The Macroeconomic Effects of Inflation Expectations: The Distribution Matters

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The views expressed are those of the authors and/or of the model and do not necessarily reflect those of De Nederlandsche Bank.

Research questions

What are the macroeconomic effects of shocks to the short-term inflation expectations distribution?

Does the whole cross-sectional heterogeneity matter for macroeconomic fluctuations?

What are the implications for Central Banks communication policies?



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- Seem to capture how policy makers think about expectations
- Such a shock can have macroeconomic consequence
- Large literature on "expectational shocks"



Surprising small attention to macro effects of inflation expectation shocks

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 - Milani (2011): NK model with learning shows that (inflation) expectation shocks have an important role in business cycle



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- Surprising small attention to macro effects of inflation expectation shocks
 - Milani (2011): NK model with learning shows that (inflation) expectation shocks have an important role in business cycle
 - de-anchoring' shock to long-term inflation expectations (e.g., Clark and Davig, 2011; Diegel and Nautz, 2021; Neri, 2021) or to the inflation target of the central bank (e.g., Ireland, 2007; Cogley et al., 2010; Haque, 2019)

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 - Ascari et al. (2023): A shock that increase short term inflation expectations is stagflationary ($\uparrow \pi, \downarrow y$). See also Adams and Barrett (2022).





Looking at the whole distribution

- Surprising small attention to macro effects of inflation expectation shocks. ⇒ In this paper we try a step forward by looking at the whole distribution.
- Pervasive cross-sectional heterogeneity in survey data (Weber et al., 2022).
 - Sensitivity to salient prices, gender and cognitive biases, low attention to MP...
 - Meeks and Monti (2023): statistical relevance of heterogeneous expectations for inflation dynamics.
 - Reis (2022): qualitative evidence about second and third moments of distribution beliefs help predict inflation → there is a lot of information from the whole distribution and especially from the tails: the distribution matters.



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- The importance of accounting for the full heterogeneity in beliefs in empirical analysis.
- The cross-feedback effects between macroeconomic dynamics and expectations. Who influences who?



We augment a parsimonious Monetary Policy BVAR with the distribution of the short-term inflation expectation to investigate the macroeconomic effects of exogenous variations of the short term inflation expectation distribution:

- The importance of accounting for the full heterogeneity in beliefs in empirical analysis.
- The cross-feedback effects between macroeconomic dynamics and expectations. Who influences who?
- The macroeconomic consequences of various shocks that modify both the location and the shape of the distribution:
 - ⇒ Mean shocks, Dispersion (Variance, Tail, ...) shocks, Communication shocks



Results

Expectations affect macroeconomic variables ... but not (very little) vice versa

The distribution matters

Models based just on the consensus underestimate the macroeconomic effects of expectation shocks

- ⇒ Mean and Variance shocks are stagflationary
- ⇒ Tail (Skewness) shocks too....
- \Rightarrow The tails matter, especially the left one.

Communication should focus on the tails

Anchoring the expectation distribution is positive ... as long as the dispersion is limited \Rightarrow **The reduction in dispersion** might be more effective than anchoring the mean of the distribution.

Methodology

METHODOLOGY

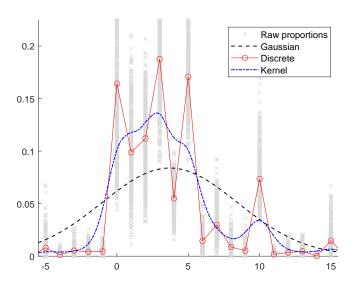
The Empirical Model



The empirical problem and our approach

- We face the problem of incorporating in the same model a time series of distributions (functions) and a time series of indicators (scalars).
- We employ a simple multi-step approach suited for the original feature of the Michigan survey, where inflation expectations are reported as integer percentage values. We consider three options:
 - \bullet "raw data" \to proportion of respondents reporting a specific inflation values
 - Rounding \to approximating with a continuous Kernel that smooths individual expectations over the cross-sectional domain \to intervals
 - parsimonious Gaussian distribution, $\sim \mathcal{N}(\mu_t, \sigma_t^2)$





Inflation Expectations Dataset and the distributional approximations

How to put the functional approximation in the BVAR? Two Challenges

Aggregating probabilities

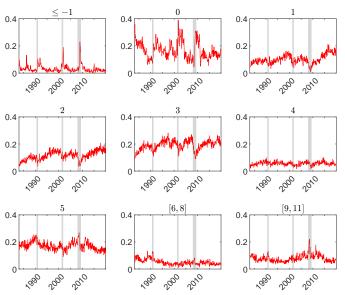
- ullet High-dimensional object o summarize with a handful of parameters
- Proportions/probabilities are positive and sum to 1, not on a vector space
- Replace discrete and Kernel distributions with their respective cumulative probabilities/distribution functions over M intervals of interest

$$w_{m,t}^{D} = \sum_{i \in \{q_{m-1}, q_m\}} p_{i,t}, \quad w_{m,t}^{K} = \int_{q_{m-1}}^{q_m} f_t^{K}(x) dx$$
 (1)

Interval selection. Data-driven

- they contain an average large amount of probability
- the last interval $(q_M, 50)$ is wide enough to guarantee that the sum of the first M is lower than 1
- the largest number of intervals such that each median is distinct





Time series of the 9 aggregated proportions, discrete approximation (\mathbf{w}_t^D)

Data and the model

- Expectation Block: $\mathbf{e}_t = [\mathbf{w}_t^a, s_t]', a \in \{D, K, G\}$. Inflation expectations proportions (\mathbf{w}_t^a) and Consumer Sentiment s_t . ¹
- Macroeconomic block: \mathbf{x}_t . Consumer Price Index, Industrial Production, Federal Funds Rate, and three controls: Real Oil Price, and the measures of uncertainty: Financial and Macroeconomic Uncertainty (Jurado et al., 2015). 2
- We model $\mathbf{y}_t = [\mathbf{e}_t', \mathbf{x}_t']'$ as a BVAR in deviation from the steady state (Villani, 2009). Minnesota shrinkages estimated in a hierarchical fashion (Chan, 2021).
- Monthly data from January 1983 to December 2019, for a total of T=422 observations.

²Fred ID: DCOILWTICO, UMCSENT, CPIAUCSL, INDPRO, ₩EDFUNDS. > ≥ 90

¹IE microdata and the sentiment from the monthly **Michigan Survey of Consumers**

Identification

- Households fill the questionnaire on the third week of the month, so at that point, they do not know the data about the industrial production or the inflation rate of the month
- Main identification assumptions: on impact the macro variables does not affect expectations (only exception oil price and uncertainty)
- Block recursive structure: extensive robustness checks ⇒ results robust to alternative ordering of the blocks

	OP + UNC	Inflation expectations	Fundamental Macro
OP + UNC	+	0	0
Expectation block		+	0
Other Macro			+

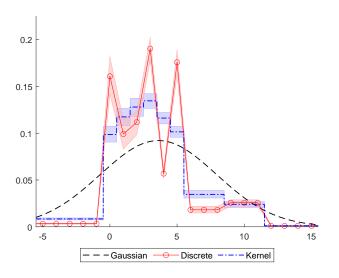


Results

RESULTS



SS-BVAR Steady States Distributions

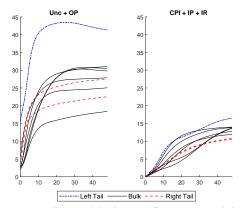




FEVD

FEVD by block of variables: Expectations block

Expectations are barely affected by macroeconomic shocks



Forecast Error Variance Decompositions

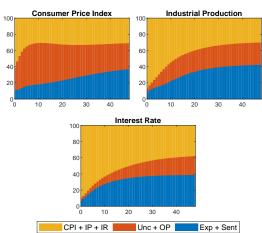
Contribution of the two exogenous variables (first subplot) and the three macroeconomic variables (second subplot) to the FEVD of the nine expectations series at different horizons (x-axis). Different colors indicate the location in the domain of the proportions. Left tail (< 0, blue dashed-dotted line), bulk (\in [1, 5], black lines), right tail (> 5, red dashed lines).



FEVD

FEVD by block of variables: Macroeconomic block

Exogenous variations in expectations affect macroeconomic variables

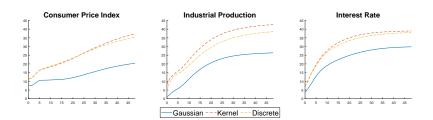


Forecast Error Variance Decompositions

Contributions of the three different identified blocks of indicators to the FEVDs of the macroeconomic variables at different horizons (x-axis). Alternative/Opposite Identification

FEVD

FEVD by block of variables: approximations comparison



Comparison among different approximations of the contribution of the expectation block to the FEVD of the macroeconomic variables

In order to isolate the contribution of the distribution the effect of the sentiment has been subtracted.



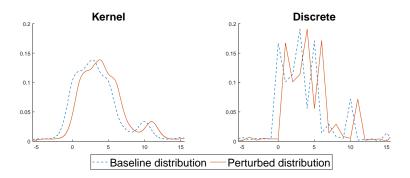
Density IRF Analysis

We now look at DENSITY IRFs to shocks to the short-term inflation expectations distribution

Mean and dispersion shocks



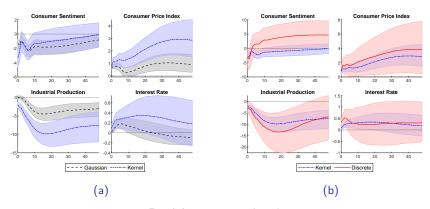
Mean shock



Positive mean shock. Baseline and perturbed distributions on impact for the discrete and Kernel approximations.



Mean shock



Positive mean shock

Panel (a): comparison between the Gaussian (black dashed lines), and the Kernel (blue dashed-dotted lines) approximations. Panel (b): comparison between the Kernel and the discrete (red continuous lines) approximation. 68% credible bands.

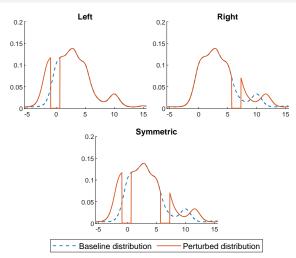


Dispersion shock

- We now analyze the macroeconomic effects of shocks increasing the dispersion of the expectations
- Unlike location, dispersion is a general concept, and there are countless ways to induce dispersion perturbation
- Among the infinite combinations, we focus on:
 - A variance shock (in a Gaussian sense) Variance Shock
 - A kurtosis shock (in a Student's t sense)
 - A tail shock Scaling, Kurtosis, and Tail Shock
 - Asymmetric shocks
- Location and Dispersion Shock when Macro variables are exogenous to expectations DIRFs: Alternative Identification

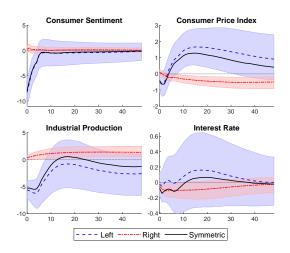


Asymmetric tail shock



Asymmetric Tail Shock. Contemporaneous effects of the left, right, and symmetric version of the tail shock.

Asymmetric tail shock



Asymmetric Tail Shock.

Median IRFs of the macroeconomic variables to the left, right, and symmetric version of the tail shock. 68% credible bands.

POLICY IMPLICATIONS

COMMUNICATION



Communication Shock

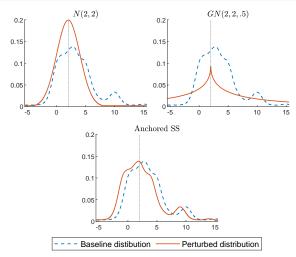
We devise a 'communication' shock \rightarrow the central bank can influence expectations with communication, such that the mean of the expectation distribution shifts to 2%.

Three different scenarios:

- Ideal. Gaussian with signal to noise = 1: $\mathcal{N}(2,2)$.
- Consensus to the target + high dispersion. Generalized Normal with heavy tails: $\mathcal{GN}(2,2,.5)$
- Consensus to the target preserving empirical shape. SS distribution centered at 2: "Anchored SS".

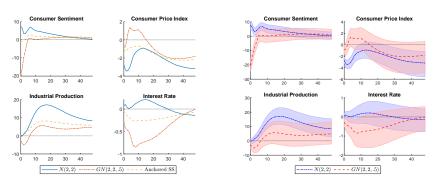
³The standard Normal is recovered by setting $\mathcal{GN}(2,2,2)$.

Communication Shock



Communication Shock. Contemporaneous effects on the inflation expectations distribution for the three communication shocks considered

Communication Shock



Communication Shock.

Panel (a): Median IRFs of the macroeconomic variables after the three shocks. Panel (b): Median IRFs and 68% credible bands of the macroeconomic variables after two selected shocks: the Normal (dotted blue) and the Generalized Normal (fat tails, dashed red).



Conclusions

- We have proposed a simple yet effective way of jointly modeling a time series of vectors and a time series of distributions to analyze the macroeconomic impact of changes in the distribution of short-term household inflation expectations
- Households inflation expectations affect macro variables (seemingly more than vice versa).
- A positive mean shock is stagflationary.
- Anchoring the expectation distribution is positive, conditionally on a limited variance
 - ⇒ Communication needs to target also the dispersion.



Additional

SLIDES

Event Studies

EVENT STUDIES

GFC AND COVID-19



GFC

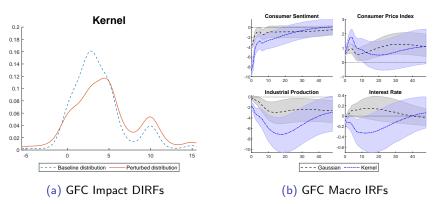


Figure: Panel (a): Baseline and perturbed distributions on impact for the Kernel approximation. Panel (b): comparison between the Gaussian (black dashed lines), and the Kernel (blue dashed-dotted lines) approximations.

Covid-19

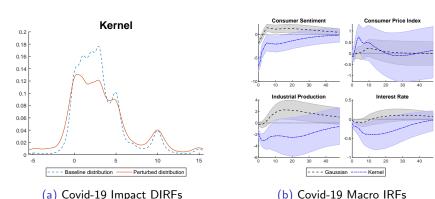


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The empirical problem

- We face the problem of incorporating in the same model a time series of distributions (functions) and a time series of indicators (scalars).
- Functional analysis approaches approximate the distribution with spline-basis functions (Chang et al., 2024; Meeks and Monti, 2023):

$$y_i = \int_{\Omega} X_i(\omega) \beta(\omega) d\omega = \sum_{k}^{\infty} x_{i,k} \beta_k \approx \sum_{k}^{K} x_{i,k} \beta_k$$

with K finite.

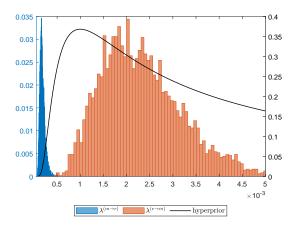
Pros: Good approximation performance. **Cons: Troublesome interpretation**: hard to endow the basis functions with economic meaning.



Why Proportions?

- Interpretability $\Rightarrow w_{j,t}^D$ simply equals the proportion of respondents who report a specific inflation value in a given interval $\{q_{j-1},q_j\}$ in a given period $t\Rightarrow$ Each series is endowed with a **direct economic** interpretation (anchored, fear of deflation, ...)
- The more general functional framework lacks parameters interpretability
- Having the proportions under control in the BVAR, we can perform DIRF by shocking these sub-classes of agents, where the perturbation of the distribution can be directly interpreted

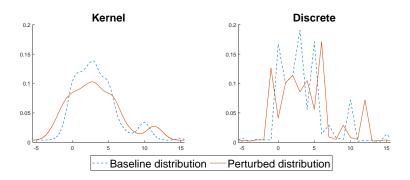
Cross-influence between blocks of variables



Hyperprior (black line, right axis scale) and posterior distribution of the two shrinkage parameters: $\lambda^{(m \to e)}$ (blue bins, left axis scale), and $\lambda^{(e \to m)}$ (orange bins, left axis scale).



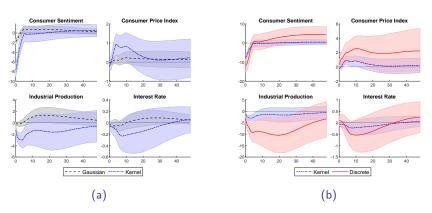
Positive standard deviation shock



Positive standard deviation shock. Baseline and perturbed distributions on impact for the discrete and Kernel approximations.



Positive standard deviation shock



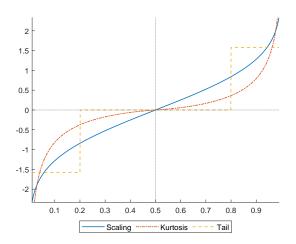
Positive standard deviation shock.

Panel (a): comparison between the Gaussian (black dashed lines), and the Kernel (blue dashed-dotted lines) approximations. Panel (b): comparison between the Kernel and the discrete (red continuous lines) approximation. 68% credible bands.





Dispersion shock

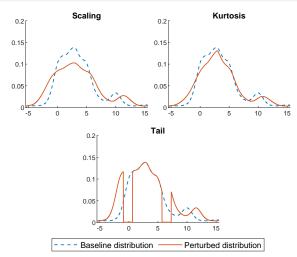


Scaling (Variance), Kurtosis, and Tail shock

Three quantile functions used to induce an increase in dispersion.



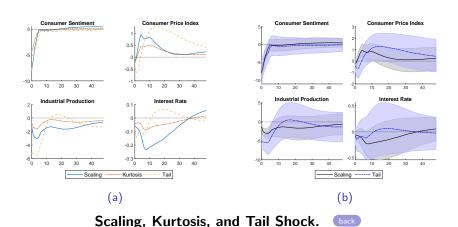
Dispersion shock



Scaling (Variance), Kurtosis, and Tail Shock Contemporaneous effects of the three different perturbations on the inflation expectations distribution.

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Dispersion shock



Panel (a): Median IRFs of the macroeconomic variables after the three shocks. Panel (b): Median IRFs and 68% credible bands of the macroeconomic variables after two selected shocks: the Scaling (blue dashed-dotted lines), and the Tail (black lines).

Who is populating the left tail? Panel regression

	A = [-50, 2]	A = [0, 2]	A = [1, 2]
Hold diploma	-1.085	-1.090	1.403**
	(0.007)	(0.007)	(0.006)
Hold degree	1.066***	1.052***	1.823***
	(0.003)	(0.003)	(0.003)
Income	1.296***	1.237***	0.727***
(quintiles)	(0.001)	(0.001)	(0.001)
If invest	1.751***	2.055***	3.937***
	(0.004)	(0.003)	(0.003)
Sentiment	8.198***	7.893***	3.718***
	(0.001)	(0.001)	(0.001)
R ²	0.0323	0.0437	0.061
Observations	114847	114847	114847
Effects	Time	Time	Time
Controls	Yes	Yes	Yes
VCV Robust	Yes	Yes	Yes

$$P[E_{i,t}(\pi_{t+12}) \in A] = \mu_t + \mathbf{x}_{i,t}' \boldsymbol{\beta} + \epsilon_{i,t}$$

- Pooled OLS regression on inflation expectations
- Dependent variables take value of 1 if individual i has an inflation expectation that falls within the set A, 0 otherwise
- For interpretability, all the coefficients has been multiplied for 100.



Inflation Expectation Shocks in Macro: Related Literature

Surprising small attention to macro effects of inflation expectation shocks

A shock that increase short term inflation expectations have a negative macroeconomic impact $(\downarrow y, \uparrow \pi)$ (Ascari et al., 2023).

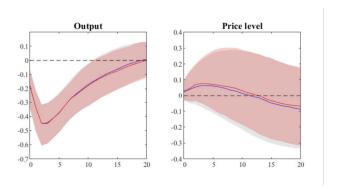
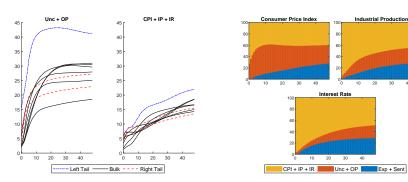


Figure: VAR Irfs to an inflation expectation shock, from Ascari et al. (2023).





FEDV: Macro vbls exogenous to expectations



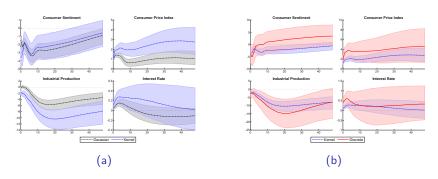
Forecast Error Variance Decompositions.

All the macroeconomic variables exogenous to Expectations, but endogenous to Uncertainty indicators and the oil price.

Contributions of the different identified blocks of indicators to the FEVDs of the expectations (left panel) and of macroeconomic variables (right panel) at different horizons (x-axis).



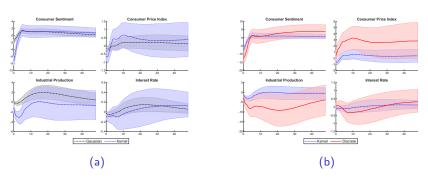
Location Shock: Macro vbls exogenous to expectations



Positive location shock with no contemporaneous identification between the fundamental macro and expectations blocks. The figure shows the effects of a shock that increases the location of the inflation-expectations distribution of 1% on impact. No zero restrictions are imposed between the macro block and the expectations blocks.



Variance shock: Macro vbls exogenous to expectations



Positive standard deviation shock with no contemporaneous identification between the fundamental macro and expectations blocks. The figure shows the effects of a positive standard deviation shock. No zero restrictions are imposed between the macro block and the expectations blocks.



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