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Returns: Reassessing the
Informational Value of Bets**

Massimiliano Castellani
Pierpaolo Pattitoni
Roberto Patuelli

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Event Clustering and Abnormal Returns: Reassessing the Informational Value of Bets

Massimiliano Castellani,^{*} Pierpaolo Pattitoni,[†] Roberto Patuelli[‡]

Abstract

We analyse the links between soccer match results, bets and stock returns of all listed European soccer teams. Using an event study approach, we measure abnormal returns following wins, ties and losses. Wins are associated with positive abnormal returns, and ties and losses with negative abnormal returns. Additionally, we analyse the role of bets in shaping market reactions to unexpected results, which we find to be non-significant. We propose an alternative econometric approach, using seemingly unrelated regression models, to take into account the problem of overlapping events. While our results concerning match results are confirmed, abnormal returns following unexpected results are found to be statistically significant and to magnify the positive (negative) effects of wins (losses).

Keywords: Information and Market Efficiency; Event Studies; Soccer; Bets; Event Clustering; Seemingly Unrelated Regression Equation (SUR)

JEL Classification: G14; L83; C30

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^{*} Department of Economics, University of Bologna, Bologna, Italy; The Rimini Centre for Economic Analysis (RCEA), Rimini, Italy. E-mail: m.castellani@unibo.it.

[†] Department of Management, University of Bologna, Italy; The Rimini Centre for Economic Analysis (RCEA), Rimini, Italy. E-mail: pierpaolo.pattitoni@unibo.it.

[‡] Department of Economics, University of Bologna, Bologna, Italy; The Rimini Centre for Economic Analysis (RCEA), Rimini, Italy. E-mail: roberto.patuelli@unibo.it.

1 – Introduction

In the literature of the last twenty years, several studies have suggested an association between sport events and individual mood: fans usually experience a strong positive reaction when their team performs well and a strong negative reaction when their team performs poorly. More specifically, sport wins are generally associated with improvements in mood and self-esteem (Hirt et al. 1992). Instead, sport losses are followed by an increase in heart attacks, crimes and suicides (White 1989; Trovato 1998; Wann 2001; Carroll et al. 2002; Berthier and Boulay 2003; Chi and Klöner 2003).

Other studies have found a relationship between investor mood and stock prices. Several contributions in behavioral finance literature investigate the effect of mood and investor sentiment on asset prices. Remarkable examples are Barberis et al. (1998), Daniel, Hirshleifer and Subrahmanyam (1998), and Baker and Wurgler (2006).¹

A number of further authors were able to join these two streams of literature and show that soccer matches may influence stock returns. Ashton et al. (2003) find a strong association between the performance of the England soccer team and the general stock market index returns. Using international soccer results as a mood variable, Edmans et al. (2007) find an economically and statistically significant market reaction to losses, but not to wins. Brini and Palomba (2011) find a relationships between the stock returns of Italian soccer teams, their sport performances in terms of match outcomes, and the importance of the matches.

Two recent studies by Bernile and Lyandres (2011) and by Palomino et al. (2009) consider the information content of pre-match bets in studying the relationship between soccer match results and soccer team market returns.² Using bets as a proxy for investor expectations, Bernile and Lyandres (2011) analyse the returns of a sample of listed European soccer teams around important matches. They find that investors are overly optimistic about their teams' prospects *ex ante*, and disappointed *ex post*. This result leads to negative post-event abnormal returns. Using a sample of soccer teams listed on the London Stock Exchange, Palomino et al. (2009) find a strong market reaction to match results and a lack of reaction to betting odds. They conclude that this lack of reaction may be due to the lack of informational content of betting information.

Our paper fits in the above stream of literature. In particular, we analyse the relationship between soccer match results, bets and stock returns of all listed European soccer teams. Since sport events affect mood, expected and unexpected soccer match results provide an unique way of studying the

¹ A review of this stream of the literature can be found in Hirshleifer (2001) and Shiller (2000).

² The use of pre-match bets as an indicator of match uncertainty is quite common in the applied literature (e.g., see Czarnitzki and Stadtmann 2002).

link between investor mood and stock market prices. Our analysis aims at answering several questions.

- i. First, splitting the matches between wins, ties and losses, we wonder if these events lead to a different market reaction.
- ii. Second, we test whether the intensity of the result, as measured by goal difference, has an effect on listed soccer team returns.
- iii. Third, we ask whether the location where a match is played (at home or away) influences the market reaction.
- iv. Fourth, we test if any type of competition or date (year and month) effect influences the market reaction following wins, ties and losses.
- v. Finally, using betting data, we divide all match results between expected and unexpected, and test whether unexpected events are associated with a larger effect on listed soccer team returns.

Our main findings show positive abnormal returns following wins and negative abnormal returns following both ties and losses. This result strongly persists after controlling for the location where the match is played, the type and the date of competition. Furthermore, using the information of pre-match betting odds, we show that abnormal returns are magnified by unexpected results.

Our study contributes to the existing literature in several respects:

- i. The data at our disposal allow us to corroborate known results with new empirical evidence and offer new insights on the effect of moods on soccer team stock prices.
- ii. The existing literature focuses on national and international matches in single countries, or on international matches in several countries. Instead, our dataset includes all national and international match results of all 23 listed European soccer teams, in the three year period 2007–09. These teams play in 9 national championships and in all UEFA³ international competitions, and represent the whole universe of the listed soccer teams.⁴ Thus, our results contribute to widen the previous empirical evidence on the effect of soccer match results on market returns.
- iii. The role of betting is analysed only by Bernile and Lyandres (2011) and Palomino et al. (2009). Both studies consider the information contents of the bets market as a proxy for in-

³ UEFA is the Union des Associations Européennes de Football.

⁴ Although these 23 soccer teams represent the entire universe of the listed soccer teams, they are clearly just a small part of all the European soccer teams.

vestor expectations and provide controversial results. Our dataset includes all pre-match betting odds of all 23 listed European soccer teams and allows us to deepen the empirical evidences on their information content.

- iv. To double check our findings, we use an alternative econometric methodology with respect to the traditional event study approach used in most of the previous studies. Standard event studies generally deal with non-overlapping events, for which it is reasonable to assume that abnormal returns on individual securities are uncorrelated. However, soccer matches are played in the same days, thus the non-overlapping hypothesis no longer holds. To accommodate this problem of event clustering, we apply a Seemingly Unrelated Regression (SUR) model, which allows the covariances between the abnormal returns to be non-null. Our results shows that the problem of overlapping events is not negligible.

The remainder of this paper is organized as follows: Section 2 describes the dataset; Section 3 presents our hypotheses and modeling framework; Section 4 shows the results of our event study analysis; in Section 5, we retest some of our hypotheses using the SUR methodology; finally, Section 6 draws the conclusions.

2 – Variable description and descriptive statistics

We base our empirical analysis on a unique dataset we collected from the archives of SNAI (Società Nazionale Agenzie Ippiche), the leading agency for sports betting in Italy. Our dataset includes all the results and the pre-match betting odds of 2,157 matches played by all 23 listed soccer teams in Europe in the period 2007–09.⁵ Market data on these soccer teams and market index returns are collected from Datastream.

In what follows, we propose a short description of the variables we use in our econometric analysis.

Abnormal returns

Ar_i is the percentage abnormal return of the soccer team i at the event date, defined as $Ar_i = r_i - r_m$, where r_i is the logarithmic return of the soccer team i and r_m is the logarithmic market return (represented by the

⁵ The 23 listed European soccer teams are Celtic, Borussia Dortmund, Tottenham Hotspur, Afc Ajax, Lazio, Brondby If B, Arhus Elite, Sport Lisboa E Benfica, Silkeborg, Futebol Clube Do Porto, Watford, Galatasaray, Aalborg Boldspilklub, Millwall Hldg, Sport, Aik Football, Parken Sport & Entertainment, Trabzonspor Sportif Yatir, Fenerbahce Sportif Hizmet, Juventus, Besiktas, As Roma, Olympique Lyonnais.

STOXX® Europe 600 index⁶). We consider a market-adjusted return to take into account the systematic component of stock returns.

Match result

Win_i, Tie_i, Loss_i are three dummy variables indicating if the soccer team *i* won, tied, or lost the match.⁷ Of 2,157 matches considered, 1,088 are wins (50%), 519 are ties (24%) and 550 are losses (26%). The prevalence of wins on losses and ties shows that listed teams generally belong to the upper tier of their respective national leagues.

Delta score

Delta_i is the goal difference of the match, defined as the goals scored by the home team minus the goals scored by the away team. This variable assumes positive, negative, or null values when the soccer team *i* wins, loses, or ties the match respectively. *Delta_i* exhibits a symmetric distribution with a mean (median) value of 0.54 (1) and a minimum (maximum) of -8 (8). Extreme positive or negative values of this variable (less than -5 or more than 5) are very rare (0.74% of the observations).

Site of competition

Away_i, Home_i are two dummy variables indicating if the soccer team *i* played the match away or at home. 1085 (50%) matches are played at home and 1078 (50%) matches are played away.

Type of competition

Ncha_i, Clea_i, Uefa_i, Icup_i, Ncup_i are a set of dummy variables indicating the type of competition played by the soccer team *i*. 1663 (76.88%) of competitions are national championship matches, 144 (6.66%) are champions league matches, 128 (5.92%) are UEFA league matches, 12 (0.55%) other international cup matches, and 216 (9.99%) are national cup matches.

Year of competition

Y07_i, Y08_i, Y09_i are a set of year dummies. Of the total number of matches, 1,062 were played in 2008 (49% of the total), 566 in 2007 (26%), and 529 in 2009

⁶ Using country-specific market indices might seem more appropriate if we consider that an investor is more likely to invest in her country. This hypothesis, called home bias, is often justified by the tendency of people to invest in what they believe to know better. To take this point into account, we recalculate all abnormal returns using country-specific indices as a robustness check. However, the results are not qualitatively different from the ones reported in the paper.

⁷ All the dummies are equal to one in correspondence of the first trading day following the match. If the match is played during the week, the first available trading day is the day following the match. If the match is played during the weekend, the first available trading day is on Monday.

(25%). The average number of matches by year is 719.

Month of competition

$Jan_i, Feb_i, Mar_i, Apr_i, May_i, Jun_i, Jul_i, Aug_i, Sep_i, Oct_i, Nov_i, Dec_i$ are a set of month dummies. The number of matches is almost uniformly distributed across months (about 8% per month). The only exceptions are the summer months when less matches are played.

Expected and unexpected results

Exp_i, Une_i are two dummy variables indicating if the match results was expected or unexpected based on the information of the last pre-match betting odds. The match result with the smaller probability of occurrence is considered an unexpected event. 952 out of 2,157 match results are unexpected (44% of the total): 142 are unexpected wins (13% of the total number of wins), 519 are unexpected ties (100% of the total number of ties), and 291 are unexpected losses (53% of the total number of losses). Therefore, while all ties are unexpected,⁸ about 90 per cent of wins are expected, while 47% of losses are expected.

3 – Hypotheses and modeling framework

We use a standard event study methodology (Campbell et al. 1997) to examine the behavior of teams' stock returns around matches, i.e., the events. The response variable used in all regression models is AR_i .⁹

We start from a basic model analyzing the market reaction following wins, ties and losses. The model equation is

$$AR_i = \alpha_1 Win_i + \alpha_2 Tie_i + \alpha_3 Loss_i + \varepsilon_i. \quad (1)$$

⁸ Since a tie is the most probable result when teams are equally skilled, the absence of unexpected ties may indicate a disparity of skills between listed and unlisted soccer teams. However, this notion is merely speculative, thus future research might consider investigating this issue.

⁹ We use the event parameter approach (also known as regression-based approach). The event parameter approach produces results that are numerically equivalent to the standard event study method à la Brown and Warner (1980, 1985). All our inference is based on heteroskedasticity consistent standard errors and is double checked by constructing bootstrap confidence intervals. Bootstrap confidence intervals are constructed using the procedure described in Greene (2007) by carrying out (with replacement) 5000 samples from the full sample of observations.

In this model, α_1 , α_2 and α_3 measure the average abnormal returns following wins, ties and losses respectively. Coherently with previous findings in the literature and in financial economic theory we expect

- H_1 : $\alpha_1 > 0$, $\alpha_2 \cong 0$, $\alpha_3 < 0$, and $|\alpha_1| < |\alpha_3|$.

The first hypothesis predicts positive abnormal returns following wins and negative abnormal returns following losses, but no abnormal returns following ties. Furthermore, given the well-known asymmetric reaction associated with positive and negative events, we expect the abnormal returns following losses to be greater in absolute terms than those following wins. All the predictions of H_1 are motivated by the effect of moods on stock prices.

As a further step in the analysis, we test whether the intensity of the result, as measured by goal difference, has a (possibly nonlinear) effect on abnormal returns. The second model is

$$AR_i = \alpha + \delta_1 \Delta_i + \delta_2 \Delta_i^2 + \varepsilon_i. \quad (2)$$

In this model, $dAR_i/d\Delta_i = \delta_1 + 2\delta_2 \Delta_i$ is the marginal effect of the goal difference on the abnormal returns. We expect

- H_2 : $\delta_1 > 0$, $\delta_2 < 0$, and $|\delta_1| > |\delta_2|$.

If the market reaction is driven by mood, we expect abnormal returns to increase with the goal difference.¹⁰ However, we expect the effect of the goal difference to be nonlinear and to reduce for large values of this variable, i.e., the positive mood associated with a win gradually decreases as Δ_i increases.

A third concern that may arise is whether the location where a match is played (at home or away) influences abnormal returns. To answer this question, we estimate the following model:

$$AR_i = \alpha_1 Win_i \times Away_i + \alpha_2 Tie_i \times Away_i + \alpha_3 Loss_i \times Away_i + \alpha_4 Win_i \times Home_i + \alpha_5 Tie_i \times Home_i + \alpha_6 Loss_i \times Home_i + \varepsilon_i. \quad (3)$$

¹⁰ Beck and Meyer (Beck and Meyer 2012) use goal differences as the outcome of a soccer match. The motivation for their decision is that marginal performance variations might be reflected by variations in observed goal differences.

In this model, α_1 , α_2 and α_3 measure the average abnormal returns following wins, ties and losses when the soccer team i plays the match away; α_4 , α_5 and α_6 measure the average abnormal returns following wins, ties and losses when the soccer team i plays the match at home. We expect

- H_3 : $\alpha_1 > \alpha_4 > \alpha_2 > 0$, and $\alpha_6 < \alpha_3 < \alpha_5 < 0$.

We expect a win away to produce a larger positive abnormal return than a win at home or a tie away. Conversely, we expect a loss at home to cause larger negative abnormal returns than a loss or a tie away. Indeed, wins away or at home and ties away are generally perceived as a good news, while losses at home or away and ties at home are commonly considered as a bad news. Playing at home is expected to provide an advantage, because the team can benefit from the support of its fans.

We also check if any type or date (year and month) of competition influences the abnormal returns following wins, ties and losses. To answer this question, we estimate four regression models (one for each effect). We setup the equation of each of the four models, not reported here to preserve space, by interacting the match result dummies with the type of competition, year, and month dummies. Joint Wald tests on the difference of the coefficients associated with one effect should evidence any influence on abnormal returns. In what follows, we state and briefly comment the hypotheses associated with these regression models.

- H_4 : The type of competition played by the soccer team i ($Ncha_i$, $Clea_i$, $Uefa_i$, $Icup_i$, $Ncup_i$), has an effect on the magnitude of abnormal returns following wins, ties and losses.

Since not all type of competitions are equally important, we expect a significant Wald test associated with this model (e.g., champion league matches are generally considered more important than UEFA league matches and can elicit larger market reactions).

- H_5 : The year of competition in which the match takes place ($Y07_i$, $Y08_i$, $Y09_i$) has an effect on the magnitude of abnormal returns following wins, ties and losses.

Since the attention of the media around soccer and the interest of the investors on the listed soccer teams may vary over time, we could expect matches played in different years to be characterized by systematic differences in terms of abnormal returns.

- H_6 : The month in which the match takes place ($Jan_i, Feb_i, Mar_i, Apr_i, May_i, Jun_i, Jul_i, Aug_i, Sep_i, Oct_i, Nov_i, Dec_i$), has an effect on the magnitude of abnormal returns following wins, ties and losses.

The period of the year in which the match takes place (beginning or end of the season) may have an effect on fan mood and, thus, on the abnormal returns (e.g., the final match of UEFA Champions League, which is one of the most important event for the soccer fans, is played in May or June).

The last hypothesis that we test concerns the market reaction following expected and unexpected match results. The equation of the estimated model is

$$AR_i = \alpha_1 Win_i \times Exp_i + \alpha_2 Loss_i \times Exp_i + \alpha_3 Win_i \times Une_i + \alpha_4 Tie_i + \alpha_5 Loss_i \times Une_i + \varepsilon_i. \quad (4)$$

In this model, α_1 and α_2 measure the average abnormal returns following expected wins and losses; α_3 , α_4 and α_5 measure the average abnormal returns following unexpected wins, ties and losses. It is worth noting that, as already pointed out, all the ties in our dataset are unexpected, thus it is not possible to identify the average abnormal return following expected ties. Our hypotheses can be stated as

- H_7 : $\alpha_3 > \alpha_1 > 0$, $\alpha_5 < \alpha_2 < 0$, and $\alpha_4 \cong 0$.

We expect larger abnormal returns following unexpected events, i.e. unexpected wins and losses should determine larger abnormal returns than expected wins and losses respectively.

4 – Event study results

In this section, we present the results of our event study analysis. The following subsections are devoted to the test of the hypotheses stated in the previous section.

4.1 – Market reaction to match results, goal difference, and site of competition

The first hypothesis we test is H_1 . The results of the estimation of Model (1) are presented in Table 1, which shows statistically significant positive market reactions following wins (0.792%) and negative ones following losses (−1.142%). Furthermore, the average abnormal return following

losses is larger in absolute terms than the average abnormal return following wins. All these results are in line with expectations (and with previous findings in the literature). However, we reject the hypothesis of no market reaction associated with ties: ties seem to have a negative effect in terms of abnormal returns, even if their effect is not as large as the one associated with losses (−0.371%). Thus, our results shed new light on the informational content of ties: investors seem to consider ties as negative rather than neutral events. This result could be explained considering that ties reduce the probability of winning a championship or a cup, and that all soccer teams in our sample are upper tier teams in their respective leagues.

Table 1. Match result and abnormal returns.

Variable	Estimate	Std. Error	p-value	Sign.
Loss	−1.142	0.169	0.000	***
Tie	−0.371	0.175	0.034	**
Win	0.792	0.138	0.000	***

***, ** and * indicate statistical significance at the 1%, 5% and 10% level, respectively.

Once we have verified that abnormal returns are affected by match results, we test the second hypothesis (H_2) by estimating Equation (2). Indeed, the results (see Table 2) seem to support our hypothesis: abnormal returns increase with the goal difference. Furthermore, as expected, the effect of the goal difference is nonlinear and decreases for larger goal differences.

Table 2. Delta score and abnormal returns.

Variable	Estimate	Std. Error	p-value	Sign.
Intercept	−0.088	0.105	0.400	
Delta	0.457	0.056	0.000	***
Squared Delta	−0.042	0.016	0.008	***

***, ** and * indicate statistical significance at the 1%, 5% and 10% level, respectively.

The results of this second model are consistent with those of the first model: match results affect abnormal returns. However, this effect is amplified (in a gradually decreasing way) by the intensity of the result. This nonlinear pattern (depicted in Figure 1) can be interpreted as follows: the positive coefficient on the linear term (0.457%) identifies win- or loss-related moods (e.g., winning is preferable to tying or losing); the negative coefficient on the quadratic term (−0.042%) may imply satiety (a greater overall goal difference at the end of a league rarely brings to significant advantages) or a disparity in the strength of the two teams, which thus reduces the importance of the win itself.

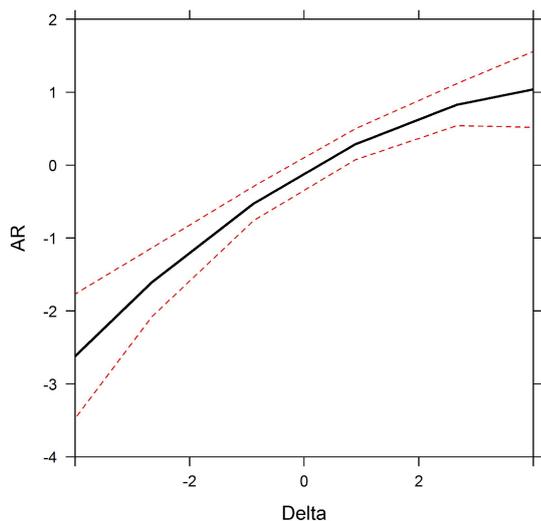


Figure 1. Non linearity in delta score along with 95-percent confidence interval around the estimated effect

Now we test the third hypothesis (H_3). The results are presented in Table 3. First, we note that the estimated coefficients of Model (3) are in line with those presented in Table 1: a positive market reaction follows wins (1.021% at home and 0.476% away), a negative market reaction follows losses (−1.317% at home and −1.045% away), and a negative market reaction is suggested for ties (a non-significant −0.300% at home and a slightly significant −0.437% away). Second, we test whether the location where a match is played (at home or away) influences abnormal returns. A joint Wald test for the site of competition, based on a re-parameterization of the model not reported for brevity, shows that there is no statistically significant difference between the abnormal returns of matches played at home or away (Wald test = 1.514, p-value = 0.209). Thus, the presence of a site of competition effect is not supported by the data. This evidence is not consistent with the stylized fact that playing at home is an important advantage for a soccer team and leads to a rejection of H_3 .

4.2 – Competition type and date effect

In this section, we test our hypotheses regarding the competition type and date. The first hypothesis (H_4) concerns the influence of the type of competition on abnormal returns. We estimate a model distinguishing among different competition types (i.e., national championship, champions league, UEFA league, other international cup, and national cup matches), and we find a strongly statistical difference in abnormal returns associated with different types of competition (Wald test = 4.0529, p-value < 0.000). Instead of reporting the results in a table (there would be a total of 15 coeffi-

cients), we prefer to plot the regression coefficients in a figure (see Figure 2). The abnormal returns registered for the national championship completely agree with those reported in Model (1): wins elicit positive abnormal returns; ties and losses are associated with negative abnormal returns. The same finding applies to Champions League and national cup matches, even if with differences in the magnitude of abnormal returns. The only anomalies we observe concern UEFA cup wins (negative abnormal returns) and international cup losses and ties (positive abnormal returns). However, this anomaly is only apparent. In fact, the coefficients associated with these events are not significant (thus no abnormal returns are detected). This statistical insignificance may be justified by the small number of observations for these competition types. However, all in all the empirical evidence seems to support our hypothesis (H_4) that not all type of competitions are equally important.

Table 3. Site of competition and abnormal returns.

Variable	Estimate	Std. Error	p-value	Sign.
Loss Away	-1.045	0.211	0.000	***
Tie Away	-0.437	0.248	0.078	*
Win Away	0.476	0.215	0.027	**
Loss Home	-1.317	0.281	0.000	***
Tie Home	-0.300	0.245	0.221	
Win Home	1.021	0.179	0.000	***

***, ** and * indicate statistical significance at the 1%, 5% and 10% level, respectively.

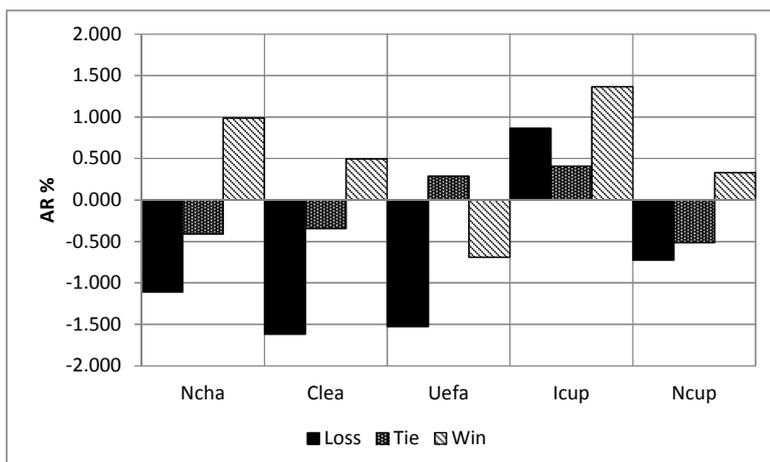


Figure 2. Competition type and abnormal returns

The fifth hypothesis (H_5) focuses on the year of competition (2007, 2008 and 2009). We hypothesized potential differences in the abnormal returns associated with these three years. Figure 3 seems to support this hypothesis. However, a joint Wald test on the year effect rejects the hypothesis of

changing abnormal returns over years on a statistical ground (Wald test = 1.647, p-value = 0.130). Thus, H_5 is not supported by the data.

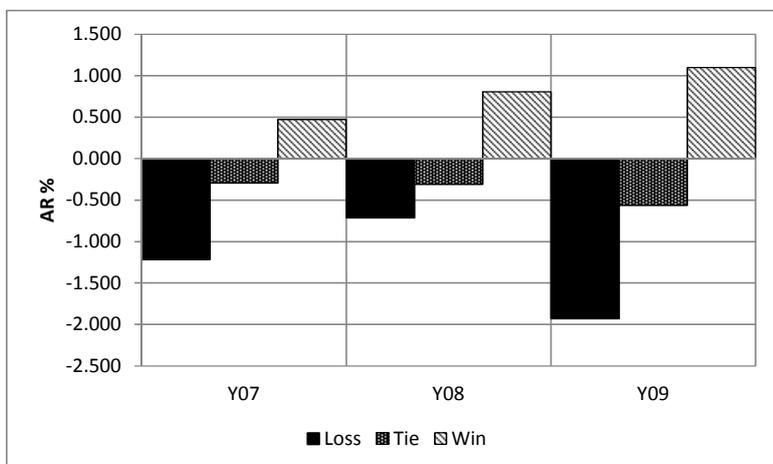


Figure 3. Year of competition and abnormal returns

In addition, we test for the presence of a month effect (H_6). Figure 4 show that the abnormal returns in different months mimic those of Model (1) (with few exceptions). Furthermore, the figure seems to suggest that the month in which the match takes place influences the magnitude of the abnormal returns. A joint Wald test on the month effect supports this hypothesis (Wald test = 2.060, p-value = 0.000). The strongest reaction is registered at the end of season in July (despite the appearance the abnormal returns in June are not statistically significant), supporting our hypothesis that the period of the year in which the match takes place affects the importance of the match itself. However, it is worth noting that only 14 matches were played in July.¹¹ Thus, caution is needed in interpreting this finding.

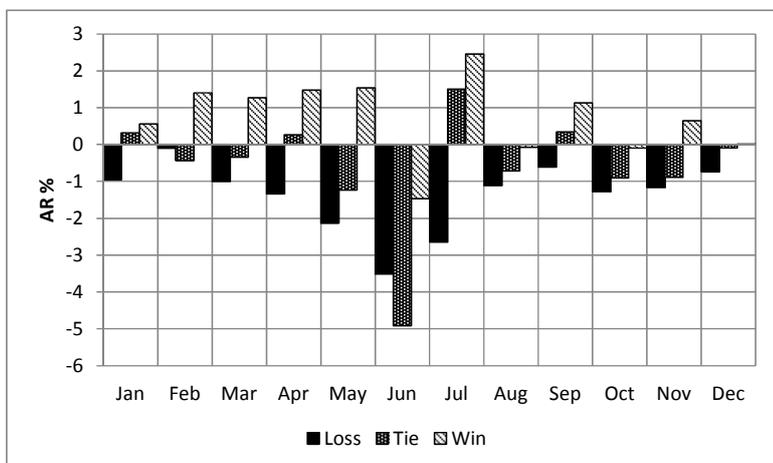


Figure 4. Month of competition and abnormal returns.

¹¹ Most championships have already ended by July.

4.3 – Informational value of bets

In this section, we focus on the eighth hypothesis (H_7), which is related to the informational value of bets. To test this hypothesis, we estimate Model (3) and present the results in table 4. The abnormal returns associate with wins and losses follow the usual pattern, both following expected and unexpected events. Furthermore, unexpected ties (the only available) register negative abnormal returns. All these results are in line with those of Model (1). Focusing more in depth on H_7 , a joint Wald test does not reject the null of no effect of unexpected events (Wald test = 0.2255, p-value = 0.7981). Thus, according to these results, we reject H_7 , which predicts a stronger market reaction following unexpected events. Since this result is counterintuitive, it needs to be analysed further.

Table 4. Informational value of bets.

Variable	Estimate	Std. Error	p-value	Sign.
Exp Loss	-1.028	0.266	0.000	***
Exp Win	0.808	0.139	0.000	***
Une Loss	-1.244	0.250	0.000	***
Tie	-0.371	0.188	0.048	**
Une Win	0.685	0.359	0.057	*

***, ** and * indicate statistical significance at the 1%, 5% and 10% level, respectively.

5 – The problem of overlapping events: a SUR approach

A problem with the standard event study methodology is that it generally deals with non-overlapping events. In case of non-overlapping events, it is reasonable to assume abnormal returns on individual securities to be uncorrelated between teams (one of the main hypothesis in event study testing procedures). However, matches are played in the same days, thus the non-overlapping hypothesis does not hold. Ignoring event date clustering would result in biased standard errors, which in turn may lead to misleading results in hypothesis testing.

In our dataset, the average number of events per day is around 12. Thus, the problem of overlapping events is a concern. To accommodate this potential problem, we apply an SUR model (Zellner 1962) to retest H_1 and H_7 . The SUR model takes into account the cross sectional correlation due to clustering of event days, allowing the covariances between the abnormal returns to be non-null (Schipper and Thompson 1985). We focus only on H_1 and H_7 because, by construction, the SUR

model does not allow to retest the remaining hypotheses. To do so, the dataset should include observations for each possible combination of variables (e.g., wins and losses for each team and for each month).

Formally, the SUR model can be seen as a generalization of a linear regression model consisting of several regression equations. For each team i , let \mathbf{r}_i denote a $T \times 1$ vector of daily returns, with generic element r_{it} ($t = 1, \dots, T$), \mathbf{X}_i a $T \times k$ matrix of k covariates, $\boldsymbol{\varepsilon}_i$ a $T \times 1$ vector of error terms and $\boldsymbol{\beta}_i$ a vector of k parameters. Then, the SUR model consists of N equations (one for each soccer team) of the form

$$\mathbf{r}_i = \mathbf{X}_i \boldsymbol{\beta}_i + \boldsymbol{\varepsilon}_i, \quad i = 1, \dots, N. \quad (5)$$

Two main assumptions on the errors of a SUR model are

$$E(\boldsymbol{\varepsilon}_i | \mathbf{X}_1, \dots, \mathbf{X}_N) = \mathbf{0}, \quad i = 1, \dots, N; \quad \text{and}$$

$$\text{Cov}(\boldsymbol{\varepsilon}_i, \boldsymbol{\varepsilon}_j | \mathbf{X}_1, \dots, \mathbf{X}_N) = \begin{cases} \sigma_i^2 \mathbf{I}_T & \text{if } i = j \\ \sigma_{ij} \mathbf{I}_T & \text{if } i \neq j \end{cases},$$

where \mathbf{I}_T is the identity matrix.

It is important to note that the system in Equation (4) does not provide a structural relationship among the different equations (i.e., the dependent variable of one equation never appears among the covariates of other equations). However, it is clear from the covariance equation above that the N equations are linked together through the error covariances σ_{ij} . This assumption is crucial because it allows the correlation across the abnormal returns of different teams to assume any values and implies unobserved variables with correlated values between equations.

The SUR model can be estimated by means of the feasible generalized least squares (FGLS) method. As T diverges, the FGLS estimator is consistent, asymptotically efficient and asymptotically normal.¹²

After presenting the main methodological feature of the SUR model, we now focus on retesting H_1 . To do so, we estimate an SUR model in which each equation (one for each team) can be formalized as follows:

¹² The model can also be estimated equation-by-equation using ordinary least squares (OLS). Even if the OLS estimator is consistent, it is not efficient because it ignores the cross-correlation among errors.

$$r_{it} = \alpha_i + \beta_i^W Win_{it} + \beta_i^T Tie_{it} + \beta_i^L Loss_{it} + \mathbf{con}'_i \boldsymbol{\gamma}_i + \varepsilon_{it}. \quad (6)$$

In Equation (6), Win_{it} , Tie_{it} and $Loss_{it}$ are dummy variables indicating if the soccer team i won, tied, or lost the match played at time t . β_i^W , β_i^T and β_i^L measure the market reaction following wins, ties and losses, respectively. Furthermore, $\boldsymbol{\gamma}_i$ is the vector of coefficients associated with the control variables included in \mathbf{con}_t .¹³ In particular, we estimate three models, which differ for the control variables included in \mathbf{con}_t . In the first case, we do not include any controls (i.e., we impose $\boldsymbol{\gamma}_i = \mathbf{0}$). In this model, α_i measures the average return that occurs when no matches are played, and the abnormal returns are defined as a deviation from α_i when matches are instead played. In the second case, to take into account the systematic component of stock returns, we add the variable Mkt_t among the controls.¹⁴ Mkt_t is the STOXX® Europe 600 index's logarithmic daily return, used as a proxy for the market return. Finally, in the third case, we consider a possible year effect by adding year dummies among the controls.

After estimating our SUR model, we test the joint significance of all the coefficients β_i^W , β_i^T and β_i^L (one for each team) by means of three tests: Wald test (F-distributed), Wald test and Likelihood ratio test (both Chi-squared-distributed).¹⁵

Since for each team we estimate a different β_i^W , β_i^T and β_i^L , as a summary measure we report the average value of all coefficients:

$$Ar^W = \frac{1}{23} \sum_{i=1}^{23} \beta_i^W, \quad Ar^T = \frac{1}{23} \sum_{i=1}^{23} \beta_i^T, \quad Ar^L = \frac{1}{23} \sum_{i=1}^{23} \beta_i^L. \quad (7)$$

where Ar^W , Ar^T and Ar^L are the average abnormal returns following wins, ties and losses, respectively.

Focusing on H_7 , we estimate an augmented SUR model in which each equation is written as

¹³ It is worth noting that the control variables are the same for each soccer team, thus we do not need the team-specific index i to characterize the vector \mathbf{con}_t .

¹⁴ More sophisticated models can be used to assess abnormal returns (e.g., the Fama-French three factor model, or Carhart's four-factor model). However, previous literature on short-term event studies (Campbell et al. 1997) points out that results are virtually unaffected by the choice of the model. This is especially true in our case, because the stocks considered belong to the same sector and are quite homogeneous.

¹⁵ While the latter two, which are asymptotically equivalent, are especially indicated for large samples, the former has better finite sample properties.

$$r_{it} = \alpha_i + \beta_i^W Win_{it} + \beta_i^T Tie_{it} + \beta_i^L Loss_{it} + \delta_i^W (Une_{it} \times Win_{it}) + \delta_i^L (Une_{it} \times Loss_{it}) + \mathbf{con}'_i \boldsymbol{\gamma}_i + \varepsilon_{it}. \quad (8)$$

In Equation (8), Win_{it} , Tie_{it} and $Loss_{it}$ indicate, as in Equation (6), if the soccer team i won, tied, or lost the match played at time t ; Une_{it} is a dummy variable indicating if the match result was unexpected based on the information of the pre-match betting odds. Based on this model specification, β_i^W and β_i^L measure the market reaction following expected wins and losses, while δ_i^W and δ_i^L measure the difference in abnormal returns when the events are unexpected. Thus, $\beta_i^W + \delta_i^W$ and $\beta_i^L + \delta_i^L$ indicate the total abnormal returns associated with unexpected wins and losses. β_i^T represents the abnormal return associated with unexpected ties. Once more, $\boldsymbol{\gamma}_i$ is the vector of coefficients associated with the control variables included in \mathbf{con}_i .

Similarly to what was done for H_1 , we test the joint significance of all the coefficients β_i^W , β_i^T , β_i^L , δ_i^W and δ_i^L by means of the three aforementioned tests, and report the average value of all coefficients as a summary statistic:

$$Ar^W = \frac{1}{23} \sum_{i=1}^{23} \beta_i^W, \quad Ar^T = \frac{1}{23} \sum_{i=1}^{23} \beta_i^T, \quad Ar^L = \frac{1}{23} \sum_{i=1}^{23} \beta_i^L, \quad (9)$$

$$\Delta UneAr^W = \frac{1}{23} \sum_{i=1}^{23} \delta_i^W, \quad \Delta UneAr^L = \frac{1}{23} \sum_{i=1}^{23} \delta_i^L,$$

where Ar^W , Ar^T and Ar^L are the average abnormal returns following expected wins, unexpected ties, and expected losses, respectively, while $\Delta UneAr^W$ and $\Delta UneAr^L$ measure the average differential abnormal returns following unexpected wins and losses.

5.1 – Reassessment of market reaction

We first retest H_1 . To do so, we estimate three SUR models, which differ according to the control variables included (indicated at the bottom of Table 5). While the first model includes no control variables, the second and the third model introduce among the covariates the market returns and the year dummies. It is worth noting that the results of the three models agree in the sign and magnitude of the average coefficients reported.

Table 5. Average market reaction to match results.

Abnormal Return (AR percent)						
	SUR 1		SUR 2		SUR 3	
Ar^W	0.589		0.627		0.631	
Wald test (F-test)	3.914	<0.001	4.230	<0.001	4.258	<0.001
Wald test (Chi-squared-test)	90.030	<0.001	97.286	<0.001	97.925	<0.001
Likelihood ratio test (Chi-squared-test)	86.280	<0.001	93.060	<0.001	93.836	<0.001
Ar^T	-0.687		-0.626		-0.631	
Wald test (F-test)	2.410	<0.001	2.262	<0.001	2.246	<0.001
Wald test (Chi-squared-test)	55.430	<0.001	52.034	<0.001	51.649	<0.001
Likelihood ratio test (Chi-squared-test)	54.155	<0.001	50.946	<0.001	50.767	<0.001
Ar^L	-1.476		-1.451		-1.443	
Wald test (F-test)	6.126	<0.001	6.165	<0.001	6.106	<0.001
Wald test (Chi-squared-test)	140.900	<0.001	141.800	<0.001	140.430	<0.001
Likelihood ratio test (Chi-squared-test)	137.200	<0.001	138.110	<0.001	137.240	<0.001
con_t						
Mkt^t	No		Yes		Yes	
Year dummies	No		No		Yes	

Table 5 presents the values of Ar^W , Ar^T and Ar^L as defined above. Consistently with Section 4.1, our results show a statistically significant positive market reaction following wins, and a statistically significant negative market reaction following ties and losses. The average abnormal return following losses is larger in absolute terms than the average abnormal return following wins. In addition, ties have a negative effect in terms of abnormal returns, even if their effect is not as large as the one associated with losses.

We now retest H_7 . Table 6 reports the estimation of our three SUR models, which agree in sign and magnitude of the average estimated coefficients: Ar^W , Ar^T , Ar^L , $\Delta UneAr^W$ and $\Delta UneAr^L$.

Our results show a statistically significant positive (negative) market reaction following expected wins (losses). Once again, the absolute average abnormal return following losses is larger than the one following wins. Furthermore, ties seem to have a negative effect in terms of abnormal returns. These results agree with the findings of Section 4.3. However, when looking at $\Delta UneAr^W$ and $\Delta UneAr^L$, we observe that unexpected events generate a larger market reaction than expected event. In particular, in all three models, $\Delta UneAr^W$ is positive and $\Delta UneAr^L$ is negative. Thus, unexpected events are associated with larger abnormal returns on average. This result is consistent with H_7 and contrasts with our findings in Section 4.3, suggesting that not accounting for simultaneous events may lead to unreliable inference.

Table 6. Average market reaction to expected and unexpected results.

	Abnormal Return (AR percent)					
	SUR 4		SUR 5		SUR 6	
Ar^W	0.509		0.550		0.553	
Wald test (F-test)	3.621	0.000	3.9169	0.000	3.941	0.000
Wald test (Chi-squared-test)	83.272	0.000	90.088	0.000	90.642	0.000
Likelihood ratio test (Chi-squared-test)	80.352	0.000	86.724	0.000	87.435	0.000
Ar^T	-0.702		-0.637		-0.641	
Wald test (F-test)	2.491	0.000	2.323	0.000	2.301	0.000
Wald test (Chi-squared-test)	57.288	0.000	53.421	0.000	52.929	0.000
Likelihood ratio test (Chi-squared-test)	56.305	0.000	52.657	0.000	52.388	0.000
Ar^L	-1.158		-1.169		-1.166	
Wald test (F-test)	3.012	0.000	2.924	0.000	2.934	0.000
Wald test (Chi-squared-test)	69.272	0.000	67.243	0.000	67.478	0.000
Likelihood ratio test (Chi-squared-test)	67.177	0.000	65.091	0.001	65.493	0.000
$\Delta UneAr^W$	0.493		0.492		0.494	
Wald test (F test)	1.572	0.040	1.726	0.017	1.704	0.019
Wald test (Chi-squared)	36.153	0.040	39.686	0.017	39.183	0.019
Likelihood ratio test (Chi-squared)	35.090	0.051	38.505	0.022	38.152	0.025
$\Delta UneAr^L$	-0.639		-0.596		-0.591	
Wald test (F test)	2.348	0.000	2.133	0.000	2.137	0.000
Wald test (Chi-squared)	53.998	0.000	49.050	0.000	49.141	0.000
Likelihood ratio test (Chi-squared)	52.455	0.000	47.765	0.002	47.979	0.002
con _{<i>t</i>}						
<i>Mkt</i> _{<i>t</i>}	No		Yes		Yes	
Year dummies	No		No		Yes	

5.2 – The distribution of individual abnormal returns

Since in the previous section we find that the SUR model leads to different findings with respect to the standard event study approach for H_7 , it is worth inspecting the entire distribution of individual abnormal returns. In order to analyse the potential heterogeneity of the individual coefficients¹⁶ Ar^W , Ar^L , $Ar^W + \Delta UneAr^W$, Ar^T and $Ar^L + \Delta UneAr^L$, in Figure 5 we report five boxplots. Boxplots show the distribution of abnormal returns conditionally to expected and unexpected match results. The extremes of the box correspond to the first and third quartiles (Q_1 and Q_3). The vertical height of the box represents the inter-quartile difference ($\delta_Q = Q_3 - Q_1$), a measure of variability. The horizontal bold line inside each box represents the median, while the black point is the average. Finally, the two external lines correspond to $Q_1 - 1.5\delta_Q$ and $Q_3 - 1.5\delta_Q$, respectively. Observations beyond these lines indicate the presence of extreme values. While the analysis of the previous section was

¹⁶ We use the coefficients estimated for the third SUR model (SUR 6, which controls for the market and year effects), but our results are virtually unaffected by this choice.

limited to providing the average abnormal return related to match results (a measure of central tendency), the boxplots are more informative.

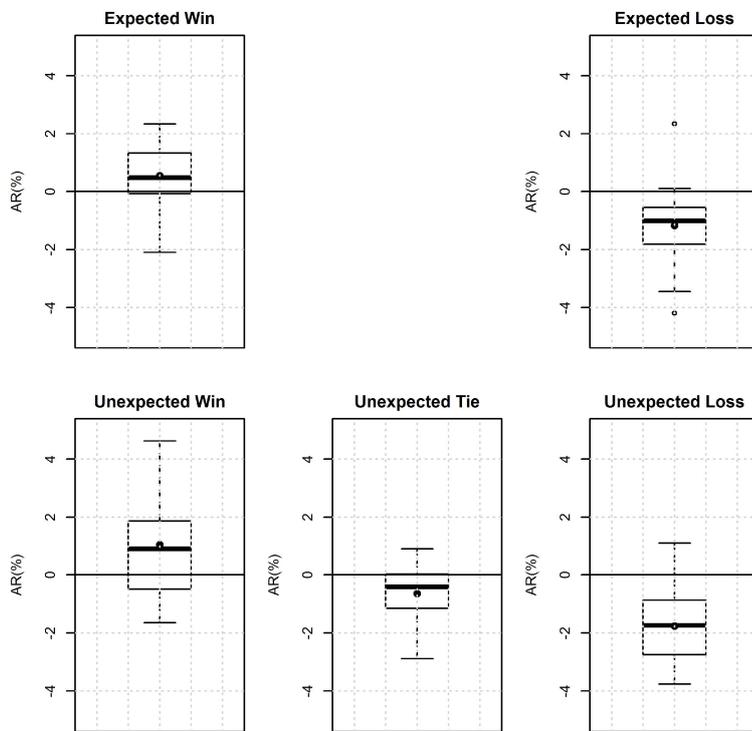


Figure 5. Distribution of individual abnormal returns among soccer teams.

The plots of the first column of Figure 5 present the abnormal returns in case of expected and unexpected wins respectively. Coherently with the result in the previous section, the average (and median) abnormal return in the two plots is positive with a larger abnormal return in case of unexpected wins. Furthermore, the two plots show that the main part of the distribution of individual teams' abnormal returns is characterized by positive abnormal returns. The second plot of the second row of Figure 5 presents the abnormal returns in case of (unexpected) ties. In this case, the average (and median) abnormal return is negative, but the distribution appears to be asymmetric: the median is close to the upper extreme of the box. Finally, the plots of the third column of Figure 5 present the abnormal returns in case of expected and unexpected losses. In this case, the negative market reaction following this event is evident: the average (and median) abnormal return is strongly negative in both plots (with a larger abnormal return in case of unexpected losses) and all the abnormal returns appear to be concentrated in the negative quadrant of the plot around the median.¹⁷ Thus, the SUR model results appear to be largely confirmed when inspecting the distribution of the team-specific coefficients.

¹⁷ The only exception is a positive (non-significant) abnormal return associated to Futebol Clube Do Porto.

6 – Conclusions

Several studies suggest that match results influence investor mood that in turn affects sport team stock returns. Wins are generally associated with a good mood and losses with a bad mood. These emotional reactions affect investor behavior on the market, which reacts positively to wins and negatively to losses.

In this work, we analyse the relationship between soccer match results, bets and stock returns of all listed European soccer teams. Our dataset includes all national and international match results and pre-match betting odds of all 23 listed European soccer teams, in the three year period 2007–09. To measure the aggregate market reaction to match results, we use an event study approach. Our main results can be summarized as follows.

- i. While wins are followed by positive abnormal returns, ties and losses are followed by negative abnormal returns. Furthermore, abnormal returns following losses are larger in magnitude with respect to those following wins. Positive and negative abnormal returns are nonlinearly related to the intensity of a match result, as measured by goal difference.
- ii. Abnormal returns are not affected by the location where the match is played (home vs. away) nor by the year of the competition. However, both the type and the month of competition influence abnormal returns.
- iii. When we use a SUR model to accommodate the problem of event clustering, positive and negative abnormal returns are magnified by unexpected match results. From this point of view, our study makes an empirical contribution to the existing literature.

The objective of this paper was not to identify a profitable trading strategy. However, albeit indirectly, our results suggest some trading rules that may be applied by investors. For example, investors may choose to rebalance their personal portfolios by buying stocks of the teams that are expected to win, and by selling those of the teams that are expected to lose. Future extensions of this paper could attempt to measure the expected profit of this investment strategy. Furthermore, since positive and negative abnormal returns are magnified by goal difference, future research could explore the relationship between the pre-match betting odds on the goal number and the actual goal difference.¹⁸

¹⁸ In this regard, Goddard and Asimakopoulou (2004) and Goddard (2005) propose a model to test betting strategies with a positive expected return.

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References

- Ashton JK, Gerrard B, Hudson R (2003) Economic Impact of National Sporting Success: Evidence from the London Stock Exchange. *Applied Economics Letters* 10 (12):783-785. doi:10.1080/1350485032000126712
- Baker M, Wurgler J (2006) Investor Sentiment and the Cross-Section of Stock Returns. *The Journal of Finance* 61 (4):1645-1680. doi:10.1111/j.1540-6261.2006.00885.x
- Barberis N, Shleifer A, Vishny R (1998) A Model of Investor Sentiment. *Journal of Financial Economics* 49 (3):307-343
- Beck N, Meyer M (2012) Modeling Team Performance. *Empirical Economics* (forthcoming). doi:10.1007/s00181-011-0463-2
- Bernile G, Lyandres E (2011) Understanding Investor Sentiment: The Case of Soccer. *Financial Management* 40 (2):357-380. doi:10.1111/j.1755-053X.2011.01145.x
- Berthier F, Boulay F (2003) Lower Myocardial Infarction Mortality in French Men the day France Won the 1998 World Cup of Football. *Heart* 89 (5):555-556. doi:10.1136/heart.89.5.555
- Brini L, Palomba G (2011) Do Match Results Affect Stock Returns? An Empirical Model for the Italian Football Clubs. Mimeo.
- Brown SJ, Warner JB (1980) Measuring Security Price Performance. *Journal of Financial Economics* 8 (3):205-258
- Brown SJ, Warner JB (1985) Using Daily Stock Returns: The Case of Event Studies. *Journal of Financial Economics* 14 (1):3-31
- Campbell JY, Lo AW, MacKinlay AC (1997) *The Econometrics of Financial Markets*. Princeton University Press, Princeton
- Carroll D, Ebrahim S, Tilling K, Macleod J, Smith GD (2002) Admissions for Myocardial Infarction and World Cup Football: Database Survey. *BMJ* 325
- Chi JS, Kloner RA (2003) Stress and Myocardial Infarction. *Heart* 89 (5):475-476. doi:10.1136/heart.89.5.475
- Czarnitzki D, Stadtmann G (2002) Uncertainty of Outcome versus Reputation: Empirical Evidence for the First German Football Division. *Empirical Economics* 27 (1):101-112. doi:10.1007/s181-002-8361-7

- Daniel K, Hirshleifer D, Subrahmanyam A (1998) Investor Psychology and Security Market Under- and Overreactions. *The Journal of Finance* 53 (6):1839-1885. doi:10.1111/0022-1082.00077
- Edmans A, García D, Norli Ø (2007) Sports Sentiment and Stock Returns. *The Journal of Finance* 62 (4):1967-1998. doi:10.1111/j.1540-6261.2007.01262.x
- Goddard J (2005) Regression Models for Forecasting Goals and Match Results in Association Football. *International Journal of Forecasting* 21 (2):331-340
- Goddard J, Asimakopoulos I (2004) Forecasting Football Results and the Efficiency of Fixed-Odds Betting. *Journal of Forecasting* 23 (1):51-66. doi:10.1002/for.877
- Greene WH (2007) *Econometric Analysis*. Prentice-Hall, Upper Saddle River
- Hirshleifer D (2001) Investor Psychology and Asset Pricing. *The Journal of Finance* 56 (4):1533-1597. doi:10.1111/0022-1082.00379
- Hirt ER, Zillmann D, Erickson GA, Kennedy C (1992) Costs and Benefits of Allegiance: Changes in Fans' Self-Ascribed Competencies after Team Victory versus Defeat. *Journal of Personality and Social Psychology* 63 (5):724-738
- Palomino F, Renneboog L, Zhang C (2009) Information Saliency, Investor Sentiment, and Stock Returns: The Case of British Soccer Betting. *Journal of Corporate Finance* 15 (3):368-387
- Schipper K, Thompson R (1985) The Impact of Merger-Related Regulations Using Exact Distributions of Test Statistics. *Journal of Accounting Research* 23 (1):408-415
- Shiller RJ (2000) *Irrational Exuberance*. Princeton University Press, Princeton
- Trovato F (1998) The Stanley Cup of Hockey and Suicide in Quebec, 1951-1992. *Social Forces* 77 (1):105-126. doi:10.1093/sf/77.1.105
- Wann DL (2001) *Sport Fans: The Psychology and Social Impact of Spectators*. Routledge, New York
- White GF (1989) Media and Violence: The Case of Professional Football Championship Games. *Aggressive Behavior* 15 (6):423-433. doi:10.1002/1098-2337(1989)15:6<423::aid-ab2480150603>3.0.co;2-1
- Zellner A (1962) An Efficient Method of Estimating Seemingly Unrelated Regressions and Tests for Aggregation Bias. *Journal of the American Statistical Association* 57 (298):348-368



Alma Mater Studiorum - Università di Bologna
DEPARTMENT OF ECONOMICS

Strada Maggiore 45
40125 Bologna - Italy
Tel. +39 051 2092604
Fax +39 051 2092664
<http://www.dse.unibo.it>