sum_{j=1}^{\infty} (1 - \omega^j) \Delta E_1 (r_{1+j}^n - \pi_{1+j})\]

<table>
<thead>
<tr>
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<th>(g)</th>
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</thead>
<tbody>
<tr>
<td>Inflation</td>
<td>+2.32</td>
<td>-(0.19)</td>
<td>-(0.52)</td>
<td>+3.03</td>
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<tr>
<td>Recession</td>
<td>-2.50</td>
<td>-(0.31)</td>
<td>-(1.67)</td>
<td>-4.49</td>
</tr>
<tr>
<td>Surplus</td>
<td>-0.08</td>
<td>-(0.46)</td>
<td>-(0.54)</td>
<td>+1.08</td>
</tr>
<tr>
<td>Disc. Rate</td>
<td>-0.12</td>
<td>-(0.40)</td>
<td>-(0.49)</td>
<td>-1.00</td>
</tr>
<tr>
<td>Surplus, no i</td>
<td>0.07</td>
<td>-(0.70)</td>
<td>-(0.30)</td>
<td>-0.92</td>
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</tbody>
</table>

\[\Delta E_1 \pi_1 - \Delta E_1 r^n_1 = -\sum_{j=0}^{\infty} \Delta E_1 s_{1+j} - \sum_{j=0}^{\infty} \Delta E_1 g_{1+j} + \sum_{j=1}^{\infty} \Delta E_1 (r_{1+j}^n - \pi_{1+j})\]

<table>
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<tr>
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<th>(g)</th>
<th>(r^n - \pi)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Inflation</td>
<td>1.00</td>
<td>-(1.92)</td>
<td>-(0.19)</td>
<td>-(0.52)</td>
</tr>
<tr>
<td>Recession</td>
<td>-1.00</td>
<td>-(2.44)</td>
<td>-(0.31)</td>
<td>-(1.67)</td>
</tr>
<tr>
<td>Surplus</td>
<td>-0.01</td>
<td>-(0.32)</td>
<td>-(0.46)</td>
<td>-(0.54)</td>
</tr>
<tr>
<td>Disc. Rate</td>
<td>-0.02</td>
<td>-(0.35)</td>
<td>-(0.40)</td>
<td>-(0.49)</td>
</tr>
<tr>
<td>Surplus, no i</td>
<td>0.07</td>
<td>-(0.01)</td>
<td>-(0.70)</td>
<td>-(0.30)</td>
</tr>
</tbody>
</table>

\[\Delta E_1 r^n_1 = -\sum_{j=1}^{\infty} \omega^j \Delta E_1 (r_{1+j}^n - \pi_{1+j}) - \sum_{j=1}^{\infty} \omega^j \Delta E_1 \pi_{1+j}\]

<table>
<thead>
<tr>
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</thead>
<tbody>
<tr>
<td>Inflation</td>
<td>-1.92</td>
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<tr>
<td>Recession</td>
<td>2.44</td>
<td>-(1.50)</td>
</tr>
<tr>
<td>Surplus</td>
<td>0.32</td>
<td>-(0.07)</td>
</tr>
<tr>
<td>Disc. Rate</td>
<td>0.35</td>
<td>-(0.09)</td>
</tr>
<tr>
<td>Surplus, no i</td>
<td>-0.01</td>
<td>-(0.00)</td>
</tr>
</tbody>
</table>

Table 7: Terms of the inflation and bond return identities. Sample 1930-2018.

The impulse-response quantiles, plotted in Figure 13, are even larger than those of the full sample, but not so large that the results are meaningless.

Overall, we see a comfortably similar picture, and many signs of weak estimation in a short sample. At least it is comforting not to see the point estimates paint a much different picture, as they do in the prewar sample studied in the next section.

I do not present results for the 1947-1980 subsample to save space, since it too paints about the same picture. The near-term (5 years) response functions are similar. However, the point estimate has an eigenvalue of the transition matrix greater than one, so one must either reduce that or make calculations based on the first few responses only, not \((I - A)^{-1}\) calculations.
Figure 10: Response to inflation shocks, sample 1980-2018.

Figure 11: Response to inflation shocks, sample 1980-2018.
Figure 12: Response to inflation shocks, sample 1980-2018.

Figure 13: Inflation shock response quantiles, sample 1980-2018.
This section presents results using the full sample of data that I have been able to collect, 1930-2018.

![Table 8: OLS VAR estimate. Sample 1930-2018. One (two) stars means the estimate is one (two) Monte Carlo standard errors away from zero.](image)

Table 8 presents OLS VAR regression coefficients, parallel to Table 1. Table 10 compiles inflation decompositions, parallel to Table 2. Figures 14, 15, and 16 plot responses to inflation shocks, paralleling Figures 2, 3, and 4. Figure 17 presents sampling quantiles, paralleling Figure 9.

Start with the impulse response function for the inflation shock, Figure 14, paralleling Figure 2. The general pattern is similar. But the magnitudes are completely different. The 1% inflation shock still corresponds to a prolonged deficit, and the deficit eventually turns to surplus. But the deficit is larger and longer, and following surpluses no longer pay off the accumulated debts. The sum of the surplus responses is -2.59, not -0.06, accounting for more than all of the 1.83% weighted sum of inflation.

Discount rates follow the same general pattern as well. But the decline in discount rate is longer lasting, and the subsequent rise much smaller, so discount rates now account for -0.52% inflation, not +1.004% inflation.

The growth response goes the other way, now rising with inflation rather than declining, and therefore contributes -0.93% inflation rather than +0.49%.
\[
\sum_{j=0}^{\infty} \omega^j \Delta E_1 \pi_{1+j} = -\sum_{j=0}^{\infty} \Delta E_1 s_{1+j} - \sum_{j=0}^{\infty} \Delta E_1 g_{1+j} + \sum_{j=1}^{\infty} (1 - \omega^j) \Delta E_1 (r_{1+j}^n - \pi_{1+j})
\]

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<th>( g )</th>
<th>( r^n - \pi )</th>
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<tr>
<td>Inflation</td>
<td>1.83</td>
<td>-(-2.59)</td>
<td>-0.93</td>
<td>+(0.17)</td>
</tr>
<tr>
<td>Recession</td>
<td>-2.00</td>
<td>-2.59</td>
<td>-2.13</td>
<td>-(1.54)</td>
</tr>
<tr>
<td>Surplus</td>
<td>0.09</td>
<td>-1.04</td>
<td>-0.04</td>
<td>+(0.91)</td>
</tr>
<tr>
<td>Disc. Rate</td>
<td>-0.05</td>
<td>-0.89</td>
<td>-0.05</td>
<td>+(1.00)</td>
</tr>
<tr>
<td>Surplus, no i</td>
<td>0.30</td>
<td>-1.27</td>
<td>0.27</td>
<td>-(0.70)</td>
</tr>
</tbody>
</table>

\[
\Delta E_1 \pi_1 - \Delta E_1 r^n_1 = -\sum_{j=0}^{\infty} \Delta E_1 s_{1+j} - \sum_{j=0}^{\infty} \Delta E_1 g_{1+j} + \sum_{j=1}^{\infty} \Delta E_1 (r_{1+j}^n - \pi_{1+j})
\]

<table>
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<th>( r^n - \pi )</th>
</tr>
</thead>
<tbody>
<tr>
<td>Inflation</td>
<td>1.00</td>
<td>-(-0.14)</td>
<td>-2.59</td>
<td>-(0.93)</td>
</tr>
<tr>
<td>Recession</td>
<td>-1.00</td>
<td>-(0.17)</td>
<td>-2.59</td>
<td>-(2.13)</td>
</tr>
<tr>
<td>Surplus</td>
<td>0.07</td>
<td>-(0.13)</td>
<td>-1.04</td>
<td>-0.04</td>
</tr>
<tr>
<td>Disc. Rate</td>
<td>-0.01</td>
<td>-(0.16)</td>
<td>-0.89</td>
<td>-0.05</td>
</tr>
<tr>
<td>Surplus, no i</td>
<td>0.26</td>
<td>-(0.01)</td>
<td>-1.27</td>
<td>-(0.27)</td>
</tr>
</tbody>
</table>

\[
\Delta E_1 r^n_1 = -\sum_{j=1}^{\infty} \omega^j \Delta E_1 (r_{1+j}^n - \pi_{1+j}) - \sum_{j=1}^{\infty} \omega^j \Delta E_1 \pi_{1+j}
\]

<table>
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<tr>
<th></th>
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<th>( \pi )</th>
</tr>
</thead>
<tbody>
<tr>
<td>Inflation</td>
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<td>-(0.83)</td>
</tr>
<tr>
<td>Recession</td>
<td>-(0.82)</td>
<td>-(1.00)</td>
</tr>
<tr>
<td>Surplus</td>
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<td>-(0.02)</td>
</tr>
<tr>
<td>Disc. Rate</td>
<td>-(0.11)</td>
<td>-(0.05)</td>
</tr>
<tr>
<td>Surplus, no i</td>
<td>0.01</td>
<td>-(0.05)</td>
</tr>
</tbody>
</table>

Table 9: Terms of the inflation and bond return identities. Sample 1930-2018.

In sum, the full sample data paint a picture more than diametrically opposite. A 1% inflation shock, drawn out to 1.83% cumulative weighted inflation, is more than accounted for by 2.53% cumulative deficits, and buffered by an 0.52% disinflationary decline in discount rates, and 0.93% disinflationary rise in growth.

The full-sample results appear to support a simple fiscal theory, which would be convenient – inflation comes from persistent deficits. Discount rates only mitigate that result.

Why then do I emphasize the postwar sample in the text, and relegate these to an online appendix? Clearly, the full sample results do not carry through the postwar period to the present. As in essentially all macroeconomics and monetary economics, which
studies the post-1947 sample, the post-1959 sample, or, increasingly, the post-1980 sample, the war and prewar data behave differently. My interest in this paper is to characterize the behavior of inflation in postwar recessions, and the peacetime inflation of the 1970s and 1980s. Making an inference about that behavior from war and prewar data, when the central results switch in a postwar-only sample would be hugely misleading.

The nature of the prewar and war regime is interesting. Alas, the 1930-1947 sample is too short for these VAR methods. An investigation of the prewar regime with a long historical time series beckons.

What are the stylized facts and influential data points behind this switch in behavior? As before, long-run forecasts are driven by slow-moving state variables. Think of a system

\[
x_{t+1} = \alpha y_t + \varepsilon_{x,t+1} \\
y_{t+1} = \rho y_t + \beta \varepsilon_{x,t+1} + \varepsilon_{y,t+1}.
\]

In the second equation, I express the \( y \) shock in terms of a component correlated with the \( x \) shock and an orthogonal component. In this system, the variable \( y \) is the persistent state variable for long-run responses. The long-response of \( x \) to the \( \varepsilon_x \) shock depends on
how much the state variable \( y \) moves, \( \beta \), and the persistence of the \( y \) variable. In response to \( \varepsilon_{x,1} = 1 \), the long-run \( x \) response is

\[
\Delta E_t \sum_{j=0}^{\infty} x_{1+j} = 1 + \frac{\alpha \beta}{1 - \rho}.
\]

With this insight, let us understand the responses of Figure 14. Three state variables matter most. From Table 8, inflation basically follows its own AR(1), unaffected by other variables, with a a persistence of 0.53, the same value as the postwar sample. The value of debt is the most important state variable for long-run responses with an 0.91 coefficient on its own lag. However, this debt to GDP ratio does respond strongly (-0.61) to surpluses, and to lagged growth (-1.1) as we would expect, so at medium runs it evolves jointly with these other variables. The surplus has a strong coefficient on its lag, 0.65, so in part any shock to surpluses coincident with the inflation shock will persist. The surplus also responds positively though with a small value 0.08 to the debt. This coefficient does not account for much of the short run dynamics, as the movements of surplus and debt are roughly the same size, but is the dominant force behind very long run surpluses which repay debts. The surplus responds and negatively -0.75 to inflation. This key coef-
ficient is only -0.25 in the postwar sample. Interest rates also have a persistent response, but they move so little in this estimate that they are not an important state variable.

So, what accounts for the long deficits in Figure 14? The surplus does not jump down by a large amount with the shock, declining only 0.25, so the surplus’ autocorrelation is not a big part of the story. The big decline in surplus follows from its -0.75 coefficient on inflation, and the inflation AR(1) response. If inflation this year forecasts deficits next year, then a very simple fiscal theory story that inflation is accounted for by deficits follows swiftly.

But deficits should raise the value of debt, and the rise in the value of debt, which is very persistent, should pull deficits back to surplus, no? Here, another difference in the full sample is key. In the full sample, the value of debt $v$ jumps down by 1.10% when inflation jumps up 1%, where in the postwar sample the value of debt jumps down half as much, 0.65%. Now, a low value of debt does not put into motion additional surpluses. So, the effect seen in the postwar sample of Figure 3, that deficits quickly give rise to higher debt which then triggers surpluses, is absent here because so much debt was wiped out by the inflation shock.

Contrasting Figures 15 and 3 help to explain the differing behavior of the discount

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**Figure 16: Response to inflation shocks, sample 1930-2018.**

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rate. In both cases, the behavior of nominal interest rates is disturbingly disconnected from the behavior of inflation. In the postwar sample, nominal rates rise immediately and very persistently. When inflation declines and passes by the higher nominal rates, real rates are higher. In the full sample, nominal rates move much less, reflecting the zero bound in the great depression and interest rate targets in WWII and the early postwar period. The resulting real rate the inverse of the inflation AR(1), and mostly negative.

The massive deficits of 1943 and 1944 are key influential data points that account for the shift in behavior of the full sample. Estimates from the 1940-2018 sample, not shown, are similar. Figure 18 plots inflation and surplus during WWII. The WWII deficits are immense. Inflation, more volatile in the pre-1947 period, was above its mean in the years prior to these immense deficits. Thus, this inflation preceding deficits of 1943 and 1944 drives the result that inflation forecasts deficits in the full sample, and thus the result that inflation shocks are accounted for by deficits. This is clearly not a robust result, or one that should be taken as evidence that inflation today is due to deficits.

The strong negative correlation between shocks to inflation and to the value of debt in the full sample comes from a different set of influential observations. The inflation of 1943 and 1944 was largely expected, according to the VAR, and preceded rather
than coincided with increased debt. Instead, the sharp and unexpected (by the VAR) postwar inflation of 1947 coincided with a sharp decline in the real value of debt, and the sharp deflation of 1932 coincided with a sharp rise in the real value of debt. These events are conventionally regarded as times in which deflation raised the value of debt, in the first, and inflated it away, in the second. But again, one is loath to let these two observations double our estimate of the correlation between shocks to inflation and the value of debt for the postwar period.

The inflation shock is already positively correlated with a growth shock in the full sample, due to a strong positive correlation in the 1930s. As a result, the response to the inflation + growth shock (not shown) is not much different from the response to the inflation shock. Again, the 1% deflation and 2% cumulative inflation corresponds to 2.6% cumulative rise in surpluses. This time a long-run decline in discount rate contributes to deflation, but an equally large decline in growth contributes to inflation.