

# Decomposing Euro-Area Sovereign Spreads: Credit and Liquidity Risks

A. Monfort<sup>1</sup> and J.-P. Renne<sup>2</sup>

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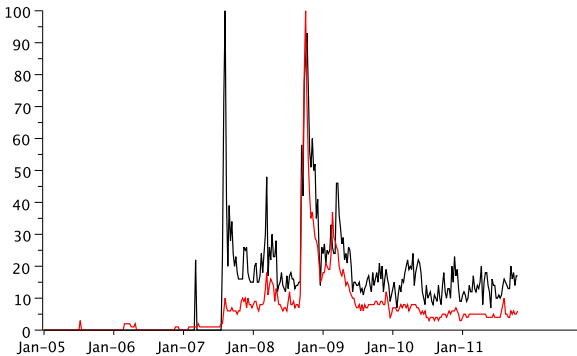
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<sup>2</sup>Banque de France. The views expressed in the following are those of the authors and do not necessarily reflect those of the Banque de France.

# Introduction

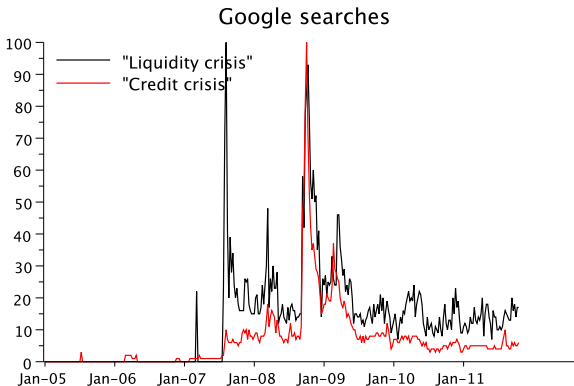
## Credit and / vs Liquidity



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# Introduction

## Credit and / vs Liquidity

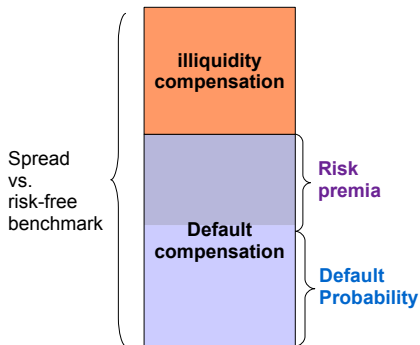


Source: Google Insights, [www.google.com/insights/](http://www.google.com/insights/)

# Introduction

## Motivations

- ① Are euro-area sovereign yields affected by **liquidity-pricing effects** (and to what extent)?
- ② **Interactions** between credit-related and liquidity-related risk factors?
- ③ How to **extract probabilities of default** (PDs) from bond prices?
- ④ Do the default compensations include **risk premia**? [*credit-spread puzzle*, Huang and Huang 2012, Chen et al. 2009]



## Introduction

### Disentangling credit and liquidity risks: what for?

- **Policy implications** of a rise in spreads depend on the source:
  - Liquidity problems  $\Rightarrow$  improve market functioning
  - Credit concerns  $\Rightarrow$  enhance the solvency of the debtors (Codogno, Favero and Missale, 2003)
- **Investment decisions** (Longstaff, 2009):
  - Buy-and-hold investors seek bonds whose price is low because of poor liquidity
  - Bond A and Bond B have the same credit quality but A is less liquid  $\Rightarrow$  buy-and-hold investors buy A because less expensive than B for the same final payoff distribution

# Introduction

## Overview of connected literature

- **Disentangling credit/liquidity risks:** (using spreads that are known to be liquidity-driven)  
Spreads between bonds of the same maturity but **different ages** (Fontaine and Garcia, 2012), between **T-bill rates and repo rates** (Liu, Longstaff and Mandel, 2006), between **govies and swaps** (Feldhütter and Lando, 2008)
- **Credit affine term-structure models:** Duffie and Singleton (1999), Geyer, Kossmeier and Pichler (2004), Longstaff, Mithal and Neis (2005), Mueller (2008)
- **Regime-switching in ATSM:** Ang, Bekaert and Wei (2008), Monfort and Pegoraro (2007), Dai, Singleton and Yang (2007), Monfort and Renne (2013), Gourieroux et al. (2013)
- **Extraction of PDs** from market prices: Litterman and Iben (1991)
- **Sovereign risk premia:** Borri and Verdelhan (2012), Longstaff et al. (2011), Ang and Longstaff (2011)

# Introduction

## Outline

- ① Model
- ② Data and estimation
- ③ Results
  - ① Credit/liquidity decomposition
  - ② Probabilities of default
- ④ Concluding remarks

# Model

## Variables

- $N$  debtors (countries) that issue defaultable and illiquid bonds
- **Credit risk**: Debtors may default
  - Default variable:  $d_t^{(n)} = 1$  if in debtor  $n$  in default at  $t$  (0 otherwise)
  - Default intensity:  $\lambda_{c,t}^{(n)}$
  - Fractional loss given default (LGD):  $1 - \zeta$
- **Liquidity risk**: Bondholders may have to liquidate hastily [Ericsson and Renault (2006)]
  - Liquidity-shock variable:  $\ell_t$
  - Liquidity-shock intensity:  $\lambda_{\ell,t}$
  - Fractional loss given liquidation (LGL):  $1 - \theta^{(n)}$
- $r_t$ : risk-free short-term rate
- $z_t$ : crisis-regime variable



Model  
Overview

- **Credit risk:** Fractional-loss credit intensity  $\lambda_{fc,t}^{(n)}$ :

$$\lambda_{fc,t}^{(n)} = \underbrace{(1 - \zeta)}_{\text{fractional LGD}} \times \underbrace{\lambda_{c,t}^{(n)}}_{\text{default intensity}}$$

- **Liquidity risk:** Fractional-loss liquidity intensity  $\lambda_{fl,t}^{(n)}$ :

$$\lambda_{fl,t}^{(n)} = \underbrace{(1 - \theta^{(n)})}_{\text{fractional loss}} \times \underbrace{\lambda_{l,t}}_{\text{liq.-shock intensity}}$$

# Model

## Historical dynamics of the regimes

- Liquidity-related stress:  $z_{\ell,t}$
- Credit-related stress:  $z_{c,t}$
- For  $i$  in  $\{\ell, c\}$ , three stress levels:

$$z_{i,t} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} : \text{low stress} \quad z_{i,t} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} : \text{distress} \quad z_{i,t} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} : \text{severe stress}$$

- 9 unobserved regimes  $z_t = z_{\ell,t} \otimes z_{c,t}$
- The two chains ( $z_{\ell,t}$  and  $z_{c,t}$ ) cause each other:

$$\begin{cases} \mathbb{P}(z_{\ell,t} | z_{\ell,t-1}) \neq \mathbb{P}(z_{\ell,t} | z_{\ell,t-1}, z_{c,t-1}) \\ \mathbb{P}(z_{c,t} | z_{c,t-1}) \neq \mathbb{P}(z_{c,t} | z_{\ell,t-1}, z_{c,t-1}) \end{cases}$$

## Model

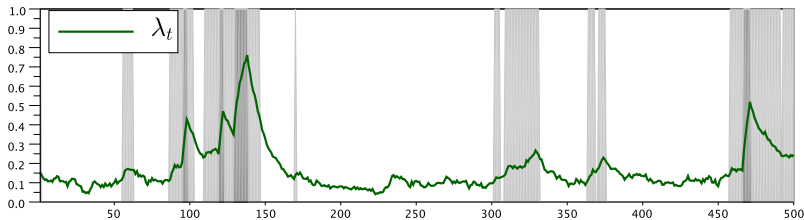
### Historical dynamics of the intensities

Generic process followed by the intensities (the  $\lambda_{c,t}^{(n)}$ 's and  $\lambda_{\ell,t}$ ):

$$\lambda_t = \mu' z_t + \rho \lambda_{t-1} + \sigma \varepsilon_t$$

Simulated example:

with  $\mu = [ 0.01 \quad 0.03 \quad 0.10 ]$ ,  $P = \begin{bmatrix} 0.98 & 0.02 & 0 \\ 0.05 & 0.90 & 0.05 \\ 0 & 0.30 & 0.70 \end{bmatrix}$ . and  $\varepsilon_t \sim \text{i.i.d. } \mathcal{N}(0, 0.01^2)$



Light grey: crisis regime ( $z_t = [0, 1, 0]'$ ), Dark grey: severe crisis regime ( $z_t = [0, 0, 1]'$ )

## Model

### S.D.F.

- Stochastic discount factor (s.d.f.) between  $t - 1$  and  $t$ :

$$M_{t-1,t} = \exp \left[ -r_{t-1} - \frac{1}{2} \nu'_t \nu_t + \nu'_t \varepsilon_t + (\delta z_{t-1})' z_t \right]$$

- The risk-sensitivity matrix  $\delta$  and the vectors  $\nu_t$  respectively price the regimes  $z_t$  and the (standardized) Gaussian innovations  $\varepsilon_t$  of  $\lambda_t$ .
- Under  $\mathbb{Q}$ :
  - $z_t$  follows a time-homogenous Markovian chain whose dynamics is described by transition probabilities  $\pi_{ij}^*$  (given by  $\pi_{ij} \exp \delta_{ij}$ )
  - The intensities follow:

$$\lambda_{i,t} = \mu_i^* z_t + \rho_i^* \lambda_{i,t-1} + \sigma_i \varepsilon_{i,t}^*$$

where  $\varepsilon_{i,t}^* \sim \mathcal{N}^{\mathbb{Q}}(0, 1)$ ,  $\mu_i^* = \mu_i + \sigma_i \nu'_{z,i}$  and  $\rho_i^* = \rho_i + \sigma_i \nu_{\lambda,i}$   
( $\nu_{i,t} = \nu_{\lambda,i} \lambda_{i,t-1} + \nu'_{z,i} z_t$ )

## Model

Risk-neutral dynamics of  $d_t$  and  $\ell_t$

$\mathbb{P}$  and  $\mathbb{Q}$  intensities in this framework

- The conditional distributions of  $d_t$  and  $\ell_t$ , given  $(\lambda_{c,t}, \lambda_{\ell,t}, \underline{W}_{t-1})$ , are the same functions of  $\lambda_{c,t}$  and  $\lambda_{\ell,t}$  under  $\mathbb{P}$  and  $\mathbb{Q}$
- The default and liquidity intensities are the same process in both worlds
- This stems from the fact that the variables  $d_t$  and  $\ell_t$  do not enter the s.d.f.

## Model

### Risk-neutral dynamics of $d_t$ and $\ell_t$

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- The default and liquidity intensities are the same process in both worlds
- This stems from the fact that the variables  $d_t$  and  $\ell_t$  do not enter the s.d.f.
- The intensities are the same processes under both measures **but their  $\mathbb{Q}$ - and  $\mathbb{P}$ -dynamics are different.** (Hence  $\mathbb{Q}$ - and  $\mathbb{P}$ -PDs are different.)

# Model

## Markov-switching State-space model

### Bond spreads formulas

$$\begin{aligned}y_{t,h}^{(n)} - r_{t,h} &= -\frac{1}{h} \ln E_t^{\mathbb{Q}} \exp \left( -\lambda_{fc,t+1}^{(n)} - \dots - \lambda_{fc,t+h}^{(n)} - \lambda_{fl,t+1}^{(n)} - \dots - \lambda_{fl,t+h}^{(n)} \right) \\ &= a_h^{(n)'} z_t + b_h^{(n)'} \lambda_t\end{aligned}$$

where the  $a_h^{(n)}$ 's and the  $b_h^{(n)}$ 's are computed recursively.

- Model in state-space form ( $S_t$ : vector of spreads):

$$\begin{cases} \lambda_t = \mu' z_t + \Phi \lambda_{t-1} + \Sigma \varepsilon_t \\ S_t = A z_t + B \lambda_t + \xi_{S,t} \end{cases}$$

# Model

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- Model in state-space form ( $S_t$ : vector of spreads):

$$\begin{cases} \lambda_t = \mu' z_t + \Phi \lambda_{t-1} + \Sigma \varepsilon_t \\ CF_t = A_{CF} z_t + B_{CF} \lambda_t + \xi_{CF,t} \\ S_t = A z_t + B \lambda_t + \xi_{S,t} \end{cases}$$

⇒ Small-sample persistence bias problem: survey-based forecast are introduced  
[Kim and Orphanides, 2012]



# Data and estimation

## Data

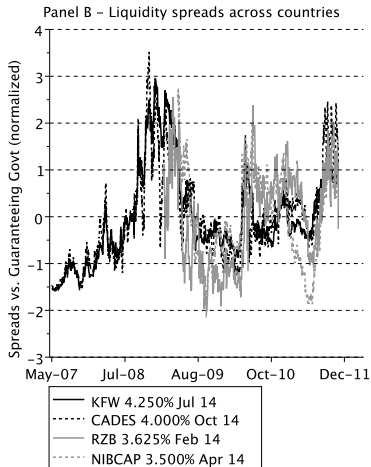
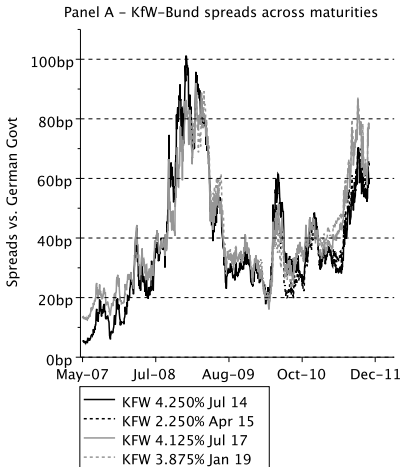
- Weekly data July 2006 – February 2013
- Eight euro-area countries: Austria, Belgium, Finland, France, Germany, Italy, the Netherlands and Spain
- Riskfree yields  $r_{t,h}$ :

$$r_{t,h} = \underbrace{\text{German yields}}_{\text{riskfree yd} + \text{German credit risk}} - \underbrace{\text{German CDS}}_{\text{German credit risk}}$$

- 12-month-ahead forecasts of 10-year sovereign yields (France, Germany, Italy, Netherlands and Spain) are exploited
- **MLE estimation**: Log-likelihood of the Markov-Switching State-Space model computed by means of Kim's (1994) algorithm
- **Liquidity-factor identification**: KfW bonds feature the same credit quality as Germany but are less liquid

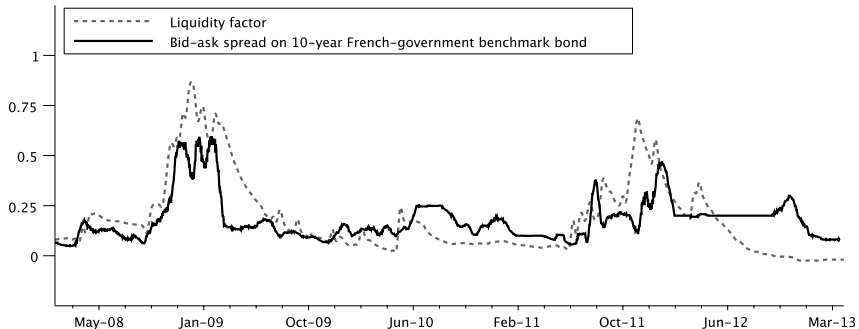
# Data and estimation

Liquidity factor: sprds between gov-guaranteed bonds and govies



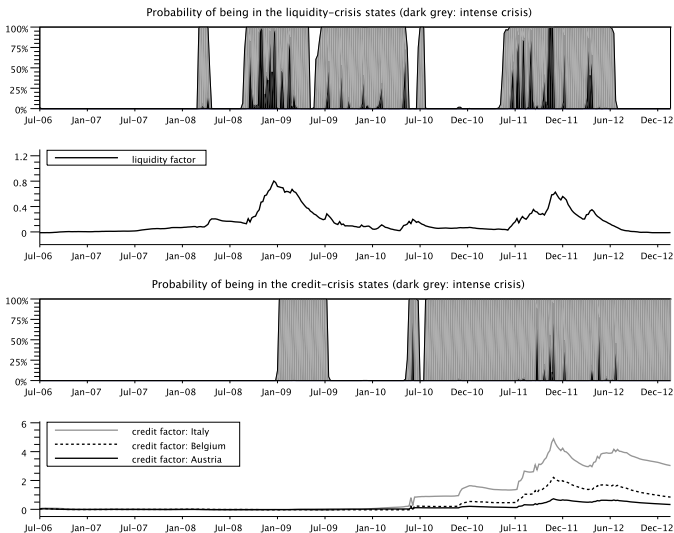
# Data and estimation

Liquidity factor  $\lambda_{\ell,t}$



# Data and estimation

## Estimated crisis regimes



# Data and estimation

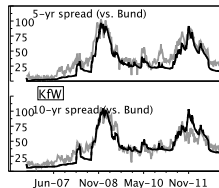
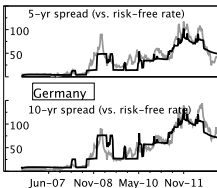
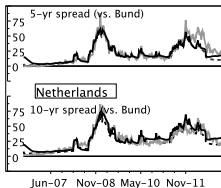
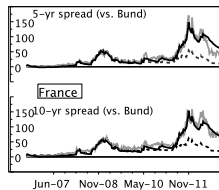
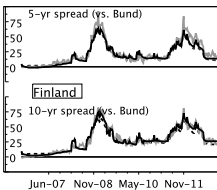
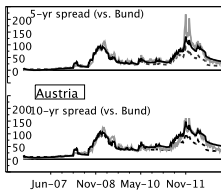
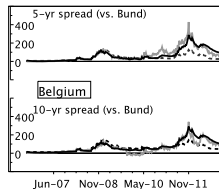
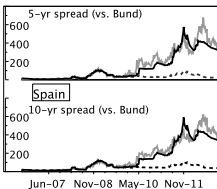
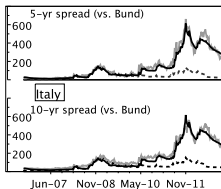
## Transition probabilities

		Under $\mathbb{P}$		
		$NL_{t+1}$	$L_{t+1}$	$LL_{t+1}$
$NL_t$	$NC_t$	0.999*** (0.0011)	0.00106 (0.0011)	- -
	$C/CC_t$	0.96*** (0.012)	0.043*** (0.012)	- -
$L_t$	$NC_t$	0.037*** (0.002)	0.85*** (0.042)	0.109*** (0.042)
	$C/CC_t$	0.0061 (0.0055)	0.89*** (0.034)	0.105*** (0.034)
$LL_t$		-	0.69***	0.31***
		-	(0.092)	(0.092)

## Data and estimation

## Transition probabilities

		Under $\mathbb{Q}$		
		$NL_{t+1}$	$L_{t+1}$	$LL_{t+1}$
$NL_t$	$NC_t$	0.999*** (0.00035)	0.0015*** (0.00035)	- -
	$C/CC_t$	0.98*** (0.0056)	0.021*** (0.0056)	- -
$L_t$	$NC_t$	0.019*** (0.004)	0.000049 (0.12)	0.98*** (0.11)
	$C/CC_t$	0.000048 (0.0046)	0.00004 (0.13)	0.9999*** (0.13)
$LL_t$		-	1***	0.000049
		-	(0.088)	(0.088)



# Credit/Liquidity

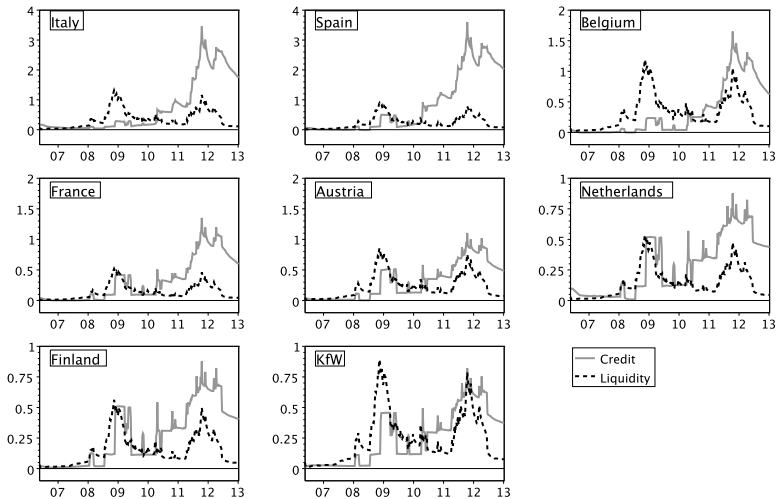
## Risk Premia Analysis (5-year maturity)

	Expectation part of the spreads				Risk premiums			
	Credit		Liquidity		Credit		Liquidity	
	Mean	SD	Mean	SD	Mean	SD	Mean	SD
Austria	0.06	0.09	0.08	0.04	0.31	0.29	0.23	0.20
Belgium	0.17	0.26	0.12	0.06	0.33	0.42	0.32	0.27
Finland	0.03	0.05	0.05	0.03	0.27	0.24	0.15	0.13
France	0.11	0.16	0.06	0.03	0.33	0.35	0.14	0.12
Germany	0.04	0.05	-	-	0.25	0.22	-	-
Italy	0.49	0.63	0.10	0.07	0.74	0.91	0.35	0.30
Netherlands	0.04	0.04	0.05	0.03	0.28	0.24	0.14	0.12
Spain	0.45	0.63	0.14	0.05	0.82	0.98	0.24	0.20



# Credit/Liquidity

## Risk Premia Analysis (5-year maturity)



## Credit/Liquidity

### Risk Premia Analysis (5-year maturity)

- Regression analysis suggests that the different parts of the spreads are related to macro-finance indicators:

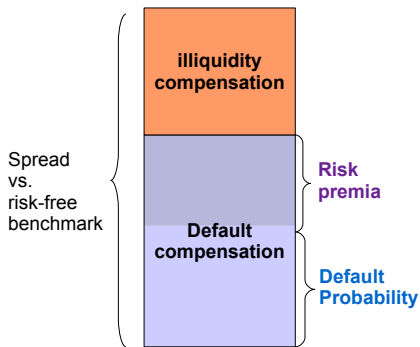
	Expectation part	Risk premium
Credit	VSTOXX, IBOR-OIS, BVOL	VSTOXX, EUROSTOXX, BVOL
Liquidity	EUROSTOXX	IBOR-OIS, EUROSTOXX

# Credit/Liquidity

## Extracting default probabilities

- Once the model is estimated, one can derive implicit default probabilities
- In the spirit of Litterman and Iben (1991), various basic methodologies result in risk-neutral PDs (Chan-Lau, 2006)
- In our framework, we can compute **both the risk-neutral and the actual (or real-world) PDs**

⇒ Results: significant differences between actual and risk-neutral PDs



## Credit/Liquidity

### Extracting default probabilities

- The actual PD between time  $t$  and time  $t + h$  is:

$$\mathbb{P}_t \left( d_{t+h}^{(n)} = 1 \mid d_t^{(n)} = 0 \right),$$

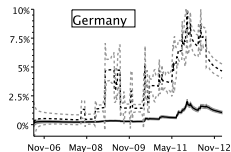
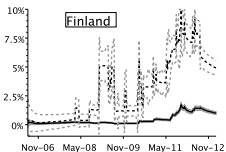
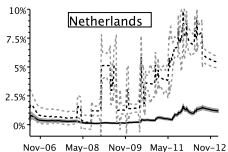
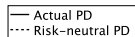
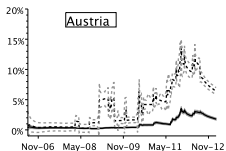
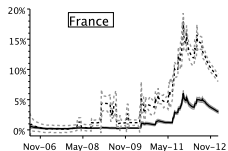
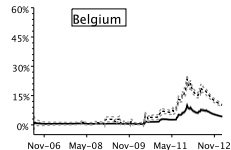
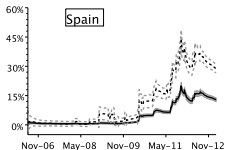
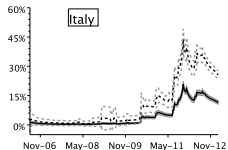
that is:

$$1 - E_t^{\mathbb{P}} \left( \exp \left\{ \frac{1}{1 - \zeta} \left( -\lambda_{fc,t+1}^{(n)} - \dots - \lambda_{fc,t+h}^{(n)} \right) \right\} \right)$$

- We use a recovery rate of 50%

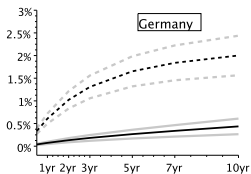
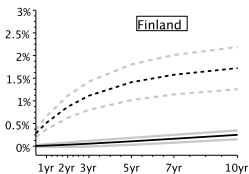
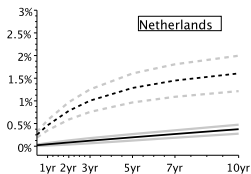
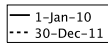
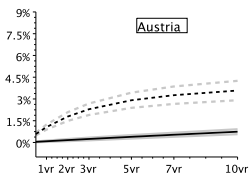
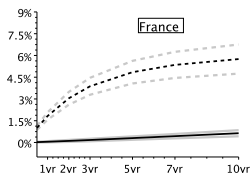
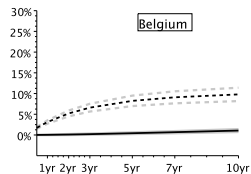
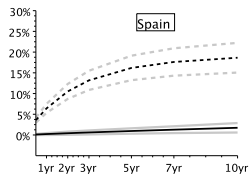
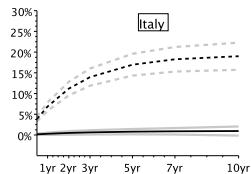
# Credit/Liquidity

## Extracting default probabilities (5-year PDs)



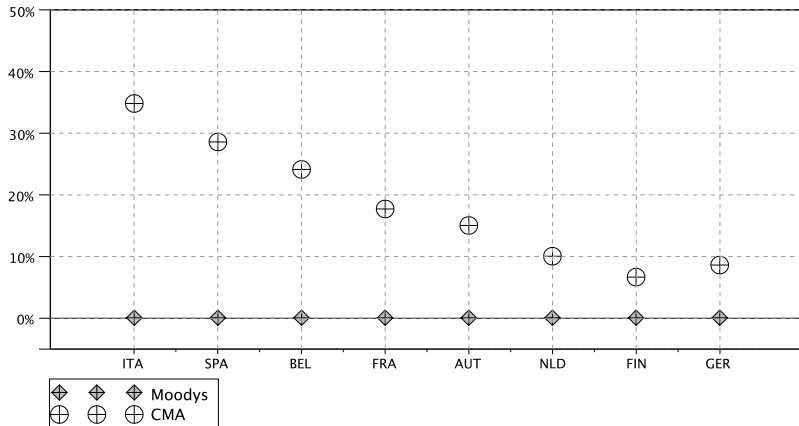
# Credit/Liquidity

## Extracting default probabilities (Term Structures of PDs)



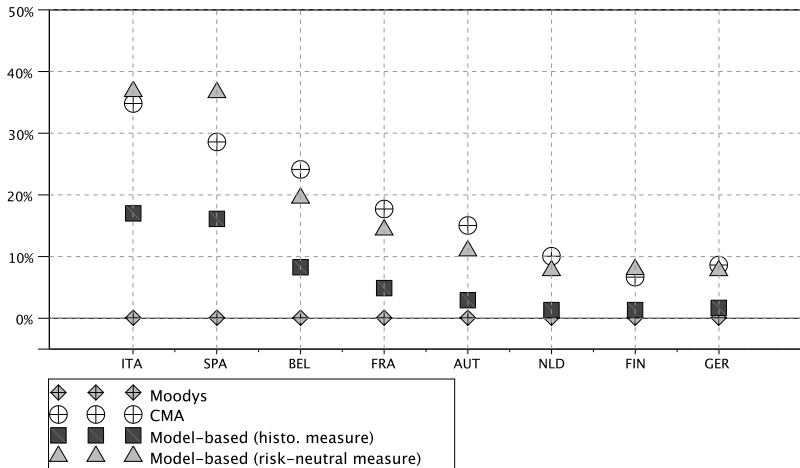
# Credit/Liquidity

## Extracting default probabilities (5-year PDs)



# Credit/Liquidity

## Extracting default probabilities (5-year PDs)



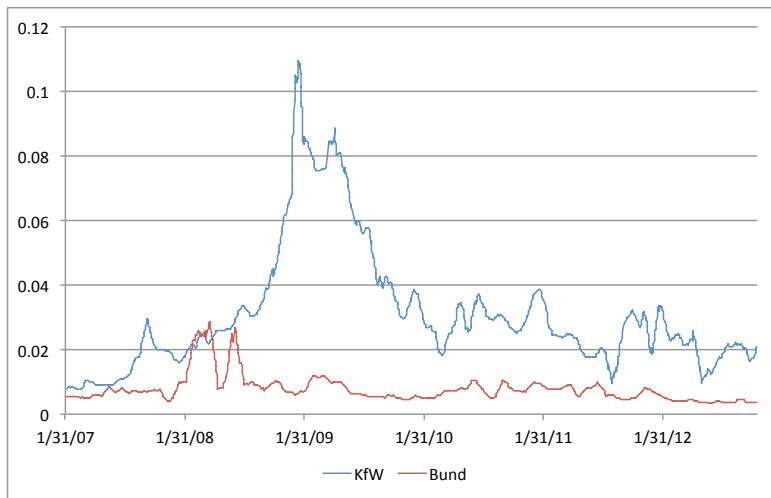


## Concluding remarks

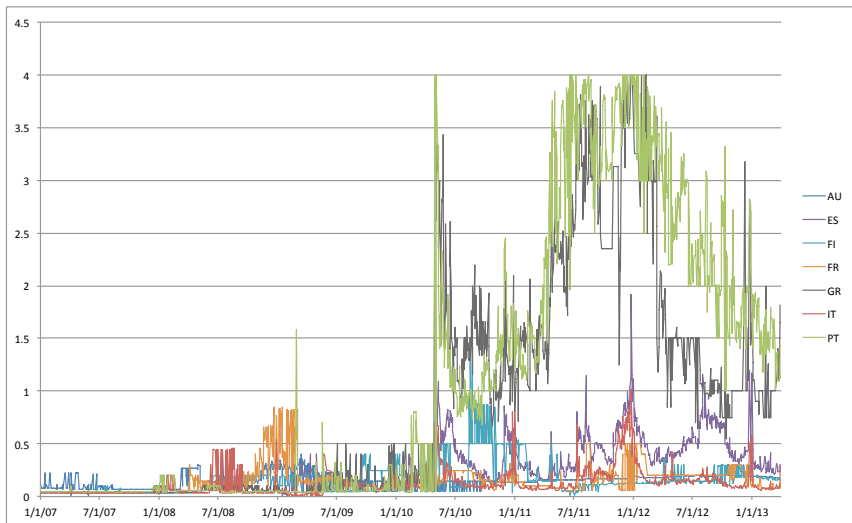
- One of the few attempts to model simultaneously sovereign EA spreads in a no-arbitrage framework
- Innovative use of regime-switching features to model interactions between credit and liquidity pricing effects
- Empirical results:
  - Liquidity effects account for a substantial part of EA-spreads fluctuations, but credit aspects have dominated over the last two years
  - The existence of risk premia results in significant differentials between risk-neutral and actual PDs

Thank you for your attention.

## Bid-Ask spreads: KfW vs Bund



# Bid-Ask spreads: Euro-area sovereigns



# A structural interpretation of the liquidity intensities

Ericsson and Renault (2006)

- The liquidity-related fractional-cost intensity:  $\lambda_{\ell,t}^{(n)} = (1 - \theta^{(n)})\lambda_{\ell,t}$
- Fractional cost:  $1 - \theta^{(n)}$ .
- $\ell_t = 1 \Rightarrow$  the investor has to exit by selling her bond holdings.
  - This liquidation has to be done in a limited period of time, between  $t$  and  $t + \varepsilon$ , say (where  $\varepsilon \ll 1$ ).
  - Random number  $K$  of offers from traders,  $K \sim \mathcal{P}(\gamma^{(n)}), \dots$
  - ... each offer is a random fraction  $\omega_i$  ( $i \in \{1, \dots, K\}$ ) of  $B_{t,h}^{(n)}$ ,  $\omega_i \sim \mathcal{U}([0, 1])$ .
- The selling price is:  $\max_{i \in \{1, \dots, K\}} (\omega_i) B_{t,h}^{(n)} = \theta(\gamma^{(n)}) \times B_{t,h}^{(n)}$ , where  $\theta$  is monotonically increasing and valued in  $[0, 1]$ .