

Great moderation at the firm level? Unconditional versus conditional output volatility

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Abstract

Aggregated output in industrialized countries has become less volatile over the past decades. Whether this “Great Moderation” can be found in firm level data as well remains disputed. We study the evolution of firm level output volatility using a balanced panel dataset on German firms that covers 35 years (1971-2005) and about 1,500 firms per year. In contrast to earlier work using firm level data, we use the multifactor residual model proposed by Pesaran (2006) to isolate the idiosyncratic component of firms’ real sales growth from macroeconomic developments. Our paper has three main findings. First, time trends in unconditional firm level and aggregated output volatility in Germany are similar. There has been a long-run downward trend, which was interrupted by the unification period. Second, the conditional, idiosyncratic firm level volatility does not exhibit a downward trend. If anything idiosyncratic volatility has been on a slow trend rise. Third, we find evidence of a positive link between growth and volatility at the firm level.

Keywords: firm level volatility, Great Moderation, multifactor residual model

JEL-Classification: E32, D21

Non-technical summary

Aggregated output has become less volatile over the past decades in industrialized countries. While this stylized fact is widely accepted, the causes for the “Great Moderation” are still the subject of lively debate. The jury is still out on the question whether “Good Policy”, in particular a more stabilizing monetary policy, or “Good Luck”, i.e. the absence of major shocks, is the main cause of the Great Moderation.

Yet, for individual households, it is the development of volatility at the firm level rather than the development of aggregate volatility that matters. If firm level volatility increases, households must find ways to diversify firm level risks and to shield their consumption against income fluctuation. Possibilities for diversification, in turn, depend on the relative importance of macroeconomic and idiosyncratic factors affecting volatility at the firm level. Idiosyncratic volatility can be diversified in developed financial markets. Macroeconomic factors, in contrast, affect all firms in a similar manner and can be diversified nationally only to the extent that some firms react differently to the same macroeconomic shock.

In this paper, we thus test whether evidence of a Great Moderation can be found at the firm level. Earlier evidence, which mostly uses US data, has partly conflicting results. In fact, the link between aggregated and firm level volatility is not clear a priori. On the one hand, developments of growth volatility at the aggregate and the firm level might differ if output changes are imperfectly correlated across firms and if these correlations change over time. On the other hand, aggregate growth rates of GDP can be seen as the averaged growth rates across all firms in an economy. This makes differences in the time series patterns of firm level and aggregated volatility unlikely. Instead, volatility at different levels of aggregation moves into the same direction if the shares of firms in aggregated output and the responses of firms to macroeconomic developments remain relatively stable.

In this paper, we go beyond earlier studies by using a comprehensive panel dataset for German firms. These data allow tracking individual firms for a maximum of 35 years (1971-2005). Our data come from the Corporate Balance Sheet Statistics and the Financial Statements Data Pool provided by the Deutsche Bundesbank.

Methodologically, we go beyond earlier literature in two regards. First, we use the multifactor residual model proposed by Pesaran (2006) to isolate the idiosyncratic growth of firms' sales from macroeconomic factors. We control for observed and unobserved macroeconomic developments affecting all firms while allowing for a heterogeneous response of firms to these factors. We find that trends in firm level and aggregate volatility are similar when looking at the unconditional volatility at the firm level. Controlling for macroeconomic developments, firm level volatility is not only lower but also exhibits no downward trend. If anything, there even is some evidence of a very slow increase in idiosyncratic firm level volatility. Second, we employ panel regressions to analyze the determinants of firm level growth volatility. Results show that higher asset growth and lower inventory-sales ratios are associated with a higher volatility of firms' sales.

Nicht-technische Zusammenfassung

Schwankungen des aggregierten Output haben in entwickelten Volkswirtschaften über die vergangenen Jahrzehnte hinweg abgenommen. Die Ursachen für diese „Great Moderation“ werden in der Literatur strittig diskutiert. Sowohl eine bessere Geldpolitik als auch geringere makroökonomische Schocks und strukturelle Veränderungen werden für diese Entwicklung verantwortlich gemacht.

Für individuelle Haushalte ist hingegen eher die Entwicklung auf Firmen- als auf Makroebene relevant. Steigt die Volatilität auf Firmenebene, müssen die Haushalte Wege finden, diese Volatilität zu diversifizieren, um ihren Konsum gegen Schwankungen des Einkommen abzusichern. Inwiefern das möglich ist, hängt wiederum von den Bestimmungsründen der einzelwirtschaftlichen Volatilität ab. Stehen makroökonomische Entwicklungen hinter den Schwankungen auf Firmenebene, kann eine Absicherung von Risiken im nationalen Rahmen nur begrenzt gelingen. Sind idiosynkratische, firmenspezifische Schocks ausschlaggebend, besteht hingegen die Möglichkeit zur Diversifizierung auf nationalen Finanzmärkten.

In diesem Papier untersuchen wir, inwiefern auf Firmenebene ein Rückgang der Volatilität zu verzeichnen ist. Frühere Evidenz, die sich vorwiegend auf die USA bezieht, kommt hier zu widersprüchlichen Aussagen: Grundsätzlich ist der Zusammenhang zwischen der Volatilität auf aggregierter und auf Firmenebene nicht eindeutig. Einerseits können sich die Entwicklungen unterscheiden, wenn die Wachstumsraten auf Firmenebene unvollständig miteinander korreliert sind und sich diese Korrelationen über die Zeit ändern. Andererseits kann das Wachstum auf aggregierter Ebene als der gewichtete Durchschnitt der Wachstumsraten auf Firmenebene aufgefasst werden. Bleiben die Anteile einzelner Firmen über die Zeit hinweg konstant und reagieren die Firmen ähnlich auf makroökonomische Entwicklungen, sollte sich somit die Entwicklung der Volatilität auf unterschiedlichen Aggregationsebenen nicht wesentlich unterscheiden.

In diesem Papier nutzen wir Daten für eine repräsentative Gruppe deutscher Firmen. Unsere Daten erlauben es, eine Gruppe von rund 1.500 Unternehmen über einen Zeitraum von 35 Jahren (1971-2005) kontinuierlich zu verfolgen. Unsere Daten sind der

Unternehmensbilanzstatistik und dem Jahresabschlussdatenpool der Deutschen Bundesbank entnommen.

In methodischer Hinsicht gehen wir über frühere Arbeiten hinaus, indem wir das Multifaktormodell von Pesaran (2006) nutzen, um die idiosynkratische Komponente des Wachstums auf Firmenebene von makroökonomischen Faktoren zu isolieren. Wir berücksichtigen beobachtete und unbeobachtete makroökonomische Faktoren und wir lassen eine heterogene Reaktion der Unternehmen auf diese Faktoren zu. Unsere Ergebnisse zeigen, dass Trends in der firmenspezifischen und der aggregierten Volatilität ähnlich sind, wenn nicht um makroökonomische Faktoren bereinigt wird. Nachdem wir makroökonomische Faktoren herausgerechnet haben, findet sich allerdings kein starker Trend mehr in der firmen-spezifischen Volatilität: wenn überhaupt, hat diese über die Zeit hinweg leicht zugenommen. Gleichzeitig zeigen diese Ergebnisse, dass makroökonomische Faktoren einen starken Einfluss auf die Entwicklungen auf Firmenebene haben. In einem zweiten Schritt untersuchen wir die Determinanten der firmenspezifischen Volatilität. Höheres Wachstum auf Unternehmensebene und geringere Lagerhaltung gehen diesen Berechnungen zufolge mit einer stärkeren Volatilität einher.

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Great Moderation at the Firm Level? Unconditional versus Conditional Output Volatility*

1 Introduction

Aggregated output has become less volatile over the past decades in industrialized countries (see, among others, Blanchard and Simon 2001, Stock and Watson 2003). While this stylized fact is widely accepted, the causes for the “Great Moderation” are still the subject of lively debate. The jury is still out on the question whether “Good Policy”, in particular a more stabilizing monetary policy, or “Good Luck”, i.e. the absence of major shocks, is the main cause of the Great Moderation. (For recent contributions, see Benati 2007, Benati and Surico 2006, Canova 2006, or Giannone et al. 2007.)

Yet, for individual households, it is the development of volatility at the firm level rather than the development of aggregate volatility that matters. If firm level volatility increases, households must find ways to diversify firm level risks and to shield their consumption against income fluctuation. Possibilities for diversification, in turn, depend on the relative importance of macroeconomic and idiosyncratic factors affecting volatility at the firm level. Idiosyncratic volatility can be diversified in developed financial markets. Macroeconomic factors, in contrast, affect all firms in a similar manner and can be diversified nationally only to the extent that some firms react differently to the same macroeconomic shock. In this paper, we thus test whether evidence of a Great Moderation can be found at the firm level.

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To date, there is little consensus reached how firm level output volatility has evolved over time, let alone the causes for changes in this volatility. For the USA, Comin and Philippon (2005) were the first to claim diverging patterns in firm level and aggregated output volatility. According to their results, firm level output volatility has increased, whereas aggregated output volatility has decreased (see also Comin and Mulani 2006). Comin et al. (2006) confirm these general patterns in the data and show that output and employment volatility display similar trends. Yet Davis et al. (2006) demonstrate that the finding of an increase in firm level output volatility crucially depends on the sample of firms chosen. According to their results, the increase in firm level volatility is a feature of large, publicly traded firms. Using information on privately-owned firms as well, Davis et al. (2006) find a downward trend in firm level volatility for the US, mirroring the macro level development.

Evidence on firm level volatility for non-US countries is scarce (Davis and Kahn 2007). For France, Thesmar and Thoenig (2004) show an upward trend in firm level volatility. For Germany, patterns at the firm level are similar to those found in aggregated data, and there is no evidence of an increase in volatility (Buch et al. 2006).¹

In this paper, we go beyond earlier studies by using a comprehensive panel dataset for German firms. These data allow tracking individual firms for a maximum of 35 years (1971-2005). Our data come from the Financial Statements Data Pool and the Corporate Balance Sheet Statistics of German firms provided by the Deutsche Bundesbank.

Methodologically, we depart from earlier literature in two regards. First, we use the multifactor residual model proposed by Pesaran (2006) to isolate the idiosyncratic growth of firms' sales from macroeconomic factors. We control for observed and unobserved macroeconomic developments affecting all firms while allowing for a heterogeneous response of firms to these factors. Our results show that differences between the unconditional volatility of firms' sales growth and the idiosyncratic, conditional volatility are quite substantial. Without distinguishing the micro- and the macro-component, inference about trends in firm level versus aggregate volatility might thus be misleading. Second, we employ fixed effects panel regressions as well as the

¹ For evidence on aggregated volatility in Germany see Aßmann et al. (2006) and Buch et al. (2004).

heteroskedastic regression model proposed by Harvey (1976) to analyze the determinants of firm level growth volatility. This method has the advantage that we can simultaneously estimate the determinants of the mean and the variance of firms' sales growth. Moreover, our results do not rely on arbitrary choices of choosing a time window for the computation of (rolling) volatilities.

Why would one expect firm level and aggregated volatility to differ? After all, aggregate growth rates of GDP can be seen as the averaged growth rates across all firms in an economy. Yet, this does not automatically imply that the same trends as in the volatility of GDP growth can be found in firm level data. Instead, developments of growth volatility at the aggregate level and the firm level might differ if output changes are imperfectly correlated across firms (Comin and Phillipon 2005). Patterns of correlation, in turn, may change due to differences in the process of deregulation across sectors, differences in R&D intensities, or different degrees of goods market competition. Also, advancing financial markets provide better opportunities for risk-sharing, which increases the ability of investors to take on risks. Depending on the degree of information asymmetries at the firm level, different firms might be affected differently. These arguments suggest that studying the development of aggregated volatility only may mask diverging trends at the firm level.

Davis et al. (2006), in contrast, argue that differences in the time series patterns of firm level and aggregated volatility are unlikely. They argue that volatility at different level of aggregation moves into the same direction if the shares of firms in aggregated output and the responses of firms to macroeconomic developments remain relatively stable.

The structure of the paper is as follows. In Part 2, we provide a brief theoretical background. In Parts 3 and 4, we describe our data and our empirical model. In Part 5, we use this model to decompose firms' growth rates into the idiosyncratic component and the component driven by macroeconomic factors, and we analyze the determinants of firm level growth volatility. Part 6 concludes. We find that trends in firm level and aggregate volatility are similar when looking at the unconditional volatility at the firm level. This supports the view of Davis et al. (2006). Controlling for macroeconomic developments, we find that firm level volatility is not only lower but also exhibits no

downward trend. In fact, there even is some evidence of a very slow increase in idiosyncratic firm level volatility, supporting the view of Comin and Phillipon (2005). We also investigate the determinants of idiosyncratic volatility and find higher asset growth and lower inventory-sales ratios to be associated with a higher volatility of firms' sales. Leverage has no significant impact.

2 Theoretical Background

In order to understand how firm level and aggregate volatility develop – and why time trends might differ – a theoretical framework which departs from the assumption of symmetric, homogenous firms is required. Using a standard Dixit-Stiglitz-type specification of household preferences, it can be shown that firm level prices and output depend on firm level and macroeconomic factors alike. As shown in Woodford (2003: Chapter 3), the first-order condition for the optimal pricing strategy of the supplier of good i is then given by:

$$\Pi_1(p_t(i), p_t^I, P_t; Y_t, \xi_t, \xi_{it}) = 0 \quad (1)$$

where Π_1 is the first derivative of the firm i 's profit function, $p_t(i)$ is the price charged for good i , p_t^I is an index of the price charged in industry I to which firm i belongs, P_t is the economy-wide price index, Y_t is an index of aggregate demand, ξ_t is a vector of exogenous macroeconomic disturbances, and ξ_{it} is a vector of firm-specific disturbances. According to equation (1), a firm's profit maximum with the corresponding prices and quantities depends on firm-specific, industry-specific, and macroeconomic developments. These variables, by definition, also have an impact on the volatility of firm level variables.

Equation (1) also implies that trends in the volatility of firm-specific and macroeconomic factors might differ if ξ_t and ξ_{it} exhibit different time trends or if the propagation channels between micro- and macroeconomic developments change systematically over time. Focusing on the micro-economic determinants of firm level volatility, we can think of four main reasons why volatility may change over time.

First, the process of financial market deregulation and integration may provide better risk-sharing possibilities for investors. Hence, *ceteris paribus*, more risky firms may be able to obtain external finance, thus raising the average risk and volatility of firms. Thesmar and Thoenig (2004) find evidence for France supporting the hypothesis that the financial market deregulation has contributed to an increase in firm level output volatility. We will account for the importance of financial frictions by including the leverage ratio of firms.

Second, greater exposure to international competition and the deregulation of product markets may increase the elasticity of demand that firms are facing. This could change the responsiveness of firms to a given shock.²

Third, changes in the process of productivity growth away from the imitation of existing technologies towards innovative research and development may increase volatility. Eichengreen (2007), for instance, notes that, up until roughly the 1970s, growth in Europe was characterized by a catching up process to the technological frontier whereas in later decades innovative R&D started to dominate.

Fourth, new information technologies have helped firms to improve their inventory management. During the “Great Moderation”, output volatility fell further than final sales volatility (Davis and Kahn 2007). This implies a change in the behavior of inventories in terms of lower volatility of inventories and/or a change in the covariance between inventories and sales. In our empirical analysis below, we will look at the volatility of real sales rather than output. From a theoretical point of view, we would expect inventories to be used to smoothen production when shocks to firms’ sales dominate (Hornstein 1998). If productivity shocks dominate, in contrast, inventories are used to stabilize sales. Hence, we use firm level information on the inventory-sales ratio to account for the fact that inventories should be adjusted counter-cyclically with regard to fluctuations in real sales.

Our data are not rich enough to study the importance of all of these factors. However, we will show trends in the data and model the determinants of volatility as closely as

² See Rodrik (1997) for a related argument as to why labor market volatility may increase in integrated markets.

possible to the above hypotheses. Our focus is on isolating micro factors and macro factors contributing to changes in firm level volatility.

3 Data and Descriptive Statistics

3.1 The Data

To document long-run trends in firm level volatility, we exploit, for the first time, a new firm level dataset that links two datasets available at the Deutsche Bundesbank: the Corporate Balance Sheets Statistics (*Unternehmensbilanzstatistik*), which covers the period from 1971 to 1996, and the Financial Statements Data Pool (*Jahresabschlussdatenpool*), which starts in 1997. (For methodological changes, see Deutsche Bundesbank (2005, 2006).) Earlier research was restricted to either of these statistics. For example, the Corporate Balance Sheets Statistics were used to trace the dynamics of firms over the cycle (Döpke and Weber 2006). The Financial Statements Data Pool was used to exploit the cross-section dimensions to analyze the link between growth and volatility at the firm level (Buch and Döpke 2007, 2008).

The Financial Statements Data Pool replaces the Corporate Balance Sheets Statistics of the Deutsche Bundesbank, as the number of corporate annual financial statements being submitted to the Bundesbank declined after the launch of the European Monetary Union in 1999 and the subsequent discontinuation of the Bundesbank's rediscount business. It has been compiled in a joint project of the Bundesbank and a number of commercial banks and other financial institutions. The Financial Statements Data Pool is somewhat broader than the Corporate Balance Sheet Statistics. The firms represented about three-quarters of the gross value added of the non-financial business sector in Germany in 2003 (Deutsche Bundesbank 2005). Both data sets can be used for statistical purposes, but under strict confidentiality on the premises of the Bundesbank only. The data have been made anonymous, and only Bundesbank staff is allowed to work with the Financial Statements Data Pool.

The total time series dimension of our new, linked data set covers 35 years (1971-2005). The unit of observation is the individual firm. Generally, our dataset contains information on all balance sheet items and financial statements. Since we are interested in tracking the evolution of a large number of firms over a long time horizon, we

retrieve information on firms' sales, assets, the legal form, the industry, and some key firm level control variables only. We use the following firm level variables:

- (Real) sales: Like Comin and Phillipon (2005) or Davis et al. (2006), we use real sales as a proxy for firm level output. For this variable, we have relatively few data losses owing to incomplete and inconsistent reporting. Developments of sales and output differ in that the inventory growth and sales growth are imperfectly correlated. We use sales in order to avoid a possible bias arising from the fact that we have only balance sheet data on inventories. Lacking information on firm level prices, we convert nominal variables into real variables by deflating each firm's sales with the industry-level price index obtained from the EU KLEMS database (see www.euklems.net) and taking the difference of the logarithm of real sales.
- Assets: Total assets of firms, deflated with the same price index as firms' sales, are used as a proxy for the size of firms. The advantage of this measure over size measures such as the stock of fixed assets, equity capital, or external debt, is that it is not influenced by differences in the financing structure or the asset structure of firms. Also, total assets are a stock measure of firm size, which is likely to fluctuate less over time than sales as a flow measure.
- Inventories: We include the inventory-sales ratio to capture possible effects of technological changes on the volatility of firms' sales. This also takes into account that sales are an imperfect proxy of output if inventories change.
- Leverage: We compute a measure of leverage as a firm's debt divided by total equity. Leverage could be one measure of credit market friction. From a theoretical point of view, we expect that more severe financial friction reduce growth and increase the volatility of investment (Aghion and Banerjee 2005). Accordingly, the expected sign of leverage would be negative.³

Our dataset contains information on about 80,000 firms per year for the 1971-2005 period. The unrestricted dataset has 2.8 million firm-year observations. The average duration in the panel is about 13 years per firm. Since our empirical method, which will be explained in more detail below, is partly based on time series regressions for each

³ Note that information on leverage is not available for a number of firms, which reduces the size of our sample from about 1,750 to 1,450 firms. However, the qualitative results that are reported in the following also hold for the larger panel.

individual firm, we retain only those firms which are present in the dataset for the full period. This reduces the total number of firms contained in the sample to 1,464 firms or 51,240 firm-year observations.

In the appendix, the distribution of the unrestricted sample (i.e. the unbalanced panel) by the number of firms and the volume of real sales by industry (Table 1a) and years (Table 1b) is given. We also provide comparative evidence of the balanced panel.

One particularly interesting feature of our dataset is that it contains information on manufacturing *and* services firms. In terms of the volume of sales, services account for roughly 19% of sales in the balanced panel. The bulk of these sales (13%) is in the wholesale trade sector. In the unbalanced panel, the largest sector is wholesale trade as well, accounting for about 24% of total in terms of sales. The share of the wholesale trade sector is even higher in terms of the number of firms (21% in the unbalanced panel and 29% in the balanced panel). These numbers are out of proportion to the importance of this sector for GDP since we refer to sales rather than to value added. None of the remaining sectors dominates as clearly. As regards the number of firms, the basic metals sector (11.6%) ranks second, and machinery and equipment (10.7%) ranks third. In terms of the volume of real sales, the largest sector is petroleum products (29.6%), followed by transport equipment (15.7%), and wholesale trade (12.7%).

A comparison of the structure of the balanced panel and the unbalanced panel shows that firms from manufacturing sectors are overrepresented and services firms are underrepresented in the balanced panel. This is not surprising since the balanced panel favors firms which had taken part in the Bundesbank's rediscount business. At the same time, this makes our findings robust to structural changes (see, e.g., Parker 2007). The disadvantage, of course, is that the inflow of younger and presumably riskier firms into the sample as well as the exit of high volatility firms is not captured. In Section 5.3 below, we discuss different sample splits, which take into account that the survivorship bias might differ across firms of different sectors, size, or legal status. Table 1b shows the allocation of firms over the years. Here, the selection bias due to our restriction on a balanced panel is less severe, as we have about 3% of all observations in each year in both samples.

3.2 Computing Firm Level Volatility

As a first step towards computing firm level volatility, we compute firms' growth rates as

$$g_{it} = \frac{y_{it} - y_{it-1}}{y_{it-1}} \quad (2)$$

where g_{it} is firm i 's sales growth rate, and y_{it} are firm i 's real sales in t . Since our data contain no information on reasons for large outliers such as mergers and acquisitions, we drop observations which fall by more than half or more than double. The same restriction is applied to the growth rate of assets. Moreover, observations are excluded if leverage is larger than 100. Table 2a in the appendix shows the descriptive statistics. The mean firm level growth rate for the full sample is 3.6%, which is slightly higher than the growth rate of real GDP (2.2%).

As regards the measurement of volatility, it has become relatively standard in the literature to compute growth volatility as the standard deviation in growth rates over a moving 5-year or 10-year window (see, e.g., Comin and Phillipon 2006, Davis et al. 2006). The volatility of real sales over a 10-year window is thus defined as:

$$\sigma_{it}^{10}(y) = \left[\frac{1}{10} \sum_{\tau=-9}^0 (g_{i,t+\tau} - \bar{g}_{it})^2 \right]^{1/2} \quad (3).$$

where \bar{g}_{it} measures the mean growth rate. Correspondingly, the volatility over a 5-year window $\sigma_{it}^5(y)$ uses observations ranging from $t = -4$ to $t = 0$.

Graph 1 (see the appendix) plots the mean, unconditional firm level volatility of sales growth against the volatility of the growth rate of GDP. The unconditional volatility has not been corrected for macroeconomic influences. The median standard deviation of firm level growth is about 12%, which is comparable to evidence found in earlier studies (Buch et al. 2006, Comin and Phillipon 2005) and 10 times higher than the volatility of aggregated GDP. Graph 1 shows that the time trends are quite similar and share two main features. First, there has been a trend decline in volatility over time. Second, volatility was temporarily higher in the period following German unification in

the early 1990s. These general patterns in the data are very similar when using the unbalanced panel of firms.

The information in Graph 1 alone, however, says little about time trends in the idiosyncratic volatility of firms' growth volatility though. We will therefore describe the methodology, in the following, how we isolate microeconomic factors from macroeconomic factors affecting firm level volatility.

4 Computing Idiosyncratic Firm Level Volatility

Previous empirical work on the determinants of firm level volatility uses the volatility of output as given by equation (3) as the dependent variable. Such regressions suffer from three potential problems. First, by construction, the dependent variable is serially correlated. Second, the choice of the 5-year or 10-year window for the computation of volatility is somewhat arbitrary, although it is often used in the empirical literature. Third, using the unconditional firm level volatility does not allow distinguishing idiosyncratic, firm level, from macroeconomic factors. The parallel evolution of firm level and aggregated volatility documented in Graph 1 could be the result of similar developments at different levels of aggregation, but it could also cloud diverging patterns at the firm level. Simply including observable macroeconomic volatilities as regressors may not fully isolate the idiosyncratic component because only observed factors are included. Also, in a pooled regression, the impact of macroeconomic factors is assumed to be homogenous across firms.

Hence, in this section, we describe the multifactor residual model recently proposed by Pesaran (2006). This model allows us to filter observed and unobserved macroeconomic factors out of the firm level data. It provides us with a measure of the conditional, idiosyncratic firm level real sales growth. To see the logic of the multifactor residual model, assume that firm level sales growth is given by

$$g_{it} = \alpha'_i d_t + \beta'_i x_{it} + \varepsilon_{it} \quad (4)$$

where $i = 1, 2, \dots, N$ is the number of firms and $t = 1, 2, \dots, T$ the number of years. α and β are parameters to be estimated. Equation (4) states that firm level growth depends on a $k \times 1$ vector of observable macroeconomic factors (d_t) and a vector of

observed firm-specific regressors (x_{it}). The errors are assumed to have a multifactor structure:

$$\varepsilon_{it} = \gamma'_i f_t + u_{it} \quad (5)$$

where f_t is an $m \times 1$ vector of unobserved macroeconomic factors and u_{it} the individual-specific (idiosyncratic) errors, which are assumed to be distributed independently of d_t and x_{it} . In Pesaran (2006), the observed and unobserved factors d_t and f_t are assumed to be covariance stationary. Generally, the unobserved factors can be correlated with (d_t, x_{it}) ; the individual-specific regressors are modelled according to:

$$x_{it} = A'_i d_t + \Gamma'_i f_t + v_{it} \quad (6)$$

where A_i and Γ_i are factor loading matrices and v_{it} those components of x_{it} which are independent of the macroeconomic factors.

In most applications, the interest is in the slope coefficients β_i in equation (4). The factor loadings α_i and γ_i are not the main interest, but they can be estimated consistently when both N and T are large. Since our panel has a time series dimension of $T = 35$, we can analyze the firm level responses to the macroeconomic factors. However, our main interest is in the variance of the idiosyncratic term u_{it} . With an estimate of u_{it} at hand, we can compute the variance of the idiosyncratic component of

firm level growth for each firm i . This variance can be proxied through the variance of

the first-stage residuals: $\frac{1}{T+1} \sum_{j=0}^T \hat{u}_{it-T+j}^2 \quad \forall i$, which converges to $\frac{1}{T+1} \sum_{j=1}^T \sigma(u_{it-T+j})^2$.

(See e.g. Gorbachev (2007) for a similar specification using household consumption data.)

The key challenge is to isolate developments at the firm level from aggregate developments while taking into account the fact that some of the macroeconomic factors affecting all firms alike are unobserved. Hence, we isolate factors which affect all firms (d_t, f_t) from those firm-specific variables, which are independent from the

macroeconomic factors, i.e. the residual of equation (6) (v_{it}). We perform this decomposition in two steps.

In a first step, we run time-series regressions for sales growth of each firm in the sample on observed and unobserved macroeconomic factors. Following Pesaran (2006), the unobserved macroeconomic factors can be proxied through the sample means of firm-specific variables. We therefore take advantage of the fact that the model described by equations (4)-(6) is quite general as it allows the unobserved common factors f_i to be correlated with the individual-specific regressors x_{it} via a multifactor structure with heterogeneous factor loadings over the cross-section units. We run these regressions for each of the firms in our balanced panel and for each of the firm level variables separately and retain the residuals from these regressions.⁴

In a second step, we use the squared residuals from these time series regressions, which are the growth rates of firm level variables cleaned of the observed and unobserved factors, in a pooled firm level panel regression:

$$u_{it}^2 = \alpha_0 + \alpha'_1 v_{it} + \alpha'_2 \delta_i + \alpha'_3 \delta_t + \alpha'_4 \delta_s + \alpha'_5 \delta_l + \eta_{it} \quad (7)$$

where u_{it}^2 is the squared residual of the first-stage regression of sales growth on macroeconomic factors, $\delta_i, \delta_t, \delta_s, \delta_l$ are firm-, time-, sector-, and legal fixed effects, and v_{it} is a set of firm-specific regressors, i.e. the first-stage residuals from equation (6). We use time fixed effects as a general specification of the time trend as well as a linear time trend. Note that the time trend also captures the fact that firm level volatility is related to firm age. Typically, older and more established firms are less volatile than younger firms. Taken in isolation, this should lead to a negative impact of the time trend on volatility but, of course, the trend also captures other factors that vary systematically over time.

To check whether there is a time trend in the variance of the idiosyncratic, conditional firm level growth rates, we also use graphical as well as regression-based evidence. We employ equation (3) above and compute the rolling standard deviation of

⁴Note that the individual-specific coefficients are consistently estimated. They are also asymptotically unbiased for $N \rightarrow \infty$ for both T fixed and $T \rightarrow \infty$ as long as the rank condition concerning the factor loadings is satisfied (Pesaran 2006).

the idiosyncratic volatility. In addition, we use fixed effects panel regressions and the heteroskedastic regression model suggested by Harvey (1976) to check whether the squared residuals as a proxy for the idiosyncratic volatility of firm level sales growth follow a systematic process. Note that the first-stage regressions account for the observed and unobserved macroeconomic factors in the mean growth of firm level sales. This does not preclude finding that macroeconomic factors and time trends to have an impact on the idiosyncratic variance of firm level growth.

5 Decomposing and Explaining Firm Level Volatility

Results for the unconditional volatility of firm level sales growth reported in Graph 1 provide evidence of a decline in firm level volatility, which largely resembles the drop in macroeconomic volatility. To what extent does this decline in firm level volatility result from lower volatility at the macro-level and to what extent does it reflect smaller idiosyncratic volatility? In this section, we report the results of the first- and second-stage regressions described above.

5.1 First-Stage Regression: Decomposing Firm Level Volatility

The goal of the first-stage regressions is to clean the growth of firm level variables of observed and unobserved macroeconomic factors. Our set of observed macroeconomic factors includes domestic and foreign as well as price and quantity measures. To capture the domestic macroeconomic environment, we include the growth rate of domestic absorption, short-term interest rates, and domestic inflation. To capture the international macroeconomic environment, we include the annual changes in the real exchange rate, world demand, and raw material prices. (Details on the data definitions are given in the Appendix.)

Table 2a in the appendix reports the growth rate of other macroeconomic factors, i.e. domestic real absorption (mean growth rate of 1.9%), world import volume (4.0%), raw material prices (6.7%), and the real effective exchange rate of the German economy against its 23 most important trading partners (-0.1%). An increase in this indicator implies a real appreciation. The average short-term interest rate and the inflation rate over this period were 5.7% and 2.7%, respectively.

We use four means of firm-specific variables as proxies for unobserved macroeconomic factors. The mean growth rate of firms' sales captures structural shifts in firm level growth. The growth rate of firms' assets captures changes in the size structure of firms. The mean inventory-sales ratio captures changes in the importance of inventories due to technological progress, and mean leverage across firms captures changing financial frictions. As expected, the inventory-sales ratio shows a downward trend over time, reflecting improvements in inventory management. Leverage, in contrast, has been increasing up until the early 1990s and falling thereafter, reflecting legislative changes affecting the financing of firms in Germany as well as the increase in stock market valuation during the 1990s.

Table 2b in the appendix shows the correlations between the observed macro-factors and the unobserved factors. All correlations are below 0.9, and most are below 0.5. We find the highest correlations between GDP growth and the growth rate of domestic absorption as well as between some firm level variables. Since domestic absorption is less correlated with the mean firm level factors than GDP growth, we use domestic absorption as our domestic demand component. In addition, the changes in the real exchange rate and in world imports as well as the interest rate and inflation show relatively high pairwise correlations. The same holds for the mean firm level variables. However, we include a full set of observed and unobserved macro-factors as these may still pick up some orthogonal component of the macroeconomic environment.

The advantage of running the first-stage regressions firm-by-firm is that we allow for the largest possible heterogeneity across firms. At the same time, results of these regressions are difficult to visualize. Therefore, we regress the firm level coefficients obtained from the time series regressions on a set of dummy variables. To account for the fact that the coefficients are measured with uncertainty, we weight these observations with the inverse of their standard deviation.

The aim of this exercise is to see whether, for instance, large and small firms react differently to the macroeconomic environment. Results reported in Table 3 (see the appendix) show some differences in the response of firms to macroeconomic factors. Generally, larger firms grow faster than smaller firms when domestic and world market demand increases and the real exchange rate appreciates. Also, an increase in raw

materials prices and in the real interest rate increases the growth of large firms relative to that of small firms. This would be consistent with large firms being oriented more towards the international market and less affected by financial constraints. Listed firms, in contrast, do not differ significantly from the rest of the sample. An increase in the real interest rate lowers growth of firms in the sectors manufacturing, services, and construction in comparison to growth in the remaining sectors (agriculture, mining, energy). Higher world import demand increases growth in the manufacturing and construction sector, while higher raw materials prices lower growth in manufacturing and services.

To assess the importance of the macroeconomic factors for firm level sales growth, we also compute the partial R^2 s for these factors. This partial R^2 indicates how much a particular macroeconomic factor contributes to explaining the variance of firm level sales growth.

On average, the macroeconomic factors individually explain about 5-6% of the variance in sales growth across firms (Table 2c). However, the explanatory power varies widely with the highest partial R^2 s being in the range of 0.45 for domestic absorption and 0.59 for the real exchange rate. Hence, for some of the firms in the sample, individual macroeconomic factors explain more than 50% of the variance in real sales over time.

Generally, the exact specification of the first-stage regressions should not be overemphasized. Essentially, these regressions serve the purpose of filtering macroeconomic information out of the firm level time series and to retain the idiosyncratic component of sales growth. In this sense, they are similar to the mean equations in an ARCH or GARCH setting, which serve the purpose of providing an estimate of the variance of the variable at hand.

The most interesting outcome of the first-stage regressions are perhaps the plots of the residual standard deviations. Graph 2 in the appendix shows the conditional volatility of firms' sales growth. As before, we use a 5-year rolling window. There are two striking observations that immediately meet the eye.

First, the conditional firm level volatility is substantially lower than the unconditional volatility. During the turbulent 1970s, in particular, macroeconomic

factors accounted for more than 50% of the volatility of firm level sales growth. In subsequent periods, the importance of macroeconomic factors has been less pronounced, but the gap between the conditional and the unconditional volatility series was still substantial. Hence, even though the impact of individual macroeconomic factors on firm level sales growth has, on average, been small, the overall impact of the macro-economy on firm level developments has been quite substantial.

Second, the downward trend in firm level volatility reported in Graph 1 is not visible once macroeconomic factors are taken into consideration. Instead, the time profile of firm level volatility is rather flat and even shows a slight upward trend. In this sense, our results confirm the conclusion reached by Davis et al. (2006) that (unconditional) firm level and aggregate volatility can be expected to behave similarly. However, our findings for the conditional volatility are also consistent with Comin and Phillipon (2006) who argue that the time trends of aggregate and firm level volatility differ.

5.2 Second-Stage Regressions: Explaining the Volatility of Sales Growth

Which are the firm-specific factors that determine the volatility of growth at the firm level? Do size and sector effects matter? Do we find support for different hypothesis as to the causes of the Great Moderation at the firm level? These are questions we will now turn to.

In Table 4 (see the appendix), we report results of firm level regressions using the squared residuals of the first-stage regressions as a measure of the volatility of the conditional firm level growth rates of sales as a dependent variable (cf. equation (7)). We use fixed-effects panel regressions with robust standard errors. In contrast to the first-stage regressions, we now estimate homogenous coefficients.

Generally, our model explains less than 5% of the variation in the volatility of sales growth across firms. In similar (unreported) regressions using the growth rate of firms' sales as a dependent variable, we obtain higher R^2 s ranging from 0.13 to 0.19. It does not come as a surprise to find that the explanatory power of the idiosyncratic, firm level volatility is relatively low, since we have stripped the variables off macroeconomic developments and since we consider only the idiosyncratic variation in the data. Moreover, these results are strongly driven by the very large amount of cross-sectional heterogeneity, which is typical for large micro-data sets like the one at hand.

(Gorbachev (2007), for instance, analyzes the determinants of household-level consumption risk for the US and reports R^2 s in a similar range.)

We find a very consistent pattern of sales volatility to be positively related to growth of total assets. The estimated coefficient is about 0.03, and it is quite robust across different specifications with regard to other control variables and the inclusion of fixed effects. Moreover, the impact of asset growth on firm level volatility builds up gradually over time. We include up to three lags of asset growth, and the first lag is significant throughout. In terms of economic significance, asset growth is the most important variable, with a beta coefficient of 8%. Hence, variation in sales growth across firms accounts for about 8% in the variation of volatility across firms.

The finding that faster-growing firms also have more volatile sales growth is consistent with earlier findings at the firm level or industry level concerning the growth-volatility nexus. It is at odds with some findings for a negative relationship between growth and volatility using aggregated data. However, it would not come as a surprise to find that the growth rates of firms are imperfectly correlated. Imbs (2007), for example, shows that the correlation between growth and volatility depends on the level of aggregation of the data. He finds a positive correlation between growth and volatility at the industry level. One explanation is that growth rates are imperfectly correlated across sectors. Our firm level data suggest a similar positive relationship.

Firms with a high inventory-sales ratio have a low volatility of real sales. The beta coefficient is -1.3% . Leverage as a measure of the financial constraints that firms are facing is insignificant, in contrast. At first sight, the negative coefficient on inventories seems at odds with the hypothesis that improved inventory management helps firms to reduce the volatility of sales. According to this explanation, technological innovations help firms to reduce the desired inventory-sales ratio. A lower inventory-sales ratio, in turn, would weaken the destabilizing, pro-cyclical impact that inventories have for final sales via a positive covariance between inventories and sales (Ramey and Vine 2004). Our results would be consistent though with inventories serving as a buffer against productivity shocks. According to this interpretation, higher inventory-sales ratios would smoothen the impact of volatile production on sales.

Generally, however, standard models of inventory management often fail to find support in the macro-data. Prominent theories such as the production smoothing model (Holt et al. 1960) predict a negative correlation between inventory investment and overall production, which was not found in the data. (See Blinder (1986) and the literature cited therein.) Rather, most macro studies find a positive correlation between inventory investment and GDP. Possible explanations of these findings point, inter alia, to the fact that inventories are still treated at a too aggregated level. In particular, a distinction between inventories of inputs and the inventories of finished goods is warranted. However, we lack such information in our dataset.

To test whether there are time trends in firm level volatility, we include a time trend and a unification dummy. The unification dummy and the time trend are insignificant if lagged dynamics are accounted for. This shows that our macroeconomic factors have filtered out impacts of the unification process on firm level volatility. The time trend is significant but relatively small, thus indicating only a relatively slow increase in volatility. Moreover, the beta coefficient is only 0.26%. In contrast to unconditional firm level volatility and aggregated volatility, the idiosyncratic volatility of firms has thus been on a slow trend rise.

In unreported regressions, we include a full set of time-fixed effects as a more flexible way to capture time trends in the data. Results for the remaining control variables are unchanged. We also interact the time trend with a dummy variable for listed firms to test the hypothesis put forward by Davis et al. (2006) that firm level volatility has increased because of an upward trend in the volatility of large, publicly listed firms. However, this interaction term is insignificant.

5.3 Robustness and Selection Issues

To check the robustness of our results, we split the sample along different dimensions, we use a heteroskedastic regression model to simultaneously model the mean and the variance of idiosyncratic sales growth, and we correct for the bias arising from the fact that we estimate our model on a reduced sample of firms.

Sample splits: The descriptive statistics reported in Tables 1a and 1b show differences between the full and the reduced sample. Hence, sample selection may affect our results since only firms which have data for the full sample are included. This

sample selection could affect our results if the residuals from the regression of the determinants of firm level output volatility were correlated with residuals from a latent regression determining the survival of firms. These survival probabilities, in turn, may be related to firm size, the sector in which firms are active, or access to the capital market. To account for such differences, we split the sample into manufacturing and services, listed and unlisted, and small and large firms (Table 4a). The cut-off for small and large firms is the median level of real sales. To test whether time trends in firm level volatility have changed over time, we break down the sample further into the 1970s, the 1980s, the 1990s, and the current decade (Table 4b). We also split the sample in 1985, the year which is often taken as the start of the Great Moderation period.

Perhaps the most interesting finding is that the qualitative results for the control variables are generally not affected by the different sample compositions. There are a few exceptions though. In the 1970s, none of our control variables is significant. Instead, there has been a positive and significant upward time trend in firm level volatility. Moreover, the inventory-sales ratio is insignificant for the sub-samples of listed firms (although the negative sign is retained). Inventories have a negative and significant effect for services firms, which would seem surprising for services which are non-storable. However, since the sector “services” also includes retail and wholesale trade firms and since these firms carry inventories, this effect seems quite plausible.

Selection bias: In unreported regressions, we have also estimated the volatility regression applying the correction for sample selection proposed by Heckman (1976). The selection equation shows that large firms, firms with lower sales growth, and firms with lower leverage are more likely to be in the balanced panel. The Mills ratio is significant, indicating that sample selection is indeed an issue. Results for the equation explaining the volatility of firm level output show very similar results with regard to the coefficient estimates and significance as regressions not accounting for the selection bias. More specifically, firms that grow faster, that have a higher inventory sales ratio, and that have lower leverage have a lower volatility. The linear time trend is again positive and significant.

Heteroskedastic regression model: To check the robustness of our results for the variance equation, we alternatively use a regression model with multiplicative

heteroskedasticity as proposed by Harvey (1976). The advantage of this model is that we can simultaneously specify a “mean” equation – explaining the growth of sales – as well as a “variance” equation – explaining the residual variance.

The heteroskedastic regression model has, to the best of our knowledge, not been applied to an analysis of firm level volatility so far. Earlier applications use it to model heteroskedasticity in the residuals, yet most papers do not focus on estimates of the variance equation. There are two exceptions. A recent paper by Cerqueiro et al. (2007) studies banks’ loan pricing decisions. Ang and Peterson (1985) estimate a capital asset pricing model and study the determinants of rates of returns as well as the variance of returns.

The mean equation of the heteroskedastic regression model gives the level of sales growth as a function of a set of explanatory variables X : $\tilde{g}_{it} = \tilde{x}'_{it} \beta + \varepsilon_{it}$ where ε_{it} is the residual with $E[\varepsilon_{it} | \tilde{x}_{it}] = 0$ and $VAR[\varepsilon_{it} | \tilde{x}_{it}] = \sigma_{it}^2 = \exp\{Z'_{ij} \phi\}$. The variance equation is given by $\sigma_{it}^2 = Z'_{it} \phi$. The coefficients β and ϕ can be obtained by maximizing the log-likelihood function. One advantage of this methodology is that the parameters of the mean and of the variance equation are uncorrelated. We essentially estimate the same model as before but we now include dummies for the different types of firms (large versus small, listed versus unlisted) in lieu of a full set of firm fixed effects.

Results are reported in Table 5 (see the appendix). Turning to the results of the mean equation first, we find that asset growth is positively correlated with the growth rate of firm level sales. There is evidence that larger firms grow faster than average. Large inventory-sales ratios are associated with lower sales growth, indicating that inventories contain mainly finished products. Since the specification of the mean equation is certainly a bit ad hoc, we also use a specification using a linear time trend and the unification dummy only.

The results for the variance equation by and large confirm the findings reported above considering the positive impact of higher asset growth and the positive impact of the time trend. The firm size dummy is negative and significant. The negative impact of

the inventory-sales ratio is significant only in the specification using a very lean specification of the mean equation (Column 2 of Table 5).

6 Summary and Conclusions

This paper has tested whether evidence of a “Great Moderation” can be found in firm level data of German firms. To answer this question, we have used a unique firm level dataset which allows a large sample of German firms to be tracked across 35 years. Our paper has three main findings.

First, the unconditional firm level volatility and aggregate volatility have developed quite similarly. This confirms Davis et al. (2006) who use US data and differs from the findings of Comin and Phillipon (2005) who find diverging trends of firm level and aggregated volatility.

Second, we show that inference about the evolution of firm level volatility requires us to take account of macroeconomic factors. Using the multifactor panel model proposed by Pesaran (2006), we have decomposed firm level sales growth and volatility into the idiosyncratic component and the component driven by observed and unobserved macroeconomic developments. Results from this disaggregation show that the overall time trends in unconditional firm level volatility are driven by macroeconomic factors. Once we take this into account and extract the idiosyncratic component of firm level volatility, we not only find volatility to be lower but also to exhibit a flat and, if anything, slightly increasing pattern.

Third, we use different panel models to analyze the determinants of the idiosyncratic component of firm level volatility. Faster-growing firms also exhibit a higher volatility of real sales growth. In this sense, higher idiosyncratic volatility could have a positive impact on overall growth and thus economic welfare. If anything, the time trend in firm level volatility has been positive. However, the time trend explains only a very small share of the variation in idiosyncratic volatility across firms and time.

Our results show the importance of a stable macroeconomic environment for firm level stability. High macroeconomic volatility such as was observed during the 1970s has roughly doubled firm level volatility. In later periods, the decline in firm level volatility was driven largely by the decline in macroeconomic volatility. To the extent

that “Good Policy” is behind the decline in aggregate volatility, this has also contributed to lower volatility at the firm level. Idiosyncratic, firm level volatility exhibits no particular time trend. Studying whether similar patterns can be found in other indicators such as the volatility of employment, income, and consumption, remains an important topic for future research.

Finally, our paper has implications for the potential to diversify risks across German firms. Whereas a substantial part of volatility at the firm level is driven by common macroeconomic factors, which limits the diversification potential across domestic firms, the relative importance of idiosyncratic risks has increased over time. These risks, however, can be diversified in developed financial markets.

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8 Data Appendix

Firm Level Data

All firm level data are taken from the Corporate Balance Sheet Statistics and the Financial Statements Data Pool provided by the Deutsche Bundesbank (Deutsche Bundesbank 2005). The data are confidential firm level data, and the Financial Statements Data Pool can be used by the staff of the Deutsche Bundesbank only.

Sales growth: Annual growth rate of a firm's real sales.

Asset growth: Annual growth rate of a firm's total real assets.

Inventory-sales-ratio: Ratio of a firm's inventory over total sales.

Large/small firms: Firms with real sales above (below) the sample median.

Leverage: Ratio of a firm's total debt to total equity.

Listed/unlisted firms: Incorporated (not incorporated) companies (*Aktiengesellschaften*).

Manufacturing firms: Chemicals, Basic metals, Machinery, Transport equipment, Electrical equipment, Wood, Pulp and paper, Textiles, Food, Computers, Non-metallic mineral products, Plastic products, Petroleum products, Furniture.

Services firms: Wholesale trade, Hotels, Transport & communication, Financial intermediation, Real estate, Public services, Retail trade, Firm-related services.

Macroeconomic Data and Sources

GDP growth: Annual growth rate of German real GDP (German Federal Statistical Office).

Growth rate of domestic absorption: Annual growth rate of German real absorption (German Federal Statistical Office).

Growth rate of world imports: Annual growth rate of world import volumes (International Monetary Fund).

Change in real exchange rate: Annual rate of change in Germany's real effective exchange rate against 23 countries in quantity notation. An increase indicates a real appreciation (European Central Bank).

Change in raw material prices: Annual rate of change in the HWWA index of raw material prices (overall index).

Short-term interest rate: Average short-term interest rate (3 months, Frankfurt money market rate; Deutsche Bundesbank).

Table 1: Distribution of Firms by Sector and Year

This table gives the distribution of firms in the sample by sector (Table 1a) and year (Table 1b). We report the distribution of the number of firms and of the volume of real sales (in million euro). The balanced panel includes all firms; the unbalanced panel includes only firms which are in the sample for the full period.

(a) Distribution of Firms by Sector

	Number of firms				Real sales			
	Unbalanced		Balanced		Unbalanced		Balanced	
	Number	Percent	Number	Percent	Mil. euro	Percent	Mil. euro	Percent
Agriculture	75,905	2.66	70	0.14	136.96	0.2	5.21	0.04
Basic Metals	162,012	5.67	5,922	11.56	3715.76	5.5	554.64	4.72
Chemicals	32,498	1.14	2,119	4.14	3007.83	4.45	709.93	6.04
Computers	4,315	0.15	24	0.05	467.71	0.69	7.72	0.07
Construction	256,831	8.99	1,710	3.34	2257.61	3.34	176.54	1.5
Electrical equipment	84,587	2.96	1,995	3.89	3339.02	4.94	530.11	4.51
Energy	17,316	0.61	997	1.95	2805.73	4.15	389.37	3.31
Financial intermediation	2,037	0.07	n.a.	n.a.	79.72	0.12	n.a.	n.a.
Firm related services	141,704	4.96	126	0.25	1171.98	1.73	2.71	0.02
Food	90,483	3.17	2,993	5.84	3416.93	5.06	307.85	2.62
Furniture	28,801	1.01	633	1.24	306.31	0.45	20.42	0.17
Hotels	19,952	0.7	n.a.	n.a.	84.67	0.13	n.a.	n.a.
Machinery	152,390	5.34	5,498	10.73	3846.20	5.69	473.37	4.03
Mining	23,341	0.82	443	0.86	826.85	1.22	267.23	2.27
Non-metallic mineral products	48,739	1.71	1,712	3.34	783.03	1.16	147.25	1.25
Petroleum products	1,728	0.06	209	0.41	6217.74	9.2	3479.59	29.6
Plastic products	56,767	1.99	1,312	2.56	1080.94	1.6	186.90	1.59
Public services	64,973	2.27	62	0.12	837.19	1.24	15.74	0.13
Pulp and paper	72,340	2.53	1,844	3.6	1406.65	2.08	154.83	1.32
Real estate	184,257	6.45	137	0.27	1187.24	1.76	4.69	0.04
Retail trade	432,062	15.13	2,872	5.6	5688.73	8.42	571.69	4.86
Textiles	92,461	3.24	3,238	6.32	1174.39	1.74	133.50	1.14
Transport & communication	107,160	3.75	458	0.89	2302.19	3.41	239.76	2.04
Transport equipment	22,795	0.8	862	1.68	4733.33	7.01	1848.69	15.73
Wholesale trade	590,360	20.67	14,965	29.21	16031.16	23.73	1498.17	12.74
Wood	78,183	2.74	1,019	1.99	656.98	0.97	30.29	0.26
Unallocated	12,009	0.42	20	0.04				
Total	2,856,006	100	51,240	100	67562.86	100	11756.18	100

(b) Distribution of Firms by Year

Year	Number of firms				Real sales			
	Unbalanced		Balanced		Unbalanced		Balanced	
	Number	Percent	Number	Percent	Mil. euro	Percent	Mil. euro	Percent
1971	47,081	1.65	1,464	2.86	1126.02	1.67	237.63	2.02
1972	47,649	1.67	1,464	2.86	1211.66	1.79	274.21	2.33
1973	47,983	1.68	1,464	2.86	1235.05	1.83	236.49	2.01
1974	50,057	1.75	1,464	2.86	1338.98	1.98	251.58	2.14
1975	61,034	2.14	1,464	2.86	1389.46	2.06	275.13	2.34
1976	70,758	2.48	1,464	2.86	1547.38	2.29	297.35	2.53
1977	78,006	2.73	1,464	2.86	1643.88	2.43	345.54	2.94
1978	84,658	2.96	1,464	2.86	1588.63	2.35	302.41	2.57
1979	87,734	3.07	1,464	2.86	1694.85	2.51	313.94	2.67
1980	83,461	2.92	1,464	2.86	1714.93	2.54	313.84	2.67
1981	77,945	2.73	1,464	2.86	1809.99	2.68	347.65	2.96
1982	78,270	2.74	1,464	2.86	1783.21	2.64	357.54	3.04
1983	79,905	2.8	1,464	2.86	1775.05	2.63	350.66	2.98
1984	80,632	2.82	1,464	2.86	1859.23	2.75	379.22	3.23
1985	79,846	2.8	1,464	2.86	1866.91	2.76	378.69	3.22
1986	79,273	2.78	1,464	2.86	1707.83	2.53	328.64	2.8
1987	78,328	2.74	1,464	2.86	1819.53	2.69	423.13	3.6
1988	76,000	2.66	1,464	2.86	1861.77	2.76	408.70	3.48
1989	73,296	2.57	1,464	2.86	1947.44	2.88	467.55	3.98
1990	71,734	2.51	1,464	2.86	2032.62	3.01	480.62	4.09
1991	72,123	2.53	1,464	2.86	2112.71	3.13	461.08	3.92
1992	71,819	2.51	1,464	2.86	1862.75	2.76	324.80	2.76
1993	71,246	2.49	1,464	2.86	1715.33	2.54	283.22	2.41
1994	71,651	2.51	1,464	2.86	1839.71	2.72	283.91	2.41
1995	68,442	2.4	1,464	2.86	1887.40	2.79	278.82	2.37
1996	66,430	2.33	1,464	2.86	1789.63	2.65	229.48	1.95
1997	104,157	3.65	1,464	2.86	2329.22	3.45	333.35	2.84
1998	97,194	3.4	1,464	2.86	2154.85	3.19	288.02	2.45
1999	98,674	3.45	1,464	2.86	2297.06	3.4	301.90	2.57
2000	100,916	3.53	1,464	2.86	2605.32	3.86	333.83	2.84
2001	106,080	3.71	1,464	2.86	2761.53	4.09	345.94	2.94
2002	111,348	3.9	1,464	2.86	2699.09	3.99	338.89	2.88
2003	123,129	4.31	1,464	2.86	2735.96	4.05	351.22	2.99
2004	134,397	4.71	1,464	2.86	2850.51	4.22	406.53	3.46
2005	124,750	4.37	1,464	2.86	2967.38	4.39	424.69	3.61
Total	2,856,007	100	51,240	100	67562.86	100	11756.18	100

Table 2: Descriptive Statistics and Correlation Matrix

This table reports descriptive statistics for the data used in the regressions. Panel (c) provides the descriptive statistics for the partial R^2 s.

(a) Descriptive Statistics

Variable	Obs	Mean	Std. dev.	Min	Max
GDP growth	49,183	0.022	0.017	-0.009	0.052
Growth rate of domestic absorption	49,183	0.019	0.021	-0.022	0.052
Growth rate of world imports	49,183	0.040	0.113	-0.211	0.305
Change in real exchange rate	49,183	-0.001	0.048	-0.099	0.113
Change in raw material prices	49,183	0.067	0.239	-0.474	0.943
Short-term interest rate	49,183	0.057	0.027	0.021	0.121
Inflation	49,183	0.028	0.019	-0.007	0.073
Mean firm level growth real sales	49,183	0.036	0.043	-0.067	0.157
Mean firm level growth of assets	49,183	0.037	0.037	-0.045	0.107
Mean inventory-sales ratio	49,183	0.162	0.007	0.147	0.173
Mean leverage	49,183	6.835	0.866	5.365	8.428
Growth rate of firm level real sales	49,183	0.036	0.162	-0.498	1.950
Growth rate of firm level assets	49,183	0.037	0.172	-0.499	1.892
Inventory-sales ratio	49,183	0.162	0.147	0.000	7.970
Leverage	49,183	6.836	9.663	-0.052	99.931
Residual growth firm level real sales	49,183	0.000	0.121	-0.763	1.566
Residual growth firm level assets	49,183	0.000	0.134	-0.836	1.612
Residual inventory-sales ratio	49,183	0.000	0.053	-1.327	3.109
Residual leverage	49,183	0.000	4.074	-44.065	66.140

(b) Correlation Matrix

	GDP growth	Growth rate of domestic absorption	Growth rate of world imports	Change in real exchange rate	Change in raw material prices	Short-term interest rate	Inflation	Mean firm level growth in real sales	Mean firm level growth in assets	Mean inventory-sales ratio
GDP growth	1.00									
Growth rate of domestic absorption	0.8431	1.00								
Growth rate of world imports	0.0218	-0.2373	1.00							
Change in real exchange rate	0.1004	0.241	-0.7115	1.00						
Change in raw material prices	0.0604	-0.1934	0.3447	-0.3028	1.00					
Short-term interest rate	0.185	-0.0319	0.013	0.0469	0.2413	1.00				
Inflation	0.0692	0.0467		0.2075	0.2064	0.7664	1.00			
Mean firm level growth in real sales	0.7958	0.5927	0.2383	-0.1727	0.415	0.0562	0.0353	1.00		
Mean firm level growth in assets	0.7329	0.6261	0.3318	-0.2735	0.2887	-0.1098	-0.1615	0.8729	1.00	
Mean inventory-sales ratio	0.1043	0.0448	0.0551	-0.2241	0.0427	0.6296	0.511	0.0767	-0.0178	1.00
Mean leverage	-0.0629	0.0179	0.2551	-0.0115	-0.189	-0.101	-0.4194	-0.2212	-0.0147	-0.1517

(c) Descriptive Statistics: Partial R^2

	Observations	Mean	Standard deviation	Minimum	Maximum
Domestic absorption	56,179	0.054	0.069	0.00	0.453
World imports	56,179	0.057	0.074	0.00	0.519
Real exchange rate	56,179	0.076	0.094	0.00	0.595
Raw material prices	56,179	0.053	0.073	0.00	0.579
Short-term interest rate	56,179	0.048	0.066	0.00	0.549
Inflation	56,179	0.061	0.079	0.00	0.507

Table 3: Firm Level Response to Macroeconomic Factors

The panel reports results of cross-section regressions using the first-stage coefficient estimates as the dependent variable. Observations are weighted by the inverse of the first-stage regression standard errors. The sectors not captured by the sector dummies are agriculture, energy, and mining. All regressions include a full set of legal fixed effects (unreported). The explanatory variables are 0-1-dummy variables. Residuals are adjusted for heteroskedasticity. ***, **, * = significant at the 1%, 5%, 10%-level.

	Domestic absorption	World import demand	Real exchange rate	Raw material prices	Short-term interest rate	Inflation
Large (0/1)	0.233* (1.67)	0.055** (2.45)	0.105** (2.06)	0.018** (2.26)	0.220** (2.23)	-0.302* (1.89)
Listed (0/1)	0.421 (0.84)	-0.004 (0.04)	-0.111 (0.51)	0.012 (0.30)	0.629* (1.73)	-0.828 (1.23)
Manufacturing (0/1)	-0.277 (0.77)	0.105** (2.55)	-0.005 (0.04)	-0.037* (1.88)	-0.639*** (2.91)	1.235*** (4.62)
Services (0/1)	-0.569 (1.50)	0.029 (0.64)	0.029 (0.21)	-0.071*** (3.39)	-0.520** (2.24)	0.856*** (2.92)
Construction (0/1)	1.048 (1.41)	0.195* (1.84)	0.029 (0.12)	-0.034 (0.79)	-1.303*** (2.88)	3.016*** (4.15)
Constant	-0.075 (0.13)	-0.095 (0.86)	-0.045 (0.19)	0.023 (0.52)	0.071 (0.18)	-0.212 (0.31)
Observations	1,446	1,446	1,446	1,446	1,446	1,446
R-squared	0.02	0.02	0.01	0.03	0.02	0.03

Table 4: Determinants of Firm Level Volatility of Real Sales

This table reports results of panel fixed effects regressions. The dependent variable is the squared residual of a regression of firm level real sales growth on macroeconomic factors. Asset growth, the inventory-sales ratio, and leverage are residuals of first-stage regressions as well. “Year” is a linear time trend, “Unification” dummy is a dummy variable which is set to “One” for the years 1991-1995 and “Zero” otherwise. ***, **, * = significant at the 1%, 5%, 10%-level.

(a) Baseline Regressions and Sample Splits by Type of Firm

	Full sample	Manu- facturing	Services	Listed	Unlisted	Small	Large
Asset growth	0.036*** (6.01)	0.028*** (3.32)	0.045*** (5.16)	0.110*** (2.66)	0.028*** (5.70)	0.027*** (4.28)	0.048*** (4.34)
Asset growth (t-1)	0.016*** (4.82)	0.014*** (3.03)	0.011*** (2.78)	0.040*** (2.70)	0.013*** (4.08)	0.012*** (3.45)	0.021*** (3.47)
Asset growth (t-2)	0.002 (0.62)	0.002 (0.36)	0.003 (1.02)	0.024 (1.37)	0.00 (0.16)	0.002 (0.86)	0.001 (0.22)
Asset growth (t-3)	-0.002 (0.99)	-0.004 (1.17)	0.004 (1.45)	0.00 (0.03)	-0.002 (1.25)	-0.003 (0.94)	-0.002 (0.56)
Inventory-sales ratio	-0.073*** (3.76)	-0.061** (1.98)	-0.072*** (3.34)	-0.062 (1.30)	-0.075*** (3.54)	-0.071*** (3.90)	-0.076** (2.42)
Leverage	0.00 (0.74)	0.00 (0.54)	0.00 (0.34)	0.00 (0.41)	0.00 (0.48)	0.00 (0.78)	0.00 (0.17)
Unification	0.00 (0.31)	0.00 (0.11)	0.00 (0.08)	-0.003* (1.79)	0.001 (0.78)	0.00 (0.10)	0.00 (0.29)
Year	0.000** (2.05)	0.00 (1.12)	0.00 (1.15)	0.000** (2.46)	0.00 (1.20)	0.00 (0.55)	0.000** (1.96)
Constant	-0.142** (2.13)	-0.032 (0.43)	-0.109 (1.11)	-0.673** (2.59)	-0.067 (1.06)	0.015 (0.20)	-0.145 (1.34)
Sector dummies	yes	yes	yes	yes	yes	yes	yes
Legal dummies	yes	yes	yes	yes	yes	yes	yes
Year dummies	no	no	no	no	no	no	no
Observations	43,799	25,291	15,550	4,857	38,942	21,930	21,869
Number of cross-sections	1,464	884	579	212	1337	731	733
R-squared	0.02	0.01	0.03	0.06	0.02	0.02	0.02

(b) Sample Splits by Time

	1970s	1980s	1990s	2000s	Pre-1985	Post-1985
Asset growth	0.001 (0.08)	0.033*** (2.77)	0.050*** (4.35)	0.049*** (4.07)	0.018** (2.15)	0.045*** (5.74)
Asset growth (t-1)	0.012 (1.39)	0.027*** (3.76)	0.016*** (2.69)	0.009* (1.80)	0.025*** (3.78)	0.014*** (3.73)
Asset growth (t-2)	0.002 (0.25)	0.003 (0.43)	0.004 (0.76)	0.005 (1.00)	0.003 (0.84)	0.002 (0.67)
Asset growth (t-3)	-0.002 (0.46)	-0.002 (0.31)	-0.002 (0.37)	0.00 (0.06)	-0.004 (0.91)	-0.001 (0.45)
Inventory-sales ratio	-0.002 (0.09)	-0.085** (2.21)	-0.125 (1.63)	-0.048 (0.85)	-0.089** (2.05)	-0.080* (1.88)
Leverage	0.00 (0.30)	0.00 (0.51)	0.00 (0.21)	0.00 (1.54)	0.00 (0.37)	0.00 (0.47)
Unification			0.005*** (3.80)			0.00 (0.34)
Year	0.001*** (3.79)	0.00 (0.30)	0.001*** (4.15)	0.001** (2.12)	0.001*** (3.13)	0.00 (1.28)
Constant	-2.002*** (3.73)	-0.102 (0.29)	-1.710*** (4.15)	-1.122** (2.01)	-1.135*** (3.11)	-0.128 (1.18)
Sector dummies	yes	yes	yes	yes	yes	yes
Legal dummies	yes	yes	yes	yes	yes	yes
Year dummies	no	no	no	no	no	no
Observations	7,088	14,245	14,097	8,369	14,221	29,578
Number of cross-sections	1,459	1,461	1,461	1,453	1,462	1,464
R-squared	0.02	0.03	0.05	0.03	0.02	0.03

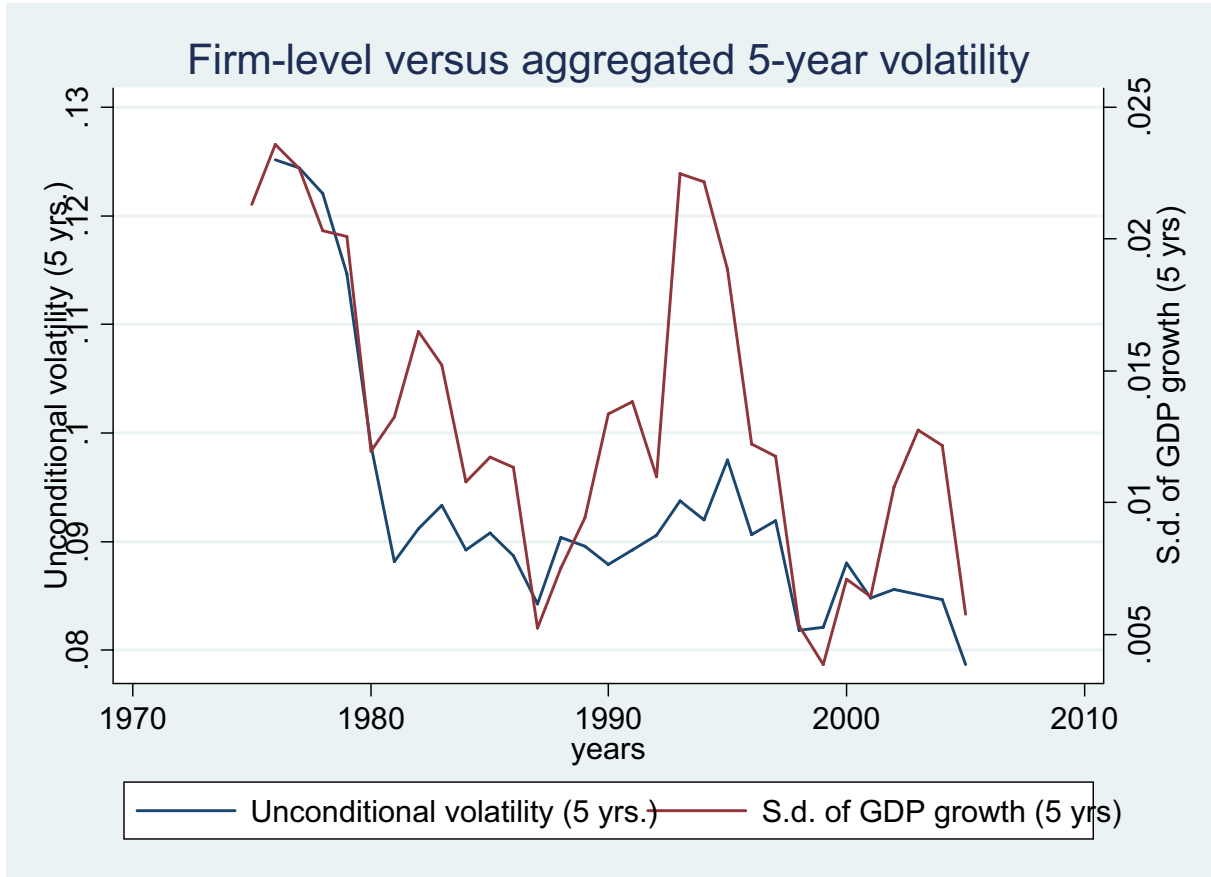
Table 5: Regression Results Heteroskedastic Regression Model

This table reports results of the heteroskedastic regression model proposed by Harvey (1976). The dependent variable is the mean firm level volatility of sales for firms in each sector. Year is a linear time trend, Unification dummy is a dummy variable which is set to “One” for the years 1991-1995 and “Zero” otherwise. ***, **, * = significant at the 1%, 5%, 10%-level.

	(1)	(2)
Mean equation		
Asset growth (t)	0.32*** (51.24)	
Asset growth ($t-1$)	0.09*** (18.55)	
Asset growth ($t-2$)	-0.02*** (-3.53)	
Listed (0/1)	-0.0004 (-0.26)	
Inventory-sales ratio (%)	-0.46*** (-11.52)	
Leverage	-0.00 (-0.98)	
Large (0/1)	0.001 (1.22)	
Manufacturing (0/1)	-0.001 (-0.57)	
Services (0/1)	-0.001 (-0.39)	
Construction (0/1)	0.005 (0.73)	
Year	-0.00 (-0.08)	-0.00 (-0.69)
Unification (0/1)	-0.001 (-0.08)	0.00 (0.09)
Variance equation		
Asset growth (t)	0.86*** (8.22)	1.01*** (9.32)
Asset growth ($t-1$)	0.57*** (5.38)	0.50*** (4.64)
Asset growth ($t-2$)	0.15 (1.44)	0.08 (0.78)
Listed (0/1)	0.13** (2.17)	0.08 (1.42)
Inventory-sales ratio (%)	0.23 (1.03)	-0.95*** (-4.66)
Leverage		
Large (0/1)	-0.18*** (-5.77)	-0.14*** (-5.10)
Manufacturing (0/1)	-0.03 (-0.42)	-0.02 (-0.30)
Services (0/1)	0.001 (0.01)	0.04 (0.49)
Construction (0/1)	1.32*** (11.52)	1.25*** (11.75)
Year	0.001 (0.97)	0.003* (1.98)
Unification	-0.05 (-0.90)	-0.04 (-0.85)

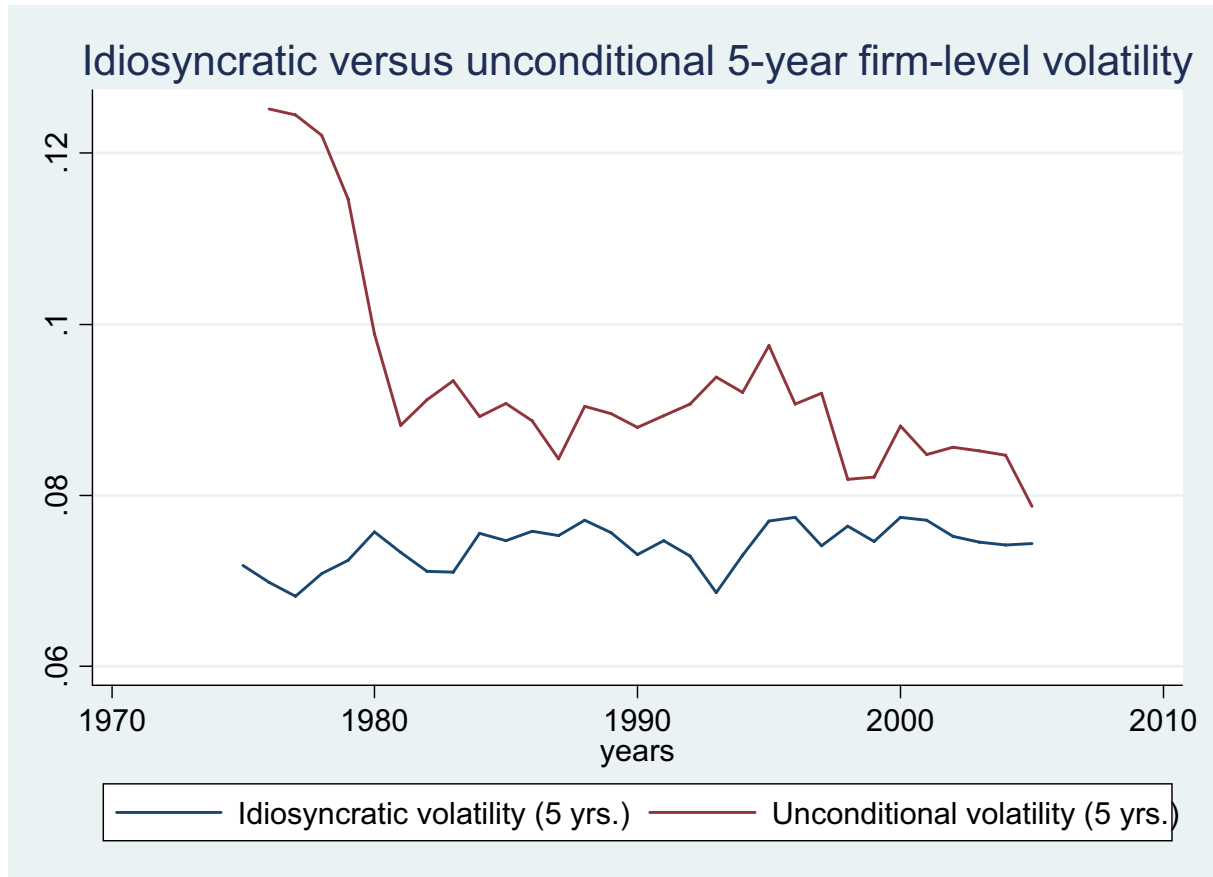
Graph 1: Unconditional Firm Level Versus Aggregated Volatility

Unconditional firm level output volatility is the median volatility of firm level sales growth over a five-year window. Aggregated output volatility is the corresponding volatility of GDP growth.



Graph 2: Conditional Versus Unconditional Firm Level Volatility

Unconditional firm level output volatility is the median volatility of firm level sales growth over a five-year window. Conditional firm level output volatility is the corresponding volatility of the residual sales growth using the method proposed by Pesaran (2006) to control for macroeconomic developments.



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