

1 Assessing the impact of a cattle risk-based trading scheme on the movement of bovine
2 tuberculosis infected animals in England and Wales

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34 Running header: Impact of bTB risk-based trading scheme
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36

37 ABSTRACT

38

39 The adoption of bovine tuberculosis (bTB) risk-based trading (RBT) schemes has the
40 potential to reduce the risk of bTB spread. However, any scheme will have cost
41 implications that need to be balanced against its likely success in reducing bTB. This
42 paper describes the first stochastic quantitative model assessing the impact of the
43 implementation of a cattle risk-based trading scheme to inform policy makers and
44 contribute to cost-benefit analyses. A risk assessment for England and Wales was
45 developed to estimate the number of infected cattle traded using historic movement
46 data recorded between July 2010 and June 2011. Three scenarios were implemented:
47 cattle traded with no RBT scheme in place, voluntary provision of the score and a
48 compulsory, statutory scheme applying a bTB risk score to each farm. For each
49 scenario, changes in trade were estimated due to provision of the risk score to
50 potential purchasers. An estimated mean of 3,981 bTB infected animals were sold to
51 purchasers with no RBT scheme in place in one year, with 90% confidence the true
52 value was between 2,775 and 5,288. This result is dependent on the estimated
53 between herd prevalence used in the risk assessment which is uncertain. With the
54 voluntary provision of the risk score by farmers, on average, 17% of movements were
55 affected (purchaser did not wish to buy once the risk score was available), with a
56 reduction of 23% in infected animals being purchased initially. The compulsory
57 provision of the risk score in a statutory scheme resulted in an estimated mean change
58 to 26% of movements, with a reduction of 37% in infected animals being purchased
59 initially, increasing to a 53% reduction in infected movements from higher risk sellers
60 (score 4 and 5). The estimated mean reduction in infected animals being purchased
61 could be improved to 45% given a 10% reduction in risky purchase behaviour by

62 farmers which may be achieved through education programmes, or to an estimated
63 mean of 49% if a rule was implemented preventing farmers from the purchase of
64 animals of higher risk than their own herd.

65

66 Given voluntary trials currently taking place of a trading scheme, recommendations
67 for future work include the monitoring of initial uptake and changes in the purchase
68 patterns of farmers. Such data could be used to update the risk assessment to reduce
69 uncertainty associated with model estimates.

70

71 Keywords: risk factors, risk-based trading, bovine tuberculosis, risk scores

72 **INTRODUCTION**

73

74 Bovine tuberculosis (bTB) is an infectious disease of cattle caused by the bacterium
75 *Mycobacterium bovis* and is one of the biggest challenges facing the cattle farming
76 industry in England and Wales. The cost of controlling bTB is the largest single
77 component of animal health related expenditure in these countries paid by the tax
78 payer, amounting to nearly £100 million in 2014 (Defra, 2014). The adoption of risk-
79 based trading (RBT) has the potential to aid the management of livestock diseases by
80 providing those participating within schemes more accurate information when
81 purchasing animals (Defra, 2013a). However, the performance of such schemes in
82 reducing the movement of infected cattle between farms is dependent on how well
83 schemes are implemented and the specific rules established to permit or prevent trade.
84 Risk scores can be implemented within assurance schemes or certification standards
85 that are managed by industry organisations with a voluntary disclosure of the score, or
86 assisted by government with statutory controls whereby disclosure is compulsory in
87 order for the legal sale of cattle. Scheme rules can dictate whether or not certain
88 batches are permitted to move between herds or zones of different risk scores, and
89 whether a herd score is affected by the purchase of animals of a lower risk status.

90

91 Discussions were facilitated with representatives from the farming community
92 (farmers, auctioneers, private veterinarians, government officials involved in
93 monitoring facilities, and farmer association representatives) at seven meetings during
94 2012-2013 in England and Wales to evaluate how informed cattle trading may vary
95 within different schemes that could be adopted. Understanding the basis of the
96 decisions made by farmers is crucial to the success of any functioning RBT scheme.

97

98 In order to parameterise the model, estimates on the expected level of RBT scheme
99 participation by farmers with the voluntary provision of the risk score was discussed
100 with stakeholders, alongside compliance levels that may be achieved within a
101 statutory scheme based on the compulsory provision of the risk score prior to
102 purchase. From 25 interested stakeholders (farmers, valuers, and representatives from
103 non-government organisations) when asked whether cattle farmers would prefer a
104 voluntary or statutory RBT scheme, 76% (19/25) expressed a preference for a
105 voluntary provision of the risk score, with all Welsh respondents opting for an initial
106 voluntary scheme. However, concerns were frequently raised that without a statutory
107 scheme the system may not be effectively carried out and that there may be
108 differences in its application in different regions. It was felt that for farmers in clean
109 areas, or those that have not experienced a recent breakdown that a statutory system
110 may be favoured. However, for those farms that had experienced a recent breakdown,
111 several stakeholders expressed the view that such farmers would not want to
112 participate in any scheme that reduced the price of their animals or where they had to
113 declare their bTB status. The engagement of farmers in RBT schemes by geographic
114 location, and the purchasing choices given different schemes, were explored and
115 quantitative estimates gained through a follow up questionnaire.

116

117 The aim of this research was to estimate the impact of farmers using risk scores to
118 make more informed choices when buying cattle. The reduction in movements of
119 infected cattle between farms over one year in England and Wales was estimated
120 under three key scenarios: (1) cattle traded with no RBT scheme, (2) voluntary
121 provision of the risk score, and (3) compulsory provision of the risk score in a

122 statutory RBT scheme. Additionally, the impact of changes in calculating the risk
123 score were evaluated together with an investigating into areas of significant
124 uncertainty in input parameters.

125

126

127 **METHODS**

128

129 A stochastic model implemented in Excel with the add on @Risk (version 6.1) was
130 used to estimate the number of infected movements under each of the three scenarios.
131 The final risk score developed using a method described in the accompanying paper,
132 that could be practically applied, is presented in Table 1.

133

134 In this risk assessment each iteration in the model represents a random year with
135 convergence to 4% of the mean value of each output parameter achieved with 5,000
136 iterations using Latin Hypercube sampling. Each individual trading farm was included
137 in the model and separately simulated for the probability of being infected (between
138 herd infection), and if infected, the within herd prevalence was sampled for that herd
139 size. All historical trading events in England and Wales recorded on the Cattle
140 Tracing System (CTS) have been used (July 2010 to June 2011) to estimate the
141 number of total movements and infected movements in one year with no RBT scheme
142 in place. Movements to slaughter have not been included as such movements would
143 not spread infection to new herds. It is assumed that all remaining movements involve
144 a trade between a selling farm and a purchasing farm. The risk assessment uses
145 distributions for certain parameters to describe any known uncertainty or variability
146 associated with input parameters. Where uncertainty could not be quantified within a

147 distribution, separate scenario simulations were carried out to investigate the impact
148 on model results of the level of participation by farmers, bTB between herd
149 prevalence and purchase behaviour by farmers as detailed in the sensitivity analysis.

150

151 **Estimating the number of infected movements per year**

152

153 The number of infected movements per year is dependent on (1) the probability each
154 farm which is selling cattle is bTB infected but the infection is undetected (farm either
155 not under restriction or with specific movement license), (2) the within herd infection
156 prevalence on that farm, (3) the proportion of animals moved from that farm in
157 batches to other farms, and (4) the sensitivity of the pre-movement test where applied.

158 The risk pathway for the movement of infected animals off farm is provided in Figure
159 1. Numerous parameter values were extracted from the National database SAM
160 RADAR bTB reception database, herein referred to as SAM.

161

162 Probability farm infected with bTB, P_{inf}

163

164 For each farm in the dataset the probability of the herd being bTB infected, P_{inf} was
165 estimated using a modified freedom from infection (FFI) model (AHVLA, 2011).

166 This model has been previously developed to estimate the probability that a given
167 herd was free of infection given its test and disease history, $P(free)$ (Martin et al.,
168 2007) and is described in the accompanying paper. There is considerable uncertainty
169 associated with the probability of a herd being infected with bTB which is
170 investigated in the sensitivity analysis. For each iteration, each selling farm is either

171 infected or not, modeled as a Bernoulli random variable, based on the probability of
172 infection per year estimated for that farm.

$$P_{inf} \sim \text{Binomial}(1, 1 - P(\text{free}))$$

173

174 Number of animals infected, N_{inf}

175 The number of infected animals in a herd is dependent on the within herd bTB
176 prevalence and the number of animals within that herd. From a review of the
177 literature, the within herd bTB prevalence applicable to undetected infected herds of
178 varying herd size in England and Wales was not available. To calculate, we first
179 estimated the annual number of infected animals in herds, Inf_{year} , where routine
180 whole herd testing had been carried out in 2011. Where disease is not suspected,
181 whole herd tests are conducted with the single intradermal comparative cervical
182 tuberculin test (SICCT) test. Given the mean sensitivity of the SICCT test, Se_{mean} ,
183 together with the total number of test positive reactors identified in whole herd tests
184 S_{year} (SAM) in England and Wales, the negative binomial distribution was used to
185 describe the total annual number of infected animals in tested herds:

$$186 \quad Inf_{year} \sim \text{Negbin}(S_{year} + 1, Se_{mean}) + S_{year},$$

187

188 The estimated within herd prevalence for individual herds, P_{prev} was then sampled
189 from the surveillance dataset, representing those herds assumed to be infected, such
190 that the cumulative estimated number of infected animals per year across herds
191 equalled the expected number infected per year Inf_{year} . This subset included herds
192 where no reactors had been found ($S=0$)

193
$$P_{prev} \sim \frac{Negbin(S+1, Se)+S}{h},$$

194 where S denotes the number of reactors per surveillance herd identified by the SICCT
 195 test in 2011 (SAM), Se is the sensitivity of the SICCT test, and h is the total number
 196 of animals tested in that surveillance herd (SAM). The negative binomial distribution
 197 was truncated to ensure that the number infected in an individual herd (reactors and
 198 false negatives) was not greater than the total number of animals tested in that
 199 surveillance herd. The distribution of bTB within herd prevalence was generated from
 200 500,000 iterations to ensure convergence to 4% of the estimated mean. Results were
 201 filtered to include only those iterations where the observed 2011 England and Wales
 202 reactor herds were included in the subset and are provided in Table 2.

203

204 The distribution of the sensitivity of the SICCT test at the herd level was described
 205 using the Beta distribution with values of $\alpha = 6.66$ and $\beta = 6.37$ (Downs et al., 2011).
 206 At the national level, Se_{mean} , a mean sensitivity of 0.511 was used for the SICCT
 207 test. The estimated prevalence of bTB on infected farms, not previously suspected of
 208 disease, decreases with increasing herd size, following the same trend as the
 209 prevalence of detected reactors on infected farms. Note, this is not the probability of a
 210 farm being infected, but the level of infectivity on farms that are infected. Separate
 211 cumulative probability distributions representing the uncertain within herd prevalence
 212 by herd size were applied in the model. Given the estimated within herd prevalence, a
 213 binomial distribution was used to estimate the variable number of infected animals on
 214 each infected farm from the total number of animals on farm:

$$N_{Inf} \sim Binomial(Herdsize, P_{prev})$$

215 where $Herdsize$ was the average number of animals on farm (SAM).

216

217 Allocation of infected animals to off movements or remaining on farm, $N_{Inf_{total}}$
 218
 219 Each selling farm may move animals off to a number of different locations during one
 220 year. Paired movements between all farms between July 2010 to June 2011 was
 221 extracted using the Cattle Tracing System (CTS). The estimated number of infected,
 222 N_{inf} being allocated to these different batches moved off farm, or remaining on the
 223 farm, was assumed not to be dependent on animal infection status. The probability of
 224 any one infected animal being allocated to a batch was therefore equal to the number
 225 of animals sold in that batch divided by the original total number of animals in the
 226 herd. For most farms there was more than one batch movement sold per year.
 227 Therefore, a multinomial distribution was implemented as a set of nested binomial
 228 distributions to describe the between year variability for allocation of infected animals
 229 to batches or remaining on farm:

$$N_{Inf_{total}} \sim Multinomial(N_{Inf}, \{P_{farm}, P_{batch1}, P_{batch2} \dots P_{batchn}\})$$

$$N_{Inf_{total}} = N_{inf_{farm}} + N_{inf_{batch1}} + N_{inf_{batch2}} \dots + N_{inf_{batchn}}$$

230 where $N_{Inf_{farm}}$ is the number of infected animals allocated to remain on farm, and
 231 $N_{Inf_{batchn}}$ the number allocated to batch n . Where the selling farm is located within
 232 an area subject to annual or bi-annual bTB tests (areas of high bTB incidence), all
 233 cattle over 42 days of age require a pre-movement test to be taken 60 days prior to
 234 movement. Within the risk assessment it is assumed that all animals originating from
 235 farms located in the high risk area are tested and. This is a simplification as there are
 236 movements which would be exempt from testing including animals under 42 days and
 237 those licensed between Approved Finishing Units (AFUs) and certain farms under
 238 restriction. It was assumed that each infected animal had the same likelihood of

239 testing positive in the absence of any latent period included in the model. A binomial
240 random variable with the number of infected animals in that batch and the sensitivity
241 of the SICCT test, Se , was sampled for the variability associated with a positive pre-
242 movement test. Given any positive results it was assumed that the entire batch was not
243 sold. Detection of positive animals in the pre-movement test would result in trading
244 restrictions placed on the farm thereafter. However, given that all movements occur in
245 one annual time step with no chronological order, the assumption was made that batch
246 results were independent from other batch results for that source farm. This
247 simplification made does not affect the comparison of RBT schemes because the entire batch
248 is removed from all schemes for that iteration.”

249

250 **Estimating the impact of a voluntary scheme**

251

252 This scheme was based on the risk score of the seller (S_{score}), being made available
253 voluntarily to auctioneers and purchasers prior to purchase by the seller. The risk
254 score of the purchaser (P_{score}), may influence which animals they buy. The risk
255 pathway for one selling farm is shown in Figure 1 and was used to estimate the
256 infected and uninfected animals in each batch. This risk pathway was extended with
257 an example batch as shown in Figure 2 to take account of whether or not the
258 purchaser participates in a scheme and, given participation, whether or not the
259 purchaser accepts the risk score of the seller. A ‘failed initial movement’ occurs when
260 the purchaser does not accept the sellers risk score.

261

262 Probability of participating in a trading scheme, $P_{scheme}(P_{score}, P_{region})$

263

264 The percentage of farmers that would be likely to purchase through a voluntary RBT
265 scheme was discussed at seven meetings with stakeholders during 2012-2013 in
266 England and Wales, with a follow up questionnaire (available from corresponding
267 author). There were 17 quantitative estimates received. Stakeholders felt that there are
268 many dependencies to be factored into estimates generated including the individual
269 bTB status and circumstances of the purchaser and how successfully the scheme was
270 rolled out. For Wales, it was deemed that the level of uptake of an RBT scheme would
271 differ by region. Therefore, different estimates for uptake were calculated for regions
272 defined as Low risk and High risk. Estimates were also stratified by purchasers risk
273 score as it was thought that incurring a breakdown in recent years would influence the
274 purchasing farmers' behaviour. The effect of differences in the purchasing relating to
275 farm herd type was also raised. For example finishing farms (animals fattened for
276 slaughter) were considered less likely to be concerned about the bTB risk of animal
277 than breeding farms, however, insufficient data were available to include stratification
278 by farm type in the model. The opinion elicited is provided in Table 3. The probability
279 of farmers purchasing through a voluntary scheme was associated with significant
280 unquantified uncertainty which was further investigated in the sensitivity analysis.
281 Over one year it was assumed that each batch purchaser elected either to participate in
282 the scheme or not for all batches destined for that farm represented by a Bernoulli
283 random variable.

284

285 Probability of purchase given risk score, $P_{buy}(P_{score}, S_{score})$

286

287 For those farmers participating in the scheme, the probability that farmers will buy
288 certain animals will depend on their own farm status, their risk appetite, and also on

289 the information provided by the score regarding the animals for sale. As with the
290 percentage of farmers using the scheme, there will be considerable variability between
291 farmer needs (breeding farmer purchasing versus farmer restocking large numbers),
292 other factors, such as the price of the animal, and on the overall ‘trust’ a farmer places
293 in the risk scores and on the local implementation of the RBT scheme including the
294 amount of educational activities rolled out with schemes. Stakeholders were asked to
295 consider a hypothetical farmer that was interested in using risk scores. For each risk
296 score pairing (seller score – purchaser score), respondents were asked to select a
297 probability ranging from “Will” to “Will not” divided into six increments. Each of the
298 boxes was associated with a probability, with a maximum of 100% representing
299 “Will” and minimum of 0% for “Will not” with 1%-25%, 26%-50%, 51%-75%, and
300 76%-99% for the middle four boxes. There were 12 quantitative responses provided
301 with 5 unknowns (5 stakeholders did not answer this question in the questionnaire). A
302 discrete distribution was then simulated until convergence for each pairing to estimate
303 the combined expert opinion mean, maximum and standard deviation of the
304 associated uncertainty. The uncertain probability of purchase for each pairing of risk
305 score between purchaser and seller was applied in the risk assessment using a fitted
306 lognormal distribution using the key statistics of the distribution shown in Table 4. A
307 Bernoulli random variable with the given probability was sampled for the variability
308 associated with the decision to purchase given the risk score.

309 **Estimating the number of infected movements within a statutory scheme**

310

311 The statutory scheme was based on the compulsory provision of the risk score to
312 auctioneers and purchasers prior to purchase. In a perfect system this would imply
313 that all purchasers would be involved in the scheme with $P_{Scheme(P_{score}, P_{region})}=1$.

314 However, the potential for purchase of animals from farmers not using the system was
315 discussed with stakeholders and a minimum of 5% and maximum 15% thought to be
316 plausible bounds for the uncertain probability of not complying with the scheme, with
317 a mean value of 10%. For those farmers participating in the statutory scheme, the
318 probability of the purchase being made ($P_{buy(P_{score}, S_{score})}$), given that the risk score
319 was made available was assumed to be the same as that estimated within the voluntary
320 scheme.

321

322 **Estimating the impact of changes to calculating the risk score**

323 The baseline risk score for each farm, as described in the accompanying paper, was
324 based on selected risk factors from a full model identified by a logistic regression. The
325 impact of including some of the removed risk factors (region risk West England and
326 Wales, and breakdowns > 10 years previously) on the performance of the score was
327 investigated together with a more simplified scheme (only 0-2 years since breakdown
328 and breakdown information without high risk movements), and finally the impact of
329 implementing a rule whereby farmers are not permitted to purchase animals of higher
330 risk status than their own herd.

331

332 **Parameter uncertainty and sensitivity analysis**

333 During development of the risk assessment several key parameters were identified as
334 being uncertain with little available information to describe that uncertainty.

335 Therefore, upper and/or lower limits of parameters were identified and implemented
336 in separate simulations of the risk assessment:

337

338 (1) The between herd prevalence of bTB $p(inf)$, calculation uses a value from the
339 literature that herds cannot achieve a probability of freedom greater than 62%
340 for 24 months post breakdown (detailed in the accompanying paper). The
341 uncertainty associated with this value is not known. To estimate the impact of
342 this uncertainty, the probability of infection for each farm was increased and
343 decreased by 5% and separately simulated.

344 (2) The level of participation of farmers in a voluntary RBT scheme,
345 $P_{scheme}(P_{score}, P_{region})$ was acknowledged in discussions as being highly
346 uncertain - relating to farmer trust in that RBT scheme and ease of use and
347 accessibility. Model scenarios were run at levels of 20%, 40%, 60%, 80% and
348 100% farmer participation to evaluate the relationship between participation
349 and performance of the scheme.

350 (3) The probability farmers would still purchase high risk animals once bTB
351 information was provided, $P_{buy}(P_{score}, S_{score})$, was associated with the
352 purchasers status and the amount of education and explanation that
353 accompanied the roll out of any scheme, which at present is uncertain. To
354 investigate the impact of RBT schemes that change the baseline probability of
355 buying higher risk animals, a scenario was simulated where all purchasing
356 farmers were 10% more likely and 10% less likely to purchase higher risk
357 animals than the values elicited for the baseline model.

358

359 A sensitivity analysis based on Analysis of Variance (ANOVA) was undertaken. An
360 ANOVA was selected as it has previously provided robust insights regarding
361 identification of key inputs in probabilistic risk assessments, for example, Mokhtari
362 and Frey, 2005. The reduction in infected movements comparing no RBT scheme and

363 a statutory RBT scheme at 90% compliance per farm was used as the response
364 variable. Predictor variables were values of each input parameter for that farm
365 represented by a range. The ANOVA was populated with half a million randomly
366 selected farms.

367

368 RESULTS

369

370 The trade in cattle between farms without a RBT scheme, with a voluntary RBT
371 scheme, and with a statutory RBT scheme, were simulated over one year for each
372 farm. The number of infected movements that, if pre-movement tested, batch tested
373 clear was summed and stratified by country and area. It was assumed in the baseline
374 and each scenario that all movements from herds in the high risk area were pre-
375 movement tested. Uncertainty and variability considered in the model was represented
376 by 5th and 95th percentiles (within parentheses), which indicate the range within which
377 90% of the results lie. Uncertainty was also considered in separate scenario runs of the
378 risk assessment. It should be emphasised that not all variability and uncertainty has
379 been estimated in the calculations and scenarios, as not all can be quantified.

380 Therefore results describe the amount of *quantified* variability and uncertainty
381 included in the assessment. Results stratified by region and by farm risk score, are
382 presented in the supplementary materials.

383

384 **Results with no RBT scheme**

385 For trade in cattle with no RBT scheme there were 379,951 batches of animals moved
386 off farm in England and Wales to another farm in England and Wales where the risk
387 score and region of the seller and purchaser was determined. As shown in Table 5,

388 this represented a total of 1.2 million animals with 18.4%, 43.3%, 5.6%, 10.8%, and
389 22.0% of animals sold by farms scoring 1, 2, 3, 4, and 5 respectively, where a score of
390 1 is the lowest risk and a score of 5 is the highest risk score. An estimated mean of
391 35,588 infected animals were on farms from which off movements occurred (farms
392 not under restriction or those restricted but with a specific licence to move to another
393 restricted facility) with 5th and 95th percentiles that this varied between 32,881 and
394 38,369. Of these infected animals, approximately 11% or 3,981 (5th 2,775, 95th 5,288)
395 were sold to purchasers in England and Wales with the majority remaining on farm.
396 Of those 3,981 infected animals per year, an estimated mean of 41.8% infected
397 animals were sold by farms scoring 5, rising to an average 60.2% for farms scoring 4
398 or 5, whilst 6.1% were estimated to be sold from the lowest risk farms scoring 1.

399

400 **Voluntary RBT scheme**

401 Uptake by farmers for a voluntary RBT scheme was estimated to vary between 40%
402 to 81%, as shown in Table 3, dependent on location and purchaser bTB status. Table 5
403 presents the estimated results from implementation of a voluntary RBT scheme with
404 approximately 17% of animals that were traded with no RBT scheme being rejected
405 by the initial buyer. It can be seen that the estimated trade from lower risk sellers was
406 found to be less affected, with trade from higher risk sellers being most affected to
407 low risk purchasers. The estimated trade was most affected in the high risk areas in
408 England and Wales (regional differences shown in supplementary materials). There
409 was an estimated mean rejection of 23% (5th 22%, 95th 25%) of infected animals by
410 purchasers based on sellers providing the risk score voluntarily.

411

412 **Statutory RBT scheme**

413 Under a statutory RBT scheme with an estimated mean compliance of 90% of
414 purchasers having access to the risk score of the seller an estimated mean of 26% of
415 animals were rejected once the risk scores were made available. The majority of
416 estimated trade to low risk purchasers (score 1) from high risk sellers (score 4 or 5)
417 was affected by the implementation of a statutory scheme. Of the estimated number of
418 infected animals on farm a mean of 37% (5th 35%, 95th 39%) of infected animals were
419 rejected by purchasers. Of those infected animals rejected from sellers, the majority
420 are estimated to be those sold by high risk farms (score 4 or 5), with on average a 53%
421 reduction in infected movements from those farms.

422

423 **Alternative schemes**

424 Figure 3 displays the boxplot of different RBT schemes according to the estimated
425 mean percentage reduction of infected movements. Results using the baseline risk
426 score are presented in dark green and highlights the linear relationship between the
427 percentage uptake by farmers and the percentage reduction achieved by that scheme.
428 The dark green dashed line through the simulation results represents the uncertainty
429 regarding the level of uptake for each scheme. The dashed black vertical lines through
430 each box plot represent the between year variability and uncertainty about the mean
431 simulation result and terminate at the estimated minimum and maximum value.

432 Variations on the baseline risk score used in an RBT scheme, adding or subtracting
433 certain risk factors from the scoring system (as described in the accompanying paper)
434 at 90% compliance has been provided together with an extrapolation of how those
435 schemes would perform. From the results it can be seen that there are only marginal

436 increases in the performance of the scheme given the addition of risk factors selected
437 from the logistic regression (region risk West England and Wales, and breakdowns >
438 10 years previously). The impact of a ban on farmers purchasing below their farm risk
439 score, assumed to be implemented with 100% compliance yields a 49% reduction the
440 initial purchase of infected animals (5th 47%, 95th 51%).

441

442 **Parameter uncertainty and sensitivity analysis**

443

444 There were three important parameters identified by the ANOVA: (1) the uncertain
445 probability of the purchaser buying the animal once the sellers score was shown
446 (derived from expert opinion) $P_{buy}(P_{score}, S_{score})$, (2) the variable risk score of the
447 seller, S_{score} , and (3) the variable risk score of the purchaser, P_{score} . It should be
448 noted that the uncertain level of compliance for the statutory scheme,
449 $P_{scheme}(P_{score}, P_{region})$ was significant but less significant than the top three. For the
450 voluntary scheme, the uncertainty associated with the probability of participating in
451 the scheme was also highly important.

452

453 In addition to the sensitivity analysis, scenarios were identified during model
454 development and parameterisation where there was limited information on parameter
455 uncertainty with results shown in Table 6 and displayed in the boxplot in Figure 3.
456 The true between herd prevalence of bTB infection, P_{inf} , the proportion of herds that
457 have at least one infected animal, is associated with considerable uncertainty from the
458 freedom from infection model (AHVLA, 2011) which is heavily reliant on input
459 assumptions. Using alternative parameterisations, the performance of RBT schemes
460 was within the convergence values for the original parameterised simulations. This is

461 due to the fact that the percentage change in infected movements is not dependent on
462 the scale of the true prevalence, only the pattern of the true prevalence across English
463 and Welsh farms. However, the absolute number of infected movements per year was
464 significantly affected. Decreasing the between herd prevalence by 5% decreased the
465 number of infected movements by a mean of 22%, whilst increasing by 5% increased
466 the average number of infected movements by 21%.

467

468 Simulations were carried out varying the percentage uptake by farmers and the
469 percentage reduction achieved. For every 10% of farmers that participated in the
470 baseline scheme there was an additional 3.8% reduction in the initial purchase of
471 infected animals until the mean estimated maximum of 38% was reached at the
472 maximum of 100% participation.

473

474 The greatest increase in performance of the score arose from a 10% decrease in the
475 baseline estimates for risky farmer behaviour (purchasing cattle at higher risk than
476 their own farms) with a 45% mean reduction in the initial purchase of infected
477 animals (5th 43%, 95th 47%). This result concurs with the identification in the
478 ANOVA of this parameter as having the highest impact on the RBT performance
479 output considering the associated quantified uncertainty and variability.

480

481 DISCUSSION

482

483 Cattle trading patterns are complex and dynamic due to seasonal factors, economic
484 factors and changes in Government controls. Nevertheless a quantitative approach to
485 estimating the impact of a RBT scheme was possible for England and Wales. It was

486 possible to estimate with a reasonable amount of confidence the impact of a specific
487 scheme over one year and show that a significant impact could be achieved with the
488 reduction of movements from high risk areas or high risk farms.

489

490 One of the major reasons for adopting a quantitative approach was the need to account
491 for the dynamic movement patterns between farm types and farm areas and regional
492 differences in the application of control measures. Historic paired movements were
493 used which linked direct farm to farm animal movements and those via markets to
494 farms. This allowed a comparison between high and low risk areas and different
495 trading schemes. The absolute results for the number of animals infected and traded
496 was dependent on the scale of the between herd and within herd prevalence. The
497 between herd prevalence was associated with uncertainty not quantified in the model.
498 However, the comparison between cattle traded with no RBT and the different RBT
499 schemes was not dependent on the magnitude of prevalence – only the regional or
500 farm characteristic pattern. It was apparent that changes in the calculation of the
501 between herd prevalence could have a significant effect on the absolute number of
502 infected movements predicted. The provision of values for the number of infected
503 animals with associated uncertainty is, however, provided as such values are
504 important for economic analyses when considering the cost benefits of establishing
505 and maintaining a RBT scheme. Before consideration could be made of a statutory
506 scheme, a cost-benefit analysis would be required estimating the full costs of
507 implementing a scheme, such as impacts on trade and adjustments of the market,
508 together with the benefits of reduced disease spread.

509

510 Analysis of the results from the risk assessment demonstrated the importance of
511 encouraging maximal uptake of schemes. The sensitivity analysis and parameter
512 uncertainty scenarios demonstrated the importance of farmer purchase
513 behaviour, $P_{buy}(P_{score}, S_{score})$ on the performance of any RBT scheme. The quantified
514 uncertainty associated with this parameter could be reduced from gathering
515 appropriate data from any pilots conducted. In addition, careful consideration should
516 be given to any programme of education of farmers which could result in reducing
517 risky purchase behaviour, thereby considerably improving the performance of RBT
518 schemes. Importantly, we repeatedly heard at stakeholder meetings that many farmers
519 believed that if an animal had been tested for bTB, then that animal was not infected,
520 i.e. they considered that the bTB test applied was 100% sensitive. This may lead to
521 the conclusion that further effective education of farmers may be warranted. The
522 England TB RBT group also identified that a voluntary scheme will only succeed if a
523 critical mass of farmers participate (Defra, 2013a). This will depend on how well any
524 scheme is rolled out, ease of use, trust, the level of understanding achieved of the risk
525 posed by purchasing cattle to herds and sufficient information being made available to
526 farmers to make an informed choice.

527

528 In the absence of any RBT scheme being piloted in England and Wales during the
529 lifetime of this research project, the values elicited by expert opinion represented a
530 ‘best guess’, however, it is the only data currently available. Should any schemes be
531 piloted, it would be advisable to monitor initial uptake and changes in farmer
532 behaviour to update the risk assessment. For example, Gates and colleagues
533 monitored the change brought about by cattle movement restrictions on Scottish farms
534 (Gates et al., 2013). Such data would be invaluable to reduce the uncertainty

535 associated with model estimates. Given sufficient data, further work could investigate
536 the most likely fate of those infected movements that initially fail from high scoring
537 sellers. The England RBT group commented that a short research project be
538 conducted after an introductory period to investigate engagement and behavioural
539 change. This may include a survey of auctioneers as to whether any risk-based trading
540 data has been included in catalogues or on screen/boards at point of sale and how
541 many buyers are asking for the risk score prior to purchase. Statutory databases could
542 also be queried as to whether any significant changes had occurred to paired
543 movements (particularly those deemed the most risky) between/into/out of selected
544 geographical/incidence based/score based categories. An alternative would be a check
545 on the average distance travelled for movements from holdings of certain categories.

546

547 A RBT scheme would reduce infection transmission attributable to cattle movements
548 which is one transmission pathway contributing to the bTB epidemic (Gopal et al.,
549 2006). This would reduce the between herd prevalence (the proportion of farms with
550 at least one infected animal). In the risk assessment, historical movements are either
551 accepted or rejected; the model makes no attempt to reallocate the movement to
552 another farm or area once the original trade is declined. However, at the market or
553 sale, another farmer may purchase the rejected batch at a lower price. Alternatively
554 farmers with high scores may seek out other purchasing farmers with the same risk
555 status for trade, for example, with the development of 'orange' markets. The model
556 indicates that, given the introduction of a RBT scheme, there would be significantly
557 less infected animals purchased by low scoring farms, particularly for those low risk
558 farm that are located in the high risk area (HRA). If those rejected movements were
559 sold to high risk farms, which may already be harbouring undetected infection, this

560 may, in the long term, increase the bTB within herd prevalence of those herds
561 engaging in this risky behaviour. Unfortunately, the risk assessment is simulated only
562 over one year and therefore cannot quantify the long-term changes that may eventuate
563 from implementation of risk-based schemes, however, if such farms resided in an area
564 of higher testing frequency, such as the HRAs in England and Wales, detection of
565 those infected animals may occur earlier due to a higher prevalence of infection on the
566 test farm, and increased frequency of testing in the form of pre-movement tests and
567 annual whole herd tests thus complementing and potentially improving the sensitivity
568 of the current regional controls in place.

569

570 CONCLUSIONS

571

572 In conclusion, this paper details the design of the first risk assessment to measure the
573 impact of theoretical risk-based animal trading schemes based on a given farm risk
574 score for bTB. If a voluntary or statutory RBT scheme was in place, a significant
575 impact could be achieved with the reduction of infected movements from high risk
576 areas or high risk farms. Key to reducing infected movements through a risk-based
577 trading scheme is promoting maximal uptake in schemes and on reducing risky farmer
578 purchase behaviour.

579

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