# The link between national football tournaments and domestic abuse - evidence from England

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# 1 Introduction

Domestic abuse is increasingly recognised as a major public policy concern in many countries, including the UK (Prpic & Rosamund, 2018). While anyone can become a victim of domestic abuse, women are disproportionately affected, with more than 25% of women, and 15% of men in England and Wales reported to have experienced some form of domestic abuse since the age of 16 (Office for National Statistics, 2018a).

The link between sporting events and domestic abuse has been the focus of a number of smaller studies (Williams & Neville, 2014), but large-scale quantitative investigations of this relationship are relatively scarce. In England, previous studies have focused on the link between football (soccer) and domestic abuse. Football's history is inextricably linked to England, and it is by far the most popular sport in the country (Parry, Jones, & Wann, 2014), with the 2018 World Cup attracting a record number of 44.5 million viewers (BBC Sport, 2018). These studies have found significant increases in the number of domestic abuse cases recorded by the police on England win, and particularly on England loss days (Brimicombe & Cafe, 2012; Kirby, Francis, & O'Flaherty, 2014). While domestic abuse is predominantly understood as a pattern of ongoing behaviour involving a series of occurrences, rather than a one-off incident triggered by football (Brooks-Hay & Lombard, 2018), these studies, and other qualitative investigations (Swallow, 2017) nevertheless suggest that national football tournaments can create an environment where domestic abuse is more likely to occur.

Why would national football tournaments, such as the World Cup or the European Championship precipitate domestic abuse? England's participation in these tournaments are times of heightened patriotic emotions and a strengthened sense of "Englishness", fuelled by media narratives that often use war references, and a "us vs. them" rhetoric to generate and represent an English national identity (Vincent & Harris, 2014). Previous qualitative research has suggested that televised contact sports can serve as vehicle for the male sports fan to redefine, and express his masculinity in a way that allows dominance, control, and can ultimately manifest in the perpetration of domestic abuse (Sabo, Gray, & Moore, 2000; Swallow, 2017), given susceptibility to such behaviours. We speculate that this observation is especially pertinent

in the context of England's participation in national football tournaments, owing to the popularity of the sport in the country, the associated media attention, and the resulting heightened sense of national consciousness. Not sure if it would be worth adding something on the link between nationalism and masculinity? It would be good to offer a story for potential pathways whilst avoiding having to say something strong which can be attacked

Previous studies found a strong link between alcohol and domestic abuse, suggesting that alcohol can either be seen as a contributing cause, an aggravating factor, or a trigger of violent behaviour in domestic (and other) settings (Leonard & Quigley, 2017). The ecological model of intimate partner violence (IPV; the most common form of domestic abuse) sees IPV as a product of societal, community, relationship, and individual influences. From a community perspective, (Graham, Wilson, & Taft, 2017) argue that social contexts where excessive drinking is encouraged (e.g., through the promotion of alcohol during sporting events) are often permissive of violent behaviour and sexism, which can increase the likelihood of alcohol-related domestic abuse. However, strong inhibitory factors (e.g., social norms, and more specifically the condemnation of peers) can offset the disinhibitory effect of alcohol consumption, and reduce the likelihood of subsequent perpetration.

Despite the evident links between alcohol and domestic abuse, the role alcohol plays in the link between football and domestic abuse has not yet been explored in a large-scale quantitative study. Given the strong association between drinking culture and football in England (Dixon, 2014), a relationship continuously reinforced by the marketing practices of the alcohol industry (Gornall, 2014), we conjecture that alcohol acts as an aggravating factor in the link between football and domestic abuse in England. Exploring the role alcohol plays in this relationship will deepen our understanding of the pathway through which football increases propensity for violence in domestic settings.

To test this hypothesis, we investigate whether the daily number of reported alcohol and non-alcohol related domestic abuse cases recorded by the third largest police force in England (West Midlands Police; WMP) between 2010 and 2019 increase on days when the England national team plays in a national football tournament (the FIFA World Cup and UEFA European Football Championship), and whether the effect, if any, depends on the result of the

match.

In these tournaments, England match days are allocated randomly (as the fixtures are determined through a random draw), making these events a suitable instrument for testing the casual effect of England matches on domestic abuse They will most certainly say that we can't say causal. Our dataset covers a decade worth of crime data, allowing us to conduct a thorough investigation of the characteristics and temporal pattern of the domestic abuse perpetrated on England match days, and extend our findings to understand various characteristics of this relationship. We also test the robustness of the effect using various model specifications, time periods, and geographical areas.

#### 2 West Midlands Crime Dataset

Our dataset comprises all incidents and crimes recorded by the West Midlands Police (WMP) in the period between January 1, 2010 and October 10, 2019 (3570 days). The WMP is the third largest police force in England (Home Office, 2018), serving an area with a population over 2.9 million in 2018 (Office for National Statistics, 2018b).

The WMP records all reported cases in two databases: incidents and crimes. Most reported cases will be initially recorded as an incident (and then later might be also recorded as a crime after investigation), with a smaller fraction immediately recorded as crime (if it is clear that it meets the criteria for notifiable offences, Home Office, 2019a). For this reason, some recorded crimes will have matching incident entries, while others will not. As an exception, specific types of reported cases (e.g., rape, domestic abuse, child abuse, racially or religiously aggravated cases) that do not qualify as a crime after being investigated (e.g., due to lack of evidence), still get recorded in the crimes database with the indicator "non crime" (e.g., "domestic violence incident - non crime").

The incidents dataset contains a list of markers capturing various key aspects of the reported case, which are recorded in the call centre when the case is being reported. Most of these markers are common across police forces in England (e.g., domestic abuse, alcohol, drugs, mental health, etc.; Home Office, 2011). These then could be used to assess the severity of

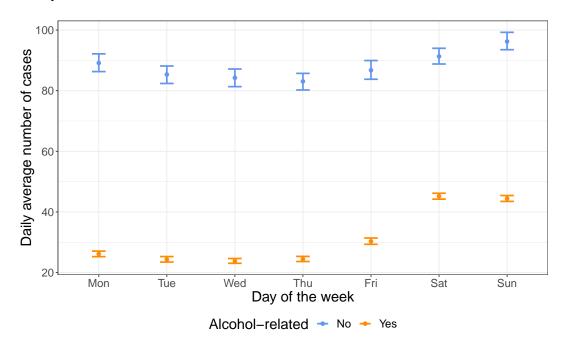
case (and support the decision as to whether police attendance is required). In contrast, each case in the crimes dataset had been attended by the police, and therefore contains detailed information about the offence that took place, the gender, age, and ethnicity of the victim and offender, and the location and exact time of the crime (the time of the first and last instance of the perpetrated case, as well as the time of reporting is recorded to the minute not sure if this grammatically correct). A separate alcohol marker exists in the crimes dataset.

Given that much more information are recorded about crimes than incidents, we decided to restrict our analysis to entries in the crimes dataset (i.e., cases that have been investigated by the police), and use only matching entries from the incidents dataset (if any), which provide additional useful call centre-recorded information about the nature of each case. This allows us to identify if a recorded serious crime (e.g., grievous bodily harm) was domestic abuse-related. However, our main results do not change if we only use information contained in the crimes dataset (i.e., domestic abuse cases that after investigated, did not qualify as a crime) or include both crimes and incidents in the analysis (i.e., domestic abuse cases that had been investigated and suspected domestic abuse cases that had not been investigated).

We coded each reported case as domestic abuse-related if (1) in the crime dataset the offence description explicitly stated so (either "domestic violence incident" or "coercive/controlling behaviour in a domestic setting") or (2) one of the linked incident markers (if any) stated "domestic violence". This method ensured that we used all information available to us to identify domestic-abuse related cases. The role of the victim and the offender was determined by the police.

To determine whether a case was alcohol-related, we again collated information from markers found in the incidents and crimes dataset and created an alcohol flag if any of these markers contained the word "alcohol". The main source of information on alcohol-involvement is the crime dataset, where the recording officer has to choose either "Yes" or "No" to indicate whether the case was alcohol-related. According to the Home Office Counting Rules for Recorded Crime, alcohol-related crimes are defined as "any notifiable offence (crime) where it is perceived, by the victim or any other person, that the effects of alcohol consumption on the offender or victim was an aggravating factor" (Home Office, 2019b). Unfortunately, it is

Figure 1 The daily average number of alcohol and non-alcohol related domestic abuse cases reported to the police, by day of the week. The error bars show bootstrapped 95% confidence intervals, stratified by month.



not specified who was under the influence of alcohol (victim, offender, or both). In addition, it is possible that this flag is an underestimation of the true number of alcohol-related cases, as one could reasonably expect it to be more likely that "No" will be selected for an alcohol-related case than the other way around, due to human error.

Overall there were 427,351 reported cases of domestic abuse in this 10-year period, comprising 17% of all recorded crimes and incidents. Of these cases, 26% were alcohol-related (in contrast, only 9% of non domestic abuse cases are alcohol-related). In the period between 2010 and 2018, the daily rate of non-alcohol related domestic incidents falls between 1.42–3.95 cases per 100,000 individuals, whereas the daily rate of alcohol-related cases falls between 0.84–1.67 cases per 100,000 individuals (the overall rate falls between 2.4 and 5.01 per 100,000 individuals).

Figure 1 shows the daily average number of alcohol and non-alcohol related domestic abuse cases reported to WMP by day of the week. The reported number of domestic abuse cases, and particularly alcohol-related cases shows strong variation throughout the week, with generally fewer reported cases mid-week, and a pronounced increase over the weekend.

In the UK, the term "domestic abuse" refers to a wide range of behavioural patterns, from physical and sexual violence to psychological, emotional, financial abuse, threatening behaviour, stalking and harassment, either within a family or an intimate relationship (Office for National Statistics, 2018a). Previous research has mostly focused on intimate partner violence (IPV), the largest subcategory of domestic abuse. While IPV is more common than abuse perpetrated by family members (Office for National Statistics, 2018a), our dataset does not contain information about the exact relationship between the victim and perpetrator (family vs current or ex-partner), therefore we cannot separate the two types of abuse, and we will refer to them collectively as "domestic abuse".

Our dataset contains all cases of domestic abuse that have been reported to the WMP between 2010 and 2019, but the vast majority of domestic abuse incidents in fact never get reported (according to the Crime Survey of England and Wales, only 17% of domestic abuse victims reported the abuse to the police between April 2017 and March 2018; Office for National Statistics, 2018a). This substantial reporting bias, and its potential correlation with other contextual factors warrants a careful interpretation of the estimates from any quantitative study investigating domestic abuse, and highlights the importance of utilising a mixed methods approach to explore the factors contributing to the prevalence of domestic abuse.

There were three World Cups (2010, 2014, 2018) and two European Championships (2012, 2016) in the period covered by our dataset. All of these tournaments took place in the months of June and July. Table 1 shows the matches of the English national football team in these tournaments. Out of 22 matches overall, England won and lost both eight times, while six matches ended in a draw.

Table 1 Matches of the English national football team in the World Cup and European Championship tournaments between 2010 and 2019, by outcome.

Date	Other Team	Tournament	Year	Outcome
2010-06-12	USA	World Cup	2010	England draw
2010-06-18	Algeria	World Cup	2010	England draw
2010-06-23	Slovenia	World Cup	2010	England win
2010-06-27	Germany	World Cup	2010	<b>England lost</b>
2012-06-11	France	European Championship	2012	England draw
2012-06-15	Sweden	European Championship	2012	England win
2012-06-19	Ukraine	European Championship	2012	England win
2012-06-24	Italy	European Championship	2012	<b>England lost</b>
2014-06-14	Italy	World Cup	2014	England lost
2014-06-19	Uruguay	World Cup	2014	<b>England lost</b>
2014-06-24	Costa Rica	World Cup	2014	England draw
2016-06-11	Russia	European Championship	2016	England draw
2016-06-16	Wales	European Championship	2016	England win
2016-06-20	Slovakia	European Championship	2016	England draw
2016-06-27	Iceland	European Championship	2016	<b>England lost</b>
2018-06-18	Tunisia	World Cup	2018	England win
2018-06-24	Panama	World Cup	2018	England win
2018-06-28	Belgium	World Cup	2018	<b>England lost</b>
2018-07-03	Colombia	World Cup	2018	England win
2018-07-07	Sweden	World Cup	2018	England win
2018-07-11	Croatia	World Cup	2018	England lost
2018-07-14	Belgium	World Cup	2018	<b>England lost</b>

# 3 Methods

To investigate the role alcohol plays in the link between England football matches and domestic abuse, we used a regression framework. Our main results are derived from Poisson and Negative Binomial regressions, where each observation is a day or a three-hour interval in the period between 2010 and 2019, and the outcome variable is the number of cases perpetrated in an observation period. To investigate certain characteristics of the domestic abuse cases perpetrated on England match days, we also used a logistic regression framework. Below is a detailed explanation of the statistical models used in our analyses.

#### 3.1 Count per day analyses

In these regressions, each day in our dataset is an observation (of which there are 3570), and the outcome variable is the number of domestic abuse cases perpetrated on that day (alcoholrelated or not). We classified each day in this period as one of the following: England won (England win, N=8), lost (England lost, N=8) or drew (England draw, N=6), a day after an England match day (After England, N=22), any other day during the months of the tournament (Tournament on, N=106), or any other day during the rest of the year (Non-match day, N=3420). Our aim is to understand the extent to which this classification variable can explain the number of alcohol-related and non-alcohol related cases reported to the police. To control for potential confounders and account for the temporal variability of the number of domestic abuse cases, we also added a set of control variables, including day of the week, month, year, Christmas (covering the period between 24th and 26th of December) and New Year's Eve (1st of January). These holidays are consistently associated with an increased number of domestic abuse cases reported to the police, therefore including these will result in a more precise estimate of the association between football and domestic abuse. Given that our outcome variable is a count, we considered using a Poisson or a negative binomial regression model framework. Formally, if  $C_i$  is the number of reported cases (alcohol or

non-alcohol related) on days i = 1...N, then these models can be formally expressed as

$$\ln(\lambda_i) = \boldsymbol{x}_i^T \boldsymbol{\beta} \tag{1}$$

where  $C_i \sim Pois(\lambda_i)$  or  $C_i \sim NegativeBinomial(\lambda_i\theta)$ ,  $\mathbf{x_i}$  is a vector of covariates, and  $\boldsymbol{\beta}$  is the vector of coefficients.

In the Poisson framework, it is assumed that the mean of the distribution  $(\lambda_i)$  is equal to the variance. However, for certain types of count data, the observed variance can exceed the mean, causing overdispersion. If overdispersion is present in the data, the Poisson model understate the standard error of the estimates, leading to erroneous conclusions. Given the complex nature of domestic abuse, it is likely that our model will not able to fully account for all the factors influencing the number of reported cases on a given day, leading to unobserved heterogeneity and increasing the likelihood of overdispersion if using a Poisson model. Testing for overdispersion in a Poisson model involves testing the null-hypothesis of equidispersion (the mean being equal to the variance).

If there is evidence for overdispersion, a negative binomial model is a more suitable modeling framework, as it includes an additional parameter  $\theta$  to allow the variance to deviate from the mean.

# 3.2 Count per three-hour interval analysis

In this analysis, we used the regression framework described in section 3.1 (either Poisson or negative binomial regression) to investigate the temporal dynamics of the number of domestic abuse cases (alcohol-related or not) perpetrated before and after England matches. To this end, we divided each day in our dataset into eight three-hour periods, the first one starting at 12am, and used these to identify specific time windows around the time of the match (six and three hours before the match, during match, and three, six, twelve, and twenty-four hours after the match). The exact time of the matches vary considerably (the earliest starting at 1pm, and the latest at 11pm). We first identified the three-hour period of the day into which each match falls. If the start and end time of the match did not fall in the same three-hour period,

we chose the three-hour period that covers the larger part of the match (e.g., a 2.5 hour long match starting at 7pm will be assigned to the 6-9pm period and not to the 9pm-12am period). In these regressions, each time window indicator is a dummy variable, therefore the exponentiated coefficients directly represent the relative change in the number of cases during these periods, compared to any other (overwhelmingly non-match) periods. To control for potential confounders, we controlled for year, month, day of week, three-hour period of the day, Christmas and New Year's Eve.

# 3.3 Binary outcome variable analyses

To further understand the characteristics of domestic abuse cases perpetrated on England football days, such as whether these cases are more likely to occur in a public location or end in serious injury, we used a logistic regression framework. Given a binary response variable ( $Y_i$  can either take the value 0 or 1), we wish to model the probability of an event occurring, i.e.,  $\pi = P(Y=1)$ . In a logistic regression framework, we can model this probability as a function of predictor variables with the equation

$$log(\frac{\pi}{1-\pi}) = \boldsymbol{x_i^T}\boldsymbol{\beta} \tag{2}$$

where  $\mathbf{x_i}$  is a vector of covariates, and  $\boldsymbol{\beta}$  is the vector of coefficients.

#### 4 Results

In the following count data analyses, we present results from negative binomial models, as we found evidence for overdispersion in all cases (after testing for overdispersion using the method described in Cameron & Trivedi, 1990). All analyses were performed in R (version 3.6.2).

#### 4.1 Core results

Table 2 shows the results from two negative binomial regressions investigating the link between football and domestic abuse by alcohol-involvement. For non alcohol-related cases, we find no evidence for an increase (or decrease) on days when the England football team plays, compared to non-match days. In contrast, we see a significant, 47%, 95% CI [26%– 71%], increase in the number of alcohol-related cases on days when the England football team wins, compared to non-match days. We find no evidence for comparable increases in the number of reported domestic abuse cases when the England national team loses. Less surprising, and more consistent with previous findings is the lack of evidence for an increase on England draw days, probably due to the fact that high-stake matches after the group-stage in the tournament cannot result in a draw. In addition, we estimate that there is an 18%, 95% CI [7%–30%], increase in the number of alcohol-related cases reported after an England match day, which could be the result of a temporal spill-over effect from the previous match day. To further understand the characteristics of this increase, we exploited the detailed time information in our dataset, which allowed us to conduct a three-hour analysis to explore the finer temporal patterns of domestic abuse cases (alcohol-related or not) perpetrated before and after an England match. Figure 2 shows a plot of the estimated percentage increase from these negative binomial regressions, revealing a stark increase in alcohol-related domestic abuse on days of an England victory, starting in the three hour period of the match, peaking in the three-hour period afterwards, and gradually declining to its original level in the twenty-four hours following the victory. While a similar pattern can be observed on days when England lost, the increase in the number of alcohol-related cases is substantially smaller and not statis-

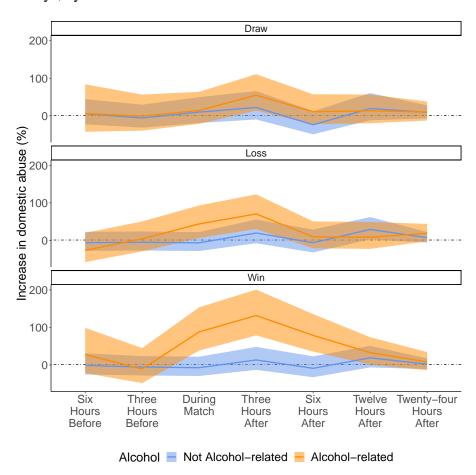
Table 2 The effect of England football days on domestic abuse by alcohol-involvement, 2010-2019

	Dependent v	variable:
	Number of	f cases
	Non-alcohol related domestic abuse	Alcohol-related domestic abuse
	(1)	(2)
Tournament on	1.024	0.984
	(0.986, 1.064)	(0.935, 1.037)
England win	0.950	1.469***
	(0.838, 1.077)	(1.259, 1.714)
England draw	1.045	1.145
_	(0.905, 1.208)	(0.955, 1.375)
England lost	1.033	1.129
	(0.915, 1.167)	(0.963, 1.324)
After England	1.056	1.181***
	(0.979, 1.140)	(1.069, 1.305)
Observations	3,570	3,570

#### Notes:

 $<sup>^{\</sup>rm a}$  \*p<0.1; \*\*p<0.05; \*\*\*p<0.01  $^{\rm b}$  Estimates are exponentiated coefficients from negative binomial regressions with non-match day as benchmark; controls include day of week, month, year, Christmas, and New Year's Eve

Figure 2
The temporal dynamics of the number of domestic abuse cases perpetrated on England draw, lose, and win days, by alcohol involvement.



Note: Estimates are from two separate negative binomial regressions with year, month, day of week, three-hour period of day, Christmas, New Year's eve controls. Shaded area is 95% CIs.

tically different from zero.

## 4.2 Other types of crimes and incidents

Our crime dataset further allows us to explore whether England games have similar effects on other types of criminal behaviours, apart from domestic abuse. Specifically, we are interested in how an England match day affects the number of reported property-related crimes (where the offence class is either theft or robbery), hate crimes (where the offence description includes the words "hate", "racial" or "racially"), and other violent crimes (where the offence class is classified as violence against the person, after excluding cases of domestic abuse). Of particular interest is the effect of football on non-domestic violent crimes, since it is possible that alcohol-fuelled violence that follows an England victory is not limited to family and intimate partner relationships.

I think we should say something about how detecting alcohol-involvement varies by crime type- e.g., for domestic abuse we have greater confidence in identifying alcohol-related cases as opposed to property-related crimes

Table 3 shows the results for these three types of criminal behaviours, by alcohol-involvement. Compared to non-match days, hate crimes and incidents with no alcohol-involvement increase by 12%, 95% CI [3%–22%], during the months of the tournaments. Similarly, compared to non-match days, alcohol-related thefts and robberies (property-related offences) increase by 10%,95% CI [1%–19%], when the tournament is on, and by 29%, 95% CI [0%–66%], when England wins, although these estimates are not very precise (demonstrated by the width of the confidence intervals), due to the relatively small number of alcohol-related theft and robbery cases. In contrast, the effect of an England win on alcohol-related cases clearly extends to other, non-domestic violent offences, resulting in a 47%, 95% CI [19%–83%], increase compared to non-match days. While it is possible that misclassified domestic abuse cases contribute to this result (e.g, if the victim chooses not to disclose any relationship to the offender), these results suggests that the increase in alcohol-related violence following an England victory is not limited to domestic settings.

The effect of England football days on hate-related, property-related, and other violent offences by alcohol-involvement Table 3

			Dependent variable: Number of cases	ependent variable: Number of cases		
	Z	Non-alcohol related	þ		Alcohol-related	
	Hate- related	Property- related	Other violent	Hate- related	Property related	Other violent
	(1)	(2)	(3)	(4)	(5)	(9)
Tournament on	1.120*** (1.030, 1.216)	1.014 (0.986, 1.043)	1.035 (0.986, 1.085)	1.099 (0.914, 1.316)	1.099** (1.011, 1.194)	1.060 (0.987, 1.139)
England win	1.002 (0.759, 1.311)	0.961 (0.879, 1.052)	1.085 (0.935, 1.263)	1.556* (0.902, 2.558)	1.291** (0.998, 1.662)	1.473*** (1.192, 1.826)
England draw	0.993 (0.709, 1.369)	0.953 (0.859, 1.059)	0.970 (0.810, 1.167)	0.848 (0.400, 1.624)	0.934 (0.701, 1.237)	1.251* (0.974, 1.613)
England lost	0.984 (0.744, 1.288)	0.974 (0.890, 1.066)	1.066 (0.917, 1.243)	0.760 (0.363, 1.427)	1.008 (0.744, 1.353)	0.988 (0.785, 1.247)
After England	1.124 (0.953, 1.323)	0.978 (0.925, 1.035)	1.085* (0.988, 1.193)	1.244 (0.872, 1.742)	1.047 (0.884, 1.237)	1.092 (0.948, 1.258)
Observations	3,570	3,570	3,570	3,570	3,570	3,570
Note:					*p<0.1; **p<(	*p<0.1; **p<0.05; ***p<0.01

 $<sup>^{</sup>a}$  \*p<0.1; \*\*p<0.05; \*\*\*p<0.01 $^{b}$  Estimates are exponentiated coefficients from a series of negative binomial regressions with non-match day as benchmark; controls include day of week, month, year, Christmas, and New Year's Eve

## 4.3 Other types of sports: rugby

Does this effect generalise to other sporting events, or is it specific to football? It has been previously suggested that other popular sports, such as rugby have similar links with domestic abuse (Brooks-Hay & Lombard, 2018). Rugby is the second most popular sport in England after football (*Rugby Union 'Britain's Second Most Popular Sport'*, 2003). Focusing on the Six Nations, a high-profile rugby tournament that takes place every year in the months of February and March with the participation of England, Wales, Scotland, Ireland, France and Italy, we explored whether the reported number of domestic abuse cases increase on days when the England national rugby team plays.

Each tournament year, England plays one match with each of the other five participating countries, and these matches take place almost always during the weekend (Saturday, N=37; Sunday, N=11; Friday, N=2). The most common result for England is to win (N=35), much less frequent are losses (N=12) and draws (N=2).

Table 4 shows no comparable effects for rugby matches on domestic abuse, potentially stemming from differences in timing, media coverage, audience numbers, and the involvement of alcohol between the two tournaments.

Table 4 The effect of England rugby days (Six Nations) on domestic abuse by alcohol-involvement

	Dependent v	variable:
	Number of	f cases
	Non-alcohol related domestic abuse	Alcohol-related domestic abuse
	(1)	(2)
Tournament on	0.980	0.969
	(0.947, 1.014)	(0.923, 1.017)
England win	0.967	1.058
_	(0.906, 1.033)	(0.973, 1.151)
England draw	1.030	0.966
	(0.810, 1.318)	(0.716, 1.308)
England lost	1.067	0.989
	(0.963, 1.184)	(0.867, 1.129)
After England	0.983	1.004
	(0.929, 1.041)	(0.931, 1.083)
Observations	3,570	3,570

#### Notes:

 $<sup>^{\</sup>rm a}$  \*p<0.1; \*\*p<0.05; \*\*\*p<0.01  $^{\rm b}$  Estimates are exponentiated coefficients from negative binomial regressions with non-match day as benchmark; controls include day of week, month, year, Christmas, and New Year's Eve

#### 4.4 Previous evidence

A previous study by (Kirby et al., 2014) found that an England loss results in the most pronounced increase in domestic abuse (38%), and a win or draw have a slightly smaller effect (26%). They used daily counts of IPV in Lancashire from the months of the 2002, 2006, 2010 World Cups. Upon re-analysing their data by treating wins and draws as two separate variables (resulting in an improved model fit, see Table 5), we see a roughly similar effect for wins (45%, 95% CI [28%–64%]) and losses (39%, 95% CI [18%–64%]), and no effect when England draws. Separating wins and draws clearly improves the fit of the model, demonstrated by the reduction in the Bayesian and Akaike Information Criteria (BIC and AIC, respectively). However, the estimated effect of an England loss is insensitive to this change. In conclusion, our reanalysis replicates the win effect seen in the West Midlands data, though the absence of a loss effect remains a stark difference between the two studies.

Table 5
Re-analysis of Kirby et al. (2014) with wins and draws as separate variables

	De	pendent variable:
	Number of reported In	timate Partner Violence cases per day
	Original Model	Win/Draw Separate
	(1)	(2)
England windraw	1.256***	
	(1.128, 1.398)	
England lost	1.382***	1.388***
	(1.151, 1.662)	(1.175, 1.639)
After England	1.111**	1.113**
-	(1.004, 1.228)	(1.014, 1.221)
England win		1.452***
		(1.282, 1.644)
England draw		1.032
		(0.893, 1.191)
Observations	92	92
BIC	747.763	739.662
AIC	714.980	704.356

a \*p<0.1; \*\*p<0.05; \*\*\*p<0.01

To explore the underlying reason for this discrepancy and test the robustness of our results,

<sup>&</sup>lt;sup>b</sup> Estimates are exponentiated coefficients from negative binomial regressions (based on tests of overdispersion) with year and day of week controls and no England match days during the tournament as benchmark; data is only available during the tournament period

we find it instructive to break our analysis into specific tournament years for the two (Lancashire and West Midlands; see Table 6). To be consistent with previous results, we present the regressions for both types of domestic abuse (alcohol and non-alcohol related), however, unfortunately we cannot do the same for the Lancashire dataset, as we have no information on alcohol-involvement. Only year 2010 overlaps in the two datasets, which allows us to at least partly evaluate the robustness of the England win effect.

An interesting common pattern in both datasets is the large effect of England's victory over Slovenia in the group stage of the 2010 World Cup, which, after much anticipation, secured their progression to the next stage of the tournament. Equally, the subsequent loss against Germany in the knockout stage resulted in a substantial increase in the number of reported domestic abuse incidents, which is the strongest loss effect in our dataset. In addition, the size of the win and loss coefficients are remarkably similar, giving us confidence that these effects are likely to generalise to other areas within England. These patterns are also in line with findings from an earlier examination by (Brimicombe & Cafe, 2012), who found an increase in the number of domestic abuse cases reported to the police across England on both victory and loss days during the 2010 World Cup. Most importantly, if the same mechanism lies behind our results, then it suggests that the increase observed in other datasets are also likely to stem from alcohol-related cases.

It also clear that World Cup wins have the strongest effect on the number of domestic abuse incidents: a clear, large increase can be observed in 2002, 2006, 2010, and 2018 (England did not have a single victory in the 2014 World Cup). The number of domestic abuse cases also increase after England losses in World Cup tournament years, but this effect is usually about half the size of the win effect (except for 2006).

As a further robustness check, we examine the sensitivity of the result to the exclusion of specific tournament years (see Table 7). This analysis shows that while the size of the England win effect varies by year, it is robust to the exclusion of specific tournament years, and is not driven by an unusually large effect in one of the tournament years.

The effect of England football days on domestic abuse by alcohol-involvement and tournament, Lancashire and West Midlands data Table 6

						Number	Dependent variable: Number of domestic abuse cases	le: se cases					
		Lancashire						West Midlands	idlands				
	All	All	All	$NA^{\circ}$	$A^c$	NA	Ą	NA	A	NA	٨	NA	٨
	2002	2006	2010	2010	2010	2012	2012	2014	2014	2016	2016	2018	2018
	(1)	(2)	(3)	(4)	(5)	(9)	(7)	(8)	(6)	(10)	(11)	(12)	(13)
Tournament on				1.047	1.038	0.881	0.936	0.976	0.999	1.041	0.855**	1.081	1.014
				(0.974, 1.127)	(0.974, 1.127) $(0.936, 1.151)$	(0.740, 1.049)	(0.780, 1.125)	(0.901, 1.057)	(0.890, 1.123)	(0.962, 1.128)	(0.757, 0.967)	(0.979, 1.194)	(0.889, 1.155)
England win	1.596***	1.297***	1.916***	1.171	1.845***	0.673**	1.106			1.062	1.302	1.006	1.529***
	(1.186, 2.153)	(1.115, 1.505)	1.186, 2.153) (1.115, 1.505) (1.528, 2.389)	(0.897, 1.514)	$\Box$	(0.456, 0.982)	(0.770, 1.576)			(0.796, 1.420)	(0.834, 1.995)	(0.844, 1.202)	(1.249, 1.869)
England draw	1.100	1.098	0.863	0.905	1.248*	1.204	1.116	0.984	0.954	1.069	1.128		
)	(0.820, 1.477)	(0.820, 1.477) $(0.802, 1.481)$	(0.713, 1.037)		(0.966, 1.611)	(0.773, 1.877)	(0.676, 1.805)	(0.723, 1.343)	(0.576, 1.527)	(0.869, 1.318)	9		
England lost	1.200	1.373***	1.568***		1.082 1.435**	0.812	0.961	0.934	1.079	0.772*	$0.620^{*}$	1.101	1.252*
	(0.759, 1.894)	(1.086, 1.721)	(0.759, 1.894) $(1.086, 1.721)$ $(1.270, 1.928)$	(0.838, 1.389)	(1.036, 1.992)	(0.508, 1.293)	(0.614, 1.491)	(0.747, 1.170)	(0.792, 1.459)	(0.568, 1.048)	(0.343, 1.058)	(0.904, 1.348)	(0.975, 1.603)
After England	1.253**	1.122*	1.024	1.049	1.132	0.814	1.108	1.037	1.080	1.013	096.0	1.147*	$1.390^{***}$
	(1.029, 1.526)	(0.977, 1.286)	1.029, 1.526)  (0.977, 1.286)  (0.900, 1.162)  (0.905, 1.194)  (0.933, 1.331)	(0.905, 1.194)	(0.933, 1.331)	(0.535, 1.092)	(0.834, 1.381)	(0.863, 1.249)	(0.832, 1.394)	(0.853, 1.172)	(0.715, 1.206)	(1.001, 1.317)	(1.173, 1.646)
Observations	30	32	30	365	365	366	366	365	365	366	366	365	365
Note:												*p<0.1; **p<0	*p<0.1; **p<0.05; ***p<0.01
1													•

a \*p<0.1; \*\*p<0.05; \*\*\*p<0.01

b Estimates are exponentiated coefficients from a series of negative binomial regressions, except for Model (2) and (3) which are from Poisson regressions (based on tests of overdispersion). Each regression is restricted to a specific tournament year. Regressions 1-3 have day of week control with non England tournament days as benchmark, regressions 4-13 have month, day of week, Christmas, New Year's eve controls, with non-match days as benchmark. c All = All cases, NA = Non-alcohol related cases, A = Alcohol-related cases.

Robustness of the result: sensitivity to the exclusion of specific years Table 7

					Dependent variable:	t variable:				
					Number of dome	Number of domestic abuse cases				
	$NA^c$	$A^c$	NA	A	NA	A	NA	А	NA	A
	2010	2010	2012	2012	2014	2014	2016	2016	2018	2018
	(1)	(2)	(3)	(4)	(5)	(9)	(7)	(8)	(6)	(10)
Tournament on	1.035	0.959	1.063*	$1.084^{**}$	0.976	1.085**	0.960	1.105***	0.948	1.069*
	(0.946, 1.133)	(0.901, 1.021)	(0.994, 1.136)	(1.014, 1.160)	(0.890, 1.073)	(1.010, 1.166)	(0.877, 1.052)	(1.029, 1.186)	(0.870, 1.034)	(0.997, 1.146)
England win	0.957	1.385***	1.143	$1.674^{***}$	0.979	1.528***	0.935	1.558***	*869.0	1.583***
	(0.727, 1.289)	(1.154, 1.666)	(0.909, 1.457)	(1.344, 2.100)	(0.753, 1.299)	(1.254, 1.873)	(0.709, 1.262)	(1.262, 1.938)	(0.486, 1.040)	(1.192, 2.127)
England draw	1.037	1.126	0.660	1.347**	0.921	1.383**	0.793	1.391**	0.950	1.304**
	(0.724, 1.545)	(0.873, 1.456)	(0.769, 1.297)	(1.056, 1.734)	(0.662, 1.324)	(1.079, 1.788)	(0.550, 1.190)	(1.051, 1.862)	(0.708, 1.306)	(1.034, 1.656)
England lost	1.102	1.090	1.140	1.218*	1.096	1.213	1.128	1.263**	0.902	1.134
	(0.837, 1.482)	(0.901, 1.322)	(0.922, 1.428)	(0.986, 1.512)	(0.813, 1.520)	(0.964, 1.538)	(0.856, 1.521)	(1.021, 1.574)	(0.655, 1.280)	(0.880, 1.475)
After England	1.095	1.198***	1.169**	1.287***	1.054	1.283***	1.043	1.316***	0.891	1.205**
	(0.914, 1.276)	(1.077, 1.319)	(1.030, 1.307)	(1.151, 1.422)	(0.874, 1.234)	(1.147, 1.418)	(0.860, 1.225)	(1.177, 1.455)	(0.693, 1.088)	(1.050, 1.360)
Observations	3,205	3,205	3,204	3,204	3,205	3,205	3,204	3,204	3,205	3,205
Note:									*p<0.1; **p<(	*p<0.1; **p<0.05; ***p<0.01

 $^{a*}p<0.1;^{**}p<0.05;^{***}p<0.01$ <sup>b</sup> Estimates are exponentiated coefficients from a series of negative binomial regressions (based on tests of overdispersion) with year, month, day of week, Christmas, New Year's eve controls  $^{c}$  NA = Non-alcohol related cases, A = Alcohol-related cases

Characteristics of domestic abuse cases perpetrated on England football days I - non-repeat cases, public perpetrated cases, serious injury Table 8

			Dependent variable:	t variable:		
			Number of dome	Number of domestic abuse cases		
	Ž	Non-Alcohol related			Alcohol-related	
	Newly Reported	Public Location	Serious Injury	Newly Reported	Public Location	Serious Injury
	(1)	(2)	(3)	(4)	(5)	(9)
Tournament on	0.993 (0.941, 1.047)	1.038 (0.967, 1.114)	1.050 (0.983, 1.121)	1.093* (0.986, 1.212)	1.042 (0.904, 1.201)	0.931 (0.831, 1.042)
England win	1.058 (0.879, 1.273)	1.165 (0.943, 1.439)	1.092 (0.883, 1.350)	1.268* (0.959, 1.676)	1.079 (0.753, 1.547)	0.793
England draw	1.139 (0.884, 1.466)	0.903 (0.686, 1.190)	1.156 (0.887, 1.507)	0.890 (0.611, 1.296)	1.045 (0.660, 1.652)	1.064 (0.723, 1.564)
England lost	0.981 (0.836, 1.150)	1.380*** (1.136, 1.678)	1.206* (0.990, 1.470)	1.055 (0.769, 1.447)	1.076 (0.742, 1.560)	0.923 (0.660, 1.291)
After England	1.042 (0.945, 1.148)	1.112 (0.976, 1.267)	0.981 (0.855, 1.126)	0.983 (0.811, 1.191)	0.967 (0.766, 1.220)	1.035 (0.850, 1.260)
Observations	260,466	282,400	282,400	76,309	89,153	89,153

 $<sup>^{</sup>a}$   $^{*}p<0.1;$   $^{***}p<0.05;$   $^{***}p<0.01$   $^{b}$  Estimates are log odds from a series of logistic regressions with year, month, day of week, Christmas, New Year's eve controls, where every observation is a reported domestic abuse case; cases that happened in 2010 were excluded from models 1 and 4; confidence intervals are constructed from standard errors clustered by victim-offender pairs

Characteristics of domestic abuse cases perpetrated on England football days II - days since last incident, days until next incident, hours elapsed between incident and reporting Table 9

			Dependen	Dependent variable:		
	Z	Non-Alcohol related	Number of domed	Number of domestic abuse cases	Alcohol-related	
	Days since last	Days until next	Hours until reported	Days since last	Days until next	Hours until reported
Tournament on	-0.007 (0.941, 1.047)	-0.007 (0.945, 1.045)	0.054 (0.940, 1.184)	-0.0001 (0.915, 1.093)	-0.008 (0.911, 1.080)	0.021 (0.760, 1.373)
England win	0.130* $(0.978, 1.327)$	-0.130 (0.741, 1.040)	-0.173 (0.600, 1.179)	-0.087 (0.704, 1.192)	0.149 (0.949, 1.418)	$-0.054 \\ (0.388, 2.314)$
England draw	0.066 (0.855, 1.335)	-0.140 (0.704, 1.073)	-0.169 (0.557, 1.279)	-0.045 (0.701, 1.305)	0.123 (0.872, 1.466)	-0.264 (0.332, 1.773)
England lost	0.006 (0.853, 1.186)	0.056 (0.907, 1.233)	$-0.470^{***}$ (0.443, 0.882)	0.080 (0.830, 1.414)	-0.039 $(0.758, 1.220)$	0.068 (0.329, 3.481)
After England	0.114** (1.014, 1.238)	0.036 (0.946, 1.137)	$-0.284^{***}$ (0.607, 0.934)	-0.027 (0.827, 1.145)	0.031 (0.885, 1.204)	-0.508** (0.367, 0.986)

a \*p<0.1; \*\*p<0.05; \*\*\*p<0.01

sion) with month, day of week, Christmas, New Year's eve controls; every observation is a reported domestic abuse case; <sup>b</sup> Estimates are exponentiated coefficients from a series of negative binomial regressions (based on tests of overdisper-

 $^*p<0.1; ^{**}p<0.05; ^{***}p<0.01$ 

confidence intervals are constructed from standard errors clustered by victim-offender pairs

Table 10 Alcohol transition in repeat domestic abuse cases

	Dependent v	variable:
	Alcohol-involvemen	nt in current case
	Previous case not alcohol-related	Previous case alcohol-related
	(1)	(2)
Tournament on	0.907 (0.806, 1.022)	0.846*** (0.745, 0.961)
England win	1.311 (0.909, 1.890)	1.168 (0.773, 1.765)
England draw	1.554** (1.066, 2.265)	1.339 (0.723, 2.481)
England lost	1.116 (0.802, 1.553)	1.225 (0.762, 1.970)
After England	1.070 (0.873, 1.311)	1.119 (0.852, 1.469)
Observations	95,813	34,156

<sup>&</sup>lt;sup>a</sup> \*p<0.1; \*\*p<0.05; \*\*\*p<0.01

b Estimates are log odds from a series of logistic regressions with year, month, day of week, Christmas, New Year's eve controls, where every observation is a reported and repeated domestic abuse case; confidence intervals are constructed from standard errors clustered by victim-offender pairs

# **Discussion**

We have shown that an England victory in a national football tournament is followed by a 61% increase in the reported number of alcohol-related domestic abuse cases, and that the temporal pattern of the increase suggests a causal mechanism. The effect is entirely limited to alcohol-related abuse, even though alcohol-related domestic abuse cases comprise only 23% of all domestic abuse in our dataset. As such, we see this as strong quantitative evidence that alcohol plays an instrumental role in the relationship between football and domestic abuse. This effect is exclusively limited to male-perpetrated domestic abuse, implicating a victoryinduced violent expression of masculinity coupled with alcohol consumption as the pathway by which football increases abuse. In line with this, anecdotal evidence suggests that alcohol consumption increases following an England victory (Davies, 2018). It had been previously suggested that the apparent link between football and domestic abuse can be explained by other factors, including high-profile events taking place around the time of the match, increased policing on England match days, and the effect of awareness campaigns before the tournaments (Brooks-Hay & Lombard, 2018). However, our three-hour analysis strongly suggests that the effect is causal, further supported by the fact that the majority of England match days were allocated randomly. In addition, we could expect that higher levels of policing on England match days would result in an increased number of recorded cases perpetrated outside, and that a successful pre-tournament awareness campaign would result in an increase in the number of newly reported cases. Our results presented in Table ?? in the Supplemental Material do not support either of these alternative hypotheses in that we do not see more newly reported, or publicly perpetrated cases of abuse on England win days. Furthermore, it is unclear why the effect of other events, different policing practices, or awareness campaigns would depend on the result of the match.

One limitation of our analysis is that it only relies on cases recorded by the police, while the vast majority of domestic abuse cases do not get reported. Future investigations of the link between football, alcohol, and domestic abuse should aim to combine data from various sources (e.g., police data, calls to shelters, hospital admissions, etc.) to address the potential bias aris-

ing from underreporting.

Our study represent the most extensive quantitative investigation of the link between football and domestic abuse up to date. We have shown that our findings are largely in line with results from previous, smaller-scale investigations from England (the main difference being the lack of an England loss effect in our sample), and substantially extend them in highlighting the instrumental role alcohol plays in the relationship between football and domestic abuse. While the effect of a victory or loss is likely to be highly specific to the context of a particular match (e.g., group stage or knockout stage, previous performance of the team, weather on the day, etc.), the estimated effect of an England victory on the number of reported domestic abuse cases is robust to different model specifications, using data from a different geographical area, and the exclusion of specific tournament years.

Based on the pre-match betting odds, all of the England victories were expected in our dataset. This suggests that in the context of England's participation in national football tournaments, it is living up to the expectations of the fans that results in largest emotional effect and affects levels of alcohol consumption. Indeed, English newspapers' narratives about the team's performance in these tournaments are characterised with high levels of optimism, expectation and yearning for the glory of the 1966 World Cup (Vincent, Kian, Pedersen, Kuntz, & Hill, 2010). Previous research has demonstrated how the vicarious experience of watching their team play can increase supporter's testosterone and cortisol levels, even when they expect their team to win, suggested to be an adaptive response to the perceived threat to one's social identity (van der Meij et al., 2012).

For victims, domestic abuse does not occur once every four years following a football match, but is a lived experience of constant fear (Brooks-Hay & Lombard, 2018). Nevertheless, our results provide a deeper understanding of the contexts that can be conducive to abuse. In particular, these findings illuminate that the experience of "national success" in a highly male-dominated sport is a breeding ground for male-perpetrated, alcohol-related domestic abuse.

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# **Supplemental Material**