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Interlocking Boards and Firm Performance: Evidence from a New Panel Database

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Abstract

An interlock between two firms occurs if the firms share one or more directors in their boards of directors. We explore the effect of interlocks on firm performance for 101 large Dutch firms using a large and new panel database. We use five different performance measures, and for each performance measure we design three different panel data models, where we allow the effect of the number of interlocks to be linear, quadratic or square root, either with or without lags.

Based on all results we conclude that current interlocks can have a negative effect on future firm performance. We show that this negative effect is jointly established by (1) interlocking directors being too busy and (2) by directors being members of a homogenous upper class group.

Keywords: interlocks, firm performance

JEL codes: C23, G34, J53, Z13

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1 Introduction

A director can hold several directorships in different firms. Such a director constitutes a link between the firms. Firms that are linked in this way are interlocked. There is much research on interlocks, ranging from a description of what the network of interlocked firms looks like to studies on the influence of interlocks on firm strategy and performance. We address this last topic by analyzing a new large and unique panel data set concerning firms in The Netherlands.

There are several views on the origin and effect of interlocks, see Mizruchi (1996) for an extensive review. Here we mention four well-known views. The first is that interlocks are a way for firms to coopt and/or monitor each other (Dooley, 1969 and Mizruchi and Stearns, 1994). The second view states that interlocks provide firms with information on business practice (Davis, 1991). The third is that interlocks merely reflect upper-class cohesion (Useem, 1984). And, fourth, the recently put forward busyness hypothesis of Ferris et al. (2003) states that multiple directorships place an excessive burden on directors (see also Fich and Shivdasani, 2006). The first two views predict a *positive* influence of interlocks on firm performance. In contrast, the busyness hypothesis predicts a *negative* influence on firm performance. Finally, the effect of interlocks can be either positive, negative or neutral when interlocks reflect upper-class cohesion. So, only the upper class cohesion theory and the busyness hypothesis can explain a negative effect of interlocks on performance, the other two views predict a positive effect¹.

When interlocks would indeed reflect upper class cohesion the effect of interlocks can be either positive, negative or neutral. The argument that interlocks in this case have a negative effect proceeds as follows. In The Netherlands there is a cohesive upper class of directors, who often have more than two directorships and meet each other regularly (see Stokman et al., 1988, and Van Hezewijk, 1986, 1988). The boards of firms with many interlocks per director apparently consist of directors who mostly belong to this particular cohesive upper class. Cohesive groups perform worse in decision making, as they strive for unanimity and often suffer from a reduction in independent critical thinking (see Janis, 1982, and Mullen et al., 1994). Moreover, the members of the Dutch cohesive upper class mostly have the same background (Van Hezewijk, 1986, 1988) and hence such a board is less diverse, while it is precisely such diversity that has been shown to *improve* firm performance (see Carter et al., 2003).

Given the different views, it is not surprising that much empirical research on the effect of interlocks on performance has been carried out. Results of these studies are mixed, see for example Bunting (1976), Pennings (1980), Burt (1983), Fligstein and Brantley (1992), and Phan et al. (2003). Note that most research is based on US data. We are aware of only two studies which concern The Netherlands, and these are Meeusen and Cuyvers (1985) and Van Ees et al. (2003). The first study documents a positive relation between interlocks within financial firms and their performance. The second study mentions a negative effect of the percentage share of outsiders on firm performance, where outsiders are defined as directors who have at least two directorships.

In our present paper we again take up the issue of interlocks and performance by presenting empirical results based on a newly created large panel database for The Netherlands. In contrast to previous studies, our database constitutes a panel for many years, instead of the commonly used cross section. Hence, we can also examine the dynamic effects of interlocks, which, as we will document, are quite prominent.

In the first part of our paper we explore the relation between the number of interlocks and firm performance. If we find an effect for some models and some combinations of variables it is a negative effect. This effect matches with the busyness hypothesis and with the cohesive group theory. In the remainder of the paper we seek to establish which of the two is most plausible, at least for the data at hand. Looking at the results of various regressions leads us to conclude that we cannot clearly distinguish between the two hypotheses, as both seem partly valid.

The outline of the paper is as follows. First, in Section 2 we give a description of our data. Next, in Section 3, we will set out the method we use to investigate the effect of the number of interlocks on firm performance. The results of applying this method to the data are presented in Section 4. In Section 5 we elaborate on possible explanations of the negative influence of interlocks on performance and in Section 6 we summarize our findings.

2 A new panel database

In The Netherlands virtually all firms have a two-tier board structure. There is an executive board, which consists of a Chief Executive Officer (CEO), a Chief Financial officer (CFO) and of other executive directors. Additional to this executive board, there also is a supervisory board. The main task of this board is to monitor the executive board itself and to monitor the major business decisions taken by the executive board. The supervisory board largely consists of retired executive directors. Almost all of the interlocks of a firm are formed by members of the supervisory board who also serve as supervisory director for other firms. It is not uncommon for a high-profile director to have four or five supervisory directorships. As interlocks are mainly formed by the supervisory board, we focus on these directors.

Defining interlocks

We gathered data on supervisory boards of 101 large, listed, Dutch firms in the period 1994 to 2004, using annual reports and the REACH database. The annual reports give us, for each firm, the directors in the supervisory board by the end of July in each year. Using this information, we count the number of interlocks of a firm with other firms in the database. For example, suppose firm A has two directors, X and Y. X also sits on the supervisory boards of firms B and C, and Y also sits on the board of firm D. As such, firm A then has three interlocks. Multiple interlocks are counted as one. Suppose Y is on the boards of A, D and also B, like X. Then, firm A still has only three interlocks, as the multiple interlock with firm B is counted as one interlock.

We divide the number of interlocks by the number of directors to correct for the size of the board². In our database there are a few firms with very large boards (15 or more members) as well as firms with smaller (3 or less members) boards. Clearly, large boards can have much more interlocks than small boards.

We acknowledge that the number of interlocks per director is the best variable

to measure the directors' experience and busyness, while the number of interlocks itself is the best variable to measure the amount of information the firm gets about its environment. We therefore did the same analysis as reported below using the number of interlocks instead of the number of interlocks divided by the size of the board, and the results are qualitatively the same. In what follows we will therefore use the term 'interlocks' for the number of interlocks divided by the size of the board. Figure 1 shows the network of interlocks for the year 1998. For other years the network looks roughly the same (although of course various little changes occur over time). We see one giant component of firms that are linked to each other and several fringe firms that have no links or are only linked to another fringe firm.

Performance measures

Additional to information on the supervisory boards, we also gathered data on the performance of the firms during 1994 to 2004 using the REACH database. We gathered data on stock returns, the price-earnings ratio and the price-to-book ratio, the return on assets and the return on equity. Furthermore, for each firm we store the BIK codes³ (four-digit level), the turnover and the growth of the turnover.

Some of the dependent variables (stock returns, the price-earnings ratio and the price-to-book ratio) show serious skewness and excess kurtosis. To accommodate this, we transform the stock return by taking the log of (1 + return/100). For the price-earnings ratio we delete all observations on firms that make losses, as the price-earnings ratio is not defined for these firms. We then transform the resulting price-earnings ratios by taking natural logs. Finally, the price-to-book ratio is also taken in natural logs.

We have the BIK codes of the firms in the sample at the four-digit level. Most firms have several BIK codes as our database concerns large firms active in several related areas. For each firm we take the main sector in which it is active and reduce the corresponding BIK code to the one-digit level. This way, firms are divided in seven different groups, like finance, transport and communication, industry, and construction. In Table 1 these categories are summarized.

In Table 2 we give some statistics for each of the firms in our panel database.

We report the BIK code of the firm and the average values (over the years) of the untransformed performance measures, the turnover, the board size and the number of interlocks per director. From this table it is clear that the firms differ widely on these features.

In Table 3 we give statistics per year, now averaged over all 101 firms. We report the averages of the untransformed performance measures, the turnover, the board size and the number of interlocks per director. For the number of interlocks per director we exclude all firms that started only after 1994. These firms have somewhat lesser interlocks per director and, consequently, when these firms are included the number of interlocks per director decreases over time. As one can see from the last column of Table 3, the number of interlocks per director does not have a clear upward or downward trend. Interestingly, the size of the board declines over time, also when the firms that start after 1994 are excluded. As expected, we see that the average firm performance is lower during 2001 and 2002 and we also note that the turnover increases over time.

For Tables 2 and 3 we used all available observations for the computations. However, for the regressions below we need to delete all firm-year observations that have either a missing value on one of the performance measures or on turnover or growth of turnover. We also removed some outliers. In Table 4 we report the available sample sizes for the five performance measures. The sample size clearly differs between performance measures. In Table 5 we give some statistics of the (transformed) data, using the samples of Table 4. From Table 5 one can see that the statistics differ only slightly between the different performance measures.

3 Methodology

We have five different financial performance measures and we analyze each of these separately. For each performance measure we estimate the parameters of three different regression models. The (panel data) models we use are a fixed effects (FE) model, a random effects (RE) model and a model based on the BIK codes. We will denote this last model as the BIK model. We use the AIC for model selection as, in contrast to the BIC, this criterion yields plausible final models. This AIC-based selected model is used to draw conclusions.

Models

The FE model is given by

$$y_{it} = \alpha_i + x_{it}\beta + \gamma_t + \varepsilon_{it},\tag{1}$$

with ε_{it} iid and normally distributed, y_{it} the dependent variable, where x_{it} collects the independent variables and γ_t measures developments over time⁴. Note that the constant α_i depends on the firm *i*. The estimator of β is the within-estimator, as it is based on differences over time within a firm, and not on differences between firms.

The RE model is written as

$$y_{it} = \alpha + x_{it}\beta + \gamma_t + \alpha_i + \varepsilon_{it}, \qquad (2)$$

with both ε_{it} and α_i iid and normally distributed. Again, y_{it} is the dependent variable, x_{it} summarizes the independent variables and γ_t concerns time. Note that here α does not depend on the firm *i*. Instead, correlation in the data of the same firm over time is captured by a random variable α_i , which is the same over time for one firm, but potentially differs across firms. The error term $\zeta_{it} = \alpha_i + \varepsilon_{it}$ is not iid and normally distributed as ζ_{it} and $\zeta_{i,t-1}$ are correlated. The resulting estimator of β is based on differences within a firm over time as well as on differences between firms.

For the BIK model, let BIK_i denote the BIK code of firm *i*. The BIK model is given by

$$y_{it} = \alpha_{BIK_it} + x_{it}\beta + \varepsilon_{it},\tag{3}$$

with ε_{it} iid and normally distributed. The idea of including α_{BIK_it} is that firms in the same sector might have the same performance over time. For example, over time the patterns in stock returns might be the same for firms in the same sector. Note that with the BIK model we allow for different patterns over time. For instance, the model allows the performance in sector 1 to increase while at the same time the performance in sector 2 decreases. Note that this is not possible with the time dummies in the FE and RE models.

Variables

As independent variables in the three models we use the variable 'interlock', defined in the previous section, as well as some control variables. These control variables are the same for each of the performance measures. We include turnover, squared turnover, turnover one year lagged, squared turnover one year lagged, growth of turnover (in short: growth), squared growth, growth one year lagged and squared growth one year lagged. With the squared variables we allow for a potential nonlinear effect of the control variables on performance. Note that turnover serves as a measure of the size of the firm.

We estimate the parameters in each of the three models three times. First, we include 'interlock' linearly, next we include 'interlock' quadratically and finally we include the square root of 'interlock'. We do this as previous research in for example Bunting (1976) indicates that the effect of 'interlock' could be nonlinear. As a theoretical explanation for a nonlinear effect one could think of a combination of the 'information on business practices' theory and the busyness hypothesis. In this case, having more interlocks would lead to more information but also to more busy directors. When the number of interlocks is low, directors are not yet timeconstrained and therefore more interlocks would lead to a better performance. On the other hand, when the number of interlocks is already high, adding more interlocks would not lead to much more information while the directors, who are already busy, would get even more time-constrained, leading to worse performance. Hence, the nonlinear relation would then be an inverted U-shape.

As we have a panel database it is also possible to estimate the models while including 'interlock' with a one year lag. This allows us to see if there is a time lag in the effect, that is, whether the effect of interlocks is immediately visible or whether it takes some time before the effect can be noticed. Another advantage of using 'interlock' one year lagged is that it enhances the robustness of the estimates. Indeed, some researchers have expressed concern on the possible reverse causality between the number of interlocks and firm performance. Not only will the number of interlocks influence performance, but good performing firms could attract more interlocking directors. In the database we use, the number of interlocks is based on the directorships halfway the year, while performance is measured over the complete year, and is quite volatile over years. Hence, perverse causality is unlikely to happen and by including lagged interlocks we can even exclude this situation.

4 The main empirical results

We only report the results on the models (FE, RE or BIK) that have the lowest AIC, which means that for the stock returns we only report the results of the BIK model, while for the other performance measures the FE model has the lowest AIC. Table 6 summarizes the AIC values of the models. It is not surprising that the BIK model is favored for stock returns. In the sample there is both a boom (before 2000) and a decline (after 2000) of stock prices. It is well known that stock prices in some sectors lead booms and declines, while others sectors follow. Allowing for different patterns over time for different sectors seems to be reasonable here, and hence the favorable AIC values for the BIK model.

In Tables 7 and 8 we report the estimation results of the method outlined in the previous section. Each panel of the tables has a different performance measure as the dependent variable in the panel model. We report parameter estimates on 'interlock', as well as the relevant p-values, the AIC and for the quadratic models the p-value of an F-test testing the joint significance of the linear and quadratic term. We estimate the FE, RE and BIK model, using a linear, quadratic and square root specification of 'interlock'. Moreover, we estimate the same models using 'interlock' one year lagged. In each model we also include the control variables mentioned in the previous section (estimation results are not reported to save space).

When looking at the results, we note that the effect of the number of interlocks on stock returns (Table 7), the price-earnings ratio (Table 7) and the return on assets (Table 8) is not significant. On the other hand, the effect on the price-tobook ratio (Table 7) and the return on equity (Table 8) is (partly) significant⁵. For the price-to-book ratio, all parameters for current interlocks are significant at the 5% or 10% level and the parameters for lagged interlocks are significant at the 1% level. For return on equity only the lagged variables are relevant. The significant estimates in the linear and square root specifications are all negative. The estimated quadratic specification has a U-shape with minimum at approximately 3.9 interlocks per director for the price-to-book ratio and an inverted U-shape with maximum at - 0.4 interlocks per director for return on equity (lagged specification). As the number of interlocks per director is a positive variable with a mean of approximately 0.65 and a standard deviation around 0.55, the significant quadratic specifications also suggest a negative effect.

In the models above both lagged and current interlocks are used as independent variables. The AIC values in the last columns of Tables 7 and 8 allow us to compare the performance of the contemporaneous specification with the performance of the lagged specification. For the price-to-book ratio and the return on equity the lagged specification gives a better AIC, while for stock returns, the price-earnings ratio and the return on assets the evidence is mixed. We note that the first two performance measures give clear-cut results, especially in the lagged specification.

All this leads us to conclude that *if* there is an effect of interlocks on performance, this effect is negative and it occurs with a lag.

5 What could explain our findings?

Now we have found a small negative effect of the number of interlocks per director on firm performance, we wonder which of the two theories could dominate in predicting a negative influence of interlocks. In this section we take a closer look at these theories.

The busyness hypothesis

First there is the busyness hypothesis of Ferris et al. (2003). Directors who gather many directorships get short of time and the performance of their firms deteriorates. This has already been noticed by Mace (1971) for the US case. Recent evidence for the US is documented by Fich and Shivdasani (2006). In The Netherlands the situation is somewhat different, as the supervisory directors almost all are retired, and a supervisory directorship is not a full-time job. It is however widely acknowledged that having three or four directorships should be the maximum. The recently introduced Dutch corporate governance code (the so-called code Tabaksblat) advises that no director in the board should have more than five directorships, with a chair position counting twice. To investigate whether having busy directors on the board has an influence on performance we construct a variable that measures the 'busyness' of the board and include this variable in the models.

For each director in our database we count his or her number of directorships. In line with the code Tabaksblat, we define a director as being busy when he has more than four directorships. The reason not to use five as a threshold is that the number of directors with more than five directorships is very limited. In addition, we have no information on chair positions and therefore cannot count these positions twice. For each firm and year we then count the number of supervisory directors in the board, and the number of these directors who are not busy (at least, according to our definition). Division of these numbers gives the fraction of non-busy directors in the board. We include this variable in a linear fashion. If the busyness hypothesis would hold true, the effect of the fraction of non-busy directors on firm performance should be positive. Before we examine this hypothesis empirically, we first address the second possible cause of our findings.

Homogeneous upper class

The second hypothesis of why interlocks have a negative effect on performance has to do with (the absence of) diversity in the board and a homogenous upper class of directors. In The Netherlands there seems to be an upper class of directors, who meet each other regularly, either in the boardroom or in all kinds of elite clubs. This has for instance been documented by Van Hezewijk (1986, 1988). The members of this upper class of directors have several directorships, mostly have the same background and as they meet regularly they often exchange opinions. They therefore can be called members of a homogenous group. It has been shown that diversity in the board enhances the performance of the firm. We propose that for a firm it is therefore best to have a mixed board, with some directors belonging to the upper class and some not. If this is indeed true, heavily interlocked firms have mostly upper class directors and so would perform worse than other firms.

To test this upper class hypothesis we define a director as belonging to the upper class when he has three or more directorships. We then calculate for each firm and year the fraction of non-upper class directors in the board. We will include this variable in the model in a quadratic way, to allow for the optimal fraction to be somewhere between 0 and 1, which indicates a positive effect of diversity.

Limitations

We acknowledge that both measures above could have errors. First, a busy director is defined as having more than four directorships in our dataset. We capture 101 large firms in The Netherlands, but there are of course other firms either in The Netherlands or abroad that a director could serve, and we do not count these directorships. Furthermore, some directors are also active in government organizations, like the Dutch central bank. This information is however not available, and so we can only use the 101 firms in our dataset. We however believe that this is a reasonable proxy as for busyness we capture almost all Dutch listed firms, and only very recently there have been some signs of internationalization in the Dutch boardrooms in the sense that Dutch directors get a position in foreign firms and foreign directors get a supervisory directorship in Dutch firms.

According to our upper class measure, a director belongs to the upper class if he or she has more than two directorships. As with the busyness measure we only look at directorships in our 101 firms in the dataset. Also, the threshold of having more than two directorships is not based on previous evidence. Therefore, we also estimated the model using a threshold of more than one directorship and using a threshold of more than three directorships, but it turns out that the model based on a threshold of more than two directorships gives easily interpretable and partly significant results. Hence, we stick to the threshold of more than two directorships.

Estimation results for the busyness hypothesis

We now turn to the estimation results when the ratio of non-busy directors in the board is included in the model. The estimation results are reported in Tables 9 and 10 As in the previous tables without the ratio of non-busy directors we report the estimation results only for the model (FE, RE or bik) that leads to the best AIC values. We again estimate each model using a linear, a quadratic and a square root specification for 'interlock', while using a lagged and a non-lagged specification. The ratio of non-busy directors is always included linearly. When 'interlock' is included in a non-lagged way, the ratio of non-busy directors is included in a non-lagged way, and when 'interlock' is included in a lagged way, the ratio of non-busy directors is always directors is always directors is always directors is included in a non-lagged way.

The first result from Tables 9 and 10 is that almost all estimates have the expected (positive) sign. For stock returns, the return on assets and the return on equity, the ratio of non-busy directors is not significant. Also the AIC including the ratio of non-busy directors is worse compared to when the ratio of non-busy directors is left out of the model. Furthermore, the effect of 'interlock' is not very different from the model without the ratio of non-busy directors.

For the price-earnings and price-to-book ratios the non-lagged specification gives a significant effect of the ratio of non-busy directors. The AIC for the non-lagged specification improves and the estimates on 'interlock' that were significant now get insignificant. The lagged specification however shows no effect of the ratio of nonbusy directors. The estimated parameters are not significant, the AIC worsens and the effect of 'interlock' does not change much. We note that the size of the estimated effect in the non-lagged specification is economically quite significant.

From the results in Tables 9 and 10 we conclude that the ratio of non-busy directors in the board does have an influence on performance when it is measured by the price-earnings and price-to-book ratios. This effect is positive, as expected, but it only occurs in the non-lagged specification. In this non-lagged specification it indeed explains the negative effect of 'interlock'. The significant effects for on 'interlock' now become insignificant. In the lagged specification however the ratio of non-busy directors cannot explain the negative effect of 'interlock'.

Estimation results for upper class hypothesis

Tables 11 and 12 give the estimation results of the model including the measure that indicates the ratio of directors in the board who do not belong to the upper class. As before, we only report results on the model (FE, RE or BIK) that has the best AIC values, we include 'interlock' linearly, quadratically and with a square root, and we use both lagged and non-lagged specifications. The ratio of directors who do not belong to the upper class is included quadratically, as we expect an inverted U-shape. When 'interlock' is not lagged, the ratio of directors who do not belong to the upper class is also non-lagged, and when 'interlock' is lagged the ratio is also lagged.

The stock return, the price-earnings ratio and the price-to-book ratio give insignificant estimates on the ratio of directors who do not belong to the upper class (using an F-test). The AIC gets worse compared to the model without the ratio of directors who do not belong to the upper class and the estimates on 'interlock' do not change much.

For the return on assets and the return on equity the results are different. The ratio of directors who do not belong to the upper class is significant (F-test), the AIC improves and the estimates on 'interlock' that were significant now are not significant anymore. These results are obtained in both the current and the lagged specification. The estimated quadratic function indeed is an inverse U. It peaks at approximately a ratio of 0.6 (current) and 0.8 (lagged) for the return on assets, which implies that it is optimal to have a board which consists of 20% to 40% of upper class directors. For the return on equity the peak is at approximately a ratio of 1 (non-lagged) and 0.9 (lagged), and so the estimated effect is positive almost everywhere. We again note that the estimated effects are quite sizable.

From the results in these tables we conclude that the ratio of directors who are not in the upper class has an effect on the return on assets and the return on equity. This effect takes the form of an inverted U-shape for the return on assets and is positive for the return on equity. This 'upper class effect' replaces the effect of 'interlock'.

6 Conclusion

In this study we explored the effect of interlocks on firm performance for The Netherlands during 1994 to 2004 while using a new and detailed panel database. We find a small negative effect that occurs with a lag. There are two hypotheses that could explain a negative effect, and these are the busyness hypothesis and an upper class cohesion hypothesis. We find empirical support for both hypotheses. Firms with more busy directors on their board perform worse, supporting the busyness hypothesis. For the upper class cohesion hypothesis we find that there is an optimal percentage (20%-40%) of upper class directors in the board. We believe that these findings have strong managerial implications.

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Notes

¹In addition to the four views mentioned here there is an abundance of alternative views on interlocks. As far as we know, all these alternative views predict a positive effect of interlocks on performance.

 2 In the rare case a firm has no supervisory directors (in one or two cases it happened that the entire supervisory board steps down by the end of July) we deleted the observation.

³BIK codes are the Dutch equivalent of SIC codes.

⁴Note that γ_t contains parameters that need to be estimated. Thus we allow for a flexible pattern over time. This is especially important for stock returns, as our database contains both years of boom and years of decline of the stock market.

⁵The non-lagged quadratic specification is significant for the price-earnings ratio. As other nonlagged specifications and also the lagged quadratic specification are clearly insignificant, the effect is not robust and we conclude that the effect of interlocks on the price-earnings ratio is insignificant.



Figure 1: Interlock network in 1998. Dots denote firms and lines denote interlocks between firms.

Table 1: Description of BIK codes. The category names in the first column are used in Table 2.

category	description	number of firms
С	oil and mineral mining	1
D	manufacturers	43
\mathbf{F}	construction	6
G	trade	12
Ι	transport and communication	8
J	financials	12
Κ	provision of services and renting	19

Table 2: Statistics per firm. The first column gives the name of the firm. The second column denotes the BIK code of the firm; see Table 1 for the description of these codes. Columns 3 to 7 give the average performance of each firm over time. 'Stock' denotes the percentage growth of the stock price (stock return), 'p/e ratio' is the price-earnings ratio (stock price divided by earnings per share), 'p/b ratio' is the price-to-book ratio (stock price divided by equity per share), 'roa' denotes the return on assets (in percentages) and 'roe' denotes return on equity (in percentages). The eighth column gives the average turnover in billions of euros. The last two columns give board characteristics: the size of the board (average number of board members) and the average number of interlocks per director.

			p/e	p/b			turn-	board	inter-
firm	BIK	stock	ratio	ratio	roa	roe	over	size	lock
Aalberts	D	24.06	15.38	4.47	10.55	31.88	0.49	3.91	0.72
ABN Amro	J	13.65	11.99	2.42	0.77	20.74	14.90	13.36	1.41
Acomo	G	12.08	9.16	1.96	13.49	25.72	0.19	3.73	0.00
Aegon	J	20.84	22.51	3.14	1.06	14.36	22.51	10.91	1.12
Ahold	G	5.51	0.27	7.34	3.71	16.58	37.88	6.73	1.39
Akzo Nobel	D	6.07	20.04	4.34	10.03	25.83	12.63	9.73	1.02
Alanheri	G	-5.00	53.02	0.75	0.31	1.33	0.13	3.00	0.15
AM	\mathbf{F}	7.07	11.05	1.79	9.45	16.64	1.56	5.36	0.43
Arcadis	Κ	11.22	9.94	1.80	10.03	19.18	0.68	6.55	0.37
ASM International	D	49.82	79.22	6.12	6.05	-3.66	0.50	4.36	0.92
ASML	D	42.11	27.50	8.39	11.51	11.57	1.38	6.30	0.46
Athlon	Κ	20.37	11.80	2.16	2.95	20.53	0.80	3.91	0.55
Ballast Nedam	\mathbf{F}	3.93	7.89	1.08	0.88	2.49	1.87	6.18	0.70
BAM	\mathbf{F}	16.72	5.31	1.52	5.26	19.46	2.74	6.55	0.55
Batenburg beheer	\mathbf{F}	4.38	8.69	1.54	13.85	18.54	0.11	2.91	0.24
Beterbed	G	12.15	13.61	10.34	18.96	52.31	0.20	3.25	0.44
Boskalis	\mathbf{F}	8.78	10.77	1.76	8.54	19.06	0.86	5.45	1.26
Brunel	Κ	-0.76	32.10	2.63	14.69	16.04	0.23	3.25	0.31
Buhrmann	G	0.62	58.29	1.12	2.29	4.57	7.56	7.55	1.56
Ten Cate	D	7.34	5.32	1.07	6.00	7.53	0.62	5.55	0.84
LogicaCMG	Κ	60.93	15.92	14.82	1.03	0.43	1.75	4.22	0.24
Coberco	D	0.20	7.61	NA	6.50	12.94	4.32	11.43	0.21
Corio	J	8.43	NA	1.03	5.19	9.76	0.19	5.91	0.73
Corus	D	-16.66	2.39	0.69	5.61	8.10	9.09	8.64	1.27
Crown van Gelder	D	2.29	2.25	0.59	9.41	7.49	0.13	4.00	0.05
Crucell	D	24.42	-7.59	2.08	-29.37	-39.54	0.01	6.00	0.00

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	DIII		p/e	p/b			turn-	board	inter-
Firm	BIK	stock	ratio	ratio	roa	roe	over	size	lock
CSM	D	4.98	14.15	4.59	13.06	41.37	2.57	6.36	1.19
CTAC	Κ	-11.12	22.64	5.91	17.00	33.80	0.02	2.71	0.43
Delft Instruments	D	5.89	11.83	1.43	4.68	4.61	0.20	5.27	0.53
Dico international	D	-11.11	1.65	0.58	-13.70	-25.55	0.03	2.82	0.33
DOCdata	D	-9.25	11.60	1.35	4.39	5.35	0.09	3.13	0.44
Draka	D	10.82	18.27	2.22	6.71	17.65	1.20	5.55	1.08
DSM	D	9.87	10.34	1.09	7.68	17.17	6.49	7.82	0.57
Econosto	D	-10.28	21.83	2.50	1.60	-4.14	0.29	4.00	0.11
Elsevier	D	3.76	-98.96	5.21	13.62	11.35	5.90	7.55	0.99
Eriks group	G	6.63	24.15	1.57	12.52	17.34	0.30	3.00	0.00
Eurocommercial									
properties	J	8.19	15.38	1.00	3.57	6.35	0.07	4.00	0.00
EVC International	D	5.52	-2.09	0.49	-2.32	-18.10	1.06	4.10	0.13
Exact	Κ	-4.23	25.57	8.70	25.19	35.42	0.18	4.29	0.12
Exendis	D	-5.12	9.19	2.25	5.41	2.27	0.02	3.36	0.30
de Porceleyne Fles	D	8.18	10.00	1.73	5.85	13.02	0.00	3.18	0.24
Fornix	D	37.76	6.72	5.19	16.98	24.65	0.04	3.18	0.00
Fortis	J	13.95	-147.55	2.81	9.09	11.55	1.79	10.80	0.16
Frans Maas	Ι	2.11	9.38	1.67	5.74	12.00	0.82	4.55	0.50
Fugro	Κ	27.57	13.24	3.60	10.68	32.41	0.67	5.73	0.61
Gamma Holding	D	0.78	9.85	1.36	8.73	15.33	0.82	6.09	1.49
Getronics	Κ	32.28	28.15	10.20	5.24	13.67	2.52	4.82	0.67
Grolsch	D	0.30	16.69	2.96	14.87	17.61	0.24	6.00	1.07
Grontmij	Κ	10.20	13.13	1.23	6.04	13.56	0.39	6.18	0.38
Hagemeijer	G	-9.37	16.12	3.43	4.90	11.60	6.17	5.91	0.85
Heijmans	\mathbf{F}	19.30	9.90	1.93	9.32	21.85	1.61	4.82	0.79
Heineken	D	7.18	23.70	4.83	12.38	22.02	6.85	7.36	1.59
Hunter Douglas	D	9.87	10.27	1.93	12.59	19.38	1.44	7.00	0.20
Imtech	Κ	14.87	9.23	2.39	9.40	34.26	2.10	6.18	1.41
ING	J	16.31	11.88	1.82	NA	18.37	9.29	11.45	1.06
Kas bank	J	15.01	10.36	1.17	NA	14.17	0.10	7.09	0.06
Kendrion	D	-11.25	6.98	3.42	2.72	-3.68	0.73	3.64	0.17
KLM	T	1.17	70.17	0.74	0.32	6.70	6.10	9.36	0.95
KPN	Ī	7.27	14.54	2.68	1.73	-3.37	10.73	7.55	0.92
Van Lansschot	Ī	4 81	13 43	$\frac{2.00}{2.27}$	NA	18.21	0.35	8 43	0.74
Macintosh Retail	Ğ	6.55	16.26	1.39	7.74	13.48	0.69	5.00	0.62
Magnus	K	-22.05	35.62	5.39	5.35	4.89	0.03	2.33	0.00
Van der Molen	.I	22.55	6 63	4 11	10.09	62.13	0.20	<u>-</u> .55 4 64	0.18
Nedan	Ď	20.58	16.38	3 36	16.63	21.36	0.10	3.91	0.26
Nedschroef	D	16.11	5.62	0.94	7.32	10.77	0.25	5.00	0.18
Neways	D	6 77	7.63	1 79	3 46	-2.15	0.20	2.60	0.06
itteways	D	0.11	1.00	1.13	0.40	2.10	0.10	2.04	0.00

Table 2 –	continued	from	previous	page

			p/e	p/b		P - 0 -	turn-	board	inter-
Firm	BIK	stock	ratio	ratio	roa	roe	over	size	lock
Numico	D	25.23	20.88	7.44	1.05	48.25	2.52	6.18	1.05
Nutreco	D	9.21	12.31	2.09	4.55	11.78	2.95	3.73	0.46
Nvloplast	D	-6.27	23.39	1.44	8.70	10.61	0.02	3.09	0.11
OCE	D	8.52	27.58	1.93	5.42	11.50	2.64	5.82	1.10
Opg groep	D	9.97	9.85	1.94	10.57	24.73	1.65	7.00	0.23
Ordina	Κ	31.21	22.45	18.20	29.11	83.72	0.27	3.27	0.52
P&O Nedlloyd	Ι	19.26	18.34	0.56	-0.26	6.78	2.37	5.73	1.44
Philips	D	22.11	3.74	2.05	4.40	14.58	32.29	7.82	0.68
Randstad	Κ	27.69	25.28	8.85	13.44	38.44	4.63	5.73	0.99
Reesink	G	-0.27	10.38	1.13	9.23	10.51	0.12	4.00	0.41
De Vries Robbe	Κ	-12.16	9.02	0.97	-30.97	-57.34	0.01	2.43	0.07
Rodamco Europe	J	8.32	12.69	0.96	5.20	13.12	0.47	6.40	0.28
SBM	D	14.85	24.17	3.50	6.09	17.06	1.16	5.55	1.65
Schuitema	G	16.30	16.76	4.55	11.54	28.01	2.25	4.45	0.27
Shell	\mathbf{C}	9.73	51.07	2.81	14.45	15.36	130.77	7.27	1.85
Simac	Κ	31.67	30.42	5.12	-3.90	-33.07	0.19	3.82	0.08
Sligro food group	G	24.79	15.27	4.54	13.80	32.92	0.88	4.00	0.05
Smit International	Ι	17.82	8.64	1.09	5.22	18.31	0.32	5.64	0.87
Nieuwe Steen									
investments	J	3.58	11.27	1.18	5.22	11.36	0.06	4.29	0.00
Stork	D	14.83	3.66	1.29	5.36	10.84	2.19	7.00	1.02
Telegraaf	D	6.07	-8.25	2.23	8.35	6.53	0.68	6.27	0.32
TNT	Ι	-3.36	23.99	4.57	10.91	22.86	10.48	8.57	0.82
Twentsche kabel	D	14.53	9.70	1.44	8.78	10.45	0.45	5.91	0.73
Unilever	D	10.66	26.93	6.15	10.81	35.58	43.06	10.00	0.43
Unit 4 Agresso	Κ	17.65	23.19	12.61	25.57	69.78	0.16	3.43	0.00
Univar	Ι	126.57	11.30	0.72	3.38	5.87	3.93	4.33	1.15
United Services	Κ	33.15	15.87	6.78	20.21	55.63	0.81	4.29	0.93
Vedior	Κ	9.43	-8.21	2.09	-2.60	-2.54	5.44	3.88	0.33
Vendex	G	-0.83	31.72	3.32	9.21	28.89	4.52	6.00	0.82
Versatel	Ι	57.13	-13.96	1.37	-34.02	-30.34	0.31	5.67	0.00
VHS onroerend goed	Κ	16.23	8.32	1.65	4.54	19.88	0.04	3.00	0.00
VNU	D	15.65	28.05	4.98	8.40	23.53	3.01	6.18	1.46
Wereldhave	J	7.08	11.81	0.99	5.39	9.35	0.14	4.45	0.46
Wessanen	D	-2.21	53.32	1.77	6.75	16.21	2.78	5.18	0.89
Wolters-Kluwer	D	4.39	22.24	7.88	9.70	24.28	3.01	6.36	1.35

Table 3: Statistics per year. The first column gives the year. Columns 2 to 6 give the average performance in each year. 'Stock' denotes the percentage growth of the stock price (stock return), 'p/e ratio' is the price-earnings ratio (stock price divided by earnings per share), 'p/b ratio' is the price-to-book ratio (stock price divided by equity per share), 'roa' denotes the return on assets (in percentages) and 'roe' denotes return on equity (in percentages). The seventh column gives the average turnover in billions of euros. The last two columns give board characteristics: the size of the board (average number of board members) and the average number of interlocks per director.

year	stock	p/e ratio	p/b ratio	roa	roe	turnover	board size	interlock
1994	2.67	11.11	1.53	6.05	7.99	NA	6.18	0.69
1995	7.26	14.14	2.83	11.32	12.18	3.06	5.98	0.70
1996	45.38	17.54	3.61	9.90	22.42	3.34	5.77	0.66
1997	27.56	22.03	4.62	9.83	28.83	3.85	5.64	0.67
1998	1.02	25.82	5.56	10.27	26.09	3.57	5.63	0.79
1999	16.66	29.15	4.52	8.19	21.91	4.18	5.46	0.77
2000	1.65	25.65	3.50	5.91	18.59	5.52	5.59	0.73
2001	-13.97	20.86	2.40	3.73	6.05	5.61	5.60	0.72
2002	-27.72	12.23	1.69	1.12	-3.38	5.66	5.60	0.72
2003	36.32	19.32	2.00	2.84	8.97	5.42	5.49	0.69
2004	21.95	28.54	2.16	5.87	9.26	5.82	5.36	0.60

Table 4: Sample sizes for the five performance measures.

stock return	734
price-earnings ratio	616
price-to-book ratio	722
return on assets	716
return on equity	734

Table 5: Summary statistics for each of the performance measure samples, where the sample sizes are given in Table 4. Each column corresponds to the sample of one of the (transformed) performance measures: the log of (1+ stock return/100), the logs of the price-earnings and price-to-book ratios, the return on assets (in percentages, no transformation applied) and the return on equity (in percentages, no transformation applied). The first panel gives the mean, the standard deviation (s.d.) and the median of the (transformed) performance measures itself. The second panel gives the same statistics on the number of interlocks per director. Panel three gives the same statistics on the percentage growth of the turnover. Note that the columns of panels two, three and four thus concern different sample sizes.

		stock return	p/e ratio	p/b ratio	roa	roe
	mean	-0.018	2.667	0.731	6.047	14.379
dependent	s.d.	0.438	0.794	0.859	13.361	31.166
	median	0.001	2.550	0.642	7.225	15.735
	mean	0.638	0.668	0.638	0.638	0.638
interlock	s.d.	0.552	0.560	0.554	0.555	0.551
	median	0.500	0.571	0.536	0.500	0.571
	mean	5.347	5.665	5.295	5.420	5.400
turnover	s.d.	17.140	18.189	17.192	17.336	17.143
	median	0.849	0.917	0.821	0.903	0.900
growth turnover	mean	10.527	11.927	10.925	10.427	10.614
	s.d.	33.980	28.160	33.899	34.267	33.709
	median	5.950	8.060	6.210	5.850	5.940

Table 6: AIC's of the different models. Each combination of performance measure
and panel data model (FE, RE, BIK) is estimated in six different forms, that is, with
interlocks incorporated linearly, quadratically and in a square root and both lagged
and not lagged. This table gives for each combination of performance measure and
panel data model the lowest and highest obtained AIC's of the six different forms.

model		lowest AIC	highest AIC
	\mathbf{FE}	1.041	1.052
growth of stock price	RE	1.177	1.206
	BIK	0.895	0.898
	\mathbf{FE}	2.057	2.068
price-earnings ratio	RE	2.123	2.135
	BIK	2.402	2.403
	\mathbf{FE}	1.265	1.295
price to book ratio	RE	1.515	1.542
	BIK	2.299	2.310
	\mathbf{FE}	7.474	7.478
return on assets	RE	7.532	7.537
	BIK	7.531	7.547
	\mathbf{FE}	9.428	9.440
return on equity	RE	9.458	9.470
	BIK	9.639	9.650

Table 7: Estimation results on the performance measures $\log(\text{stock return}/100 + 1)$, $\log(\text{price-earnings ratio})$ and $\log(\text{price-to-book ratio})$, p-value in parentheses. The column 'model' denotes which model is used (FE, RE or BIK model), which transformation of interlock is used (linear, quadratic or square root) and whether interlock is lagged. The column 'F-test' gives the p-value of the F-test of joint significance of interlock and interlock-squared.

measure	model	interlock	$interlock^2$	F-test	AIC
stock	BIK, linear	-0.004 (0.888)	-	-	0.896
	BIK, quadr.	-0.007(0.930)	$0.002 \ (0.968)$	0.989	0.898
	BIK, sq.root	$0.000\ (0.999)$	-	-	0.896
	BIK, linear, lag	-0.009(0.757)	-	-	0.896
	BIK, quadr., lag	-0.089(0.268)	$0.048 \ (0.285)$	0.538	0.897
	BIK, sq.root, lag	-0.016(0.653)	-	-	0.895
p/e ratio	FE, linear	-0.134 (0.200)	-	-	2.064
	FE, quadr.	$0.353\ (0.137)$	-0.268(0.023)	0.033	2.057
	FE, sq. root	-0.028(0.839)	-	-	2.067
	FE, linear, lag	$0.038\ (0.719)$	-	-	2.067
	FE, quadr., lag	$0.285\ (0.227)$	-0.134(0.241)	0.471	2.068
	FE, sq.root, lag	$0.156\ (0.257)$	-	-	2.065
p/b ratio	FE, linear	-0.158 (0.016)	-	-	1.292
	FE, quadr.	-0.200(0.169)	$0.025\ (0.743)$	0.053	1.295
	FE, sq. root	-0.201(0.014)	-	-	1.292
	FE, linear, lag	-0.311 (0.000)	-	-	1.265
	FE, quadr., lag	-0.401 (0.004)	$0.053 \ (0.466)$	0.000	1.267
	FE, sq.root, lag	-0.351(0.000)	-	-	1.270

Table 8: Estimation results on the performance measures return on assets and return on equity, p-value in parentheses. The column 'model' denotes which model is used (FE, RE or BIK model), which transformation of interlock is used (linear, quadratic or square root) and whether interlock is lagged. The column 'F-test' gives the p-value of the F-test of joint significance of interlock and interlock-squared.

measure	model	interlock	$interlock^2$	F-test	AIC
roa	FE, linear	-1.035(0.465)	-	-	7.476
	FE, quadr.	-0.403(0.899)	-0.365(0.824)	0.747	7.478
	FE, sq. root	-1.410(0.428)	-	-	7.475
	FE, linear, lag	-1.033(0.465)	-	-	7.476
	FE, quadr., lag	3.450(0.265)	-2.572(0.104)	0.203	7.474
	FE, sq.root, lag	$0.544\ (0.755)$	-	-	7.476
roe	FE, linear	-4.193(0.268)	-	-	9.439
	FE, quadr.	2.797(0.741)	-4.033(0.356)	0.353	9.440
	FE, sq. root	-2.294(0.629)	-	-	9.440
	FE, linear, lag	-10.483(0.005)	-	-	9.428
	FE, quadr., lag	-3.286(0.689)	-4.145(0.325)	0.013	9.429
	FE, sq.root, lag	-9.071(0.051)	-	-	9.435

Table 9: Estimation results including a 'busyness' parameter for performance measures $\log(\operatorname{stock} \operatorname{return}/100 + 1)$, $\log(\operatorname{price-earnings ratio})$ and $\log(\operatorname{price-to-book ratio})$, p-value in parentheses. The column 'model' denotes which model is used (FE, RE or BIK model), which transformation of interlock is used (linear, quadratic or square root) and whether interlock is lagged. The column 'F-test' gives the p-value of the F-test of joint significance of interlock and interlock-squared.

measure	model	interlock	$interlock^2$	F-test	ratio non-busy	AIC
stock	BIK, linear	-0.011(0.760)	-	-	0.192(0.488)	0.898
	BIK, quadr.	-0.021(0.794)	$0.023\ (0.662)$	0.868	$0.261 \ (0.414)$	0.900
	BIK, sq.root	$0.013\ (0.742)$	-	-	$0.179\ (0.473)$	0.898
	BIK, linear, lag	-0.017(0.641)	-	-	-0.096(0.718)	0.898
	BIK, quadr., lag	-0.091(0.263)	$0.052\ (0.308)$	0.533	$0.051 \ (0.866)$	0.899
	BIK, sq.root, lag	-0.023(0.577)	-	-	-0.082(0.731)	0.898
p/e ratio	FE, linear	-0.011 (0.923)	-	-	1.594(0.030)	2.058
	FE, quadr.	0.334(0.159)	-0.208(0.094)	0.244	1.185(0.124)	2.056
	FE, sq.root	$0.101 \ (0.486)$	-	-	1.790(0.009)	2.057
	FE, linear, lag	0.094(0.427)	-	-	0.747(0.294)	2.068
	FE, quadr., lag	$0.270\ (0.253)$	$0.105\ (0.390)$	0.504	0.520(0.494)	2.070
	FE, sq.root, lag	0.208(0.149)	-	-	0.789(0.237)	2.065
p/b ratio	FE, linear	-0.100(0.178)	-	-	0.770(0.107)	1.291
	FE, quadr.	-0.210(0.149)	0.069(0.380)	0.275	0.909(0.071)	1.292
	FE, sq.root	-0.147(0.089)	-	-	0.827(0.063)	1.289
	FE, linear, lag	-0.283 (0.000)	-	-	0.380(0.403)	1.266
	FE, quadr., lag	-0.412 (0.004)	0.083(0.281)	0.000	0.557(0.249)	1.267
	FE, sq.root, lag	-0.309 (0.000)	-	-	$0.691 \ (0.105)$	1.269

Table 10: Estimation results including a 'busyness' parameter for performance measures return on assets and return on equity, p-value in parentheses. The column 'model' denotes which model is used (FE, RE or BIK model), which transformation of interlock is used (linear, quadratic or square root) and whether interlock is lagged. The column 'F-test' gives the p-value of the F-test of joint significance of interlock and interlock-squared.

measure	model	interlock	$interlock^2$	F-test	ratio non-busy	AIC
roa	FE, linear	-0.494(0.763)	-	-	$6.850\ (0.508)$	7.478
	FE, quadr.	-0.472(0.882)	-0.014 (0.994)	0.956	$6.821 \ (0.534)$	7.480
	FE, sq.root	-0.947(0.617)	-	-	$6.760\ (0.479)$	7.477
	FE, linear, lag	$-0.214\ (0.895)$	-	-	$10.340\ (0.301)$	7.477
	FE, quadr., lag	3.354(0.278)	-2.279(0.176)	0.397	5.362(0.614)	7.476
	FE, sq.root, lag	1.450(0.434)	-	-	$13.426\ (0.147)$	7.476
roe	FE, linear	-1.963(0.652)	-	-	28.150(0.303)	9.440
	FE, quadr.	2.565(0.762)	-2.879(0.533)	0.744	22.330(0.439)	9.442
	FE, sq.root	$0.062\ (0.990)$	-	-	$34.356\ (0.174)$	9.440
	FE, linear, lag	-9.069(0.035)	-	-	17.785(0.498)	9.430
	FE, quadr., lag	-3.460(0.674)	-3.585(0.425)	0.079	$10.036\ (0.720)$	9.432
	FE, sq.root, lag	-6.825(0.166)	-	-	$33.391\ (0.171)$	9.434

Table 11: Estimation results including an upper class parameter for performance measures $\log(\text{stock return}/100 + 1)$, $\log(\text{price-earnings ratio})$ and $\log(\text{price-to-book ratio})$, p-value in parentheses. The column 'model' denotes which model is used (FE, RE or BIK model), which transformation of interlock is used (linear, quadratic or square root) and whether interlock is lagged. The first column 'F-test' gives the p-value of the F-test of joint significance of 'interlock' and 'interlock-squared', the second column 'F-test' does the same for the ratio non-upper class.

measure	model	interlock	$interlock^2$	F-test	ratio non-upper class	$ratio^2$	F-test	AIC
stock	BIK, linear	$0.039\ (0.503)$	-	-	0.718(0.204)	0.390(0.299)	0.370	0.898
	BIK, quadr.	-0.068(0.590)	$0.059\ (0.337)$	0.504	$1.195\ (0.112)$	-0.729(0.157)	0.233	0.899
	BIK, sq.root	$0.030 \ (0.652)$	-	-	$0.615 \ (0.276)$	-0.349(0.381)	0.414	0.899
	BIK, linear, lag	0.003(0.953)	-	-	0.405(0.494)	-0.244(0.535)	0.784	0.900
	BIK, quadr., lag	-0.215(0.086)	0.119(0.050)	0.146	1.372(0.075)	-0.935(0.076)	0.202	0.897
	BIK, sq.root, lag	-0.028(0.668)	_	-	0.445(0.456)	-0.310(0.459)	0.755	0.900
p/e ratio	FE, linear	-0.146(0.394)	-	-	1.701(0.202)	-1.217 (0.171)	0.390	2.067
	FE, quadr.	0.393(0.250)	-0.292(0.068)	0.131	-0.392(0.823)	0.267(0.824)	0.975	2.063
	FE, sq.root	$0.056\ (0.775)$	_	-	1.944(0.135)	-1.130 (0.221)	0.242	2.068
	FE, linear, lag	$0.057 \ (0.740)$	-	-	-0.129(0.928)	0.134(0.887)	0.979	2.074
	FE, quadr., lag	0.526(0.106)	-0.259(0.090)	0.225	-2.144(0.249)	1.538(0.219)	0.462	2.071
	FE, sq.root, lag	$0.266\ (0.157)$	_	-	-0.374(0.785)	0.468(0.627)	0.674	2.070
p/b ratio	FE, linear	-0.211 (0.049)	_	-	-0.007(0.993)	-0.113 (0.844)	0.807	1.297
	FE, quadr.	-0.317(0.125)	$0.061 \ (0.548)$	0.120	$0.440 \ (0.700)$	-0.424(0.584)	0.712	1.299
	FE, sq.root	-0.235(0.041)	-	-	$0.536\ (0.527)$	-0.418(0.486)	0.775	1.296
	FE, linear, lag	-0.304(0.004)	-	-	-0.953(0.291)	0.676(0.255)	0.521	1.268
	FE, quadr., lag	-0.302 (0.119)	-0.001(0.989)	0.015	-0.964(0.409)	0.683(0.383)	0.680	1.271
	FE, sq.root, lag	-0.256(0.019)	-	-	-0.120(0.891)	$0.280\ (0.650)$	0.441	1.273

Table 12: Estimation results including an upper class parameter for performance measures return on assets and return on equity, p-value in parentheses. The column 'model' denotes which model is used (FE, RE or BIK model), which transformation of interlock is used (linear, quadratic or square root) and whether interlock is lagged. The first column 'F-test' gives the p-value of the F-test of joint significance of 'interlock' and 'interlock-squared', the second column 'F-test' does the same for the ratio non-upper class.

measure	model	interlock	$interlock^2$	F-test	ratio non-upper class	$ratio^2$	F-test	AIC
roa	FE, linear	-3.122(0.179)	-	-	$29.227 \ (0.117)$	-24.907(0.045)	0.078	7.473
	FE, quadr.	-8.914(0.048)	$3.318\ (0.133)$	0.131	$53.671\ (0.030)$	-41.948(0.013)	0.026	7.472
	FE, sq.root	-4.437(0.076)	-	-	$38.240\ (0.037)$	-31.015(0.017)	0.042	7.470
	FE, linear, lag	$0.854 \ (0.708)$	-	-	$49.345\ (0.012)$	-30.007(0.021)	0.042	7.471
	FE, quadr., lag	1.388(0.745)	-0.309(0.882)	0.922	$46.941 \ (0.066)$	-28.343(0.099)	0.155	7.473
	FE, sq.root, lag	2.132(0.375)	-	-	$45.911 \ (0.016)$	-26.773(0.046)	0.024	7.469
roe	FE, linear	4.681(0.449)	-	-	99.999 (0.045)	-49.529(0.135)	0.057	9.435
	FE, quadr.	3.131(0.794)	$0.888 \ (0.880)$	0.742	106.538(0.107)	-54.088(0.229)	0.086	9.438
	FE, sq.root	5.136(0.440)	-	-	$87.708\ (0.072)$	-42.665(0.217)	0.034	9.435
	FE, linear, lag	-2.819(0.642)	-	-	$133.176\ (0.012)$	-74.769(0.031)	0.029	9.422
	FE, quadr., lag	-7.977(0.484)	$2.973\ (0.593)$	0.778	$156.380\ (0.022)$	-90.876(0.048)	0.041	9.424
	FE, sq.root, lag	-2.602(0.683)	-	-	$141.306\ (0.006)$	-78.915(0.028)	0.004	9.422