

Nonlinearities in Monetary Policy Transmission

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Abstract

First order approximations of a model's impulse response functions cannot capture non-linear behavior in shock transmission. Namely, it's possible that big shocks are not a scaled version of small shocks (size effect) or that responses following a positive shock do not mirror those from a negative shock (asymmetry). Using a Local Projection Instrumental Variable (LP-IV) framework, I estimate impulse responses of standard macro variables to monetary policy shocks and look for evidence of nonlinearities. Specifically, I use monetary policy surprise series as an instrument and decompose the measured shock into regimes based on how the Fed Funds Rate changed in a given month to allow for nonlinearity. I find evidence against the notion that larger shocks amplify the effects of smaller shocks and some instances of broader asymmetry. Future work will gauge how well this result holds up with other estimation approaches (e.g., VAR-based, non-parametric, Bayesian methods) and investigate if enriching standard model frictions helps produce impulse responses that align with US data.

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Introduction

Much of the evidence on the effects of monetary policy has been produced from model-based and empirical approaches that impose a linear relationship between shocks and real variables. Namely, while standard impulse response functions (IRFs) can be a non-linear function of time, it's often the case that if the impulse response to a shock s_t is $\{y_{t+h}\}$, the impulse response to αs_t is simply $\{\alpha y_{t+h}\}$. This rules out non-linearity from a shock's *size* (big and small don't have the same effect) and *sign* (positive and negative don't have symmetric effects). To produce any meaningful policy recommendations, it must be determined whether this simplification is consequential.

The first stage of this project will consist of estimating IRFs from US data of standard macro variables to monetary shocks in a way that allows for non-linearity. This will be accomplished by using monetary surprise measures that are well-established in the literature (e.g., [Romer and Romer, 2004](#); [Nakamura and Steinsson, 2018](#)), in combination with specifying regimes that correspond to the relative size and sign of change in the Federal Reserve's target federal funds rate in a given month. Then, we will see how well "off-the-shelf" models are able to match the nonlinearities we do observe. By construction, typical formulation and solution methods for standard models produce near or exactly linear IRFs. To make this model comparison exercise consequential, we modify the default constraints faced by firms (e.g., price and wage setting rigidities) and households (e.g., investment adjustment costs) to accommodate potential asymmetries, which also shed light on the extent to which the status quo structure dictates the results these models produce. Finally, we will consider further departures from the representative agent and full information paradigm to account for and ultimately rationalize what we observe.

The results gathered so far imply a substantial degree of non-linearity. For both negative and positive shocks, we find evidence of size effects for consumption, output (industrial production), and inflation (CPI growth). Specifically, we find evidence that modest, expansionary changes to monetary policy increase our macro variables of interest relatively more than a large change. Relativity in this context means that the effects of a small interest rate cut (a movement smaller than 10% of the Fed's previous target) are proportionally larger than a big cut. We also find that larger, counteractions changes have relatively less effect on inflation, while simultaneously producing a stronger depression in the level of consumption and output. To put these results in perspective, [Table 1](#) gives a comparison in levels for a horizon of 12 months. For sign effects, we find evidence that there are asymmetries between a big cut and a big hike but the evidence for small changes is inconclusive.

Table 1: Relative difference in year over year growth

	Cut Size Effect	Hike Size Effect
Consumption	-3.5%	-2.2%
Output	-8.4%	-3.8%
Inflation	-2%	0.9%

All estimates significant at 90% level

The bedrock of the parametric approach will be a Local Projection Instrumental Variables (LP-IV) framework advanced by [Jordà et al. \(2015\)](#), which offers several advantages over alternatives. For deciding between a LP or a vector autoregressive estimation (VAR), [Plagborg-Møller and Wolf \(2021\)](#) clarifies that in finite samples, this decision comes down to a trade-off between flexibility (LP) and efficiency (VAR). Notable advantages for LP are the insensitivity to the presence of unit roots or non-stationary variables as both outcome and control ([Plagborg-Møller and Montiel Olea, 2021](#); [Plagborg-Møller et al., 2024](#)) and the ability to relax the assumptions imposed by VAR structure (namely, invertibility and recursiveness). For deciding between LP and LP-IV, the various monetary shock series that have been developed still have unresolved interpretability issues (e.g., [Brennan et al., 2024](#)), but what is consistent across all approaches is that they are developed in order to capture relevant changes in monetary policy while maintaining exogenous variation. Thus, it's more appropriate to treat these shock series as instruments. Moreover, by instrumenting actual changes to the interest rate target, we can argue the coefficients produced in our estimation are representative of the effects of monetary policy in general, rather than just policy surprises. Finally, we could in principle use a different shock series for each regime of our decomposition, so we can simply use whichever one is the strongest instrument for a given regime. Doing so with a pure LP estimation would yield incomparable parameters.

The main disadvantage of LP-IV relative to LP is the inability to incorporate state dependence. It should be noted that the virtually all past approaches claiming to estimate impulse responses from state-dependent local projection estimation (as well as many VAR-based approaches) likely suffer from substantial bias for non-marginal shock sizes ([Gonçalves et al., 2024](#)). [Jordà \(2023\)](#) proposes a fix using a very specific type of interaction term framework, but it cannot be extended to LP-IV without suffering from the curse of dimensionality. In any event, [Gonçalves et al. \(2024\)](#) provides a simple non-parametric procedure to properly estimate state-dependent effects, which we can augment with instruments and additivity to account for sample size¹ and shock interpretability issues. Because we would expect ex ante for the efficacy of policy to be at least in part dependent of the world (e.g., the economy may not grow after a rate cut, but perhaps it would have contacted much more without one), we should consider this an important benchmark. Another previously referenced disadvantage of LP is the relatively large size of standard errors. We can consider Bayesian extensions to our approach in the spirit of [Ferreira et al. \(2024\)](#). Moreover, I plan to try all described estimation procedures to provide a frame of referenced. So far, the LP and state-dependent LP results have yielded qualitatively the same results for nearly all variables and various shock series (we also consider [Bauer and Swanson \(2023\)](#) and [Bu et al. \(2021\)](#)).

¹Sample size is a constraining issue in general, given the need to include lagged outcome and control variables for all specifications. This, in addition to the desire to include the entire zero lower bound episode and subsequent tightening, prevents the use of series with arguably superior identification strategies such as [Aruoba and Drechsel \(2023\)](#), [Acosta \(2023\)](#) and [Miranda-Agrippino and Ricco \(2021\)](#)

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