# Identifying structural shocks to volatility through a proxy-MGARCH model

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### 1. Motivation

- Multivariate volatility models (MVMs) are good at describing stylized facts of asset returns and widely used in modeling and forecasting second order moment dynamics.
- As a drawback, they do not provide an intelligible interpretation of the shock system driving asset returns – they are "reduced form models" as opposed to structural models as e.g. SVAR models.
- We identify the reduced form MGARCH model through a structural approach using proxy variables and Givens rotations.

# 3. Structural Identification Problem in MVMs

- We consider a system of n speculative log returns  $r_t = \mu_t + \varepsilon_t$ ,  $t \in I \coloneqq \{1, \ldots, T\}$  where  $\mu_t = E[r_t | \mathcal{F}_{t-1}]$ with  $\mathcal{F}_t$  the  $\sigma$ -algebra generated by the returns up to and including time t.
- The reduced form innovations follow an MGARCH model:  $\varepsilon_t | \mathcal{F}_{t-1} \sim (0, H_t)$ . They do not bear an economic interpretation. Let  $\varepsilon_t$  be generated by

 $\varepsilon_t | \mathcal{F}_{t-1} \sim Q_t \xi_t ,$ 

where  $(\xi_t)_{t \in I}$  is an *n*-dimensional vector of structural shocks with  $E[\xi_t] = 0$  and  $E[\xi_t \xi_t^{\top}] = I_n$ .

- $Q_t$  denotes the unknown structural matrix decomposition of  $H_t$  which satisfies  $Q_t Q_t^{\top} = H_t$ .
- Given the principal matrix square root as initial decomposition  $Q_t$ , we identify the true structural matrix decomposition  $Q_t$  by identifying the unique rotation R such that  $Q_t R = Q_t$ . The structural model parameters are given by the rotation matrix.
- Our identification problem differs from the SVAR case with modeled heteroscedasticity as we account for complex dynamics in the conditional covariance process of  $\varepsilon_t$ . E.g.  $Q_t$  thus varies with  $H_t$  over time.

## 4. Identification by Proxy and Orthogonalization by Givens Rotations

• Identification similar to the proxy-SVAR approach of Mertens and Ravn (2013); Stock and Watson (2012). Assume there exists a centered (n-1)-dimensional proxy variable process  $Z = (Z_t)_{t \in I}$  such that, for all  $i=1,\ldots,n-1,$ 

$$E[\xi_{it}Z_{it}] = \phi_i \in \mathbb{R} \setminus \{0\}$$
$$E[\xi_{jt}Z_{it}] = 0 \ (j \neq i)$$

• Then one can estimate the columns of the rotation matrix by

$$\tilde{R}_{\bullet i} = \pm \operatorname{E}[u_t Z_{it}] \left( \operatorname{E}[Z_{it} u_t^{\top}] \operatorname{E}[u_t Z_{it}] \right)^{-1/2}$$

where  $u_t$  denote the MGARCH residuals standardized with the principal matrix square root of  $H_t$ .

- To identify the full rotation matrix we exploit that *n*-dimensional rotations can be expressed as sequences of Givens rotations which guarantees orthogonality of the structural shocks.
- This is of particular importance, as it is a priori not clear which proxy variables deliver orthogonal shocks.

### 2. Data

- We analyse the return system of S&P500, 10 year treasury note yield and Finex USD index from 1998 to 2014.
- Proxy variables are news indicators from Thompson Reuters: U.S. market and U.S. bond market sentiment.

(2)(relevance) (3)(exogeneity)

Set

(1)

(4)

- Inferred structural shocks can be narratively corroborated in detail, they are economically meaningful, interpretable shocks.
- The volatility spillover mechanism is asymmetric. Labelled structural shocks allow us to monitor spillover patterns and directions.
- Structural volatility models open new doors, e.g. to meaningful impulse response analyses.

### 5. Rotation & Structural Model Results

- Every  $\tilde{R} \in \mathbb{R}^{n \times n}$  can be expressed as a composition of  $\frac{n(n-1)}{2}$  "two-dimensional" rotations  $R^{ij}(\theta_{ij})$ .
- $\tilde{R} = \prod \prod R^{ij}(\theta_{ij}) \quad (\theta_{ij} = \text{rotation angle})$  $i=1 \ j=i+1$  $= R^{12}(\theta_{12}) \cdots R^{1n}(\theta_{1n}) R^{23}(\theta_{23}) \cdots R^{n-1,n}(\theta_{n-1,n})$
- Knowledge of the first column of  $\tilde{R}$  allows us to infor the first (n-1) rotation angles and we are left with the identification of a (n-1)-dimensional rotation using the remaining (n-2) proxy variables. We can identify the full rotation recursively.

# 6. Volatility Spillover Analysis

### 7. Conclusion

• Provide a fully identified asset return system.

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• Volatility reception = proportional impact of the structural shocks (equity, bond & currency shock) on each asset (row-wise for S&P500, 10 yr note yield and USD index).

• Equity shock accounts for the largest share in S&P500 volatility, but the bond market shock shows notable contributions especially during calm market periods (see (1)).

• Return on the yield shows pronounced volatility reception from equity shock ((2)).

• Volatility transmission = proportional impact of one structural shock on all assets (columnwise for each structural shock).

• Long-lasting trend in volatility transmission between equity and fixed income markets (see(3) and (4))

