

GDP-at-Risk: Research and Policy Agenda

David Aikman

Qatar Centre for Global Banking and Finance, King's College London

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Outline

- The concept of GDP-at-Risk and its potential use in macroprudential policy
- Two research papers:
 - “Credit, Capital and Crises: A GDP at Risk Approach” with Jonathan Bridges (BoE), Sinem Hacioglu-Hoke (BoE), Cian O’Neill (BoE) and Akash Raja (LSE)
 - “A Tale of 3 Occasionally-Binding Constraints”, with Kristina Bluwstein (BoE) and Sudipto Karmakar (BoE)
- Concluding thoughts on where this literature is heading

GDP-at-Risk and macroprudential policy

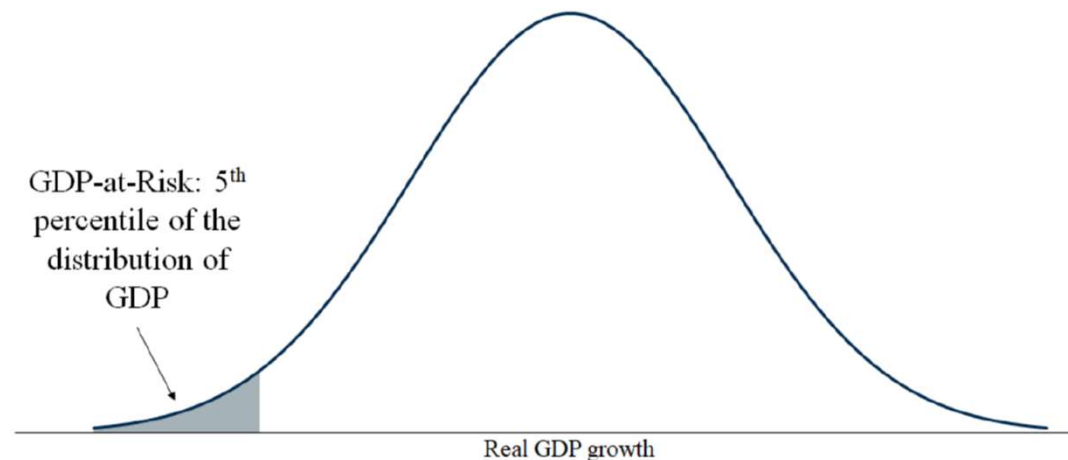
- Macropru as a policy framework is severely hampered by the lack of a quantitative objective
- This was not a problem in the immediate aftermath of the Global Financial Crisis as the direction of policy was clearly to build resilience (Barwell, 2021)
- But a decade plus on from that episode, the countercyclical component of the macroprudential framework is rudderless

Problems this creates

- Accountability is impeded
- Decision making is poorer
 - Ambiguity fosters the status quo bias
 - The attention of policymakers may gravitate to more measurable goals like supporting competition and growth
- Imagine the task of Monetary Policy Committees right now if we didn't have inflation targets in place!

What is GDP-at-Risk?

- GDP-at-Risk is the most promising idea we have developed so far for quantifying the objectives of macroprudential policy
- It's an estimate of the most severe recession we might reasonably expect to experience over a defined period ahead, eg the next 3 years



Attractive concept for several reasons:

- It provides a numeraire, ie it allows us to express financial stability risks in common units
- GDP is a well understood concept
- The focus on “tail risk” dovetails with the focus of monetary policy on the central outlook
 - A GDP-at-Risk forecast could play a role akin to that of the inflation forecast in monetary policy deliberations

But it also has drawbacks

- Understanding the relationship between financial sector developments and macro tail risk is a significant challenge – is it beyond our ability?
- As we're seeing, there are many other sources of risk for the real economy beyond financial crises – so a financial stability “target” for GDP-at-Risk doesn't make much sense

Some key contributions to the literature

- Early literature: Cecchetti (2006), Cecchetti and Li (2008), De Nicolo and Lucchetta (2012)
- Seminal paper: Adrian et al. (2019), “Vulnerable Growth”, AER
- Voluminous literature following this contribution:
 - Adrian et al (2019): Term structure of Growth at Risk
 - Duprey and Ueberfeldt (2020): Managing GDP at Risk
 - Kiley (2019): Unemployment Risk
 - Chavleishvili and Manganelli (2020): Quantile VAR
 - Boyarchenko et al (2020)
 - Covi et al (2022)
 - Suarez (2021)
- Plagborg-Moller et al (2020): When is growth at risk? BPEA

Some of my work in this area

- “Credit, capital and crises: A GDP-at-Risk approach”
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- “A tale of three occasionally-binding constraints: a modelling approach to GDP-at-Risk”
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What this paper does

- This paper examines the relationship between a wide set of macroprudential indicators and GDP-at-Risk
- 2 novelties relative to existing literature:
 - We include measures of banking system resilience in the analysis, allowing us to account for the impact of Basel 3 on GDP-at-Risk
 - We focus on a horizon 3-5 years ahead, which is arguably the relevant policy horizon for considering prophylactic macroprudential policy actions

Data

- Cross-country panel dataset, 16 advanced economies, 1980Q4-2017Q4
- 5 macroprudential indicators:
 - 3-year change in private non-financial sector credit to GDP ratio
 - 3-year growth in real house prices
 - Current account deficit (% of GDP)
 - Realised volatility in equity prices
 - Banking system tangible common equity to total assets ratio
- Macroeconomic controls: policy rate, inflation, lagged GDP growth

Quantile regression approach

We follow Canay (2011) 2-step approach:

First step: Estimate the following linear pooled panel model by OLS

$$y_{i,t+h} = \alpha_i^h + \gamma^h X_{i,t} + \epsilon_{i,t},$$

where $y_{i,t+h} = \frac{(Y_{i,t+h} - Y_{i,t})}{h/4}$ and X are conditioning variables

The fixed effects can be estimated as: $\hat{\alpha}_i^h = \frac{1}{N} \sum_{i,t} (y_{i,t+h} - \hat{\gamma}^h X_{i,t})$

Quantile regression approach

Second step: We estimate the quantile regression coefficients:

$$\hat{\beta}_{\tau}^h = \operatorname{argmin}_{\beta^h} \sum_{i,t} \rho_{\tau}(y_{i,t+h}^* - X_{i,t}\beta_{\tau}^h)$$

where $y_{i,t+h}^* = y_{i,t+h} - \hat{\alpha}_i^h$.

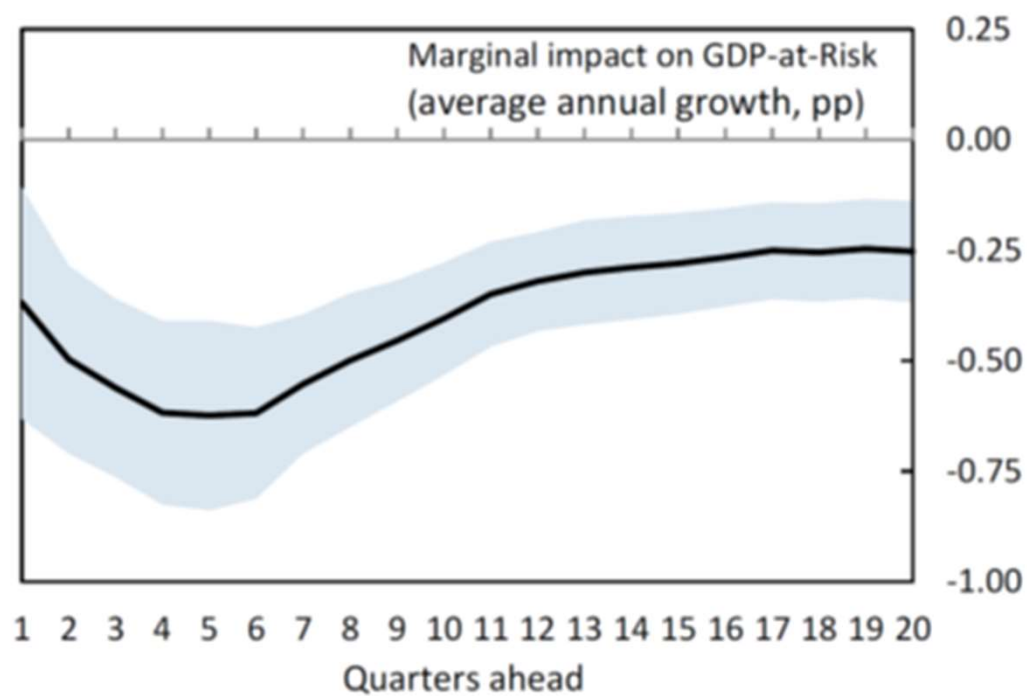
We estimate betas for 1-20 quarters ahead using local projections

Predicted 5th percentile of GDP growth = **“GDP-at-Risk”**

GDP-at-Risk response following 1 std increase in:

(A) Credit-to-GDP (3 year pp change)

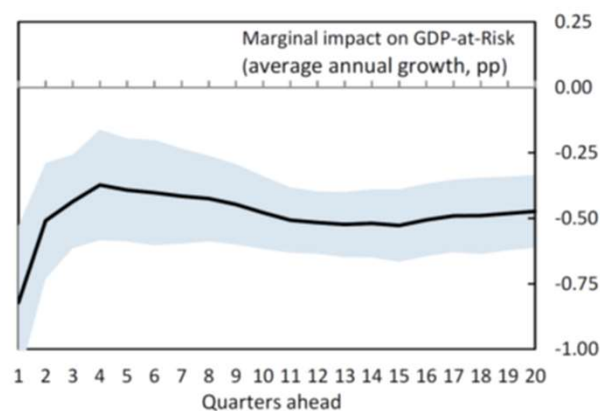
Note: +-1 standard error confidence intervals



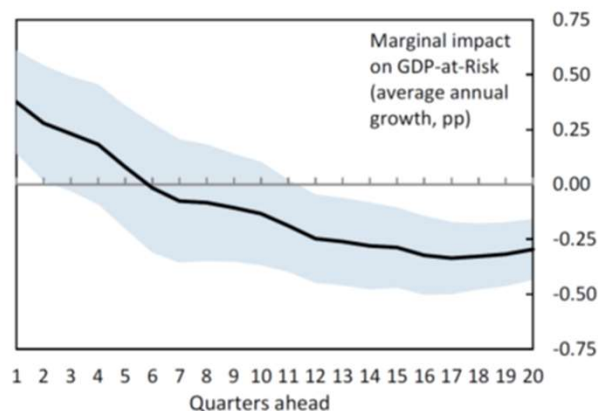
GDP-at-Risk response following 1 std increase in:

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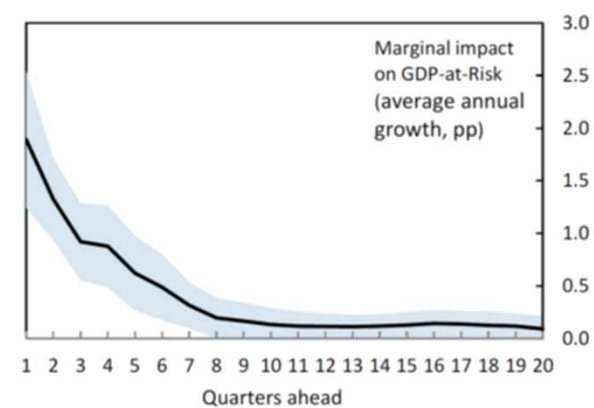
(B) Current account deficit



(c) Real house price growth (3 year)



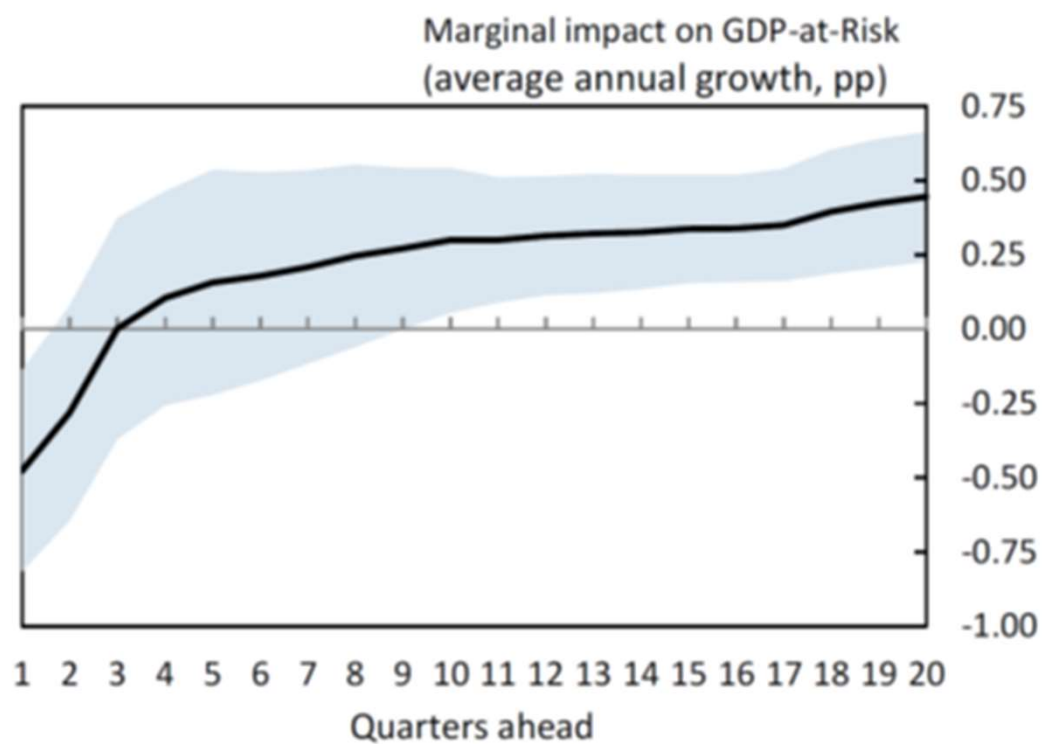
(D) Volatility



GDP-at-Risk response following 1 std increase in:

Note: +-1 standard error confidence intervals

(E) Bank capital (TCE) ratio

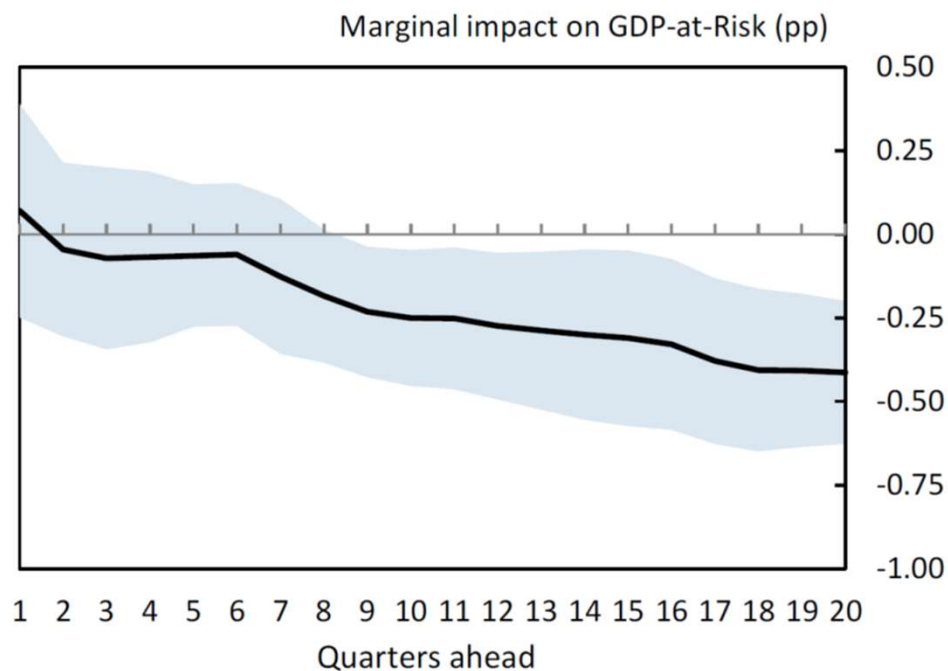


Summary of GDP-at-Risk impacts over 3-years

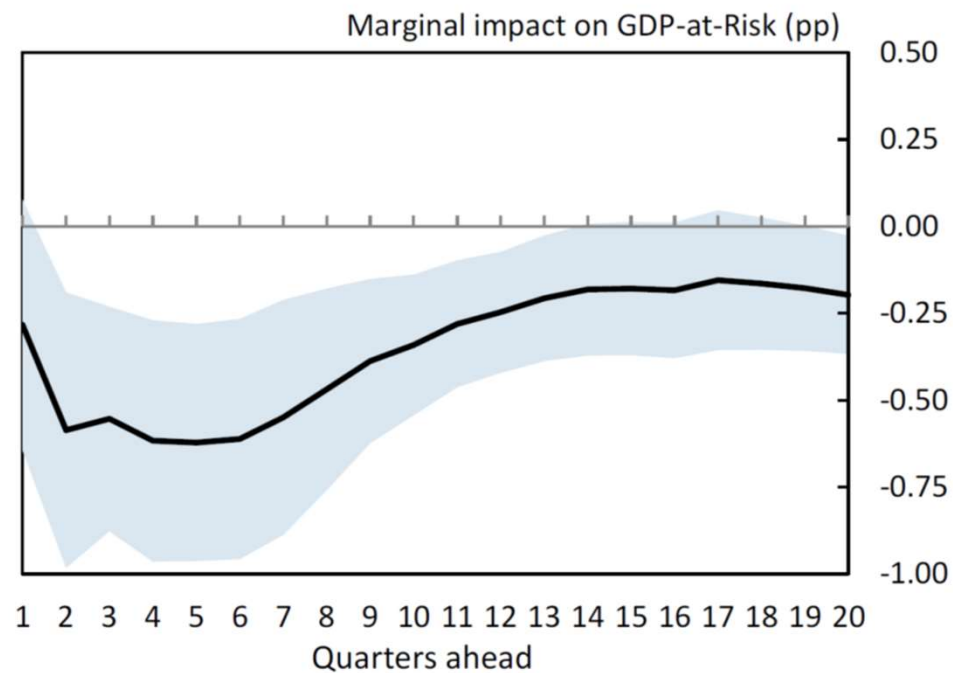


Household vs firm credit impact on GDP-at-Risk

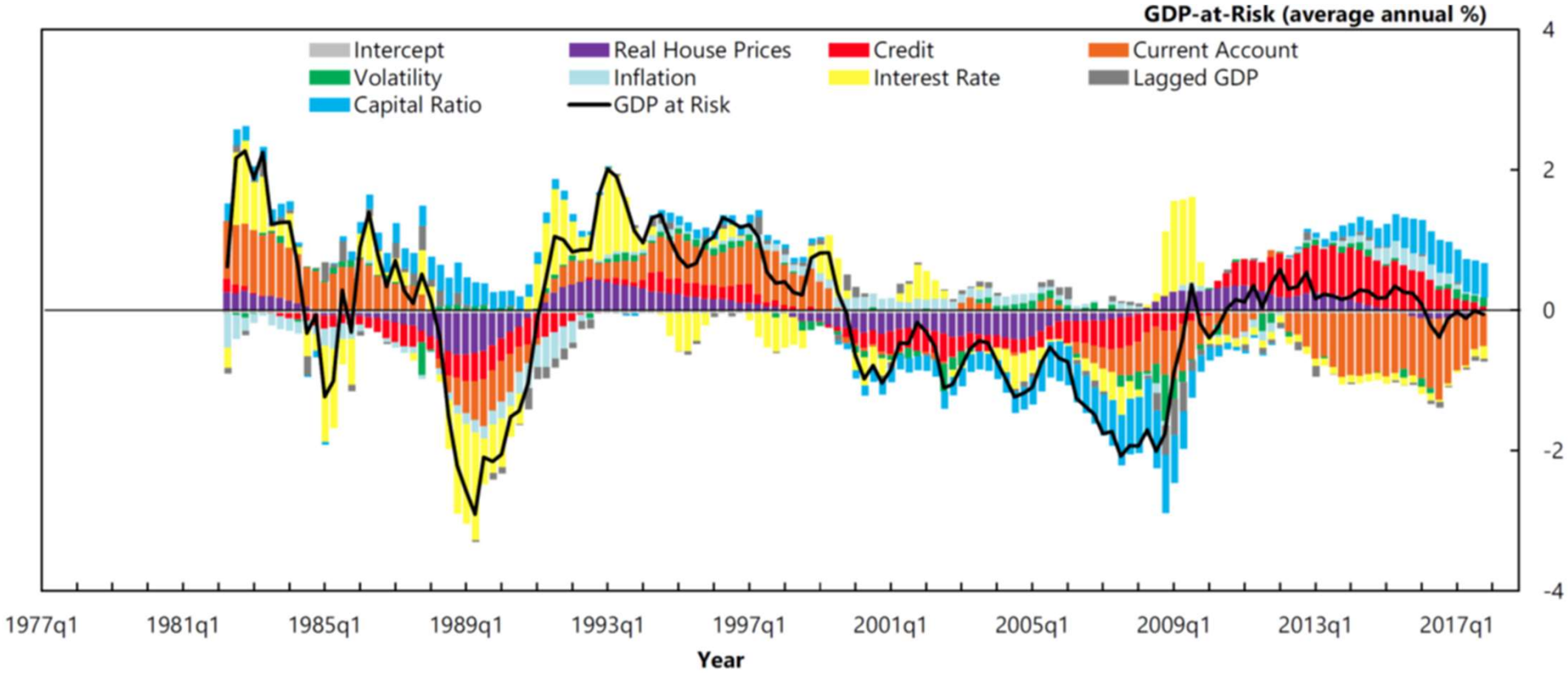
(A) Household credit-to-GDP (3 year pp change)



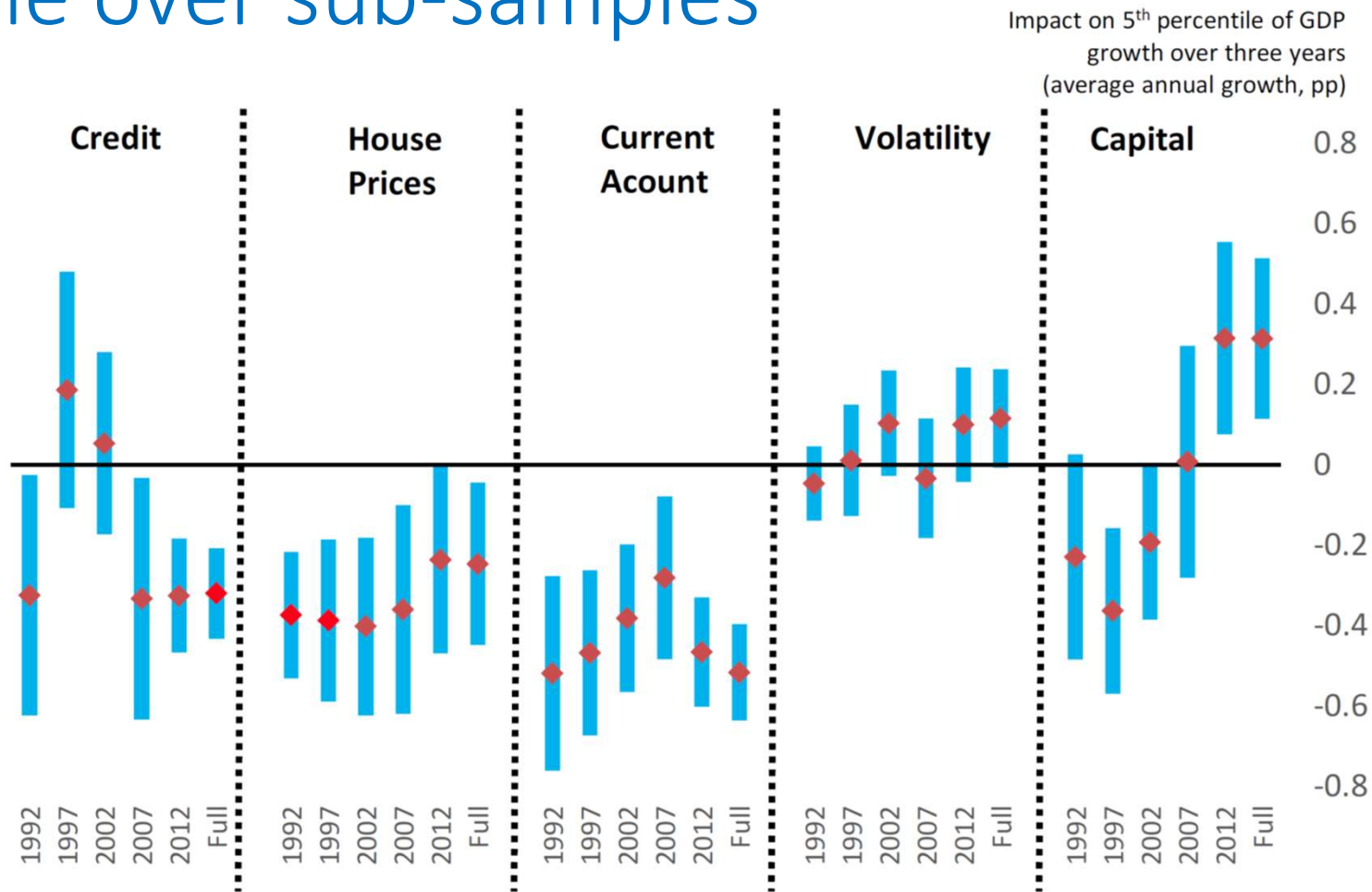
(B) Corporate credit-to-GDP (3 year pp change)



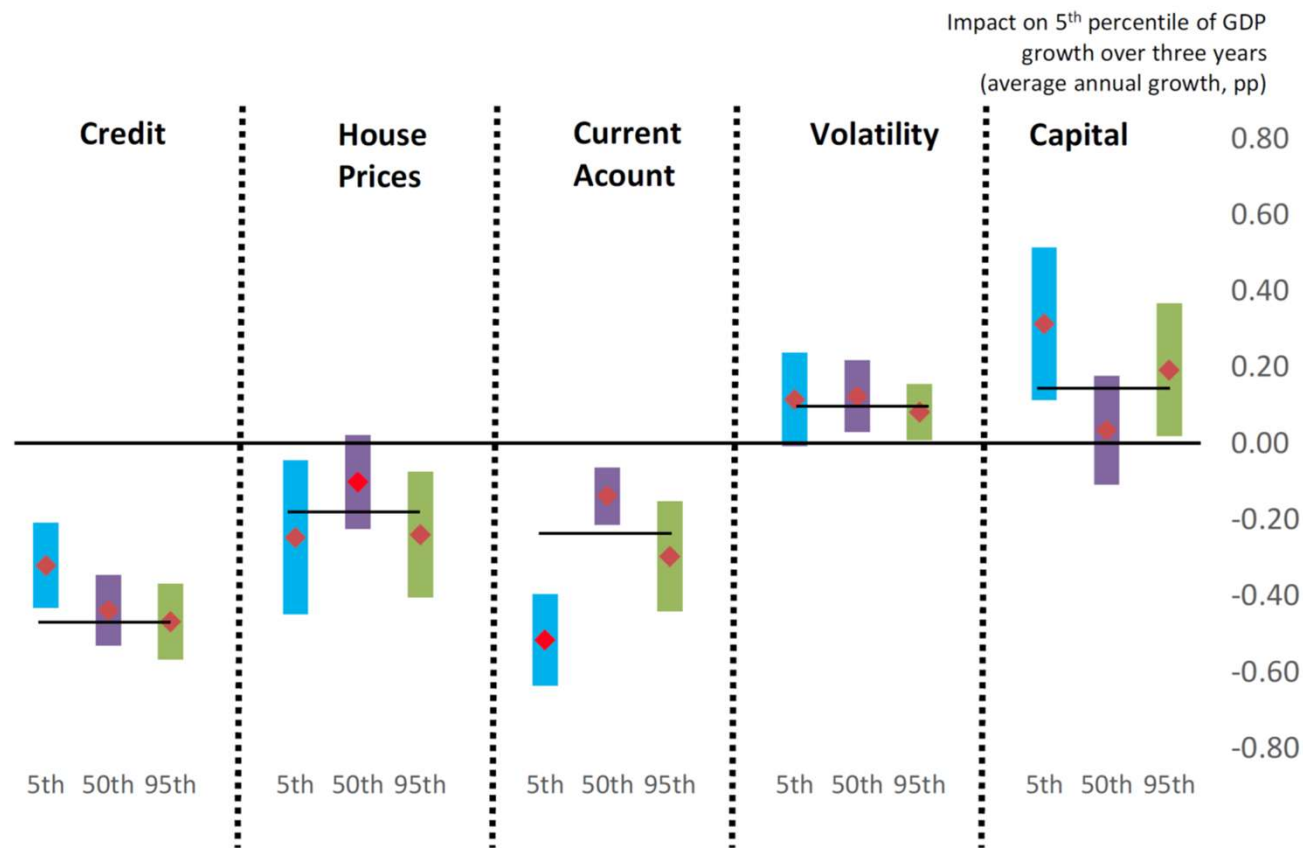
Decomposition of GDP-at-Risk 3-years ahead: UK



Some GDP-at-Risk coefficient estimates are unstable over sub-samples

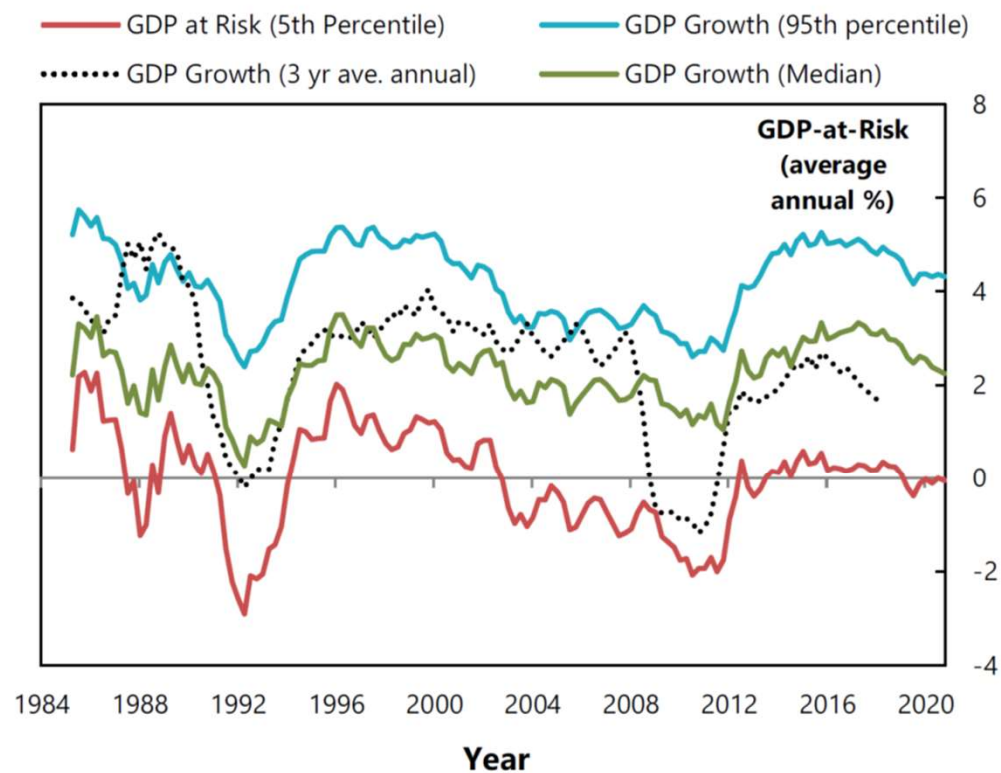


Some variables change the location rather than the shape of the growth distribution

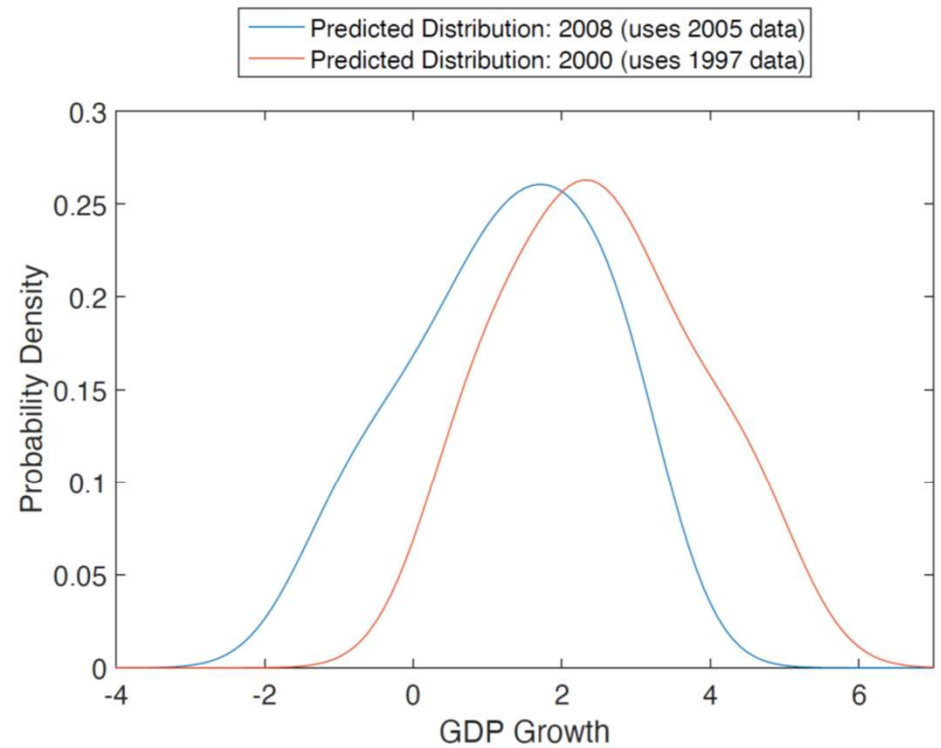


Predicted GDP growth density

(A) Forecast from 3 years previously vs. actual outturn



(B) Predicted density (3 years ahead)



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- **“A tale of three occasionally-binding constraints: a modelling approach to GDP-at-Risk”**
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A tale of 3 occasionally-binding constraints

- Hitherto, most studies of GDP-at-Risk have been purely empirical
- Few attempts to provide a more structural interpretation of the concept
- A model of GDP-at-Risk is useful for:
 - Running “what if” scenarios, eg what’s the impact of a persistent inflation shock?
 - Understanding interaction effects, eg if the economy is further away from the ELB, what does that mean for the vulnerability of the banking system?
 - Analysing policy strategies

What this paper does

- We build a semi-structural New Keynesian model
- The model incorporates three non-linear constraints:
 - An effective lower bound on nominal interest rates
 - A credit crunch in credit supply when bank capital is depleted
 - Deleveraging by borrowers when debt service burdens become excessive
- We calibrate the model for the UK and use simulations to understand its properties

The model in words

- The macro block is pretty conventional: IS curve, Phillips curve, Taylor rule
- The financial block includes:
 - Credit demand, linking credit to output and the (spread adjusted) real interest rate
 - Credit supply, linking credit to the bank capital ratio
 - Bank capital dynamics, via which bank net worth varies autoregressively and depends on the level and change in nominal interest rates plus (procyclical) credit losses

Nonlinear constraints

1. Effective lower bound on policy rate

2. Bank capital constraint

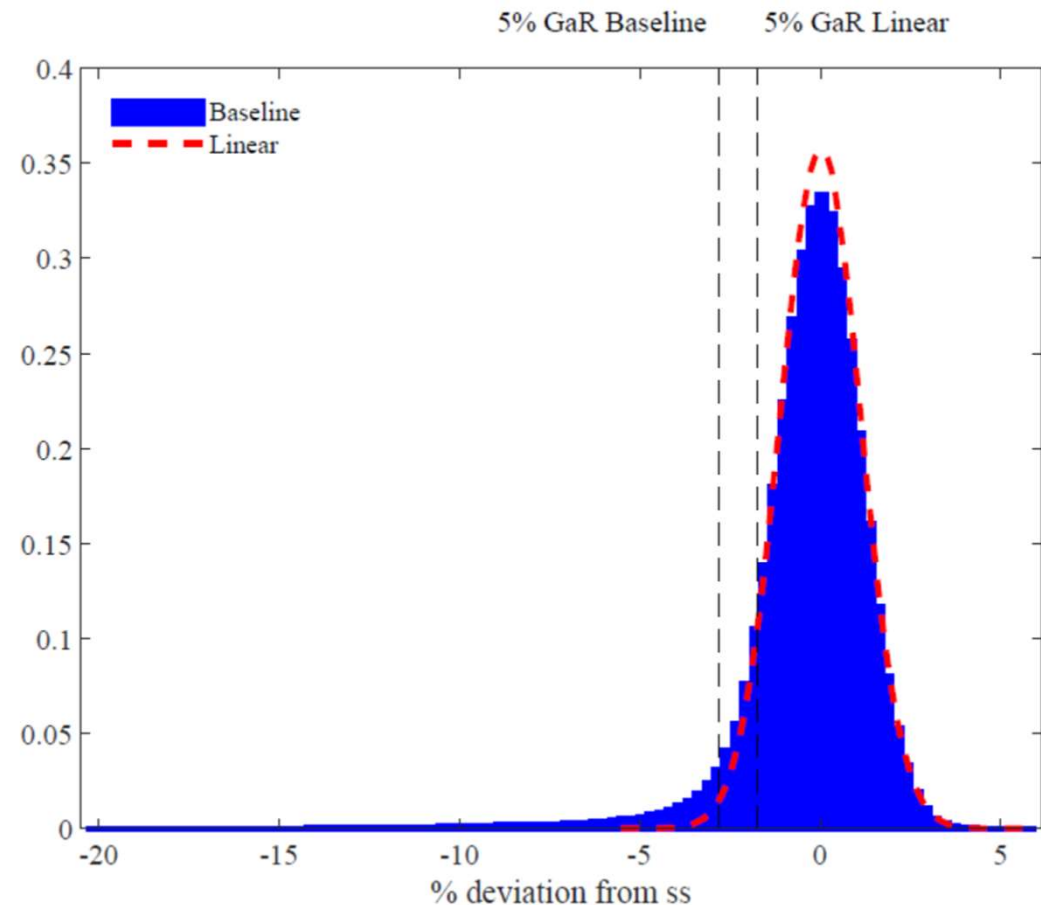
- In normal times, loan supply is insensitive to bank capital
- But when capital ratios drop below a threshold, loan supply falls abruptly and becomes highly sensitive to bank capital

3. Debt deleveraging constraint

- Borrowers begin deleveraging when their debt service ratios (DSRs) cross a boundary
- Deleveraging is a simultaneous shock to the IS and credit demand equations

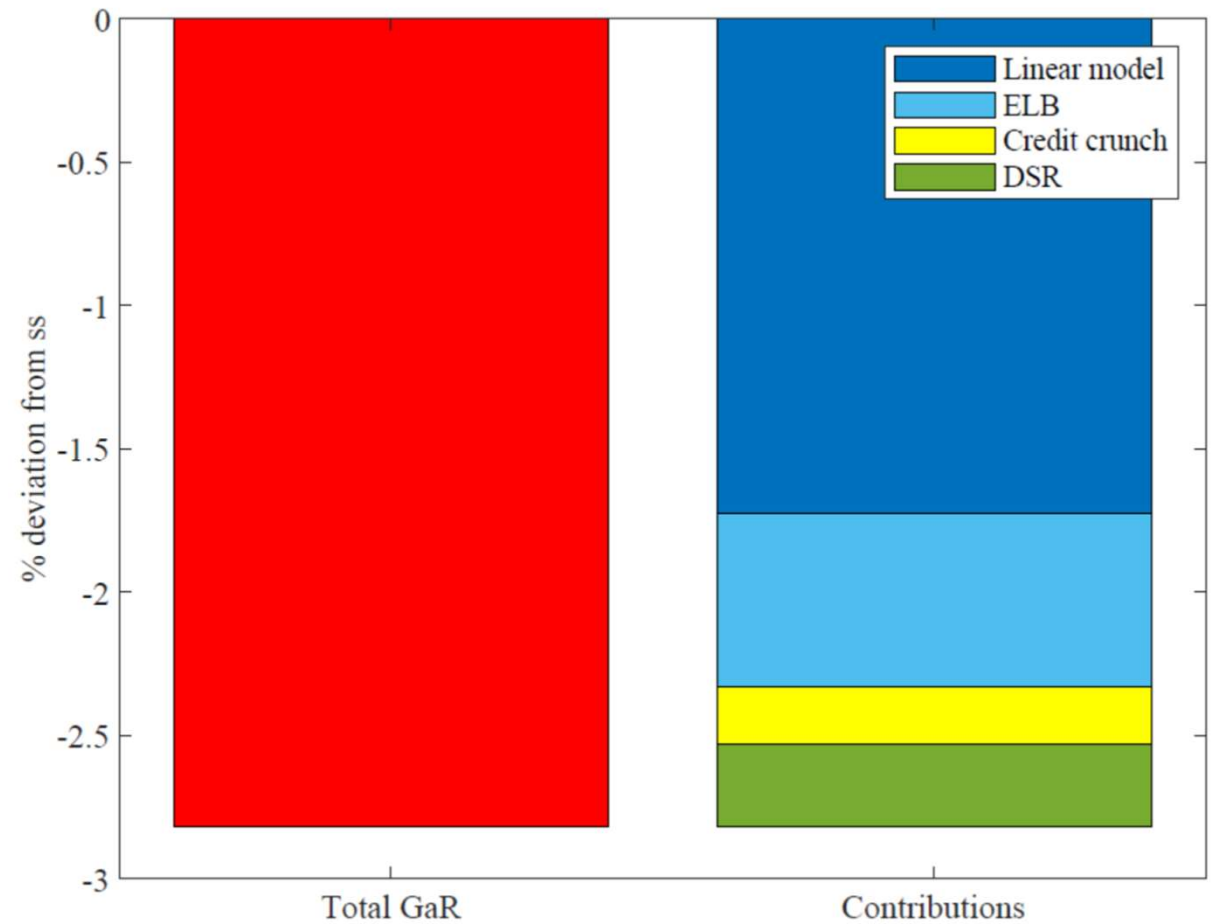
Unconditional distribution of (detrended) GDP

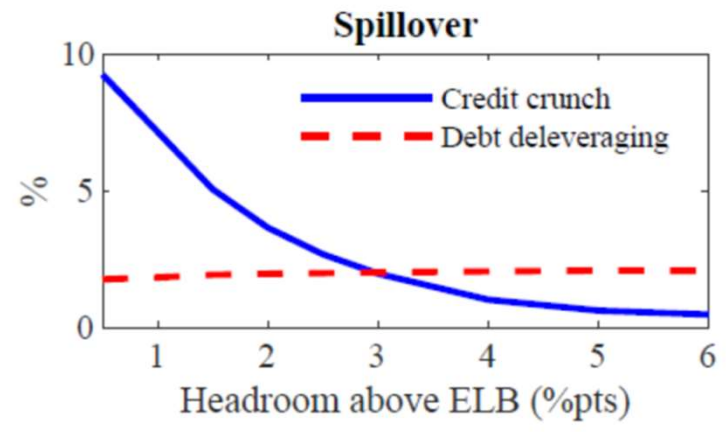
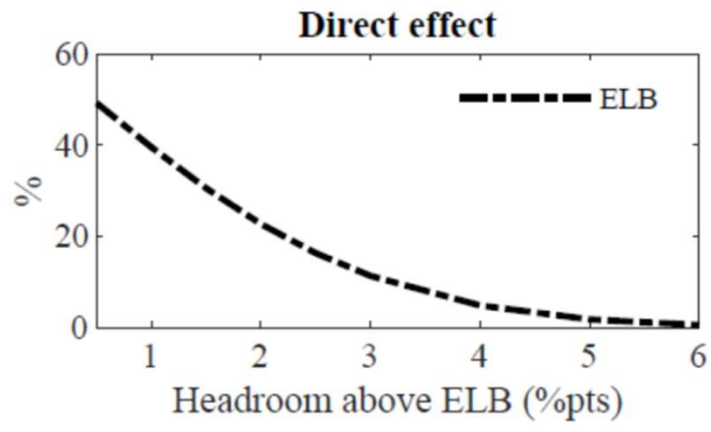
- 5000 simulations of 140 quarters (first 40 discarded)
- GDP-at-Risk: **-2.8%**
- NB: GDP-at-Risk is
 - **-1.7%** in the model without constraints
 - **-2.9%** in the data

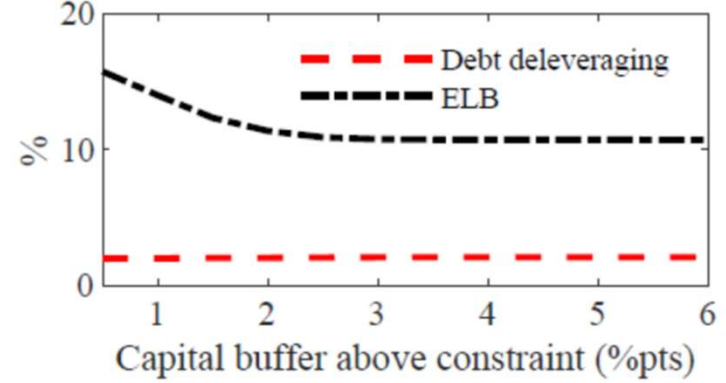
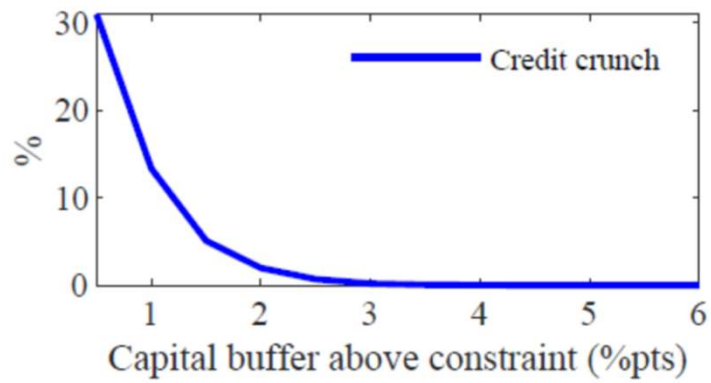
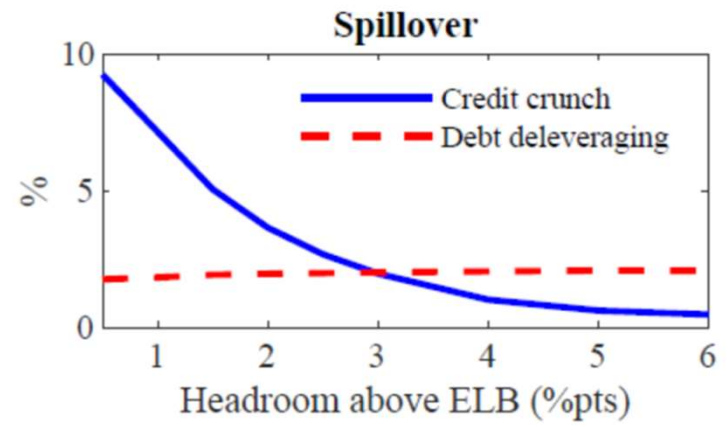
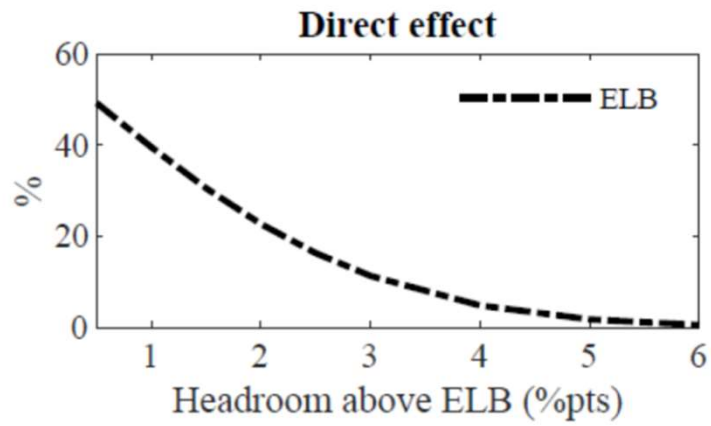


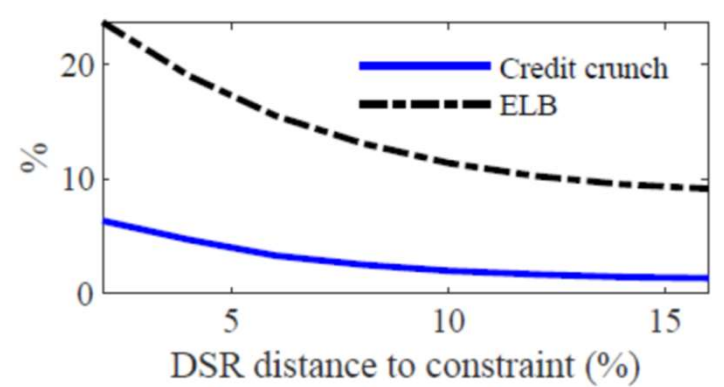
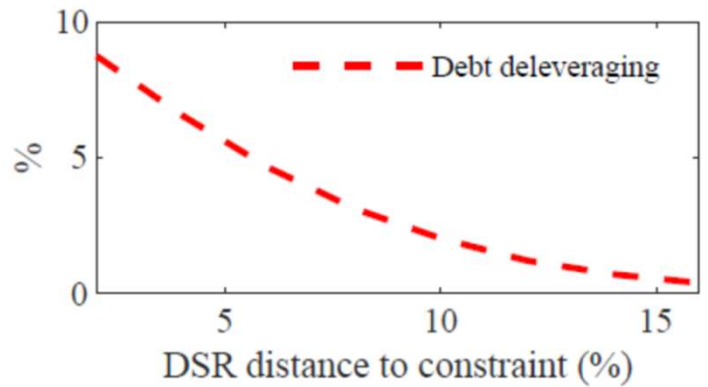
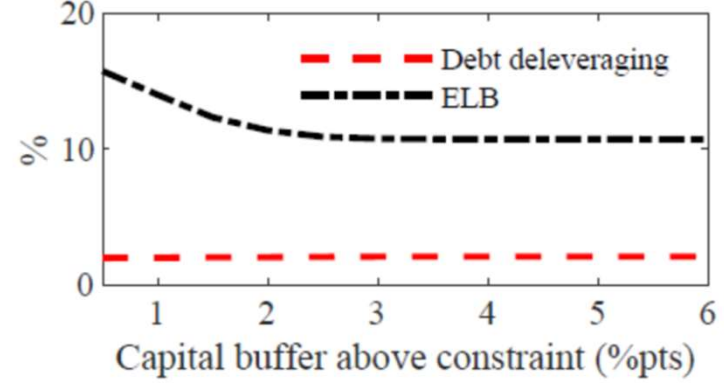
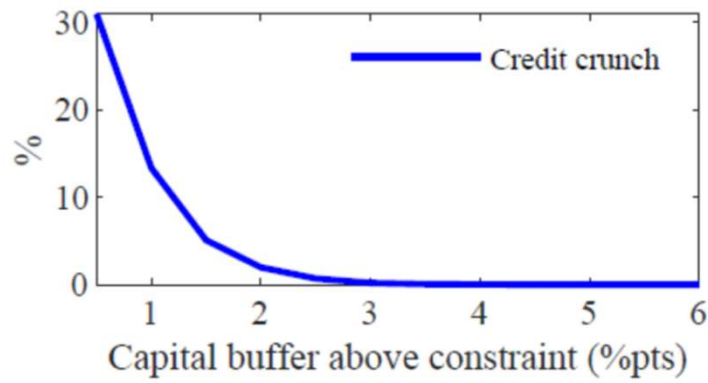
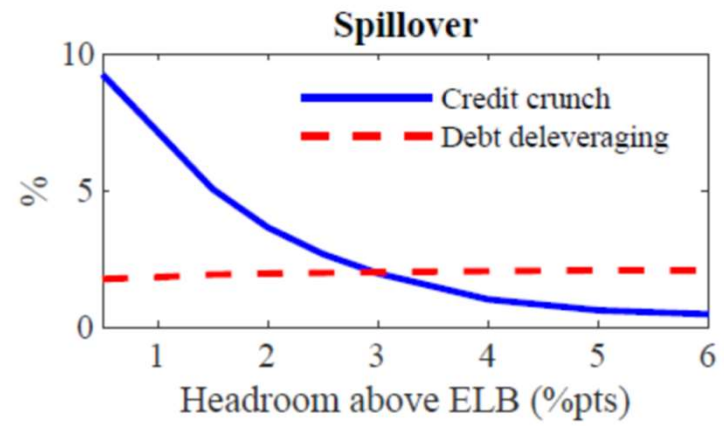
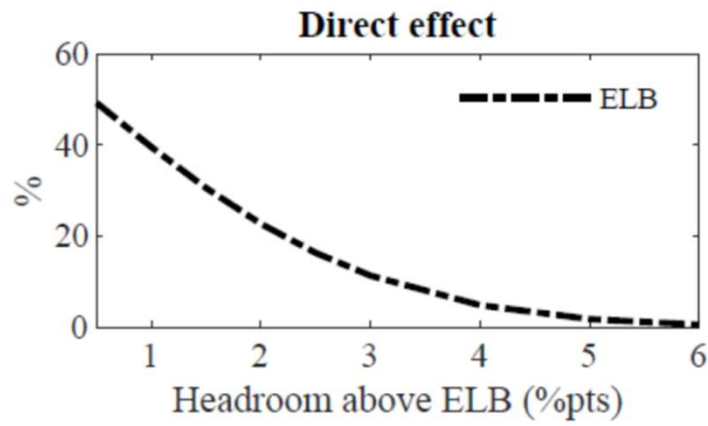
Shapley value contributions to GDP-at-Risk

- Frequency with which constraints bind:
 - ELB: 11.1%
 - Credit crunch: 1.8%
 - DSR: 1.9%

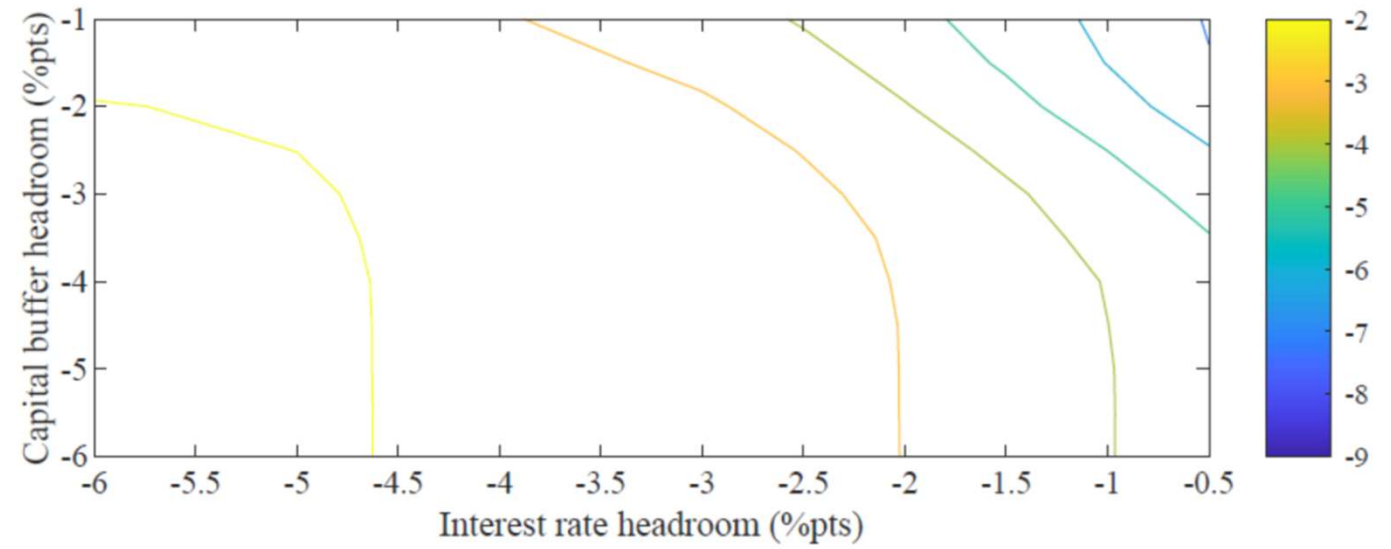




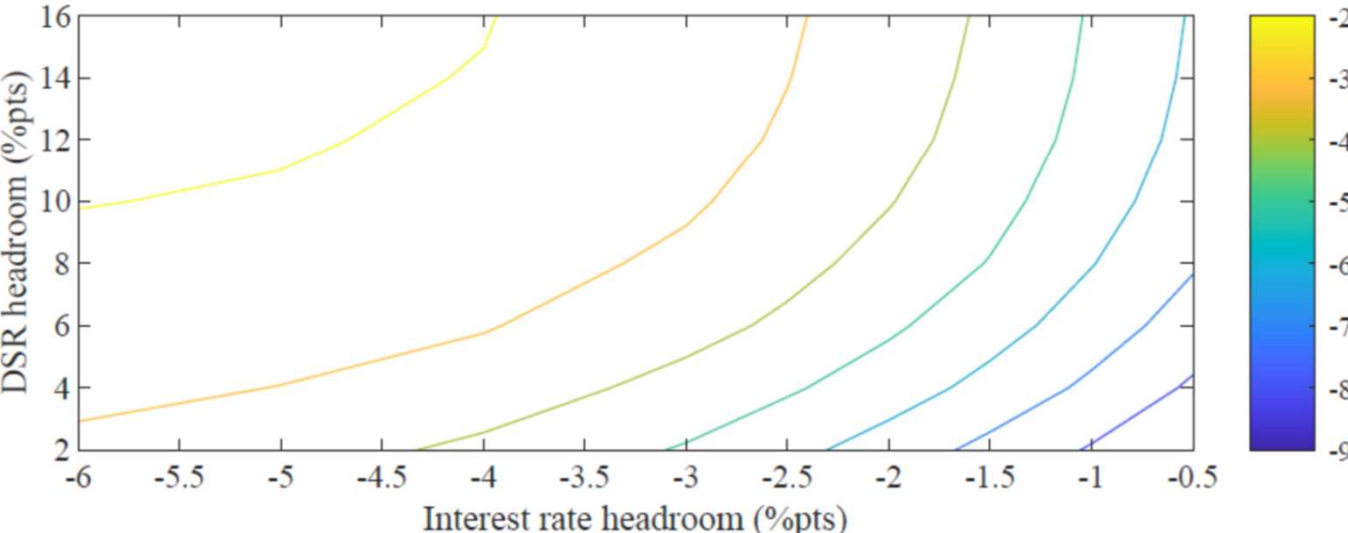
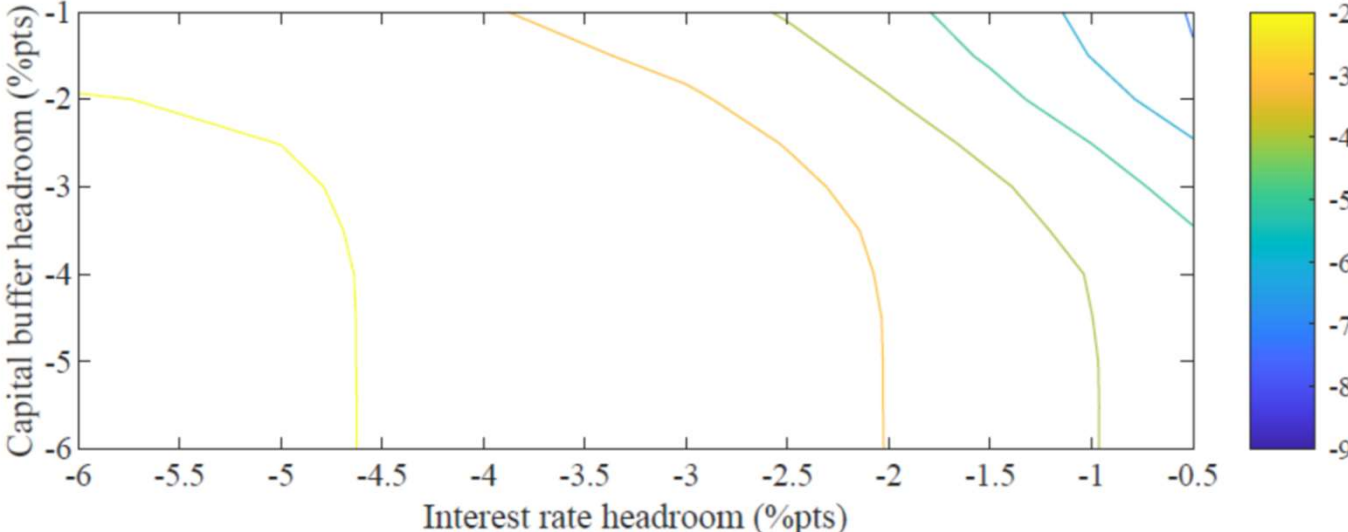




Iso-risk curves



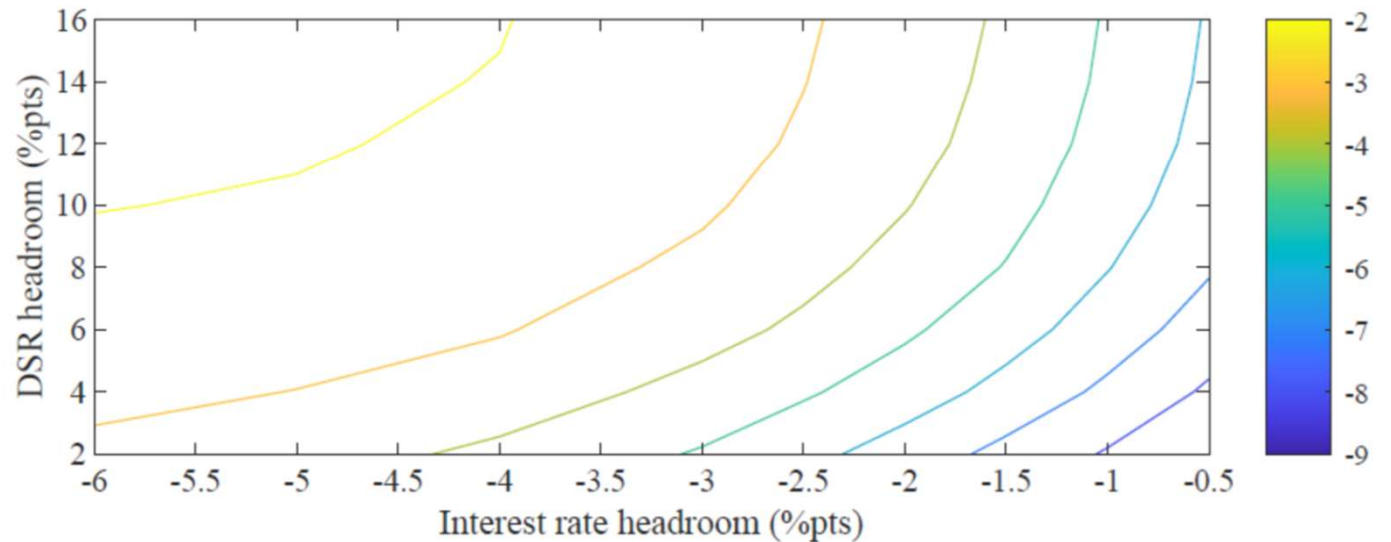
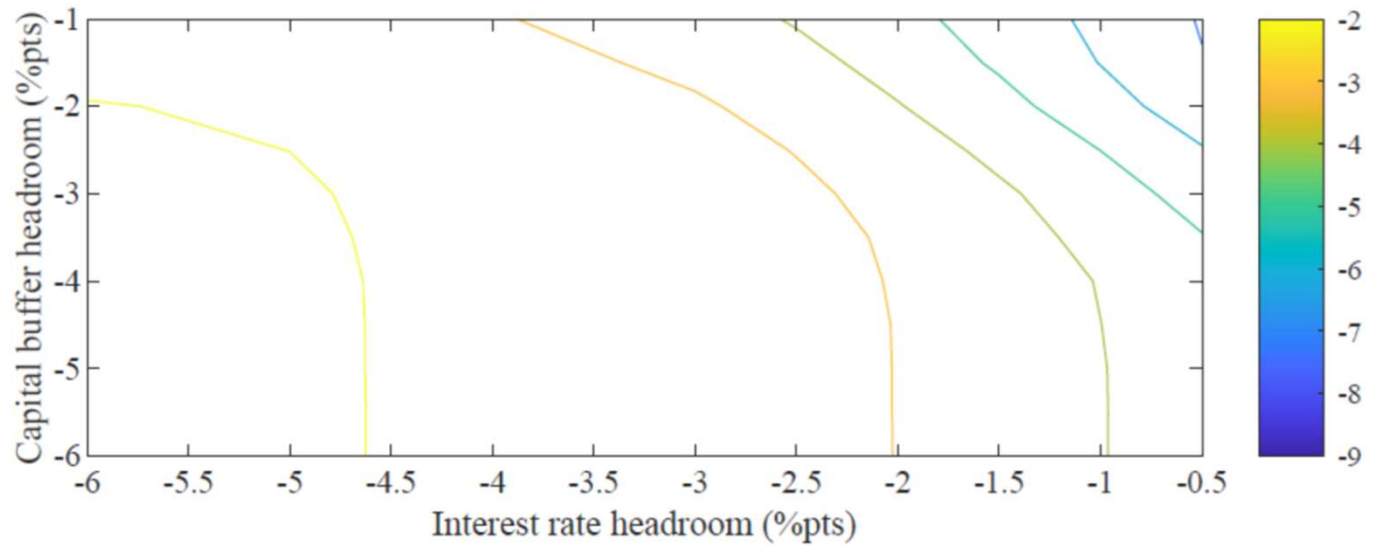
Iso-risk curves



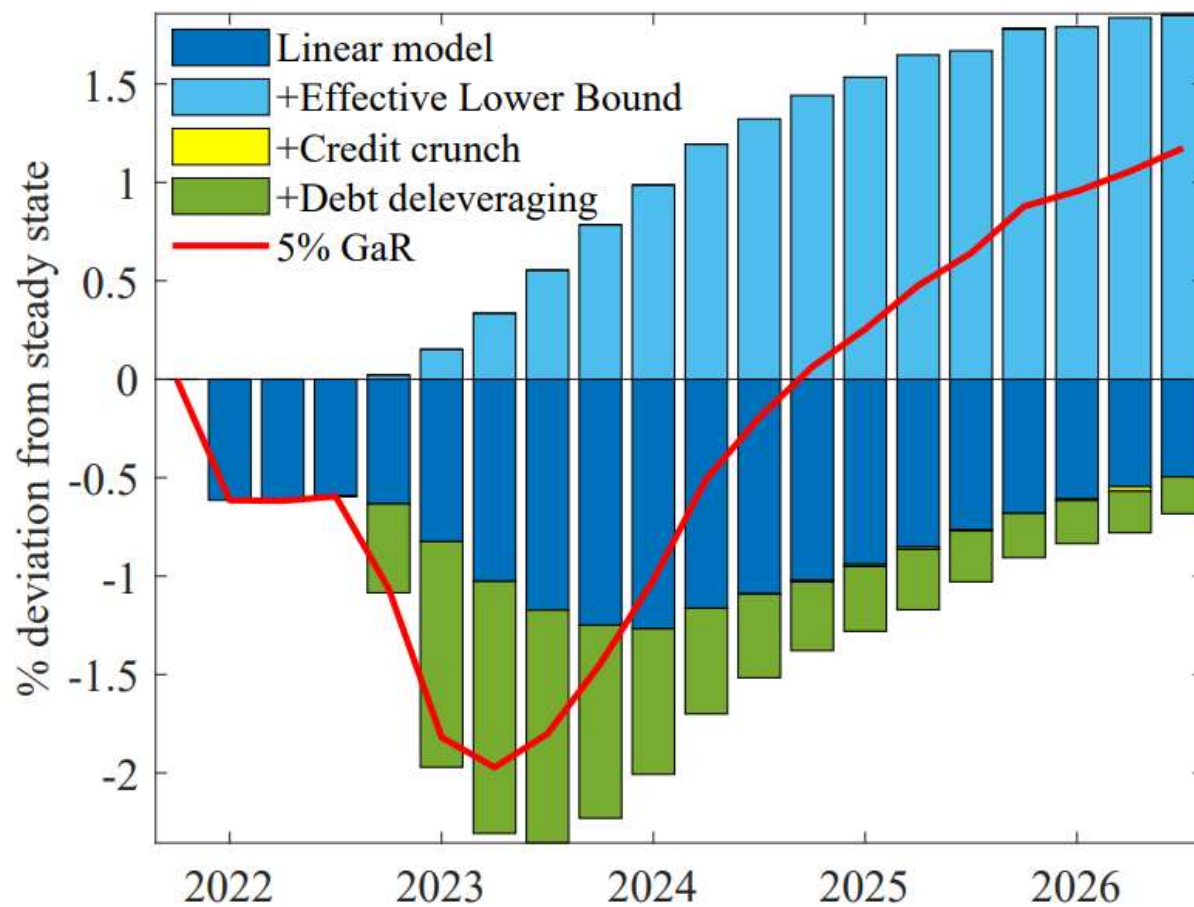
Iso-risk curves

50 basis point decline in monetary policy headroom requires:

- Additional capital buffers of 1% of assets
- Lower debt service burdens of 2-2.5% pts



Impact of a persistent inflation shock on GDP-at-Risk



Thoughts on what next for this literature

1. We need more empirical work on the reliability of GDP-at-Risk estimates at the policy horizon relevant for macropru: 2-3 years+
2. We need richer structural models of GDP-at-Risk, which incorporate a role for macropru policy tools such as the CCyB
3. Is this the right concept or should we focus on a metric of credit supply at risk?
4. Whether / how to integrate GDP-at-Risk with macro stress testing and in an integrated policy framework for macropru, ie with a loss function

Additional slides

Model: the macro block

- **IS equation** extended with spreads [Curdia & Woodford, 2010]:

$$\underbrace{y_t}_{\text{Output}} = \theta^y y_{t-1} - \theta^r \left(\underbrace{r_t}_{\text{interest}} - \underbrace{\pi_{t-1}}_{\text{inflation}} + \underbrace{s_t}_{\text{spread}} \right) + \varepsilon_t^y + \varepsilon_t^d$$

- **Phillips curve** extended with spreads:

$$\pi_t = \beta^\pi \pi_{t-1} + \beta^y y_{t-1} + \beta^s s_{t-1} + \varepsilon_t^\pi$$

- **Taylor rule** with effective lower bound:

$$r_t = \max[\bar{r}, (1 - \phi^r)(\phi^\pi \pi_t + \phi^y y_t) + \phi^r r_{t-1} + \varepsilon_t^r]$$

- Shocks $\varepsilon_t^y, \varepsilon_t^\pi, \varepsilon_t^r$ are AR(1) processes with iid shocks.
- ε_t^d is a deleveraging shock

Model: the financial block

- **Credit demand:**

$$\underbrace{b_t}_{\text{credit}} = \gamma^y y_t + \gamma^b b_{t-1} - \gamma^r (r_t - \pi_{t-1} + s_t) + \varepsilon_t^b + \varepsilon_t^d$$

- **Borrowing spreads/Credit supply:**

$$s_t = f^s s_{t-1} + f^b b_t - \tilde{f}^{kb} \underbrace{k_{t-1}}_{\text{capital ratio}} + \varepsilon_t^s$$

- **Bank capital ratio evolution:**

$$k_t = \delta^k k_{t-1} - \delta^r \Delta r_t + \delta^s (r_{t-1} + s_{t-1}) + \varepsilon_t^k$$

- Shocks $\varepsilon_t^b, \varepsilon_t^s$ are AR(1) processes with iid shocks.
- The variable ε_t^k captures banks' (pro-cyclical) credit losses
 $\varepsilon_t^k = v^y y_{t-1} + \rho^k \varepsilon_{t-1}^k + u_t^k$

Nonlinear constraints

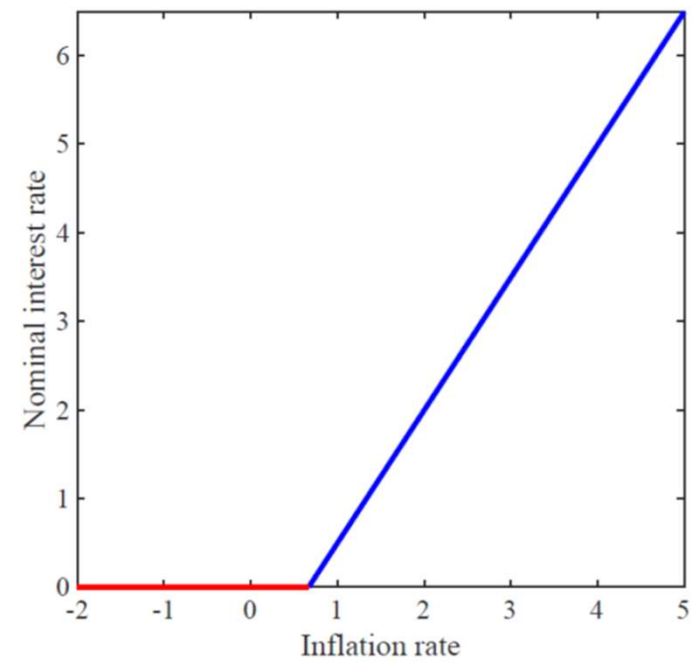
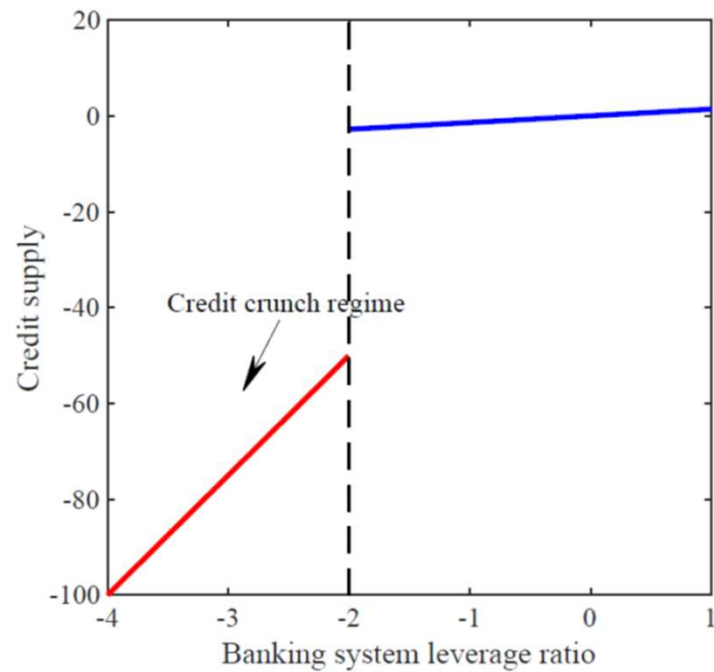
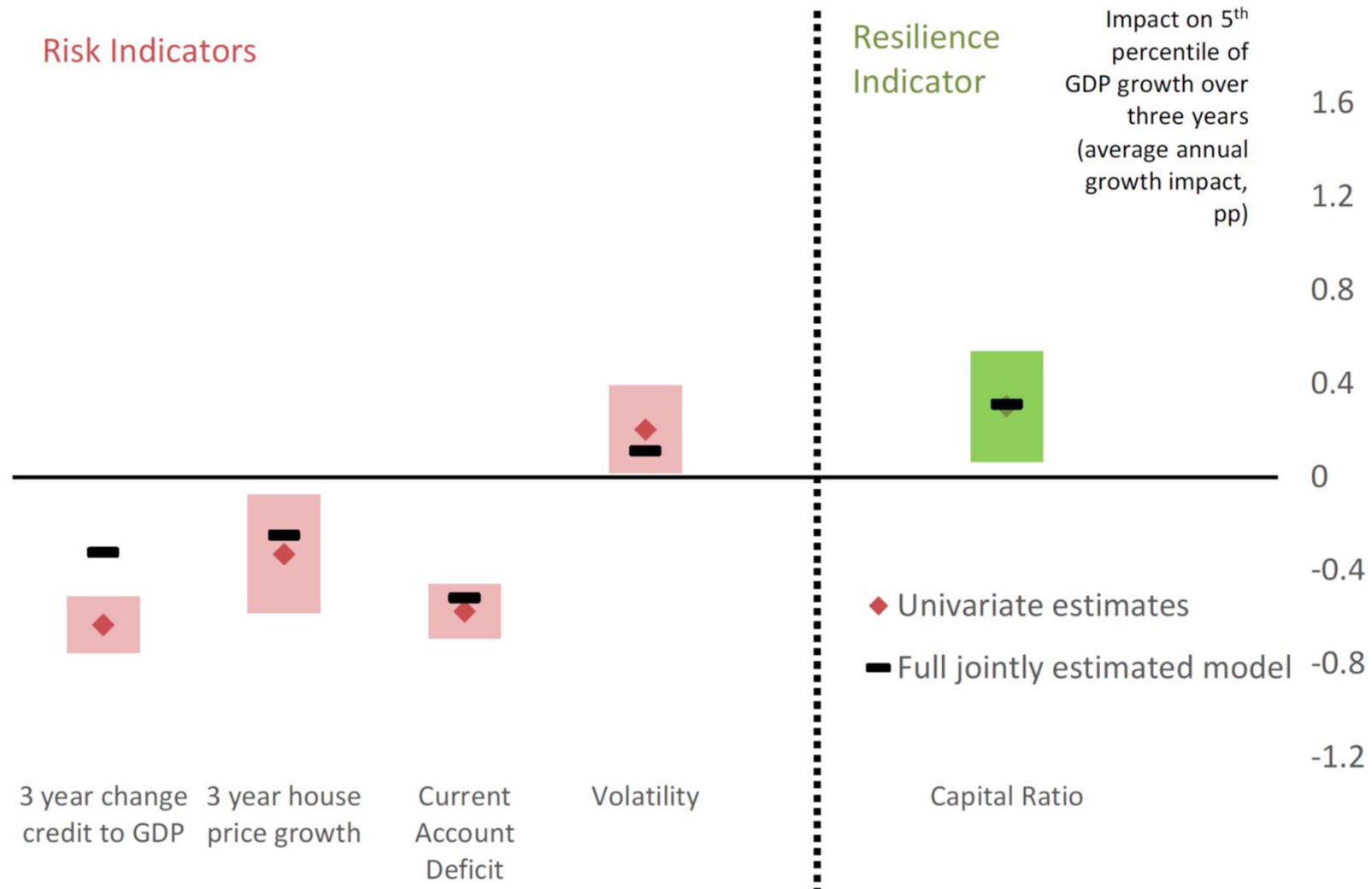


FIGURE A.III: Baseline results and single-indicator model



Decomposition of GDP-at-Risk 3-years ahead: US

