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THE IMPACT OF COMPETITION ON BANK ORIENTATION

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The Impact of Competition on Bank Orientation

Abstract

How do banks react to increased competition? Recent banking theory offers conflicting predictions about the impact of competition on bank orientation – i.e., the choice of relationship based versus transactional banking. We empirically investigate the impact of interbank competition on bank branch orientation. We employ a unique data set containing detailed information on bank-firm relationships. We find that bank branches facing stiff local competition engage considerably more in relationship-based lending. Our results illustrate that competition and relationships are not necessarily inimical.

Keywords: bank orientation, bank industry specialization, competition, lending relationships. JEL: G21, L11, L14.
I. Introduction

In their seminal paper in the *Quarterly Journal of Economics* Petersen and Rajan (1995) investigate the effects of competition between banks on the loan rate and the availability of bank credit to firms. Petersen and Rajan model how especially lower quality firms are negatively affected by competition between banks, as banks may be unwilling to invest in relationships by incurring initial loan losses when firms can later on obtain low loan rates in a competitive banking market. Petersen and Rajan document that young firms in more concentrated banking markets obtain more relationship benefits, i.e., lower loan rates and easier access to bank credit, than firms in more competitive banking markets.

However, recent work is starting to question whether credit market competition is always inimical to the formation of mutually beneficial relationships between firms and banks. Boot and Thakor (2000), for example, revisit the presumed incompatibility between competition and relationship finance and argue that the source of competition matters in the determination of *bank orientation* (i.e., relationship-based versus transactional lending). In their model, *interbank* competition actually increases the relative amount of relationship lending whereas *capital market* competition reduces relationship-lending chosen by banks. Their reasoning is that banks when faced with stiffer interbank competition have greater incentives to offer relationship loans. Relationship lending (compared to transactional lending) allows banks to shield rents more effectively, as relationship banking differentiates the lending bank better from competing banks.

A recent paper by Elsas (2005), published in this *Journal*, is the first to empirically study the determinants of relationship lending. Elsas employs a cross-sectional data set containing bank credit files on 122 large German firms to investigate the relationship between local bank market concentration and the likelihood a bank assesses itself to be the “Hausbank” of a firm. He documents a mostly decreasing relationship between concentration and the incidence of the Hausbank status.

However, his study suffers from serious shortcomings we aim to address in our empirical analysis. Indeed, his Hausbank measure is tenuous and subjective, affected by *framing* directly related to bank market concentration. First, Elsas analyzes a small sample of 122 German firms that are quite large: the median firm in his sample is around 500 times larger in sales than the median U.S. firm in the widely employed 1993 NSSBF sample for example. Large firms may have their seats in cities where many different banks have located their branches. However large firms may seek funds nationwide from a handful of large banks or obtain funding locally where their production plants are based. His empirical bank market concentration however employs the number of bank branches, measured in the local debt market at the registered seat of the firm. For
firms borrowing regionally or nationally, bank market concentration measured in the local debt market at the registered seat of the firm may not be all that relevant. Second, banks may be quicker to claim to be the Hausbank of these larger and more prestigious firms (an interpretation not incompatible with some of the size results in Elsas). On this account his Hausbank – market concentration findings could be spurious.

For firms borrowing locally, a 30% share of total debt financing for example may lead a bank to assess itself to be a Hausbank in a market with 20 equally sized banks but not in an equally shared duopoly. If banks know other banks’ market shares but not individual borrower shares, the case in Elsas (2005), a 30% share seems large in the 20-bank case, but is known for sure to be the smallest share in the duopoly. As a result the hurdle above which a bank may be reporting Hausbank status increases in concentration and also on this account his Hausbank – market concentration findings could be spurious.

Despite these serious shortcomings, the empirical approach taken by Elsas (2005) has two fundamental merits. First, any comprehensive measure of the lending arrangement is better suited for a cross-sectional exploration of the determinants of relationship banking and such a measure can capture relationship formation far better than any individual loan rate or proxy of credit availability (because of bank fees and cross-selling for example). Second, his study of the lending arrangement per se can be further motivated by the salient observation that in “relationship models” the ultimate effects on loan rates and welfare of changes in competition can be ambiguous, but the effects on bank orientation typically are not (as in Boot and Thakor (2000)).

Our paper follows the astute empirical approach taken by Elsas (2005), but addresses its original shortcomings. Rather than relying on a single subjective measure of the lending arrangement, we employ multiple objective measures. We also analyze a much larger data set containing loans to 13,098 small firms, mainly single-person businesses, for which the informational asymmetry building blocks of theory are more relevant. This data set allows us for the first time to study how local and national competition affects the lending orientation at the local level (which we will argue later is actually the relevant level to study in this case).

We find, in line with Boot and Thakor (2000), that when local interbank competition is fiercer a borrower is more likely to be engaged in relationship banking. In particular the presence in the

1 Even if all banks would know borrower shares and even if there is only one other lender covering the remaining 70%, the 30%-bank may be more likely to designate itself a (second) Hausbank in the 20-bank market than in the duopoly.

2 In this regard Elsas’ study complements Petersen and Rajan (1995) who employ cross-sectional data to infer loan rate smoothing and increased availability of credit over the lifetime of their sample firms (Black and Strahan (2002), p. 2812, f. 4).
postal zone of the borrower of many other banks with equal market shares or the presence of banks with multiple contacts across other postal zones results in substantially more relationship lending. These findings hold for different subsamples and proxies for relationship banking and are seemingly not spurious.

We further document that borrowers located closer to the bank branch are more likely to be engaged as relationship borrowers. Finally, we report that larger bank branches lend substantially more on a transactional basis, a result suggestive of organizational size effects modeled by Stein (2002).

We organize the rest of the paper as follows. Section II reviews the theoretical predictions and recent empirical findings regarding interbank competition and bank orientation. Section III introduces the data and discusses the variables used in our empirical analysis. Section IV displays and discuss the empirical results. Section V concludes.

II. Theoretical Predictions and Recent Empirical Findings

A. Interbank Competition and Bank Orientation

Theory offers conflicting views on the relation between interbank competition and a bank’s willingness to engage in relationship lending (Figure 1 summarizes the predictions of the different theoretical models). A first set of theories argues that competition and relationships are incompatible. Mayer (1988) is the first to apply this insight to banking competition and relationship formation. Mayer hypothesizes that long-term relationships, allowing firms to intertemporally share risks with their banks, only arise if banks enjoy the possibility to extract profits over time, i.e. when the flexibility of the borrowing firms to switch banks is limited. Vigorous competition in the banking market undermines a firm’s ability to commit towards future compensation of a bank’s initial investments.

Petersen and Rajan (1995) model the impact of bank market power on the possibilities to intertemporally share risks. Market power is exogenous in their framework and a monopolistic bank extracts the high future surplus generated by the firm by backloading interest payments. A bank in a competitive (future) market does not have the same latitude to share surplus intertemporally and consequently the bank may be less willing to initiate a relationship and offer credit. Especially lower quality firms are negatively affected by competition, as banks are
unwilling to incur losses that can never be recouped. Hence, credit will be more widely available in banking markets where banks enjoy market power.³

Boot and Thakor (2000) extensively revisit the presumed incompatibility between competition and the nature of relationship financing. They argue that more interbank competition leads to more relationship lending. Boot and Thakor distinguish between two sources of competition, i.e., capital market competition and interbank competition, and they allow banks to choose between relationship lending and transactional lending. In their model stiffer capital market competition reduces relationship lending, while interbank competition actually increases the relative amount of relationship lending. A bank offering a relationship loan augments a borrower’s success probability. Relationship lending then allows extracting higher rents from the borrower. Fiercer interbank competition pushes banks into offering more relationship lending, as this activity permits banks to shield their rents better. In their model interbank competition further reduces bank industry specialization in relationship loans as on the margin the returns to industry specialization decline. Hence, the value added of the relationship loan for the borrower also decreases.⁴


To conclude, how interbank competition affects bank orientation seems ultimately an empirical question.

³ Market power is exogenous in Petersen and Rajan (1995) and the crucial information asymmetry is between borrowing firms and banks. Firms initially know their own quality, but banks do not. Banks learn the borrowers’ type over time. In contrast Fischer (1990), Rajan (1992), Sharpe (1990), and von Thadden (2004) highlight the information asymmetry between banks. By lending repeatedly “inside” banks gather proprietary repayment information, an informational advantage vis-à-vis “outside” competing banks, and some degree of monopoly power over the borrowing firms. Bank relationships arise endogenously in these models, even in perfectly competitive banking markets, and the “learning by lending” does not require relationship specific investments as in Anand and Galetovic (2006) for example. In Dell’Ariccia (2001) banks combine market power from product differentiation (exogenous) with informational monopoly power (endogenous). The contours of the informational asymmetry per se determine both the choice of banking type and the resulting market structure. Abatement in the informational problem in his model may lead to more banks operating in the market and more transactional banking, resulting in a similar correspondence (though not causality) between market structure and banking choice as in Petersen and Rajan (1995). More product differentiation on the other hand leads, for a given number of banks, to more price discrimination in the second period and higher loan rates in the first period.

B. Empirical Findings on Interbank Competition and Bank Orientation

Most empirical work so far has investigated the effects of interbank competition on indirect measures of bank orientation (Figure 1 also summarizes the main empirical findings). In their seminal paper Petersen and Rajan (1995) investigate the effect of local interbank competition on the loan rate and the availability of bank credit for credit-constrained (e.g., young or distressed) firms in the 1988 U.S. National Survey of Small Business Finance dataset. They employ a Herfindahl – Hirschman Index (HHI) in the local market for deposits to measure concentration. Petersen and Rajan find that young firms in more concentrated markets (HHI > 0.18) obtain lower loan rates and take more early (trade credit) payment discounts (i.e., have easier access to bank credit) than firms in more competitive banking markets. Banks seemingly smooth loan rates in concentrated markets and as a result provide more financing, in line with the predictions of their theoretical model.5

Black and Strahan (2002) revisit the local competition – bank orientation issue exploring an alternative measure of local credit availability. In particular, they investigate the rate of new business incorporations across U.S. states. They find that deregulation of bank branching restrictions positively affects new incorporations and, more importantly, that in contrast to Petersen and Rajan (1995) deregulation reduces the negative effect of banking market concentration on new incorporations. They also find that the widespread presence of small banks decreases business formation.6

Recent papers by Fischer (2000) and Elsas (2005) investigate the local competition – bank orientation correspondence using German data. Fischer (2000) focuses on the transfer of information and the availability of credit and finds that both are higher in more concentrated markets. Elsas (2005) studies the determinants of relationship lending. He documents a non-monotonic relationship between local bank market concentration and the probability a bank is designated as “Hausbank”. In particular, he finds that the incidence of Hausbank status is actually the lowest for an intermediate range of market concentration with an HHI of around 0.2, though he notes that most observations of the HHI are also in that low range. Nevertheless his findings broadly suggest the presence of more relationship banking in more competitive markets.

5 Recent work by Zarutskie (2005) and Scott and Dunkelberg (2001) analyzing other U.S. datasets broadly confirm these findings. On the other hand preliminary work by Montoriol Garriga (2005) and Ogura (2005) using the NSSBF dataset show competition and relationship banking are not necessarily incompatible.

6 Cetorelli (2001), Cetorelli (2003), and Cetorelli and Strahan (2005) also find that banking market power may represent a financial barrier to entry in product markets. However Bonaccorsi di Patti and Dell’Ariccia (2004) find opposite results for Italy.
To conclude, many empirical papers have investigated the effects of interbank competition on indirect measures of bank orientation. However, none of the aforementioned papers employs direct measures of bank orientation (with the exception of Elsas (2005)) and controls for both local and nation-wide competition jointly.

III. Data and Variables

A. Data

The unique data set we analyze consists of loans granted to 13,098 firms by an important Belgian bank that operates all over Belgium. The sample includes all existing loans at the bank as of August 10, 1997 that were initiated after January 1, 1995. Degryse and Van Cayseele (2000) and Degryse and Ongenae (2005) employ the same data set. For each borrower we take the characteristics at the time of the first contract observed in the bank’s loan portfolio.

Critical elements of both the Belgian financial landscape and the bank itself make this data ideally suited to investigate the effect of local and nation-wide interbank competition on bank orientation. Previous studies show that the highly developed Belgian banking market is very representative for many other banking markets around the world in terms of concentration and competition. In addition to a representative banking sector, the Belgian financial system was characterized by a high degree of capital market stability before and during the sample period and most firms we study did not have access during the sample period; hence we can safely abstract from changes in capital market competition in our empirical work.

The bank we study is one of a handful of truly national and general-purpose banks operating in Belgium in 1997. As such the bank lends to firms located in most postal zones and is active in 50 different industries (according to a two-digit NACE classification). Around 83% of the firms in its portfolio are single-person businesses and most borrowers obtain just one, relatively small, loan from this bank. Small Belgian firms typically do not tap into the equity or bond markets for their external financing, and typically apply for loans at local bank branches.

Table 1 provides summary statistics for the 13,098 fully identifiable borrowers, showing the definition, mean, standard deviation, minimum, and maximum of our variables, broken down into seven sets of characteristics: (1) dependent variables measuring bank orientation, (2) competition

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7 For example, Barth, Caprio and Nolle (2004) report that the three largest banks in the market account for 57% of all bank assets, while foreign banks hold 24%. The average percentages in their sample covering 55 countries are 50% and 36% respectively, ranking (from high to low) the Belgian banking sector in 21st and 23rd position. Claesens and Laeven (2004) report that the Panzar and Rosse (1987) H-statistic for Belgium equals 0.73 for the years 1994-2001 (H<0 indicates a monopoly, while H=1 indicates perfect competition), while the average for the 50 countries in their sample is 0.67.
measures, (3) distance variables, (4) the bank branch characteristic, (5) postal zone variables, (6) firm size and legal form dummies, and (7) other firm characteristics. We turn to each of these sets of variables in the next subsections.

B. Level of Analysis: Local and Branch

We a priori choose to analyze the competition - orientation correspondence at the local level. Most sample firms are small, implying that loan applications by firms and loan decisions by banks are taken locally both at the data-granting bank as well as at rival banks. Formalized interviews we conducted within the data-granting bank indeed indicated that loan officers located in the bank’s branches enjoyed considerable autonomy granting and pricing small business loans. Upon deciding loan conditions officers merely file a mostly qualitative summary report with the headquarters in Brussels (calculated credit scores were surprisingly only used as inputs to determine bank level re-insurance). Crucial for our purposes, the officers’ own assessment of the development of the relationship with the firm played a key role in motivating past lending decisions.

According to this apparent de jure local autonomy in lending orientation, in Degryse and Ongena (2005) we document substantial variation in loan rates across bank branches and patterns of spatial price discrimination at the branch level. Such pricing behavior strongly substantiates also de facto local autonomy and even branch profit maximization. Ultimately though, the interpretation of our competition – orientation specifications never truly hinges upon branch autonomy: even if “Brussels” would optimally determine lending orientation vis-à-vis each individual borrower taking into account the intensity of local competition and distance (which we find rather unlikely given the inside information we have about the bank), our results would still remain interpretable as such. The only place where the empirical analysis would veer somewhat off the theoretical track is when we investigate branch industry specialization (further down in the paper).

Consequently given the characteristics of our sample firms, our inside information and the results of our earlier empirical work it seems reasonable and actually essential to test the relevant theoretical predictions at the local and branch level, though the theoretical modeling mostly relates to decision units that are labeled ‘banks’.

C. Dependent Variables Measuring Orientation

We employ as our main dependent variable measuring bank orientation a dummy Relationship Banking we define to equal one if the length of the relationship with the borrower exceeds one year and if the bank considers itself as the Main Bank, and to equal zero otherwise. Hence our main dependent variable reflects both the duration and the scope of the engagement between bank and
borrower. Boot (2000) and Ongena and Smith (2000a) argue that both duration and scope characterize relationship banking. In addition we note that Petersen and Rajan (1995) focus on intertemporal pricing in a relationship, while in Boot and Thakor (2000) the scope of an existing relationship may enhance project success.

A firm - bank relationship starts when a firm buys for the first time a product from that bank. The average **duration** of the relationship in the sample is around eight years. Duration proxies for the increased time for a firm to experience the banks’ products and to appreciate the added flexibility the bank has to maintain and fulfill implicit contracts. While the bank gains private information about a firm to tailor its products, the firm may also become locked-in (for example, Boot and Thakor (1994), Sharpe (1990), and Rajan (1992)).

We find justification for using a duration cut-off of only one year in Angelini, Di Salvo and Ferri (1998) and Cole (1998), who document that credit availability does not increase much beyond the first years of a relationship (we replace one year by three years in robustness exercises). We also note that the repayment duration of more than 60% of the observed loans is shorter than or equal to one year. Hence it seems likely that for the majority of the borrowers rollovers of loans take place within the first year of the relationship.

The variable Main Bank captures the **scope** of the relationship and indicates whether this bank considers itself as the main-bank of the firm or not. The definition used by the bank to determine whether it is the main-bank is the firm is “having a monthly ‘turnover’ on the current account of at least BEF 100,000 (€ 2,500), and is buying at least two products from the bank.” Only 54% of all borrowers are classified as Main Bank customers. Banks may obtain an important informational advantage from observing turnover on checking accounts (Nakamura (1993), Vale (1993), Mester, Nakamura and Renault (2005)). But as de Bodt, Lobez and Statnik (2005) document that small Belgian firms employ on average two banks, our Main Bank variable captures variation beyond the mere mechanical outcome of the firms’ preference for a single checking account and relationship.

On the other hand we recognize that the competition – orientation models we highlight disregard the option of multiple bank relationships. However we conjecture that the main hypotheses in these models may unfold qualitatively unaffected if this additional choice is introduced, a topic we leave for future theoretical research.

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8 The **intensity** of the relationship, reflected in for example the frequency of loan rollovers, could be a relationship dimension usefully separable from **scope** (Bodenhorn (2003)). **Trust or quality**, build on **personal contacts** between the loan officer and the entrepreneur, could also be relevant (Harhoff and Körting (1998); Lehmann and Neuberger (2001)).

9 We use Belgian Francs (BEF) throughout the paper but indicate equivalent amounts in Euros. Belgium switched to the Euro on January 1, 1999.
We frame the dependent variable as a dummy variable because theory suggests a dichotomy between relationship and transactional lending. However we will employ the duration of the relationship and Main Bank separately as dependent variables in robustness exercises.

Additional advantages of our dummy approach are that: (1) given our definition about half the firms are engaged as relationship borrowers (i.e., the mean of our independent variable is close to 50%); (2) the reported partial derivatives allow for a straightforward percentage interpretation; and (3) comparison with results in other papers, in particular Elsas (2005), is possible.

D. Herfindahl – Hirschman Index of Market Concentration

As of December 31st, 1994, we identify 7,477 branches operated by 145 different banks and located in 837 different postal zones. Each postal zone carries a postal code between 1,000 and 9,999 (the first digit in the code indicates a geographical region, which we call “postal area” and which in most cases coincides with one of the ten Provinces in Belgium). A postal zone covers on average 26 sq km and contains approximately six bank branches. A postal area covers 3,359 sq km on average. Not surprisingly borrowers are often located in more densely banked areas, with on average more than 17 bank branches per postal zone.

Previous research has argued that the relevant loan market is local in nature for small businesses. Branch proximity continues to play an important role in determining bank choice by borrowers in both the US (Hannan (1991)) and Europe (Sapienza (2002)). Results reported in Degryse and Ongena (2005) show that loan rates in Belgium are not uniform across borrowers or across branches. In addition, physical distance between borrower and local financier affects loan conditions.

The median borrower in our sample is located less than 2.5 kilometers from the lending bank branch. This distance seemingly hasn’t increased by much since the mid-seventies (Degryse and Ongena (2005)). As the number of bank branches decreased by only around 4% during the last two decades and few nationwide entries or M&As took place, local branch configurations most likely hardly changed as banking consolidation through a number of major within-market mergers took place only at the end of the nineties. We therefore a priori select each postal zone as the relevant market.10

Our main measure of competition is the Herfindahl – Hirschman Index (HHI). This variable is widely used as a measure of concentration in the literature; Petersen and Rajan (1995) for example employ the HHI as a measure of competition in their empirical work, while in Boot and Thakor
(2000) the number of banks and hence by extension the HHI delineates the degree of competition in their model. We define HHI as the summed squares of bank market shares by the number of branches in each postal zone. U.S. bank concentration studies always use deposit market shares. However, Elsas (2005) also employs branch market shares for Germany while Fischer (2001) shows that for U.S. Metropolitan Statistical Areas the “branch HHI” is highly correlated with the “deposit HHI”.

We also employ the total Number of Branches and the Number of Banks in each postal zone as competition measures. The former measure assumes no coordination can occur between the branches of the same bank, while the latter measure presupposes coordination effectively takes place. We invert both variables to account for the decreasing effects of additional bank branches and banks. Inversion also facilitates the interpretation of the estimated coefficients and comparisons across the competition variables, in particular with the HHI measure. Both transformed Number measures are bound between zero and one, with zero indicating no market concentration and one indicating maximum concentration.

E. Multi-Market Contact

The postal zone is our a priori chosen banking market. However, many banks are operating in more than one postal zone and often compete with other multi-location banks across zones (Barros (1999) or Park and Pennacchi (2004)). Edwards (1955) introduced the “linked oligopoly” hypothesis that predicts cross-market contacts among banks to increase the incentives for banks to collude. The hypothesis implies that banks compete less when geographical market-overlap increases. Multi-market contact facilitates anti-competitive “mutual forbearance”, as the punishment for deviation from collusion becomes large (Heggestad and Roades (1978), Bernheim and Whinston (1990)), and coordination between banks then fosters relationship banking as in Anand and Galetovic (2006).

However, other theoretical work points towards a possible pro-competitive effect of multi-market contact (Scott (1982)). Mester (1987), for example, presents a Cournot competition model

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10 An incorrect a priori choice of the relevant geographical market cuts against finding significant results for the simple reason that with inappropriate market delineation we expect the resulting “markets” not to be relevant in determining competitive conditions.

11 For postal zones without bank branches we set the HHI equal to one to facilitate decomposing the concentration index later in the paper (by corollary the Number of Banks, another competition variable introduced shortly, is also set equal to one). However, as a robustness check, we remove branchless postal zones in part of the exercises.

12 As some borrowers reside in postal zones without bank branches (i.e., the lending bank branch is located in another, possibly adjacent, postal zone), we add one to the Number of Branches before inverting (remember that for postal zones without bank branches we already set the Number of Banks equal to one).
in which banks have incomplete information about their rivals’ marginal costs. As a result banks
claim to have low marginal costs to sway competitors to produce less. If costs are imperfectly
correlated across markets, multi-market banks have an incentive to put larger quantities on the
market than the profit-maximizing level. “In markets with high concentration, control is in the
hands of a few banks. Thus incentives for these [banks] to mislead other [banks] are greater since
they stand to gain more” (p. 540). Similarly, but in a different setting, Park and Pennacchi (2004)
show that the presence of large multi-market banks promotes local competition, in particular in
highly concentrated markets.

We construct a state-of-the-art Multi-Market Contact measure as proposed in Evans and
Kessides (1994). The variable can be defined succinctly as the sum of all bank pairs in the
borrower’s postal zone weighted by the relative frequency of their bilateral contacts in other postal
zones. The variable is bounded between zero (banks in the postal zone have no contact elsewhere)
and one (all banks in the zone have contact with all other banks across all other postal zones).

F. Distance Variables

Location may determine the degree of competition for a borrower when either borrower
(Hotelling (1929); Salop (1979)) or lender (Sussman and Zeira (1995)) face transportation costs. In
standard spatial models, borrowers select the closest bank and the location of the median borrower
determines the intensity of competition. However, there is no distinction in these models between
“relationship” and “transactional” banking, as borrowers seek only one bank product in a single
period.

In multi-product spatial models, firms in need of multiple products may engage a single bank,
most likely the closest one, in order to minimize transportation and search costs (Armstrong and
Vickers (2001)). Consequently in multi-product spatial models firms close to the lender may be
more likely to opt for “relationship banking” (in scope).

Alternatively, in multi-period spatial models in which borrowers can switch lender, “close”
borrowers may be more likely to stay than the borrowers located farther away from their first-
period lender (Dell’Ariccia (2001)). Again, close borrowers are destined to be “relationship

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13 We consolidate the branches in 104 banks (sometimes banks comprise distinctly incorporated sets of
branches in Brussels, Flanders, and Wallonia). There are 837 postal zones with bank branches. Let \( D_{ij} = 1 \) if
bank \( i \) operates in postal zone \( j \), and \( = 0 \) otherwise, for \( i = 1, \ldots, 104; j = 1, \ldots, 837 \). Let
\( a_{kl} = \sum_{j=1}^{837} D_{kj}D_{lj} \), and \( f_j \): the number of different banks offering service in postal zone \( j \). The Multi-Market Contact measure is
then defined as: \( MMC_j = \frac{2}{837f_j(f_j-1)} \sum_{k=1}^{104} \sum_{l=1}^{104} a_{kl}D_{kj}D_{lj} \).
borrowers” (now in duration) on the basis of their proximity to the lender. In addition, this effect may actually strengthen (Hauswald and Marquez (2006)) if the number of local banks increases.

To control for the effects of transportation costs, we calculate the distance between the borrower and both the lending bank and the branches of all other, competing banks located in the same postal zone as the borrower (Degryse and Ongena (2005) provide details). Address recording errors, incomplete map coverage, changes in street names and borrower relocation cut in our sample. We further conservatively remove the 1-% borrowers located farthest from their lending bank and drop borrowers located in postal zones without competing banks. We end up with Distance to Lender and Distance to Closest Competitors measures for 11,222 borrowers (we call this reduced sample the “Distance sample”).

We transform both measures to (1 + Distance to Lender)$^{-1}$ and [1 – (1+Distance to Closest Competitors)$^{-1}$], respectively. Again, both transformations account at once for the possibly decreasing effects of distance and constrain the variables to be between zero (“around the corner from the lender”) and one (“really far away from the lender”) enabling easier benchmarking. For example, if both distance measures equal one, the borrower is located close to the observed lender but really far from a competing bank. Conditioning on the fact that we observe the close lender granting the loan, we expect, as in a multi-product problem or in a multi-period setting as in Dell’Ariccia (2001), that the engagement is more likely to be relationship-based. On the other hand, if both distance measures equal zero, the borrower is located far from the observed lender but really close to a competing bank. Conditioning on the fact that we observe a far-away lender granting the loan, we can expect the engagement to be transactional.

G. Control Variables

We introduce bank branch size, postal zone variables, and firm size, legal form and industry dummies in the base regressions. We include additional firm characteristics in robustness exercises.

Start with the variable Branch Size. Berger, Demsetz and Strahan (1999) argue that organizational diseconomies of engaging in different type of lending activities may prevent large banks from efficiently providing both transaction-based lending to large corporations and relationship-based lending to small businesses. Large hierarchical banks in Stein (2002) only succeed when information is “hard” enough to flow freely inside the bank. On the other hand, only loan officers at small banks may have the proper incentives to collect and take advantage of “soft” information (that cannot “travel” so easily up the chain of command), precisely the type of information that could be needed to advance relationship banking (see also Berger, Miller, Petersen, Rajan and Stein (2005), Liberti (2004)).
We conjecture that Stein’s arguments are also relevant when assessing the lending activities of autonomous bank branches. Large branches may have one or two hierarchical layers. As a result, loan officers employed in large branches then are less willing to engage in the collection of soft information and relationship lending suffers. We include Branch Size to control for the size differences across the branches. In effect, we pursue a more stringent test of some of the “size” implications of Stein’s model as all branches belong to the same bank. We measure Branch Size by the proportion of the bank’s business loan portfolio at each branch (we take the number of loans at a branch over the total number of loans). There are substantial differences in Branch Size across the bank. The mean bank branch accounts for 44 loans (0.25%) while the largest branch reports 161 of all 17,776 loans (0.91%). The smallest branch has only one business loan.

To control for regional variation in corporate demand for banking services, we introduce a set of postal zone variables that also includes eight Postal Area Dummies. The variable Number of Firms measures the number of registered firms in the borrower’s postal zone, while the variable Assets of Firms averages the amount of assets of registered firms in the borrower’s postal zone. Both variables are constructed using Belfirst, a database containing end-of-1994 information on 176,382 Belgian firms. We similarly construct Industry Concentration to measure the proportion of registered firms in the borrower’s postal zone in the industry of the borrower. Finally, we introduce a dummy variable Urban to control for general differences between businesses located in rural and urban communities. Urban may further capture heterogeneity in information available to banks. For example, banks in urban areas may rely more on hard information while rural banks may collect more soft information (Klein (1992)). Urban equals one when the borrower is located in an agglomeration with more than 250,000 inhabitants, and zero otherwise.

To control for other firm characteristics, we include two firm size,\textsuperscript{14} four legal form and as many as 49 industry dummies (in addition to the base case). We can distinguish between Single-Person Businesses (82.8% of the sample), Small (16.0%), and Medium and Large (1.6%) Firms; and between Sole Proprietorships (82.1%), Limited Partnerships (12.1%), Limited Partnerships with Equal Sharing (1.0%), Corporations (3.9%), and Temporary Arrangements (0.9%). In the regressions, we exclude the dummies for Single-Person Businesses and Sole Proprietorships.

To control further for firm characteristics we also focus on the 9,213 (70.3%) of the borrowers that are both Single-Person Business and Sole Proprietorship (this reduced sample we call the “SPB & SP sample”), collect Age for 1,991 firms (the “Age sample”), and glean Assets, Earnings /

\textsuperscript{14} It may be more profitable for banks to reserve relationship lending for loans of larger size (Stanton (2002)) and for large firms. We employ firm size dummies, as the full dataset does not contain any other measures of firm size.
Assets, and Short-Term Debt / Assets from Belfirst for 645 firms (the “Augmented sample”). We will employ each of these samples in robustness exercises. We display some key sample statistics in Table 2.

IV. Empirical Results

In this section we analyze the regressions of the dependent variable(s) measuring bank orientation on the set of competition and control variables. The correlations displayed in Table 3 between the main dependent and the discussed competition variables already indicate the direction of some of our results. We start discussing the effects of the competition variables and return to a discussion of all the control variables at the end of the section. We first discuss the results for the dependent variable Relationship Banking and turn to the alternative measures of bank orientation, i.e., Duration and Main bank in robustness checks.

A. Postal Zone Competition and Relationship Banking

1. Various Measures of Competition

Since Relationship Banking is a binary dependent variable, we employ a Probit model. In Table 4 we report the partial derivatives, in percent, at the means and significance levels based on t-ratios for the coefficients. To conserve space we neither display partial derivatives for most of the control variables nor the standard errors in Table 4, but return to these estimates later.

In Model I we start with the commonly used (and previously detailed) measure of market concentration, i.e., the Herfindahl – Hirschman Index (HHI). The coefficient on this measure is statistically insignificant and economically small. For example, an increase of 0.1 in the HHI, say from a “competitive” (HHI < 0.10) to a “highly concentrated” (HHI > 0.18) market, would only increase the probability of Relationship Banking by around 0.3%. We replace HHI by respectively \((1 + \text{Number of Branches})^{-1}\), \((1 + \text{Number of Adjacent Branches})^{-1}\), or \((\text{Number of Banks})^{-1}\), but none of the coefficients is statistically significant or economically relevant (we chose not to tabulate the results).

In Model II we add \(HHI^2\) to capture the non-monotonicity present in for example Dinç (2000), Yafeh and Yosha (2001), or Anand and Galetovic (2006). Both coefficients are statistically significant, though in sign opposite to the non-monotonicity predictions, and economically modest but relevant. An increase in the HHI from 0.05 to 0.50 decreases the probability of observing

\[15 \text{ “On theoretical grounds it is difficult to justify this choice” (Greene (1997), p. 875) of a Probit model, hence we also rerun all exercises using a Logit model. Given that the mean of the dependent variable is close to 50%, not surprisingly results are almost unaffected.} \]
relationship banking by close to 5%. Replacing HHI and $HHI^2$ by a set of dummies that equal one if HHI is situated in a certain range (we employ 0.1 intervals) and are zero otherwise yields similar results. Adding squared terms to the specifications featuring $(1 + \text{Number of Branches})^{1/4}$, $(1 + \text{Number of Adjacent Branches})^{1/4}$, or $(\text{Number of Banks})^{1/4}$ yield statistically insignificant and economically irrelevant results.

The regressions so far left two possibly important factors determining borrower engagement unaccounted for. First, banks may take into account exactly whom their competitors are in the postal zone given contact in other postal zones, i.e., banks may care about multi-market contact. Second, as argued above, proximity could encourage firms to frequent the same bank for multiple services during a longer time period.

2. **Controlling for Multi-Market Contact and Distance**

To control for either pro- or anti-competitive effects arising from Multi-Market Contact, we introduce the contact variable in Model III in Table 4. To control for spatial effects, we add the two distance measures in Model IV. Removing Multi-Market Contact in Model IV does not alter the results and we center our discussion on Model IV (even though it is employing a somewhat smaller sample).

The coefficients on both HHI variables remain significant and actually become substantially larger in Model IV. Figure 2 displays the resulting schedule (at the means of the other variables). The percentage probability of observing Relationship Banking is measured along the vertical axis, while HHI is on the horizontal axis. The scale on the horizontal axis is proportionate to the number of observations with particular values for HHI. Increasing HHI from 0.10 to 0.18, indicated by vertical lines in the Figure, decreases Relationship Banking by 3.1% (from 55.0 to 51.9) while increasing the HHI from 0.05 to 0.50 decreases the probability by almost 10%. We further note (jumping ahead somewhat in the discussion of the coefficients) that a 10% change is similar in magnitude to the effect of distance: a borrower located around the corner from the bank is 11.3% more likely to be engaged as a relationship customer than a borrower located (infinitely) far away. Consequently the effect of local market concentration (admittedly measured with some error) on relationship banking is similar to the effect of distance (probably measured more precisely) and hence the concentration effect should be assessed to be sizeable and economically relevant.

These findings confirm a key result in Boot and Thakor (2000) but are at odds with Petersen and Rajan (1995). Branches seemingly engage in more relationship banking when facing fiercer banking competition. These findings may also be at odds with the correspondence between market power in banking and market power in the firms that banks lend to (Cetorelli and Strahan (2005)). Concentrated product markets show less firm entry, hence contain older firms with potentially
longer and broader relationships with their banks. Concentration in the banking sector through concentration in the product market would consequently be connected with more relationship banking. But that is not what we find.

However at this point it is useful to note that our findings regarding the HHI – Relationship Banking correspondence are qualitatively similar to the (somewhat stronger) non-monotonicity documented in Elsas (2005). In his paper the incidence of the Hausbank status drops from 80% to 40% as HHI increases from zero to 0.2, and then sharply increases to 100% for an HHI equal to 0.45. We conjecture that the differences in firm size and the corresponding number of bank engagements between his and our sample are responsible for this result.\textsuperscript{16} The 11,222 firms in our “distance” sample are much smaller than the 122 firms in his sample;\textsuperscript{17} hence our firms are possibly more opaque and may seek to engage fewer – sometimes one – banks to satisfy their credit needs.\textsuperscript{18} As a result, an increase in the number of banks on the market may result in a smaller increase in the degree of competition for the firms in our sample than for the large firms in Elsas (2005) that had engaged many (all) banks in the local market already or had engaged only large banks.

3. Very Concentrated Markets

The substantial increase in Relationship Banking for HHI values close and equal to one requires further exploration. Replacing HHI and HHI\textsuperscript{2} by a set of dummies that equal one if HHI is situated in a certain range and are zero otherwise (to partly neutralize the effects of these observations) yields qualitatively similar results. Similarly, the partial derivative on HHI equals -20.5\textsuperscript{**} and -27.3\textsuperscript{***} respectively, if we drop HHI\textsuperscript{2} and remove observations for HHI equal to one or larger than 0.9. Hence the derivative remains statistically significant and economically relevant in both exercises.

If Relationship Banking decreases with concentration in less concentrated markets, why then do we observe more relationship banking in very concentrated markets? Physical proximity, as pointed out earlier, could compel a firm to frequent a close-by bank for all its needs. A monopolist in a postal zone then simply satisfies this demand by providing all services, in particular when

\textsuperscript{16} The local markets in his paper are also substantially larger than in ours. The average postal zone in Belgium contains less than 10,000 inhabitants, while the mean Landkreise in Germany counts around 175,000 people.

\textsuperscript{17} The average firm in Elsas (2005) has annual sales of approximately 4,000 million BEF, while the average firm in our Augmented sample reports 14 million BEF in total assets.

\textsuperscript{18} German and Belgian corporations seem to maintain a similar number of bank relationships (Ongen and Smith (2000b)), but small firms in general are found to have fewer bank relationships (the empirical evidence is reviewed in Ongen and Smith (2000a)). The average small Belgian firm surveyed by de Bodt, et al. (2005) employs two banks. The firms in the latter sample are on average more than three times larger and 7 years older than the firms in our sample.
banks in other postal zones are far away. An increase in Relationship Banking for high HHI values then merely affirms our a-priori choice of the postal zone as the relevant geographical market.

Alternatively, we note that Boot and Thakor (2000) predict that a monopoly bank should engage in little or no Relationship Banking. However, the monopolist bank may become an industry specialist at zero or low cost (by servicing all firms in the vicinity) and hence supply relationship banking nevertheless. This is not modeled in Boot and Thakor, as in their model even a monopolist incurs specialization costs (that are not a function of market structure).

4. Multi-Market Contact

Next we focus on the coefficient of Multi-Market Contact in Model IV. Multi-Market Contact carries a positive sign, is statistically significant, and economically relevant. An increase in the variable from 0 to 0.33 (the observed range) increases the probability of observing Relationship Banking by almost 10%. However, removing both HHI variables causes the coefficient on Multi-Market Contact to become insignificant, possibly indicating the need to control for market concentration and multi-market contact simultaneously.

The contact variable is significantly and negatively correlated with HHI (see Table 3), and this is partly by construction. Indeed, an increase in the number of banks in a postal zone increases the likelihood that some bank pairs also meet in another postal zone hereby increasing Multi-Market Contact. However, an increase in the number of banks also decreases market concentration as measured by HHI. We further investigate the coefficient on this variable in the robustness section.

5. Distance Measures

Now we turn to the distance measures in Model IV in Table 4. The coefficient on the transformed Distance to Lender is positive, statistically significant, and economically relevant, confirming either a multi-product or multi-period switching hypotheses emanating from spatial models. The probability of observing Relationship Banking for a borrower close to the Lender (i.e., \((1 + \text{Distance to Lender})^{-1} = 1\)) is more than 11% higher than for a far-away borrower (i.e., \((1 + \text{Distance to Lender})^{-1} = 0\)). On the other hand, the transformed Distance to Closest Competitor is not statistically significant.

These results are unaffected if we remove either one of the two HHI and/or Multi-Market Contact variables. Similarly, removing both distance variables in Model IV leaves the other coefficients unaffected. Motivated by Hauswald and Marquez (2006) we further interact HHI and/or \(\text{HHI}^2\) with our distance measures. The coefficients on HHI and \(\text{HHI}^2\) remain broadly the same in sign and magnitude, but are no longer significant. The interaction terms are insignificant as well. We suspect multicollinearity problems.

As an alternative, we split the sample in firms that are closer to the lender than to the closest
bank competitor (we call these firms the “relatively close” firms) and those firms that are closer to
the closest bank competitor than to the lender (the “relatively far” firms). The coefficients on our
competition measures in both subsamples retain the same sign, significance, and magnitude. The
distance measures are only significant for the firms that are “relatively far”.

Taken together, these results suggest the distance variables may proxy for other factors
(transportation costs as in Degryse and Ongena (2005)?) than those picked up by our postal zone
and national competition measures. By introducing branch effects we will shortly (in the
robustness section) corroborate that Relationship Banking and Distance are seemingly not driven
jointly by an omitted variable and that Distance (marked as a complement and not a substitute to
Relationship Banking) may not proxy for the intensity and ease of informational flows as in
Petersen and Rajan (2002). Given the short distances involved these findings probably shouldn’t
come as a surprise.

To conclude, the observed lender engages more borrowers in relationship banking if many other
banks (possibly with equal market shares) operate in the same postal zone or if the banks in the
postal zone have multiple contacts across other postal zones. More relationship banking is also
being observed when firms are located close to the bank.

B. Robustness Checks

1. Subsample of Single-Person Businesses and Sole Proprietorships

Model V in Table 4 focuses on the 9,213 firms that are both Single-Person Businesses (SPB)
and Sole Proprietorships (SP). There are a number of reasons to believe that the possible
correspondence between competition and bank orientation will appear sharpest in this subsample.
First, remember that we are looking at the loan portfolio of one single bank and that we now retain
just one type of firm. Consequentially, important firm characteristics potentially clouding our
previous results are controlled for. Second, Single-Person Businesses / Sole Proprietorships are the
smallest (possibly most opaque and locally restricted) firms that are affected most by the structure
of the local banking market.

The findings in Model V confirm this expectation and strengthen our earlier results. The non-
monotonicity in HHI is again economically relevant. For example, increasing HHI from zero to 0.4
decreases the probability of Relationship Banking by almost 15%, from 60 to 45%. We again
replace HHI and HHI^2 by range dummies and confirm these findings.

2. Additional Independent Variables and Branch Effects

Models VI and VII in Table 4 add Age and other firm characteristics (Assets, Earnings / Assets,
Short-Term Debt / Assets) to the specification. The main results go through almost unaffected,
even though the samples are substantially reduced and quite different in their composition (for example, the Distance sample contains 83% single-person businesses, 16% small and 1% medium and large firms, the Age sample only 6% single-person businesses, 89% small and 5% medium/large, and the Augmented sample 5%, 87% and 8% respectively).

We further add Multi-Market Contact\(^2\) to Model III and all possible combinations of Multi-Market Contact\(^2\), (1 + Distance to Lender)\(^2\), \([1 - (1 + \text{Distance to Closest Competitors})^{-1}]^2\) to specifications IV to VII. Admittedly we know of little theoretical justification for doing so (hence we choose not to tabulate the results). However, the coefficients of HHI, HHI\(^2\), Multi-Market Contact, and (1 + Distance to Lender)\(^{-1}\) are virtually unaffected in significance, sign and size in all specifications and only the coefficient on the newly added (1 + Distance to Lender)\(^{-2}\) becomes negative and significant at a 10% level in a few specifications.

We further replace Branch Size by random branch effects,\(^{19}\) remove Industry Dummies (to avoid multicollinearity problems), and employ OLS to re-estimate the main specifications. Branch effects could capture omitted variables that could be correlated with bank orientation, such as branch service quality and local firm presence and/or competition (Cetorelli (2001)), for which we couldn’t construct reasonable proxies. However, results are unaffected if anything they are even more “striking” in statistical significance and economic relevance.

3. Multi-Market Contact Deconstructed

Multi-Market Contact between banks across postal zones stimulates Relationship Banking. Hence, the contact variable possibly captures a pro-competitive effect if this variable would cut in the same direction as HHI. However, to shed further light on this issue we first examine more closely what occurs at the postal zone level (following Anand and Galetovic (2006)) and then turn to interacting HHI with Multi-Market Contact (as in Mester (1987) and Park and Pennacchi (2004)).

Recall that in Anand and Galetovic (2006) only coordination between a few banks with equal market shares fosters relationship banking. To test whether the effect of concentration on Relationship Banking arises through a decrease in the number of banks or through the inequality of bank market shares, we decompose HHI in \((\text{Number of Banks})^{-1}\) and \([\text{HHI} - (\text{Number of Banks})^{-1}]\). The results (we choose not to tabulate) suggest that it is only the change in the number of banks, and not the change in their market shares, that is driving our results (though admittedly our measure based on the number of bank branches is rather coarse when measuring market shares). An increase in the number of banks from for example 3 to 37 increases the probability of Relationship Banking.

\(^{19}\) A Hausman test cannot reject at a 1-% level that random effects should be favored.
Banking by 8.5% (from 40.9 to 49.4%). Consequently the observed lender seemingly doesn’t coordinate with other banks at the local level in offering relationship banking (as in Anand and Galetovic (2006)).

Alternatively we decompose HHI in $(Branch \ Share \ of \ the \ Lender)^2$ and $[HHI – (Branch \ Share \ of \ the \ Lender)^2]$ to check for possible coordination between branches of the observed lender. And indeed, a variety of specifications suggest that a larger relative presence of the lender increases Relationship Banking at about the same rate as the relative presence of other lenders decreases it, though the coefficients are not always statistically significant. Taken together these results suggest that within one postal zone, branches of the lending bank may coordinate (independently) among themselves but not with the branches of the other banks present there.

Now, given the local discretion in setting loan conditions (an assessment that is, as already mentioned, based on formal interviews and loan rate variation), it would be surprising if the bank would succeed in coordinating with other banks at the national level to achieve relationship orientation at the local level. To test for the occurrence of national coordination versus a pro-competitive effect more directly, we interact HHI and $HHI^2$ with Multi-Market Contact. Mester (1987), for example, argues that if the Contact variable measures “mutual forbearance” then the Contact variable itself should have the same sign as HHI (a result we did not have so far) while the interaction terms should equal zero.

Results (again unreported) are somewhat mixed. The coefficients on the interaction terms suggest no coordination takes place, but multicollinearity robs the coefficients of their significance. The coefficient on the Multi-Market Contact variable is still positive and opposite the coefficient on HHI but much smaller than in earlier specifications. To conclude, coordination may occur between branches of our bank, but none of the exercises suggests coordination takes place either locally or nationally between banks.

4. Alternative Definitions of the Dependent Variable

As the duration cutoff of one year in the construction of the dependent variable Relationship Banking was somewhat arbitrarily chosen (remember however that results in Angelini, et al. (1998) and Cole (1998) suggested a short duration cutoff), we also run all specifications with a three-year cut-off. Results are virtually unaffected.

Next we employ our other two variables capturing bank orientation, i.e., Duration and Main Bank. Elsas (2005), for example, argues that duration may be a poor proxy for the significance of the relationship; hence we employ Main Bank (by itself) as the dependent variable. We report the almost unaffected results in Table 5. We also estimate a Tobit model (censored at zero) with $\ln(\text{Duration of Relationship})$ as the dependent variable and report the results in Table 6. Again the
results are very similar to the ones reported above, seemingly contradicting the claim of non-relevancy of duration as a measure of relationship importance by Elsas (2005). We again conjecture that the differences in firm size and the corresponding number of bank relationships between his and our sample are responsible for this result. The firms in our sample are much smaller and may have fewer bank relationships. As a result, for the firms in our sample the observed duration of a relationship may capture or at least be correlated with relationship orientation.

5. Omitted Factors and Endogeneity

We are further concerned that duration caused or is affected by factors that caused current market concentration. For example, a pool of high-quality firms in the postal zone 10 years ago may have contributed to the longevity of the observed relationships as both relationships and firms survived (on the other hand high-quality firms may have had less need for a relationship lender). But circumstances in the postal zone 10 years ago that led to the high quality of the firm pool may also have attracted other banks to set up branches there in the period since then. To deal with this pernicious problem we toss out all observations with durations exceeding 10 (7) years and rerun most specifications. Even though we loose more than one third (one half) of the sample, the competition results are almost unaffected.

We further collect bank branch information from the end of the year 1985, recompile the HHI, and rerun all relevant exercises. This newly constructed “HHI 1985” is highly correlated with the HHI for 1995, the correlation coefficient equals 0.92***, and therefore not surprisingly results are almost unaffected when the new HHI 1985 measure replaces the HHI 1995.

C. Control Variables

Finally, we return to the coefficients on the control variables, starting with Branch Size. We reported the coefficient on Branch Size in all Tables discussed so far. The coefficient is almost always significant at a 1% level and economically quite relevant. The partial derivative at the means for both Relationship Banking and Main Bank varies around -14, indicating that an increase from the smallest to the largest branch (0.006 to 0.905) decreases the incidence of relationship banking by around 13%. The partials in the Duration Tobit models (Table 6) suggest an equivalent decrease by around 3 years in the length of the observed relationship for a similar increase in branch size. Hence, ceteris paribus, larger bank branches pursue more transactional banking.

20 Unfortunately we could not find a comprehensive listing of branches of savings banks for 1985, but the total number of savings branches we know has remained constant during this time period. Consequently we employed the 1995 list to impute the 1985 HHI. Savings banks operate around one half of all branches.
Berger, et al. (2005) document that larger banks have less exclusive and shorter relationships than smaller banks. To make our results better comparable to theirs, we replace Branch Size by $\ln(\text{Branch Loan Volume})$ defined as the natural logarithm of the loan portfolio of the branch in 1000s of US$ (they employ the log of bank assets). We estimate Logit and OLS models with Relationship Banking and $\ln(1 + \text{Duration of Relationship})$ as the dependent variables (we don’t tabulate the results). The resulting coefficients are comparable in magnitude, for duration as the dependent variable but not for Main Bank. However then the definition of their scope variable, dummy = 1 if only lender, differs from our Main Bank variable.

Coefficients on the other control variables are reported in Table 7. We report the representative coefficients from Model IV, VI, and VII. None of the four postal zone coefficients are consistent in sign, size, or statistical significance. The legal form dummies in Model IV are highly significant. Banks engage Sole Proprietorships less likely in a Relationship and profitable firms more likely, possibly because of bankruptcy risks. As such the specifications highlight the need to control carefully for firm characteristics, as we do in Models V to VII.

D. Bank Industry Specialization

1. Dependent Variable

Theory also provides hypotheses concerning the relation between interbank competition and bank industry specialization. In Boot and Thakor (2000), for example, competition affects the banks’ investment in industry expertise and hence the “value” of bank-firm relationships. In their model interbank competition reduces bank industry specialization in relationship loans as on the margin the returns to industry specialization decline. Hence, the value added of the relationship loan for the borrower also decreases.

We construct a new dependent variable measuring bank industry specialization. We classify the borrowers in the 50 two-digit NACE code classes and for each branch calculate a variable $\text{Industry Specialization}$ as the proportion of loans of the bank branch loan portfolio in the same industry as the borrower, in percentage. The mean of industry specialization is 18.2% with a standard deviation of 0.21.

**Note:** Our measure assumes that the bank’s knowledge about a particular industry flows from observing the loan repayments by other bank borrowers active in that industry. However, in Boot and Thakor (2000) the degree of industry specialization is chosen ex-ante and is not derived from the actual composition of the realized loan portfolio. In addition, the degree of industry specialization is observable in their model by the individual borrower upon their first contact with the bank. We doubt the bank branches we study ever recorded their “ex ante” choices and costs of industry specialization, but we consider these costs unlikely to be observable by the borrowers in any case. In that sense our $\text{ex post}$ measure relying on portfolio composition may be a reasonable proxy for the branch’s selected degree of specialization as the bank’s clientele in an industry may have been partially observable by (and even “advertised” by the branch to) interested firms from that industry.
deviation of 13.9%. The cost of industry specialization is bank-specific in Boot and Thakor (2000) and further depends on the diversity of borrower types served by the bank. Consequently when estimating the impact of competition on industry specialization we again control for branch characteristics such as branch size.\textsuperscript{22}

2. Estimation Results

We now analyze the regressions of the dependent variable(s) measuring bank industry specialization on the same set of competition and control variables. We first employ ordinary least squares. The dependent variable, Industry Specialization, is by construction always larger than zero, but it is censored at 100. However, as the variable is equal to 100 for only 19 borrowers we disregard this minor censoring issue. We follow the same line-up of exercises as for bank orientation and report the results in Table 8. Overall our results indicate that market concentration is hardly economically relevant in explaining industry specialization and that any reported statistical relationship is weak and seemingly not robust to model alterations.

We start by focusing on the full sample. In Model I in Table 8, we introduce $HHI$ as the measure for concentration. The coefficient turns out to be both statistically and economically insignificant. Theory suggests potential non-monotonicity; hence, we incorporate $HHI^2$ in Model II. The results remain insignificant providing no evidence in favor of banks specializing in an industry when competition is high (Dell’Ariccia and Marquez (2004)) or intermediate (Boot and Thakor (2000)). Model III in Table 8 incorporates the Multi-Market Contact variable. If more contact implies a pro-competitive effect, Boot and Thakor (2000) hypothesize less industry specialization should be observed, whereas according to Dell’Ariccia and Marquez (2004) more industry specialization should be observed. Our empirical results are in line with the former suggesting that more competition leads to less specialization. But the effects seem rather modest. For example, an increase in the contact variable from 0 to 0.33 (minimum to maximum) decreases Industry Specialization by around 3%.

\begin{itemize}
\item\textsuperscript{22} The bank chooses first the degree of industry specialization followed by its orientation in Boot and Thakor (2000). Its specialization decision is conditioned on the assessed probability distribution over borrower quality, while the orientation decision is made on the basis of the actual (representative) borrower’s quality. For any outside observer (uninformed about the bank’s initial assessment and decision) the two decisions appear inseparable and estimating two reduced-form equations containing orientation and specialization respectively (as dependent variables) seems appropriate. The borrower can decide to go to either the capital market or the banking market in between the bank’s industry specialization and orientation decision. The corresponding selection issue appears minimal for most firms in our sample that have no access to the capital market anyway. The first-round matching with the banks is stochastic (borrowers also don’t know the degree of bank specialization in the model) and is followed by the orientation decision and random competitive bidding by outside banks. Hence no additional selection takes place.
\end{itemize}
We again arrive at our Base Model (IV) by incorporating the two distance measures. Our transformed Distance to Lender is again statistically significant, but only at a 10% level, and negative. The closer the borrower the less specialization we observe. But the effects also seem modest. Industry Specialization for a close borrower is only 1.4% lower than for a borrower far away. The transformed Distance to Closest Competitors is not significant.

The Base Model also suggests a concave relationship between HHI and specialization, but the coefficients are seemingly small. Figure 3 plots the resulting schedule (at the means of the other variables) using a similar setup as in Figure 2. An increase in HHI from 0.10 to 0.18 (the vertical lines marking the regions with varying degree of competition), for example, increases industry specialization by only 0.4% (from 17.8 to 18.2%). Figure 3 broadly confirms that competition reduces industry specialization at the branch level, but also suggests small economic relevance.

To conclude, the branches of the analyzed bank engage somewhat fewer borrowers in the same industry if local market concentration decreases or when banks in the postal zone have more contacts across other postal zones. Branches possibly reduce industry specialization as competition intensifies. But the effects seem at best rather modest, both in terms of statistical significance and economic relevance. Less industry specialization is also being observed when firms are located closer to the bank. In that case, industry specialization may become less prevalent because borrowers are less discriminate about their choice of bank branch.

3. Robustness Checks and Control Variables

In Model V in Table 8 we again restrict the sample to the 9,213 firms that are both Single-Person Businesses (SPB) and Sole Proprietorships (SP). However, we continue to assume that Industry Specialization is based on the entire loan portfolio of the branch. As expected (as these firms are possibly more opaque), results are statistically somewhat more significant and economically relevant. Next we add Age in Model VI and other Firm Characteristics in Model VII. Now all coefficients on the Competition variables become insignificant confirming our earlier assessments of relatively weak statistical significance.

In Boot and Thakor (2000) competition affects bank industry specialization only for relationship borrowers. We run all models on the set of borrowers we identified as relationship borrowers, (i.e., Relationship Banking = 1). We first assume that industry specialization should be measured only for the portfolio containing these relationship borrowers. We rerun all specifications but choose not to report. Most coefficients are similar in sign and size, but somewhat less statistically significant. Next we measure industry specialization for the entire loan portfolio of the branch (assuming some positive knowledge spillovers from transactional lending) and re-run all seven models for the same sets of relationship borrowers. Results are virtually
unaffected and again we choose not to tabulate them.

Next we are concerned about overweighing industry specialization by large branches (by definition many borrowers belong to those industries that large branches specialize in). We weigh all observations by the inverse of the number in each industry – branch group. None of the coefficients on the competition variables are statistically significant or economically relevant any longer indicating that only large branches adjust their degree of specialization in their focused industries somewhat to competition. This interpretation may also explain the percentage-wise small adjustments we pick up.

Our “linear” industry specialization measure may further fail to accommodate for the possible presence of fixed specialization costs and/or learning. We take the natural log of Industry Specialization and rerun all OLS and WLS exercises. Results are qualitatively unaffected.

Finally, we discuss the control variables. The coefficient on Branch Size is always negative, significant, and economically relevant in Table 8. Increasing Branch Size from the smallest to the largest branch decreases Industry Specialization by around 6.5% to 12.5%. The other control variables are hardly statistically significant (and left unreported).

V. Conclusion

Are competition and relationships necessarily inimical? We address this issue employing a unique data set containing detailed information on both bank-firm relationships and the local banking market structure. Interbank competition seemingly affects bank orientation (and to a much lesser extent bank industry specialization). Fiercer competition results in more relationship banking (in most observed cases). Borrowers located closer to the bank branch are more likely to consume other bank services and to be engaged over a longer time period. Finally, larger bank branches lend substantially more on a transactional basis but are less likely to be specialized in particular industries. We cannot fail to notice that bank entry in a spatial model of competition decreases market concentration, bank – borrower distance, and bank branch size, and that the estimated coefficients on all three variables are negative, significant, and of almost equal economic relevancy.

Taken at face value these results cannot reject hypotheses proposed by Boot and Thakor (2000), among others, suggesting competition and relationships are not necessarily inimical. However the results seem at odds with insights and results by for example Petersen and Rajan (1995). Reconciling both sets of hypotheses and results seems a natural but challenging task for future research.
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Ogura, Y., 2005, Does Lending Competition Promote Relationship Banking? Evidence from the US Small
Montoriol Garriga, J., 2005, Relationship Lending: Doe the Number of Banks Matter? Evidence from the US,
Mayer, C., 1988, New Issues in Corporate Finance,
**Figure 1. Competition and Bank Orientation**

**Panel A. Theoretical Predictions**

- **Degree of Competition in the Banking Sector**
  - High: Many Banks, Low Concentration
  - Low: Few Banks, High Concentration

- **Transactional Banking**
  - Many Banks: Many competitors, high competition
  - Few Banks: Few competitors, low competition

- **Relationship Banking**
  - Many Banks: Close relationships, high trust
  - Few Banks: Less personalized service

- **References**
  - Dinç (2000); Yafeh and Yosha (2001); Anand and Galetovic (2006)
  - Boot and Thakor (2000); Dell’Ariccia and Marquez (2004); Hauswald and Marquez (2006)

**Panel B. Empirical Findings**

- **HHI in Local Market for Deposits**

- **HHI in Local Market, by Number of Bank Branches**
  - Fischer (2000): Germany IfO, 403 Firms, 1996

- **Observations**
  - Relationship Banking: Lower loan rate & more early trade credit discounts taken (= more bank credit available) by young firms
  - Relationship Banking: Probability of business formation.
  - Relationship Banking: Higher % of Hausbank status
  - Transactional Banking: More information transfer & more credit
  - Transactional Banking: Higher % of Hausbank status
FIGURE 2. BANK MARKET CONCENTRATION AND BANK ORIENTATION

- Competitive
- Intermediate Concentration
- Concentrated

Relationship Banking, in %

Herfindahl Hirschman Index (HHI)
Figure 3. Bank Market Concentration and Bank Industry Specialization
**Table 1. Data Description**

# Obs is the number of observations. *The definition used by the bank to determine whether it is the main bank is: for Single-Person Businesses and Small Firms, have a "turnover" on the current account of at least BEF 100,000 per month and buy at least two products from that bank. †We set HHI = 1 and (Number of Banks)\(^1\) = 1 if the Number of Branches = 0. *40 Belgian Francs (BEF) are approximately equal to 1 Euro. ‡The dummies for Single-Person Businesses and Sole Proprietorships are suppressed in the regressions, hence not included in the Table.

<table>
<thead>
<tr>
<th>Variables</th>
<th>Definition</th>
<th># Obs</th>
<th>Mean</th>
<th>St.dev</th>
<th>Min</th>
<th>Max</th>
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</thead>
<tbody>
<tr>
<td><strong>Dependent Variables</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Relationship Banking</td>
<td>= 1 if the length of the relationship with the borrower exceeds one year and if the bank considers itself as main bank(^a)</td>
<td>13,098</td>
<td>0.524</td>
<td>0.499</td>
<td>0</td>
<td>1</td>
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<tr>
<td>Duration of Relationship</td>
<td>Length of relationship with current lender, in years</td>
<td>13,098</td>
<td>7.8</td>
<td>5.5</td>
<td>0</td>
<td>26.3</td>
</tr>
<tr>
<td>Main Bank</td>
<td>= 1 if bank considers itself as main bank(^a)</td>
<td>13,098</td>
<td>54.3</td>
<td>49.8</td>
<td>0</td>
<td>1</td>
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<tr>
<td></td>
<td>ln(1 + Duration of Relationship)</td>
<td>13,098</td>
<td>1.9</td>
<td>0.8</td>
<td>0.0</td>
<td>3.3</td>
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<td><strong>Competition Variables</strong></td>
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<td></td>
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<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Number of Branches</td>
<td>Number of bank branches in borrower’s postal zone</td>
<td>13,098</td>
<td>16.4</td>
<td>15.6</td>
<td>0</td>
<td>103</td>
</tr>
<tr>
<td>Number of Adjacent Branches</td>
<td>Number of bank branches in borrower’s and adjacent postal zones</td>
<td>13,098</td>
<td>70.9</td>
<td>47.1</td>
<td>0</td>
<td>471</td>
</tr>
<tr>
<td>Number of Banks</td>
<td>Number of banks in borrower’s postal zone</td>
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<td>8.3</td>
<td>4.8</td>
<td>0</td>
<td>37</td>
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<tr>
<td>HHI</td>
<td>Herfindahl – Hirschman Index, i.e. the summed squares of bank market shares by number of branches in borrower’s postal zone</td>
<td>13,098</td>
<td>0.205</td>
<td>0.194</td>
<td>0.057</td>
<td>1(^b)</td>
</tr>
<tr>
<td>Multi-Market Contact</td>
<td>Sum of the bank pairs in borrower’s postal zone weighted by the relative frequency of their bilateral contacts in other postal zones (see Appendix).</td>
<td>13,098</td>
<td>0.174</td>
<td>0.080</td>
<td>0</td>
<td>0.335</td>
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<tr>
<td></td>
<td>(1 + Number of Branches)(^1)</td>
<td>13,098</td>
<td>0.123</td>
<td>0.178</td>
<td>0.009</td>
<td>1</td>
</tr>
<tr>
<td></td>
<td>(1 + Number of Adjacent Branches)(^1)</td>
<td>13,098</td>
<td>0.047</td>
<td>0.175</td>
<td>0.001</td>
<td>1</td>
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<tr>
<td></td>
<td>(Number of Banks)(^1)</td>
<td>13,098</td>
<td>0.183</td>
<td>0.199</td>
<td>0.027</td>
<td>1(^b)</td>
</tr>
<tr>
<td></td>
<td>HHI(^2)</td>
<td>13,098</td>
<td>0.079</td>
<td>0.214</td>
<td>0.003</td>
<td>1</td>
</tr>
<tr>
<td></td>
<td>HHI – (Number of Banks)(^1)</td>
<td>13,098</td>
<td>0.021</td>
<td>0.023</td>
<td>0</td>
<td>0.875</td>
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<td>(Number of Banks)(^2)</td>
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<td>0.214</td>
<td>0.000</td>
<td>1(^b)</td>
</tr>
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<td></td>
<td>[ HHI – (Number of Banks)(^1) ](^2)</td>
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<td>0.001</td>
<td>0.010</td>
<td>0</td>
<td>0.765</td>
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<tr>
<td></td>
<td>(Number of Banks)(^1) [ HHI – (Number of Banks)(^1) ](^2)</td>
<td>13,098</td>
<td>0.002</td>
<td>0.004</td>
<td>0</td>
<td>0.140</td>
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</table>
### Distance Variables

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<th>Variable</th>
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<th>Observations</th>
<th>Mean</th>
<th>Median</th>
<th>Std. Dev.</th>
<th>Min</th>
<th>Max</th>
</tr>
</thead>
<tbody>
<tr>
<td>Distance to Lender</td>
<td>Shortest traveling time, in minutes</td>
<td>11,222</td>
<td>6.7</td>
<td>7.2</td>
<td>0</td>
<td>51</td>
<td></td>
</tr>
<tr>
<td>Distance to Closest Competitors</td>
<td>Shortest traveling time to closest quartile competitor in borrower’s postal zone, in minutes</td>
<td>11,222</td>
<td>3.7</td>
<td>2.3</td>
<td>0</td>
<td>24</td>
<td></td>
</tr>
</tbody>
</table>

\[(1 + \text{Distance to Lender})^{-1}\]

\[1 - (1 + \text{Distance to Closest Competitors})^{-1}\]

### Bank Branch Characteristic

<table>
<thead>
<tr>
<th>Characteristic</th>
<th>Description</th>
<th>Observations</th>
<th>Mean</th>
<th>Median</th>
<th>Std. Dev.</th>
<th>Min</th>
<th>Max</th>
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</thead>
<tbody>
<tr>
<td>Branch Size</td>
<td>Proportion of bank loan portfolio at the bank branch, in percent</td>
<td>13,098</td>
<td>0.249</td>
<td>0.152</td>
<td>0.006</td>
<td>0.905</td>
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### Postal Zone Variables

<table>
<thead>
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<th>Variable</th>
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<th>Observations</th>
<th>Mean</th>
<th>Median</th>
<th>Std. Dev.</th>
<th>Min</th>
<th>Max</th>
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</thead>
<tbody>
<tr>
<td>Number of Firms</td>
<td>Number of registered firms in the borrower’s postal zone, in thousands</td>
<td>13,098</td>
<td>0.749</td>
<td>0.891</td>
<td>0.002</td>
<td>6.103</td>
<td></td>
</tr>
<tr>
<td>Assets of Firms</td>
<td>Average amount of assets of registered firms in the borrower’s postal zone, in billions of BEF</td>
<td>13,098</td>
<td>0.068</td>
<td>0.131</td>
<td>0.000</td>
<td>3.739</td>
<td></td>
</tr>
<tr>
<td>Industry Concentration</td>
<td>Proportion of registered firms in borrower’s postal zone in industry of borrower, in percent</td>
<td>13,098</td>
<td>1.9</td>
<td>3.4</td>
<td>0</td>
<td>66.6</td>
<td></td>
</tr>
<tr>
<td>Urban</td>
<td>= 1 if located in agglomeration &gt; 250,000 inhabitants</td>
<td>13,098</td>
<td>0.099</td>
<td>0.298</td>
<td>0</td>
<td>1</td>
<td></td>
</tr>
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</table>

### Firm Dummies

<table>
<thead>
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<th>Description</th>
<th>Observations</th>
<th>Mean</th>
<th>Median</th>
<th>Std. Dev.</th>
<th>Min</th>
<th>Max</th>
</tr>
</thead>
<tbody>
<tr>
<td>Small Firm</td>
<td>= 1 if &lt; 10 employees and turnover &lt; 250 million BEF</td>
<td>13,098</td>
<td>0.160</td>
<td>0.367</td>
<td>0</td>
<td>1</td>
<td></td>
</tr>
<tr>
<td>Medium and Large Firm</td>
<td>= 1 if &gt; 10 employees or turnover &gt; 250 million BEF</td>
<td>13,098</td>
<td>0.012</td>
<td>0.111</td>
<td>0</td>
<td>1</td>
<td></td>
</tr>
<tr>
<td>Limited Partnership</td>
<td>= 1 if firm is limited partnership</td>
<td>13,098</td>
<td>0.121</td>
<td>0.326</td>
<td>0</td>
<td>1</td>
<td></td>
</tr>
<tr>
<td>Limited Partnership w/ ES</td>
<td>= 1 if firm is limited partnership with equal sharing</td>
<td>13,098</td>
<td>0.010</td>
<td>0.103</td>
<td>0</td>
<td>1</td>
<td></td>
</tr>
<tr>
<td>Corporation</td>
<td>= 1 if firm is corporation</td>
<td>13,098</td>
<td>0.039</td>
<td>0.194</td>
<td>0</td>
<td>1</td>
<td></td>
</tr>
<tr>
<td>Temporary Arrangement</td>
<td>= 1 if firm is a temporary arrangement</td>
<td>13,098</td>
<td>0.009</td>
<td>0.095</td>
<td>0</td>
<td>1</td>
<td></td>
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</tbody>
</table>

### Firm Characteristics

<table>
<thead>
<tr>
<th>Characteristic</th>
<th>Description</th>
<th>Observations</th>
<th>Mean</th>
<th>Median</th>
<th>Std. Dev.</th>
<th>Min</th>
<th>Max</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age</td>
<td>in years</td>
<td>1,991</td>
<td>16.4</td>
<td>24.3</td>
<td>0</td>
<td>96.2</td>
<td></td>
</tr>
<tr>
<td>Assets</td>
<td>in billions of BEF</td>
<td>645</td>
<td>0.014</td>
<td>0.049</td>
<td>0.000</td>
<td>0.878</td>
<td></td>
</tr>
<tr>
<td>Earnings / Assets</td>
<td>in percent</td>
<td>645</td>
<td>0.117</td>
<td>0.148</td>
<td>-0.528</td>
<td>1.252</td>
<td></td>
</tr>
<tr>
<td>Short-Term Debt / Assets</td>
<td>in percent</td>
<td>645</td>
<td>0.406</td>
<td>0.216</td>
<td>0.001</td>
<td>0.957</td>
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</table>
**Table 2. Samples’ Characteristics**

<table>
<thead>
<tr>
<th>Sample</th>
<th>All</th>
<th>Distance</th>
<th>SPB &amp; SP</th>
<th>Age</th>
<th>Augmented</th>
</tr>
</thead>
<tbody>
<tr>
<td>Number of Observations</td>
<td>13,098</td>
<td>11,222</td>
<td>9,213</td>
<td>1,991</td>
<td>645</td>
</tr>
<tr>
<td>Number of Postal Zones</td>
<td>922</td>
<td>737</td>
<td>717</td>
<td>509</td>
<td>309</td>
</tr>
<tr>
<td>Average Relationship Banking, in %</td>
<td>52.4</td>
<td>53.0</td>
<td>51.4</td>
<td>60.5</td>
<td>65.7</td>
</tr>
</tbody>
</table>

**Table 3. Correlation Table**

The number of observations is 13,098 in the area (1) – (6) and 11,222 elsewhere. *, **, and *** = significant at 10%, 5% and 1% level, using Pearson-correlation.

<table>
<thead>
<tr>
<th></th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(4)</th>
<th>(6)</th>
<th>(7)</th>
<th>(8)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Relationship Banking</td>
<td>0.963***</td>
<td>0.361***</td>
<td>-0.010</td>
<td>-0.003</td>
<td>0.003</td>
<td>0.034***</td>
<td>0.008</td>
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<tr>
<td>ln(Duration of Relationship)</td>
<td>1</td>
<td>0.291***</td>
<td>-0.030***</td>
<td>-0.023***</td>
<td>0.028***</td>
<td>0.098***</td>
<td>0.014</td>
</tr>
<tr>
<td>Main Bank</td>
<td>0.980***</td>
<td>-0.420***</td>
<td>-0.149***</td>
<td>-0.180***</td>
<td>-0.046***</td>
<td>-0.017*</td>
<td></td>
</tr>
<tr>
<td>HHI</td>
<td></td>
<td>1</td>
<td>-0.420***</td>
<td>-0.149***</td>
<td>-0.180***</td>
<td>-0.046***</td>
<td>-0.017*</td>
</tr>
<tr>
<td>HHI²</td>
<td></td>
<td></td>
<td>1</td>
<td>-0.045***</td>
<td>-0.153***</td>
<td>-0.281***</td>
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<td>Multi-Market Contact</td>
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<td>1</td>
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<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(1+Distance to Lender)¹</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
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<td></td>
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<tr>
<td>1−(1+Distance to Closest Competitors)¹</td>
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<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
TABLE 4. BANK ORIENTATION

The dependent variable is Relationship Banking. The definition of the variables can be found in Table 1. The table reports the partial derivatives at the means, in percent, from binary Probit models. *, **, and *** = significant at 10%, 5% and 1% level, two-tailed. SPB & SP: Single-Person Businesses and Sole Proprietorships. The Pseudo R squared is calculated as in Zavoina and McElvey (1975).

<table>
<thead>
<tr>
<th>Model</th>
<th>I</th>
<th>II</th>
<th>III</th>
<th>IV</th>
<th>V</th>
<th>VI</th>
<th>VII</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sample</td>
<td>All</td>
<td>All</td>
<td>All</td>
<td>Distance</td>
<td>SPB &amp; SP</td>
<td>Age</td>
<td>Augmented</td>
</tr>
<tr>
<td>Number of Observations</td>
<td>13,098</td>
<td>13,098</td>
<td>13,098</td>
<td>11,222</td>
<td>9,213</td>
<td>1,991</td>
<td>645</td>
</tr>
<tr>
<td>HHI</td>
<td>3.1</td>
<td>-23.1*</td>
<td>-44.8***</td>
<td>-56.0***</td>
<td>-64.3***</td>
<td>-52.8</td>
<td>-118.1*</td>
</tr>
<tr>
<td>HHI²</td>
<td>23.8**</td>
<td>46.0***</td>
<td>64.1***</td>
<td>67.4***</td>
<td>72.2*</td>
<td>158.7**</td>
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<tr>
<td>Multi-Market Contact</td>
<td>17.5*</td>
<td>28.0***</td>
<td>26.4**</td>
<td>47.4**</td>
<td>112.8***</td>
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<td></td>
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<tr>
<td>(1+Distance to Lender)⁻¹</td>
<td>11.3***</td>
<td>12.6***</td>
<td>11.9</td>
<td>33.0**</td>
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<tr>
<td>1 – (1 + Distance to Closest Competitors)⁻¹</td>
<td>3.8</td>
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<td>8.3</td>
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<td>Firm Characteristics</td>
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<td>Pseudo R squared</td>
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<td>0.406</td>
<td>0.406</td>
<td>0.406</td>
<td>0.404</td>
<td>0.392</td>
<td>0.404</td>
</tr>
</tbody>
</table>
**Table 5. Bank Orientation: Main Bank**

The dependent variable is Main Bank. The definition of the variables can be found in Table 1. The table reports the partial derivatives at the means, in percent, from binary Probit models. *, **, and *** = significant at 10%, 5% and 1% level, two-tailed. SPB & SP: Single-Person Businesses and Sole Proprietorships. The Pseudo R squared is calculated as in Zavoina and McElvey (1975).

<table>
<thead>
<tr>
<th>Model</th>
<th>I</th>
<th>II</th>
<th>III</th>
<th>IV</th>
<th>V</th>
<th>VI</th>
<th>VII</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sample</td>
<td>All</td>
<td>All</td>
<td>All</td>
<td>Distance</td>
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**TABLE 6. BANK ORIENTATION: DURATION OF RELATIONSHIP**

The dependent variable is ln(Duration of Relationship). The definition of the variables can be found in Table 1. The table reports the partial derivatives at the means from Tobit models. *, **, and *** = significant at 10%, 5% and 1% level, two-tailed. SPB & SP: Single-Person Businesses and Sole Proprietorships.

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The dependent variable is Relationship Banking. The definition of the variables can be found in Table 1. The table reports the partial derivatives at the means, in percent, from binary Probit models (RB), or the coefficients from ordinary least squares models (IS). *, **, and *** = significant at 10%, 5% and 1% level, two-tailed. SPB & SP: Single-Person Businesses and Sole Proprietorships.

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### Table 8. Bank Industry Specialization

The dependent variable is Industry Specialization. The definition of the variables can be found in Table 1. The table reports the coefficients from ordinary least squares models. *, **, and *** = significant at 10%, 5% and 1% level, two-tailed. SPB & SP: Single-Person Businesses and Sole Proprietorships.

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