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Sovereign bond and CDS market contagion: A story from the Eurozone crisis*

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Abstract

We examine the asymmetric and nonlinear nature of the cross- and intra-market linkages of eleven EMU sovereign bond and CDS markets during 2006-2018. By adopting the excess correlation concept of [Bekaert *et al.* \(2005\)](#) and the local Gaussian correlation approach of [Tjøstheim and Hufthammer \(2013\)](#), we find that contagion phenomena occurred during two major phases. The first, extends from late 2009 to mid 2011 and concerns the outright contagion transmission from EMU South bond markets towards all European CDS markets. The second, is during the revived fears of a Greek exit in November 2011 and is characterized by contagion from (i) CDS spreads in the EMU South towards bond yields in the same bloc and Belgium, and (ii) from Italian and Spanish CDS spreads towards all European CDS spreads. Consistent with their “too big to bail out” status, Italy and Spain emerge as pivotal for the evolution of sovereign credit risk across the Eurozone. Our examination of the relevant mechanisms, highlights the importance of credit risk over liquidity risk, and the containment effect of the naked CDS ban.

Keywords: sovereign bond market, sovereign CDS market, nonlinear dependence, contagion, local Gaussian correlation

JEL Classification: G01; G14; G15; C1; C58

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“Contagion occurs when cross-country correlations increase during crisis times relative to correlations during tranquil times”. — World Bank

1 Introduction

Contagion emerges in times of crisis and was a prominent feature of the European sovereign debt crisis. Recent evidence suggests that negative shocks were diffused in a different way across EMU (Economic and Monetary Union) member states (Claeys and Vařicek, 2014; Broto and Pérez-Quirós, 2015; Caporin *et al.*, 2018); countries with higher debt/deficits (Greece, Italy, Ireland, Portugal and Spain, henceforth GIIPS) were immediately affected, whereas the direct impact on the rest of the European economies has appeared less severe.¹ This issue has revived the discussion on the transmission of shocks and contagion in the euro area, which has led to ambiguous conclusions (see, e.g., Metiu, 2012; De Santis, 2012; Beirne and Fratzscher, 2013; Blatt *et al.*, 2015; Caporin *et al.*, 2018). This paper considers all relevant adverse financial and economic shocks stemming from the GIIPS during the Eurozone crisis and analyse their impact on the intra- and cross-market linkages of the European sovereign bonds and CDS markets. Having established that, we examine whether these linkages can generate contagion. If the answer to the latter is yes, we also ask whether this contagion is synchronous across different EMU countries.

Testing for contagion is not a straightforward exercise, as there is no broad consensus on what exactly constitutes contagion.² In this study, we adopt the approach of Bekaert *et al.* (2014) and use a factor model to examine unexpected factor exposures with respect to changes in fundamentals. Following Bekaert *et al.* (2005) we employ the residuals of the factor model to test for ‘residual contagion’, where contagion is defined as ‘correlation between markets in excess of that implied by economic fundamentals’. We complement this with a measure of local Gaussian correlation by Tjøstheim and Hufthammer (2013); this enables us to examine asymmetric and nonlinear changes in the dependence structure and test for contagion effects transmitted from the GIIPS to the entire Eurozone. We identify

¹An additional classification of Eurozone countries is that between the EMU South (Greece, Ireland, Italy, Portugal, Spain) and the EMU North (Austria, Belgium, Finland, France, Germany, the Netherlands). Instead of the terms “EMU South” and “EMU North” the terms “EMU periphery” and “EMU core” respectively are also common. Throughout this paper these terms are used interchangeably.

²See King and Wadhvani (1990); Bekaert and Hodrick (1992); Forbes and Rigobon (2002); Bekaert *et al.* (2005); reviews are provided by Pericoli and Sbracia (2003), Dungey *et al.* (2005) and Forbes (2012).

the dates where these effects occurred and provide a timeline of all the events that triggered financial contagion during the Eurozone crisis. We also reveal the direction of this contagion and the counterparties affected.

In specific, our dataset includes 5-year sovereign bond yields and CDS spreads covering the period from January 2006 to April 2018. In the first step, our analysis employs a factor model, where the bond and CDS data are conditioned on certain state variables (see [Bekaert *et al.*, 2005](#); [De Bruyckere *et al.*, 2013](#); [Bekaert *et al.*, 2014](#); [Fontana and Scheicher, 2016](#)). We employ these variables within a Kalman filter model and calculate the residuals of each of our bond and CDS series. By estimating time-varying coefficients from the entire sample, we ensure that any findings of contagion are not erroneously attributed to higher correlations due to volatility bias.

In the second step, we employ the method of [Tjøstheim and Hufthammer \(2013\)](#) to study the local Gaussian correlation dynamics of the residual bond and CDS series. We adopt the approach of [Demetrescu and Wied \(2019\)](#) to detect endogenous break-points in the correlation structure of our series and identify the dates of changes in their intra- and cross-asset (inter)dependence. The next step is to estimate the local Gaussian correlation for each country-pair around the break date and test for contagion effects via a bootstrap test (for an application see, [Støve *et al.*, 2014](#)). If the local correlation has increased significantly after the break date, this is evidence of “pure contagion”, i.e., contagion over and above what one would expect from economic fundamentals.

We subsequently replicate the analysis for our initial bond and CDS data without employing the first-stage factor model. By contrasting the estimates from the analysis of the residuals with those from the analysis of the original series, we can further distinguish between cases of “pure contagion” and cases of contagion due to changes in economic fundamentals (“wake-up call contagion”) or short-lived contagion episodes (“limited contagion”).

Our results indicate that contagion is neither a single-source nor a single-market phenomenon in the context of the European sovereign debt crisis. “Pure contagion” phenomena in the sovereign bond and CDS markets have undergone two major phases. The first phase extends from late 2009 to mid 2011 and concerns the outright transmission of contagion from the sovereign bond markets of the EMU South towards the CDS markets of both EMU blocs.

Most correlation break-points that indicate contagion are concentrated around important economic events. These include the Greek deficit's upward revision in November 2009 and the consequent fears about a possible Greek default, which led to the Greek government's official request of its first rescue plan in April 2010. Other events include the agreements on the rescue plans of Ireland and Portugal (November 2010 and April 2011 respectively) and the negative economic developments in Italy and Spain (July-August 2011).

On the other hand, the European sovereign bond market appears immune to extreme negative developments in the periphery countries' bond yields (a finding partly documented in [Missio and Watzka, 2011](#); [Metiu, 2012](#); [De Santis, 2012](#); [Beirne and Fratzscher, 2013](#); [Claeys and Vařicek, 2014](#); [Caporin *et al.*, 2018](#)). There is opposite responsiveness between the European sovereign bond and CDS markets with regards to the periphery bond-stemming contagion. This reveals an additional difference between the two markets during that period: the capacity of the bond market to reduce shocks and/or losses. This is in contrast to the CDS market where such shocks – reflected by the higher yields in the periphery bonds – are preserved and even amplified. The analysis of the original series reveals that European sovereign bond yields were subject to a “wake-up call” during the late 2008–early 2009 period. This “wake-up call” prompted investors to reassess the vulnerability of Eurozone countries, thereby leading to a repricing of the factors that affect sovereign credit risk ([Bekaert *et al.*, 2014](#); [Claeys and Vařicek, 2014](#)).

The second phase is around November 2011 and Greece's plan to hold a referendum on Eurozone membership. This prolonged political uncertainty has revived the fears of the country's Eurozone exit. During this period, CDS-bond contagion is mainly a “periphery” phenomenon: contagion stems from the CDS markets of the EMU South and is directed to the bond markets of the same bloc and Belgium. When intra-CDS contagion is concerned, we find evidence of outright contagion from the Italian and Spanish CDS markets towards the CDS markets of both blocs. The outbreak of CDS-stemming contagion during the respective period marks the development of the Greek debt crisis into a European debt crisis. In this regard, Italy and Spain appear to be the key countries for the evolution of euro area sovereign credit risk. This finding is consistent with their “too big to bail out” status and the fears that a Greek default would cause a domino effect across the Eurozone. The exposure of

euro area CDS spreads to Italian and Spanish CDS spreads stands in contrast to previous findings within the intra-CDS context where contagion was found to be either non-existent (see Caporin *et al.*, 2018) or only a Eurozone periphery phenomenon (see Broto and Pérez-Quirós, 2015) and also to arguments about the limited capacity of the periphery CDS markets to generate contagion (see Kalbaska and Gatkowski, 2012).

We complement the above studies by offering a timeline of the magnitude and direction of contagion within the European sovereign debt and credit markets. Contagion is decomposed into its “pure” and “wake-up call” components. We show that the two Eurozone blocs were following a divergent path; the fiscal shocks across the EMU South were the driving cause of such divergence. On the contrary, the notion that contagion phenomena in the Eurozone were a consequence of the US financial meltdown and the resulting global financial crisis is rather weak. The occurrence of the vast majority of “pure contagion” episodes in the late 2009 – late 2011 period confirm these arguments.

Importantly, we identify the potential mechanisms for the emergence of contagion. As such, we focus on liquidity risk, which was an important determinant of European sovereign bond yields during the Eurozone crisis (see Beber *et al.*, 2009; Brunnermeier and Pedersen, 2009; Monfort and Renne, 2014). By distinguishing between contagion transmission due to credit risk considerations and due to liquidity risk considerations, we show that liquidity risk is an important source of risk in the European CDS market, but only when stemming from the bond market. We thus content that credit risk is a stronger determinant of price discovery and contagion transmission between the two markets relative to liquidity risk.

On the same line, we examine the role of arbitrage opportunities. Our findings reveal that although arbitrage forces may be present (see Fontana and Scheicher, 2016; Gyntelberg *et al.*, 2018), they were not able to fully close the pricing gaps between the two markets and affect contagion dynamics. We thus demonstrate that the different regimes and adjustment speeds which characterize the correcting mechanisms between the sovereign bond and CDS markets (also evident in Gyntelberg *et al.*, 2018), further restrict their contagion capacity.

Moreover, our study is the first to link the important policy changes on the regulatory and monetary policy fronts during the Eurozone crisis with sovereign contagion dynamics. We provide evidence that the voting of the naked CDS ban exerted an easing effect on the

transmission of CDS-bond contagion within the EMU South. We further consider the ECB's large-scale monetary policy interventions (in the form of government bond purchases and one-off liquidity injections) and find no sign of bond contagion either before or after their conduct; if anything, these interventions aimed at easing periphery countries' borrowing costs and restoring monetary policy transmission, rather than targeting contagion.

Finally, our paper relates to the literature on tail market events and financial contagion. For this purpose, several estimation and testing strategies have been applied, including the analysis of the coincidence of tail returns (Bae *et al.*, 2003), the multivariate distribution modeling of tail returns (Longin and Solnik, 2001), the estimation of the expected market crashes after a crash in one market (Hartmann *et al.*, 2004) or the exceedance correlation concept (Ang and Bekaert, 2002; Ang and Chen, 2002; Hong *et al.*, 2007). The latter refers to the concept of conditional correlation, where the ordinary product-moment correlation is computed for certain regions of the distribution. However, the conditional correlation in a local region is not equal to the global correlation for a pair of jointly Gaussian variables.³ Moreover, this approach produces a measure of linear dependence locally, which is questionable in a nonlinear and non-Gaussian framework.

In this paper we use the local Gaussian correlation concept to assess the nonlinear dependence behavior of our series. The local Gaussian correlation does not suffer from the bias problem of the conditional correlation (e.g., as in the exceedance correlation concept), while the latter can be considered as a special case of the former. More closely related to our approach are the recent proposals for measuring and testing nonparametric dependence based on auto-distance correlation functions (Székely and Rizzo, 2009; Zhou, 2012), copula models (Oh and Patton, 2017) and the cross-quantilogram of Han *et al.* (2016). However, these test dependence under the assumption of independent and identically distributed series or estimate time series serial dependence. Little has been done in the perhaps most important situation of estimating the strength of cross-dependence, and testing for contagion between two general stationary time series. In this respect, the local Gaussian correlation is easier to interpret than auto-distance and copulas, providing a direct measure of both average and upper-lower tail dependence and a complete characterization of dependence structure.

³This is known as the bias problem for conditional correlation (see Forbes and Rigobon, 2002) and refers to the fact that in a Gaussian distribution, dependence is completely characterized by the correlation coefficient.

The remaining of the paper is organised as follows: Sections 2 and 3 present the empirical methodologies and Section 4 describes the dataset. Section 5 examines contagion transmission from the sovereign bond and CDS markets of the EMU South, while Section 6 identifies the mechanisms and assesses the sensitivity of our findings. Section 7 concludes.

2 Methodology

The concept of contagion is not unequivocal given the number of alternative methodological approaches to analyze it. In this paper, we follow [Bekaert *et al.* \(2014\)](#); we take an asset pricing perspective on measuring economic fundamentals and identify contagion through the correlation of an asset pricing model’s residuals. These residuals are obtained from a regression in which the dependent variable is the daily change in the bond yield (CDS spread) of a given country and the explanatory variables are six factors related to the bond and CDS markets. In this respect, our six-factor model encompasses two EMU-, two global financial- and two US-specific factors. Given this, we set a benchmark for what sovereign bond and CDS market comovements should be, based on these fundamentals.

We use Kalman filtering to estimate time-varying coefficients for our factor model. We calculate the standardized signal errors from the Kalman filter (the residuals from the factor model) and test for increasing correlations between the residual bond/CDS series by using the local correlation approach. We endogenously detect the break dates in correlation after considering changes in the variance of the residual series following the method of [Demetrescu and Wied \(2019\)](#). Subsequently, we test for contagion via a bootstrap test for each intra-market pair (i.e., bond-bond pair or CDS-CDS pair) and cross-market pair (i.e., bond-CDS or CDS-bond pair) that we observe an increase in local Gaussian correlation after the correlation break date. Effectively, our approach is in conceptual proximity with the “excess correlation” concept of [Bekaert *et al.* \(2005\)](#), who define contagion as “correlation over and above what one would expect from economic fundamentals”. Finally, we perform the correlation break test and the bootstrap test for each possible pair from our original series (i.e., without accounting for fundamentals through our factor model). Ultimately, this leads to three different types of contagion:

- an increase in the exposure of bond yields and/or CDS spreads to common factors,

which is labelled as “wake-up call contagion”. Practically, this occurs when there is evidence of increased local correlation on the original series. The hypothesis of contagion between a given country-pair is verified for the original series but rejected for the residual series.

- an increase in the exposure of bond yields and/or CDS spreads to common factors, which is further accompanied by an increase in the correlation across and/or between unexplained bond yields and CDS spreads. This is labelled as “pure contagion” and occurs when there is evidence of increased local correlation on both the original and the residual series. Accordingly, the hypothesis of contagion is confirmed for both the original and the residual series. Generally, such cases refer to potent and persistent shocks, whose impact is not limited to economic fundamentals but goes beyond them.

- an increase in the correlation across and/or between unexplained bond yields and CDS spreads without a corresponding increase in the exposure of bond yields and/or CDS spreads to common factors. This is labelled as “limited contagion” and occurs when there is evidence of increased local correlation on the residual series but not on the original ones. The hypothesis of contagion for a given pair is confirmed by the estimation of the residual series but rejected by that of the original series. This form of contagion generally refers to short-lived shocks that quickly recede: due to their magnitude they are reflected on the unexplained series, however their temporary nature is not reflected on economic fundamentals.

2.1 The factor model

We employ a six-factor model that controls for developments at the EMU-level, global risk aversion and conditions in the US. Our first EMU-specific factor concerns the idiosyncratic equity returns in each country (*EqR*), to control for market-wide changes in business climate. This is defined as the difference between the country’s equity-index return and the benchmark Eurozone index return (STOXX Europe 600). We expect equity returns in each country to be negatively related to that country’s credit spreads. We further control for general developments in the sovereign bond markets by using the returns on a synthetic Euro benchmark bond in the regression for the country’s bond yields (*SynBond*). Similarly, in the regression for the country’s CDS spread, we include a sovereign industry-specific index constructed for European countries based on composite CDS data provided by Thomson Reuters (*IndCDS*).

This follows [Antón *et al.* \(2018\)](#) and enables us to disentangle normal developments in the bond and CDS markets (as opposed to abnormal, that are the focus of our study).

Global risk factors constitute an additional driver of sovereign credit spreads, as higher volatility is associated with higher economic uncertainty ([Longstaff *et al.*, 2011](#)). We control for systemic risk with the 3-month Euribor-OIS spread (*3mEUR*). We employ this variable (in logarithmic form) as it is considered a reliable measure of systemic risk (see, e.g., [Rodríguez *et al.*, 2013](#)). Since systemic credit risk has become a considerably large fraction of total credit risk (for evidence from the subprime crisis, see [Bhansali *et al.*, 2008](#)), we expect that an increase in systemic risk is reflected in growing bond yields and CDS spreads. We further control for global uncertainty with the VIX index (*VIX*). We employ this variable (in logarithmic form) as it exhibits a strong relationship with sovereign credit risk ([Pan and Singleton, 2008](#)). The fact that the VIX, is also highly correlated with spreads on sovereign entities suggests that VIX is a proxy for global “event risk” in credit markets. Given that, we expect a positive relationship between the VIX and sovereign CDS spreads.

Turning to US-specific factors, we control for exchange rate uncertainty, by using the 30-day implied EURO/USD exchange rate volatility index (*EVZ*).⁴ In the event of higher uncertainty about the future path of the bilateral exchange rate, USD-quoted protection should be more expensive than the equivalent euro-quoted protection. This is due to the currency hedge provided by the USD-quoted protection against a potential sovereign default and a consequent euro depreciation; this indicates a positive correlation between bilateral exchange rate volatility and CDS spreads ([Fontana and Scheicher, 2016](#)). Finally, we control for overall economic conditions in the U.S. by employing the S&P500 index (*SP500*).

The regression specification of the factor model takes the following form:

$$\begin{aligned} \Delta Y_{it} = & \beta_{0,t} + \Delta\beta_{1,t}EqR_{it} + \Delta\beta_{2,t}Dev_{it} + \Delta\beta_{3,t}3mEur_t + \Delta\beta_{4,t}VIX_t + \Delta\beta_{5,t}EVZ_t \\ & + \Delta\beta_{6,t}SP500_t + e_{it} \end{aligned} \quad (1)$$

where ΔY_{it} is a vector representing the change in bond yield or in the CDS spread of a country i at time t , ΔEqR_{it} is the change in idiosyncratic equity returns, ΔDev_{it} denotes the changes in the bond (*SynBond*) or CDS’s (*IndCDS*) market developments, $\Delta 3mEur_t$

⁴The index is provided by the CBOE and follows the methodology of the VIX index.

is the change in the 3-month Euribor-OIS spread, ΔVIX_t is the change in the VIX index, ΔEVZ_t is the change in the Euro/USD exchange rate volatility index and $\Delta SP500_t$ is the change in the S&P500 stock market index.⁵

2.2 Evolution and stability of model factors

However, our selected factors may have a varying explanatory power for our bond and CDS series; this is partly attributed to the fact that the key drivers of sovereign credit risk have affected CDS premia and bond spreads differently during the global financial crisis (Groba *et al.*, 2013; Fontana and Scheicher, 2016). In fact, as volatilities tend to dramatically increase during crises, increased correlations are not necessarily indicative of contagion, a concept introduced by Forbes and Rigobon (2002). Hence, not controlling for the possibility that our bond and CDS series exhibit a time-varying exposure to fundamentals (as in Bekaert *et al.*, 2014; Bekaert and Hoerova, 2016), might lead to erroneous findings of contagion.⁶

To this end, we employ the Kalman filter and estimate time-varying coefficients for our set of bond and CDS factors. This approach enables us to determine the optimal degree of coefficient persistence and smoothing, thereby capturing any variation in the exposure of our bond and CDS series to our factors. The Kalman filter model with time varying-coefficients is specified as:

$$s_{t+1} = l_t + T s_t + R_t \eta_t \quad (2)$$

$$y_t = \alpha_t + Z_t s_t + \epsilon_t \quad (3)$$

where Equation (2) is the transition or state equation and Equation (3) is the measurement equation, $\eta_t \sim N(0, Q_t)$, $\epsilon_t \sim N(0, H_t)$ are assumed to be independent and Q_t , H_t are positive definite. In the empirical application of linear regression, regression coefficients are presented by the state vector. The state vector of the Kalman filter is expressed as:

⁵A battery of unit root tests for credit spreads and for the state variables expressed in levels, has not rejected the unit root hypothesis. Therefore we estimate this equation in changes only. Results are omitted for reasons of space but are available from the authors upon request.

⁶Bekaert *et al.* (2014) analyze the transmission of the 2007-09 crisis to 415 country-industry equity portfolios in 55 countries. They employ a three-factor model, to distinguish between a US-specific factor, a global financial factor and a domestic factor respectively. Bekaert and Hoerova (2016) employ techniques of time-varying risk aversion and uncertainty for Germany and the US from January 1992 to March 2008. They find that the variance premium contains a substantial amount of information regarding risk aversion in both countries, while the credit spread primarily contains information about economic uncertainty.

$$s_t = \begin{bmatrix} \beta_{0,t} \\ \beta_{1,t} \\ \beta_{2,t} \\ \beta_{3,t} \\ \beta_{4,t} \\ \beta_{5,t} \\ \beta_{6,t} \end{bmatrix}$$

and the states (the regression coefficients) evolve according to random walk:

$$\beta_{i,t+1} = \beta_{i,t} + \eta_{i,t}, \quad \eta_{i,t} \approx N(0, \sigma_{\eta_i}^2), \quad i \in [0, 6].$$

We use the Kalman filter for each country's bond yield vis-à-vis each of the different bond market factors and present results in Figure 5 of the Appendix. Similarly, in Appendix Figure 6 we present results from the Kalman filter for each country's CDS spread and each of the relevant CDS market factors. Turning to Figure 5, we observe that the assumption of parameter constancy holds for almost all countries. According to our estimates, the sign and the magnitude of the correlations between bond yields and the six factors in each country remain stable over our sample period. This is further evident for the CDS series in Figure 6.

The analysis from the Kalman filter further points to a substantial asymmetry in the importance of each of the bond and CDS market fundamentals in our factor model. Moreover, we observe that the pricing of these factors is different between the bond and the CDS markets, which further motivates our examination of cross-asset contagion in Section 5; for example, an increase in the level of systematic risk is accompanied by an increase in the Greek CDS spreads but not in the Greek bond yields.⁷ Overall, findings from this exercise indicate that there are only small time variations in the exposures of our bond and CDS series to our six model factors (the β 's in our factor model).

Having tested the stability of our factor model's coefficients, we subsequently obtain the standardized signal errors through Kalman filtering (i.e., the residuals from the factor model). To identify increasing correlations between the residual bond/CDS series, we use a local correlation approach and apply a bootstrap test for contagion; we discuss these methods in Section 3 and provide more technical details in Sections A.1 and A.2 of the Appendix.

⁷Additionally, we would expect that when risk aversion is high, spreads are relatively high. This is not confirmed by our time-varying estimations for none of the two markets. This is in line with the arguments about limited pricing of risk in the run-up to the global financial crisis (see D'Agostino and Ehrmann, 2014).

3 Contagion identification scheme

Tjøstheim and Hufthammer (2013) introduce a new measure of nonlinear dependence inherent to the concept of local correlation (cf. also Teräsvirta *et al.*, 2010). The central idea of the new approach is to approximate an arbitrary bivariate return distribution by a family of Gaussian bivariate distributions. At each point of the return distribution there is a Gaussian distribution that approximates that point (approximate the density locally rather than the correlation). The correlation of the approximating Gaussian distribution is taken as the local correlation in that neighbourhood.⁸

3.1 A bootstrap test for contagion

This section presents a test for contagion that uses the measure of local Gaussian correlation to examine whether cross-market linkages have increased (for more discussion and an application in financial markets see, Støve *et al.*, 2014). Contagion is confirmed if the local correlation for the crisis period has increased significantly compared to that before the crisis. The test was proposed by Støve *et al.* (2014) and is a bootstrap procedure.⁹

Denote $Z_t, t = 1, \dots, T$ as the sovereign bond yields in country where the crisis started and $X_t, t = 1, \dots, T$ the bond yields in another country. Let the yields be written as $d_t = (X_t, Z_t)$. We then split the data in a pre-crisis period (NC) and a post-crisis period (C). If the local correlation for the post-crisis period is significantly above the pre-crisis one, contagion is confirmed. Fixed gridpoints (x_i, z_i) for $i = 1, \dots, n$ are used to estimate the local correlations. Thus, the null and the alternative hypothesis can be written as:

$$H_0 : \rho_{NC}(x_i, z_i) = \rho_C(x_i, z_i) \text{ for } i = 1, \dots, n \text{ (no contagion)}$$

$$H_1 : \sum_{i=1}^n (\rho_C(x_i, z_i) - \rho_{NC}(x_i, z_i)) > 0 \text{ (contagion)}$$

The bootstrap works by drawing observations $\{d_1, \dots, d_T\}$ at random and replacing them in $\{d_1^*, \dots, d_T^*\}$. Next, this resample is divided in NC and C and $\hat{\rho}_{NC}^*(x_i, z_i)$ and $\hat{\rho}_C^*(x_i, z_i)$ is computed on the grid (x_i, z_i) for $i = 1, \dots, n$. The diagonal grid $(x_i = z_i)$ is employed in the subsequent analysis to minimize the computational time. The next step is to calculate:

⁸For brevity, a detailed description of the local Gaussian correlation procedure is provided in Appendix A.2.

⁹Similar ones are often used in a nonparametric setting, e.g., to for differences between quantities in nonparametric regressions (see, for example, Hall and Hart, 1990; Vilar-Fernandez *et al.*, 2007).

$$D_1^* = \frac{1}{n} \sum_{i=1}^n [\hat{\rho}_C^*(x_i, x_i) - \hat{\rho}_{NC}^*(x_i, x_i)] w_i(x_i, x_i),$$

where w_i denotes a weight function that allows to concentrate on a certain region. The weight function is chosen to minimize the distance between the gridpoints and the observations. In other words, we avoid the estimation of local correlation in a gridpoint far away from any observations. Repeated resampling allows us to compute D_1^* for these resamples and to construct its distribution. Last, we calculate $\hat{\rho}_{NC}(x_i, x_i)$, $\hat{\rho}_C(x_i, x_i)$ and D_1 from the real filtered observations $\{d_1, \dots, d_T\}$. The p -value in terms of the D_1^* distribution is found and implies a rejection of H_0 if it is below a chosen significant level α .¹⁰

4 Data

4.1 Bond yields and CDS spreads

The dataset consists of bond yields and CDS spreads for sovereign bonds and CDS contracts with 5-year maturity from January 2, 2006, to April 5, 2018.¹¹ The sample includes all EMU member states at the time of the euro's introduction, i.e., Austria (AT), Belgium (BE), Finland (FI), France (FR), Greece (GR), Germany (DE), Ireland (IE), Italy (IT), the Netherlands (NL), Portugal (PT), and Spain (ES). The 5-year tenor constitutes the most liquid and frequently quoted part of the credit curve and thus, the most traded maturity for CDS contracts. Daily bond yields are from Thomson Reuters Datastream and daily CDS spreads are from Markit. We further convert the change in bond yields into basis points (bps), due to the CDS spread changes (calculated as: $\Delta Y_{it} = Y_{it} - Y_{it-1}$) being already in basis points. We employ daily frequency given that comovements in the bond and CDS markets are not constant as investors shift their assets; in addition arbitrage opportunities, which represent a significant driver (and consequence) of these comovements are likely to be diminished at lower frequencies. In sensitivity analysis, we further use the bid-ask spread (calculated as: *ask price - bid price*) on the sovereign bonds and CDSs with 5-year maturity and the CDS-bond basis for the same maturity from January 2 2006, to December 31, 2014.

¹⁰The authors would like to thank Dag Bjarne Tjøstheim and Bård Støve for providing the R codes for the contagion bootstrap test.

¹¹For Greece the sample ranges from January 2, 2006 to March 8, 2012 since the next day, on March 9, 2012, after the agreement on sovereign-debt restructuring, the Greek bonds stopped trading. Data for Luxembourg is available from February 2009 onwards, except for the period from January 2010 to February 2011, and therefore is not included in the sample. The two-letter country codes that are used in abbreviation in our analysis are taken from the International Standard for country codes (ISO 3166).

4.2 Descriptive statistics

Figure 1 (Panels A and C), graphs the EMU periphery bond yields, which have been soaring since late-2009 and after mid-2010 when Greece reached an agreement with the European Union (EU) and the International Monetary Fund (IMF) for a €110 billion financing package to recover from its debt crisis. Bond yields in the EMU core follow the opposite course; they have been falling during the post-2008 period, with only a slight rise in the first half of 2011, appearing to have benefited from the skyrocketing borrowing costs of the periphery countries. As shown in Figure 1 (Panels B and D), CDS spreads in the periphery match closely the upward trend of the periphery bond yields, pointing to a close association between the two asset-markets. However, CDS spreads also increased in the core, indicating an overall rise in sovereign default probabilities across the entire Eurozone.

[Insert Figure 1 about here.]

Table 1, presents descriptive statistics for daily changes in bond yields and CDS spreads. The average bond yield change (standard deviation), expressed in basis points, is -0.121 (4.54) for the core and 0.395 (17.87) for the periphery. Similarly, the mean CDS spread change (standard deviation), is 0.005 (8.15) and 3.086 (137.2) for the core and periphery respectively. Overall, GIIPS bond yields and CDS spreads are higher and more volatile than their core counterparts. Table 2 presents the response of bond yields and CDS spreads in each country to a number of negative events in the EMU South. We observe a sharp rise in the bond yields and CDS spreads of the source country (in bold), which exceeds the equivalent rise in the remaining countries. Importantly, this is accompanied by a rise in the cross-asset correlations between the source country's bond yields and CDS spreads after each event: the degree of comovement in this period ranges between 0.50 and 0.75, indicating that sovereign bond and CDS markets do not respond uniformly to the same shocks.

[Insert Tables 1 and 2 about here.]

5 Empirical results

We proceed to the examination of the degree of pair-wise conditional correlations between and across the European sovereign bond and CDS markets considering as source of conta-

gion each of the European periphery countries, namely Greece, Ireland, Italy, Portugal and Spain. We detect endogenous break dates on which contagion transmission is initiated by using the algorithm for correlation change-point inference of [Demetrescu and Wied \(2019\)](#).¹² This is a test for constant correlations that allows for breaks at unknown times in the marginal moments (means and variances). This also enables us to locate and identify a change in the correlation between the bond and CDS markets and thus determine the exact date associated with a fundamental change in the relationship between them. The identification of a structural change in the cross- and intra-asset correlations further allows us to split the sample into a pre- and post-event period, or more properly into a pre- and post-contagion period. We then quantify the impact of the structural change by estimating the transmission of a shock to the bond and CDS markets during the respective periods.

In line with our discussion in Section 2, we conduct the analysis by employing the bond and CDS series after accounting for fundamentals via the factor model of Equation (1); we present results from this exercise in Tables 3 and 4. We then conduct the same analysis for the original bond and CDS series and present results in Appendix A.4 (Tables 8 and 9).¹³ By contrasting the results from the two methods, we can differentiate between cases of “pure contagion” and cases of “wake-up call contagion” and “limited contagion”. In this respect, if the hypothesis of contagion between a given pair is verified by the estimation of the original series but rejected by the estimation of the residual series, this would serve as evidence of “wake-up call contagion”. In contrast, findings of contagion under both the original and residual series should be interpreted as evidence of “pure contagion”. Last, if contagion is only verified by the residual series, this points to a short-lived episode of “limited contagion”.

5.1 Transmission of contagion from the GIIPS bond markets to the EMU bond and CDS markets

5.1.1 Intra-bond contagion

Table 3, presents the results on the transmission of contagion from the bonds of the EMU South towards the bonds (Panel A) and CDSs (Panel B) in each of the eleven EMU member

¹²The [Demetrescu and Wied \(2019\)](#) methodology for the identification of structural changes in correlation is described in detail in Appendix A.1.

¹³As in [Forbes and Rigobon \(2002\)](#), we have performed heteroscedasticity-filtering in the original series. If heteroscedasticity bias is ignored when testing for changes in correlation, then contagion is over-accepted.

states. When the analysis is concentrated solely within the bond market, estimates in Panel A reveal that bonds yields across the Eurozone have been immune to contagion phenomena during the pre- as well as post-financial crisis period: the bootstrap test for contagion is not able to reject the null hypothesis of no-contagion in practically all pairs (columns 1 to 5 in Panel A). This absence of contagion stemming from the distressed periphery countries (also partly documented in [Missio and Watzka, 2011](#); [Metiu, 2012](#); [De Santis, 2012](#); [Beirne and Fratzscher, 2013](#); [Claeys and Vařicek, 2014](#); [Caporin *et al.*, 2018](#)) is not necessarily a sign of widespread disassociation between the sovereign debt markets.¹⁴ It can serve as (i) supporting evidence of the investors' flight(s) away from the risky periphery bonds to the safer core bonds and therefore of a negative correlation between bond yield movements in the two blocs (see [Beber *et al.*, 2009](#)) and (ii) an indication that the common shifts of periphery bond yields can be explained in terms of the – enduring– interdependence between them.

[Insert Table 3 about here.]

5.1.2 Bond-CDS contagion

This disassociation is not confirmed for the bond-CDS nexus: results in Panel B of Table 3, point to transmission of bond-originated contagion from all countries of the EMU South to nearly all European countries. These results are derived from the analysis of the residual series, thus indicating contagion over and above what can be explained by fundamentals (“pure contagion”). Interestingly, this transmission occurs over a series of different phases, all coinciding with major economic and political events during the late 2009–mid 2011 period.

The first phase is during November 2009, shortly after the upward revision of the Greek government's budget deficit: estimates in the first column of Panel B, indicate the transmission of contagion from the Greek bond to each of the European CDS markets.¹⁵ The Greek-bond stemming contagion does not appear to be a one-off phenomenon as it further emerges in April 2010. The respective month is characterized by the growing fears about a

¹⁴The aforementioned studies find no evidence of shift contagion across the European sovereign bond markets when the role of contagion originator is primarily assumed to be Greece.

¹⁵The Greek budget deficit was initially revised upward from 6.0% of GDP to 12.5% on October 19, 2009 by the new Greek Minister of Finance in his first Eurogroup meeting. In the budget draft for fiscal year 2010 submitted to the Hellenic Parliament for consideration in the November 5, 2009, the 2009 deficit was revised to 12.7%, while in the final draft of November 16, 2009 (voted by the Parliament), it was revised to 13.6%.

possible Greek default, which ultimately led to the Greek government's official request for the activation of the joint EU/IMF aid package.¹⁶

The next phase occurs in November 2010 and includes the transmission of contagion from the Irish bond yields to almost all European CDS spreads (second column of Panel B). This period marks the re-eruption of the Irish crisis (back from the 2008 banking crisis and the €64 billion bailout of Irish banks) and resulted in a joint EU/IMF financial assistance programme. The Irish government's request for the programme was made on November 21, 2010, while on November 24, 2010 the government outlined €15 billion in spending cuts and tax increases to reduce its budget deficit from 31% of GDP to 3% by 2014. These were followed by massive rallies and protests in Dublin three days later, before the €67.5 billion bailout package being accepted on November 28, 2010.

The fourth and more intense phase of contagion occurred during the second and third quarter of 2011. Starting from April 2011, the Portuguese government requested financial assistance from the EU, which in early May was agreed to be provided jointly by the EU and the IMF.¹⁷ According to estimates in the fourth column of Panel B, the period shortly before the Portuguese request (late March 2011) until the final bailout agreement (early May 2011) was characterized by the transmission of contagion from Portuguese bond yields to nearly all EMU CDS spreads. Contagion phenomena were even more prevalent during the third quarter of 2011. Results in the third and fifth column of Panel B, point to contagion from the bond markets of Italy and Spain respectively towards each of the EMU CDS markets. Most of the correlation break-dates that indicate contagion are observed between mid July and early August of 2011, a period of significant turmoil in the Italian and Spanish economies.

In specific, on July 14, 2011, Italy raised €3 billion from selling government bonds, albeit at a record interest rate of 5.9%. One day later, the European Banking Authority (EBA) announced that five Spanish banks failed its "stress tests", while seven other Spanish banks barely passed.¹⁸ During the same period, talks abounded that Greece would become the first

¹⁶On April 23, 2010, Greek Prime Minister George Papandreou made a live broadcast announcement from the Greek island of Kastelorizo on the request of the €60 billion EU/IMF financial aid programme.

¹⁷On April 6, 2011, Portuguese Prime Minister José Sócrates extended a request to the EU for a financial assistance programme. On May 6, 2011, Portugal reached an agreement for a €78 billion EU/IMF programme.

¹⁸The banks that failed the EBA's stress tests were Catalunya Caixa, Caja de Ahorros de Mediterraneo, Banco Pastor, Unnim, and Group Caja3. Seven banks, i.e., Banco Sabadell, Banco Popular, and Bankinter and the savings banks Novacaixagalicia, Caja Ontinyent, Banca Civica, and Bankia, just achieved the minimum requirement of core equity Tier 1 (CET1) capital ratio of 5.0%.

country to be forced to exit the Eurozone. Indeed, a few breaking points are observed on July 20, just one day before the agreement between EU and IMF for a second bailout package totaling €109 billion. However, this was not able to contain speculations on a potential Greek default, which were echoed in the European Commission President’s warning that the sovereign debt crisis was spreading beyond the Eurozone periphery. According to our findings, developments in Italy and Spain as well as continuing speculations about a looming Greek exit were diffused across the Eurozone CDS markets, via the rising Italian and Spanish bond yields. This period of contagion transmission is further consistent with the onset of the upward trend in the Italian and Spanish bond yields, evidenced in Panel B of Figure 1.

We subsequently contrast the results in Table 3 with those from the original series in Table 8 of the Appendix. Contrary to Panel A in Table 3, estimates in Panel A of Table 8 provide evidence of widespread contagion from each of the GIIPS bond markets to practically all European bond markets. Since these findings are derived from the analysis of the original series, but not confirmed from that of the residual ones, they point to a form of “wake-up call contagion”. Importantly, most correlation change-dates fall within the late 2008–early 2009 period. Thus, we can infer that movements in the European bond markets were part of the general repricing of sovereign credit risk after the global financial crisis and during the onset of the Eurozone crisis. Turning to Panel B in Table 8, we observe that estimates exhibit only marginal deviations from those of Panel B in Table 3, confirming the transmission of “pure contagion” from the GIIPS bond markets to all European CDS markets.

5.2 Transmission of contagion from the GIIPS CDS markets to the EMU bond and CDS markets

5.2.1 Bond-CDS contagion

Table 4, presents results when contagion is assumed to be stemming from the CDS market. Estimates from the examination of the CDS-bond transmission in Panel A, reveal that contagion phenomena are restricted only within the EMU South and Belgium. The majority of correlation change-points indicating contagion are observed during the third quarter of 2011 and in particular, November 2011. Initially, contagion appears to originate from Greece in early November (column 1 in Panel A). The Greek CDS-stemming contagion could be linked

to the political developments in Greece during that period, which had implications for the Eurozone's viability. On October 31, 2011, the Greek Prime Minister called for a referendum on the EU/IMF rescue plan for Greece agreed only days earlier, which on November 2, 2011 was modified to be a referendum on Greece's Eurozone membership.¹⁹ Shortly after the referendum call, efforts for the formation of a national unity government in Greece temporarily collapsed (November 4, 2011), only to resume successfully seven days later.²⁰

[Insert Table 4 about here.]

According to our estimates, this heightened period generated contagion from the Greek CDS spreads towards the periphery bond yields. Shortly after, contagion phenomena further emerged from the periphery economies. Results in columns 2-5 of Panel A show that the CDS markets of Ireland, Italy, Portugal and Spain transmitted contagion to the bond markets in each country of the EMU South and Belgium. The dates of this transmission are located in the days right after the formation of the Greek government unity. Evidently, the political uncertainty in Greece sparked a contagion wave within the EMU South. Importantly, this transmission is not verified by the analysis of the original series: most of the break dates surrounding the Greek developments of November 2011 do not appear or enter with a non-significant sign (Panel A of Table 9 in the Appendix). This points to a short-lived episode of "limited contagion" owing to the abrupt political uncertainty in Greece that shortly receded.

5.2.2 Intra-CDS contagion

The analysis at the intra-CDS market level (see Panel B in Table 4), provides evidence of contagion stemming from the EMU South's biggest economies, i.e., Italy and Spain to practically all European CDS markets. Estimates in columns 3 and 5 of Panel B, show that contagion transmission took place over two phases: the first and more intense was during November 2011, where Italian and Spanish CDS spreads directed contagion towards each

¹⁹The rescue plan included a 50% debt write-off for private sector investors and €130bn of new bailout loans to Greece. The initial referendum call on the proposed EU/IMF rescue plan was modified after pressures from the French President Nicolas Sarkozy and the German Chancellor Angela Merkel in the G20 Cannes summit of November 2, 2011. The call for referendum was abandoned on November 4, 2011.

²⁰On November 4, 2011, the leaders of the two largest political parties engaged in talks for the formation of a government of national unity, but talks collapsed within the same day. The following day, Greek Prime Minister George Papandreou resigned and succeeded by Lucas Papademos on November 11, 2011, who led a new government of national unity.

Eurozone country. These results complement those of Panel A in Table 4, where the same period – and the accompanied political developments in Greece – were found responsible for the transmission of CDS-stemming contagion to periphery bond markets. Based on our estimates, this period further establishes Italy and Spain as the absolute transmitters of CDS contagion towards all European CDS markets. Indeed, during the respective period we observe a surge in the CDS spreads of both EMU blocs (Panels C and D in Figure 1).

The second phase of this transmission is observed in the first weeks of 2012, the period leading to the Greek debt restructuring of March 2012.²¹ Although contagion phenomena are not as intense as in late 2011, they are still evident and mainly directed towards the largest economies of the core (France, Germany, the Netherlands) and Belgium. Notwithstanding their size, the increasing influence of Italian and Spanish CDSs can also be explained when considering the nature of the CDS contracts. By construction, CDSs mainly reflect sovereign credit risk. Hence, a possible default by either Italy and/or Spain could trigger domino effects that could eventually lead to the collapse of the EMU. Panel B, shows that these concerns and the consequent contagion transmission were more prevalent in late 2011 (primarily) and early 2012 (secondarily). The exposure of European CDS markets to the Italian and Spanish CDS spreads stands in contrast to previous findings within the intra-CDS context, where contagion was found to be either non-existent (see Caporin *et al.*, 2018) or only a European periphery phenomenon (see Broto and Pérez-Quirós, 2015), and to arguments about the limited capacity of the GIIPS to generate contagion (see Kalbaska and Gatkowski, 2012).

We further compare the results from the residual series in Table 4 with those from the original series in Table 9 of the Appendix. Panel A of Table 9, points to certain cases of CDS-bond contagion from the periphery (Italy, Portugal, Spain) towards the core (Germany, Finland, France, the Netherlands) that are not evident in Panel A of Table 4. As such, these cases, which are mainly concentrated in the second half of 2010, cannot be classified as “pure contagion”. They can be rather perceived as a “wake-up call” for government bond

²¹On February 12, 2012, the Greek parliament voted in favour of a second bailout package for Greece totalling €130 billion. The parliamentary approval of this package was a prerequisite for the debt restructuring of March that occurred between March and April 2012. Under the restructuring, the Greek government amended the conditions of bonds under Greek law with a total face value of €177 billion. The restructuring included the private sector’s involvement, where investors were required to accept a 53.5% haircut of the face value of Greek government bonds. As a result, the nominal value of Greek debt was reduced by €107 billion, approximately 50% of GDP. It constituted the world’s biggest debt restructuring, involving securities of €206 billion.

yields across the core and attributed to the investors' upward repricing of those countries' fundamentals, prompted by the rising CDS spreads across the periphery. Last, estimates from Panel B of Table 9, do not reveal additional cases of intra-CDS contagion that are not consequently confirmed by the main estimations in Panel B of Table 4.

5.3 Discussion

Our analysis identifies two key phases for the transmission of contagion within the European sovereign financial market framework. The first extends from late 2009 to mid 2011 and is characterized by the outright transmission of contagion from the bond markets of the EMU South towards the CDS markets of all member states. Contagion phenomena emerge in the periods corresponding to the Greek deficit's upward revision (November 2009) and the fears of a Greek default that culminated in the country's first bailout package (April 2010), the rescue programmes in Ireland and Portugal (November 2010 and April 2011 respectively) and the negative economic developments in Italy and Spain (July-August 2011).

However, during these intense periods there is absence of bond-stemming contagion towards any of the European government bonds whatsoever. The rising CDS spreads in both the periphery and core during these periods (also evident in Panels C and D of Figure 1) indicate that the negative developments in the periphery were only transmitted across the CDS market. It therefore appears that the European sovereign bond and CDS markets differ with regards to their capacity to preserve or amplify potential shocks occurring in the periphery.

The next phase is around November 2011, and includes Greece's referendum announcement and the prolonged political uncertainty that revived the fears of the country's Eurozone exit. In this period, we observe a) the emergence of CDS-bond contagion in the periphery, and b) the transmission of contagion from the Italian and Spanish CDS spreads to either blocs' CDS spreads. This period marks the onset of the triggering capacity of the CDS market; for the first time during our examination, rising sovereign default probabilities are transmitted to the periphery bond yields as well as to both blocs' CDS spreads. Evidently, this is the critical point where the Greek debt crisis developed into a European debt crisis.

Our estimates further elevate Italy and Spain to key determinants of sovereign credit risk across the Eurozone: Italian and Spanish CDS spreads are the only transmitters of contagion

to all countries' CDS spreads. Importantly, this transmission mainly occurs during the late 2011-early 2012 period. In addition to speculations about a Greek Eurozone exit, this period was also characterized by increasing market concerns over the economic outlook and debt sustainability of Italy and Spain.²² Due to their economic size (third and fourth EMU economy respectively), both countries are considered too expensive to be realistically bailed out. Only for Italy, where public debt stands at approximately €2 trillion (the world's fourth largest), a 3-year rescue support program was estimated by the IMF at €600 billion. The European Financial Stability Fund evidently did not have enough funds to accomplish this, thereby rendering either country as too big to bail out. Hence, this period reflected the fears that a Greek default would cause a domino effect, causing Italy and Spain to fall as well, with resulting implications for Eurozone's stability.

Interestingly, this heavy influence of the Italian and Spanish CDS markets appears to ease after March 2012. According to our estimates, the presence of contagion phenomena regardless of their source is rather limited thereafter. A potential explanation can be offered by an important development in the beginning of 2012, namely the Greek debt restructuring of March 9, 2012. This historical deal prevented a Greek default and demonstrated the EU's willingness to preserve the Eurozone. For what matters, this deal is associated with the minimization of contagion phenomena across the Eurozone entering the second quarter of 2012, particularly those directed from the periphery CDS markets.

The analysis of the original series (Tables 8 and 9 of the Appendix), provides evidence of a "wake-up call" in the European bond markets in late 2008-early 2009. This period was marked by the spillovers of the financial crisis and the early seeds of the Eurozone crisis. Either crises provided new information, prompting investors to reassess the vulnerability of other countries, leading to a repricing of sovereign bonds (Bekaert *et al.*, 2014; Claeyns and Vařicek, 2014). The consequent rise in GIIPS long-term government bond yields was further fueled by recession and government announcements of bank rescue operations that exacerbated investor perceptions of sovereign credit risk. This was reflected in the opposite

²²In Italy, the low rates of productivity and output growth were not keeping up with an increasing debt load of almost 120% of GDP, the second highest in the Eurozone, behind Greece. In Spain, the property bubble eventually turned to bust, resulting in the country's banks accumulating a mounting pile of bad mortgage debts, and the highest unemployment rate in the EMU. Along these lines, Italy's credit rating was decreased by three notches in late 2010, while Spain's credit rating was cut three times (one notch each time) from 2010 to 2011.

evolution of government bond yields between the periphery and core (Monfort and Renne, 2014; Fontana and Scheicher, 2016), a fact also evident in Panels A and B of Figure 1. The concentration of correlation break-dates in this period, indicates that these developments were priced by the bond markets, only to the extent that affected bond fundamentals.

6 Further analysis

The previous section revealed that the countries of the EMU South acted as transmitters of contagion not only within their own country-bloc but importantly, towards the EMU North. This transmission took place over two distinct phases of the European sovereign debt crisis. Having established that, in this section we perform a number of tests to further enlighten our findings and assess their heterogeneity around certain fiscal and regulatory events. Moreover, we examine the role of liquidity and basis deviations for contagion transmission.

6.1 The developments in Greece

Estimates in section 5.1 pointed to a surge in contagion transmission from the Greek bond yields to almost all European CDS spreads in the days surrounding two important events: the November 2009 Greek upward deficit revision (primarily), and the April 2010 Greece's request for financial assistance (secondarily). Since these essentially marked the onset of the Eurozone crisis, we examine the transmission of contagion from the Greek bonds to the rest EMU CDSs in the different subperiods associated with these developments. In specific, we estimate our bootstrap test for the period between the deficit revision and the financial assistance request, as well as for the periods preceding and succeeding each event.

Table 5 presents estimates for the period extending from January 2, 2006 (the beginning of our sample) until April 23, 2010 (the Greek Prime Minister's announcement on the EU/IMF financial aid request), with November 16, 2009 being the break-date. We observe that following the deficit revision and until the financial assistance request, the negative movements in the Greek bond yields are transmitted to each of the remaining countries' CDS spreads: all p -values are statistically significant at conventional levels, pointing to the contagious nature of the Greek bonds during this period. Importantly, the negative movements in the Greek bonds led to a rise in default probabilities across the Eurozone as reflected in

euro area CDS spreads. We subsequently test for contagion from November 16, 2009 until April 5, 2018 (the end of our sample), with April 23, 2010 as break-date. We observe that contagion phenomena recede in the period following the Greek government's request for the EU/IMF economic adjustment programme (non-statistically significant p -values).

[Insert Table 5 about here.]

This pattern is further confirmed when examining the evolution of the dependence structure between our series. Figure 2 illustrates the local Gaussian correlation estimates for the Greek bond yields vis-à-vis each of the remaining countries' CDS spreads, for the period before November 2009 (green line) and for the period between November 2009 and April 2010 (red line). Overall, we observe an increased sensitivity of the European CDS market to the negative developments in Greece. The estimates between the Greek sovereign bond and the European CDS markets provide strong evidence of increased dependence for all country pairs during the Eurozone crisis. For all the Greek bond–European CDSs pairs, the entire local correlation curve for the pre-crisis period has moved up. It should further be noted that in most cases there is a similar uniform increase in local correlation over the different segments of the distribution. This points to the existence of strong linkages between the Greek bond market and the European CDS markets during the post-crisis period.

[Insert Figure 2 about here.]

On the same line, we contrast local Gaussian correlation estimates for the period between November 2009 and April 2010 (green line) against estimates for the period after April 2010 (red line) and plot them in Figure 3. Again, we observe a positive correlation across all country-pairs in the pre-April 2010 period, that declines thereafter. This decline is more potent for the core relative to the periphery (including Belgium); for the latter, it lies above zero even after April 2010. This points to the existence of strong linkages (albeit not contagious) between the Greek bond and these countries' CDSs after April 2010.

Overall, results from this exercise highlight the dominant role of the November 2009 Greek developments and the contagious capacity of the Greek bond (due to a potential Greek default) for the rising sovereign credit risk across the Eurozone. Moreover, they demonstrate the easing effect exerted by the agreement on the EU/IMF aid programme in April 2010.

[Insert Figure 3 about here.]

6.2 The naked CDS ban

In an attempt to curb destabilizing speculation on distressed Eurozone countries' default the EU banned the purchase of naked CDS contracts, effective November 1, 2012.²³ Because bond and CDS markets are complementary, the naked CDS buyers might inflate sovereign CDS spreads, thereby driving up sovereign bond yields (Silva *et al.*, 2016; Gyntelberg *et al.*, 2018). As Section 5.2 revealed, contagion from the CDS market to the bond market was mainly observed during the Greek Eurozone exit discussions in November 2011.

Although the Greek-stemming contagion subsided following the cancellation of the referendum and the formation of a national unity government in Greece, contagion phenomena were further evident for the remaining countries of the EMU South. During the same period, the regulation on naked CDS contracts was voted into law by the European Parliament (November 15, 2011). Since these events coincide, it may be that the intense political events in Greece curtail the easing effect of the regulatory ban brought about by the passage of the law, especially on the bonds of Ireland, Italy, Portugal and Spain.

To adequately isolate the effect of this regulatory change, we examine its impact on contagion transmission. We expect that the main transmission channel is from the CDS market to bonds. Moreover, although the regulation's effective date was known in advance, the voting outcome was fairly unanticipated. To this end, Table 6 examines differences in contagion transmission from the GIIPS CDSs to all European bonds in the periods before and after the regulation's passage in November 15, 2011. We find no evidence of contagion from the periphery CDS markets to any of the two blocs following the EU Parliament's voting: all p -values generated by our test are non-statistically significant after November 15, 2011.

[Insert Table 6 about here]

To fully detect whether this reflects a change from the pre-November 2011 period, we further examine the evolution of the dependence structure between our series. Figure 4,

²³The Regulation on Short Selling and certain aspects of Credit Default Swaps was published in the Official Journal of the European Union on March 24, 2012. The regulation prohibits any person or legal entity in the European Union from entering into uncovered ("naked") CDSs on sovereign debt and restricting uncovered short sales on shares and sovereign debt after November 1, 2012. Effectively, the Regulation bans CDS contracts on sovereign debt that do not hedge exposure to the sovereign debt itself, or to assets or liabilities whose value is correlated to the value of the sovereign debt.

illustrates the local Gaussian correlation estimates for the GIIPS CDS spreads vis-à-vis each of the remaining countries' bond yields, for the period before November 15, 2011 (green line) and the period after (red line). Evidently, there is a strong positive correlation between the CDSs and bonds of GIIPS pre-November 2011, which recedes close to zero in the period after. On the other hand, the correlation between GIIPS CDSs and core bonds is very close to zero in both subperiods. Taken together, results from this exercise confirm the containment effect of the regulatory ban on the transmission of CDS-bond contagion within the EMU South. The passage of the regulation appears to have strengthened the easing effect exerted by the reversal of the negative developments in Greece in early November 2011, namely the country's political instability and intention about a Eurozone membership referendum.

[Insert Figure 4 about here]

6.3 The role of liquidity

In periods of market distress investors tend to rebalance their portfolios towards less risky and more liquid securities, a phenomenon referred to as “flight-to-quality” or “flight-to-liquidity” (Beber *et al.*, 2009; Brunnermeier, 2009; Monfort and Renne, 2014; Fontana and Scheicher, 2016; Gyntelberg *et al.*, 2018). This is particularly important in the context of the European sovereign bond market, since the destination of large flows into (and out of) this market is determined almost exclusively by liquidity (Beber *et al.*, 2009; Monfort and Renne, 2014).

In practice, it is difficult to disentangle the two phenomena in the Eurozone crisis setting. If investors decrease their periphery bond holdings in favour of core countries' bonds, it is not clear whether they do so because of concerns about credit risk or liquidity risk (also given the strong correlation between the two). However, if contagion transmission from the EMU South (due to rising bond yields and/or CDS spreads) is also accompanied by a general drop in liquidity (an illiquidity contagion) across the same bloc, this would be a supporting argument that liquidity is an additional driver of investors' actions during the crisis.

To examine this premise, we test for contagion between sovereign bond and CDS liquidity by replacing our bond yields and CDS spreads with a measure of liquidity, namely the quoted bid-ask spread (see Goyenko and Ukhov, 2009).²⁴ To this end, we estimate Equation

²⁴The quoted bid-ask spread of the bond (CDS) is equal to $(Ask - Bid)/(0.5(Ask + Bid))$, where *Ask* and *Bid* are the ask price and bid price respectively of the bond (CDS); see Goyenko and Ukhov (2009).

(1) with the 5-year sovereign bond (CDS) quoted bid-ask spread as dependent variable, and use the residuals to test for contagion between our liquidity measures. In essence, we assume that if the negative developments in the periphery affect investors' liquidity risk considerations (raising bid-ask spreads), this should also be reflected in the transmission of illiquidity across that bloc. Furthermore, if investors search for liquidity in the core, the latter should be relatively immune to liquidity developments in the periphery.

Results from this exercise are presented in Table 7, where we initially focus on the transmission of illiquidity from the GIIPS bonds to all European bonds (Panel A) and CDSs (Panel B). According to Panel A, an increase in quoted bid-ask spreads in any of the GIIPS bonds is not transmitted to the bid-ask spreads in either of the two country-blocs; almost all p -values are non-statistically significant at conventional levels.

[Insert Table 7 about here]

When turning to Panel B, we notice that a rise in bid-ask spreads in any of the GIIPS bonds is transmitted to that bloc's CDSs; importantly, the transmission dates match closely those of the transmission of price contagion in Panel B of Table 3. Nevertheless, this rise in periphery bond illiquidity is not transmitted to the core bloc's CDSs. If this is combined with the results about the transmission of bond-CDS price contagion in Section 5.1.1, we can argue that a rise in credit risk in the periphery has contagious effects on both blocs' CDS markets, whereas a rise in liquidity risk is only confined within the periphery CDS markets.

On the other hand, testing for illiquidity contagion stemming from the GIIPS CDS markets (either to all bond markets or to all CDS markets) did not yield significant results (not presented for brevity). We conclude, that if anything, liquidity risk is a material source of risk in the context of the European sovereign CDS market only when stemming from the sovereign bond market. Moreover, credit risk emerges as a stronger determinant of price discovery between the two markets relative to liquidity risk.

6.4 Additional tests

We conduct additional sensitivity tests, the results of which are available on request. First, we consider the role of arbitrage opportunities as explained by the deviations of the CDS-bond basis from its zero equilibrium value. In fact, arbitrage opportunities during the Eurozone

crisis were primarily present due to “funding frictions” and “short-selling frictions”. While the former type of frictions made it difficult for arbitrageurs to finance the purchase of the bond (via repo transaction) for implementing a “negative basis trade”, the latter prevented arbitrageurs to short-sell the bond (in a “positive basis trade”) in order to profit from the relative mispricing (see [Fontana and Scheicher, 2016](#)).

Since arbitrage opportunities affect the equilibrium relationship between bonds and CDS contracts, we test for contagion transmitted from the GIIPS bond yields towards the CDS-bond basis of all countries. We do so, by estimating Equation (1) with the 5-year sovereign CDS-bond basis as dependent variable, and calculate the residuals. We subsequently test for contagion between our residual bond yields and the residual basis series. Across our sample period, we find no dates that indicate contagion transmission from the GIIPS bonds to the basis of any country in the two blocs. This is further evident, when the reverse direction is considered, i.e., from the GIIPS basis to all European bond yields.

We conclude that although arbitrage forces might have been present in the context of the Eurozone crisis, they have nevertheless been unable to affect contagion dynamics. In fact, frictions and imperfections such as illiquidity and high trading costs often prevent arbitrage forces from fully closing the pricing gaps between the two markets; if markets are subject to such frictions, it is possible that the correcting mechanisms may have different regimes with different adjustment speeds ([Gyntelberg et al., 2018](#)).

Second, we investigate the impact of the ECB’s government bond purchases under the Securities Markets Programme (SMP). These purchases aimed at lowering yields and liquidity premia in the distressed countries’ sovereign bonds and restoring monetary policy transmission ([Eser and Schwab, 2016](#)). As such, we expect that these purchases contributed to the minimization of contagion transmission from the GIIPS bond markets. To examine this, we test for differences in intra-bond contagion before and after the SMP’s implementation in May 10, 2010. We find no evidence of contagion following the May 2010 period. We then test for changes in the correlation of our series with May 10, 2010 as break-date; again the correlation is around zero in both subperiods. Since the programme was reactivated in August 7, 2011 to enable the purchases of Italian and Spanish bonds, we further test for contagion before and after this date; again, results confirm the absence of intra-bond contagion

between almost all country-pairs.

Finally, we consider the impact of the two LTRO auctions (December 8, 2011 and February 29, 2012) on contagion transmission from the GIIPS bonds to all European bonds. If the massive liquidity injections to the euro area banking sector eased sovereign default concerns via the sovereign-bank nexus, we expect the minimization of contagion phenomena following these auctions. We fail to find any contagion phenomena within the European sovereign bond market in the periods before and after each of the auction dates.

6.5 Summing up

Overall, results in this section reveal certain heterogeneities in the sovereign bond and CDS contagion dynamics during the Eurozone crisis. These are mainly associated with the occurrence of certain fiscal and regulatory events during the main phase of the crisis. Moreover, contagion phenomena primarily arise due to credit risk concerns, although liquidity risk also plays a non-trivial role, particularly when stemming from the GIIPS bond markets.

7 Conclusion

By adopting the correlation concept of contagion by [Bekaert *et al.* \(2005\)](#), and employing a new measure of local Gaussian correlation by [Tjøstheim and Hufthammer \(2013\)](#), we examine asymmetric and nonlinear changes in dependence structure and test for contagion in the European sovereign bond and CDS markets during the 2006-2018 period. Our empirical findings suggest that contagion phenomena in the European sovereign bond and CDS markets have undergone two major phases.

The first phase extends from late 2009 to mid 2011 and concerns the outright transmission of contagion from the bond markets of the EMU South towards the CDS markets of both EMU blocs. Most correlation break-points that indicate contagion are concentrated around important economic events. These include the Greek deficit's upward revision (November 2009) and the consequent first rescue plan for Greece (April 2010), the financial assistance programmes in Ireland and Portugal (November 2010 and April 2011 respectively) and the negative economic developments in Italy and Spain (July-August 2011). In contrast, during these intense periods, bond markets in both blocs appear immune to contagion stemming

from the periphery bond markets; they were rather subject to a “wake-up call” during the late 2008–early 2009 period. This “wake-up call” prompted investors to reassess the vulnerability of Eurozone countries, leading to a repricing of the factors that affect sovereign bonds (Bekaert *et al.*, 2014; Claeys and Vařicek, 2014)

The second phase is during November 2011 and the Greece’s referendum announcement that prolonged political uncertainty and revived the fears of the country’s Eurozone exit. This period is characterized by contagion stemming a) from periphery CDS spreads towards bond yields in the periphery and Belgium and b) from Italian and Spanish CDS spreads towards CDS spreads in both EMU blocs. The outbreak of CDS-stemming contagion during the respective period marks the development of the Greek debt crisis into a European debt crisis. During this escalation, Italy and Spain emerge as key countries for the evolution of sovereign credit risk across the Eurozone. This is consistent with the “too big to bail out” status of either country. Arguably, the fears that a Greek default would generate a domino effect, causing Italy and Spain to fall as well, were well-founded.

Our examination of the mechanisms of contagion transmission shows that liquidity risk is an important source of risk in the European CDS market, but only when stemming from the bond market. We content that credit risk is a stronger determinant of price discovery and contagion transmission between the two markets relative to liquidity risk. We further show that although arbitrage forces may be present, they were unable to fully close the pricing gaps between the two markets and affect contagion dynamics. Last, by studying the regulatory response during the crisis, we provide evidence that the voting of the naked CDS ban exerted a containment effect on the transmission of CDS-bond contagion within the EMU South.

A clear implication from our analysis concerns the management of the financial institutions’ exposures, which should correspond to the nature and channels of contagion during crises. In addition to first-order/direct exposures, regulation should encourage financial institutions to also manage second-order risks, such as those related to intra- and cross-asset correlations, particularly when being largely exposed to sovereign debt issues by countries under fiscal strain. To this end, the identification of the extent to which contagion phenomena depend on bilateral and multilateral exposures between countries would be of interest. We leave that to future research.

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Table 1: This table reports means, standard deviations, maximums and minimums for bond yield and CDS spread changes (in basis points). Bond yields is the yield on 5-year on-the-run sovereign bonds. CDS spread is the change in 5-year sovereign CDSs. The sample includes Austria, Belgium, Finland, France, Ireland, Italy, Germany, Hellenic Republic (Greece), the Netherlands, Portugal, Spain. Bond yields are from Datastream, and CDS spreads are from Markit. The sample period for Greece is January 2, 2006 to March 8, 2012.

	Δ (Bond yields)				Δ (CDS spreads)				Obs.#
	Mean	Std. Dev	Max.	Min.	Mean	Std. Dev	Max.	Min.	
AT	-0.119	4.41	37.32	-34.43	0.003	4.23	44.28	-28.11	3090
BE	-0.122	4.79	43.77	-50.30	0.003	5.38	40.55	-57.38	3090
DE	-0.124	3.73	19.12	-22.51	0.002	2.41	20.81	-20.27	3090
ES	-0.113	7.08	60.51	-88.94	0.008	10.94	96.88	-102.66	3090
FI	-0.121	4.63	51.17	-41.18	0.003	2.64	25.49	-25.73	3090
FR	-0.119	3.93	34.54	-46.41	0.003	4.66	46.79	-44.93	3090
GR	2.420	45.86	516.57	-551.57	15.398	643.86	9025.77	-9888.18	1505
IE	-0.120	10.86	102.04	-148.00	0.004	11.48	113.79	-152.45	3090
IT	-0.102	7.25	69.20	-91.05	0.018	16.32	162.73	-161.76	3090
NL	-0.122	5.77	85.29	-89.16	0.014	29.56	274.84	-275.94	3090
PT	-0.108	18.30	409.19	-222.46	0.002	3.67	79.10	-74.23	3090
GIIPS	0.395	17.87	231.50	-220.40	3.086	137.26	1895.65	-2075.85	
Core	-0.121	4.54	45.20	-47.33	0.005	8.15	75.46	-75.39	

Table 2: This table presents the changes in basis points on 5-year sovereign bond yields and CDS spreads after a specific adverse event on GIIPS (Greece, Ireland, Italy, Portugal and Spain). (1) On November 5, 2009, the Greek government revealed a revised budget deficit of 12.7% of GDP for 2009. (2) On January 15, 2009, Ireland abandons plans to inject €1.5 bn into third largest bank Anglo Irish Bank and nationalises the commercial lender amid fears it could collapse. (3) On April 18, 2012, the Italian government cut its growth forecast for the economy in 2012, predicting a further shrink by 0.8%. (4) On April 6, 2011, Portugal requests a bailout from the EU. (5) On April 27, 2012, the rating agency Standard and Poor's has lowered the rating of Spain by two notches to triple B plus and maintained a negative outlook. The last two columns present the correlation coefficient between changes on bond yields and CDS spreads in basis points before and after a specific event.

		AT	BE	DE	ES	FI	FR	GR	IE	IT	NL	PT	Correlation $\Delta(\text{Bond})$ vs $\Delta(\text{CDS})$	
													pre-event	post-event
Panel A: $\Delta(\text{Bond})$														
GR	(1) Budget revision	0.04	0.08	-1.06	1.75	0.08	-2.16	40	-2.33	-2.51	-3.94	5.3	0.36	0.59
IE	(2) Rescue AIB	-3.29	-1.74	-5.54	-1.27	-0.95	1.22	-3.39	5.32	-5.94	4.29	-2.52	-0.13	0.55
IT	(3) Growth forecast revision	-1.64	-2.86	-2	-8.83	-2.89	-0.7	-4.77	-0.51	-0.25	-2.33	-18.45	0.57	0.72
PT	(4) Bailout request	0.79	1.27	1.42	0.38	1.36	1.24	-2.53	-21.43	-0.67	1.61	10.21	0.48	0.51
ES	(5) S&P downgrade	-1.85	-0.33	0.55	11.65	-2.16	0.26	-15.32	0.51	4.54	-2.12	-62.04	0.53	0.75
Panel B: $\Delta(\text{CDS})$														
GR	(1) Budget revision	2.99	8.66	6.80	23.74	2.01	6.94	127.69	24.15	20.67	3.36	75.53		
IE	(2) Rescue AIB	4.43	-0.15	2.10	3.27	-0.17	1.35	5.69	35.54	1.48	2.11	1.24		
IT	(3) Growth forecast revision	4.54	9.72	4.50	6.56	2.52	1.10	//	5.25	18.99	6.79	-0.68		
PT	(4) Bailout request	-0.74	-2.54	-0.74	-3.00	-0.91	-1.47	2.31	-15.46	-0.28	-0.58	10.94		
ES	(5) S&P downgrade	-3.42	-2.67	-1.60	4.31	-0.82	-1.86	//	-4.84	0.07	-4.66	-37.97		

Table 3: Bootstrap test for contagion considering GIIPS as the countries of origin. This table shows p -values from the bootstrap test for contagion according to the time dates of correlation change-points estimated with the algorithm of Demetrescu and Wied (2019) based on the residuals from the Kalman filter model. The null hypothesis indicates no contagion between the GIIPS (GR, IE, IT, PT, ES) bonds and the European bond (Panel A) and CDS (Panel B) markets. The bootstrap test for contagion is based on 1000 replications. The sample period starts on January 2, 2006 and ends on March 8, 2012 for Greece and on April 5, 2018 for the rest EMU South countries.

	GR (1)	Date	IE (2)	Date	IT (3)	Date	PT (4)	Date	ES (5)	Date
<i>Panel A : Bond Markets</i>										
AT	0.995	13/10/2008	0.099	29/1/2009	0.612	20/7/2011	0.999	9/10/2009	0.694	8/12/2009
	0.987	30/11/2009	0.871	25/3/2010			0.677	22/3/2010	0.230	11/6/2012
BE	0.993	6/8/2009	0.450	3/3/2010			0.451	12/4/2011		
DE	0.998	16/10/2008	0.958	25/2/2009	0.997	22/10/2008	0.999	26/9/2008	0.853	29/12/2008
	0.962	26/11/2009	0.348	2/12/2010	0.995	27/4/2010	0.954	26/11/2009	0.768	3/8/2011
					0.965	7/7/2011				
ES	0.412	23/12/2009	0.045	1/12/2010	0.000	20/7/2011	0.013	24/3/2011		
					0.887	30/6/2015	0.899	3/7/2015		
FI	0.975	26/10/2009	0.931	2/12/2010	0.999	2/7/2008	0.985	22/4/2011	0.974	3/12/2009
			0.812	5/3/2014	0.996	19/5/2010			0.063	20/7/2011
FR	0.870	16/11/2009	0.765	1/4/2010	0.986	18/7/2011	0.956	29/1/2010	0.350	14/7/2009
			0.631	9/12/2010			0.349	20/4/2011	0.512	18/7/2011
GR			0.112	9/12/2009	0.371	15/10/2008	0.350	17/2/2009	0.630	23/12/2009
					0.968	23/11/2009				
IE	0.034	9/12/2009					0.980	30/11/2010	0.154	1/12/2010
IT	0.941	15/10/2008					0.662	10/2/2010	0.012	20/7/2011
	0.855	23/11/2009					0.635	20/5/2011	0.312	30/6/2015
NL	0.991	14/10/2009	0.978	23/1/2009	0.830	24/6/2011	0.891	30/9/2009	0.794	21/6/2011
							0.208	1/4/2011		
PT	0.944	17/12/2009	0.750	30/11/2010	0.332	10/2/2010			0.614	24/3/2011
					0.428	20/5/2011			0.431	3/7/2015
<i>Panel B : CDS Markets</i>										
AT	0.039	2/11/2009	0.035	15/11/2010	0.008	18/7/2011	0.000	28/3/2011	0.000	2/8/2011
BE	0.000	6/11/2009	0.000	25/11/2010	0.019	22/7/2011	0.000	1/4/2011	0.000	3/8/2011
	0.000	21/4/2010								
DE	0.004	5/11/2009	0.041	23/11/2010	0.033	20/7/2011	0.000	18/4/2011	0.032	20/7/2011
	0.662	22/04/2010								
ES	0.000	9/11/2009	0.006	2/12/2010	0.000	20/7/2011	0.000	11/4/2011		
	0.000	22/4/2010	0.540	25/07/2011						
FI	0.000	13/11/2009			0.002	2/8/2011	0.037	4/5/2011	0.000	3/8/2011
FR	0.002	5/11/2009	0.001	12/11/2010	0.000	20/7/2011	0.014	18/4/2011	0.000	18/7/2011
	0.431	21/04/2010								
GR			0.269	2/1/2008	0.212	6/11/2009	0.961	17/11/2009	0.881	9/11/2009
			0.057	12/11/2009	0.571	20/4/2010	0.412	22/4/2010	0.417	22/4/2010
IE	0.502	2/1/2008			0.615	3/12/2010	0.312	30/11/2010	0.770	2/12/2010
	0.009	12/11/2009			0.000	15/7/2011	0.041	29/3/2011	0.000	25/7/2011
IT	0.000	6/11/2009	0.008	3/12/2010			0.000	12/4/2011	0.000	20/7/2011
	0.000	20/4/2010	0.415	15/7/2011						
NL	0.083	5/11/2009	0.000	10/12/2010	0.016	2/8/2011	0.012	4/5/2011	0.016	18/7/2011
PT	0.001	17/11/2009	0.000	30/11/2010	0.000	12/4/2011	0.000	5/5/2011	0.045	11/4/2011
	0.000	22/4/2010	0.565	29/3/2011						

Table 4: Bootstrap test for contagion considering GIIPS as the countries of origin. This table shows p -values from the bootstrap test for contagion according to the time dates of correlation change-points estimated with the algorithm of Demetrescu and Wied (2019) based on the residuals from the Kalman filter model. The null hypothesis indicates no contagion between the GIIPS (GR, IE, IT, PT, ES) CDS and the European bond (Panel A) and CDS (Panel B) markets. The bootstrap test for contagion is based on 1000 replications. The sample period starts on January 2, 2006 and ends on March 8, 2012 for Greece and on April 5, 2018 for the rest EMU South countries.

	GR (1)	Date	IE (2)	Date	IT (3)	Date	PT (4)	Date	ES (5)	Date
<i>Panel A : Bond Markets</i>										
AT	0.233	2/11/2009	0.148	13/10/2011	0.912	31/10/2011			0.064	25/10/2011
	0.411	4/11/2011							0.000	21/11/2011
BE	0.000	3/11/2011	0.000	14/11/2011			0.137	30/11/2010		
DE	0.872	22/4/2010	0.341	12/10/2011	0.832	26/9/2008	0.985	26/9/2008	0.946	26/9/2008
					0.307	25/10/2011	0.204	15/5/2013	0.742	26/2/2010
ES	0.018	8/11/2011	0.000	15/11/2011	0.000	9/11/2011	0.885	6/4/2010		
			0.996	7/8/2015	0.991	22/6/2015	0.000	16/11/2011		
FI	0.311	13/11/2009	0.962	9/1/2008	0.865	23/4/2010	0.994	10/9/2008	0.807	23/4/2010
	0.411	15/11/2011	0.981	23/4/2010	0.411	21/10/2013	0.962	15/5/2013	0.639	15/5/2013
FR	0.452	21/4/2010	0.948	17/7/2008	0.757	13/11/2009	0.912	26/9/2008	0.720	6/4/2010
			0.127	10/10/2011	0.412	18/11/2011	0.002	2/8/2011	0.518	8/11/2011
GR			0.011	9/11/2011	0.000	10/11/2011	0.011	9/11/2011	0.000	8/11/2011
IE	0.009	9/11/2011			0.012	12/10/2010	0.000	18/11/2011	0.050	15/11/2011
					0.997	17/11/2011			0.619	7/8/2015
IT	0.000	10/11/2011	0.912	12/10/2010			0.000	14/11/2011	0.000	9/11/2011
			0.003	17/11/2011					0.004	22/6/2015
NL	0.981	5/11/2009	0.965	1/7/2008	0.992	23/1/2009	0.975	10/9/2008	0.963	26/9/2008
	0.757	7/11/2011	0.438	3/8/2010	0.659	26/3/2013			0.589	26/3/2013
PT	0.046	9/11/2011	0.000	18/11/2011	0.018	14/11/2011			0.589	6/4/2010
									0.000	16/11/2011
<i>Panel B : CDS Markets</i>										
AT			0.992	4/1/2013	0.001	22/11/2011	0.822	15/9/2008	0.988	15/9/2008
							0.996	1/2/2012	0.021	3/1/2012
BE			0.994	28/1/2012	0.005	10/1/2012	0.963	13/1/2012	0.033	9/1/2012
DE			0.788	13/1/2012	0.000	17/1/2012	0.975	15/9/2008	0.019	19/1/2012
							0.992	13/1/2012		
ES	0.034	16/11/2011			0.010	22/11/2011	0.987	31/10/2011		
FI	0.411	25/9/2008	0.899	19/1/2012	0.001	30/11/2011	0.641	2/1/2012	0.010	14/11/2011
FR					0.005	10/1/2012	0.418	12/11/2009	0.011	5/1/2012
					0.966	1/2/2013	0.847	13/1/2012	0.968	1/2/2013
GR			0.411	11/11/2011					0.000	18/11/2011
IE	0.015	9/11/2011			0.003	16/11/2011	0.872	5/1/2012		
					0.958	31/8/2015				
IT	0.983	14/9/2010	0.672	15/11/2011			0.868	14/11/2012	0.041	22/11/2011
			0.784	31/8/2015			0.994	3/9/2012	0.885	
NL	0.311	22/9/2008	0.760	7/12/2012	0.033	19/1/2012	0.478	5/9/2008	0.912	15/9/2008
	0.471	22/9/2010					0.988	13/1/2012	0.020	16/1/2012
PT			0.967	5/1/2012	0.000	15/11/2012			0.017	7/11/2011

Table 5: This Table presents the estimated p-values from the bootstrap test for contagion from the Greek bond yields (Column 2) and CDS spreads (Column 4) to the rest of the European sovereign CDS markets. Greece (GR) is considered as the country of origin for the European sovereign debt crisis. Significance levels at 10%, 5% and 1% are denoted by *, **, ***. Yes indicates that the null of no contagion is rejected at 5% level. The bootstrap test for contagion is based on 1000 replications.

Origin:Greece	Deficit Revision November 2009		Financial Request April 2010	
	<i>Bonds</i> (GR)	Contagion?	<i>CDS</i> (GR)	Contagion?
<i>CDS Markets</i>				
Austria	0.002***	Yes	0.391	No
Belgium	0.007***	Yes	0.845	No
Finland	0.001***	Yes	0.601	No
France	0.000***	Yes	0.559	No
Germany	0.008***	Yes	0.713	No
Ireland	0.041**	Yes	0.681	No
Italy	0.022**	Yes	0.000	No
Netherlands	0.007***	Yes	0.770	No
Portugal	0.011**	Yes	0.423	No
Spain	0.000***	Yes	0.718	No

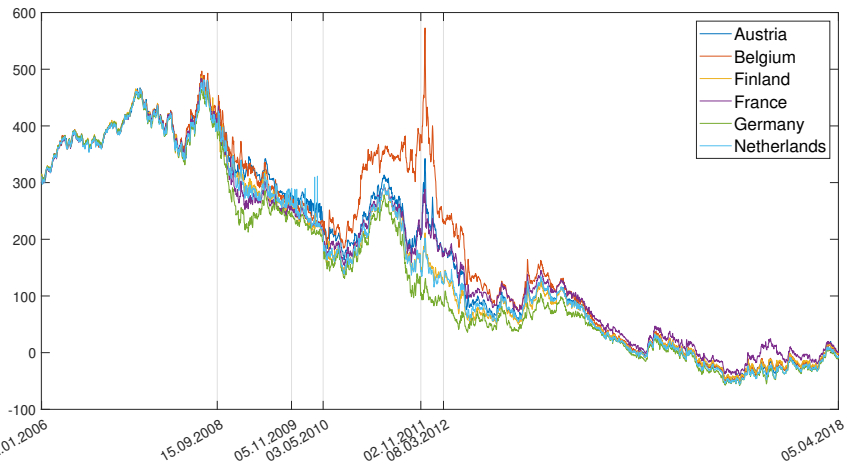
Table 6: This Table presents the estimated p-values from the bootstrap test for contagion from the GIIPS CDS spreads to the rest of the European sovereign bond markets. The bootstrap test for contagion is based on 1000 replications.

	GR (1)	IE (2)	IT (3)	PT (4)	ES (5)
<i>Bond Markets</i>					
AT	0.401	0.741	0.451	0.711	0.499
BE	0.431	0.293	0.694	0.441	0.792
DE	0.213	0.575	0.923	0.242	0.891
ES	0.415	0.641	0.952	0.921	–
FI	0.471	0.488	0.435	0.578	0.398
FR	0.542	0.790	0.944	0.414	0.813
GR	–	0.412	0.214	0.545	0.931
IE	0.551	–	0.741	0.225	0.611
IT	0.425	0.899	–	0.418	0.704
NL	0.405	0.216	0.902	0.601	0.915
PT	0.490	0.290	0.331	–	0.444

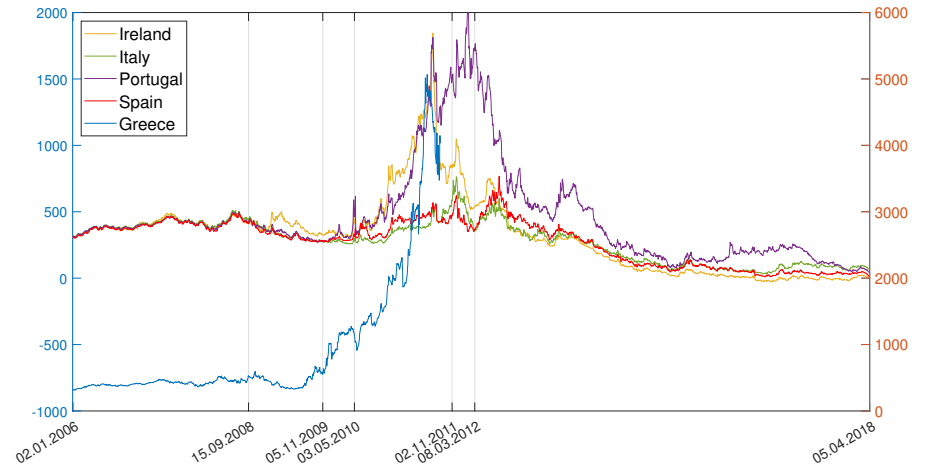
Table 7: Bootstrap test for contagion considering GIIPS as the countries of origin. This table shows p -values from the bootstrap test for contagion according to the time dates of correlation change-points estimated with the algorithm of Demetrescu and Wied (2019) based on the quoted bid-ask spreads. The null hypothesis indicates no contagion between the GIIPS (GR, IE, IT, PT, ES) bonds and the European bond (Panel A) and CDS (Panel B) markets. The bootstrap test for contagion is based on 1000 replications. The sample period starts on January 2, 2006 and ends on March 8, 2012 for Greece and on April 5, 2018 for the rest EMU South countries.

	GR (1)	Date	IE (2)	Date	IT (3)	Date	PT (4)	Date	ES (5)	Date
<i>Panel A : Bond Markets</i>										
AT	0.912	7/11/2011					0.890	22/9/2008		
BE	0.918	6/10/2009	0.640	4/1/2010	0.741	29/6/2011	0.240	6/10/2010	0.914	22/06/2011
DE	0.790	2/10/2009	0.454	14/12/2009	0.496	7/7/2011			0.246	5/7/2011
ES	0.915	17/09/2009	0.642	18/3/2010	0.784	13/7/2011	0.442	12/1/2011		
FI	0.948	3/11/2011	0.773	10/2/2010						
FR	0.974	21/09/2009	0.881	5/3/2010			0.821	11/11/2009		
GR			0.090	14/10/2009	0.470	29/6/2011	0.570	22/09/2009	0.593	17/09/2009
IE	0.531	14/10/2009			0.525	29/3/2010	0.997	28/1/2009	0.524	18/3/2010
IT	0.929	29/6/2011	0.752	2/2/2010	0.489	16/11/2009	0.754	25/4/2011	0.649	13/7/2011
NL	0.925	24/11/2011	0.430	10/12/2009	0.835	10/5/2010	0.604	6/10/2009		
PT	0.920	22/09/2009	0.809	28/1/2009	0.382	25/4/2011			0.295	12/1/2011
<i>Panel B : CDS Markets</i>										
AT	0.995	12/10/2009	0.344	25/11/2010	0.511	21/7/2011	0.878	21/3/2011	0.691	25/7/2011
BE	0.198	20/11/2009					0.996	30/3/2011	0.785	15/8/2011
DE	0.443	30/11/2009	0.675	8/11/2010	0.997	11/7/2011	0.994	5/4/2011	0.887	15/8/2011
ES	0.005	18/11/2009	0.011	25/11/2010	0.001	3/8/2011	0.000	14/4/2011		
FI	0.558	24/11/2009	0.487	2/11/2010			0.412	28/4/2011	0.908	12/8/2011
FR	0.778	23/11/2009			0.812	4/7/2011			0.411	2/8/2011
GR			0.002	29/10/2009	0.009	3/11/2009	0.021	11/11/2009	0.031	17/11/2009
IE	0.000	29/10/2009			0.017	22/6/2011	0.041	8/12/2010	0.031	25/11/2010
IT	0.001	3/11/2009	0.010	10/12/2010			0.012	20/7/2011	0.012	2/8/2011
NL	0.487	19/11/2009	0.511	1/12/2010	0.915	8/7/2011	0.191	9/5/2011	0.951	20/7/2011
PT	0.028	11/11/2009	0.015	24/11/2010	0.014	19/7/2011			0.021	12/4/2011

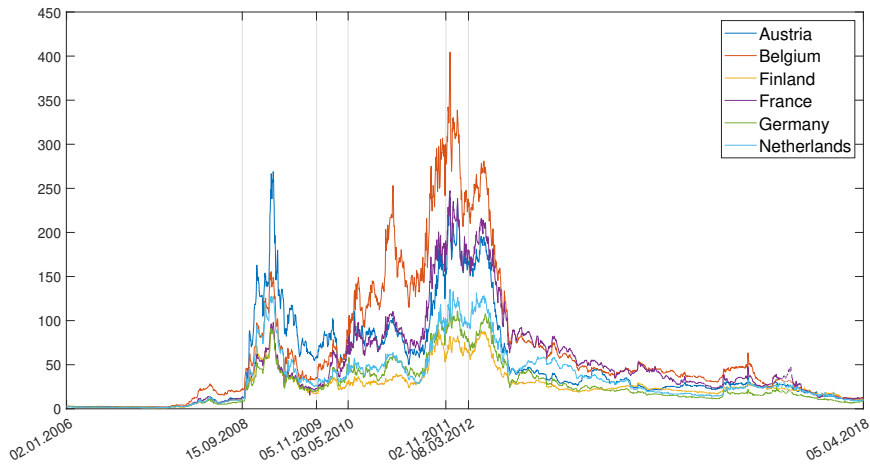
(A) EMU North 5-year sovereign bond yields



(B) EMU South 5-year sovereign bond yields



(C) EMU North 5-year sovereign CDS spreads



(D) EMU South 5-year sovereign CDS spreads

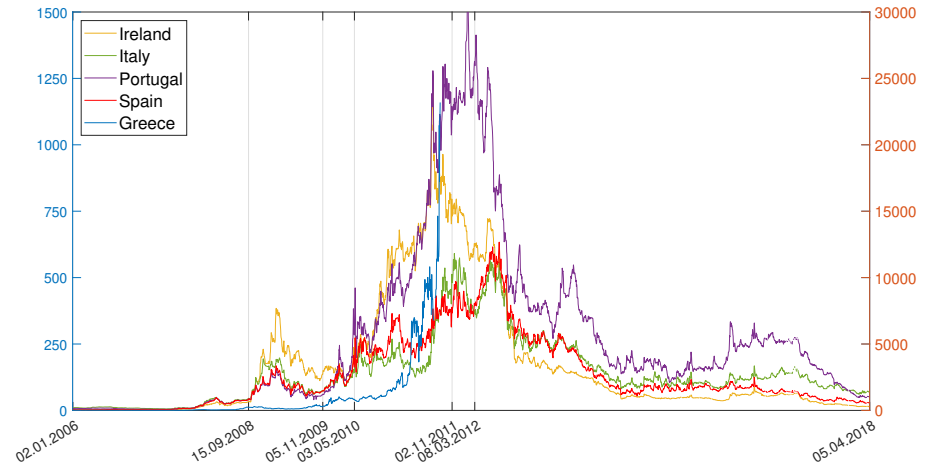


Figure 1: Evolution of sovereign bond yields and CDS spreads (in basis points). Panels A and C show the 5-year sovereign bond yields and CDS spreads for the countries of the EMU North (Austria, Belgium, Finland, France, Germany, Netherlands) and Panels B and D show the 5-year sovereign bond yields and CDS spreads for the countries of the EMU South (Greece, Ireland, Italy, Portugal, Spain). In Panels B and D, Greece is shown on the right axis while the rest EMU South countries are shown on the left axis.

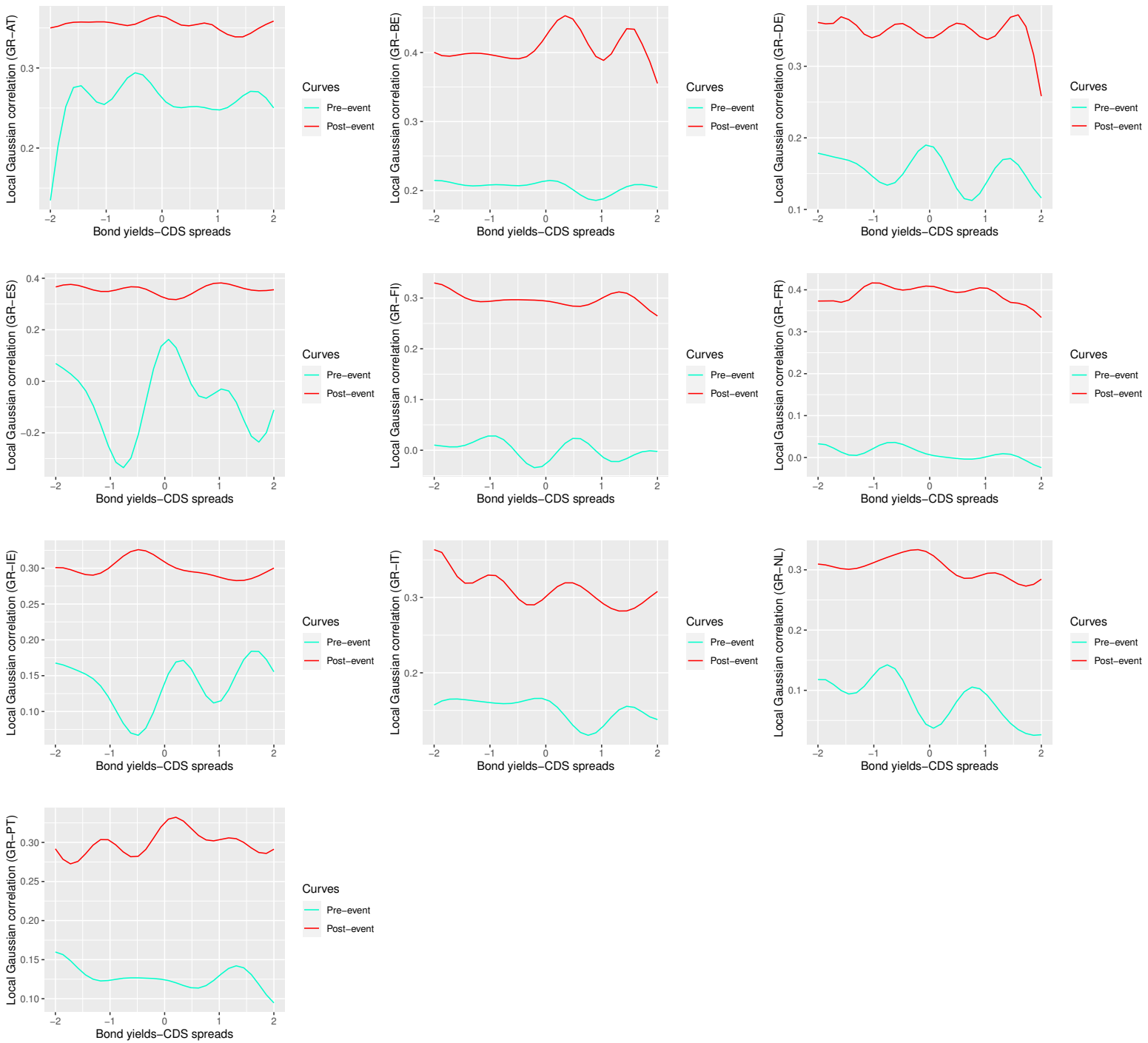


Figure 2: Local Gaussian correlation estimates between the Greek sovereign bond yields and each of the European sovereign CDS yields following Greece's upward deficit revision on November 16, 2009.

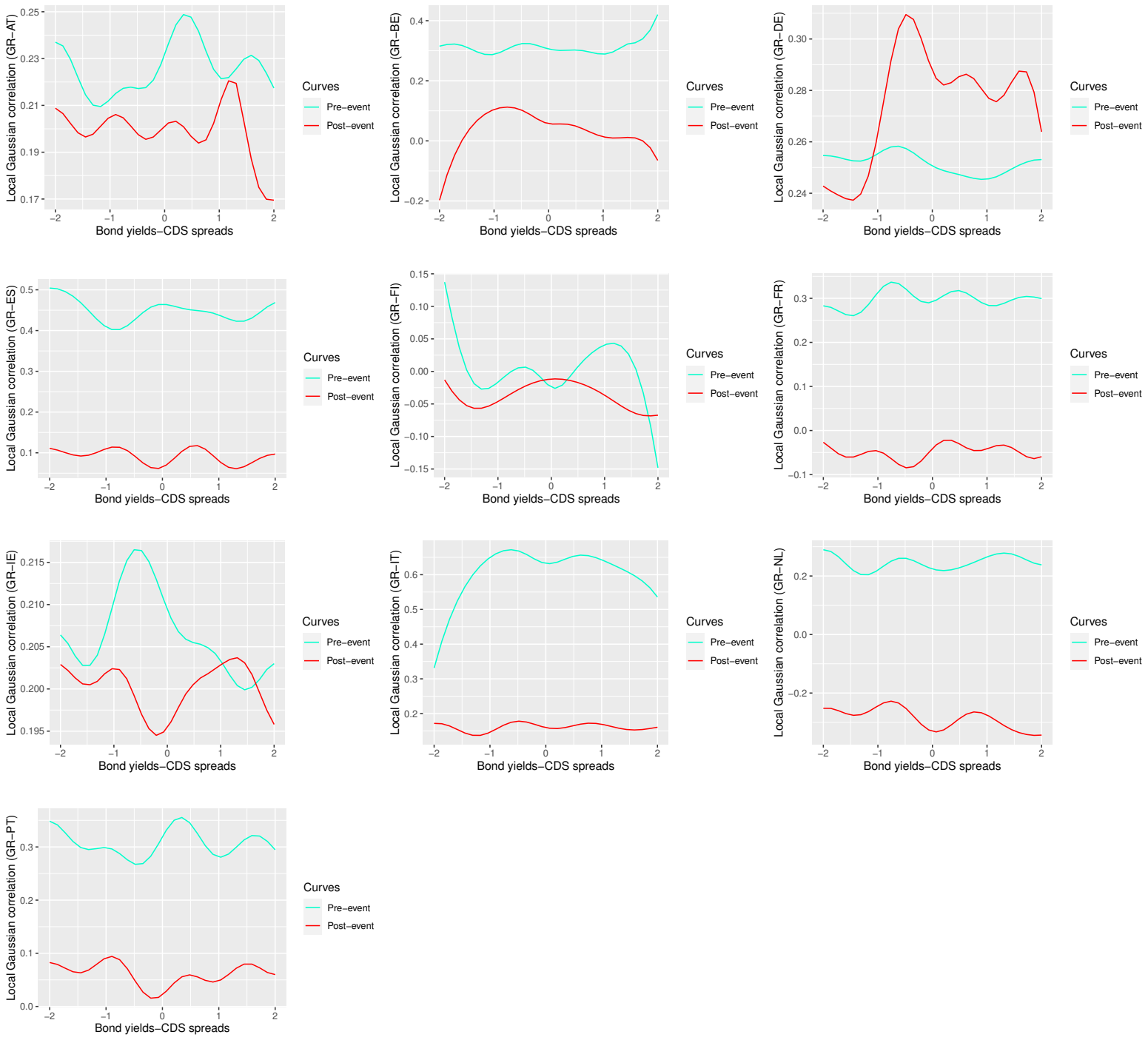


Figure 3: Local Gaussian correlation estimates between the Greek sovereign bond yields and each of the European sovereign CDS yields following Greece’s request for financial assistance in April 23, 2010.

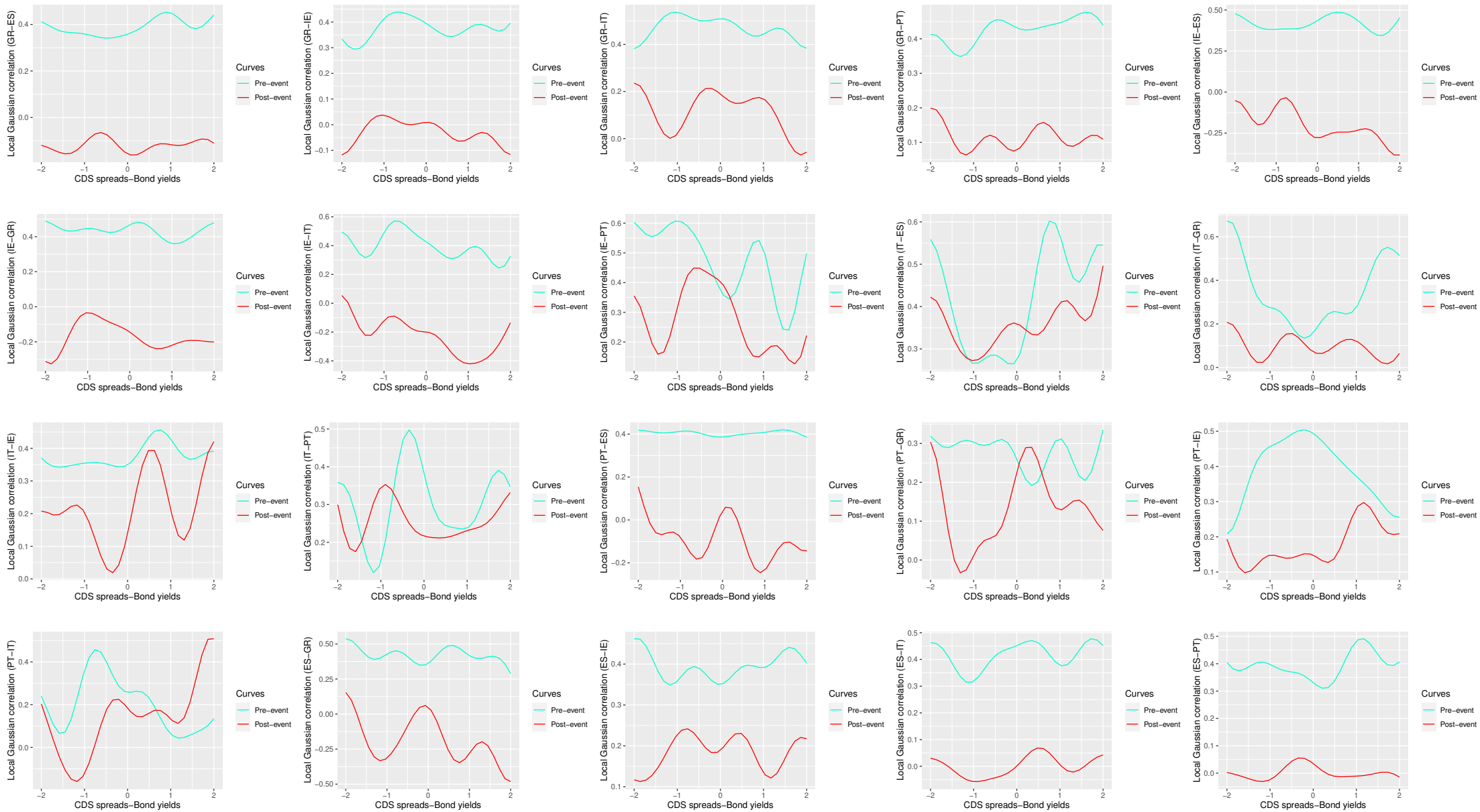


Figure 4: Local Gaussian correlation estimates between the GIIPS CDS spreads and each of the European sovereign bond yields following the voting of the naked CDS ban on November 15, 2011.

A Appendix

A.1 Demetrescu and Wied (2019) algorithm for the identification of structural changes in correlation

Suppose Y_t is a sequence of random variables for $t = 1, 2, \dots$ with finite absolute moments. Our purpose is to test whether the variance of Y_t is constant over time.

For example we test, $H_0 : Var(Y_t) = \sigma^2 \forall t \in T$ against $H_1 : \exists t \in \{1, \dots, T-1\} : Var(Y_t) \neq Var(Y_{t+1})$ for a constant σ^2 . The test statistic is then formed as,

$$Q_T(Y) = \max_{1 \leq i \leq T} \left| \hat{D} \frac{i}{\sqrt{T}} ([VarY]_i - [VarY]_T) \right|$$

where the empirical variance is calculated from the first s observations as,

$$[VarY]_s = \frac{1}{s} \sum_{t=1}^s Y_t^2 - \left(\frac{1}{s} \sum_{t=1}^s Y_t \right)^2 = \overline{Y_s^2} - (\overline{Y_s})^2.$$

The scalar $\hat{D} = \left((1, -2\overline{Y_T}) \hat{D}_1 (1, -2\overline{Y_T})' \right)^{-1/2}$, which mainly captures the fluctuations resulting from estimating the expected values is necessary for the asymptotic null distribution. The test rejects the null hypothesis of constant variance in the case that the empirical variances fluctuate too much, given by:

$$\max_{1 \leq i \leq T} |[VarY]_i - [VarY]_T|.$$

The weighting factor $\frac{i}{\sqrt{T}}$ is scaling down deviations at the beginning of the sample when the $[VarY]_i$ are more volatile.

A transformation is needed in order to examine the large sample properties,

$$Q_T(Y) = \sup_{l \in [0,1]} \left| \hat{D} \frac{\tau(l)}{\sqrt{T}} \left([VarY]_{\tau(l)} - [VarY]_T \right) \right| \text{ with } \tau(l) = [1 + l(T-1)]$$

Under the null hypothesis and assumptions A.1 – A.4 in ?,

$$Q_T(Y) \rightarrow \sup_{l \in [0,1]} |B(l)|, \text{ where } B(l) \text{ is a one-dimensional Brownian Bridge.}$$

The quantiles and the limit distribution of $Q_T(Y)$ provide an asymptotic test.

In our paper we first apply the aforementioned test of ? in each series to take the estimated variance change point locations, in combination with a binary segmentation algorithm applied in a similar way as in Galeano and Weid (2014). Applied on the two time series, the test yields a variance change-point at January 2, 2008 for the Greek bond yields series and at November 5, 2007 for the German bond yield series. Next, the data are split into the interval before the change-point(including the point) and after so as to test in both segments again. If the smallest of the two p-values is smaller than $1-0.95^{1/2}$, then a new change point is found at the argmax of the corresponding series, which is split at this point again. The procedure is repeated until no further change points can be found. The variance breaks found are considered as fixed at the following the correlation break test.

The Demetrescu and Wied (2019) test for constant correlation employs a multivariate CUSUM statistic given by,

$$\hat{Q}_n = \max_{1 \leq j \leq n} \frac{j}{\sqrt{n}} \sqrt{\left(\hat{\mathbf{S}}_j - \hat{\mathbf{S}}_n \right)' \hat{\Omega}^{-1} \left(\hat{\mathbf{S}}_j - \hat{\mathbf{S}}_n \right)} \text{ with } \hat{\mathbf{S}}_j = \frac{1}{j} \sum_{t=1}^j \mathbf{g}(\hat{Z}_t)$$

in combination with the robustified constant-correlation test based on \hat{Q}_n ,

$$g(\hat{Z}_{t1}, \hat{Z}_{t2}) = \hat{Z}_{t1} \hat{Z}_{t2} \text{ and } \hat{Z}_{ti} = \frac{X_{ti} - \hat{\mu}_{1,i}(1-D_{t,\lambda}) - \hat{\mu}_{2,i}D_{t,\lambda}}{\sqrt{\hat{\sigma}_{i,1}^2(1-D_{t,\lambda}) + \hat{\sigma}_{i,2}^2D_{t,\lambda}}}$$

Under some assumptions it holds that as $n \rightarrow \infty$,

$$\hat{Q}_n \xrightarrow{d} \sup_{s \in [0,1]} \sqrt{\left(\hat{\mathbf{\Gamma}}(s) - s\hat{\mathbf{\Gamma}}(1) \right)' \left(\hat{\mathbf{\Gamma}}(s) - s\hat{\mathbf{\Gamma}}(1) \right)} \text{ given that } \hat{\lambda} = \lambda_0 + O_p(n^{-1}), \text{ where}$$

$$\hat{\mathbf{\Gamma}}(s) = \mathbf{\Gamma}(s) + \Omega^{-\frac{1}{2}} \tau_{\lambda_0}(s) \left(\Pi'_{\lambda_0}(1) W \Pi_{\lambda_0}(1) \right)^{-1} \Pi'_{\lambda_0}(1) W \sum_{\lambda_0}^{1/2} \Theta(1)$$

The above test allows for a two regime model in the variances. Once we eliminate variance breaks stationarity is desirable under the null hypothesis of constant correlations. To perform the test we rely on a bootstrap approximation with 4,999 bootstrap replications.

A.2 Local Gaussian approximation and local correlation

Assume two return series with observed values $\{(X_t, Z_t) \ t = 1, \dots, T\}$. The correlation between them conditionally on being in the region S can be written as:

$$\hat{\rho}_c(S) = \frac{\sum_{(X_t, Z_t)} (X_t - \hat{\mu}_{X,c}) (Z_t - \hat{\mu}_{Z,c})}{\left[\sum_{(X_t, Z_t) \in S} (X_t - \hat{\mu}_{X,c})^2 \right]^{\frac{1}{2}} \left[\sum_{(X_t, Z_t) \in S} (Z_t - \hat{\mu}_{Z,c})^2 \right]^{\frac{1}{2}}} \quad (4)$$

where $\hat{\mu}_{X,c} = \frac{1}{n_S} \sum_{(X_t, Z_t) \in S} X_t$ and $\hat{\mu}_{Z,c} = \frac{1}{n_S} \sum_{(X_t, Z_t) \in S} Z_t$ with n_S being the number of pairs with $(X_t, Z_t) \in S$. For ergodic series $\{X_t, Z_t\}$ as $n_S \rightarrow \infty$, $\hat{\rho}_c(S)$ would converge to $\hat{\rho}_c(S) = \text{corr}(X, Z \mid (X, Z) \in S)$.

A general bivariate density f for the variables (X, Z) would be fitted locally in a neighborhood of each point $y = (x, z)$, by a bivariate Gaussian density,

$$\phi(u, \theta(y)) = \frac{1}{2\pi\sigma_1(y)\sigma_2(y)} \times \exp \left\{ -\frac{1}{2(1-\rho^2(y))} \left[\left(\frac{u_1 - \mu_1(y)}{\sigma_1(y)} \right)^2 + \left(\frac{u_2 - \mu_2(y)}{\sigma_2(y)} \right)^2 - 2\rho(y) \left(\frac{u_1 - \mu_1(y)}{\sigma_1(y)} \right) \left(\frac{u_2 - \mu_2(y)}{\sigma_2(y)} \right) \right] \right\} \quad (5)$$

where $u = (u_1, u_2)^\top$ is the running variable in the Gaussian distribution and $\theta(y) = \phi(\mu_1(y), \mu_2(y), \sigma_1(y), \sigma_2(y), \rho(y))$, with $\mu_i(y)$, $i = 1, 2$ the local means, $\sigma_i(y)$, $i = 1, 2$ the local standard deviation and $\rho(y)$, the local correlation at the point $y = (x, z)$. The population values of the local parameters $\theta_b(y) = \theta(y)$ are obtained by minimizing the local penalty function,²⁵

$$q = \int K_b(u - y) [\phi(u, \theta(y)) - \log \phi(u, \theta(y)) f(u)] du \quad (6)$$

where $K_b = (b_1 b_2)^{-1} K(b_1^{-1}(u_1 - y_1)) K(b_2^{-1}(u_2 - y_2))$ is a product kernel with bandwidth $b = (b_1, b_2)$, and the local Gaussian correlation $\rho_b(y) = \rho(y)$ is defined as the last element of the vector $\theta(y)$ that minimizes q . Moving to another point $y' = (x', z')$ of f another Gaussian $\phi(u, \theta(y'))$ is required to approximate f in a neighbourhood S' of y' . In this way f may be represented by a family of Gaussian bivariate densities as y varies and in each specific neighborhood of y , the local dependence properties are described by $\rho(y)$. The (local) dependence may be defined to be positive (negative) if $\rho(y) > 0$ ($\rho(y) < 0$). The bias of conditional correlation is accommodated since the same Gaussian f fits every point.

Given the observations $Y_i = (X_i, Z_i)$, $i = 1, \dots, n$ from f the corresponding estimates $\hat{\theta}(y)$ are obtained by maximizing the local log-likelihood function (see Hjort and Jones, 1996),

$$L(Y_1, \dots, Y_n, \theta_b(y)) = n^{-1} \sum_i K_b(Y_i - y) \log \phi(Y_i, \theta_b(y)) - \int K_b(u - y) \phi(u, \theta_b(y)) du \quad (7)$$

²⁵This type of penalty function q was used in Hjort and Jones (1996) for density estimation purposes and later by Tjøstheim and Hufthammer (2013) in the development of local Gaussian correlation. The former argue that q can be interpreted as a locally weighted Kullback-Leibler criterion for measuring the distance between $f(\cdot)$ and the chosen parametric distribution (in our case $\phi(\cdot, \theta(y))$).

For the local likelihood function (4) to be consistent with the penalty function q , the $\theta(y)$ is chosen to minimize q , such that it satisfies the following 5-dimensional set of equations:²⁶

$$\frac{\partial q}{\partial \theta_j} = \int K_b(u-y) \frac{\partial}{\partial \theta_j} \{\log(\phi(u, \theta(y)))\} [\phi(u, \theta(y)) - f(u)] du, \quad j = 1, \dots, 5 \quad (8)$$

Using the notation,

$$\gamma_j(\cdot, \theta) = \frac{\partial}{\partial \theta_j} \{\log \phi(\cdot, \theta)\}, \quad (9)$$

and assuming that $E\{K_b(Y_i - y) u_j(Y_i, \theta(y))\} < \infty$, the law of large numbers gives,

$$\begin{aligned} \frac{\partial L}{\partial \theta_j} &= n^{-1} \sum_i K_b(Y_i - y) \gamma_j(Y_i, \theta(y)) - \int K_b(u - y) \gamma_j(u, \theta(y)) \phi(u, \theta(y)) du \\ &\rightarrow \int K_b(u - y) \gamma_j(u, \theta(y)) [f(u) - \phi(u, \theta(y))] du = -\frac{\partial q}{\partial \theta_j}, \quad j = 1, \dots, 5 \end{aligned} \quad (10)$$

as $n \rightarrow \infty$, we see that (7) can be identified with (5). Also note that as $b \rightarrow \infty$ (4) reduces to the ordinary log-likelihood for a Gaussian distribution ϕ plus a constant, and hence $\rho(y)$ reduces to the ordinary global Gaussian correlation. For more details about the local $b_i \rightarrow 0, i = 1, 2$ and estimation of standard errors, we refer to [Tjøstheim and Hufthammer \(2013\)](#). The numerical maximization of the local likelihood (4) leads to local likelihood estimates $\theta_{n,b}(y)$, including estimates $\rho_{n,b}(y)$ of the local correlation. It is shown in [Tjøstheim and Hufthammer \(2013\)](#) that under relatively weak regularity conditions $\theta_{n,b}(y) \rightarrow \theta_b(y)$ for b fixed, and $\theta_{n,b}(y) \rightarrow \theta(y)$ almost surely for $b = b_n$ tending to zero.²⁷

[Berentsen and Tjøstheim \(2014, Section 3.4\)](#) argue that the bandwidth choice depends on the nature of the question. To investigate the local dependence structure in the dataset can be quite informative to compute several bandwidths to obtain information about the dependence structure on different scales of locality. In some cases it would be beneficial to have a data-driven bandwidth choice similar to a bandwidth choice for density kernel estimation. In our empirical analysis, we employ two methods for bandwidth selection, the normal-reference rule-of-thumb as in [Støve et al. \(2014\)](#) and the methodology of likelihood cross-validation proposed by [Hall et al. \(2004\)](#).²⁸ Since both approaches provide qualitatively similar results, we present the bandwidth choice based on the normal-reference rule-of-thumb.²⁹

²⁶More details concerning the local Gaussian theory can be found in [Berentsen and Tjøstheim \(2014\)](#) and [Tjøstheim and Hufthammer \(2013\)](#).

²⁷The R-package 'localgauss' has been used for estimating $\rho_{n,b}(y)$. An introduction to the R package 'localgauss' for estimation and visualization of local dependence is available in [Berentsen et al. \(2014\)](#).

²⁸The R-package 'MASS' and 'np' have been used for rule-of-thumb and the data-driven bandwidth selection methods respectively.

²⁹For further discussion regarding bandwidth selection, see [Tjøstheim and Hufthammer \(2013\)](#).

A.3 Kalman filter model estimations



Figure 5: Correlations between bond yields to the six factors generated by the Kalman filter model.

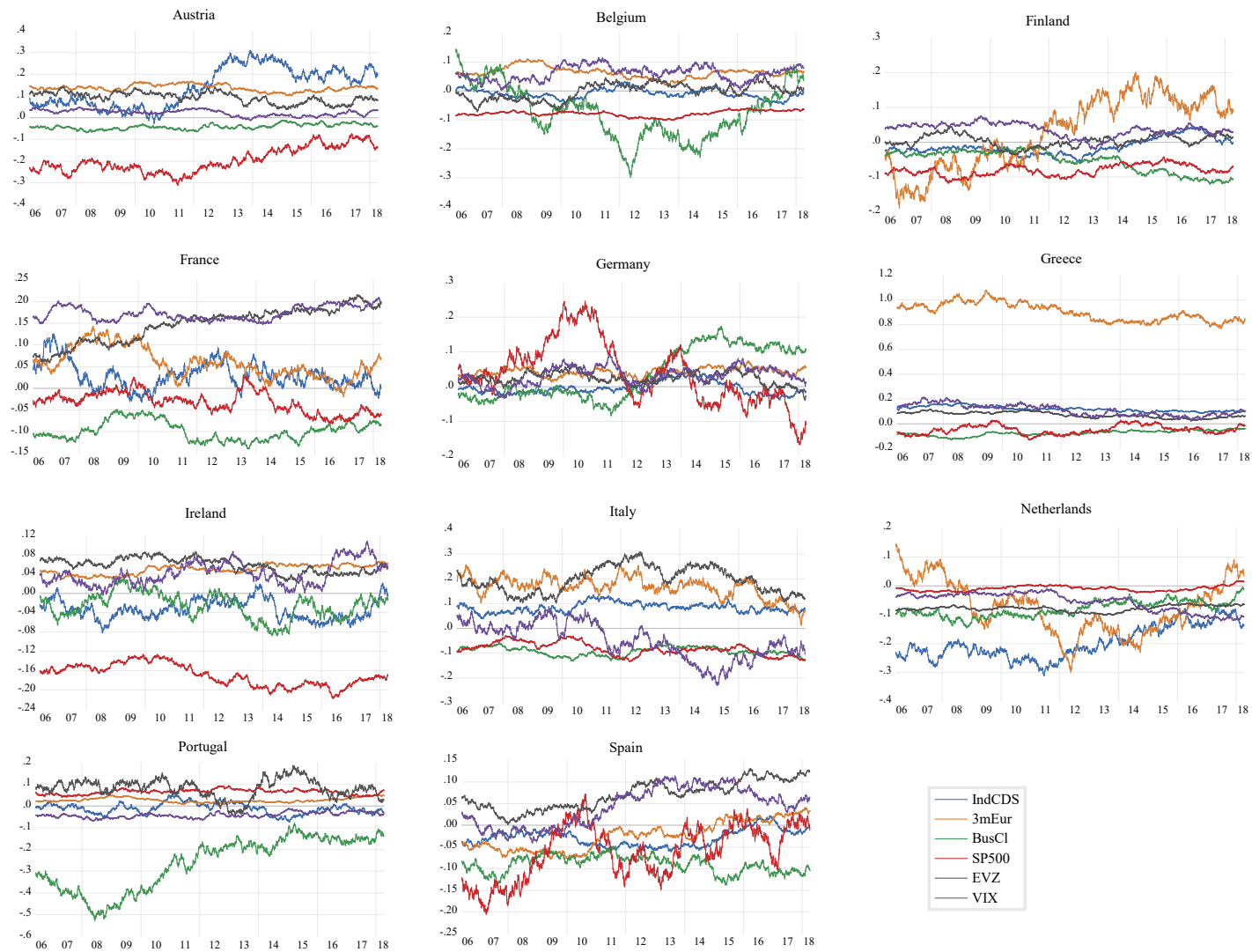


Figure 6: Correlations between CDS spreads to the six factors generated by the Kalman filter model.

A.4 Bootstrap test for contagion using the original series

Table 8: This table shows p -values from the bootstrap test for contagion according to the time dates of correlation change-points estimated with the algorithm of Demetrescu and Wied (2019) based on the original series. The null hypothesis indicates no contagion between the GIIPS (GR, IE, IT, PT, ES) bonds and the European bond (Panel A) and CDS (Panel B) markets. The bootstrap test for contagion is based on 1000 replications. The sample period starts on January 2, 2006 and ends on March 8, 2012 for Greece and on April 5, 2018 for the rest EMU South countries.

	GR (1)	Date	IE (2)	Date	IT (3)	Date	PT (4)	Date	ES (5)	Date
<i>Panel A : Bond Markets</i>										
AT	0.000	13/10/2008	0.020	30/1/2009	0.029	20/7/2011	0.031	9/10/2009	0.000	8/12/2009
	0.037	30/11/2009	0.008	25/3/2010			0.012	22/3/2010	0.230	11/6/2012
BE	0.003	6/8/2009	0.007	3/3/2010			0.451	12/4/2011		
DE	0.010	16/10/2008	0.011	25/2/2009	0.000	21/10/2008	0.013	26/9/2008	0.042	30/12/2008
	0.009	25/11/2009	0.009	2/12/2010	0.046	27/4/2010	0.019	26/11/2009	0.011	3/8/2011
					0.965	7/7/2011				
ES	0.002	23/12/2009	0.045	1/12/2010	0.000	21/7/2011	0.013	24/3/2011		
					0.587	1/7/2015	0.591	2/7/2015		
FI	0.000	26/10/2009	0.931	2/12/2010	0.016	2/7/2008	0.011	22/4/2011	0.000	3/12/2009
			0.721	7/3/2014	0.019	19/5/2010			0.063	20/7/2011
FR	0.004	16/11/2009	0.010	1/4/2010	0.519	20/7/2011	0.029	29/1/2010	0.032	14/7/2009
			0.548	1/12/2010			0.009	21/4/2011	0.512	18/7/2011
GR			0.021	9/12/2009	0.048	15/10/2008	0.030	17/2/2009	0.044	23/12/2009
					0.000	23/11/2009				
IE	0.034	9/12/2009					0.980	30/11/2010	0.005	1/12/2010
IT	0.000	15/10/2008					0.033	10/2/2010	0.012	20/7/2011
	0.012	23/11/2009					0.045	20/5/2011	0.425	29/6/2015
NL	0.007	14/10/2009	0.007	23/1/2009	0.008	24/6/2011	0.040	30/9/2009	0.009	21/6/2011
PT	0.048	17/12/2009	0.012	30/11/2010	0.003	10/2/2010			0.027	24/3/2011
					0.042	20/5/2011			0.582	2/7/2015
<i>Panel B : CDS Markets</i>										
AT	0.045	2/11/2009	0.045	15/11/2010	0.011	18/7/2011	0.000	28/3/2011	0.000	2/8/2011
BE	0.010	5/11/2009	0.005	26/11/2010	0.042	21/7/2011	0.000	1/4/2011	0.007	3/8/2011
	0.011	21/4/2010								
DE	0.021	6/11/2009	0.045	23/11/2010	0.027	21/7/2011	0.004	18/4/2011	0.035	20/7/2011
	0.452	20/04/2010								
ES	0.003	9/11/2009	0.030	3/12/2010	0.010	20/7/2011	0.027	11/4/2011		
	0.008	22/4/2010	0.398	22/7/2011						
FI	0.005	12/11/2009			0.009	2/8/2011	0.041	5/5/2011	0.012	3/8/2011
FR	0.002	5/11/2009	0.001	12/11/2010	0.019	20/7/2011	0.017	18/4/2011	0.000	18/7/2011
	0.712	22/04/2010								
GR			0.422	3/1/2008	0.490	4/11/2009	0.873	18/11/2009	0.730	19/11/2009
			0.091	14/11/2009	0.731	21/4/2010	0.554	22/4/2010	0.497	21/4/2010
IE	0.249	3/1/2008			0.457	3/12/2010	0.354	29/11/2010	0.831	2/12/2010
	0.013	12/11/2009			0.000	15/7/2011	0.045	28/3/2011	0.010	25/7/2011
IT	0.000	6/11/2009	0.012	3/12/2010			0.014	12/4/2011	0.002	20/7/2011
	0.000	20/4/2010	0.539	15/7/2011						
NL	0.072	5/11/2009	0.000	10/12/2010	0.019	3/8/2011	0.037	5/5/2011	0.009	19/7/2011
PT	0.000	17/11/2009	0.023	30/11/2010	0.018	12/4/2011	0.029	5/5/2011	0.034	11/4/2011
	0.005	21/4/2010	0.348	30/3/2011						

Table 9: This table shows p -values from the bootstrap test for contagion according to the time dates of correlation change-points estimated with the algorithm of Demetrescu and Wied (2019) based on the original series. The null hypothesis indicates no contagion between the GIIPS (GR, IE, IT, PT, ES) CDS and the European bond (Panel A) and CDS (Panel B) markets. The bootstrap test for contagion is based on 1000 replications. The sample period starts on January 2, 2006 and ends on March 8, 2012 for Greece and on April 5, 2018 for the rest EMU South countries.

	GR (1)	Date	IE (2)	Date	IT (3)	Date	PT (4)	Date	ES (5)	Date
<i>Panel A : Bond Markets</i>										
AT	0.233	2/11/2009	0.148	13/10/2011	0.912	31/10/2011			0.089	25/10/2011
	0.411	4/11/2011							0.152	21/11/2011
BE	0.122	3/11/2011	0.187	20/10/2011			0.137	30/11/2010		
DE	0.872	22/4/2010	0.341	12/10/2011	0.012	15/7/2010	0.005	22/7/2010	0.000	29/7/2010
					0.307	25/10/2011	0.204	15/5/2013	0.742	26/2/2010
ES	0.471	10/10/2011	0.561	5/10/2011	0.745	9/11/2011	0.885	6/4/2010		
			0.996	7/8/2015	0.991	22/6/2015				
FI	0.311	13/11/2009	0.962	9/1/2008	0.015	1/9/2010	0.022	2/9/2010	0.004	8/9/2010
	0.411	15/11/2011	0.981	23/4/2010	0.411	21/10/2013	0.962	15/5/2013	0.639	15/5/2013
FR	0.452	21/4/2010	0.948	17/7/2008	0.007	22/7/2010	0.032	23/7/2010	0.030	5/8/2010
			0.127	10/10/2011	0.412	18/11/2011	0.041	4/8/2010	0.518	8/11/2011
GR			0.712	9/11/2011	0.301	10/11/2011	0.541	9/11/2011	0.738	8/11/2011
IE	0.497	9/11/2011					0.430	4/11/2011	0.351	15/11/2011
					0.997	17/11/2011			0.619	7/8/2015
IT	0.230	10/11/2011	0.912	12/10/2010			0.631	14/11/2011	0.350	9/11/2011
			0.259	10/11/2011					0.341	15/6/2015
NL	0.981	5/11/2009	0.965	1/7/2008	0.025	15/7/2010	0.001	28/7/2010	0.047	1/7/2010
	0.757	7/11/2011	0.438	3/8/2010	0.659	26/3/2013			0.589	26/3/2013
PT	0.329	9/11/2011	0.341	18/11/2011	0.338	14/11/2011			0.589	6/4/2010
<i>Panel B : CDS Markets</i>										
AT			0.812	26/1/2012	0.001	23/11/2011	0.722	15/9/2008	0.458	15/9/2008
							0.716	2/2/2012	0.021	3/1/2012
BE			0.291	25/1/2012	0.005	10/1/2012	0.370	12/1/2012	0.033	9/1/2012
DE			0.489	10/1/2012	0.000	18/1/2012	0.341	18/9/2008	0.019	19/1/2012
							0.992	13/1/2012		
ES	0.003	16/11/2011			0.000	23/11/2011	0.579	1/11/2011		
FI	0.411	25/9/2008	0.899	19/1/2012	0.013	29/11/2011	0.511	3/1/2012	0.010	14/11/2011
FR					0.005	10/1/2012	0.420	18/11/2009	0.015	4/1/2012
					0.966	1/2/2013	0.847	13/1/2012	0.968	1/2/2013
GR			0.741	10/11/2011					0.020	18/11/2011
IE	0.015	9/11/2011			0.003	16/11/2011	0.482	5/1/2012		
					0.958	31/8/2015				
IT	0.983	14/9/2010	0.712	16/11/2011			0.581	15/11/2012	0.000	22/11/2011
			0.349	31/8/2015			0.918	5/9/2012		
NL	0.401	19/9/2008	0.311	12/12/2012	0.009	18/1/2012	0.558	10/9/2008	0.421	16/9/2008
	0.391	23/9/2010					0.820	12/1/2012	0.020	16/1/2012
PT			0.571	12/1/2012	0.004	14/11/2012			0.008	8/11/2011