Corporate Sport Sponsorship and Stock Returns: Evidence from the NFL

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ABSTRACT

Most home stadiums/arenas of major-sport teams in the U.S. are sponsored by large publicly traded companies. Using NFL data we find that stock returns to the sponsoring firms are affected by the outcomes of games played in their stadiums. Wins in Monday night games generate next-day abnormal returns 50 basis points higher than losses. The effect is 80 basis points in the post-season and when the game outcome is unexpected. This does not revert over the next few days. Outcomes of NFL games could serve as a reasonably exogenous instrument for investor sentiment.

Keywords: Stock returns; Sport sponsorship

JEL Classifications: A12, G12, G14

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1. Introduction

Consider a major sport team that is hosting an important and highly rated televised match, at which it is expected to win. The match is played at the team's home stadium, which is named after the team's sponsor, a large and well-known publically traded corporation. The name of the sponsor is therefore repeatedly mentioned and seen during the match. Now suppose that the team is unexpectedly losing the match. Will this loss be reflected in the return on the sponsor's stock in the next trading day? If so, is it because the sponsor has been associated with the prevailing disappointment in the team? Or is the market reacting to the financial implications of the loss – a shorter season, fewer opportunities to promote the sponsor? Would the market have had a similar reverse reaction after a win? In this study, we document that stock prices of companies that sponsor National Football League (NFL) stadiums *are* affected by the outcomes of important individual games played in the stadiums. Our evidence suggests that this effect is partially driven by investor sentiment.

Several recent studies have explored the association between the outcomes of sport matches and stock market returns. Boyle and Walter (2002) find no relation between the success of the New Zealand national rugby team and the stock market reaction in the country. Ashton, Gerrard, and Hudson (2003) find a strong relationship between the performance of the English national soccer team and the change in the price of shares traded on the London stock exchange, where good (bad) performances by the national team are followed by positive (negative) market returns. Edmans, Garc'ia, and Norli (2007) conduct a cross-country analysis and find that losses in soccer (and other sports) matches have an economically and statistically significant negative effect on the losing country's stock market; yet they find no evidence of a corresponding effect after wins. Scholtens

and Peenstra (2009) analyze matches of eight publically traded European soccer teams. They find that the stock market response is significant and positive for victories and negative for defeats.

Our study is the first to examine the effect of professional sport match outcomes on stock returns of the teams' sponsors.¹ Although corporate sponsorship of professional sports stadia can be traced back to the early 1900s, the number of major league teams playing in corporate-named stadiums and arenas has sharply increased in the last two decades. As of the end of 2013, 62 percent of the home stadiums/arenas of the four major league sports (football, baseball, basketball, and hockey) were sponsored by publically traded companies.

Sponsoring a sport team is a major decision for a corporation. It is typically a long-term commitment that requires a significant investment. The average price for acquiring naming rights of a team's stadium in the U.S. National Football League in recent years is 120 million dollars for an average period of 17 years (see details in Table A1 in the appendix). In turn, the sponsoring company is provided an opportunity to tie the company's brands with a successful and popular sport organization, an opportunity to establish a strong relation with a large fan base and the local community, and a range of effective marketing tools – the sponsor typically gets branding and signage inside and outside the stadium, product placement rights within the stadium, exclusivity for use of its products by the team (official sponsor status), and has access to the team's coaches and players for promotions.²

¹ Extant stadium naming literature focuses on market reaction to the initiation of stadium sponsorship (see, for example, Clark, Cornwell, and Pruitt (2002), Becker-Olsen (2003), and Leeds, Leeds, and Pistolet (2007)).

² For example, the NFL team the Dallas Cowboys has recently signed a 25-year, 500 million dollar stadium naming rights deal with AT&T. As part of the deal, AT&T will continue to invest in improvements to wireless technology at the stadium, the city of Dallas will get 5 percent of revenue from the deal to help pay off the city's debt, and even the Legends Drive near the stadium will become AT&T Blvd. (DallasNews.com, July 25, 2013).

In light of these potential advantages of sport sponsorship, attaching the company's name to a sport team can also be risky. If the team does not perform well, it will likely suffer less exposure in the media, lower demand for the team's games and merchandises, and damaged reputation, all of which can affect the team's sponsoring company. The nature of competitive sports suggests that a single event or game can largely determine the success versus failure of the team, and thereby the value of its sponsoring company.³

In an efficient market therefore post-game stock returns to the sponsoring firms should reflect rational market reaction to the financial implications of the game outcomes. Yet there may also be a behavioral effect, which is driven by the unique ability of sport events to generate strong sentiment at the moment of time among large populations.⁴ Edmans et al. (2007) argue that national sport events (especially soccer matches) can produce substantial and correlated mood swings in a large proportion of a country's population, which is translated into the stock market movement in the country. Analyzing publically traded European soccer clubs, Bernile and Lyandres (2011) conclude that investors are overly optimistic about their teams' prospects ex ante and, on average, end up disappointed ex post, leading to negative post-game abnormal returns. This conclusion is consistent with the study of Brown and Hartzell (2001) on the impact of basketball game results on the stock price of the NYSE listed Boston Celtics Limited Partnership. Palomino, Renneboog, and Zhang (2009) find evidence that the abnormal returns for the winning British soccer teams do not reflect rational expectations but are high due to overreactions induced

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³ For example, an article in BusinessWeek observes: "BMW Oracle (ORCL) team sailboat, eliminated in May from the America's Cup qualifying competition in Valencia, Spain, before the main event even started. German press reports put the cost of the failed Cup bid at nearly \$200 million." (Jack Ewing, June 7, 2007).

⁴ Other effects on investor sentiment include sunshine (Saunders (1993) and Hirshleifer and Shumway (2003)), changes to and from daylight saving hours (Kamstra, Kramer, and Levi (2000)), amount of daylight across seasons (Kamstra, Kramer, and Levi (2003)), nonsecular holidays (Frieder and Subrahmanyam (2004)), temperature (Cao and Wei (2005)), lunar cycles (Yuan, Zheng, and Zhu (2006)), and aviation disasters (Kaplanski, and Levy (2010)). See Hirshleifer (2001) on psychological biases in asset pricing.

by investor sentiment. We ask therefore whether sudden change in investor mood as a result of game outcomes is translated into the market value of the teams' stadium sponsors. Because the outcomes of sport events are clear and decisive, we have a reasonable cause to assume that their effects aggregate across fans/investors, unlike more individual effects, such as prospect theory, loss aversion, skewness, or habit preferences.

We concentrate on the NFL for several reasons. First, football is the most popular sport in the U.S. and has been for many years by a wide margin. Football games attract the largest crowds and achieve the highest television ratings among all major sports. Second, a fairly large proportion of NFL stadiums are sponsored by publically traded companies, 21 out of total of 32 NFL teams (see details in Table A1). Third, the importance of a single game in the NFL is very high, relative to the other major sports in the U.S. (baseball, basketball, and hockey). This is because the NFL season is very short (16 games, compared to at least 82 games in the other sports). Furthermore, the NFL post-season (playoff) system is based on one game at each progression towards the championship, compared to a series of games (typically best-of-seven) in the other sports.

The NFL therefore provides a unique setting to assess the impact of outcomes of important and popular sport matches on the stock price of the sponsoring companies. We manually collected detailed data on all NFL games for teams with stadiums that are sponsored by publically traded companies. The sample contains 3,399 games (1,710 home games) during the pre-seasons, regular seasons, and post-seasons of 21 teams with 26 sponsoring companies, over the period 1997-2013.

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⁵ A 2009 ESPN Sports poll asked respondents to name their favorite spectator sport (defined as one in which the responder attends games or matches, watches them on TV, listens to them on the radio, or reads about them). The poll results showed that professional football is the most popular sport with 24.4%, where professional baseball is the second with 11.0%. The Harris Poll has recently released the results of its annual survey of the favorite sports of Americans; professional football was the most popular with 35%, followed by professional baseball with 14% (BusinessInsider.com, January 27, 2014). According to the NFL, during the fall of 2013, NFL games accounted for 34 of the 35 most-watched TV shows among all programming (Bloomberg.com, February 5, 2014).

We begin our analysis by looking at home games of NFL teams and comparing the next-day abnormal stock returns of their stadium sponsors after wins and losses. We measure abnormal returns using six different models. When all home games are included in the sample, the results do not indicate any abnormal return after wins or losses. This is not very surprising given that most games are played simultaneously with other games, are not played in prime time hours, are not nationally televised, and for the most part their outcome (i.e., the game winner) is expected. We therefore focus our examination on subgroups of games that attract the highest interest: regular season games that are played on Monday nights, post-season games, and games with unexpected outcome. Monday night games and post-season games attract high attention because they are stand-alone (no other games played at the same time), nationally broadcasted by a major television network, played at prime time hours (the Monday night games), and are critical for the team success (the post-season elimination games). An unexpected outcome of a game naturally attracts more media attention and also carries more "news" to the fans/investors, thus is likely to generate a stronger post-game stock market reaction. We classify a game outcome of a team as unexpected using two criteria: if the game outcome is contrary to the pre-game betting spread prediction or if it comes after a sequence of games with the opposite outcome (i.e., a loss after a series of wins).

The results indicate that games' outcomes affect the market value of the stadium sponsoring companies. Wins in Monday night games have an economically and statistically significant positive effect on sponsor stock price, an average of 0.51 percent across all models of abnormal return, whereas losses have virtually no effect (average of 0.01 percent). The difference however between the win and loss returns shows relatively weak statistical significance. For post-season games the difference is greater and somewhat more significant statistically; the losing teams' sponsors earn average abnormal return lower by 82 basis points than that of the winning teams'

sponsors, although the returns are negative both after wins and losses (-0.19 and -1.01 percent, respectively). This can be explained by the crucial effect a playoff game loss has on the team, as it eliminates the team from the playoff contention. This may also indicate a surprise effect, i.e., the home team is expected to win in the post-season. A playoff game is typically played at the stadium of the team with the better record; thus a win does not have much effect while a loss leads to a significant negative return of the home team's sponsor.

Games with unexpected outcome generate a similar economic effect to that of playoff games, but stronger statistically likely due to the larger sample size. The sponsor of the home team earns on average a positive abnormal return after wins and a negative abnormal return after losses, yielding a significant win-loss return difference of 81 basis points. This emphasizes the importance of the element of surprise in moving the stock price of the sponsoring companies.

We confirm the post-game abnormal returns using a pooled regression. We estimate abnormal returns for all sponsor-days in the sample (game and no-game days) and regress them against dummy variables indicating a first trading day after a win/loss of the sponsored teams. The coefficients of the win and loss indicators are typically positive and negative, respectively, where the differences between the coefficients are fairly similar to the cross-sectional differences in abnormal returns after wins and losses. Team-specific analysis shows a positive effect for the majority of the teams, where there is no clear relation between the magnitude of the effect and the local market size of the team.

We next explore the drivers for the effect of match outcomes on sponsors' stock returns. As discussed above, we distinguish between two mechanisms. On the one hand post-game returns to the sponsoring companies should reflect changes in expected cash flows due to the financial

implications of the game outcomes. On the other hand outcomes of sport events are associated with investor sentiment, which is often reflected in stock price movements.

We assess the presence of investor sentiment using two tests. First, we look at returns on the sponsors' stock after away games (games played at the opponent's stadium). Rational expectations suggest a similar effect of wins/losses in away games and home games, as both should have similar implications for the team's success (i.e., wins/losses count the same in home and away games) and thereby the exposure of its sponsor for the remaining of the season. Behavioral bias suggests a stronger effect for home games; this is because the stadium sponsoring company is only visible during home games, for which investors may associate the game outcome with the sponsor. The results generally show substantial reductions in sponsor abnormal returns after away games, which is consistent with investor sentiment.

Second, the effect of investor sentiment on the stock return in the first trading day after the game can also be associated with subsequent returns; i.e., an initial overreaction to game outcome is more likely to reverse in the following days. Our results however are mixed; while in certain cases the next-day abnormal return tends to reverse, other cases exhibit return continuation. Considering the results of the two tests, we conclude that investor sentiment plays a partial role in shaping the market value of the sponsoring firms after important games.

In the final part of the study we investigate whether the sponsors' post-game stock return patterns provide profit opportunities. We form a weekly zero-investment portfolio of buying the stocks of all sponsoring companies whose teams won that week and selling the stocks of all sponsoring companies whose teams lost that week. We hold this portfolio from the second trading day to the fifth trading day after the game (as the first-day profit cannot be earned). The portfolio generates abnormal profits, particularly for home games whose outcomes are hard to predict and

by nature attract more attention and provide a higher element of news: mean excess return and factor-model alphas of approximately 28 percent per NFL season.

The paper contributes to the literature in several aspects. To the best of our knowledge, this is the first study to highlight the sensitivity of sponsoring corporations' value to the outcome of individual sporting events. Although NFL stadium sponsors are typically among the largest and most well-known firms listed on the U.S. stock exchanges, the outcome of a single match played by their sponsored teams can lead to significant swings in sponsor firm stock prices. Second, we provide new evidence of the presence of investor sentiment in stock pricing. Identifying exogenous factors that can affect the mood among large populations is a key ingredient in studying investor sentiment. Prior studies use such factors typically at the aggregate level, for example: sunshine (Saunders (1993) and Hirshleifer and Shumway (2003)), temperature (Cao and Wei (2005)), and aviation disasters (Kaplanski, and Levy (2010)). We believe that the highly popular NFL games serve well as an exogenous factor that can generate massive mood swings. Furthermore, NFL games and stadium sponsorship allow for analysis at the firm level. Our results indicate that stock market reaction to game outcomes is driven, at least to a certain extent, by investor sentiment. This finding is consistent with prior studies attributing countrywide market price movement after sport matches to changes in investor mood (e.g., Edmans et al. (2007)), and also with the documented emotional reactions that NFL matches can illicit (see White (1989) and Chang et al. (2012)). Lastly, in a broader asset pricing view, we show that the effect of game outcome on sponsoring firm value embeds various profitable stock trading opportunities.

The paper proceeds as follows. Section 2 describes the data and game samples, Section 3 examines the effect of home game outcome on sponsoring firm return, Section 4 tests for the

presence of investor sentiment, Section 5 offers post-game stock trading opportunities, and Section 6 concludes.

2. Data and game samples

We manually collected data on all NFL games over the years for teams that are sponsored by publically traded companies. Using official NFL team websites and stadium websites, we identified teams who are or have engaged in stadium naming rights agreements with publicly traded firms and obtained key characteristics of the agreements. We then use team websites as well as secondary sports websites to gather data on game schedules across the sample period, including game date, location, score, television coverage, and more.⁶ Based on the game date, we identified the first subsequent day of stock market activity for the sponsoring firms. Sunshine Forecasts' database was used to identify historic betting spreads for each game. Our sample contains 3,399 games (1,710 home games) during the pre-season, regular season, and post-season, representing 21 NFL teams and 26 sponsoring companies over the period 1997-2013. Table A1 lists the sample teams, stadiums, and sponsoring companies. Table A2 shows the game distribution over the sample period.

We combine the NFL data with CRSP and Compustat to draw accounting variables and stock return data on the sponsoring companies. Table 1 presents descriptive statistics for the sponsoring companies and for all CRSP/Compustat firms over the same sample period. Not surprisingly, firms that sponsor home stadiums of NFL teams are typically much larger than the average firm, have less growth opportunities (indicated by higher book-to-market ratios), and higher leverage ratios.

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⁶ In addition to NFL.com, secondary websites included Sports Illustrated (sportsillustrated.cnn.com), ESPN (espn.go.com), and ProFootball Weekly (profootballweekly.com).

Stocks of sponsoring firms are highly traded and highly liquid; trading volume, Amihud's illiquidity measure, and bid-ask spread are all significantly different than those of the average firm. Sponsors' stocks are also less volatile (measured by the standard deviation of daily stock returns over a month) but with market beta somewhat higher than that of the full sample.

Considering the full sample of home games, which includes all pre-season, regular season, and post-season games, we do not expect a strong effect. Most NFL games are played simultaneously with other games, are not played in prime time hours, are not nationally televised, and do not attract particular post-game media attention unless their outcome is really unexpected. Hence, the stadium sponsoring companies are not visible to a large, national audience for a typical game. We therefore focus on several samples of home games that provide very high visibility and are highly important for the sponsoring companies.

The first sample contains all regular season games that are played on Monday night (92 home games). Monday night games are distinct. A game held on Monday night is the last game played in the NFL week (Thursday to Monday), receives exclusive game-day publicity at a national level, is always played in prime-time hours (typically at 8:30pm EST), and is nationally televised. In addition, Monday night games are usually chosen based on the importance and the general interest of the game. For many years 'Monday Night Football' has been one of the highest-rated television shows in the U.S. This means that sponsoring companies are more visible for games held on Monday night in their stadiums, and thus are more likely to be affected by the outcome of the games. Important to this study, there is also typically a 24 hour gap between the start time of the last game on Sunday and the Monday night game. This allows the market reaction measured to be isolated from the reaction to other NFL-week games.

Our second tested sample includes games that are also very visible, but much more important - the post-season (playoff) elimination games (57 home games). As with Monday night games, post-season games are stand-alone and are typically nationally broadcasted by a major television network, thus providing high visibility to the stadium's sponsoring firm. But more importantly, post-season games are the most meaningful games for the teams, as their outcome solely determines if the team will continue to compete for the championship (in case it wins) or will be eliminated from the competition (if it losses). The perception of a successful season versus a failed one is often determined by a single post-season game. The outcome of post-season games therefore can have a direct impact on the visibility of the sponsors in the rest of the season.

Our third and final sample consists of all games whose outcome is unexpected. These games are likely to get more media attention after the game, making the stadium sponsoring companies more visible. In addition, if a game win/loss is unexpected, it creates news of larger magnitude for the team and thereby for its sponsor. An unexpected win or loss can be viewed as any other corporate news that carries value for the company, and thus can have stronger stock market reaction. To determine whether a game outcome is expected or unexpected we use information obtained from pre-game betting spreads and prior team performance. A game outcome is considered expected if it is consistent with the sign of the betting spread or if it comes after a sequence of at least three games with the same outcome. In the same way, a game outcome is classified as unexpected if it is against the sign of the betting spread or if it comes after a sequence of at least three games with the opposite outcome. To eliminate trivial game predictions, we consider only spreads of at least 5 points. Game outcomes for which the two categories of betting spread and prior win/loss runs conflict are not considered as either expected or unexpected. And

⁷ The post-season sample does not include the Super Bowl game as it is played at a neutral stadium.

games with spreads lower than 5 points and without prior runs are considered as unpredictable ex ante. Our sample of interest, games with unexpected outcome, contains 278 home games.

Table A2 shows a fairly uniform distribution of the samples' games over the years, which provides a solid ground for our examination. That is, the results are not likely influenced by an unusual effect in a specific season, or by cross-sectional dependence induced by same-day clustering (see Brown and Warner (1985)). This is verified by robustness tests accounting for time-clustering.

3. Effect of home game outcome on sponsoring firm return

We investigate the effect of game outcome by estimating the abnormal return to the sponsoring companies in the first trading day after each game played in their stadium and comparing the abnormal returns after wins and losses. We verify the robustness of the results using pooled regression, wherein we calculate abnormal returns for all sponsor-days in the sample and regress them against variables indicating first-trading days after home game wins/losses. Details and results are below.

3.1 Calculating abnormal return

We begin our empirical analysis by estimating abnormal stock returns of the stadium sponsoring companies in the first trading day after the game (referred to as 'post-game day'). The firm's abnormal stock return is the difference between its raw return and its expected return for that day. To mitigate the sensitivity of the results to a specific model of expected return, we employ six different models that are commonly used in the literature (for detailed analyses of the models see Brown and Warner (1985) and Barber and Lyon (1997)). The first model is the *mean-adjusted model*: expected return is estimated by averaging the firm's raw returns during the past 250 trading

days prior to the game. The second is *market-adjusted model*: expected return is estimated by the value-weighted market index on the post-game day. The third is *market model*: expected return is estimated by the fitted value of the stock return on the post-game day, based on a regression of the sponsoring firm's raw return on the value-weighted market index return during the past 250 days prior to the game. The fourth is *factor model*: expected return is estimated in a similar way to the market model with a larger set of explanatory variables, including the Fama and French (1993) three factors, the momentum factor (all factor returns are downloaded from Kenneth French's website), dummy variables indicating January and Monday, and lagged return on the sponsor's stock. The fifth is reference portfolio: expected return is estimated by the equal-weighted average return of firms in a size/book-to-market portfolio that includes the sponsoring firm; portfolios are formed by first sorting all stocks into ten equal deciles according to the firm's size as of the beginning of the post-game day, and then within each decile, sorting all stocks into five equal book-to-market quintiles. The sixth is *matched (control) firm*: expected return is measured by the return of the firm with the closest book-to-market ratio within the same size decile as the sponsoring firm.

3.2 Post-game day abnormal return

In Table 2 we look first at the effect of "typical" NFL games, i.e., games that are comparably not highly visible or important at the onset, and that do not produce unexpected outcomes that attract media attention. We present the abnormal returns according to the six models described above for the game samples, as well as the results based on the average across all models. The *t*-statistic of the average abnormal return is thus calculated using the cross-sectional properties of

the average, which should capture the dependency structure across the individual models.⁸ All returns are reported in percent.

As expected, when all home games are included in the sample, the results do not indicate any abnormal return after wins or losses. In fact, the average abnormal return after a loss of the home team is higher by 10 basis points than that after a win, yet is not statistically significant. Outcomes of unpredictable games also do not lead to any significant effect on the sponsors' stock price, but interestingly, expected outcomes lead to a negative effect: a win-loss return difference of -0.60 percent with a *t*-statistic of -3.07.

The negative effect of expected outcomes may seem puzzling, i.e., why would expected wins lead to negative returns and expected losses to positive returns? A possible explanation for this result can be related to fans' behavior prior to sport matches. Several studies argue that fans often tend to overreact to the expected performance of their teams in certain events (see, e.g., Krizan and Windschitl (2007) and Bernile and Lyandres (2011)). This means that fans can be overly optimistic for a game that their team expects to win and overly pessimistic if the team expects to lose. Because a game with an expected outcome should not have a material effect on the sponsor stock price, a post-game correction to pre-game overreaction will lead by itself to a negative effect of the game outcome.

Turning our focus to highly visible and important games, the results show that game outcome has a positive effect on sponsor stock price (reported in Table 3). After Monday night games, the home team's sponsor earns a positive abnormal return if the team had won, an average of 0.51

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⁸ For robustness we also applied Bayesian model averaging on the six models (see Avramov (2002)). Assuming a prior uniform distribution of the validity of the different models, we regressed daily raw returns on the models' benchmarks of expected return, calculated the posterior probabilities that each benchmark is the correct one given the data, and used these posterior probabilities as weights for averaging the models' abnormal returns. The Bayesian averaging yields almost identical results as the equal-weighted averaging (not reported).

percent with a *t*-statistic of 2.18, and no abnormal return if the team had lost (average of 0.01 percent). Although the average difference between the win and loss post-game day abnormal returns is large in economic terms, its *t*-statistic is only 1.40. This is partially due to the relatively small sample of Monday night games (92 home games).

For post-season games the impact is larger economically and somewhat more significant statistically. Averaging across all models, the losing teams' sponsors earn an abnormal return lower by 0.82 percent than that of the winning teams' sponsors (t-statistic of 1.72). All six abnormal return models yield meaningful effects (differences in win-loss abnormal returns of 58 to 127 basis points). Unlike the Monday night games, the effect of playoff games seems driven by negative returns, as the average post-game day abnormal return is negative both after wins and losses (-0.19 and -1.01 percent, respectively). This can be explained by the crucial effect a playoff game loss has on the team. A playoff loss eliminates the team from the playoff contention, effectively ending the team's season. A playoff win only guarantees to keep the team in the competition for one more game. While wins and losses in the regular season serve to seed the team for playoff matchups, wins and losses in the post-season ultimately determine the success or failure of the team's entire season. The post-season results may also indicate that there is a surprise effect. As a playoff game is typically played at the stadium of the team with the better record, there is a greater expectation that the home team will win in a post-season game; thus a win does not have much effect while a loss leads to a significant negative return to the home team's sponsor.9

Games with unexpected outcome show a similar impact to that of the post-season games, but results are more significant statistically, likely due to the larger sample size. The sponsor of the

⁹ Note that although post-season games are played during the weekend, their post-game day negative abnormal returns are not driven by the weekend effect (see, French (1980)). This is because for most models of abnormal return, the expected returns are estimated from Monday returns as well.

home team earns on average a positive abnormal return if the team had won and a negative return after a loss, yielding a win-loss post-game day return difference of 0.81 percent with a *t*-statistic of 2.50. This result indicates the importance of the element of surprise in moving the stock price of the sponsoring company.

The mean differences in abnormal returns after wins and losses reported in Table 3 are meaningful. Clark et al. (2002) find that the sponsorship agreement announcement increases the sponsor's stock price by 1.65 percent on average in the four major sports in the U.S. We show that the outcome of a single game affects the sponsor's stock by an average return of 0.50 to 0.82 across the samples of Monday night, post-season, and unexpected outcome games. These effects are also comparable and even stronger than those reported in Edmans et al. (2007). They find that at the national level a loss in the soccer World Cup elimination stage leads to a next-day abnormal stock return of -0.49 percent, whereas a win does not lead to a significant positive return.

3.3 Team analysis and market size

An interesting aspect of the potential impact of game outcomes is the size of the media market of the team. An argument could be made for a stronger effect in various markets. The effect could be stronger in smaller markets, where the fans are generational, teams are more involved in the community, and media coverage is more likely to be supportive. Yet areas with a lower population density may have a weaker influence on the stock market and generally have fewer media channels to report on team activities and mention stadium sponsors. The effect could be stronger in larger markets. These markets have a greater population density and more media outlets/reporters. Yet large markets usually have multiple sports teams, creating a dilutive effect to game and media coverage.

To examine the effect of the local market size, we look separately at each individual team. Because many teams have only a few observations, we combine for each team the three groups of Monday night games, post-season games, and games with unexpected outcome, and we include only teams with at least five wins and five losses across these groups. This provides us with a total of 15 teams. Figure 1 shows the team-specific mean difference in the sponsoring companies' abnormal stock returns after home game wins and losses, averaged across the six models of abnormal return. The teams are ordered by the size of their media market, measured by population in TV households within a 75-mile radius of the team's stadium.¹⁰

The figure indicates that home game outcome has a positive effect on sponsor stock price for the majority of the teams, reinforcing the general results: nine teams show a large positive effect (win-loss returns between 0.6 and 2.0 percent); three teams show a positive, but low effect; two teams show a low negative effect; and one team shows a negative effect of -0.53 percent. The impact of the game outcome however is not clearly related with the local market size of the team. Teams located in the middle of the scale seem to have relatively small effects, and the two teams with the largest markets in the sample (New England Patriots and Oakland Raiders) do not exhibit any significant effect. This is not surprising given the aforementioned opposing effects of market size.

As a final note, we acknowledge that New York is a special case. Both the Giants and the Jets play in MetLife Stadium, which is close to a market epicenter and a high population density. In addition, the stock market is located in New York thus local sports fans may be more involved in stock trading. The New York market and MetLife Stadium also have the highest exposure to dilution as eight major sports teams claim affiliation with the city. Unfortunately, the MetLife

¹⁰ The NFL defines a team's "local" market as all the TV markets that lie within a 75-mile radius of the stadium.

sponsorship started in August of 2011, providing us with too few observations for a city-specific analysis at this time. Perhaps in the future, these data could provide a unique opportunity to examine investment behavior of regionally specific populations.

3.4 Robustness test: pooled regression

To confirm the impact of game outcomes on stadium sponsors' stock returns documented above we employ a pooled time-series regression (for similar specifications see Edmans et al. (2007) and Bernile and Lyandres (2011)). Using the six models of expected return described above, we estimate the abnormal return for each of the sponsor-days in the sample. For the full sample of sponsor-days (game and no-game days), we run a pooled regression of the six-model average abnormal return on two dummy variables: *Win* indicates a first trading day after a home game win of the sponsored team, and *Loss* is the equivalent indicator for a home loss of the sponsored team. The loadings on these dummy variables thus capture the marginal effects of win and loss on the sponsoring company's stock price.

This regression offers the advantage of exploiting the full time-series of the sponsors' stock returns, it allows estimating the effects of wins and losses independently, and it controls for potential systematic biases in the estimates of expected returns. For example, if a specific sponsor is relatively small in its reference portfolio, then a positive abnormal return to this sponsor's stock after a game win can be partially attributed to its relatively small size. Comparing the sponsor's post-game day abnormal return to its time series of abnormal returns that are estimated in the same way should eliminate this size effect.

The regression model also allows us to address a potential effect of cross-sectional dependence. Our samples generally show a lower level of time-clustering. The playoff sample consists of 57 games that are played over 42 different weekends, where no more than three games are held in any

of the weekends. The sample of Monday night games has almost no clustering because there is usually a single Monday night game every week, and the sample of games with unexpected outcome is not highly clustered as well (278 games played over 184 different weekends). Yet, as prior studies show that even relatively small cross-correlation can lead to over-rejection of the null hypothesis (see e.g., Kolari and Pynnönen (2010)), we examine the robustness of the results to game-day clustering. In addition to the OLS regression we estimate a Generalized Least Squares (GLS) regression that accounts for group-wise clustering by trading day.

The OLS regression estimates reported in Panel A of Table 4 are consistent with the abnormal return averages in Table 3. As the results above, Monday night games show a strong win effect and no significant loss effect (the *t*-statistics of the win and loss coefficients are 2.37 and 0.08, respectively), whereas post-season game effect is driven solely by home team losses (*t*-statistics of win and loss of -0.54 and -2.26, respectively). For games with unexpected outcome, a win has a strong positive effect (*t*-statistic of 3.03), where a loss has a negative effect but not as significant (*t*-statistics of -1.22).

Moreover, across the three samples, the differences between the coefficients of the win and loss dummy variables, which represent the marginal influence of the game outcome on the sponsor's stock return, are very similar to the win-loss average abnormal return differences reported in Table 3; difference in coefficients of 0.50 to 0.84 percent (with *p*-values of 0.06 and lower), compared with win-loss abnormal return difference of 0.50 to 0.82 percent. Finally, the GLS regressions reported in Panel B show similar and even slightly more significant results than the OLS regressions, confirming that the *t*-statistics are not inflated by time-clustering.

The results in this section consistently indicate that the market values of NFL sponsors are affected directly by the outcomes of meaningful games played in their stadiums.

4. Testing for investor sentiment

The findings in Section 3 raise the question regarding the mechanism driving the impact of home game outcomes on the sponsoring firms' stock price. On the one hand, outcomes of important games can have real financial implications for the sponsoring companies. A team that just lost an important regular season game that affects its chances to compete for the championship in the current season will naturally attract less attention in the remainder of the season. And a team that just lost a post-season elimination game has also ended the NFL-affiliated media mentions of their stadium, effectively ending the sponsor's naming rights campaign until the next game is played in the stadium. These losses mean reduced media coverage, TV ratings, home game attendance, and demand for team products – all of which should affect the sponsoring firm's future cash flows. A reverse effect is expected had the firm won the game. A rational expectations argument suggests therefore that the post-game day change in the sponsor's stock price reflects changes in expected cash flows.

On the other hand, outcomes of sport events are correlated with sudden change in investor mood, which is often reflected in stock price movements (see Boyle and Walter (2002) and Edmans et al. (2007)). This argument can apply specifically to the NFL due to its high popularity and the strong emotions it generates. For example, Chang et al. (2012) find that game outcome of NFL teams affects the value of locally headquartered stocks, where White (1989) documents that elimination from NFL playoff games leads to a significant increase in homicides in the cities following the games. The question is whether the ability of NFL games to generate strong sentiment among the teams' fans lead to stock market overreaction for the sponsoring companies attached to the teams.

Note that we address the behavior of the stock investor and not that of the consumer. One can argue that outcomes of NFL games can create consumer sentiment, resulting in over-consumption of the products of the sponsoring companies. However, if the sponsor stock price increases because investors expect over-consumption, this would qualify as rational stock trading behavior because this consumption builds real value to the sponsor.

The results so far can provide some indication for whether the post-game day abnormal stock return to the sponsoring firm is driven by rational expectations or investor sentiment. For example, the evidence that Monday night home games affect sponsors' return, but not necessarily any regular season home game, is more consistent with the investor sentiment argument. This is because the main difference between Monday night games and all other regular season games is visibility, not the level of importance. That is, if two games are equally important for their teams, they should create the same impact on the sponsoring firms' value. But if the more visible game generates a stronger impact, it is likely driven by correlated change in fan sentiment towards the team and thereby its sponsor. In contrast, the evidence that outcomes of post-season elimination games create stronger impact than those of Monday night games is consistent with the rational expectations argument. This is because both types of games are very visible, but the post-season games are, on average, much more important than Monday night games (and any other regular season games).

To further explore the presence of investor sentiment, we look at whether there is an effect also in away games (games played at the opponent's stadium) and whether the next-day sponsor return tends to reverse in the following days.

4.1 Effect of away games

Rational expectations suggest a similar effect of wins/losses in away games and home games, as both should have similar implications for the team and the sponsor for the remaining of the season (i.e., wins/losses have the same count in home and away games). Behavioral bias suggests a stronger effect for home games than for away games; this is because the stadium sponsoring company is only visible during home games, and thus investors may associate the game outcome with the sponsor.

We replicate the main results of the prior tables for away games, and summarize the comparison of home/away games in Figure 2. Both mean difference and pooled regression results indicate that Monday night games and games with unexpected outcome that are played away lead to much weaker and insignificant abnormal return on the sponsors' stocks; an effect of 19 to 22 basis points after away games, compared to 50 to 84 basis points after home games. These results strongly support the role of investor sentiment.

In the post-season, however, the effect of away games is more similar to that of home games. The average difference in abnormal returns between wins and losses is actually higher in away games than in home games (0.87 percent compared with 0.82 percent). The difference between the pooled regression coefficients of the win and loss indicators is reduced from 0.82 to 0.52 percent in away games. This may suggest that the effect of playoff games on the teams' sponsors is driven mostly by rational expectations and only to a minor degree by investor sentiment.

Figure 2 captures much of the message of the findings so far: outcomes of important home games affect the market value of the team stadiums' sponsors; away games generate a significantly weaker effect for the most part, leading to the conclusion that abnormal returns to the sponsors' stocks are driven at least to a certain extent by investor sentiment.

4.2 Subsequent return reversal

We examine whether the post-game day abnormal return tends to reverse in the following days, as evidence of initial overreaction to game outcome. We first estimate the cumulative average abnormal return during the four days following the first post-game trading day (days 2 to 5). We choose a window of four days as it gives sufficient time for return reversal, yet is not affected by the next game outcome (NFL games are typically played once a week). In addition, we estimate a cross-sectional regression of the cumulative abnormal return in days 2 to 5 on the first trading day abnormal return. A negative slope coefficient would suggest a return reversal, where a slope coefficient equals to -1 would indicate a complete reversal.

The results in Panel A of Table 5 do not indicate any evidence for return reversal. For playoff games, the difference between abnormal returns after wins and losses continues to be positive in the subsequent days; a cumulative abnormal return of 3.0 percent in days 2 to 5 with a *t*-statistics of 1.28. For the other samples the cumulative win-loss abnormal return difference in the subsequent days is very low and insignificant.

The cross-sectional regression in Panel B however yields mixed results. An indication for partial return reversal appears for games with unexpected outcome, where the slope coefficient is negative and significant for the sample of game wins. The post-season games show weak evidence for return reversal after wins, yet a strong evidence for return continuation after losses. After Monday night games the results indicate mainly return continuation. While these results do not support an initial overreaction for the most part, we cannot rule out the presence of behavioral biases, as almost all cases exhibit significant relations (positive or negative) between the next-day and subsequent days returns.

Considering the results of the two tests distinguishing between rational market reaction and investor sentiment (home/away games and subsequent returns), we can only conclude that investor sentiment plays a partial role in shaping the market value of the sponsoring firms after important games.

Attempts to conduct additional tests to explore the presence of investor sentiment do not yield definitive results. For example, prior studies have argued that investor sentiment is correlated with trading characteristics, including the timing and volume of trading, the type of traders (institutional vs. individual), the extent of order imbalance, and others. Building on these studies, we examine whether the effect of home game outcome on the stadium sponsor market value is associated with unusual levels of several trading variables. Our results provide only marginal evidence of behavioral trading after home games (not reported).

4.3 Discussion

Investor sentiment has been discussed theoretically and explored empirically in the literature mostly at the aggregate or market level; Saunders (1993), Hirshleifer and Shumway (2003), Kamstra, Kramer, and Levi (2000, 2003), Baker and Wurgler (2007), Edmans et al. (2007) and Kaplanski and Levy (2010) to name only a few. We believe that extending marketwide or countrywide analyses such as in Edmans et al. (2007) to the firm-level can enhance our understanding of investor sentiment.

First, analyzing cross-sectional variation in sentiment among stocks traded in the same market eliminates marketwide or countrywide sensitivity to sentiment. For instance, stock markets around the world are likely to exhibit different levels of market efficiency, and thereby have different exposures to investor sentiment. Moreover, sentiment can affect individual stocks in different ways (see, e.g., Baker and Wurgler (2006)).

In addition, and perhaps most importantly, we explore the presence of investor sentiment in a unique setting that puts together two extreme sides of the scale of sentiment. On the one hand there are firms that are least likely to be affected by behavioral biases, namely the sponsoring companies. These companies are among the largest companies traded in the U.S. stock market, they are highly liquid, highly traded, highly visible, and are less affected by market frictions (e.g., Baker and Wurgler (2006) suggest that sentiment is more likely to affect small and young companies with relatively low levels of asset tangibility and profitability). On the other hand, due to sport stadium sponsorship agreements, these firms are directly linked to frequent events that generate strong emotions at the moment of time among large populations, especially the highly popular NFL games. Having this contrast, our findings indicate that investor mood and emotions do affect the market values of the sponsoring firms. We show therefore that investor sentiment is not limited only to small and less popular stocks, it can also affect the largest and most transparent stocks on the exchange.

It is important to acknowledge however that although this study presents new aspects of investor sentiment, the magnitude of the impact of sentiment as well as the mechanisms triggering sentiment, especially across firms, are far from being completely understood. For example, while post-game trading behavior and the reduced effect in away games indicate behavioral biases, the patterns of subsequent returns are quite puzzling. Specifically, next-day abnormal return tends to reverse for certain samples, while other samples exhibit return continuation. This can raise several questions; for example, does return continuation reflect initial underreaction to true value generated by the game outcome, or continued overreaction driven by change in investor mood? More generally, is the impact of investor sentiment on stock prices as a result of isolated events limited to one or only a few trading days, or reflected in longer horizons? We do not provide

answers to these questions in this study. We do however explore in the next section whether these unusual patterns in subsequent returns embed profitable trading opportunities, which are often used as evidence for stock mispricing/behavioral bias.

5. Post-game trading opportunities

The findings above suggest that the market reaction to game outcome is not limited to the next trading day, but rather seems, at least for some of the samples, to continue in the following days. Assuming that the next-day return cannot be realized by investors, we examine whether one can earn abnormal profits by trading sponsors' stocks in the subsequent days.

Because profitable trading rules require a sufficiently large number of traded securities at the moment of time, we do not limit our examination to the game samples analyzed in the previous section as they cannot provide more than one or only a few games in a given week. We rather consider two large groups of games. The first group contains all home games, capturing the general impact of games played in sponsored stadiums. The second group contains the home games whose outcome is most unpredictable ex ante. The motivation for this distinction is that any outcome in unpredictable games, win or loss, provides some element of news that can lead to sharper stock market reaction. We therefore include in this group only games with very low betting spreads, 3 points and lower in absolute value.¹¹

We examine the performance of the following long-short investment strategy. Every week during the NFL season we form a value-weighted portfolio of buying the stocks of all sponsoring companies whose teams won that week and selling the stocks of all sponsoring companies whose teams lost that week. We hold this portfolio from the second trading day after the game until the

¹¹ A cutoff of 3 is natural as it represents the points awarded for a field goal in an NFL game.

fifth trading day. Table 6 shows the portfolios' mean excess weekly returns (in excess of the risk-free rate) and alphas from factor models over the years 1997-2013. The CAPM one-factor model uses the market factor. The three factors in the 3-factor model are the Fama and French (1993) factors. The 4- and 5-factor models include the Fama-French factors augmented with momentum and reversal factors. All returns and alphas are in percent per week.

When applied to all home games, the portfolio shows some evidence of abnormal profit; the mean weekly excess return is around 0.4 percent, and factor-model alphas are slightly higher, where the *t*-statistics are 1.70 to 1.83. The results are very strong when the investment is applied to most unpredictable home games. The mean excess return and factor-model alphas are 1.45 to 1.52 percent per week with *t*-statistics between 3.23 and 3.34. Considering that a regular NFL season runs for 17 weeks, this investment strategy yields an abnormal profit of approximately 28 percent per season. Abnormal returns following home games, especially unpredictable ones, can thus be translated into profitable trading opportunities.

To verify the importance of game outcomes for the success of NFL sponsor trading, we examine if one can achieve abnormal profits by trading sponsors' stocks prior to the game. Every Friday during the NFL season we form a value-weighted portfolio of buying the stocks of all sponsoring companies whose teams expect to win over the weekend and selling the stocks of all sponsoring companies whose teams expect to lose. We hold the portfolio until the next Friday. This pre-game trading strategy however does not generate positive profits (results are not reported), as opposed to the post-game strategy.

The abnormal profit generated by the winning-losing trading strategy is informative in several aspects. First, the magnitude of the profit is comparable and even higher than those generated by the most puzzling return patterns, such as momentum (see Jegadeesh and Titman (1993)) and

distress effect (see Dichev (1998); Campbell, Hilscher, and Szilagyi (2008)). Second, while we do not have a convincing explanation for this abnormal profit, we acknowledge that it can be driven by behavioral bias. This can provide further indication for the presence of investor sentiment in the sponsors' stock price movements.

Lastly, an important aspect of the post-game effect is that it involves only very large companies, the teams' stadium sponsors. These companies are typically highly traded, highly liquid, have less information asymmetry, and are strong financially. This means that the portfolios' profits document above are not likely driven by market frictions and are easier to implement, compared to other pricing effects. For example, Griffin and Lemmon (2002) document that the value anomaly is stronger among distressed stocks; Avramov et al. (2007) point to momentum being present mostly among low credit rating stocks; and Eisdorfer (2008) shows that approximately 40 percent of the momentum profit is generated by delisting returns. The NFL teams' stadium sponsors are not likely to default or be delisted from the stock exchange. In addition, the profits generated by anomalous stock return portfolios typically require massive short selling, which is not always feasible for many stocks (e.g., low liquid stocks, distressed stocks). In that respect, therefore, it is fairly easy to form the long-short NFL sponsors' portfolios.

6. Conclusions

We document that stock prices of companies that sponsor NFL stadiums are affected by the outcomes of important games played in the stadiums. Employing six different models of abnormal return shows that the mean difference between next-day return after a win and after a loss of the home team is 50 to 82 basis points. Pooled regressions exploiting the full time-series of the sponsors' stock returns yield very similar estimates.

We explore whether the post-game day abnormal returns represent rational changes in expectations of future cash flows, or overreaction by investors associating the team's performance with its sponsoring company. We find that the effect of game outcomes on sponsors' stock prices is significantly reduced after away games, which is consistent with investor sentiment. We do not find however a clear evidence of return reversal that is assumed to correct for initial overreaction. We conclude that the impact of home game outcomes is partially driven by investor sentiment.

The post-game return patterns provide profit opportunities. We form a weekly zero-investment portfolio of buying the stocks of all sponsoring companies whose teams won that week and selling all sponsoring companies whose teams lost that week. Applying this investment strategy to home games considered as unpredictable ex ante (using pre-game betting spread data) generates mean excess return and factor-model alphas of approximately 28 percent per NFL season.

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Table 1. Descriptive statistics

The table presents descriptive statistics for the sample of NFL stadium sponsoring firms and for all CRSP/Compustat firms over the period 1997-2013. For all variables, observations outside the top and the bottom percentiles are excluded. Size is market equity value (in millions of dollars). Book-to-market is book equity value divided by market equity value. Leverage is the ratio of book value of total debt to book value of total assets. Daily stock return is the daily return over the sample period, and Stdev of stock returns is the standard deviation of the daily stock returns in a calendar month (both are reported in percent). Market beta is measured by regression of stock return on market return over the past 60 months. Trading volume is the monthly volume over the sample period (in thousands). Amihud's illiquidity is the monthly average of daily ratios of absolute return to dollar trading volume (in millions). Bid-ask spread is the difference between the stock's closing ask price and bid price, divided by the average bid and ask prices (reported in percent).

Interpretation: Firms that sponsor stadiums of NFL teams are larger, have less growth opportunities, and are more leveraged than the average firm. Sponsors' stocks are also highly liquid and highly traded, and are less volatile.

| | NFL s | oonsors (26 | ifirms) | Full sam | Full sample (18,426 firms | | | |
|------------------------|----------|-------------|----------|----------|---------------------------|---------|--|--|
| | Mean | Median | Stdev | Mean | Median | Stdev | | |
| Size | 27,740.6 | 12,919.7 | 40,677.8 | 1,786.2 | 207.9 | 5,485.3 | | |
| Book-to-market ratio | 0.667 | 0.417 | 1.028 | 1.023 | 0.541 | 1.497 | | |
| Leverage ratio | 0.295 | 0.311 | 0.197 | 0.217 | 0.166 | 0.214 | | |
| Daily stock return | 0.032 | 0.000 | 2.508 | -0.120 | 0.000 | 3.781 | | |
| Stdev of stock returns | 2.115 | 1.775 | 1.268 | 2.949 | 2.438 | 2.021 | | |
| Market beta | 1.151 | 0.988 | 0.757 | 1.036 | 0.905 | 0.895 | | |
| Trading volume | 138.27 | 38.46 | 389.41 | 10.80 | 1.45 | 29.10 | | |
| Amihud's illiquidity | 0.049 | 0.004 | 0.185 | 11.475 | 0.320 | 43.722 | | |
| Bid-ask spread | 0.430 | 0.116 | 0.712 | 2.101 | 0.714 | 3.457 | | |

Table 2. Post-home game day abnormal returns: Typical games

The table presents the average abnormal stock returns of NFL stadium sponsoring companies in the first trading day after a home game of their sponsored teams over the period 1997-2013. Abnormal returns are presented for all home games and separately for wins and for losses of the home teams, within three samples. The first sample contains all games (including pre-season, regular season, and post-season games). The second sample contains all games with expected outcome; a game outcome is classified as expected if it either meets the prediction of a pre-game betting spread of at least 5 points or comes after at least three consecutive games with the same outcome. The third sample contains all games whose outcomes are unpredictable ex ante; these includes games with betting spreads smaller than 5 points and with no prior runs of at least three wins or losses. Abnormal return is the difference between the raw return and the expected return, as measured by the six models described in Section 3.1. The table also reports the results based on the average of all models. The *t*-statistic of the average abnormal return is calculated using the cross-sectional properties of the average. All returns are in percent and *t*-statistics are in parentheses.

Interpretation: For NFL home games that are not highly visible or that do not produce unexpected outcomes, wins (losses) do not increase (decrease) the market value of the sponsoring companies. The samples of all games and unpredictable games do not exhibit any significant return after wins or losses, whereas games with expected outcome generate a negative effect of 60 basis points (*t*-statistic of -3.07).

| | | Model of abnormal return | | | | | | | |
|---------------|-----------|--------------------------|----------|---------|---------|-----------|---------|------------|--|
| | | Mean- | Market- | Market | Factor | Reference | Matched | Average | |
| | N | adjusted | adjusted | model | model | portfolio | firm | Abn return | |
| All games | | | | | | | | | |
| Wins | 1,019 | -0.014 | -0.029 | -0.042 | -0.041 | -0.005 | -0.028 | -0.027 | |
| | | (-0.16) | (-0.39) | (-0.63) | (-0.74) | (-0.08) | (-0.29) | (-0.40) | |
| Losses | 691 | 0.108 | 0.127 | 0.073 | -0.013 | 0.099 | 0.014 | 0.068 | |
| | | (0.85) | (1.33) | (0.82) | (-0.18) | (1.11) | (0.11) | (0.77) | |
| Win-Loss | | -0.123 | -0.156 | -0.116 | -0.028 | -0.105 | -0.042 | -0.095 | |
| | | (-0.80) | (-1.31) | (-1.05) | (-0.31) | (-0.93) | (-0.27) | (-0.87) | |
| Expected gam | ne outcom | ne | | | | | | | |
| Wins | 472 | -0.176 | -0.197 | -0.178 | -0.090 | -0.170 | -0.264 | -0.179 | |
| | | (-1.41) | (-2.18) | (-2.33) | (-1.16) | (-2.00) | (-1.95) | (-2.13) | |
| Losses | 183 | 0.456 | 0.498 | 0.431 | 0.223 | 0.556 | 0.363 | 0.421 | |
| | | (1.52) | (2.00) | (1.77) | (1.32) | (2.36) | (1.28) | (1.85) | |
| Win-Loss | | -0.633 | -0.695 | -0.609 | -0.313 | -0.726 | -0.627 | -0.600 | |
| | | (-2.31) | (-3.28) | (-3.13) | (-1.92) | (-3.62) | (-2.23) | (-3.07) | |
| Unpredictable | games | | | | | | | | |
| Wins | 396 | -0.106 | -0.057 | -0.078 | -0.068 | -0.023 | 0.102 | -0.039 | |
| | | (-0.77) | (-0.52) | (-0.75) | (-0.77) | (-0.22) | (0.73) | (-0.38) | |
| Losses | 338 | 0.023 | 0.077 | 0.024 | -0.011 | 0.029 | -0.003 | 0.023 | |
| | | (0.14) | (0.70) | (0.24) | (-0.12) | (0.28) | (-0.02) | (0.22) | |
| Win-Loss | | -0.130 | -0.134 | -0.102 | -0.057 | -0.053 | 0.106 | -0.062 | |
| | | (-0.62) | (-0.85) | (-0.70) | (-0.44) | (-0.35) | (0.50) | (-0.41) | |

Table 3. Post-home game day abnormal returns: Highly visible and important games

The table presents the average abnormal stock returns of NFL stadium sponsoring companies in the first trading day after a home game of their sponsored teams over the period 1997-2013. Abnormal returns are presented for all home games and separately for wins and for losses of the home teams, within three samples. The first sample contains all regular season games played on Monday night. The second sample contains all post-season (playoff) games. The third sample contains all games with unexpected outcome; a game outcome is classified as unexpected if it is either against the prediction of a pre-game betting spread of at least 5 points or comes after at least three consecutive games with the opposite outcome. Abnormal return is the difference between the raw return and the expected return, as measured by the six models described in Section 3.1. The table also reports the results based on the average of all models. The *t*-statistic of the average abnormal return is calculated using the cross-sectional properties of the average. All returns are in percent and *t*-statistics are in parentheses.

Interpretation: The outcomes of highly visible and important NFL home games affect the stock price of the stadium sponsoring companies. Wins in Monday night games have a significant positive effect on sponsor stock price, an average of 0.51 percent across all models of abnormal return, whereas losses have virtually no effect (average of 0.01 percent). The difference however between the win and loss returns shows weak statistical significance (*t*-statistic of 1.40). Post-season games generate a larger win-loss return difference of 82 basis points (*t*-statistic of 1.72), and similarly, games with unexpected outcome yield a return difference of 81 basis points (*t*-statistic of 2.50).

| | | | Model of abnormal return | | | | | | | |
|---------------|----------|----------|--------------------------|---------|---------|-----------|---------|------------|--|--|
| | | Mean- | Market- | Market | Factor | Reference | Matched | Average | | |
| | N | adjusted | adjusted | model | model | portfolio | firm | Abn return | | |
| Monday night | games | | | | | | | | | |
| Wins | 52 | 0.866 | 0.552 | 0.211 | 0.353 | 0.537 | 0.530 | 0.508 | | |
| | | (2.07) | (2.10) | (0.97) | (1.82) | (2.20) | (1.55) | (2.18) | | |
| Losses | 40 | 0.120 | -0.028 | 0.071 | 0.079 | -0.081 | -0.093 | 0.011 | | |
| | | (0.33) | (-0.10) | (0.29) | (0.32) | (-0.31) | (-0.21) | (0.04) | | |
| Win-Loss | | 0.746 | 0.580 | 0.140 | 0.274 | 0.618 | 0.623 | 0.497 | | |
| | | (1.31) | (1.52) | (0.43) | (0.89) | (1.71) | (1.13) | (1.40) | | |
| Post-season g | ames | | | | | | | | | |
| Wins | 39 | -0.353 | -0.198 | -0.284 | -0.246 | -0.241 | 0.188 | -0.189 | | |
| | | (-1.48) | (-0.80) | (-1.27) | (-1.01) | (-1.11) | (0.67) | (-0.90) | | |
| Losses | 18 | -1.290 | -1.090 | -0.930 | -0.839 | -0.817 | -1.080 | -1.010 | | |
| | | (-1.63) | (-1.68) | (-1.77) | (-2.29) | (-1.57) | (-1.39) | (-1.86) | | |
| Win-Loss | | 0.937 | 0.888 | 0.646 | 0.593 | 0.576 | 1.270 | 0.819 | | |
| | | (1.47) | (1.56) | (1.34) | (1.36) | (1.21) | (1.90) | (1.72) | | |
| Unexpected ga | ame outo | come | | | | | | | | |
| Wins | 124 | 0.821 | 0.679 | 0.556 | 0.222 | 0.663 | 0.540 | 0.580 | | |
| | | (2.24) | (2.05) | (1.71) | (1.26) | (2.07) | (1.43) | (1.98) | | |
| Losses | 154 | -0.128 | -0.161 | -0.179 | -0.230 | -0.240 | -0.409 | -0.225 | | |
| | | (-0.48) | (-0.82) | (-1.09) | (-1.71) | (-1.44) | (-1.43) | (-1.34) | | |
| Win-Loss | | 0.950 | 0.841 | 0.734 | 0.452 | 0.904 | 0.949 | 0.805 | | |
| | | (2.13) | (2.29) | (2.14) | (2.07) | (2.64) | (2.05) | (2.50) | | |

Table 4. Pooled time-series regression of abnormal return on home game wins and losses

Using the six models of expected return described in Section 3.1, we estimate the abnormal return for each of the sponsor-days in the sample. For the full sample of sponsor-days (game and no-game days), we run a pooled regression of the six-model average abnormal return on two dummy variables: *Win* indicates a first trading day after a win of the sponsored home team, and *Loss* is the equivalent indicator for a loss of the home team. Panel A shows estimates of OLS regression, and Panel B shows estimates of Generalized Least Squares (GLS) regression accounting for group-wise clustering by trading day. The results are presented for the three samples of home games described in Table 3. All coefficients are multiplied by 100 and *t*-statistics are in parentheses.

Interpretation: The pooled regression estimates corroborate the mean abnormal returns. For all three samples, the difference between the coefficients of the win and loss dummy variables is very similar to the win-loss average abnormal return difference reported in Table 3. The GLS regression estimates verify that the results are not driven by time-clustering.

| | Monday night games | Post-season games | Unexpected outcome games | | | | | | |
|-----------------|--------------------|-------------------|--------------------------|--|--|--|--|--|--|
| Panel A: OLS | | | | | | | | | |
| Intercept | -0.005 | -0.030 | -0.004 | | | | | | |
| | (-1.14) | (-5.02) | (-0.57) | | | | | | |
| Win | 0.513 | -0.159 | 0.617 | | | | | | |
| | (2.37) | (-0.54) | (3.03) | | | | | | |
| Loss | 0.017 | -0.977 | -0.223 | | | | | | |
| | (0.08) | (-2.26) | (-1.22) | | | | | | |
| Win-Loss | 0.496 | 0.818 | 0.840 | | | | | | |
| <i>p</i> -value | (0.049) | (0.059) | (0.001) | | | | | | |
| | Panel | B: GLS | | | | | | | |
| Intercept | -0.005 | -0.030 | -0.004 | | | | | | |
| | (-0.59) | (-4.21) | (-0.46) | | | | | | |
| Win | 0.539 | -0.146 | 0.599 | | | | | | |
| | (2.59) | (-0.50) | (2.95) | | | | | | |
| Loss | -0.007 | -1.027 | -0.236 | | | | | | |
| | (0.05) | (-2.38) | (-1.30) | | | | | | |
| Win-Loss | 0.546 | 0.881 | 0.835 | | | | | | |
| <i>p</i> -value | (0.015) | (0.045) | (0.001) | | | | | | |

Table 5. Subsequent abnormal returns

Panel A shows the mean difference in the sponsoring companies' abnormal stock returns in the first trading day after home game wins and after home game losses (averaged across the six models of abnormal return described in Section 3.1), and the cumulative average abnormal return during the following four days (days 2 to 5). Panel B shows the slope coefficient estimates of cross-sectional regression of the cumulative abnormal return (CAR) in days 2 to 5 on the first trading day average abnormal return (AR). The panel also reports the Wald test p-value for the hypothesis that the coefficient equals -1. Results are presented for the three samples described in Table 3. Returns are in percent and t-statistics are in parentheses.

Interpretation: Cumulative average abnormal returns in subsequent days do not show return reversal. Cross-sectional regressions show mixed relations. These results do not support an initial overreaction to game outcome. Yet the strong relations (positive or negative) between returns in consecutive days may indicate other forms of behavioral bias.

| | | Panel A: Cumulative average abnormal return | | | | | | | | |
|------------------------------|--------|---|---------|----------|-----------|------------|--|--|--|--|
| | Mond | ay night | Post- | season | Unexpecte | ed outcome | | | | |
| | day 1 | days 2-5 | day 1 | days 2-5 | day 1 | days 2-5 | | | | |
| Wins | 0.508 | -0.482 | -0.189 | 0.182 | 0.580 | -0.252 | | | | |
| t-statistic | (2.18) | (-0.72) | (-0.90) | (0.30) | (1.98) | (-0.58) | | | | |
| Losses <i>t</i> -statistic | 0.011 | -0.502 | -1.010 | -2.850 | -0.225 | -0.067 | | | | |
| | (0.04) | (-0.56) | (-1.86) | (-1.24) | (-1.34) | (-0.19) | | | | |
| Win-Loss <i>t</i> -statistic | 0.497 | 0.021 | 0.819 | 3.030 | 0.805 | -0.185 | | | | |
| | (1.40) | (0.02) | (1.72) | (1.28) | (2.50) | (-0.33) | | | | |

| | Panel B: $\hat{\beta}_1$ from the regression $CAR(days\ 2-5)_i = \beta_0 + \beta_1 AR(day\ 1)_i + \varepsilon_i$ | | | | | | |
|---|--|-------------|--------------------|--|--|--|--|
| | Monday night | Post-season | Unexpected outcome | | | | |
| All games t -statistic p -value for β_1 =-1 | 0.993 | 1.602 | -0.128 | | | | |
| | (3.31) | (3.53) | (-1.32) | | | | |
| | (<0.001) | (<0.001) | (<0.001) | | | | |
| Wins t -statistic p -value for β_1 =-1 | 0.495 | -0.489 | -0.251 | | | | |
| | (1.24) | (-1.05) | (-2.03) | | | | |
| | (0.001) | (0.278) | (<0.001) | | | | |
| Losses t -statistic p -value for β_1 =-1 | 1.674 | 2.926 | 0.158 | | | | |
| | (3.65) | (3.81) | (0.93) | | | | |
| | (<0.001) | (<0.001) | (<0.001) | | | | |

Table 6. Post-game trading rule

Every week during the NFL season we form a value-weighted portfolio of buying the stocks of all sponsoring companies whose teams whose teams whose teams whose teams lost that week. We hold the portfolio from the second trading day after the game until the fifth trading day. We apply this long-short trading strategy to two samples: all home games and home games whose outcomes are most unpredictable ex ante (games with betting spreads of 3 points and lower in absolute value). The table shows the portfolios' mean excess weekly returns (in excess of the risk-free rate) and alphas from factor models. The CAPM one-factor model uses the market factor. The three factors in the 3-factor model are the Fama and French (1993) factors. The 4- and 5-factor models include the Fama-French factors augmented with momentum and reversal factors. All returns and alphas are in percent per week and the corresponding *t*-statistics are in parentheses. The sample period is 1997 to 2013.

Interpretation: Abnormal sponsor stock returns following home games can be translated into profitable trading opportunities, particularly if applied to the most unpredictable games. The trading strategy generates mean excess return and factor-model alphas of approximately 28 percent per NFL season.

| | Excess return | CAPM alpha | 3-factor alpha | 4-factor alpha | 5-factor alpha |
|---------------------------------------|---------------|---------------|-------------------|-------------------|-------------------|
| All home games (N=1,710) | | | | | |
| Sponsors of winning teams | 0.577 | 0.253 | 0.231 | 0.234 | 0.278 |
| | (2.58) | (1.52) | (1.40) | (1.42) | (1.70) |
| Sponsors of losing teams | 0.163 | -0.191 | -0.204 | -0.187 | -0.139 |
| | (0.66) | (-1.03) | (-1.11) | (-1.01) | (-0.76) |
| Long-short portfolio | 0.414 | 0.444 | 0.436 | 0.421 | 0.417 |
| | (1.70) | (1.83) | (1.79) | (1.73) | (1.71) |
| Most unpredictable home games (N=537) | | | | | |
| Sponsors of winning teams | 1.525 | 0.750 | 0.770 | 0.782 | 0.859 |
| | (3.99) | (2.51) | (2.57) | (2.61) | (2.88) |
| Sponsors of losing teams | 0.073 | -0.750 | -0.746 | -0.730 | -0.603 |
| | (0.17) | (-2.13) | (-2.11) | (-2.07) | (-1.72) |
| Long-short portfolio | 1.452 | 1.500 | 1.516 | 1.512 | 1.461 |
| | (3.23) | (3.32) | (3.34) | (3.33) | (3.20) |

Figure 1. Team-specific effect by market size

The sample includes all teams with at least five home wins and five home losses across Monday night games, post-season games, and games with unexpected outcome (as described in Table 3). The figure shows the mean difference in the sponsoring companies' abnormal stock returns after game wins and losses, averaged across the six models of abnormal return described in Section 3.1. The teams are ordered by the size of their media market, measured by population in TV households within a 75-mile radius of the team's stadium. All returns are in percent.

Interpretation: Most teams show a positive effect of game outcome on the market value of their stadium sponsoring companies. There is no monotonic relation between the magnitude of the effect and the local market size of the team.

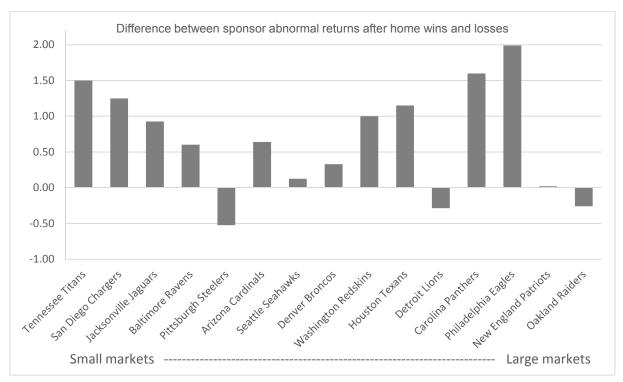
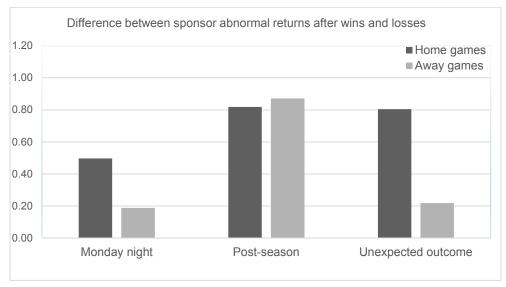
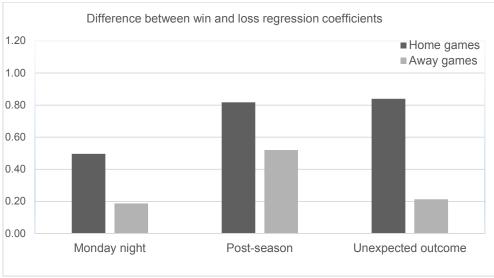


Figure 2. Effect of game outcome on sponsor stock return in home and away games

The upper figure shows the mean difference in the sponsoring companies' abnormal stock returns after game wins and losses, averaged across the six models of abnormal return described in Section 3.1. The lower figure shows the coefficients of the win and loss dummy variables of the regression model described in Section 3.3. Results are presented for home and away games, applied to the three samples described in Table 3: Monday night games, post-season games, and games with unexpected outcome. Returns are in percent and regression coefficients are multiplied by 100.

Interpretation: Outcomes of important home games affect the market value of the team stadiums' sponsors, whereas away games typically generate a much weaker effect. This suggests that abnormal returns to the sponsors' stocks are driven at least to a certain extent by investor sentiment.





Appendix

Table A1. Publically traded companies sponsoring NFL teams' stadiums

The table lists all NFL teams' home stadiums sponsored by publically traded companies over the years 1996-2013.

| | | | | | | Sponsorship | | | |
|-----------|---------------|---------|-------------------------------|------------|--------------------------|----------------|-------|--------------|--|
| NFL Team | ı City | State | Stadium Name | Start Date | Sponsoring Firm | Total | Years | Avg. Annual | |
| Bills | Toronto | Ontario | Rogers Centre | Jan 2008 | Rogers Communications | n/a | n/a | n/a | |
| Broncos | Denver | CO | Invesco Field at Mile High | Sep 2001 | Invesco Ltd. | \$ 120,000,000 | 20 | \$ 6,000,000 | |
| Browns | Cleveland | ОН | FirstEnergy Stadium | Feb 2013 | First Energy Corporation | n/a | n/a | n/a | |
| Cardinals | Glendale | ΑZ | University of Phoenix Stadium | Aug 2006 | Apollo Group | \$ 154,000,000 | 26 | \$ 5,923,077 | |
| Chargers | San Diego | CA | Qualcomm Stadium | Jan 1997 | Qualcomm | \$ 18,000,000 | 20 | \$ 900,000 | |
| Dolphins | Miami | FL | Sun Life Stadium | Jan 2010 | Sun Life Financial | \$ 37,500,000 | 5 | \$ 7,500,000 | |
| Eagles | Philadelphia | PA | Lincoln Financial Field | Aug 2003 | Lincoln National Corp | \$ 139,000,000 | 21 | \$ 6,619,048 | |
| Giants | E. Rutherford | NJ | MetLife Stadium | Aug 2011 | MetLife Inc. | \$ 400,000,000 | 25 | \$16,000,000 | |
| Jaguars | Jacksonville | FL | Alltel Stadium | Jan 1997 | Alltel Corp | \$ 6,200,000 | 10 | \$ 620,000 | |
| Jaguars | Jacksonville | FL | EverBank Field | Aug 2010 | EverBank Financial Corp | \$ 16,600,000 | 5 | \$ 3,320,000 | |
| Jets | E. Rutherford | NJ | MetLife Stadium | Aug 2011 | MetLife Inc. | \$ 400,000,000 | 25 | \$16,000,000 | |
| Lions | Detroit | MI | Ford Field | Aug 2002 | Ford Motor Company | \$ 40,000,000 | 20 | \$ 2,000,000 | |
| Panthers | Charlotte | NC | Ericsson Stadium | Sep 1996 | LM Ericsson | \$ 20,000,000 | 10 | \$ 2,000,000 | |
| Panthers | Charlotte | NC | Bank of America Stadium | Jan 2004 | Bank of America | \$ 140,000,000 | 20 | \$ 7,000,000 | |
| Patriots | Foxborough | MA | Gillette Stadium | May 2002 | The Gillette Company | n/a | 15 | n/a | |
| Patriots | Foxborough | MA | Gillette Stadium | Oct 2005 | Proctor & Gamble | \$ 105,000,000 | 15 | \$ 7,000,000 | |
| Raiders | Oakland | CA | McAfee Coliseum | Jan 2005 | McAfee Inc. | \$ 13,700,000 | 5 | \$ 2,740,000 | |
| Raiders | Oakland | CA | Overstock.Com | Apr 2011 | Overstock.com | \$ 7,200,000 | 6 | \$ 1,200,000 | |
| Ravens | Baltimore | MD | M&T Bank Stadium | May 2003 | M&T Bank | \$ 75,000,000 | 15 | \$ 5,000,000 | |
| Redskins | Landover | MD | FedEx Field | May 1999 | FedEx | \$ 205,000,000 | 27 | \$ 7,592,593 | |
| Saints | New Orleans | LA | Mercedes-Benz Superdome | Oct 2011 | Daimler AG | n/a | 10 | n/a | |
| Seahawks | Seattle | WA | Qwest Field | Jun 2004 | Qwest | \$ 75,000,000 | 15 | \$ 5,000,000 | |
| Seahawks | Seattle | WA | CenturyLink Field | Jun 2011 | CenturyLink Inc. | \$ 75,000,000 | 15 | \$ 5,000,000 | |
| Steelers | Pittsburgh | PA | Heinz Field | Aug 2001 | H.J. Heinz Company | \$ 57,000,000 | 20 | \$ 2,850,000 | |
| Texans | Houston | TX | Reliant Stadium | Aug 2002 | Reliant Energy | \$ 320,000,000 | 32 | \$10,000,000 | |
| Texans | Houston | TX | Reliant Stadium | Aug 2010 | NRG Energy | \$ 320,000,000 | 32 | \$10,000,000 | |
| Titans | Nashville | TN | LP Field | Jun 2006 | Louisiana-Pacific | \$ 30,000,000 | 10 | \$ 3,000,000 | |

Table A2. Game sample distribution

The table presents the yearly distribution of all NFL games (home and away) of teams with publically traded stadium sponsors. Distributions are presented for the full sample and separately for three subsamples: regular season games played on Monday night, all post-season (playoff) games, and games with unexpected outcome, where a game outcome is classified as unexpected if it is either against the prediction of a pregame betting spread of at least 5 points or comes after at least three consecutive games with the opposite outcome.

| | Full sample # games Home Away | | | Monday night | | | Post-season # games Home Away | | | Unexpected outcome # games Home Away | | |
|-------|----------------------------------|--------|-------|--------------|--------|------|----------------------------------|-------|------|---|------------|------|
| | # garries | TIOTIC | Away | # garries | TIOTHC | Away | # garries | HOHIC | Away | # garries | , i loille | Away |
| 1997 | 31 | 16 | 15 | 1 | 1 | 0 | 1 | 0 | 1 | 2 | 0 | 2 |
| 1998 | 30 | 15 | 15 | 2 | 2 | 0 | 0 | 0 | 0 | 4 | 1 | 3 |
| 1999 | 47 | 23 | 24 | 2 | 1 | 1 | 2 | 1 | 1 | 7 | 2 | 5 |
| 2000 | 49 | 26 | 23 | 5 | 2 | 3 | 4 | 3 | 1 | 18 | 9 | 9 |
| 2001 | 89 | 44 | 45 | 5 | 2 | 3 | 0 | 0 | 0 | 17 | 12 | 5 |
| 2002 | 161 | 80 | 81 | 8 | 4 | 4 | 0 | 0 | 0 | 24 | 14 | 10 |
| 2003 | 195 | 97 | 98 | 8 | 4 | 4 | 2 | 1 | 1 | 27 | 15 | 12 |
| 2004 | 228 | 113 | 115 | 10 | 5 | 5 | 6 | 5 | 1 | 35 | 14 | 21 |
| 2005 | 242 | 128 | 114 | 14 | 7 | 7 | 9 | 7 | 2 | 33 | 21 | 12 |
| 2006 | 275 | 138 | 137 | 18 | 11 | 7 | 14 | 5 | 9 | 55 | 27 | 28 |
| 2007 | 286 | 144 | 142 | 17 | 8 | 9 | 10 | 5 | 5 | 46 | 22 | 24 |
| 2008 | 285 | 144 | 141 | 15 | 10 | 5 | 10 | 5 | 5 | 56 | 22 | 34 |
| 2009 | 262 | 131 | 131 | 16 | 6 | 10 | 15 | 7 | 8 | 44 | 21 | 23 |
| 2010 | 267 | 133 | 134 | 14 | 5 | 9 | 7 | 3 | 4 | 43 | 24 | 19 |
| 2011 | 309 | 157 | 152 | 16 | 7 | 9 | 8 | 5 | 3 | 60 | 24 | 36 |
| 2012 | 319 | 160 | 159 | 18 | 10 | 8 | 11 | 5 | 6 | 43 | 22 | 21 |
| 2013 | 324 | 161 | 163 | 16 | 7 | 9 | 10 | 5 | 5 | 56 | 28 | 28 |
| Total | 3,399 | 1,710 | 1,689 | 185 | 92 | 93 | 109 | 57 | 52 | 570 | 278 | 292 |