How Does Liquidity Affect Government Bond Yields?

Carlo Favero, Marco Pagano, and Ernst-Ludwig von Thadden*

Abstract

The paper explores the determinants of yield differentials between sovereign bonds, using euro-area data. There is a common trend in yield differentials, which is correlated with a measure of aggregate risk. In contrast, liquidity differentials display sizeable heterogeneity and no common factor. We propose a simple model with endogenous liquidity demand, where a bond's liquidity premium depends both on its transaction cost and on investment opportunities. The model predicts that yield differentials should increase in both liquidity and risk, with an interaction term of the opposite sign. Testing these predictions on daily data, we find that the aggregate risk factor is consistently priced, liquidity differentials are priced for a subset of countries, and their interaction with the risk factor is in line with the model's prediction and crucial to detect their effect.

I. Introduction

What determines the yield differentials between bonds? Even though research has shown that the moments of bonds' return distribution and liquidity typically both play a role, we still understand imperfectly the relative importance of these two determinants and their possible interactions.

The European Monetary Union (EMU) offers a particularly good arena to examine these issues and sharpen our understanding because in the euro area we can observe bonds issued by several sovereign issuers, without the complications arising from different currencies and different bond conventions but with variation in liquidity. Even though yields on euro-area government bonds converged

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significantly since EMU's inception in 1999, these bonds are still not regarded as perfect substitutes by market participants: Nonnegligible differences in yields across countries have remained, to different extents for different issuers and maturities, and they fluctuate over time. Even the bonds issued by AAA-rated issuers are not regarded as perfect substitutes, so that, for example, French bonds traded in the cash market are not seen as a perfect hedge for positions in Bund futures.¹

A possible reason for these persistent differentials is persistent risk differences. Different sovereign issuers are perceived as having different solvency risks in spite of the provisions of the Stability Pact. A second possible explanation is liquidity. This is indeed the explanation that is often advanced by practitioners.

However, a look at the time-series behavior of euro-area yield differentials suggests that neither one of these two factors in isolation is likely to provide the full answer. First, as shown below, the yield differentials relative to the German Bund tend to fluctuate together much more than measures of liquidity such as bidask spreads do. This suggests that liquidity alone cannot be the full answer and that other factors must be driving the differentials' time-series behavior. Such factors are likely to be related to international investment opportunities or global risk perceptions. For instance, even if the default risk of the Italian and French governments relative to the German one were very stable over time, a changing world price for risk could induce the implied yield differentials to fluctuate together. But this cannot be the full story either. Sizable yield differentials have been observed for several years even within the group of AAA-rated euro-zone countries: As late as 2002, 10-year AAA-rated Finnish debt yielded on average 20 basis points (bp) more than the 10-year German Bund. This suggests that liquidity differences indeed may play a role, as practitioners claim.²

To analyze these issues, we develop a simple asset pricing model with exogenous transaction costs and endogenous liquidity demand. The model is deliberately kept simple in order to isolate the implications of our key assumption of endogenous liquidity demand. The most important insight of our theoretical analysis is that liquidity matters for pricing, but that it interacts with aggregate factors in a way that is different from what models like the traditional capital asset pricing model (CAPM) would predict.

Our model is based on the idea that the demand for liquidity responds both to the magnitude of trading costs and to the availability of outside investment opportunities. First, investors are less inclined to trade securities with larger trading costs. Second, they are less likely to liquidate securities when outside investment

¹See Pagano and von Thadden (2004) for an account of the integration of European bond markets and a survey of the relevant literature.

²For instance, the increase of yield differentials relative to the Bund rate in late 1999 was explained as follows: "After having tested the waters of Europe's smaller bond markets, institutional investors are deciding they've had enough. . . . Declining liquidity in the smaller debt markets is boosting the premiums these countries are having to pay investors compared with the core euro-zone nations" (*Wall Street Journal Europe* (November 3, 1999)). Market practitioners clearly attribute remaining yield differentials to liquidity premiums, which are held to be larger in thinner markets, irrespective of their credit rating: "'Peripheral issuers in Europe are in trouble: They're paying a huge liquidity premium,' says Steven Mayor, chief bond strategist at ING Barings in London. He says that their problem comes down to the fact that some still only represent 1% to 2% of the euro-zone issuance" (ibid.).

opportunities are less attractive, a situation that is assumed to coincide with increased aggregate risk. As a result, when risk is expected to increase, investors' demand for liquidity abates, and the premium they place on more liquid securities declines. Therefore, although in general investors value liquidity, they value it less when risk increases.

A key ingredient of our model is the insight of Merton's (1971) intertemporal CAPM (I-CAPM) that future investment opportunities matter for individual portfolio choice. In our model, when investors choose their portfolio, they know that they may want to rebalance it later when they receive new information about their exposure to an aggregate factor. Since an increase in aggregate risk worsens the risk-return profile of nonmarketable investments (such as private equity, for which no price effect compensates for increased risk), at times of high aggregate risk investors are less likely to liquidate marketable assets in order to move into alternative investments. This asymmetry between marketable and nonmarketable assets together with the I-CAPM insight that heterogenous future investment opportunities impact current prices implies the cross effect that is specific to our theory. As in any sensible theory, yields increase in expected transaction costs. But investors in our model understand that these transaction costs will only be incurred when they rebalance their portfolio later on, and that the frequency of this rebalancing decreases in aggregate risk. Hence, the direct effect of transaction costs (illiquidity) on yields is positive, but this effect is decreasing in aggregate risk.

These ideas contrast with those embedded in the model by Vayanos (2004), where the demand for liquidity is not sensitive to trading costs and increases at times of high risk. In Vayanos's model, fund managers are subject to withdrawals due to mutual fund liquidations (or binding margin requirements) if their performance falls below a given threshold. Therefore they are more likely to liquidate at times of high volatility, and these forced sales are naturally effected through the cheapest channel (i.e., by using the most liquid assets). So increases in volatility induce a flight to liquidity and an increase in the liquidity premium. Hence, while anticipated transaction costs increase an asset's required yield, as in our setting, this effect is amplified by aggregate risk: The two models have the same prediction regarding the direct effect of transaction costs but the opposite prediction with respect to the interaction term.

Our model also differs from the CAPM model of liquidity by Acharya and Pedersen (2005), who implicitly assume that the demand for liquidity is inelastic to the magnitude of trading costs and that it does not vary predictably with the fundamental risk of securities. In their model, equilibrium returns compensate investors for anticipated trading costs (as in our setting), but also for liquidity risk: Unanticipated changes in liquidity are priced to the extent that they generate covariance risk (due to commonality in liquidity) or exacerbate fundamental covariance risk. But the premium that investors are willing to pay for *anticipated* trading costs does not depend on fundamental risk, in contrast to both our setting and that of Vayanos (2004).

The empirical implication of our model is that, while both increases in risk and illiquidity should reduce asset prices and drive up their returns, their interaction should have the opposite sign (i.e., it should increase asset prices and dampen the increase in their required return). An estimation that ignores the effect caused

by the interaction of liquidity with aggregate risk is likely to underestimate the direct impact of liquidity (as well as that of risk) on prices. We bring these ideas to the data using 2 years of daily observations on yields and liquidity variables for euro-area sovereign bonds at 5- and 10-year maturities. The results show that a standard proxy for aggregate risk, the yield difference between U.S. fixed interest on swaps and U.S. government bonds at the corresponding maturity, is the single most important explanatory variable for euro-area yield differentials. Liquidity differentials, as proxied by the difference between the local and the relevant reference bid-ask spread, play a role only in a subset of countries. Whenever it appears with a statistically significant coefficient, the bid-ask spread positively impacts the corresponding yield relative to that of the benchmark, as any asset pricing model would predict. However, unlike other models, its interaction with the aggregate risk factor is negative and precisely estimated. In other words, i) illiquidity appears to command a premium, as in most of the literature following Amihud and Mendelson (1986), and ii) the size of such a premium is reduced by covariation between the cost of illiquidity and aggregate risk.

To get an idea of the economic magnitude of our estimated effects, consider the case of the 10-year yield differentials on German government bonds. In the sample that we consider, which covers daily observations in 2002 and 2003, yield differentials fluctuated in a range between -8 bp and 25 bp. Our proxy for aggregate risk instead fluctuated in a range between 25 bp and 80 bp, while liquidity differentials fluctuated between -4 bp and 6 bp. Our estimates show that in periods of high aggregate risk (where the U.S. swap-bond differential is higher than 70 bp) the effect of liquidity on yield differentials is not significantly different from 0, and fluctuations in yield differentials depend exclusively on the risk factor. But in periods of low aggregate risk (U.S. swap-bond differentials below 30 bp) liquidity differentials have a positive and significant impact on yield differentials: On average, an increase of 1 bp in liquidity differentials is associated with an increase of 2 bp in yield differentials (which in these periods are below 10 bp). Similar qualitative results hold for the 5-year differentials on French government bonds, keeping in mind that these differentials are on average half the 10-year differentials, so that all measures of impact have to be reduced proportionately.

The structure of the paper is as follows. Section II relates the paper to the relevant literature. Section III presents the data and describes the stylized facts that emerge from them. Section IV lays out the model and its predictions. Section V presents and discusses the estimation results. Section VI concludes. In Appendix A, we generalize the basic model used in the main text to include uncertainty and risk aversion. Appendix B describes our data.

II. Related Literature

This paper adds to a considerable literature on the relation between returns and liquidity. At a theoretical level, two main views have been advanced to explain why liquidity should be priced by financial markets: Illiquidity i) creates trading costs, and ii) can itself create additional risk. These views are not mutually exclusive, although they have emerged sequentially in the literature. This

paper builds on the first view and develops it in a new direction that is similar in spirit to that of the second view.

The "trading cost view" holds that illiquid securities must provide investors with a higher expected return to compensate them for their larger transaction costs, controlling for fundamental risk. The prediction here is cross-sectional: Risk-adjusted expected returns must be higher for less liquid securities. This view, first proposed and tested by Amihud and Mendelson (1986), has been the basis of a vast empirical literature. Many subsequent studies of stock market data have confirmed a significant cross-sectional association between liquidity (as measured by the tightness of the bid-ask spread or trading volume) and asset returns, controlling for risk: among these are Brennan and Subrahmanyam (1996), Chordia, Roll, and Subrahmanyam (2000), and Datar, Naik, and Radcliffe (1998).

Other studies have focused on liquidity effects in fixed-income security markets. Here, too, the initiators were Amihud and Mendelson (1991), who showed that the yield to maturity of Treasury notes with 6 months or less to maturity exceeds the yield to maturity on the more liquid Treasury bills. Studies on U.S. public debt (e.g., Warga (1992), Daves and Ehrhardt (1993), Kamara (1994), and Krishnamurthy (2002)) confirmed these findings, although using more recent data Strebulaev (2003) found that the yield spread between bills and matched notes is much smaller than previously found, especially for on-the-run bills. Goldreich, Hanke, and Nath (2005) have refined this line of analysis by investigating the impact of expected liquidity on securities' prices. They analyze the prices of Treasury securities as their liquidity changes predictably in the transition from onthe-run to the less liquid off-the-run status and show that the liquidity premium depends on the expected future liquidity over their remaining lifetime rather than on their current liquidity.

The "liquidity risk view," developed in particular by Pástor and Stambaugh (2003), highlights that liquidity is priced not only because it creates trading costs, but also because it is itself a source of risk, since it changes unpredictably over time. Since investors care about returns net of trading costs, the variability of trading costs affects the risk of a security. In an important paper, Acharya and Pedersen (2005) show theoretically in a CAPM framework with overlapping generations of investors that liquidity risk should be priced to the extent that it is correlated across assets and with asset fundamentals, and they uncover evidence consistent with this prediction. Similarly, Ellul and Pagano (2006) show that the initial underpricing of initial public offering shares should also compensate investors for the expected illiquidity and for the liquidity risk that investors face in after-market trading, and not only for fundamental risk and adverse selection problems. Also Gallmeyer, Hollifield, and Seppi (2006) propose a model of liquidity risk where traders have asymmetric knowledge about future liquidity, so that less informed investors try to learn from the amount of current trading volume how much liquidity there may be in the future. They show that current liquidity is a predictor of future liquidity risk and therefore is priced.

Our paper puts forward what may be labeled the "risk-liquidity interaction view": We point out that liquidity alters the impact of changes in risk on current prices and yields. So here the emphasis is not on liquidity risk (indeed, in this approach future liquidity is perfectly anticipated) but rather on the interaction

between liquidity and aggregate risk. In our model, changes in aggregate risk affect the liquidity premium that assets with lower transaction costs command. As noted in the introduction, this parallels the work by Vayanos (2004), who also uses constant exogenous transaction costs to model illiquidity. Our model has the same prediction as his regarding the direct effect of transaction costs, but the opposite prediction with respect to the interaction term, as explained in the previous section: In our setting, higher aggregate risk makes investors less eager to trade away from the existing portfolio of marketable assets and therefore reduces liquidity demand and liquidity premiums, while Vayanos (2004) predicts that high volatility amplifies the effect of illiquidity and therefore increases liquidity premiums.³

On the empirical front, our analysis adds to a growing recent literature on euro-area yield differentials. Codogno, Favero, and Missale (2003) estimate models of euro-area differentials with both monthly and daily data. Their estimates based on monthly data show that for most countries, only international risk factors, and not domestic ones, have explanatory power (the former being proxied by U.S. bond yield spreads and the latter by ratios of national debt to gross domestic product (GDP)). In their estimates of daily data (that refer to 2002 only), macroeconomic variables are not included because they move too slowly to allow the estimation of the impact of the domestic risk factor. Again, the international factor is statistically significantly for most countries, while liquidity (as measured by trading volume) is significant only for France, Greece, the Netherlands, and Spain.

Geyer, Kossmeier, and Pichler (2004) estimate with weekly data a multiissuer state-space version of the Cox-Ingersoll-Ross (1985) model of bond yield spreads (over Germany) for 4 EMU countries (Austria, Belgium, Italy, and Spain). They find that idiosyncratic country factors have almost no explanatory power, and yield-spread data reflect mainly a single ("global") factor, whose variation can, to a limited extent, be explained by EMU corporate bond risk (as measured by the spread of EMU corporate bonds over the Bund yield) but by nothing else in particular not by measures of liquidity. Their measurement of liquidity variables is, however, at best indirect, as they do not use data on bid-ask spreads, but rather derived measures of liquidity such as issue size and the yield differential between on-the-run and off-the-run bonds. Despite the considerable differences in the methodology and data used, both Geyer et al. (2004) and Codogno et al. (2003) agree on the finding that yield differentials under EMU are driven mainly by a common risk (default) factor related to the spread of corporate debt over government debt, and they suggest that liquidity differences have at best a minor role in the time-series behavior of yield spreads. As we shall see, our results, which rely on a more direct measure of liquidity (daily bid-ask spreads), confirm the former result but also highlight that the effect of liquidity cannot be properly gauged without taking into account its interaction with changes in the common risk factor.

³It is useful to emphasize that this effect arises from the interaction of liquidity and the common risk factor of government bonds. Concerning asset-specific risk, our model does not deliver more than the standard result that higher default risk increases the yield.

Beber, Brandt, and Kavajecz (2009) complement the MTS data on market prices and bid-ask spreads with daily data on credit default swaps for European sovereign bonds to disentangle empirically the problems of flight to quality and flight to liquidity. They show that credit risk and liquidity are positively correlated across European sovereign issuers and find, consistent with previous studies, that liquidity typically is of minor importance in explaining yields compared to credit quality. Yet liquidity becomes an important determinant of yields in times of high market volatility, measured by a U.S. index (the Chicago Board Options Exchange Volatility Index (VIX)) or by a European index (the Dow Jones EURO STOXX 50 Volatility Index (VSTOXX)). This result contrasts with our evidence that the liquidity premium on euro government debt tends to be lower when aggregate risk is higher. This difference in results may arise from the fact that Beber et al. (2009) control for country-specific risk but do not consider aggregate risk factors.⁴ In contrast, we model risk as being driven by a common factor, consistent with the results by Codogno et al. (2003) and Geyer et al. (2004).⁵

III. Data and Stylized Facts

The data that we use in the empirical analysis concern benchmark bonds' prices and liquidity indicators for the euro area, observed at daily frequencies for the period from January 1, 2002 to December 23, 2003. The data are collected from the Euro MTS Group's European Benchmark Market (EBM) trading platform and refer to a snapshot taken at 11AM (Central European Time (CET)) in all market days for the Telematico cash markets. The database contains: i) the best 5 bid and ask prices across all markets, ii) the aggregate quantity of all the outstanding proposals made at the best bid and ask prices, and iii) the daily trading volume of each bond on the EBM. From these data we calculate redemption yields, maturities, and a range of liquidity-related variables described in Appendix B. We consider Austria, Belgium, Finland, France, Germany, Italy, the Netherlands, Portugal, and Spain. We do not include Greece and Ireland in the sample, because in 2002 the convergence process to EMU was still ongoing for Greece, while the Euro MTS data for Ireland became available only at a very late stage of our sample.

Table 1 provides descriptive statistics for the yield differentials relative to Germany (for 10-year bonds) and France (for 5-year bonds) and the bid-ask spreads by country. For 10-year benchmark bonds, average yield differentials range from 4.16 bp and 6.94 bp for France and the Netherlands to 14.47 bp and 15.50 bp for Italy and Portugal, respectively, while the range of variation is smaller for 5-year bonds. In both cases the standard deviation indicates that yield differentials feature considerable time-series variability. The statistics reported in Panel B indicate that bid-ask spreads are all very tight and stable over time. For 10-year benchmark bonds, average bid-ask spreads range from 2.52 and 2.86

⁴Other work has investigated the high-frequency impact of credit risk on corporate bond spreads (see, e.g., Longstaff, Mithal, and Neis (2005), Ericsson and Renault (2006)).

⁵Another difference between the two studies lies in the data set, which refers to a largely nonoverlapping different time period, and to a set of securities that is not identical.

ticks for Italy and France to 4.60 and 4.87 for Austria and Finland, respectively. German Bunds are the third most liquid bonds after Italian and French bonds in the cash market, with a spread of 3.25 ticks. The situation is similar for 5-year bonds.

TABLE 1 Descriptive Statistics by Country

Panel A of Table 1 reports summary statistics (in basis points) for 10-year yield differentials between euro-area benchmark government bonds and German Bunds, and for 5-year yield differentials between euro-area benchmark government bonds and the French OAT. Panel B shows summary statistics (in ticks) for bid-ask spreads for euro-area 10- and 5-year government bonds. The statistics in both panels are based on daily data collected from the Euro MTS Groups EBM trading platform from January 1, 2002 to December 23, 2003.

Panel A. Euro-Area Y	Yield Differentials
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		10-Year Benchmark Bon	ds		5-Year Benchmark Bon	ds
Country	Mean	Median	Std. Dev.	Mean	Median	Std. Dev.
Austria	10.05	9.46	7.19	3.35	0.74	9.22
France	4.16	5.62	4.36	3.57	2.37	4.70
Netherlands	6.94	6.92	4.48	6.07	5.60	6.87
Belgium	13.45	11.79	6.80	4.78	4.40	8.09
Spain	9.72	8.06	7.44	-2.16	-0.42	10.13
Finland	10.88	9.34	8.30	6.48	5.82	11.18
Italy	14.47	15.70	4.88	7.97	8.34	8.01
Portugal	15.50	14.48	7.73	6.46	12.03	16.76
Panel B. Bid-Ask	Spreads					
		10-Year			5-Year	
		Benchmark Bon	ds		Benchmark Bon	ds
Country	Mean	Median	Std. Dev.	Mean	Median	Std. Dev.
Austria	4.60	4.40	1.10	4.11	4.00	0.64
France	2.86	2.80	0.46	2.52	2.60	0.34
Netherlands	3.55	3.60	0.50	3.75	3.80	0.45
Belgium	3.47	3.40	0.53	2.71	2.60	0.31
Spain	3.47	3.20	0.80	2.94	2.60	0.78
Finland	4.87	4.60	1.09	4.07	3.80	0.81
Italy	2.52	2.40	1.37	2.12	2.00	0.43
Portugal	4.33	4.40	0.69	3.16	3.00	0.51
Germany	3.25	3.00	0.67	3.20	3.20	0.45

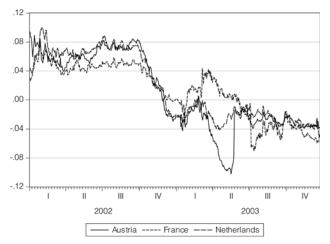
Figure 1 illustrates the time variation of 10-year yield differentials between each country in our sample and Germany, taken as the reference country. For clarity, we report the data for the Netherlands, France, and Austria in Graph A and for all the remaining countries in Graph B. Yield differentials have a clear tendency to comove. The presence of comovement is confirmed by Table 2, which reports the correlation between yield differentials over the sample period and presents a principal-components analysis. Correlations are very high both within and between groups, and the principal-components analysis shows that the first principal component explains more than 90% of the variance of the series. Liquidity indicators behave differently. Figure 2 shows the difference in bid-ask spreads observed for benchmark bonds relative to German ones for the same groupings of countries as those used in Figure 1. The figure shows 5-day moving averages of the daily observations to smooth volatility. Clearly, liquidity indicators have a different time pattern from yield differentials. This is confirmed by the correlations and principal-components analysis shown in Table 3. The correlation between differentials in liquidity indicators is much lower than that between yield differentials. Moreover, the principal-components analysis reveals that for liquidity indicators, at least 6 components are needed to explain the same proportion of the total variance as that explained by the first component in the case of yield differentials.

The principal-components analysis of Table 2 shows clearly that there is a common international factor in yield differentials in Europe. In Figure 3 we display the behavior of a variable that is often proposed in the literature as a proxy for this factor: the spread between the yield on 10-year fixed interest rates on

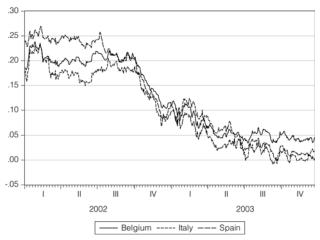
FIGURE 1 Yield Differentials in the Euro Area

Figure 1 displays the 10-year yield differentials between euro-area benchmark government bonds and German Bunds from January 1, 2002 to December 23, 2003. For clarity, we report the data for Austria, France, and the Netherlands in Graph A; for Belgium, Italy, and Spain in Graph B; and for Finland and Portugal in Graph C.

Graph A. Austria, France, and the Netherlands



Graph B. Belgium, Italy, and Spain



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FIGURE 1 (continued) Yield Differentials in the Euro Area

Graph C. Finland and Portugal

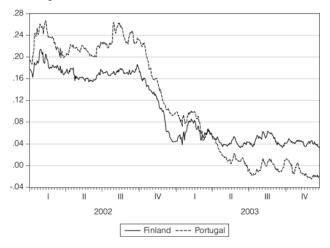


TABLE 2

Correlation and Principal Components of Euro-Area Yield Differentials

Panel A of Table 2 reports correlations among 10-year yield differentials between euro-area benchmark government bonds and German Bunds. Panel B shows the principal components of these differentials. The statistics in both panels are based on daily data collected from the Euro MTS Groups EBM trading platform from January 1, 2002 to December 23, 2003.

Panel A. Correlation Matrix

Country	Austria	Tranco e	_	Netherlands	Belgium	Spain	Finland	Italy	Portugal
Austria	1.00	_	-	_	_	_	_	_	_
France	0.65	1.0	0	_	_	_	_	_	_
Netherlands	0.51	0.4	8	1.00	_	_	_	_	_
Belgium	0.88	0.7	2	0.61	1.00	_	_	_	_
Spain	0.88	0.6	7	0.58	0.94	1.00	_	_	_
Finland	0.84	8.0	1	0.73	0.93	0.90	1.00	_	_
Italy	0.75	3.0	4	0.52	0.82	0.80	0.89	1.00	_
Portugal	0.92	0.6	9	0.61	0.87	0.89	0.88	0.78	1.00
Panel B. Principal Components									
Component			_2_	_3_	_4_	_5_	6	7	8
Eigenvalue Proportion of vari	ance	7.28 0.91	0.26 0.03	0.16 0.02	0.13 0.016	0.06 0.008	0.05 0.006	0.03 0.004	0.01 0.001
Cumulative propo	ortion	0.91	0.94	0.96	0.98	0.988	0.99	0.998	1.00

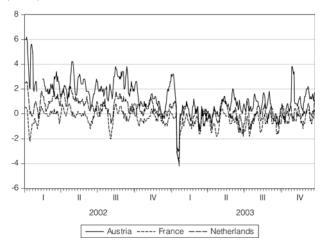
swaps and the yield on 10-year U.S. government bonds. There is ample evidence of a common trend in international bond spreads (see, e.g., Dungey, Martin, and Pagan (2000)). The empirical literature on sovereign bond spreads in emerging markets shows that the yield of U.S. government bonds, the slope of the U.S. yield curve, and risk indicators on the U.S. bond markets are the main determinants of sovereign spreads (see, e.g., Eichengreen and Mody (2000), Barnes

and Cline (1997), Kamin and Von Kleist (1999), and Arora and Cerisola (2001)). Blanco (2001) and Codogno et al. (2003) use proxies for global credit risk derived from the U.S. yield curve in their models of euro-zone government security yields. Consistent with these findings and with the results of Geyer et al. (2004), Figure 3 shows that this international risk factor is strongly correlated with the first principal component of yield differentials in the euro area.

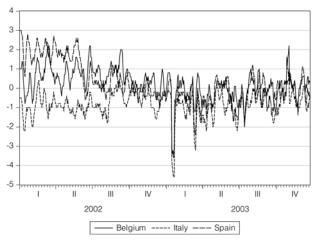
FIGURE 2 Bid-Ask Spread Differentials in the Euro Area

Figure 2 shows 5-day moving averages of the daily differentials between euro-area bid-ask spreads for benchmark government bonds and those for German Bunds (in cents) from January 1, 2002 to December 23, 2003. For clarity, we report the data for Austria, France, and the Netherlands in Graph A; for Belgium, Italy, and Spain in Graph B; and for Finland and Portugal in Graph C.

Graph A. Austria, France, and the Netherlands



Graph B. Belgium, Italy, and Spain



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FIGURE 2 (continued) Bid-Ask Spread Differentials in the Euro Area

Graph C. Finland and Portugal

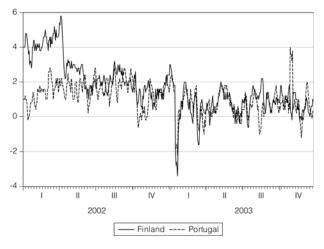


TABLE 3

Correlation and Principal Components of Euro-Area Bid-Ask Spread Differentials

Panel A of Table 3 reports correlations among differentials between euro-area bid-ask spreads for benchmark government bonds and those for German Bunds. Panel B shows the principal components of these bid-ask differentials. The statistics in both panels are based on daily data collected from the Euro MTS Group's EBM trading platform from January 1, 2002 to December 23, 2003.

Panel A. Correlation Matrix

Country	Austria	France	Netherlands	Belgium	_	Spain	Finland	Italy	Portugal
Austria	1.00	_	_	_		_	_	_	_
France	0.22	1.00	_	_		_	_	_	_
Netherlands	0.49	0.51	1.00	_		_	_	_	_
Belgium	0.39	0.46	0.44	1.00)	_	_	_	_
Spain	0.58	0.26	0.60	0.35	5	1.00	_	_	_
Finland	0.48	0.21	0.54	0.26	3	0.76	1.00	_	_
Italy	0.09	0.37	0.19	0.26	3	-0.08	-0.11	1.00	_
Portugal	0.22	0.50	0.29	0.56	3	0.20	0.19	0.24	1.00
Panel B. Principal Components									
Component	t	_1_	_2_	_3_	_4_	_5_	_6_	_7_	_8_
Eigenvalue		3.46	1.76	0.80	0.61	0.48	0.38	0.29	0.19
Proportion of var	iance	0.43	0.22	0.10	0.08	0.06	0.05	0.04	0.03
Cumulative prop	ortion	0.43	0.65	0.75	0.83	0.89	0.93	0.97	1.00

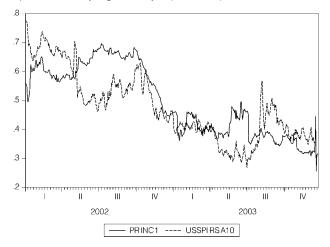
IV. Theoretical Framework

We consider a partial-equilibrium model with three dates, t = 0, 1, 2. There are 2 traded bonds, denoted by A and B, and a riskless asset that yields a net return of r per period. Bond i = A, B pays its face value V with probability q_i and 0 with probability $1 - q_i$ at date 2 and nothing at date 1. Without loss of generality we assume that $q_B \le q_A$ and interpret bond A as the benchmark. The repayment

FIGURE 3

First Principal Components of Euro-Area Yield Differentials and the Spread between the Fixed Interest Rate on Swaps and U.S. Government Bond Yield

Figure 3 plots the first principal component of 10-year yield differentials between euro-area benchmark government bonds and German Bunds from January 1, 2002 to December 23, 2003 (PRINC1) and the spread between the 10-year fixed interest rate on swaps and the U.S. 10-year government yield (USSPIRSA10).



probabilities of the 2 bonds are driven by a common factor α . When this factor increases, both bonds are less likely to repay: $dq_i/d\alpha \leq 0.6$ The purchase price of bond i at date 0 is p_{0i} . The bonds can be retraded at date 1 at a bid price of $(1-t_i)p_{1i}$ and ask price of p_{1i} , where t_i is a proportional transaction cost.

A continuum of investors $h \in [0,1]$ are ready to invest at date 0 in order to consume at date 2. Investors are risk neutral and maximize date-2 consumption. Investors may want to liquidate their bonds at date 1 to invest in alternative investment opportunities, which are comprised of nonmarketable private equity and human capital. The expected return of these real investments is negatively affected by the aggregate factor α :

$$\bar{r} - e^h \alpha$$
.

where $\bar{r} > 0$ is the maximal expected return and $e^h \in [0,e]$ is a person-specific parameter that captures investor h's exposure to the aggregate factor. In our riskneutral setting, the aggregate factor can be anything that depresses the profitability of real investment opportunities while reducing the solvency of marketable debt, such as the likelihood or severity of macroeconomic downturns. Accordingly, the individual factor e^h can be thought of as the sensitivity of a person's private investment opportunities to this common factor, as determined for instance by the sector of her family firm. In a model with risk-averse investors and risky investments, the common factor α would instead capture aggregate risk, while the individual

⁶Note that since we work in a risk-neutral setting, there is no need to specify the variance-covariance structure of bond returns. We describe such an extension for the case of risk-averse investors in Appendix A.

factor e^h would measure the investor's exposure to it. We sketch such an extension of the model in Appendix A.

The assumption that the common factor α affects both the returns to marketable assets (in our case, the occurrence of default) and those to nonmarketable investments accords with a substantial body of evidence that finds a strong positive correlation between them, starting with the seminal paper on private equity returns by Moskowitz and Vissing-Jorgensen (2002). In our empirical analysis, we will proxy the factor α by the difference between the U.S. corporate swap rate and the Treasury yield, a measure typically used to describe aggregate corporate risk (although it also captures other macroeconomic factors).

As the investor's exposure to the aggregate factor e^h increases over its support, the expected return of h's investment opportunity falls. We assume that e^h satisfies the law of large numbers, so that uncertainty washes out in the aggregate and individual probabilities are the same as aggregate frequencies. Let us denote by G the differentiable cumulative distribution function of e^h and by g = G' its density function.

At date 1, an investor will liquidate his holdings of bond *i* to invest in his outside investment opportunity if and only if the latter's expected return exceeds the expected return of the bond over its residual life, net of transaction costs:

(1)
$$\overline{r} - e^h \alpha > \frac{q_i V}{(1 - t_i) p_{1i}}.$$

Hence, investor h will sell his holdings of bond i with ex ante probability

(2)
$$\pi_{i} = \operatorname{prob}\left(e^{h} < \frac{\overline{r}}{\alpha} - \frac{q_{i}V}{\alpha(1 - t_{i})p_{1i}}\right)$$
$$= G\left(\frac{\overline{r}}{\alpha} - \frac{q_{i}V}{\alpha(1 - t_{i})p_{1i}}\right).$$

Note that here investors liquidate their assets as a function of changing investment opportunities and therefore as a function of market prices and yields. In other words, in this model demand for liquidity is price elastic.

At the interim date, investors will buy bond i only if expression (1) does not hold and if the bond does not yield less than the safe investment (i.e., if its price is sufficiently low):

$$p_{1i} \leq \frac{q_i V}{1+r} \equiv v_i.$$

Hence, $p_{1i} = v_i$ is the maximum possible price for an equilibrium at date 1 to exist. It is therefore necessary for the existence of an equilibrium that aggregate demand for liquidity at this price is positive (i.e., that \bar{r} is sufficiently large). If \bar{r} were too small, even the investors with the most profitable investment opportunity would value them less than their bond holdings and no investor would ever want to sell bonds at date 1.

No arbitrage implies that the expected yields of all assets are equalized in equilibrium. Hence expression (3) must hold with equality, implying:

$$(4) p_{1i} = v_i, i = A, B,$$

(5)
$$\pi_i = G\left(\frac{\overline{r}}{\alpha} - \frac{1+r}{\alpha(1-t_i)}\right).$$

In particular, the discount rate between dates 1 and 2 is r, the same as between 0 and 1. The expected payoff of bond i at date 0 therefore is

(6)
$$p_{0i} = \pi_i \frac{(1-t_i)p_{1i}}{1+r} + (1-\pi_i)\frac{q_iV}{(1+r)^2} = \frac{(1-\pi_it_i)q_iV}{(1+r)^2},$$

because investors have rational expectations over their future decision to either sell (which occurs with probability π_i) or hold on to the bond (which occurs with probability $1 - \pi_i$). So the bond's pledged yield to maturity is

$$1 + Y_i = \frac{V}{p_{0i}} = \frac{(1+r)^2}{(1-\pi_i t_i)q_i}.$$

The yield ratio between the two bonds is simpler to calculate than the yield differential. It is given by

$$\frac{1+Y_B}{1+Y_A} = \left(\frac{1-\pi_A t_A}{1-\pi_B t_B}\right) \left(\frac{q_A}{q_B}\right).$$

By using the approximation $ln(1 + x) \approx x$ the yield differential at date 0 can be approximated simply by

(7)
$$\Delta Y = Y_B - Y_A \approx \pi_B t_B - q_B - \pi_A t_A + q_A.$$

This expression for the yield differential is intuitive. First, there is a direct positive impact of transaction costs. If π_i were constant (i.e., if the probability of liquidation were not endogenously chosen by investors), then transaction costs would, as usual, drive up yields. Equivalently, greater bond liquidity would be associated with lower required yields. This is clear from expression (7): If π_i is constant and the cost t_B of trading bond B increases, its yield increases relative to that of bond A. The reason is that the buyer of the asset anticipates trading costs that must be compensated. As these costs only materialize if the holder trades at date 1, t_B is weighed with the probability of this event, which is π_B .

The second feature of expression (7) is that the yield differential increases in fundamental risk: The higher the risk of default of, say, bond B, which is $1-q_B$, the higher its required yield compared to bond A. Given the absence of risk aversion in our model, this is not a risk premium but simply reflects the reduction in the discounted expected payoff of the bond. The impact of fundamental

⁷In a closed economy with no short-selling allowed, there could be an equilibrium in which expression (3) holds with strict inequality. Clearly, this is of no interest in our setting.

risk and of liquidity costs on yield spreads in (7) is standard and in line, for example, with the empirical findings of Beber et al. (2009) for euro government bonds.

However, the response of bond yields to changes in liquidity and risk becomes more interesting once one takes into account that the probability of liquidation is endogenous in this model. This probability itself reacts to changes in both transaction costs (the terms at which the "supply of liquidity" is available) and outside investment opportunities (that determine the "demand for liquidity" by market participants). Indeed, as we shall see, it also depends nontrivially on the interaction between the two.

As one would expect, both higher transaction costs (higher t_i) and less attractive aggregate investment opportunities (higher α) reduce a given bond's probability of liquidation (π_i):

$$(8) \qquad \frac{\partial \pi_i}{\partial t_i} = -\frac{1+r}{\alpha(1-t_i)^2} g\left(\frac{\overline{r}}{\alpha} - \frac{1+r}{\alpha(1-t_i)}\right) < 0,$$

(9)
$$\frac{\partial \pi_i}{\partial \alpha} = -\frac{1}{\alpha^2} \left(\overline{r} - \frac{1+r}{1-t_i} \right) g \left(\frac{\overline{r}}{\alpha} - \frac{1+r}{\alpha(1-t_i)} \right) < 0.$$

Intuitively, if transaction costs increase (i.e., the bond becomes less liquid), then liquidating it becomes less attractive, hence the probability of selling decreases. Similarly, if the aggregate risk factor increases, the investors' market investment opportunities become less attractive, which again decreases their probability of selling the bond.⁸

Equipped with these results, we can now explore how risk and liquidity affect bond yields in our simple model. For simplicity, we perform the comparative statics of the yield differential (7) with respect to t_B only (the effect of t_A is analogous):

(10)
$$\frac{\partial \Delta Y}{\partial t_B} = \pi_B + \frac{\partial \pi_B}{\partial t_B} t_B,$$

(11)
$$\frac{\partial \Delta Y}{\partial \alpha} = \frac{\partial \pi_B}{\partial \alpha} t_B - \frac{\partial \pi_A}{\partial \alpha} t_A + \frac{\partial (q_B - q_A)}{\partial \alpha},$$

(12)
$$\frac{\partial^2 \Delta Y}{\partial t_B \partial \alpha} = \frac{\partial \pi_B}{\partial \alpha} + \frac{\partial^2 \pi_B}{\partial t_B \partial \alpha} t_B.$$

The first term in equation (10) is positive, the second negative. Hence, the sign is a priori ambiguous. This ambiguity reflects two opposing effects: direct and indirect. The direct effect of higher transaction costs is to drive up the price by a factor π_B . With an exogenous liquidation probability, this would be the only

⁸We do not report the joint effect of an increase in transaction costs and in aggregate risk, $\partial^2 \pi_i/\partial t_i \partial \alpha$. In our model, this cross-derivative cannot be signed unambiguously, but it is positive for several examples of distributions G, including the uniform distribution. If it is positive, an increase in transaction costs (i.e., a decrease in liquidity) reduces the absolute value of the impact of the aggregate factor on trading volume π_i , hence, there is a flight to liquidity. Yet, as we shall see, this term will be of second order in the derivation of the relevant derivatives for yield spreads and can be neglected in a first approximation.

effect, as observed above in discussing expression (7). But when liquidation is endogenously determined, this effect is at least partly counteracted by a reduction of the trading probability, as described in expression (8). As this probability decreases, the relative price of the bond increases, because liquidation costs are incurred less often. Hence this effect is proportional to t_B . To the extent that bidask spreads in the bond market are empirically very small (see Table 1), we expect the latter effect to be small in absolute terms and the overall effect to be positive.

The impact of risk on yield differentials in equation (11) has the first two terms of small size and opposite sign, so that one would expect its overall sign to depend mainly on the sign of the third term, $\partial(q_A-q_B)/\partial\alpha$, which measures the impact of aggregate risk on the fundamental risk of the two bonds. Two possibilities exist in principle. First, the effect of aggregate risk is similar for both bonds, in which case the term is approximately 0. Alternatively, the riskier bond is more sensitive to aggregate risk than the safer one, in which case we have $\partial(q_A-q_B)/\partial\alpha>0$. In this case we have the well-known "flight to quality": An increase in aggregate risk makes the safer bond more attractive than the riskier one and hence drives up the yield differential. This is the typical effect stressed by practitioners. Our empirical strategy is to let the data tell us the sign of the derivative. In Appendix A we show how the positive sign arises in a simple model of imperfect but constant correlation of bond returns.

Finally, the joint effect of changes in risk and liquidity is again ambiguous but has one dominant element. As seen above, while the direct effect of aggregate risk on the probability of liquidation—the first term in equation (12)—is always negative, the second term is ambiguous. This second term is the product of the flight to liquidity in terms of order flow and the transaction cost t_B . Since t_B is small, we expect the negative first term to dominate.

We summarize the previous discussion in the following testable predictions regarding the effects of risk and illiquidity on the yield differential, where, recalling that changes in t_B are mirror images of those in t_A , we measure illiquidity by the transaction cost differential Δt :

Hypothesis 1. The yield differential is insensitive to aggregate risk if aggregate risk affects bonds of different fundamental value identically. It is increasing in aggregate risk if the fundamentals of riskier bonds react more to aggregate risk than those of less risky ones:

$$\frac{\partial \Delta Y}{\partial \alpha} \geq 0.$$

Hypothesis 2. The yield differential between the two bonds depends positively on their transaction cost differential:

$$\frac{\partial \Delta Y}{\partial \Delta t} > 0.$$

Hypothesis 3. The positive effect of transaction costs on the yield differential is dampened by aggregate risk:

$$\frac{\partial^2 \Delta Y}{\partial \Delta t \partial \alpha} < 0.$$

While the first two predictions are in line with those of other existing models of asset pricing and liquidity, the third prediction is specific to our model. If, for example, bond trades are driven by consumption or endowment shocks, as in the overlapping generations (OLG) model of Acharya and Pedersen (2005) or the theoretical liquidity literature building on Diamond and Dybvig (1983), then π_i would be constant, and from expression (7) yield differentials would only depend on liquidity and aggregate risk directly. The cross effect identified in expression (15) would be 0 in this case. If bond sales were triggered by changes in consumption opportunities or other "supply side" considerations as in the delegated-portfolio management model of Vayanos (2004) rather than from changes in the risk-return profile of alternative investments, π_i would depend positively on α rather than negatively, as in our model, and the cross-effect expression (15) would be positive. Then aggregate risk would amplify rather than dampen liquidity effects.

V. Empirical Evidence

The empirical strategy used to test the predictions of Section IV is based on the estimation of a simultaneous equation model for yield differentials in the euro area at different maturities.

We measure the aggregate risk factor by the spread between j-year fixed interest rates on U.S. swaps, $R_{SWUS,t}^{j}$, and the yield on j-year U.S. government bonds, $R_{\text{US},t}^{j}$. We opt for this measure because of its high correlation with all U.S.-based measures of risk and because of its availability at different maturities. In the next section we report estimates obtained using alternative measures of risk and show that our results are robust to the choice of risk measure. The robustness of our estimates to different proxies for risk is important, since the swap-Treasury differential could arguably reflect the taking on and unwinding of carry trades in the market: These trades (which are generally trades with a U.S. dollar leg) may explain why factors based on U.S. interest rates become international risk factors. While in principle this argument does not make this differential less suitable as a proxy for aggregate risk in the euro area, it raises the doubt that our proxy itself may be affected by liquidity. But our robustness analysis indicates that this is unlikely; the strong and positive comovements of all the measures of risk at different maturities make it difficult to interpret them as affected by carry trades, which should have different effects on prices at different maturities. As for the liquidity factor, we consider a range of alternative liquidity indicators and select the bid-ask spread as the most significant measure.

In taking the model to the data, we take the following specification strategy. First, we choose German bonds as benchmarks for 10-year maturity and French bonds for 5-year maturity. This choice is supported by the econometric evidence provided by Dunne, Moore, and Portes (2007) and by the fact that traders commonly regard the 10-year Bunds as the 10-year euro-area benchmark and French OATs (obligations assimilable du trésor) as the 5-year euro-area benchmark, because French bonds are seen as particularly liquid for the 5-year maturity bucket.

Second, as yield spreads in the euro area are very persistent and our predictions are derived within a static framework, we posit the following dynamic partial adjustment model for yield differentials:

$$Y_{i,t} - Y_{b,t} = \rho_i (Y_{i,t-1} - Y_{b,t-1}) + (1 - \rho_i) (Y_{i,t} - Y_{b,t})^* + u_{i,t},$$

where $u_{i,t}$ are independent and identically distributed shocks and $(Y_{i,t} - Y_{b,t})^*$ is the theory-consistent long-run equilibrium value for yield differentials. Third, we augment the specification with the differentials in the residual maturity of the benchmark bonds in country i and the benchmark country in order to filter out of the data the effect introduced by the different maturity of benchmark bonds and the effect of changes in benchmarks occurring at different dates for different countries in the sample period.

A. The Baseline Model

To sum up, we estimate as a baseline model the following 8-equation model, where the dependent variables are the yield differentials relative to a benchmark government bond for the other 8 countries listed in Section III:

$$\begin{split} Y_{i,t}^{j} - Y_{b,t}^{j} &= \rho_{i} \left(Y_{i,t-1}^{j} - Y_{b,t-1}^{j} \right) + (1 - \rho_{i}) \left(Y_{i,t}^{j} - Y_{b,t}^{j} \right)^{*} \\ &+ \delta_{i} \left(M_{i,t}^{j} - M_{b,t}^{j} \right) + u_{i,t}, \\ \left(Y_{i,t}^{j} - Y_{b,t}^{j} \right)^{*} &= \beta_{1,i} \left(c_{i,t}^{j} - c_{b,t}^{j} \right) + \beta_{2,i} \left(R_{\text{SWUS},t}^{j} - R_{\text{US},t}^{j} \right) \\ &- \beta_{3,i} \left(c_{i,t}^{j} - c_{b,t}^{j} \right) \left(R_{\text{SWUS},t}^{j} - R_{\text{US},t}^{j} \right), \end{split}$$

where $(Y_{i,t}^j-Y_{b,t}^j)^*$ is the long-run equilibrium yield differential consistent with the theory; $M_{i,t}^j-M_{b,t}^j$ are differentials in the residual maturity of the benchmark bonds in country i and the benchmark country; $R_{\mathrm{SWUS},t}^j-R_{\mathrm{US},t}^j$ is the spread between j-year fixed interest rates on U.S. swaps, $R_{\mathrm{SWUS},t}^j$, and the yield on j-year U.S. government bonds, $R_{\mathrm{US},t}^j$; and $c_{i,t}^j-c_{b,t}^j$ is the differential between the bidask spread of bonds in country i and the bid-ask spread of bonds in the benchmark country. The index i varies across countries and the index j varies across maturities (5 and 10 years). The estimation is performed by seemingly unrelated regression (SUR), and the empirical results are shown in Panels A and B of Table 4.

The estimates for the 10-year maturity yield differential are presented in Panel A of Table 4. The coefficient of the lagged dependent variable is always

⁹We also tried different methods of dealing with these problems such as omitting them from the sample dates in which benchmarks are changed or constructing constant maturity yields. We favor the use of the maturity differentials in that it is a natural way of correcting the differentials and it allows our liquidity indicator to operate during episodes in which liquidity might highly matter, such as at dates when benchmarks are changed. An alternative to the maturity differential is the duration differential. However, the difference between these two measures is not very relevant in our case given that they both act as dummy variables to model the same jump in duration and maturity occurring in the occasion of benchmark changes.

TABLE 4
Estimation of a System of Simultaneous Equations for Euro-Area Yield Differentials

The equations are estimated by the seemingly unrelated regression (SUR) method on a sample of daily observations from January 1, 2002 to December 23, 2003. Panel A in Table 4 shows the coefficient estimates for the 10-year maturity; spreads are on German bonds. Panel B shows coefficients estimates results for the 5-year maturity; spreads are on French bonds. Standard errors are reported within parentheses below the coefficient estimates. ** and * indicate that the corresponding coefficient is significantly different from 0 at the 5% and 10% levels, respectively.

Variable	Constant	Own Lag	Maturity	Risk Factor	Bid-Ask Spread	Interaction		
Panel A. 10-Year Yield Differentials								
Austria	-0.167**	0.857**	0.280**	0.546**	0.043**	-0.077**		
	(0.026)	(0.016)	(0.034)	(0.060)	(0.014)	(0.026)		
Belgium	-0.129**	0.936**	0.357**	0.497**	0.052**	-0.099**		
	(0.021)	(0.007)	(0.040)	(0.043)	(0.022)	(0.048)		
Spain	-0.135**	0.867**	0.349**	0.485**	0.007	-0.009		
	(0.034)	(0.018)	(0.061)	(0.077)	(0.024)	(0.047)		
Finland	-0.159**	0.956**	0.207**	0.467**	-0.01	-0.025		
	(0.049)	(0.006)	(0.045)	(0.118)	(0.024)	(0.079)		
France	-0.119**	0.945**	0.184**	0.321**	0.016	-0.025		
	(0.038)	(0.01)	(0.077)	(0.072)	(0.038)	(0.079)		
Italy	-0.077**	0.912**	0.288**	0.290**	0.017	-0.042		
	(0.021)	(0.01)	(0.037)	(0.047)	(0.018)	(0.043)		
Netherlands	-0.076**	0.891**	0.314**	0.305**	0.034**	-0.052*		
	(0.019)	(0.012)	(0.029)	(0.042)	(0.016)	(0.032)		
Portugal	-0.150**	0.920**	0.384**	0.633**	0.080**	-0.139**		
	(0.044)	(0.010)	(0.052)	(0.099)	(0.033)	(0.07)		
Panel B. 5-Year	Yield Differential	<u>'s</u>						
Austria	-0.251**	0.833**	0.170**	0.679**	0.079**	-0.184**		
	(0.039)	(0.017)	(0.011)	(0.09)	(0.023)	(0.048)		
Belgium	-0.082**	0.774**	0.214**	0.297**	-0.022	-0.033		
	(0.015)	(0.016)	(0.008)	(0.034)	(0.018)	(0.042)		
Spain	-0.143**	0.693**	0.210**	0.337**	0.048**	-0.095**		
	(0.013)	(0.021)	(0.007)	(0.03)	(0.020)	(0.041)		
Finland	-0.106**	0.606**	0.205**	0.258**	-0.018	-0.025		
	(0.017)	(0.022)	(0.005)	(0.041)	(0.012)	(0.024)		
Germany	-0.017	0.742**	0.168**	0.01	-0.007	0.004		
	(0.015)	(0.018)	(0.007)	(0.03)	(0.012)	(0.026)		
Italy	-0.043**	0.584**	0.172**	0.231**	0.107**	-0.208**		
	(0.016)	(0.03)	(0.006)	(0.028)	(0.017)	(0.032)		
Netherlands	-0.123**	0.563**	0.191**	0.317**	0.017**	-0.045**		
	(0.016)	(0.021)	(0.04)	(0.036)	(0.009)	(0.020)		
Portugal	-0.122**	0.853**	0.240**	0.458**	0.052**	-0.125**		
	(0.022)	(0.010)	(0.006)	(0.05)	(0.018)	(0.04)		

significant and close to unity, which indicates strong persistence of yield differentials. Also, the coefficient of the maturity differential variable is uniformly positive and significant, confirming the importance of this correction. The corresponding results for the 5-year maturity are shown in Panel B of Table 4. Again, the coefficient of the lagged dependent variable is positive and significant, but it is smaller for all 8 countries, indicating lower persistence in the time-series behavior of the 5-year yield differentials. Also, the maturity correction coefficient stays positive and significant for all 8 countries.

More importantly, the coefficient of the aggregate risk factor is positive and significantly different from 0 for all 8 countries in both maturities, in accordance with Hypothesis 1 of Section IV. It ranges between 0.3 and 0.6 for the 10-year bonds and between 0.23 and 0.68 for the 5-year bonds (except in the latter case for Germany, where the coefficient is virtually 0). Interpreting these positive

coefficients from the perspective of the theory summarized in Hypothesis 1 leads us to conclude that the fundamental risk of nonbenchmark bonds is perceived to be more strongly affected by changes in aggregate risk than that of benchmark bonds. So higher aggregate risk, as proxied by our U.S. swap-yield differential, is correlated with wider euro-area yield differentials relative to the Bund (or, respectively, the OAT).

Turning to the liquidity variable, we note that at the 5-year maturity the coefficients of the liquidity variables are positive and significant in the case of Austria, Belgium, Spain, Italy, Netherlands, and Portugal, in line with Hypothesis 2 of Section IV.

Finally, in line with Hypothesis 3, the interaction between liquidity and the aggregate risk factor always has a negative impact on the yield differentials. This finding is substantially confirmed at the 10-year maturity level, although the results are weakened in the case of Spain and Italy. This evidence illustrates the importance of nonlinearities in the effect of liquidity indicators on yield differentials. Interestingly, the coefficient of the liquidity differential variable becomes significant only when the interaction between liquidity indicators and the aggregate risk factor is also included in the regression. If the coefficient of the interaction is constrained to 0, then the level of the liquidity indicator also becomes insignificant.

This evidence does not simply reflect the fact that for less liquid bonds, prices take more time to absorb the change in risks. In fact, we control for different dynamic effects across countries of the variables included in our model by having potentially different coefficients on the lagged dependent variable. Moreover, a simple check, effected by adding further lags of the included variables, delivers nonsignificant parameters for higher order dynamics.

It could be observed that our SUR estimation is inefficient when valid crossequation restrictions can be imposed on our model. This argument is strengthened in the context of our theoretical model, where the cross-equation restrictions on the coefficients on the measure of liquidity and on the interaction between the aggregate risk factor and this measure are indeed implied by theory.

In Table 5 we explore this possibility by imposing cross-equation restrictions on our estimated models for 5- and 10-year differentials. We test for the validity of cross-equation restrictions on each coefficient separately and on the full set of coefficients. Interestingly, the Wald statistics reported in Table 5 illustrate that the panel restrictions can only be validly imposed on the liquidity indicators at the 10-year maturity. When these restrictions are imposed, the effect of the liquidity variables is significant and in line with the prediction of the theory. Yet this result does not carry over to the 5-year maturity, where the panel restrictions cannot be imposed on the liquidity variables. Furthermore, the panel restrictions on the coefficient of the aggregate risk factor are always rejected, in line with the predictions of the model where the impact of international risk on fundamental risk is heterogenous across different bonds.

Summing up, our empirical results are generally supportive of the implications of our theoretical framework: In particular, the aggregate risk factor always has a positive and significant effect on yields, and there is an important negative interaction between liquidity and the international factor.

TABLE 5 Testing Panel Restrictions

Table 5 is based on fixed-effects panel estimates for the 10- and 5-year yield differentials. The *p*-value of the Wald test of the identity restriction of individual coefficients for all 8 countries is shown on the right of the relevant coefficient. The *p*-value of the Wald test of the identity restriction of all the coefficients for all 8 countries is shown in the bottom row. Standard errors (SE) are reported within parentheses below the coefficient estimates. ** indicates that the corresponding coefficient is significantly different from 0 at the 5% level.

	10-Year Yield E	Differentials	5-Year Yield Differentials			5-Year Yield Differentials	
Variable	Coefficient and SE	Wald p-Value	Coefficient and SE	Wald p-Value			
Own lag	0.956** (0.006)	0.000	0.853** (0.006)	0.000			
Maturity	0.269** (0.041)	0.000	0.232** (0.003)	0.000			
Risk factor	0.172** (0.063)	0.000	0.372** (0.032)	0.000			
Bid-ask spread	0.047** (0.021)	0.207	0.039** (0.008)	0.000			
Interaction	-0.064** (0.033)	0.192	-0.087** (0.018)	0.000			
Panel restriction		0.000		0.000			

B. Robustness

Swap spreads can be considered a good measure of risk for a number of reasons. First, they are effectively the difference between the yields to maturity of 2 bonds with the same maturity. By no arbitrage, long-term yields to maturity can be exactly decomposed in a component reflecting the discounted sum of future 1-period bond yields and in a component reflecting term premiums. Taking the difference between two yields of exactly the same maturity, we are left with a quantity that exclusively reflects term premiums. Second, swap spreads are available at the different maturities relevant to our study, thus enabling us to account for a nonflat term structure of risk premiums. Third, U.S. swap spreads provide a measure of risk that does not refer to the European market, and therefore they are much more likely to be an exogenous variable in the estimation of the parameters of interest than any measure based on European yields. Fourth, as a spread between homogenous types of bonds, swap spreads are a superior measure of risk compared to the spread between Treasury bonds and corporate bonds.

However, it must be recognized that swap spreads are a special measure of risk in that they include the counterparty risk of swap dealers¹² and occasionally might reflect factors that are not related to aggregate risk. A close examination of Figure 3 reveals that the positive and strong comovement between the first principal component of yield differentials in the euro area and our measure of

¹⁰Note that spreads can still depend on the level of interest rates, but only to the extent that the level of interest rates is differently reflected in the term premiums associated to the two bonds on which yield differentials are computed.

¹¹Duffee (1998) notes that the spread between Treasury bonds and corporate bonds is a spread between callable bonds and a mixture of callable and noncallable bonds. Given that the response of callable and noncallable bonds to shocks in the level of the term structure is different, the government-corporate spread is sensitive to the level of the term structure.

¹²However, in practice this effect is minimal (see, e.g., Duffie and Huang (1996)).

risk has a clear exception in late July 2003. At this date, swap spreads suddenly increased for reasons related to the hedging of mortgage-backed securities and hence little related to international factors. It is therefore important to assess the robustness of our results.

We provide the relevant evidence in Table 6, where we report the results of reestimating our model for the 10-year yield differentials on a shorter sample that

TABLE 6 Robustness Analysis

Table 6 reports robustness analysis for the SUR system on 10-year yield differentials. We consider 3 alternative risk factors: RF1 is the swap spread; RF2 is the differential between yields on seasoned U.S. BAA bonds and seasoned U.S. AAA bonds (drawn from the Federal Reserve Economic Data (FRED)); and RF3 is the implied volatility in options on the EURO STOXX 50. The source for non-U.S. data is Datastream. Standard errors are reported within parentheses below the coefficient estimates. ** and * indicate that the corresponding coefficient is significantly different from 0 at the 5% and 10% levels, respectively.

Sample	Constant	Own	Maturity	RF1	RF2	BE3	Bid- Ask	Interaction
								-0.089**
Jan. 2002–Dec. 2003 Jan. 2002–Dec. 2003	(0.021) -0.324** (0.034) -0.199** (0.021)	(0.022) 0.811** (0.018) 0.813** (0.018)	(0.027) 0.237** (0.026) 0.308** (0.027)	(0.046) 0.388** (0.052) 0.492** (0.047)	0.193** (0.034)	0.178** (0.032)	(0.011) 0.035** (0.010) 0.031** (0.010)	(0.020) -0.062** (0.019) -0.051** (0.020)
Jan. 2002–June 2003 Jan. 2002–Dec. 2003	-0.081** (0.018) -0.252**	0.906** (0.011) 0.925**	0.317** (0.034) 0.327**	0.433** (0.035) 0.384**	0.148**		0.081** (0.019) 0.042**	-0.14** (0.039) -0.08** (0.039)
Jan. 2002-Dec. 2003	-0.177** (0.019)	0.917**	0.343** (0.031)	0.488** (0.033)	(0.041)	0.174** (0.003)	0.042** (0.019)	-0.077** (0.037)
Jan. 2002-June 2003	-0.10** (0.017)	0.77** (0.02)	0.324** (0.045)	0.465** (0.059)			0.034* (0.019)	-0.06* (0.034)
Jan. 2002-Dec. 2003 Jan. 2002-Dec. 2003	-0.33** (0.05) -0.17**	0.82** (0.02) 0.83**	0.239** (0.05) 0.343** (0.048)	0.284** (0.073) 0.417**	0.242** (0.053)	0.203**	0.02 (0.02) -0.002	0.006 (0.035) 0.025 (0.04)
Jan. 2002-June 2003	-0.110**	0.953**	0.127*	0.433**		(0.000)	-0.02	-0.017 (0.06)
Jan. 2002-Dec. 2003	-0.377** (0.07)	0.944** (0.01)	0.166** (0.043)	0.312** (0.101)	0.250** (0.07)		-0.006 (0.02)	0.02 (0.03)
Jan. 2002-Dec. 2003	-0.190** (0.045)	0.950** (0.006)	0.282** (0.05)	0.527** (0.107)		0.01 (0.08)	-0.01 (0.02)	0.03 (0.04)
Jan. 2002–June 2003 Jan. 2002–Dec. 2003	-0.035 (0.039) -0.176**	0.945** (0.02) 0.944**	0.053 (0.073) 0.212**	0.184** (0.069) 0.293**	0.057		0.042 (0.034) 0.01	-0.062 (0.067) -0.02
Jan. 2002-Dec. 2003	(0.069) -0.169** (0.033)	(0.01) 0.930** (0.012)	(0.08) 0.163** (0.06)	(0.09) 0.307** (0.056)	(0.07)	0.188** (0.056)	(0.04) 0.01 (0.03)	(0.07) -0.02 (0.06)
Jan. 2002–June 2003	-0.027 (0.021)	0.88**	0.263** (0.040)	0.242** (0.043)	0.114**		0.02 (0.016)	-0.042 (0.039) -0.03
Jan. 2002-Dec. 2003	(0.038) -0.06** (0.015)	(0.02) 0.86** (0.01)	(0.040) 0.237** (0.02)	(0.043) 0.292** (0.03)	(0.03)	0.187** (0.023)	(0.01) 0.014 (0.01)	(0.03) -0.036 (0.028)
Jan. 2002-June 2003	-0.076** (0.02)	0.88** (0.012)	-0.07** (0.026)	0.329** (0.037)			0.046** (0.016)	-0.072** (0.032)
Jan. 2002–Dec. 2003	-0.183** (0.03)	0.86**	0.28** (0.025)	0.180** (0.04)	0.130** (0.03)	0.222**	0.028** (0.013)	-0.042* (0.026)
	(0.02)	(0.012)	(0.031)	(0.033)		(0.043)	(0.034)	-0.031** (0.014)
	(0.013)	(0.038)	(0.044)	(0.083)	0 224**		(0.029)	-0.173** (0.06) -0.10*
Jan. 2002-Dec. 2003	(0.057) -0.215** (0.029)	(0.012) 0.870** (0.013)	(0.046) 0.386** (0.032)	(0.095) 0.589** (0.061)	(0.06)	0.293** (0.038)	(0.027) 0.049** (0.020)	(0.057) -0.082** (0.04)
	Jan. 2002–Dec. 2003	Jan. 2002–June 2003	Sample Constant (0.021) Lag Jan. 2002–June 2003 -0.135** 0.775** Jan. 2002–Dec. 2003 -0.324*** 0.811** Jan. 2002–Dec. 2003 -0.199** 0.813** Jan. 2002–June 2003 -0.081** 0.906** (0.018) (0.011) 0.906** (0.039) (0.099) 0.925** (0.019) (0.008) 0.906** (0.019) (0.009) 0.925** (0.019) (0.009) 0.925** (0.019) (0.008) 0.906** (0.019) (0.008) 0.909* Jan. 2002–June 2003 -0.10** 0.77** (0.017) (0.02) 0.92* Jan. 2002–Dec. 2003 -0.33** 0.82** (0.027) (0.048) (0.009) Jan. 2002–Dec. 2003 -0.110** 0.953** (0.027) (0.048) (0.009) Jan. 2002–Dec. 2003 -0.110** 0.953** (0.047) (0.049) (0.009) Jan. 2002–Dec. 2003 -0.16	Sample Constant Lag Maturity Jan. 2002–June 2003 -0.135** 0.775** 0.285** Jan. 2002–Dec. 2003 -0.324** 0.811** 0.237** Jan. 2002–Dec. 2003 -0.199** 0.813** 0.308** Jan. 2002–June 2003 -0.199** 0.813** 0.308** Jan. 2002–Dec. 2003 -0.282** 0.996** 0.317** (0.018) (0.011) (0.034) Jan. 2002–Dec. 2003 -0.252** 0.925** 0.327** (0.019) (0.009) (0.043) Jan. 2002–June 2003 -0.17** 0.91** 0.343** (0.019) (0.009) (0.043) Jan. 2002–Dec. 2003 -0.17** 0.91** 0.343** (0.017) (0.029) (0.045) Jan. 2002–Dec. 2003 -0.17** 0.82** 0.239** (0.027) (0.048) (0.049) (0.045) Jan. 2002–Dec. 2003 -0.110** 0.953** 0.127* (0.027) (0.048) (0.049) (0.07)	Sample Constant Lag Maturity RF1 Jan. 2002–June 2003 -0.135** 0.775** 0.285** 0.503** Jan. 2002–Dec. 2003 -0.324** 0.811** 0.237** 0.388** Jan. 2002–Dec. 2003 -0.199** 0.813** 0.308** 0.492** Jan. 2002–June 2003 -0.199** 0.813** 0.308** 0.492** Jan. 2002–Dec. 2003 -0.252** 0.906** 0.317** 0.433** (0.018) (0.011) (0.034) (0.035) Jan. 2002–Dec. 2003 -0.252** 0.925** 0.327** 0.384** (0.018) (0.019) (0.043) (0.053) Jan. 2002–Dec. 2003 -0.17** 0.91** 0.343** 0.465** Jan. 2002–Dec. 2003 -0.17** 0.91** 0.343** 0.465** Jan. 2002–Dec. 2003 -0.33** 0.82** 0.234** 0.465** Jan. 2002–Dec. 2003 -0.33** 0.82** 0.234** 0.45** Jan. 2002–Dec. 2003 -0.110** 0.953** <td< td=""><td>Sample Constant Lag Maturity RF1 RF2 Jan. 2002–June 2003 -0.135** 0.775** 0.285** 0.503** -0.193** Jan. 2002–Dec. 2003 -0.324** 0.811*** 0.237** 0.388** 0.193** Jan. 2002–Dec. 2003 -0.199** 0.813** 0.308** 0.492** Jan. 2002–June 2003 -0.081** 0.906** 0.317** 0.433** Jan. 2002–Dec. 2003 -0.252** 0.995** 0.327** 0.384** 0.148** Jan. 2002–Dec. 2003 -0.252** 0.995** 0.327** 0.384** 0.148** Jan. 2002–Dec. 2003 -0.177** 0.917** 0.343** 0.488** 0.148** Jan. 2002–Dec. 2003 -0.10** 0.77** 0.324** 0.485** 0.242** Jan. 2002–Dec. 2003 -0.10** 0.77** 0.324** 0.45** 0.242** Jan. 2002–Dec. 2003 -0.110** 0.95** 0.239** 0.24** 0.242** Jan. 2002–Dec. 2003 -0.110** 0.95** 0.12**</td><td>Sample Constant Lag Maturity RF1 RF2 RF3 Jan. 2002–June 2003 -0.135** 0.775** 0.285** 0.503** 1</td><td> Sample Constant Lag Maturity RF1 RF2 RF3 Ask </td></td<>	Sample Constant Lag Maturity RF1 RF2 Jan. 2002–June 2003 -0.135** 0.775** 0.285** 0.503** -0.193** Jan. 2002–Dec. 2003 -0.324** 0.811*** 0.237** 0.388** 0.193** Jan. 2002–Dec. 2003 -0.199** 0.813** 0.308** 0.492** Jan. 2002–June 2003 -0.081** 0.906** 0.317** 0.433** Jan. 2002–Dec. 2003 -0.252** 0.995** 0.327** 0.384** 0.148** Jan. 2002–Dec. 2003 -0.252** 0.995** 0.327** 0.384** 0.148** Jan. 2002–Dec. 2003 -0.177** 0.917** 0.343** 0.488** 0.148** Jan. 2002–Dec. 2003 -0.10** 0.77** 0.324** 0.485** 0.242** Jan. 2002–Dec. 2003 -0.10** 0.77** 0.324** 0.45** 0.242** Jan. 2002–Dec. 2003 -0.110** 0.95** 0.239** 0.24** 0.242** Jan. 2002–Dec. 2003 -0.110** 0.95** 0.12**	Sample Constant Lag Maturity RF1 RF2 RF3 Jan. 2002–June 2003 -0.135** 0.775** 0.285** 0.503** 1	Sample Constant Lag Maturity RF1 RF2 RF3 Ask

excludes the July 2003 episode. The table also reports the evidence obtained by augmenting the baseline regression with two alternative measures of risk. The first is the yield spread between BBB long-term corporate bonds and AAA long-term corporate bonds; the second is an indicator based on the European equity market: the implied volatility from options on the EURO STOXX 50. The results show that our estimates are robust both to the choice of the sample size and to the use of different measures of risk. In particular, the results on the shorter sample wholly confirm the evidence from our full sample with some slight modification of the original point estimates. Augmentation of the model with alternative measures of risk shows that although all alternative measures of risk are significant, their inclusion does not affect the significance of all variables included in the original model. Overall, the significance of risk factors is more robust than that of the liquidity factor and of the interaction term. We perform similar robustness checks for the 5-year differentials, but for brevity we do not report the corresponding results, which confirm those obtained for the 10-year spreads.

In the case of the 5-year bonds we also reestimate the model with the German Bund as a benchmark instead of the French OAT. This modification leads to much less precise estimates of all relevant parameters and to a set of results that are much less consistent with those obtained for the 10-year differentials. We take this as confirmatory evidence of the econometric analysis by Dunne et al. (2007) that indicates the OAT as the preferred choice of benchmark for the 5-year maturity.

Having chosen a U.S.-based measure of aggregate risk, a final robustness check concerns the inclusion of the dollar-euro exchange rate as an independent source of risk. In all specifications, this variable is insignificant.

VI. Conclusions

This paper explores the determinants of observed yield differentials between sovereign bonds. It does so by drawing on data for long-term bond yields in the euro area. Inspection of daily data for the early EMU period indicates that there is a strong comovement in yield differentials of benchmark bonds, and that their first principal components explain more than 90% of the variance. This common trend appears highly correlated with measures of aggregate risk. In contrast, liquidity differentials—proxied, for example, by bid-ask spread differentials—feature sizeable heterogeneity and no common factor.

To generate testable predictions about the relation between yield differentials, fundamental risk, and liquidity, we develop a simple model of the interplay between aggregate risk and transaction costs. Our model has two key novel features. First, the demand for liquidity responds to the magnitude of transaction costs and to changes in investors' opportunity set (rather than being determined by exogenous liquidation needs). Specifically, investors value liquidity less when alternative investment opportunities become less attractive. Second, if the payoff of alternative investments covaries with that of marketable assets, both driven by a common factor, the equilibrium value of liquidity tends to be lower in worse aggregate states, as determined by this common factor.

The model predicts that bond yield differentials should increase in both liquidity and aggregate risk, with a nonlinear term that captures the effect on required

bond returns due to the interaction between liquidity and the aggregate factor. We test these predictions on daily data for the euro-area sovereign yield differentials in 2002 and 2003. The econometric results show that the aggregate risk factor is consistently priced, while liquidity differentials are priced only for a subset of 9 country/maturity pairs (out of a total of 16), and that the interaction of liquidity differentials with the risk factor is always negative when significant, in line with the prediction of the model. Our findings are also consistent with those of recent papers that find that the effect of liquidity variables on bond returns is not economically important when considered in isolation. In fact, when the coefficient of the interaction term between aggregate risk and liquidity is constrained to be 0, liquidity becomes insignificant throughout our regressions.

Our data do not allow us to draw conclusions regarding the effect of cross-sectional differences in macroeconomic conditions or fiscal policies. For example, simple cross-country regressions show that on average over the sample period, (average) yield spreads are positively correlated with (average) government debt/GDP ratios, which in turn are negatively correlated with (average) bid-ask spreads. But being based on 9 data points only, these cross-sectional estimates have little reliability, and in our time-series analysis we do not have sufficient variation in macroeconomic variables (such as debt/GDP ratios) to draw conclusions about possible macroeconomic determinants of the variables that we observe.

The implications of our analysis for policymakers and for portfolio managers are rather more subtle. From a policymaking standpoint, the empirical estimates highlight the importance of the aggregate risk factor in determining bond yield spreads, and they thus underscore the importance of good macroeconomic fundamentals to minimize exposure to the aggregate risk factor. Instead, there seems to be little need for further action on the liquidity side, because bid-ask spreads are already rather uniform and very small across European bond markets, at least for benchmark bonds. Instead, the lesson for portfolio management is that liquidity can affect the risk sensitivity of the assets being held, and that this interaction depends on the covariance of illiquidity costs with aggregate risk. In this sense, the lesson of our model, in spite of its simplicity, may be considerably more general than our specific application to euro-area bond markets.

Appendix A. Extension to Risky Portfolios

In Appendix A, we sketch an extension of the model in which the investors' investment opportunities are risky and imperfectly correlated. As in the main text of the paper, the two bonds can either repay $(\widetilde{V}_i = V)$ or default $(\widetilde{V}_i = 0)$. The probabilities are

$$(\widetilde{V}_A,\widetilde{V}_B) = \begin{cases} (V,V) & \text{with probability } \gamma + \alpha \\ (V,0) & \text{with probability } p_{V0} \\ (0,V) & \text{with probability } 1 - \alpha - \gamma - p_{V0} \\ (0,0) & \text{with probability } \alpha \end{cases}.$$

It is straightforward to show that the correlation between the two returns is

$$\rho = \frac{(\alpha + \gamma + p_{V0})(p_{V0} + \alpha) - p_{V0}}{\sqrt{(\alpha + \gamma + p_{V0})(1 - \alpha - \gamma - p_{V0})(p_{V0} + \alpha)(1 - p_{V0} - \alpha)}}.$$

Holding the correlation constant, for small values of α (which is the relevant case), an increase in α increases the variance of both bonds, hence α can be viewed as an indicator of aggregate risk. If we hold ρ constant, this structure has two free parameters: namely, the aggregate factor α and the relative attractiveness of bond A, which can be measured by γ .

The assumption $q_A \ge q_B$ (bond A is the benchmark) implies that $\alpha + \gamma + 2p_{V0} \ge 1$. Under this assumption, for ρ constant, it is straightforward (though tedious) to calculate that $(\partial/(\partial\alpha))(q_A - q_B) \ge 0$ for small values of α , as postulated in Section IV.

The alternative investment opportunity of investor h pays off as follows:

$$\widetilde{R}^h \quad = \quad \left\{ \begin{array}{ll} \overline{r} - \sqrt{e^h} & \text{with probability } \alpha \\ \overline{r} & \text{with probability } s \\ \overline{r} + \sqrt{e^h} & \text{with probability } \alpha \end{array} \right. ,$$

where s > 0, $s + 2\alpha = 1$, $0 \le e^h \le e$. For each h we have $E\widetilde{R}^h = \overline{r}$ and $var(\widetilde{R}^h) = 2\alpha e^h$. Hence, higher α means higher risk, and the investor bears more of the risk the greater her e^h . If she is risk-averse, she will therefore behave as if she maximized the certainty equivalent $\overline{r} - \alpha e^h$ considered in the main text.

If risk-averse, the investor's objective is to maximize Eu(c), where u is increasing and concave,

$$c = x_A \widetilde{V}_A + x_B \widetilde{V}_B + x_C (1+r) + x_B \widetilde{R}^h$$

is final consumption, and x_i are the investor's end-of-period-1 holdings of the 4 assets.

Because of transaction costs, the first-order condition of this problem depends on whether the investor buys or sells in equilibrium. As an example, consider the case where the investor sells both bonds in order to invest alternatively. In this case, her first-order conditions are

$$\begin{split} \mathrm{E}(\widetilde{V}_A - (1+r)(1-t_A)p_{1A})u'(c) &= 0, \\ \mathrm{E}(\widetilde{V}_B - (1+r)(1-t_B)p_{1B})u'(c) &= 0, \\ \mathrm{E}(\widetilde{R}^h - (1+r))u'(c) &= 0. \end{split}$$
 and

Hence.

$$(1-t_i)p_{1i}\mathrm{E}[\widetilde{R}^hu'(c)] = \mathrm{E}[\widetilde{V}_iu'(c)], \qquad i=A,B$$

$$\Leftrightarrow \operatorname{cov}\left(\frac{\widetilde{V}_i}{p_{1i}} - (1-t_i)\widetilde{R}^h, u'(c)\right) = \left(\frac{\overline{V}_i}{p_{1i}} - (1-t_i)\overline{r}\right)\mathrm{E}u'(c).$$

This condition on how the excess return of each bond over the investor's (transaction costs adjusted) alternative return must covary with marginal utility is a standard CAPM-type condition, with the new feature that the alternative return is investor-specific. This "investor-specific β " gives rise to a supply curve of bonds that is increasing in p_{1i} , i = A, B, and decreasing in aggregate risk α , as used in the main text.

Appendix B. Data Description

The data for 5- and 10-year-maturity bonds from January 1, 2002 to December 23, 2003 are collected from Euro MTS Group's European Benchmark Market (EBM) trading platform at 11AM CET during all market days in the Telematico cash markets. The database contains the best bid or ask prices across all markets, the aggregate quantity of all of the outstanding proposals on the basis of the best bid and best ask prices, and the daily trading volume of each bond on the EBM. From these data we calculate redemption yields, maturities, and a set of time-varying liquidity variables for the benchmark bonds of each country in our sample. The countries are Austria, Belgium, Finland, France, Germany,

Italy, the Netherlands, Portugal, and Spain. We constructed the following liquidity variables (in all cases as the difference between the relevant country's value and the value observed for the benchmark, which was Germany for the 10-year bucket and France for the 5-year bucket):

- i) the 5-day-moving-average of the bid-ask spread (in ticks);
- ii) trading volume for the benchmark bond, in million of euros;
- iii) bid-side market depth, defined as the difference between bid and mid price, divided by the bid quantity;
- iv) ask-side market depth, defined as the difference between mid price and ask price, divided by the ask quantity; and
- v) maximum quantity available at the best 5 prices.

After experimentation, we selected the bid-ask spread as the most significant liquidity indicator and reported the estimates of our nonlinear empirical model only for this liquidity indicator.¹³

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¹³This is confirmed for example by Beber et al. (2009), who find no significant difference in the impacts of the four liquidity variables they construct from the MTS database.

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