FX Interventions Rules for Central Banks International Experiences and a Novel Risk-Based Framework

Romain Lafarguette Romain Veyrune

IMF Monetary and Capital Markets Department Central Bank Operations Division

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Goals and Intermediate Objectives of FX Interventions

Goals

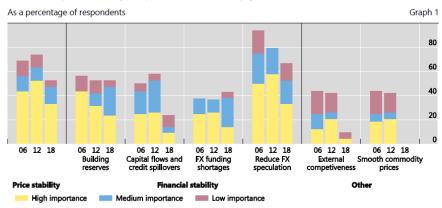
- Price stability: disanchored inflation expectations due to exchange-rate pass-through
- Financial stability: preserving market functionning, curbing excessive volatility

Intermediate Objectives

- Mitigating excessive volatility and limit pressure due to volatile capital flows
- Liquidity provision to thin markets

BIS Survey: Goals

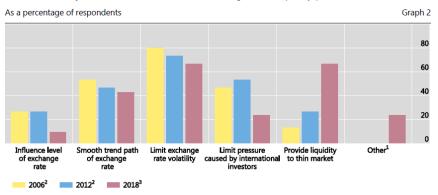
Price stability and curbing FX speculation remain key goals for FX intervention



Source: BIS surveys from 2012, 2018 and 2019.

BIS Survey: Intermediate Objectives

Intermediate objectives of FX intervention: Increasing role of liquidity provision



Mostly non-floating exchange rate arrangements. The "Other" option was not provided in 2006 and 2012. ² 15 central banks.
³ 19 central banks.

Source: BIS surveys in 2012 and 2018.

FX Interventions Rules

 An intervention is deemed rule-based when it reacts to predetermined parameters to deliver predictable responses

Intervention rules have many advantages:

- Consistent approach with the policy objectives
- Robust operational framework
- If public: anchors agents' expectations and influence their behavior to reach the policy objectives
- Supports market functionning and market development
- Strengthens the central bank credibility and independence

The large majority of FX interventions rules implemented by central banks are based on **fixed-volatility thresholds**: cf. Mexico, Columbia and Guatemala

Bank of Mexico FX Operations

The BM put in place two types of FX operations:

- 1. To accumulate or sell foreign reserves
 - Via pre-determined put options for purchases
 - And auctions for selling foreign reserves

- 2. To intervene on the market to limit extreme exchange rate volatility and provide liquidity to ensure proper market operating conditions
 - Using minimum-prices auctions (rule-based)
 - No-minimum prices auctions (discretionary prices)

We focus on the second type of operations, the FX interventions

Minimum Bid Price Mechanism

- The central bank auctions a prespecified amount of US dollars every day and sets a minimum accepted price
- The trigger to sell US dollars to the market is predetermined by market conditions without meddling with the free-floating exchange rate regime
 - The minimum price accepted was set equal to the **previous** day closed value plus a 2 percent depreciation factor (was changed to 1.5 and then 1 after 2014 and 2015)

No Minimum Bid Price Mechanism

- Sometimes, a constant demand for US dollars that could distort operating conditions and generate volatility is expected to persist
- The total amount offered corresponded to the **expected** accumulation of international reserves during the following quarter (most of the accumulation came from the national oil company sales to the CB)
 - Sometimes, the amount was directly decided by the central bank, on a discretionary basis

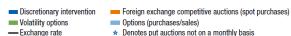
Columbia: Volatility Options

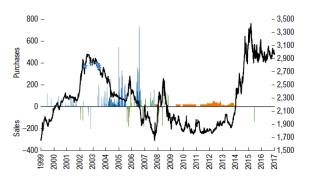
- The central bank publicly committed to offer, through auctions, call (put) options with one-month maturity
- Whenever the representative bilateral rate (TRM) was higher/lower than its 20-day moving average +/-5 percent (then changed to 2 percent in 2011 then 7 percent later)
- The strike price was the TRM, and the condition for being able to exercise the option was the same as for the activation of the auction.
- Objective to mitigate extreme fluctuations: amount auctionned large compared to market turnover
- However, the central bank struggled to find the right threshold (5-7-2 percent), and had a lot of try and miss

Columbia FX Operations

Figure 9.1. Colombian Central Bank Foreign Exchange Intervention, 1999–2017

(Millions of US dollars, left scale; US dollar/Colombian peso official exchange rate, right scale)





Source: Central Bank of Colombia. Note: Data are from September of each year.

Guatemala: Auctions With Maximum Allotment

- Intervention if the reference exchange rate reaches or exceeds the fluctuation band of +/- 0.80% around the moving average of the reference exchange rates from the past five business days
- Auction to buy or sell up to US\$10 million in foreign currency
- ▶ No more than five auctions a day are allowed
- The results of public auctions are disclosed to the public by publication on the website of the BOG and through print mass media.

Fixed-Volatility Rules in Perspective

Fixed-volatility rules are simple, transparent and consistent with inflation targeting. However:

- ▶ The fixed thresholds are difficult to decide and have to be revised often (cf Columbia and Mexico)
- Rigid to market conditions: might prove inadequate in extreme stress periods
- Prone to market manipulation, and don't properly align the incentives of market participants with CB objectives
- Fixed triggers lead to asymmetric interventions, and can put the FX reserves at risk
- Often, tend to over-hedge FX risk for market participants, leading to moral hazard and constraining FX market development
- ▶ Interventions are not financially optimal from a cost perspective

We have developed a new framework to address these shortcomings

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Contributions

- Design a rule to address tail-risks related to direct and indirect FX exposures in the economy
- ▶ Provides guidance on **when** to intervene ("triggers")
- Appropriate for **floating exchange rate regimes** with FX macrofinancial risks (e.g. dollarization)
- Consistently target FX risk in the economy
- A risk management framework for central banks' financial stability mandate: aligned with industry's best practices in risk management

Desirable Properties of FXI Rules

Foreign Exchange intervention rules should be:

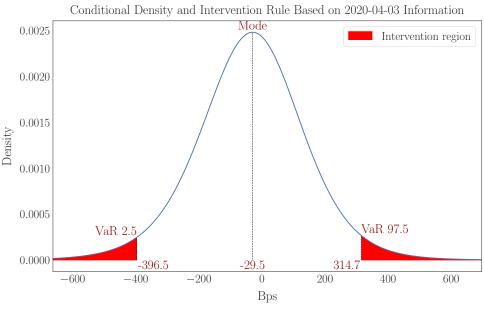
- Adaptative, depend on market conditions
- Objective, anchored to a risk tolerance level rather than an aribtrary FX level threshold
- Capture FX non-linearities and asymmetries between appreciation and depreciation
- Be easily operationalizable, and financially viable

We propose an FX intervention rule based on Conditional Value-at-Risk

Concept: Value-at-Risk FXI Rule

- Rather than using a fixed volatility rule (e.g. intervene if daily exchange rate varies by more than 2%)
- Use a **risk-based rule**: intervene when the daily exchange rate log-returns fall within the tails of the conditional distribution
- Measure the tail-risk via the concept of Value-at-Risk (the conditional quantile of the log returns distribution)
- The conditional distribution is estimated daily with a standard financial GARCH model and varies with market conditions
- ▶ The central bank decides on the **risk tolerance**: e.g. intervene in the tail at 1%, 5%, 10%, etc.

VaR FXI Rule



A Risk-Management Approach to FX Interventions

- Tail-risks hedge not always available: incomplete markets
- The central bank is transferring FX risk from the market to its balance sheet. It buys a risky asset (FX) and issues a risk-free asset (local currency)
- Provide a public good to address market failure. Leave a fix share of risk for the market to hedge
- ullet Risk tolerance should depend on the macrofinancial risk
- The financial stability mandate of the central bank is properly formalized and quantified via VaR metric

Main Features

- 1. Allows flexible exchange rate to act as a **shock absorber**: more flexibility in crisis time => **avoid overshooting**
- 2. **No excessive interventions** in crisis time, often ineffective and costly (exhaust FX reserves)
- 3. No free insurance to the market: avoid **moral hazard**, foster the **development of hedging market**
- 4. Prevent market manipulation and windfall effects
- 5. Guarantees **fixed-frequency** interventions:
 - Certainty about interventions: the central bank can intervene with larger FX amounts, making the interventions more efficient
 - Budget neutrality with symmetric risk preference
- 6. Financially optimized: always buy/sell at the best price

Operational Implementation

- Standard data requirements, easily accessible for a central bank, can be customized
- Parsimonious GARCH model featuring embedded heteroskedasticity, asymmetries (appreciation/depreciation), non-linearities (exponential volatility) and parametric density forecasting
- We created a Python wrapper, **free and open-source** (soon on Github): estimation, forecasting, out-of-sample evaluation, benchmarking, etc. Results are **fully replicable**
- Can be readily used by central banks and deployed during Technical Assistance (TA) missions

Challenges

- Some central banks might be reluctant to use a VaR-rule: more difficult to communicate to the public
 - However, FXI occur on the wholesale FX market, where market participants are fully aware of the VaR concept
- Some policymakers might prefer to keep discretion over FXI
 - Trade-off: a transparent rule anchors better market expectations, maximize efficiency and strengthen central bank's independence

The Framework Extends Beyond FXI triggers

- 1. Determine FX Intervention triggers
- 2. Conduct market monitoring and provide policy guidance
- 3. Benchmark FX interventions, including discretionary interventions

- We present below an application of the toolkit to the Mexican Peso, based on publicly available data
- More than 4500 daily observations, from 2009 to 2018, with Bank of Mexico (public) FX interventions, mostly concentrated in 2009 and 2016

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Specification

- Non-linear, Exponential GARCH (EGARCH) model
- The dependent variable is the FX log-returns, $r_t = \log(\frac{e_t}{e_{t-1}})$, where e_t is the bilateral market exchange rate against the major currency (e.g. USD)
- **Drift AR-X(1):** $r_{t+1} = \alpha_d + \rho r_t + \beta X_{t+1} + \epsilon_{t-1}$
- Exponential volatility: $\log \sigma_{t+1}^2 = \omega + \beta g(r_t)$ where $g(r_t) = \alpha_v r_t + \gamma(|r_t| \mathbb{E}|r_t|)$
- Error term distribution $\epsilon_t = \sigma_t \varepsilon_t$, $\varepsilon_t \sim \text{TSK}(0, 1, \nu)$
- The forecasted conditional probability distribution function is defined as:

$$\hat{f}(r_{t+1}|r_t, X_{t+1}) = \text{TSK}(\hat{r}_{t+1}, \hat{\sigma}_{t+1}^2, \hat{\nu})$$

Estimation

- The GARCH estimation is standard and done with maximimum likelihood
- Selection of parameters is done via AIC/BIC criteria.
- Our Python package allows to flexibly select:
 - The set of exogeneous regressors
 - ► The number of lags
 - The volatility specification (exponential, RiskMetric, standard GARCH, etc.)
 - The distribution family of the error-terms (Gaussian, Student, Tskew, Generalized Gaussian, etc.)
- More complex models (e.g. copulas, non-parametric kernels, etc.) can be used within the same VaR framework.
 However, more difficult to understand and to implement

Exogeneous Regressors

- 1. **FX microstructure**: FX bid-ask spread (averaged over the day)
- 2. CIP: daily interest rate differential with the US Libor
- 3. **Hedging costs**: one-month forward exchange rate
- 4. Past policy interventions: lagged amount of central bank FX intervention
- 5. Global risk sentiment: The VIX, implied volatility on the S&P 500
- 6. Global FX factor: The EURUSD exchange rate

Regression Table

	Microstructure	CIP	Dollar move	Risk Appetite	Baseline
Intercept	-2.33***	-2.23	-1.84	-2.55	-1.63
Lag FX log returns	-0.07***	-0.08***	-0.08***	-0.08***	-0.08***
Bid ask abs	5.73***	24.55	-35.84	-2.48	3.43
Min max abs	35.55***	34.27	34.36***	34.44*	26.16*
Forward points first difference	23.29***	17.85***	26.44***	19.82***	19.44***
Interbank rate vs Libor		33.7***	39.31***	34.76***	33.87***
EURUSD log returns			-0.14***	-0.17***	-0.16***
VIX first diff				15.66***	15.37***
FX intervention dummy lag					2.23
Oil prices log returns					-0.02***
Omega	0.13***	0.13***	0.12***	0.11***	0.12***
Alpha	0.17***	0.17***	0.16***	0.16***	0.15***
Gamma	0.07***	0.06***	0.06***	0.05***	0.05***
Beta	0.98***	0.99***	0.99***	0.99***	0.99***
Nu	8.33***	8.67***	8.92***	8.71***	8.54***
Lambda	0.08***	0.08	0.09***	0.07*	0.08***
R2	5.8 %	6.7 %	10.4 %	27.3 %	27.6 %
R2 adjusted	5.8 %	6.6 %	10.4 %	27.2 %	27.5 %
Number of observations Significance *10%, **5%, ***1%	5986	5986	5682	5682	5680

Formalization of the Intervention Rule

• Consider the estimated conditional distribution of the exchange rate log returns r_t defined as

$$\mathbb{P}[r_t \leqslant x] = \int_{-\infty}^x \hat{f}(r_t | r_{t-1}, X_t) dr_t$$

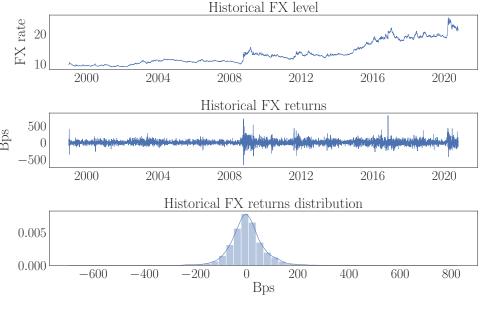
• The Conditional Value-at-Risk at threshold τ is simply defined as the conditional τ -quantile

$$Q(r_t, \tau) \equiv \mathbb{P}[r_t \leqslant Q(r_t, \tau)] = \tau, \text{ for } \tau \in (0, 1)$$

The FXI intervention rule is a simple boolean rule, based on two risk-thresholds $(\underline{\tau}, \overline{\tau})$, for depreciation and appreciation, potentially risk-symmetric $(\overline{\tau} = 1 - \underline{\tau})$

$$\mathbb{1}\left[\left\{r_t \leqslant Q(r_t, \underline{\tau})\right\} \cup \left\{r_t > Q(r_t, \overline{\tau})\right\}\right]$$

Dynamics of the Mexican Peso against USD



Conditional In-Sample Volatility of the Mexican Peso

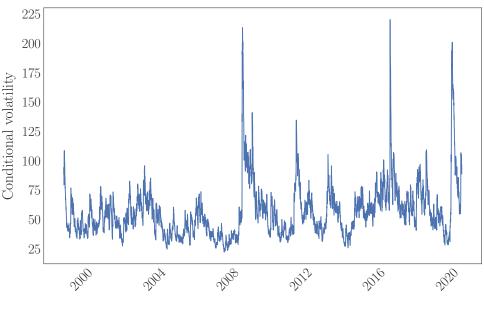


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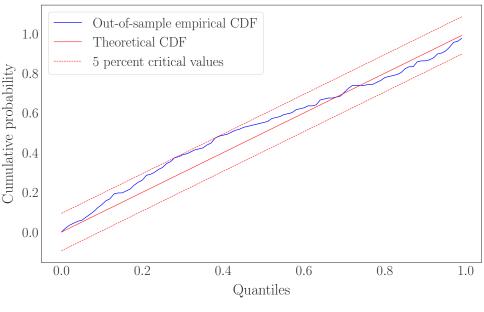
Benchmarking

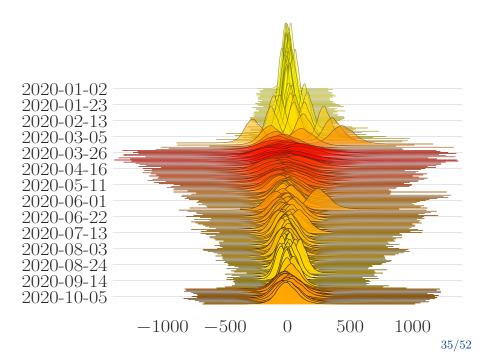
Policy Implications and Future Work

Forecasting

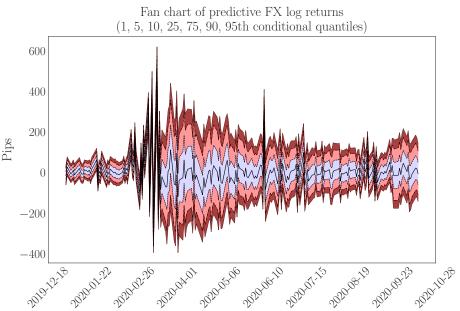
- Real-time forecasting based on market conditions
- Estimate the GARCH and derive the forecasted drift and volatility
- Infer the full-fledged conditional distribution of FX log returns for any point in time
- Assess model accuracy via (i) in-sample metrics and (ii) out of sample performance (probability integral transform test)
- The probability integral transform test assess on whether the random variable defined as $PIT(R) \equiv F_R R$ is uniformally distributed $F_R R \sim U(0,1)$, where R is the stochastic process of the FX log returns $r_t, \forall t \in [0,T]$

Density Evaluation

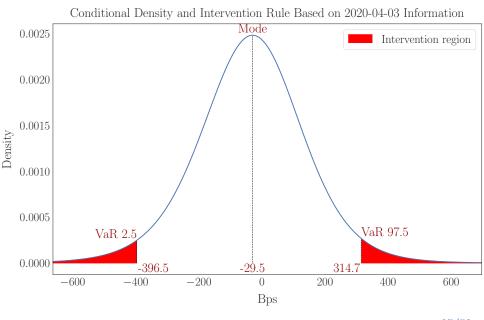


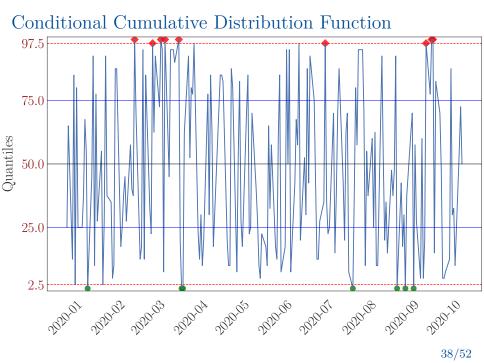


Fan Chart



VaR FXI Rule





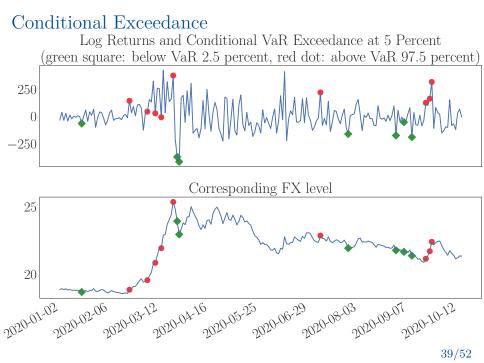


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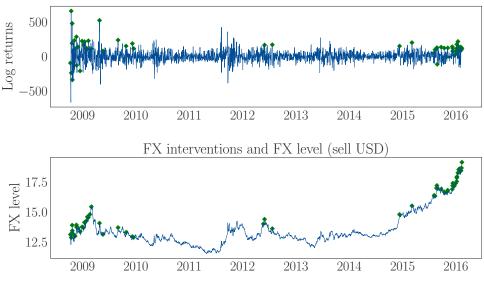
Benchmarking

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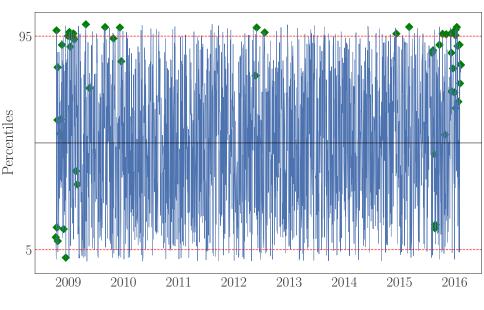
Bank of Mexico FX Interventions Setup

- ► The Banco Mexico (BM) implemented both ex-ante, transparent FX auctions and discretionary-rate auctions
- Different reservation rates:
 - Rule-based setting: BM operated an auction every day with a pre-announced a minimum rate for eligible bids
 - Discretionary setting: the auction was organized at the BM's discretion without reservation rate
- Often, no demand for the ruled-based auction as the market rate was below the reservation rate
- No-minimum price auctions could be motivated by other considerations than the exchange rate level
- What was the risk level when the FXI occurred?

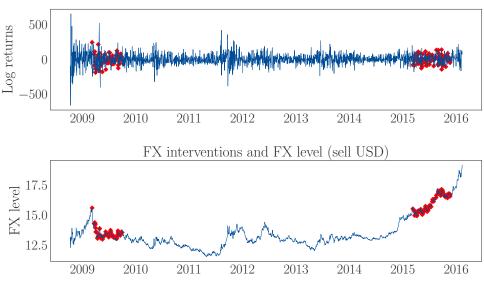
Rule-Based Benchmarking



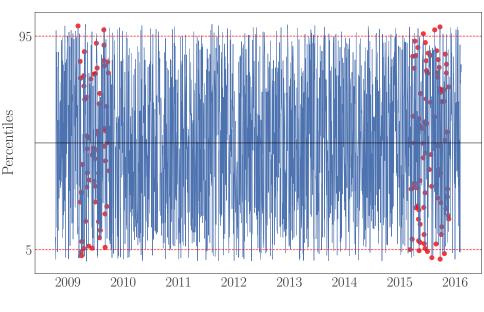
Rule-Based Benchmarking: Risk-Level



Discretion-Based Benchmarking



Discretion-Based Benchmarking: Risk-Level



Financial Performances

- We compared the financial performance of the three different interventions strategies: VaR, minimum price, without minimum price
- Using an entire year where the frequency of intervention was the same for auctions with and without minimum price
- We construct the counterfactual VaR interventions by calibrating the risk tolerance so that the frequency of intervention is the same as for auctions with and without minimum price
- Auctions with minimum price outperforms the no-minimum price auctions by 6.9% on average
- ► The VaR rule outperforms the no-minimum price auctions by 8.3% on average

Financial Performances: Results

	No minimum price	Minimum price	VaR rule
Daily variation bps	-6.1	108.8	164.8
Average exchange rate	16.6	17.8	18
FX Perf. vs. discretionary	0 %	6.9 %	8.3 %
Total volume bn USD	3.32	3.6	3.32
Number of interventions	18	18	18

Benchmarking Results

1. FX auctions with ex-ante minimum price ("rule-based")

- The minimum price auctions did not fully prevented BM to intervene outside of the tails of the conditional distribution
- In that respect, VaR-based intervention would have been better to mitigate tail-risks

2. FX auctions with no ex-ante minimum price ("discretion-based")

- No minimum prices interventions occurred at almost any risk level
- Discretion triggers are not identifiable based on risk

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Policy Implications

- Useful for floating rate regimes to operationalize their financial stability mandate with a risk-management framework
- The VaR-based rule could be considered as one option to improve the rules that central banks currently use
- ▶ Let the nominal exchange rate acts as a **shock absorber**
- Could be used to accompany the transition to exchange rate flexibility, with gradually less and less interventions
- More generally, could be used by central banks for market and risk monitoring

Future Work

- 1. Use expected shortfall (ES) instead of VaR, as ES has better risk properties
- 2. Look **beyond spot FX markets** and apply a similar and consistent approach to:
 - FX derivatives, e.g. forward spreads
 - Offshore/onshore interest rate markets
 - Fixed income market
- 3. Determine the risk tolerance by **identifying vulnerabilities** and their impact to the economy. Align with the "at-risk" work done in MCM

Alternative Models: Benchmarking

	PIT	Logscore diff against Baseline	Diff pvalue
Baseline	Pass		
Unconditional			
Quantile Reg			
Gaussian EGARCH	Fail	1.54	0.938
TSkew GARCH	Fail	1.768	0.961
Gaussian GARCH	Fail	1.755	0.96