

**Discussant comments on  
Macro stress testing of credit risk focused on the tails**

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Prepared for the BIS CCA Conference on  
“Systemic risk, bank behaviour and regulation over the business cycle”

Buenos Aires, 18–19 March 2010

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\* These comments reflect the views of the author and not necessarily those of the BIS or of central banks participating in the meeting.

Discussion of “*Macro stress testing  
of credit risk focused on the tails*”  
by Wagner Piazza Gaglianone and  
Ricardo Schechtman

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# The paper

- Reduced form macro model (VAR):
  - GDP
  - Unemployment
  - Inflation
  - Interest rate
  - Credit volume
- Credit risk equation dependent on contemporaneous macro variables.
  - Credit risk proxied by non-performing loans (NPL).
- Stress testing based on bad macro scenario:
  - 1, 2, or 3 s.d. shocks to the macro forecasts of the VAR.
- Focus on quantiles of NPL:
  - Indirect: NPL is an additional equation of the VAR.
  - Direct: NPL is modelled via regression quantiles

# The Model

$$\mathbf{y}_t = \boldsymbol{\mu} + \mathbf{A}_0 \mathbf{y}_t + \sum_{i=1}^m \mathbf{A}_i \mathbf{y}_{t-i} + \boldsymbol{\varepsilon}_t$$

$$\mathbf{y}_t \equiv \begin{bmatrix} \text{NPL}_t \\ \text{GDP}_t \end{bmatrix} \quad \mathbf{A}_0 \equiv \begin{bmatrix} 0 & a \\ 0 & 0 \end{bmatrix}$$

$$\boldsymbol{\varepsilon}_t \sim \text{N}(\mathbf{0}, \boldsymbol{\Sigma})$$

plus additional restrictions on  $\mathbf{A}_i$ ,  $i > 0$

# Quantile estimation

Two strategies:

**Indirect**: Estimate previous model and obtain the quantiles from the parametric distribution of  $\varepsilon_{1,t}$

**Direct**: Model first equation of previous model via regression quantile:

$$Q(NPL_t, \tau | \Omega_t, GDP_t) = \mu_1(\tau) + a(\tau)GDP_t + \sum_{i=1}^m [b_i(\tau)NPL_{t-i} + c_i(\tau)GDP_{t-i}]$$

# Stress testing

- Assume bad realization for GDP at time  $T$  (1, 2, or 3 standard deviation shock).
- Look at the effect of this realization on the mean and quantile of NPL.
- Compare conditional (on bad realization of GDP at time  $T$ ) and unconditional means and quantiles.

# Comment 1: Structural VAR

$$\mathbf{y}_t = \boldsymbol{\mu} + \mathbf{A}_0 \mathbf{y}_t + \sum_{i=1}^m \mathbf{A}_i \mathbf{y}_{t-i} + \boldsymbol{\varepsilon}_t$$

Assume a diagonal variance-covariance matrix for  $\boldsymbol{\varepsilon}_t$  and give a structural interpretation to the VAR:

Macro shocks contemporaneously affect the NPL but not vice versa.

$$\mathbf{y}_t = \Lambda \boldsymbol{\mu} + \Lambda \sum_{i=1}^m \mathbf{A}_i \mathbf{y}_{t-i} + \Lambda \boldsymbol{\varepsilon}_t$$

$$\Lambda \equiv (\mathbf{I} - \mathbf{A}_0)^{-1} \quad \text{is upper triangular}$$

# Stress testing

$$1) E(y_{1,t} | \Omega_t, \varepsilon_{2,t}) - E(y_{1,t} | \Omega_t)$$

$$2) Q(y_{1,t}, \tau | \Omega_t, \varepsilon_{2,t}) - Q(y_{1,t}, \tau | \Omega_t)$$

$$3) \Pr[y_{1,t} < Q(y_{1,t}, \tau | \Omega_t) | \Omega_t, \varepsilon_{2,t}]$$

$$\hat{\tau} \quad \text{s.t.} \quad Q(y_{1,t}, \hat{\tau} | \Omega_t, \varepsilon_{2,t}) = Q(y_{1,t}, \tau | \Omega_t)$$



# Comment 2: Quantile Simulation

$$y_{1,t} = Q(y_{1,t}, \tau | \Omega_t, \varepsilon_{2,t}) + \tilde{\varepsilon}_{1,t} \quad \text{where} \quad Q(\tilde{\varepsilon}_{1,t}, \tau | \Omega_t, \varepsilon_{2,t}) = 0$$

$$Q(y_{1,t}, \tau | \Omega_t, \varepsilon_{2,t}) = \alpha_0(\tau) + \sum_{i=0}^p \alpha_i(\tau) y_{t-i} + \sum_{j=1}^m \gamma_j(\tau) z_{t-j}$$

For instance you could assume:

$\tilde{\varepsilon}_{1,t} \sim N(-k_\tau \sigma, \sigma)$  where  $k_\tau$  is the  $\tau$ -quantile of the normal distribution

$$\begin{aligned} \Pr(\tilde{\varepsilon}_{1,t} < 0) &= \Pr(\tilde{\varepsilon}_{1,t} + k_\tau \sigma < k_\tau \sigma) \\ &= \Pr[(\tilde{\varepsilon}_{1,t} + k_\tau \sigma) / \sigma < k_\tau] = \tau \end{aligned}$$

If you don't like the normality assumption, you could use the skewed Laplace distribution.

# Comment 3: Uncertainty

- Careful about the impact on risk measurement of:
  - Model misspecification
    - After the summer 2007 turmoil Goldman Sachs admitted that its models suggested their portfolios were hit by a 25 standard deviation shock.
    - This is an event that occurs once every  $10^{138}$  times...
    - What was the shock implied by GS models after September 2008?
  - Estimation error (DeMiguel et al., RFS 2009)
    - Show that no estimated mean-variance model can consistently outperform an equally weighted portfolio.
    - Exercise limited to 20 assets.
    - Typical portfolio of a bank includes many more assets.
    - Attempt to model joint macro and credit risks may suffer of similar problems.
- Rules of thumb may be not too bad after all.

# Comment 4: The Decision Problem

- What is the assessment? Did banks have enough capital to face the worst case scenario?
- What is the decision variable? Given your macro stress test exercise, how much capital buffer would you recommend?
- To answer this question you need first to introduce into the model:
  - Decision variable
  - Objective function
- Impulse-responses with two instruments:
  - Interest rate
  - Macro-prudential tool
- Tightening the macro-prudential tool would reduce credit risk, but what about its impact on GDP? Need to define the optimal trade-off.
- The endogeneity of the decision variables adds complexity.