

1 **Environmental and field characteristics associated with lameness**  
2 **in sheep: A study using smartphone lameness app for data**  
3 **recording**

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## 1 **Abstract**

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2 Background: Sheep lameness is a major concern among farmers and policy makers with significant  
3 impacts on animal welfare standards as well as financial and production performance. The present study  
4 attempts to identify the relative importance of environmental and farm-level management  
5 characteristics on sheep lameness.

6 Method: To address this objective, data was derived from the SPiLaMM project from 18 farms that  
7 used smartphone app to collect data, the British Geological Survey (BGS) and the Meteorological  
8 Office (Met Office) over 2016-2018. Data was analysed using a Multilevel Poisson Regression model.

9 Results: Temperature and higher length of pasture had a positive relationship with lameness while  
10 concentration of Selenium in soil and flock size had a negative relationship with lameness. In addition,  
11 results showed lower lameness levels for the bedrock class Mudstone, siltstone, limestone and  
12 sandstone (MSLS) in comparison to Sandstone (SD) and finally, lambs and ewes younger than 1 year  
13 old had lower levels of lameness than older ewes.

14 Conclusion: Findings of the present approach show the potential use of data collected via a smartphone  
15 app to study the epidemiology of disease. Furthermore, factors identified could be validated in  
16 intervention studies and generate data driven disease predictive models.

# 1 Introduction

2 Sheep lameness has been acknowledged by both farmers and veterinarians as one of the most significant  
3 issues for animal welfare (Kaler et al., 2010) mainly due to the pain that affected animals experience  
4 (O Callaghan et al., 2003). Furthermore, animal health problems relate not only to poor animal welfare  
5 but also to reduction of productivity (Gelasakis et al., 2015; King et al., 2016) along with financial  
6 losses (Wassink et al., 2010a). It has been estimated that lameness triggers annual losses of £24 to £84  
7 million within the UK sheep industry (Nieuwhof and Bishop, 2005; Wassink et al., 2010b).

8 Most of the lameness incidents in the UK are triggered by footrot (FR) (Kaler and Green, 2008) which  
9 is an infectious disease and is caused by the anaerobic bacterium, *Dichelobacter nodosus* (Beveridge,  
10 1941). FR has two clinical forms, in the first there is inflammation of the interdigital skin of the foot  
11 (interdigital dermatitis) while in the second the hoof horn gets separated from the underlying tissue  
12 (severe FR) (Witcomb et al., 2014). In the UK, the majority of the observational epidemiological  
13 research in the past decade has focussed on identifying farm-level management factors affecting the  
14 levels of lameness. The latter resulted in the development of current best practice on managing lameness  
15 in sheep. More specifically, management factors including isolation of purchased and returning sheep,  
16 vaccination of ewes with Footvax (MSD, n.d.) as well as prompt treatment of lame sheep within 3 days  
17 (without trimming) have been reported to be associated with lower prevalence while, routine trimming  
18 and foot bathing of ewes associate with higher prevalence of lameness (Kaler et al., 2012, 2010; Kaler  
19 and Green, 2008b; Winter et al., 2015). Interestingly, in an intervention study by Wassink et al. (2010b),  
20 the best practice recommendations were able to reduce the prevalence (number of existing cases at a  
21 particular time point) but unable to reduce the incidence (number of new cases during a specific time  
22 period) of lameness in one group. This highlights the importance of other factors possibly environmental  
23 factors that could influence the susceptibility of sheep and contribute to transmission of *D. nodosus* and  
24 thus, leading to disease. Unlike the increased knowledge that has emerged on management risk factors,  
25 the understanding of environmental characteristics that affect prevalence levels remains less studied in  
26 the UK and worldwide in general.

27 A review by Green and George (2008) highlighted that very little is known on survival times of *D.*  
28 *nodosus* and how soil characteristics such as type and chemical composition influence this. Within this  
29 context, recent literature has examined *D. nodosus* survival in reference to soil type (sand, silt, clay)  
30 and weather conditions (temperature) (Cederlöf et al., 2013; Muzafar et al., 2016). Furthermore,  
31 experimental approaches demonstrate that presence of micronutrients such as Selenium (Se) enhance  
32 the restoration of sheep immune functions hindered by FR and it has been suggested that exposing sheep  
33 to forage with increased Se is a sufficient way to achieve this natural restoration (Hall et al., 2013, 2009;  
34 Hujeriletu et al., 2013). However, little is known about the extent and the ways that variation in  
35 concentration of such chemicals in the soil (including Se) along with the type of soil (bedrock) associate

1 with sheep lameness levels. Furthermore, observational studies have examined the effects of seasonality  
2 on prevalence demonstrating that the seasons of Autumn and Summer relate to increased lameness  
3 levels (Angell et al., 2018, 2015). Precision livestock farming offers a solution to record data on  
4 individual animals (e.g. smartphone app, EID technologies and sensors) and also enable the combination  
5 of data with other digital information systems (e.g. GIS) which could be beneficial for understanding  
6 the epidemiology of diseases (Kaler and Ruston, 2019). The aim of this study is collect data on lameness  
7 using lameness smartphone app and to explore associations between sheep lameness and potential risk  
8 factors among farm and field level features, chemical characteristics of soil, weather conditions and  
9 seasonality from a longitudinal study. Findings from this study enable a novel understanding of  
10 lameness in relation to the importance of environmental influencing factors.

## 11 2 Materials and methods

### 12 2.1 Dataset and variables

13 Farm characteristics and lameness levels were derived from the dataset constructed by the SPILLAM  
14 project<sup>2</sup>. SPILLAM is a project aiming to address the challenge of sheep lameness through developing  
15 hardware and software systems for lameness data collection. The software used in the study was  
16 developed by Farmwizard (Agriwebb company) in collaboration with University of Nottingham and  
17 Dunbia as part of a project funded by Innovate UK. The lameness smartphone app had the ability to a)  
18 record: information on individual sheep, lameness in field for groups, lameness treatments given, b)  
19 inform: videos and photos of lesions and lameness for farmers to support lameness recording, lameness  
20 best practice and seasonal advice and c) alert: reminders to record, significant changes in lameness  
21 levels (<https://www.farmwizard.co.uk/sheep-manager>). A total of 18 farms participated in this study.  
22 These farms were given training on using the smartphone lameness app. Also, training on lameness  
23 recording was provided to the farmers by a trained researcher who then also validated independently  
24 the lameness estimates. Lameness levels were validated as per methodology in King and Green (2011).  
25 Briefly, the researcher visited the farm and independently assessed lameness. A sample of lame (up to  
26 30) sheep were turned and lesions were recorded where the most common cause of lameness was found  
27 to be footrot (>97%).

28 Farmers logged the lameness levels for their fields from 2016 to 2018 onto their smartphone app, while  
29 sheep needed to be in the field for 14 days for a recording to occur. Minimum interval for lameness to  
30 be recorded was set to be one week. Farmers recorded number of animals in the field and also recorded  
31 lameness scores (0-6) for all animals based on Kaler et al. (2008). Based on that scale, sheep with  
32 locomotion lameness scores of 2 and above were considered lame in the context of the present analysis.

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<sup>2</sup> Further information on the SPILLAM project can be found in the following link:  
<https://spilamm2017.wixsite.com/spilamm>

1 In terms of management of lameness, all farms treated lameness based on the best practice (Kaler et al.,  
2 2010). With regards to grass length measurements, farmers were given a sward stick and pasture length  
3 classed were explained. Farmers were asked to use this, however this was not validated. Even though  
4 there could be an error in the measurements, it does not affect the results per se as we don't expect these  
5 measurements to be biased. Further information on the raw data for pasture length in relation to count  
6 lame sheep is presented in fig. 1.

7 Using the farm location, environmental data (bedrock classification, soil chemical composition, weather  
8 characteristics) was derived from the British Geological Survey (BGS) (Rawlings et al. 2012, Smith,  
9 2013) and the Meteorological Office (Met office) (Met Office, 2012). Weather variables included  
10 precipitation, mean, maximum and minimum temperature. The Met office variables comprised daily  
11 recordings and thus, were averaged in three different ways using three, seven and fourteen day rolling  
12 average windows representing the days immediately prior to the readings. Different combinations of  
13 the derived averages were modelled and the final selection was subject to performance of the model  
14 considering the AIC score.

15 The geographic reference of the farm businesses was obtained from farm addresses and thus, data from  
16 the three different sources (SPILaMM, BGS, Met Office) were merged geographically into a  
17 Geographic Information Systems (GIS) environment by spatial overlap. Detailed information on the  
18 examined variables is presented in Table 1. Finally, through a data cleaning process, ten recordings  
19 were omitted from the subset due to duplicates or missing values and errors. The final dataset includes  
20 521 observations of sheep lameness at field level for the 18 farms capturing information for the years  
21 2016, 2017 and 2018. Specifically, data collection started in August 2016 and finished in September  
22 2018 and number of observations were 87, 374 and 60, respectively for the three years.

23 Regarding the study area of the analysis, farm businesses under consideration spanned geographically  
24 across England and Wales while the majority of them was located in the regions of Midlands and  
25 Southwest along with Wales (Fig. 2).

## 26 2.2 Multilevel Poisson Regression model

27 The employed dataset comprised of repeated lameness recordings over time for the various farms and  
28 fields and thus, multilevel statistical analysis was incorporated to account for the nested observations  
29 (Solano et al., 2015). Furthermore, a Poisson model was employed as count data on lame sheep per field  
30 has been used as the dependent variable in the modelling (Winter et al., 2015). Additionally, the  
31 assumption of mean being equal to variance was violated as variance was greater than the mean  
32 indicating that the data was over dispersed. To account for this, the model was fitted using a quasi-  
33 Poisson method. Furthermore, a check for autocorrelation was conducted by considering the Variance

1 Inflation Factor (VIF) to check all predictor variables. No issues were identified as the estimates for all  
2 variables were below the threshold of 3 as discussed by Zuur et al., (2010).

3 More specifically, the Multilevel Poisson Regression model was constructed to explore the explanatory  
4 power of environmental and farm-level variables in variations of sheep lameness prevalence. The model  
5 was constructed using the statistical software R and specifically the package “lme4” which provides  
6 functions for fitting mixed models (Bates et al., 2015; R Core Team, 2018). The dependent variable of  
7 the model was offset by the natural logarithm of total sheep in the field that was imported to adjust for  
8 differences in population sizes (here flock size). In addition, the log link function was used and the  
9 model is in the following form:

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$$\text{Number of lame sheep in field }_{fkd} \sim a + \beta_{fkd}X_{fkd} + e_{fkd} + offset$$

11 Where  $\sim$  is the log link function,  $a$  is the intercept,  $\beta_{fkd}$  are the coefficients for a vector of  $X_{fkd}$   
12 explanatory variables which vary by the levels  $fkd$ ,  $e_{fkd}$  is the residual and offset is the natural  
13 logarithm of the number of sheep in each field. Three levels were applied in the current approach as  
14 random effects representing observations of farms ( $k = 1, \dots, K$ ) containing fields ( $f = 1, \dots, F_K$ ) for  
15 the different dates of recordings of lameness ( $d = 1, \dots, D_{FK}$ ).

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1 Table 1 List of variables considered in the analysis

Variable category	Variable	Levels / Range
Sheep	Sheep category	Ewe (1 year and older)
		Lamb and Ewe (younger than 1 year)
		Ram
Farm	Pasture Length	Short - up to 5cm
		Medium - 6 to 10 cm
		Long - longer than 10 cm
	Pasture Type	Forage crop
		New leys
		Permanent
	Flock size (class)	I - up to 50 sheep
		II - 51 to 100 sheep
		III - 101 to 200 sheep
		IV - 201 to 400 sheep
Environment	Seasonality	Autumn (Sep/Oct/Nov)
		Spring (Mar/Apr/May)
		Summer (Jun/Jul/Aug)
		Winter (Dec/Jan/Feb)
	Topography (Altitude)	0 to 250 m
	Chemical Soil condition - Se concentration	0.2 to 1.3 mg kg <sup>-1</sup>
	Bedrock type (Rock classification scheme)	Limestone with subordinate sandstone and argillaceous rocks (LSSA)
		Mudstone, siltstone and sandstone (MDSS)
		Mudstone, siltstone, limestone and sandstone (MSLS)
		Sandstone, limestone and argillaceous rocks (SLAR)
Climate	Precipitation - fourteen day rolling average	0 to 8.5 mm
	Minimum Temperature - three day rolling average	-5 to 16 °C
	Maximum Temperature - three day rolling average	3 to 26 °C

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### 1 3 Results

2 Results of the multivariable multilevel model are presented in Table 2. Results indicate that there were  
3 statistically significant associations between sheep lameness and the variables of sheep category, length  
4 of pasture, seasonality, Se concentration in soil, temperature, flock size and bedrock type. The median  
5 observation of the response variable in the model was 0.02 (or equivalently, 2% prevalence). More  
6 specifically, the latter indicates that of all the sheep in all the fields, 2% were lame. Detailed information  
7 on the results of the regression is presented in Table 2. In interpreting these results it is important to  
8 take into consideration that the estimates are related to the dependent variable by the log link and thus,  
9 it is essential to first exponentiate the coefficients. Following, in the cases that the exponentiated value  
10  $j_n$  for variable  $n$  is less than one then the effect of the independent variable is negative and for each  
11 extra unit of it, the dependent variable decreases by  $(1 - j_n) * 100$  percent. The same estimation was  
12 used for coefficients above one and taking the absolute value of this calculation gives the percentage of  
13 increase. Furthermore, as several explanatory variables were categorical, the modelling results were  
14 interpreted in relation to a reference group.

15 With regards to sheep category, for Lamb and Ewe (younger than 1 year), sheep lameness decreased  
16 by 25% (while keeping the rest of the predictors fixed) in reference to Ewe category. Furthermore,  
17 considering pasture length, findings suggest that for medium (6 – 10 cm) and short (up to 5 cm) length  
18 lameness decreased by 17% and 32% respectively in comparison to long (>10 cm) length. Regarding  
19 seasonality, results show that for Spring, Summer and Winter lameness is increased in comparison to  
20 Autumn by 44%, 64% and 100% respectively. Concerning Se soil concentration results show that for  
21 each extra unit of Se in the soil, lameness decreased by 84%. Concentration values of Se range from  
22 0.2 to 1.3 mg kg<sup>-1</sup> and thus, it may be more appropriate to suggest that for each extra 0.1 Se units in the  
23 soil (e.g. increasing from 0.2 to 0.3 mg kg<sup>-1</sup>), lameness decreased by 8.4%. With regards to flock size,  
24 results demonstrate that flock sizes of 51 to 100, 101 to 200 and 201 to 400 sheep have decreased levels  
25 of lameness in comparison to the reference category (flock size of up to 50 sheep) by 25%, 36% and  
26 51% respectively.

27 The three day rolling average of maximum temperature has a positive relationship with lameness levels.  
28 Specifically, it is estimated that for one extra degree there is an increase of 3% of lameness. The  
29 fourteen-day rolling average of rainfall had no statistically significant effect on lameness. The  
30 maximum and minimum temperature averages were highly correlated and thus only the rolling average  
31 of maximum temperature was left in the final model to test whether warmer conditions have an effect.

32 A statistically significant negative association was identified between bedrock type and lameness.  
33 Specifically, the class MSLS showed decreased lameness levels by a factor of 80% when compared to



1 the reference category of Sandstone. The variable of altitude was initially considered in the modelling  
 2 however, no statistical association was observed between them and lameness levels.  
 3 Finally, concerning the goodness of fit of the model, the results indicate that the proportion of variance  
 4 explained by the fixed factors (marginal) is 0.31 while proportion of variance explained by both the  
 5 fixed and random factors (conditional) is 0.94.

6 Table 2 Multilevel Poisson Regression model results

	N Farms	N obs	Estimate	Std. Error	z value	Exponentiated estimate	Effect	Pr(> z )	
(Intercept)	18		-1.130	0.919	-1.229	0.323		0.219	
Sheep Category Ewe (>=1 year)	18	291	Ref.						
Sheep Category Lamb and Ewe younger than 1 year	16	96	-0.290	0.090	-3.214	0.748	-25%	0.001	**
Sheep Category Ram	5	13	-0.177	0.280	-0.631	0.838		0.528	
Pasture Length >10 cm (long)	11	145	Ref.						
Pasture Length 6-10 cm (medium)	16	192	-0.183	0.089	-2.054	0.832	-17%	0.040	*
Pasture Length up to 5 cm (short)	15	182	-0.384	0.114	-3.375	0.681	-32%	<0.001	***
Pasture Type Forage Crop	2	9	Ref.						
Pasture Type New Leys	6	27	0.487	0.471	1.035	1.628		0.301	
Pasture Type Permanent	18	487	0.388	0.451	0.860	1.474		0.390	
Season Autumn	10	177	Ref.						
Season Spring	11	115	0.366	0.138	2.658	1.442	44%	0.008	**
Season Summer	9	100	0.494	0.119	4.135	1.639	64%	<0.001	***
Season Winter	12	127	0.694	0.164	4.232	2.002	100%	<0.001	***
Selenium concentration	18	521	-1.850	0.735	-2.516	0.157	-84%	0.012	*
Precipitation (2 week mean)	18	521	-0.034	0.024	-1.452	0.966		0.147	
Maximum temperature (3 day mean)	18	521	0.029	0.013	2.244	1.029	3%	0.025	*
Flock size up to 50	15	171	Ref.						
Flock size 51 to 100	17	173	-0.292	0.122	-2.392	0.746	-25%	0.017	*
Flock size 101 to 200	14	102	-0.443	0.149	-2.971	0.642	-36%	0.003	**
Flock size 201 to 400	7	73	-0.713	0.173	-4.114	0.490	-51%	<0.001	***
Bedrock type SLAR	2	10	Ref.						
Bedrock type LSSA	2	31	-0.880	0.992	-0.887	0.415		0.375	
Bedrock type MDSS	2	81	-0.873	0.624	-1.399	0.418		0.162	
Bedrock type MSLS	13	397	-1.623	0.816	-1.990	0.197	-80%	0.047	*
R squared: 0.31 (marginal), 0.94 (conditional)									
Significance codes: '***' 0.001 '**' 0.01 '*' 0.05									
Random effects:									
Groups	Name	Variance	Std.Dev.						
Farm	(Intercept)	0.4923	0.7016						
Field	(Intercept)	0.4426	0.6653						
Date	(Intercept)	0.3802	0.6166						

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### 2 3.1.1 Model diagnostics

3 The regression analysis was followed by assumption testing to examine whether assumptions were met  
4 in the modelling. Furthermore, it was assumed that the variance of the residuals is equal among the  
5 various model levels which was satisfied in the model. Finally, the distribution of the residuals was  
6 examined for the assumption of normality and a normal probability QQ plot and a histogram of the  
7 model residuals suggested that the assumption was met.

## 8 4 Discussion

9 In the current study a range of potential environmental risk factors for sheep lameness were examined.  
10 This is the first study using observational data of sheep lameness from multiple farms in the UK that  
11 quantifies and signifies the relationship between lameness and Se concentration in the soil. Findings of  
12 the current study suggest that Se has a negative relationship with lameness prevalence. This finding is  
13 in line with relevant experimental research showing that Se as a micronutrient has a positive effect in  
14 restoring sheep immune functions negatively affected by footrot (one of most common cause of  
15 lameness) (Hall et al., 2013; Hujeriletu et al., 2013). Furthermore, (Hall et al., 2009) a positive effect  
16 is to be identified for sheep exposed to increased-Se forage, that have been infected by footrot.

17 This is also the first attempt to examine whether soil type has an impact on sheep lameness prevalence.  
18 According to the findings of this approach, the bedrock class MSLS (Mudstone, siltstone, limestone and  
19 sandstone) is associated with a decrease in lameness levels when compared to soil type SD (Sandstone).  
20 In a report by Agriculture and Horticulture Development Board, UK (2016) it is recommended that  
21 application of limestone can work as a prevention method for sheep lameness. However, interestingly  
22 the quantified findings and recommendations remain fairly limited on the importance of this parameter.  
23 Through a different context, Muzafar et al. (2016) indicate that clay soils (here classes containing  
24 mudstone) provide fairer conditions for the survival of *D. nodosus* in comparison to sand soils (here  
25 sandstone). However, the rock classification scheme provides a range of classes where each represents  
26 a composition of various materials in the soil (Table 1). Specifically, the classes MSLS and SD have  
27 some common chemical features (Sandstone and Limestone) while they differ in the rest of their  
28 materials (Mudstone, Argillaceous rocks and Conglomerate) (Table 1). Thus, further investigation is  
29 needed to enable the identification of the particular materials in the soil that have an association with  
30 sheep lameness.

31 Additionally, in our approach a positive relationship was identified between pasture length and  
32 lameness levels (pasture length higher than 10 cm had increased lameness in comparison to up to 5cm  
33 and 5 to 10 cm). The latter is in line with Angell et al. (2018) who suggest that sheep are more likely to

1 have footrot when grazing longer swards. Pastures of longer length tend to retain higher levels of  
2 moisture when compared to shorter length pastures. As a result, sheep standing in longer pastures have  
3 their feet exposed in more humid conditions in comparison to those standing on shorter length pastures  
4 and thus, providing adequate conditions for the development of lameness.

5 Towards this direction, findings of the current approach demonstrate that the flock size has a negative  
6 association with sheep lameness prevalence. Specifically, it is demonstrated that larger flocks relate to  
7 lower levels of lameness. This finding is in accordance to Winter et al., (2015) who also report that  
8 increased flock size is linked to decreased lameness. On the contrary, similar approaches find either a  
9 positive relationship (Angell et al., 2018) or no association (Kaler and Green, 2008b) between flock  
10 size and lameness. According to Dickins et al. (2016), lower lameness percentages for larger flock sizes  
11 as a finding may indicate a density dependency in the systems under consideration. A potential  
12 interpretation is that larger flocks may be managed with more effective and commercially oriented  
13 production strategies (Gelasakis et al., 2013) where biosecurity measures may be more precise and  
14 standardised (Dickins et al. 2016).

15 A positive association was observed between maximum temperature (of the last 3 days prior to  
16 recording) and sheep lameness. Our findings are in line with Green and George (2008) and Smith et al.  
17 (2014) who have suggested the impacts of climate on the transmission of footrot. Furthermore, Raadsma  
18 and Egerton (2013) argue that footrot transmission within infected sheep depends on presence of  
19 adequate temperatures and previous hydration of the interdigital skin. With regards to moisture, studies  
20 have identified precipitation as a significant parameter triggering increased levels of prevalence (Abbott  
21 and Lewis, 2005). However, in the current study there was no such association identified. In a relevant  
22 context, Wassink et al. (2003) have also reported a lower prevalence of lameness in areas with higher  
23 levels of rainfall indicating that variations in prevalence are perhaps driven more drastically by  
24 management factors at the farm level.

25 Results of the current study point out a relationship between seasons and lameness. Particularly, in the  
26 current study all seasons had increased lameness when compared to the reference group which was  
27 Autumn. Among the seasons, Spring estimated the smallest increase in comparison to Autumn. This  
28 finding is in contrary to Angell et al., (2015b) who reported increased prevalence during Autumn and  
29 Summer. To an extent, seasonal increase in lameness imply effects of fair climatic conditions for the  
30 survival of the bacterium (warm and wet conditions). However, Wassink et al., (2003) reported that  
31 within the UK, even areas with such conditions (Southwest England and South Wales) had lower levels  
32 of FR in comparison to areas with lower average temperatures. In that sense, it seems likely that  
33 lameness may be affected by a combination of parameters in addition to weather conditions, such as  
34 management practices at the farm level.

1 Furthermore, in this study, differences were identified among the different sheep categories (here  
2 relating to age groups). The categories of Lamb and Ewe (younger than one year) had lower lameness  
3 levels when compared to the Ewe category. This result is in line with Angell et al. (2018) who showed  
4 decreased lameness levels for yearling sheep in comparison to lambs and adult ewes. However, there is  
5 possibility that older ewes are more likely to have poor foot confirmation and poor conformation  
6 increases susceptibility to footrot due to reduction in resistance with increasing age (Kaler et al., 2010).

7 Although this study gave us insights into various environmental factors associated with lameness, it  
8 does not imply causality. The impact of variables identified in the current study need to be studied  
9 further and validated on farms with wider geographical area. No animal management factors were  
10 included in the current study as all farmers followed best practice to manage lameness, however, other  
11 variables such as the length of the stay in a particular field and whether new sheep were introduced  
12 could impact on infection pressure. Unfortunately, this was unknown in the current study. This study  
13 used farmers reporting to identify lameness; despite this lameness levels are unlikely to be biased given  
14 farmers were trained and independently validated before the start of the study, In addition, previous  
15 studies suggest farmers can identify lameness accurately (Kaler and Green, 2008a; King and Green,  
16 2011).

17 In this study a smartphone app was used to record lameness levels, the study demonstrates its usefulness  
18 in line with other precision livestock technologies to be able to generate long term disease data on farm,  
19 something vastly lacking on sheep farms (Kaler and Green, 2013; Kaler and Ruston, 2019). In addition,  
20 use of the app improved data recording on farm (results not shown).

21 In conclusion, the findings of the present analysis enable a novel understanding of the ways and the  
22 extent that environmental and farm-level management characteristics associate with sheep lameness  
23 prevalence. Using technology such as the one used in this study for data gathering alongside data on  
24 environmental and field variables could help us build predictive models for lameness as well as a tool  
25 for managing lameness. This could be of essential value for policy makers as well as farmers aiming to  
26 tackle and reduce lameness prevalence levels.

27

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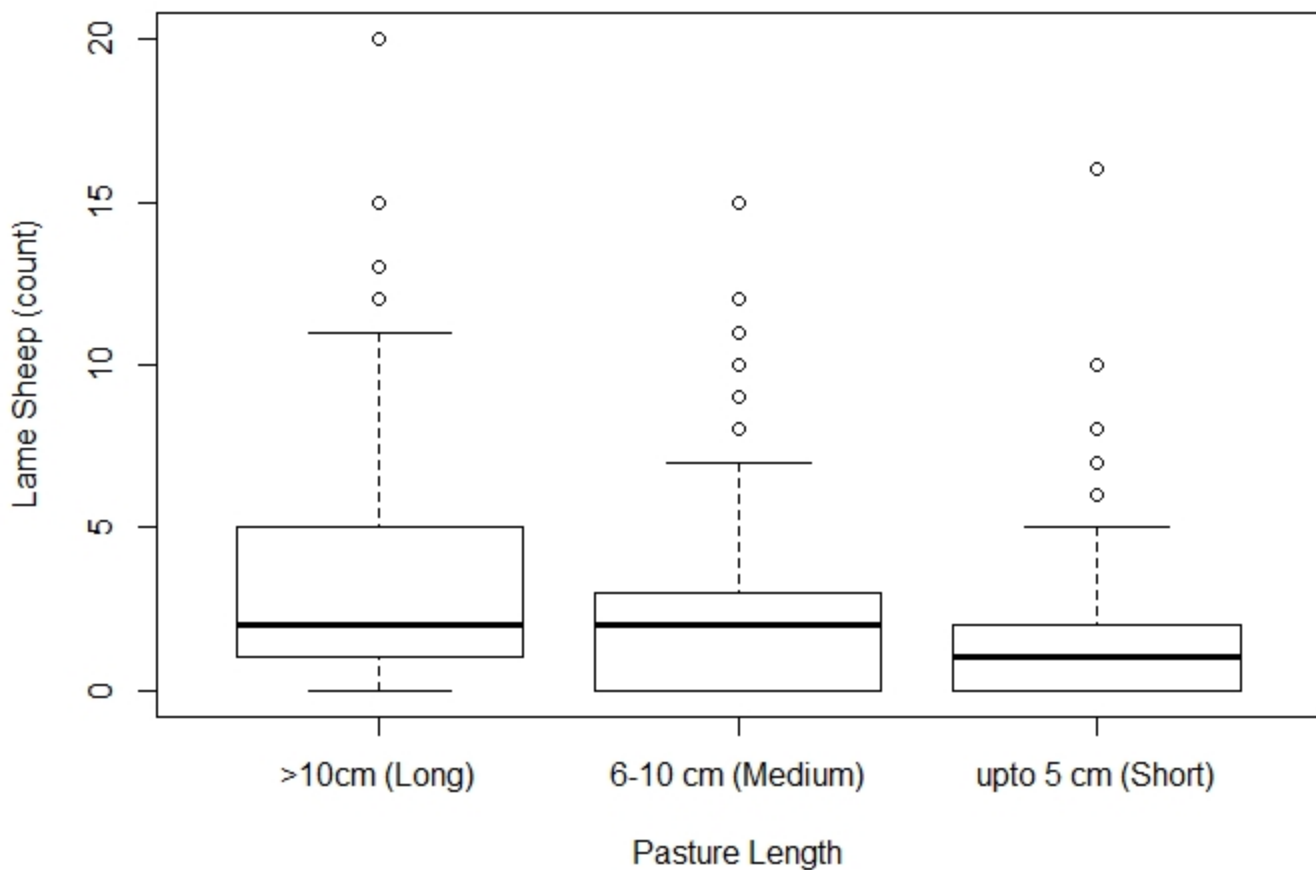
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## Pasture Length



N



# Legend

▲ Farm businesses

■ Wales

■ England

