# What Role Do Tech Companies' R\&D Expenditures Play in Analysts Sales and Earnings Forecasts? 

Vijaykumar Gandapodi<br>Georgia State University

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# What Role Do Tech Companies' R\&D Expenditures Play in Analysts' Sales and Earnings Forecasts? 

## By

Vijaykumar Gandapodi

A Dissertation Submitted in Partial Fulfillment of the Requirements for the Degree

Of

Executive Doctorate in Business

In the Robinson College of Business
Of

Georgia State University

GEORGIA STATE UNIVERSITY

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This dissertation was prepared under the direction of Sergio Quinones-Romandia Dissertation Committee. It has been approved and accepted by all members of that committee, and it has been accepted in partial fulfillment of the requirements for the degree of Executive Doctorate in Business in the J. Mack Robinson College of Business of Georgia State University.

Richard Phillips, Dean

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## TABLE OF CONTENTS

LIST OF TABLES ..... vi
LIST OF FIGURES ..... x
ABSTRACT ..... xi
I CHAPTER 1: INTRODUCTION ..... 1
II CHAPTER 2: LITERATURE SYNTHESIS ..... 4
III CHAPTER 3: THEORETICAL BACKGROUND AND HYPOTHESES ..... 11
III. 1 Research and Development Expense (R\&D) ..... 13
III. 2 Current Year EPS Estimate Revision ..... 14
III. 3 Next Year EPS Estimate Revision ..... 16
III. 4 Current Year Sales Estimate Revision ..... 17
III. 5 Next Year Sales Estimate Revision ..... 18
IV CHAPTER 4: DATA COLLECTION AND ANALYSIS ..... 19
IV. 1 Distribution of Samples over Time, by Size and by Industry ..... 24
IV.1.1 Time ..... 24
IV.1.2 Size ..... 25
IV.1.3 Industry ..... 26
IV.1.3.1 Software and Services group ..... 26
IV.1.3.2 Technology Hardware and Equipment ..... 26
IV.1.3.3 Semiconductor. ..... 27
IV. 2 Data analysis ..... 28
V CHAPTER 5: RESULTS ..... 31
V. 1 Descriptive Statistics ..... 31
V. 2 Correlation Matrix of Independent Variables ..... 34
V. 3 Hypothesis 1: A change in R\&D expenditure impacts analysts' sales estimate forecasts for the current year ..... 38
V. 4 Hypothesis 2: A change in R\&D expenditure impacts analysts' sales estimate forecasts for the next year. ..... 39
V. 5 Hypothesis 3: A change in R\&D expenditure impacts analysts' EPS estimate forecasts for the current year ..... 51
V. 6 Hypothesis 4: A change in R\&D expenditure impacts analysts' EPS estimate forecasts for the following year. ..... 51
VI CHAPTER 6: DISCUSSION ..... 73
VI. 1 Correlation analysis ..... 79
VI. 2 Regression Analysis ..... 87
VI. 3 Summary of Hypothesis Results ..... 100
VI. 4 Contribution to Practice ..... 101
VI. 5 Contribution to Theory ..... 102
REFERENCES ..... 104
VITA ..... 109

## LIST OF TABLES

Table 1 R\&D and Company-Related Information ..... 21
Table 2 Analyst Data from IBES ..... 22
Table 3 Data Calculated and Consolidated in Excel Spread Sheet ..... 23
Table 4 Year Wise Number of Firms ..... 25
Table 5 Industry Grouping ..... 26
Table 6 Number of Records Industry Wide ..... 27
Table 7 Sub Industries and Company Examples. ..... 28
Table 8 Descriptive Statistics ..... 32
Table 9 Correlation Matrix for All Independent Variables ..... 34
Table 10 Correlation Matrix for Large Size Firms ..... 35
Table 11 Correlation Matrix for Medium Size Firms ..... 36
Table 12 Correlation Matrix for Small Size Firms ..... 37
Table 13 Correlation R\&D and Analyst Sales Estimate ..... 38
Table 14 Regression of R\&D, Revenue, ... with Analyst Sales Estimate. ..... 39
Table 15 Correlation of R\&D with Analyst Sales - Market Cap Segmentation ..... 42
Table 16 Regression of R\&D, Revenue, ... with Analyst Sales Estimate (Large Size companies) ..... 43
Table 17 Regression of R\&D, Revenue, ... with Analyst Sales Estimate (Medium Size companies) ..... 44
Table 18 Regression of R\&D, Revenue, ... with Analyst Sales Estimate (Small size companies) ..... 45
Table 19 Correlation of R\&D with Analyst Sales - Industry Group Segmentation ..... 47
Table 20 Regression of R\&D...... with Analyst Sales Estimate (Software \& Services). 48
Table 21 Regression of R\&D...... with Analyst Sales Estimate (Technology Hardware)49
Table 22 Regression of R\&D.......with Analyst Sales Estimate (Semiconductor) ..... 50
Table 23 Correlation R\&D with Analyst EPS Estimate ..... 51
Table 24 Regression of R\&D, Revenue, ... with Analyst EPS Estimate ..... 52
Table 25 Regression of R\&D, ... with Analyst EPS Estimate (Without Revenue) ..... 53
Table 26 Correlation of R\&D with Analyst EPS - Market Cap Segmentation. ..... 54
Table 27 Regression of R\&D, Revenue, ... with Analyst EPS Estimate (Large Size
$\qquad$companies)55
Table 28 Regression of R\&D, ... with Analyst EPS Estimate, Large Size companies (No
Revenue) ..... 56
Table 29 Regression of R\&D, Revenue, ... with Analyst EPS Estimate (Medium Size
companies) ..... 57
Table 30 Regression of R\&D... with Analyst Sales Estimate, (Medium Size companies)(No Revenue). ..... 58
Table 31 Regression of R\&D, Revenue, .. with Analyst EPS Estimate ( Small Size companies) ..... 59
Table 32 Regression of R\&D... with Analyst EPS Estimate (Small Size companies) (No
Revenue) ..... 60
Table 33 Correlation of R\&D with Analyst EPS - Industry Group Segmentation ..... 61
Table 34 Regression of R\&D......with Analyst EPS Estimate (Software and Services
companies) ..... 62
Table 35 Regression of R\&D.......with Analyst EPS Estimate (Software and service companies) (No Revenue)........................................................................................ 63

Table 36 Regression of R\&D...... with Analyst EPS Estimate (Technology Hardware companies)64
Table 37 Regression of R\&D......with Analyst EPS Estimate (Technology hardware companies) (No Revenue). ..... 65
Table 38 Regression of R\&D......with Analyst EPS Estimate (Semiconductor companies) ..... 66
Table 39 Regression of R\&D......with Analyst Sales Estimate (Semiconductor companies)(No Revenue). ..... 67
Table 40 Correlation R\&D with Analyst Sales Estimate for R\&D increase > 5\% ..... 68
Table 41 Regression of R\&D, Revenue.

$\qquad$
with Analyst Sales Estimate with R\&D
increase $>5 \%$ ..... 69
Table 42 Correlation R\&D with Analyst EPS Estimate for R\&D increase > 5\% ..... 70
Table 43 Regression of R\&D, Revenue .with Analyst EPS Estimate for R\&D increase
$\qquad$$>5 \%$.71
Table 44 Regression of R\&D.

$\qquad$
.with Analyst EPS Estimate (No Revenue) filtered for
R\&D increase > 5\% ..... 72
Table 45 Consolidated Correlation of R\&D with Analyst Sales and EPS Revision ..... 74
Table 46 Consolidated Correlation of R\&D with Analyst Sales and EPS Revision for first
10 years (1995-2004) ..... 77
Table 47 Consolidated Correlation of R\&D with Analyst Sales and EPS Revision for first
10 years (2005-2014) ..... 78
Table 48 Distribution of R\&D Spending ..... 80
Table 49 Correlation Summary of All Tests, Year Wise View ..... 84
Table 50 Correlation tests for analyst sales estimate with R\&D decade wise compared
with 20 years data ..... 86
Table 51 Consolidated Regression of R\&D with Analyst Sales and EPS Revision ..... 90
Table 52 Consolidated Regression of R\&D with Analyst Sales and EPS Revision (No
Revenue) ..... 95
Table 53 Regression Summary of All Tests, Year Wise View. ..... 97
Table 54 Number of R\&D Firms ..... 99
Table 55 Earnings of R\&D Firms ..... 100
Table 56 Summary of Hypothesis Results ..... 100

## LIST OF FIGURES

Figure 1 U.S. R\&D as Percent of Gross Domestic Product, 1953-2008 ..... 7
Figure 2 American Economic Association ..... 9
Figure 3 Hypothesis 1 ..... 12
Figure 4 Hypothesis 2 ..... 12
Figure 5 Hypothesis 3 ..... 12
Figure 6 Hypothesis 4 ..... 13
Figure 7 Input to Analyst Report ..... 16

ABSTRACT<br>What Role Do Tech Companies' R\&D Expenditures Play in Analysts' Sales and Earnings Forecasts?

by
Vijaykumar Gandapodi
August 2016

## Committee Chair: Conrad Ciccotello

Major Academic Unit: Executive Doctorate in Business
Many top market capitalization companies are information technology (IT) firms, including Apple, Google, Microsoft, and Facebook, each of which is valued at more than $\$ 300$ billion. Facebook is less than 10 years old and is one of the top 10 companies in the world in terms of market capitalization. However, technologies change rapidly; website revenue-which once grew at a brisk rate-has slowed down, while mobile technology growth is increasing and technology trends are shifting toward cloud hosting and big data analytics. IT companies that have increased their R\&D spending remain leaders throughout periods of technology change. Companies such as Facebook and Google have doubled and tripled their profits, respectively over the past decade. In this dynamic environment, analysts play a critical role in evaluating IT company financial statements and estimating company sales and earnings per share (EPS). This study examines how changes in R\&D spending are related to analysts' sales and earnings estimate revisions. An analysis of data over a 20-year period shows that analysts typically revise their sales estimates based on changes in a company's R\&D expenditures. The correlation between analyst earnings estimates and $R \& D$ expenditures, however, varies based on company size and industry within the IT sector. Analysts play a particularly important role in small companies, where the correlation between R\&D and sales changes is not as high as in
large companies. Analysts are thus critical to the functioning of capital markets in the IT sector.

## I CHAPTER 1: INTRODUCTION

In this research study, I explore whether analysts' revisions in sales and earnings per share (EPS) estimates are related to R\&D expenditure changes by technology companies. This issue is of interest to both stock market participants and managers of these firms. With the rising pressure to create and sustain competitive advantages through technological innovation, IT companies increasingly depend on the efficient management of research and development (R\&D) activities (Bone \& Saxon, 2000). R\&D investments are a critical element of growth in firms (Chan, Martin \& Kensinger, 1990). Market participants use analysts' forecasts because analysts process and transform the information contained in financial statements-along with additional information about the industry, firm strategy, and economy-into future earnings predictions (Wieland, 2011). Analysts' forecast revisions promote market price discovery (Gleason \& Lee, 2003) and market participants react to forecast revisions.

The problem statement for this study is: "What role do technology companies' R\&D expenditures play in analysts' sales and earnings forecasts?" This study zeros in on IT companies, extracting records filtered for that sector (group 45) from the Wharton Research Data Services (WRDS) Compustat database, with a focus on IT companies that have R\&D expenditures on their balance sheets. I extracted analysts' forecasts for IT company sales and EPS for the current and the following year from the Institutional Brokers' Estimate System (I/B/E/S) database.

This research focuses on companies trading in US exchanges. I categorize companies by size into small, midsized, and large based on market capitalization. I also examine various industries within the IT sector; as one study found, $R \& D$ intensity that is higher
than the particular industry's average leads to larger stock-price increases for firms in high-tech industries (Chan, Martin \& Kensinger, 1990).

I observe that analysts do change their sales forecasts in response to changes in R\&D expenditures by technology firms. The relation is robust across three different sectors of the IT industry. Analysts also change their EPS estimates in response to changes in $R \& D$, although this relation is not as strong when I consider changes in sales. Interestingly, I find that in the smallest capitalization firms, analysts change EPS estimates in response to changes in $R \& D$, even when sales changes are considered. I consider this as evidence of the importance of analysts to capital allocation in the technology industry.

The findings I captured during my analysis add to the academic research related to R\&D expenditure and analyst estimates. Research spending is heavily concentrated in technology and science-oriented industries. The computer programming, software, and services industry represents about 17 percent of the sales and two times the earnings compared to other companies in these sectors. Other research (Chan, Lakonishok \& Sougiannis, 2001) focuses on R\&D, but does not concentrate on the analyst aspect. This study will benefit practitioners, allowing them to make smarter investments based on R\&D expenditure. Further, analysts' recommendations on EPS and sales have a correlation with stock prices. Huo and Hung (2014) show that stock price drift emerges after analysts' revise their earnings forecasts.

The study will also benefit managers of technology companies by clarifying how analysts make revisions. For small firms, analysts play a particularly important role, as these firms tend to have less market coverage.

Analyst forecasts are superior to time-series forecasts because analysts possess both an information advantage and a timing advantage (Brown, 1987). Keung's study (2010) finds that earnings forecast revisions supplemented with sales forecast revisions have a greater impact on security prices than stand-alone earning forecast revisions. He further found that financial analysts are more likely to supplement their earnings forecasts with sales forecasts when they have better information. As Keung's study discusses in detail, supplementary sales forecasts appear to lend credibility to earnings forecasts because financial analysts provide better sales forecasts when they are more informed. These findings help us understand the characteristics of analysts' sales and EPS forecasts.

## II CHAPTER 2: LITERATURE SYNTHESIS

R\&D in technology companies is important for several reasons. For example, it influences executives' incentives, compensation, and firm performance. Currim, Lim, and Kim (2012) found the increase in equity-to-bonus compensation ratio for top executives is positively associated with an increase in R\&D spending. Further, in their work on analyst forecast revisions and market price discovery, Gleason and Lee found a postrevision price drift associated with these forecast revisions (Gleason \& Lee, 2003). Also, changes in R\&D expenditure in either direction indicate transitions between exploitative and exploratory R\&D and are associated with increased firm performance (Mudambi \& Swift, 2014). Such examples illustrate the importance of R\&D expenditure and how it impacts top executives' compensation and the stock price performance of firms.

This association is more intense in high-growth firms and is especially significant in the high-tech sector (García-Manjón \& Romero-Merino 2012). This study further found that policymakers and business leaders in the high-tech sector maintain R\&D expenditures even when facing a recession. Graham and Frankenberger (2008) report that increases in R\&D spending in recessions increase firm profit and intangible value. Even during the recession, companies reduce $\mathrm{R} \& \mathrm{D}$ spending to meet their quarterly results, which in turn impacts the growth of the firm. As this literature synthesis shows, R\&D spending is a key metric for increasing or decreasing firm's earnings.

R\&D is also important to market participants. The study by Kumar, Charurvedula, Rastogi, and Bang (2009) found that buy recommendations issued by analysts help investors generate abnormal returns on the day of the recommendation. On the other hand, sell recommendations do not show significant negative abnormal returns.

An analyst forecast revision and market price discovery study by Gleason and Lee (2003) found that post-revision price drift is associated with analyst forecast revisions. That study documents the following four significant factors:

1. The market does not sufficiently distinguish between revisions that provide new information and revisions that merely move toward the consensus.
2. The price adjustment process is faster and more complete for celebrity analysts than for more obscure yet highly accurate analysts.
3. The price adjustment process is faster and more complete for firms with greater analyst coverage.
4. A substantial portion of the delayed price adjustment occurs around subsequent earnings-announcement and forecast-revision dates.

The above studies confirm that investors use analyst information for investing in the stock market. Hillary and Hsu (2013) empirically showed that analysts with a lower standard deviation of forecast errors have a greater ability to move prices. These results have three implications:

1. Consistent analysts are less likely to be demoted and more likely to be nominated as all-star analysts.
2. Analysts strategically deliver downward-biased forecasts to increase their consistency (sometimes at the expense of stated accuracy).
3. The benefits of consistency and of "lowballing" (accuracy) are to increase (or decrease) the institutional investor's presence.

These findings help us understand that analyst reports are used not only by individual investors, but also by institutional investors.

Fama and French (1992) also found positive abnormal returns associated with high Earnings to Price (E/P) stocks, but they found an even stronger relationship between book value to price $(\mathrm{B} / \mathrm{P})$ ratios and abnormal returns. Bauman and Dowen (1988) discovered mixed results between high growth stocks and stock returns; during their study, they found long-term, low growth stocks with low $\mathrm{P} / \mathrm{E}$ had higher return than higher growth stocks with higher P/E. These studies help us to understand the importance of the EPS; the Fama and French (1992) study confirms the significance of a company's earnings for the share price being traded. This literature synthesis illuminates how the investment community uses analyst forecast reports.

Analysts offer significantly greater coverage for firms with larger R\&D and advertisement expense relative to their industry, as well as for firms in industries with large R\&D expenses (Barth, Kasniz \& McNicholas, 1999). As Figure 1 shows, the US National Science Foundation offers a reliable source for R\&D trends for US companies and government, with a steady transfer of R\&D spending from government to the business sector.
U.S. R\&D as Percent of Gross Domestic Product, 1953-2008
(Total, Industrial, and Federal R\&D)


Figure 1 U.S. R\&D as Percent of Gross Domestic Product, 1953-2008
Source - National Science Foundation, (Hirschey et al. 2012)
Similar to current cash flow, growth, risk, and market share, advertising and R\&D expenditures are key determinants of a firm's market value (Chauvin \& Hirschey 1993). Chauvin and Hirschey's also found that the market value effects of advertising and R\&D are broadly operative throughout both manufacturing and nonmanufacturing sectors. They suggest that advertising and $\mathrm{R} \& \mathrm{D}$ are an attractive alternative means of investment in valuable intangible capital that have differing degrees of relevance in different economic sectors.

Sougiannis (1994) found that, on average, a one-dollar increase in R\&D leads to a two-dollar increase in profit over a seven-year period, with a five-dollar increase in
market value. Companies with high R\&D-to-equity-market value (which tend to have poor past returns) earn large excess returns (Chan, Lakonishok \& Sougiannis, 2001). Analyst accrual and forecast revision strategies generate returns of 15.5 percent and 5.5 percent, respectively, when implemented independently (Bath \& Hutton, 2003). Bath and Hutton add that a combined strategy that uses forecast revisions to refine the accrual strategy generates a return of 28.55 percent. They further discuss many studies pertaining to analyst earnings forecast; some of these studies argue that analysts don't account for key accounting data (Stober, 1992; Abarbanell \& Bushee, 1997). However, other studies point out that analyst forecasts are more accurate than time-series models in predicting future earnings (Brown, Griffin, Hagerman \& Zmijewski, 1987). Such research suggests that analyst forecasts have the potential to give investors value-relevant information about earnings (Bath \& Hutton, 2003).

Bath and Hutton note that a second stream of analyst forecast research focuses on whether investors actually heed the information in analyst forecast revisions. Numerous studies (e.g., Givoly \& Lakonishok, 1980; DeBondt, 1991; Mendenhall, 1991; Stickel, 1991; Gleason \& Lee, 2000; and Elgers, Lo, \& Pfeiffer, 2001) have found that analyst forecast revisions predict future returns, indicating that investors do not fully utilize the information reflected in the forecasts on a timely basis.

In their study, Hirschey et al. (2012) found that R\&D spending continues to grow faster than advertising and capital expenditures. As Figure 2 shows, they found that IT companies and small market capitalization companies that spend on R\&D increased their spending by 57 percent between 1976 and 2010, while non-IT sector companies and large
companies (which constitute most of the NYSE) reduced such spending by 12 percent.
The data in the Figure 2 does not include financial or utilities sector companies.

| Yar | CRSP <br> Industrial <br> firms | NYSEindustrialsM | AMEX,NASDAindostrials | All Industrial firms |  |  | NSSE Sirms |  |  | AMEX, XASDAQ f frus |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  | R\&D | Advertising | C.apes | R\&D | Advertising | Capes | R\&D | Adverising | Cupes |
| 1976 | 2663 | $9 \% 6$ | 1757 | 1,203 | 1284 | 2632 | 455 | 44 | 899 | 708 | 840 | 1,733 |
| 1977 | 2649 | 900 | 1749 | 1,160 | 1270 | 2622 | 480 | 430 | 893 | 680 | 840 | 1,729 |
| 1978 | 2589 | 898 | 1691 | 1,122 | 1225 | 2552 | 459 | 439 | 888 | 653 | 786 | 1,664 |
| 1979 | 2574 | 881 | 1693 | 1,114 | 1220 | 2542 | 458 | 431 | 875 | 656 | 789 | 1,667 |
| 1980 | 2640 | 874 | 1766 | 1,135 | 1191 | 2607 | 451 | 444 | 865 | 684 | 787 | 1,742 |
| 1981 | 2864 | 858 | 2006 | 1,241 | 1210 | 2799 | 439 | 389 | 849 | 802 | 821 | 1,950 |
| 1982 | 2903 | 836 | 2067 | 1,290 | 1225 | 2846 | 429 | 374 | 830 | 861 | 851 | 2,016 |
| 1983 | 3289 | 841 | 2448 | 1,495 | 1453 | 3195 | 429 | 383 | 830 | 1,066 | 1,070 | 2,365 |
| 1984 | 3442 | 826 | 2016 | 1,570 | 1551 | 3332 | 413 | 385 | 813 | 1,157 | 1,166 | 2519 |
| 1985 | 3506 | 804 | 2702 | 1,001 | 1556 | 3378 | 398 | 355 | 785 | 1,203 | 1201 | 2593 |
| 1986 | 3710 | 782 | 2928 | 1,673 | 1606 | 3576 | 378 | 340 | 759 | 1,295 | 1,326 | 2,817 |
| 1987 | 3864 | 799 | 3065 | 1,704 | 1691 | 3724 | 383 | 341 | 780 | 1,121 | 1,350 | 2,44 |
| 1988 | 3695 | 793 | 2902 | 1,659 | 1609 | 3576 | 379 | 334 | 779 | 1,280 | 1,275 | 2,797 |
| 1989 | 3567 | 794 | 273 | 1616 | 1515 | 3436 | 37 | 329 | 775 | 1,239 | 1,186 | 2,661 |
| 190 | 3525 | 830 | 2695 | 1590 | 1463 | 3407 | 389 | 326 | 814 | 1,201 | 1,137 | 2,593 |
| 1991 | 3591 | 878 | 2713 | 1663 | 1412 | 3452 | 412 | 341 | 880 | 1,251 | 1,101 | 2592 |
| 1902 | 3770 | 951 | 2819 | 1779 | 1423 | 3667 | 439 | 361 | 933 | 1,340 | 1,062 | 2,734 |
| 1593 | 4187 | 1013 | 3174 | 1939 | 1500 | 4084 | 455 | 385 | 991 | 1,484 | 1,115 | 3,093 |
| 1944 | 4010 | 1066 | 3340 | 1998 | 734 | 4314 | 473 | 204 | 1,47 | 1,226 | 530 | 3,267 |
| 1905 | 4602 | 1098 | 3504 | 2154 | 741 | 4495 | 488 | 206 | 1,076 | 1,666 | 535 | 3,419 |
| 1996 | 4986 | 1186 | 3800 | 2385 | 945 | 4866 | 498 | 254 | 1,157 | 1,887 | 691 | 3,709 |
| 1997 | 5000 | 1235 | 3765 | 2416 | 1034 | 4880 | 513 | 273 | 1200 | 1,903 | 761 | 3,680 |
| 1998 | 4723 | 1241 | 3482 | 2292 | 1032 | 4607 | 520 | 285 | 1,211 | 1,7\% | 747 | 3,3\% |
| 1999 | 4551 | 1184 | 3367 | 2275 | 1198 | 444 | 508 | 309 | 1,156 | 1,767 | 889 | 3,291 |
| 2000 | 4391 | 1091 | 3300 | 2308 | 1301 | 4294 | 463 | 309 | 1,603 | 1,445 | 992 | 3,231 |
| 2001 | 342 | 1071 | 2871 | 2138 | 1281 | 3844 | 450 | 340 | 1,44 | 1,678 | 941 | 2,800 |
| 2002 | 3662 | 1071 | 2591 | 1972 | 1294 | 3608 | 452 | 369 | 1,603 | 1,510 | 925 | 2545 |
| 2003 | 344 | 1056 | 2368 | 1840 | 1307 | 3403 | 465 | $3 \%$ | 1,050 | 1,375 | 908 | 2,353 |
| 2004 | 3441 | 1056 | 2385 | 1872 | 1411 | 3410 | 471 | 432 | 1,053 | 1,401 | 974 | 2,357 |
| 2005 | $339 \%$ | 1043 | 2353 | 1836 | 1433 | 3357 | 453 | 440 | 1,40 | 1,373 | 993 | 2,17 |
| 2006 | 3363 | 1040 | 2323 | 1796 | 1446 | 3313 | 453 | 456 | 1,036 | 1,43 | 990 | 227 |
| 2007 | 3338 | 1019 | 2319 | 1787 | 1415 | 3274 | 438 | 441 | 1,014 | 1,44 | 974 | 2260 |
| 2008 | 3155 | 998 | 2157 | 1686 | 137 | 3102 | 436 | 437 | 992 | 1,250 | 940 | 2,110 |
| 2009 | 3003 | 99\% | 2007 | 1592 | 1307 | 2973 | 430 | 429 | 992 | 1,162 | 878 | 1,981 |
| 2010 | 2902 | 991 | 1911 | 1543 | 1255 | 2882 | 432 | 423 | 989 | 1,111 | 832 | 1,893 |
| Absolute change over 1976-2010 | 239 | 85 | 154 | 340 | $-29$ | 250 | 63 | -21 | 9 | 403 | 8 | 160 |
| Percent change over 1976-2010 | 9.0\% | 9.4\% | 8.8\% | 28.3\% | -2.3\% | 9.5\% | -12.\% | 4.7\% | 10.\% | 56\%\% | -1.0\% | 9.2\% |

Figure 2 American Economic Association
(Hirshey et al. 2008)

Although various studies examine analyst forecasts and $\mathrm{R} \& \mathrm{D}$, no existing studies have examined how R\&D impacts analysts' sales and EPS estimates, nor have researcher's analyzed segmentation based on company size and industry type. This information can be very helpful for practitioners as they consider moving their companies into different industry segments or consider buying the stock of IT companies of different sizes. This quantitative study will fill this gap and contribute both to practitioners and the academic literature.

## III CHAPTER 3: THEORETICAL BACKGROUND AND HYPOTHESES

Based on the previous studies in my literature synthesis, it is clear that R\&D spending has an impact on what aspect of high-tech companies. Much existing literature focuses on analysts and their impact on stock price. Prior studies have shown that analysts' estimates of earnings or sales are quite close to the actual sales and earnings of the company-and hence the market reacts when analysts change their estimates. The stock's price moves on the day of an analyst's revision, especially if that analyst has maintained a strong reputation over the years in covering that sector. In their study, Kumar et al. (2009) found that buy recommendations issued by analysts on public domains help investors generate abnormal returns on the day of the recommendation, while sell recommendations show no significant negative abnormal returns.

Analysis done through this quantitative study will expand on the earlier studies and focus on the association between the R\&D spending and analysts' sales and EPS estimate forecasts. This research will compare the generated results between various industries within the IT sector. It will also review the impact of R\&D change on analysts’ estimate forecasts based on the size of the firms. Practitioners will benefit from the findings through a better understanding of how analysts modify their estimates for different industry groupings and company sizes in the IT sector. Further, by understanding how estimates are reflected in the current year versus the next year, management can better plan their R\&D spending budget to enhance the organization's future.

Hypothesis 1: A change in R\&D expenditure impacts analysts' sales estimate forecasts for the current year.


Figure 3 Hypothesis 1

Hypothesis 2: A change in R\&D expenditure impacts analysts' sales estimate forecasts for the next year.

| Research and <br> development year <br> over year | $\longrightarrow$ | Analyst EPS estimate <br> revision for next year |
| :--- | :--- | :--- |
|  |  |  |

Figure 4 Hypothesis 2

Hypothesis 3: A change in R\&D expenditure impacts analysts' EPS estimate forecasts for the current year.


Figure 5 Hypothesis 3

Hypothesis 4: A change in R\&D expenditure impacts analysts' EPS estimate

## forecasts for the following year.

| Research and |
| :--- | :--- | :--- |
| development year |
| over year |$\longrightarrow$| Analyst Sales |
| :--- |
| estimate revision for |
| next year |

This study primarily focuses on these four hypotheses, reviewing the correlation and regression results of R\&D year-over-year expenditure differences associated with analysts' sales and EPS estimates for the current year and the next year. The difference between $\mathrm{R} \& \mathrm{D}$ expenditure from the prior to the current year is the independent variable. Analysts provide sales or EPS estimates for a company for multiple years based on the information they gather during company earning calls and/or through reports filed by the company to regulating agencies. The analysts' sales and EPS estimates for the current year and the next year are the dependent variables.

## III. 1 Research and Development Expense (R\&D)

According to the Frasacti Manual, "Research and experimental development (R\&D) comprise creative work undertaken on a systematic basis in order to increase the stock of knowledge, including knowledge of man, culture and society, and the use of this stock of knowledge to devise new applications" (OECD, 1993).

R\&D covers three activities: basic research, applied research, and experimental development. Basic research is experimental or theoretical work undertaken primarily to acquire new knowledge of the underlying foundations of phenomena and observable facts, without any particular application or use in view. Applied research is also an
original investigation, undertaken to acquire new knowledge. It is, however, directed primarily toward a specific practical aim or objective. Experimental development is systematic work, drawing on existing knowledge gained from research and/or practical experience and directed towards producing new materials, products, and devices; installing new processes, systems, and services; or improving substantially those already produced or installed (OECD, 1993).

R\&D spending has grown sharply as a percentage of sales. In 1975, R\&D expenditure stood at 1.70 percent, but it more than doubled by 1995 to 3.75 percent (Chan, Lakonishok \& Sougiannis, 2001). R\&D expenditure is provided by companies in their income statement, which I extracted from the WRDS database for this study.

## III. 2 Current Year EPS Estimate Revision

Current year EPS is defined as the company's total profit in a fiscal year, divided by the number of outstanding shares. Analysts gather this information based on the financial statements, company regulatory filing statements and by interacting with company management. Agarwal et al. (2012) found that earnings forecasts strongly respond to macroeconomic releases that signal changes in overall business conditions after controlling for analysts' learning from firm and industry-specific earnings surprises. They also found that medium-term forecasts respond much more strongly to macroeconomic news than forecasts for the current fiscal year. On average, macroeconomic surprises lead analysts to revise their current year earnings forecasts for cyclical firms by three cents and the following year's forecast by five cents. Such revisions might be made when the company's leadership changes, or when the company faces new competition.

Analysts develop expertise in obtaining and analyzing information from various sources, including the following:

- Earnings and other information from SEC filings, such as proxy statements and periodic financial reports
- Industry and macroeconomic conditions
- Conference calls and other management communications

Using this information, analysts produce earnings forecasts, target price forecasts, and stock recommendations, along with qualitative reports describing a firm's prospects. Ramnath (2002) showed that analysts revise their earnings forecasts in response to the earnings announcement of other firms in the same industry. Based on the above studies, I can confirm that analysts include specific industry and/or market sector factors in their earnings forecasts. EPS is also impacted when there is an increase in advertisement spending, as it will attract more customers. With additional customers, the firm's revenues will rise, which in turn increases the bottom line of the company.


Figure 7 Input to Analyst Report
(Ramnath, Rock, Shane 2003)

## III. 3 Next Year EPS Estimate Revision

Next year EPS is defined as the profit a company generates per share in the following year. Analysts update this estimate at the same frequency as they update their current year EPS; the factors that impact the current year EPS estimates might also impact the next year EPS estimates. However, in some scenarios, the current year EPS might be reduced, while the next year EPS is increased. For example, a company might spend more on advertising in the current year to market a new product, reducing advertising in the year following, which will increase the EPS estimate for that following
year. Also, the next year's EPS estimate might be revised up, as the resources hired to support the new product might not be needed in the following year. As the new product stabilizes, customer support will be reduced, which improves the EPS of the organization.

## III. 4 Current Year Sales Estimate Revision

The current year sales forecast is defined as the company revenue that analysts predict for the current year. Analysts' forecasts and the revisions that follow influence price-relevant trades (Givoly \& Lakonishok, 1979; Lys \& Sohn, 1990; Park \& Stice, 2000). Analysts' estimating activities should cause prices to reflect the market and industry information, resulting in larger return synchronicity (Piotroski \& Roulstone 2003). Clement (1999) and Jacob, Lys, and Neale (1999) found that analyst accuracy improves with industry specialization, while Gilson et al. (2001) illustrated that the analyst coverage composition is impacted after spin-offs and equity carve-outs. The macro economy also plays an important part in analysts' sales estimates. For example, when consumer confidence falls, people reduce their spending, which impacts a company's revenue. Also, when the companies are not hiring and wages stagnate, consumer spending also declines, which also impacts a company's sales. When jobs are impacted, people delay buying computers, software, and other technology products that impact computer manufacturers and software development companies. The same is true for corporations: companies review the outlook of the economy and decide on their spending; if the outlook is bleak, they postpone capital purchase for few years and wait for market conditions to improve. Corporate analysts thus revise their revenue forecasts for technology companies based on job market conditions, consumer confidence, and economic outlook.

Historically, the technology industry has experienced considerable disruption. Blackberry-once the leader in the smart phone industry-was replaced by Apple and Samsung, while Yahoo's leadership in search engine and email technology was usurped by Google. Analysts thus look at the industry, review the competition in various areas, and revise the revenue forecasts for companies accordingly.

## III. 5 Next Year Sales Estimate Revision

Analysts' next year sales estimate forecasts are defined as the revenue a company is expected to make the following year. Analysts update their estimates for the next year at the same frequency as they update their sales estimate forecasts for the current year, as factors that impact the current year's sales might also impact those of the following year. Similar to EPS, the next year's sales estimate might differ from the current year's sales estimate. New product launches or new marketing initiatives might influence next year's revenue, which in turn is captured by analysts when they estimate the next year's sales.

## IV CHAPTER 4: DATA COLLECTION AND ANALYSIS

For this research, I extracted two sets of data from the WRDS website:

- R\&D expenditure and company-related information
- Analyst estimate information

I extracted the R\&D expenditure and company-related information from the Compustat database. Because my research focuses on the IT sector, I specifically extracted the data pertaining to this sector by querying the database with Global Industry Classification (GIC) sector code 45. I gathered the R\&D/company-related data from the yearly database section, which annually consolidates this data for IT companies.

I extracted analyst estimates from the WRDS IBES database for the following categories:

- Sales estimate for the current year (" $1^{\text {st }}$ year" as per WRDS database)
- EPS estimate for the current year
- Sales estimate for the next year (" 2 nd year" as per WRDS database)
- EPS estimate for the next year

Most analysts make changes to their estimates following the company's quarterly earnings release; analysts attend the quarterly earnings release conference and ask questions needed to update their forecast estimate. Their questions typically focus on the existing quarterly results, as well as the company's forecast in terms of sales, earnings, expenditure on advertisement, and R\&D. This information gives them the input needed to model their earnings estimate forecasts for sales and EPS. After gathering this information, analysts generate a detailed report about the target company. In that report, they cover the company's future sales, earnings, and ideal stock price, and offer a
recommendation to buy, sell, or hold the stock. Finally, some analysts change their estimate forecasts in the wake of macro information or changes related to the company. Although analysts publish their estimates about a target company on different days in a month, the IBES database consolidates the information on a monthly basis. Because multiple analysts cover particular stocks-typically those of large or popular companies-in a given month, multiple analysts might change their estimates. I thus use the median information for this study.

The data I collected for $\mathrm{R} \& \mathrm{D}$ expenditures and analyst estimates is for 20 years, from 1995 to 2014. I extracted the data in Excel format from the Compustat and IBES databases. I then cleaned the data in the Excel database before loading it into IBM Statistical Package for Social Sciences (SPSS). I extracted data related to all IT companies, but not all such companies have R\&D expenditures so I filtered out those companies who did not spend on R\&D in the Excel spreadsheet.

Table 1 R\&D and Company-Related Information

| Data extracted | Description of the data. |
| :--- | :--- |
| Stock Symbol | Symbol representing the stock, (Apple, symbol will |
| Fiscal year | Financial year (1995, 1996....) Microsoft MSFT) |
| Company Name | Full name of the company, Apple will be Apple |
|  | INC, IBM will be INTL BUSINESS MACHINES |
| Revenue | CORP |
| Asset | Total revenue for the year. |
|  | Total Asset the company reported during end of |
| Gross Profit | their financial year in their reports. |
|  | Profit company makes after deducting the costs |
| R\&D spending for current year | Research and development spending for the current |
| R\&D spending for prior year | Research and development spending for the prior |
| Long-term debt | year |
|  | Debt obligations such as bank loans, mortgage, |
| bonds which matures more than one year |  |

## Notes: R\&D information and company-related information from Compustat database.

I calculated the following information using the data in Table 1:

- Gross margin: gross profit/revenue
- The gross margin data point (used during the regression)
- Debt-to-sales: long-term debt/revenue
- The debt-to-sales data point (used during the regression)
- Revenue difference: the revenue for the current year minus the revenue for the prior year
- Revenue difference percentage: revenue difference/revenue for prior year Because the revenue for the prior year is not available in the WRDS, I derived it from the previous year's data using Excel advanced programming.

I generated the second set of data from the IBES database.
Table 2 Analyst Data from IBES
Data extracted from IBES database
Symbol
Period end data
Measurement (EPS or Sales)
Forecast period (1-current year, 2 - Next year)
Median estimate

I extracted the IBES data pertaining to analyst estimates into Excel as monthly summary data, then transformed it into analysts' beginning of the year and end of the year estimates. I did this using Excel functions and pivot tables. A company's starting month and ending month are identified at the start of the data construction process; such
information varies among companies as they have different fiscal year closing months (most companies use December as the fiscal year closing month, but some use March, June, or September).

I used advanced Excel functions to calculate analyst data in four areas-current year sales percentage, next year sales percentage, current year EPS percentage, and next year EPS percentage - then merged the data with the records generated from the Compustat database. The R\&D data, company data, and analyst data are normalized into a single record for each company per calendar year. For example, for IBM, there is one record per year from 1995 to 2014; this contains all the data needed for my analysis. The data massaged in Excel is then uploaded and analyzed in SPSS software.

Table 3 Data Calculated and Consolidated in Excel Spread Sheet
Data consolidated in Excel spread sheet
Analyst Sales estimate forecast for current year
Analyst Sales estimate forecast for Next year
Analyst EPS estimate forecast for current year
Analyst EPS estimate forecast for Next year
Analyst Sales estimate forecast percentage change for current year
Analyst Sales estimate forecast percentage change for Next year
Analyst EPS estimate forecast percentage change for current year
Analyst EPS estimate forecast percentage change for Next year

## IV. 1 Distribution of Samples over Time, by Size and by Industry

## IV.1.1 Time

The data in Table 4 provides the year-wise breakup of the number of companies in each calendar year for this study. This research is a 20-year study of IT companies, with data gathered for the years 1995 to 2014. The records identified are for companies that had $\mathrm{R} \& \mathrm{D}$ expenses allocated in their balance sheets that were also covered by analysts. The number of companies in Table 4 for each calendar year increased from 1995 to 1999, then, following dot-com burst, the number declined, as many technology companies went bankrupt. After the dot-com crash, the market value of many companies decreased dramatically, and many were either bought by or merged with other (often larger) companies. In 2000, 526 companies IT companies had R\&D allocations on their income statements and were covered by analysts. In 2013, the number declined to 327a reduction of 199 companies over 13 years due to bankruptcy, mergers, or purchase by other companies. However, the study's final year (2014) showed a slight increase in IT companies.

Table 4 Year Wise Number of Firms

| No of <br> companies <br> in this |  |  |
| :--- | ---: | ---: |
| Year | study | Percentage |
| 1995 | 301 | 3.8 |
| 1996 | 387 | 4.9 |
| 1997 | 413 | 5.2 |
| 1998 | 423 | 5.4 |
| 1999 | 462 | 5.8 |
| 2000 | 526 | 6.7 |
| 2001 | 489 | 6.2 |
| 2002 | 448 | 5.7 |
| 2003 | 407 | 5.2 |
| 2004 | 407 | 5.2 |
| 2005 | 392 | 5.0 |
| 2006 | 378 | 4.8 |
| 2007 | 398 | 5.0 |
| 2008 | 363 | 4.6 |
| 2009 | 346 | 4.4 |
| 2010 | 337 | 4.3 |
| 2011 | 339 | 4.3 |
| 2012 | 332 | 4.2 |
| 2013 | 327 | 4.1 |
| 2014 | 341 | 4.3 |

## IV.1.2 Size

Market capitalization data provides company size: the larger the market capitalization, the larger the company. The firms are divided into three categories according to their market capitalization:

- Large: greater than or equal to $\$ 10$ billion (USD)
- Midsized: greater than $\$ 1$ billion, but less than $\$ 10$ billion
- Small: less than $\$ 1$ billion

I categorize the data to show how analysts capture R\&D expenditure changes in their estimate forecasts. Large companies (greater than $\$ 10$ billion) are covered by more analysts; typically those companies have been in business for a long time and have grown over the years. Small companies (less than $\$ 1$ billion) are typically newer firms that are covered by fewer analysts.

## IV.1.3 Industry

After extracting data for the IT sector using GIC sector code 45, three industry groupings emerged as identified by GIC group codes: Software and Services (which had the most records), Technology Hardware and Equipment, and Semiconductor.

Table 5 Industry Grouping

| GIC group code | Description of the group |
| :---: | :---: |
| 4510 | Software and Services |
| 4520 | Technology Hardware and equipment |
| 4530 | Semiconductor |

## IV.1.3.1 Software and Services group

The Software and Services group encompasses application software, systems software, Internet software and services, data processing and outsourced services, IT consulting, and home entertainment software companies. As Table 6 shows, this group had the most records $(3,074)$.

## IV.1.3.2 Technology Hardware and Equipment

The Technology Hardware and Equipment group consists primarily of communications equipment, computer hardware, computer storage and peripherals, electronic equipment and instruments, electronic components, electronic manufacturing
services, technology distributors, and office electronics companies. This group had 2,961 records (see Table 6).

## IV.1.3.3 Semiconductor

The Semiconductor group is made up of semiconductor and semiconductor equipment firms, and was once part of the Technology Hardware and Equipment group. It had the fewest number of records, with 1,863 (Table 6).

Table 6 Number of Records Industry Wide

|  | No of Records | Percentage |
| :--- | ---: | ---: |
| Type of industry | 3074 | 38.9 |
| Toftware and Services | 2961 | 37.5 |
| Semicondogy Hardware and equipment | 1863 | 23.6 |

Table 7 shows the industry groupings, sub-industry descriptions, and sample companies.

Table 7 Sub Industries and Company Examples.

| Industry | Sub Industries description | Examples of companies in these industry |
| :---: | :---: | :---: |
| Software and Services | Internet Software \& Services | Yahoo, AOL, |
|  | IT consulting and other services | Teradata, IBM |
|  | Data processing and outsourced services | Xerox |
|  | Application Software | Intuit, Adobe systems |
|  | Systems Software | Oracle, Microsoft |
|  | Home Entertainment Software | Take-Two, Zynga |
| Technology Hardware \& equipment | Communications Equipment | Cisco, Qualcom |
|  |  |  |
|  | Computer Hardware | Dell |
|  | Computer Storage and Peripherals | Apple, Sandisk |
|  | Electronic Equipment and | Itron, Zebra |
|  | Instruments | technologies |
|  | Electronic Components | Corning |
|  | Electronic Manufacturing Services | Flextronics |
|  | Technology distributors | Richardson electronics |
|  | Office electronics | General Scanning |
| Semiconductor | Semiconductor Equipment | Lam Research, Teradyne |
|  | Semiconductors | Texas Instrument, First solar |

## IV. 2 Data analysis

A single record for each company per fiscal year was created in Excel and imported into SPSS to generate the results. Separate results were generated for analysts' sales
estimates and analysts' EPS estimates; to show analysts' estimate changes for current year and next year, I ran the tests separately for each of those years.

A summary of the analysis done in SPSS to generate the results is as follows:

- Descriptive statistics were generated for all variables used in this study.
- A correlation matrix was generated for all independent variables and for different market capitalization companies.
- Correlation tests were performed to identify the association between R\&D spending and the changes in analysts' current year sales estimates; these tests were run for the three different industry groups and for the three different company sizes. A correlation test was also run for R\&D spending increases that were greater than five percent (eight tests were performed in this category).
- Correlation tests were performed to identify the association between R\&D spending and changes in analysts' next year sales estimates, as well as the above-mentioned six segment tests and those for $\mathrm{R} \& \mathrm{D}$ spending increases greater than five percent (eight tests were also performed in this category).
- Correlation tests were performed to identify the association between R\&D spending and analysts' current year EPS estimates, along with the same six segment tests and those for R\&D spending increases greater than five percent (again, eight tests were performed in this category).
- The same correlation tests were performed to identify the association between R\&D spending and changes in analysts' next year EPS estimates.
- Regression tests were run on five independent variables-R\&D spending differences, revenue differences, assets log, long-term debt/sales, and gross margin- association with changes in analysts' current year sales estimates. As with correlation, regression tests were run for the three industry groups and three company sizes, as well as for R\&D spending increases greater than five percent. I also ran an additional test without revenue difference regression (performing nine regression tests in this category).
- Regression tests were run on the five independent variables association with changes in analysts' next year sales estimates, the six segmentation tests, R\&D greater than 5 percent, and regression without revenue differences (nine tests).
- The same regression tests were performed with the above-mentioned independent variables association with changes in analysts' current year EPS estimates.
- The same regression tests were performed with the above-mentioned independent variables association with changes in analysts' next year EPS estimates.

Overall, 32 correlation and 40 regression tests were conducted to analyze the four different hypotheses in this study. The outcomes of these tests will help to determine whether the results support the four hypotheses.

## V CHAPTER 5: RESULTS

## V. 1 Descriptive Statistics

Table 8 shows summary statistics of data gathered for 7,900 samples related to IT firms with group ID 45. The variable assets, gross profit (loss), long-term debt, revenue, gross margin, research and development current year, research and development prior year, and market value were extracted from the Compustat database. Revenue prior year, revenue difference, revenue difference percentage, $R \& D$ difference, and $R \& D$ difference percentage were calculated.

I extracted analysts' current and next year EPS estimates from the end and the beginning of each year from the IBES database. I extracted the same data for analysts' sales estimates from IBES. The actual difference and the difference in percentage variables for EPS and sales were calculated in Excel before uploading the values into SPSS.

## Table 8 Descriptive Statistics

|  | Minimum | Maximum | Mean | Std. Dev |
| :---: | :---: | :---: | :---: | :---: |
| Assets - Total | 0.03 | 290479.00 | 1990.70 | 10510.24 |
| Gross Profit (Loss) | -719.76 | 104126.00 | 756.84 | 4213.38 |
| Long-term debt | 0.00 | 39959.00 | 232.63 | 1572.32 |
| Revenue |  |  |  |  |
| Revenue current year | 0.00 | 233715.00 | 1430.16 | 7936.83 |
| Revenue Prior year | 0.00 | 182795.00 | 1518.93 | 7873.99 |
| Revenue difference | -12519.00 | 50920.00 | 114.47 | 1293.41 |
| Revenue difference percentage | -1.00 | 65.87 | 0.20 | 1.14 |
| Gross Margin | -141.41 | 1.00 | 0.38 | 3.49 |
| Market Value - Total - Fiscal | 0.60 | 626550.35 | 5072.62 | 28018.82 |
| Research \& Development |  |  |  |  |
| Research \& Development Expense | 0.00 | 12128.00 | 146.98 | 675.72 |
| Research \& Development Prior year | 0.00 | 11537.00 | 134.53 | 629.37 |
| Research spending difference | -1595.00 | 2209.00 | 12.45 | 94.22 |
| Research difference percentage | -1.00 | 345.84 | 0.36 | 4.31 |
| Analyst estimate current year EPS end of the year | -645.00 | 53.20 | 0.00 | 9.33 |
| Analyst estimate current year EPS beginning of the year | -510.00 | 57.00 | 0.32 | 7.04 |
| Analyst estimate current year EPS difference | -318.20 | 27.75 | -0.32 | 5.00 |
| Analyst estimate current year EPS difference \% | -116.25 | 205.00 | -0.18 | 4.75 |

Analyst estimate next year EPS end of the year

Analyst estimate next year EPS beginning of the year

Analyst estimate next year EPS difference

$$
-141.43
$$

Analyst estimate next year EPS difference

$$
-276.57
$$

Analyst estimate next year EPS difference

$$
\%
$$

Analyst estimate current year sales end of the year

Analyst estimate current year sales year beginning of the year

Analyst estimate current year sales

## difference

Analyst estimate current year sales
difference \%

Analyst estimate next year sales current year end of the year

Analyst estimate next year sales current year
beginning of the year
$\begin{array}{llllll}\text { Analyst estimate next year sales difference } & -85452.37 & 33495.50 & -68.99 & 1726.58\end{array}$
Analyst estimate next year sales difference

| $\%$ | -1.00 | 5.87 | -0.03 | 0.29 |
| :--- | :---: | :---: | :---: | :---: |
| logofassets | -3.38 | 12.58 | 5.53 | 1.74 |
| Long-term debt/sales | 0.00 | 2.03 | 0.07 | 0.13 |

## V. 2 Correlation Matrix of Independent Variables

I generated the correlation matrix for all the variables used in the regressions.
The matrix shows which independent variables are strongly correlated and which are not correlated. As Table 9 shows, the research difference percentage is strongly correlated with the revenue difference percentage. The correlation between research spending difference and revenue difference is .205 , with .000 significance.

Table 9 Correlation Matrix for All Independent Variables

| Correlations matrix |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Research difference percentage | Longterm debt/sales | Revenue difference percentage | Log of Assets | Gross <br> Margin |
| Research spending | Pearson Correlation | X | -. 011 | . 205 | -. 018 | -. 005 |
| difference <br> in | Sig. (2tailed) |  | . 317 | . 000 | . 117 | . 651 |
| percentage <br> (year over year) | N | 7898 | 7767 | 6650 | 7898 | 7881 |
| Long-term debt/sales | Pearson Correlation | -. 011 | X | -. 019 | . 221 | -. 004 |
|  | Sig. (2tailed) | . 317 |  | . 122 | . 000 | . 746 |
|  | N | 7767 | 7767 | 6562 | 7767 | 7753 |
| Revenue difference percentage (year over year) | Pearson <br> Correlation | . 205 | -. 019 | X | -. 012 | -. 026 |
|  | Sig. (2tailed) | . 000 | . 122 |  | . 336 | . 033 |
|  | N | 6650 | 6562 | 6650 | 6650 | 6647 |
| Log of Assets | Pearson Correlation | -. 018 | . 221 | -. 012 | X | . 073 |
|  | Sig. (2tailed) | . 117 | . 000 | . 336 |  | . 000 |
|  | N | 7898 | 7767 | 6650 | 7898 | 7881 |
| Gross <br> Margin | Pearson | -. 005 | -. 004 | -. 026 | . 073 | X |
|  | Correlation | -.005 | -. 004 | -. 026 | . 073 | X |
|  | Sig. (2tailed) | . 651 | . 746 | . 033 | . 000 |  |
|  | N | 7881 | 7753 | 6647 | 7881 | 7881 |

Notes : Correlation Matrix between all independent variables used in regression.

In Table 10, the correlation matrix shows the results generated for the companies with market capitalization greater than or equal to $\$ 10$ billion. As with Table 9's results, the research difference percentage is strongly correlated with the revenue difference percentage. The correlation between research spending difference and revenue difference is .586 , with .000 significance. This is much higher than the correlations for the sample at large, indicating that large companies exhibit a very strong relationship between R\&D and sales changes. All other independent variables are negatively correlated to the research difference percentage.

Table 10 Correlation Matrix for Large Size Firms

| Correlations Matrix for market value >=10 billion |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Research difference percentage | $\begin{gathered} \text { Long- } \\ \text { term } \\ \text { debt/sales } \end{gathered}$ | Revenue difference percentage | Log of Assets | Gross <br> Margin |
| Research spending | Pearson Correlation | X | -. 122 | . 586 | -. 321 | -. 095 |
| difference <br> in | Sig. (2tailed) |  | . 013 | . 000 | . 000 | . 051 |
| percentage (year over year) | N | 424 | 412 | 413 | 424 | 424 |
| Long-term debt/sales | Pearson <br> Correlation | -. 122 | X | . 030 | . 176 | -. 076 |
|  | Sig. (2tailed) | . 013 |  | . 555 | . 000 | . 121 |
|  | N | 412 | 412 | 401 | 412 | 412 |
| Revenue difference | Pearson <br> Correlation | . 586 | . 030 | X | -. 366 | . 116 |
| percentage (year over | Sig. (2tailed) | . 000 | . 555 |  | . 000 | . 018 |
| year) | N | 413 | 401 | 413 | 413 | 413 |
| Log of | Pearson | -. 321 | . 176 | -. 366 | X | -. 085 |
|  | Sig. (2tailed) | . 000 | . 000 | . 000 |  | . 079 |
|  | N | 424 | 412 | 413 | 424 | 424 |
| Gross <br> Margin | Pearson <br> Correlation | -. 095 | -. 076 | . 116 | -. 085 | X |

Sig. (2-
tailed)
. 051
$\mathrm{N} \quad 424$
$\mathrm{N} \quad 424$
. 121
.018 . 079
412
413
424
424
In Table 11, the correlation matrix illustrates the results for companies with less than $\$ 10$ billion and more than $\$ 1$ billion in market capitalization. Similar to large companies, the research difference percentage is strongly correlated with the revenue difference percentage. The correlation between research spending difference and revenue spending is .627 .

Table 11 Correlation Matrix for Medium Size Firms

| Correlations Matrix for Market value > 1 billion and < 10 billion |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Research difference percentage | $\begin{gathered} \text { Long- } \\ \text { term } \\ \text { debt/sales } \\ \hline \end{gathered}$ | Revenue difference Percentage | Log of Assets | Gross <br> Margin |
| Research spending difference in percentage (year over year) | Pearson <br> Correlation | X | -. 092 | . 627 | -. 201 | -. 069 |
|  | Sig. (2tailed) |  | . 000 | . 000 | . 000 | . 004 |
|  | N | 1716 | 1693 | 1580 | 1716 | 1716 |
| Long-term debt/sales | Pearson Correlation | -. 092 | X | -. 095 | . 298 | -. 017 |
|  | Sig. (2tailed) | . 000 |  | . 000 | . 000 | . 487 |
|  | N | 1693 | 1693 | 1558 | 1693 | 1693 |
| Revenue difference percentage (year over year) | Pearson Correlation | . 627 | -. 095 | X | -. 269 | -. 107 |
|  | Sig. (2tailed) | . 000 | . 000 |  | . 000 | . 000 |
|  | N | 1580 | 1558 | 1580 | 1580 | 1580 |
| Log of Assets | Pearson | -. 201 | . 298 | -. 269 | X | -. 023 |
|  | Correlation |  |  | -. 26 |  | -. 023 |
|  | Sig. (2tailed) | . 000 | . 000 | . 000 |  | . 336 |
|  | N | 1716 | 1693 | 1580 | 1716 | 1716 |
| Gross <br> Margin | Pearson | -. 069 | -. 017 | -. 107 | -. 023 | X |
|  | Correlation | -. 069 | -. 017 | -. 107 | -. 023 | X |
|  | Sig. (2tailed) | . 004 | . 487 | . 000 | . 336 |  |
|  | N | 1716 | 1693 | 1580 | 1716 | 1716 |

In Table 12, the correlation matrix shows results for companies with market capitalization that is less than or equal to $\$ 1$ billion. Although these results show that the research difference percentage is correlated with revenue difference percentage, the correlation is not as strong as it is for companies with more than $\$ 1$ billion in market capitalization. The correlation between research spending difference and revenue spending is .167 although this is much lower than in large companies. All other independent variables are not significant at normal levels.

Table 12 Correlation Matrix for Small Size Firms

|  | Correlations Matrix for Market value <= 1 billion |
| :--- | :--- | ---: | ---: | ---: | ---: | ---: |

I analyzed sales estimate forecasts and research spending, then followed the same procedure for the EPS estimates. Within the sales tests, I analyzed both the current year and the following year. The tests were initially focused on all IT companies; I then drilled down into market size and industry segments.

## V. 3 Hypothesis 1: A change in R\&D expenditure impacts analysts' sales

 estimate forecasts for the current year.To test this hypothesis, a correlation test was performed between the R\&D spending difference percentage and the analysts' sales estimates difference percentage for the current year (see Table 13).

Table 13 Correlation R\&D and Analyst Sales Estimate

| Years | Sector description | Pearson | Sig | N |
| :--- | :--- | :--- | :--- | :--- |
|  |  | Correlation |  |  |
| Current year estimate | All IT companies | .153 | .000 | 6830 |
| Next year estimate | All IT companies | .183 | .000 | 6690 |

Notes: Correlation of Analyst Sales estimate percentage difference (dependent
variable) with R \& D expenditure percentage difference (independent variable)

As Table 13 shows, there is a positive correlation between $R \& D$ expenditure and analysts' sales estimates for the current year. The Pearson correlation is .153 , with .000 significance (Table 13); the number of records used for the current year analysis was 6,830. Based on the correlation results, Hypothesis 1 is supported.

## V. 4 Hypothesis 2: A change in R\&D expenditure impacts analysts' sales estimate forecasts for the next year.

To test this hypothesis, I performed correlation analysis for the next year (Table 13). The results show a correlation between $R \& D$ expenditure and analysts' sales estimates. The Pearson correlation is .183 , with .000 significance; the number of records used for the next year analysis was 6,690 . Based on the correlation results, Hypothesis 2 is also supported.

Although the correlation tests support hypotheses 1 and 2, I conducted regression tests to confirm whether changes in R\&D have a major influence on changes to analysts’ sales estimates. I executed the regression tests using the five independent variablesR\&D expenditure difference percentage, gross margin, log of assets, revenue difference, and long-term debt/sales-with the dependent variable being analysts' sales estimate percentage difference.

Table 14 Regression of R\&D, Revenue, ... with Analyst Sales Estimate

| Years | $\mathrm{R}^{2}$ | Sig | F | Research | Gross | Log | Revenue | Long- |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets | difference | term |
|  |  |  |  |  | /Sig |  | /Sig |  |
| debt/Sales |  |  |  |  |  |  |  |  |
| Current | .075 | .000 | 98.09 | $.221 / .000$ | $.006 / .607$ | $.095 / .000$ | $.092 / .000$ | $-.015 / .242$ |
| year |  |  |  |  |  |  |  |  |
| estimate |  |  |  |  |  |  |  |  |
| Next year | .069 | .000 | 87.99 | $.159 / .000$ | $.005 / .686$ | $.123 / .000$ | $.142 / .000$ | $-.007 / .575$ |
| estimate |  |  |  |  |  |  |  |  |

## Notes : Regression of Independent variables (R\&D expenditure percentage difference, Ratio of gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst sales estimate percentage difference (dependent variable).

For the current year, the $\mathrm{R}^{2}$ value is .07 ; reviewing the standards coefficient beta values shows that research spending difference has the highest influence in the regression with .221 and .000 significance (Table 14). The revenue difference percentage standard coefficient beta value is .092 , with .000 significance, and the asset $\log$ is .095 , with .000 significance, which illustrates that these independent variables also influence the regression. Other independent variables did not prove significant in this regression. For the next year, the regression results (Table 14) are quite similar to the current year, with $R^{2}$ at .069 . In the next year results, research spending is the highest contributing independent variable: .159 , with .000 significance. Similar to the current year, in the next year, both revenue difference and assets have a positive coefficient beta value, with .000 significance.

I further analyzed the analysts' sales estimate percentage difference and R\&D expenditure percentage difference based on company size. For all three categories, the current year and next year correlations are positive and the significance is .000 (see Table 15). For companies with $\$ 10$ billion or more in market value for the current year, the correlation is quite strong: .569 , with .000 significance and a sample size of 421 . The sample size increases for companies with less than $\$ 10$ billion in market capitalization; it increases even more for companies with less than $\$ 1$ billion in market capitalization. For current year analyst sales estimates for companies with market capitalization of more than $\$ 1$ billion and less than $\$ 10$ billion, the correlation is .350 with a significance of .000
and a sample size of 1,685 . For companies with less than $\$ 1$ billion in market capital, the current year correlation is not as strong as in other results, but still has a positive correlation of .128 , with significance of .000 and a sample size of 4,502 .

The correlation for analysts' next year sales estimate differences and the R\&D spending differences follows a pattern identical to the current year results. The correlation is stronger for larger market capitalization companies. The next year results reveal that, for companies with $\$ 10$ billion in market value, the correlation is strong: .479, with .000 significance. The sample size for this analysis is 421 (the sample size increases as the market value decreases, similar to the current year pattern). For companies with market capitalization greater than $\$ 1$ billion and less than $\$ 10$ billion for the next year, the correlation is .361 , with a .000 significance and a sample size of 1,674 . For firms with a market capitalization of less than $\$ 1$ billion, the correlation between the analysts' next year sales estimates and the R\&D spending difference is not as strong as in other segments: the correlation value is .128 , with .000 significance and a sample size of 3,969 .

Table 15 Correlation of R\&D with Analyst Sales - Market Cap Segmentation

| Years | Market value | Pearson | Sig | N |
| :--- | :--- | :--- | :--- | :--- |
|  |  | Correlation |  |  |
| Current year estimate | $>=10$ billion | .569 | .000 | 421 |
| Current year estimate | $>1$ billion \& | .350 | .000 | 1685 |
|  | $<10$ billion |  | .000 | 4502 |
| Current year Estimate | $=<1$ billion | .128 | .000 | 421 |
| Next year Estimate | $>=10$ billion | .479 | .000 | 1674 |
| Next year Estimate | $>1$ billion \& | .361 |  |  |
| Next year Estimate | $=<1$ billion | .102 | .000 | 3969 |

Notes: Market cap wise correlation of Analyst Sales estimate percentage difference (dependent variable) with $\mathbf{R} \& \mathbf{D}$ expenditure percentage difference (Independent variable).

Positive correlation exists for all results in Table 15; to confirm that R\&D expenditures are the major influencer for analysts to update their sales estimate, regression tests were performed. The regression results in Table 16 demonstrate that $\mathrm{R}^{2}$ is 65 percent, but that research spending is not significant in this regression. Other values also are not significant in this regression, apart from the revenue difference percentage, which had a strong coefficient value of .839 . The same pattern is observed in the next year estimate results: $R^{2}$ is 37 percent, the research spending standard coefficient beta is .009 , with .851 significance. The results show that, in the current year, the revenue percentage difference's standard coefficient beta is .636 , with .000 significance.

Table 16 Regression of R\&D, Revenue, ... with Analyst Sales Estimate (Large Size companies)

| Years | R2 | Sig | F | Research | Gross | Log of | Revenue | Long- |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets | difference | term |
|  |  |  |  | /Sig |  | /Sig |  | debt/Sales |
| Current | .657 | .000 | 150.2 | - | $-.024 / .434$ | $.047 / .151$ | $.839 / .000$ | $.020 / .513$ |
| year |  |  |  | $.020 / .583$ |  |  |  |  |
| estimate |  |  |  |  |  |  |  |  |
| Nextyear | .379 | .000 | 47.95 | - | $-.030 / .468$ | $.047 / .285$ | $.636 / .000$ | $.034 / .400$ |
| estimate |  |  |  | $.009 / .851$ |  |  |  |  |

## Notes : Regression of independent variables (R\&D expenditure percentage difference, Ratio of gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst Sales estimate difference percentage for companies with market capitalization >= $\mathbf{1 0}$ billion.

As Table 17 shows, for companies with market capitalization greater than \$1 billion and less than $\$ 10$ billion, the results are similar to companies with market capitalization greater than $\$ 10$ billion. In Table $17, R^{2}$ is 49 percent with an $F$ value of 299, but research spending is negatively correlated, with significance higher than .005 . The other variables in this regression are not significant, apart from revenue difference percentage, which is significant with the coefficient value of .722 .

The same pattern occurs in the next year estimate results (Table 17), where $\mathrm{R}^{2}$ is 65 percent; the research spending standard coefficient beta is not significant, but the revenue percentage difference standard coefficient beta is .636 , with .000 significance.

Table 17 Regression of R\&D, Revenue, ... with Analyst Sales Estimate (Medium Size companies)

| Years | R2 | Sig | F | Research | Gross | Log of | Revenue | Long- |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets | difference | term |
|  |  |  |  | /Sig |  | /Sig |  | debt/Sales |
| Current | .493 | 000 | 299.5 | - | $-.055 / .004$ | - | $.722 / .000$ | $.027 / .181$ |
| year |  |  |  | $.037 / 120$ |  | $.009 / .641$ |  |  |
| estimate |  |  |  |  |  |  |  |  |
| Next year | .649 | .000 | 222 | $.016 / .524$ | $-.104 / .000$ | $.006 / .772$ | $.622 / .000$ | $.012 / .568$ |
| estimate |  |  |  |  |  |  |  |  |

## Notes : Regression of independent variables (R\&D expenditure percentage difference, Ratio of gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst Sales estimate difference percentage for companies with market capitalization < 10 billion and market capitalization > 1 billion.

The pattern changes for companies with market capitalization of $\$ 1$ billion or less. In this segment, the research spending estimate difference has the highest standard coefficient beta value of the independent values at .217 , with a .000 significance. The value for next year research spending estimate difference standard coefficient beta is .103. In both current year and next year, the $\mathrm{R}^{2}$ is .05 and .02 , respectively, and the F values are 42.20 and 16.8 , respectively, with .000 significance. Gross margin and longterm debt sales values (Table 18) are not significant, and revenue difference percentage is significant only in the next year results section. Assets values are significant at .059 for the current year and .068 for the next year.

Table 18 Regression of R\&D, Revenue, ... with Analyst Sales Estimate (Small size companies)

| Years | R2 | Sig | F | Research | Gross | Log of | Revenue | Long- |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  |  | spending | Margin/Sig | assets | difference | term |
|  |  |  |  |  | Sig |  | /Sig |  | debt/Sales |
| Current | .055 | .000 | 42.20 | $.217 / .000$ | $.003 / .863$ | $.059 / .000$ | $.029 / .079$ | $-0.14 / .381$ |  |
| year |  |  |  |  |  |  |  |  |  |
| estimate |  |  |  |  |  |  |  |  |  |
| Next year | .024 | .000 | 16.8 | $.103 / .000$ | $.001 / .955$ | $.068 / .000$ | $.074 / .000$ | $-.011 / .534$ |  |
| estimate |  |  |  |  |  |  |  |  |  |

## Notes : Regression of independent variables (R\&D expenditure percentage difference, Ratio of gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst Sales estimate difference percentage for companies with market capitalization < 1 billion.

Table 19 shows the analysts' sales estimates percentage difference compared to the R\&D expenditure difference based on the IBES database's three GIC industry groupings: Software and Services, Technology Hardware and Equipment, and Semiconductor. As in all three company size categories, for current year and next year the correlation is positive for all three industry groups, with a .000 significance. For Software and Services firms, the correlation is strong and significant with .210 and the sample size of 2,719.

The sample size is the largest for Software and Service group. Correlation results for Technology and Hardware companies for the current year analyst sales estimate difference compared to $\mathrm{R} \& \mathrm{D}$ expenditure difference is .088 , with .000 significance and a
sample size of 2,458 . The sample size for the current year Semiconductor category is 1,683 - the smallest of the three sectors-yet the correlation is still significant at .136.

The next year's results for the industry groups demonstrate the association of analysts' sales estimate percentage difference to R\&D expenditure difference: the correlation for Software and Services is .205 , with .000 significance and a sample size of 2,657. For Technology Hardware and Equipment, the correlation with the same constructs is .211 , with .000 significance and a sample size of 2,369 . As with the current year, the sample size is lowest for the semiconductor group at 1,664 , with a .120 correlation.

Table 19 Correlation of R\&D with Analyst Sales - Industry Group Segmentation
$\left.\begin{array}{lllll}\hline \text { Years } & \text { Market value } & \text { Pearson } & \text { Sig } & \text { N } \\ \text { Correlation }\end{array}\right)$

Notes: Industry wise correlation of Analyst Sales estimate percentage difference (dependent variable) with $\mathbf{R} \& \mathbf{D}$ expenditure percentage difference (independent variable).

Table 20's regression results for Software and Services shows that $\mathrm{R}^{2}$ is .08 , but the research spending standard coefficient beta is positive at .235 , with significance below . 01 . For the revenue difference percentage and assets log, the coefficient values are both significant at .106 and .057 . The next year results are identical to the current year
results, where $R^{2}$ is .11 with an $F$ value of 60.11 and .000 significance. The research spending standard coefficient beta is positive at .168 , while the revenue difference and assets $\log$ are at .229 and .130 , respectively, and all three values have a significance of less than .01. Other values are not significant in this regression.

Table 20 Regression of R\&D...... with Analyst Sales Estimate (Software \& Services)

| Years | R2 | Sig | F | Research | Gross | Log of | Revenue | Long- |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets | difference | term |
|  |  |  |  |  | /Sig |  | /Sig |  |
| Current | .083 | .000 | 41.92 | $.235 / .00$ | $.004 / .860$ | $.057 / .005$ | $.106 / .000$ | $.001 / .960$ |
| year |  |  |  |  |  |  |  |  |
| estimate |  |  |  |  |  |  |  |  |
| Next year | .118 | .000 | 60.11 | $.168 / .00$ | $.008 / .697$ | $.130 / .000$ | $.229 / .000$ | $.007 / .734$ |
| estimate |  |  |  |  |  |  |  |  |

Notes: Regression of independent variables (R\&D percentage difference, Ratio of gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst sales estimate percentage difference (dependent variable) for Software and Service industry group.

In the Hardware and Equipment group, the regression results illustrate that the standard coefficient beta for the current year is .195 ; for the next year, the value is the same, with a .000 significance for both years. The analysis for current year results showed no significance value for revenue difference, but the gross margin and asset log values are significant, with coefficient values of .061 and .025 , respectively. Results for
the next year reveal that, apart from research spending and long-term debt/sales, all other values are not significant.

Table 21 Regression of R\&D...... with Analyst Sales Estimate (Technology Hardware)

| Years | R2 | Sig | F | Research | Gross | Log of | Revenue | Long- |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets | difference | term |
|  |  |  |  | /Sig |  | /Sig |  | debt/Sales |
| Current | .073 | .000 | 34.48 | $.195 / .000$ | $.061 / .004$ | $.159 / .000$ | $.025 / .227$ | $-.001 / .962$ |
| year |  |  |  |  |  |  |  |  |
| estimate |  |  |  |  |  |  |  |  |
| Next year | .60 | .000 | 27.09 | $.195 / .000$ | $.045 / .035$ | $.113 / .000$ | $.044 / .041$ | $.027 / .218$ |
| estimate |  |  |  |  |  |  |  |  |

Notes: Regression of independent variables (R\&D percentage difference, Ratio of gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst sales estimate percentage difference (dependent variable) for Technology Hardware industry group.

Finally, in the Semiconductor Equipment group (Table 22), the results show that research spending values are not significant. In both years, the revenue difference has a significant standard coefficient value of .494 and .390 . The $\mathrm{R}^{2}$ is 51 percent for the current year and 16 percent for next year.

Table 22 Regression of R\&D......with Analyst Sales Estimate (Semiconductor)

| Years | R2 | Sig | F | Research | Gross | Log of | Revenue | Long- |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets | difference | term |
|  |  |  |  |  | /Sig |  | /Sig |  |
| Current | .516 | .000 | 112.8 | - | $.08 / .000$ | $.117 / .000$ | $.494 / .000$ | $-.038 / .092$ |
| year |  |  |  | $0.43 / .064$ |  |  |  |  |
| estimate |  |  |  |  |  |  |  |  |
| Next year | .166 | .000 | 61.16 | - | $.007 / .769$ | $.119 / .000$ | $.390 / .000$ | $.030 / .217$ |
| estimate |  |  |  | $0.19 / .436$ |  |  |  |  |

## Notes: Regression of independent variables (R\&D percentage difference, Ratio of gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst sales estimate percentage difference (dependent variable) for semiconductor industry group.

The correlation and regression results for the analysts' sales estimate percentage difference compared to the $\mathrm{R} \& \mathrm{D}$ percentage difference primarily shows a positive correlation; a few results are negative in regression but not significant. The next few tables show results for analysts' EPS estimate revisions for the current year and the next year; these results were gathered using correlation and regression tests.

For EPS, I applied a filter on the analysts' EPS estimate percentages that were greater than -100 percent or less than 100 percent to avoid data outliers. Also, a few records had more than 9,000 percent positive or negative differences. These high percentage differences were due to the EPS estimate change from a negative value at the beginning of the year to a positive value at the end of the year, or vice versa. Filtering
greater than -100 percent and less than 100 percent removed these extreme values and reduced the data sample size by 5 percent.

## V. 5 Hypothesis 3: A change in R\&D expenditure impacts analysts' EPS estimate

 forecasts for the current year.The correlation results (Table 23) for all IT companies between analysts' EPS estimate differences and R\&D differences do not show a significant value for the current or next year; the current year sample is 6,387 and its correlation is positive at .008 , but it is not significant (the significance value is .536).

Table 23 Correlation R\&D with Analyst EPS Estimate

| Years | Sector description | Pearson | Sig | N |
| :--- | :--- | :--- | :--- | :--- |
|  |  | Correlation |  |  |
| Current year estimate | All IT | .008 | .536 | 6387 |
| Next year estimate | All IT | .021 | .084 | 6515 |

Notes: Correlation of Analyst EPS estimate percentage difference (dependent variable) with $\mathbf{R} \& \mathbf{D}$ percentage difference (independent variable) filtered on Analyst EPS estimate percentage > -100 and Analyst EPS estimate percentage $<\mathbf{1 0 0 \%}$.

## V. 6 Hypothesis 4: A change in R\&D expenditure impacts analysts' EPS estimate forecasts for the following year.

The correlation results (Table 23) are positive for the next year estimate, with a coefficient value at .021 ; the significance is slightly above .05 , with a sample size of 6,515. Although results for the next year are better than the current year, the hypothesis cannot be confirmed as the significance is greater than .05 .

Similar to the regression test for analyst sales estimate forecasts to confirm the correlation results, regression tests were performed for analyst EPS estimate differences as the dependent variable and $\mathrm{R} \& D$ expenditure difference, gross margin, assets log, revenue difference, and long-term debt/sales as the independent variables.

The regression results (Table 24) for all IT companies using R\&D difference, gross margin, asset logs, revenue difference, and long-term debt/sales associated with analysts' EPS estimate differences show that research spending standard coefficient beta is negative, but the value is not significant. The revenue difference has the highest positive standard coefficient beta value, with .000 significance.

Table 24 Regression of R\&D, Revenue, ... with Analyst EPS Estimate

| Years | R2 | Sig | F | Research | Gross | Log of | Revenue | Long- |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets | difference | term |
|  |  |  |  |  | /Sig |  | /Sig |  |
| Current | .023 | .000 | 24.15 | - | $-.011 / .000$ | $.093 / .000$ | $.124 / .000$ | debt/Sales |
| year |  |  |  | $.023 / .113$ |  |  |  |  |
| estimate |  |  |  |  |  |  |  |  |
| Nextyear | .072 | .000 | 83.86 | - | $-.027 / .037$ | $.131 / .000$ | $.246 / .000$ | $-.021 / .116$ |
| estimate |  |  |  | $.035 / .013$ |  |  |  |  |

[^0]We removed revenue difference from the regression testing to understand research spending's influence on the regression in the absence of the revenue difference variable. As Table 25 shows, the standard coefficient beta value was positive but not significant. The next year estimate is .023 , with .06 significance, and the $\mathrm{R}^{2}$ values are quite small.

Table 25 Regression of R\&D, ... with Analyst EPS Estimate (Without Revenue)

| Years | R2 | Sig | F | Research | Gross | Log of | Long-term |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets /Sig | debt/Sales |
|  |  |  |  | /Sig |  |  |  |
| Current year | .005 | .000 | 7.220 | $.009 / .493$ | $-.017 / .191$ | $-.014 / .263$ | $-.014 / .263$ |
| estimate |  |  |  |  |  |  |  |
| Next year | .010 | .000 | 15.89 | $.023 / .060$ | $-.016 / .197$ | $.099 / .000$ | $-.028 / .029$ |
| estimate |  |  |  |  |  |  |  |

Notes : Regression of independent variables (R\&D percentage difference, gross margin, Log(Assets), Long-term debt/sales) with Analyst EPS estimate percentage difference(dependent variable) filtered on Analyst EPS estimate \% > $100 \%$ and Analyst EPS estimate $<\mathbf{1 0 0 \%}$

I conducted market segmentation analysis to understand how analysts' EPS estimate difference is associated with R\&D difference for various market sizes. As noted earlier, the data was filtered for analysts' EPS estimate percentage difference greater than -100 percent and less than 100 percent. The current year results illustrate (Table 26) that the correlation is positive and significant for large and midsized companies, with .134 and .050 , respectively; for companies with less than $\$ 1$ billion in market capitalization, the correlation is not significant.

However, reviewing results for the next year shows a standard coefficient beta value of .223 for large, .133 for medium, and .055 for small companies, and a significance value for all three segments of less than .01 with the next year analyst EPS estimate as an independent variable. The current year sample sizes for large, medium, and small companies are $404,1,533$, and 3,370 , respectively, with next year sample sizes of $405,1,558$, and 3,425 , respectively.

Table 26 Correlation of R\&D with Analyst EPS - Market Cap Segmentation

| Years | Market value | Pearson | Sig | N |
| :--- | :--- | :--- | :--- | :--- |
|  |  | Correlation |  |  |
| Current year Estimate | $>=10$ billion | .134 | .007 | 404 |
| Current year Estimate | $>1$ billion \& <10billion | .050 | .051 | 1533 |
| Current year Estimate | $=<1$ billion | .013 | .457 | 3370 |
| Next year Estimate | $>=10$ billion | .223 | .000 | 405 |
| Next year Estimate | $>1$ billion \& <10billion | .133 | .000 | 1558 |
| Next year Estimate | $=<1$ billion | .055 | .001 | 3425 |

Notes: Market capitalization wise Correlation of Analyst EPS estimate percentage difference(dependent variable) with $\mathbf{R} \& \mathbf{D}$ percentage difference (independent variable) filtered on Analyst EPS estimate percentage difference > $100 \%$ and $<100 \%$.

Regression results for large companies (Table 27) show $\mathrm{R}^{2}$ at 11 percent for current year and 9 percent for next year. The standard coefficient beta value for research spending is negative and not significant; the revenue difference standard coefficient beta is .386 and .384 for the current and next year, respectively, with less than .01 significance.

Table 27 Regression of R\&D, Revenue, ... with Analyst EPS Estimate (Large Size companies)

| Years | $\mathrm{R}^{2}$ | Sig | F | Research | Gross | Log of | Revenue | Long- |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets | difference | term |
|  |  |  |  | /Sig |  | /Sig |  | debt/Sales |
| Current | .119 | .000 | 10.24 | - | $-.048 / .328$ | $.008 / .886$ | $.386 / .000$ | $-.017 / .726$ |
| year |  |  |  | $.076 / .222$ |  |  |  |  |
| estimate |  |  |  |  |  |  |  |  |
| Next year | .095 | .000 | 7.975 | - | $-.049 / .331$ | - | $.304 / .000$ | $-.013 / .815$ |
| estimate |  |  |  | $.004 / .953$ |  | $.028 / .573$ |  |  |

Notes: Regression of independent variables (R\&D percentage difference, gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst EPS estimate percentage difference (dependent variables) filtered on Analyst EPS estimate $\%$ > $\mathbf{- 1 0 0 \%}$ and Analyst EPS estimate < $\mathbf{1 0 0 \%}$ for companies with market capitalization > 10 billion.

The regression test (Table 28) was performed using four constructs: research spending difference, gross margin, assets log, and long-term debt/sales; revenue difference was not included as it had a strong influence in the previous regression test. The results from the regression without the revenue difference reveal that $\mathrm{R}^{2}$ for the current and next year is 2 percent and 5 percent, respectively. The research difference standard coefficient beta for companies with market capitalization of more than $\$ 10$ billion has positive values of .114 and .206 for the current and next year, respectively. The significance for both years is less than .05 mean, while other independent variables in this regression have significance greater than .05 .

Table 28 Regression of R\&D, ... with Analyst EPS Estimate, Large Size companies (No Revenue)

| Years | R2 | Sig | F | Research | Gross | Log of | Long-term |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets /Sig | debt/Sales |
|  |  |  |  | /Sig |  |  |  |
| Current year | .022 | .075 | 2.142 | $.114 / .036$ | $.035 / .491$ | $-.068 / .209$ | $.001 / .986$ |
| estimate |  |  |  |  |  |  |  |
| Next year | .053 | .000 | 5.470 | $.206 / .000$ | $.029 / .563$ | $-.049 / .353$ | $-.029 / .563$ |
| estimate |  |  |  |  |  |  |  |


#### Abstract

Notes: Regression of independent variables (R\&D percentage difference, gross margin, Log(Assets), Long-term debt/sales) with Analyst EPS estimate percentage difference (dependent variables) filtered on Analyst EPS estimate \% > $\mathbf{1 0 0 \%}$ and Analyst EPS estimate $<\mathbf{1 0 0 \%}$ for companies with market capitalization $>=10$ billion.


Midsized companies regression results (Table 29) follow the same pattern as large companies: standard coefficient beta value for research spending is negative and not significant, whereas the revenue difference standard coefficient beta is at .386 and .384 for the current and next year, respectively, with significance less than .01 . The $\mathrm{R}^{2}$ for the current year and next year are .02 and .05 respectively. The gross margin is positive, with a significance value of less than .05 .

Table 29 Regression of R\&D, Revenue, ... with Analyst EPS Estimate (Medium Size companies)

| Years | R2 | Sig | F | Research | Gross | Log of | Revenue | Long- |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets | difference | term |
|  |  |  |  |  | /Sig |  | /Sig |  |
| Current | .069 | .000 | 20.42 | - | $.069 / .014$ | - | $.284 / .000$ | $-.049 / .067$ |
| year |  |  |  | $.132 / .000$ |  | $.060 / .038$ |  |  |
| estimate |  |  |  |  |  |  |  |  |
| Nextyear | .150 | .000 | 50.22 | - | $.116 / .000$ | - | $.449 / .000$ | $-.003 / .912$ |
| estimate |  |  |  | $.138 / .000$ |  | $.022 / .423$ |  |  |

Notes: Regression of independent variables (R\&D percentage difference, gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst EPS estimate percentage difference (dependent variables) filtered on Analyst EPS estimate $\%>\mathbf{- 1 0 0 \%}$ and Analyst EPS estimate $<\mathbf{1 0 0 \%}$ for companies with market capitalization < 10 billion and market capitalization > 1 billion.

Results without revenue difference for midsized companies indicate (Table 30) that research spending for the current year is not significant; however, the standard coefficient of research spending for next year is significant, with the coefficient value at .106. The $R^{2}$ is weak when regression is run without the revenue difference: the $R^{2}$ is 2 percent and 3 percent for the current and next year, respectively.

Table 30 Regression of R\&D... with Analyst Sales Estimate, (Medium Size companies)(No Revenue)

| Years | R2 | Sig | F | Research | Gross | Log of | Long-term |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets /Sig | debt/Sales |
|  |  |  |  | /Sig |  |  |  |
| Current year | .022 | .000 | 8.292 | $.026 / .329$ | $.024 / .35$ | $-1.697 / .09$ | $-.045 / .090$ |
| estimate |  |  |  |  |  |  |  |
| Next year | .039 | .000 | 15.71 | $.106 / .000$ | $.031 / .21$ | $-.146 / .000$ | $.003 / .921$ |
| estimate |  |  |  |  |  |  |  |

Notes: Regression of independent variables (R\&D percentage difference, gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst EPS estimate percentage difference (dependent variables) filtered on Analyst EPS estimate $\%>\mathbf{- 1 0 0} \%$ and Analyst EPS estimate $<\mathbf{1 0 0 \%}$ for companies with market capitalization < 10 billion and market capitalization > 1 billion.

The last category for the regression study under the market capitalization is for small market capitalization companies (Table 31). The regression results pattern is the same here as for large and midsized companies: the standard coefficient values of research spending are negative, with significance slightly above .05 for the current and next years. As in the results for the large and midsized companies, the revenue difference standard coefficient beta value is positive, with a significance of less than .01 .

Table 31 Regression of R\&D, Revenue, ... with Analyst EPS Estimate ( Small Size companies)

| Years | R2 | Sig | F | Research | Gross | Log of | Revenue | Long- |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets | difference | term |
|  |  |  |  | $/$ Sig |  | /Sig |  | debt/Sales |
| Current | .014 | .000 | 7.882 | - | $-.010 / .578$ | - | $.101 / .000$ | $.028 / .146$ |
| year |  |  |  | $.036 / .063$ |  | $.061 / .002$ |  |  |
| estimate |  |  |  |  |  |  |  |  |
| Next year | .042 | .000 | 25.76 | - | $-.042 / .021$ | - | $.209 / .000$ | $-.005 / .787$ |
| estimate |  |  |  | $.036 / .056$ |  | $.018 / .332$ |  |  |

Notes: Regression of independent variables (R\&D percentage difference, gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst EPS estimate percentage difference (dependent variables) filtered on Analyst EPS estimate \% > -100\% and Analyst EPS estimate $<\mathbf{1 0 0 \%}$ for companies with market capitalization <= 1 billion.

In Table 32, the regression results without revenue difference for small companies demonstrate that the $\mathrm{R}^{2}$ and F values are weak for both years. The research spending standard coefficient beta is .05 , with significance less than .01 ; however, for the current year, the results are not significant.

Table 32 Regression of R\&D... with Analyst EPS Estimate (Small Size companies) (No Revenue)

| Years | R2 | Sig | F | Research | Gross | Log of | Long-term |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets /Sig | debt/Sales |
|  |  |  |  | /Sig |  |  |  |
| Current year | .006 | .001 | 4.605 | $.011 / .521$ | $-.015 / .393$ | $-.071 / .000$ | $.023 / .189$ |
| estimate |  |  |  |  |  |  |  |
| Next year | .004 | .000 | 3.706 | $.054 / .002$ | $-.022 / .200$ | $-.026 / .140$ | $-.007 / .682$ |
| estimate |  |  |  |  |  |  |  |


#### Abstract

Notes: Regression of independent variables (R\&D percentage difference, gross margin, Log(Assets), Long-term debt/sales) with Analyst EPS estimate percentage difference (dependent variables) filtered on Analyst EPS estimate \% > $100 \%$ and Analyst EPS estimate $<\mathbf{1 0 0 \%}$ for companies with market capitalization <= 1 billion.


The next set of results is based on industry grouping. I performed correlation tests to understand the association of R\&D spending with analysts' EPS estimate revisions and how the results vary for different industry groups. Correlation results for sales demonstrated a positive correlation, with a significance of less than .01. However, reviewing the EPS results (Table 33) shows that only the Semiconductor industry grouping has a positive correlation of .078 for the current year and .017 for the next year, with a significance of less than .05 . The other industry results show positive correlation, but with a significance value that is greater than .05 .

Table 33 Correlation of R\&D with Analyst EPS - Industry Group Segmentation

| Years | Market value | Pearson <br> Correlation | Sig | N |
| :---: | :---: | :---: | :---: | :---: |
| Current year <br> Estimate | Software and Services (4510) | . 014 | . 486 | 2555 |
| Current year <br> Estimate | Technology Hardware and equipment (4520) | . 004 | . 844 | 2388 |
| Current year <br> Estimate | Semiconductor and Semiconductor equipment(4530) | . 078 | . 003 | 1475 |
| Next year <br> Estimate | Software and Services (4510) | . 036 | . 072 | 2568 |
| Next year <br> Estimate | Technology Hardware and equipment (4520) | . 017 | . 390 | 2476 |
| Next year <br> Estimate | Semiconductor and Semiconductor equipment(4530) | . 110 | . 000 | 1492 |

Notes: Industry wise Correlation of Analyst EPS estimate percentage difference (dependent variable) with $\mathbf{R} \& D$ percentage difference (independent variable) Analyst EPS estimate \% > -100\% and Analyst EPS estimate < $\mathbf{1 0 0 \%}$.

The next six tables (34 to 39) show regression results of analysts' EPS estimate percentage difference based on industry segmentation. As with other EPS tests, the data is filtered by analysts' EPS estimate to avoid the outliers with extreme values. The same pattern is observed in research spending, with a negative standard coefficient beta value for the Software and Services industry group for both years and values that are not
significant (Table 34). The revenue difference standard coefficient is positive at .115 and .214 for the current year and next year, respectively, with less than .01 significance. The $R^{2}$ values are 2 percent and 6 percent for the current and next years, respectively, whereas $F$ values are 8 and 28, respectively.

Table 34 Regression of R\&D......with Analyst EPS Estimate (Software and Services companies)

| Years | R2 | Sig | F | Research | Gross | Log of | Revenue | Long- |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets | difference | term |
|  |  |  |  |  | /Sig |  | /Sig |  |
| Current | .021 | .000 | 8.46 | - | $-.004 / .873$ | $.094 / .000$ | $.115 / .000$ | $.001 / .949$ |
| year |  |  |  | $.028 / .226$ |  |  |  |  |
| estimate |  |  |  |  |  |  |  |  |
| Next year | .066 | .000 | 28.89 | - | $-.049 / .021$ | $.145 / .000$ | $.214 / .000$ | $.035 / .114$ |
| estimate |  |  |  | $.035 / .139$ |  |  |  |  |


#### Abstract

Notes: Regression of independent variables (R\&D percentage difference, gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst EPS estimate percentage difference (dependent variables) filtered on Analyst EPS estimate $\%>\mathbf{- 1 0 0 \%}$ and Analyst EPS estimate $<\mathbf{1 0 0 \%}$ for software and services industry group.


Software and Services regression results without revenue difference (Table 35) reveal that the research spending standard coefficient for next year is positive at .044 , with significance below .05 . The current year result is not significant even though values are positive. The $R^{2}$ values are quite weak for both years, and the $F$ values are 2 and 11 for the current and next year.

Table 35 Regression of R\&D......with Analyst EPS Estimate (Software and service companies) (No Revenue)

| Years | R2 | Sig | F | Research | Gross | Log of | Long-term |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets /Sig | debt/Sales |
|  |  |  |  | /Sig |  |  |  |
| Current year | .005 | .000 | 2.976 | $.015 / .446$ | $-.012 / .555$ | $.067 / .001$ | $.003 / .882$ |
| estimate |  |  |  |  |  |  |  |
| Next year | .018 | .000 | 11.20 | $.044 / .026$ | $-.041 / .039$ | $.115 / .000$ | $.024 / .000$ |
| estimate |  |  |  |  |  |  |  |

Notes: Regression of independent variables (R\&D percentage difference, gross margin, Log(Assets), Long-term debt/sales) with Analyst EPS estimate percentage difference (dependent variables) filtered on Analyst EPS estimate \% > $\mathbf{1 0 0 \%}$ and Analyst EPS estimate $<\mathbf{1 0 0 \%}$ for software and services industry group.

Technology Hardware and Equipment regression results reveal a negative standard coefficient for research spending, which is significant as the values are less than .01 for the current year and .06 for the next year. Revenue difference is positive at .150 and .260 for the current and next year, with less than .01 significance. The $R^{2}$ values are 2 percent and 8 percent for the current and next year, respectively, and the $F$ values are 11 and 35 , respectively.

Table 36 Regression of R\&D......with Analyst EPS Estimate (Technology Hardware companies)

| Years | R2 | Sig | F | Research | Gross | Log of | Revenue | Long- |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets | difference | term |
|  |  |  |  | /Sig |  | /Sig |  | debt/Sales |
| Current | .029 | .000 | 11.4 | - | $-.031 / .167$ | $.103 / .000$ | $.150 / .000$ | $-.007 / .763$ |
| year |  |  |  | $.067 / .006$ |  |  |  |  |
| estimate |  |  |  |  |  |  |  |  |
| Next year | .080 | .000 | 35.89 | - | $-.002 / .916$ | $.133 / .000$ | $.260 / .000$ | $.133 / .000$ |
| estimate |  |  |  | $.043 / .060$ |  |  |  |  |


#### Abstract

Notes: Regression of independent variables (R\&D percentage difference, gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst EPS estimate percentage difference (dependent variables) filtered on Analyst EPS estimate $\%>-\mathbf{1 0 0 \%}$ and Analyst EPS estimate $<\mathbf{1 0 0 \%}$ for technology hardware industry group.


In Table 37, results without the revenue spending difference for Technology Hardware and Equipment companies' show that the standard coefficient is not significant. In this regression, the assets log seems to have a positive standard coefficient with a significance of less than .05 . Similar to other regressions without revenue, the $\mathrm{R}^{2}$ values are 1 percent or less for both years.

Table 37 Regression of R\&D.......with Analyst EPS Estimate (Technology hardware companies) (No Revenue)

| Years | R2 | Sig | F | Research | Gross | Log assets | Long-term |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | /Sig | debt/Sales |
|  |  |  |  |  |  |  |  |
| Current year | .005 | .018 | 2.969 | $.004 / .851$ | $-.022 / .284$ | $.071 / .001$ | $-.013 / .539$ |
| estimate |  |  |  |  |  |  |  |
| Next year | .014 | .000 | 8.542 | $.018 / .362$ | $.018 / .374$ | $.101 / .000$ | $-.07 / .001$ |
| estimate |  |  |  |  |  |  |  |

Notes: Regression of independent variables (R\&D percentage difference, gross margin, Log(Assets), Long-term debt/sales) with Analyst EPS estimate percentage difference (dependent variables) filtered on Analyst EPS estimate \% > $\mathbf{1 0 0 \%}$ and Analyst EPS estimate $<\mathbf{1 0 0 \%}$ for technology hardware industry group.

Semiconductor Equipment regression results (Table 38) also have revenue difference as a strong influence, with the standard coefficient at .306 and .284 for the current year and the next year, respectively, and a significance of less than .01 . The research spending values are not significant for both years and the $R^{2}$ value is 9 percent for both years.

Table 38 Regression of R\&D......with Analyst EPS Estimate (Semiconductor companies)

| Years | R2 | Sig | F | Research | Gross | Log of | Revenue | Long- |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets | difference | term |
|  |  |  |  | /Sig |  | /Sig |  | debt/Sales |
| Current | .091 | .000 | 25.73 | - | $-.05 / .062$ | - | $.306 / .000$ | $.081 / .004$ |
| year |  |  |  | $.048 / .093$ |  | $.027 / .330$ |  |  |
| estimate |  |  |  |  |  |  |  |  |
| Next year | .090 | .000 | 25.99 | - | $-.028 / .286$ | $.119 / .000$ | $.284 / .000$ | $-.029 / .291$ |
| estimate |  |  |  | $.015 / .596$ |  |  |  |  |

Notes: Regression of independent variables (R\&D percentage difference, gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst EPS estimate percentage difference (dependent variables) filtered on Analyst EPS estimate \% > - $\mathbf{1 0 0 \%}$ and Analyst EPS estimate < $\mathbf{1 0 0 \%}$ for semiconductor industry group.

Regression results for the Semiconductor industry group without the revenue spending difference show a positive standard coefficient value for research spending, with a significance of less than .01 . The research spending has the most influence in these regression results, with standard coefficient values at .075 and .116 for the current and next year, respectively. When compared with other industries, this research spending difference for the Semiconductor group has consistent positive coefficient values for both years.

Table 39 Regression of R\&D......with Analyst Sales Estimate (Semiconductor companies)(No Revenue)

| Years | R2 | Sig | F | Research | Gross | Log of | Long-term |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets /Sig | debt/Sales |
|  |  |  |  | /Sig |  |  |  |
| Current year | .011 | .003 | 4.069 | $.075 / .004$ | $-.020 / .438$ | $.07 / .010$ | $-.031 / .255$ |
| estimate |  |  |  |  |  |  |  |
| Next year | .021 | .000 | 7.930 | $.116 / .000$ | $.028 / .274$ | $.088 / .001$ | $-.024 / .377$ |
| estimate |  |  |  |  |  |  |  |

Notes: Regression of independent variables (R\&D percentage difference, gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst EPS estimate percentage difference (dependent variables) filtered on Analyst EPS estimate $\%>\mathbf{- 1 0 0 \%}$ and Analyst EPS estimate $<\mathbf{1 0 0 \%}$ for semiconductor industry group.

## Analysis for R\&D expenditure more than 5 percent

In their study on long-term abnormal stock returns and operating performance following R\&D increases, Eberhart, Maxwell, and Siddque (2004) used companies with research spending greater than 5 percent as their sample data. Their study found consistent evidence of abnormal stock returns experienced by the firm shareholders following the R\&D increases. They also observed positive long-term abnormal operating performance following their R\&D increases and suggested that the market is slow to recognize the extent of this benefit. One of their sample criteria for selecting data for their study was a 5 percent increase in R\&D. My sample data was generated by filtering the data for research spending greater than 5 percent; I then ran correlation and regression tests
against this generated sample data. Tables 40 to 44 show the results of these tests: the sample sizes for $R \& D$ spending percentage greater than 5 percent are 4,241 and 4,176, for the current year and next year, respectively, whereas the complete sample without the filter of 5 percent was 6,830 for the current year and 6,690 for the next year. So, using the data filter reduced the sample size by 38 percent for both the current and next year. The correlation results (Table 40) for all companies with research spending greater than 5 percent year over year shows a positive correlation, with .201 and .167 for the current and next year, respectively, and a significance of less than .01 for both years. The correlation results were positive for all companies, with values at .153 and .183 for the current and next year, respectively (Table 13).

Table 40 Correlation R\&D with Analyst Sales Estimate for R\&D increase > 5\%

| Years | Sector | Pearson | Sig | N |
| :--- | :--- | :--- | :--- | :--- |
|  | description | Correlation |  |  |
| Current year estimate | All IT | .201 | .000 | 4241 |
| Next year estimate | ALL IT | .167 | .000 | 4176 |

## Notes: Correlation of Analyst sales estimate percentage difference (dependent variable) with $\mathbf{R} \& D$ percentage difference (independent variable) filtered on Analyst EPS estimate percentage > -100 and Analyst EPS estimate percentage < $100 \%$ for R\&D spending difference year over year >5\%.

The regression results (Table 41) for all companies using the same constructs as in my earlier tests demonstrates that the standard coefficient beta for research spending is .310 and .146 for current year and next year, respectively, with a significance of less than .01. The values of research spending difference are higher than revenue spending
difference. The standard coefficient values for research spending for all IT companies without the R\&D spending filter were .221 and .159 for the current and next year, respectively (Table 14).

Table 41 Regression of R\&D, Revenue......with Analyst Sales Estimate with R\&D increase > 5\%

| Years | R2 | Sig | F | Research | Gross | Log | Revenue | Long- |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets | difference | term |
|  |  |  |  | /Sig |  | /Sig |  | debt/Sales |
|  |  |  |  |  |  |  |  |  |
| Current | .130 | .000 | 111.3 | $.310 / .000$ | $.078 / .000$ | $.128 / .000$ | $.098 / .000$ | $-.018 / .184$ |

year
estimate
Next year $\quad .057 \quad .000 \quad 43.15 \quad .146 / .000 \quad .007 / .654 \quad .131 / .000 \quad .123 / .000 \quad .000 / .977$
estimate

## Notes : Regression of Independent variables (R\&D expenditure percentage difference, Ratio of gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst sales estimate percentage difference (dependent variable) filtered on R\&D spending difference year over year > 5\%.

The correlation results (Table 42) for analysts' EPS estimate percentage difference compared to $\mathrm{R} \& \mathrm{D}$ percentage difference shows that the correlation is not significant in either year. These results are consistent with the results for all IT companies without the $\mathrm{R} \& \mathrm{D}$ spending filter (Table 23).

Table 42 Correlation R\&D with Analyst EPS Estimate for R\&D increase > 5\%

| Years | Sector description | Pearson | Sig | N |
| :--- | :--- | :--- | :--- | :--- |
|  |  | Correlation |  |  |
| Current year estimate | All IT | .005 | .726 | 4217 |
| Next year estimate | All IT | .016 | .281 | 4304 |

Notes: Correlation of Analyst EPS estimate percentage difference (dependent variable) with $\mathbf{R} \& D$ percentage difference (independent variable) filtered on Analyst EPS estimate percentage > -100 and Analyst EPS estimate percentage < $100 \%$ for on R\&D spending difference year over year > 5\%.

Regression results (Table 43) for analysts' EPS estimate percentage difference with the independent variables (research spending, gross margin, assets log, revenue difference, and long-term debt/sales) shows that, for the current year, the correlation in not significant, while for the next year, it is negatively correlated with a value of -.047 and a significance of less than .01 . The correlation results for all IT companies (Table 24) for the current year is not significant, while the next year value is -.035 . The revenue difference (Table 43) has a positive standard coefficient with a significance of less than .01 . The $\mathrm{R}^{2}$ values are 3 percent and 8 percent and the F values are 26 and 61 for the current and next year, respectively.

Table 43 Regression of R\&D, Revenue......with Analyst EPS Estimate for R\&D increase > 5\%.

| Years | R2 | Sig | F | Research | Gross | Log | Revenue | Long- |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Sig | assets | difference | term |
|  |  |  |  |  | /Sig |  | /Sig |  |
| Current | .039 | .000 | 26.70 | - | $-.008 / .000$ | $.147 / .000$ | $.141 / .000$ | $-.028 / .115$ |
| year |  |  |  | $.029 / .104$ |  |  |  |  |
| estimate |  |  |  |  |  |  |  |  |
| Next year | .081 | .000 | 61.09 | - | $-.031 / .053$ | $.157 / .000$ | $.250 / .000$ | $-.008 / .625$ |
| estimate |  |  |  | $.047 / .007$ |  |  |  |  |

Notes: Regression of independent variables (R\&D percentage difference, gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst EPS estimate percentage difference (dependent variables) filtered on Analyst EPS estimate $\%>-\mathbf{1 0 0 \%}$ and Analyst EPS estimate $<\mathbf{1 0 0 \%}$ for on R\&D spending difference year over year >5\%.

I performed the regression again by removing the revenue difference construct for the R\&D spending difference of greater than 5 percent year over year. The research spending standard coefficient values are not significant (Table 44), the $R^{2}$ values for both years are 1 percent, and the F values are 12 and 13 for the current and next year, respectively.

Table 44 Regression of R\&D......with Analyst EPS Estimate (No Revenue) filtered for R\&D increase > 5\%

| Years | R2 | Sig | F | Research | Gross | Log assets | Long-term |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  | spending | Margin/Si | $/ \mathrm{Sig}$ | debt/Sales |
|  |  |  |  | $/ \mathrm{Sig}$ | g |  |  |
| Current year | .011 | .000 | 12.21 | $.008 / .600$ | $-.020 / .202$ | $.110 / .000$ | $-.033 / .038$ |
| estimate |  |  |  |  |  |  |  |
| Next year | .013 | .000 | 13.86 | $.020 / .913$ | $-.018 / .227$ | $.115 / .000$ | $-.018 / .240$ |
| estimate |  |  |  |  |  |  |  |

Notes: Regression of independent variables (R\&D percentage difference, gross margin, Log(Assets), Long-term debt/sales) with Analyst EPS estimate percentage difference (dependent variables) filtered on Analyst EPS estimate \% > -100\% and Analyst EPS estimate $\mathbf{<} \mathbf{1 0 0 \%}$ for on R\&D spending difference year over year > 5\%.

## VI CHAPTER 6: DISCUSSION

After reviewing the correlation test results for all the IT companies, I found that R\&D expenditure has a positive correlation with analysts' sales estimates. These results are consistent for the current year and the next year, but the next year results show higher correlation than the current year. The correlation results for analysts' EPS estimates and R\&D expenditure are different than for the analysts' sales estimate correlation results. The results for all IT companies show that the R\&D spending and the analysts' EPS estimates do not have a significant correlation. The next year correlation values are close to significant, and they improved from the current year correlation values. The regression results for all IT companies using the independent variables (R\&D expenditure difference percentage, gross margin, assets log, revenue difference, longterm debt/sales) and the dependent variable (analysts' sales estimate percentage difference) reveal a strong $\mathrm{R}^{2}$ for both years and a significant positive standard coefficient beta for both years. The revenue and assets have a significant standard coefficient beta for both years, whereas the research spending standard coefficient reduces year over year.

## Summary of all Correlation of analyst sales and EPS with R\&D

Table 45 Consolidated Correlation of R\&D with Analyst Sales and EPS Revision


| EPS | Current year | All IT companies | . 008 | . 536 | 6387 | Not Significant |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | > $=10$ billion | . 134 | . 007 | 404 | Yes |
|  |  | >1 billion \& < 10billion | . 050 | . 051 | 1533 | Yes |
|  |  | $=<1$ billion | . 013 | . 457 | 3370 | Not Significant |
|  |  | Software and Services | . 014 | . 486 | 2555 | Not Significant |
|  |  | Technology Hardware and | . 004 | . 844 | 2388 | Not Significant |
|  |  | equipment |  |  |  |  |
|  |  | Semiconductor | . 078 | . 003 | 1475 | Yes |
|  |  | R\&D spending percentage | . 005 | . 726 | 4217 | Not Significant |
|  |  | > 5 |  |  |  |  |
|  | Next year | All IT companies | . 021 | . 084 | 6515 | Not Significant |
|  |  | > $=10$ billion | . 223 | . 000 | 405 | Yes |
|  |  | >1 billion \& < 10billion | . 133 | . 000 | 1558 | Yes |
|  |  | $=<1$ billion | . 055 | . 001 | 3425 | Yes |
|  |  | Software and Services | . 036 | . 072 | 2568 | Not Significant |
|  |  | Technology Hardware and | . 017 | . 390 | 2476 | Not Significant |
|  |  | equipment |  |  |  |  |
|  |  | Semiconductor and | . 110 | . 000 | 1492 | Yes |
|  |  | Semiconductor equipment |  |  |  |  |
|  |  | R\&D spending percentage | . 016 | . 281 | 4304 | Not Significant |
|  |  | > 5 |  |  |  |  |

The same correlation tests that were executed for analyst sales estimates and R\&D spending difference were performed for the first 10 years to observe analysts' pattern of
revising analyst sales and EPS estimates. Table 46 shows the results for the first 10 years (1995-2004). The results for next 10 years (2005-2014) are captured in Table 47.

## Summary of all Correlation of analyst sales and EPS with R\&D for first 10 years

Table 46 Consolidated Correlation of R\&D with Analyst Sales and EPS Revision for first 10 years ( 1995 - 2004)


Summary of all Correlation of analyst sales and EPS with R\&D for last $\mathbf{1 0}$ years
Table 47 Consolidated Correlation of R\&D with Analyst Sales and EPS Revision for first 10 years (2005-2014)


## VI. 1 Correlation analysis

A review of all of the companies in Table 45 shows a strong correlation that supports Hypothesis 1.

## Hypothesis 1: A change in R\&D expenditure impacts analysts' sales estimate

 forecasts for the current year.The market capitalization segmentation analysis shows that correlation strength varies depending on company size: large companies have a strong positive correlation, midsized companies have a positive correlation, and small companies have a relatively weak correlation. Srinivasan's (2007) study shows the contingent effect of a firm's advertising and R\&D expenditures on the dispersion in analysts' forecast of earnings, suggesting that analysts pay attention to firms' marketing activities. Srinivasan further adds that firms with a record of past performance, have decreased dispersion in analyst forecasts. Typically, large and midsized companies have been in business longer and have past experience that gives analysts more confidence when they update their estimates. The greater the firm's R\&D expenditure, the lower the support for dispersion in analysts' forecast hypotheses (Srinivasan 2007). Over the past two decades, smaller firms have begun allocating more money to R\&D (see Table 48). Hirschey et al. (2012) discuss how the $R \& D$ share is being distributed among large, medium, and small companies. For example, companies that Hirschey et al. ranked above 1,000 spent more
than $\$ 1$ billion on R\&D in 2010. (Their data does not include financial or utility companies.)

Table 48 Distribution of R\&D Spending
(Hirschey et al)

| Ranking | R\&D Sbare |  | Cumulative R\&D Sbare |  | Real R\&D (2010 Smillions) |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 1976 | 2010 | 1976 | 2010 | 1976 | 2010 |
| top 100 | 79.1\% | 75.7\% | 79.1\% | 75.7\% | \$39,611.5 | \$170,647.1 |
| 101-200 | 10.7\% | 9.6\% | 89.8\% | 85.2\% | 5,343.7 | 21,638.8 |
| 201-300 | 4.1\% | 4.5\% | 93.9\% | 89.7\% | 2,055.1 | 10,110.8 |
| 301-400 | 2.2\% | 2.9\% | 96.1\% | 92.6\% | 1,092.5 | 6,500.0 |
| 401-500 | 1,3\% | 2.1\% | 97.4\% | 94.7\% | 675.5 | 4,730.8 |
| 501-600 | 0.9\% | 1.5\% | 98.4\% | 96.2\% | 462.5 | 3,453.4 |
| 601-700 | 0.6\% | 1.1\% | 99.0\% | 97.4\% | 308.0 | 2,534.5 |
| 701-800 | 0.4\% | 0.8\% | 99.4\% | 98.2\% | 215.4 | 1,893.9 |
| 801-900 | 0.3\% | 0.6\% | 99.7\% | 98.8\% | 139.0 | 1,390.0 |
| 901-1000 | 0.2\% | 0.4\% | 99.9\% | 99.3\% | 89.5 | 999.9 |
| 1001-1100 | 0.1\% | 0.3\% | 100.0\% | 99.6\% | 46.4 | 690.6 |
| 1101-1200 |  | 0.2\% |  | 99.8\% | 15.7 | 448.6 |
| 1201-1300 |  | 0.1\% |  | 99.9\% |  | 276.4 |
| 1301-1400 |  | 0.1\% |  | 100.0\% |  | 173.2 |
| 1401-1500 |  | 0.0\% |  | 100.0\% |  | 75.5 |
| 1501-1600 |  | 0.0\% |  | 100.0\% |  | 7.5 |
| Total for all firms | 100.0\% | 100.0\% | 100.0\% | 100.0\% | \$50,054.9 | \$225,571.0 |

From a comparative review of industry segmentation, the Software and Services group has a stronger positive correlation than the Technology Hardware or Semiconductor groups, but all three industry groups are positively correlated. In a 2014 Market Realist article, Hirschey et al. discuss the software company cost structure of IT companies, noting that the majority of their operating expenditures consist of R\&D costs and marketing spending. With the emergence and adoption of cloud computing and open source software, companies are finding it very difficult to maintain the high margins once associated with the industry (Market Realist, 2014). They note that software companies spend more on R\&D investment to differentiate those products that are hard to replicate or that are protected by intellectual property rights or patents. Patents serve as armor for
software companies. Small companies and start-ups find patents hard to penetrate; establishing a strong customer base makes switching costs very high and adds to the challenges of start-ups.

As per Bloomberg's data, between 2013 and 2014, the Internet companies Google and Amazon increased their R\&D spending by 17 and 43 percent, respectively; hardware companies IBM and Cisco increased their R\&D spending by -1.2 and 8.3 percent, respectively; and the Semiconductor company Intel increased its R\&D spending by 4.6 percent. The growth of cloud technology, which has impacted hardware companies' revenues for the past decade, has not given analysts the confidence to update that sector's forecast. As other researchers have noted, past performance of companies decreases dispersion in analyst forecasts (Srinivasan, 2007). In my analysis of companies that have increased their R\&D spending by 5 percent, I found a strong positive correlation between R\&D spending and analyst sales estimate revisions for the current year.

## Hypothesis 2: A change in R\&D expenditure impacts analysts' sales estimate

 forecasts for the next year.I ran correlation and regression tests for only one year following the current year; testing beyond that reduces the analyst forecast quality. The O'Brien study (1988) compares consensus analyst forecasts with time-series forecasts from one- to fourquarters ahead. The analyst forecasts outperform the time-series model for one- and two-quarter-ahead forecasts, are approximately the same for three-quarter ahead forecasts, and perform worse for four-quarter ahead forecasts. Thus, the advantage analysts gain from firm-specific information seems to deteriorate as the time horizon for forecasting is
extended. In valuation, the focus is more on long-term growth rates in earnings than on next quarter's earnings. There is little evidence to suggest that analysts provide superior earnings forecasts when those forecasts span three or five years. An early study by Malkiel \& Cragg (1980) compared long-term forecasts by five investment management firms in 1962 and 1963 with actual growth over the following three years; they concluded that analysts were poor long-term forecasters.

In my study, the correlation for next year is positive in all categories: for all IT companies, it was slightly higher and, for hardware companies, the next year correlation was higher than the current year. Another noticeable result was that, in the smaller companies, the next year correlation is not as strong as the current year. The fact that the analyst forecasts are not accurate over the longer term might be due to other unknown factors, such as interest rate changes and industry-specific fluctuations. These characteristics that are not firm-specific impact long-term analyst forecasts, which reduces the correlation for smaller companies as they often lack the past performance information that analysts use to update their next year estimates. Fairfield, Ramnath, and Yohn (2009) describe how industry-specific models generate more accurate forecasts of sales growth in firms because "firms' sales growth depends on product demand, which are determined at the industry level."

## Hypothesis 3: A change in R\&D expenditure impacts analysts' EPS estimate

 forecasts for the current year.When reviewing the correlation between analysts' current year EPS estimates and R\&D expenditures for all companies, the results are not significant. The companies with
more than $\$ 10$ billion market value show positive correlation, as do midsized companies (between $\$ 1$ billion and $\$ 10$ million in market value), whereas the correlation is not significant with smaller companies. This might be due to various other expenses that impact the EPS, despite a positive correlation with the sales. For example, Company A might increase its R\&D budget and see sales increase in the current year or next year, but expenses such as additional marketing expenses or hiring more customer service personal to support the new products might impact the EPS and hence analysts might not change the EPS based on R\&D expenditure revisions.

Past performance of the companies can also influence analyst updates of the EPS. Large and midsized companies with more of an R\&D history might inspire analysts to modify the EPS estimates. Semiconductor companies show a positive correlation here, whereas Software and Services and Technology Hardware and Equipment companies do not show significant results.

## Hypothesis 4: A change in R\&D expenditure impacts analysts' EPS estimate

 forecasts for the following year.The next year results for analysts' EPS estimate forecasts are much better than the current year results. This is because other costs associated with products are reduced as the years go by, so analysts increase the earnings. When comparing for companies with market cap greater than $\$ 10$ billion, the correlation value is .223 - much higher than the current year value of .134. Midsized companies show a next year correlation value of .133, compared to a current year value of .050 . For small capitalization companies, the next year results show a positive correlation compared to the current year, where the
results were not significant. In terms of industry groups, Technology Hardware correlation for the next year is not significant; the values are similar to the current year. As with the current year, Software and Services results are not significant for the next year, though the significance value is lower at .07 . For the Semiconductor industry, the correlation results for the next year are higher than the current year, but both are positively correlated. The correlation results for $\mathrm{R} \& \mathrm{D}$ spending greater than 5 percent over the previous year is not significant.

## Table 49 Correlation Summary of All Tests, Year Wise View

|  | Sales |  | EPS |  |
| :--- | :--- | :--- | :--- | :--- |
|  | Current year | Next year | Current year | Next year |
| All IT companies | .153 | .183 | Not Sig | Not Sig |
| $>=10$ billion | .569 | .469 | .134 | .223 |
| $>1$ billion \& <10billion | .350 | .361 | .050 | .133 |
| $=<1$ billion | .128 | .102 | Not Sig | .055 |
| Software and Services | .210 | .205 | Not Sig | Not Sig |
| Technology Hardware and | .088 | .211 | Not Sig | Not Sig |
| Semiconductor | .136 | .120 | .078 | .110 |
| R\&D spending percentage >5 | .201 | .167 | Not Sig | Not Sig |

## Notes : Correlation summary of R\&D difference to Analyst sales and EPS estimate changes.

Table 49 consolidates the summary of all correlation results. As a review of this table shows, Hypothesis 1 and Hypothesis 2 are strongly supported. Hypotheses 3 and 4 are not significant for all tests, but Hypothesis 4 has better results than Hypothesis 3. Givoly, Hayn, and D'Souza's (1999) study shows that there is a stronger correlation between firm sales and industry sales than between firm profits and industry. Sales
increases are not impacted by other expenses that normally impact the EPS, such as interest expenses and tax law changes. So, R\&D spending has direct association with the sales estimate, whereas for the EPS is impacted by many other factors.

It is also clear that the current year sales correlation results are stronger than the next year's results, but for EPS, the next year correlation results are stronger than the current year. This phenomenon is due to the fact that R\&D spending by IT might show revenue benefit the same year as the IT industry is moving at a rapid pace. The next year EPS results are better because the current year includes additional marketing and training costs to launch the product; these costs reduce in the years that follow.

When I analyzed the results for first 10 years and compared them with next 10 years, I noticed a change in analyst's behavior. The decade wise results were gathered by running correlation tests for the analyst sales estimate difference with R\&D spending changes. Table 50 shows the results from both tests. These decade wise results are compared against the test results from 20 years.

Table 50 Correlation tests for analyst sales estimate with R\&D decade wise compared with 20 years data

|  | Current/Next | Data filtered on | 1995- | 2005- | 1995-2014 |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  | 2004 | 2014 |  |
| Sale | Current year | All IT companies | . 232 | . 046 | . 153 |
| s |  |  |  |  |  |
|  |  | $>=10$ billion | . 584 | . 247 | . 569 |
|  |  | >1 billion \& <10billion | . 370 | . 278 | . 350 |
|  |  | $=<1$ billion | . 208 | Not Sig | . 128 |
|  |  | Software and Services | . 361 | . 009 | . 210 |
|  |  | Technology Hardware | . 093 | . 132 | . 088 |
|  |  | and equipment |  |  |  |
|  |  | Semiconductor | . 144 | . 130 | . 136 |
|  |  | R\&D spending | . 223 | . 070 | . 201 |
|  |  |  |  |  |  |
|  | Next year | All IT companies | . 221 | . 108 | . 183 |
|  |  | $>=10$ billion | . 473 | . 258 | . 479 |
|  |  | >1 billion \& <10billion | . 385 | . 320 | . 361 |
|  |  | $=<1$ billion | . 138 | . 062 | . 102 |
|  |  | Software and Services | . 251 | . 093 | . 205 |
|  |  | Technology Hardware | . 230 | . 201 | . 211 |
|  |  | and equipment |  |  |  |
|  |  | Semiconductor and | . 162 | . 065 | . 120 |
|  |  | Semiconductor equipment |  |  |  |
|  |  | R\&D spending | . 200 | . 075 | . 167 |
|  |  | percentage > 5 |  |  |  |

The results for the decade from 1995-2004 illustrate that the next year correlation results are lower than the current year results; however, for midsized companies, the Technology Hardware and Semiconductor industries' next year results are higher than the current year. The results for 1995-2004 follow the same pattern as the 20-year correlation results, except for IT and semiconductor companies.

My analysis also showed that the pattern for 2005-2014 was different than that for 1995-2004. The correlation results for 2005-2014 are positive, but they are much lower than the 1995-2004 results. Also the results indicate that the next year results for 2005-2014 are higher than current year results. Based on these results, there is strong evidence that the $\mathrm{R} \& D$ spending differences in the past decade (2005-2014) are associated more with the next year analyst sales estimate revisions than those of the current year.

## VI. 2 Regression Analysis

I performed regression analysis using the five independent variables (R\&D expenditure difference percentage, gross margin, assets log, revenue difference, and longterm debt/sales) and the analysts' sales estimate percentage difference as the dependent variable. For sales and EPS regression tests, R\&D, revenue, and assets log had significant positive standard coefficient, while gross margin and long-term debt/sales did not.

## Hypothesis 1: A change in R\&D expenditure impacts analysts' sales estimate

## forecasts for the current year.

I analyzed regression results to see if Hypothesis 1 is supported by R\&D changes. For all IT companies, $\mathrm{R}^{2}$ is .07 and the standard coefficient beta is positive. Segment analysis
shows that, for market capitalization greater than $\$ 10$ billion, the $R^{2}$ is .65 and $R \& D$ is not significant, but the revenue coefficient is high at .839 . For midsized companies, $\mathrm{R}^{2}$ is .50 and, similar to large companies, the revenue coefficient is high. Based on this information, I conclude that the revenues are influencing the regression. For large companies, the revenue changes are not volatile and $\mathrm{R} \& \mathrm{D}$ spending change influences will be low. For small companies, the R\&D standard coefficient is positive at .217 , whereas the revenue is not significant. Also, revenue changes are volatile for small companies, and quite hard to predict, hence the R\&D has a greater influence in analysts' sales estimate changes.

Studying industry group segmentation shows that Software and Services and Technology Hardware and Equipment have a positive standard coefficient for R\&D expenditure changes. Semiconductor revenue, however, shows a higher coefficient in the regression. As the Bloomberg report showed, Software and Services companies like Google spend more on $R \& D$, and hence analysts are influenced to increase their sales estimates over their revenue estimates. The same report showed that Cisco, a Technology Hardware Company, increased its R\&D by a high single digit, whereas the Semiconductor company Intel increased it by only 4 percent. Analysts reviewing the percentage increase of $R \& D$ spending are influenced by the industry practice of $R \& D$ spending, and will use the latter to update their sales estimate.

For companies with abnormal R\&D increases, the R\&D year-over-year standard coefficient is .310 compared to .098 for revenue difference. This result shows that, the higher the R\&D increase, the more likely analysts are to change their sales forecasts. Based on my analysis of these regression, I conclude that Hypothesis 1 is supported by
the regression test results, even though some results showed that revenue has a considerable influence.

Summary of all regression of Analyst sales and EPS estimate with R\&D
Table 51 Consolidated Regression of R\&D with Analyst Sales and EPS Revision


|  |  | =< 1 billion | . 02 | 17 | . 103 | Not Sig | Asset |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Software and Services | . 12 | 60 | . 168 | . 229 | Asset |
|  |  | Technology Hardware and | . 06 | 27 | . 195 | Not Sig | Asset, Gross |
|  |  | equipment |  |  |  |  | Margin |
|  |  | Semiconductor and | . 16 | 62 | Not | . 390 | Asset |
|  |  | Semiconductor equipment |  |  | Sig |  |  |
|  |  | R\&D spending percentage | . 06 | 43 | . 146 | . 123 | Asset |
|  |  | > 5 |  |  |  |  |  |
| EPS | Current | All IT companies | . 02 | 24 | Not | . 124 | Asset |
|  | year |  |  |  | Sig |  |  |
|  |  | > $=10$ billion | . 12 | 10 | Not | . 386 |  |
|  |  |  |  |  | Sig |  |  |
|  |  | $>1$ billion \& < 10billion | . 07 | 20 | -. 132 | . 284 |  |
|  |  |  |  |  |  |  | Gross Margin |
|  |  | $=<1$ billion | . 01 | 8 | Not | . 101 |  |
|  |  |  |  |  | Sig |  |  |
|  |  | Software and Services | . 02 | 8 | Not | . 115 | Asset |
|  |  |  |  |  | Sig |  |  |
|  |  | Technology Hardware and | . 02 | 11 | -. 067 | . 150 | Asset |
|  |  | equipment |  |  |  |  |  |
|  |  | Semiconductor | . 09 | 25 | Not | . 306 |  |
|  |  |  |  |  | Sig |  |  |
|  |  | R\&D spending percentage | . 04 | 26 | Not | . 141 | Asset |
|  |  | $>5$ |  |  | Sig |  |  |


|  | All IT companies | . 07 | 83 | -. 035 | . 246 | Asset |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| year |  |  |  |  |  |  |
|  | >=10 billion | . 09 | 8 | Not | . 304 | Asset |
|  |  |  |  | Sig |  |  |
|  | >1 billion \& < 10billion | . 15 | 50 | -. 138 | . 449 | Gross Margin |
|  | $=<1$ billion | . 04 | 26 | Not | . 209 |  |
|  |  |  |  | Sig |  |  |
|  | Software and Services | . 06 | 29 | Not | . 214 | Asset |
|  |  |  |  | Sig |  |  |
|  | Technology Hardware and | . 08 | 36 | Not | . 260 | Asset |
|  | equipment |  |  | Sig |  |  |
|  | Semiconductor and | . 09 | 26 | Not | . 284 | Asset |
|  | Semiconductor equipment |  |  | Sig |  |  |
|  | R\&D spending percentage | . 08 | 61 | Not | . 250 | Asset |
|  | > 5 |  |  | Sig |  |  |

## Hypothesis 2: A change in R\&D expenditure impacts analysts' sales estimate

## forecasts for the next year.

Next year regression results (Table 51) are similar to current year regression results. The tests that had a positive standard coefficient for the current year also had a positive standard coefficient for the next year. The next year's standard coefficient is slightly less than the current year results. As I noted earlier, the O'Brien study (1988) confirms that analyst forecasts that are more than four quarters ahead are worse than those derived using the time-series model. Other macro and industry factors impact analysts' sales estimates for the next year. For example, Skyworks Solutions Company, a supplier to

Apple, might increase its R\&D spending, but if Apple revenue is expected to reduce due to China's slowing growth, analysts might not increase their sales estimate forecasts for Skyworks. The next year results show that, for large companies, midsized companies, and Technology Hardware companies, the revenue independent variable influences the regression more than R\&D expenditures. Thus, when comparing the next year values, the revenue coefficient is slightly lower than the current year.

R\&D spending increases of more than 5 percent have a positive standard coefficient for all companies and a greater influence on regression than on revenue. Out of eight regression tests, five show that $\mathrm{R} \& \mathrm{D}$ has bigger influence on the regression than the revenue, with a positive coefficient. These results support Hypothesis 2-that R\&D expenditure change impacts analysts' next year sales estimate forecasts.

## Hypothesis 3: A change in R\&D expenditure impacts analysts' EPS estimate

 forecasts for the current year.Analyst EPS regression results for all IT companies-using analysts' EPS estimate forecasts as the dependent variable-show that R\&D spending differences have little influence on the regression. However, regression test results show that revenue spending's standard coefficient is positive and has as strong influence on regression. Further, analysts take other costs into consideration, and hence will be unlikely to modify their EPS estimate. For example, strong competition requires more spending on marketing the product, which impacts the bottom line and might prevent analysts from modifying their EPS estimates. Also, companies typically hire customer service reps to
support new products, and this cost also impacts the bottom line and might be another reason why analysts are hesitant to update EPS estimates.

I performed regression without the revenue as a dependent variable, while keeping the other four independent variables and the analysts' EPS estimate revisions as the dependent variable. The results (Table 52) show that, for large and midsized market capitalization companies, the $R \& D$ change is significant, with a positive standard coefficient. In the Semiconductor group, companies have a positive standard coefficient. For all these tests, the $R^{2}$ and $F$ values are quite low. Thus, regarding Hypothesis 3, the regression test shows that revenue is a strong influencer on analysts' EPS estimates, but when revenue is removed from the regression, three out of eight tests demonstrate a positive standard coefficient. Thus, hypothesis 3 is partially supported when revenue is not part of the regression.

## Summary of all regression of Analyst EPS estimate with R\&D, without

## Revenue

Table 52 Consolidated Regression of R\&D with Analyst Sales and EPS Revision (No Revenue)

|  | Current | Data filtered on | $\mathrm{R}^{2}$ | F | R\&D |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | /Next |  |  |  |  |
| EPS | Current year | All IT companies | . 005 | 7.2 | Not Sig |
|  |  | $>=10$ billion | . 02 | 2.1 | . 114 |
|  |  | $>1$ billion $\&<10$ billion | . 02 | 8.2 | . 026 |
|  |  | $=<1$ billion | . 01 | 7.8 | Not Sig |
|  |  | Software and Services | . 00 | 2.9 | Not Sig |
|  |  | Technology Hardware and equipment | . 00 | 2.9 | Not Sig |
|  |  | Semiconductor | . 01 | 4.0 | . 07 |
|  |  | R\&D spending percentage $>5$ | . 01 | 12.21 | Not Sig |
|  | Next year | All IT companies | . 01 | 15 | Not Sig |
|  |  | $>=10$ billion | . 05 | . 00 | . 206 |
|  |  | $>1$ billion $\&<10$ billion | . 04 | 15 | . 106 |
|  |  | $=<1$ billion | . 00 | 3.7 | . 054 |
|  |  | Software and Services | . 01 | 11 | . 044 |
|  |  | Technology Hardware and equipment | . 01 | 8.5 | Not Sig |
|  |  | Semiconductor and Semiconductor | . 02 | . 00 | 7.9 |
|  |  | equipment |  |  |  |
|  |  | R\&D spending percentage $>5$ | . 01 | 13.8 | Not Sig |

## Hypothesis 4: A change in R\&D expenditure impacts analysts' EPS estimate

 forecasts for the following year.EPS next year regression test results (Table 51) are similar to the current year regression results: revenue difference has more influence in the regression for all eight tests and R\&D is not significant; where it is significant, the values are negative standard coefficients. The same reasons as for current year analysts' using revenue over R\&D apply for the next year test results. I reran regression without revenue, keeping other independent variables, to understand the behavior of $\mathrm{R} \& \mathrm{D}$ change on the regression. Results show that, in three out of eight tests, the R\&D differences had a positive standard coefficient for the next year. Five out of the eight tests had a positive standard coefficient, while the next year results had a higher coefficient than the previous year's values.

Large capitalization company values were .206 , while midsized capitalization company values were .106 . The current year values were .114 for large companies and .026 for midsized companies. I also found a significant increase in the coefficient in the next year results. The values of small companies and software companies were not significant in the current year, but have a positive standard coefficient in the next year. As stated in the correlation analysis, analysts modify EPS estimates based on R\&D differences in the: previous year. Analyst are recognizing that R\&D spending in large companies is based on past $\mathrm{R} \& \mathrm{D}$ investment experiences and their impact on the company's financial results.

## Regression summary of R\&D difference to Analyst sales and EPS estimate changes.

 Table 53 Regression Summary of All Tests, Year Wise View|  | Sales |  | EPS |  | Without revenue |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  |  | difference |  |
|  | Current | Next | Current | Next | Current | Next |
|  | year | year | year | year | year | year |
| All IT companies | .221 | .159 | Not Sig | -.035 | Not Sig | Not Sig |
| >=10 billion | Not Sig | Not sig | Not Sig | Not Sig | .114 | .206 |
| $>1$ billion \& <10billion | Not Sig | Not sig | -.132 | -.138 | .026 | .106 |
| $=<1$ billion | .217 | .103 | Not Sig | Not Sig | Not Sig | .054 |
| Software and Services | .235 | .168 | Not Sig | Not Sig | Not Sig | .044 |
| Technology Hardware and | .195 | .195 | -.067 | Not Sig | Not Sig | Not Sig |
| Semiconductor | Not Sig | Not Sig | Not Sig | Not Sig | .07 | 7.9 |
| R\&D spending percentage | .310 | .146 | Not Sig | Not Sig | Not Sig | Not Sig |
| $>5$ |  |  |  |  |  |  |

Table 53 shows all the results, summarizing all of the regression tests in a single snapshot. The regression results are similar to correlation in terms of sales estimate revisions: five out of the seven tests are supported in the current and next year results. The pattern of R\&D impact on analysts' sales estimates reduces in the next year, which matches the pattern observed in analysts' sales estimate revisions and R\&D difference correlation results. Based on the earlier observations, I conclude that hypotheses 1 and 2 are supported. Summarizing analysts' EPS estimates shows no significance for R\&D difference in the regression, and revenue difference is the primary influencer on the estimates. When I remove revenue difference from the regression, large, midsized, and

Semiconductor companies' R\&D change results are significant; the past R\&D performance of these companies influence analysts' to update their EPS estimates. A January 2009 survey conducted by the US National Science Foundation and the US Census Bureau found that companies with fewer than 500 employees accounted for 11 percent of worldwide sales for all US companies, but 19 percent of research spending. Companies with more than 25,000 employees accounted for 42 percent of global sales, but just 36 percent of research outlays. These very small companies (those with 5 to 24 employees) spent $\$ 3$ billion-that is, at least 50 percent more than all other small firms (those with fewer than 500 employees) - in payments to others to do research. The tinest of the companies also had the most research ( $\$ 5$ billion worth) paid for by others (Courtney Rubin, Inc., 2009).

Even though the large companies with more than 25,000 employees are smaller in number, they account for 36 percent of the total research spending. Companies gradually increase their research spending; for large companies, this creates a trail of past performance that lets analysts revise their estimates based on R\&D spending over time. The next year results for EPS estimate revision impact shows five positive coefficients, and their values are higher than the current year, which follows the pattern of EPS correlation results. Regression patterns match correlation patterns when the results are compared year over year. The current year results in sales estimate regression are stronger than the next year results; however, for EPS estimates, the regression without revenue difference shows stronger next year results than current year results. R\&D revisions that impact analysts' sales revisions for IT companies are seen in the same year, while analysts' EPS estimate changes due to R\&D revisions are seen in the next year due
to the additional costs involved in launching a new product. Table 54 shows the results of the Hirschey et al. study that compared research spending changes from 1976 to 2010. Although smaller companies are spending more, the number of firms that spent less than $\$ 25$ million increased by 53 percent, while large ( $\$ 250$ million or more in spending) and midsized companies (between $\$ 25$ and $\$ 250$ million) showed 490 percent and 134 percent growth, respectively.

Table 54 Number of R\&D Firms
(Hirschey et al, 2012)

| Panel A: R\&D |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $\begin{aligned} & \text { Real spending (2010 } \\ & \text { dollars) } \end{aligned}$ | Number of firms 1976 | Number of firms 20110 | Change <br> (1976- <br> 20101 | Percentage change $(1976$. $2010)$ | $\begin{aligned} & \text { Real } \\ & \text { R\& } \\ & 1976 \end{aligned}$ | $\begin{gathered} \text { Real R\&D } \\ 2010 \\ \hline \end{gathered}$ | Change <br> (1976- <br> 2010) | $\begin{gathered} \hline \text { Percentage } \\ \text { change } \\ (1976- \\ 2010) \\ \hline \end{gathered}$ |
| \$2 billion or more | 2 | 19 | 17 | 850.0\% | \$7,422, | \$102,888.8 | \$95,466.1 | 1286.1\% |
| \$1 to \$2 billion | 5 | 24 | 19 | 380.0\% | 6,660.2 | 35,179.4 | 28,519,2 | 428.2\% |
| \$500 million to \$1 bill | 11 | 31 | 20 | 181.8\% | 7,825.9 | 21,292, 3 | 13,466.4 | 172.1\% |
| \$ $\$ 50$ to $\$ 500$ million | 23 | 54 | 31 | 134.8\% | 8,458.2 | 20,013.7 | 11,555.5 | 136.\% |
| \$100 to \$250 million | 54 | 120 | 66 | 122.2\% | 8,766.3 | 18,464,5 | 9,698.2 | 110.6\% |
| \$ $\$ 0$ to \$100 million | 54 | 184 | 130 | 240.7\% | 3,969,9 | 12,714,7 | 8,744,9 | 220.3\% |
| \$25 to \$50 million | 70 | 221 | 151 | 215.7\% | 2,357.5 | 7,970.0 | 5,612,5 | 238.1\% |
| \$10 10 $\$ 25$ million | 143 | 297 | 154 | 107.7\% | 2,298.9 | 4,925.5 | 2,626,6 | 114.3\% |
| less than \$10 million | 841 | 594 | -247 | -29,4\% | 2,295.3 | 2,122,1 | -173.3 | -7.5\% |
| Total | 1,203 | 1,544 | 341 | 28.3\% | \$50,054.9 | \$225,571.0 | \$175,516.1 | 350.\% |
| \$ $\$ 50$ million and above | 41 | 128 | 87 | 212.2\% | 30,367.0 | 179,374.2 | 149,007,2 | 490.7\% |
| Less than \$ $\$ 250$ million | 1,162 | 1,416 | 254 | 21.9\% | 19,687.9 | 46,196.8 | 26,508.9 | 134.6\% |
| Less than 525 million | 984 | 891 | -93 | -9.5\% | 4,594.2 | 7,047.6 | 2,453.3 | 53.4\% |

Table 55 shows the results of Hirschey et al.'s study in terms of R\&D spending to total earnings as reported by the top 100 R\&D spenders, the next 100 largest, and so on. (Again, these figures do not include financial or utilities companies.)

Table 55 Earnings of R\&D Firms
(Hirschey et al, 2012)
Panel A: Earnings of R\&D Firms

| Ranking | Share of total earnings, R\&D spending industrials |  | Cumulative share of total earnings, R\&D spending industrials |  | Real earnings of R\&D spending industrials (2010 \$millions) |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 1976 | 2010 | 1976 | 2010 | 1976 | 2010 |
| top 100 | 70.1\% | 77.0\% | 70.1\% | 77.0\% | \$108,991.8 | \$346,495.6 |
| 101-200 | 14.9\% | 10.3\% | 85.1\% | 87.3\% | 23,201.6 | 46,464.3 |
| 201-300 | 5.2\% | 3.6\% | 90.3\% | 90.9\% | 8,140.4 | 16,083.6 |
| 301-400 | 2.8\% | 2.8\% | 93.1\% | 93.7\% | 4,391.9 | 12,602.8 |
| 401-500 | 2.0\% | 1.0\% | 95.1\% | 94.7\% | 3,066.4 | 4,672.4 |
| 501-600 | 1.2\% | 1.5\% | 96.3\% | 96.2\% | 1,863.4 | 6,684.5 |
| 601-700 | 1.1\% | 0.8\% | 97.4\% | 97.0\% | 1,678.3 | 3,484,0 |
| 701-800 | 0.9\% | 0.8\% | 98.3\% | 97.8\% | 1,413.8 | 3,672.7 |
| 801-900 | 0.8\% | 0.6\% | 99.1\% | 98.4\% | 1,260.3 | 2,925.3 |
| 901-1000 | 0.4\% | 0.2\% | 99.5\% | 98.7\% | 639.0 | 1,005.5 |
| 1001-1100 | 0.2\% | 0.1\% | 99.8\% | 98.8\% | 365.9 | 564.5 |
| 1101-1200 | 0.2\% | 0.8\% | 100.0\% | 99.6\% | 358.7 | 3,651,6 |
| 1201-1300 | 0.0\% | 0.2\% | 100.0\% | 99.8\% | 10.9 | 898.0 |
| 1301-1400 |  | 0.1\% |  | 99.9\% |  | 571.5 |
| 1401-1500 |  | 0.0\% |  | 99.9\% |  | 123.7 |
| 1501-1600 |  | 0.1\% |  | 100.0\% |  | 226.9 |
| Total for all firms | 100\% | 100\% | 100\% | 100\% | \$155,382.5 | \$450,127.0 |

## VI. 3 Summary of Hypothesis Results

Table 56 Summary of Hypothesis Results

|  | Correlation | Regression | Regression without <br> Revenue difference |
| :--- | :--- | :--- | :--- |
| Hypothesis 1 | Supported | Supported |  |
| Hypothesis 2 | Supported | Supported |  |
| Hypothesis 3 | Supported | Not Significant | Partially Supported |
| Hypothesis 4 | Supported | Not Significant | Supported |

Table 56 summarizes the hypothesis findings in the correlation and regression tests. Hypotheses 1 and 2 are supported by both correlation and regression. Hypotheses 3 and 4 are supported by correlation testing. During regression testing, when the revenue difference variable is removed, Hypothesis 4 is supported and Hypothesis 3 is partially supported.

## VI. 4 Contribution to Practice

This study finds a statistically significant relationship between the changes in R\&D spending and analysts' sales estimate forecasts. Although there is strong belief in the industry that $\mathrm{R} \& \mathrm{D}$ spending impacts company revenues, analysts recognize the benefits of R\&D differently based on a company's size and IT sector. Further, analysts update their sales and EPS estimates in different ways for the current and the next year. This research breaks down the relationship between the R\&D spending and analysts' sales estimate revisions and EPS estimate revisions using 20 years' worth of IT sector data. Practitioners, company leaders and chairmen, and chief executive officers (CEOs) can review this report when making decisions about R\&D spending. Previous studies have proven a strong correlation between analyst revisions of company revenue and earnings to stock price movement. Top managers at all firms are interested in increasing their stock price.. This research paper will help the CEOs and chief financial officers (CFOs) of IT companies make decisions about how to allocate R\&D expenditures for their company in a way that benefits them in both the short and long term.

As these research results show, a strong correlation exists between R\&D spending and analysts' sales estimates, and how the EPS revisions increase next year over current year based on R\&D spending. The economic significance of the estimates is large. I find that
analysts typically adjust their EPS estimates in a range of about 5-30 percent of the percentage change in $R \& D$ expenditures. So if $R \& D$ expenditures were to change by $50 \%$, for example, analyst EPS revisions would change by about 2.5 to 15 percent. This understanding is especially important for small companies, since they have fewer market participants who follow their stock.

Companies that are planning to diversify from one industry to another can also benefit from this research, including hardware companies that are planning to move into the software business. IBM, which is a primarily a hardware company, has moved into the software industry over the past two decades. Companies wanting to take the path of IBM or vice versa can review this report to get an overview of their R\&D allocation to different segments.

Finally, companies that are growing in market capitalization can benefit from this study, which can help them understand how R\&D spending impacts the analysts' estimates in sales and earnings.

## VI. 5 Contribution to Theory

My study fills a gap in the literature by examining how analysts use R\&D expenditures to adjust their sales and EPS estimates. In theory, increased R\&D should lead to increased sales, which should translate to increased earnings. Academic research has traditionally focused on R\&D spending and its impact on earnings. For example, Chan et al.'s study (2001) focused on R\&D spending's financial impact on high-tech companies. My paper is the first to shed light on how analysts perceive the linkage between R\&D, sales, and earnings. In so doing, I provide the bridge to other strands of the $\mathrm{R} \& \mathrm{D}$ and analyst literature. One of these strands is the relationship between C-level
executives' compensation and R\&D spending. Another is a line of literature that links analyst estimates and their impact on stock price.

While Keung (2010) has discussed the importance of analyst forecast sales and earnings forecast revisions, little research has been done on R\&D and analysts' estimates, and none has studied R\&D spending and its impact on analysts' sales and EPS estimates. This research helps to fill that academic gap by examining companies' $\& \& D$ expenditures in relation to these key analyst estimates.

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[^0]:    Notes: Regression of independent variables (R\&D percentage difference, gross margin, Log(Assets), Revenue difference, Long-term debt/sales) with Analyst EPS estimate percentage difference (dependent variables) filtered on Analyst EPS estimate $\%>\mathbf{- 1 0 0 \%}$ and Analyst EPS estimate $<\mathbf{1 0 0 \%}$.

