

# Drivers of price volatility in Brazilian pork market, internal and external effects - DCC GARCH approach

Gabriel Rosero<sup>1</sup>, Tinoush Jamali Jaghdani<sup>2</sup>, Bernhard Brümmmer<sup>1</sup>

<sup>1</sup>Georg-August-Universität Göttingen • <sup>2</sup>Leibniz Institute of Agricultural Development in Transition Economies (IAMO)

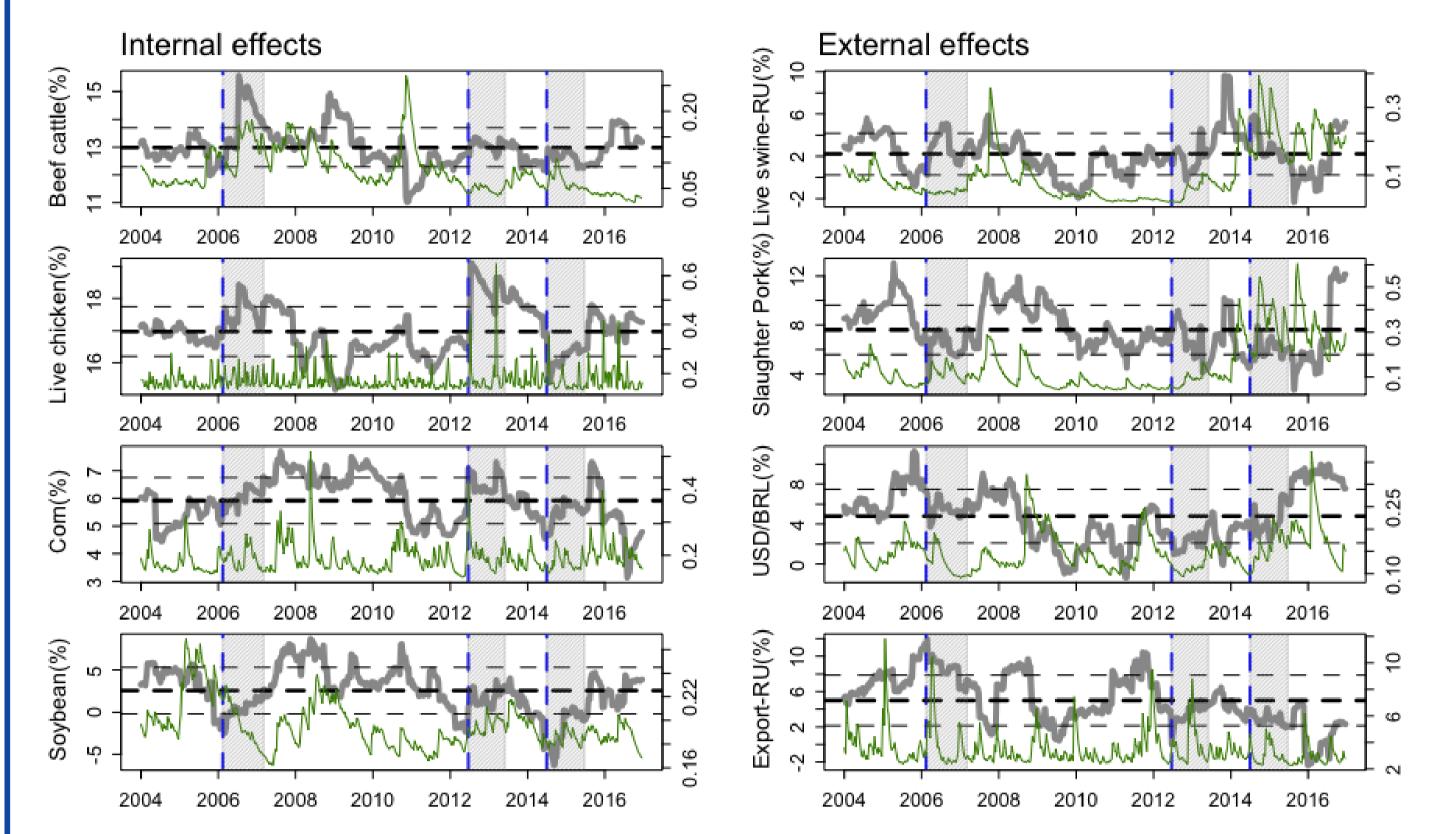
## **1. Background**

 Pork is the most consumed meat worldwide, and Brazil is one of the major players in the global meat industry.

- The pork market is likely affected by production issues, domestic and macroeconomic conditions and policy interventions (Brümmer et al., 2013; Prikhodko and Davleyev, 2014) that cause price volatility at national and international level.
- Food price volatility generates economic and political instability, affects food and nutrition security and reduces investment in agriculture production with a negative

## **5.** Results

#### 5.1 Dynamic Conditional Correlation (DCC-GARCH)



impact on food supply (Kalkuhl et al., 2016).

• The Russian Federation has been the primary pork market for Brazilian exports; however, the Russian government has imposed trade policy interventions that have affected Brazilian meat markets.

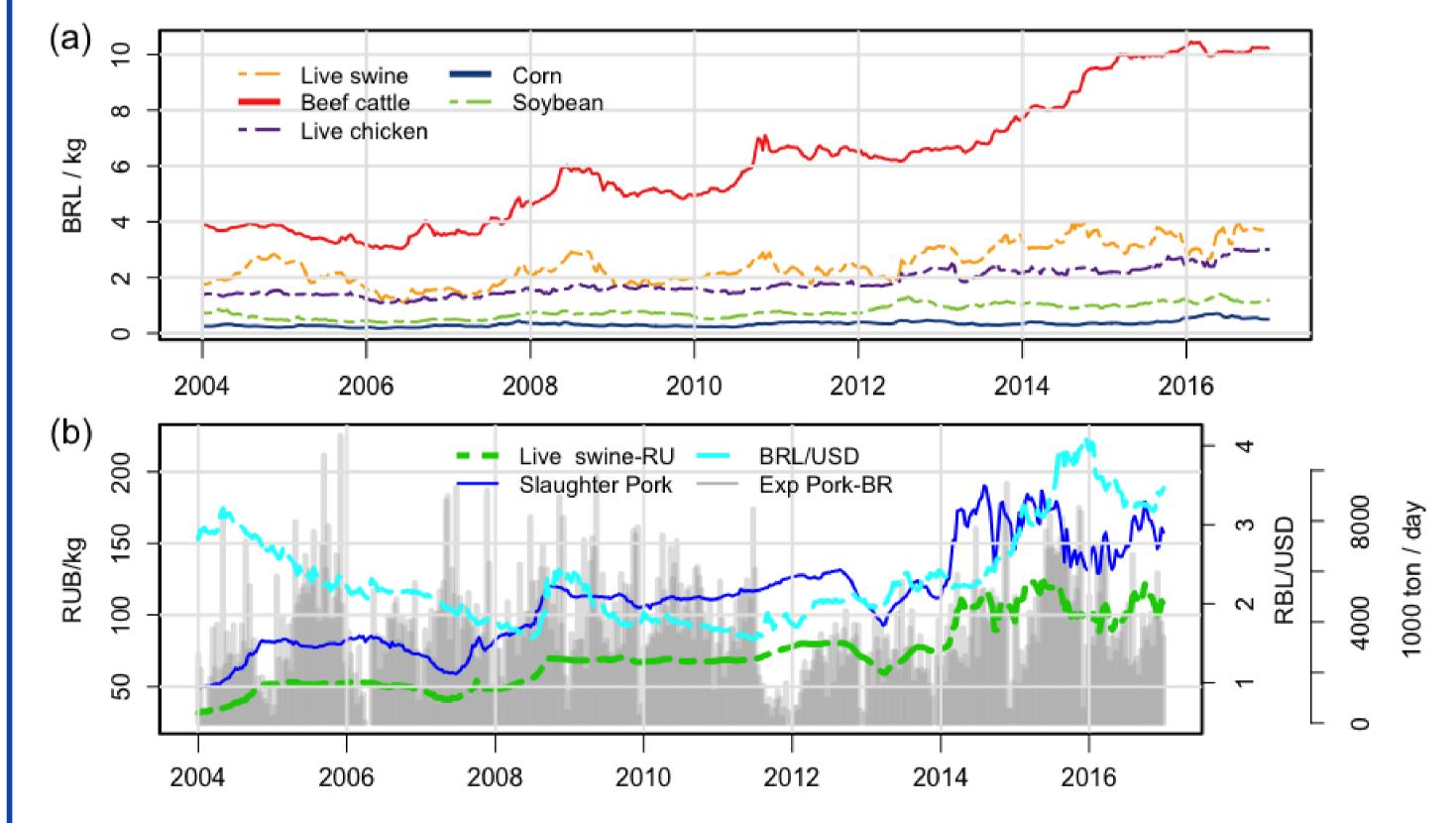
## **2. Research question**

What are dynamics and drivers of the price volatility in the Brazilian pork market?
How do trade policies affect the price volatility and volatility transmission in the short-run and long-run?

### **3.** Data

• Time series from January 2004 to December 2016 from SEAB (2017), Rosstat (2017) and CEPEA (2017) with a 10 days' frequency

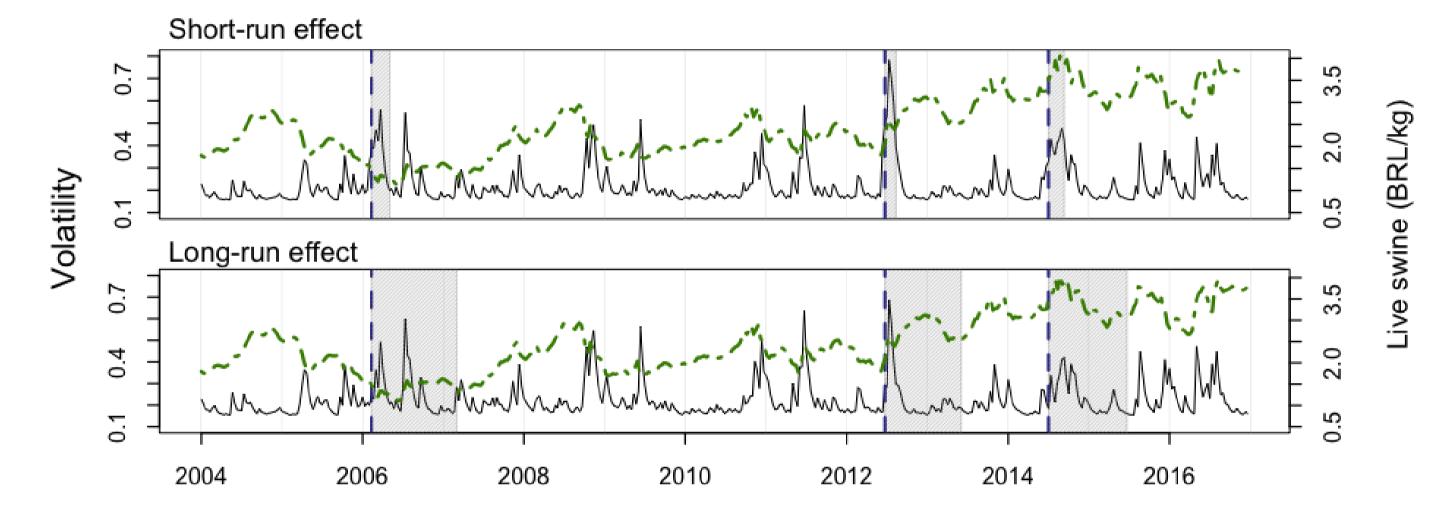
#### Price development - Brazilian pork market



Note: The green line represents the price volatility whereas the grey line denotes the dynamic conditional correlation in percentage. Beef cattle, live chicken and corn are estimated by the EGARCH (1,1) model.

#### **5.2 Impact of Russian trade policies**

Impact on the conditional variance: FMD (112%), WTO (223%) and WC (90%)



Note: The dotted blue vertical lines represent the Russian trade policy interventions: Foot Mouth Disease import ban (FMD), WTO accession, and Western Countries import ban (WC) Source: Own estimation.

Note: Tested variables for volatility drivers on the Brazilian pork market are divided into two groups: a) internal effects (domestic factors) and b) external effects. Source: Own illustration.

## 4. Methodology

• Dynamic Conditional Correlation (DCC) model (Engle, 2002) is used to estimate the multivariate price volatility transmission.

#### **Conditional mean equations**

## 6. Conclusions

- Domestic Brazilian markets show higher price volatility than external factors.
  Brazilian beef and corn react stronger to positive shocks than to negative; indicating the use of "market power".
- Russian trade policies have significantly increased Brazilian live swine price volatilities in the short-run (two months).
- Volatility spill-over across the analysed markets are found. Internal effects have a stronger effect on the live swine price volatility.
- Pork substitutes increase the Brazilian live swine price volatility more than the input markets.
- The correlation between feed grains and live swine is low but the literature suggest that changes in input prices affect producer profitability directly.
- Russian trade policies have affected the price volatility transmission in live chicken, corn and soybean markets.



• Log returns:  $r_{it} = ln(P_{it}/P_{it-1})$ ,  $r = u_t + a_t$ • ARMA model:  $u_t = \phi_o + \sum_{i=1}^p \phi_i r_{t-i} - \sum_{i=1}^q \theta_i a_{t-i} + \sum_{i=1}^l ODx_t$ • Seasonal-ARMA model:  $(1 - \phi L)(1 - \phi_s L^s)r_t = (1 - \theta L)(1 - \theta_s L^s)a_t$ 

**Conditional variance equation** 

 $\begin{array}{ll} \bullet \text{ GARCH (1,1) model:} & \sigma_t^2 = \alpha_o + \alpha_1 a_{t-1}^2 + \beta_1 \sigma_{t-1}^2; & a_t = \sigma_t \epsilon_t \\ \bullet \text{ EGARCH (1,1) model:} & h_t = \alpha_o + (\alpha_1 \epsilon_{t-1} + \gamma(|\epsilon_{t-1}| - E(|\epsilon_{t-1}|)) + \beta_1 h_{t-1}) \end{array} \end{array}$ 

Dynamic Conditional Correlation (DCC-GARCH)

 $egin{aligned} &ullet H_t = D_t R_t D_t; \ \ D_t = diag\{\sigma_1 t, \dots, \sigma_1 t\} \ &ullet R_t = diag\{Q_t\}^{-1/2} \ Q_t \ diag\{Q_t\}^{-1/2} \ &ullet Q_t = (1 - lpha - eta) ar Q + lpha (E_{t-1} E'_{t-1}) + eta Q_{t-1}; \ \ E_t = D_t^{-1} \ a_t \end{aligned}$ 

Impact of Russian trade policies

• GARCH-X (1,1) model:  $\sigma_t^2 = lpha_o + \sum_{i=1}^p lpha_i a_{t-i}^2 + \sum_{j=1}^q eta_j \sigma_{t-j}^2 + \sum_{k=1}^k \delta z_t$ 

Brümmer, B., Korn, O., Schlüßler, K., Jaghdani, T.J., Saucedo, A., 2013. Volatility in the after crisis period – A literature review of recent empirical research. CEPEA, 2017. Brazilian Agricultural Prices. Engle, R., 2002. Dynamic Conditional Correlation. Journal of Business & Economic Statistics 20, 339–350. https://doi.org/10.1198/073500102288618487 Kalkuhl, M., Braun, J. von, Torero, M. (Eds.), 2016. Food Price Volatility and Its Implications for Food Security and Policy. Springer International Publishing, Cham. https://doi.org/10.1007/978-3-319-28201-5 Prikhodko, D., Davleyev, A., 2014. Russian Federation: Meat sector review. Food and Agriculture Organization of the United Nations (FAO), Rome. Rosstat, 2017. Agricultural commodity prices. SEAB, 2017. Information System of Agricultural Market - Paraná, Brazil.

## Contacts

<sup>1</sup> gabriel.rosero-asqui@uni-goettingen.de,

<sup>2</sup> Jaghdani@iamo.de,

<sup>1</sup> bbruemm@gwdg.de.

