

Raknerud, Arvid; Vatne, Bjørn Helge; Rakkestad, Ketil

Working Paper

How do banks' funding costs affect interest margins?

Discussion Papers, No. 665

Provided in Cooperation with:

Research Department, Statistics Norway, Oslo

Suggested Citation: Raknerud, Arvid; Vatne, Bjørn Helge; Rakkestad, Ketil (2011) : How do banks' funding costs affect interest margins?, Discussion Papers, No. 665, Statistics Norway, Research Department, Oslo

This Version is available at:

<https://hdl.handle.net/10419/192647>

Standard-Nutzungsbedingungen:

Die Dokumente auf EconStor dürfen zu eigenen wissenschaftlichen Zwecken und zum Privatgebrauch gespeichert und kopiert werden.

Sie dürfen die Dokumente nicht für öffentliche oder kommerzielle Zwecke vervielfältigen, öffentlich ausstellen, öffentlich zugänglich machen, vertreiben oder anderweitig nutzen.

Sofern die Verfasser die Dokumente unter Open-Content-Lizenzen (insbesondere CC-Lizenzen) zur Verfügung gestellt haben sollten, gelten abweichend von diesen Nutzungsbedingungen die in der dort genannten Lizenz gewährten Nutzungsrechte.

Terms of use:

Documents in EconStor may be saved and copied for your personal and scholarly purposes.

You are not to copy documents for public or commercial purposes, to exhibit the documents publicly, to make them publicly available on the internet, or to distribute or otherwise use the documents in public.

If the documents have been made available under an Open Content Licence (especially Creative Commons Licences), you may exercise further usage rights as specified in the indicated licence.

*Arvid Raknerud, Bjørn Helge Vatne and
Ketil Rakkestad*

How do banks' funding costs affect interest margins?

Abstract:

We use a dynamic factor model and a detailed panel data set with quarterly accounts data on all Norwegian banks to study the effects of banks' funding costs on their retail rates. Banks' funds are categorized into two groups: customer deposits and long-term wholesale funding (market funding from private and institutional investors including other banks). The cost of market funding is represented in the model by the three-month Norwegian Inter Bank Offered Rate (NIBOR) and the spread of unsecured senior bonds issued by Norwegian banks. Our estimates show clear evidence of incomplete pass-through: a unit increase in NIBOR leads to an approximately 0.8 increase in bank rates. On the other hand, the difference between banks' loan and deposit rates is independent of NIBOR. Our findings are consistent with the view that banks face a downward-sloping demand curve for loans and an upward-sloping supply curve for customer deposits.

Keywords: interest rates, NIBOR, pass-through, funding costs, bank panel data, dynamic factor model

JEL classification: C13, C22, C51, G10

Acknowledgements: The opinions expressed here are solely those of the authors and do not necessarily reflect the views of Statistics Norway and Norges Bank. We thank Farooq Akram, Sigbjørn Atle Berg, Terje Skjerpen and Bjørne-Dyre Syversten for many useful comments and suggestions.

Address: Arvid Raknerud, Statistics Norway, Research Department. E-mail: rak@ssb.no

Bjørn Helge Vatne, Norges Bank, Division of Financial stability.
E-mail: Bjorn-Helge.Vatne@Norges-Bank.no

Ketil Rakkestad, Norges Bank, Division of Financial stability.
E-mail: Ketil-Johan.Rakkestad@Norges-Bank.no

Discussion Papers

comprise research papers intended for international journals or books. A preprint of a Discussion Paper may be longer and more elaborate than a standard journal article, as it may include intermediate calculations and background material etc.

© Statistics Norway

Abstracts with downloadable Discussion Papers
in PDF are available on the Internet:

<http://www.ssb.no>

<http://ideas.repec.org/s/ssb/dispap.html>

For printed Discussion Papers contact:

Statistics Norway

Telephone: +47 62 88 55 00

E-mail: Salg-abonnement@ssb.no

ISSN 0809-733X

Print: Statistics Norway

Sammendrag

Vi bruker paneldata for banker for 2001Q2-2010Q3 basert på kvartalsvise resultatregnskap og balanse for å studere hvordan bankenes finansieringskostnader påvirker deres innskudds- og utlånsrenter. Vi estimerer en økonometrisk modell på mikrodata aggregert opp til syv bankgrupper, definert slik at bankene i samme gruppe har felles OMF-foretak. Et system av ligninger for husholdningers innskuddsrente og to utlånsrenter (husholdninger og ikke-finansielle foretak) blir estimert simultant for tre typer renter og syv bankgrupper. Forklaringsvariablene i modellen er 3-månders interbank rente (NIBOR), volatiliteten i denne (definert som standardavviket i dagsobservasjonene i NIBOR, beregnet kvartalsvis) og en indikativ spread på usikrede 3-års bankobligasjoner. Våre estimater viser klar evidens for ikke-komplett pass-through og vi finner liten grad av heterogenitet i hvordan bankene responderer på endringer i forklaringsvariablene når det gjelder innskuddsrenter og utlånsrenter til husholdninger både på kort og lang sikt. Det er imidlertid betydelig større heterogenitet mellom bankene når det gjelder lån til foretak. I en estimert likevektsrelasjon finner vi at 10 basispunkter økning i 3-månders NIBOR leder til 8 basispunkter økning i både innskudds- og utlånsrenter. Dette er et gjennomsnitt over syv bankgrupper, men parametrene er svært like på tvers av alle de syv bankgruppene. Videre finner vi at kredittpåslaget (indikativ spread) har hatt stor betydning for utlånsmarginene og totalmarginene under og etter finanskrisen. I vår modell ser vi på de indikative spreadene på 3-års usikrede bankobligasjoner rapportert av DnB NOR Markets. Mens spreaden var nesten konstant frem til 2007Q4, økte den kraftig under finanskrisen og bidro til økte utlånsmarginer relativt til 3-månders NIBOR.

1 Introduction

In this paper we investigate the degree of pass-through from banks' funding costs to their deposit and loan rates. We do so by estimating an econometric model utilizing quarterly microdata on individual Norwegian banks from 2001Q2 to 2010Q3. Our focus is on the transmission mechanism from changes in the level and volatility of market interest rates to *retail rates* (in this paper defined as all interest rates on deposits and loans set by the bank). Traditionally this issue has been examined either by employing time series econometrics on aggregate bank interest rate data (e.g., Chong, 2010; Hofmann and Mizen, 2004) or panel data methods on individual banks' interest rates (e.g., Hannan and Berger, 1991; De Graeve *et al.*, 2007). In contrast, we consider the retail rates of various banks, when aggregated into seven groups, as jointly dependent variables. Each bank group is defined so that the banks in the same group have a common covered bond mortgage (OMF) company.

There is a related empirical literature on interest rate margins using banks' net interest income relative to total assets as the main dependent variable, either at the aggregate level (for a representative bank) (e.g., Saunders and Schumacher, 2000; Andersen *et al.*, 2008) or, more rarely, at the bank level (e.g., Maudos and Guevara, 2004). These approaches suffer from the weakness that the dependent variable is a mixture of price (interest rate) and volume effects. It is not straightforward to infer anything from these studies about how banks' interest rates and interest margins respond to changes in exogenous variables, because different effects are entangled.

We employ a detailed panel data set with quarterly accounts data on all Norwegian banks from 2001Q2 until 2010Q3. This period is particularly interesting because it is characterized by increased competition between banks, as well as pro-

ductivity growth due to wider use of Internet-based payment services. In the data, volumes and interest rates over a quarter are specified for various types of deposits and loans, according to sector (such as households or nonfinancial firms) and type of loan (mortgage, other loans). Microdata allow us to study heterogeneity between banks, for example, whether the interest margins of different banks react differently to exogenous shocks. Moreover, we are able to analyze differences in interest margins between loans to businesses and households, and in the speed of adjustment of banks' interest rates to changes in exogenous variables. The existence of bank-specific dynamics in retail rates implies that estimates of long-run coefficients from aggregate data will be biased, even if the primary interest is in the parameters of the (long-run) relation between retail rates and funding costs of an average (or representative) bank (cf. Pesaran and Smith, 1995). This paper addresses this problem by estimating a flexible model with heterogeneous, bank group-specific equations. By aggregating the estimated bank-group specific equations, the corresponding empirical relation for a representative bank can be obtained.

More specifically, we focus on (i) loans to households, (ii) loans to corporations in the nonfinancial sector and (iii) households' bank deposits. The corresponding interest rates are collected from all banks (or bank groups), placed in one system of equations and analyzed within the framework of dynamic factor modeling. The comovements among various bank groups' retail rates are then captured by a small number of common factors. As a result, we are able to separate the effect on retail rates of common observed variables (such as interbank market rates) from the effect of unobserved common variables (reflecting, for example, changes in bank regulations, competition and productivity). In accordance with most empirical literature on interest margins (e.g., Saunders and Schumacher, 2000), our model includes an interbank market rate; the three-month Norwegian Inter Bank Offered Rate (NI-

BOR), as a main explanatory variable.

Our empirical framework allows us to test particular hypotheses about both the short run and the long run (“steady state”) relation between changes in market rates on banks’ retail rates. In particular, we investigate three types of interest margins at the bank group level: the difference between (i) the household loan rate and the deposit rate, (ii) the corporate loan rate and the deposit rate and (iii) the average net interest margin: the ratio of total funding costs to total loans. The latter is a measure of banks’ earnings per unit in total outstanding loan, not the difference between two interest rates. In a competitive bank market, a permanent change in the marginal cost of wholesale funding should be passed fully over to loan and deposit rates (see Hannan and Berger, 1991). However, if banks have market power, they are faced with a trade-off between conflicting goals: a high (low) interest rate on loans (deposits) on the one hand and a high volume of loans and deposits on the other. The spread between the price of market funding and retail rates may therefore change as a result of a change in the former; that is, there may be incomplete pass-through in both the long and the short run. The completeness hypothesis will be formally tested.

The remainder of the paper is organized as follows. Section 2 discusses the main concepts and describes the empirical model, Section 3 presents the data, Section 4 presents the results and Section 5 concludes.

2 The modeling framework

Funding costs Banks need to raise funds to provide loans. We organize these funds into two categories: customer deposits and wholesale funding (market funding from private and institutional investors, including other banks). According to lenders, the marginal funding source for loans to households and businesses is whole-

sale funding. Banks may need to raise a large amount of funding over a short period. This cannot be done through raising retail deposits by increasing the rates on deposits, because bank customers (households and firms) typically do not react quickly to changes in interest rates. Wholesale funding is typically bonds, which to varying degrees match the expected maturity of the loans provided. There may be considerable heterogeneity between banks with regard to the extent and composition of market funding (e.g., regarding the maturity of debt). In this paper we interpret the marginal funding cost as that of raising senior unsecured bonds. An unsecured bond may be issued with a fixed or variable interest rate. In the first case, a Norwegian bank typically enters into an interest rate swap to achieve a level of variable rate exposure that matches the variable rate loans. The banks' costs may be expressed by two components: the variable rate cash flows paid in the interest rate swap (normally three-month NIBOR) and the fixed cash flow due to the issuer-specific credit spread over the swap rate¹. In this paper we include both the three-month NIBOR and the spread of unsecured senior bonds issued by Norwegian banks as measures of the cost of market funding. The spread of unsecured senior bonds is represented by the indicative spreads on three-year bonds as reported by DnB NOR Markets.

The net interest margin is the difference between the interest that a lender receives on all loans and the interest it pays on all funding of those loans divided by total loans. In our analysis we distinguish between two categories of loans: those to businesses (in the nonfinancial sector) and those to households. If L_H , L_B and D denote loans to households, loans to businesses and bank deposits, respectively, with corresponding interest rates r^H , r^B and r^D , and r denotes the (average) market

¹See https://www2.sparebank1.no/portal/1001/3_privat?_nfpb=true&_pageLabel=page_privat_innhold&aId=1201861729341 for examples of bank bonds with varying maturity and with interest payments equal to the three-month NIBOR plus a fixed credit spread.

interest rate, the average net interest margin, π , can be expressed as

$$\pi = \frac{(r^H - r)L_H + (r^B - r)L_B + (r - r^D)D}{L_H + L_B}. \quad (1)$$

π thus involves three interest spreads relative to the NIBOR rate: $r^H - r$, $r^B - r$ and $r - r^D$. It is obviously important for banks' profitability how the market rate (r) is passed through to the retail rates (r^H , r^B and r^D). Assuming that the residual outstanding loan $L_H + L_B - D$ is financed by variable rate market funding at the rate r , π will be a measure of the *average* profitability per NOK in outstanding loans. A complementary issue, deferred for later study, is how interest rate changes affect demand for loans and supply of deposits.

Econometric model As mentioned above, our analysis distinguishes between two categories of loans: loans to businesses and households (the personal market). The corresponding loan rates for bank group i ($i = 1, \dots, 7$) at time t ($t = 1, \dots, T$) are denoted by r_{it}^B and r_{it}^H , respectively, where t refers to the end of a particular quarter in the given year. The interest rate on bank deposits is denoted by r_{it}^D . At this level of aggregation, r_{it}^D , r_{it}^H and r_{it}^B can be calculated as weighted averages of more disaggregate interest rates, where the weights are available from the outgoing balance in the bank accounts data (see Section 3).

Our explanatory variables represent the exogenous funding costs of banks and the associated risk. The main variable is the three-month NIBOR rate, r_t , which is a key determinant of external funding costs, as explained above. For the individual banks, it is reasonable to assume that r_t is exogenous; that is, the individual bank cannot influence NIBOR through its supply of and demand for credit in the interbank market. The rationale behind this assumption is that (major) banks can borrow and lend NOK through the foreign exchange rate markets such as the NOK–USD exchange swap market. Covered interest rate parity implies that the NIBOR rate is

determined by international lending and swap exchange rates, which are exogenous to individual Norwegian banks.²

We also include the volatility of r_t , σ_t , as an explanatory variable. This variable is a proxy for interest rate risk, as described by Ho and Saunders (1981). In the Ho and Saunders model, banks finance the difference between the demand for loans and the supply of deposits by wholesale funding. If banks are risk averse, the interest margin between the loan rate and the deposit rate will be increasing in the volatility of the market rate.³ As described in detail in Section 3, in our empirical implementation, r_t and *squared* volatility, σ_t^2 , are weighted averages of daily interest rates and daily squared interest rate *deviations* (from the mean), calculated each quarter, with geometrically decaying weights.

Our econometric model specifies a stochastic relation between the retail rates $(r_{it}^D, r_{it}^H, r_{it}^B)$ and the exogenous variables (r_t, σ_t, s_t) for each bank group. It accommodates the following important features:

- asymmetries in the relation between the retail rates and r_t , depending on the sign of $\Delta r_t = r_t - r_{t-1}$;
- flexible short-term dynamics, where different r_{it}^X , $X = D, H, B$, are allowed to react differently to exogenous shocks;
- bank group-specific parameters; stochastic shocks that are common across different bank groups (i) and type of interest rate (X);
- stochastic shocks specific to a particular bank group and interest rate.

²For an example, see equation (1) in Akram and Christophersen (2011): http://www.norges-bank.no/upload/publikasjoner/staff%20memo/2011/staff_memo_0111.pdf

³This model has been developed further, for example by Allen (1988), to incorporate different types of deposits and loans, and by Angbazo (1997), to allow both credit and interest rate risk. Empirical models of interest rate volatility and implications for interest rate risk are examined by Chan *et al.* (1992).

Conditional on the common explanatory variables, we model the individual retail rates as univariate autoregressive processes, augmented with common dynamic factors to account for joint dependencies. The use of common dynamic factors is a parsimonious way of capturing the comovements among variables. In contrast, the number of parameters in a VAR models increases exponentially with the number of equations. Examples of dynamic factors are the so-called diffusion index models (see Forni *et al.*, 2000, and Stock and Watson, 2002) and the factor-augmented VAR model, FAVAR (see Bernanke *et al.*, 2005). Our approach has more in common with the tradition of multivariate structural time series models than with the approximate dynamic factor models most commonly favored in the literature.⁴

In our most general specification we assume that, for $X = D, H, B$; $i = 1, \dots, 7$ (seven bank groups); and $t = 1, \dots, T$:

$$r_{it}^X = \mu_i^X + \alpha_{i,0}^X r_t + \alpha_{i,1}^X r_{t-1} + \alpha_0^X \max(\Delta r_t, 0) + \alpha_1^X \max(\Delta r_{t-1}, 0) + \beta_{i,0}^X \sigma_t + \beta_{i,1}^X \sigma_{t-1} + \gamma_0^X s_t + \sum_{j=1}^{p_i} \phi_{ij}^X r_{i,t-j}^X + \sum_{k=1}^m \theta_{ik}^X f_{kt} + e_{it}^X, \quad (2)$$

where μ_i^X is a bank group- and interest rate-specific fixed effect, the α -parameters capture the effects of the NIBOR rate by allowing both the current NIBOR rate, r_t (through $\alpha_{i,0}^X$), and the lagged NIBOR rate, r_{t-1} (through $\alpha_{i,1}^X$), to affect the current interest rate on loans ($X = H, B$) and deposits ($X = D$). One lag is allowed in order to capture the effect of notification rules that restrict the speed at which banks are allowed to increase their loan rates. Moreover, asymmetries in the effects of positive and negative changes are captured by the term $\alpha_0^X \max(\Delta r_t, 0)$ and the lagged term, $\alpha_1^X \max(\Delta r_{t-1}, 0)$. For example, if α_0^D is negative, the bank deposit rate r_{it}^D is changed more slowly as a result of a given positive change ($\Delta r_t > 0$) than for the corresponding negative change. The effects of changes in the volatility, σ_t , of the NIBOR rate are determined by the bank group-specific β -parameters, whereas

⁴See Harvey (1989) for a general exposition of structural time series models.

the credit spread measure, s_t , is assumed to affect each bank group through common parameters γ_0^X ($X = D, H, B$). The latter restriction is imposed in view of the very limited variation in s_t before 2008 (see Section 3).

The autoregressive parameters ϕ_{ij}^X , $j = 1, \dots, p_i$, determine how the effects of a shock in any of the exogenous variables evolve over time. The number of lags, p_i , is allowed to differ from bank group to bank group. Using the Akaike information criterion, we find that $p_i = 2$ is adequate in most of the equations. Finally, the unobserved stochastic terms consist of m dynamic factors, f_{kt} , $k = 1, \dots, m$, which pick up the dependencies across banks due to common, unobserved variables (e.g., effects of the business cycle, credit market regulations and competition) and the idiosyncratic error term, e_{it}^X , that is, independent across banks (i) and over time (t). The vector $(e_{it}^D, e_{it}^H, e_{it}^B)$ is assumed to have a trivariate normal distribution, with covariance matrix Σ , whereas the dynamic factors, f_{kt} , are assumed to be independent, Gaussian AR(1) processes:

$$f_{kt} = \psi_k f_{k,t-1} + \eta_{kt}, \eta_{kt} \sim \mathcal{IN}(0, 1); k = 1, \dots, m. \quad (3)$$

Thus, f_{1t}, \dots, f_{mt} are latent stochastic processes that capture the comovements between the interest rates of different banks not accounted for by the observed explanatory variables. The impact of the dynamic factors on the individual banks is determined by bank group-specific impact coefficients, θ_{ik}^X . In our model the factors play a similar role to that of the “risk factor contributions” in Rosen and Saunders (2010), in the context of portfolio risk analysis. Our model is estimated by employing a version of the maximum likelihood algorithm described in Raknerud *et al.* (2010).

Partial effects Our econometric framework allows us to disentangle partial effects of changes in exogenous variables. In particular, we are interested in the effects of

changes in market rates. Assume now that the system is in a steady state at t defined by $r_{t-j} = r^*$, $\sigma_{t-j} = \sigma$, $s_{t-j} = s$, $f_{k,t-j} = f_k$ (i.e. arbitrary fixed values) and $e_{i,t-j}^X = 0$ for all i and $j > 0$. Let $\Delta r_{i,t+j}^X(\delta)$ denote the *causal* effect $\Delta r_{i,t+j}^X = r_{i,t+j}^X - r_{i,t+j-1}^X$ due to a permanent change in r_t from $r_{t-1} = r^*$ to $r_{t+j} = r^* + \delta$ for $j \geq 0$, Then

$$\begin{aligned} \Delta r_{it}^X(\delta) &= \alpha_{i,0}^X \delta + \alpha_0^X \max(\delta, 0) \\ \Delta r_{i,t+1}^X(\delta) &= \alpha_{i,1}^X \delta + \alpha_1^X \max(\delta, 0) - \alpha_0^X \max(\delta, 0) + \phi_{i1}^X \Delta r_{it}^X(\delta) \\ \Delta r_{i,t+2}^X(\delta) &= \phi_{i1}^X \Delta r_{i,t+1}^X(\delta) + \phi_{i2}^X \Delta r_{it}^X(\delta) - \alpha_1^X \max(\delta, 0) \\ &\vdots \\ \Delta r_{i,t+k}^X(\delta) &= \phi_{i1}^X \Delta r_{i,t+k-1}^X(\delta) + \dots + \phi_{ip_i}^X \Delta r_{i,t+k-p_i}^X(\delta) \quad (k \geq p_i). \end{aligned}$$

The effect of a permanent change, initiated at time t , on $r_{i,t+h}^X$ is then given by the cumulative sum $\sum_{j=0}^h \Delta r_{i,t+j}^X(\delta)$. Moreover, in an equilibrium path where all *observed* exogenous variables are assumed to be have permanent values, $r_t = r$, $\sigma_t = \sigma$, $s_t = s$, we have

$$r_{it}^X = \frac{\mu_i^X}{1 - \sum_{j=1}^{p_i} \phi_{ij}^X} + \left(\frac{\alpha_{i,0}^X + \alpha_{i,1}^X}{1 - \sum_{j=1}^{p_i} \phi_{ij}^X} \right) r + \left(\frac{\beta_{i,0}^X + \beta_{i,1}^X}{1 - \sum_{j=1}^{p_i} \phi_{ij}^X} \right) \sigma + \left(\frac{\gamma_0^X}{1 - \sum_{j=1}^{p_i} \phi_{ij}^X} \right) s + d_t + \varepsilon_{it}^X, \quad (4)$$

where d_t captures the effects of the present and lagged dynamic factors, f_{js} , $s \leq t$, and ε_{it}^X is a moving average of the error terms e_{is}^X , $s \leq t$. Equations (4) therefore determines the long-term relation between the retail rates and *permanent* levels of the exogenous variables. We see that if the bank-group specific parameters differ across bank-groups, the parameters of the weighted average $\sum_{i=1}^7 w_i r_{it}^X$, where w_i is the share of total assets (see Table 1), generally differ from parameters obtained by aggregating the variables first and then estimating the aggregate version of equation (2). This fact motivates an analysis of a more disaggregate level even if the main interest should be on aggregate results for the whole banking sector or for a representative bank.

3 Data

The sample consists of quarterly accounts data on all Norwegian banks from 2001Q2 until 2010Q3 and is based on the accounts statistics for financial corporations assembled by Statistics Norway.⁵ Bank-level data are aggregated into seven bank groups, as listed in Table 1. The grouping is done so that all banks in each group have a common covered bond mortgage (OMF) company. Covered bonds (OMFs) were introduced in Norway in June 2007 and have become an important source of funding for Norwegian financial services groups and banking alliances.⁶ Key statistics for the seven bank groups are given in Table 1.

Table 1: **Descriptive statistics for seven bank groups.**

	Percentage of market				Percentage of bank loans	
	Total assets	Loans to households	Loans to businesses	Deposits	Households	Firms
DnB NOR	42	33	32	35	64	35
Subsidiaries of foreign banks	14	13	17	12	60	40
Branches of foreign banks	13	11	17	10	54	46
SpareBank1-alliansen	15	20	17	19	68	32
Terra Gruppen	4	7	4	6	77	23
Other savings banks	10	14	11	13	70	30
Other commercial banks	3	3	3	4	69	31

Since 2001Q2, Norwegian banks have been obliged to report interest rates at the end of each quarter. We calculate the average interest rate of the banks in a group as a value-weighted average of the reported interest rates. From the bank statistics we get interest rates and volumes of various loans in each bank. The interest rates are weighted by the corresponding nominal book values to obtain a value-weighted average interest rate. The three-month effective Norwegian Inter Bank Offered Rate (NIBOR) reported by Norges Bank is a proxy for the cost of long- and medium-term market financing. Illustrations of these interest rates are provided in Figure 1. The

⁵See <http://www.ssb.no/skjema/finmark/rapport/orbof/> (in Norwegian).

⁶See the following article by Rakkestad and Dahl in Penger og Kredit 1/2010 (in Norwegian): http://www.norges-bank.no/Upload/80111/OMF_marked_i_vekst_PK_1_10_nov.pdf

graph labeled “NIBOR +/- sigma” shows the range of *daily* NIBOR rates that lie within one standard deviation of the mean within the corresponding quarter.

In our econometric model, the quarterly NIBOR rate, r_t , and the corresponding volatility, σ_t , are operationalized as follows: Let $r_{t,j}$, $j = 1, \dots, M_t$, denote the NIBOR rate of day j in quarter t , where M_t is the number of trading days in quarter t . Then

$$r_t = \frac{1}{k_t} \sum_{j=0}^{M_t-1} \lambda^j r_{t,M_t-j},$$

with $k_t = \sum_{j=0}^{M_t-1} \lambda^j$. To measure squared volatility, σ_t^2 , we calculate the weighted mean of the *squared deviations* $(r_{t,j} - r_t)^2$:

$$\sigma_t^2 = \frac{1}{\tilde{k}_t} \sum_{j=0}^{M_t-1} \xi^j (r_{t,M_t-j} - r_t)^2,$$

where $\tilde{k}_t = \sum_{j=0}^{M_t-1} \xi^j$. In our application we use $\lambda = 0.9$ and $\xi = 0.5$, which means that the weight attached to the first observation in the quarter relative to the latest observation is about 10 percent when calculating r_t , whereas only the latest 4–5 observations have nonnegligible weight when σ_t^2 is calculated. These parameter values approximately maximize the in-sample fit of the model when a grid search is conducted over possible λ - and ξ - values.

As explained above, the cost of market funding depends on the credit spread over the swap rate, denoted s_t . We use an index consisting of indicative bid spreads based on average trading levels over the swap rate (three-year fixed/three-month NIBOR) for senior bonds issued by a range of Norwegian banks since 2001. The series includes DnB NOR Bank, Nordea Bank Norge and a representative selection of banks of various sizes and ratings. It is clear from Figure 2 that until 2008 the variation in funding cost was dominated by variations in NIBOR. However, from 2008Q1 to 2008Q4, the spread increased dramatically, and was still much higher than its pre-2008 level at the end of 2010. The (issuer-specific) spread may consist of compensation for types of risk other than credit risk. During the financial crisis, a

substantial part of the spread was compensation for liquidity risk; that is, it occurred largely because of reduced liquidity in funding markets.⁷ The combined effect on banks' funding costs is difficult to assess. This is the motivation for our choice to include the credit spread as a separate variable in the econometric model, rather than adding it to the NIBOR rate to obtain an estimate of total long-term funding costs.

Data for issuance indicate that the banks reduced their ordinary funding activity dramatically during the period 2008Q1 to 2008Q4, when the the credit spread soared. At the same time, several authority measures to support banks' funding took effect. In particular, a marked reduction in the policy rate led to a sharp fall in the NIBOR rate. The combined effect is that from 2008Q2 we observe a marked fall in deposit margins and an (offsetting) increase in the margins of loans to households (relative to NIBOR). A possible explanation is that when the policy rate becomes very low, banks' opportunity to lower their deposit rates is limited, and the deposit margin falls. To compensate for the reduced margins on deposits, the banks may increase their margins on loans. Moreover, banks have a limited ability to quickly adjust the rates on loans because of notification rules, which may contribute to temporary high margins on loans during periods of falling policy rates.

There is considerable heterogeneity in the funding sources of banks. Small national banks tend to have more deposits than foreign or large national banks, while the latter banks rely more on market funding. For example, Terra Gruppen, which is a group of small banks, had the highest average ratio of household deposits over total loans during 2001-2010: 42 percent. The two foreign bank groups had the lowest ratio—18 percent—while Norway's largest bank, DnB NOR, had a ratio of 29 percent. Figure 3 shows the difference in average deposit interest rates between a

⁷See Chapter 3 (especially Figure 3.16) in Bank of England's Financial Stability Report, Issue 27, June 2010: <http://www.bankofengland.co.uk/publications/fsr/2010/fsrfull1006.pdf>

group of small banks and a group of large banks. While the figure reveals considerable short-term fluctuations, there appears to be no systematic long-term difference between the deposit rates of these two bank groups.

Examining the stationarity of r_t To perform statistical tests and assess estimation uncertainty, it is important to know whether the NIBOR rate is a unit root process or not, because this affects the asymptotic distribution of the maximum likelihood estimator of the empirical model (2). Taking as a starting point the assumption that r_t is *not* a unit root process, we can test this hypothesis using the test proposed by Choi (1994) in combination with Andrews' (1991) automatic lag truncation procedure, as recommended by Choi and Ahn (1999). We conducted the test both on daily data (yielding 2724 observations) and quarterly data. In neither case did we reject the null hypothesis of stationarity. For example, based on daily data the test statistic becomes 0.21, which is far from significant (P-value=0.45). This result is consistent with Choi and Ahn (1999), who do *not* reject that the real interest rate is stationary, using monthly data for several countries for 1980–1991 (Norway not included). Supporting evidence is provided by Anundsen and Jansen (2011), who reject the null hypothesis that the real interest rate is integrated of order one against the alternative of stationarity, using an Augmented Dickey-Fuller test on quarterly NOK interest rate data for 1986–2008. Although we use nominal interest rates in our analysis, our data come from a period with inflation targeting and a low and stable inflation rate. Note that stationarity of r_t does not necessarily imply stationarity of the retail rates, r_{it}^X , because the common dynamic factors, f_{kt} , could be nonstationary. Thus there may be a (stochastic) trend in the spread between retail rates and the NIBOR rate and/or in the loan and deposit interest margins, for example due to changes in competition or regulatory measures. This

flexibility of our model is empirically important, because Figure 4 reveals a distinct decreasing trend for the interest margins between loans and deposits over the sample period.

4 Results

The results presented below are based on a version of the model (2) with $\alpha_0^X = \alpha_1^X = 0$ for $X = H, B$. These zero restrictions were imposed because the estimates of α_0^X and α_1^X were almost zero and clearly insignificant for both types of loans. On the other hand, we obtain a significant negative estimate of α_0^D equal to -0.22 (std. error = 0.06) and a negative, but insignificant, estimate of α_1^D equal to -0.08 (std. error = 0.06). Thus the deposit rate falls more quickly than it rises, but the loan rate does not.

Equations (5) are estimates of the long-run (steady-state) equations (4) for a representative bank. These estimates are obtained from the corresponding bank-group specific equations, using the estimates shown in Table 2. Standard errors in parentheses are obtained by the delta method. We see that for the representative bank, the coefficient of r in the steady state is close to 0.8 , and is significantly below one for all three retail rates. Thus the hypothesis of complete pass-through in the long run is clearly rejected. If we examine the bank group-specific estimates in Table 2, they are remarkably close to 0.8 across bank groups and interest rates. A formal test of whether all the steady-state coefficients of r are equal across all bank groups is provided by the Wald tests reported in the last row of Table 2. Evidently, we cannot reject the hypothesis of homogeneous long-run parameters. The corresponding equations for the differences between the household loan rate and the deposit rate, and between the business loan rate and the deposit rate, are shown in (6). We see that in the steady state, these two interest margins do not

depend on r , because both estimates are almost zero. Note that the marked fall in deposit margins and increase in the margins of loans to households observed from 2008Q2-2008Q4 (cf. Figure (2) and the discussion in Section 3) are consistent with our estimated model: a fall in the NIBOR rate will lead to an increase in the loan margin which is offset by a decrease in the deposit margin.

With regard to the intercept of the steady-state equation, Table 2 shows that the bank group-specific parameters vary considerably across bank groups, but also that the estimation uncertainty is considerably larger than for the steady-state coefficients of r . The results from the Wald test show that we cannot reject the hypothesis that banks have the same steady-state intercept with regard to deposits, although we do reject it with regard to both types of loans.

Turning to the credit spread measure, the results in (5) show that permanent changes in s_t have no significant long-run effect on the deposit rate. On the other hand, an increase in the credit spread induces a significant positive pressure on the loan rates, especially loans to businesses. Recall that the underlying (short-run) parameters γ_0^X are common across banks, so we only report common long-run effects (i.e. a weighted average of the coefficient of s in (4)). The estimated effects on the margins between the loan rates and the deposit rate in (6) indicate that a permanent unit increase in the credit spread leads to a long-term increase in these interest rate margins from roughly 0.3 to 0.4. It should be noted, however, that the estimated effects are identified mainly by events immediately before and after the onset of the financial crisis in 2008Q3 and must be interpreted with care, as discussed in Section 3.

$$\begin{aligned}
\sum_{i=1}^7 w_i r_{it}^D &= d_t + \underset{(0.05)}{0.77}r + \underset{(0.24)}{0.58}\sigma - \underset{(0.08)}{0.08}s + \text{residual} \\
\sum_{i=1}^7 w_i r_{it}^H &= d_t + \underset{(0.05)}{0.80}r + \underset{(0.27)}{0.74}\sigma + \underset{(0.12)}{0.22}s + \text{residual} \\
\sum_{i=1}^7 w_i r_{it}^B &= d_t + \underset{(0.07)}{0.84}r + \underset{(0.33)}{0.70}\sigma + \underset{(0.14)}{0.32}s + \text{residual} \tag{5}
\end{aligned}$$

$$\begin{aligned}
\sum_{i=1}^7 w_i (r_{it}^H - r_{it}^D) &= d_t + \underset{(0.03)}{0.03}r + \underset{(0.15)}{0.16}\sigma + \underset{(0.07)}{0.30}s + \text{residual} \\
\sum_{i=1}^7 w_i (r_{it}^B - r_{it}^D) &= d_t + \underset{(0.4)}{0.07}r + \underset{(0.24)}{0.12}\sigma + \underset{(0.09)}{0.40}s + \text{residual} \tag{6}
\end{aligned}$$

Let us now examine the impact of interest rate volatility, σ . The bank group-specific parameter estimates shown in Table 2 reveal a high degree of statistical uncertainty regarding the impact of σ . Nevertheless, as predicted by economic theory (e.g., Ho and Saunders, 1981), the aggregate equations (5) show a significant positive relation between σ and the retail rates.

Figures 5 and 6 display the partial predictive power of r and σ , respectively, when all the other variables in the model (observed and unobserved) are kept constant over time. When the graphs are constructed, all variables except that on the horizontal axis are kept constant at the sample average, whereas the data points are ordered according to the variable on the horizontal axis. Comparing the actual data and the fitted interest curves in Figure 5, we see that the partial predictive power of r is quite good. On the other hand, using σ as (the sole) explanatory variable results in large prediction errors, as evident in Figure 6.

Table 2: Estimates of key parameters in the steady-state equation for six of the bank groups. Standard errors in parentheses.

Equation (X):	Intercept			Coefficients of					
	D	H	B	r	σ			B	
				D	H	B	D	H	B
Bank group A*	-.14 (.24)	2.82 (.23)	2.38 (.06)	.72 (.06)	.76 (.05)	.84 (.06)	1.04 (.39)	.87 (.26)	.46 (.30)
Bank group B	-.26 (.16)	2.32 (.22)	2.05 (.06)	.79 (.03)	.82 (.05)	.86 (.06)	.52 (.19)	.75 (.26)	.40 (.27)
Bank group C	-.27 (.34)	1.78 (.17)	1.33 (.35)	.91 (.04)	.85 (.04)	.93 (.07)	-.34 (.21)	.31 (.22)	-.17 (.55)
Bank group D	-.30 (.21)	2.52 (.22)	2.45 (.39)	.79 (.04)	.82 (.05)	.82 (.10)	.70 (.22)	.80 (.25)	1.48 (.51)
Bank group E	-.17 (.19)	2.59 (.22)	3.55 (.29)	.78 (.04)	.80 (.05)	.73 (.05)	.69 (.21)	.63 (.26)	.79 (.26)
Bank group F	-.22 (.23)	2.52 (.22)	2.83 (.49)	.76 (.04)	.81 (.05)	.77 (.07)	.58 (.19)	.71 (.25)	1.08 (.37)
Common estimate**	-.33 (.18)	2.34 (.20)	2.60 (.28)	.81 (.04)	.82 (.05)	.83 (.06)	.57 (.19)	.72 (.25)	.71 (.31)
P-value of Wald-test***	.15	.001	.000	.15	.53	.14	.004	.73	.20

* The bank groups reported here are not identified by name for confidentiality reasons. One bank-group (G) is omitted

** Average of seven bank group-specific coefficients weighted by inverse variance of estimator

***Wald test of the restriction that all parameters are equal (6 d.f.).

The estimated autoregressive parameters, ϕ_{ij}^X , corresponding to the bank group-specific retail rates, and the AR(1) coefficients ψ_k , corresponding to the three common dynamic factors f_{kt} , are displayed in Table 3. The number of lags is equal to two in most cases. These choices were made by applying the Akaike information criterion (see Raknerud *et al.*, 2010, for details regarding model selection in a similar model). All the lag polynomials $1 - \phi_{i1}^X L - \phi_{i2}^X L^2$ (where L is the lag operator) have roots well outside the unit circle, so the individual retail rates clearly evolve as stationary processes after subtracting the effects of the three common dynamic factors f_{kt} , $k = 1, 2, 3$. However, two of the dynamic factors are estimated to be unit root processes, so the retail series themselves are not stationary but evolve around a stochastic trend. These trends pick up, among other things, the decrease in average loan-deposit interest margins over time that is evident from Figure 4. The downward trend in both household and business interest margins over time may occur because of increased competition and increased productivity in the banking sector, e.g. due to Internet-based payment services.

The Wald tests reported in Table 3 reveal that there is significant bank group-specific heterogeneity in the interest rate dynamics with regard to the first-lag parameter, ϕ_{i1}^X . On the other hand, the hypothesis that ϕ_{i2}^X has a common value across banks could not be rejected for any retail rate. The estimated autoregressive parameters are typically less than 0.2 in absolute value, with $\phi_{i1}^X > 0$ and $\phi_{i2}^X < 0$. These estimates imply that retail rates adjust quickly to exogenous shocks.

Figures 7–8 depict the estimated response curves for the representative bank, that is, the *increase (decrease)* in retail rates as a function of time, given a permanent positive (negative) unit change in NIBOR ($\Delta r = \pm 1$). We see that all three curves are quite close, and stabilize at around 0.8, that is, $|\Delta r^X| \simeq 0.8|\Delta r|$. Moreover, the loan–deposit interest margins displayed in Figures 9–10 are not significantly

changed at any point in time following the increase in NIBOR. The confidence intervals in the figures reflect the statistical uncertainty in the estimates of the interest rate response functions of the representative bank. As noted above, there exists significant asymmetry between the short-run effects of a unit increase and a unit decrease in the NIBOR rate. The speed of adjustments following a permanent unit increase ($\delta = 1$) and a decrease ($\delta = -1$) are further displayed in Table 4. Almost all of the adjustment is completed by the end of the first quarter after the change in NIBOR (quarter 1), and approximately one-third to a half of the full adjustment is conducted in the same quarter (quarter 0). The exception is the deposit rate when NIBOR increases; then, the adjustment in the same quarter is estimated to approximately one-fifth on average, reflecting some rigidity in deposit rates in the case of a positive shock in the market rate. Table 4 reveals little, if any, systematic differences across bank groups.

Viewed in conjunction with the expression for banks' average net interest margin π in (1), our estimates reveal that π decreases with the level of the market rate, r , when r increases: the margins $r^H - r^D$ and $r^B - r^D$ remain unchanged, but the spreads on the loan rates relative to NIBOR, $r^H - r$ and $r^B - r$, decrease (because the coefficients of r are significantly less than one in the steady state). On the other hand, $r - r^D$ increases. In a perfectly competitive market, any increase in marginal funding costs, r , should be passed through to all retail rates. However, faced with a downward-sloping demand curve for loans, banks balance the positive price effect and the negative effect on the demand for loans when increasing their loan rates. Similarly, when faced with an upward-sloping supply curve for deposits, banks will take into consideration that deposits will decrease when the deposit rate is lowered. The presence of such effects is confirmed by our finding that the coefficients of r in (5) are clearly below one for all retail rates. This is in line with De Graeve *et al.*

(2007), who also analyze microdata, but contrary to most evidence from aggregate bank data (see De Bondt, 2002, for an overview).

The development in the average net interest margin of a representative bank when the NIBOR rate increases is illustrated in Figure 11. Here it is assumed that the average price of market funding in the quarter is equal to the three-month NIBOR. This assumption is not entirely realistic. First, the credit spread is ignored. Second, the *average* cost of market funding will not follow the NIBOR rate (the *marginal* cost) in the short run. The upper chart examines a scenario where the NIBOR rate is (*cet. par.*) gradually doubled from 2.6 at the beginning of quarter 0 to 5.2 percent at the end of quarter 0, and then remains permanently at this level. A tripling of the NIBOR rate to 7.8 during quarter 0 is illustrated in the lower chart of Figure 11. The immediate negative impact on the net interest margin is clearly visible. In the short run the banks can only partially adjust their retail rates, while (floating rate) market funding immediately becomes more costly. After 4–5 quarters, π stabilizes at a new but significantly lower level than the initial level because of incomplete pass-through. The peculiar nonmonotonic pattern in π in Figure 11 occurs because of the catching-up effect on the interest margins that follow the immediate decrease in π displayed in Figures 9–10. There is considerable heterogeneity between banks with regard to the effects of an increase the NIBOR rate. Bank groups with a large share of market financing (such as DnB NOR) are more vulnerable when NIBOR increases rapidly than are banks with a smaller share of market financing (such as Terra Gruppen). Banks' access to deposit financing makes them less vulnerable to short-run fluctuations in the NIBOR rate. On the other hand, to increase its market share a bank needs to rely more on market funding, which makes it more vulnerable to shocks in the market rate.

Table 3: Estimates of autoregressive parameters for six of the bank groups. Standard errors in parentheses

Equation: (X)	ϕ_{i1}^X		ϕ_{i2}^X		Common dynamic factors		
	D	H	D	H	ψ_1	ψ_2	ψ_3
Bank group A*	.56 (.08)	.24 (.06)	.35 (.08)	-.18 (.04)	-.13 (.05)		
Bank group B	.10 (.05)	.19 (.06)	.42 (.06)	-.14 (.04)	-.14 (.04)		
Bank group C	-.09 (.09)	.11 (.06)	.22 (.14)	-.13 (.03)	–		
Bank group D	.19 (.06)	.13 (.05)	.61 (.08)	-.13 (.04)	-.10 (.04)		
Bank group E	.14 (.05)	.11 (.05)	.24 (.05)	-.13 (.04)	-.08 (.04)		
Bank group F	.08 (.05)	.12 (.05)	.41 (.08)	-.12 (.04)	–		
Common estimate**	.18 (.05)	.20 (.05)	.44 (.07)	-.13 (.03)	-.09 (.04)	.59 (.04)	1.00 (.02)
P-value Wald-test***	.000	.005	.005	.70	.40	.22	

* The bank groups reported here are not identified by name for confidentiality reasons.

** The results for ϕ_{ij}^X are averages of seven bank group-specific coefficients weighted by their inverse variance.

***Wald test of the restriction that all bank groups have equal parameters (6 d.f.).

Table 4: **Adjustment speed after a permanent change equal to δ in the NIBOR rate.** Standard errors in parentheses.

Equation (X):	$\delta = 1$ (increase)			$\delta = -1$ (decrease)			$\delta = 1$			$\delta = -1$		
	D	H	B	D	B	D	D	H	B	D	H	B
Bank group A**	.34 (.08)	.42 (.06)	.41 (.06)	.59 (.08)	.71 (.07)	.99 (.06)	.71 (.07)	.99 (.06)	.89 (.04)	.97 (.05)	.89 (.04)	.97 (.05)
Bank group B	.16 (.05)	.35 (.07)	.49 (.05)	.39 (.05)	.92 (.08)	.95 (.06)	.92 (.08)	.95 (.06)	.89 (.04)	1.06 (.03)	.89 (.04)	1.06 (.03)
Bank group C	.20 (.06)	.29 (.06)	.40 (.09)	.41 (.06)	1.05 (.08)	.99 (.06)	1.05 (.08)	.99 (.06)	.77 (.05)	1.13 (.05)	.77 (.05)	1.13 (.05)
Bank group D	.20 (.05)	.32 (.07)	.72 (.08)	.43 (.05)	.82 (.08)	.98 (.05)	.82 (.08)	.98 (.05)	.88 (.06)	.98 (.04)	.88 (.06)	.98 (.04)
Bank group E	.16 (.06)	.33 (.07)	.38 (.06)	.39 (.06)	.87 (.08)	.98 (.05)	.87 (.08)	.98 (.05)	.87 (.03)	1.01 (.03)	.87 (.03)	1.01 (.03)
Bank group F	.14 (.06)	.33 (.06)	.57 (.06)	.37 (.06)	.89 (.08)	.97 (.05)	.89 (.08)	.97 (.05)	.83 (.04)	1.03 (.04)	.83 (.04)	1.03 (.04)
Common estimate***	.18 (.05)	.33 (.06)	.48 (.07)	.41 (.06)	.88 (.07)	.97 (.05)	.88 (.07)	.97 (.05)	.85 (.03)	1.02 (.03)	.85 (.03)	1.02 (.03)
P-value Wald-test****	.72	.73	.006	.34	.28	.13	.28	.13	.55	.29	.55	.29

* That is, $\Delta r_{i,t}^X(\delta) / \sum_{j=0}^{\infty} \Delta r_{i,t+j}^X(\delta)$, $X = D, H, F$

**The bank groups reported here are not identified by name for confidentiality reasons.

*** Average of bank group-specific coefficients weighted by inverse variance.

****Wald test of the restriction that all bank groups have equal parameters (6 d.f.).

5 Conclusion

We have used a dynamic factor model and a detailed panel data set with quarterly accounts data on all Norwegian banks to study how banks' funding costs affect their interest rates. We found clear evidence of incomplete pass-through from the market rate to retail rates, with lending rates increasing more and deposit rates increasing less than the market rate. In our analysis the cost of market funding was estimated by the three-month Norwegian Inter Bank Offered Rate (NIBOR) and the spread of unsecured bonds issued by Norwegian banks. .

Our estimates show that a unit increase in NIBOR leads to an approximately 0.8 increase in banks' retail rates (both loan rates and deposit rates) in the long run. These findings are consistent with each bank facing a downward-sloping demand curve for loans and an upward-sloping supply curve for customer deposits. While the margin between loan and deposit rates remains unchanged when the NIBOR rate increases, the spread between the loan rate and the NIBOR rate decreases. The results indicate that banks balance a positive price effect and a negative effect on the demand for loans when deciding on an increase in lending rates. There is also a significant positive relation between the indicative credit spread of uncovered bonds issued by banks and loan rates, especially regarding loans to businesses. The estimated effects on the interest rate margin between loan and deposit rates indicate that a permanent unit increase in credit spread leads to a long-term increase in the interest rate margins of roughly 0.3 to 0.4.

The econometric relations established in this paper should be useful in a stress test framework, where one is typically interested in how shocks in market rates or policy rates affect banks' lending rates and net interest margins. Another topic, which is currently of great policy importance, is how the effect of tighter capital and liquidity requirements, as proposed in the Basel III reform, will affect bank rates (see

Angelini *et al.*, 2011). For example, the reform is expected to increase the average maturity of banks' wholesale funding, which will increase the credit spread relative to NIBOR if the yield curve is increasing. To the extent that the direct impact of these regulatory measures on the (indicative) credit spread can be assessed, our econometric framework can be used directly to estimate the impact of such changes on lending rates and interest margins.

References

- [1] Allen, L (1988). “The determinants of bank interest margins: A note”. *Journal of Financial and Quantitative Analysis*, 23, 231–235.
- [2] Andersen, H., S.A. Berg and E. Jansen (2008). “The dynamics of operating income in the Norwegian banking sector”. Working paper 2008/13, Norges Bank.
- [3] Angelini, P., L. Clerk, V. Curdia, L. Gambacorta, A. Gerali, A. Locarno, R. Motto, W. Roeger, S. van den Heuvel and J. Vlcek (2011). “Basel III. Long term impact on economic performance and fluctuations”. BIS Working paper 338, Bank of International Settlements.
- [4] Akram, Q. F. and C. Christophersen (2011). “Norwegian overnight interbank interest rates”. Staff memo 1/2011, Norges Bank.
- [5] Andrews, L. (1991). “Heteroskedasticity and Autocorrelation Consistent Covariance Matrix Estimation”. *Econometrica*, 59, 817–858.
- [6] Angbazo, L. (1997). “Commercial bank net interest margins, default risk, interest-rate risk and off-balance sheet banking”. *Journal of Banking and Finance*, 21, 55–87.
- [7] Anundsen, A.K.. and E.S. Jansen (2011). “Self-reinforcing effects between housing prices and credit. Evidence from Norway ”. Discussion papers 651, Statistics Norway.
- [8] Bernanke, B.S., Boivin, J. and P. Elias (2005). “Measuring the effects of monetary policy: A factor-augmented vector autoregressive (FAVAR) approach”. *Quarterly Journal of Economics*, 120, 387–422.

- [9] Chan, K.C., A. Karolyi, F.A. Longstaff and A.B. Sanders, I. (1992). “An empirical comparison of alternative models of the short-term interest rate”. *The Journal of Finance*, 47, 1209–1227.
- [10] Choi, I.(1994). “Residual-Based Tests for the Null of Stationarity with Applications to U.S. Macroeconomic Time Series”. *Econometric Theory*, 10, 720–746.
- [11] Choi, I. and B.C. Ahn (1999). “Testing the Null of Stationarity for Multiple Time Series”. *Journal of Econometrics*, 88, 41–77.
- [12] Chong, B.S. (2010). “Interest rate deregulation: Monetary policy efficacy and rate rigidity”. *Journal of Banking and Finance*, 34, 1299–1307.
- [13] De Bondt, G. (2002). “Retail bank interest rate pass-through: New evidence at the Euro area level”. ECB Working Papers, No 136.
- [14] De Graeve, F., O. De Jonghe and R.V. Vennet (2007). “Competition, transmission and bank pricing policies: Evidence from Belgian loan and deposit markets”. *Journal of Banking and Finance*, 31, 259–278.
- [15] Forni, M., Hallin, M., Lippi, M. and L. Reichlin (2000). “The generalized dynamic factor model: Identification and estimation”. *Review of Economics and Statistics*, 82, 540–554.
- [16] Hannan, T. and A. Berger (1991). “The rigidity of prices: Evidence from banking industry” *American Economic Review*, 81, 938–945.
- [17] Harvey, A.C. (1989). *Forecasting, Structural Time Series Models and the Kalman Filter*. Cambridge. Cambridge University Press.
- [18] Ho, T and A. Saunders (1981). “The determinants of banks’ interest margins:

- Theory and empirical evidence” *Journal of Financial and Quantitative Analysis*, 16, 581–600.
- [19] Hofmann, B and P. Mizen (2004). “Interest rate pass-through and monetary transmission: Evidence from individual financial institutions’ retail rates” *Economica*, 71, 99–123.
- [20] Maudos, J and J. Fernandez de Guevara (2004). “Factors explaining the interest margin in the banking sectors of the European Union”. *Journal of Banking and Finance*, 28, 2259–2281.
- [21] Pesaran, M. H. and R.P. Smith (1995). “Estimating long-run relationships from dynamic heterogeneous panels”. *Journal of Econometrics*, 68, 79–113.
- [22] Raknerud, A., T. Skjerpen and A.R. Swensen (2010). “Forecasting key macroeconomic variables from a large number of predictors: A state space approach”. *Journal of Forecasting*, 29, 367–387.
- [23] Rosen, D and D. Saunders (2010). “Risk factors contributing in portfolio credit risk models”. *Journal of Banking and Finance*, 34, 336–349.
- [24] Saunders, A. and L. Schumacher (2000). “The determinants of bank interest rate margins: An international study”. *Journal of International Money and Finance*, 19, 813–832.
- [25] Stock, J.H. and M.W. Watson (2002). “Forecasting using principal components from a large number of predictors”. *Journal of the American Statistical Association*, 97, 1167–1179.

Figures

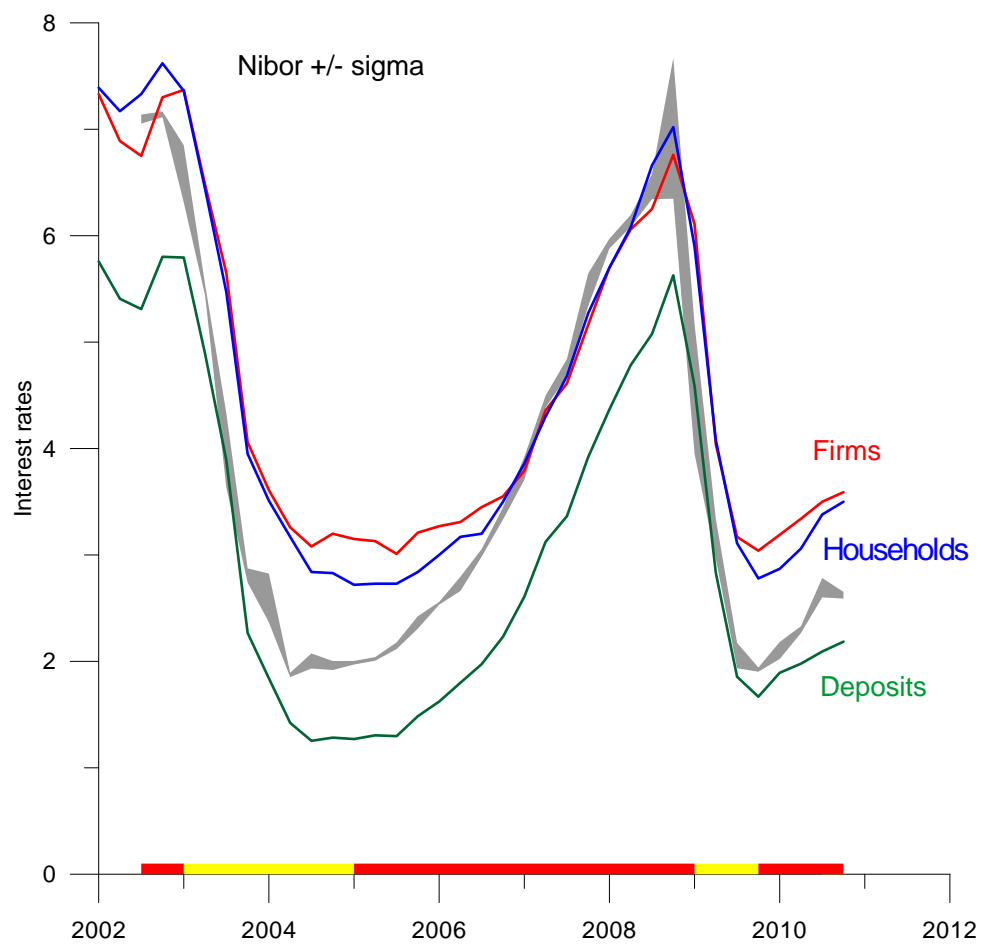


Figure 1: Three month NIBOR rates and average bank interest rates on deposits and loans to firms and households

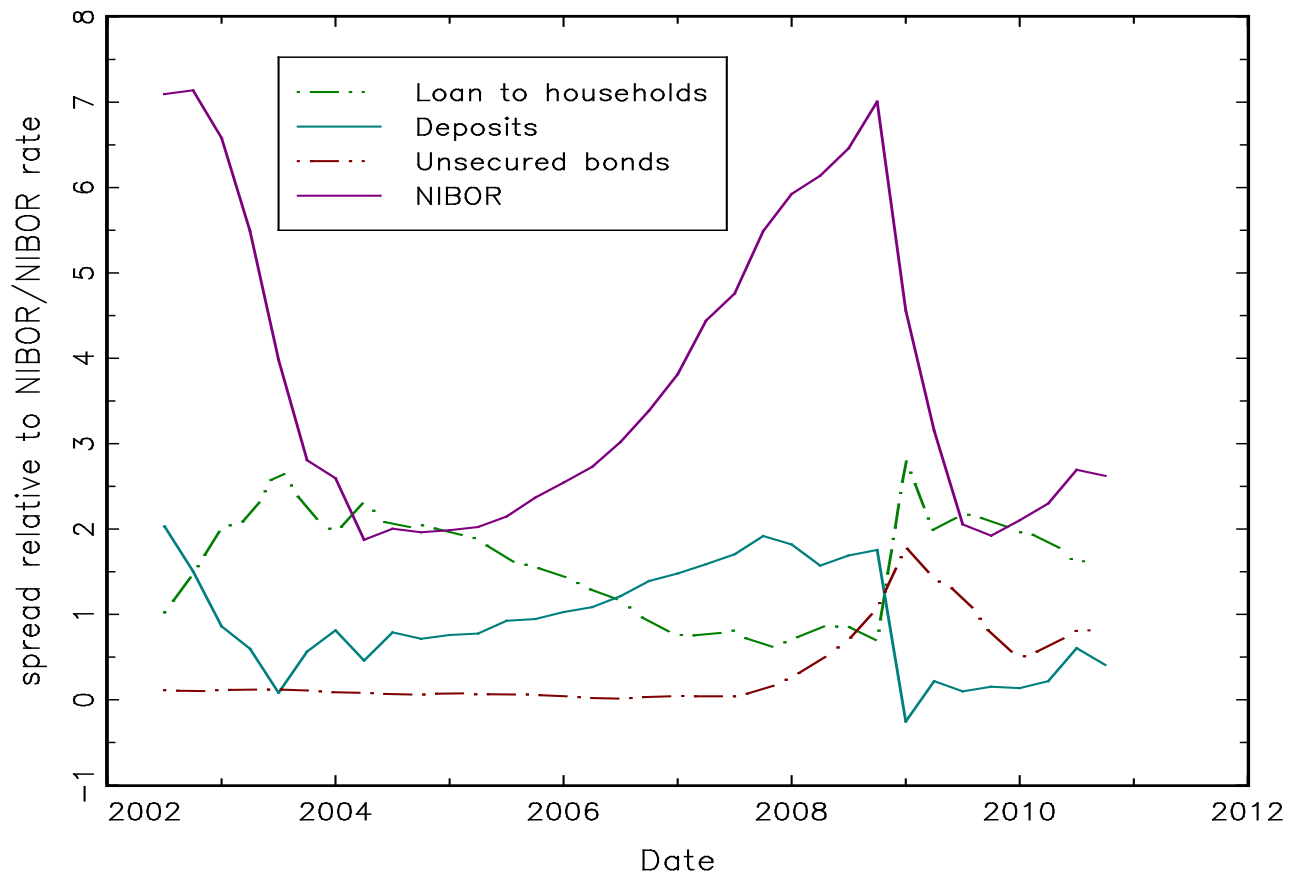


Figure 2: The NIBOR rate and interest rate spreads relative to NIBOR on i) loan to households ($r^H - r$), ii) bank deposits ($r - r^D$) and iii) unsecured bank bonds (s)

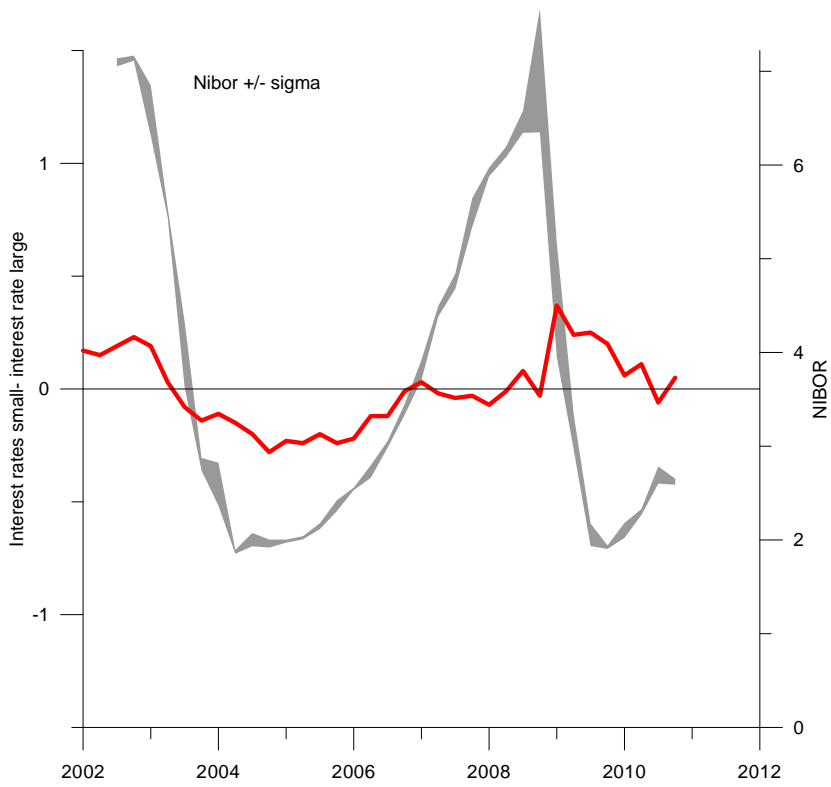


Figure 3: The deposit interest rate difference between a group of small banks and a group of large banks

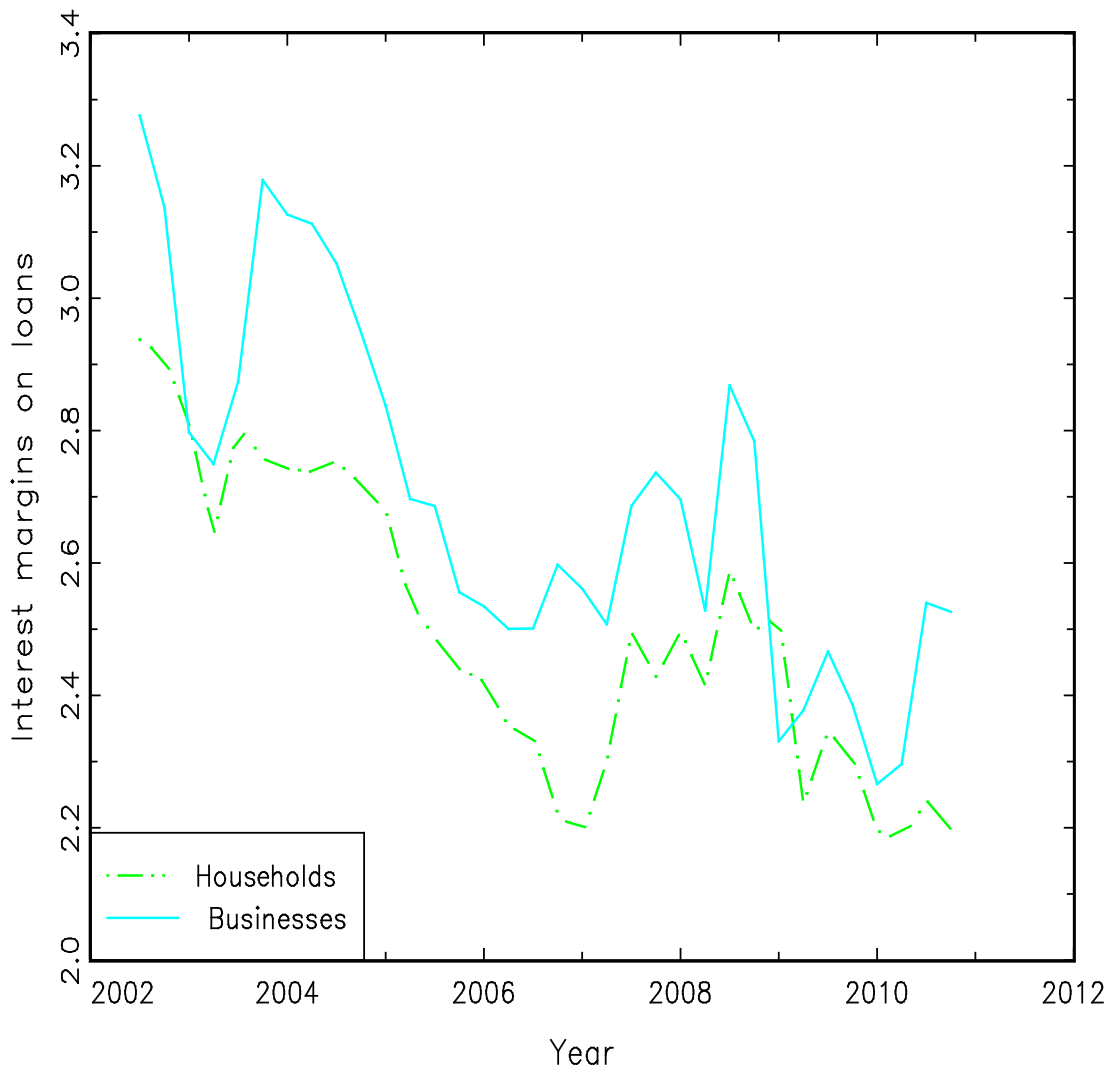
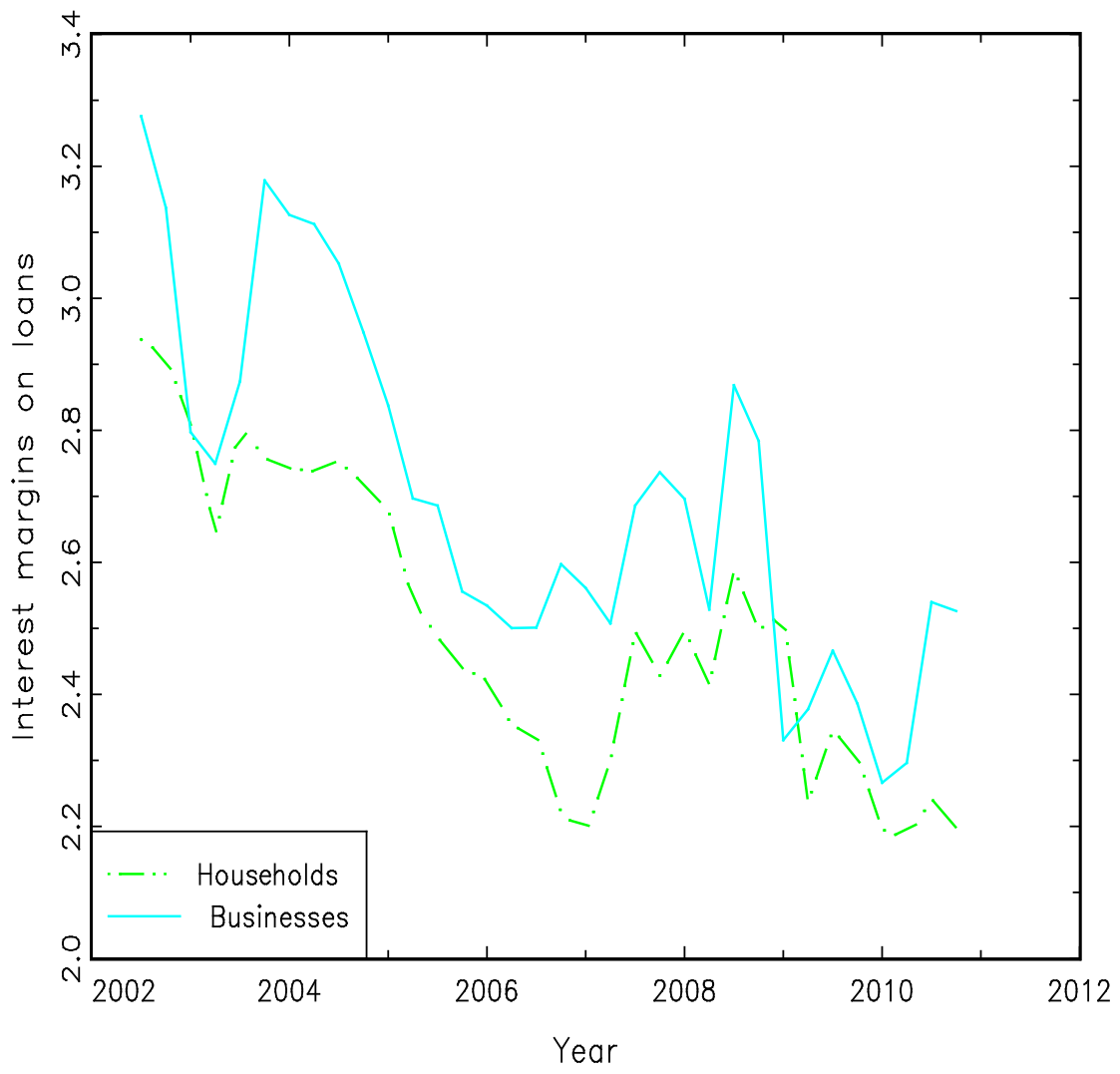


Figure 4: Interest rate margins between loans and deposits 2002-2010. Weighted average across banks



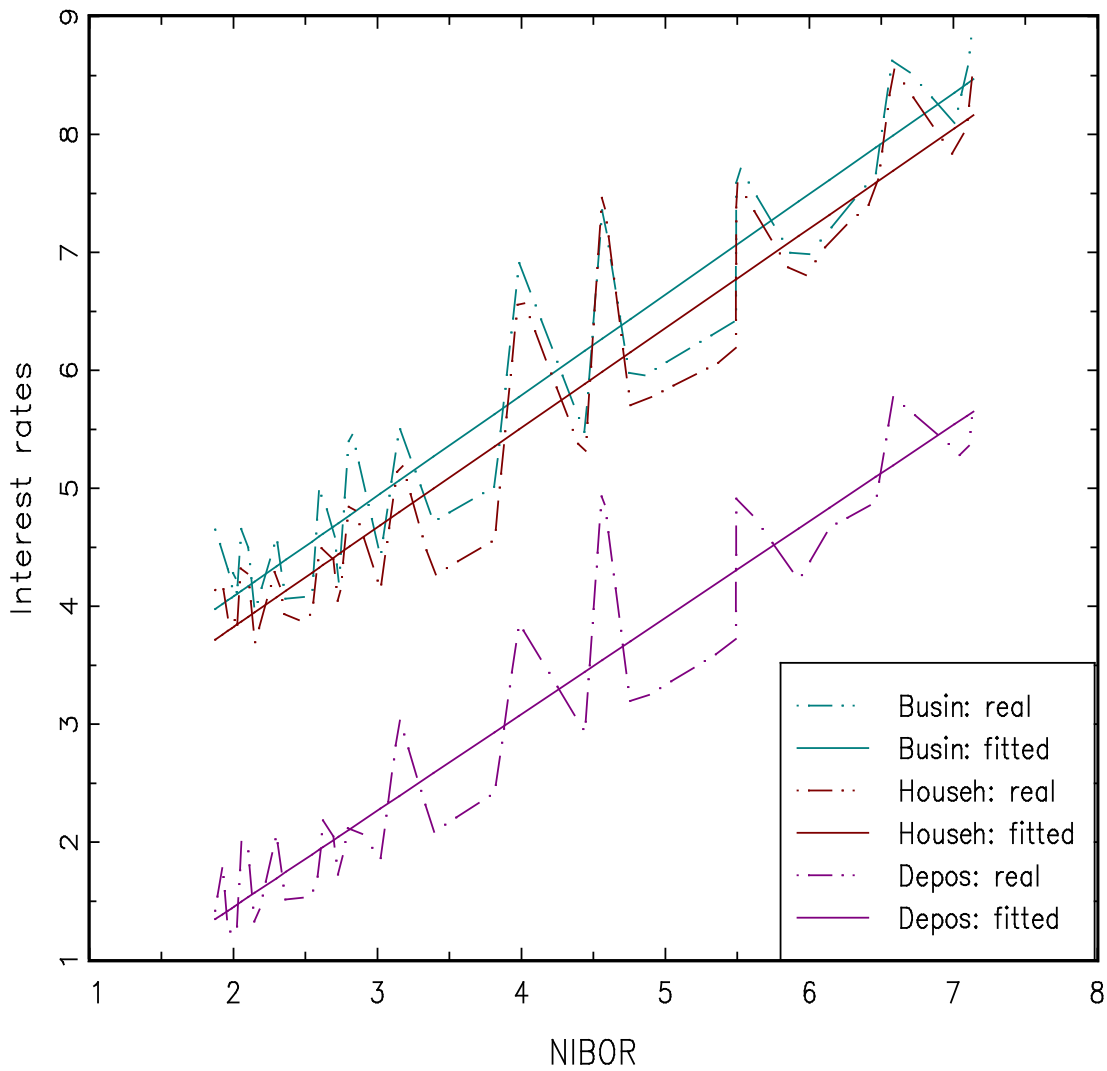


Figure 5: Actual data and fitted interest rate curves from estimated steady state equation using NIBOR as sole predictor

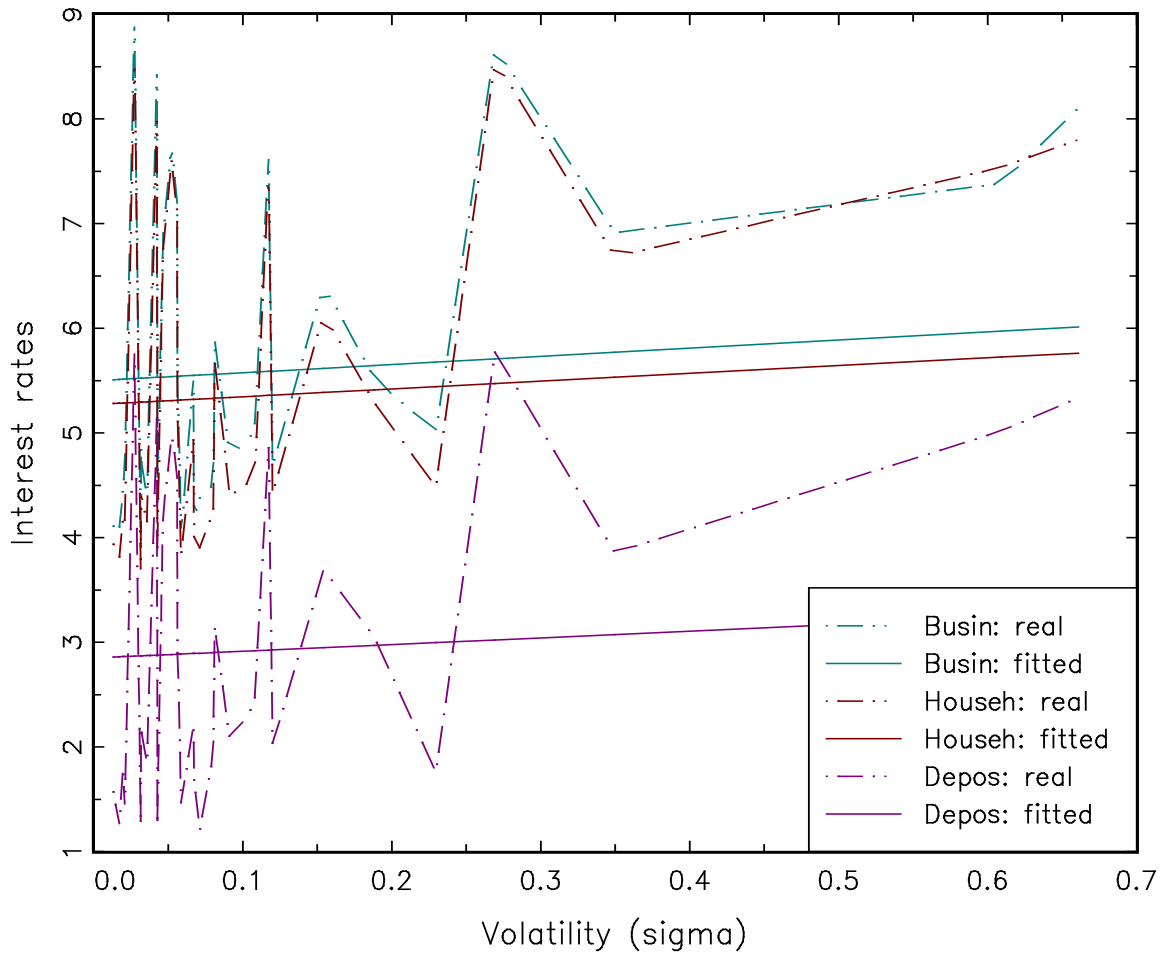
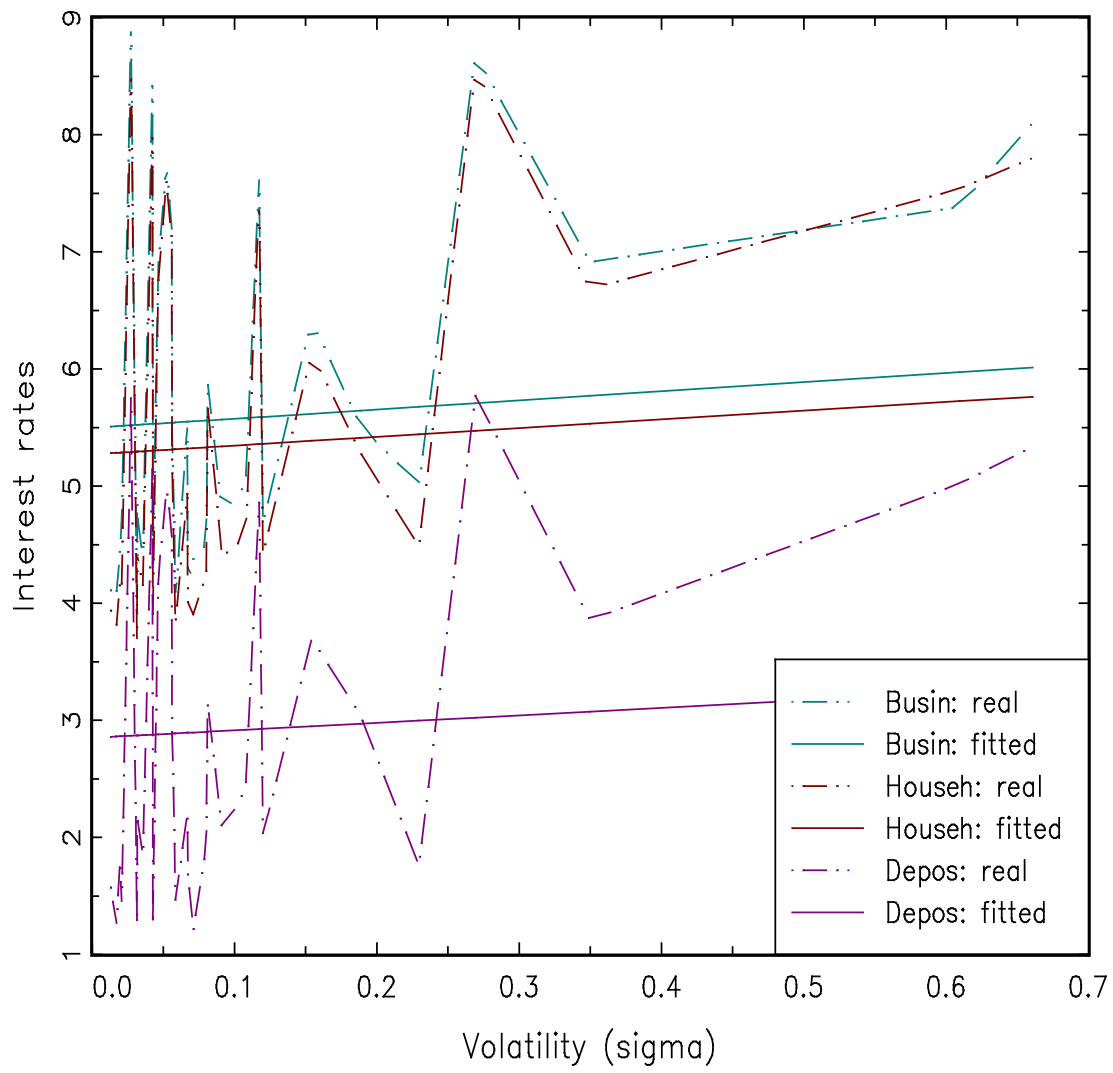


Figure 6: Actual data and fitted interest rate curves from estimated steady state equation using volatility as sole predictor



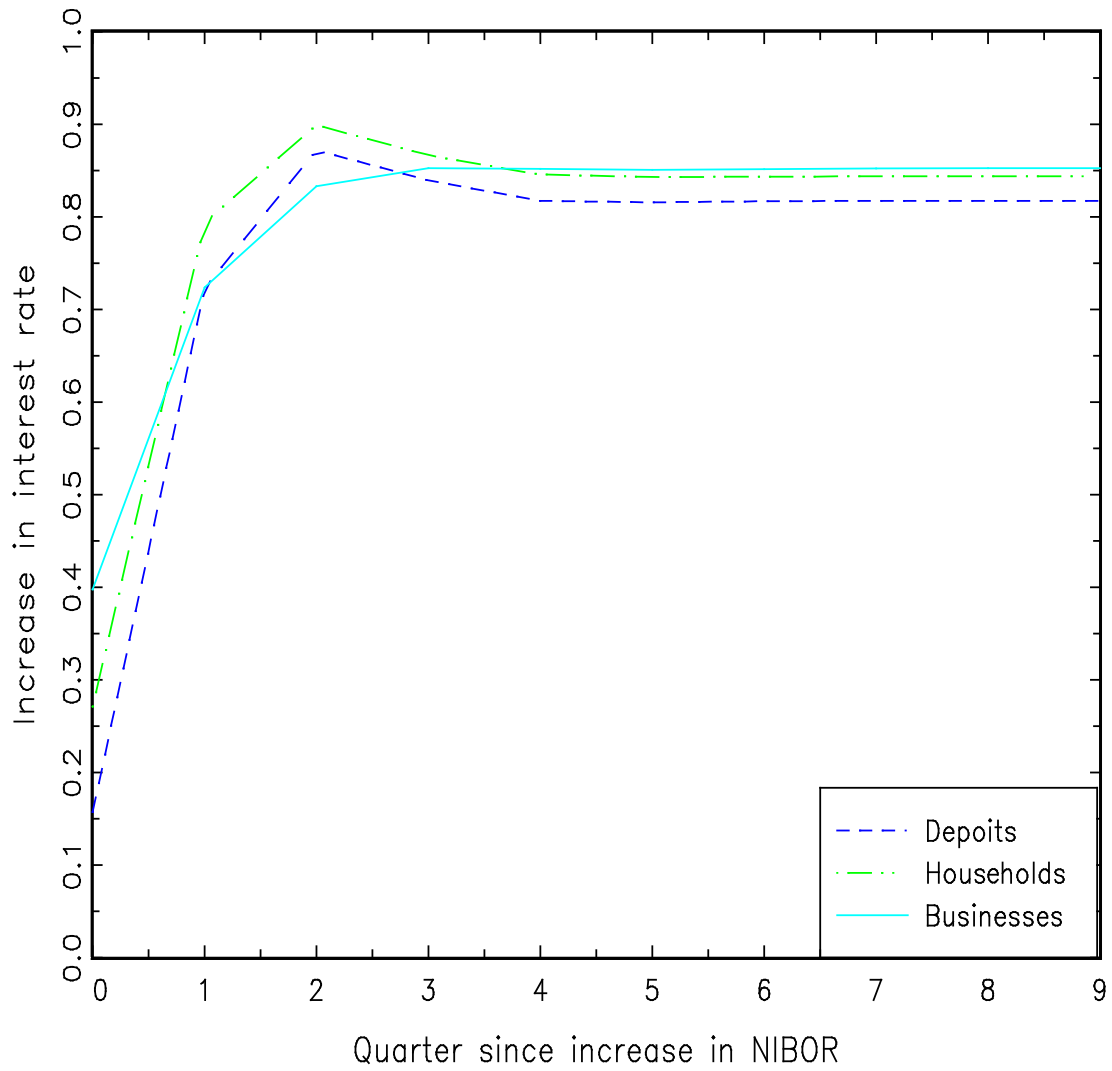


Figure 7: Estimated interest rates response functions: effects of a unit increase in NIBOR (in quarter 0). Average across banks

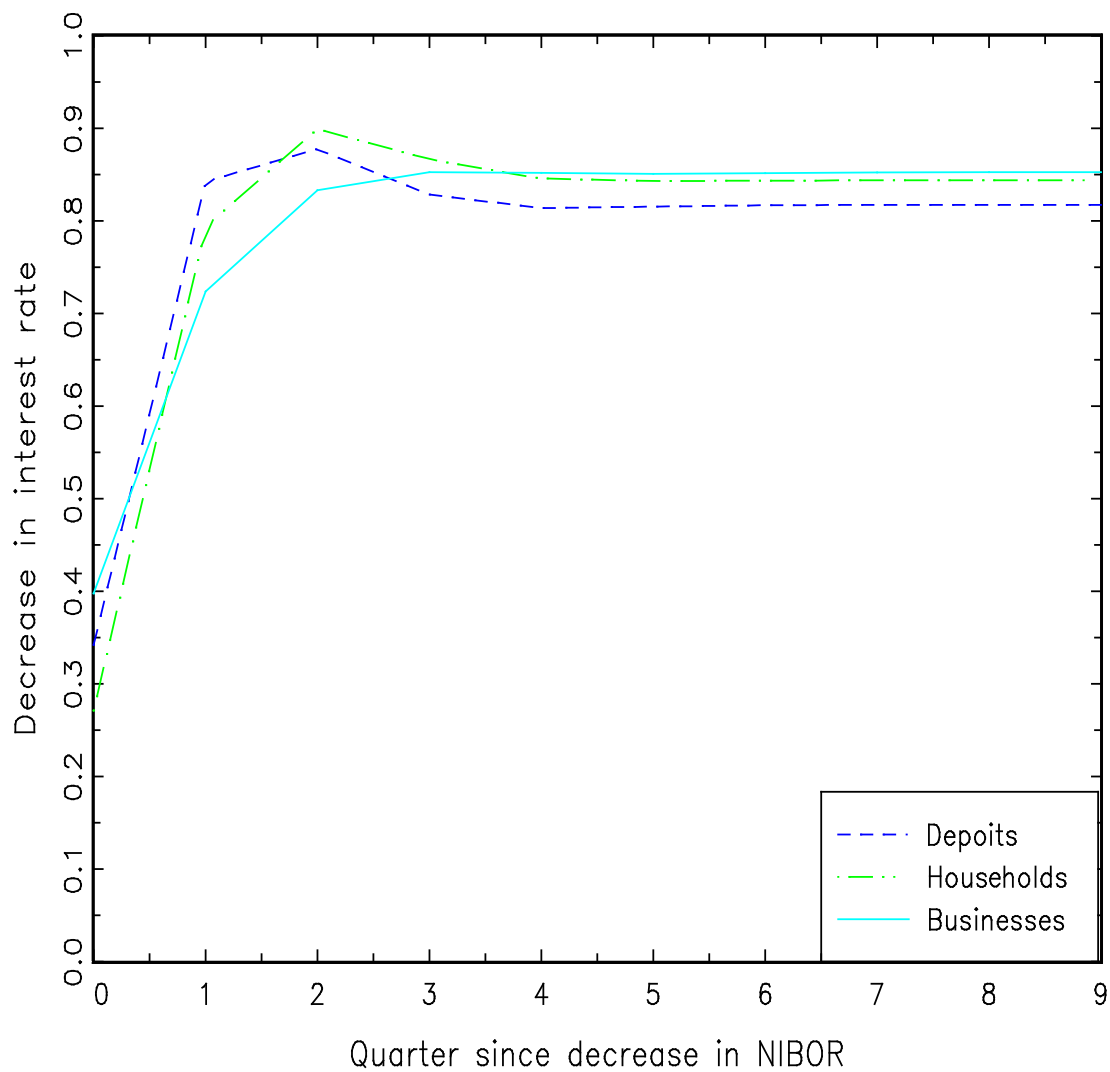
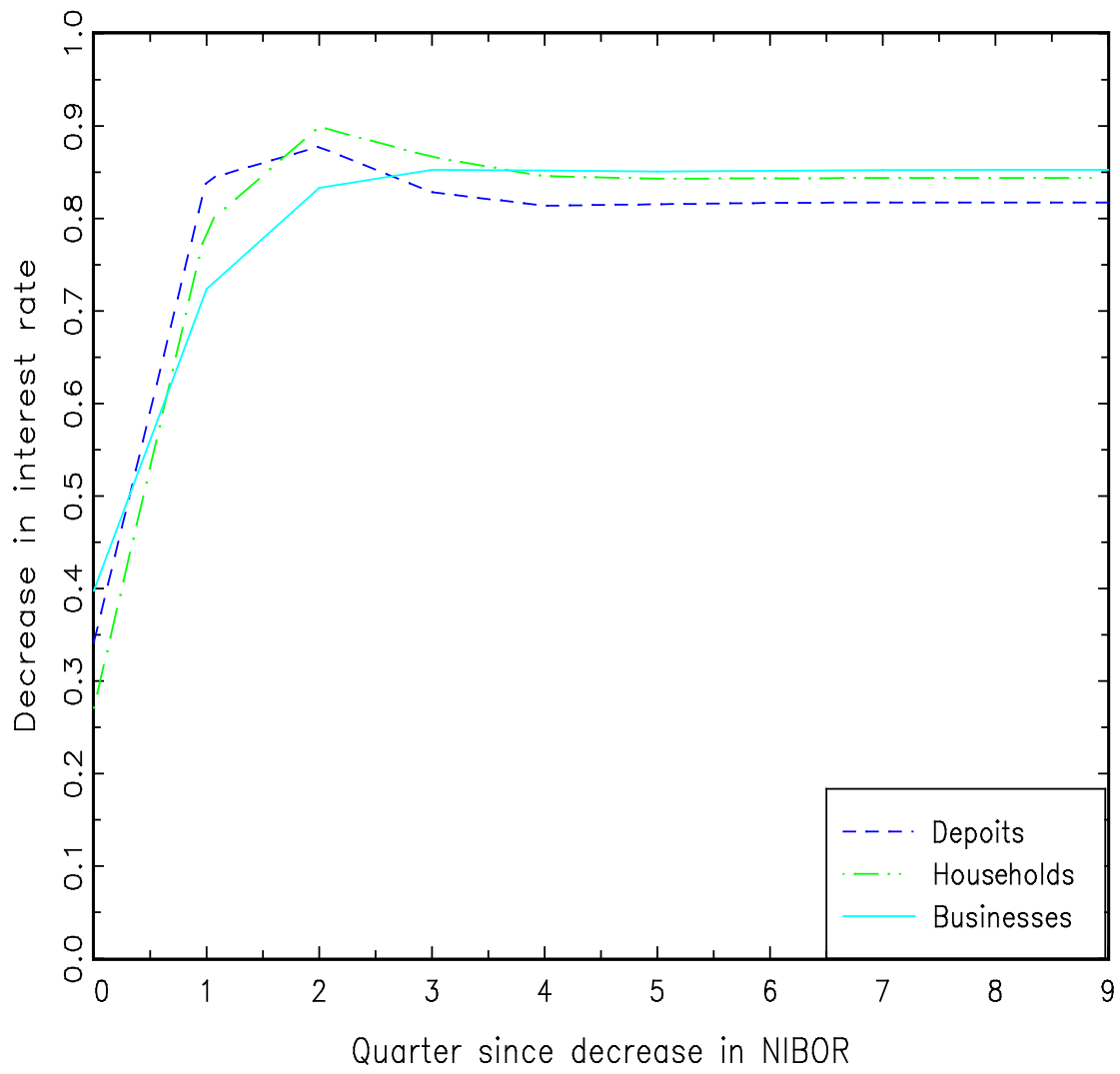


Figure 8: Interest rates response functions: effects of a unit decrease in NIBOR (in quarter 0)



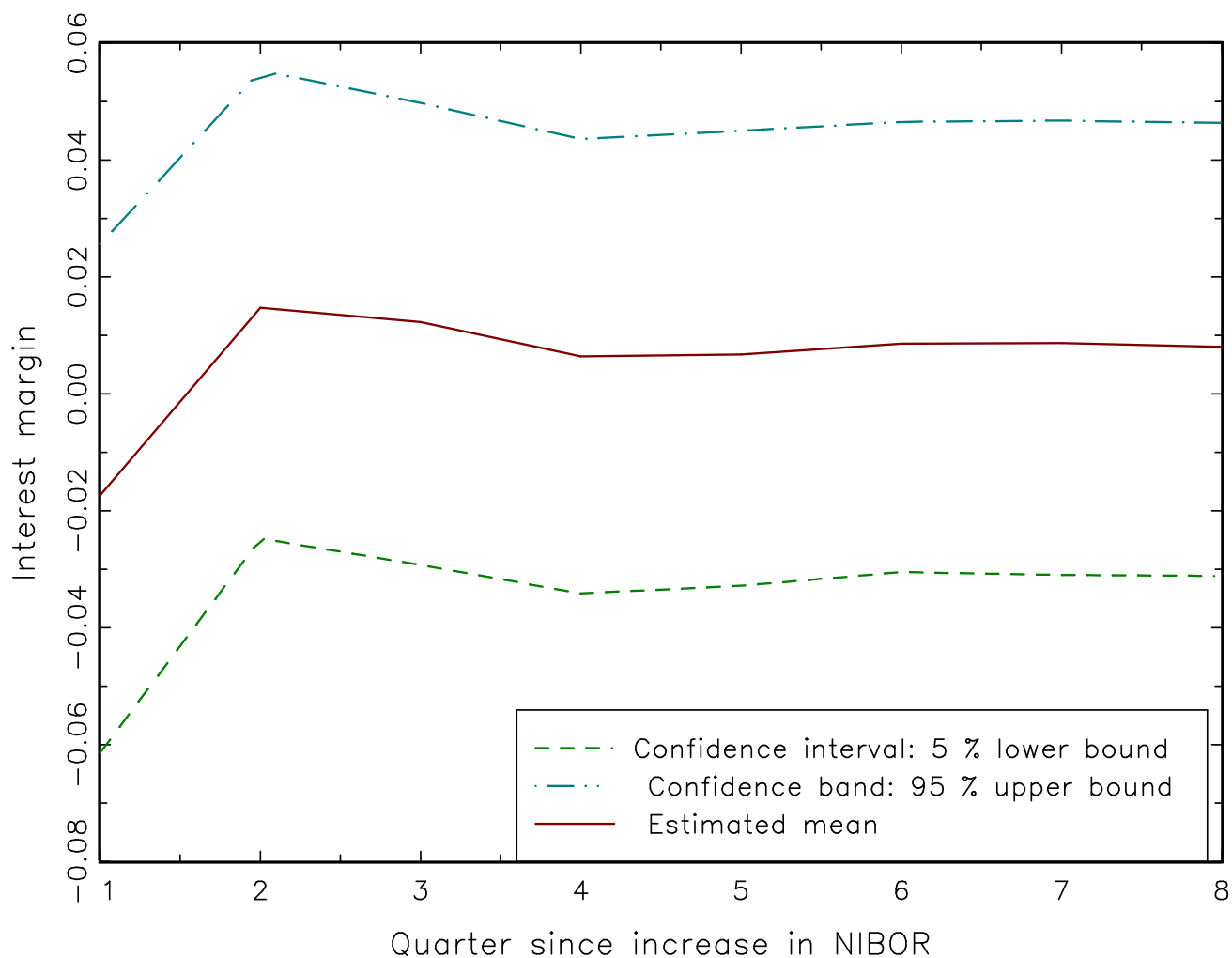


Figure 9: Changes in the difference between the household loan rate and the deposit rate following a unit increase in NIBOR for a representative bank. Zero corresponds to no change relative to the initial (quarter 0) interest rate margin.

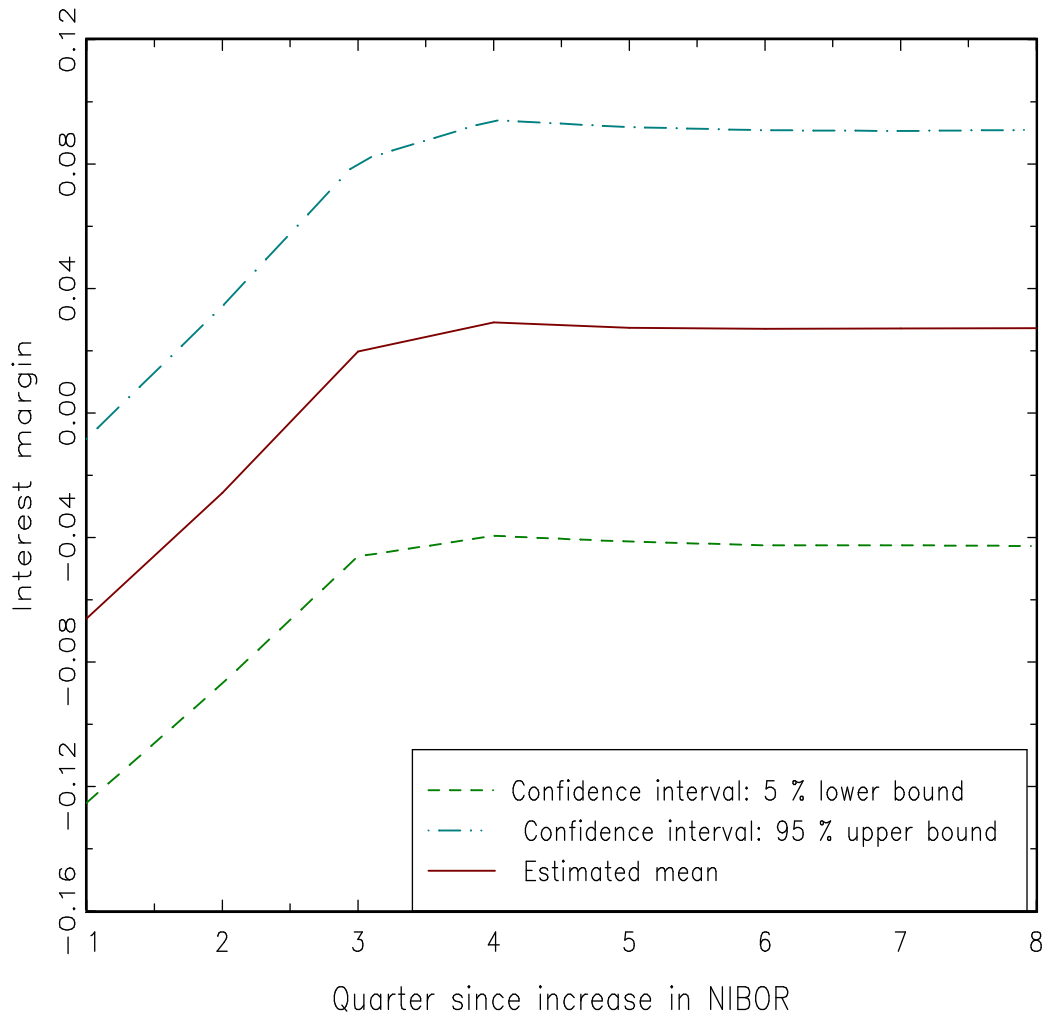


Figure 10: Changes in the difference between the business loan rate and the deposit rate following a unit increase in NIBOR for a representative bank. Zero corresponds to no change relative to the initial (quarter 0) interest rate margin.

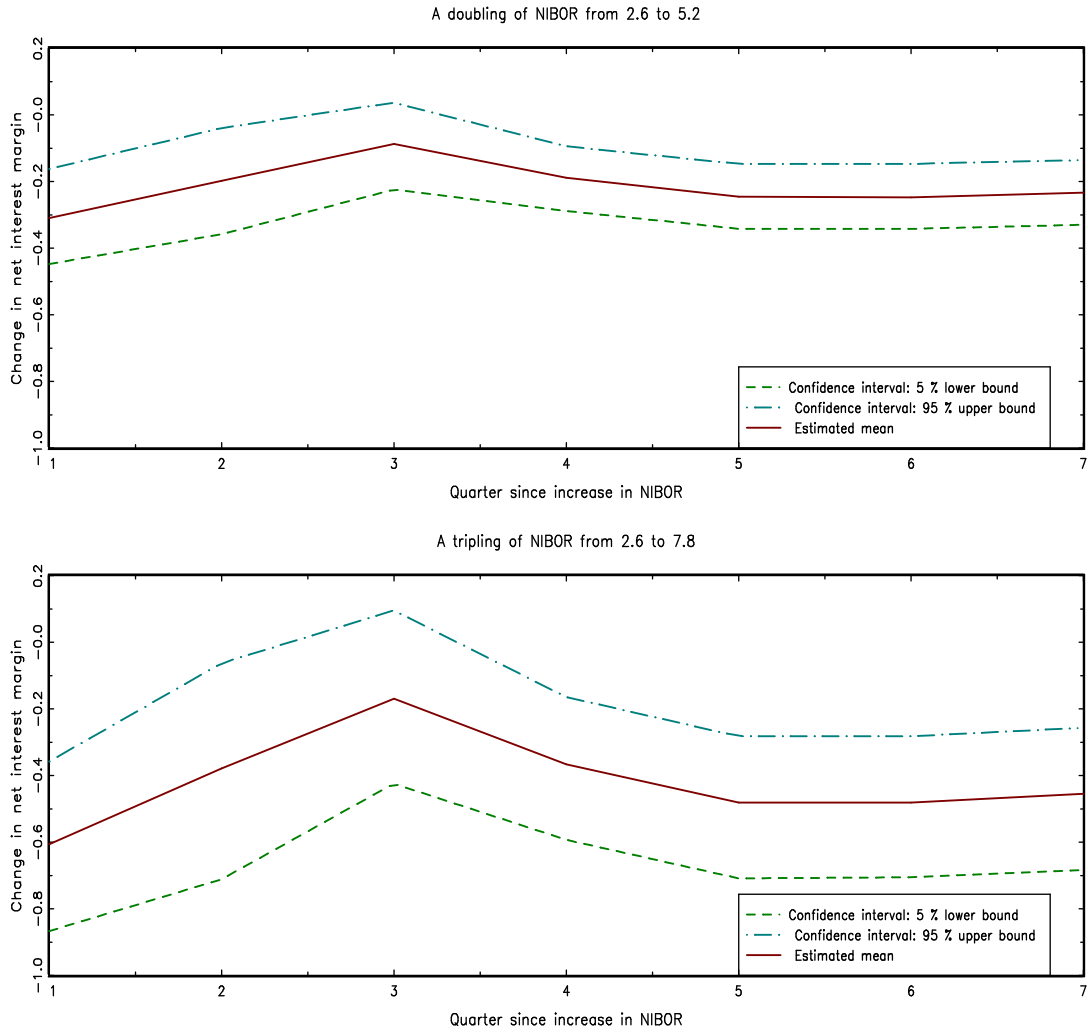


Figure 11: The estimated change in net interest rate margin when increasing NIBOR from 2.6 to 5.2 and 7.8 per cent, respectively. Weighted average across bank groups