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The Dynamic Progression of the Redenomination and Sovereign Risk in the Price Discovery Process of Italian Banks' CDSs

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Abstract

The recent global financial crisis and the subsequent sovereign debt crisis of the Eurozone peripheral countries have generated historic levels of volatility and instability in the financial markets. In particular, during the sovereign debt crisis market operators have begun to focus on the so-called "redenomination risk", that is the hypothesis of exit from the EMU (Euro Monetary Union) by one or more countries and the consequent redenomination of their debt in the past national currency. This type of risk constitutes a form of additional credit risk premium due to expected systemic failure of the Eurozone. The effects of the economic-financial crisis, the weak economic growth and the political instability that have characterized especially the Italian system in recent years provide the ideal starting point to analyze the evolution of the redenomination risk in the pricing process of the Italian banks' CDSs (Credit Default Swaps).

The contribution of this work is to evaluate the dynamic evolution of sovereign and redenomination risk in the price discovery process of the Italian banks' CDS spreads (or premia) by using rolling window regressions. Results show that redenomination risk explains a great part of the variance in the CDS spreads during periods of financial distress. The sovereign risk component explains a large part of the variance for almost the entire considered period.

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1. Introduction

The recent global financial crisis and the subsequent sovereign debt crisis of the peripheral countries of the Eurozone (the so-called "*PIIGS*") have generated historic levels of volatility and instability in the financial markets, putting a strain on the entire European institutional structure. The following uncertainty has made it interesting to observe the effects on the financial markets and on the national banking systems that have been directly involved due to exposure to sovereign debts (Li and Zinna, 2014). The 2008 global financial crisis, in fact, has made it necessary the action of many European States in the rescue of national banking systems. The bailout operations have consisted of transforming private debt into public debt, causing a strengthening of the link between State and baking systems. With the sovereign debt crisis, this link of interdependence has been further strengthened (so-called "*doom loop*" or "*deadly embrace*") (Farhi and Tirole, 2017) because of the banks' exposure to the credit risk of domestic sovereigns, especially as for Italian and Spanish banks⁵ (Li and Zinna, 2014).

In this context, the cross-border relations of national banking systems have increased exposure to non-domestic sovereign risk and strengthened the links between European sovereigns (Korte and Steffen, 2014). The links have become so strong that they have generated the fear that the failure of a State could cause the breakup of the Euro Area (Li and Zinna, 2014).

Speculation on the irreversibility of the Euro Area has induced market operators to put increasing attention to the redenomination risk, the risk that a monetary union country could redenominate its debt in the national currency. According to De Santis (2019), redenomination risk can be defined as: "the compensation demanded by market participants for the risk that a euro asset will be redenominated into a devalued legacy currency".

Although several bailouts that have avoided the bankruptcy of the peripheral countries of the Eurozone and have ensured the solidity of the Euro. Despite the ECB's (European Central Bank) interventions aimed at reassuring investors of the irreversibility of the single currency⁶ (Busetti and Cova, 2013), speculation on the irreversibility of Euro, measured by the redenomination risk, periodically reveals itself during periods of greater political-economic stress.

⁵ In order to avoid the eventuality of default of these two States, in December 2011, the respective Italian and Spanish banks purchased the domestic public debt using the liquidity made available by the ECB (European Central Bank) through the VLTRO operation (Very Long-Term Refinancing Operations), with which over one trillion Euro of resources were provided to the Euro Area banking systems (Cesaratto, 2016).

⁶ It's remarkable the speech of Mario Draghi, ECB's President, on the occasion of the Global Investment Conference on 26 July 2012, in which he stated: "*Within our mandate, the ECB is ready to do whatever it takes to preserve the euro. And believe me, it will be enough*".

The crucial point, from which our analysis starts, is that the redenomination risk has led to a rethinking of the CDS's characteristics. The 2003 definitions did not contemplate the possibility that a country could leave the Euro Area and redenominate its debt in the local currency. Precisely, the 2003 definitions defined the currencies in which the redenomination was allowed, although this did not trigger a credit event, and therefore the CDS (Kremens, 2019). However, following the Greek default, the circumstance that a country could exit from the Eurozone was no longer unthinkable and, due to investors pressure, in 2014 the ISDA (International Swaps and Derivatives Association) introduced a series of new standards for CDS contracts in order to take into account the possibility of redenomination and the consequent losses (ISDA, 2014). The 2014 definitions include the CACs (Collective Action Clauses), which establish that any sovereign debt restructuring action, including redenomination, must be approved by at least 75% of investors (Cesaratto, 2015). It was precisely their activation by the Greek government that led the ISDA (International Swaps and Derivatives Association) to declare on March 9, 2012 the credit restructuring event for CDS referring to the sovereign debt of Greece (De Santis, 2019).

The Italian case offers an interesting opportunity to analyze the impact of these dynamics on the Italian banks' CDS spreads. In fact, it's great to underline that, despite Italy is the founding country, third economy and second manufacturing in the EU (European Union), it has been hit harshly by the sovereign debt crisis due to the high debt-to-GDP ratio, but it has also been characterized by weak economic growth and political instability in recent years. In addition to that, as a member country of the Eurozone of the so-called "*PIIGS*", it is generally exposed with greater intensity to volatility and speculation during periods of financial stress.

The contribution of this work is to evaluate the dynamic evolution of sovereign risk and redenomination risk in the price discovery process of the Italian bank's CDS spreads for 2008-2020, using a "rolling window regressions" approach.

The redenomination risk is measured by the differential between the CDS contracts signed under the 2014 Definitions and those defined according to the 2003 Definitions. The sovereign risk is measured by the differential between the ten-year Italian government bond (BTP) yield and the respective German government bond (Bund) yield. Results show that the redenomination risk represents a key variable during the most relevant periods of political-financial stress, while sovereign risk explains much of the variance in CDS spreads for almost the entire period considered. The structure of the document is the following: Section 2 reports the literature review, Section 3 describes the model and data, Section 4 reports the results of the analysis and Section 5 reports the economic discussion and conclusions.

2. Literature Review

In a recent literature, several important papers analyze the determinants of banks' CDS spreads and the links with sovereign risk. Below are presented some of the important studies. In particular, the first two papers are focused on the banks' CDS premia price discovery process.

Chiaramonte and Casu (2010) analyze the determinants of CDS spreads using specific balance sheet ratios and evaluate their possible use as proxy of bank default risk. The analysis focuses on three sub-periods: pre-crisis (January 2005 - June 2007), crisis (July 2007 - March 2009) and low crisis (April 2009 - June 2011). The study shows that the pre-crisis phase reflects the risk measured by the balance sheet ratios, while Tier1 ratio and leverage are non-significant in all three sub-periods, and liquidity ratios are significant during the crisis period.

The balance sheet ratios are also used in the analysis of *Samaniego-Medina et al.* (2016), who investigate the determinants of CDS spreads for a sample of 45 European banks during the period 2004-2010, using not only balance sheet but also market ratios. The authors highlight that the market variables have a greater explanatory capacity during the crisis than in the pre-crisis period.

Unlike these two studies, the analysis of *Avino and Cotter* (2014) is more focused on the interconnectedness of bank and sovereign CDS markets during the period before the financial crisis started in mid-2007. Their research shows that spreads on sovereign CDS incorporate more quickly the evolution of expectations about the default probability of European banks than corresponding bank CDS spreads in times of crisis.

In recent years, due to the events that have characterized the financial markets, the literature has begun to investigate in detail the redenomination risk. However, we can identify two streams: the first, characterized by a large part of the papers, focuses on the evaluation and the measurement of redenomination risk in the price discovery process of sovereign CDS premia. Instead, the second stream, characterized by few works, is aimed at reaching the same objective with reference to the price discovery process of bank CDS premia.

The main reference work of the first stream is that of De Santis (2019). The main aim of his paper is to demonstrate how the redenomination risk shocks are able to influence sovereign yield spreads. In particular, the author uses a dynamic countryspecific measure of the redenomination risk for the countries of the Euro Area, defined as the "quanto CDS", calculated as the difference between the CDS spreads on bonds denomiated in US dollar and the CDS spreads on equivalent bonds denominated in Euro. He uses the difference between the quanto CDS for a member country and for a benchmark country (eg Germany). By focusing on Italy, Spain and France and using Germany as a benchmark for the Eurozone sovereign debt market, De Santis demonstrates that the redenomination risk was able to influence sovereign yield spreads, especially of Italy and Spain.

In addition, Kremens (2019) investigates the effects of the currency redenomination risk in the price process of sovereign yields, taking into consideration France, Italy

and Germany. The measure of the redenomination risk is constructed using the pricing difference between the CDS contracts signed under the 2014 Definitions and the CDS contracts signed under the 2003 Definitions (so-called ISDA basis) (Nolan, 2018). The author tries to explain how the negative effects of an Italian exit from the Euro Area can be sufficiently controlled, while a French exit from the monetary union could have a domino effect on the rest of the Eurozone.

The main reference paper of the second stream is that of Anelli et al. (2020). The authors analyze the price discovery process of the Italian banks' CDS spreads by integrating the variables of the Merton model (1974). In particular, the authors include in their model the quanto CDS as measure of the redenomination risk, as in De Santis (2019). The main aim of the paper is to evaluate the evolution of the explanatory power of the redenomination risk and the classic variables of the Merton model in the price discovery process of the Italian banks' CDS spreads during the most volatile phases of the recent financial crisis. In particular, the authors split the analysis on three specific periods: the financial crisis (August 2008 - October 2009), the sovereign debt crisis (October 2009 - July 2012) and the phase of confrontation between the Italian government and the European Union (March 2018 - September 2018). The authors firstly studied the lead-lag structure between the bank and sovereign CDS series, and then focused on the evaluation of the determinants of bank CDS spreads. Their work demonstrates, in line with the results of this paper, that the redenomination risk played a decisive role in the price evolution of CDS spreads during the sovereign debt crisis and especially in 2018.

Our paper is more related to the second stream and in particular to the last paper. This document contributes to the literature by analyzing the dynamic evolution of the redenomination risk (measured by the differential between the CDS contracts signed under the 2014 definitions and those defined according to the 2003 definitions) and of the sovereign risk (measured by the differential between the tenyear Italian government bond (BTP) yield and the respective German government bond (Bund) yield) in the price discovery process of the main Italian banks' CDS spreads from 2008 to 2020 using a "rolling window regressions" model.

The case of Italy is particularly interesting for this analysis, as it is a member country of the Eurozone of the so-called "*PIIGS*", so it is generally exposed with greater intensity to volatility and speculation during periods of financial stress. Moreover, it's a country characterized by a high debt-to-GDP ratio, low economic growth and the presence of declared anti-European political parties with growing electoral consensus. Unlike what Kremens (2019) suggests, the idea of this work, in line with the study of Anelli et al. (2020), is that a possible exit from the Euro Area of Italy, that is the founding country, the third economy and the second manufacturing of the EU (European Union), could deeply undermine the entire European institutional structure and definitively question the irreversibility of monetary union.

3. Model and Data description

The main objective of this paper is to evaluate the dynamic evolution of the sovereign and redenomination risk in the price discovery process of the major Italian banking groups' CDS by using rolling window regressions approach. According to this approach, we take into account the size of each sliding window starting from the overall sample size. The sliding window (equal to one-year observations) represents the number of observations for each subsample. We estimate the whole model using each subsample from 2008 to 2020 by means of rolling OLS regressions.

The analysis is performed on the Italian banks' CDS spreads (*Italian banks proxy CDS SR 5y D14* or *IBP*), using as proxy of Italian banking system the weighted average values of the CDS spreads of the most capitalized banking groups (Intesa San Paolo, Unicredit and Monte dei Paschi di Siena⁷). Specifically, each series respectively includes the *five-years senior (modified-modified restructuring) CDS* contracts⁸ of Intesa San Paolo, Unicredit and Monte dei Paschi di Siena weighted by their market capitalization⁹, for the period 2008-20, using daily Euro denominated data from Bloomberg (3132 observations). Figure 1 shows the overtime movements of the series referring to dependent variables.

⁷ Intesa San Paolo and Unicredit are the largest Italian banks in terms of market capitalization and total assets (Sirletti and Salzano, 2018). In 2018, they accounted for about 45% of the total assets of the Italian banking system (Anelli et al., 2020). Therefore, the idea behind the approximation is expressed by the concept of too-big-to-fail: if one of these groups goes into crisis, it would be logical to think that the entire Italian banking system would be involved. The choice of sample also depends on the fact that the mentioned banking groups represent the "specialists" in Government Securities. As reported by the Ministry of Economy and Finances (MEF, 2011): "Dealers that are market makers (primary dealers) have obligations as to subscriptions in government bond auctions and trading volumes on the secondary market. These give rise to some privileges, among which is the right to exclusive participation in supplementary placements of the issuance auctions".

⁸ In the analysis we chose CDS contracts with a maturity of five years because they are the most liquid, so the data are easily (Kremens, 2019).

⁹According to Bloomberg data, Intesa San Paolo's market capitalization is \notin 41.24 bn., \notin 29.30 bn. that of Unicredit and \notin 1.58 bn. the capitalization of Monte dei Paschi di Siena. Therefore, only considering Intesa San Paolo and Unicredit, the market capitalization is approximately \notin 70 bn., about 98% of the total capitalization.

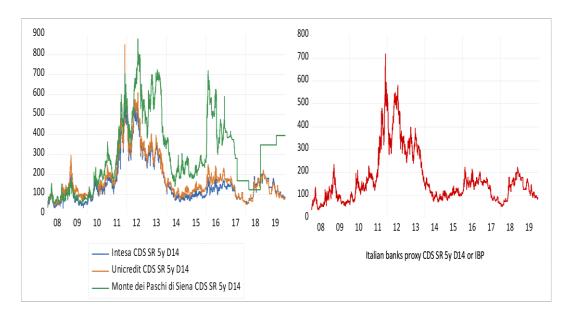


Figure 1: Intesa San Paolo CDS SR 5y D14, Unicredit CDS SR 5y D14, Monte dei Paschi CDS SR 5y D14 and Italian banks proxy CDS SR 5y D14 or IBP series: period 2008-2020.

Source: authors' calculations in Eviews 11 based on Bloomberg data.

We consider as a measure of redenomination risk (namely *Italy Redenomination Risk*) the differential between the CDS spreads of five-years Italian government bond CDS contracts according to the ISDA 2014 definitions and the CDS spreads of five-years Italian government bond CDS contracts according to the ISDA 2003 definitions, for the period 2008-20, using daily data from Bloomberg (3132 observations).

In the model we also include as a masure of sovereign risk for Italy (namely *Spread It-Ge*) the differential between the ten-years Italian government bond (BTP) yield and the respective German government bond (Bund) yield, for the period 2008-20, using daily data from Bloomberh (3132 observations). Figure 2 shows the overtime movements of the series referring to independent variables.

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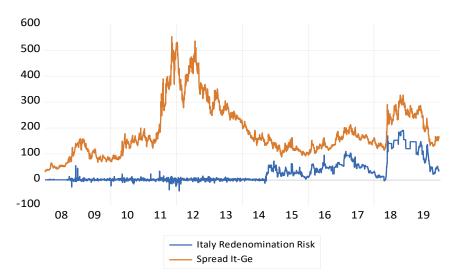


Figure 2: Italy Redenomination Risk and Spread It-Ge series: period 2008-2020.

Source: authors' calculations in Eviews 11 based on Bloomberg data.

The basic econometric model is defined by the following equations:

$$Intesa_t = \beta_{10} + \beta_{1i}X_{1t} + \alpha_{1i}X_{2t} + u_{1t}$$
(1)

$$Unicredit_{t} = \beta_{20} + \beta_{2i}X_{1t} + \alpha_{2i}X_{2t} + u_{2t}$$
(2)

$$Monte \ dei \ Paschi_t = \beta_{30} + \beta_{3i} X_{1t} + \alpha_{3i} X_{2t} + u_{3t}$$
(3)

$$IBP_{t} = \beta_{40} + \beta_{4i}X_{1t} + \alpha_{4i}X_{2t} + u_{4t}$$
(4)

where:

- t = 1,2,3,...,(T-1),T is the time horizon;
- i = 1, 2, 3, ..., n is the observations' number;
- Intesa_t, Unicredit_t, Monte Dei Paschi_t and IBP_t are, respectively, Intesa San Paolo CDS SR 5y D14 at time t, Unicredit CDS SR 5y D14 at time t, Monte dei Paschi CDS SR 5y D14 at time t and Italian banks proxy CDS SR 5y D14 at time t;
- β_{10} , β_{20} , β_{30} and β_{40} are, respectively, the *constant terms* of the equation (1), (2), (3) and (4);
- β_{1i} , β_{2i} , β_{3i} and β_{4i} are, respectively, *the coefficients* of the first regressor of the equation (1), (2), (3) and (4);

- α_{1i} , α_{2i} , α_{3i} and α_{4i} are, respectively, *the coefficients* of the second regressor of the equation (1), (2), (3) and (4);
- X_{1t} is the *first regressor* and represents *Italy Redenomination Risk* at time *t*;
- X_{2t} is the second regressor and represents Spread It-Ge at time t;
- u_{1t} , u_{2t} , u_{3t} and u_{4t} are, respectively, the *error terms* of the equation (1), (2), (3), and (4).

The model can be also represented through a first difference transformation¹⁰. This transformation is usually made to solve the non-stationarity problem, typical of financial data. To overcome this problem and avoid misinterpretation of results of the regression analysis, the dependent and independent variables have been transformed in first differences. The performed model, therefore, is the following:

$$\Delta Intesa_t = \beta_{10} + \beta_{1i} \Delta X_{1t} + \alpha_{1i} \Delta X_{2t} + u_{1t}$$
(5)

$$\Delta Unicredit_t = \beta_{20} + \beta_{2i} \Delta X_{1t} + \alpha_{2i} \Delta X_{2t} + u_{2t}$$
(6)

$$\Delta Monte \ dei \ Paschi_t = \beta_{30} + \beta_{3i} \Delta X_{1t} + \alpha_{3i} \Delta X_{2t} + u_{3t}$$
(7)

$$\Delta IBP_t = \beta_{40} + \beta_{4i} \Delta X_{1t} + \alpha_{4i} \Delta X_{2t} + u_{4t}$$
where:
$$(8)$$

- $\Delta Intesa_t$, $\Delta Unicredit_t$, $\Delta Monte Dei Paschi_t$ and ΔIBP_t are, respectively, the first difference for Intesa San Paolo CDS SR 5y D14, Unicredit CDS SR 5y D14, Monte dei Paschi CDS SR 5y D14 and Italian banks proxy CDS SR 5y D14;
- ΔX_{1t} is the first difference for Italy Redenomination Risk and ΔX_{2t} is the first difference for Spread It-Ge.

¹⁰ The first difference of a time series is the variation of Y between the period t-1 and the period t. The first difference in formal terms is expressed in the following way: $\Delta Y_t = Y_t - Y_{t-1}$ (Stock and Watson, 2016).

4. Main Results

This section presents the results of the analysis. Results refer to the CDS spreads of the Italian banking system proxy (ΔIBP_t), while the results for each single bank of the sample are reported in Appendix A.2. The augmented Dickey-Fuller Test (see Appendix A.1) shows that series are stationary in each rolling window. The *Durbin-Watson*¹¹ statistic is shown in the results' table (see Table 1) and proves the absence of serial correlation. Table 1 shows the estimated rolling coefficients for the January 2008 - January 2020 period.

where:

• u_{t-1} represents the OLS residual at the time t-1

- Under the null hypothesis $\hat{\rho} = 0$, so DW = 2
- In presence of positive correlation $\hat{\rho} = 1$, so $DW \approx 0$
- In presence of negative correlation $\hat{\rho} = -1$, so $DW \approx 4$

¹¹ Formally, following Wooldridge (2010), the Durbin-Watson statistic is computed as follows: $DW = \frac{\sum_{t=2}^{n} (u_t - u_{t-1})^2}{\sum_{t=1}^{n} u_t^2}$

[•] u_t represents the OLS residual at time t

The null hypothesis is $H_0: \rho = 0$, which implies the absence of *serial correlation*, against the alternative $H_1: \rho \neq 0$, which implies the presence of *serial correlation* (Palomba, 2018). Knowing that, according to a simple relation $DW = 2(1 - \hat{\rho})$, we have:

Period January 2008-December 2008 (261 Obs.)					
Variable	Coefficient	Std. Error	t-Statistic	Prob.	
βο	0.060827	0.411635	0.147769	0.8826	
Δ (Italy Red. Risk)	-0.020212	0.059040	-0.342339	0.7324	
Δ (Spread It-Ge)	0.571365***	0.130568	4.376001	0.0000	
R ²		0.0692			
Adj. R ²		0.0620			
Durbin-Watson Stat.		1.8322	221		
	anuary 2009-D	ecember 2009	(261 Obs.)		
Variable	Coefficient	Std. Error	t-Statistic	Prob.	
βο	-0.060636	0.468906	-0.129315	0.8972	
Δ (Italy Red. Risk)	0.107017	0.155358	0.688842	0.4915	
Δ (Spread It-Ge)	0.400421***	0.107409	3.728002	0.0002	
R ²		0.0529	986		
Adj. R ²		0.0456	545		
Durbin-Watson Stat.		2.3824	139		
Period J	anuary 2010-D	ecember 2010	(261 Obs.)		
Variable	Coefficient		t-Statistic	Prob.	
βο	0.072799	0.413710	0.175966	0.8605	
Δ (Italy Red. Risk)	0.341940***	0.080123	4.267686	0.0000	
Δ (Spread It-Ge)	0.733912***	0.064327	11.40912	0.0000	
R ²		0.3901	61		
Adj. R ²		0.3854	134		
Durbin-Watson Stat.		2.5865	532		
Period J	anuary 2011-D	ecember 2011	(260 Obs.)		
Variable	Coefficient	Std. Error	t-Statistic	Prob.	
βo	0.470936	1.143037	0.412004	0.6807	
Δ (Italy Red. Risk)	0.125658	0.154435	0.813662	0.4166	
Δ (Spread It-Ge)	0.610463***	0.074083	8.240217	0.0000	
R ²		0.2228			
Adj. R ²		0.2167			
Durbin-Watson Stat.		2.079.7			
	anuary 2012-D		, ,		
Variable	Coefficient	Std. Error	t-Statistic	Prob.	
βο	-0.202094	0.738555	-0.273634	0.7846	
Δ (Italy Red. Risk)	-0.073379	0.114770	-0.639358	0.5232	
Δ (Spread It-Ge)	0.742077***	0.053825	13.78680	0.0000	
R ²		0.425705			
Adj. R ²	0.421253				
Durbin-Watson Stat.	2.241493				

Table 1: OLS estimates: period January 2008 – January 2020 (3131 Obs.)

Period January 2013-December 2013 (261 Obs.)							
Variable	Coefficient	Coefficient Std. Error t-Statistic Prob.					
βo	-0.339116	0.557757	-0.607999	0.5437			
Δ (Italy Red. Risk)	0.195840	0.147629	1.326570	0.1858			
Δ (Spread It-Ge)	0.591549***	0.069344	8.530636	0.0000			
R ²		0.2311	86				
Adj. R ²		0.2252	227				
Durbin-Watson Stat.		2.4248	345				
Period Ja	nuary 2014-D	ecember 2014	(261 Obs.)				
Variable	Coefficient	Std. Error	t-Statistic	Prob.			
βo	-0.069667	0.310606	-0.224295	0.8227			
Δ (Italy Red. Risk)	0.123090*	0.067696	1.818.293	0.0702			
Δ (Spread It-Ge)	0.289545***	0.060534	4.783176	0.0000			
R ²		0.0903	396				
Adj. R ²		0.0833	345				
Durbin-Watson Stat.		2.4895	532				
Period Jo	nuary 2015-D	ecember 2015	(261 Obs.)				
Variable	Coefficient	Std. Error	t-Statistic	Prob.			
βo	0.088744	0.275560	0.322048	0.7477			
Δ (Italy Red. Risk)	0.126272**	0.055469	2.276465	0.0236			
Δ (Spread It-Ge)	0.367238***	0.050609	7.256380	0.0000			
R ²		0.2313	364				
Adj. R ²		0.2254	106				
Durbin-Watson Stat.		2.5794	13				
Period Ja	nuary 2016-D	ecember 2016	6 (261 Obs.)				
Variable	Coefficient	Std. Error	t-Statistic	Prob.			
βo	-0.044992	0.394257	-0.114120	0.9092			
Δ (Italy Red. Risk)	0.211982***	0.072254	2.933850	0.0036			
Δ (Spread It-Ge)	0.800585***	0.088937	9.001679	0.0000			
R ²		0.3168	371				
Adj. R ²		0.3115	576				
Durbin-Watson Stat.		2.2322	209				
Period Jo	nuary 2017-D	ecember 2017	' (260 Obs.)				
Variable	Coefficient	Std. Error	t-Statistic	Prob.			
βo	-0.345190	0.248925	-1.386723	0.1667			
Δ (Italy Red. Risk)	0.140806***	0.052512	2.681380	0.0078			
Δ (Spread It-Ge)	0.137357**	0.060870	2.256569	0.0249			
R ²		0.0578		-			
Adj. R ²	0.050550						
Durbin-Watson Stat.	2.259061						

Period January 2018-December 2018 (261 Obs.)						
Variable	Coefficient	Std. Error	t-Statistic	Prob.		
βο	0.408706	0.429392	0.951825	0.3421		
Δ (Italy Red. Risk)	0.039496	0.052791	0.748147	0.4551		
Δ (Spread It-Ge)	0.082891	0.051592	1.606647	0.1094		
R ²		0.0197	704			
Adj. R ²	0.012104					
Durbin-Watson Stat.		2.054012				
Period J	anuary 2019	January 2020	(262 Obs.)			
Variable	Coefficient	Std. Error	t-Statistic	Prob.		
βο	-0.227905	0.225315	-1.011496	0.3127		
Δ (Italy Red. Risk)	0.215370***	0.045034	4.782336	0.0000		
Δ (Spread It-Ge)	0.179134*** 0.041540 4.312294 0.0000					
R ²	0.196956					
Adj. R ²	0.190755					
Durbin-Watson Stat.		2.010344				

Note: *** signals parameter significance at 1%, ** signals parameter significance at 5%,* signals parameter significance at 10%. Source: authors' calculations in Eviews 11 based on Bloomberg data.

Table 1 shows that the redenomination risk had a significant impact on the Italian banks' CDS reaching a peak during the begin of the sovereign debt crisis (2010) and started to increase its statistical frequency starting from the launch of the QE (Quantitative Easing) by the ECB. It is statistically significant during the referendum Brexit period (2015-2016), before the formation of the Italian anti-establishment government (2017) and during the first Conte's Italian government (2019), differently from the result obtained by Anelli et al. (2020) using the "quanto CDS" spreads. Regarding to the sovereign risk, Table 1 suggests that it had a significant role during the entire considered period.

During the 2008-2009 period, the relationship between the sovereign risk variable and the Italian banks' CDS spreads is positive, reporting a coefficient of about 0.57. This means that an increase of 100 basis points of the regressor corrisponded to an increase of about 57 basis points of the Italian banks' CDS spreads variation. The redenomination risk's coefficient is not statistically significant.

In the following year, the sovereign risk and the redenomination risk proxy resulted to be statistically significant at a 1% threshold. The sovereign risk reported a coefficient of about 0.73, while the redenomination risk reported a coefficient of about 0.34, reaching the peak. This confirms that market begun to perceive a growing redenomination risk in a context of financial distress caused by the sovereign debt crisis. However, at this phase, sovereign risk explains much of the variance in CDS spreads.

The sovereign risk is also statistically significant in the 2011-2014 period, while the

redenomination risk becomes again statistically significant only in 2014¹².

The sovereign risk and the redenomination risk proxy resulted to be statistically significant respectively at a 1% and 5% threshold in 2015, with the launch of the QE (Quantitative Easing) by the ECB (European Central Bank), rspectively reporting a coefficient of about 0.37 and 0.13.

In 2016, however, other international turmoils make the relationship between the sovereign risk and the redenomination risk with the CDS spreads positive: the *Brexit* referendum. Regarding the estimated coefficients, the sovereign risk and the redenomination risk proxy are statistically significant at a 1% threshold and the weight of both coefficients on the variance of CDS spreads increases compared to the previous year. The sovereign risk reported a coefficient of about 0.80, reaching the peak, while the redenomination risk reported a coefficient of about 0.21.

Due to the country's structural issues, in 2017 the estimated coefficients of the sovereign risk and the redenomination risk remain statistically significant respectively at a 5% and 1% threshold but the weight of both decreases, resulting about 0.14.

In the last phase (2018-2019), the markets perceived a growing redenomination risk for Italy. Similarly, the spread started growing again from mid-2018. During the January 2019-January 2020 period, the sovereign risk and the redenomination risk proxy resulted to be statistically significant at a 1% threshold. The sovereign risk reported a cofficient of about 0.18, while the redenomination risk reported a coefficient of about 0.21, reaching Brexit-level.

The Figure 3 suggests that the redenomination risk reached its maximum in 2010, then decreased and reached its minimum in 2015. The weight of the coefficient returned to growth in 2016, due to the Brexit referendum, and, after a gradual decrease, returned to Brexit-level in the last period. The trend of the sovereign risk coefficient, on the other hand, was much more linear before 2016, year in which it reached its peak, and soon after decreased, remaining at contained levels, in contrast to the redenomination risk.

¹² The redenomination risk coefficient is positive and statistically significant only for Monte dei Paschi di Siena (See Appendix A.2. Table A4). Although it represents the least capitalized bank among the three considered in the analysis, the precarious condition of the bank has generated the fear of a possible bankruptcy which, in turn, has led the markets to price a redenomination risk in the bank's CDS spreads. This has had an impact on the Italian banking system. However, the State's entry into the bank's capital as the largest shareholder (Telara, 2017) has caused the loss of significance for redenomination risk coefficient, since, as long as the State will be present in the bank's capital, there is no reason for insuring against banks's bankruptcy. For this reason, from now on, in the analysis of the impact of redenomination risk on the Italian banking system, the coefficient will be significant only for the other two main banking groups.

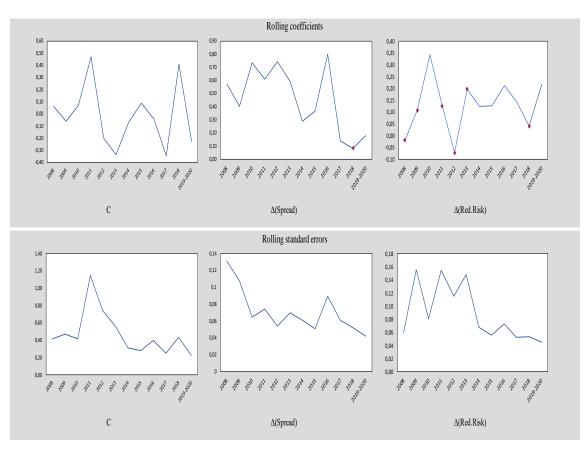


Figure 3: Rolling coefficient and standard errors.

Source: authors' calculations in Excel based on Bloomberg data. Note: red points signal the years of statistical non-significance.

5. Economic Discussion

Before the global financial crisis, the Eurozone countries were considered stable and economically reliable and the possibility that a country could declare its default and consequently exit from the Euro Area was far from the market perception. For this reason, the interest around the concepts of spread on sovereign yields and CDSs was almost limited. However, the market perception of a growing sovereign risk for the so-called "PIIGS" (Portugal, Ireland, Italy, Greece and Spain) played a significant role starting from 2009 (Li and Zinna, 2014). Looking at the Italian case, the BTP-Bund spreads rose, as can be seen in Figure 4. The risk of a possible breakup of the Eurozone¹³ appeared for the first time in 2010, following the events that brought Greece to the brink of bankruptcy¹⁴. The redenomination risk component, up to that time almost inexistent (see Figure 5), progressively played an important role in the price discovery process of the Italian banks' CDS spreads, thus pricing the risk of a possible Italexit already during the sovereign debt crisis (Anelli et al., 2020). The perception of a hypothetical Italy's currency redenomination is explained by the debt sustainability problems. Due to the high debt-to-GDP ratio, the market fear was that Italy could exit from the Euro Area and redenominate the public debt in the old national currency to repay its nominal value at maturity and, at the same time, honour the currents interest payments (Cesaratto, 2015).

¹³ When the sovereign debt crisis broke out, it was perceived the risk that the resilience of the sovereign debt of a Eurozone peripheral country could damage not only its own banking system, but also that of other Euro Area countries through bank holdings of other countries' debt (Bolton and Jeanne, 2011). Cross-border relations have strengthened the links between sovereign States and have widened the chain of contagion: if a State fails, it is very likely that this could generate a domino effect that could lead to the breakup of the Euro Area (Li and Zinna, 2014).

¹⁴ It became clear that Greece, with failing public finances, would have not been able to refinance its debt on the market. Subsequently, the other European countries intervened with bilateral loans, as well as the European Financial Stability Facility and the International Monetary Fund (Cesaratto, 2015).

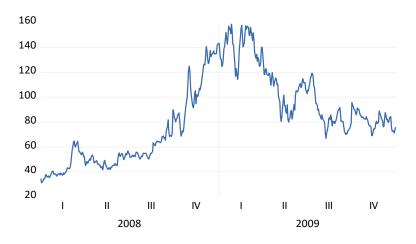


Figure 4: Btp-Bund spread: period 2008-2009.

Source: authors' calculations in Eviews 11 based on Bloomberg data.

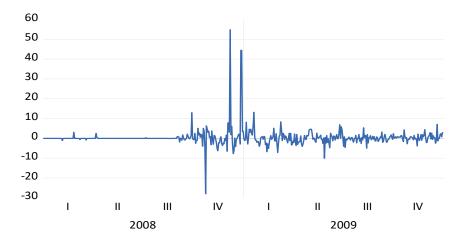


Figure 5: Redenomination risk: period 2008-2009.

Source: authors' calculations in Eviews 11 based on Bloomberg data.

Fears about the reversibility of the Eurozone were then contained in 2012 thanks to the Mario Draghi's "Corageous Leap" speech in May 2012 and the "Whatever it takes" speech in July of the same year. Finally, the introduction of OMT¹⁵ (Outright Monetary Transactions) favored a gradual reduction of spreads and sovereign risk in the Euro Area (Li and Zinna, 2014).

¹⁵ According to Li and Zinna (2014), with the fourth phase of the OMT the market has no longer priced in Euro systemic sovereign risk.

In the following years characterized by the launch in March 2015 of QE (Quantitative Easing)¹⁶ by the ECB (European Central Bank), sovereign risk and redenomination risk played a more restrained role in the price discovery process of CDS spreads.

However, the apparent stability on the markets was interrupted due to the referendum on Brexit, an event in which, in quantitative terms, both components had a strong impact on the variance of the Italian banks' CDS spreads. This event caused systemic strain in the markets: observing Italy, the spread started to rise again and the redenomination risk perceived by the markets significantly grew compared to 2015.

The Brexit problem, as said by the Governor of the Bank of Italy Visco (2016), is linked to the prolonged uncertainty that the event generated in the European Union, which had consequences on the financial markets. In the case of Italy, it was not so much the commercial aspect that generated the redenomination risk (Italy is not very integrated from the commercial point of view with the United Kingdom such as Luxembourg, Ireland, Germany or France, which maintain close commercial and financial relationships), as well as the political situation. The perceived risk was what the Minister of Economy and Finance Padoan defined as the risk of "political emulation", as to say, the risk that also in Italy could form political currents in favour of leaving the EU (European Union), following the experience of the United Kingdom (Bricco et al., 2016).

In the wake of the underlying political idea of the Brexit, a pro-deficit government coalition, supported by populist parties, has formed in Italy in March 2018. During the anti-establishment government, the redenomination risk component assumed the role of the main driver of the variance of the CDS spreads of Italian banks, the sovereign risk component assumed a lower weight. However, an interesting aspect to analyze is that towards the end of 2018, while interest rates on government bonds across the Eurozone generally decreased (Longo, 2019), in Italy it was the redenomination risk perceived by the markets that caused the increase of the BTP-Bund spreads (Gros, 2018). As can be seen in Figure 6, starting from mid-2018 the CDS contracts signed under the 2014 definitions clearly grew, a symptom of the fact that the markets began to perceive not only a pure default risk, but also the risk of a possible Italexit and a consequent redenomination in a devalued currency (Gros, 2018). The cause is mainly attributable to political uncertainty generated by the new

¹⁶ The QE, initially called "Expanded Asset Purchase Program (APP)" and later "Public Sector Purchase Program (PSPP)", consists in the purchase by the ECB (European Central Bank) of long-term public and private securities in definitive way. It is an expansionary monetary policy measure with which the central bank essentially injects liquidity into the system, aimed at keeping long-term interest rates low and, therefore, at supporting aggregate demand (Cesaratto, 2015). The tensions associated with this intervention are reflected in the increased redenomination risk. One of the limits of QE, as Cesaratto said (2015), is that this has not been accompanied by expansionary fiscal policies. Considering the Italian case, austerity policies continued to exist in those years and, although one effect of the program was to reduce interest rates on debt, the debt-to-GDP ratio continued to grow. Therefore the market attention on the country's ability to refinance its debt has remained high.

government coalition (anti-establishment or populist) in favor of the growth of the deficit (pro-deficit) and openly against the technocracy of the European Union. This attitude, combined with the will of the coalition parties to leave the Euro Area, declared during and after the electoral campaign, generated the perception that the Brexit experience could also happen in Italy (Anelli et al., 2020).

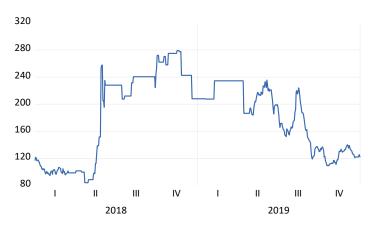


Figure 6: The 2014 Italian CDS Definitions: period 2018-2019. Source: authors'calculations in Eviews 11 based on Bloomberg data.

The fundamental difference between the two periods in which redenomination risk component assumed greater statistical significance is given by the various factors that influenced the market behavior. In 2010, in fact, the market fears about a possible Italexit were caused by the precarious condition of the country's public finances. On the contrary, in the last phase the markets began to perceive a pure redenomination risk, regardless of the health of public finances, especially due to the Italian political situation (Gros, 2018). As reported by Reed (2018), the high volatility that hit the Italian banking sector was caused in particular by the country's political uncertainty rather than by the economic situation.

In this context, the market have taken into consideration political events, while the economic fundamentals of the Italian banking system have been overshadowed (Anelli et al., 2020).

6. Conclusions

In this paper, we analyzed the dynamic evolution and impact of redenomination risk and sovereign risk on the CDS spreads of the main Italian banks from 2008 to 2020 with a "rolling window regressions" model.

The results of the analysis highlight the statistical significance of redenomination risk during periods of greatest political-financial stress: starting from the genesis of the sovereign debt crisis (2010), during the lauch of the ECB's QE (2015), on the occasion of the referendum on Brexit (2016), before the formation of the Italian

anti-establishment government (2017) and during the first Conte's Italian government (2019), differently from the result obtained by Anelli et al. (2020) using the "quanto CDS" spreads. The sovereign risk component explains a large part of the variance in bank CDS spreads for almost the entire considered period.

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Appendix

Augmented Dickey-Fuller Test					
Residuals					
Period	t-Statistic	Prob.*			
2008	-10.78183	0.0000			
2009	-11.52749	0.0000			
2010	-10.30006	0.0000			
2011	-12.14997	0.0000			
2012	-13.27087	0.0000			
2013	-9.181330	0.0000			
2014	-10.72865	0.0000			
2015	-9.568894	0.0000			
2016	-10.70846	0.0000			
2017	-12.30533	0.0000			
2018	-13.51737	0.0000			
2019-2020	-12.03188	0.0000			

Table 2: Augmented Dickey-Fuller test

*MacKinnon (1996) one-sided p-values. Soure: authors' own calculations in Eviews 11.

• Intesa San Paolo

Period January 2008-December 2008 (261 Obs.)							
Variable	Coefficient	CoefficientStd. Errort-StatisticProb.					
βο	0.053855	0.470134	0.114553	0.9089			
Δ (Italy Red. Risk)	-0.045905	0.067431	-0.680776	0.4966			
Δ (Spread It-Ge)	0.577953***	0.149124	3.875664	0.0001			
R ²		0.0562	209				
Adj. R ²	0.048893						
Durbin-Watson Stat.		2.0981	34				
Period Ja	nuary 2009-De	cember 2009	(261 Obs.)				
Variable	Coefficient	Std. Error	t-Statistic	Prob.			
βο	-0.106364	0.461950	-0.230249	0.8181			
Δ (Italy Red. Risk)	0.082447	0.153054	0.538681	0.5906			
Δ (Spread It-Ge)	0.365106*** 0.105816 3.450398 0.0007						
R ²	0.045297						
Adj. R ²	0.037896						
Durbin-Watson Stat.	2.537520						

Period January 2010-December 2010 (261 Obs.)							
Variable	Coefficient	Coefficient Std. Error t-Statistic Prob.					
βο	0.074037	0.508534	0.145589	0.8844			
Δ (Italy Red. Risk)	0.371204***	0.098488	3.769041	0.0002			
Δ (Spread It-Ge)	0.687374***	0.079071	8.693145	0.0000			
R ²		0.2809	45				
Adj. R ²		0.2753	571				
Durbin-Watson Stat.		2.6558	514				
Period Ja	nuary 2011-De	cember 2011	(260 Obs.)				
Variable	Coefficient	Std. Error	t-Statistic	Prob.			
βo	0.488129	1.183.558	0.412425	0.6804			
Δ (Italy Red. Risk)	0.067791	0.159910	0.423931	0.6720			
Δ (Spread It-Ge)	0.586834***	0.076710	7.650061	0.0000			
R ²		0.1946	532				
Adj. R ²		0.1883	65				
Durbin-Watson Stat.		2.2762	.89				
Period Ja	nuary 2012-De	ecember 2012	(261 Obs.)				
Variable	Coefficient	Std. Error	t-Statistic	Prob.			
βo	-0.203377	0.889909	-0.228537	0.8194			
Δ (Italy Red. Risk)	-0.039619	0.138290	-0.286493	0.7747			
Δ (Spread It-Ge)	0.710763***	0.064856	10.95913	0.0000			
R ²		0.3183	80				
Adj. R ²		0.3130	96				
Durbin-Watson Stat.		2.5261	.83				
Period Ja	nuary 2013-De	cember 2013	(261 Obs.)				
Variable	Coefficient	Std. Error	t-Statistic	Prob.			
βo	-0.320095	0.573635	-0.558011	0.5773			
Δ (Italy Red. Risk)	0.323188**	0.151832	2.128595	0.0342			
Δ (Spread It-Ge)	0.551735***	0.071318	7.736262	0.0000			
R ²		0.2094	-90				
Adj. R ²		0.2033					
Durbin-Watson Stat.		2.4151	46				
	nuary 2014-De		(261 Obs.)				
Variable	Coefficient	Std. Error	t-Statistic	Prob.			
βο	-0.129210	0.359644	-0.359271	0.7197			
Δ (Italy Red. Risk)	0.117668	0.078383	1.501187	0.1345			
Δ (Spread It-Ge)	0.244316***	0.070091	3.485690	0.0006			
R ²		0.0517					
Adj. R ²	0.044379						
Durbin-Watson Stat.	2.663919						

Period January 2015-December 2015 (261 Obs.)					
Variable	Coefficient	Std. Error	t-Statistic	Prob.	
βο	0.087833	0.279390	0.314375	0.7535	
Δ (Italy Red. Risk)	0.094636*	0.056240	1.682733	0.0936	
Δ(Spread It-Ge)	0.362762***	0.051313	7.069654	0.0000	
R ²	0.209806				
Adj. R ²	0.203680				
Durbin-Watson Stat.	2.665585				

Period January 2016-December 2016 (261 Obs.)

Variable	Coefficient	Std. Error	t-Statistic	Prob.	
βo	-0.028613	0.383173	-0.074674	0.9405	
Δ (Italy Red. Risk)	0.186989***	0.070223	2.662808	0.0082	
Δ (Spread It-Ge)	0.717512***	0.086437	8.300979	0.0000	
R ²	0.281842				
Adj. R ²	0.276275				
Durbin-Watson Stat.		2.374418			

Period January 2017-December 2017 (260 Obs.)

Variable	Coefficient	Std. Error	t-Statistic	Prob.	
βο	-0.274319	0.244281	-1.122964	0.2625	
Δ (Italy Red. Risk)	0.166097***	0.051533	3.223142	0.0014	
Δ (Spread It-Ge)	0.164326***	0.059734	2.750960	0.0064	
R ²	0.082452				
Adj. R ²	0.075312				
Durbin-Watson Stat.	2.353011				

Period January 2018-December 2018 (261 Obs.)

Variable	Coefficient	Std. Error	t-Statistic	Prob.	
βο	0.422540	0.538737	0.784316	0.4336	
Δ (Italy Red. Risk)	0.069520	0.066235	1.049596	0.2949	
Δ(Spread It-Ge)	0.005254	0.064730	0.081166	0.9354	
R ²	0.005638				
Adj. R ²	-0.002070				
Durbin-Watson Stat.		2.1573	2.157390		

Period January 2019-January 2020 (262 Obs.)							
Variable	Coefficient	Coefficient Std. Error t-Statistic Prob.					
βο	-0.220111	0.239295	-0.919832	0.3585			
Δ (Italy Red. Risk)	0.238062***	0.047829	4.977411	0.0000			
∆(Spread It-Ge)	0.188002***	0.044118	4.261384	0.0000			
R ²	0.202238						
Adj. R ²	0.196078						
Durbin-Watson Stat.		1.984132					

Note: *** signals parameter significance at 1%, ** signals parameter significance at 5%,* signals parameter significance at 10%. Source: authors' calculations in Eviews 11 based on Bloomberg data.

• Unicredit

Period January 2008-December 2008 (261 Obs.)				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
βο	0.076751	0.515482	0.148892	0.8818
Δ (Italy Red. Risk)	0.014261	0.073935	0.192887	0.8472
Δ (Spread It-Ge)	0.556289***	0.163508	3.402222	0.0008
R ²		0.0432	207	
Adj. R ²		0.0357	'90	
Durbin-Watson Stat.		2.0743	800	
Period January 2009-December 2009 (261 Obs.)				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
βο	0.000109	0.706747	0.000154	0.9999
Δ (Italy Red. Risk)	0.134674	0.234160	0.575139	0.5657
Δ (Spread It-Ge)	0.450204***	0.161889	2.780937	0.0058
R ²		0.0304	52	
Adj. R ²	0.022936			
Durbin-Watson Stat.	2.436740			
Period January 2010-December 2010 (261 Obs.)				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
βο	0.055275	0.558998	0.098883	0.9213
Δ (Italy Red. Risk)	0.301087***	0.108261	2.781126	0.0058
Δ (Spread It-Ge)	0.791775***	0.086917	9.109520	0.0000
R ²	0.278650			
Adj. R ²	0.273059			
Durbin-Watson Stat.	2.849291			

Table 4: OLS estimates: Period January 2008 – January 2020 (3131 Obs.)

Period January 2011-December 2011 (260 Obs.)					
Variable	Coefficient	Std. Error	t-Statistic	Prob.	
βο	0.460812	1.652408	0.278873	0.7806	
Δ (Italy Red. Risk)	0.211353	0.223256	0.946681	0.3447	
Δ (Spread It-Ge)	0.642284***	0.107097	5.997207	0.0000	
R ²		0.1358	96		
Adj. R²		0.1291	72		
Durbin-Watson Stat.		2.2577	'95		
Period Jo	anuary 2012-D	ecember 2012	(261 Obs.)		
Variable	Coefficient	Std. Error	t-Statistic	Prob.	
βo	-0.239350	0.797670	-0.300061	0.7644	
Δ (Italy Red. Risk)	-0.122290	0.123956	-0.986563	0.3248	
Δ (Spread It-Ge)	0.777340***	0.058133	13.37164	0.0000	
R ²		0.4120	40		
Adj. R ²		0.4074	-83		
Durbin-Watson Stat.		2.3715	21		
Period Ja	anuary 2013-D	ecember 2013	(261 Obs.)		
Variable	Coefficient	Std. Error	t-Statistic	Prob.	
βο	-0.372424	0.699221	-0.532627	0.5948	
Δ (Italy Red. Risk)	0.019963	0.185072	0.107864	0.9142	
Δ (Spread It-Ge)	0.645924***	0.086932	7.430248	0.0000	
R ²		0.1781	97		
Adj. R ²		0.1718	26		
Durbin-Watson Stat.		2.6768	344		
Period Jo	anuary 2014-D	ecember 2014	(261 Obs.)		
Variable	Coefficient	Std. Error	t-Statistic	Prob.	
βo	0.026960	0.337863	0.079796	0.9365	
Δ (Italy Red. Risk)	0.119713	0.073636	1.625731	0.1052	
Δ (Spread It-Ge)	0.337007***	0.065846	5.118108	0.0000	
R ²		0.0989	13		
Adj. R ²	0.091928				
Durbin-Watson Stat.					
	Durbin-Watson Stat. 2.445232 Period January 2015-December 2015 (261 Obs.)				
Variable	Coefficient Std. Error t-Statistic Prob.				
βο	0.088486	0.412178	0.214678	0.8302	
Δ (Italy Red. Risk)	0.171621**	0.082969	2.068503	0.0396	
Δ (Spread It-Ge)	0.370168***	0.075700	4.889929	0.0000	
R ²	0.132515				
Adj. R ²	0.125791				
Durbin-Watson Stat.	2.635660				

Period January 2016-December 2016 (261 Obs.)				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
βο	-0.082030	0.471465	-0.173990	0.8620
Δ (Italy Red. Risk)	0.257340***	0.086403	2.978357	0.0032
Δ (Spread It-Ge)	0.882333***	0.106354	8.296194	0.0000
R ²		0.2893	78	
Adj. R ²		0.2838	570	
Durbin-Watson Stat.		2.3009	62	
Period Jo	anuary 2017-D	ecember 2017	(260 Obs.)	
Variable	Coefficient	Std. Error	t-Statistic	Prob.
βο	-0.413290	0.317892	-1.300095	0.1947
Δ (Italy Red. Risk)	0.111189*	0.067062	1.658011	0.0985
Δ (Spread It-Ge)	0.106176	0.077734	1.365880	0.1732
R ²		0.0225	48	
Adj. R ²		0.0149		
Durbin-Watson Stat.		2.2149	45	
Period Jo	anuary 2018-D	ecember 2018	(261 Obs.)	
Variable	Coefficient	Std. Error	t-Statistic	Prob.
βo	0.377143	0.402603	0.936762	0.3498
Δ (Italy Red. Risk)	-0.002744	0.049498	-0.055445	0.9558
Δ (Spread It-Ge)	0.189874***	0.048374	3.925146	0.0001
R ²		0.0679		
Adj. R ²		0.0607		
Durbin-Watson Stat.		1.9947	68	
	anuary 2019-J		(262 Obs.)	
Variable	Coefficient	Std. Error	t-Statistic	Prob.
βo	-0.260091	0.246618	-1.054632	0.2926
Δ (Italy Red. Risk)	0.195668***	0.049292	3.969550	0.0001
Δ (Spread It-Ge)	0.178588***	0.045468	3.927792	0.0001
R ²	0.155905			
Adj. R ²	0.149387			
Durbin-Watson Stat.	2.065946			

Note: *** signals parameter significance at 1%, ** signals parameter significance at 5%,* signals parameter significance at 10%. Source: authors' calculations in Eviews 11 based on Bloomberg data.

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Period January 2008-December 2008 (261 Obs.)				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
βο	-0.052306	0.427699	-0.122295	0.9028
Δ (Italy Red. Risk)	0.011072	0.061345	0.180483	0.8569
Δ(Spread It-Ge)	0.678795***	0.135663	5.003525	0.0000
R ²		0.0887	45	
Adj. R ²		0.0816	81	
Durbin-Watson Stat.		1.8149	77	
Period Ja	unuary 2009-D	ecember 2009	(261 Obs.)	
Variable	Coefficient	Std. Error	t-Statistic	Prob.
βο	0.006267	0.452747	0.013842	0.9890
Δ (Italy Red. Risk)	0.235202	0.150004	1.567965	0.1181
Δ (Spread It-Ge)	0.398969***	0.103707	3.847066	0.0002
R ²		0.0631	93	
Adj. R ²		0.0559	31	
Durbin-Watson Stat.		2.3414	-33	
Period Ja	unuary 2010-D	ecember 2010	(261 Obs.)	
Variable	Coefficient	Std. Error	t-Statistic	Prob.
βο	0.364905	0.429510	0.849584	0.3963
Δ (Italy Red. Risk)	0.335726***	0.083183	4.035992	0.0001
Δ (Spread It-Ge)	0.875291***	0.066784	13.10636	0.0000
R ²		0.4448	22	
Adj. R ²	0.440518			
Durbin-Watson Stat.		2.4214	-40	
Period Ja	anuary 2011-D	ecember 2011	(260 Obs.)	
Variable	Coefficient	Std. Error	t-Statistic	Prob.
βο	0.210384	1.305947	0.161097	0.8721
Δ (Italy Red. Risk)	0.047028	0.176446	0.266530	0.7900
Δ (Spread It-Ge)	0.637062***	0.084642	7.526546	0.0000
R ²	0.188178			
Adj. R ²	0.181860			
Durbin-Watson Stat.	2.327086			

Period January 2012-December 2012 (261 Obs.)						
Variable	Coefficient	Std. Error	t-Statistic	Prob.		
βο	0.520969	1.016753	0.512385	0.6088		
Δ (Italy Red. Risk)	-0.047552	0.158001	-0.300961	0.7637		
Δ (Spread It-Ge)	0.905192***	0.074100	12.21580	0.0000		
R ²		0.3671	91			
Adj. R ²		0.3622	.86			
Durbin-Watson Stat.		2.0393	45			
Period J	anuary 2013-D	ecember 2013	(261 Obs.)			
Variable	Coefficient	Std. Error	t-Statistic	Prob.		
βo	-0.218126	1.010416	-0.215878	0.8293		
Δ (Italy Red. Risk)	0.133619	0.267440	0.499621	0.6178		
Δ (Spread It-Ge)	0.622287***	0.125621	4.953671	0.0000		
R ²		0.0900	81			
Adj. R ²		0.0830	28			
Durbin-Watson Stat.		2.2720	26			
Period J	anuary 2014-D	ecember 2014	(261 Obs.)	-		
Variable	Coefficient	Std. Error	t-Statistic	Prob.		
βo	-0.307036	0.694943	-0.441814	0.6590		
Δ (Italy Red. Risk)	0.326882**	0.151461	2.158195	0.0318		
Δ (Spread It-Ge)	0.589334***	0.135437	4.351339	0.0000		
R ²		0.0818	46	I		
Adj. R ²		0.0747	28			
Durbin-Watson Stat.		2.4302	.97			
Period J	anuary 2015-D	ecember 2015	(261 Obs.)			
Variable	Coefficient	Std. Error	t-Statistic	Prob.		
βο	0.117235	0.409669	0.286171	0.7750		
Δ (Italy Red. Risk)	0.111054	0.082464	1.346699	0.1793		
Δ (Spread It-Ge)	0.429644***	0.075239	5.710363	0.0000		
R ²		0.1473	91			
Adj. R ²		0.1407	81			
Durbin-Watson Stat.	2.232043					
Period J	anuary 2016-D	ecember 2016	(261 Obs.)			
Variable	Coefficient	Std. Error	t-Statistic	Prob.		
βo	0.213865	1.483779	0.144136	0.8855		
Δ (Italy Red. Risk)	0.023505	0.271926	0.086438	0.9312		
Δ (Spread It-Ge)	1.451684***	0.334714	4.337086	0.0000		
R ²	0.076399					
Adj. R ²	0.069239					
Durbin-Watson Stat.		2.3119	000	2.311900		

Period January 2017-December 2017 (260 Obs.)					
Variable	Coefficient	Std. Error	t-Statistic	Prob.	
βo	-0.931053	0.494420	-1.883122	0.0608	
Δ (Italy Red. Risk)	0.030107	0.104301	0.288658	0.7731	
Δ (Spread It-Ge)	0.011894	0.120901	0.098381	0.9217	
R ²		0.0004	33		
Adj. R ²		-0.0073	346		
Durbin-Watson Stat.	2.009834				
Period Ja	Period January 2018-December 2018 (261 Obs.)				
Variable	Coefficient	Std. Error	t-Statistic	Prob.	
βο	0.632516	0.824633	0.767026	0.4438	
Δ (Italy Red. Risk)	0.039162	0.101384	0.386276	0.6996	
Δ (Spread It-Ge)	0.125259	0.099081	1.264198	0.2073	
R ²	0.010318				
Adj. R ²	0.002646				
Durbin-Watson Stat.	2.009978				
Period J	anuary 2019-J	Ianuary 2020	(262 Obs.)		
Variable	Coefficient	Std. Error	t-Statistic	Prob.	
βo	0.164807	0.183304	0.899090	0.3694	
Δ (Italy Red. Risk)	-0.011161	0.036637	-0.304639	0.7609	
∆(Spread It-Ge)	-0.041809	0.033795	-1.237136	0.2172	
R ²	0.008217				
Adj. R ²	0.000558				
Durbin-Watson Stat.	2.011289				

Note: *** signals parameter significance at 1%, ** signals parameter significance at 5%,* signals parameter significance at 10%. Source: authors' calculations in Eviews 11 based on Bloomberg data.