Model

Joint and cond risk

SMP and EFSF

Conclusion

Conditional probabilities for euro area sovereign default risk

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"Dynamic score models" conference, Tinbergen Institute Amsterdam, 17-18 Jan 2013 email: bernd.schwaab@ecb.int Disclaimer: Not necessarily the views of ECB, Eurosystem, or Riksbank.



We propose a **novel modeling framework** to infer **conditional** and **joint probabilities** for sovereign default risk from observed CDS.

Novel framework? Based on a *dynamic GH skewed-t* multivariate density/copula with time-varying volatility and correlations.

Multivariate model is sufficiently flexible to be **calibrated daily** to credit market expectations. Not an "official opinion".

Analysis is based on **Euro area** CDS data, daily from 2008 to end-2012. **Event study**: SMP/EFSF announcement & initial impact on risk.



- Sovereign credit risk: e.g. Pan and Singleton (2008), Longstaff, Pan, Pedersen, and Singleton (2011), Ang and Longstaff (2011).
- Risk contagion, see e.g. Forbes and Rigobon (2002), Caporin, Pelizzon, Ravazzolo, Rigobon (2012).
- Observation-driven time-varying parameter models, see Creal, Koopman, and Lucas (2011, 2012), Zhang, Creal, Koopman, Lucas (2011), Creal, Schwaab, Koopman, Lucas (2011), Harvey (2012).
- 4. Non-Gaussian dependence/copula/credit modeling, see e.g. Demarta and McNeil (2005), Patton and Oh (2011).

Conclusion

Empirical questions

 $(\mathbf{Q1})$ Financial stability information: Based on credit market expectations, what is ...

Pr(two or more credit events in Euro area)? Pr(i|j)-Pr(i), for any i,j? Spillovers, e.g. Pr(PT|GR) - Pr(PT|not GR)? Corr_t(i,j) at time t?

(Q2) Model risk: For answering (a), how important are parametric assumptions? *Normal* vs *Student-t* vs *GH skewed-t*.

(Q3) Event study: did the May 09, 2010 Euro area rescue package change risk dependence? How?

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Copula framework

Sovereign defaults iff benefits (v_{it}) exceed a cost (c_{it}) , where

$$\mathbf{v}_{it} = (\boldsymbol{\varsigma}_t - \boldsymbol{\mu}_{\boldsymbol{\varsigma}}) \tilde{\boldsymbol{L}}_{it} \boldsymbol{\gamma} + \sqrt{\boldsymbol{\varsigma}_t} \tilde{\boldsymbol{L}}_{it} \boldsymbol{\epsilon}_t, \quad i = 1, ..., n,$$

 $\epsilon_t \sim N(0, I_n)$ is a vector of risk factors, \tilde{L}_{it} contains risk factor loadings, $\gamma \in \mathbb{R}^n$ determines skewness, $\varsigma_t \sim IG$ is an additional scalar risk factor for, say, *interconnectedness*.

A default occurs with probability p_{it} , where

$$p_{it} = \Pr[v_{it} > c_{it}] = 1 - F_i(c_{it}) \iff c_{it} = F_i^{-1}(1 - p_{it}),$$

where F_i is the CDF of v_{it} .

Focus on *conditional* probability $\Pr[v_{it} > c_{it} | v_{jt} > c_{jt}]$, $i \neq j$.

GH skewed-t dependence

$$y_t = \mu + L_t e_t, \quad t = 1, ..., T, e_t \sim \text{GHST}, \ \text{E}[e_t e_t'] = I_n,$$

$$p(y_t;\cdot) = \frac{v^{\frac{v}{2}}2^{1-\frac{v+n}{2}}}{\Gamma\left(\frac{v}{2}\right)\pi^{\frac{n}{2}}\left|\tilde{\Sigma}_t\right|^{\frac{1}{2}}} \cdot \frac{K_{\frac{v+n}{2}}\left(\sqrt{d(y_t)\cdot(\gamma'\gamma)}\right)e^{\gamma'\tilde{L}_t^{-1}(y_t-\tilde{\mu}_t)}}{(d(y_t)\cdot(\gamma'\gamma))^{-\frac{v+n}{4}}d(y_t)^{\frac{v+n}{2}}},$$

where

$$\begin{array}{lll} d(y_t) &=& v + (y_t - \tilde{\mu}_t)' \tilde{\Sigma}_t^{-1} (y_t - \tilde{\mu}_t), \\ \tilde{\mu}_t &=& -v / (v - 2) \ \tilde{L}_t \gamma, \\ \tilde{\Sigma}_t &=& \tilde{L}_t \tilde{L}_t' & \text{is scale matrix} \end{array}$$

If $\gamma = 0$, then GH skewed-t simplifies to Student's t density. If in addition $v^{-1} \rightarrow 0$, then multivariate Gaussian density. $\tilde{\Sigma}_t(f_t) = \tilde{L}_t(f_t)\tilde{L}_t(f_t)'$ is driven by 1st and 2nd derivative of the pdf. ◆□▶ ◆□▶ ◆臣▶ ◆臣▶ 臣 の�?

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Time varying parameters: score

Important: first two derivatives are available in closed form.

$$abla_t = \partial \ln p(y_t; \tilde{\Sigma}(f_t), \gamma, v) / \partial f_t$$

$$= \frac{\partial \operatorname{vech}(\Sigma_t)'}{\partial f_t} \frac{\partial \operatorname{vech}(L_t)'}{\partial \operatorname{vech}(\Sigma_t)} \frac{\partial \operatorname{vech}(\tilde{L}_t)'}{\partial \operatorname{vech}(L_t)} \frac{\partial \ln p_{GH}(y_t|f_t)}{\partial \operatorname{vech}(\tilde{L}_t)}$$

$$= \dots$$

$$= \Psi_t' H_t' \operatorname{vec} \left\{ w_t y_t y_t' - \tilde{\Sigma}_t - \left(1 - \frac{v}{v-2} w_t\right) \tilde{L}_t \gamma y_t' \right\}$$

where
$$\Psi_t = \partial \operatorname{vech}(\Sigma_t) / \partial f'_t$$

 $H_t = \operatorname{messy}$
 $w_t = \frac{v+n}{2 \cdot d(y_t)} - \frac{k'_{\frac{v+n}{2}}\left(\sqrt{d(y_t) \cdot (\gamma'\gamma)}\right)}{\sqrt{d(y_t)/\gamma'\gamma}}; k'_a(b) = \frac{\partial \ln K_a(b)}{\partial b}.$

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Time varying parameters: scaling

Scaling is by the inverse Fisher information matrix for a symmetric-t.

$$S_t = \left\{ \Psi'(I \otimes \tilde{L}_t^{-1})' \left[gG - \operatorname{vec}(I) \operatorname{vec}(I)' \right] (I \otimes \tilde{L}_t^{-1}) \Psi \right\}^{-1},$$

where

$$\begin{split} \Psi_t &= \partial \text{vech}(\Sigma_t) / \partial f'_t, \\ g &= (v+n)(v+2+n) \\ G &= \mathsf{E}[x_t x'_t \otimes x_t x'_t] \text{ for } x_t \sim \mathsf{N}(0, I_n) \end{split}$$

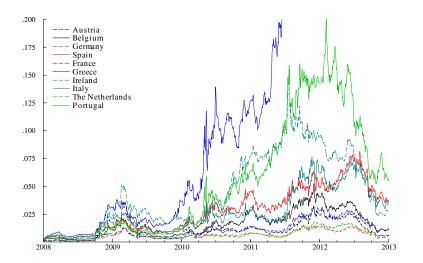
Extracting marginal pd's from CDS

CDS fee equates a premium leg and a default leg given a default intensity, see Duffie (1999), O'Kane and Turnbull (2003).

- We use a nonlinear solver to find the default intensity that matches E[PV premium leg] = E[PV default leg].
- Intensity and annual pd are nonlinearly related.
- Use 25% recovery rate and interest rate flat at 1%, ignore counterparty credit risk.
- Overall, not that difficult.

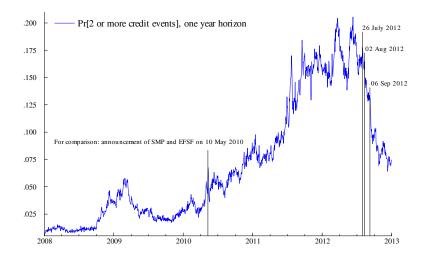
Conclusion

Marginal pd's from CDS



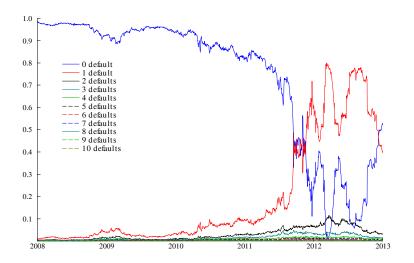
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Pr[2 or more credit events]

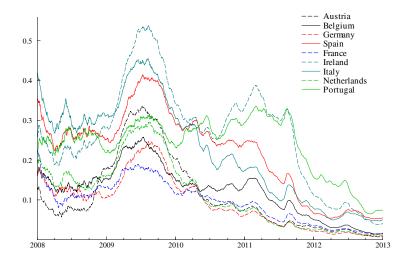


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The probability of k=0,1,2,... failures

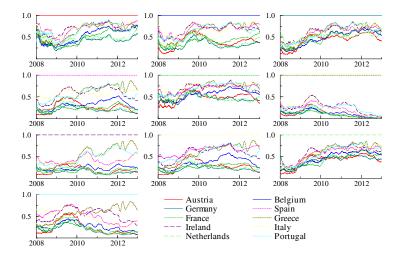


Conditional pds: Pr(country i|GR)



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Conditional pds: Pr(all i | all j)



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The May 09, 2010 package

Joint risk, Pr(i ∩ j)							
	Thu 06 May 2010			Tue 11 May 2010			
	PT	GR	DE	PT	GR	DE	
AT	1.1%	1.1%	0.6%	0.6%	0.7%	0.4%	
BE	1.2%	1.4%	0.7%	0.9%	1.0%	0.6%	
DE	1.0%	1.1%		0.8%	0.8%		
ES	3.0%	3.3%	0.9%	1.5%	1.6%	0.6%	
FR	1.0%	1.0%	0.6%	0.8%	0.9%	0.6%	
GR	4.8%		1.1%	2.3%		0.8%	
IR	2.6%	3.1%	0.8%	1.4%	1.8%	0.6%	
IT	2.8%	2.9%	0.9%	1.4%	1.5%	0.6%	
NL	0.9%	0.9%	0.5%	0.6%	0.7%	0.5%	
РТ		4.8%	1.0%		2.3%	0.8%	
Avg	2.0%	2.2%	0.8%	1.1%	1.2%	0.6%	

The May 09, 2010 package

Conditional risk, Pr(i j)								
	Thu 06 May 2010			Tue 11 May 2010				
	PT	GR	DE	PT	GR	DE		
AT	17%	8%	53%	22%	10%	46%		
BE	20%	10%	60%	32%	15%	61%		
DE	16%	8%		26%	12%			
ES	49%	25%	78%	50%	23%	63%		
FR	16%	8%	58%	28%	12%	62%		
GR	78%		99%	80%		86%		
IR	43%	23%	75%	49%	26%	68%		
IT	45%	22%	77%	49%	21%	64%		
NL	14%	7%	49%	21%	10%	50%		
РТ		36%	91%		33%	81%		
Avg	33%	16%	71%	40%	18%	64%		

Bottom line: joint risks $\downarrow \downarrow$, but dependence \uparrow . "Firewall"-analogy?

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We proposed a **novel modeling framework** to infer **conditional** and **joint probabilities** for sovereign default risk from observed CDS.

Based on a *dynamic skewed-t* multivariate density with time-varying volatility and correlations. Application to euro area CDS.

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Thank you						

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