# What Explains the Stock Market's Reaction to Federal Reserve Policy? 

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#### Abstract

This paper analyzes the impact of unanticipated changes in the Federal funds target on equity prices, with the aim of both estimating the size of the typical reaction, and understanding the reasons for the market's response. On average over the June 1989 to December 2002 sample, a typical unanticipated 25 basis point rate cut is associated with a roughly 1 percent increase in the level of stock prices, as measured by the S\&P 500 composite or the CRSP value-weighted index. There is some evidence of a weaker response to unexpected decisions to keep the funds rate unchanged, and a stronger response to reversals in the direction of rate changes. The estimated response also varies widely across industries, with the greatest sensitivity observed in high tech and telecommunications, and the smallest in energy and utilities. Very little of the market's reaction can be attributed to policy's effects on the real rate of interest. The majority of the response of the current excess return on equities can be traced to policy's impact on expected future excess returns, with some effect also coming through expected future dividends. JEL codes: E44, G12.


## 1 Introduction

The reaction of the stock market to monetary policy is clearly a topic of intense interest both to market participants and policymakers. Those holding equities would obviously like to

[^0]know how possible Federal Reserve actions might affect the value of their portfolios. Similarly, an estimate of the likely effect of policy on asset prices is an important ingredient in assessing the transmission of monetary policy through the "wealth effect." The size of and reasons for the market's response to policy are not yet clearly understood, however, in part because of the interdependence between asset prices, monetary policy, and macroeconomic conditions.

The overall goal of this paper is to analyze the reaction of equity prices to monetary policy, using a market-based measure of unanticipated policy actions to isolate more cleanly the effects of policy. One specific objective is to document and quantify the stock market's response to monetary policy actions, both in the aggregate and for a variety of industry classifications. A second objective is to determine the reasons for the stock market's response, i.e., the extent to which the reaction can be traced to the impact of policy on (expectations of) future real interest rates and dividends.

Estimating the response of equity prices to monetary policy actions is not as easy as it may seem, however, as the market is unlikely to respond to policy actions that were already anticipated. Distinguishing between expected and unexpected policy actions is therefore essential for discerning their effects. A natural way to do this is to use the technique proposed by Kuttner (2001), which uses Fed funds futures data to construct a measure of "surprise" rate changes. ${ }^{1}$ Explaining the market's response is harder still, as it requires an assessment of how those policy surprises affect expectations of future interest rates and excess returns. To do this, we adapt the procedure developed by Campbell (1991) and Campbell and Ammer (1993), which uses a vector autoregression (VAR) to calculate revisions in expectations of future interest rates, stock returns, and dividends.

[^1]The results presented in section 2 of the paper show that the market does indeed react strongly to surprise funds rate changes. Estimated over a sample consisting of the union of days with a change in the target funds rate target and FOMC meeting days, the CRSP valueweighted index registers a one-day gain of roughly 1 percent in response to a surprise 25 basis point easing. The market reacts little, if at all, to the component of funds rate changes that are anticipated by futures market participants. A comparable reaction is observed at a monthly unit of observation.

Section 3 takes up the question of what explains equity prices' response. It turns out that only a small portion of equities' excess return variance can be explained by the effects of real interest rates or dividends; most is attributed to the induced time variation in expected excess equity returns. So while an unanticipated rate cut generates an immediate increase in equity prices, it is followed by an extended period of lower-than-normal excess returns. One interpretation of this result is that monetary policy surprises are for some reason associated with changes in the equity premium. But in the absence of a fully-developed asset pricing model, it is impossible to distinguish this interpretation from a market overreaction.

Naturally, this is not the first paper to estimate the impact of monetary policy on equity prices. Using several different gauges of the stance of monetary policy, including shocks from an identified VAR and one-day changes in the Fed funds rate, Thorbecke (1997) documented a connection between expansionary policy and above-average stock returns. Another paper in this vein is Jensen, Johnson and Mercer (1996), Jensen and Mercer (1998), who examined the disaggregated response of stock prices to changes in the discount rate. Rigobon and Sack (2002) found a robust response of the stock market to interest rate surprises derived from eurodollar futures, using a novel estimator exploiting the heteroskedasticity introduced by unexpected policy actions to correct for possible endogeneity. Taking a less structured approach, Fair (2002) identified the largest changes in equity prices at 1to 5-minute intervals, and found that nearly one-third of those were associated with news
about monetary policy.
Fewer papers have tackled the question of why the stock market reacts to monetary policy. The results of Goto and Valkanov (2000) suggest that the covariance between equity prices and inflation induced by policy shocks from an identified VAR may be one reason for the response. Boyd, Jagannathan and Hu (2001) focus on equity prices' response to unemployment news, rather than monetary policy; but their explanation for the market's perverse reaction (i.e., the association of higher-than-expected unemployment with increases in equity prices) involves a presumed expansionary response of monetary policy to bad unemployment news. Patelis (1997) addresses the question using identified VAR shocks within the framework of (1991) and Campbell and Ammer (1993).

This paper uses a common approach to estimating the size of the stock market's reaction, and to understand the cause of that reaction. The approach is similar to that of Rigobon and Sack (2002) in its use of futures data to calculate policy surprises. It is also closely related to the method used by Patelis (1997) to address the same question. The main difference is in the use of policy surprises based on futures data, rather than shocks from an identified VAR.

## 2 Estimating equities' reaction to target rate changes

This section focuses on the immediate impact of monetary policy on equity prices, both for broad stock market indices, and for individual industries. As noted in the introduction, however, one difficulty inherent in measuring policy's effects is that asset markets are forward looking, and hence tend to incorporate any information about anticipated policy changes. Some effort is therefore required to isolate the unexpected, or "surprise" policy change which might plausibly generate a market response. This does not say that asset prices respond to monetary policy only when the Fed surprises the markets, of course. Nat-
urally, asset prices will also respond to revisions in expectations about future policy, which in turn may be driven by news about changing economic conditions. Unexpected policy actions merely represent convenient "natural experiments" which allow us to gauge the market reaction.

One convenient, market-based way to identify unexpected funds rate changes relies on the prices of Fed funds futures contracts, which embody expectations of the effective Fed funds rate, averaged over the settlement month. ${ }^{2}$ Krueger and Kuttner (1996) found that the Fed funds futures rates did a good job of forecasting, efficiently incorporating available data on the likely policy actions. Kuttner (2001) subsequently used this approach to estimate the response of the term structure to monetary policy. The analysis in this section uses a similar approach to gauging the response of equity prices to unanticipated changes in the Fed funds rate from June 1989 through December 2002. For comparison, a parallel set of results is presented below using policy surprises defined at a monthly frequency.

### 2.1 Measuring the surprise element of policy actions

A measure of the surprise element of any specific change in the Fed funds target can be derived from the change in the futures contract's price relative to the day prior to the policy action. Specifically, the "surprise" target rate change can be calculated as the change in the "spot month" (i.e., for the month in which the target is changed) Fed funds futures rate on the day of the rate change, scaled up by a factor related to the number of days in the month affected by the change,

$$
\begin{equation*}
\Delta \hat{r}_{t}^{u}=\frac{m}{m-t}\left(f_{s, t}^{0}-f_{s, t-1}^{0}\right), \tag{1}
\end{equation*}
$$

[^2]where $\hat{r}^{u}$ is the unexpected target rate change, $f_{s, t}^{0}$ is the spot-month futures rate on day $t$ of month $s$, and $m$ is the number of days in the month. ${ }^{3}$ The expected component of the rate change is simply defined as the actual minus the surprise, or
\[

$$
\begin{equation*}
\Delta \hat{r}_{t}^{e}=\Delta \hat{r}_{t}-\Delta \hat{r}_{t}^{u} . \tag{2}
\end{equation*}
$$

\]

Getting the timing right is crucial for the analysis of daily data. Before 1994, when the Fed instituted its current policy of announcing changes in the funds rate target, markets generally became aware of policy actions on the day after the FOMC's decision, when it was implemented by the Open Market Desk. Following Rudebusch (1995) and Hilton (1994), most rate changes prior to 1994 are assigned to the date of the Desk's implementation. As documented in Kuttner (2003), however, the sample contains several minor deviations from this pattern. Six of these correspond to days on which the Desk allowed the funds rate to drift downward in advance (and presumably in anticipation) of the FOMC's decision, with the full awareness that its inaction would be interpreted as an easing of policy. A seventh exception occurred on December 18, 1990, when the Board of Governors made an unusual late-afternoon announcement of a cut in the discount rate, from which market observers (correctly) inferred a 25 basis point rate cut.

The policy of announcing target rate changes, which began in February 1994, eliminates virtually all of the timing ambiguity associated with rate changes in the earlier part of the sample. Moreover, because the change in the target rate is generally announced prior to the close of the futures market, the futures rate generally incorporates any news about monetary policy. The only exception is October 15,1998 , when a 25 basis point rate cut was announced after the close of the futures markets. In this case, the difference between the opening rate on the 16th and the closing rate on the 15 th is used to calculate the surprise.

[^3]
### 2.2 Baseline event study results

One approach to measuring the impact of Federal Reserve policy on the stock market is to calculate the market's reaction to funds rate changes on the day of the change. The market may of course also react to the lack of a change in the target, if a change had been anticipated. Because this approach involves looking at the response to specific events, it might be described as an "event-study" style of analysis. For the purpose of this paper, the relevant sample of "events" is defined as the union of all days when the funds rate target was changed, and days corresponding to FOMC meetings. The time period analyzed spans May 1989 through December 2002, which includes a total of 131 observations. ${ }^{4}$

Table 1 contains the baseline regression results for the event-study analysis. Column (a) reports the results from a simple regression of the CRSP value-weighted return on the raw change in the Fed funds target,

$$
\begin{equation*}
H_{t}=\alpha+\beta \Delta \hat{r}_{t}+u_{t}, \tag{3}
\end{equation*}
$$

making no distinction between surprise and expected changes; $H$ represents the stock return, and $\hat{r}_{t}$ is the funds rate target. The response to the raw target rate changes is negative, as expected, but small and insignificant.

Column (b) shows the response when the funds rate change is broken down into the expected and surprise components. The regression used here is

$$
\begin{equation*}
H_{t}=\alpha+\beta_{1} \Delta \hat{r}_{t}^{e}+\beta_{2} \Delta \hat{r}_{t}^{u}+u_{t}, \tag{4}
\end{equation*}
$$

where the expected and unexpected changes, $\Delta \hat{r}_{t}^{e}$ and $\Delta \hat{r}_{t}^{u}$, are computed as described above in section 2.1. The highly significant negative coefficient on the surprise coefficient implies a -4.68 percent one-day return in response to a 1 percentage point surprise rate cut. ${ }^{5}$ The

[^4]Table 1: The response of equity prices to funds rate changes

|  | Full sample |  |  | Excluding outliers |  |
| :--- | :---: | :---: | :--- | :---: | :---: |
| Regressor | (a) | (b) |  | (c) | (d) |
| Intercept | 0.23 | 0.12 |  | 0.17 | 0.11 |
|  | $(2.58)$ | $(1.35)$ |  | $(2.14)$ | $(1.37)$ |
| Raw funds rate change | -0.61 | $\ldots$ |  | -0.11 | $\ldots$ |
|  | $(1.06)$ |  |  | $(0.31)$ |  |
| Expected change | $\ldots$ | 1.04 |  | $\ldots$ | 0.67 |
|  |  | $(2.17)$ |  | $(1.62)$ |  |
| Surprise change | $\ldots$ | -4.68 |  | $\ldots$ | -2.55 |
|  |  | $(3.03)$ |  | $(2.79)$ |  |
| Adjusted R-squared | 0.007 | 0.171 |  | -0.007 | 0.048 |

Notes: The dependent variable is the one-day CRSP value-weighted return, in percent. Parentheses contain $t$-statistics, calculated using heteroskedasticity-consistent estimates of the standard errors. The full sample consists of the 55 target rate changes and the 77 FOMC meeting dates over the period from June 1989 through December 2002, excluding the 17 September 2001 observation, for a total of 131 observations. The outliers excluded from the regressions in columns (c) and (d) correspond to the six observations with influence statistics in excess of 0.3 , depicted in figure 2 , leaving 125 usable observations.
strength of the measured reaction is remarkable, given that much of the Fed funds surprises measured at a daily frequency represent shocks to the timing or rate changes, rather than to the medium-term path of interest rates. ${ }^{6}$ Also noteworthy is the $R^{2}$ of 0.17 , which implies that 17 percent of the variance in equity prices on these "event" days is associated with monetary policy actions (or inactions, in the case of FOMC meetings that left the funds rate target unchanged).

The negative impact of surprise funds rate increases on the stock market is readily visible in the scatterplot shown in Figure 1. Also apparent are a number of observations characterized by very large changes in equity prices - many well in excess of 2 percent. This naturally raises the question of whether the results reported in the first two columns of

[^5]Figure 1: Scatterplot of equity returns and Fed funds surprises, daily data


Table 1 are sensitive to the inclusion of these observations.
To determine which observations might be having the largest effect on the regression results, influence statistics were computed for each observation in the sample. These statistics are calculated as the quadratic form $\Delta \beta_{t}^{\prime} \Sigma^{-1} \Delta \beta_{t}$, where $\Delta \beta_{t}$ is change in the regression coefficient resulting from dropping observation $t$, and $\Sigma$ is the estimated covariance matrix of the coefficients. The distribution of these statistics, plotted in figure 2, confirms that six observations, all with statistics in excess of 0.3 , exert an unusually large influence on the estimates; the comparable statistics for the remaining observations are all well below 0.20 and most are less than 0.10 . The six observations associated with the large influence statistics are labeled in figure 1: 8 August 1991, 2 July 1992, 15 October 1998, 3 January 2001, 20 March 2001, and 18 April 2001. The first two of these are associated with events

Figure 2: Distribution of influence statistics

other than monetary policy actions, while the most recent four arguably represent unusual reactions to monetary policy actions. Each is in its own way is revealing.

Interestingly, three of the six candidate outliers occurred during the easing cycle that began in 2001. The surprise 50 basis point intermeeting rate reductions on 3 January and 18 April were both greeted euphorically, with one-day returns of 5.3 and 4.0 percent respectively. The 50 basis point rate cut on 20 March was received less unenthusiastically, however. Even though the cut was more or less what the futures market had been anticipating, financial press reported that many equity market participants were "disappointed" the rate cut hadn't been an even larger 75 basis point action. Consequently, the market lost more than 2 percent.

Another unusually vehement reaction to a Fed action is associated with the 25 basis point intermeeting rate cut on 15 October 1998, which was taken in response to unsettled conditions in the financial markets - specifically, the deteriorating situations in Asia and Russia. For whatever reason, the unexpected intermeeting cut lifted equities over 4 percent.

The stock market fell less than 0.3 percent on 2 July 1992. What makes this reaction unusual, however, is the fact that it came on a day when the Fed unexpectedly cut the
funds rate target by 50 basis points. The decision to cut was no doubt influenced by that day's unusually bleak employment report, in which reported payroll employment fell by 117 thousand. This raises the issue that some of the "surprise" rate changes in the sample may in fact represent endogenous responses to economic news, such as the employment report. This possibility is investigated in greater detail below in section 2.3.

The final candidate outlier is 21 August 1991, when the CRSP value-weighted index rose 2.7 on a day associated with an FOMC meeting. The futures market had apparently priced in some possibility of a rate cut on that day, but the FOMC's decision to leave rates unchanged generated a small, positive surprise. The financial press reported that the stock market jump was a response to the resolution of the attempted coup in Russia - clearly an event with no direct relation to that day's FOMC decision.

Two other observations are highlighted in Figure 1: 17 May and 16 August 1994. While the influence statistics associated with these two dates are not extraordinarily high (0.05 and 0.04 respectively), they stand out as unusual instances in which equities rose in spite of significant, positive funds rate surprises. As noted in Kuttner (2001), a similarly anomalous response is observed in the response of bond yields on those dates. The reason seems to be that both of these larger-than-expected 50 basis point rate hikes were accompanied by statements by the FOMC suggesting that further rate increases were not imminent. Thus, expectations of the level of the funds rate in the future may have fallen, despite the increase in the current setting of the funds rate target.

Columns (c) and (d) of Table 1 show what happens when the six candidate outliers associated with the large influence statistics are dropped from the sample. (The two observations from 1994 are retained.) The measured response to funds rate surprises is still negative and significant, but smaller in magnitude: -2.55 , as opposed to -4.68 . The response to the expected component is smaller (and now no longer significant at the 0.05 level), as is the response to the raw funds rate change in column (c). Interestingly, ex-
cluding the six outliers actually decreases the adjusted $R$-squared, presumably because the fitted regression was doing a good job of accounting for the variance contributed by the three observations in the extreme northwest of the scatterplot.

### 2.3 Endogeneity and subsample stability

One concern is the possibility of a nonzero correlation between the funds rate surprise and the error term in equation 4 , which would bias the parameter estimates reported above. Such a correlation could arise from one of at least three sources. Perhaps the most obvious source would be a contemporaneous (in this context, same-day) response of the funds rate of the funds rate to equity prices. Such a response has been hard to detect using conventional time series methods, such as those employed by Bernanke and Gertler (1999) and by Fuhrer and Tootell (2003). However using a novel estimator that achieves identification by exploiting the heteroskedasticity in the data, Rigobon and Sack (2003) find some evidence for a response at a monthly frequency. There is no evidence of such a response at a daily frequency, however.

A second possible source of bias is a joint response of policy and the stock market to new information. For example, the release of data indicating weaker-than-expected economic growth might plausibly lead to both a decline in the stock market, and to a cut in the target Fed funds rate. One obvious example of such a joint response is 17 September 2001, which we had dropped from the analysis on a priori grounds. Here, the Fed's 50 basis point rate cut and the stock market's sharp drop were both clearly spurred by the previous week's terrorist attacks.

The use of daily data will tend to minimize the endogeneity problem, as same-day responses of the FOMC to new economic data have been rare in recent years. During the earlier pre-1994 part of the sample, however, it was not uncommon for the FOMC to cut rates on the heels of weaker-than-expected employment data. In fact, 8 of the 23 rate cuts
from June 1989 through July 1992 were intermeeting actions that coincided with the release of the employment report.

These employment report-related observations are distinguished in Figure 1 by (green) triangles. These observations are characterized by little, if any, correlation between the funds rate surprise and the stock return. In these instances, the "good news" for the stock market represented by the Fed's actions seems to have been almost exactly offset by the "bad news" about economic activity contained in the employment report.

A correlation between the regression's error term and funds rate surprises could also be generated by market participants' belief that the Fed's actions incorporated a response to private information. Stocks could fall on a rate cut, for example, if investors inferred from the Fed's actions that the economy was weaker than they had previously thought.

Whatever the source, endogeneity would call into question the validity of the OLS estimates presented above, and argue for using some sort of instrumental variables procedure, such as the heteroskedasticity-based estimator proposed by Rigobon and Sack (2002). To the extent that this endogeneity exists, however, it would tend introduce a downward bias in the stock market's estimated response to monetary policy. Therefore, the stock market's true response to funds rate changes is quite likely to be larger than the estimates presented above.

Another issue concerns the stability of the estimated relationship. One plausible reason to suspect a change has to do with the endogeneity problem discussed above, which is likely to have been more severe before 1994, when same-day responses to employment data were relatively routine. After 1994, with rate changes more or less dictated by the exogenous scheduling of FOMC meetings, this becomes less of an issue. Because of this possible linkage between endogeneity and the coefficients' stability over time, the two are analyzed jointly below.

To explore the possibility of different responses either post- 1994, or on days associated

Table 2: Tests for subsample stability and endogeneity

|  | Full sample |  |  | Excluding outliers |  |
| :--- | :---: | :---: | :---: | :---: | :---: |
| Regressor | (a) | (b) |  | (c) | (d) |
| Intercept | 0.16 | 0.16 |  | 0.12 | 0.12 |
|  | $(1.80)$ | $(1.76)$ |  | $(1.43)$ | $(1.43)$ |
| Expected change | 1.09 | 1.09 |  | 0.69 | 0.69 |
|  | $(2.26)$ | $(2.24)$ |  | $(1.68)$ | $(1.67)$ |
| Surprise change | -1.25 | -2.55 |  | -2.29 | -3.57 |
|  | $(1.14)$ | $(1.70)$ |  | $(2.28)$ | $(3.77)$ |
| Surprise change $\times$ |  |  |  |  |  |
| post-1994 | -6.87 | -5.58 |  | -0.78 | 0.50 |
|  | $(3.59)$ | $(2.61)$ |  | $(0.45)$ | $(0.29)$ |
| employment report | $\ldots$ | 2.67 |  | $\ldots$ | 3.33 |
|  |  | $(1.82)$ |  | $(2.55)$ |  |
| Adjusted R-squared | 0.280 | 0.283 |  | 0.042 | 0.054 |

Notes: The dependent variable is the one-day CRSP value-weighted return, in percent. Parentheses contain $t$-statistics, calculated using heteroskedasticity-consistent estimates of the standard errors. The post-1994 dummy is set to 1 for observations beginning with 4 February 1994. The employment dummy is set to 1 for pre-1994 observations when a change in the target funds rate coincided with an employment release. See also notes to table 1 .
with employment releases, we interact the surprise rate change with dummy variables: one equal to 1 starting with the 4 February 1994 observation, and another equal to 1 on the days of pre-1994 employment releases. Table 2 reports the response of the CRSP valueweighted index to surprise rate changes in the presence of these interactive dummies. Like Table 1, columns (a) and (b) give the results for the full sample, and columns (c) and (d) give the results for the sample excluding the six candidate outliers identified above.

Judging from the results in column (a), it would appear that the entire equity price response can be traced to the post-1994 period. The coefficient on the surprise itself is only -1.25 and insignificant; that on the surprise interacted with the post-1994 dummy is a highly significant -6.87 . But this regression neglects the possibility of endogeneity in the policy response prior to 1994. Including the surprise interacted with the employment
release dummy as in column (b), increases the magnitude of the surprise response, and the positive interaction term implies a near-zero response to policy when it coincides with an employment release. These coefficients statistically significant at only the 0.10 level however, and the post-1994 interaction term remains large and highly significant.

These results turn out to be heavily influenced by the six outliers identified above. With those observations excluded, as in column (c), post-1994 rate surprises have only a slightly larger effect, and the difference is not statistically significant. The coefficient on the surprise itself -2.29 , and significant at the 0.05 level. But when the employment interaction term is included, the surprise coefficient grows to -3.57 . This effect is almost exactly offset for employment release days by the 3.33 coefficient on the interaction term. Both are now highly significant. The post-1994 dummy remains insignificant. Thus, if the six candidate outliers are discarded as unrepresentative of the sample, there is no evidence of a break in 1994. Furthermore, the results confirm that the endogeneity problem discussed above does indeed attenuate the measured response of equity prices to Fed funds surprises.

### 2.4 Asymmetries

A final set of questions concerns asymmetries, broadly defined: the possibility that the equity price response to monetary policy depends somehow on the direction of the action, or on the context in which it occurred. As above in section 2.3, interactive dummy variables are used in the regression to investigate these questions.

One possibility is that the magnitude of the stock market response depends on whether the surprise itself was positive or negative. To allow for this, a dummy variable was set to 1 for those observations with positive surprises. An interaction term involving this dummy and the surprise rate change was then included in the regression. The interactive term involving the employment release is also included, in order to pick up the smaller impact of funds rate surprises on employment release days. As above, the regressions are run with

Table 3: Tests for asymmetries

| Regressor | Full sample |  |  | Excluding outliers |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | (a) | (b) | (c) | (d) | (e) | (f) |
| Intercept | $\begin{gathered} -0.02 \\ (0.17) \end{gathered}$ | $\begin{gathered} 0.12 \\ (1.34) \end{gathered}$ | $\begin{gathered} 0.14 \\ (1.72) \end{gathered}$ | $\begin{gathered} 0.12 \\ (1.32) \end{gathered}$ | $\begin{gathered} 0.12 \\ (1.42) \end{gathered}$ | $\begin{gathered} 0.13 \\ (1.63) \end{gathered}$ |
| Expected change | $\begin{gathered} 0.84 \\ (1.58) \end{gathered}$ | $\begin{gathered} 1.56 \\ (3.24) \end{gathered}$ | $\begin{gathered} 1.03 \\ (2.24) \end{gathered}$ | $\begin{gathered} 0.72 \\ (1.67) \end{gathered}$ | $\begin{gathered} 0.97 \\ (2.00) \end{gathered}$ | $\begin{gathered} 0.72 \\ (1.76) \end{gathered}$ |
| Surprise change | $\begin{gathered} -7.57 \\ (4.67) \end{gathered}$ | $\begin{gathered} -8.34 \\ (5.73) \end{gathered}$ | $\begin{gathered} -3.97 \\ (2.98) \end{gathered}$ | $\begin{gathered} -3.26 \\ (3.30) \end{gathered}$ | $\begin{gathered} -4.49 \\ (4.91) \end{gathered}$ | $\begin{gathered} -3.67 \\ (3.14) \end{gathered}$ |
| Surprise change $\times$ ( $\times$ ( |  |  |  |  |  |  |
| employment | $\begin{gathered} 7.05 \\ (4.43) \end{gathered}$ | $\begin{gathered} 8.11 \\ (5.26) \end{gathered}$ | $\begin{gathered} 2.54 \\ (1.47) \end{gathered}$ | $\begin{gathered} 3.05 \\ (2.36) \end{gathered}$ | $\begin{gathered} 4.14 \\ (3.19) \end{gathered}$ | $\begin{gathered} 3.46 \\ (2.35) \end{gathered}$ |
| positive surprise | $\begin{gathered} 7.39 \\ (1.59) \end{gathered}$ | $\ldots$ | ... | $\begin{gathered} -0.34 \\ (0.10) \end{gathered}$ | $\ldots$ | ... |
| no rate change | ... | $\begin{aligned} & 10.42 \\ & (3.81) \end{aligned}$ | $\ldots$ | ... | $\begin{gathered} 4.00 \\ (2.25) \end{gathered}$ | $\ldots$ |
| positive rate change | $\cdots$ | $\begin{gathered} 3.05 \\ (0.76) \end{gathered}$ | $\cdots$ | $\ldots$ | $\begin{gathered} 0.58 \\ (0.15) \end{gathered}$ | $\cdots$ |
| FOMC meeting | $\ldots$ | ... | $\begin{gathered} 4.25 \\ (2.75) \end{gathered}$ | $\ldots$ | . | $\begin{gathered} 0.67 \\ (0.39) \end{gathered}$ |
| reversal | $\ldots$ | $\ldots$ | $\begin{gathered} -6.33 \\ (3.09) \end{gathered}$ | $\ldots$ | . | $\begin{array}{r} -17.62 \\ (4.08) \end{array}$ |
| post-1994 | $\ldots$ | $\ldots$ | $\begin{gathered} -4.61 \\ (2.48) \end{gathered}$ | $\ldots$ | $\ldots$ | $\begin{gathered} 0.80 \\ (0.44) \end{gathered}$ |
| Adjusted R-squared | 0.260 | 0.323 | 0.369 | 0.053 | 0.065 | 0.098 |

Notes: The dependent variable is the one-day CRSP value-weighted return, in percent. Parentheses contain $t$-statistics, calculated using heteroskedasticity-consistent estimates of the standard errors. The positive surprise dummy is set to 1 when the surprise change in the funds rate is greater than zero. The no rate change and positive rate change dummies equal 1 when the funds rate target is unchanged or increased. The FOMC meeting dummy is set to 1 for those observations coinciding with FOMC meetings. The reversal dummy equals 1 for rate changes that reverse the direction of the previous change. See also notes to tables 2 and 1 .
and without the six candidate outliers identified earlier. The results reported in Table 3 provide weak support at best for this form of asymmetry. For the full sample, in column (a) of the table, the coefficient on the interaction term indicates a smaller effect of positive surprises, but the difference is not statistically significant. There is virtually no difference for the no-outlier sample, shown in column (d).

A related kind of asymmetry can be modeled by including interactive dummies for rate changes associated with increases in the funds rate, and with surprises associated with no change in the funds rate (i.e., when the futures market expected a change, but none occurred). The results of this exercise appear in columns (b) and (e) of the table for the full and no-outlier samples. Again, the statistically insignificant coefficient in the rate increase interaction variable suggests the direction of movement is not an important determinant of the market's reaction. The positive and statistically significant estimated coefficient on the "no change" interaction variable does, however, indicate that the market responds very little, if at all, to policy "inactions." This presumably means that the failure to move at any specific FOMC meeting may be viewed largely as postponing the inevitable.

A third sort of asymmetry has to do with the context of the rate decision: whether it was taken at an FOMC meeting, or represented a change in the direction of short-term interest rates. Again, these effects are captured through the use of suitable dummy variables, interacted with the surprise rate change variable. The sign of the FOMC interaction term is unclear a priori. Decisions taken at FOMC meetings may be less subject to the sort of endogeneity issues discussed above, which would tend to increase the impact of rate changes on these days. On the other hand, intermeeting changes (at least those not associated with employment reports) may convey an urgency on the part of the FOMC which would tend to increase the size of the response. What is clear, however, is that interest rate reversals - the first rate increase after a series of decreases, or vice versa - may have larger implications for future interest rates than other rate changes, and thus elicit a larger stock
market response. These "reversal" and FOMC observations are distinguished in Figure 1 by (orange) squares and (blue) circles, respectively.

Columns (c) and (f) of Table 3 show the results from a regression containing these two interaction terms, along with the employment interaction variable and (just for good measure) the post-1994 interaction term. The coefficient on the surprise term remains an economically and statistically significant -3.97 or -3.67 , depending on whether the full or no-outlier sample is used. Relative to this baseline response, there is no meaningful difference in either case between FOMC and non-FOMC days, controlling for the employment release. But reversals seem to have a large additional impact on the stock market: -6.33 for the full sample, and -17.62 for the no-outlier sample. In the latter case, the implausibly large estimate is driven almost entirely by the single observation in the southeast corner of the diagram, corresponding to the first rate increase in 1994. (The inclusion of the 3 January 2001 "outlier" actually shrinks the estimate to something more reasonable.) Thus, it appears that the stock market response to interest rate reversals may be larger than for other changes - although with so few of these reversals in the sample, this result should be treated with caution.

Taken together, the results presented above provide compelling evidence for a strong one-day reaction of the stock market to unanticipated changes in Federal funds rate. Just how strong this response is depends on whether the handful of potentially anomalous observations are viewed as representative, or discarded as outliers. The estimated response is stable over time, once the tendency for the FOMC to react to employment news in the early part of the sample is controlled for. And finally, there is some evidence of a larger response to reversals in the direction of rate changes, and a smaller response to unanticipated decisions to leave the funds rate target unchanged.

### 2.5 Results based on monthly data

An alternative way to define the policy surprise is to focus on the expected change in policy at a regular, monthly horizon. Unlike the event study approach, the regular timing is amenable to the time series analysis employed below in section 3 to assess the causes of the market's response. It is worth noting that in this approach, any month could potentially contain a surprise policy action, and that a failure to change the funds rate target in any month could represent a policy surprise. Consequently, the monthly time-series approach is probably less susceptible to any sample selection issues that might arise in the context of the event-study methodology.

The use of monthly data calls for a slightly different gauge of unanticipated policy actions. Since the price of the Fed funds futures contract is based on the monthly average Fed funds rate, a natural definition would be

$$
\begin{equation*}
\bar{\Delta} \hat{r}_{s}^{u} \equiv \frac{1}{m} \sum_{i=1}^{m} \hat{r}_{s, i}-f_{s-1, m}^{1} \tag{5}
\end{equation*}
$$

where $\hat{r}_{s, i}$ is the funds rate target day $i$ of month $s$, and $f_{s-1, m}^{1}$ is the rate corresponding to the one-month futures contract on the last ( $m$ th) day of month $s-1 .{ }^{7}$ The expected funds rate change is defined analogously as

$$
\begin{equation*}
\bar{\Delta} \hat{r}_{s}^{e} \equiv f_{s-1, m}^{1}-\hat{r}_{s-1, m} . \tag{6}
\end{equation*}
$$

The sum of the two is the average funds rate target in month $s$ minus the target on the last day of month $s-1$. (Since this is not the first difference of either the average or the month-end funds rate target, the slightly non-standard notation $\bar{\Delta}$ is used.)

This definition of the funds rate surprise raises a time aggregation issue. Measuring the surprise in terms of the average funds rate will tend to attenuate the size of the policy

[^6]surprises, as discussed in detail in Evans and Kuttner (1998). Unfortunately, without making specific assumptions about the days of possible rate changes, there is no clean way to correct for this problem. ${ }^{8}$ Consequently, some caution is required when interpreting the magnitude of the surprises measured in this way.

It is also important to note that the endogeneity issue discussed above in section 2.3 is almost certainly going to be more relevant to monthly funds rate surprises than it was for the day-ahead surprises. Rate changes that were unanticipated as of the end of the prior month may well include a systematic response to economic news, such as employment, output and inflation.

The results shown in Table 4 confirm this suspicion. The table reports the parameter estimates and R-squared from a regression of the monthly policy surprises on the surprise element of key economic reports, calculated as the difference between the number released and the consensus expectation for that number, compiled by Money Market Services. ${ }^{9}$ Over the full May 1989 through December 2002 sample, there appears to be a significant withinmonth impact of several data releases on the funds rate target: nonfarm payrolls, industrial production, retail sales, and core PPI (although these latter two oddly, have the "wrong" sign).

This relationship seems to be much stronger in the early part of the sample, however. The second column of the table shows the results for the same regression estimated over the May 1989 through September 1992 sample. The Fed's reaction to bad payroll employment news is now particularly pronounced. Moreover, the regression accounts for nearly half of the variance of the funds rate surprises. By contrast, in the more recent February 1994 through December 2002 subsample, there is very little evidence of a within-month reaction

[^7]Table 4: The impact of economic news on funds rate surprises

|  |  | Subsample |  |
| :--- | :---: | :---: | :---: |
| Data surprise | Full sample | $5 / 89-9 / 92$ | $2 / 94-12 / 02$ |
| Headline CPI | 0.016 | -0.124 | -0.010 |
| Core CPI | -0.058 | -0.012 | 0.152 |
| Headline PPI | 0.001 | -0.027 | -0.024 |
| Core PPI | $-0.085^{* *}$ | $-0.304^{* * *}$ | -0.022 |
| Nonfarm payrolls | $0.203^{* *}$ | $0.624^{* * *}$ | -0.009 |
| Industrial production | $0.069^{*}$ | 0.136 | 0.028 |
| Retail sales | $-0.031^{* *}$ | -0.061 | $-0.035^{* * *}$ |
| Retail sales, x autos | 0.023 | 0.093 | 0.021 |
| R-squared | 0.128 | 0.454 | 0.087 |
| Adjusted R-squared | 0.082 | 0.304 | 0.012 |

Notes: Asterisks denote statistical significance: *** for the 0.01 level, ${ }^{* *}$ for the 0.05 level, and $*$ for the 0.01 level. The dependent variable is the unexpected change in the funds rate, defined by equation 5. The Money Market Services data are used to calculate the surprise elements of the economic releases.
to economic news, as shown in the third column of the table. Only retail sales is significant, and the regression now accounts for a much smaller share of the variance of funds rate surprises. ${ }^{10}$

Table 5 reports the results from a regression of the monthly CRSP value-weighted return the expected and unexpected components of monthly funds rate changes,

$$
\begin{equation*}
H_{t}=\alpha+\beta_{1} \bar{\Delta} \hat{r}_{s}^{e}+\beta_{2} \bar{\Delta} \hat{r}_{s}^{u}+u_{t} . \tag{7}
\end{equation*}
$$

Column (a) reports the estimates for the full sample, consisting of all 164 months from May 1989 through December 2002. As in the earlier results, there is a strong, statistically significant negative response to unanticipated rate increases, and little or no response to the anticipated actions. The $\bar{R}^{2}$ indicates that nearly seven percent of the monthly stock return

[^8]Figure 3: Scatterplot of equity returns and Fed funds surprises, monthly data

variance can be traced to unanticipated policy actions.
It is interesting to note that the magnitude of the response, -11.43 , is over twice that found in the event-study analysis. This difference in magnitudes is readily explained by the time aggregation issue alluded to earlier. In fact, if funds rate changes on average take place in the middle of the month (for example, if rate changes were distributed uniformly over the days of the month), then the magnitude of the estimated monthly surprises will be attenuated by one-half, and this would explain the doubling of the estimated response of the stock price.

The negative relationship between policy surprises and stock returns is also evident in the scatterplot of the data in Figure 3. As in the daily data, a number of observations stand out as potential outliers, again raising the question of whether the results are sensitive to their inclusion. As above, influence statistics were calculated for each observation in the

Table 5: The monthly response of equity prices to funds rate surprises

| Regressor | Full sample <br> (a) | No outliers <br> (b) | Tests for asymmetries |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  | (c) | (d) | (e) |
| Intercept | $\begin{gathered} 0.13 \\ (0.32) \end{gathered}$ | $\begin{gathered} -0.03 \\ (0.09) \end{gathered}$ | $\begin{gathered} -0.01 \\ (0.02) \end{gathered}$ | $\begin{gathered} -0.07 \\ (0.16) \end{gathered}$ | $\begin{gathered} 0.10 \\ (0.24) \end{gathered}$ |
| Expected change | $\begin{gathered} -1.11 \\ (0.37) \end{gathered}$ | $\begin{gathered} 0.96 \\ (0.35) \end{gathered}$ | $\begin{gathered} -1.07 \\ (0.36) \end{gathered}$ | $\begin{gathered} -2.72 \\ (0.72) \end{gathered}$ | $\begin{gathered} -1.09 \\ (0.36) \end{gathered}$ |
| Surprise change | $\begin{array}{r} -11.43 \\ (3.95) \end{array}$ | $\begin{array}{r} -14.26 \\ (5.43) \end{array}$ | $\begin{array}{r} -12.46 \\ (3.69) \end{array}$ | $\begin{array}{r} -11.01 \\ (3.46) \end{array}$ | $\begin{array}{r} -10.49 \\ (2.53) \end{array}$ |
| Surprise change $\times$ |  |  |  |  |  |
| positive surprise | $\ldots$ | $\ldots$ | $\begin{gathered} 6.82 \\ (0.63) \end{gathered}$ | $\ldots$ | $\ldots$ |
| no rate change | $\ldots$ | $\ldots$ |  | $\begin{gathered} -4.88 \\ (0.75) \end{gathered}$ | $\ldots$ |
| positive rate change | $\ldots$ | $\ldots$ | $\ldots$ | $\begin{gathered} 6.59 \\ (0.52) \end{gathered}$ | $\ldots$ |
| reversal | $\ldots$ | $\ldots$ | $\ldots$ | ... | $\begin{gathered} 3.52 \\ (0.50) \end{gathered}$ |
| post-1994 | $\ldots$ | $\ldots$ | $\ldots$ | $\ldots$ | $\begin{gathered} -3.77 \\ (0.50) \end{gathered}$ |
| Employment surprise | $\ldots$ | $\ldots$ | $\ldots$ | $\ldots$ | $\begin{gathered} -0.69 \\ (0.10) \end{gathered}$ |
| Adjusted R-squared | 0.065 | 0.096 | 0.061 | 0.056 | 0.049 |
| Standard error | 4.28 | 3.85 | 4.30 | 4.30 | 4.31 |
| Durbin-Watson statistic | 2.02 | 2.09 | 2.02 | 2.02 | 2.03 |

Notes: Parentheses contain $t$-statistics. The dependent variable is the monthly return on the CRSP value-weighted portfolio, expressed in percent. The unanticipated and anticipated components of the change in the Fed funds rate are given by equations 5 and 6 , and are expressed in percent. The full sample includes 161 observations spanning May 1989 through December 2002. The no-outlier sample contains 151 observations.
sample; those with statistics in excess of 1.5 are flagged as outliers in the plot. (The most conspicuous of these is the data point deep in the southwest quadrant, which corresponds to September 2001.) Dropping these ten observations makes little difference to the results, however. In fact, as shown in column (b) of Table 5, the estimated coefficient of -14.26 is somewhat larger than it is for the full sample, and the $\bar{R}^{2}$ rises to 0.096 .

The monthly data contain very little evidence for the sorts of asymmetries uncovered in the daily data. As shown in columns (c) and (d), there is no indication that the stock price response depends on the sign of the surprise, or on the direction of the rate change. Nor is there any evidence of a different response to policy reversals, or to the MMS employment surprises. ${ }^{11}$

There is considerable cross-industry variation in the response of stock prices to monetary policy. Table 6 reports estimates of equation 7 for the ten industry portfolios constructed from CRSP returns by Fama and French. Perhaps surprisingly, the most responsive industries are high tech and telecommunications, with coefficients half again as large that for the overall value-weighted index. Less surprisingly, energy and utilities are only half as responsive as the overall market, and the relevant coefficients are statistically insignificant. Very little of those industries' variance is associated with unexpected policy actions.

## 3 Policy, fundamentals and stock prices

Having documented the reaction of equity returns to Federal Reserve policy in section 2 above, we now turn to the more difficult question of what explains the observed reaction. There are three broad reasons why an unexpected funds rate increase may lead to a decline in stock prices: it may be because of a decline in expected future dividends, an increase

[^9]Table 6: Response of Fama-French industry portfolios

| Index | Response to target change: |  | $\bar{R}^{2}$ | SE | DW |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | anticipated | unanticipated |  |  |  |
| CRSP value weighted | $\begin{gathered} 0.13 \\ (0.32) \end{gathered}$ | $\begin{array}{r} -11.43 \\ (3.95) \end{array}$ | 0.065 | 4.28 | 2.02 |
| Nondurables | $\begin{gathered} -0.85 \\ (0.25) \end{gathered}$ | $\begin{gathered} -9.65 \\ (2.88) \end{gathered}$ | 0.046 | 4.17 | 2.00 |
| Durables | $\begin{gathered} -1.47 \\ (0.38) \end{gathered}$ | $\begin{array}{r} -12.45 \\ (3.04) \end{array}$ | 0.048 | 5.56 | 1.97 |
| Manufacturing | $\begin{gathered} -2.02 \\ (0.61) \end{gathered}$ | $\begin{gathered} -8.82 \\ (2.81) \end{gathered}$ | 0.035 | 4.26 | 2.03 |
| Energy | $\begin{gathered} 0.20 \\ (1.02) \end{gathered}$ | $\begin{gathered} -4.03 \\ (1.24) \end{gathered}$ | -0.003 | 4.71 | 2.12 |
| High tech | $\begin{gathered} 0.06 \\ (0.01) \end{gathered}$ | $\begin{array}{r} -14.73 \\ (2.72) \end{array}$ | 0.025 | 8.22 | 2.00 |
| Telecommunications | $\begin{gathered} 0.35 \\ (0.60) \end{gathered}$ | $\begin{array}{r} -16.10 \\ (3.31) \end{array}$ | 0.065 | 6.16 | 1.85 |
| Wholesale/retail | $\begin{gathered} -4.75 \\ (1.47) \end{gathered}$ | $\begin{array}{r} -11.97 \\ (3.64) \end{array}$ | 0.056 | 4.85 | 1.95 |
| Health care | $\begin{gathered} -1.04 \\ (0.29) \end{gathered}$ | $\begin{gathered} -8.04 \\ (1.80) \end{gathered}$ | 0.017 | 4.96 | 2.15 |
| Utilities | $\begin{gathered} -1.24 \\ (0.48) \end{gathered}$ | $\begin{gathered} -5.42 \\ (1.55) \end{gathered}$ | 0.006 | 4.21 | 1.97 |
| Other | $\begin{gathered} -1.21 \\ (0.35) \end{gathered}$ | $\begin{array}{r} -11.08 \\ (3.61) \end{array}$ | 0.051 | 4.62 | 2.09 |

Notes: Parentheses contain $t$-statistics. The dependent variable is the monthly return on the FamaFrench industry portfolios, expressed in percent. The unanticipated and anticipated components of the change in the Fed funds rate are given by equations 5 and 6, and are expressed in percent. The regressions also include an intercept (not reported). The sample includes 161 observations spanning May 1989 through December 2002.
in the expected real interest rate used to discount those dividends, or it may increase the expected excess return (i.e., the equity premium) associated with holding stocks. Simple regressions of equity returns on surprise changes in the funds rate target are silent on the question, so a more structured approach is required to disentangle the various effects.

The approach in this paper is an adaptation of the method used by Campbell (1991), and Campbell and Ammer (1993). In brief, the first element of their method is a log-linear decomposition of excess equity returns into components attributable to news about real rates, dividends, and future excess returns; the second element is the use of a vector autoregression (VAR) to calculate the relevant expectations. ${ }^{12}$ We take the Campbell-Ammer framework one step further, however, by relating these components in turn to the news about the path of monetary policy embodied in the surprises derived from Fed funds futures. This allows us to estimate the impact of Fed funds surprises on expected future dividends, real interest rates, and expected future excess returns. It turns out that the largest effects come from revisions to expectations of future excess returns, and to expectations of future dividends. Real interest rates have a very small direct impact.

The object of this analysis is the $(\log )$ excess return on equities, denoted $y_{t+1}$. This is defined as the total return on equities (price change plus dividends), minus the risk-free rate (taken to be the one-month Treasury bill yield). The return dated $t+1$ is measured over period $t$, i.e., from the beginning of period $t$ to the beginning of period $t+1$. Let $e_{t+1}^{y}$ represent the unexpected (relative to expectations formed at the beginning of period $t$ ) excess return during period $t$, i.e., $y_{t+1}-E_{t} y_{t+1}$.

Using the linearization developed by Campbell and Shiller (1988), the period $t$ unexpected excess return on equity can be expressed in terms of the revision the expectation of discounted future dividends, the real interest rate, and future excess returns. (A sketch of

[^10]the derivation can be found in the appendix.) The decomposition can be written as:
\[

$$
\begin{equation*}
e_{t+1}^{y}=\tilde{e}_{t+1}^{d}-\tilde{e}_{t+1}^{r}-\tilde{e}_{t+1}^{y} \tag{8}
\end{equation*}
$$

\]

where the es represent the revision in expectations between periods $t$ and $t+1$, and the tilde denotes a discounted sum, so that

$$
\begin{aligned}
& \tilde{e}_{t+1}^{d}=\left(E_{t+1}-E_{t}\right) \sum_{j=0}^{\infty} \rho^{j} d_{t+j} \\
& \tilde{e}_{t+1}^{r}=\left(E_{t+1}-E_{t}\right) \sum_{j=0}^{\infty} \rho^{j} r_{t+j} \\
& \tilde{e}_{t+1}^{y}=\left(E_{t+1}-E_{t}\right) \sum_{j=1}^{\infty} \rho^{j} y_{t+1+j} .
\end{aligned}
$$

The $\rho$ discount factor, which comes out of the linearization, represents the steady-state ratio of the equity price to the price plus dividend; following Campbell and Ammer (1993), this is set to 0.9962 . As emphasized by Campbell (1991), equation 8 is really nothing more than a dynamic accounting identity relating the current excess return to revisions in expectations. As such, it contains no real economic content, much less any specific asset pricing model; such a model would be required to provide a link between the conditional expectations of future returns and economic variables (e.g., consumption).

Implementing this decomposition obviously requires empirical proxies for the expectations appearing in equation 8. The approach of Campbell (1991) and Campbell and Ammer (1993) is to model expectations using a Vector Autoregression (VAR) involving the variables of interest (excess returns and the real interest rate) along with any other indicators that might be helpful in forecasting those variables. Calculating the discounted sum of the revisions in expectations is straightforward; to do so involves requires writing the $n$ variable, $p$ lag VAR as a first-order system,

$$
\begin{equation*}
z_{t+1}=A z_{t}+\varepsilon_{t+1} \tag{9}
\end{equation*}
$$

where $z_{t+1}$ is an appropriately stacked $n p \times 1$ vector containing the excess equity return, the real interest rate, and any additional indicators. With the VAR expressed in this form, the ingredients of the identity 8 are given by

$$
\begin{aligned}
& e_{t+1}^{y}=s_{y} \varepsilon_{t+1}, \\
& \tilde{e}_{t+1}^{y}=s_{y} \rho A(1-\rho A)^{-1} \varepsilon_{t+1}, \\
& \tilde{e}_{t+1}^{r}=s_{r}(1-\rho A)^{-1} \varepsilon_{t+1} \text { and } \\
& \tilde{e}_{t+1}^{d}=e_{t+1}^{y}+\tilde{e}_{t+1}^{y}-\tilde{e}_{t+1}^{r},
\end{aligned}
$$

where $s_{y}$ and $s_{r}$ are appropriate $1 \times n p$ selection matrices.
Two features of the Campbell-Ammer method deserve further comment. One is its parametric approach to constructing long-horizon expectations of stock returns: one has to assume that the dynamics of equity returns many years in the future are adequately captured by a parsimonious VAR model. To a large extent, this parametric approach is forced upon us, as the relatively short, 13-year experience with Fed funds futures is not sufficient to directly estimate the long-horizon impact on stock asset returns, particularly in light of the questionable small-sample properties of long-horizon regressions [see Nelson and Kim (1993)]. But as discussed below, the use of the VAR does allow us to estimate the dynamics of stock returns over a longer sample than the period for which futures data are available.

A second important feature of the approach is that dividends are not included explicitly as a variable to be forecast; given $e_{t+1}^{y}, \tilde{e}_{t+1}^{y}$ and $e_{t+1}^{r}, e_{t+1}^{d}$ is backed out from the identity (8). In principle, it would be possible to forecast dividends directly in the VAR, and instead back out an implied $\tilde{e}_{t+1}^{y}$. In practice, however, this is complicated by a strong seasonal pattern, and a root near unity in the dividend process. It is important to note that to the extent that the VAR understates the predictability of excess returns, treating dividends as a residual means that the method will end up attributing too much of the return volatility to
dividends. ${ }^{13}$

### 3.1 The forecasting VAR

The first step is to set up a VAR to capture the dynamic correlations between the excess equity return and the real interest rate (calculated as the one-month bill yield minus the log difference in the non-seasonally-adjusted CPI). The VAR must therefore include these two variables at a minimum, plus whatever other variables that might be useful in forecasting them. (One important constraint, of course, is that these variables are available in real time.) We follow Campbell and Ammer (1993) in using a six-variable one-lag system that included, besides the real rate and equity return: the relative bill rate (defined as the threemonth bill rate minus its 12 -month lagged moving average), the change in the bill rate, the (smoothed) dividend price ratio, and the spread between the 10 -year and one-month Treasury yields.

### 3.2 A variance decomposition of equity returns

Equation 8 expresses this month's excess equity returns into three components, which may be correlated with one another. The variance of the current excess return can therefore be broken down into the sum of the three variances, plus (or minus) the relevant three covariances,

$$
\begin{align*}
\operatorname{Var}\left(e_{t+1}^{y}\right)= & \operatorname{Var}\left(\tilde{e}_{t+1}^{d}\right)+\operatorname{Var}\left(\tilde{e}_{t+1}^{r}\right)+\operatorname{Var}\left(\tilde{e}_{t+1}^{y}\right)- \\
& 2 \operatorname{Cov}\left(\tilde{e}_{t+1}^{d}, \tilde{e}_{t+1}^{r}\right)-2 \operatorname{Cov}\left(\tilde{e}_{t+1}^{d}, \tilde{e}_{t+1}^{y}\right)+2 \operatorname{Cov}\left(\tilde{e}_{t+1}^{y}, \tilde{e}_{t+1}^{r}\right), \tag{10}
\end{align*}
$$

giving a sense of the relative contributions of news about real interest rates, dividends, and expected future excess returns to fluctuations in the current excess return. The results of

[^11]this decomposition appear in table 7. For comparison, the table displays results for the full 1973-2002 sample and for the subsample beginning in May 1989, when the Fed funds futures data became available. The columns labeled "total" show the total contribution, and those labeled "share" expresses that contribution as a percentage of the excess return variance, i.e., normalizing by $\operatorname{Var}\left(e_{t+1}^{y}\right)$.

The results for the full 1973-2002 sample are similar to those reported by Campbell and Ammer (1993) for their 1973-87 sample. In particular, the variance in expected future excess returns accounts for the majority of the variance of the current equity return: 76 percent, compared with Campbell and Ammer's 101 percent. Dividends make a correspondingly larger contribution of 24.5 percent, as opposed to Campbell and Ammer's 14 percent. In both cases, the contribution of the real interest rate is negligible ( 0.3 and 3 percent respectively) and statistically insignificant.

The truncated 1989-2002 subsample yields somewhat different results, as shown in the right-hand portion of the table. Considerably less variance is attributed to revisions in expectations of future excess returns, and the dividend component now plays a somewhat larger role. The main reason for this seems to be a decline in the forecastability of equity returns in recent years, consistent with the observed fall in the adjusted $R$-squared from 0.04 to basically zero. With returns less forecastable, the Campbell-Ammer methodology by default assigns more of the excess return variance to dividend news.

### 3.3 The effects of Fed funds surprises

The most straightforward way to analyze the impact of monetary policy within the framework introduced above is to include the Fed funds surprises in the VAR as an exogenous variable

$$
\begin{equation*}
z_{t+1}=A z_{t}+\phi \bar{\Delta} \hat{r}_{t+1}^{u}+\varepsilon_{t+1}^{*} \tag{11}
\end{equation*}
$$

Table 7: Variance decomposition of excess equity returns

|  | 1973-2002 |  | 1989-2002 |  |
| :---: | :---: | :---: | :---: | :---: |
|  | Total | Share (\%) | Total | Share (\%) |
| $\operatorname{Var}$ (excess return) | 21.5 |  | 19.0 |  |
| $\operatorname{Var}$ (dividends) | 5.3 | $\begin{aligned} & 24.5 \\ & (6.2) \end{aligned}$ | 6.1 | $\begin{aligned} & 31.9 \\ & (1.8) \end{aligned}$ |
| Var(real rate) | 0.3 | $\begin{gathered} 1.4 \\ (2.4) \end{gathered}$ | 0.1 | $\begin{gathered} 0.6 \\ (1.5) \end{gathered}$ |
| $\operatorname{Var}$ (future returns) | 16.4 | $\begin{aligned} & 76.0 \\ & (1.8) \end{aligned}$ | 7.2 | $\begin{aligned} & 38.0 \\ & (1.2) \end{aligned}$ |
| $-2 \operatorname{Cov}($ dividends, real rate) | -0.4 | $\begin{gathered} -2.1 \\ (0.8) \end{gathered}$ | -0.6 | $\begin{gathered} -3.2 \\ (0.7) \end{gathered}$ |
| $-2 \operatorname{Cov}$ (dividends, future excess return) | 0.2 | $\begin{array}{r} 1.0 \\ (0.0) \end{array}$ | 7.2 | $\begin{array}{r} -37.7 \\ (2.3) \end{array}$ |
| $2 \operatorname{Cov}$ (future excess return, real rate) | -0.2 | $\begin{gathered} 0.8 \\ (0.1) \end{gathered}$ | 1.0 | $\begin{array}{r} 5.1 \\ (1.1) \end{array}$ |
| $\bar{R}^{2}$ from excess return equation | 0.040 |  | -0.003 |  |

Notes: The equity return used is the CRSP value-weighted index. Parentheses contain $t$-statistics, calculated using the delta method.
where $\phi$ is an $n \times 1$ vector capturing the contemporaneous response of the elements of $z_{t+1}$ to the unanticipated rate change period $t+1$. The new disturbance term $\varepsilon_{t+1}^{*}$ is by construction orthogonal to the funds rate surprise. This effectively breaks the VAR's one-month-ahead forecast error into a component having to do with news about monetary policy, $\phi \bar{\Delta} \hat{r}_{t+1}^{u}$ and an orthogonal component incorporating information about things other than policy.

An important point is that because $\Delta \hat{r}_{t+1}^{u}$ represents a prediction error from a rational forecast made at time $t$, it should be orthogonal to $z_{t} .{ }^{14}$ Consistent estimates of both $A$ and $\phi$ can therefore be obtained by first estimating the usual VAR (equation 9), and then regressing the VAR's one-step-ahead forecast errors on the funds rate surprises. Normally,

[^12]there would be no advantage to the two-step procedure over simply estimating equation 11 directly. But in our case, using the two-step procedure allows us to estimate the VAR dynamics (i.e., the coefficients in the $A$ matrix) over a sample longer than the period for which Fed funds futures are available. ${ }^{15}$ This will of course tend to improve the precision of our estimates.

### 3.3.1 The dynamic response to funds rate surprises

Incorporating the Fed funds surprises into the VAR in this way allows us to do two things. First, because it extracts an orthogonal element from the $\varepsilon_{t}$ forecast error, we can use it to calculate the dynamic responses of the variables in the VAR to the orthogonal component. The $k$-month response to a one-percentage-point surprise increase in the funds rate can be calculated quite simply as $A^{k} \phi$.

An obvious question to arise at this point concerns the relationship between these futures-based funds rate surprises and the more familiar monetary policy shocks derived from an identified VAR. The methods used to construct the one-month-ahead funds rate forecasts differ, of course, with one using the futures market's implicit forecast, and the other using a reduced-form econometric model. Forecast methodologies aside, however, the orthogonalization procedure described above is conceptually equivalent to ordering the funds rate first in a VAR system. Since this precludes any contemporaneous reaction of the funds rate to economic news, the surprises calculated in this way may well incorporate an endogenous policy response to information arriving within the month. Consequently, the impulse responses may represent the effects of things other than monetary policy per se.

One way to minimize this problem would be to purge the futures-based funds rate surprises of any contemporaneous response to the economy by projecting them onto the rele-

[^13]vant information variables, such as the data news obtained from the MMS survey. Alternatively, since the results above in section 2.5 indicate there has been little, if any, correlation between the funds rate surprises and data news since 1994, the $\phi$ estimated only on the post-1994 subsample should be relatively free from this endogeneity problem. This is the approach taken in the results presented below.

The upper-left-hand panel of figure 4 displays the dynamic response of excess returns calculated in this way. The initial decline of 11.6 percent (not shown, because of the difference in scale) is followed by another month of negative returns, and then by several months of near-zero excess returns. ${ }^{16}$ After six months, equities start to exhibit small positive excess returns, peaking at 0.16 percent per month ( 1.9 percent at an annual rate), and continuing for a period measured in years.

The contractionary funds rate surprise also leads to a sizable increase in the relative bill rate, which persists several months (by construction). The real T-bill rate rises sharply at first, but the increase is relatively short-lived, and all but disappears after four months. In the near term, the dynamics of equity excess returns are dominated by the effects of rising interest rates. But as these effects die out, the long-run effect of the dividend-price ratio, which rises as a result of the fall in equity prices, reasserts itself. It is this that leads to the highly persistent, positive excess returns visible in the impulse response function.

### 3.3.2 Explaining the stock market's reaction to Fed policy

The second thing this approach allows us to do is calculate the impact of the Fed funds surprises on the discounted sums of expected future excess returns, interest rates, and dividends. And since it is these present values that are related to the current excess return through equation 8, this provides a natural way to determine the source (or sources) of the stock market's reaction to monetary policy.

[^14]Figure 4: The estimated responses to monetary policy surprises


Notes: Each panel represents the response to a 1 percent funds rate surprise, as defined in equation 5, calculated from the VAR using value-weighted CRSP returns and estimated over the 1973-2002 sample. The period-zero response to the funds rate surprises $(\phi)$ is estimated on the February 1994 to December 2002 subsample. The initial excess return response is not shown, because of the large difference in scale.

One way to assess policy's effect on these discounted sums is simply to use the VAR to calculate $\tilde{e}_{t+1}^{d}, \tilde{e}_{t+1}^{r}$, and $\tilde{e}_{t+1}^{y}$, which represent the revisions in expectations of the relevant present values, and regress these variables in turn on $\hat{r}_{t+1}^{u}$. Although this would provide the answer we are after, the standard errors would be misleading, as they would fail to take into account the dependence of the $\tilde{e} s$ on the estimated parameters of the VAR.

An alternative way to do the same calculation is to write out the $\tilde{e} s$ in terms of the VAR coefficients. Taking $\tilde{e}_{t+1}^{y}$ as an example:

$$
\begin{aligned}
\tilde{e}_{t+1}^{y} & =s_{y} \rho A(1-\rho A)^{-1} \varepsilon_{t+1} \text { or } \\
& =s_{y} \rho A(1-\rho A)^{-1}\left(\phi \bar{\Delta} \hat{r}_{t+1}^{u}+\varepsilon_{t+1}^{*}\right) .
\end{aligned}
$$

The response of the present value of expected future excess returns to the FF surprise is just

$$
s_{y} \rho A(1-\rho A)^{-1} \phi .
$$

Thus, the response of expected future excess returns depends not only on the $\phi$ vector, but also on the VAR dynamics represented by $A$. Similarly, the response of the present value of current and expected future real returns is

$$
s_{r}(1-\rho A)^{-1} \phi,
$$

and the implied response of the present value of current and expected future dividends is

$$
s_{y} \phi+s_{y} \rho A(1-\rho A)^{-1} \phi+s_{r}(1-\rho A)^{-1} \phi
$$

or alternatively

$$
\left(s_{y}+s_{r}\right)(1-\rho A)^{-1} \phi .
$$

The standard errors for these responses can be calculated in the usual way, using the delta method.

Table 8: The impact of monetary policy on dividends, interest rates, and future returns

|  | Sample used for VAR |  |
| :--- | :---: | :---: |
|  | $1 / 73-12 / 02$ | $5 / 89-12 / 02$ |
| Current excess return | -11.55 | -11.01 |
|  | $(3.87)$ | $(3.72)$ |
| Future excess returns | 6.10 | 3.29 |
|  | $(1.74)$ | $(1.10)$ |
| Real interest rate | 0.64 | 0.77 |
|  | $(1.03)$ | $(1.87)$ |
| Dividends | -4.82 | -6.96 |
|  | $(1.73)$ | $(2.35)$ |

Notes: Parentheses contain $t$-statistics. The VAR is estimaed over the sample indicated, and the period-zero response to the funds rate surprises $(\phi)$ is estimated on the February 1994 to December 2002 subsample.

The results of these calculations appear in table 8 . With the VAR estimated over the entire 1973-2002 sample, funds rate surprises have a large, marginally significant impact on the discounted sum of future excess returns, accounting for just over half of the -11.55 contemporaneous excess return response. The reason for this large contribution is readily understood in terms of the impulse responses plotted in figure 4. Though small, funds rate shocks are estimated to have a highly persistent positive effect on excess returns. Discounting these future positive excess returns back using a discount factor near unity yields a large negative impact on the current excess return.

The -4.82 impact on dividends is nearly as large as that of future excess returns, at it too is significant at the 0.10 level. The impact on the discounted sum of real rates is very small, however, and accounts for less than one percentage point of the excess return response. This is to be expected, however, as real rates play a small role overall in the Campbell-Ammer variance decomposition.

The results are qualitatively similar when the VAR is estimated over the shorter 1989-

2000 sample. The main difference is the smaller impact on expected future excess returns, which now account for a statistically insignificant 3.29 percentage points of the -11.01 percent response. The reason for this can be traced to the smaller amount of long-run forecastability in excess returns in the post-1989 sample. In fact, the impulse response functions from this truncated sample (not shown) are nearly identical to those for the full 1973-2002 sample, shown above. The main difference is that the response of the excess return is negligible after six months or so, and it is this difference that accounts for the smaller contribution of future excess returns.

## 4 Conclusions

One contribution of this study has been to document the strong response of the stock market to unexpected monetary policy actions, using Fed funds futures data to gauge policy expectations. For broad stock market gauges like the CRSP value-weighted index or the S\&P 500 composite, an unexpected 25 basis point rate cut would typically lead to an increase in stock prices on the order of 1 percent. The average response has grown larger in recent years, primarily because there is less of a tendency for policy actions to accompany the release of economic news which would tend to offset policy's impact. There is also some evidence of a larger response to reversals in the direction of funds rate movements, and a smaller response to unexpected inaction on the part of the FOMC. Another finding is that the reactions differ considerably across industries, with the most sensitive high-tech and telecommunications sectors exhibiting a response half again as large as that of the broad market indices. Other sectors, such as energy and utilities, seem not to be significantly affected by monetary policy.

A more difficult question is why stock prices respond as they do to monetary policy whether through the effects on real interest rates, expected future dividends, or expected
future stock returns. The results presented in this paper showed, perhaps surprisingly, that the reaction of equity prices to monetary policy is not directly attributable to policy's effects on the real interest rate. This is a result of the relatively transitory movements in real interest rates induced by surprise policy actions.

Instead, the impact of monetary policy surprises on stock prices seems to come either through its effects on expected future excess returns, or on expected future dividends. The exact breakdown between these two depends somewhat on the choice of sample to the extent that this affects the long-horizon forecastability of excess returns. But at the very least, these results suggest that understanding the impact of monetary policy on expected future equity returns - and by extension, the equity premium - may hold the key to explaining the stock market's reaction to Federal Reserve policy.

## Appendix: deriving equation 8

This appendix provides a brief sketch of the derivation of the log-linearized relationship between the current excess return, expected future excess returns, dividend growth, and real interest rates given in equation 8 . The derivation roughly follows Campbell and Shiller (1988) and Campbell (1991).

The starting point is simply the definition of the stock return, $H_{t+1}$ :

$$
1+H_{t+1} \equiv \frac{P_{t+1}+D_{t}}{P_{t}}
$$

where $P$ is the stock price and $D$ is the dividend. Taking logs and letting $h_{t+1}=\ln \left(1+H_{t+1}\right)$ yields:

$$
h_{t+1}=\ln \left(P_{t+1}+D_{t}\right)-\ln \left(P_{t}\right) .
$$

The next step is to derive a log-linear approximation to $\ln \left(P_{t+1}+D_{t}\right)$. One way to do this is to first-difference, and express the change in the log of the sum as the weighted sum of the log differences

$$
\Delta \ln \left(P_{t+1}+D_{t}\right) \approx \rho \Delta p_{t+1}+(1-\rho) \Delta d_{t}
$$

where $\rho$ is the steady-state $P /(D+P)$. "Integrating" this expression gives

$$
\ln \left(P_{t+1}+D_{t}\right) \approx k+\rho p_{t+1}+(1-\rho) d_{t}
$$

substituting this into the expression for $h_{t+1}$, substituting $\delta_{t}$ for $d_{t-1}-p_{t}$, and combining terms gives

$$
\begin{align*}
h_{t+1} & \approx k-\rho \delta_{t+1}+\delta_{t}+\Delta d_{t}  \tag{12}\\
& \approx k+\left(1-\rho L^{-1}\right) \delta_{t}+\Delta d_{t} .
\end{align*}
$$

The next step is to solve forward, giving

$$
\begin{aligned}
\delta_{t} & =\left(1-\rho L^{-1}\right)^{-1}\left(h_{t+1}-\Delta d_{t}-k\right) \\
& =\sum_{i=0}^{\infty} \rho^{i}\left(h_{t+1+i}-d_{t+i}\right)-k /(1-\rho) .
\end{aligned}
$$

Substituting this, and a similar expression for $\delta_{t+1}$, into equation 12 and collecting terms yields:

$$
h_{t+1}-E_{t} h_{t+1}=-\sum_{i=1}^{\infty} \rho^{i}\left(E_{t+1}-E_{t}\right) h_{t+1+i}+\sum_{i=0}^{\infty} \rho^{i}\left(E_{t+1}-E_{t}\right) \Delta d_{t+1+i}
$$

which corresponds to equation 1 in Campbell (1991).

A breakdown of excess returns can then be derived by expressing the equity return $h_{t+1}$ as the sum of a risk-free rate and an excess return

$$
h_{t+1}=r_{t+1}+y_{t+1} .
$$

Because it is assumed that $r_{t+1}$ is known at time $t$, the "excess return surprise" $y_{t+1}-E_{t} y_{t+1}$ is the same as the overall return surprise $h_{t+1}-E_{t} h_{t+1}$. So the risk-free rate can be included in the two-way breakdown as follows:

$$
y_{t+1}-E_{t} y_{t+1}=-\sum_{i=1}^{\infty} \rho^{i}\left(E_{t+1}-E_{t}\right)\left(y_{t+1+i}+r_{t+1+i}\right)+\sum_{i=0}^{\infty} \rho^{i}\left(E_{t+1}-E_{t}\right) \Delta d_{t+1+i}
$$

or as

$$
\begin{align*}
y_{t+1}-E_{t} y_{t+1}= & -\sum_{i=1}^{\infty} \rho^{i}\left(E_{t+1}-E_{t}\right) y_{t+1+i}- \\
& \sum_{i=1}^{\infty} \rho^{i}\left(E_{t+1}-E_{t}\right) r_{t+1+i}+\sum_{i=0}^{\infty} \rho^{i}\left(E_{t+1}-E_{t}\right) \Delta d_{t+1+i} . \tag{13}
\end{align*}
$$

Again, because $E_{t} r_{t+1}=r_{t+1}$, it doesn't matter whether the summation involving the $r \mathrm{~s}$ begins at 0 or 1 . Finally, letting $e_{t+1}^{y}$ represent the "excess return surprise" and replacing the summations with the corresponding $\tilde{e} s$ yields equation 8 .

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[^1]:    ${ }^{1}$ Cochrane and Piazzesi (2002) propose using the change in term eurodollar rates, while Rigobon and Sack (2002) utilize the eurodollar futures rate. While these measures provide informative gauges of interest rate expectations over a slightly longer horizon, Gürkaynak, Sack and Swanson (2002) show that Fed funds futures do the best job of forecasting target rate changes one to five months ahead.

[^2]:    ${ }^{2}$ The Federal funds rate was either implicitly or explicitly the operating instrument of Federal Reserve policy over the period analyzed.

[^3]:    ${ }^{3}$ As discussed in Kuttner (2001), when the rate change comes on the first day of the month, $f_{s-1, m}^{1}$ would be used instead of $f_{s, t-1}^{0}$. Also, to avoid amplifying any month-end noise, when the rate change falls on one of the last three days of the month, the unscaled change one-month futures rate is used instead of the change in the spot month rate.

[^4]:    ${ }^{4}$ The 17 September 2001 observation is dropped from the sample, however, as that was the first day of trading following the September 11 terrorist attacks.
    ${ }^{5}$ This estimate is somewhat smaller than that the -7.1 reported in Rigobon and Sack (2002). That study's use of Eurodollar futures, rather than Fed funds futures, accounts for most of the difference.

[^5]:    ${ }^{6}$ This "timing hypothesis" was suggested by Kuttner (2001) as a likely explanation for the smaller-thanexpected response of interest rates to Fed funds surprises. Some support for this view can be found in the results of Demiralp and Jordá (2002) and Gürkaynak, Sack and Swanson (2002).

[^6]:    ${ }^{7}$ The settlement price of the Fed funds futures contract is determined by the average over the calendar month, carrying the prior business day's rate over to weekends and holidays.

[^7]:    ${ }^{8}$ One solution would have been to assume that post-1994 rate changes were always expected to occur at scheduled FOMC meetings. The three intermeeting rate cuts in 2001 have made this assumption less plausible, however.
    ${ }^{9}$ We are indebted to Eric Swanson for his assistance with these data.

[^8]:    ${ }^{10}$ Again, retail sales is significant with the "wrong" sign. But this result is due entirely to an anomalous 7 percent jump in retail sales in November 2001, which happened to occur in a month in which the Fed also cut the funds rate target.

[^9]:    ${ }^{11}$ Interestingly, the employment surprise is negative and significant in a univariate regression (not reported), but becomes insignificant once the Fed funds surprise is included. This is consistent with the findings of Boyd et al. (2001), and corroborates their conjecture that the policy response accounts for equities' perverse response.

[^10]:    ${ }^{12}$ Because VARs require periodic time series data, the subsequent analysis will use the monthly measure of the funds rate surprises.

[^11]:    ${ }^{13}$ A useful check on the Campbell-Ammer procedure would be to compare its implied dividend forecasts with the observed behavior of dividends. Such a comparison is beyond the scope of the present paper, however.

[^12]:    ${ }^{14}$ Krueger and Kuttner (1996) showed that in practice, the Fed funds futures prediction errors are generally uncorrelated with lagged information.

[^13]:    ${ }^{15}$ Faust, Swanson and Wright (2002) employed a similar trick: they estimate the VAR parameters over the full sample, but choose an orthogonalization based on the response of interest rates over the post-1989 subsample.

[^14]:    ${ }^{16}$ This 11.6 percent response differs slightly from the results in section 2.5 because the dependent variable is the forecast error in the log excess return, rather than the raw nominal return.

