

ENDOGENOUS PRODUCT DIFFERENTIATION IN CREDIT MARKETS: WHAT DO BORROWERS PAY FOR?*

Moshe Kim[†] Eirik Gaard Kristiansen[‡] Bent Vale[§]

May 14, 2001

Abstract

This paper studies strategies pursued by banks in order to differentiate their services from those of their rivals. In that way the level of competition in the industry is reduced. More specifically we analyze whether the bank size, a bank's ability at avoiding losses, or its capital ratio can be used as strategic variables to make banks different and increase the interest rates banks can charge their borrowers in equilibrium. Using a panel of data covering Norwegian banks between 1993 and 1998 we find empirical support for the the ability at avoiding losses measured by the ratio of loss provisions as such a variable. Borrowers in the market for credit line loans may discipline banks into avoiding losses. We also find evidence that banks pass on parts of increases in their operating costs to credit line borrowers. However, we do not find any strong evidence for the use of high capital ratio as a strategic variable that borrowers are willing to pay for. This implies that strategic competition does not lead banks to hold more capital than their cost minimizing level. Thus, our paper gives support to the rationale for imposing capital requirements on all banks.

JEL code: G21 L15

*Views and conclusions expressed are the the responsibility of the authors alone and cannot be attributed to Norges Bank.

[†]University of Haifa, Department of Economics, Haifa 31905 Israel, and Humboldt University, Berlin. Fax: +972 4 824 0059, e-mail: kim@econ.haifa.ac.il

[‡]Norges Bank (The central bank of Norway). Adress: Norges Bank, C51, Box 1179, Sentrum, N-0107 Oslo Norway. Fax: +47 22 42 40 62, e-mail: eirik.kristiansen@norges-bank.no

[§]Norges Bank (The central bank of Norway) and Norwegian School of Management. Adress: Norges Bank, C51, Box 1179, Sentrum, N-0107 Oslo Norway. Fax: +47 22 42 40 62, e-mail: bent.vale@norges-bank.no

1. Introduction

There exists a vast literature investigating the nature of competition in markets with differentiated products. In this paper we focus on *endogenous* differentiation among banks. More precisely, how banks strategically choose different “quality” characteristics (equity ratios, monitoring capacities, screening efforts, size etc.) in order to differentiate themselves from competing banks given that customers exhibit different characteristics, and thereby soften competition. This focus enables the analysis of why banks become different and not why different borrowers choose different contracts, which is extensively analyzed in the literature.

Different groups of borrowers have different characteristics or needs. For instance, some borrowers face large lock-in effects due to the fact that their current bank has an informational advantage compared to competing banks (see Sharpe (1990)). These borrowers are inclined to choose banks that they anticipate are able to extend credit lines or provide new loans in future periods (switching to another bank is costly). Hence, they may be willing to pay for loans which originate from a bank that is well diversified and therefore has a small probability of experiencing future losses and be less able to serve their customers in the future.¹ Usually diversification will increase with the number of bank loans, and hence the size of the bank (see for instance Diamond (1984)). On the other hand large banks may face larger principal agent problems in their organizations than small banks do, and thus perform monitoring and screening of less quality than smaller banks. For instance, Williamson (1967) and Qian (1994) discuss theories that may explain why agency costs may increase with the size of a hierarchical firm. Cerasi and Daltung (????) provides an interesting combination of these principal agency theories and the theory of diversification a la Diamond (1984) to arrive at a theory of the optimal size of the banking firm. Borrowers for whom tight monitoring and screening is important may hence be willing to pay for being served by a smaller bank. For borrowers willing to accept low quality of monitoring and screening functions, it may turn out to be the opposite. This example of bank size as a quality characteristic of the lender illustrates that theory alone cannot provide a definite answer as to what characteristics

¹Peek and Rosengren (1997) provide empirical evidence for a negative relation between loan losses at banks and their concurrent supply of loans.

borrowers are willing pay for; this is an empirical issue, which we explore in this paper. We also consider other quality characteristics of banks in addition to size.

The fact that borrowers are different opens up an opportunity for banks to differentiate their quality characteristics (provide different services) in order to soften competition.

The majority of the literature dealing with debt structure analyzes the decision made by firms of whether to utilize arm's length (publicly traded) versus bank (privately held) debt. A comprehensive review of the issue is given in Boot (2000). An interesting recent paper by Cantillo and Wright (2000) investigates the characteristics determining which companies finance themselves through intermediaries and which borrow directly from arm's length investors. In the present paper however, we focus on and restrict our attention to debt taken from the banking sector only, and analyze why and how banks choose different characteristics to differentiate themselves in order to soften competition. The reasons for focusing our analysis to debt taken from the banking sector alone are two-fold. First, in most European countries markets for arm's length debt (bonds and certificates) are still relatively thin.² Second, since the major reasons for firms choice between arm's length and bank debt are quite well known (see e.g., James (1987) about value enhancing bank-originated debt), it would be of much interest to further analyze and investigate the type of banks and their characteristics that borrowers are willing to pay for once they have decided to raise bank debt.

Before conducting the empirical analysis, we provide a simple theoretical model which can shed some light on ways banks can utilize borrower-heterogeneity in order to differentiate themselves. In the empirical part, we use data from the Norwegian banking industry to show along which dimensions banks find it most profitable to differentiate and soften competition.

The paper is organized in the following way: section 2 presents the theoretical model; section 3 describes the data used and variables calculations, and the empirical model is presented; section 4 presents empirical results and discussion. Section 5 concludes the paper.

²OECD statistics show that bond and certificates as of 1995 comprized only around 4.0%-5.0% of total funding for the private non-financial firms in Europe.

2. A theoretical model

A two-stage model is introduced to illustrate how banks can differentiate their services in order to attract borrowers with different financial needs. In general, banks can pursue two kinds of differentiation strategies. A bank can differ from the other banks in a way that *all* borrowers consider as better than its competitors (e.g. better services). In the literature this kind of differentiation is denoted *vertical product differentiation* which is the kind of product differentiation we consider here. In contrast, horizontal product differentiation does not imply that all borrowers agree about whether or not a bank offers better services than its competitors. For example, a bank may move a branch from one city to another. We will briefly discuss horizontal product differentiation in the section containing our empirical results.

In the theoretical model we are deliberately vague about exactly which strategic variable banks use in their vertical product differentiation strategy. In the empirical part we analyze different potential "quality" variables that banks can use to differentiate.

For simplicity, we study the case with two banks, bank A and bank B. At stage 1, the banks choose their quality variables, q_i , $i = A, B$ and, at stage 2, the banks choose interest rates, r_i , $i = A, B$ (Bertrand competition). This two-stage structure captures that some characteristics are strategic variables that are more costly or difficult to alter than interest rates. Figure 1 presents a schematic diagram of the two-stage game

Stage 1

Banks choose quality variables, q_A and q_B , simultaneously.

Stage 2

Banks choose interest rates, r_A and r_B , simultaneously.

Borrowers accept an offer from one of the banks.

Figure 1: Competition in a two-stage game

There are numerous potential ways a bank can distinguish itself from its competitors. If bank relationships are important, borrowers may be concerned about the

capabilities or characteristics of their main bank. Let us here briefly point out some potential quality variables in banking. Which quality variables that are important is an empirical question examined in the following empirical part of the paper.

- *Monitoring/screening*: Some banks may be well-known for having high-quality staff that is experienced evaluators of investment projects. A low level of losses on loans may indicate high monitoring/screening capabilities. A borrower that faces switching costs may favor such a bank since it increases the probability of correct evaluation of future loan applications (profitable projects obtain loans). This idea has been explored in Chemmanur and Fulghieri (1994).
- *Signalling*: Bank loans may signal the quality of the borrowing firm to stock owners, buyers, suppliers and other creditors.³ A loan commitment from a high-quality bank may provide a more favorable signal than a similar loan commitment from low-quality banks. This is so because high-quality banks provide a more thorough screening and monitoring of its borrowers. Furthermore, some banks have a reputation for being more risk averse than others. Consequently, a loan commitment from a bank with a reputation for being highly risk averse can signal that its borrowers have low-risk for going bankrupt. In this way, a bank loan can be used to levitate the asymmetric information problems a firm may face in negotiations with for example suppliers and buyers. Signalling quality of a bank can for instance be negatively associated with the extent to which the bank has suffered losses on its loan portfolio.
- *Bank solvency* (equity or profitability): Empirical literature has shown that borrowers may suffer if their main bank is forced to restrict its lending capacity (see Slovin, Sushka, and Polonchek (1993)). Consequently, a borrower may be concerned about their main bank's solvency or, more precisely, how likely it is that their bank may face difficulties in providing loans in the future.
- *Size*: Large banks tend to have a taller hierarchical structure than small banks. Generally, Qian (1994) shows that this may lead to weak incentives for workers

³The empirical study of Billett et al. (1995) show loans from high-quality lenders are associated with more positive stock price reactions than loans from low-quality lenders.

in the lowest hierarchical tiers. Consequently, in large organization workers are frequently guided by rules rather than a more flexible incentives system. Along these lines, Cerasi and Daltung (1994) focus on how diversification benefits in banking may be countervailed by losses due to weaker incentives in the lowest tiers in large banks. Incentive problems may lead to restriction on local branch managers' authority to provide loans and to the use of inflexible rules.⁴ Some borrowers may fear that future loan applications may be turned down in rules oriented banks.⁵ Hence, some borrowers may prefer to use small and more flexible banks.

Our very simple model structure is able to capture all these potentially important bank characteristics.

In the first period borrowers are assumed to have access to an investment project with present value, V (not including financing costs).

In the following analysis we take into account that there are many different qualities of a bank that borrowers may consider when they choose their main bank. For simplicity, we denote borrower f 's value of a bank relationship with a bank of quality q ,

$$q \cdot R_f$$

where R_f represents borrower f 's need for a long term bank relationship and q represents the (current and expected) quality of the bank. For example, a borrower which does not need additional funding in the future has a low R_f (possibly 0). In contrast, a borrower that is locked into a relationship with a particular bank (high switching cost) and need high quality service today as well as in the future, would have a high R_f .⁶ Note that R_f cannot be observed by the bank and hence a bank must charge the same interest rate from all its borrowers. On the other side, banks may also differ in many ways. A well capitalized bank will very unlikely be forced to introduce credit restrictions. Furthermore, borrowers can be ensured that banks that have suffered low losses in the past due to high-quality monitoring and screening

⁴For example, the local manager may face rules regarding borrowers equity or cash flow.

⁵Besides, weak incentives may lead to less thorough evaluation of loan applications.

⁶See for example Sharpe (1990) for a discussion of switching costs due to information asymmetries between lenders.

of loan applicants are able to do skillful evaluation of a new projects in the future. A high-quality bank is characterized by a high q . In the theory model, we do not allow the banks to differentiate along many dimensions at the same time. However, our simple framework suffices to show that heterogenous borrowers enable the banks to pursue differentiation strategies and thereby reduce the level of competition.

We make three simplifying assumptions:

Assumption 1. R_f is uniformly distributed on $[0, 1]$.

Assumption 2. Total market demand is normalized to 1.

Assumption 3. Let the costs related to deviation from a banks' cost minimizing quality level, q_o , be quadratic

$$e(q_i) = \beta (q_i - q_o)^2 \quad i = A, B$$

Note that the cost minimizing quality level, q_o , can be interpreted as the quality level that would have been chosen in the absence of strategic interactions among the banks. The banks deviate from this level in order to ease the level of competition in interest rates in period 2 (with identical banks, $q_o = q_A = q_B$ competition would be fierce and there would be no profit).

To find the sub-game perfect Nash equilibrium in the two-stage game we start with stage 2.

2.1. Competition at stage 2

First, let us examine the demand for loan given q_A , q_B , r_A , and r_B . Without loss of generality assume $q_A \geq q_B$, which implies that $r_A \geq r_B$ (otherwise bank B's offer dominates bank A's offer). Borrower f compares the net benefits from using bank A and bank B:⁷

$$\text{Bank A:} \quad V - r_A + q_A \cdot R_f$$

⁷For simplicity we have assumed that the project has a certain outcome. However, we could have assumed that there is a probability $p < 1$ for success. In case of failure the project is worthless. Then, the expected value of the project would have been: $p[V - r_i + q_i R_i]$. The choice between the two banks would not change.

$$\text{Bank B:} \quad V - r_B + q_B \cdot R_f$$

A borrower of type \widehat{R} , is indifferent between using bank A and bank B.

$$\begin{aligned} V - r_A + q_A \cdot \widehat{R} &= V - r_B + q_B \cdot \widehat{R} \\ \widehat{R} &= \frac{r_A - r_B}{q_A - q_B} \end{aligned}$$

Consequently, bank A and bank B face demand $D_A(r_A, r_B)$ and $D_B(r_A, r_B)$, respectively

$$\begin{aligned} D_A(r_A, r_B) &= 1 - \widehat{R} \\ D_B(r_A, r_B) &= \widehat{R} \end{aligned}$$

and the banks' profit levels are

$$\begin{aligned} \pi_A(r_A, r_B) &= (r_A - r_0) D_A(r_A, r_B) \\ \pi_B(r_A, r_B) &= (r_B - r_0) D_B(r_A, r_B) \end{aligned} \tag{2.1}$$

where r_0 is the banks' cost of funding. From the two banks' profit maximizing choice of interest rates, we get the Nash equilibrium at stage 2.

$$\begin{aligned} r_A &= (q_A - q_B) \frac{2}{3} + r_0 \\ r_B &= (q_A - q_B) \frac{1}{3} + r_0 \end{aligned} \tag{2.2}$$

From, equation (2.1) and (2.2) we have

$$\begin{aligned} \pi_A(q_A, q_B) &= \frac{4}{9}(q_A - q_B) \\ \pi_B(q_A, q_B) &= \frac{1}{9}(q_A - q_B) \end{aligned}$$

2.2. Competition at stage 1

At stage 1 the banks decide their strategic variables (q_A and q_B) taking as given the profit maximization behavior at stage 2.

Seen from stage 1 the banks' profit maximization problems are:

$$\text{Bank A: } \underset{q_A}{Max} \{ \pi_A(q_A, q_B) - e(q_A) \}$$

$$\text{Bank B: } \underset{q_B}{Max} \{ \pi_B(q_A, q_B) - e(q_B) \}$$

From the first order conditions we get

$$q_A^* = q_o + \frac{2}{9} \frac{1}{\beta}$$

$$q_B^* = q_o - \frac{1}{18} \frac{1}{\beta}$$

Proposition 1 sums up our predictions from the theoretical model

Proposition 1.

i) *If the banks become more differentiated, their interest rates and profitability increase*

$$\frac{dr_i}{d(q_A^* - q_B^*)} > 0, \quad i = A, B,$$

$$\frac{d\pi_i}{d(q_A^* - q_B^*)} > 0, \quad i = A, B, .$$

ii) *The bank with the highest level of the strategic quality variable is most profitable. Since $q_A^* > q_B^*$, $\pi_A > \pi_B$.*

Proposition 1 ii) implies that both banks would prefer to be the high quality bank but i) implies that both would loose if both become high quality banks (i.e. $q_A - q_B$ is small).⁸

3. Empirical model

In this section we present the empirical model that can facilitate a test of the prediction of Proposition 1 i); as banks are more dispersed in terms of a certain bank quality variable that borrowers appreciate and hence may be willing to pay for, competition is softened and banks are able to charge borrowers higher interest rates. More specifically we want to find out empirically what characteristics of a bank

⁸In this model as in all other models with ex ante symmetric agents and ex post asymmetric profit levels, there is a potential coordination problem.

borrowers are willing to pay for, and hence along what characteristics banks can distinguish themselves from each other in order to soften competition.

The general structure of our empirical model is:

$$s_{i,r,t} = f(s_{i,r,t-1}, v(\mathbf{q})_{i,r,t-h}, g(\mathbf{q})_{r,t-h}, \mathbf{x}_{i,t-h}, \mathbf{f}_{r,t-h}, \mathbf{z}_{t-h}, \epsilon_{i,r,t}) \quad , \quad (3.1)$$

where $s_{i,r,t}$ is the spread over the period t money market interest rate on loans from bank i in market r in period t . $v(\mathbf{q})_{i,r,t-h}$ is a vector representing the difference between the value of bank i 's quality variables and the cross-sectional median of the corresponding bank quality variables in market r in period $t-h$. $h \in [0, T]$ is the appropriate lag length for the various explanatory variables. $g(\mathbf{q})_{r,t-h}$ is a vector containing for each bank quality variable a measure of the inequality in that variable across banks in market r in period $t-h$. $\mathbf{x}_{i,t-h}$ is a vector of other bank and period specific variables that can influence the interest rate spread $s_{i,r,t}$ (e.g., the riskiness of a bank's loan portfolio). $\mathbf{f}_{r,t-h}$ is a vector of variables specific to market r . \mathbf{z}_{t-h} is a vector of macro economic variables. $\epsilon_{i,r,t}$ is the error term.

The type of interest rates we consider are the interest rates banks charge on credit lines to firms. Hence $s_{i,r,t}$ is the spread of interest rates on credit lines over the money market rate. Credit lines are usually considered as the most information intensive type of loans, see Berger and Udell (1995)⁹. Thus problems of lock-in and high switching costs are likely to be more profound in markets for credit lines than in other loan markets. Therefore the quality of a bank should be more important for credit line customers than for other loan customers. We therefore apply and test the hypothesis that credit line borrowers are willing to pay extra for borrowing from a bank of high quality.

The theory model in section 2.1 predicts that the better bank i is compared to its competitors in terms of a certain quality variable, the higher equilibrium interest rate it will be able to charge its borrowers. That is the motivation for specifying the variables contained in $v(\mathbf{q})_{i,r,t-h}$ as differences from the cross-sectional median of the corresponding bank quality variables in the market in which bank i operates. I.e. these variables represent what in the IO literature is referred to as vertical

⁹Mester (1992) estimates a cost function based on information-theoretic considerations, realizing the different costs entailed in the provision of different information-intensive outputs.

differentiation. However, when more than two banks are competing in the same market it is not just how much better or worse bank i is, that matters for its competitive position, i.e. how much it is able to charge its borrowers. The overall differentiation of all competitors in terms of the quality variable will also matter. A larger dispersion will soften the overall competition in the market and enable all banks to charge their borrowers a higher margin. That is the motivation for including $g(\mathbf{q})_{r,t-h}$ that represents the cross-sectional inequality or dispersion of the quality variables in each market. As will be shown below, the inequalities are measured by Gini coefficients.

Markets are defined by geography, and the country is divided into 18 regions, as is explained in more details in subsection 3.1.

3.1. Data

We use a panel of Norwegian bank data covering the years 1993 to 1998. This is the period immediately following the banking crisis in Norway. In the crisis three of the four largest banks failed and were recapitalized by the government subject to trimming of the banks' balances and operating costs. Smaller banks that failed were acquired by sounder banks by the help of guarantees from the deposit insurance funds. Only one small bank was forced to close. Thus, all other problem banks were allowed to continue their operations but, as mentioned, on certain conditions. It can therefore be assumed that in the years covered by our data both banks and their borrowers had learnt about possible consequences of a bank running into solvency problems.

The data are annual and include banks from small local savings banks to large nationwide banks. This large variety in the data ensures a relatively large dispersion of various characteristics of the banks. The data consists both of balance items, items from the banks' result account and average interest rates by the end of the year on some specific loan aggregates. The number of banks in the sample used varies between a maximum of 98 in 1997 and 71 in 1993.¹⁰ Norway is divided into 19 counties. Loans outstanding for each bank are also reported by county.

Markets are defined by geography, and the country is divided into 19 counties.

¹⁰Only banks reporting the necessary data are included in the sample.

We define each county as one market. The capital Oslo, which itself is a county, and the county surrounding it, Akershus, are defined as one market, leaving us with a total of 18 markets. The majority of Norwegian banks only operate in one or two counties. Only the three largest banks are represented in all of the 18 loan markets defined here in the whole period covered. The fourth largest bank is represented in all 18 markets in three of six years.¹¹

As the data on interest rates charged by the banks are not specified by county we have to maintain the hypothesis that there is no systematic variation in the interest rates on credit lines across counties, thus any variation is random and is captured by the error term of the model. However, we have data on total loans by all banks by county, which allows us to define what banks operate in what county. Characteristics of the banks other than loans are not specified by county. However, most of the characteristics of a banking firm that matters to the borrowers (its solvency, capital ratio, overall size, its overall ability to screen and monitor etc.) would be constant across counties. Hence for our purpose this can not be considered a severe limitation of the data.

3.2. Specification of the empirical model

We estimate the following log-linear version of 3.1¹²:

$$s_{i,t} = \alpha_0 + \alpha_1 s_{i,t-1} + v(\mathbf{q})_{i,r,t-h} \boldsymbol{\beta} + g(\mathbf{q})_{r,t-h} \boldsymbol{\gamma} + \mathbf{x}_{i,r,t-h} \boldsymbol{\delta} + \mathbf{f}_{r,t-h} \boldsymbol{\mu} + \mathbf{z}_{t-h} \boldsymbol{\theta} + \epsilon_{i,r,t} \quad (3.2)$$

where $h \in [0, 1]$.

The model is not specified as a fixed effects model, as is common in many analysis of panel data. Fixed effects would capture influence on the LHS variable that are specific to each bank but stays constant over time. In the present context where the LHS variable is the spread of the interest rate on loans, such factors would typically arise from the composition of the bank's loan portfolio. It is well known

¹¹In cases where a bank has less than 0.1 pct. of the loan market in a county, it is considered not represented in that county, and that particular combination of bank and county is not included in the data set. If this was not done, small banks, having a few borrowers that physically have moved to another county and maintained their loans in the original bank, would have been considered as actively competing for loans in that county. This also implies that a few very small banks are not included in the sample nor as competing banks to those in the sample.

¹²A linear version of the same model did not pass the RESET test for functional form.

from other empirical work that the composition of a bank’s borrowers changes over time but only slowly (see for instance Ongena and Smith (1998), Degryse and van Cayseele (2000) and Kim, Kliger, and Vale (1999)). The influence on $s_{i,t}$ from this sluggishness in the bank’s loan portfolio is therefore better captured by allowing a lagged value of $s_{i,t}$ among the RHS variables than just using a set of fixed effects.

We specify a log-linear version of 3.2 with the following RHS variables:

Variable	Description
$s_{i,t-1}$	Lag of the spread of interest rate on credit lines
Bank quality $v(\mathbf{q})_{i,r,t-h}$:	
$v(assets)_{i,r,t-1}$	Total assets of bank i end of year $t - 1$
$v(cap88)_{i,r,t-1}$	Capital ratio(Basel 88) of bank i end of year $t - 1$
$v(loss)_{i,r,t-1}$	Ratio of accumulated loss provisions for bank i end of year $t - 1$
Gini coefficients of quality $g(\mathbf{q})_{i,r,t-h}$:	
$g(assets)_{r,t-1}$	
$g(cap88)_{r,t-1}$	
$g(loss)_{r,t-1}$	
Controls ($\mathbf{x}_{i,t-h}$; \mathbf{z}_{t-h} ; $\mathbf{f}_{r,t}$):	
$costrat_{i,t}$	Ratio of materials- and wagecost to loans outstanding for bank i in year t
$cl_lossrat_{i,t}$	Ratio of accumulated loss provisions on credit line loans to credit line loans outstanding for bank i in year t
$realrate_t$	3 month money market interest rate annual average minus annual growth in CPI in year t
$herfin_{r,t}$	Herfindahl index of the bank to business credit market in county r in year t

$v(q)_{i,r,t-h}$ is a vector representing the difference between the value of bank i ’s quality variables and the cross-sectional median of the corresponding bank quality variables in market r in period $t - h$. $h \in [0, T]$ is the appropriate lag length for the various explanatory variables. $g(q)_{r,t-h}$ is a vector containing for each bank quality variable a measure of the inequality in that variable across banks in market r in period $t - h$.

All lagged stock variables are aggregated backwards, i.e. the bank structure of year t is forced upon the variable in year $t - 1$.

The variables listed under the heading ‘bank quality variables’ are variables that borrowers are likely to take into account as signals by banks when choosing a bank. The operator v represents the cross-sectional difference of a quality variable q in the

following way:

$$v(q)_{i,r,t} = q_{i,t} - \frac{1}{n_r} \sum_{i \in r} q_{i,t} \quad ,$$

where n_r is the number of banks operating in county r , and $i \in r$ indicates that summation is done only over banks operating in county r . Note that even if $q_{i,t}$ only varies across banks and years $v(q)_{i,r,t}$ in addition varies across counties.

The Gini coefficients are calculated as:

$$g(q)_{r,t} = 1 + \frac{1}{n_r} - \frac{2}{n_r^2 \bar{q}_{r,t}} \sum_{i \in r} j \cdot q_{i,t}$$

$$\text{where } \bar{q}_{r,t} = \frac{1}{n_r} \sum_{i \in r} q_{i,t} \quad , \quad j = 1, 2, 3, \dots, n_r \quad ,$$

and j is a rank number assigned to each $q_{i,t}$ in decreasing order of size.

assets represents the size of a bank. The larger the bank is the more diversified its portfolio is likely to be, and *cet.par.* the less likely it is that the bank will get huge losses and be forced to reduce its lending activity. Furthermore a larger bank is also more likely to be considered as “too big to fail” by the government. On the other hand large banks may face more severe principal/agent problems in their organizations than small banks do, and thus perform monitoring and screening of less quality than smaller banks. For instance, Williamson (1967) and Qian (1994) discuss theories that can explain why agency costs may increase with the size of a hierarchical firm. One way to reduce these agency costs is for a large bank to be more governed by rules rather than discretion, which for the borrowers may result in more rigid loan conditions. Cerasi and Daltung (????) provides an interesting combination of these principal/agent theories and the theory of diversification a la Diamond (1984) to arrive at a theory of the optimal size of the banking firm. Borrowers for whom tight monitoring and screening and more flexible loan conditions is important may hence be willing to pay for being served by a smaller bank. For borrowers willing to accept low quality of monitoring and screening functions or more rigid loan conditions, it may turn out to be the opposite. Hence the expected impact of $v(\text{assets})_{i,r,t-1}$ on the left hand side is ambiguous.

cap88 represents the solvency of a bank in terms of being able to withstand large loan losses without being forced to cut its lending in order to satisfy the

capital requirements. This variable can have a positive impact on the left hand side, as borrowers may be willing to pay for this sign of quality. A positive sign can also indicate that banks under imperfect competition are able to pass on to borrowers the higher total costs of finance due to a higher capital ratio. However, differentiation among banks in terms of their risk aversion can be reflected in a negative sign of $v(cap88)_{i,r,t-1}$ in this reduced form model. Banks with high risk aversion are expected to operate both with a high capital ratio – to minimize the possibility of coming below the minimum requirement – and at the same time lend to safe borrowers, borrowers from which they only can charge a low interest rate spread. Also differentiation among banks along their attitude towards risk can serve to soften competition and enable banks to charge a higher spread on the interest rate. The extreme case of this story would be a bank very close to or even below the minimum capital requirement that behaves like a risk lover or ‘gambles for resurrection’ by lending to high risk borrowers from which it can charge a high interest rate spread. Thus the expected sign of the coefficient $v(cap88)_{i,r,t-1}$ is indeterminate.

loss as a quality variable represents the results of the bank’s ability to screen and monitor, as well as borrowers’ need to signal their low riskiness to other creditors and to their customers, by lending from a bank that has suffered few loan losses (c.f. the discussion in the two first bullet points of section 2). To the extent borrowers are willing to pay for this quality variable, the expected impact of $v(loss)_{i,r,t-1}$ on the left hand side is negative.

The expected sign of the estimated parameters for all the Gini coefficients are positive. More dispersion among banks in terms of variables borrowers care about, serves to soften competition and hence increase the interest rate banks can charge their borrowers. Recall that a key assumption – and a fairly realistic one – in this paper is the heterogeneity of borrower preferences. One of the Gini coefficients of an underlying variable insignificantly different from zero indicates that more dispersion among competing banks along this variable does not soften competition. Under our heterogeneity assumption this also implies that the underlying variable is not a quality variable as defined in section 2.

Among the control variables $costrat_{i,t}$ represents the banks’ ability under imperfect competition to pass their operating costs on to their credit line borrowers.

$cl_lossrat_{i,t}$ represents the idiosyncratic part of the credit risk in the bank's portfolio of credit line loans. $realrate_t$, the real interest rate, represents the macroeconomic part of the banks' credit risk. This variable passes a Dickey-Fuller test and hence is a stationoray macro variable. A higher real interest rate is associated with higher credit risk due to an expected slowdown in the overall economic activity. Hence the expected sign on the coefficient of this variable is positive.

The regional Herfindahl index $herfin_{r,t}$ controls for the competitive environment, as measured by market concentration, in which a bank operates. The more concentrated the market is the higher is the value of the Herfindahl index. A more concentrated market is usually considered a less competitive market, and banks should be able to charge a higher interest rate. Hence the expected sign of this variable should be positive. However, as problems of asymmetric information and adverse selection are expected to be more severe in the market for lines of credit, the 'winner's curse' problem may be considerable in this market: The more banks that operate in the market, and borrowers shop around, the more likely it is that the lowest quality borrowers mistakenly are assumed to be good borrowers by some banks. Broecker (1990) and Shaffer (1998) study the theory of winner's curse in banking. Moreover Shaffer (1998) also finds empirical support for the hypothesis that banks operating in less concentrated markets have higher loan losses. The bank competing most aggressively for borrowers in such a market by offering low interest rates (the 'winner') will almost certainly capture a larger fraction of the low quality borrowers. Banks being aware that the winner's curse problem is more severe in a less concentrated market will compete less aggressively. Thus, banks may charge higher interest rates the less concentrated is the market, if the winner's curse problem is large.¹³¹⁴ All in all the sign of $herfin_{r,t}$ can be both positive and negative, a negative sign indicating the presence of a large winner's curse problem in the market for credit lines.

¹³Bulow and Klemperer (1999) construct a theory model of auctions where fewer bidders actually raises the price when bidders are asymmetric. This is exactly so because the winner's curse effect is shown to dominate the traditional effect of less competition. In our setting this will correspond to fewer banks bidding for borrowers leading to lower interest rate.

¹⁴Furthermore, to the extent the variables controlling for the riskiness of the banks' loan portfolio only do so incompletely, the effect on loan losses identified by Shaffer (1998) may show up as higher ineterest rate margin in our model. This effect will also imply a positive sign on $herfin_{r,t}$.

We use a lag of one year for all the quality variables. Borrowers have to base their evaluation of the bank on the values published in the bank’s annual report and financial statements for the last year. These are usually more comprehensive and more scrupulously audited statements than the quarterly statements made during the year.

4. Empirical results

The model in 3.2 is estimated using two-stage least square. Both $costrat_{i,t}$ and $cl_lossrat_{i,t}$ are endogenous, they may be partially determined by the LHS variable $s_{i,t}$. They are instrumented using their own one year lags, not aggregated backwards and aggregated backwards respectively¹⁵. The correlation between $\ln costrat_{i,t}$ and its lag is 0.91, and between $\ln cl_lossrat_{i,t}$ and its lag is 0.74.

We start by estimating the general model including all the RHS variables listed in section 3.2, a model that satisfies certain misspecification tests regarding lack of serial correlation in the residuals and functional form. The results are presented in Table 4.1 column (a).

Model (a) passes the misspecification tests for functional form and for no positive serial correlation in the residuals. However inclusion of the size as a quality variable does not pass a test of complying with economic theory as the coefficient of $\ln g(assets)_{r,t-1}$ is negative and the coefficient of $\ln v(assets)_{i,r,t-1}$ is insignificant. Hence we find no evidence of size as a quality variable.¹⁶ Thus we proceed by specifying a model without size as quality variable, model (b). Model (b) also passes the misspecification tests for functional form and for no positive serial correlation in the residuals. Furthermore exclusion of the two insignificant variables $\ln realrate_t$ and $\ln herfin_{r,t}$ from model (b) is statistically valid and we get the parsimonious model (c) which also passes the tests for functional form and for no positive serial correlation in the residuals. Note that due to the log-linear specification all coefficients of the model can be interpreted as elasticities.

The negative and significant coefficient of $\ln v(loss)_{i,r,t-1}$ and the positive and

¹⁵Backward aggregation of a variable means that the bank structure of year t is forced upon the variable in year $t - 1$.

¹⁶The same qualitative results regarding size as quality variable appeared also when total loans, number of employees or number of branches were used as measures of size.

Table 4.1: Empirical results

LHS variable $\ln s_{i,t}$ (spread over money market interest rate on credit line loans)			
Variable	(a)	(b)	(c)
$\ln s_{i,t-1}$	0.3008 (6.26)	0.3045 (6.71)	0.3010 (7.13)
Bank quality $v(\mathbf{q})_{i,r,t-h}$:			
$\ln v(\text{assets})_{i,r,t-1}$	-0.0072 (-0.94)	—	—
$\ln v(\text{cap88})_{i,r,t-1}$	-0.1435 (-4.25)	-0.1232 (-5.23)	-0.1233 (-5.39)
$\ln v(\text{loss})_{i,r,t-1}$	-0.1064 (-5.36)	-0.1196 (-6.54)	-0.1204 (-6.74)
Gini coefficients of quality $g(\mathbf{q})_{i,r,t-h}$:			
$\ln g(\text{assets})_{r,t-1}$	-0.3473 (-3.40)	—	—
$\ln g(\text{cap88})_{r,t-1}$	0.2563 (6.93)	0.1877 (5.53)	0.1913 (6.07)
$\ln g(\text{loss})_{r,t-1}$	0.4285 (7.28)	0.4263 (7.89)	0.4360 (8.34)
Controls ($\mathbf{x}_{i,t-h}$; \mathbf{z}_{t-h} ; $\mathbf{f}_{r,t}$):			
$\ln \text{constrat}_{i,t}$	0.1284 (2.53)	0.1277 (2.51)	0.1286 (2.60)
$\ln \text{cl_lossrat}_{i,t}$	0.0233 (2.09)	0.0275 (2.50)	0.0293 (2.79)
$\ln \text{realrate}_t$	0.0313 (2.02)	0.0174 (1.15)	—
$\ln \text{herfin}_{r,t}$	-0.0845 (-2.50)	-0.0246 (-0.87)	—
F -test, (b) – (c)	—	—	0.26
AR(1)	-0.3191 (-6.23)	-0.3399 (-6.43)	-0.3276 (-6.21)
RESET	-0.2511 (-0.24)	-0.1140 (-0.11)	-0.0486 (-0.05)
R^2 adj.	0.3537	0.3481	0.3482

Number of observations is 984. $v(\mathbf{q})_{i,r,t-h}$ is a vector representing the difference between the value of bank i 's quality variables and the cross-sectional median of the corresponding bank quality variables in market r in period $t-h$. $h \in [0, T]$ is the actual lag length for the various explanatory variables. $g(\mathbf{q})_{r,t-h}$ is a vector containing for each bank quality variable a measure of the Gini coefficient of that variable across banks in market r in period $t-h$. Numbers in parantheses are White robust t -values. The F -test is a test of the joint significance of the variables excluded from model (b), the p-value is reported. AR(1) is a test for auto correlation of order 1 in the residuals. An alternative test for serial correlation in models with the lagged dependent variable on RHS is applied, $u_{i,r,t}$ is regressed on $u_{i,r,t-1}$ controlling for all the RHS variables of the model. The coefficient of $u_{i,r,t-1}$ and its t -value are reported, (see Greene (1993) p. 428. RESET is the test for functional form using the square of the predicted value as RHS. Its coefficient and t -value are reported.

significant coefficient of the corresponding $\ln g(loss)_{r,t-1}$ supports the hypothesis that borrowers are willing to pay for the ability of a bank at avoiding losses (c.f. the two first bullet points in section 2). In addition, when banks to a different degree specialize in being low loss banks, they soften the competition by dividing the market as borrowers have different needs for banks with low losses. A bank at sample mean with an interest rate spread on its credit line loans of 4.85 pct. will be ‘punished’ by approx. 0.5 pct. points on its spread if its loss provisions relative to its competitors double. This indicates that there may be a market discipline effect at work not only in the money market, but also in the market for credit line loans, for banks reporting higher loan loss provisions.

The negative sign for $\ln v(cap88)_{i,r,t-1}$ supports the claim that banks with high risk aversion tend to both have high capital ratio and lend to safe borrowers implying a low interest rate spread. The positive and significant sign of the corresponding Gini coefficient indicates that as banks differ in their capital ratio they soften competition. This latter effect may be due both to capital ratio being a quality variable along which banks can differentiate themselves, and to banks’ risk aversion being a parameter along which they differentiate themselves provided more risk averse banks have higher capital ratio. The negative sign of $\ln v(cap88)_{i,r,t-1}$ combined with a positive sign for the Gini coefficient implies that we cannot rule out a high capital ratio as a quality variable by itself that borrowers are willing to pay for. However, if there is such an effect it is too weak not to be dominated by the high bank risk aversion and high capital ratio effect.

Among the control variables note that the elasticity of the $costrat_{i,t}$ is positive and significant. Thus banks operating under imperfect competition in the market for credit line loans are able to pass some of their operating costs over to these borrowers. The estimated elasticity implies that a bank at sample mean with NOK mill. 540 in operating costs and credit line loans of NOK mill. 2160 facing a ten pct. increase in its operating costs would be able to pass approx. NOK mill. 1.3 of the cost increase of NOK mill. 54 on to its credit line borrowers.

As the Herfindahl index does not get a significant coefficient we can neither give support to the traditional view of more concentrated credit markets leading to higher interest rates nor to the theories of winner’s curse.

All in all our empirical results give support to the hypothesis that there are certain characteristics of banks that borrowers evaluate and hence are willing to pay for. If banks differentiate themselves along these characteristics, banks are able to soften competition and make borrowers pay them a higher interest rate.

5. Concluding remarks

In this paper we have studied strategies pursued by banks in order to differentiate their services from those of their rivals. In that way the level of competition in the industry can be reduced. More specifically we have analyzed whether the bank size, a bank's ability at avoiding losses, or its capital ratio can be used as strategic variables to make banks different and increase the interest rates banks can charge their borrowers in equilibrium. Using a panel of data covering Norwegian banks between 1993 and 1998 we found empirical support for the ability at avoiding losses measured by the ratio of loss provisions as such a variable. In fact borrowers in the market for credit line loans may discipline banks into avoiding losses. We also found evidence that banks pass on parts of increases in their operating costs to credit line borrowers. However, we did not find any strong evidence for the use of high capital ratio as a strategic variable that borrowers are willing to pay for. This implies that strategic competition does not lead banks to hold more capital than their cost minimizing level. This level is normally below the socially optimal level due to explicit and implicit deposit insurance. Only the more risk averse among banks would hold capital well in excess of their cost minimizing level. Thus, our paper gives support to the rationale for imposing capital requirements on all banks.

References

- Berger, A. N., and G. F. Udell (1995): “Relationship lending and lines of credit in small firm finance,” *Journal of Business*, 68, 351–381.
- Boot, A. W. (2000): “Relationship banking: What do we know?,” *Journal of Financial Intermediation*, 9, 7–25.
- Broecker, T. (1990): “Credit-worthiness tests and interbank competition,” *Econometrica*, 58, 429–452.
- Bulow, J., and P. Klemperer (1999): “Prices and winner’s curse,” Working paper, <http://www.nuff.ox.ac.uk/users/klemperer/papers.html>.
- Cantillo, M., and J. Wright (2000): “How Do Firms Choose Their Lenders? An Empirical Investigation,” *The Review of Financial Studies*, 13, 155–189.
- Cerasi, V., and S. Daltung (????): “The optimal size of a bank: costs and benefits of diversification,” Discussion paper 231, LSE, Financial Markets Group.
- (1994): “The optimal size of a bank: costs and benefits of diversification,” FMG discussion paper, London School of Economics.
- Chemmanur, T. J., and P. Fulghieri (1994): “Reputation, renegotiation and the choice between bank loans and publicly traded debt,” *Review of Financial Studies*, 7, 475–506.
- Degryse, H., and P. van Cayseele (2000): “Relationship lending within a bank based system: evidence from European small business data,” *Journal of Financial Intermediation*, 9, 90–109.
- Diamond, D. (1984): “Financial Intermediation and Delegated Monitoring,” *Review of Economic Studies*, 51, 393–414.
- Greene, W. H. (1993): *Econometric Analysis*. Macmillan, New York.
- James, C. (1987): “Some evidence on the uniqueness of bank loans,” *Journal of Financial Economics*, 19, 217–235.
- Kim, M., D. Klinger, and B. Vale (1999): “Estimating switching costs and oligopolistic behavior,” Arbeidsnotat 1999/4, Norges Bank, Oslo.
- Mester, L. J. (1992): “Traditional and nontraditional banking: An information-theoretic approach,” *Journal of Banking and Finance*, 16, 545–566.

- Ongena, S., and D. C. Smith (1998): “Quality and duration of banking relationships,” in *Global Cash Management in Europe*, ed. by D. Birks. MacMillan Press, London.
- Peek, J., and E. S. Rosengren (1997): “The international transmission of financial shocks: the case of Japan,” *American Economic Review*, 87, 495–505.
- Qian, Y. (1994): “Incentives and loss of control in an optimal hierarchy,” *Review of Economic Studies*, 61, 527–544.
- Shaffer, S. (1998): “The winner’s curse in banking,” *Journal of Financial Intermediation*, 7, 359–392.
- Sharpe, S. A. (1990): “Asymmetric information, bank lending, and implicit contracts: A stylized model of customer relationships,” *Journal of Finance*, 45, 1069–1087.
- Slovin, M. B., M. E. Sushka, and J. A. Polonchek (1993): “The Value of bank durability: Borrowers as Bank Stakeholders,” *Journal of Finance*, 48, 247–266.
- Williamson, O. (1967): “Hierarchical control and optimal firm size,” *Journal of Political Economy*, 75, 123–138.