

Article

# *Mycobacterium bovis* population structure in cattle and local badgers: co-localisation and variation by farm type

Georgina Milne <sup>1\*</sup>, Adrian Allen <sup>1</sup>, Jordon Graham <sup>1</sup>, Raymond Kirke <sup>2</sup>, Carl McCormick <sup>3</sup>, Eleanor Presho <sup>1</sup>, Robin Skuce <sup>1,4 †</sup> and Andrew Byrne <sup>4,5 †</sup>

<sup>1</sup> Veterinary Sciences Division, Agri-food and Biosciences Institute (AFBI), 12 Stoney Road, Stormont, Belfast BT4 3SD, UK; [Georgina.milne@afbini.gov.uk](mailto:Georgina.milne@afbini.gov.uk) (GM); [Adrian.Allen@afbini.gov.uk](mailto:Adrian.Allen@afbini.gov.uk) (AA);

[Jordon.Graham@afbini.gov.uk](mailto:Jordon.Graham@afbini.gov.uk) (JG); [Eleanor.Breadon@afbini.gov.uk](mailto:Eleanor.Breadon@afbini.gov.uk) (EP); [Robin.Skuce@afbini.gov.uk](mailto:Robin.Skuce@afbini.gov.uk) (RS)

<sup>2</sup> Department of Agriculture, Environment and Rural Affairs, Veterinary Epidemiology Unit, Dundonald House, Belfast, BT4 3SB, UK; [Raymond.Kirke@daera-ni.gov.uk](mailto:Raymond.Kirke@daera-ni.gov.uk) (RK)

<sup>3</sup> Department of Agriculture, Environment and Rural Affairs, Veterinary Service Animal Health, Crown Buildings, Artillery Road, Coleraine BT52 2AJ, UK; [Carl.McCormick@daera-ni.gov.uk](mailto:Carl.McCormick@daera-ni.gov.uk) (CMcC);

<sup>4</sup> School of Biological Sciences, Queen's University Belfast, 19 Chlorine Gardens, Belfast, BT9 5DL, UK

<sup>5</sup> One-Health Unit, Department of Agriculture, Food and the Marine, Agriculture House, Dublin, Ireland; [AndrewW.Byrne@agriculture.gov.ie](mailto:AndrewW.Byrne@agriculture.gov.ie) (AB)

\* Correspondence: [georgina.milne@afbini.gov.uk](mailto:georgina.milne@afbini.gov.uk)

† Joint senior authorship

**Abstract:** Bovine tuberculosis surveillance in Northern Ireland includes Multiple-Locus Variable number tandem repeat Analysis (MLVA) to determine the *Mycobacterium bovis* genetic type present in both cattle, and the predominant wildlife host, the European badger (*Meles meles*). These data are informative for investigating clusters of infection and understanding the scale at which interspecific transmission may occur. We utilised a comprehensive dataset of routinely sampled isolates from infected cattle and from badgers killed in road-traffic accidents to investigate the spatial co-location of MLVA types in, and between, the badger and cattle populations. Furthermore, we investigate the hypothesis that the farming enterprise type might explain some variation in this relationship. MLVA types were spatially co-localised in cattle and RTA badger hosts, indicative of a shared epidemic. Dairy herds were more likely to have at least one MLVA type in common with nearby RTA badgers, compared to non-dairy herd types. Marginally more MLVA spatial clustering was observed in non-dairy herds, which may be a consequence relatively more between-herd movements. For the cattle population, local transmission mechanisms such as infection from contiguous herds, infectious wildlife and short-range between-herd cattle movements appear primarily to drive the epidemic: there appears to be a more limited role for long-range movements. Animal management practices are likely the driving force behind this observation, as beef rearing is associated with elevated numbers of animal movements compared to dairy herds.

**Keywords:** Bovine tuberculosis; molecular epidemiology; spatial; badgers; MLVA; Northern Ireland

## 1. Introduction

The wildlife-livestock interface presents a conduit through which pathogens can be exchanged [1]. In the UK and Ireland, the presence of wildlife reservoirs is implicated in the persistence of *Mycobacterium bovis*, the principal causative agent of bovine tuberculosis (bTB) in cattle [2, 3]. *M. bovis* can infect a wide range of hosts, both wild and domesticated [4]. In the United Kingdom (UK) and the Republic of Ireland (ROI), the most important wildlife maintenance host is the European badger, *Meles meles* [5-8]. Infection transmission may be as a result of direct contact between species [9], or potentially through contaminated shared environments and fomites [10, 11].

In Northern Ireland (NI), the costs of bTB control in cattle have exceeded £365 million over a recent twelve year period [12], and despite an intensive and costly state-led programme focusing on the cattle population, eradication has not yet been achieved [13]. NI is a relatively small area (approx. 13,500km<sup>2</sup>), yet sustains a badger population of approximately 33,500 individuals (95%CI 26,000-41,200); [14]. A passive road-traffic accident (RTA) surveillance programme for *M. bovis* in badgers has been ongoing since 1998. This survey estimated *M. bovis* prevalence in sampled badgers to be 15.3% (95%CI 13.10%-17.5%) [15, 16], and revealed elevated bTB risk in cattle herds in close proximity to infected RTA badgers, compared to herds proximal to uninfected badgers [15]. Routine surveillance efforts for *M. bovis* in cattle and badger hosts also include spoligotyping and Multiple-Locus Variable number tandem repeat Analysis (MLVA) typing [17, 18]. These data revealed clear spatial structuring (i.e. clustering) of *M. bovis* genetic types in cattle herds [18, 19], and have shown that both cattle and badger hosts with the same MLVA type tend to be closer together than hosts with a different MLVA type [20, 21]. This observed structure in the *M. bovis* population in NI indicates that the bTB epidemic is co-localised between both wild and domestic hosts, consistent with some degree of transmission between wild and domestic species [7]. However, cattle sources alone also contribute to maintenance of infection at both the local and national scales; e.g. via within-herd amplification, (regardless of source), contact with nearby infected herds, or between-herd cattle purchases involving infected animals [7, 22-24].

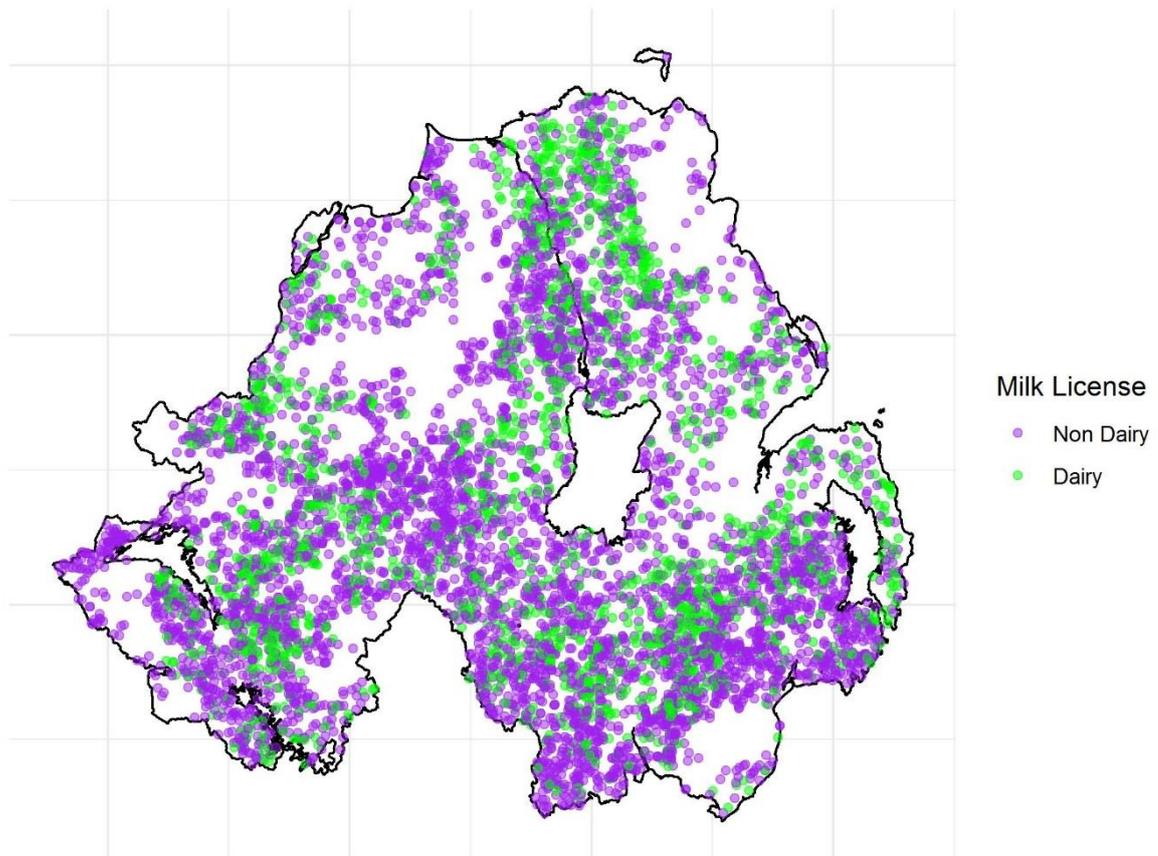
However, the influence of cattle management and trade, as a risk factor, on the *M. bovis* population structure in both cattle and badgers has been given little attention to date. Beef and dairy production systems differ across a number of factors which may influence transmission. For example, we showed recently that beef herds in NI were generally associated with more between-herd cattle movements than dairy herds, and also experience elevated MLVA richness at the herd-level [25]. Different herd types have specific between-herd contact patterns potentially linked to different infection pathways [22], with beef fattening herds appearing more susceptible to infection introduced by bought-in cattle than dairy herds, and indeed, in NI, there is elevated risk of bTB infection associated with the purchase of beef animals [26]. These differences in herd management may subsequently manifest in different spatial relationships in the clustering of *M. bovis* in, and between, infected cattle and badger populations. We therefore aimed to analyse patterns in the spatial relationships in the *M. bovis* MLVA types found both in RTA badgers and cattle herds in NI to ultimately gain insight into the spread of the epidemic, both within and between hosts.

## 2. Results

### 2.1. Summary Statistics

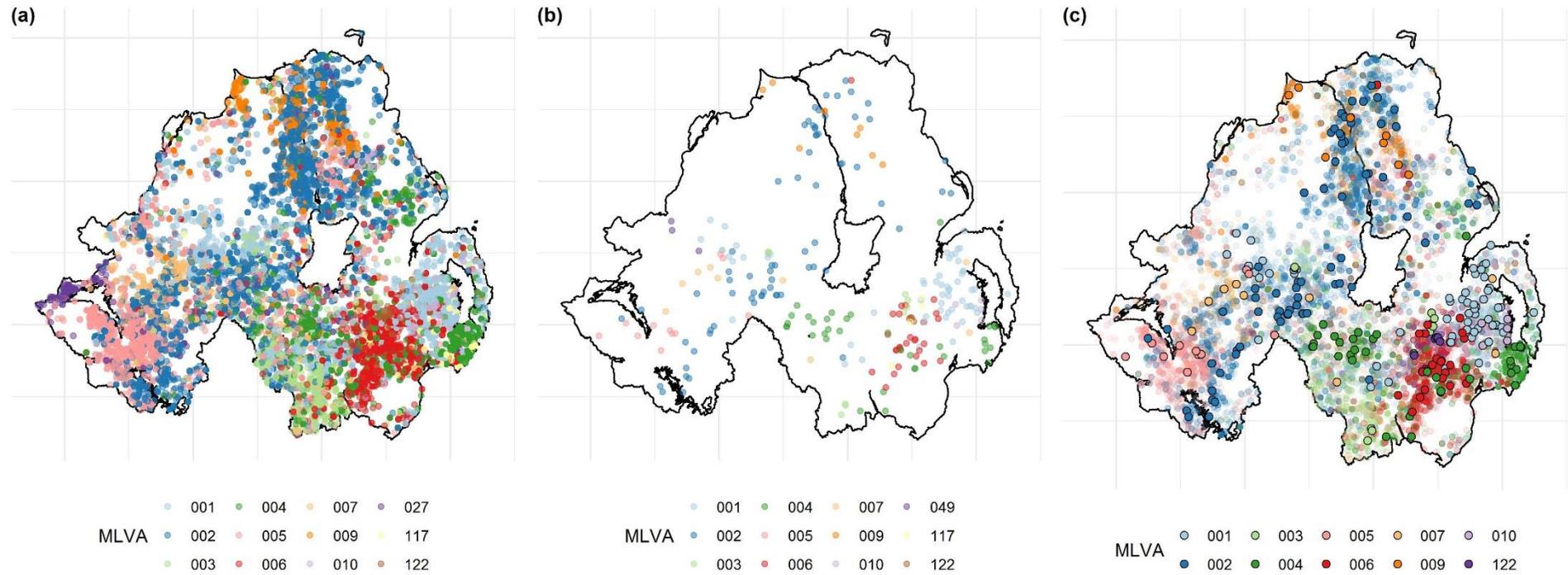
The final cattle dataset contained information on 9,208 bTB breakdowns occurring between 2008 and 2016, in 6,954 herds (herds with a milk license = 1,822, 27.6%; without a milk license = 4,772, 72.4%; see Figure 1). In total, 364 *M. bovis* MLVA types were isolated from the cattle population, 135 were found in both herds with and without milk licenses, 62 were found only in herds with milk licenses, and 167 were isolated only from herds without milk licenses. At herd level (i.e. the yearly herd-level incidence per MLVA type), the 12 most common MLVA types represented 77.8% of the total cattle isolates (Figure 2A; 001, 002, 003, 004, 005, 006, 007, 009, 010, 027, 117 and 122). The final

RTA badger dataset contained data on 271 RTA badgers collected between 2008 and 2016, inclusive. 30 different MLVA types were identified in this population, with the 12 most common MLVA types representing 90% of the total (Figure 2B). Visual inspection of the distribution of the most common herd-level *M. bovis* MLVA types present in both cattle and RTA badgers revealed spatial co-localisation in both hosts (Figure 2C). In total, 26 (83.3%) of the 30 MLVA isolates found in badgers were also found in cattle. Herds with and without milk licenses differed across a number of epidemiologically relevant criteria, including herd size (herd size of herds with a milk license, median = 234; IQR: 140-365; without a milk license = 96; IQR: 47-178; wilcoxon signed rank test  $p < 0.001$ ), outwards movements (herds with a milk license = 44; IQR: 25-73; without a milk license = 32; IQR: 12-89;  $p < 0.001$ ), and inwards movements (herds with a milk license = 1; IQR: 0-6; n without a milk license = 15; IQR: 1-79;  $p < 0.001$ ).



**Figure 1.** The spatial distribution of study herds with (green), and without (purple) a milk license in NI.

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**Figure 2.** The spatial distribution of the twelve most common MLVA types in (a) cattle, (b) RTA badgers, and (c) MLVA types common to both cattle and RTA badgers.

## 2.2. Assessment of MLVA clustering within the cattle population

Figure 3 illustrates the distribution of the most common MLVA type (002) in cattle in (i) all cattle herds; Figure 3A, (ii) herds with milk licenses; Figure 3B, and (iii) herds without milk licenses ; Figure 3C (See Supplementary Material 1, Fig 1-11 for the remaining plots). The permutation analysis revealed that herds from which the same *M. bovis* MLVA type was isolated were significantly closer compared to distances obtained from the sampling distribution (Figure 4A). Whilst similar results were obtained when considering only herds without milk licenses (Figure 4B), for some MLVA types (001, 002, 010 and 027), the actual median distance between infected herds did not appear to differ significantly from the median distances obtained in the sampling distribution (Figure 4C).

The between-herd pairwise distances for both all study herds and herds without a milk license show that the 50% of herds lay within 62km each other, where 62km represents 35.5% of the maximal pairwise distance between any two herds (174.8km). For only those herds with a milk license, 50% of herd locations were within 63.3km of each other, i.e. 41.3% of the maximal pairwise distance between any two herds with a milk license (152.3km). The frequency distributions of MLVA-specific pairwise distances (See Supplementary Material 2, Fig 1-12) show that for all MLVA types in all herds, 50% of infected herds with a given MLVA type were within 35.5% of the maximal pairwise distance between the two most distal herds, and the largest 75% of pairwise distances were within 55% of the maximal, (Table1). However, the maximal distance between any two herds sharing the same MLVA type could be considerable, with observed distances of 162km between infected herds (e.g. type 009). This reflects a general trend of localised spatial clustering, with the majority of infected herds within relative close proximity to each other, and a smaller number of infected herds disproportionately widely distributed. These patterns were exemplified by type 122, wherein 50% of herds were within 16.3km of each other, which represented only 11.1% of the recorded maximal distance between any two herds from which it was isolated (147.5km). Some 75% of herds infected with type 122 were within 32.1km of each other, which represented only 21.7% of the maximal extent. The remaining 25% of values involved between-herd distances ranging from 32.2km to 147.5km, and contributed towards the remaining 78.3% of the distribution. In all but one MLVA type (type 006), the overall maximal extent was smaller in herds with milk licenses than in herds without, by between 8km (type 010) and 69km (type 122). Despite this, there was marginally less localised clustering in dairy herds compared to non-dairy herds for eight MLVA types (001, 002, 003, 004, 122, 010, 027 and 117). Here, the values associated with the 25th and 50th percentiles were proportionally larger compared to non-dairy herds, ranging in a difference of 0.3% at the 50th percentile (type 117) to 18.9% (type 027).

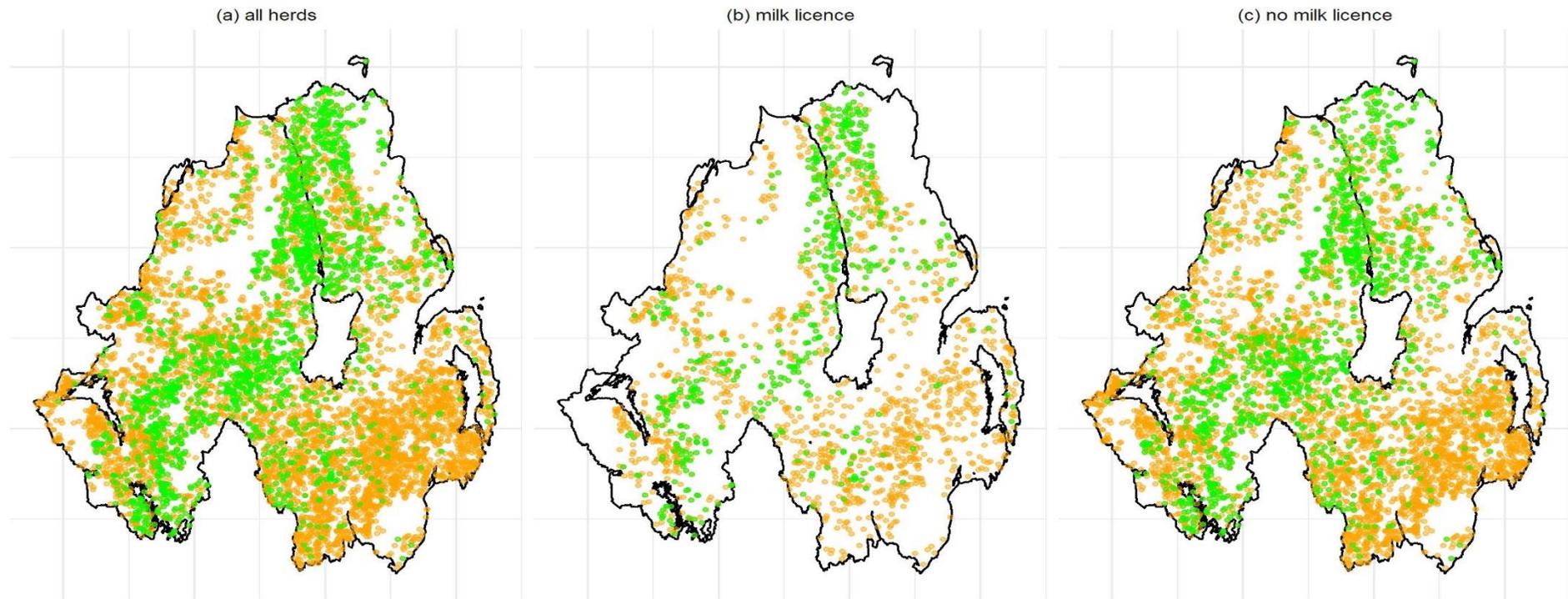
**Table 1:** The distances associated with the 25th, 50th, 75th and 100th percentiles from a frequency distribution of pairwise distances between herds infected with a given MLVA type. The values of the 25th, 50th and 75th percentiles are shown as a percentage of the maximum extent (100th percentile).

| MLVA | Dataset   | Total infected herds | 25th Percentile (km) | 50th Percentile (km) | 75th Percentile (km) | 100th percentile (km) |
|------|-----------|----------------------|----------------------|----------------------|----------------------|-----------------------|
| 001  | All       | 1141                 | 28.7 km (19.3%)      | 52.7 km (35.5%)      | 77.6 km (52.2%)      | 148.5 km              |
| 001  | Dairy     | 347                  | 29.2 km (19.2%)      | 52.8 km (34.7%)      | 76.4 km (50.1%)      | 152.4 km              |
| 001  | Non Dairy | 794                  | 27.8 km (16.9%)      | 52.5 km (31.9%)      | 78.2 km (47.5%)      | 164.5 km              |
| 002  | All       | 1885                 | 30.9 km (18.5%)      | 52 km (31.3%)        | 76.5 km (46%)        | 166.4 km              |
| 002  | Dairy     | 545                  | 28.3 km (18.9%)      | 54.1 km (36.1%)      | 80.9 km (54%)        | 149.9 km              |

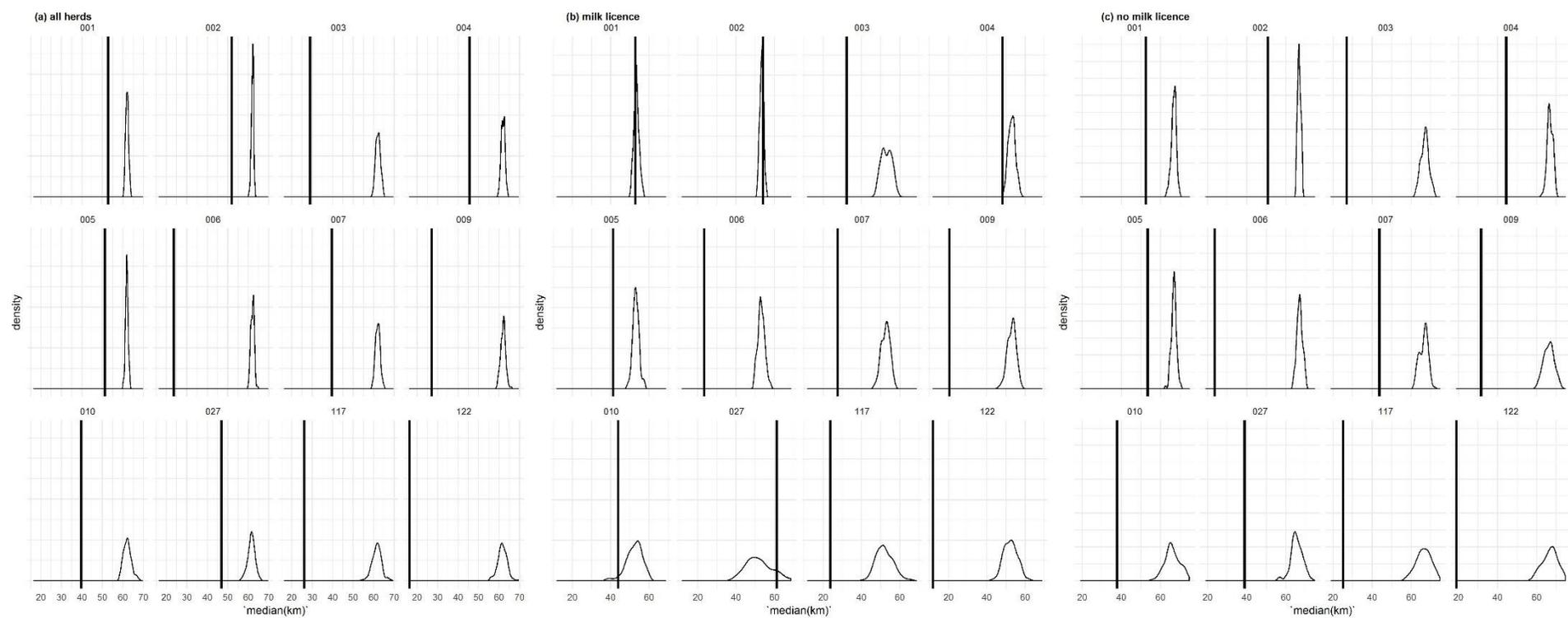
|     |           |      |                    |                    |                    |          |
|-----|-----------|------|--------------------|--------------------|--------------------|----------|
| 002 | Non Dairy | 1340 | 31.1 km<br>(18.7%) | 50.8 km<br>(30.6%) | 74.5 km<br>(44.8%) | 166.4 km |
| 003 | All       | 460  | 14.9 km<br>(10%)   | 29 km<br>(19.4%)   | 51.2 km<br>(34.2%) | 149.6 km |
| 003 | Dairy     | 118  | 17.4 km<br>(13.6%) | 32.2 km<br>(25.1%) | 51.5 km<br>(40.1%) | 128.4 km |
| 003 | Non Dairy | 342  | 13.5 km<br>(9.1%)  | 27 km<br>(18.1%)   | 50.7 km<br>(33.9%) | 149.6 km |
| 004 | All       | 736  | 26.1 km<br>(16.7%) | 45.6 km<br>(29.1%) | 64.2 km<br>(41%)   | 156.5 km |
| 004 | Dairy     | 204  | 27.6 km<br>(19.7%) | 48.1 km<br>(34.3%) | 68.3 km<br>(48.7%) | 140.5 km |
| 004 | Non Dairy | 532  | 25.4 km<br>(16.3%) | 44.4 km<br>(28.6%) | 62.7 km<br>(40.4%) | 155.3 km |
| 005 | All       | 1001 | 23.8 km<br>(14.3%) | 51.3 km<br>(30.7%) | 79.3 km<br>(47.4%) | 167.3 km |
| 005 | Dairy     | 266  | 17.9 km<br>(12%)   | 41.2 km<br>(27.4%) | 77.6 km<br>(51.7%) | 150.1 km |
| 005 | Non Dairy | 735  | 26.1 km<br>(15.6%) | 53.4 km<br>(31.9%) | 79.8 km<br>(47.7%) | 167.3 km |
| 006 | All       | 810  | 13.9 km<br>(9.5%)  | 23.6 km<br>(16.1%) | 40.9 km<br>(27.9%) | 146.4 km |
| 006 | Dairy     | 241  | 12.9 km<br>(8.8%)  | 23.2 km<br>(15.9%) | 44.7 km<br>(30.5%) | 146.4 km |
| 006 | Non Dairy | 569  | 14 km<br>(10.3%)   | 23.5 km<br>(17.3%) | 39.6 km<br>(29.2%) | 135.8 km |
| 007 | All       | 388  | 19.8 km<br>(13.8%) | 39.6 km<br>(27.6%) | 65.6 km<br>(45.6%) | 143.8 km |
| 007 | Dairy     | 120  | 14.3 km<br>(11%)   | 27.6 km<br>(21.3%) | 57 km<br>(44.1%)   | 129.4 km |
| 007 | Non Dairy | 268  | 22.7 km<br>(15.8%) | 43.7 km<br>(30.4%) | 67.9 km<br>(47.2%) | 143.8 km |
| 009 | All       | 355  | 17.1 km<br>(10.5%) | 27.1 km<br>(16.7%) | 44.9 km<br>(27.7%) | 162.0 km |
| 009 | Dairy     | 140  | 12 km (9.4%)       | 20.4 km<br>(15.9%) | 30.6 km<br>(23.9%) | 128.1 km |
| 009 | Non Dairy | 215  | 19.3 km<br>(11.9%) | 31.7 km<br>(19.5%) | 52.5 km<br>(32.4%) | 162.0 km |
| 122 | All       | 157  | 8.5 km<br>(5.8%)   | 16.3 km<br>(11.1%) | 32.1 km<br>(21.7%) | 147.5 km |
| 122 | Dairy     | 60   | 6.6 km<br>(8.5%)   | 11.9 km<br>(15.2%) | 22.7 km<br>(29.1%) | 77.9 km  |
| 122 | Non Dairy | 97   | 10.4 km<br>(7.1%)  | 19 km<br>(12.9%)   | 36.7 km<br>(24.9%) | 147.5 km |
| 010 | All       | 148  | 14.2 km<br>(10.2%) | 39.7 km<br>(28.6%) | 59.6 km<br>(42.9%) | 138.9 km |
| 010 | Dairy     | 46   | 15 km (12%)        | 43.7 km<br>(35%)   | 59.9 km<br>(48.1%) | 124.7 km |
| 010 | Non Dairy | 102  | 12.9 km<br>(9.7%)  | 37.7 km<br>(28.3%) | 59.3 km<br>(44.4%) | 133.5 km |
| 027 | All       | 157  | 18.7 km<br>(11.4%) | 46.9 km<br>(28.6%) | 88.5 km<br>(54.1%) | 163.8 km |

|     |           |     |                    |                    |                    |          |
|-----|-----------|-----|--------------------|--------------------|--------------------|----------|
| 027 | Dairy     | 26  | 27.8 km<br>(20.3%) | 61.2 km<br>(44.8%) | 84.1 km<br>(61.6%) | 136.4 km |
| 027 | Non Dairy | 131 | 13.8 km<br>(9.2%)  | 38.9 km<br>(25.9%) | 84.1 km<br>(56.1%) | 150.0 km |
| 117 | All       | 116 | 15 km (10%)        | 26.2 km<br>(17.5%) | 38.8 km<br>(25.9%) | 149.7 km |
| 117 | Dairy     | 37  | 12.3 km (9%)       | 23.7 km<br>(17.3%) | 37.4 km<br>(27.3%) | 136.8 km |
| 117 | Non Dairy | 79  | 14.6 km<br>(9.8%)  | 25.2 km<br>(17%)   | 39.5 km<br>(26.6%) | 148.5 km |

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**Figure 3.** The spatial distribution of the most common MLVA type (002), shown in (a) all herds, (b) herds with a milk license, and (c) herds without a milk license. Green dots represent herds from which the MLVA type was isolated at least once, orange dots represent herds from which the MLVA type was never isolated



**Figure 4.** The median distance between herds infected with the same MLVA type (thick black lines) compared to the median distances derived from sample distribution, following 999 random permutations. Results are shown for a) all herds, (b) herds with a milk license, and (c) herds without a milk license.

### 2.3. Intra and interspecific nearest neighbour (NN) analysis

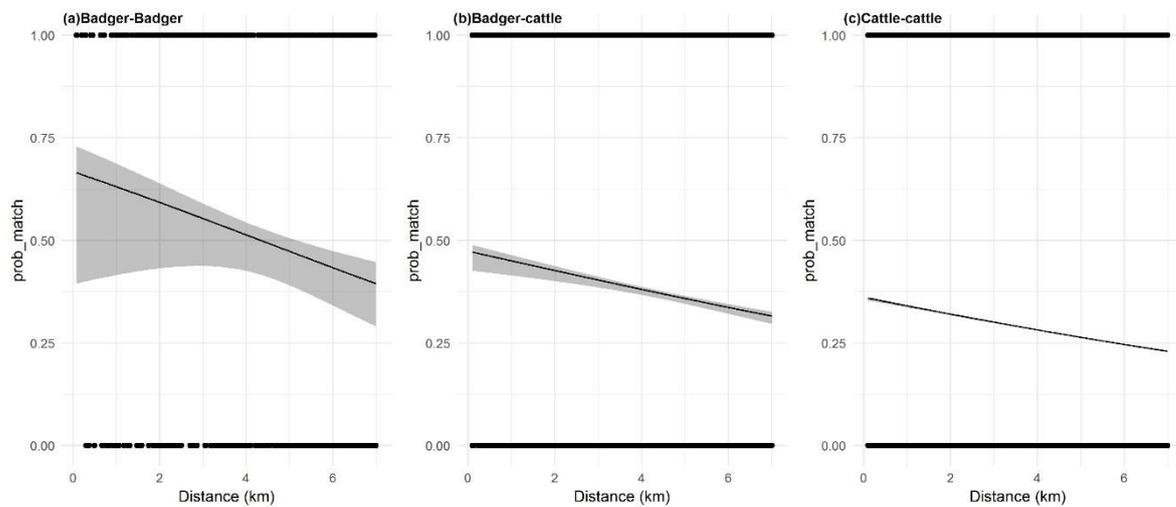
The median distance between NN RTA badgers which shared an MLVA genotype was 2.4km (Inter-Quartile Range; IQR: 1.2km-4.0km), whilst the median distance between nearest neighbour RTA badgers which did not share an MLVA genotype was 3.3km (IQR: 1.9km-5.2km;  $V = 2853$ ,  $p = 0.011$ ). RTA badgers were recorded a median distance of 0.61km (0.37km-1.1km) from cattle homesteads, and were generally found in closer proximity to homesteads associated with dairy herds (0.8km; IQR: 0.46km-1.4km) than non-dairy herds (1.4km; IQR: 0.78km-2.1km); for 182 RTA badgers (67.2%), the closest homestead was associated with a dairy herd. The distance between an RTA badger and the NN herd in which the MLVA type was not isolated (1.5km; IQR: 0.83km-2.2km) was almost 45% greater than the distance to a NN herd with the same MLVA type (0.84km; IQR: 0.50km-1.6km;  $V = 9469$ ,  $p < 0.001$ ). Similar findings were obtained for RTA badgers in proximity to both beef and dairy herds, see Table 2. The NN distance for cattle herds sharing an MLVA type was over 25% greater (1.1km; IQR: 0.61km-2.3km) than the NN distance between herds which did not share the same MLVA type (0.83km; IQR: 0.51km-1.3km). This was also observed in non-dairy herds and to a lesser extent, dairy herds; see Table 2. All findings were also replicated in the sensitivity analysis (all  $p < 0.05$ ), indicating that the spatial resolution of the data has no substantive impact on the interpretation of the results.

**Table 2:** Nearest-neighbour (NN) distances for pairs of RTA badgers and cattle herds which share MLVA types, and which do not share MLVA types

|                           | NN distance for hosts which share an MLVA type |         |         | NN distance for hosts which do not share an MLVA type |         |         | Difference in medians (km) |
|---------------------------|--|---------|---------|---|---------|---------|----------------------------|
|                           | Median (km)                                    | Q1 (km) | Q4 (km) | Median (km)   | Q1 (km) | Q4 (km) |                            |
| Badger-Badger             | 2.44   | 1.22    | 4.04    | 3.33  | 1.94    | 5.22    | 0.89                       |
| Badger-Cattle (all herds) | 0.82   | 0.50    | 1.64    | 1.49  | 0.83    | 2.25    | 0.67                       |
| Badger-Cattle (non-dairy) | 1.14   | 0.61    | 1.94    | 2.55  | 1.62    | 3.74    | 1.41                       |
| Badger-Cattle (dairy)     | 1.70   | 0.85    | 2.73    | 2.88  | 1.85    | 4.03    | 1.18                       |
| Cattle-Cattle (all herds) | 1.12   | 0.61    | 2.32    | 0.82  | 0.51    | 1.27    | 0.29                       |
| Cattle-Cattle (dairy)     | 1.60   | 0.85    | 2.95    | 1.39  | 0.82    | 2.02    | 0.21                       |
| Cattle-Cattle (non-dairy) | 1.28   | 0.67    | 2.66    | 0.93  | 0.58    | 1.50    | 0.35                       |

### 2.4. Distance-based similarity analysis

Each RTA badger was surrounded by an average of 4 other badgers (IQR: 2-7) within a 7km radius. The probability of two RTA badgers sharing the same MLVA type dropped by 17% for every km increase in distance between them (Odds Ratio; OR: 0.86; 95% lower and upper confidence limits; 95%CI: 0.80-0.92; Inverse OR: 1.17); Figure 5A. There were 63 cattle herds (IQR: 39-81) in the 7km radius around each RTA badger, and the probability of RTA badgers and herds sharing MLVA types fell by 9% with every km (OR: 0.91; 95%CI: 0.90-0.92; Inverse OR: 1.09); Figure 5B. When stratified by herd type, we found that each RTA badger was surrounded by 17 dairy herds (IQR: 10-27) and 41 (IQR: 23-59) non-dairy herds, respectively. The probability of RTA badgers and dairy herds sharing MLVA types fell by 7% per km increase in distance (OR: 0.94; 95%CI: 0.92-0.95; Inverse OR: 1.07, and by 11% for non-dairy herds (OR: 0.90; 95%CI: 0.89-0.91; Inverse OR: 1.11). In the cattle-cattle context, each cattle herd was surrounded by 63 others (IQR: 40-88), and we found that the probability of two cattle herds sharing the same MLVA type dropped by 9% for every km increase in distance (OR: 0.91; 95%CI: 0.91-0.92; Inverse OR: 1.09); Figure 5C. There was a slight decrease when considering the dairy herd population independently (OR: 0.93; 95%CI: 0.93-0.94; Inverse OR: 1.07) from non-dairy herds (OR: 0.92; 95%CI: 0.91-0.92; Inverse OR: 1.09).



**Figure 5.** The relationship between probability of an MLVA match between two hosts as a function of distance in the (a) badger-badger context, (b) badger-cattle context, and (c), cattle-cattle context.

#### 2.5. Determining the factors associated with RTA badgers and nearby cattle herds sharing MLVA types

In the final dataset, there were 9,471 RTA badger-cattle pairs with at least one MLVA type in common, and 5,964 RTA badger-cattle pairs with no MLVA types in common. The univariable analysis revealed that the 'number of MLVA types isolated from a bTB breakdown' (OR per every additional type: 1.98; 95%CI: 1.83-2.13) and 'herd size' (OR per every additional ten animals: 1.01; 95%CI: 1.01-1.02) were positively associated with RTA badgers and nearby cattle herds sharing MLVA types. 'Distance between hosts' (OR per km: 0.89, 95%CI: 0.87-0.91), 'the absence of a milk license' (OR: 0.72, 95%CI: 0.67-0.77), 'inwards movements' (OR per every additional ten animals: 0.992; 95%CI: 0.990-0.994) and 'outwards movements' (OR per every additional ten animals: 0.98; 95%CI: 0.98-0.99) were negatively associated with RTA badgers and nearby cattle herds sharing MLVA types. The final model is presented in Table 3, and shows that the factors associated with RTA badgers and cattle sharing MLVA types were increasing 'number of MLVA types during a breakdown', 'decreasing distance between hosts', 'the absence of a milk license' and 'fewer inwards movements'. No significant, biologically relevant interactions were identified. Herd size was correlated with the number of inwards movements ( $r = 0.63$ ), was confounded with herd type and was therefore omitted from the final model; this did not impact AIC scores or model coefficients, notwithstanding that inwards movements was deemed to be the more relevant factor associated with the outcome.

**Table 3:** Model coefficients for the factors associated with RTA badgers sharing at least one MLVA in common with nearby cattle herds. The median (Med) and Inter-Quartile Range (IQR) are reported for continuous variables, and where appropriate the maximum (Max) value is also included. The number of instances and percentage are reported for binary variables

| Variable                                  | Match<br>n = 9,471 (61.4%)        | No match<br>n = 5,964 (38.6%)  | Est.  | Std. Error | z value | OR   | 95%CI Lower | 95%CI Upper |
|---|-----------------------------------|--------------------------------|-------|------------|---------|------|-------------|-------------|
| Intercept                                 | -                                 | -                              | 0.19  | 0.19       | 0.99    | 1.21 | 0.83        | 1.75        |
| Number of MLVA types (per type)           | Med = 1<br>IQR:1-1; Max:10        | Med:1;<br>IQR:1-1; Max:5       | 0.85  | 0.04       | 20.28   | 2.34 | 2.16        | 2.54        |
| Distance (per km)                         | Med = 4.4 km<br>IQR:2.9 km-5.7 km | Med = 4.9 km<br>IQR:3.3 km-6km | -0.12 | 0.01       | -11.61  | 0.89 | 0.87        | 0.91        |
| Milk license (absent)                     | 6,083 (64.2%)                     | 4,261 (71.5%)                  | -0.28 | 0.04       | -7.42   | 0.75 | 0.70        | 0.81        |
| Inwards cattle movements (per 10 animals) | Med = 4<br>IQR:2-9                | Med = 5<br>IQR:2-11            | -0.02 | 0.001      | -11.00  | 0.99 | 0.98        | 0.99        |

### 3. Discussion

Spatial clustering in *M. bovis* molecular types, at various genetic and geographic scales, has previously identified co-localisation of infection between infected livestock and wildlife hosts [20, 27-31], however little attention has been given to what the patterns in spatial distribution of *M. bovis* genetic types in NI reveal about the processes driving the epidemic in, and between, hosts. It is already understood that the *M. bovis* population in Northern Irish cattle herds is characterized by marked spatial structuring and spatial clusters of MLVA types at the herd level [18, 19]. Here, we additionally show that the distribution of infection within clusters is not homogeneous, and that clusters consist of central foci, where 50% of infected herds lie within 35% of the cluster extent. This is consistent with “anchoring” influences in driving spatially restricted epidemics [32]; in Great Britain (GB), some 75% of infection was attributed to local spread [33]. Such processes act over relatively short distances, and can include infection from contiguous herds [26, 34, 35], infected wildlife [6, 36, 37], or the predominance of short-range, between-herd movements over longer-range movements, as observed in GB [38] and Ethiopia [39] (but not in Uruguay, where infection clusters change location year on year, suggesting long-distance spread of disease [40]). Spatial correlation has been reported in disease transmission coefficients at scales <14km, suggesting that a highly localised contact network is an important epidemiological driver of bTB [32]. However, we also identified pairwise distances of over 130km in herds infected with the same *M. bovis* MLVA type, and thus the spatial distribution of MLVA types also has an expansive element. Long-range cattle movements, or moving cattle between distal land-parcels, may drive the wider dissemination of infection [41, 42], but our evidence is consistent with such processes being relatively less important than local factors investigated in driving the overall epidemic. This is consistent with results from France, where spatial proximity to another infected herd was more strongly associated with bTB infection and inwards movements [43].

In our study, there was little compelling evidence of differences in the spatial dissemination of MLVA types between herds with and without a milk license, notwithstanding that dairy herds were associated with outwards moves compared to non-dairy herds. This observation may further reflect the diminished role of long-distance cattle movements in disease spread, compared to other sources. However, the data do tentatively indicate slightly less localised clustering in MLVA types in dairy herds, which could be explained by the fewer short-range, between-herd movements in dairy production. However, as yet the frequencies and Euclidian distances associated with the full cattle movement network in different herd types in NI are unknown [44]; whilst this study confirms that there are more inwards animal movements in non-dairy production, the distribution of trading distances between herds (as in Vernon, 2011) has not yet been derived.

While there was greater richness in MLVA types in cattle herds compared to RTA badgers, the cattle population of 1.6 million is over 40 times larger than badger population, estimated at 33,500 individuals (95%CI 26,000-41,200) [14], and may therefore be able to harbor a larger, more diverse microbial population [30]. Whilst the drivers of this are not yet clear, super-spreading, or historical expansions of the *M. bovis* population, may be implicated [45]. Over 80% of the *M. bovis* MLVA types identified in the RTA badger isolates were also found in cattle, and furthermore, the infection was spatially co-localised in both hosts. This is consistent with previous findings using more limited herd-level data from NI [20], and data from GB, and ROI [27-29]. This association was clear, despite accepted limitations with both the badger and cattle data; farmstead locations are unlikely to represent actual land-parcel (or herd) locations, badgers killed in RTAs may not represent the background badger population, and the RTA dataset is spatially biased to the south-east of NI, and under-sampled in the north-west [16]. Given this, detecting associations despite these disruptive factors means that the actual spatial associations may be even stronger than observed. However, this would require a more thorough systematic sample of *M. bovis* infection in the extant NI badger population. While there is some work ongoing in this area [46], that study is limited to a small area in NI. The probability of RTA badgers and cattle herds sharing *M. bovis* MLVA types decreased by

approximately 9% for every km between hosts, up to a 7km cut-off. Not only does this further confirm the co-localisation of infection in both host systems, but the relatively small rate of change also alludes to the strength of localised influence in maintaining the *M. bovis* population structure, and by extension, the epidemic. There was less dissimilarity when looking at the likelihood of RTA badgers and herds with milk licenses sharing MLVA types (decrease of 7% per km) compared to RTA badgers and herds without milk licenses (decrease of 11% per km). Whilst this difference is small, it suggests greater between-herd homogeneity in the *M. bovis* population in herds with milk licenses, compared to herds without, possibly due to inwards movements driving accumulations of within-herd MLVA type diversity in non-dairy herds [25].

RTA badgers were more likely to have at least one MLVA type in common with herds with milk licenses than herds without. We posit that this reflects animal management practices, as beef herds are likely to operate by purchasing larger volumes of animals, retaining these animals in the herd for only a short period before sending cattle onwards or to slaughter. Beef herds have been linked to the presence of multiple reactors [26] arising from the purchase of animals with undetected infection, possibly from many different geographical locations. Indeed, we identified the inwards movement of animals as a negative influence on whether RTA badgers and cattle share MLVA types. We hypothesise that high cattle turnover in these herds means that the local MLVA types are less likely to become established in the immediate environment, limiting the opportunity for a co-localisation of *M. bovis* genotypes between badgers and cattle. The findings may also suggest that inter-species transmission is perhaps not particularly efficient [9] given that a shared *M. bovis* epidemic is less likely to be observed in herds associated with a high animal turnover. Conversely, animals in dairy settings may be more likely to be exposed to local *M. bovis* genotypes, resulting in the repeated emergence of *M. bovis* genotypes via introduction from local sources, possibly exacerbated by within-herd amplification. The SICCT test may also be less effective in dairy herds [47] and infected dairy cattle may be less likely to exhibit visible lesions post mortem [48]. This is highly suggestive that dairy herds may be at elevated risk of within-herd recrudescence of the same *M. bovis* genotypes compared to beef herds.

### 3.1. Limitations

As alluded to, the main limitation of this study is the nature of the RTA badger survey. The general limitations of these data are well acknowledged [16] but the bias in collection localities was presumed to have the most impact on these results. Nevertheless, this does not limit the utility of the dataset to make some inferences about the spatial relationships in MLVA types in, and between, the badger and cattle populations, with the caveat that a more representative RTA badger dataset would enable more robust inferences to be drawn, to include for example, badger population structure or spatially explicit environmental heterogeneity associated with badger presence [49]. Ideally, a more comprehensive sampling of *M. bovis* genotypes in the badger population in NI would be undertaken. Furthermore, it is understood the cattle farms in NI are highly fragmented, and can consist of multiple, distal land parcels. The use of land parcel locations may provide a more accurate and precise indication of true herd locations and quantify the opportunity for herds to interact spatially with neighbouring herds. This could shed more light on true spatial association of *M. bovis* genotypes in cattle herds. While co-localisation of molecular types between hosts, in various settings and at differing genetic and geographical scales have been reported previously, our analyses extend our understanding of associated risk factors. Further, while molecular epidemiology using MLVA is being superseded by phylodynamics using whole-genome sequencing and modelling to investigate transmission dynamics [7], our results, in a different system, are consistent with previous literature.

## 4. Materials and Methods

### 4.1. Study Area

The area of NI is approximately 13,500 km<sup>2</sup>. The bTB programme is administered across 10 Divisional Veterinary Offices (DVOs) and 123 administrative patches. There are approximately 1.6 million cattle in NI, distributed throughout approximately 20,000 herds. This includes some 2,500 dairy herds (313,549 cattle) and 14,000 beef herds (247,009 cattle), amongst others [50].

## 4.2. Study Data

### 4.2.1. *M. bovis* molecular typing data

Whilst herd-level MLVA surveillance (MLVA typing on the first reactor) has been ongoing since 2003, from 2008, animal-level *M. bovis* MLVA typing is carried out on every SICCT reactor and lesioned animal identified at routine slaughter (LRS). MLVA analysis was carried out using established high resolution methods [18, 19, 51]. The eight *M. bovis* VNTR loci genotyped were MV2163B/QUB11B, MV4052/QUB26A, MV2461/ETRB, MV1955/Mtub21, MV1895/QUB1895, MV2165/ETRA, MV2163/QUB11A and MV3232/QUB3232. VNTR results were concatenated into a Multi-Locus VNTR Analysis (MLVA) string which constituted the molecular type of the isolate; this string was further simplified in a local laboratory nomenclature which reflected the previously assessed herd-level prevalence of MLVA types.

### 4.2.2. Cattle data

Animal-level MLVA profiles of isolates were associated with anonymised breakdown-level data made available from the Animal and Plant Health Information System database (APHIS) [52], administered by the Department of Agriculture, Environment and Rural Affairs (DAERA). This enabled the determination of the number of *M. bovis* MLVA types present in each confirmed bTB breakdown; this dataset has been described in full elsewhere [25]. Additional relevant epidemiological variables included in these data were the breakdown start and end dates, presence of a milk license (dairy herds) or no milk license (non-dairy herds), herd size at the time of bTB breakdown, the number of inwards and outwards cattle movements in the year before breakdown, and the herd DVO. The geo-referencing of registered homestead locations (here referred to as herd locations) were available in the form of the first four digits of the six figure Irish grid reference, which provides reasonable estimations of cattle herd density and distribution.

### 4.2.3. Badger data

The RTA badger dataset has been described previously [16, 20, 53]. Briefly, from 1998, the carcasses of badgers suspected to have died from accidental causes (e.g. road traffic accidents; RTA's) were collected by a wildlife officer from the Department of Agriculture, Environment and Rural Affairs (DAERA). Badger carcasses were checked for the presence of visible lesions consistent with tuberculosis, and defined tissues and bodily fluid samples were also collected for bacterial culture. Culture-confirmed *M. bovis* underwent further MLVA analysis by the Agri-Food and Biosciences Institute (AFBI) to determine the bTB spoligotype and MLVA genotype of the isolate. The geo-referenced collection location for RTA badgers was recorded to within 100m of the actual location. This study included only badgers collected after 2008 to align temporally with the cattle data.

## 4.3. Assessment of MLVA clustering within the cattle population

To visualise the extent of spatial clustering of MLVA types, the distribution of the twelve most common MLVA types were plotted in geographic space. The presence of spatial clustering in MLVA in cattle herds was confirmed by a permutation analysis. Firstly, the median Euclidean distance between herds infected with a given MLVA type was derived. Next, a random sample of herds was selected from the cattle herd population, with the sample size equal to the number of herds infected with the MLVA type of interest. The median Euclidean distance between the herds in this sample was then calculated. This process was repeated 999 times to generate a sampling distribution of distances. Finally, the actual median distance between herds infected with each

MLVA type was compared to those derived from the sampling distribution using Wilcoxon signed rank tests.

Next, a pairwise Euclidean distance matrix between all herds infected with the same MLVA type was generated. The cumulative frequency of these distances was used to investigate the spatial spread of herds within clusters; if the majority of pairwise distances lay below the distance represented by the 50th percentile, this indicated spatial clustering. However, if the pairwise distances were broadly distributed (i.e. 50% of pairwise distances equal to or greater than the 50th percentile), this instead suggested that infected herds are widely distributed across space. This process was also conducted separately for herds with and without milk licenses.

#### 4.4. Intra-and interspecific nearest neighbour analysis

Following similar methods to Trewby (2016) and Abernethy et al., (2011), the Euclidian Nearest Neighbour (NN) distance was calculated between each RTA badger to the closest RTA badger with the same *M. bovis* MLVA type, and to the closest RTA badger with a different *M. bovis* MLVA type. Only badgers within 7km of each other were included, as this represents the approximate 95 percentile of the dispersal movement kernel of badgers in the ROI [54]. Furthermore, the RTA collection dates had to fall within two years of each other; it is reasonable to assume that a two year window will adequately captured co-localisation of infection without introducing ambiguity from associating entities across longer temporal windows. Similar NN distance measures were calculated for RTA badger-cattle herd pairs, repeated separately for only those herds with a milk license, and those without a milk license. Again, only badger and cattle hosts within 7km of each other were included, and the RTA collection date and herd breakdown period had to be within two years of each other. Finally, NN distances were calculated between cattle-cattle pairs, under the same criteria. The null hypothesis was that distances between hosts which share *M. bovis* MLVA types were not significantly different to distances between hosts that do not. This was tested using paired Wilcoxon signed rank tests, and  $p$  was established at  $\leq 0.05$ . To interrogate any limitations in the resolution of spatial data (i.e. the RTA badger and cattle herd co-ordinates were each subject to a 100m error), a sensitivity analysis was conducted whereby the analyses were re-run 100 times, with values between 1m and 99m added to, or subtracted from, each of the cattle and badger latitude and longitude co-ordinates.

#### 4.5. Distance based similarity analysis

Using the approach established by Goodchild et al (2012), the Odds Ratio (OR) of an *M. bovis* MLVA type match was calculated as a function of distance between hosts that yielded MLVA-typed *M. bovis*, up to a distance of 7km. The two-year temporal association was again applied to this analysis. These data were used in the construction of logistic Generalised Linear Mixed Models (GLMMs), with a binary outcome indicating whether or not the *M. bovis* MLVA types matched. The single explanatory fixed-variable was distance, and herd DVO was allowed to vary with a random intercept.

#### 4.5. Determining the factors associated with RTA badgers and nearby cattle herds sharing MLVA types

We modelled factors associated with RTA badgers and nearby cattle sharing the same MLVA type. The outcome of interest was 'whether any cow in a herd within a 7km radius of an RTA badger shared the badger MLVA type', and was entered as a binary variable (1 = yes, 0 = no), and was modelled via a binomial GLMM. This analysis was again limited to breakdowns occurring within two years before or after an RTA collection date. Explanatory variables were: the presence or absence of a milk license, the distance between an RTA badger and registered cattle homestead, herd size, the number of inwards and outwards cattle movements, and the number of MLVA types isolated from a bTB breakdown. Initial univariable analysis involved visual assessments of each predictor, including Cleveland dotplots and boxplots, and fitting loess curves to assess linearity in the logit. Covariates were then assessed for collinearity using both multi-panel scatterplots and

correlation values; variables with a correlation coefficient greater than 0.5 or less than -0.5 were considered for removal, with the aim of retaining the most biologically relevant predictor(s). DVO was allowed to vary with a random intercept. The final model was arrived at via a backwards stepwise routine [55], and the impact of variable removal at each stage was assessed by comparing model AIC values, examining changes in model coefficients and assessing confounding. Where potential confounding was identified, it was investigated by running separate analyses on the suspected confounders. The influence of outliers and influential points was assessed by re-running models with potential influential points removed, and comparing the model coefficients.

All data processing and analyses were carried out using Microsoft Excel and R version 3.4.4 (R Core Team 2014). The packages *rgdal* [56], *rgeos* [57] and *ggplot2* [58] were used to create maps and figures, models were built using *lme4* [59], and *dplyr* [60] was used for data handling.

**Supplementary Materials:** The following are available online; Supplementary Material 1 and Supplementary Material 2.

**Author Contributions:** Conceptualization, Adrian Allen, Robin Skuce and Andrew Byrne; Data curation, Jordon Graham; Formal analysis, Georgina Milne; Funding acquisition, Robin Skuce and Andrew Byrne; Methodology, Georgina Milne, Adrian Allen, Carl McCormick, Eleanor Presho, Robin Skuce and Andrew Byrne; Project administration, Robin Skuce and Andrew Byrne; Resources, Raymond Kirke, Carl McCormick and Eleanor Presho; Software, Georgina Milne and Jordon Graham; Supervision, Adrian Allen, Robin Skuce and Andrew Byrne; Writing – original draft, Georgina Milne, Adrian Allen, Robin Skuce and Andrew Byrne; Writing – review & editing, Georgina Milne, Adrian Allen, Robin Skuce and Andrew Byrne.

**Funding:** This research was funded by the Department of Agriculture, Environment and Rural Affairs for project 16-3-06.

**Acknowledgments:** The authors wish to extend particular gratitude towards the following individuals: Dr Hannah Trewby, whose PhD research inspired some of the methodology, Drs Emily Courcier and Maria O'Hagan who were associated with managing the DAERA badger RTA data, and Prof Darrell Abernethy who previously established methods for investigating RTA badger-cattle relationships in NI. The authors also wish to extend thanks and acknowledgement to all of the AFBI staff involved in bovine tuberculosis histology, pathology, bacteriology, molecular microbiology, case confirmation and MLVA typing, along with the Epidemiology, Molecular biology and Immunology cluster ([https://twitter.com/AFBI\\_EMBI](https://twitter.com/AFBI_EMBI)) for technical support and discussion. We are also thankful to the DAERA Veterinary Service and abattoir staff who deliver the NI bTB eradication programme, those staff who maintain and manage the DAERA APHIS database, and all staff involved the DAERA badger RTA scheme.

**Conflicts of Interest:** The senior author, Dr Andrew Byrne, is a member of the *Pathogens* journal editorial board. Aside from this, the authors declare no conflict of interest. The funders had no role in the design of the study; in the collection, analyses, or interpretation of data; in the writing of the manuscript, or in the decision to publish the results.

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