Inventory changes and future returns

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ABSTRACT

We find that the negative relation between accruals and future abnormal returns documented by Sloan (1996) is due mainly to inventory changes. We propose three explanations for this result, derived from the prior literature, but find evidence inconsistent with all three explanations. To assist future investigations in formulating additional explanations, we document several empirical regularities for extreme inventory change deciles. We speculate that demand shifts explain our results, and examine the feasibility of alternative reasons for the stock market's apparent inability to recognize the impending profitability reversals. Our evidence is consistent with earnings management masking the implications of demand shifts.

Inventory changes and future returns

Sloan (1996) documents a startling finding: investing long/short in firms in the bottom/top decile of accruals (scaled by average beginning and ending total assets) generates a hedge portfolio return of about 10 percent in the following year, and about 5 percent and 3 percent in the two years after that. He concludes that the stock market fails to recognize that accruals and cash flows, the two components of reported earnings, have different persistence. As a result, firms with high (low) accruals, or low (high) cash flows, report earnings in the following year that are predictably lower (higher) than market expectations, and stock prices move accordingly. This result has been investigated extensively in the recent literature, and the collective evidence suggests that while managers recognize the implications of accruals (e.g., Beneish and Vargus 2001), stock prices and relatively sophisticated stock market participants do not (e.g., Bradshaw, Richardson, and Sloan 2001).

Our first objective is to identify the components of Sloan's accrual measure that are primarily responsible for this apparent market inefficiency. We find that inventory changes represent the one component that exhibits a consistent and substantial relation with future returns. Our second objective is to understand how inventory changes, a seemingly innocuous item (since inventory acquisitions represent accruals that affect operating cash flows but not earnings), are linked to subsequent abnormal returns. We first propose three explanations with testable predictions, which are derived from results documented in the prior literature, but find evidence inconsistent with all three explanations. We then switch to documenting empirical regularities and fashioning an explanation that is potentially consistent with our evidence. Since limitations of available data hamper our efforts to directly test this conjecture, our contribution lies in laying the groundwork for subsequent investigations. The following are some key empirical regularities we document for extreme inventory change firms. First, firms with inventory increases (decreases) experience higher (lower) profitability, growth, and stock returns over the prior five years, but those trends reverse after the extreme inventory change. Second, firms with inventory increases (decreases) experience inventory decreases (increases) in the prior year, even though profitability increases (decreases) in both years. Third, the abnormal returns observed after the inventory changes are concentrated at subsequent quarterly earnings announcements and are related to predictable earnings "surprises" reported at those announcements. Fourth, quarterly COGS/Sales and SG&A/Sales exhibit similar patterns, including unusual fourth quarter changes.¹ Fifth, LIFO firms with inventory increases represent one subgroup of extreme inventory change firms that exhibits abnormal returns are observed for extreme deciles of changes in all three inventory components—raw material (RM), finished goods (FG) and work-in process (WIP)—with the highest abnormal returns observed for changes in RM inventory.

We conjecture that demand shifts cause both the inventory changes and related profitability reversals we observe. Firms with prior increases (decreases) in profitability and demand are projected to continue that trend, but for some of these firms actual demand may fall short of (exceed) projected demand, and this imbalance between sales and production/purchases results in inventory increases (decreases). Demand shifts for these firms presage a reversal in profitability trends, but the stock market does not fully recognize this reversal until the following year because the implications of the demand shift are not revealed in contemporaneous reported profitability. We consider earnings management and the impact of varying production levels on

¹ COGS, represents Cost of Goods Sold (or manufacturing costs, incurred in-house and/or paid to suppliers), and SG&A, represents Selling, General, and Administrative (or non-manufacturing) expenses.

fixed manufacturing overhead absorbed in COGS (e.g., Jiambalvo, Noreen, and Shevlin 1995) as potential reasons why the impending reversals are masked in reported profitability.

This paper is related to the extensive recent literature that has probed various aspects of the Sloan (1996) result and generally confirmed its robustness (e.g., Ahmed, Nainar, and Zhou 2001, Ali, Hwang, and Twombley 2000, Barth and Hutton 2001, Bradshaw, Richardson, and Sloan 2001, Collins and Hribar 2001 and 2002, Fairfield, Whisenant and Yohn 2001, Richardson and Tuna 2001, Richardson et al. 2001, Tarpley 2000, and Xie 2001). Three studies, Hribar (2000), Chan et al. (2001) and Zach (2001), also consider the impact of inventory changes.²

We describe our samples and variables in section 1 and document support in section 2 for our claim that Sloan's result is due primarily to inventory changes. In section 3, we examine the three explanations that are derived from results in prior studies. Section 4 contains the empirical regularities we uncover for extreme inventory change firms, and section 5 discusses our demand explanation. Section 6 concludes.

1. Samples and variables

Our sample extends from 1970 to 1997 (years as defined by Compustat) and consists of 39,315 observations. Financial statement (stock return) data are obtained from the 1998 (1999) edition of Compustat (CRSP). To maintain consistency with Sloan (1996), we include only NYSE and AMEX firms, and use the same variable definitions (details of all variables used in this study are provided in the Appendix).³ We require that the following data items be non-missing for a firm-year to be included in our sample: accruals as defined in Sloan (1996), change in accounts payable, change in accounts receivable, depreciation expense, change in inventory,

² Despite differences across studies in the samples studied and methodologies employed, the robustness of our primary finding regarding the importance of inventory changes is confirmed in all studies.

³ Exchange membership provided in Compustat relates only to the status as of 1998, whereas the CRSP event file provides the history of exchange membership. We use CRSP membership data when building our sample.

next year's earnings and 12-month size-adjusted returns (described below). For many of our tests, we use decile ranks for the different independent variables, to reduce the impact of outliers and to maintain consistency with the prior literature examining accrual anomalies. These decile ranks are constructed based on each year's distribution.⁴

Size-adjusted return (SAR) represents the difference between the firm's buy-and-hold return and the buy-and-hold return on a value-weighted portfolio of firms in the same CRSP size decile. Size deciles are determined by the distribution of market values of all NYSE/AMEX firms at the beginning of the calendar year. SARs are computed over two holding periods: a) 12-month holding periods, beginning four months after the fiscal year end, and b) 3-day windows around quarterly earnings announcements including the day before, the day after, and the day of the earnings announcement (as reported in Compustat).

Given the volume of results generated, we only report details of the most important results. Other results are summarized where relevant and details of those results are available upon request. Also, given the descriptive nature of the empirical regularities in section 4, we eschew tests of statistical significance for that evidence and find it convenient to present some of it as plots, rather than in tabular form.

2. Importance of inventory changes

The first row in Table 1, Panel A, provides the mean size-adjusted abnormal returns (SAR_{t+1}) earned by different accrual deciles over the 12 month period beginning 4 months after year 0. The 4 columns labeled "lowest" to "highest" provide the mean SAR_{t+1} earned over the pooled sample for the two bottom and two top deciles. (The six intermediate deciles exhibit

⁴ These distributions, which combine firms from the same year as defined by Compustat (fiscal years ending between June of this year and May of the following year), are not available to form portfolios until all firms have reported their annual results. While this is a potential source of bias, extensive sensitivity analysis we conducted suggests that in fact there is little bias created. See also Tarpley (2000).

abnormal returns that vary monotonically between the two extremes, and are not reported here for brevity.) These SAR_{t+1} range from a mean of 3.3 percent for the most negative accruals to a mean of -5.98 percent for the most positive accruals. The mean SAR_{t+1} for the lowest decile minus the mean SAR_{t+1} for the highest decile, which is 9.28 percent, is reported under the "hedge" column. The value of 8.84 percent reported under the "year-by-year hedge" column provides the mean of the 28 annual hedge returns earned in each sample year, and the value of 22 under the "# years" column indicates that the hedge returns are positive in 22 out of 28 years. The indicated statistical significance of the mean annual hedge return (# of years), which is determined by a t-test (sign test) based on the magnitude (sign) of the 28 annual hedge returns, confirm that the returns are significant at the 1 percent level. These results are consistent with those reported in Sloan (1996).

To identify the components of Sloan's accrual measure that are relatively more important for his result, we report SAR_{t+1} for deciles of different accrual components. Items that appear with a negative sign in the accrual measure (e.g. depreciation) are multiplied by -1 to align the sign of hedge returns across components. Given potential correlation among the different accruals components, these univariate results are for illustrative purposes and help to identify the component that provides the largest and most consistent abnormal returns; the partial effect of each component is better described by the regression results discussed later.⁵

The next three rows in Panel A consider an initial partition of Sloan's accrual measure: changes in non-cash current assets (Δ CurAss), minus changes in current liabilities other than taxes payable and the current portion of long-term debt (- Δ CurLia), and minus depreciation and

⁵ Examination of pairwise correlations among accruals and the different accrual components considered suggest that accruals are most closely related to changes in accounts receivable and inventories, and depreciation. While most pairwise correlations between accrual components are significant, the strongest correlations observed are between changes in accounts receivable and inventories.

amortization (-Deprec), all scaled by average total assets. Changes in current assets generate the highest hedge returns in year +1, and appear to be the component that is the primary source of abnormal returns for Sloan's accrual measure. The hedge returns and year-by-year results are similar to those reported for accruals. While the sign of the hedge returns for changes in current liability are consistent with those observed for accruals, the smaller magnitudes reported (hedge return of 1.34 percent), the fewer # of years (17 out of 28 years) with positive hedge returns, and the lack of statistical significance suggest this component is relatively unimportant for Sloan's result. Finally, while the depreciation deciles exhibit statistically significant returns consistent with the accrual strategy, the magnitude of the hedge portfolio returns and the # of years with consistent hedge returns are less convincing than those for current asset changes.

To further probe the source of abnormal returns earned by accrual portfolios, we examine portfolios based on deciles of the following individual current asset and current liability accounts: changes in accounts receivable (Δ AccRec), inventory (Δ Invent), and all other items of current accruals (Other Acc), and minus changes in accounts payable (- Δ AccPay). The results reported in the next four rows of Table 1, Panel A, suggest that inventory changes represent the component of accruals that provides the largest hedge returns. Not only are these hedge returns larger than those for accruals (11.39 percent versus 9.28 percent), the hedge return is positive in 27 of the 28 years examined.⁶ While the negative hedge returns earned by the accounts payable deciles appear inconsistent with the Sloan hypothesis, the regression results reported next indicate that this is due to omitted correlated accrual components.

⁶ The lone year with negative hedge returns (of –4 percent) is 1983. We were unable to discern a relation between the magnitude of hedge returns in each year and various potentially relevant factors, such as economic conditions and the cutoff values used to form extreme inventory change deciles.

The results of estimating regressions of SAR_{t+1} on different accrual components are reported in Panel B, of Table 1.0 allow comparisons with the hedge portfolio results in Panel A, the deciles of the different regressors are transformed to range between 0 (lowest) and 1 (highest). Ignoring correlations among regressors, this regression analysis is equivalent to examining hedge returns (with the signs reversed, since hedge returns are based on SAR_{t+1} for the lowest minus SAR_{t+1} for the highest decile), provided abnormal returns are related linearly to the different decile ranks. Since the abnormal returns/decile rank relations are in fact non-linear, the hedge portfolio returns deviate from the regression coefficients. For example, estimating a bivariate regression of abnormal returns on the inventory change deciles results in a lower slope (7.26 percent) than the hedge return reported in Table 1 (11.39 percent). In Panel B, we report the coefficients from pooled regressions, the mean of the coefficients from annual regressions, and the # of years that the coefficient in the annual regressions has a negative sign.

The results of regression 1 confirm the importance of changes in current assets, relative to changes in current liabilities and depreciation, when explaining variation in SAR_{t+1}. The pooled coefficients are consistent with the year-by-year results (mean annual coefficients and # of years with negative sign) and these results are generally comparable to the corresponding hedge results reported in Panel A. Moving to the accrual components considered in regression 2, again the overall tenor of the results reported in Panel A is maintained. While the coefficient of 7.5 percent on change in inventory is considerably lower than the 11.39 percent hedge return in Panel A, that difference is due entirely to the above-mentioned nonlinear relation between abnormal returns and inventory deciles, not due to the presence of other accrual components in the multiple regression. Note that the coefficient on changes in accounts payable is now negative

(though insignificant) indicating that the negative hedge returns obtained in Panel A for this component are due to correlation with other accrual components.

We repeat the regression analyses reported in regression 2 using the Mishkin (1983) test framework. Sloan (1996) and Hribar (2000) both use this framework to a) determine the implications of accruals and its components in year 0 for earnings in year +1, and b) check if market prices reflect correctly those implications. Those papers provide additional details of the test and the inferences that can be drawn. Our results confirm our conclusion that the market misinterprets the implications of inventory changes, and abnormal returns are best explained by this accrual component. Specifically, the permanence of inventory changes, measured by the coefficient in the earnings regression (=0.0236), is much smaller than the market's assessment of that permanence, measured by the coefficient in the returns regression (=0.183). In effect the market assumes that the earnings impact of inventory changes is substantially more permanent than it actually is. While some of the comparisons for other accrual components are also statistically significant, the inventory change component is clearly the most significant (largest test statistic) and appears to be the most important driver of the mispricing documented by Sloan.

3. Testing explanations for inventory results derived from prior research

We consider three potential explanations for our inventory result, each of which is based on results of prior research that identify a less visible factor which is associated with profitability reversals and future abnormal returns. We consider the possibility that the observed relation between inventory changes and subsequent abnormal returns is because inventory changes are correlated with these factors. While the first explanation is derived from research unrelated to the mispricing of accruals, the remaining two explanations are derived from research that proposes alternative explanations for Sloan's finding regarding accruals predicting abnormal returns. The first explanation is based on Titman, Wei, and Xie (2001), which documents a negative relation between capital expenditures and subsequent abnormal returns. They conclude that firms with high (low) profitability in prior periods generate more (less) free cash flows, which results in reduced (increased) future profitability and negative (positive) abnormal returns because of increased (reduced) investments in negative net present value capital expenditures. Second, the results reported in Fairfield, Whisenant, and Yohn (2001) suggest that accruals (primarily working capital accruals) are related positively to changes in net operating assets, which are in turn negatively related to future profitability. Finally, Tarpley (2000) investigates the possibility that positive (negative) accruals follow periods of high (low) growth, that the stock market erroneously extrapolates those growth rates into the future, and stock returns decline (increase) predictably when growth rates mean revert.

To investigate the first two explanations, we report in Panel A of Table 2 the abnormal returns earned by decile portfolios formed using growth in net plant, property and equipment (PPE) and growth in net assets other than PPE and working capital (ONA). We observe fairly large hedge portfolio returns (7.67 percent and 6.05 percent for PPE and ONA, respectively) and remarkably consistent year-by-year performance (23 and 24 years out of 28). We confirm that these two measures of noncurrent accruals are in fact positively correlated with inventory changes. To test the separate ability of inventory changes to predict future abnormal returns, after controlling for these two non-operating accruals, we estimate regression 3, reported in Panel B of Table 2, which includes deciles of these two noncurrent accrual measures to the components of accruals considered in regression 2 (reported in Table 1, Panel B). While the pooled coefficient, mean of annual coefficients, and the number of years with significant mean coefficients for the inventory change variable decline slightly from the levels reported in regression 2, they remain

large and significant. Interestingly, growth in PPE and growth in other net assets are also equally important. These results suggest that although growth in both PPE and other net assets predict future abnormal returns, those relations are separate from the relation between inventory changes and abnormal returns we claim drives the Sloan result.

Consistent with the second explanation that relies on a negative relation between changes in entity size and profitability shifts, Zach (2001) argues that a portion of the Sloan result is due to increases (decreases) in working capital accruals derived from balance sheet numbers reflecting mergers/acquisitions (divestitures), which in turn appear to be related to subsequent increases (decreases) in profitability.⁷ To investigate the possibility that our inventory change results are due to acquisitions and divestitures, we examine abnormal returns for inventory deciles for three subgroups (based on Compustat footnote AFTNT 1): those with mergers/acquisitions, those with divestitures, and those with neither. Consistent with the more extensive analysis of this potential methodological bias carried out by Zach (2001) for accrual deciles, we find very similar negative inventory change/abnormal return relations across all three subgroups. We also formed deciles based on changes in the ratio of inventory to total assets, to eliminate the correlation between our inventory change measure and entity changes. Despite the dampening effect caused by changes in noncurrent assets being negatively related to future abnormal returns (since noncurrent assets are reflected in total assets, the scaling variable), we find significant abnormal returns that are about half the returns for Δ Invent in Table 1, Panel A.

To investigate the third explanation, relating to the stock market overestimating (underestimating) future growth for high (low) growth firms, we considered the both historic and forecasted growth rates but could only discern a weak ability to predict future returns: the hedge

⁷ We confirm the results of prior research (e.g. Hribar 2000) that similar abnormal returns are observed when portfolios are formed based on inventory changes from the cash flow statement (available only for years after 1987), which are less likely to be biased by entity changes than the balance sheet changes we consider.

returns for extreme growth deciles were uniformly low, and in regression analyses, we observed insignificant coefficients on growth. More important, there is little impact on the regression coefficient on inventory changes. As an aside, we find that the relation between inventory changes and abnormal returns is a function of forecast growth. Specifically, hedge returns for extreme inventory deciles are smaller for firms with low forecasted growth in earnings per share. Tarpley (2000) reports a similar finding regarding the interaction between growth and the accruals/abnormal return relation.

4 Empirical regularities associated with extreme inventory change deciles

In the absence of promising explanations for the relation we document between inventory changes and abnormal returns, we focus on extreme inventory change deciles and seek to establish some descriptive features of those firms. These regularities should assist future research in formulating additional explanations. We provide evidence on the annual (between years -5 and +2) and quarterly (between years -1 and +2) time-series of profitability, growth, and abnormal returns for these two groups in sections 4.1 and 4.2, respectively. Some other regularities are reported in Section 4.3.

4.1 Annual time-series

The abnormal returns reported in the first two rows of Table 3 indicate a reversal around year 0 in the abnormal returns of firms with extreme inventory changes. Note that the abnormal returns refer to the 12-month period that begins 4 months *after* the end of that year. For example, the abnormal returns under year 0 correspond approximately to the period referred to as year +1. The extreme inventory increase (decrease) group exhibits large positive (negative) abnormal returns in the 12-month period after years -4 and -3, which then decline for the 12-month period after year 0. While the magnitudes of these abnormal returns for the lowest decile are similar to those

reported for the period following years -2 and -1, the magnitudes for the highest decile are slightly higher than (similar to) those following year -2 (year -1). The magnitudes of abnormal returns then decline over the 12-month period following years +1 and +2.

To understand the determinants of this sharp reversal in stock returns, we provide in Table 3 the median values for two growth measures, for sales and capital expenditures, and two profitability measures, return on equity (ROE) and return on assets (ROA), for the two extreme inventory deciles.⁸ Note that the year 0 numbers, especially for sales growth and capital expenditures, are unreliable because the selection of extreme inventory change groups tends to include a disproportionate number of firms making acquisitions (divestitures) in the inventory increase (decrease) group. The two profitability measures exhibit trends similar to those reported for the abnormal returns, especially the reversal around year 0. (Recall that the abnormal returns refer to the 12-month period following that year and should be aligned with the growth and profitability measures for the next year.) The two growth measures also exhibit similar trends prior to year 0, but the growth trends seem to extend for one more year, before reversing.

4.2 Quarterly time-series

To obtain a better understanding of the stock price movements around year 0, we plot in Figure 1, Panel A, seasonally differenced quarterly earnings, scaled by end of quarter stock price, for the quarters around year 0. This measure serves as a simple proxy for the unexpected component of earnings reported in those quarters. To smooth the effect of outliers, we Winsorize values above +1 (below -1) to +1 (-1). The inventory increase group exhibits values slightly above zero indicating small but sustained increases in quarterly earnings during years -1 and 0. The inventory decrease group exhibits values considerably below 0, indicating sustained declines

⁸ The mean results are fairly similar and not reported here. Also, we examined a variety of other financial ratios and obtained results similar to the representative measures reported here.

of considerable magnitude over those two years. The two patterns reverse after year 0, with the inventory increase (decrease) firms showing profitability declines (sharp increases) over year +1, culminating in the largest changes in the fourth quarter of year +1. The trend continues over year +2, but the magnitudes are considerably lower than those in year +1.

Examination of the size-adjusted returns around 3-day quarterly earnings announcement windows, plotted in Panel B of Figure 1, provides the following results. When calibrating the 3-day announcement window returns, note that announcement window returns are positive on average (e.g., Ball and Kothari 1991); i.e. the appropriate benchmark to determine positive and negative news is the unconditional mean of about 0.33 percent, not zero. First, the patterns observed for these 3-day returns resemble approximately the patterns in Panel A, indicating that seasonally-differenced quarterly earnings correspond roughly to the stock market's perceived earnings surprises. Second, even though the abnormal returns during year 0 continue their historic trend, stock prices for inventory decreases begin to anticipate the upcoming reversal by the fourth quarter of year 0. Finally, the stock price reversal observed during years +1 and +2 appears to be related to disclosures made during the corresponding quarterly earnings announcement window, because a disproportionately high fraction (over 30 percent) of the annual abnormal returns (reported as SAR_{t+1} under years 0 and +1 in Table 3) is concentrated in the corresponding four quarterly windows.

We turn next to the quarterly time-series of unexpected changes in COGS and SG&A, scaled by Sales. In addition to describing quarterly trends in profit margins around year 0, where margins represent one component of the earnings surprises reported in Figure 1, Panel A, these

two measures offer evidence of potential earnings management.⁹ Specifically, earnings management is supported by unusual patterns consistent with hypothesized incentives, especially if they occur around fourth quarters and if similar patterns are observed for both COGS and SG&A. Based on our finding that a seasonal random walk model (without drift) consistently describes both ratios better than other time-series models we considered, we use seasonal differences to measure unexpected variation in quarterly levels of both ratios.¹⁰

Seasonally differenced quarterly COGS/Sales and SG&A/Sales could deviate from zero because of factors other than earnings management. First, both ratios are affected by variation in selling prices, which could reflect exogenous demand shifts or endogenous pricing decisions designed to influence demand. Second, to the extent that COGS and SG&A contain fixed elements that do not vary much from year to year with variation in units produced and sold, respectively, changes in units produced and units sold affect seasonal differences in both ratios. Third, to the extent that some items of COGS and SG&A for interim quarters reflect estimates, variation in interim quarters should be less than that in the fourth quarter, when adjustments are made for deviations between estimates and actual amounts for such items. Note that while these factors might often cause random variation in seasonal differences for both ratios, in certain cases these factors could cause patterns similar to those expected for earnings management: unusual fourth quarter variation and similar patterns for COGS and SG&A.

⁹ Traditional measures of earnings management that are based on estimated discretionary accrual measures are unlikely to be useful in this case because of the high correlation between our inventory change measure and those measures of discretionary accruals.

¹⁰ We examined plots of both ratios for different subsamples and determined a distinct seasonal pattern with the ratios being relatively similar across the interim quarters for each firm, but substantially lower in the fourth quarter. We estimated a variety of seasonal quarterly time-series models, and examined the distribution of forecast errors for bias and precision.

Figure 2, Panels A and B provide the median ratio of seasonal differences at the quarterly level for COGS/Sales and SG&A/Sales for the two extreme inventory deciles.¹¹ The results in Panel A suggest an increase (decrease) in COGS/Sales for the inventory decrease (increase) group that extends to the fourth quarter of year 0, and a reversal after that.¹². Consistent with the abnormal return patterns noted in Panel B of Figure 1, the profit reversal for inventory decreases is evident slightly earlier: it occurs in the fourth quarter of year 0, rather than in the first quarter of year +1.The most striking aspect of these plots is the accentuated patterns observed at the fourth fiscal quarters, especially in years –1 and +1.

To understand better the interpretation of these plots, consider the COGS/Sales plot for inventory increases. The ratio declined in each quarter of year -1, relative to the same fiscal quarter in year -2, with the largest decline (of approximately 0.5 percent of sales) occurring in the fourth quarter of year -1 (labeled quarter -4). The ratio continued to decline in all four quarters of year 0. While the magnitudes of declines are smaller than those in year -1, this plot indicates that profit margins continued to increase in year 0 from their elevated levels in year -1, which in turn were higher than those in year -2. In particular, even though the changes observed during the fourth quarter of year 0 are small, about -0.2 percent of sales, this implies that the unusually low COGS/Sales ratio achieved in the fourth quarter of year -1 and 0 are for the most part surrendered in the four quarters of year +1, indicated by the large increases in COGS/Sales in quarters +1 to +4.

¹¹ To determine the statistical significance of these unexpected changes in the ratios, we computed the p-values associated with the sign-rank test for each point in Figure 2. Most of the points are associated with very significant p-values. Similar results are observed when examining mean results (suitably Winsorized to mitigate the effect of outliers) and associated t-tests.

¹² Examination of these ratios over years before year -2 indicate a trend that begins as far back as year -5.

The patterns observed for SG&A/Sales are similar to those observed for COGS/Sales. Again, the one exception is that the switch from positive to negative deviations for the inventory decrease decile does not occur in the first quarter of year +1; in this case it is delayed and occurs in the third quarter of year +1. While the magnitudes of unexpected changes in SG&A/Sales are considerably smaller than those for COGS/Sales they are commensurate with the relative magnitudes of SG&A and COGS (the levels of COGS/Sales are on average about 72 percent, which is about four times the average level of SG&A/Sales of about 19 percent). As with COGS/Sales, the sharpest changes are observed in the fourth quarters of years –1 and +1.

To place the observed magnitudes of deviations for COGS in context, we examine changes in the level of inventories. Since the components of inventory (FG, RM, and WIP inventory) are available only at the annual level, not at the quarterly level, we report the medians of annual levels of these components (in Figure 3, Panel A) and medians for the quarterly seasonal differences for total inventory (in Figure 3, Panel B). Whereas the inventory change deciles are formed by first computing annual differences in inventory levels and then scaling by average total assets, in these plots we first scale inventory levels by total assets and then examine levels and changes for that ratio to minimize the effect of changes in the scale of the firm. While the entire sample is represented in Panel B, only firm-years with available inventory component data are represented in Panel A (see Table 4, Panel A, for sample sizes for these components).

The results in Figure 3, Panel A, suggest that median levels for FG, RM, and WIP inventory are about 11, 8, and 6 percent of total assets, respectively. Prior to year –2, the corresponding component levels for extreme inventory increases and decreases are similar. The year 0 changes for all three components are large, corresponding to approximately 1 percent of total assets for all cases except for WIP inventory for the inventory increase decile, which

exhibits a smaller change. Assuming that sales is approximately equal to total assets, the average of the four quarterly values of median unexpected changes in COGS/Sales documented for year 0 in Panel A of Figure 2, which is less than 0.5 percent of sales, is much smaller than the changes in inventory documented here (about three percent of total assets). Note that the changes in year -1 and +1 are in the opposite direction to those of the year 0 changes.

The trends reported in Panel B confirm that while inventory levels decrease (increase) slightly for inventory increases (decreases) in year -1, that trend reverses sharply in year 0, and then reverses again in year +1. To interpret the seasonal changes that occur after year 0, note that although it appears that the extreme inventory increase decile exhibits an increase in inventory in quarter +1, that is an increase over the level in quarter -3, but a decrease relative to the level in quarter 0. In sum, the quarterly plots suggest smooth changes during the years, rather than abrupt changes in inventory levels during fourth quarters.

4.3 Other regularities

We discuss next some other interesting features of extreme inventory change deciles. To identify the relative importance of the three components of inventory—RM, WIP and FG—for the prediction of future abnormal returns, we report in the first three rows of Panel A of Table 4, the mean abnormal returns earned by deciles formed using changes in each of the three components. Our results reveal higher hedge returns for changes in RM and FG inventories (10.82 and 9.35 percent, respectively), which exceed substantially the hedge returns for changes in WIP inventory (6.45 percent). Note that the sample sizes are reduced considerably when inventory components are examined, because of missing data.¹³ To confirm that these findings

¹³ Since the subsamples used for the three inventory component portfolios are different, and differ also from the sample for inventory change portfolios, the four sets of results are not strictly comparable. To identify potential differences in these four samples, we computed the inventory change portfolio abnormal returns for each of the three inventory component subsamples. The abnormal returns earned are 10.23 percent, 11.11 percent, and

are not due to correlation with other accrual components, we report in Table 4, Panel B, the results of regression 4, which is obtained by replacing inventory change deciles in regression 2 (see Table 2, Panel B) with deciles of changes in RM, WIP, and FG inventory. The relative importance of RM inventory changes for future abnormal returns is even more exaggerated in the regression results.

To provide evidence on the importance of over/underproduction (the effect of varying production levels on fixed manufacturing overhead per unit absorbed in COGS), we focus on abnormal returns for inventory deciles for firms in the retail and wholesale industry (SIC code=5xxx). Since over/underproduction is not relevant for this group, any differences between the features of this group and those for firms with manufactured inventory could potentially be attributed to over/underproduction effects. Our results indicate no differences between this sample and our overall sample. The hedge returns earned by deciles of inventory change for retailers, reported in the fourth row of Table 4, Panel A, appear similar to those reported in Table 1, Panel A, for the overall sample. We also find that the patterns for seasonally differenced quarterly COGS/Sales (and SG&A/Sales) resemble substantially the results reported for the overall sample in Figure 2.

Since inventory changes based on year-end values could reflect unusual inventory changes that occur in the fourth quarter, we replicated the analysis of mean SAR_{t+1} for inventory change deciles using seasonal differences for inventory levels as of the first, second, and third fiscal quarters. We observe levels of hedge returns that are similar to those reported in Table 1, Panel A, for inventory changes based fourth quarter numbers. These results are inconsistent with unusual fourth quarter inventory changes being responsible for the observed abnormal returns.

^{11.37} percent for the RM, WIP, and FG subsamples. As these returns are similar to those observed for the larger sample, we believe meaningful comparisons can be made across all four samples.

We also repeated the analysis of seasonally differenced quarterly COGS/Sales and SG&A/Sales, similar to Figure 2, for extreme inventory change deciles based on second quarter numbers and observed evidence of unusual patterns, especially for COGS/Sales, around fiscal fourth quarters before and after the inventory change quarter. These results suggest that there is considerable overlap among firms represented in the extreme inventory change deciles based on year-end values and those included in extreme deciles based on interim quarter values.

Our final analysis, relating to a comparison of LIFO and non-LIFO firms, is motivated by the findings in Hribar (2000) regarding the transparency of the low permanence of earnings created by LIFO liquidations (inventory decrease firms) and the lower mispricing he observes for those firms. While LIFO liquidations are highlighted more clearly and thus their income effects are less likely to be misinterpreted by the stock market, the income effect in year 0 should be positive for LIFO firms with inventory decreases (since COGS/Sales should decrease when older, lower inventory costs are expensed), which is contrary to the year 0 income effect noted for inventory decreases in Figures 1 and 2. Our results, described below, indicate that LIFO firms with inventory decreases exhibit abnormal returns and COGS/Sales patterns similar to the patterns observed for other firms with inventory decreases. Quite unexpectedly, LIFO firms with inventory increases exhibit positive abnormal returns that are markedly different from those for other firms with inventory increases.

The bottom three rows of Panel A in Table 4 provide the abnormal returns earned in year +1 by each of the following three groups of firms: a) LIFO firms, that use only the LIFO valuation method, b) mixed firms, that use LIFO and other valuation methods, and c) non-LIFO firms, that use methods other than LIFO. While the inventory decrease firms (under the "lowest" column) exhibit positive abnormal returns that are similar in magnitude across all three groups,

the inventory increase firms (under the "highest" column) are clearly different across the three groups: the non-LIFO group exhibits the most negative abnormal returns and the LIFO group actually earns positive abnormal returns.

To understand better this unexpected result observed for LIFO firms with inventory increases, we report in Figure 4 the time series of median COGS/Sales for the LIFO and non-LIFO partitions of extreme inventory change firms. (The mixed partition exhibits patterns in between these two partitions, and is deleted from the plot.) The results reported in Figure 4 are consistent with the abnormal return results reported above. Whereas the two non-LIFO groups and LIFO firms with inventory decreases exhibit patterns very similar to those presented for our overall sample (Figure 2, Panel A), LIFO firms with inventory increases exhibit patterns that are considerably muted: the deviations are generally closer to zero, relative to the other three groups.

5. Conjectures relating to demand shifts and earnings management

Many features of the patterns reported in Figures 1 through 4 are generally inconsistent with the three explanations considered in section 3, where inventory changes are merely symptoms of correlated underlying factors that cause profitability changes. The very distinct profitability patterns (SUE, COGS and SG&A) observed for fourth quarters of years –1 and +1 are not easily reconciled with a one-time shift in reported profitability caused in year 0 by some other underlying factor. Also, while the profitability changes are positively related to inventory changes in year 0, the two are inversely related in year –1. Finally, if the results for inventory changes reflect correlation with omitted underlying factors, why should this correlation be different for LIFO firms with inventory increases?¹⁴ We then considered other possible

¹⁴ Other features of our results are also inconsistent with the three explanations. For example, we find evidence suggesting that the abnormal returns are explained more by the dollar magnitude of inventory changes, rather than percent changes or changes relative to other firms in the same industry (which would be more relevant under those three explanations).

explanations and investigated their consistency with our findings. The most promising of those explanations is one that combines demand shifts with two potential reasons why the stock market fails to recognize fully the impending reversal in profitability.

The economics and financial statement analysis literatures have long emphasized the relations among demand shifts, inventory changes, and profitability reversals (e.g., Lev and Thiagarajan 1993, Abarbanell and Bushee 1997 and 1998). To illustrate the demand explanation, consider two groups of firms, labeled past winners and past losers, that have experienced demand and profitability increases and decreases, respectively, over the recent past. Assume that prior trend is maintained for another year, corresponding to year -1 in our analysis. The stock market's reaction is positively related to these demand shifts, and the demand increases (decreases) in year -1 for past winners (losers) cause a decline (increase) in year-end inventory. Winners (losers) then anticipate continued growth (declines) in demand into year 0, and increase (decrease) production in year 0 to satisfy this demand shift and also to compensate for the inventory declines (increases) in year -1. The subset of these winners (losers) with actual demand that is less (more) than anticipated demand in year 0 experiences inventory increases (decreases). In effect, past winners (losers) with extreme demand decreases (increases) in year 0 constitute our extreme inventory increase (decrease) deciles, and this shift in demand is also associated with profitability reversals.

In the absence of reliable data on sales that is adjusted for the considerable changes in entity and scale experienced by extreme inventory change deciles, we are unable to directly examine the role played by demand shifts. To be sure, some of our results, such as the strong relation between RM inventory changes and abnormal returns, cannot be reconciled with our demand explanation without imposing additional assumptions and structure.¹⁵ Rather than discuss the details of those assumptions, we focus on why the profitability numbers might not reflect the reversals until year +1. To recap, while the patterns for inventory levels, profitability, and abnormal returns for past winners (losers) in year –1 can be reconciled with the increases (decreases) in demand hypothesized to occur in that year, the continued increases (decreases) in profitability observed during year 0 are inconsistent with the hypothesized unexpected decreases (increases) in demand and observed increases (decreases) in inventory. In essence, why are the profitability reversals predicted for year 0 by our demand shift explanation observed with a lag, in year +1 reported numbers?

One possibility is that earnings are managed in year 0 to mask these profitability reversals. While earnings could be managed in a variety of revenue and expense accounts, one type of earnings management that is unique to inventory balances is related to inventory "misstatements". The following cost balance illustrates the impact of inventory misstatement: COGS equals beginning inventory plus inflows (inventory purchases and manufacturing costs) less ending inventory. Misstatement of inventory balances is measured relative to inventory valuations that are observed subsequently. Although it implies bias relative to this ex post value, misstatement does not imply fraud and any bias is likely to be within the discretion allowed by GAAP (which is based on reasonable ex ante judgment).¹⁶ In essence, overstated (understated) ending inventory generates understated (overstated) COGS and overstated (understated) profitability in year 0, and this effect is reversed in year +1.¹⁷

¹⁵ For example, the RM inventory increases (decreases) for past winners (losers) experiencing unexpected decreases (increases) in demand could be explained by relatively inflexible contracts with suppliers to purchase RM inventory based on anticipated demand.

¹⁶ See evidence in Beasley, Carcello, and Hermanson (1999) for incidence of fraudulent inventory misstatement.

¹⁷ Misstatements could be unintentional and due to biased management expectations. For example, in the recent boom and bust cycle experienced by technology firms, some managers may maintain optimistic expectations of

Most of our results for years 0 and +1, including the unusual fourth quarter patterns observed and the similarity in trends observed for COGS and SG&A, are generally consistent with the following description: managers of past winners (losers) facing a reversal of profitability caused by demand shifts in year 0 seek to delay reporting that reversal and maintain the prior trend by managing earnings upward (downward) in year 0. The incentives for past winners to defer reporting bad news could arise from a number of sources, including the sharp disappointment expressed by the market when growth stocks fail to maintain their prior growth (e.g., Skinner and Sloan 2001). The incentives for past losers to defer reporting good news are harder to enumerate. Despite some indication that margins for the inventory decrease decile begin to reflect the profit reversal in the fourth quarter of year 0 (Figure 2, Panel A) and the stock market partially recognizes this reversal (Figure 1, Panel B), a large fraction of the profitability reversal and abnormal return occurs in year +1. Perhaps, managers of past losers believe that the stock market reacts less negatively (than average) to additional bad news, and these negative accruals are available to boost earnings in future periods.¹⁸ Similarly, the absence of abnormal returns and muted patterns for COGS and SG&A observed for LIFO firms with inventory increases are not easily explained by earnings management.¹⁹

The second reason we consider for delayed recognition of profitability reversals, which relates to absorption of fixed manufacturing costs, is also unique to inventories. To the extent that manufacturing costs contain a fixed component that does not vary much with production

inventory valuations even after clear signs of a slowdown in that sector, because of the success they experienced in the past. While the predictions for COGS remain similar to those for the intentional misstatement case, the patterns for SG&A/Sales are not expected to resemble the patterns predicted for COGS/Sales.

¹⁸ While these conjectures refer to corporate-level incentives to mislead investors, our evidence is also consistent with earnings management occurring at the divisional level, in response to budgetary incentives.

¹⁹ We are unaware of evidence supporting the view that the costs of overstating profits are relatively higher for LIFO firms. To the extent that overstated profits are due to overstated inventory, perhaps LIFO firms avoid adding additional LIFO layers caused by overstated inventory.

levels, increases (decreases) in production levels will result in lower (higher) COGS. While this effect would cause seasonally-differenced COGS/Sales for interim quarters to vary based on *expected* production levels for the two fiscal years, seasonally differenced COGS/Sales for the fourth fiscal quarter is also affected by differences between actual and expected production in each of the two years (e.g., Jiambalvo, Noreen, and Shevlin 1995). To illustrate, consider the following description for year –1: past winners (losers) anticipate higher (lower) production levels in year –1, relative to year –2, and that results in lower (higher) COGS in interim quarters of year –1. In the fourth quarter of year –1, COGS declines (increases) even further since actual production was higher (lower) than the original estimates because of the positive (negative) unexpected demand. In essence, the patterns observed for COGS/Sales in Panel A of Figure 2 are due to innocent production decisions, and over/under production causes profitability reversals to be masked in year 0 reported numbers.

While the evidence for COGS/Sales is generally consistent with over/underproduction some other results suggest that this explanation alone is insufficient. First, we observe similar quarterly patterns for SG&A, which unlike COGS are not affected by over/underproduction. Second, we observe similar patterns for our retailer subsample for abnormal returns as well as for COGS/Sales and SG&A/Sales. Again, over/underproduction is not relevant for these firms. The different pattern observed only for LIFO firms with inventory increases is also not easily reconciled with over/underproduction.

Overall, it is feasible that demand shifts cause inventory changes and profit reversals, and the profit reversals are not clearly revealed to the stock market in reported year 0 profitability because of earnings management and over/underproduction.

6. Conclusions

We find that inventory change is the component of the accrual measure used by Sloan (1996) that is most strongly related to next year's abnormal returns. We also find that firms with inventory increases (decreases) have experienced higher (lower) levels of profitability, growth and abnormal returns over the prior 5 years, and those trends reverse immediately after the inventory change. Why inventory changes are related to profitability reversals, and why the stock market does not recognize the impending profitability shifts are two important questions that we are unable to provide definitive answers for.

We are, however, able to reject some explanations for our result. We are also able to document several interesting empirical regularities associated with extreme inventory changes. Finally, we fashion an explanation that is based on demand shifts, which we hypothesize cause both the profitability reversals and the inventory changes, and suggest two possible reasons why the impact of the profitability shift is masked in contemporaneous reported profitability. One possibility is related to earnings management, potentially by misstating inventory balances, and the other is related to variation in production levels altering COGS by affecting the amount of fixed manufacturing overhead absorbed into each unit produced. Inventory changes are related naturally to demand shifts and both possibilities rely on features unique to inventories. We hope that the exploratory evidence and demand explanation we offer will serve as a jumping off point for future research.

Appendix

Variable definitions

Compustat data item #s are indicated in parentheses next to each item, where appropriate.

Accruals = change in current assets (4) - change in cash (1) - change in current liabilities (5) + change in debt included in current liabilities (34) + change in income taxes payable (71) - depreciation and amortization expense (14), all deflated by average total assets (6).

 Δ **CurLia** = change in current liabilities (5) - change in debt included in current liabilities (34) - change in income taxes payable (71), all divided by average total assets (6). (This variable is multiplied by -1 to get the same directional effect as other components of accruals.)

 $\Delta CurAss$ = change in current assets (4) - change in cash (1), scaled by average total assets (6).

Deprec = depreciation expense (14) divided by average total assets (6). (This variable is multiplied by -1 to get the same directional effect as other components of accruals.)

 Δ AccRec = change in accounts receivable (2), deflated by average total assets (6).

 Δ AccPay = change in accounts payable (70), deflated by average total assets (6). (This variable is multiplied by -1 to get the same directional effect as other components of accruals.)

 Δ **Invent** = change in total inventory (3), deflated by average total assets (6).

Other Acc = Accruals (as defined above) minus change in accounts receivable (2) plus depreciation and amortization expense (14) minus change in inventory (3) plus change in accounts payable (70) deflated by average total assets

 ΔRM Invent = change in raw material inventory (76), deflated by average total assets (6).

 Δ WIP Invent = change in work-in-process inventory (77), deflated by average total assets (6).

 Δ FG Invent = change in finished goods inventory (78), deflated by average total assets (6).

Size adjusted return is computed by taking the raw buy and hold return, inclusive of dividends and subtracting the buy and hold return on a size matched, value-weighted portfolio of firms (obtained from CRSP). The size portfolios are based on market value of equity deciles of NYSE and AMEX firms, as measured at January 1, each year.

 $SAR_{i} = \prod_{s} (1 + r_{is}) - \prod_{s} (1 + r_{ps})$, where r_{is} and r_{ps} are the returns in *s* for firm *i* and size portfolio *p*

For annual periods monthly returns are used, and the portfolio holding periods begin 4 months after the fiscal year-end. For 3-day earnings announcement windows, daily returns are used, and the holding periods are from day -1 to +1 of the earnings announcement date (from Compustat).

Growth in Sales = current year's sales (12) minus previous fiscal year's sales, divided by previous year's sales.

Capital expenditure = Capital expenditure (30) divided by average total assets (6).

ROE = Income before extraordinary items, available for common stocks, (237) divided by ending common equity (60).

ROA = Operating income after depreciation (178) divided by average total assets (6).

SUE = The seasonal difference in quarterly primary EPS excluding extraordinary items (19), adjusted for stock splits using the Compustat adjustment factor (17), deflated by the closing price as of the third month of the quarter (14).

Growth PPE = change in net property, plant and equipment (8) divided by average total assets (6).

Growth ONA = change in ONA divided by average total assets (6). ONA is calculated by subtracting working capital (WC) and net PPE (NPPE) from net operating assets (NOA). NOA is defined as (total assets (6) – cash/equivalents (1)) – (total liabilities (181) – long term debt (9) – debt in current liabilities (34)). WC is defined as (current assets (4) – cash/equivalents (1)) – (current liabilities (5) – current debt (34) – taxes payable (71)). NPPE is net property, plant and equipment (8).

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Table 1

Analysis of accrual components (39315 firm-years from 1970 to 1997)^a

			Decile	ranking		Year	-by-year					
	Ν	lowest	2	9	highest	hedge ^b	hedge ^c # y	vears ^d				
Panel A: SAR _{t+1} for deciles of accruals and its components												
Accruals	39315	3.30%	2.25%	-2.91%	-5.98%	9.28% *	8.84%*	22*				
ΔCurAss	39315	3.23%	1.80%	-2.74%	-5.86%	9.09% *	8.82%*	23*				
-∆CurLia	39315	0.35%	0.45%	-0.86%	-0.99%	1.34%	1.27%	17				
-Deprec	39315	1.05%	1.18%	-1.23%	-3.69%	4.74% *	4.43%**	18				
∆AccRec	39315	0.69%	0.53%	-1.81%	-2.97%	3.67% *	3.62%**	18				
-AAccPay	39315	-2.09%	-0.33%	1.26%	0.65%	-2.74% **	-2.64%	13				
ΔInvent	39315	4.50%	1.81%	-1.52%	-6.89%	11.39% *	11.05%*	27*				
Other Acc	39315	2.41%	0.12%	-1.11%	-1.37%	3.78% *	3.74%*	22*				

Panel B: Regression of $\ensuremath{\mathsf{SAR}}_{t+1}$ on deciles of accrual components

Regression 1 (Pooled regression adjusted R^2 = 0.0034)

	Intercept	ΔCurAss	-∆CurLia	-Deprec
Pooled	0.077*	-0.083*	-0.035*	-0.035*
mean annual ^e		-0.080*	-0.035*	-0.032**
# of years ^f		23*	19	20**

Regression 2 (Pooled regression adjusted R^2 = 0.0042)

	-	-	-			
	Intercept	∆AccRec	-AAccPay	ΔInvent	-Deprec	Other Acc
Pooled	0.081*	-0.016**	-0.006	-0.075*	-0.034*	-0.043*
mean annual		-0.017	-0.007	-0.071*	-0.032**	-0.042*
# of years		19	16	23*	20**	22*

Notes:

- **a**. The variables are calculated as indicated below (see appendix for details) and all regressors are sorted into deciles and assigned values between 0 and 1. All changes are scaled by average total assets, unless indicated otherwise.
- $SAR_{t+1}=12$ -month buy and hold return beginning 4 months after fiscal year-end of inventory change year less corresponding size portfolio buy and hold return.
- Accruals = the change in non-cash current assets, less the change in current liabilities (exclusive of short-term debt and taxes
- payable), less depreciation expense.
- Δ CurAss = change in non-cash current assets
- -\Delta current liabilities (exclusive of short-term debt and taxes payable)
- -Deprec = minus depreciation and amortization expense
- $\Delta AccRec = change in accounts receivable$
- $-\Delta$ AccPay = minus change in accounts payable
- Δ Invent = change in total inventory
- Other Acc = accruals minus change in accounts receivable and inventory plus change in accounts payable plus depreciation and amortization expense
- **b**. Hedge is the difference in size-adjusted return between the lowest and highest decile.
- c. Mean hedge reports the mean value of annual hedge portfolio returns. Statistical significance is based on a t-test using magnitudes of the 28 annual hedge returns.
- **d**. # of years reports the number of years in which the hedge portfolio return is positive. Statistical significance is based on a sign test using signs of the 28 annual hedge returns.
- e. Mean annual reports the mean value of annual coefficient estimates. Statistical significance is based on a t-test using magnitudes of the 28 annual coefficient estimates.
- **f**. # of years reports the number of years in which the coefficient is negative. Statistical significance is based on a sign test using signs of the 28 annual coefficient estimates.
- *: significant at the 1% level;
- **:significant at the 5% level;

Table 2

Analysis of three explanations: (39,315 firm-years from 1970 to 1997)^a

			Decile	ranking		Year-by-year				
	Ν	lowest	2	9	highest	hedge ^b	hedge ^c # years ^d			
Panel A: Mean SAR _{t+1} for deciles of growth in PPE and ONA										
Growth PPE	39290	3.11%	3.10%	-0.58%	-4.56%	7.67% *	7.46%* 23*			
Growth ONA	39253	2.30%	2.11%	-2.08%	-3.75%	6.05% *	6.03%* 24*			

Panel B: Multivariate regression SAR_{t+1} on accrual components and deciles of growth in PPE and ONA

Regression 3 (Pooled regression adjusted R^2 = 0.0071)

	Intercept	∆AccRec	-AccPay	ΔInvent	-Deprec	Other Acc	Growth PPE	Growth ONA
Pooled	0.126*	-0.01	-0.014***	-0.061*	-0.037*	-0.045*	-0.057*	-0.057*
mean annual ^e		-0.011	-0.017	-0.057*	-0.036**	-0.044*	-0.059*	-0.058*
# of years ^f		15	17	22*	18	22*	23*	25*

Notes:

a. The variables are calculated as indicated below (see appendix for details) and all regressors are sorted into deciles and assigned values between 0 and 1. All changes are scaled by average total assets, unless indicated otherwise.

 $SAR_{t+1}=12$ -month buy and hold return beginning 4 months after fiscal year-end of inventory change year less corresponding size decile buy and hold return.

Growth PPE = change in net property, plant and equipment.

Growth ONA = change in other (net) long-term assets and non-interest bearing liabilities.

Accruals = the change in non-cash current assets, less the change in current liabilities (exclusive of short-term debt and taxes

payable), less depreciation expense.

-Deprec = minus depreciation and amortization expense

 $\Delta AccRec = change in accounts receivable$

 $-\Delta AccPay = minus$ change in accounts payable

 Δ Invent = change in total inventory

Other Acc = accruals minus change in accounts receivable and inventory plus change in accounts payable plus depreciation and amortization expense

b. Hedge is the difference in size-adjusted return between the lowest and highest decile.

c. Mean hedge reports the mean value of annual hedge portfolio returns. Statistical significance is based on a t-test using

magnitudes of the 28 annual hedge returns.

d. # of years reports the number of years in which the hedge portfolio return is positive. Statistical significance is based on a sign test using signs of the 28 annual hedge returns.

e. Mean annual reports the mean value of annual coefficient estimates. Statistical significance is based on a t-test using

magnitudes of the 28 annual coefficient estimates.

f. # of years reports the number of years in which the coefficient is negative. Statistical significance is based on a sign test using signs of the 28 annual coefficient estimates.

*: significant at the 1% level.

**: significant at the 5% level..

Table 3

Time-series of key financial variables around the year with extreme inventory changes

	-5	-4	-3	-2	-1	0	1	2				
Next year's	s size adjusted	return (SAR	R _{t+1}), over the	e 12 months b	eginning 4 n	nonths after	the fiscal yea	r end				
lowest	-6.72%	-17.11%	-9.44%	-4.19%	-4.32%	4.50%	1.47%	1.86%				
highest	3.05%	13.08%	9.60%	6.39%	3.82%	-6.89%	-1.66%	-0.36%				
median gro	owth in sales											
lowest	9.85%	9.22%	9.58%	8.02%	5.31%	-3.68%	2.76%	6.92%				
highest	11.99%	12.49%	12.68%	13.50%	14.87%	18.52%	15.02%	9.64%				
median car	oital expenditu	ire										
lowest	5.26%	5.18%	5.10%	4.91%	4.60%	3.81%	3.99%	4.39%				
highest	5.42%	5.42%	5.25%	5.42%	5.59%	6.37%	5.52%	5.07%				
median RC	DE											
lowest	10.39%	10.02%	9.54%	8.61%	6.69%	5.02%	8.04%	9.06%				
highest	12.41%	12.52%	12.61%	12.92%	13.37%	13.11%	11.50%	11.16%				
median RC	DA											
lowest	10.17%	9.73%	9.33%	8.47%	6.92%	5.63%	7.34%	8.02%				
highest	12.57%	12.64%	12.71%	12.88%	13.37%	13.05%	11.40%	10.84%				

Year relative to year 0, the year with the extreme inventory change

Notes:

 $SAR_{t+1} = 12$ -month buy and hold return, beginning 4 months after the fiscal year-end, in excess of the corresponding size decile

buy and hold returns

Lowest= lowest decile of inventory change

Highest= highest decile of inventory change

Growth in Sales = Percent change in sales.

Capital expenditure = Capital expenditures divided by average total assets

ROE = Income before extraordinary items- available for common stocks divided by book value of common equity

ROA = Operating income after depreciation divided by average total assets.

Table 4

Descriptive results for SAR_{t+1} (39,315 firm-years from 1970 to 1997)^a

			Decile 1	anking		Year-by-year					
	Ν	lowest	2	9	highest	hedge ^b	hedge ^c # years ^d				
Panel A: Mean SAR _{t+1} for deciles based on inventory components, and for deciles of inventory changes for subgroups based on retailers and inventory valuation method											
ΔRM Invent	21673	3.89%	3.31%	-2.82%		10.82% *	10.72%* 20**				
Δ WIP Invent	18829	3.38%	-1.15%	-2.36%	-3.07%	6.45% *	6.62%* 21**				
∆FG Invent	20576	3.42%	2.19%	-1.73%	-5.93%	9.35% *	8.99%* 23*				
Retailers	4,769	5.64%	4.31%	0.94%	-5.96%	11.60% *	9.27%** 21**				
LIFO	2,937	5.31%	3.10%	-0.53%	1.34%	3.97%	6.44% 17				
Mixed	9,683	4.19%	0.97%	-0.67%	-5.31%	9.50% *	10.21%* 24*				
Non-LIFO	21,959	3.92%	2.77%	-2.97%	-8.82%	12.74% *	11.89%* 26*				

Panel B: Regression of SAR_{t+1} on deciles based on inventory components and accrual components

Regression 4 (Pooled regression adjusted R^2 = 0.0071)

	Intercept	ΔAccRec	-AAccPay	ΔRM Invent	∆WIP Invent	ΔFG Invent	-Deprec	Other Acc
Pooled	0.122*	-0.017	-0.020	-0.079*	-0.028**	-0.042*	-0.054*	-0.042*
mean annual ^e		-0.015	-0.019	-0.079*	-0.031**	-0.037**	-0.053*	-0.040*
# of years ^f		14	17	23*	21**	20**	21**	22*

Notes:

- **a**. The variables are calculated as indicated below (see appendix for details) and all regressors are sorted into deciles and assigned values between 0 and 1. All changes are scaled by average total assets, unless indicated otherwise.
- $SAR_{t+1}=12$ -month buy and hold return beginning 4 months after fiscal year-end of inventory change year less corresponding size portfolio buy and hold return.

-Deprec = minus depreciation and amortization expense

 $\Delta AccRec = change in accounts receivable$

 $-\Delta$ AccPay = minus change in accounts payable

Other Acc = accruals minus change in accounts receivable and inventory plus change in accounts payable plus depreciation and

amortization expense

Retailers includes all firms in the 5xxx SIC industry group (expected to have no manufacturing costs)

 Δ RM Invent = change in raw material inventory

 Δ WIP Invent = change in work-in-process inventory

 Δ FG Invent = change in finished goods inventory

Inventory method is coded as LIFO if the LIFO method alone is used, as non-LIFO if the LIFO method is not used, and as mixed for all other firm-years.

b. Hedge is the difference in size-adjusted return between the lowest and highest decile.

c. Mean hedge reports the mean value of annual hedge portfolio returns. Statistical significance is based on a t-test using magnitudes of the 28 annual hedge returns.

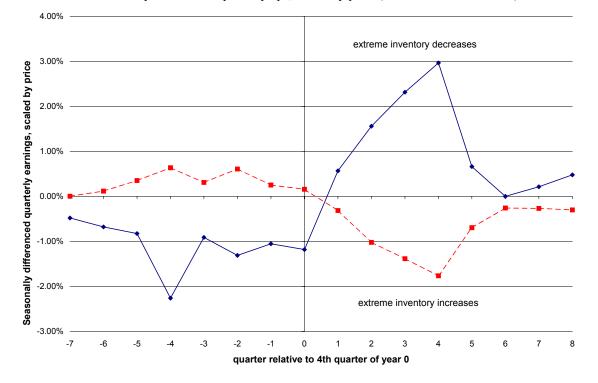
d. # of years reports the number of years in which the hedge portfolio return is positive. Statistical significance is based on a sign test using signs of the 28 annual hedge returns.

e. Mean annual reports the mean value of annual coefficient estimates. Statistical significance is based on a t-test using magnitudes of the 28 annual coefficient estimates.

f. # of years reports the number of years in which the coefficient is negative. Statistical significance is based on a sign test using signs of the 28 annual coefficient estimates.

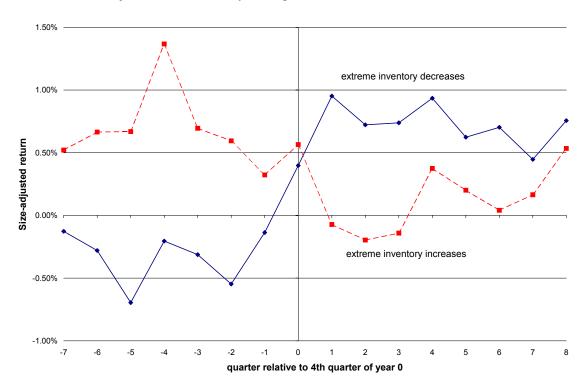
*: significant at the 1% level;

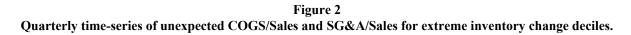
**:significant at the 5% level;

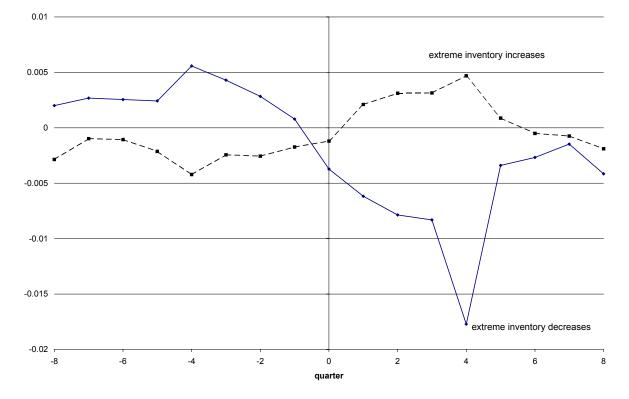


Panel A: Mean seasonally-differenced quarterly eps, scaled by price (Winsorized at -1 and +1)

Panel B: Mean size-adjusted returns for 3-day earnings announcement windows.







Panel A: Median seasonally-differenced COGS/Sales around the 4th quarter of the inventory change year (quarter 0)

Panel B: Median seasonally-differenced SG&A/Sales around the 4th quarter of the inventory change year (quarter 0)

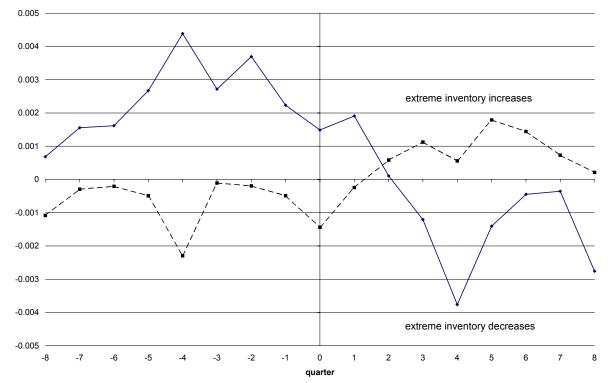
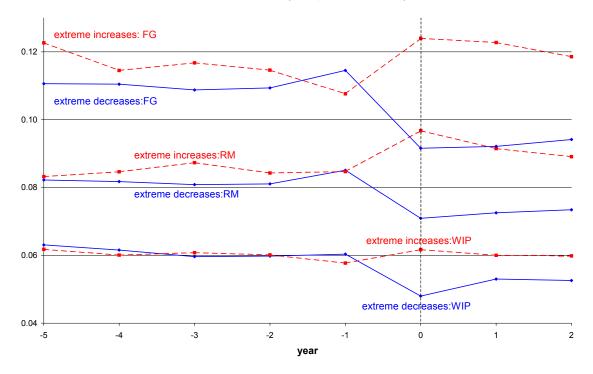


Figure 3 Time-series of inventory levels for extreme inventory change deciles



Panel A: median level of inventory components, scaled by total assets

Panel B: median seasonal changes in total inventory, scaled by total assets

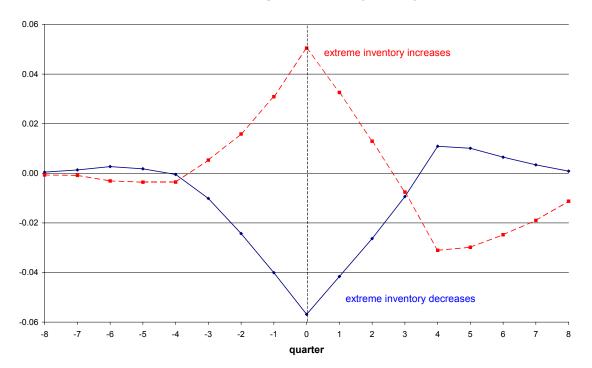


Figure 4 Median seasonally-differenced COGS/Sales for extreme inventory change deciles for LIFO and non-LIFO firms

