Systemic risk and financial connectedness: empirical evidence

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Introduction

One of the main contributions of literature on financial networks is the property of financial system called robust-yet-fragile. The term was first coined by chief economist of Bank of England, Andrew Haldane (Haldane (2009)). He posits that financial connections can serve at the same time as shock-absorbers and shock-amplifiers to the financial sector. This makes the system robust, when the magnitude of shock is relatively small, but fragile, when the shock is large.

A seminal paper by Acemoglu et al. (2015), provides a formal model, in which an extent of financial contagion exhibits a form of regime transition. When the shocks are small, the damages are dissipated through large number of financial institutions. On the other hand, when the shock is above some threshold, the properties of the system changes markedly. The damages are no longer dissipated, but amplified through the network. This makes the effect of connectedness on the system regime-dependent. Similar works that provide a theoretical mechanisms this property are Callaway et al. (2006) and Gai and Kapadia (2010)

This research aims at providing empirical evidence for the regime-dependent effect of connectedness on financial stability.

Research design

The research design is a two step process. Using stock prices data of the biggest banks from eurozone and US, various statistical measures of system connectedness are estimated. The measures are calculated on a rolling basis, providing the time series of connectedness.

In the second step, in order to assess the regime-dependent effect of connectedness on financial stability, the time series of connectedness is used as an explanatory variable in a Markov switching ARCH model.

Measures of financial connectedness

Because of lack of transaction-level data among banks, the econometric literature provides several methods that aim at estimating a level of connectedness

of financial sector. The measures are all estimated based on stock prices data, therefore providing a relatively high frequency variables.

1. Average correlation

$$\bar{\rho}(R) = \frac{\sum_{i \neq i}^{N} \sum_{j \neq j}^{N} \rho_{i,j}(R)}{N^2 - N}$$

Where $\rho_{i,j}(R)$ is a correlation matrix of rate of returns, estimated with a Ledoit-Wolf estimator Ledoit and Wolf (2003). This estimator is shown to have smaller estimation error than sample covariance when the number of observations is relatively small. This is the case for the research design, where the measures of financial connectedness are estimated on a rolling basis.

2. Covariance eigenvalue

Given the sample covariance matrix of rates of returns Σ , the covariance eigenvalues λ are obtained solving the below equation:

$$|\mathbf{A} - \lambda \mathbf{I}| = \mathbf{0}$$

The eigenvalues based measure of connectedness is then defined as:

$$\frac{\sum_{i}^{k}}{\sum_{i}^{N}}$$

This measure captures the proportion of variance of the system that is explained by the first k eigenvalues, as in principal component analysis. The higher the proportion, the more connected the system is.

3. Granger causality network degree

The measure is based on the Granger defined causality Granger (1969). In which a time-series variable x_t "granger" cause variable y_t , when it contains enough information at time t to predict value of y_{t+1} .

Specifically, the granger causality is investigated with following regression:

$$r_{i,t+1} = \beta_0 + \beta_1 r_{m,t} + \beta_2 r_{i,t}$$

Where $r_{j,t}$ is a rate of return of bank j at time t, which allegedly granger causes the rate of return of bank i. The regression also controls for the market rate of return $r_{m,t}$, which is a proxy for the systematic risk.

The banks j and i are said to be connected when the coefficient β_2 is statistically significant.

Given the procedure above we can define an adjacency matrix G describing the relationship between the bansk:

$$G_{i,j} = \begin{cases} 1 & \text{if } j \text{granger cause } i \\ 0 & \text{otherwise} \end{cases} \forall i \neq j$$

Then the measure of granger connectedness is defined as:

$$\frac{\sum_{i \neq j}^{N} \sum_{j \neq i}^{N} G_{i,j}}{N \times (N-1)}$$

The last two of the above connectedness measures are as described in Billio et al. (2012).

Modeling the regime-dependent effect of connectedness

As the theory suggests, the effect of connectedness on financial stability is regime-dependent. In order to capture this property, a Markov switching ARCH model is used to describe the time-varying volatility of banking sector. The proxy of the banking sector being the apropriate banking index (eurozone or US banking sector index, S&P or STOXXX).

The mean specification of the model controls for the first order autocorrelation of the broad market rate of returns (S&P 500 or STOXX600) and an autoregressive component:

$$\mu_i, t = \beta_0 + \beta_1 r_{b,t-1} + \beta_1 r_{m,t-1} + \epsilon_t$$

Where $r_{b,t}$ is the rate of return of the banking sector index and $r_{m,t}$ is the rate of return of the broad market index. The Markov-switching ARCH specification is then:

$$\sigma_t^2 = \alpha_{0,s} + \alpha_1 \epsilon_{t-1}^2 + \alpha_{2,s} \kappa_{t-1}$$

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