

# Expectations Are Observables. And We Haven't Even Started Yet ...

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CESifo, ifo

Keynote – University of Hamburg – Workshop “Household  
Surveys in Macroeconomics”

October 6, 2016.

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Even though macroeconomics has benefitted tremendously from such data.

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- Expectations.
- Subjective reasons (unfortunately on the firm side).



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Particles do not have a sense of future.

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Saying it with Heidegger (*Sein und Zeit*): *Ein Existenzial des Daseins ist Zeitlichkeit. Dasein ist Sorge, ist Sich-vorweg-sein.*

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Studies empirically with survey data how firms form and update their expectations.



# Critique I

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Aside: recall the revealed preference approach to microeconomics.

## Critique II

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In a sense, rational expectations took expectations *as economic data* off the table, because the models took care of it.

## Both Strands of Critique Together

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This was not a strictly necessary development, because one could have tested rational expectations plus the model assumptions jointly against expectational and other economic data – but the behaviorist streak in economics was quite happy to get rid of expectations as data.

So, we ended up with testing big rational expectations (often DSGE) models on “objective” outcome data. Aside: this is orthogonal to the estimation-calibration distinction.

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Both household and firm level expectation data are reasonably predictive of the business cycle, and contain often a strong news component about future productivity. (Barsky and Sims, 2012, American Economic Review: “Information, Animal Spirits, and the Meaning of Innovations in Consumer Confidence”.)



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- Rational expectations is still an important benchmark / first pass / default – but no longer the Alpha and Omega of economics.
- Economists see value again in testing not entire large models, but certain key elements / modules of them (the way they had been doing it in earlier times – think of all the PIH tests in the literature).

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This development has certainly been reinforced if not triggered by recent macroeconomic events and a resulting general openness / willingness to rethink the foundations of the field.

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*We need to encourage those who are trying to learn more about how people actually form expectations. [...] At the same time, we need to be a lot more flexible in our thinking about models and theory, so that they can be firmly grounded in this improved empirical understanding.*

# “What Can Survey Forecasts Tell Us about Information Rigidities?”

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- Expectations react gradually to news, ruling out full-information models.
- Disagreement in inflation forecasts does not seem to respond to shocks, which means that *noisy* information models are favored over *sticky* information models.

# “Is The Phillips Curve Alive and Well After All? Inflation Expectations and the Missing Disinflation”

Coibion and Gorodnichenko, 2015, American Economic  
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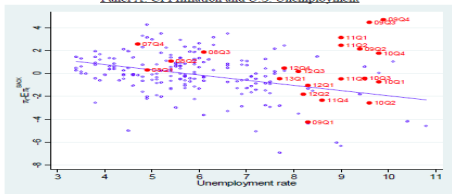
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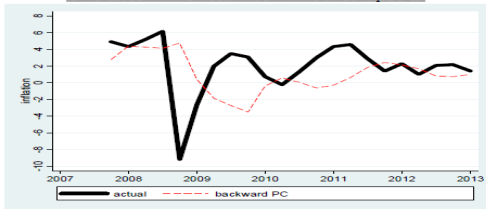
Use direct inflation expectations data to “save” the *Phillips Curve*, an important ingredient for monetary macroeconomics.

# “Is The Phillips Curve Alive and Well After All? Inflation Expectations and the Missing Disinflation”

FIGURE 1: THE MISSING DISINFLATION  
Panel A: CPI Inflation and U.S. Unemployment



Panel B: CPI Inflation and Predicted Inflation from Phillips Curve



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- Reason: oil price spikes during the time.

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- Taylor rule type reasoning especially prevalent when labor markets are weak (rational inattention story?).

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- Forecasts of future inflation: very uncertain, dispersed and volatile.

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- Can thus estimate directly the *Euler equation* and the corresponding *elasticity of intertemporal substitution*, a key macroeconomic parameter.
- Recall, that the Euler equation features expected consumption growth, while the literature traditionally has estimated Euler equations on realized consumption growth (Attanasio and Weber in many papers), essentially presupposing rational expectations.

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  - Inflation anchoring.
- Test the Euler equation and estimate key structural parameters: elasticity of intertemporal substitution.

# Background: Stabilization Policy and Transmission

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In the Neokeynesian model:

monetary policy  
(at the ZLB)

fiscal policy  
 $\Delta G, \Delta VAT$



inflation expectations



demand / spending

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- Economic policy trust / confidence / uncertainty

## Some Quotes

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- *“Another possible effect is a temporary climb in inflation expectations. Ordinarily, this would be undesirable. But in the current situation, where nominal interest rates are constrained because they can't go below zero, a small increase in expected inflation could be helpful. It would lower real borrowing costs, and encourage spending on big-ticket items like cars, homes and business equipment.” – Christina Romer, New York Times, November 2011*

# Literature

*“But he could have paid the balance of 25 marks at any time and thus have made the teeth his own. If he did not do so, it was because he had heard from many people that the accession of the Nationalists to power would be followed by inflation of currency, [...]. And yet business was better than one might have expected during this rather quiet winter season. The talk of inflation induced many people to spend their money on household needs instead of putting it in the savings bank.”*

From: Lion Feuchtwangers “The Oppermanns” (in the translation by Ruth Gruber), about the business dealings of the furniture salesman Markus Oppermann with his dentist and his clients right after the rise to power of the Nazi party in January 1933.

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- Quantity expectations matter positively (Old Keynesianism appears to be alive and well).
- Panel dimension: for “good” inflation forecasters / informed households, we get a significantly positive sign.
- Households understand relative prices: expected house price increases lead to a higher readiness to buy a house now.

# Other Literature

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  - Ichiue and Nishiguchi (2015): Japanese households have a positive sign (long life under a ZLB regime?)



# “The Effect of Unconventional Fiscal Policy on Consumption Expenditure”

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- Use a pre-announced VAT increase in Germany to instrument increases in inflation expectations to uncover a causal effect of inflation expectations on spending.

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- Use GFK consumer survey data.

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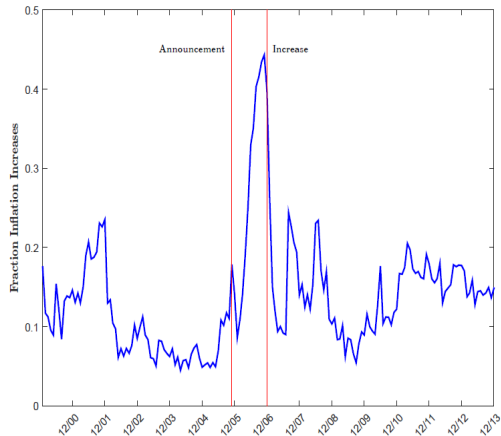
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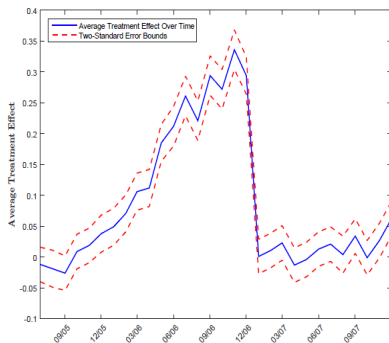
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- Germany part of Euro zone and no independent monetary policy.
- Nominal rate did not increase to offset inflation expectations.

# Inflation Expectations Increased in Germany ...



# ... and So Did Readiness to Spend Relative to Other European Countries



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- Communication of policy is key.

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- ... helps us test a key part of an important stabilization policy mechanism;
- ... tells us ways to conduct the management of inflation expectations - communication and salience seem key.

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“What Drives Aggregate Investment? Evidence from German Survey Data,” joint with Peter Zorn, 2016, working paper.

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- Even more specifically: the fluctuations of the year-over-year investment growth rate.
- Novel approach: narrative, survey-based.

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- In other words: we move from investment determinants to economic shocks.

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- 1 First distinguishing technology shocks versus non-technological shocks (need a minimal set of assumptions).
- 2 Making more assumptions and putting more structure on the empirical model: extract also aggregate demand and finance shocks.

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We see the advantage of a survey-based approach towards identifying shocks in its putative *directness*: the survey respondents (*decision makers*) directly report whether their investment activity in a given year was influenced by, for instance, technological considerations and, if so, how strongly.

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See, for instance, Romer (2004, 2010).

Also: these data are confidential, so there is probably little danger of decision makers strategically lying.

# Preview of Results

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- ② But: technology shocks explain (only) roughly *one third* of the variance of aggregate (manufacturing) investment growth.
- ③ Aggregate demand shocks: explain roughly *one half*. Find suggestive evidence that these demand shocks are sentiment shocks.



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- Boom in the 1990s and the slump in the early 2000s clearly related to technological factors.
- Recovery from the slump in the latter half of the 2000s is a positive sentiment shock.
- Great Recession shock looks like a combination of a negative sentiment shocks and a significant technological slow down.

# Some Background on the Survey

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- Manufacturing.
- Starts in 1955, but the for us relevant questions start only in 1989. Our baseline sample period: 1989-2008, to focus on regular year-to-year business cycle fluctuations first.

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- Drawbacks:
  - Investment determinants only annually asked (fall).
  - Relatively short time series, few data, though sectoral disaggregation can help here.

# Our Two Questions

## Q1. Gross Fixed Capital Formation in Fiscal Year *[Last Year]*

*[Last Year]* \_\_\_\_\_  
(in 1000 Euro)

## Q2. Investment Determinants *[This Year]*

Our investment activity in the Old Laender in *[This Year]* was positively/negatively affected by:

Investment Determinant	<i>[This Year]</i>				
	strongly positive influence	weakly positive influence	no influence	weakly negative influence	strongly negative influence
Sales Situation and Expectation	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Finance	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Profit Expectation	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Technical Factors	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Macro Policy Environment	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Other	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
<i>[Codification]</i>	<i>[+2]</i>	<i>[+1]</i>	<i>[0]</i>	<i>[-1]</i>	<i>[-2]</i>

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Quantification: -2 (strongly negative influence), -1 (weakly negative influence), 0 (no influence), +1 (weakly positive influence), or +2 (strongly positive influence)

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Then the aggregate investment growth rate,  $\Delta I_t^{IFO}$ , is given by:

$$\Delta I_t^{IFO} = \sum_{i=1}^{N_t} \omega_{it-1} \frac{inv_{it} - inv_{it-1}}{inv_{it-1}}$$

# Aggregation

Similarly, let  $x_{it}$  denote one of the six firm-level investment determinants.

# Aggregation

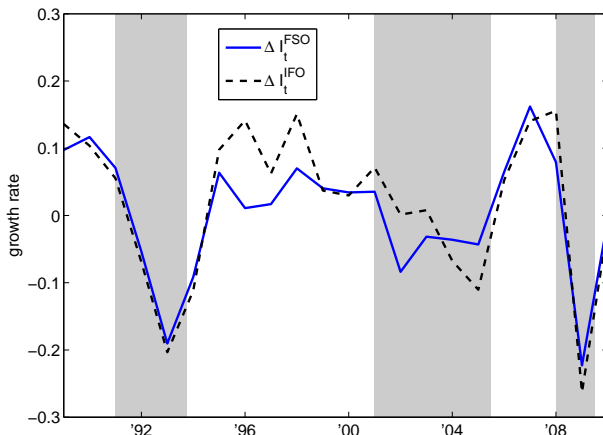
Similarly, let  $x_{it}$  denote one of the six firm-level investment determinants.

Then, for every investment determinant, we aggregate up to an investment determinant index,  $X_t$ , as follows:

$$X_t = \sum_{i=1}^{N_t} \omega_{it} x_{it}$$

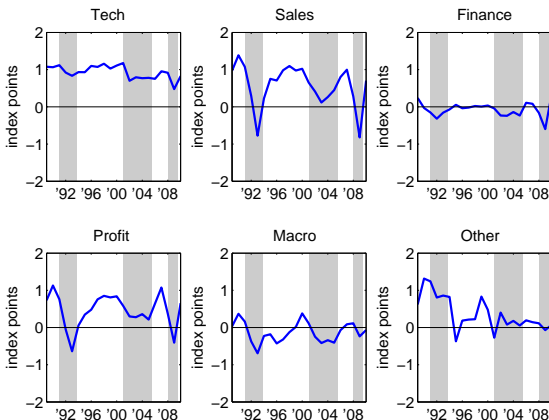
# A First Look at the Data - Investment Growth Rate

Measures of aggregate investment growth ( $\rho = 0.91$ )



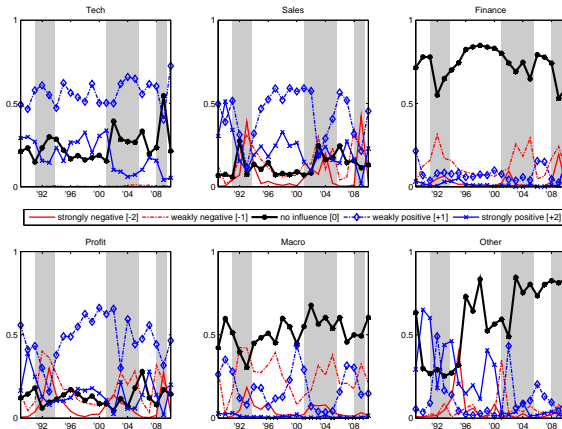
# A First Look at the Data - Investment Determinant Indices

## Aggregate investment determinant indices





# A First Look at the Data - Investment Determinant Indices



# A First Look at the Data

	Tech	Sales	Finance	Profit	Macro	Other	$\Delta I_t^{FSO}$
<b>Baseline Sample Results (1989-2008)</b>							
<i>Panel A:</i>							
Tech	1						
Sales	0.6071***	1					
Finance	0.4574**	0.5801***	1				
Profit	0.5434***	0.9434***	0.5920***	1			
Macro	0.5253***	0.7337***	0.4674***	0.7746***	1		
Other	0.1676	0.0879	-0.1100	0.0241	0.2073	1	
<i>Panel B:</i>							
$\Delta I_t^{FSO}$	0.5029***	0.8392***	0.6279***	0.8849***	0.7601***	-0.1073	1
<i>Panel C:</i>							
$\mu$	0.9602	0.6347	-0.0641	0.4947	-0.1275	0.4062	0.0166
$\sigma$	0.1490	0.4889	0.1391	0.4173	0.2846	0.4567	0.0832

# Economic Content: Tech

Mean of Tech, conditional on maintenance investment:

Tercile	N	Mean(Tech)
1	9521	1.048428
2	10765	0.8196647
3	8156	0.7707991

## Economic Content: Tech

Mean of Tech, conditional on investment in restructuring and rationalization:

Tercile	N	Mean(Tech)
1	11341	0.7818501
2	7690	0.9721699
3	9411	1.125008

Difference in means statistically significant at the 1% level.

# Economic Content: Tech

Mean of |Tech|, conditional on Eurostat's Technology Classification:

Industries	N	Mean( Tech )
Low-technology	10911	0.8956025
Medium-low-technology	8448	0.9669374
Medium-high/High-technology	8645	0.9783141

# Economic Content: Finance

Mean of  $|\text{Finance}|$ , conditional on share of external finance (IFO survey):

Share of External Finance	N	Mean( $ \text{Finance} $ )
up to 33.33%	11564	0.2520984
33.33% to 66.66%	2194	0.5049183
above 66.66%	1982	0.5344153

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- 2 Within non-technological shocks:
  - 1 Orthogonalize Profit, Macro and Other with respect to Technology, Finance and Sales.
  - 2 Baseline: Orthogonalize Finance with respect to Sales.  
External finance is not that important in Germany.

# Orthogonalization - Regression Framework

$$\begin{aligned}
 \text{Tech}_t &= v_1 + \widehat{\text{Tech}}_t \\
 \text{Sales}_t &= v_2 + \delta_{21} \widehat{\text{Tech}}_t + \widehat{\text{Sales}}_t \\
 \text{Finance}_t &= v_3 + \delta_{31} \widehat{\text{Tech}}_t + \delta_{32} \widehat{\text{Sales}}_t + \widehat{\text{Finance}}_t \\
 \text{Profit}_t &= v_4 + \delta_{41} \widehat{\text{Tech}}_t + \delta_{42} \widehat{\text{Sales}}_t + \delta_{43} \widehat{\text{Finance}}_t + \widehat{\text{Profit}}_t \\
 \text{Macro}_t &= v_5 + \delta_{51} \widehat{\text{Tech}}_t + \delta_{52} \widehat{\text{Sales}}_t + \delta_{53} \widehat{\text{Finance}}_t + \delta_{54} \widehat{\text{Profit}}_t + \widehat{\text{Macro}}_t \\
 \text{Other}_t &= v_6 + \delta_{61} \widehat{\text{Tech}}_t + \delta_{62} \widehat{\text{Sales}}_t + \delta_{63} \widehat{\text{Finance}}_t + \delta_{64} \widehat{\text{Profit}}_t + \delta_{65} \widehat{\text{Macro}}_t + \widehat{\text{Other}}_t
 \end{aligned}$$

Remark: we verify that orthogonalized series are not autocorrelated.

# Final Regression

$$\Delta I_t^{FSO} = c + \beta_1 \widehat{\text{Tech}}_t + \beta_2 \widehat{\text{Sales}}_t + \beta_3 \widehat{\text{Finance}}_t + \beta_4 \widehat{\text{Profit}}_t + \beta_5 \widehat{\text{Macro}}_t + \beta_6 \widehat{\text{Other}}_t + u_t$$

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Residuals do not display significant autocorrelation.

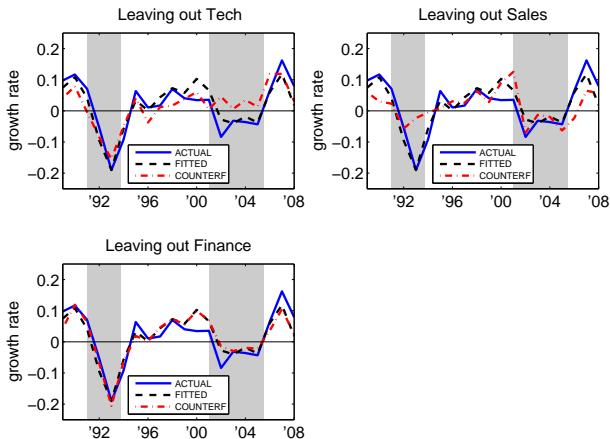
# Variance Decomposition

Relative contributions to the  $R^2$  (in percent) with different orthogonalizations of Tech - Total  $R^2$ : 84 percent!

<i>Orthogonalization:</i>	Tech	Tech
	Sales	Finance
	Finance	Sales
	Profit	Profit
	Macro	Macro
	Other	Other
$\widehat{\text{Tech}}$	30.19	30.19
$\widehat{\text{Sales}}$	53.89	33.73
$\widehat{\text{Finance}}$	3.73	23.89
$\widehat{\text{Profit}}$		7.65
$\widehat{\text{Macro}}$		1.67
$\widehat{\text{Other}}$		2.87
$R^2$		0.8377



# Counterfactuals



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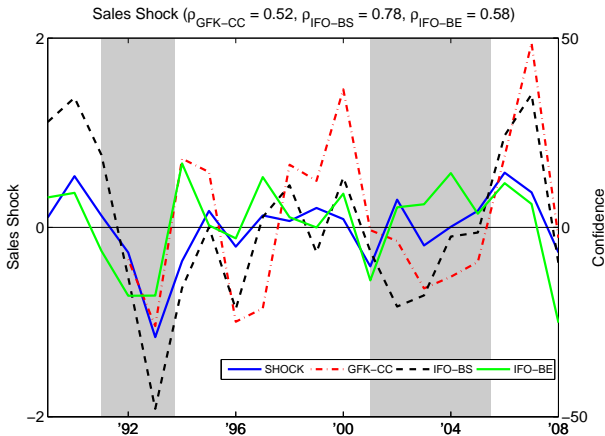
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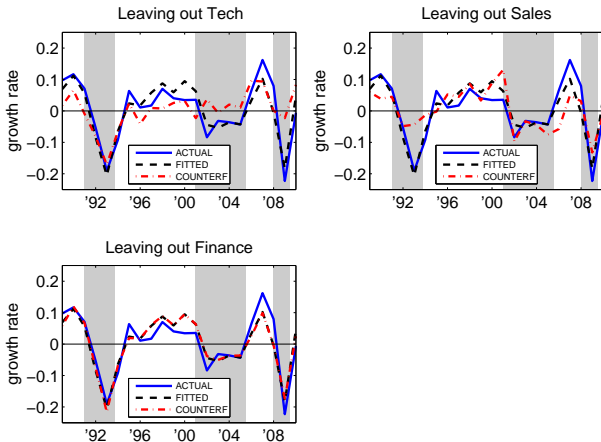
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- Disaggregate results for Laender and 2-digit industries tell the same story.



# What Are the Demand Shocks?



# Counterfactuals - Extended Sample



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- Especially expectation data have a lot to teach us about important macroeconomic ideas and issues.
- I would argue we should go a step further and ask economic agents why they did what they did.
- I applied this idea of looking at “subjective reasons” to study the ultimate business cycle question: What drives aggregate fluctuations?