What Drives Variation in the U.S. Debt/Output Ratio? The Dogs that Didn't Bark*

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Abstract

If the U.S. is on a fiscally sustainable path, then higher U.S. government debt/output ratios should reliably predict higher future surpluses or lower real returns on Treasurys. We find no evidence for this. Neither future cash flows nor discount rates account for the variation in the current debt/output ratio. Instead, the future debt/output ratio accounts for most of the variation. Systematic surplus forecast errors can account for part of these findings. Since the start of the GFC, surplus projections have anticipated a large fiscal correction that failed to materialize.

Key words: fiscal policy, debt sustainability, government budget constraint JEL codes: E6, H6, G1

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There is an ongoing debate in the U.S. and other advanced economies about fiscal sustainability (see, e.g., Chernov, Schmid, and Schneider, 2020; Jiang, Lustig, Van Nieuwerburgh, and Xiaolan, 2019; Brunnermeier, Merkel, and Sannikov, 2020; Reis, 2021; Mian, Straub, and Sufi, 2021). Some economists have pointed out that lower inflation-and growth-adjusted returns on government debt can rationalize the recent increase in the U.S. debt/output ratio (Blanchard, 2019; Furman and Summers, 2020; Cochrane, 2021a), while others have argued that the run-up in debt could reflect higher future surpluses (Bohn, 1998; Cochrane, 2020).

To analyze the empirical validity of these claims, we approach the valuation of the U.S. federal government debt using standard tools from asset pricing. We apply a Campbell and Shiller (1988)-style decomposition to the market value of the U.S. federal government's debt divided by the U.S. gross domestic product. A higher debt/output ratio has to be followed by lower real growth-adjusted returns, higher surplus/output ratios, or, absent adjustment in fundamentals, higher future debt/output ratios.

At horizons of up to 10 years, in the post-war U.S. sample, our variance decomposition attributes none of the variation in the current debt/output ratio to variation in future surpluses or future returns. Instead, the future debt/output ratio 10 years from now accounts for 100% of the variation in today's debt/output ratio. We extend this horizon to 20 years in a longer sample.

First, higher debt/output does not predict lower real growth-adjusted returns on U.S. government debt in sample. Discount rates do not explain variation in the U.S. debt/output ratio. Second, higher debt/output ratio does not forecast larger surpluses. Cash flows do not explain variation in the debt/output ratio either. Instead, we cannot rule out that the U.S. debt/output ratio follows a unit-root process. Fundamentals fail to push the debt/output ratio back to the mean, which violates fiscal sustainability.¹

We implement a forward-looking decomposition of the government debt/output ratio which decomposes the valuation into components due to future returns and future surpluses (see Berndt, Lustig, and Yeltekin, 2012; Cochrane, 2021a).² In earlier work, Hall and Sargent (2011); Berndt and Yeltekin (2015); Hall and Sargent (2021) implement an ex post, backward-looking decomposition of the variation in the U.S. federal debt/GDP ratio, imputing changes in the ratio to contemporaneously realized inflation, growth and returns. Our ex ante approach provides a different, complementary perspective.

Our findings are also different from those in the literature.³ For example, Bohn (1998), study-

¹The small role for fundamentals (cash flows and discount rates) is implied by the persistence of the debt/output ratio. The first-order auto-correlation of the debt/output ratio is 0.99 at annual frequencies. As we get closer to a unit root, the fraction of the variance accounted for by fundamentals decreases to zero % at all horizons.

²In earlier work, Gourinchas and Rey (2007) decompose the variation in the U.S. net foreign asset position using the same approach.

³There is a large literature in macro-economics that addresses the question of government debt sustainability (Hamilton and Flavin, 1986; Trehan and Walsh, 1988, 1991; Bohn, 1998, 2007), starting with the seminal work by Hansen and Sargent (1980); Hansen, Roberds, and Sargent (1991) (see Sargent, 2012, for a comprehensive review). This literature

ing a shorter sample that ends in the mid-90s, finds evidence that the primary surplus increases when the debt/output ratio is high, while Cochrane (2021a,b) finds evidence that the debt/output ratio predicts nominal returns on the government debt portfolio. We replicate these findings, but we find no evidence that the debt/output ratio predicts real growth-adjusted returns or surpluses. Bond prices today are not sufficiently responsive to news about future macro fundamentals, returns and surpluses.

We reach a different conclusion because of a statistical inference challenge, well known in asset pricing, that is pervasive in the fiscal sustainability literature. There is a small-sample bias in the slope coefficients of the return and surplus predictability regressions due to (i) the high persistence of the debt/output ratio, the predictor, and (ii), the high correlation between the innovations to the predictor and the predicted variables (Stambaugh, 1999). For example, an increase in bond risk premia will tend to lower realized returns, the dependent variable, and lower the debt/output ratio, the predictor. As a result, the expectation of the residual in the return predictability regression conditional on the debt/output ratio this year and last year cannot be zero. Thus, the classical OLS assumptions are violated. The OLS estimator of the slope coefficient will tend to be too high as a result. Because of this mechanical link between realized returns and the predictor, OLS will find evidence of return predictability, even where there is none. The same small-sample bias also affects the surplus predictability regressions. Governments must issue more debt when they run large deficits. The negative correlation between the innovation in the debt/output ratio and the surplus similarly produces spurious evidence of surplus predictability.

We cannot rule out that the debt/output ratio is subject to permanent innovations, which is inconsistent with fiscal sustainability. As a check, we simulate data from a model in which the debt/output ratio has a unit root and there is no role for fundamentals. This model exactly replicates our empirical findings. We find spurious evidence of surplus and return predictability using unadjusted OLS slope estimates, but bias-corrected estimates that recover the true values indicating no predictability.

Structural breaks are one source of permanent shocks. There may have been a structural break in the relation between bond returns, surpluses and the debt/output ratio.⁴ We find evidence of a structural break in the debt/output ratio around 2007. Bond investors may have revised their estimate of the long-run debt/output ratio upwards around the start of the GFC. If we allow for a permanent 36 log points increase in the debt/output ratio in 2007, and use the debt/ouput ratio adjusted for a different subsample mean before and after 2007, we find substantially stronger evidence for surplus and return predictability. Fundamentals now account for about 50% of the

largely sidesteps the issue of discount rate variation. Instead, most of the models assume constant discount factors.

⁴In stock markets, there also is evidence of structural breaks in the relation between returns and the dividend yield (see Lettau and Van Nieuwerburgh, 2008; Smith and Timmermann, 2020). This may help to account for the poor outof-sample performance of these predictors (Welch and Goyal, 2008).

variation in the transitory component of the debt/output ratio at the 10-year horizon.⁵ Of course, this analysis leaves the large, permanent increase in the debt/output ratio (as well as its timing) unexplained.

Mis-pricing is a candidate explanation for these structural breaks. An econometrician with access to the U.S. sample does not predict higher surpluses or lower returns when the debt/output ratio rises, but the bond investor may. If investors systematically over-predict surpluses and under-predict returns as the debt/output ratio increases, this could impute a unit root the debt/output ratio and extra measure, while under the subjective measure, the debt/output ratio is stationary. Our paper connects to the growing literature on the role of subjective beliefs in asset pricing. de La O and Myers (2021) implement a Campbell-Shiller decomposition under the subjective measure of U.S. stock valuations. Under the subjective measure, they impute a large share of variation in the price/dividend ratio and price/earnings ratio on U.S./ stocks to the cash flow component. Renxuan (2020) finds that investors in the stock market all overstate the importance of expected cash flow growth fluctuations. Mistakes in interest rate and growth forecasts also contribute to our findings (see Piazzesi, Salomao, and Schneider, 2015; Cieslak, 2018, for evidence on the gap between statistical bond risk premia measured by econometricians and subjective bond risk premia expected by investors).

We show evidence from CBO (Congressional Budget Office) projections that the permanent break in the debt/output ratio at the start of the GFC (Great Financial Crisis) has been partly fueled by persistently biased forecasts of surpluses, imputing more mean-reversion to the predicted debt/output ratio under the subjective than the actual measure.⁶ Throughout and after the GFC, the CBO consistently overestimated future surpluses and underestimated the government's effective cost of funding. Even by 2010, the CBO projections of 10-year surpluses were still overshooting realized surpluses by roughly 6.8% of GDP per annum. Under the investors' subjective measure, more of the variation in the debt/output ratio can be attributed to perceived future surpluses.

Our paper builds on the growing literature that examines the predictability of returns in asset markets. There is substantial evidence that stock returns are predictable (see Koijen and Nieuwerburgh, 2011, for a survey of the literature on stock return predictability). When the valuations are high in stock markets, they revert back to the mean mostly through lower subsequent returns (Campbell and Thompson, 2007; Cochrane, 2008), partly through higher dividend growth (Binsbergen and Koijen, 2010; Golez and Koudijs, 2018). Overall, the discount rates on stocks are remarkably volatile (Hansen and Jagannathan, 1991), and the valuation of stocks seems excessively

⁵This structural break reduces the debt/output ratio's persistence and creates more room for fundamentals in explaining the–now more transitory nature of the—variation in the debt/output ratio.

⁶These projections do not take into account any future legislative action to implement a fiscal correction, which may have been anticipated by investors. The source of bond market investors optimism may have been the expectation of future fiscal rectitude through legislation.

volatile compared to its fundamentals (LeRoy and Porter, 1981; Shiller, 1981).

Typically, in asset markets, high valuations revert back to the mean through low subsequent returns. Not so for the whole U.S. government bond market. Even though there is evidence of return predictability for individual bonds (see Fama and Bliss, 1987; Campbell and Shiller, 1991; Cochrane and Piazzesi, 2005; Ludvigson and Ng, 2009; Cochrane, 2011), we find no evidence that return predictability for the entire bond portfolio pushes the debt/GDP ratio back to its long-run mean. The valuation of the entire U.S. government bond portfolio is too smooth compared to its fundamentals. The cash flows and the discount rates are the dogs that did not bark here. The valuation of all Treasurys is not sensitive enough to news about future surpluses or returns. Instead, we only find a statistically significant role for the future debt in accounting for the value of debt today.

Our paper does not impose no-arbitrage restrictions, but only uses accounting identities to understand variation in the debt/output ratio, rather than the level. In other work, Jiang, Lustig, Van Nieuwerburgh, and Xiaolan (2019) price a claim to future surpluses in a no-arbitrage model, and they conclude that the debt is overpriced.⁷

There is a large literature on the pitfalls of predicting returns with persistent predictors, going back to Nelson and Kim (1993); Stambaugh (1999); Lewellen (2004); Torous, Valkanov, and Yan (2004); Campbell and Yogo (2006); Boudoukh, Israel, and Richardson (2020). The Stambaugh correction can be refined if the econometrician rules out ex ante that the predictor variable has a true autocorrelation larger than one (Lewellen, 2004).⁸ We do not rule out non-stationary behavior of the debt/output ratio, because we are interested in testing fiscal sustainability.⁹ We also implement the Campbell and Yogo (2006) testing procedure, valid under general assumptions, and we conclude there is neither evidence of surplus or return predictability.

We start in section 1 by deriving and implementing the decomposition of the debt/output ratio. Section 2 looks at a longer sample and longer horizons. The results are robust. Section 3 allows for permanent shocks to the debt/output ratio. Finally, section 4 argues that systematic forecast errors may play a role in accounting for debt/output dynamics.

⁷In no-arbitrage models, all of the information for forecasting bond returns is embedded in the yield curve, except when the underlying factors cannot be inverted from the yields (see Duffee, 2011). Some authors report evidence that macro factors have incremental forecasting ability for bond returns (Cooper and Priestley, 2008; Ludvigson and Ng, 2009; Joslin, Priebsch, and Singleton, 2014). Bauer and Hamilton (2017) conclude that the evidence for macro variables predicting bond returns, after controlling for bond yields, against the spanning hypothesis, is weaker than previously thought, citing similar small-sample biases in the presence of persistent predictors.

⁸If the econometrician imposes that the true autocorrelation of the debt/output ratio is smaller than one, that effectively implies that either surpluses or returns are predictable.

⁹In a Monte-Carlo experiment, we implement the Stambaugh-correction on simulated data generated from a unitroot data generating process for the debt/ouput ratio. We find spurious evidence of surplus and return predictability. However, the bias-corrected estimated slope coefficients recover the true values.

1 What Drives Variation in the U.S. Debt/Output Ratio?

To illustrate the Stambaugh bias, we use the log-linear Campbell-Shiller decomposition of the government's debt/output ratio in Cochrane (2021a,b). This approach decomposes the variation in the debt/output ratio into three components: expected future government surpluses, expected future discount rates, and expected terminal value of the debt/output ratio in the future. After we correct for the Stambaugh bias, we find that most of the variation in the debt/output ratio cannot be attributed to subsequent government surpluses or discount rates even 10 years out.

1.1 Campbell-Shiller Decomposition of the Debt/Output Ratio

Let r_{t+j} denote the return on the government debt portfolio, x_{t+j} real GDP growth (in logs), and π_{t+j} log inflation. We use annual data. The government debt data is from CRSP Treasurys. We compute the market value of all marketable U.S. Treasurys. Following Hall and Sargent (2011), we compute the return on government debt as the sum of the principal and coupon payments less new issuance, plus the market value at the end of the period, divided by the market value at the end of the previous period. We exclude non-marketable debt which is mostly held in intragovernmental accounts. The inflation rate is the annual log change in the CPI, taken from the BLS. Nominal GDP is from the Bureau of Economic Analysis. Our sample of annual data comprises the period from 1949 to 2019.

We implement Cochrane (2021a)'s version of a Campbell-Shiller decomposition for the log of the debt/GDP ratio v_t . Let $\tilde{r}_{t+j} = r_{t+j} - x_{t+j} - \pi_{t+j}$ denote the adjusted log bond return, which is the nominal log return on the government bond portfolio minus output growth and inflation (in logs). Let $s_{t+j} = sy_{t+j}/e^v$ denote the adjusted government surplus/output ratio, where sy_t denotes the actual surplus/output ratio and v denotes the average log debt/output ratio.

We start from the log-linearized return equation implied by the government budget constraint:

$$r_{t+1} - \pi_{t+1} - x_{t+1} = \rho v_{t+1} - v_t + s_{t+1},$$

where ρ is a constant of linearization. We will choose $\rho = \exp(-(r - x - \pi)) = 1$. This decomposition expands the debt/output ratio around the unconditional average $r = x + \pi$. ¹⁰ We back out the surplus variable s_t from this equation:

$$s_{t+1} \equiv \widetilde{r}_{t+1} - \Delta v_{t+1}. \tag{1}$$

By iterating this equation forward and taking expectations, we obtain the following expression for

¹⁰We relax this in section 2.3. The details of the derivation are in the appendix of Cochrane (2021a) on page 3 and 4.

the debt/output ratio:

$$v_t = \mathbb{E}_t \sum_{j=1}^T \left(s_{t+j} - \widetilde{r}_{t+j} \right) + \mathbb{E}_t v_{t+T}.$$
(2)

The debt/output ratio reflects either expected future surpluses or expected future returns after adjusting for inflation and growth. Our decomposition allows for convenience yields on government debt.¹¹ These convenience yields would lower \tilde{r} in our decomposition. This accounting identity leaves a future debt/output ratio term in the Campbell-Shiller decomposition.¹² If the TVC is satisfied and if v_t is stationary, then $\mathbb{E}_t v_{t+T}$ converges to its unconditional mean.

Taking covariances with v_t on both sides of the previous equation, we obtain the following expression for the variance of the debt/output ratio :

$$var(v_t) = cov\left(\sum_{j=1}^T s_{t+j}, v_t\right) - cov\left(\sum_{j=1}^T \widetilde{r}_{t+j}, v_t\right) + cov(v_t, v_{t+T}).$$
(3)

The log debt/output ratio varies because it either predicts future surpluses, future returns, or the future debt/output ratios. In other words, the adjustment to an increase in the debt/output ratio either happens through an increase in future surpluses or a decrease in real growth-adjusted returns, or the debt is simply rolled over *T* periods from now. If v_t is non-stationary, then we replace the unconditional population moments with the finite sample moments in this variance decomposition.¹³

$$v\hat{a}r_N(v_t) = c\hat{o}v_N\left(\sum_{j=1}^T s_{t+j}, v_t\right) - c\hat{o}v_N\left(\sum_{j=1}^T \widetilde{r}_{t+j}, v_t\right) + c\hat{o}v_N(v_t, v_{t+T}).$$
(4)

To compute the variance decomposition, we directly estimate a system of univariate forecasting regressions for $\sum_{j=1}^{T} s_{t+j}, \sum_{j=1}^{T} \tilde{r}_{t+j}, v_{t+j}$ using the lagged debt/output ratio as a predictor:

$$\sum_{j=1}^T s_{t+j} = a_s + b_T^s v_t + \epsilon_{t+T}^s,$$

¹¹In models developed by Bassetto and Cui (2018); Brunnermeier, Merkel, and Sannikov (2020); Reis (2021); Chien and Wen (2019, 2020); Kocherlakota (2021); Mian, Straub, and Sufi (2021), government debt allows agents to self-insure against idiosyncratic risk and provides liquidity services. The resulting convenience yield contribute a separate component to the valuation of public debt.

¹²We do not impose the transversality condition (TVC). That requires taking a stand on what the right discount rate is to eliminate the terminal value v_{t+T} term (see Giglio, Maggiori, and Stroebel, 2016; Jiang, Lustig, Van Nieuwerburgh, and Xiaolan, 2019). There are equilibrium models that generate violations of the TVC including Samuelson (1958); Diamond (1965); Blanchard and Watson (1982); Hellwig and Lorenzoni (2009). Most of these models abstract from aggregate risk premia which would be priced into the terminal value and are likely to enforce the TVC (Jiang, Lustig, Van Nieuwerburgh, and Xiaolan, 2020; Barro, 2020).

¹³If v_t is non-stationary, the unconditional variance is infinite. However, in any finite sample, we can still compute the sample moments, and compute this in-sample variance decomposition using the sample moments.

$$\sum_{j=1}^{T} \widetilde{r}_{t+j} = a_r + b_T^r v_t + \epsilon_{t+T}^r,$$

$$v_{t+T} = \phi_0 + \phi_T v_t + \epsilon_{t+T}^v.$$
(5)

Cochrane (2008); Lettau and Van Nieuwerburgh (2008) adopt the same approach to implementing a Campbell-Shiller decomposition of the price/dividend ratio for stocks. Our claim is not that the debt/output ratio is the only predictor of bond returns or surpluses. We simply want to compute covariance between the debt/output ratio and future returns/surpluses.

Just as their counterparts from predictability regressions using stock returns, these regression coefficients can be interpreted as the variance of the explanatory variable explained by each component for a certain horizon *j*:

$$\frac{cov(\sum_{j=1}^{T} s_{t+j}, v_t)}{var(v_t)} = b_T^s,$$

$$\frac{cov(-\sum_{j=1}^{T} \tilde{r}_{t+j}, v_t)}{var(v_t)} = -b_T^r,$$

$$\frac{cov(v_{t+T}, v_t)}{var(v_t)} = \phi_T.$$

As discussed, the slope coefficients measure the fraction of the variance in the debt/output ratio at each horizon that is attributable to each component, the future returns, the future surpluses and future debt/output ratios respectively. Note that the coefficients sum to one at each horizon. The cross-equation restriction $b_T^s - b_T^r + \phi_T = 1$ is automatically satisfied, so that these three components jointly explain 100% of the variation in the debt/output ratio v_t .

To develop intuition for how the persistence in the debt/output ratio impacts the variance decomposition, consider the simpler case in which the debt/ouput ratio follows an AR(1): The variance decomposition is given by $b_1^s(1-\phi_1)^T/(1-\phi_1)$, $-b_1^r(1-\phi_1)^T/(1-\phi_1)$ and ϕ_1^T . In this case, if the debt/output approaches a unit-root process (namely, $\phi_1 \rightarrow 1$), v_{t+T} accounts for all of the variation in v_t at horizon T.

What do we expect to find in this variance decomposition? If the debt portfolio has zero beta, the real risk-free rate is constant and the growth rate of the economy is i.i.d, a reasonable benchmark, then we expect 100% of the variance to be attributed to cash flows. The only channel pushing the debt/output ratio back to the mean is through larger surpluses, following an increase in the debt/output ratio. If there is no predictability of the surpluses, then the deflated returns have to predictable. If there is no predictability of the adjusted returns or surpluses, then the debt/output ratio inherits a unit root, which violates fiscal sustainability.

Figure 1 plots the variables we want to forecast: the adjusted log returns \tilde{r}_t in the left panel and the surplus/output ratio in the panel on the right. Returns were low from the start of the sample

Figure 1: Returns and Surpluses



This figure plots the inflation-and-growth-adjusted log returns \tilde{r}_t and the surplus/output ratio s_t .

until the 80s. After 2000, real growth-adjusted returns declined again. The average adjusted return is -71 bps per annum. In the benchmark case, we approximate the budget constraint around $\tilde{r} = 0$. We do robustness analysis for $\tilde{r} = -1\%$ in section 2.3. Interestingly, the last decade does not look like an outlier. Low interest rates have coincided with low growth and low inflation to produce average adjusted returns.

The full line in Figure 2 is the predictor variable, the debt/output ratio, demeaned and logged. There was a marked decline in the debt/output ratio between 1948 and 1974, followed by a gradual increase in the eighties and nineties, and a surge in the wake of the Great Financial Crisis.

1.2 Results without Bias Correction

Figure 3 plots the baseline regression results. These figures plot the slope coefficients in regressions of future returns, surpluses and debt/output ratios on the current debt/output ratio (see regressions in eqn. 5). At the one-year horizon, 99% of the variance is attributed to the next year's debt/output ratio. That is to say, the log debt/output ratio is highly persistent. The first-order autocorrelation is 0.99. In fact, using the Augmented Dickey-Fuller test, we cannot reject the null hypothesis of the presence of the unit root in the log debt/output ratio in our sample period. Even at the 5-year horizon, 82% of the debt/output ratio fluctuations can be attributed to the future debt/output ratio. The R^2 in this debt/output predictability regression exceeds 0.5 at the 5-year horizon – see Table 1, which reports these predictability results for horizons from 1 to 10 years.

At the 10-year horizon, the other two channels start to matter: 46% of the variation is at-





The full line is the demeaned log debt/output ratio. The dashed line is the demeaned log debt/output ratio, demeaned by two different sub-sample means before and after 2007. The null hypothesis of no structural break is rejected in 2007 at the 1% level with F-statistic of 8.81.

tributed to variations in expected government surpluses and 25% of the variation is attributed to variations in discount rates. However, we cannot reject the null that the fraction is zero. The terminal debt/output ratio in the 10-year horizon only accounts for 30% of the variation.

Why is the discount rate channel so weak? The debt/output ratio does predict nominal bond returns with the right sign, as emphasized by Cochrane (2021a): higher debt/output predicts lower nominal returns. This effect is mechanical because bond prices are determined by the nominal log returns over the maturity of the bond: $p_t^N = -\sum_{i=0}^{N-1} hpr_{t+i+1}^{N-i}$. This effect is almost completely offset by a similar effect on inflation, as shown in Table 2. The top panel looks at forecasts of nominal bond returns, the second panel looks at forecasts of real GDP growth, the third panel considers inflation, and the fourth panel considers real-growth adjusted returns. The net effect of the debt/output ratio on future real returns is small and statistically insignificant as a result.

To generate CIs for these coefficients, we construct the standard errors by bootstrapping 10,000 samples from the joint residuals in a regression of (r_t, v_t) on two lags of v_t . We draw the surplus innovation such that the identity is enforced. Here is the full system of equations for the bootstrap:

$$s_{t+1} = \tilde{r}_{t+1} - \Delta v_{t+1},$$

$$\tilde{r}_{t+1} = a_r + b_1^r v_t + b_2^r v_{t-1} + u_{t+1}^r,$$

$$v_{t+1} = \psi_0 + \psi_1 v_t + \psi_2 v_{t-1} + u_{t+1}^v.$$
(6)

For bootstrapping purposes, we use two lags because the debt/output ratio fits an AR(2) structure

well, which delivers white noise estimated residuals to bootstrap from. The surplus residual is backed out from the cross-equation restriction. The CIs imply considerable uncertainty in the point estimate. Even at the 10-year horizon, the two-standard-deviation CIs for the b_j^s and b_j^r contain zero, so that the explanatory power of future government surpluses and discount rates are indistinguishable from zero.



Figure 3: Variance Decomposition of Log Debt/Output Ratio

This figure reports the variance decomposition of the log debt/output ratio v_t into components due to future government surpluses $\sum_{j=1}^{T} s_{t+j}$, future discount rates $\sum_{j=1}^{T} \tilde{r}_{t+k}$, future log debt/output ratio v_{t+T} , and the combination of the future government surpluses and discount rates. Sample is annual, 1948—2019. We plot 1 s.e. (dark) and 2 s.e. (light) CIs. Standard errors are generated by bootstrapping 10,000 draws from the joint residuals in a regression of (r_t, v_t) on two lags of v_t .

Table 1: Forecasting Returns and Surpluses with log Debt/Output ratio

OLS Regression of $\sum_{j=1}^{T} s_{t+j}, \sum_{j=1}^{T} \tilde{r}_{t+j}, v_{t+T}$ on v_t . Annual data. Sample: 1948-2019. Standard errors generated by bootstrapping 50,000 draws from the joint residuals in a regression of (r_t, v_t) on two lags of v_t , and we set $s_{t+1} = \tilde{r}_{t+1} - \Delta v_{t+1}$.

Horizon	1	2	3	4	5	6	7	8	9	10
	No Break									
	<i>Forecasting</i> $\sum_{j=1}^{T} \widetilde{r}_{t+j}$									
$-b_T^r$	0.01	0.03	0.05	0.06	0.07	0.09	0.13	0.16	0.2	0.25
s.e.	[0.02]	[0.05]	[0.07]	[0.09]	[0.11]	[0.12]	[0.13]	[0.15]	[0.15]	[0.16]
R^2	0.01	0.03	0.03	0.03	0.03	0.04	0.06	0.07	0.08	0.1
unbiased	-0.01	-0.01	-0.02	-0.03	-0.04	-0.04	-0.03	-0.02	0	0.02
	Forecasting $\sum_{i=1}^{T} s_{t+i}$									
b_T^s	-0.01	0.01	0.04	0.07	0.11	0.16	0.22	0.29	0.37	0.46
s.e.	[0.02]	[0.06]	[0.09]	[0.13]	[0.16]	[0.19]	[0.21]	[0.23]	[0.24]	[0.26]
R^2	0	0	0.01	0.02	0.03	0.05	0.06	0.09	0.12	0.14
unbiased	-0.04	-0.05	-0.05	-0.06	-0.05	-0.03	-0.01	0.03	0.08	0.14
					Forecast	ing v++T				
ϕ_T	0.99	0.95	0.91	0.87	0.82	0.75	0.66	0.55	0.43	0.3
s.e.	[0.03]	[0.09]	[0.14]	[0.19]	[0.23]	[0.27]	[0.29]	[0.31]	[0.33]	[0.34]
R^2	0.95	0.85	0.73	0.63	0.52	0.41	0.29	0.18	0.1	0.04
unbiased	1.05	1.06	1.07	1.09	1.09	1.08	1.04	0.98	0.92	0.84

1.3 Bias Correction

The high persistence in the explanatory variable v_t raises concern of small-sample bias. In particular, the OLS slope coefficients, $|b_j^r|$, are biased upwards in absolute value, because the innovations to the returns ϵ^r are positively correlated with the regressor innovations ϵ^v , and the regressor is highly persistent (Stambaugh, 1999). For example, an increase in bond risk premia will tend to induce lower realized returns and lower debt/output ratio. That gives rise to positive correlation between the regressor innovations and the return innovations. These positive biases tend to increase for long-horizon predictability regressions (Boudoukh, Israel, and Richardson, 2020).¹⁴

We find a similar positive bias for the surplus predictability regression. An increase (decrease) in debt issuance tends to coincide with the government running large deficits (surpluses). As a result, there is a strong, negative correlation between ϵ^r and the regressor innovations ϵ^v . This, combined with the persistence of the debt/output ratio, induces a large upward bias in b_T^s as well. In both case, the bias leads us to find too much predictability.

The unbiased coefficients were obtained by applying the Stambaugh (1999); Boudoukh, Israel, and Richardson (2020) small-sample bias correction for the OLS coefficients in the predictability

¹⁴Note that we plot minus the slope coefficient $-b_T^r$ in the figures.

Table 2: Forecasting Nominal Returns and Inflation with Log Debt/Output Ratio

This table further decomposes the adjusted bond return \tilde{r}_{t+j} into its three components: nominal bond return r_{t+j} , GDP growth x_{t+j} , and inflation π_{t+j} . The adjusted bond return \tilde{r}_{t+j} is equal to $r_{t+j} - \pi_{t+j} - \pi_{t+j}$. We run OLS Regression of $\sum_{j=1}^{T} r_{t+j}, \sum_{j=1}^{T} x_{t+j}, \sum_{j=1}^{T} \pi_{t+j}$ on \tilde{v}_t . Annual data. Sample: 1948-2019. Standard errors generated by bootstrapping 50,000 draws from the joint residuals in a regression of (r_t, x_t, π_t, v_t) on two lags of v_t .

Horizon	1	2	3	4	5	6	7	8	9	10
	Forecasting $\sum_{j=1}^{T} r_{t+j}$									
b_T^r	-0.05	-0.11	-0.16	-0.22	-0.28	-0.35	-0.43	-0.52	-0.6	-0.69
s.e.	[0.02]	[0.03]	[0.05]	[0.06]	[0.07]	[0.08]	[0.09]	[0.1]	[0.11]	[0.13]
R^2	0.19	0.29	0.34	0.38	0.4	0.43	0.47	0.5	0.52	0.54
unbiased	-0.04	-0.08	-0.12	-0.16	-0.21	-0.26	-0.33	-0.4	-0.47	-0.55
	Forecasting $\sum_{i=1}^{T} x_{t+i}$									
b_T^x	0	0	0.01	0	0	Ó	0.01	0.01	0.02	0.03
s.e.	[0.01]	[0.02]	[0.03]	[0.04]	[0.05]	[0.06]	[0.06]	[0.07]	[0.08]	[0.08]
D2	0	0	0	0	0	0	0	0	0.01	0.02
K-	0	0	0	0	0	0	0	0	0.01	0.02
unbiased	-0.01	-0.01	-0.02	-0.03	-0.04	-0.05	-0.05	-0.05	-0.05	-0.05
	<i>Forecasting</i> $\sum_{i=1}^{T} \pi_{t+i}$									
b_T^{π}	-0.04	-0.08	-0.12	-0.16	-0.21	-0.26	-0.31	-0.37	-0.42	-0.48
s.e.	[0.01]	[0.01]	[0.02]	[0.02]	[0.03]	[0.04]	[0.05]	[0.06]	[0.08]	[0.09]
R^2	0.51	0.53	0.56	0.59	0.63	0.65	0.66	0.67	0.67	0.67
unbiased	-0.04	-0.08	-0.12	-0.16	-0.21	-0.26	-0.31	-0.37	-0.42	-0.47
	<i>Forecasting</i> $\sum_{i=1}^{T} \tilde{r}_{t+i}$									
$b_T^{\tilde{r}}$	-0.01	-0.03	-0.05	-0.06	-0.07	-0.09	-0.13	-0.16	-0.2	-0.25
s.e.	[0.01]	[0.02]	[0.03]	[0.04]	[0.05]	[0.06]	[0.06]	[0.07]	[0.08]	[0.09]
<i>R</i> ²	0.01	0.03	0.03	0.03	0.03	0.04	0.06	0.07	0.08	0.1
unbiased	0.01	0.01	0.02	0.03	0.04	0.04	0.03	0.02	0	-0.02

regression with horizon *j*:

$$\begin{aligned} bias_T^r &= \mathbb{E}\left(\widehat{b}_T^r - b_T^r\right) = \frac{1}{N}\left[T(1+\phi) + 2\phi\frac{1-\phi^T}{1-\phi}\right] \times -\frac{cov(\epsilon^v, \epsilon^r)}{var(\epsilon^v)},\\ bias_T^s &= \mathbb{E}\left(\widehat{b}_T^s - b_T^s\right) = \frac{1}{N}\left[T(1+\phi) + 2\phi\frac{1-\phi^T}{1-\phi}\right] \times -\frac{cov(\epsilon^v, \epsilon^s)}{var(\epsilon^v)},\end{aligned}$$

where ϕ denotes the first-order autocorrelation of v_t , T denotes the horizon, and N denotes the size of the sample. We note that $corr(\epsilon^v, -\epsilon^r) = -0.75$ and $corr(\epsilon^v, \epsilon^s) = -0.85$, so the implied biases for the coefficient b_T^s associated with surplus and for the coefficient $-b_T^r$ associated negative debt are both positive. Fixing the volatilities of these innovations, the small-sample bias grows as the true autocorrelation $\phi \rightarrow 1$, and as $corr(\epsilon^v, -\epsilon^r) \rightarrow 1$ and $corr(\epsilon^v, \epsilon^s) \rightarrow -1$. Given the persistence of the debt/output ratio, and the size of the residual correlations, it is fair to say that the bias is close to its upper bound.

To better understand the bias, we can restate the bias of the coefficients at horizon T = 1 as follows :

$$bias_1^r = \mathbb{E}[\widehat{\phi} - \phi] \times -\frac{cov(\epsilon^v, \epsilon^r)}{var(\epsilon^v)},$$

$$bias_1^s = \mathbb{E}[\widehat{\phi} - \phi] \times -\frac{cov(\epsilon^v, \epsilon^s)}{var(\epsilon^v)},$$

where $\mathbb{E}[\hat{\phi} - \phi]$ is roughly $-(1 + 3\phi)/N$. This expectation is taken over all possible values of true autocorrelation ϕ . If instead we are willing to restrict $\phi \leq 1$, then the bias attains an upper bound at $\phi = 1$ (Lewellen, 2004). In the case of stock return predictability, Lewellen (2004) shows that the null of no predictability can be rejected more often if this stationarity restriction is imposed. This is not surprising: imposing that $\phi \leq 1$ is equivalent to imposing $b_1^r + b_1^s \geq 0$, assuming ex ante that there is return or cash flow predictability. We do not impose any ex ante restrictions on the true autocorrelation of the the debt/output ratio. We do not want to rule out that the U.S. is on a fiscally unsustainable path.

Figure 4 reports the bias-adjusted regression coefficients, which still sum to one. The biascorrected variance decomposition attributes -4% and -5% of the debt/output ratio variance to the discount rate and cash flow channel respectively at the 5-year horizon. As a result, 109% is accounted for by the future debt/output ratio. At the 10-year horizon, we still attribute 84% of the variance to the future debt/output ratio, after correcting for the small-sample bias.

There is no statistical evidence that the debt/output ratio predicts either future government surpluses or future bond returns. The null hypothesis that there is no predictability in future government surpluses or future bond returns cannot be rejected at any horizon. At the 10-year horizon, we cannot even reject the joint null that the debt/output ratio does not predict the sum of surpluses and returns.

1.4 Tests of Predictability Under Local-to-Unity Asymptotics

Standard return predictability tests, such as the *t*-test of the slope of the OLS predictability coefficient may be inappropriate when the predictor is persistent and its innovations are highly correlated with returns. In such cases, large-sample theory provides a poor approximation to the finite-sample distribution of tests statistics. Campbell and Yogo (2006) offer an alternative test that is valid under general assumptions about the predictor dynamics, even when the largest root is larger than one, and the distribution of innovations. Campbell and Yogo (2006) develop a pre-test to diagnose whether the conventional *t*-test leads to valid inference. Their Dickey-Fueller Generalized Least Squares test is based on the CI for the largest autoregressive root of the predictor variable. Applying their method to our context, we obtain a 95% CI of [0.941, 1.059] for the persistence of the debt/output variable v_t . Since this CI contains the unit root, this indicates that

Figure 4: Variance Decomposition of Log Debt/Output Ratio after Bias Correction



This figure reports the variance decomposition of the log debt/output ratio v_t into components due to future government surpluses $\sum_{j=1}^{T} s_{t+j}$, future discount rates $\sum_{j=1}^{T} \tilde{r}_{t+k}$, future log debt/output ratio v_{t+T} , and the combination of the future government surpluses and discount rates. We generate bias-corrected slope coefficients using the Stambaugh (1999); Boudoukh, Israel, and Richardson (2020) small-sample bias correction for the OLS coefficients in long-horizon predictability regressions. Sample is annual, 1948—2019. We plot 1 s.e. (dark) and 2 s.e. (light) CIs. Standard errors are generated by bootstrapping 10,000 draws from the joint residuals in a regression of (r_t, v_t) on two lags of v_t .

standard t-tests are not valid.

Campbell and Yogo go on to develop an asymptotically valid and efficient *Q*-test, which results in a Bonferroni CI for the predictive coefficient of interest. When we apply their procedure to the surplus predictability regression, we find a 90% CI of [-0.0465, 0.0287]. This CI includes zero, so that we fail to reject the null hypothesis that the lagged debt/gdp ratio v_t does not forecast the surplus/gdp ratio s_{t+1} ($H_0 : b_1^s = 0$). We repeat the analysis for the adjusted return on the debt portfolio, \tilde{r}_{t+1} and find a 90% Bonferroni CI of [-0.0130, 0.0484]. We fail to reject the null hypothesis that the lagged debt/gdp ratio does not forecast the government debt return (H_0 : $b_1^r = 0$). This analysis confirms our findings: once the high degree of persistence of the debt/gdp ratio is taken into account, the evidence for predictability of future surpluses or future returns by the lagged debt/gdp ratio is very weak.

2 Robustness

We conduct robustness by looking at the shorter Bohn (1998) sample, a longer sample, and increasing the horizon. We also consider different steady-state values for the adjusted returns on the debt portfolio.

2.1 Shorter Bohn Sample

In a classic paper on this topic, Bohn (1998) interprets his finding that the debt/output ratio predicts surpluses as evidence against the unit root hypothesis, but our results show that smallsample bias will give rise to large point estimates at longer horizons. We repeated our exercise on Bohn (1998)'s 1948—1995 sample. Before correcting for the bias, there is a much larger role for fundamentals in this subsample, as shown in Figure 5, consistent with Bohn's findings. However, this evidence mostly disappears after the bias correction.

2.2 Longer Sample

Using the Hall, Payne, and Sargent (2018) historical data, we construct a longer U.S. sample that starts in 1842. With the longer sample, we also report results with forecast horizon up to 25 years instead of 10 years. The variance decomposition with and without bias adjustment is plotted in Figure 6. We focus on the results after the bias correction. The longer sample shrinks the CIs around the point estimates, but the point estimates are similar to those obtained in the post-war sample. There is no statistical evidence of either surplus or return predictability. Furthermore, we cannot reject the null that the fundamentals do not explain any variation, even after 25 years. After the bias correction, we find that future v still explains about 70% of the variation in today's v even 25 years out.

2.3 Approximation around Different Steady-State Values

Recall that we back out the linearized government surplus s_{t+1} from the following relation,

$$\tilde{r}_{t+1} = \rho v_{t+1} - v_t + s_{t+1},$$

where $\rho = \exp(-(r - x - \pi))$ is a constant of linearization. In the benchmark case, we chose $\rho = 1$ to linearize the equation system at $r = x + \pi$. Here, we re-derive our equations for general

Figure 5: Variance Decomposition of Log Debt/Output Ratio: Shorter Bohn Sample 1948—1995



Panel A: Before Bias Correction

Panel B: After Bias Correction



This figure reports the variance decomposition of the log debt/output ratio v_t into components due to future log debt/output ratio v_{t+T} and the combination of the future government surpluses and discount rates. We generate bias-corrected slope coefficients using the Stambaugh (1999); Boudoukh, Israel, and Richardson (2020) small-sample bias correction for the OLS coefficients in long-horizon predictability regressions. Sample is annual, 1948—1995. Standard errors are generated by bootstrapping 10,000 draws from the joint residuals in a regression of (r_t, v_t) on two lags of v_t . We plot 1 s.e. (dark) and 2 s.e. (light) CIs.



Panel A: Before Bias Correction

Panel B: After Bias Correction



This figure reports the variance decomposition of the log debt/output ratio v_t into components due to future log debt/output ratio $\rho^T v_{t+T}$, and the combination of the future government surpluses and discount rates. We generate bias-corrected slope coefficients using the Stambaugh (1999); Boudoukh, Israel, and Richardson (2020) small-sample bias correction for the OLS coefficients in long-horizon predictability regressions. Sample is annual, 1842—2020. Standard errors are generated by bootstrapping 10,000 draws from the joint residuals in a regression of (r_t, v_t) on two lags of v_t . We plot 1 s.e. (dark) and 2 s.e. (light) CIs.

 $\rho = \exp(-(r - x - \pi))$, and report a robustness result using $r - (x + \pi) = -1\%$ so that the risk-free rate is below the output growth rate by 1% per annum, in the Blanchard (2019) region of the paremeter space.

For the case of an arbitrary ρ , equation (2) becomes

$$v_t = \mathbb{E}_t \sum_{j=1}^T \rho^{j-1} \left(s_{t+j} - \widetilde{r}_{t+j} \right) + \mathbb{E}_t \rho^T v_{t+T}.$$
(7)

The corresponding regression equations become

$$\sum_{j=1}^{T} \rho^{j-1} s_{t+j} = a_s + b_T^s v_t + \epsilon_{t+T}^s,$$

$$\sum_{j=1}^{T} \rho^{j-1} \tilde{r}_{t+j} = a_r + b_T^r v_t + \epsilon_{t+T}^r,$$

$$\rho^T v_{t+T} = \phi_0 + \phi_T v_t + \epsilon_{t+T}^v.$$
(8)

so that the cross-equation restriction $b_T^s - b_T^r + \phi_T = 1$ is still satisfied.

Since we use the same persistent predictor v_t on the right-hand side, we apply the same Stambaugh-bias adjustment formula as in the main text. Figure 7 reports the estimates with and without bias adjustment for $\rho = \exp(1\%)$. Figure 8 reports the estimates for $\rho = \exp(-1\%)$. The results after the bias correction are quite similar to the ones obtained in the benchmark $\rho = 0$ case.

3 Permanent Shocks to the Debt/Output Ratio.

Our evidence is consistent with a unit root in the debt/output ratio. Next, we evaluate the accuracy of the small-sample bias correction term by simulating 10,000 samples with the same length as the actual sample. We simulate under the null that there is no mean reversion in the debt/output ratio. As a result, there is no contribution from return/surplus predictability (the fundamentals) either $(b_T^s - b_T^r = 0 = 1 - \phi_T)$ at all horizons *T*.

We assume that returns are i.i.d. and that the debt/output ratio follows a unit root process. We estimate an ARIMA(1,1,0) process for the debt/output dynamics. Hence, the system is given by the following equations:

$$\begin{split} \Delta v_{t+1} &= \psi_0 + \psi_1 \Delta v_t + \epsilon^v_{t+1}, \\ \tilde{r}_{t+1} &= r_0 + \epsilon^r_{t+1}, \\ v_{t+1} &= v_t + \Delta v_{t+1}. \end{split}$$

Figure 7: Variance Decomposition of Log Debt/Output Ratio: with $\rho = \exp(1\%)$



Panel A: Before Bias Correction





This figure reports the variance decomposition of the log debt/output ratio v_t into components due to future log debt/output ratio $\rho^T v_{t+T}$, and the combination of the future government surpluses and discount rates. We generate bias-corrected slope coefficients using the Stambaugh (1999); Boudoukh, Israel, and Richardson (2020) small-sample bias correction for the OLS coefficients in long-horizon predictability regressions. Sample is annual, 1948—2019. Standard errors are generated by bootstrapping 10,000 draws from the joint residuals in a regression of (r_t, v_t) on two lags of v_t . We plot 1 s.e. (dark) and 2 s.e. (light) CIs.

We infer the government surplus from the accounting identity:

$$s_{t+1} \equiv \widetilde{r}_{t+1} - \Delta v_{t+1}.$$

In each round of simulation, we draw with replacement from the joint distribution of the debt/output residual and the bond return ($\epsilon_{t+1}^v, \epsilon_{t+1}^r$). Then, we run the forecasting regressions in (5) in each

Figure 8: Variance Decomposition of Log Debt/Output Ratio: with $\rho = \exp(-1\%)$



Panel A: Before Bias Correction





This figure reports the variance decomposition of the log debt/output ratio v_t into components due to future log debt/output ratio $\rho^T v_{t+T}$, and the combination of the future government surpluses and discount rates. We generate bias-corrected slope coefficients using the Stambaugh (1999); Boudoukh, Israel, and Richardson (2020) small-sample bias correction for the OLS coefficients in long-horizon predictability regressions. Sample is annual, 1948—2019. Standard errors are generated by bootstrapping 10,000 draws from the joint residuals in a regression of (r_t, v_t) on two lags of v_t . We plot 1 s.e. (dark) and 2 s.e. (light) CIs.

simulated sample. We plot the mean slope coefficients obtained running the forecasting regressions and the two-standard-deviation CIs around these mean slope coefficient estimates.

In the simulated data, we find that future government surpluses and discount rates appear to explain variations in the debt/output ratio, even though we know that they do not. Figure 9 reports the variance decomposition implied by the simulated samples. The average slope coefficients obtained from the unit root model imply variance decompositions between the fundamen-

Figure 9: Variance Decomposition of Log Debt/Output Ratio under Unit Root



This figure reports the variance decomposition of the log debt/output ratio v_t into components due to future government surpluses $\sum_{j=1}^{T} s_{t+j}$, future discount rates $\sum_{j=1}^{T} \tilde{r}_{t+k}$, future log debt/output ratio v_{t+T} , and the combination of the future government surpluses and discount rates. The samples are generated by simulation under the null that the debt/output ratio has a unit root. We plot the mean of the small-sample slope coefficients in red. We also plot the long-sample slope coefficients in blue from a single simulation of 100,000 periods. We plot 1 s.e. (dark) and 2 s.e. (light) CIs.

tals and the future debt/output ratios that are very close to our point estimates in Figure 3. The OLS estimates for the autoregressive coefficient are severely biased downwards in small samples when the true model has a unit root (Hamilton, 1994, p. 217). As a result, we find spurious evidence of mean reversion that creates a large role for fundamentals over longer horizons, in cases where there is no mean-reversion. The true slope coefficients are one at all horizons.¹⁵

Quantitatively, the simulation result suggests that the downward bias in the variance explained by the future log debt/output ratio v_{t+T} is about 50% at the 10-year horizon. If we adjust

¹⁵The variance decomposition itself is not well defined for a unit root process, because the unconditional variance is not well defined. However, we can still run the estimation in the simulated small samples.

the variance decomposition result in Figure 3, the bias-adjusted variance explained by the future log debt/output ratio is about 30% + 50% = 80% in the 10-year horizon. This result is quite close to Figure 4 that adjusts the bias using the Stambaugh (1999) formula, and it fails to reject the null that the future log debt/output ratio explains 100% of the variance in today's log debt/output ratio.

3.1 Structural Breaks and Variance Decomposition of the Transitory Component of Debt/Output

One way of allowing for permanent shocks to the debt/output ratio, consistent with the unit-root evidence, is to allow for a structural break. There may have been structural shifts in the relation between the valuation of debt and the fundamentals. A major contributor to the small role of fundamentals is the large run-up in government debt during the GFC which was not followed by commensurate increases in surpluses or decreases in returns. There is statistical evidence of a structural break in the log debt/output ratio in 2007. The Chow test for structural breakpoints rejects the null hypothesis of no structural break at the 1% level in 2007 (and at no other date).

Following Lettau and Van Nieuwerburgh (2008)'s work on stock return predictability, we allow for a structural break in the log debt/output ratio by demeaning the log debt/output ratio $\tilde{v}_t = v_t - \bar{v}_t$ with a lower pre-2007 sample mean ($\bar{v}_t, t < 2007$) and a higher post-2007 sample mean ($\bar{v}_t, t \ge 2007$). The structural break introduces a 36 log point permanent increase in the debt/output ratio. Figure 2 plots the resulting series. This approach removes a low-frequency component from the debt/output ratio.¹⁶ Obviously, when we allow for this break, we introduce permanent innovations in the debt/output ratio.

We re-estimate the forecasting regressions using the new predictor. The slope coefficients now provide a variance decomposition of the transitory variation in debt/output. Taking covariances with v_t on both sides of the previous equation, we obtain the following expression for the variance of the transitory component of debt/output ratio :

$$var(\widetilde{v}_t) = cov\left(\sum_{j=1}^T s_{t+j}, \widetilde{v}_t\right) - cov\left(\sum_{j=1}^T \widetilde{r}_{t+j}, \widetilde{v}_t\right) + cov(\widetilde{v}_t, \widetilde{v}_{t+T}).$$
(9)

The resulting break-adjusted series is much less persistent than the original series, and hence, because of the cross-equation restrictions $b_T^s - b_T^r + \phi_T = 1$, will impute a larger role to fundamentals in explaining the variation in the transitory component.

The transitory component of the debt/output ratio now explains 23% of the variation in surpluses at the 5-year horizon (R^2). Figure 10 decomposes the variance of the transitory component into its fundamental components. When the transitory component of the debt/output ratio is

¹⁶We recompute the surpluses by feeding \tilde{v}_t into eqn. (1).

high, surpluses tend to increase to push the debt/output ratio back down.

As reported in Table 3, the fundamentals now explain 21% (48%) of the variation in transitory component of debt/output at the 5-year (10-year) horizon. We can reject the null that fundamentals do not play a role at 5 and 10 years. This approach restores a role for fundamentals after accounting for the small-sample bias, but only in explaining the transitory variation in the debt/output ratio. However, this variance decomposition punts on the large and permanent increase in the debt/output ratio in the post-2007 sample.

Table 3: Forecasting Returns and Surpluses with Break-Adjusted log Debt/Output ratio

OLS Regression of $\sum_{j=1}^{T} s_{t+j}, \sum_{j=1}^{T} \tilde{r}_{t+j}, v_{t+T}$ on v_t . \tilde{v}_t with structural break in 2007. Annual data. Sample: 1948-2019. Standard errors generated by bootstrapping 50,000 draws from the joint residuals in a regression of (r_t, v_t) on two lags of v_t , and we set $s_{t+1} = \tilde{r}_{t+1} - \Delta v_{t+1}$.

Horizon	1	2	3	4	5	6	7	8	9	10	
	Structural Break										
	Forecasting $\sum_{j=1}^{T} \tilde{r}_{t+j}$										
$-b_T^r$ s.e.	0.02 [0.03]	0.04 [0.05]	0.05 [0.08]	0.04 [0.1]	0.04 [0.12]	0.06 [0.13]	0.09 [0.15]	0.14 [0.17]	0.17 [0.18]	0.22 [0.2]	
R^2	0.01	0.02	0.02	0.01	0.01	0.01	0.02	0.04	0.06	0.08	
unbiased	0	0.01	0.01	-0.01	-0.02	-0.01	0.01	0.04	0.07	0.1	
	Forecasting $\sum_{i=1}^{T} s_{i+1}$										
b_T^s	0.08	0.18	0.27	0.35	0.41	0.48	0.53	0.58	0.64	0.7	
s.e.	[0.04]	[0.08]	[0.11]	[0.14]	[0.16]	[0.18]	[0.19]	[0.2]	[0.22]	[0.23]	
R^2	0.06	0.15	0.24	0.31	0.37	0.45	0.5	0.55	0.58	0.62	
unbiased	0.04	0.1	0.16	0.2	0.23	0.27	0.29	0.32	0.34	0.38	
	Econocating m										
ϕ_T	0.91	0.78	0.68	0.61	0.55	0.47	0.38	0.28	0.19	0.09	
s.e.	[0.05]	[0.1]	[0.14]	[0.17]	[0.19]	[0.21]	[0.22]	[0.23]	[0.24]	[0.24]	
R^2	0.87	0.71	0.56	0.46	0.38	0.27	0.18	0.1	0.05	0.01	
unbiased	0.96	0.88	0.83	0.81	0.79	0.74	0.69	0.64	0.59	0.52	

Figure 10: Variance Decomposition of Log Debt/Output Ratio with Break



This figure reports the variance decomposition of the transitory component of the log debt/output ratio $v_t - \bar{v}_t$ with a lower pre-2007 sample mean ($\bar{v}_t, t < 2007$) and a higher post-2007 sample mean ($\bar{v}_t, t \ge 2007$) into components due to future government surpluses $\sum_{j=1}^{T} s_{t+j}$, future discount rates $\sum_{j=1}^{T} \tilde{r}_{t+k}$, future log debt/output ratio $v_{t+T} - \bar{v}_{t+T}$, and the combination of the future government surpluses and discount rates. Sample is annual, 1948—2019. We impose a structural break in the log debt/output ratio by shifting its average level before and after 2007. We generate bias-corrected slope coefficients using the Stambaugh (1999); Boudoukh, Israel, and Richardson (2020) small-sample bias correction for the OLS coefficients in long-horizon predictability regressions. Standard errors are generated by bootstrapping 10,000 draws from the joint residuals in a regression of (r_t, \tilde{v}_t) on two lags of \tilde{v}_t . We plot 1 s.e. (dark) and 2 s.e. (light) CIs.

4 Systematic Forecast Errors as a source of Permanent Shocks to the Debt/Output Ratio

We find a large Stambaugh bias in the fiscal predictability regression. After we correct this bias, future government surpluses and discount rates do not explain variation in the level of outstanding debt/output ratio. Our results suggest that the debt is mis-priced, perhaps because investors have been overly optimistic about the U.S. fiscal situation. Suppose investors evaluate Eq. (2) using their own subjective expectation, denoted by the operator \mathbb{F} , which is not the conditional expectation operator \mathbb{E} :

$$v_{t} = \mathbb{F}_{t} \sum_{j=1}^{T} \left(s_{t+j} - \widetilde{r}_{t+j} \right) + \mathbb{F}_{t} v_{t+T} = \mathbb{E}_{t} \sum_{j=1}^{T} \left(s_{t+j} - \widetilde{r}_{t+j} \right) + \underbrace{\left(\mathbb{F}_{t} v_{t+T} + \left(\mathbb{F}_{t} - \mathbb{E}_{t} \right) \sum_{j=1}^{T} \left(s_{t+j} - \widetilde{r}_{t+j} \right) \right)}_{\mathbb{E}_{t} v_{T}},$$

where $(\mathbb{F}_t - \mathbb{E}_t) \sum_{i=1}^T (s_{t+i} - \tilde{r}_{t+i})$ denotes the investors' forecast errors about fundamentals.

We conjecture that government bond market forecasters turn to the CBO's detailed and sophisticated projections to make their forecasts about future debt/GDP. While the CBO forecasts GDP, inflation, and interest rates, it makes projections of future surpluses based on current law. There is a remarkably large and consistently negative bias in the debt/GDP projections, even well after the GFC. Note that if the CBO had assumed future fiscal corrections when making its debt/output projections in the past, its projections would have been even farther from the realized debt/output ratios.

Panel A of Figure 11 plots the CBO 10-year projections $\mathbb{F}_t v_{t+T}$ for the debt/GDP ratio against the actual time series. This bias arises mainly because the CBO has been over-predicting future surpluses as a fraction of GDP, as shown in panel B. As a result, the CBO projections consistently impute too much mean-reversion to the debt/output ratio. The GFC and the COVID-19 pandemic are large unanticipated shocks with a huge fiscal impact, but these projections remain biased throughout the GFC and even well after the GFC. Given the size and persistence of the CBO projection errors, it is conceivable that private investors have similarly biased forecasts of future surpluses.

We reconstruct the Campbell-Shiller decomposition of the debt/output ratio using the CBO projections. We compute the projected $s_{t+j} = sy_{t+j}/e^v$ adjusted government surplus/output ratio using the CBO projections. For v, we use average log debt/output ratio between 2007 and 2019. We then back out the implied future discount rates $\sum_{j=1}^{T} \tilde{r}_{t+k}$, for T = 10 from the projected v_{t+T} and the projected sum of the surpluses.

Consider the benchmark of no predictability under the real measure: $cov(\sum_{j=1}^{T} (s_{t+j} - \tilde{r}_{t+j}), v_t) = 0$. As a result, in this case, the debt/output ratio has a unit root. Below, we compute sample mo-

ments:

$$c\hat{o}v_N(v_t, v_{t+T}) = c\hat{o}v_N(v_t, \mathbb{F}_t v_{t+T}) + c\hat{o}v_N\left(v_t, (\mathbb{F}_t - \mathbb{E}_t)\sum_{j=1}^T s_{t+j}\right) = v\hat{a}r_N(v_t),$$

where we assumed (i) $\mathbb{F}_t \sum_{j=1}^T \tilde{r}_{t+j}$ is constant, and (ii) $cov(\sum_{j=1}^T (s_{t+j} - \tilde{r}_{t+j}), v_t) = 0$, implying that in sample $c\hat{o}v_N(v_t, v_{t+T}) = v\hat{a}r_N(v_t)$. Assumption (ii) implies a unit root under the real measure. Assumption (i) implies no predictability under the subjective measure. To explain our findings that the terminal value accounts for most of the variation, without any predictability under the real measure, we need a large positive covariance between surplus forecast errors on the part of bond market investors and the debt/output ratio. As the fiscal situation worsened, we find that the CBO projections persistently over-projected surpluses and under-projected deflated debt returns. Bond market investors were presumably pricing in a much larger surpluses than realized throughout the GFC.

Panel A in Figure 12 plots the entire sample for the 10-year sum of realized surpluses s_t and deflated returns, \tilde{r}_t . Panel B focusses on the 2007-2020 sample. The realized surplus series declines steeply as the GFC enters the forward-looking sum in 1997-1998. Just before the start of the GFC, the CBO projections were consistently too optimistic: $(\mathbb{F}_t - \mathbb{E}_t) \sum_{j=1}^T s_{t+j} \gg 0$. In 2007, at the start of the GFC, the CBO projected a positive 10-year surplus of 29 log points or roughly 2.9% per annum, shown in the top left graph of Panel B in Figure 12. The units on the y-axis are log points of output. The 2007 projection was one standard deviation above the average of realized surpluses over the entire U.S. post-war history. The actual realized number was -82 log points, or -8.2% per annum. There is a 111 log points forecast error, or 11.1% per annum. The sign of the gap is not surprising, but the size is. Throughout the GFC, the projections were much too optimistic. In 2008, there is still a 108 log points gap between the projected 10 year surplus and the realized one. The gap shrinks only slowly. In 2010, well after the end of the GFC, there is still a 68 log points or 6.8% per annum gap between projected and realized surpluses.¹⁷

The bottom left graph in panel B plots the projected S - R, the projected decrease in the log debt/output ratio over the next 10 years. There is a total 135 log points gap in 2007 between the projected and realized fundamentals. As a result, the realized debt/output ratio in 2017 was 145 log points higher than projected in 2007. In 2007, the CBO projected a more than 50 log points decline in the debt/output ratio. By 2009, it still projected a constant debt/output ratio over the next 10 years.

¹⁷Similarly, the return projections were too optimistic. The CBO-projected deflated returns over the next 10 years were -32 log points or -3.2% per annum in 2007, shown in the top right figure of Panel B in Figure 12, but the realized returns were 1.3 log points, mainly because realized growth was much lower than projected. Hence, there is an additional 22 log points gap contributed by the discount rate components. The CBO projected $r - x - \pi$ to be 2.2% lower per annum than realized. However, the CBO has significantly raised its projections for the effective cost of government funding since 2010.



Figure 11: CBO Projections for Debt/GDP and Surplus/GDP

This figure reports the CBO 10-year U.S. federal government debt/GDP and total federal government surplus/GDP projections for each forecast date, plotted against the actual debt/GDP and surplus GDP/ratio.

Our empirical evidence is certainly consistent with mis-pricing. The CBO seems to overestimates future surpluses when the economy enters a recession, and the debt/output ratio increases. This evidence is consistent with a large positive covariance between forecast errors and the debt. If investors' expectations were ligned with CBO projections, they think they live in a world without permanent shocks to debt/output, but their forecast errors induce permanent shocks in debt/output, at least from an econometrician's perspective.



Panel A: 1948-2020

Panel A and B show the decomposition of the log debt/output ratio v_t into components due to CBO-projected (and realized) future government surpluses $\sum_{j=1}^{T} s_{t+j}$, future discount rates $\sum_{j=1}^{T} \tilde{r}_{t+k}$, for T = 10. We also report future real growth $\sum_{j=1}^{T} \tilde{x}_{t+k}$.

5 Conclusion

The U.S. bond market's valuation of a claim to U.S. surpluses is surprisingly insensitive to news about future surpluses or returns, at least from the perspective of an econometrician. This is a direct result of the debt/output ratio's persistence. Our model-free exercise suggests that the bond market's assessment of future surpluses may diverge from the econometrician's. The econometrician does not forecast larger surpluses or lower discount rates when the debt/output ratio rises, not based on the entire U.S. sample, but, obviously, the bond market investor does, unless there is a violation of the no-bubble condition. We show evidence from CBO projections that investors may have been over-predicting surpluses and under-predicting returns on debt throughout and well after the GFC. As a result, bond investors anticipated mean reversion in the debt/output ratio that failed to materialize.

Even though the pricing of individual bonds does not allow for arbitrage opportunities, except in times of market disruptions, the U.S. federal government debt as a whole may be persistently mis-priced. It is plausible that limits to arbitrage (Shleifer and Vishny, 1997) would prevent rational arbitrageurs from taking advantage of potential over-pricing in the last decades. Treasurys benefit from safe asset demand. Safe assets tend to appreciate in high marginal utility states of the world, when arbitrageurs are more likely to be constrained.

Of course, we cannot definitely rule out the possibility of a peso-event, a large fiscal correction, not observed in our sample, but anticipated and priced in by investors. However, even the most sophisticated budget projections which abstract from future fiscal corrections have been far too optimistic over the past decades.

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