#### Estimating Inflation Expectation Comovement Across Countries

Shi Chen Wolfgang K. Härdle Weining Wang

Ladislaus von Bortkiewicz Chair of Statistics C.A.S.E.-Center for Applied Statistics and Economics Humboldt–Universität zu Berlin King's College London http://lvb.wiwi.hu-berlin.de





#### Inflation Expectation

- Important to financial investors and policy makers
- Inflation surprise can make real impact
- Important to estimate underlying state of the economy
   ...



#### **General Goal**

- Joint modeling of inflation expectation cross regions (states/countries)
  - Providing informative estimates of inflation expectations
- □ Accounting for the non-linear dependency among countries
  - ▶ A GeoCopula Model
- □ Cross sectional forecast and forecast of inflation expectation



#### **Measure of Inflation**

 $\boxdot$  Break-even inflation rate (BEIR) with maturity au,

$$BEIR_t(\tau) = y_t^N(\tau) - y_t^R(\tau)$$

• 
$$y_t^N(\tau)$$
 - nominal yield  
•  $y_t^R(\tau)$  - real yield

Decompose BEIR,

$$BEIR_t(\tau) = \pi_t^e(\tau) + else$$

- $\pi_t^e(\tau)$  is expected inflation
- else includes risk premium, convexity effects



#### **BEIR of European Countries**



Figure 1: Observed BEI rates (percent) of UK, France, Italy, Sweden and Germany, maturity five years.

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- ☑ Co-movement exists between countries
- Data from different countries are available at different maturities and month

#### Challenges

- Joint yield curve modeling across multiple maturities
- BEIR decomposition
- Panel model of inflation expectations
- Non linear dependency among countries
- □ New estimation and forecast of inflation expectation



#### Outline

- 1. Motivation  $\checkmark$
- 2. Yield Curve Modeling
- 3. BEIR decomposition
- 4. Dynamics of Inflation Expectation
- 5. Empirical Results
- 6. Conclusion

#### Model Approach– Single Country

☑ Nelson and Siegel (1987): Classical NS model

$$y( au) = eta_0 + eta_1\left(rac{1-e^{-\lambda au}}{\lambda au}
ight) + eta_2\left(rac{1-e^{-\lambda au}}{\lambda au} - e^{-\lambda au}
ight)$$

Diebold and Li (2006): Dynamic NS (DNS) model

$$y_t(\tau) = L_t + S_t\left(rac{1-e^{-\lambda au}}{\lambda au}
ight) + C_t\left(rac{1-e^{-\lambda au}}{\lambda au} - e^{-\lambda au}
ight)$$

 Christensen et.al (2011): Arbitrage-free DNS (AFNS) model The closest match to DNS yield is

$$y_t(\tau) = L_t + S_t \left(\frac{1 - e^{-\lambda \tau}}{\lambda \tau}\right) + C_t \left(\frac{1 - e^{-\lambda \tau}}{\lambda \tau} - e^{-\lambda \tau}\right) - \frac{A(\tau)}{\tau}$$

state variable  $X_t^{\top} = (L_t, S_t, C_t)$ Estimating Inflation Expectation Co-movement Across Countries —



#### Joint AFNS model

 $\Box$  The separate AFNS models of nominal and inflation-indexed type for a specific country i,

$$y_{it}^{N}(\tau) = \mathcal{L}_{it}^{N} + S_{it}^{N} \left(\frac{1 - e^{-\lambda\tau}}{\lambda\tau}\right) + C_{it}^{N} \left(\frac{1 - e^{-\lambda\tau}}{\lambda\tau} - e^{-\lambda\tau}\right) - \frac{A_{i}^{N}(\tau)}{\tau}$$
$$y_{it}^{R}(\tau) = \mathcal{L}_{it}^{R} + S_{it}^{R} \left(\frac{1 - e^{-\lambda\tau}}{\lambda\tau}\right) + C_{it}^{R} \left(\frac{1 - e^{-\lambda\tau}}{\lambda\tau} - e^{-\lambda\tau}\right) - \frac{A_{i}^{R}(\tau)}{\tau}$$

⊡ We assume,

$$S_{it}^{R} = \alpha_{i}^{S} S_{it}^{N}$$
$$C_{it}^{R} = \alpha_{i}^{C} C_{it}^{N}$$



## Joint AFNS model

 $\boxdot$  The joint AFNS yield curve for country i with maturity  $\tau$  is

$$\begin{pmatrix} y_{it}^{N}(\tau) \\ y_{it}^{R}(\tau) \end{pmatrix} = \begin{pmatrix} 1 & \frac{1-e^{-\lambda_{i}\tau}}{\lambda_{i}\tau} & \frac{1-e^{-\lambda_{i}\tau}}{\lambda_{i}\tau} - e^{-\lambda_{i}\tau} & 0 \\ 0 & \alpha_{i}^{S} \frac{1-e^{-\lambda_{i}\tau}}{\lambda_{i}\tau} & \alpha_{i}^{C} (\frac{1-e^{-\lambda_{i}\tau}}{\lambda_{i}\tau} - e^{-\lambda_{i}\tau}) & 1 \end{pmatrix} \\ \begin{pmatrix} L_{it}^{N} \\ S_{it}^{N} \\ L_{it}^{N} \end{pmatrix} + \begin{pmatrix} \varepsilon_{it}^{N}(\tau) \\ \varepsilon_{it}^{R}(\tau) \end{pmatrix} - \begin{pmatrix} \frac{A_{i}^{N}(\tau)}{A_{i}^{T}(\tau)} \\ \frac{A_{i}^{R}(\tau)}{\tau} \end{pmatrix}$$



## **Yield Curve Modeling**

 $\boxdot$  For country *i* with maturity  $au = ( au_1, au_2, \dots, au_n)$  is



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#### **BEIR decomposition**

#### 🖸 That is,

 $BEIR_t(\tau) = y_t^N(\tau) - y_t^R(\tau) = \pi_t^e(\tau) + \eta_t(\tau) + \phi_t(\tau)$ 



$$\pi^{e}_{t}(\tau) = -\frac{1}{\tau} \log \mathsf{E}^{\mathcal{P}}_{t} \left[ \exp\left\{ -\int_{t}^{t+\tau} (r^{\mathcal{N}}_{s} - r^{\mathcal{R}}_{s}) ds \right\} \right]$$



# Inflation Expectation Estimates– Single Country



Figure 2: Model-implied inflation expectation for different countries - UK, France, Italy, Sweden and Germany. Estimating Inflation Expectation Co-movement Across Countries

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# IE Dynamics with Credit Risk Factor (with a fixed $\tau$ )



☑ The dynamics of the common factor,

 $\Pi_t = p + q \Pi_{t-1} + \nu_t$ 

- where m<sub>i</sub>, n<sub>i</sub>, p and q are unknown parameters
- the errors ν<sub>t</sub>s are i.i.d white noises and μ<sub>it</sub> follows a GeoCopula model
- d<sub>it</sub> is the CDS varying over time



## Copulae

A continuous function  $C:[0,1]^d \rightarrow [0,1]$ ,

$$C(u_1,\ldots,u_d) = F\{F_1^{-1}(u_1),\ldots,F_d^{-1}(u_d)\}, \quad u_1,\ldots,u_d \in [0,1],$$

where  $F_1^{-1}(\cdot),\ldots,F_d^{-1}(\cdot)$  the quantile functions.

- Separate dependency and marginal distributions
- Represent general dependency



#### A GeoCopulae Model

$$\mu_{it} = \alpha_{it} + \xi_{it},\tag{1}$$

 $\alpha_{it}$  the spatial temporal variation,  $\xi_{it}$  the i.i.d. noise with mean 0 and variance  $\sigma_{\xi}$ , see Bai et. al. 2014.

$$F_{it}(\alpha) = \Phi_{NT} \{ \Phi^{-1}(F_{11}(\alpha_{11})), \cdots, \Phi^{-1}(F_{N,T}(\alpha_{NT})) | \Sigma \}, \quad (2)$$

 $\Phi_{NT}(\cdot)$  the cumulative distribution function (c.d.f.) of a multivariate Gaussian distribution with a variance covariance matrix  $\Sigma$ , which models the spatiotemporal dependence. Details



#### A Spatialtemporal Variagram

To understand the spatial temporal correlation, define

$$\Gamma(t_1 - t_2, n_1 - n_2) \stackrel{\text{def}}{=} \frac{1}{2} E(\alpha_{t_1 n_1} - \alpha_{t_2 n_2})^2.$$
(3)

Closely related to covariance function. For stationary process with  $\sigma_2$  are the variance,

$$\Gamma(t_1 - t_2, n_1 - n_2) \stackrel{\text{def}}{=} \sigma_2 - Cov(\alpha_{t_1 n_1}, \alpha_{t_2 n_2}). \tag{4}$$



| France | Germany | Italy | Sweden | UK   |
|--------|---------|-------|--------|------|
| 2.2    | 13.3    | 12.3  | 18.0   | 0.09 |
| 48.5   | 52.3    | 41.5  | 59.1   | 51.3 |

Table 1: The Coordinates of countries, in Degree.

#### Empirical variagram is defined as

$$\hat{\Gamma}(d_1, d_2) \stackrel{\text{def}}{=} \frac{1}{N_{d_1, d_2}} \sum_{t_1, t_2, n_1, n_2 : \|t_1 - t_2\| \le d_1, \|n_1 - n_2\| \le d_2} (\alpha_{t_1 n_1} - \alpha_{t_2 n_2})^2,$$
(5)

where  $N_{d_1,d_2}$  is the number of pairs, which has the spatial and temporal distance in range.



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#### Data

Bloomberg: monthly zero-coupon government bond yield.
 Type: nominal y<sup>N</sup>, inflation-indexed y<sup>R</sup>.

| Data    | Span                  | Maturity     |  |
|---------|-----------------------|--------------|--|
| UK      | 30.06.2006-31.12.2014 | 3,4,5 years  |  |
| France  | 30.06.2006-31.12.2014 | 3,5,10 years |  |
| Italy   | 29.06.2007-31.12.2014 | 3,5,10 years |  |
| Sweden  | 30.04.2007-29.08.2014 | 3,5,10 years |  |
| Germany | 30.06.2009-31.12.2014 | 5,7,10 years |  |



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#### **Estimated IE- Single Country**





## 3-year IE forecast- Single Country



Figure 3: Model-implied inflation expectation for different countries - UK, France, Italy, Sweden and Germany.



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#### Common effect with $d_{it}$





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#### **Residuals of common effect**



 Figure 5: Model residual for IE dynamics without macroeconomic factor 

 UK, France, Italy, Sweden and Germany.

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#### Estimates with d<sub>it</sub>

| Country-specific equations |                        |                             |
|----------------------------|------------------------|-----------------------------|
| UK                         | $\pi_{1t}^{e}(\tau) =$ | $-0.358d_{it} + 0.798\Pi_t$ |
| France                     | $\pi_{2t}^{e}(\tau) =$ | $0.085d_{it} + 0.714\Pi_t$  |
| Italy                      | $\pi_{3t}^{e}(\tau) =$ | $1.078d_{it} + 0.531\Pi_t$  |
| Sweden                     | $\pi^{e}_{At}(\tau) =$ | $-0.621d_{it} + 0.805\Pi_t$ |
| Germany                    | $\pi^{e}_{5t}(\tau) =$ | $0.045d_{it} + 0.700\Pi_t$  |
| Common Effect equation     |                        |                             |
|                            | $\Pi_t =$              | $0.382 + 0.976\Pi_{t-1}$    |

Table 2: Estimates for the dynamics of IE.



#### Variance decomposition

- According to the joint model of IE dynamics, decompose the variation of IE  $\pi_{it}^e$  into parts driven by,
  - common effect variation
  - country-specific variation
  - sovereign risk variation

 $\operatorname{Var}(\pi_{it}^{e}) = n_{i}^{2}\operatorname{Var}(\Pi_{t}) + l_{i}^{2}\operatorname{Var}(d_{it}) + \operatorname{Var}(\mu_{it})$ 



#### Joint IE dynamics with $d_{it}$

|                         | U.K.  | France | ltaly | Sweden | Germany |
|-------------------------|-------|--------|-------|--------|---------|
| Common effect           | 36.08 | 33.59  | 11.54 | 31.87  | 32.84   |
| Country-specific effect | 56.66 | 65.88  | 40.92 | 49.17  | 67.02   |
| Sovereign risk effect   | 7.26  | 0.53   | 47.55 | 18.96  | 0.14    |

Table 3: Variations explained in percentage



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Figure 6: The empirical fitted variogram (left) and the parametrically fitted variogram (right).

 $\hat{\eta} = 0.5$ , given  $\hat{\beta} = 0.1028$ ,  $\hat{a} = 0.001795$  means that the marginal temporal correlation decreases by around 8% with 1 month increase in time, and  $\hat{b} = 0.000298$  indicates that the marginal space correlation decays by around 9% with a 100-km increase. Estimating Inflation Expectation Co-movement Across Countries —

#### Forecast



Figure 7: The forecast of common inflation factor derived from the joint model of IE dynamics with CDS. Estimating Inflation Expectation Co-movement Across Countries —

#### Forecast with Geocopula



Figure 8: The comparision of two forecasts with (dotted) and without(solid) Geocopula, the estimation results are derived from the joint model of IE dynamics with default factor.



#### Comparison



Figure 9: Comparison of model-implied level, the observed inflation level, 1year SPF forecast and 2-year SPF forecast of inflation (Survey Professional Extimating Inflation Expectation Co-movement Across Countries —

# Forecast inflation and cross sectional forecast

|                          | U.K.  | France | ltaly | Sweden | Germany |
|--------------------------|-------|--------|-------|--------|---------|
| Without GeoCopula        | 0.413 | 0.381  | 0.463 | 0.386  | 0.298   |
| With GeoCopula           | 0.116 | 0.056  | 0.145 | 0.122  | 0.142   |
| Cross sectional forecast | 0.345 | 0.312  | 0.487 | 0.344  | 0.201   |

Table 4: Averaged one year ahead forecast starting from 201401(squared mean squared error)



## Conclusion

- Common inflation factor Π<sub>t</sub> is an important drivers of country-specific inflation expectations.
- The model extracts informative estimates of inflation expectations
- ☑ Will provide good implications for monetary policies



#### References



Nelson, C.R., Siegel, A.F. Parsimonious modeling of yield curves. Journal of Bussiness, 60:473-489, 1987.

📄 Francis X Diebold and Li. C. Forecasting the term structure of government bond yields. Journal of Econometrics, 130: 337-364, 2006,



🔈 Cochrane, J.H.

Asset Pricing.

Princeton university press, 2005.



#### References

 Jens HE Christensen, Francis X Diebold and Glenn D Rudebusch
 The affine arbitrage-free class of Nelson-Siegel term structure models.
 Journal of Econometrics, 164(1):4-20, 2011.

Francis X Diebold, Canlin Li and Vivian Z Yue Global yield curve dynamics and interactions: a dynamic Nelson-Siegel approach. Journal of Econometrics, 146(2):351-363, 2008.



#### References

- Francis X Diebold, Glenn D Rudebusch and S Boragan Aruoba The macroeconomy and the yield curve: a dynamic latent factor approach. Journal of Econometrics, 115(1):32-39, 2012.
- Jens H.E. Christensen, Jose A. Lopez and Glenn D. Rudebusch Inflation expectations and risk premiums in an Arbitrage-Free model of nominal and real bond yields. Journal of Money, Credit and Banking, 11:143-178, 2010.



#### Dynamics of state variable

 Derived from affine AF model of Duffie&Kan(2002), the real world P-dynamics is

$$dX_{it} = K_i^P(t)[\theta_i^P(t) - X_{it}]dt + \Sigma_i(t)dW_{it}^P$$

$$X_{it} = \Phi^0_{i,\Delta t} + \Phi^1_{i,\Delta t} X_{i,t-1} + \eta_{it}$$

with

$$\Phi^0_{i,\Delta t} = I - \exp(-K^P_i \Delta t) \theta^P_i$$
  
 $\Phi^1_{i,\Delta t} = \exp(-K^P_i \Delta t)$  (Return)



$$F(\alpha_{t_1,n_1}, \alpha_{t_2,n_2}) = \Phi_2(\Phi^{-1}(F_{t_1,n_1}(\alpha_{t_1,n_1})), \Phi^{-1}(F_{t_2,n_2}(\alpha_{t_2,n_2}))|\Sigma_{t_1,n_1,t_2,n_2})$$
(6)

where  $\sum_{t_1,n_1,t_2,n_2}$  is a submatrix of  $\sum$ .

$$\sigma(n_2-n_1,t_2-t_1)$$

$$= \sigma(v, u) \stackrel{\text{def}}{=} \left\{ \begin{array}{c} \frac{2\sigma^2\beta}{(a^2u^2+1)^{\eta}(a^2u^2+\beta)\gamma(\eta)} (\frac{b}{2}(\frac{a^2u^2+1}{a^2u^2+\beta})^{1/2}v)^{\eta} K_{\eta}(b(\frac{a^2u^2+1}{a^2u^2+\beta})^{1/2}v) \text{ if } v > 0 , \\ \frac{\sigma^2\beta}{(a^2u^2+1)^{\eta}(a^2u^2+\beta)\gamma(\eta)} \text{ if } v = 0, \end{array} \right.$$
  
where  $a, b, \beta, \eta$  are parameters,  $\gamma(\eta)$  is the gamma function and  $K_{\eta}(\cdot)$  is the Bessel function of the second kind.  $\checkmark$  Go Back



Appendix

$$I(\theta, d_{1}, d_{2}) = \sum_{\substack{t_{1}, t_{2}, n_{1}, n_{2}: ||t_{1} - t_{2}|| \leq d_{1}, ||n_{1} - n_{2}|| \leq d_{2}} \log f_{\alpha_{t_{1}n_{1}}, \alpha_{t_{2}n_{2}}}, \quad (7)$$
where  $\theta \stackrel{\text{def}}{=} (a, b, \beta, \eta)^{\top}$ .
$$f_{\alpha_{t_{1}, n_{1}}, \alpha_{t_{2}, n_{2}}} \stackrel{\text{def}}{=} c_{\Phi} \{F(\alpha_{t_{1}, n_{1}}), F(\alpha_{t_{2}, n_{2}})\} f(\alpha_{t_{1}, n_{1}}) f(\alpha_{t_{2}, n_{2}})$$
with

$$c_{\Phi}\{F(\alpha_{t_1,n_1}),F(\alpha_{t_2,n_2})\} = |\Sigma_{t_1,n_1,t_2,n_2}|^{-1/2} \exp\{q^{\top}(I_2 - \Sigma_{t_1,n_1,t_2,n_2}^{-1})q\},$$

$$q\stackrel{\mathrm{def}}{=}(q_{t_1,n_1},q_{t_2,n_2})$$

and

$$q_{t_i,n_i} = \Phi^{-1}\{\hat{F}(x_{t_i,n_i})\}.$$

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#### **BEIR decomposition**

Cochrane (2005), the price of the zero-coupon bond that pay one unit of consumption basket at time t,

$$P_t^N(\tau) = \mathsf{E}_t \left( M_{t+1}^N M_{t+2}^N \cdots M_{t+\tau}^N \right)$$
$$P_t^R(\tau) = \mathsf{E}_t \left( M_{t+1}^R M_{t+2}^R \cdots M_{t+\tau}^R \right)$$

*M<sup>N</sup><sub>t</sub>* and *M<sup>R</sup><sub>t</sub>* are SDFs.
 Under assumption of no arbitrage,

$$\frac{M_t^N}{M_t^R} = \frac{Q_{t-1}}{Q_t}$$



 $Q_t$  is the overall price level of consumption basket.



## **BEIR decomposition**

⊡ Converting equations,

$$y_t(\tau) = -\frac{1}{\tau} \log P_t(\tau)$$
$$\pi_{t+1} = \log \frac{Q_{t+1}}{Q_t}$$

☑ The BEIR can be defined as follows,

$$y_t^N(\tau) - y_t^R(\tau) = \frac{1}{\tau} \mathsf{E}_t \left( \log \frac{M_{t+1}^N}{M_{t+1}^R} \cdots \frac{M_{t+\tau}^N}{M_{t+\tau}^R} \right) - \frac{1}{2\tau} \mathsf{Var}_t \left( \log \frac{M_{t+1}^N}{M_{t+1}^R} \cdots \frac{M_{t+\tau}^N}{M_{t+\tau}^R} \right) \\ + \frac{1}{\tau} \mathsf{Cov}_t \left( \log \frac{M_{t+1}^N}{M_{t+1}^R} \cdots \frac{M_{t+\tau}^N}{M_{t+\tau}^R}, \log M_{t+1}^R \cdots M_{t+\tau}^R \right)$$

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