



The University of
Nottingham

UNITED KINGDOM · CHINA · MALAYSIA

McAloon, Conor G. and Doherty, Michael L. and Whyte, Paul and O'Grady, Luke and More, Simon J. and Messam, Locksley L. McV. and Good, Margaret and Mullaney, Peter and Strain, Sam and Green, Martin J. (2016) Bayesian estimation of prevalence of paratuberculosis in dairy herds enrolled in a voluntary Johne's Disease Control Programme in Ireland. *Preventive Veterinary Medicine*, 128 . pp. 95-100. ISSN 1873-1716

Access from the University of Nottingham repository:

<http://eprints.nottingham.ac.uk/38558/1/Martin%20Prev%20final%20sub.pdf>

Copyright and reuse:

The Nottingham ePrints service makes this work by researchers of the University of Nottingham available open access under the following conditions.

This article is made available under the Creative Commons Attribution Non-commercial No Derivatives licence and may be reused according to the conditions of the licence. For more details see: <http://creativecommons.org/licenses/by-nc-nd/2.5/>

A note on versions:

The version presented here may differ from the published version or from the version of record. If you wish to cite this item you are advised to consult the publisher's version. Please see the repository url above for details on accessing the published version and note that access may require a subscription.

For more information, please contact eprints@nottingham.ac.uk

1 **Title**

2 Bayesian estimation of prevalence of paratuberculosis in dairy herds enrolled in a voluntary
3 Johne's Disease Control Programme in Ireland.

4 **Author names and affiliations**

5 Conor G. McAloon^a, Michael L. Doherty^a, Paul Whyte^a, Luke O'Grady^a, Simon J. More^a, Locksley L.
6 McV. Messam^a, Margaret Good^b, Peter Mullaney^b, Sam Strain^c, Martin J. Green^d

7 ^aSchool of Veterinary Medicine, University College Dublin, Belfield, Dublin 4, Ireland

8 ^bDepartment of Agriculture, Food and the Marine, Kildare Street, Dublin 2, Ireland

9 ^cAnimal Health Ireland, Carrick-on-Shannon, Co. Leitrim, Ireland

10 ^dSchool of Veterinary Medicine and Science, University of Nottingham, Sutton Bonington, United
11 Kingdom

12 **Corresponding author**

13 Conor G. McAloon

14 E-mail: mcaloonconor@gmail.com

15 Section of Herd Health and Animal Husbandry,

16 School of Veterinary Medicine, University College Dublin, Belfield, Dublin 4, Ireland

17 00 353 1 716 6268

18 **Abstract**

19 Bovine paratuberculosis is a disease characterised by chronic granulomatous enteritis which
20 manifests clinically as a protein-losing enteropathy causing diarrhoea, hypoproteinaemia,
21 emaciation and, eventually death. Some evidence exists to suggest a possible zoonotic link and a
22 national voluntary Johne's Disease Control Programme was initiated by Animal Health Ireland

23 in 2013. The objective of this study was to estimate herd-level true prevalence (HTP) of
24 paratuberculosis in Irish herds enrolled in the national voluntary JD control programme during
25 2013-14. Two datasets were used in this study. The first dataset had been collected in Ireland
26 during 2005 (5,822 animals from 119 herds), and was used to construct model priors. Model
27 priors were updated with a primary (2013-14) dataset which included test records from 99,101
28 animals in 1,039 dairy herds and was generated as part of the national voluntary JD control
29 programme. The posterior estimate of HTP from the final Bayesian model was 0.23 - 0.34 with a
30 95% probability. Across all herds, the median animal-level true prevalence was found to be
31 0.032 (0.009, 0.145). This study represents the first use of Bayesian methodology to estimate
32 the prevalence of paratuberculosis in Irish dairy herds. The HTP estimate was higher than
33 previous Irish estimates but still lower than estimates from other major dairy producing
34 countries.

35 “Paratuberculosis”; “Dairy”; “Ireland”; “Prevalence”; “Bayesian”

36 **1. Introduction**

37 Bovine paratuberculosis is a disease characterised by chronic granulomatous enteritis which
38 manifests clinically as a protein-losing enteropathy causing diarrhoea, hypoproteinaemia,
39 emaciation and, eventually death (Sweeney, 2011). Adverse effects on animal productivity in
40 terms of lower milk yield, higher cull rates, reduced value for culled animals, possible adverse
41 effects on fertility and losses due to continued spread of infection are key drivers in the attempt
42 to control the disease at farm level. In addition some research exists to suggest that the
43 aetiologic pathogen *Mycobacterium avium* subspecies *paratuberculosis* (MAP) may pose a
44 zoonotic risk (Chiodini et al., 2012). Consequently, many major dairy producing countries have
45 introduced control programmes aimed at reducing overall prevalence (Geraghty et al., 2014).

46 Animal Health Ireland (AHI) was formed as a not-for-profit organisation providing national
47 leadership and coordination of non-regulatory animal health issues in Ireland (More et al.,
48 2011). The AHI Johne’s Disease Control Programme was developed and introduced as a

49 voluntary programme in 2013. Irish herd-level true prevalence (HTP) on dairy farms in 2005
50 was estimated at 20%, based on the results of a serological survey (Good et al., 2009),
51 considerably lower than estimates across Europe of greater than 50% (Nielsen and Toft, 2009).
52 In common with trends across the EU, the number of dairy herds in Ireland has been gradually
53 decreasing whilst herd sizes have increased. It is therefore possible that HTP has altered in the
54 intervening years.

55 Measuring the impact of control programmes requires an initial baseline estimation of the
56 occurrence of infection. In the context of chronic diseases of slow or insidious onset such as
57 paratuberculosis, incidence may be difficult to calculate and prevalence is often used instead
58 (Messam et al., 2008). A review of the prevalence of paratuberculosis across countries in Europe
59 identified critical issues in a number of studies (Nielsen and Toft, 2009), primarily these issues
60 related to the incorrect values for test sensitivity (Se) and specificity (Sp) in the analysis.

61 Estimates of Se and Sp of diagnostic tests for paratuberculosis vary considerably (Nielsen and
62 Toft, 2008). Much of this variation can be attributed to differences among reference populations
63 and sampling strategies that have been used for the test validation procedure (Greiner and
64 Gardner, 2000). However estimates of Se and Sp may also vary according to prevalence
65 (Brenner and Gefeller, 1997) and therefore between herds (Greiner and Gardner, 2000).
66 Consequently, the relationship between true prevalence (TP) and apparent prevalence (AP) can
67 be expected to vary between populations. It may therefore be unreasonable to assume a fixed,
68 constant, Se and Sp over different populations (Berkvens et al., 2006). In Bayesian analyses, all
69 parameters are considered random variables and can be modelled using probability
70 distributions. Uncertainty and variability associated with estimates of test Se and Sp may
71 therefore be incorporated in the analysis. In addition, in this instance, a Bayesian posterior
72 probability will provide inference on a prevalence estimate, conditional on both currently
73 observed data and previous information about the disease. This methodology has not yet been
74 applied to the estimation of the prevalence of paratuberculosis in Irish dairy herds, but has been

75 used extensively to estimate the prevalence in other countries (Pozzato et al., 2011; Lombard et
76 al., 2013; Verdugo et al., 2015)

77 The aim of this study, therefore was to estimate the HTP and overall animal-level true
78 prevalence (ATP) of paratuberculosis among herds enrolled in a national voluntary control
79 programme.

80 **2. Materials and Methods**

81 Two datasets were analysed in this study. The primary analysis utilised test data collected from
82 the national control programme between 2013 and 2014. Model priors for this analysis were
83 constructed by analysing a secondary (2005) dataset.

84 **2.1. Study Population**

85 The primary (2013-2014) dataset for the current study was obtained from herds voluntarily
86 enrolled in the national voluntary Johne's Disease control programme. Herds enrolled in the
87 voluntary programme are required to have all animals that are 24 months of age and older
88 serologically tested using either serum or milk samples. Diagnostic testing is conducted in both
89 government and commercial laboratories using one of 3 commercial ELISA kits approved for
90 use in the AHI programme; Parachek, Prionics, Switzerland (kit A), Paratuberculosis Antibody
91 Screening Test, Idexx, USA (kit B) and ID Screen, IDVet, Montpellier, France (kit C). Producers
92 that elect to test using blood or milk sample are required to test all eligible animals once or
93 twice per year respectively. Test data, including follow up testing, are stored centrally in the
94 Irish Cattle Breeding Federation computer database. Data were extracted for the period
95 beginning 1st November 2013 and ending 30th December 2014 and included anonymised cow
96 and herd identifiers, test-date, sample-to-positive (S/P) ratio, laboratory interpretation
97 (negative, suspect, positive), sample type (blood or milk), testing laboratory (test kit) and
98 county.

99 Test data also included follow up testing data on subsamples of animals within herds. Herd test
100 data were available for 1,040 herds, 436 of these had conducted 2 or more additional rounds of
101 testing. In order to avoid bias that may have been introduced by some herds conducting greater
102 than 1 herd screen, only one test per animal was used. The first recorded test result for each
103 animal was used for the purpose of this analysis and Se and Sp values were based on a single
104 test strategy. The “herd” in this study was therefore defined as the number of unique and
105 eligible animals on the farm within the 14 month sampling frame.

106 **2.2. Statistical Analysis**

107 **2.2.1. Analytical model**

108 Prevalence was estimated with a Bayesian model extended from that proposed by Branscum
109 (2004), which was based on methodology introduced by Hanson (2003). The number of animals
110 testing positive in each herd was considered to be binomially distributed. A binomial rather
111 than a hypergeometric distribution was used because all adult animals in each herd were
112 sampled. The model was constructed as;

$$113 \quad n_{pos_i} \sim \text{Binomial}(\pi_i, n_{herd_i}) \quad (1)$$

$$114 \quad \pi_i = Se_{jk} \times ATP_i + (1-ATP_i) \times (1-Sp_{jk}) \quad (2)$$

$$115 \quad ATP_i = HTP_i \times CWHP_i \quad (3)$$

$$116 \quad HTP_i \sim \text{Bernoulli}(\mu) \quad (4)$$

$$117 \quad CWHP_i \sim \text{Beta}(a_{CWHP}, b_{CWHP}) \quad (5)$$

$$118 \quad Se_{jk} \sim \text{Beta}(a_{Se}, b_{Se}) \quad (6)$$

$$119 \quad Sp_{jk} \sim \text{Beta}(a_{Sp}, b_{Sp}) \quad (7)$$

$$120 \quad \mu \sim \text{Beta}(a_{\mu}, b_{\mu}) \quad (8)$$

121 where n_{pos_i} equals the number of animals testing positive in the i -th herd ($herd_i$), given a
122 probability of each animal testing positive (π_i) and number of animals in the herd (n_{herd_i}). The
123 probability of a randomly chosen animal from a herd testing positive was a function of the
124 animal-level true prevalence (ATP) within $herd_i$, and the diagnostic test characteristics; Se and
125 Sp , which varied according to kit (j) and test medium (k). The ATP for a given herd was
126 modelled as a mixture distribution: the product of HTP and conditional-herd prevalence
127 (CWHP). The HTP was modelled as a Bernoulli distribution. The Bernoulli distribution is used to
128 model random variables with two possible outcomes, in this case a herd was considered to be
129 “infected” with probability μ to indicate the probability of a randomly chosen herd containing
130 one or more truly infected animals and “uninfected” with a probability $1-\mu$. Then, conditional on
131 the herd being infected, the conditional within-herd prevalence (CWHP) was modelled as beta
132 distribution. Beta distributions are a relatively flexible family of distributions on the real
133 number line from 0-1 and are a common method of modelling prevalence.

134 The effect of ELISA kit and test medium used was assessed using random and fixed effects,
135 however the change in the animal-level apparent prevalence due to the effect of these variables
136 was found to be low (<0.005) and they were removed again from the model.

137 **2.2.2. Model Priors – Test Characteristics**

138 Nielsen and Toft (2008) proposed the case definitions “infected”, “infectious” and “affected” in
139 an attempt to reduce variability between reported estimates of test Se . The subgroup “infected”
140 also includes animals that are “infectious” and “affected”, and is the population of interest in this
141 prevalence study.

142 To estimate the Se and Sp of each commercial kit, a published review of the literature (Nielsen
143 and Toft, 2008) was examined and supplemented with searches in PubMed and CABdirect of all
144 literature published between 2007 and 2015 on paratuberculosis diagnostic test evaluation.
145 Test characteristics for each test kit used in Ireland evaluating the “infected” sub group, were

146 extracted from each peer-reviewed article from this search and from the 2008 review
147 publication (Table 1).

148 The first study was limited to a population of cull cows (McKenna et al., 2005) and the second
149 study (Norton et al., 2010) was carried out on herds with a history of clinical disease and with
150 relatively high ATP. A third study (Nielsen et al., 2013), was removed because the target
151 condition “infected”, was in this case, defined based on the longitudinal interpretation of the
152 evaluated serological test. A final study (Aly et al., 2014) was removed which was based on the
153 evaluation of the test on a single herd.

154 After removing these estimates, 2 evaluation studies were available for kit A with no
155 appropriate published values available for kits B and C. When test characteristics were
156 presented by age group, a weighted mean of the test Se was calculated relative to the age
157 distribution of the present study. A sample size weighted mean was next calculated for the Se of
158 kit A (0.224) using the two estimates extracted from the study. A previously constructed
159 estimate for the Se and Sp of kit B was available (Nielsen and Toft, 2009) which has been used in
160 subsequent prevalence estimates (Pozzato et al., 2011), kits B and C are known to have similar
161 ancestry, therefore the same values were adopted for kit C. The parameters for the beta-
162 distribution were found using “betabuster” software (Chun-Lung 2010) based on a given mode
163 and either upper or lower 95th bound. The Se of individual milk ELISA relative to serum ELISA
164 has been shown to be approximately 0.87 (van Weering et al., 2007), therefore in the absence of
165 a Se estimate for milk, the Se of the serum ELISA was multiplied by a factor of 0.87. Final values
166 and associated beta distribution parameters are shown in Table 2.

167 **2.2.3. Model priors – HTP and CWHP**

168 Prior distributions for HTP and CWHP in Irish dairy herds were required. In order to construct
169 these priors, data (secondary dataset) from a previously published prevalence survey (Good et
170 al., 2009) were used as follows. Data were removed from animals less than 24 months of age,
171 from animals without a recorded date of birth and from non-dairy enterprises. This dataset

172 included a much higher proportion of small herds relative to the primary dataset, therefore,
173 farms containing less than 20 animals were removed to prevent possible overestimation of
174 CWHP priors due to small herd sizes.

175 The CWHP was estimated for each positive herd using the Rogan-Gladen estimator (Rogan and
176 Gladen, 1978), i.e., $CWHP = (AP+Sp-1)/(Se+Sp-1)$, where, AP = Apparent Prevalence. All serum
177 samples in this survey were tested using the Pourquier ELISA, this kit is now sold as Kit B, and
178 therefore, the test characteristics given for Kit B (Table 2) were used to calculate the prior
179 distribution of within-herd prevalences. The distribution of CWHPs in this dataset were plotted
180 and the mean and mode used to fit a beta distribution using the betabuster programme.

181 A number of priors were trialled for HTP including the herd-level apparent prevalence based on
182 a varying number cut point reactors. However, after it was observed that the primary model
183 was extremely insensitive to the prior for HTP, it was decided to use a flat distribution from 0 -
184 1 as the prior for this variable.

185 **2.2.4. Sensitivity Analysis**

186 Sensitivity analysis of the final estimate to the priors used in the model was assessed by varying
187 the point estimate and confidence intervals of the each prior by 10%, 25% and 50% in either
188 direction and repeating the analysis. In addition, the prior for HTP was modelled as a uniform
189 distribution from 0 – 1 and the analysis repeated. The posterior HTP was compared with the
190 estimate from the default priors and the percentage deviation calculated as; $(HTP_S -$
191 $HTP_D)/HTP_D$, where HTP_S and HTP_D represent the posterior estimates of HTP from the
192 sensitivity analysis and the default prior analysis respectively. The model was implemented in
193 WinBUGS Version 1.4.1 with the first 10,000 iterations discarded as burn-in and 50,000
194 iterations used for posterior inference. Convergence was assessed by visual inspection of the
195 time series trace plots and autocorrelation plots and by running multiple (n=3) chains from
196 different starting values. Figures were constructed using the “ggplots2” package in R.

197 **3. Results**

198 **3.1. Descriptive Statistics**

199 **3.1.1. Secondary dataset (2005); Formulation of priors**

200 In total, there were 20,323 test results available from the 2005 dataset. After removing non-
201 relevant results, 5,822 test results from 119 herds were available in the final dataset. The modal
202 value for the prior for HTP was 0.32. The 95% confidence intervals were 0 - 0.92. The beta
203 distribution was fitted with a mode of 0.32 and a 95th percentile of 0.92. The resulting
204 distribution had alpha and beta parameters of 1.18 and 1.25 and 10th, 50th and 90th
205 percentiles of 0.12, 0.48 and 0.86 respectively. Within infected herds, the CWHP was 0.151 with
206 a mode at 0.1, the resulting beta distribution used for the prior had alpha and beta parameters
207 of 2.37 and 13.31 and 10th, 50th and 90th percentiles of 0.051, 0.136 and 0.272 respectively.

208 **3.1.2. Primary dataset (2013-14)**

209 Descriptive statistics are shown in Table 3. After removing error records, data were available
210 for 99,101 animals in 1,039 dairy herds. Average herd size was 95.4 animals, the majority of the
211 herds were located in Leinster (n=249) and Munster (n=719) provinces and these herds also
212 had the greatest average herd sizes (108.5 and 102.1 respectively). Four hundred and forty
213 eight herds (43.1%) had an apparent prevalence of 0, i.e. no animals testing positive. The
214 distribution of apparent prevalence for herds with 1 or more animals testing positive is shown
215 in Figure 1.

216 **3.2. Model outcomes**

217 The median posterior estimate for HTP (95% posterior probability interval) was 0.28 (0.23,
218 0.32). Across all herds, the median ATP was found to be 0.032 (0.009, 0.145), whilst within
219 infected herds, the median CWHP was 0.137 (0.033, 0.348). Figure 2 shows the probability
220 distribution for HTP, along with the distribution of the probability of infection for all of the
221 herds.

222 **3.3. Sensitivity analysis**

223 Overall, the model was reasonably robust to each of the priors used in the analysis. Varying the
224 mode and upper 95th percentile of each prior by up to 50% in either direction resulted in
225 posterior median estimates for the HTP of between 0.265 – 0.323, which were within the 95%
226 posterior probability interval of the original estimate. The posterior distribution for HTP was
227 most sensitive to the prior for CWHP and to the Se estimate for the ELISA. In both cases, the
228 direction of the change of the posterior was counter to the direction of the change for the prior.
229 The model appeared to be relatively insensitive to variation around the prior for HTP and
230 varying this prior by up to 50% in either direction resulted in deviations of less than 0.1% in
231 HTP. Increasing test specificity led to a decrease in the posterior HTP whereas the converse was
232 noted when the specificity was reduced. However, even when the specificity estimate was
233 increased by 50%, the posterior estimate remained very similar, increasing from 0.280 to 0.288.

234 **4. Discussion**

235 This study represents the first use of Bayesian methodology to estimate the true prevalence of
236 paratuberculosis in Irish dairy herds. The posterior estimate of HTP of paratuberculosis among
237 dairy herds enrolled in the national control programme was 0.23 - 0.34 with a 95% probability.
238 Care must be taken when comparing prevalence studies which may have been conducted on
239 different populations using different tests evaluating different target conditions. Previous to this
240 study, only one HTP estimate had been published for paratuberculosis in Ireland (Good et al.,
241 2009). The posterior HTP estimate from the present study was higher than that reported in the
242 2009 study (0.206) (Good et al., 2009). However, the earlier study utilised frequentist methods
243 to estimate the true prevalence of herds with at least one infectious (shedding) animal and was
244 based on a serological test Se of 0.278-0.289. The Bayesian methodology used in the current
245 study however, incorporated uncertainty and variability associated with the test Se by
246 modelling this variable as a probability distribution, the target condition in the present study
247 was “infected” rather than “infectious” and the mode of the distribution used to model test Se
248 was 0.15 and 0.22 depending on the test used. Finally, the previous study was based on data

249 collected in 2005. In the presence of a decline in the number of dairy herds, an increase in herd
250 sizes, and in the absence of a nationally co-ordinated control programme, it is likely that HTP
251 may have increased in the intervening years, resulting in the increased estimate observed in the
252 present study.

253 It is noteworthy that within the population of herds enrolled in the national Control
254 Programme, the estimated overall HTP is significantly lower in comparison to that reported for
255 other countries. Nielsen and Toft (2009) estimated that HTP across Europe was likely to be
256 greater than 0.5 based on limited information available at that time. More recently, Pozzato et
257 al. (2011) found that HTP was likely to be approximately 0.7 in two regions of Northern Italy
258 whilst Verdugo et al. (2015) found a trend of decreasing HTP over a 3-year period in Denmark
259 from 0.92 to 0.75. Finally, Lombard et al. (2013) estimated the HTP in US dairy herds to be
260 approximately 0.91.

261 However, the results of the present study should be interpreted with some caution in the wider
262 context of the disease in Ireland. The primary (2013-2014) dataset used for the current study
263 was based on test results collected from herds enrolled in a voluntary control programme with
264 an average herd size of 95 cows, whereas the national average dairy herd size in 2014 was
265 around 60 cows (Central Statistics Office, 2015; DAFM, 2015). Furthermore, given that
266 herdowners join the national control programme voluntarily, it is likely that herds enrolled
267 within the control programme may differ from the wider population of dairy herds in Ireland.
268 Herd owners may have enrolled in the belief that their herd is free from the disease, with the
269 aim of demonstrating freedom of their herd through the control programme. In this case it
270 might be expected that HTP among herds enrolled in the scheme may be lower than that in the
271 general population. However at the time of this study, a herd classification system was not yet
272 introduced for the scheme, meaning that the benefit for the herd owner when the herd tested
273 negative was not attainable by the farmer in the short term. Conversely herd owners may have
274 joined the scheme in the belief or knowledge that their herd was infected in order to take

275 advantage of tools developed for control of the disease in infected herds. We might expect this
276 to increase the HTP in the study in relation to the national herd level prevalence.

277 The results of the sensitivity analysis (Table 4) suggest that the model was reasonably robust to
278 the selection of priors. Varying the priors by up to 50% had only a modest effect on the primary
279 outcome of interest. Overall, the model was most sensitive to the prior for CWHP and diagnostic
280 test Se.

281 Whilst conducting this research, a previously reported method for modelling CWHP was
282 considered (Branscum et al., 2004). This method utilised a combination of a beta distribution
283 and gamma distribution in order to model CWHP with the form; $\text{Beta}(\mu\psi, \psi(1-\mu))$ where μ is a
284 beta distribution and ψ is a gamma distribution. However, in attempting to use this method in
285 the present study, we noted that the low CWHP and high degree of between-herd variability
286 frequently pushed the parameters of this prior less than 1. The resulting beta distribution
287 became increasingly clustered at 0 when increased variability was introduced. We therefore
288 concluded that this method would not be appropriate for the present study. A single beta
289 distribution was used to model CWHP which combined uncertainty and variability associated
290 with this variable.

291 **5. Conclusion**

292 Paratuberculosis test records from 99,101 animals in 1,039 herds between November 2013 and
293 December 2014 were used to produce a Bayesian estimate of HTP in Irish dairy herds. The
294 median posterior estimate for HTP (i.e. the probability of a randomly selected herd containing
295 at least one truly positive animal), among dairy herds enrolled in the national Johne's Disease
296 Control Programme, was 0.28 (95% posterior probability interval; 0.23, 0.34).

297 **Acknowledgements**

298 This study was carried out as part of the ICONMAP multidisciplinary research programme
299 funded by the Irish Department of Agriculture, Food and the Marine. The authors wish to

300 acknowledge the assistance of Animal Health Ireland and the Irish Cattle Breeding Federation in
301 providing data for the study.

302 **References**

303 Alinovi, C. A., Ward, M. P., Lin, T. L., Moore, G. E., & Wu, C. C., 2009. Real-time PCR, compared to
304 liquid and solid culture media and ELISA, for the detection of *Mycobacterium avium* ssp.
305 paratuberculosis. *Vet Microbiol*, 136, 177-179.

306 Aly, S.S., Anderson, R.J., Whitlock, R.H., Adaska, J.M., 2014. Sensitivity and specificity of two
307 enzyme-linked immunosorbent assays and a quantitative real-time polymerase chain reaction
308 for bovine paratuberculosis testing of a large dairy herd. *Int J Appl Res Vet Med* 12, 1-7.

309 Berkvens, D., Speybroeck, N., Praet, N., Adel, A., Lesaffre, E., 2006. Estimating disease prevalence
310 in a Bayesian framework using probabilistic constraints. *Epidemiology* 17, 145-153.

311 Branscum, A., Gardner, I., Johnson, W., 2004. Bayesian modeling of animal-and herd-level
312 prevalences. *Prev Vet Med* 66, 101-112.

313 Brenner, H., Gefeller, O., 1997. Variation of sensitivity, specificity, likelihood ratios and
314 predictive values with disease prevalence. *Stat Med* 16, 981-991.

315 Chiodini, R.J., Chamberlin, W.M., Sarosiek, J., McCallum, R.W., 2012. Crohn's disease and the
316 mycobacterioses: a quarter century later. Causation or simple association? *Crit Rev Microbiol*
317 38, 52-93.

318 Chun-Lung, S. 2010. BetaBuster programme, Department of Medicine and Epidemiology,
319 University of California, Davis.

320 Central Statistics Office (CSO). 2015. Selected livestock numbers in December.

321 <http://www.cso.ie/multiquicktables/quickTables.aspx?id=aaa06>. Accessed 11/12/2015.

322 DAFM, 2015. Animal Health Computer System. Department of Agriculture, Food and the Marine,
323 Dublin, Ireland.

324 Enøe, C., Georgiadis, M.P., Johnson, W.O., 2000. Estimation of sensitivity and specificity of
325 diagnostic tests and disease prevalence when the true disease state is unknown. *Prev Vet Med*
326 45, 61-81.

327 Geraghty, T., Graham, D.A., Mullaney, P., More, S.J., 2014. A review of bovine Johne's disease
328 control activities in 6 endemically infected countries. *Prev Vet Med* 116, 1-11.

329 Good, M., Clegg, T., Sheridan, H., Yearsely, D., O'Brien, T., Egan, J., Mullaney, P., 2009.
330 Prevalence and distribution of paratuberculosis (Johne's disease) in cattle herds in Ireland. *I Vet*
331 *J* 62, 597.

332 Greiner, M., Gardner, I.A., 2000. Epidemiologic issues in the validation of veterinary diagnostic
333 tests. *Prev Vet Med* 45, 3-22.

334 Hanson, T., Johnson, W.O., Gardner, I.A., 2003. Hierarchical models for estimating herd
335 prevalence and test accuracy in the absence of a gold standard. *J Agr Biol Envir St* 8, 223-239.

336 Jubb, T. F., Sergeant, E. S. G., Callinan, A. P. L., & Galvin, J. W., 2004. Estimate of the sensitivity of
337 an ELISA used to detect Johne's disease in Victorian dairy cattle herds. *Aus Vet J* 82, 569-573.

338 Lombard, J., Gardner, I., Jafarzadeh, S., Fossler, C., Harris, B., Capsel, R., Wagner, B., Johnson, W.,
339 2013. Herd-level prevalence of *Mycobacterium avium* subsp. paratuberculosis infection in
340 United States dairy herds in 2007. *Prev Vet Med* 108, 234-238.

341 McKenna, S.L.B., Keefe, G.P., Barkema, H.W. and Sockett, D.C., 2005. Evaluation of three ELISAs
342 for *Mycobacterium avium* subsp. paratuberculosis using tissue and fecal culture as comparison
343 standards. *Vet Microbiol*, 110, 105-111.

344 Messam, L.McV., Branscum, A.J., Collins, M.T., Gardner, I.A., 2008. Frequentist and Bayesian
345 approaches to prevalence estimation using examples from Johne's disease. *Anim Health Res Rev*
346 9, 1-23.

347 More, S., Doherty, M., Downey, L., McKenzie, K., Devitt, C., O'Flaherty, J., 2011. Animal Health
348 Ireland: providing national leadership and coordination of non-regulatory animal health issues
349 in Ireland. *Revue Scientifique et Technique-OIE* 30, 715-723.

350 Norton, S., Johnson, W.O., Jones, G. and Heuer, C., 2010. Evaluation of diagnostic tests for Johne's
351 disease (*Mycobacterium avium* subspecies *paratuberculosis*) in New Zealand dairy cows. *J Vet*
352 *Diagn Invest*, 22, 341-351.

353 Nielsen, S.S., Toft, N., 2008. Ante mortem diagnosis of paratuberculosis: a review of accuracies of
354 ELISA, interferon- γ assay and faecal culture techniques. *Vet Microbiol* 129, 217-235.

355 Nielsen, S.S., Toft, N., 2009. A review of prevalences of paratuberculosis in farmed animals in
356 Europe. *Prev Vet Med* 88, 1-14.

357 Nielsen, S.S., Toft, N., Okura, H., 2013. Dynamics of specific anti-*Mycobacterium avium* subsp.
358 *paratuberculosis* antibody response through age. *PLoS ONE*, 8 (2013), p. e63009

359 Pozzato, N., Capello, K., Comin, A., Toft, N., Nielsen, S.S., Vicenzoni, G., Arrigoni, N., 2011.
360 Prevalence of paratuberculosis infection in dairy cattle in Northern Italy. *Prev Vet Med* 102, 83-
361 86.

362 Rogan, W.J., Gladen, B., 1978. Estimating prevalence from the results of a screening test. *Am J*
363 *Epidemiol* 107, 71-76.

364 Sweeney, R.W., 2011. Pathogenesis of paratuberculosis. *Vet Clin N Am: Food Anim Pract* 27,
365 537-546.

366 van Weering, H., van Schaik, G., van der Meulen, A., Waal, M., Franken, P., van Maanen, K., 2007.
367 Diagnostic performance of the Pourquier ELISA for detection of antibodies against
368 *Mycobacterium avium* subspecies *paratuberculosis* in individual milk and bulk milk samples of
369 dairy herds. *Vet Microbiol* 125, 49-58.

370 Verdugo, C., Toft, N., Nielsen, S.S., 2015. Within-and between-herd prevalence variation of
371 *Mycobacterium avium* subsp. *paratuberculosis* infection among control programme herds in
372 Denmark (2011–2013). *Prev Vet Med* 121, 282-287.

373 Youden, W. J., 1950. Index for rating diagnostic tests. *Cancer*, 3: 32–35

374