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# 1 Assessing the environmental impact of ruminant production systems

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## 5 1. Introduction

6 As discussed elsewhere in this book, one of the most common methods to evaluate environmental  
7 footprints of farming systems is life cycle assessment (LCA). Although LCA itself is suitable for and  
8 indeed adopted by a wide range of industries far beyond agriculture, what separates agriculture, and  
9 in particular pasture-based ruminant production systems, is the high degree of uncertainties  
10 associated with physical, chemical and biological processes that underpin production (McAuliffe *et al.*,  
11 2018a). In the presence of uncertainties, point-estimates provided by LCA models are unlikely to be  
12 informative enough to offer robust policy implications (Chen and Corson, 2014); when this is the case,  
13 the resultant environmental burdens must be expressed in the form of probability distributions and  
14 interpreted accordingly (McAuliffe *et al.*, 2017).

15 For carbon footprint (CF) analysis of ruminant systems, one significant challenge of collating a  
16 life cycle inventory is uncertainty associated with emission factors (EF), or parameters linking nutrient  
17 inputs into the system to greenhouse gas (GHG) emissions arising from the system (Pouliot *et al.*,  
18 2012). On commercial livestock farms, various factors can affect their relationships; for example,  
19 weather, soil, plant/animal genetics, management practice and interactions between them. Despite  
20 these variabilities, the majority of LCA studies adopt EFs derived outside the actual system boundary,  
21 most commonly as parameters defined as part of Intergovernmental Panel on Climate Change  
22 guidelines (IPCC, 2006). As these “generic” EFs are designed to be applicable to a wide spectrum of  
23 production environments within a single agroecological zone, a considerable level of uncertainty  
24 surrounds each of these values (Dudley *et al.*, 2014). As a case in point, the two parameters for nitrous  
25 oxide (N<sub>2</sub>O) emissions suggested by IPCC (2006), commonly known as  $EF_1$  (% fertiliser N lost as N<sub>2</sub>O)  
26 and  $EF_{3PRP}$  (% urine and dung N deposited on pasture lost as N<sub>2</sub>O), are both deemed to have a 95%  
27 confidence interval between -67% and +300% of the respective point estimates.

28 To facilitate evidence-based debates about the environmental impact of ruminant production  
29 systems and, by extension, the role of ruminants in global food security, it is therefore imperative to  
30 improve reliability of EFs in a more location-specific context. This, however, requires a significant  
31 investment into field-based research, something that is not always feasible for practical reasons.

32 Through a review of recent literature and a quantitative case study, this chapter explores how this  
33 practical trade-off between feasibility and scientific rigour should be addressed.

## 34 **2. LCA applied to ruminant production systems**

35 LCA has been applied to all major ruminant production systems (beef, dairy, lamb and wool), albeit  
36 with different degrees of scrutiny into system-wide uncertainties. Frequently cited examples of works  
37 on the sheep sector include [Biswas \*et al.\* \(2010\)](#), [Ripoll-Bosch \*et al.\* \(2013\)](#) and [Wiedemann \*et al.\* \(2015a\)](#). For dairy, [Üçtuğ \(2019\)](#) provides an extensive literature review encompassing 31 studies. In  
38 addition, [Poore and Nemecek \(2018\)](#) is accompanied by a global database of agri-food LCA results  
39 covering both plant-based and animal-based commodities.  
40

41 For the beef sector, [de Vries \*et al.\* \(2015\)](#) review and compare findings from 14 studies based  
42 on contrasting farming systems from around the world. It is noteworthy, however, that the popularity  
43 of beef LCA research has substantially increased over the last three years, likely due to reports  
44 indicating that the sector, and in particular grazing systems, are extremely heavy contributors to global  
45 GHG emissions ([Herrero \*et al.\*, 2016](#); [Springmann \*et al.\*, 2016](#)). In response to such a rapid rise in  
46 attention paid to the industry, the remainder of this section will give a summary of 14 beef LCA studies  
47 that have been published between 2015 and 2018. Papers that are not written in English and that  
48 focus on end-point modelling are excluded here.

49 In Brazil, [Dick \*et al.\* \(2015\)](#) conducted an LCA of beef cattle in two grassland systems. The first  
50 system was based on traditional grazing practices where animals can wander freely and receive little  
51 or no supplementation. The second system, termed “improved”, involved weekly rotational grazing  
52 and the introduction of winter forage species. The system boundary was from raw material extraction  
53 to farm gate and the functional unit (FU) was 1 kg liveweight gain (LWG). Data on beef production  
54 within the two systems were sourced from published literature, and GHG emissions were calculated  
55 according to IPCC guidelines. Global warming potential (GWP) for the traditional system was found to  
56 be 22.5 kg CO<sub>2</sub>-eq per kg LWG, while GWP for the improved system was 9.2 kg CO<sub>2</sub>-eq per kg LWG.  
57 This dramatic reduction in GWP was attributed to higher quality forage with increased digestibility in  
58 the alternative system, resulting in faster LWG.

59 [Mogensen \*et al.\* \(2015\)](#) estimated CFs of beef production systems in Denmark and Sweden. The  
60 system boundary was from cradle to farm gate and the FU was 1 kg carcass weight (CW). Five Danish  
61 and four Swedish beef farming scenarios were developed, which were categorised depending on  
62 intensive or extensive production, and dairy or beef breeds of cattle. For feed production (pasture,  
63 silage and concentrates), carbon sequestration was considered based on IPCC guidelines and

64 published literature. Grass-clover swards were included as part of an arable rotation where the swards  
65 remained for two to three years in a five-year rotation. GHG emissions were estimated using a  
66 combination of IPCC values and published data for Nordic conditions, the latter to estimate methane  
67 (CH<sub>4</sub>) emissions from enteric fermentation and manure management. The resultant CF ranged from  
68 8.9 to 17.0 kg CO<sub>2</sub>-eq per kg CW for the dairy-bull fattening systems, while the CF for cow-calf systems  
69 ranged from 23.1 to 29.7 kg CO<sub>2</sub>-eq per kg CW. Carbon sequestration resulted in GWP mitigation  
70 across all scenarios; amongst them, CO<sub>2</sub> reduction was largest in the grass-based systems, although  
71 these systems still generated the highest CF values despite elevated carbon uptake.

72 Wiedemann *et al.* (2015b) utilised LCA to examine the environmental impact of Australian beef  
73 and lamb being exported to the US. The system boundary was from cradle (in Australia) to the  
74 distribution warehouse in the US, and the FU was 1 kg retail ready meat. For beef systems, the study  
75 considered beef cattle bred in rangelands and finished on pasture, and dairy steers finished on grain  
76 feedlots for either 115 days or, for specialised breeds such as Wagyu, 330 days. Farm level data were  
77 obtained from governmental surveys and published case studies. Regionally tailored herd models  
78 were used to calculate feed intake and for predicting GHG emissions. Data on slaughtering and  
79 processing (such as cutting and chilling) were derived from an industry survey of meat processing  
80 plants in Australia. GWP ranged from 23.4 to 27.2 kg CO<sub>2</sub>-eq per kg beef, with the grass-finished cattle  
81 performing least favourably. Across the three scenarios, the farming phase generated the highest  
82 GWP (93%), meat processing accounted for 4%, transportation 3%, while the warehousing had  
83 negligible impacts. However, the authors also considered differences in human edible protein  
84 conversion efficiency. Under this FU, pasture-based beef production performed considerably better  
85 than grain-fed beef by converting more non-human-edible protein into human-edible protein.

86 In an effort to capture temporal variations in on-farm GHG emissions, Hyland *et al.* (2016)  
87 assessed the CF of 15 livestock enterprises over two time periods three years apart (2009/10) and  
88 2012/13). In addition to calculating farm-level emissions intensities, the authors also used a range of  
89 sensitivity analyses to investigate potential mitigation strategies. The system boundary of the study  
90 was set as cradle to farm gate, and the FU as 1 kg liveweight (LW). Across the 15 livestock enterprises  
91 examined, five specialised in lamb, four specialised in beef, while six were mixed beef and sheep farms.  
92 Emissions intensities were calculated according to IPCC (2006) Tier 1 and 2 guidelines. In tackling the  
93 issue of mixed farming allocation, where possible Hyland *et al.* (2016) used system expansion;  
94 however, in certain cases this was not possible due to a lack of differentiation and economic allocation  
95 was used instead. Between two data periods, lamb emissions were found to increase by 12%, while  
96 beef emissions decreased by 12%. However, these differences were not found to be statistically  
97 significant. Unsurprisingly, CH<sub>4</sub> emissions primarily resulting from enteric fermentation were the

98 greatest GHG burdens across all enterprises. Regarding the scenario analysis aimed at reducing on-  
99 farm emissions, the authors suggested that the primary focus for farmers should be on improved  
100 resource use efficiency. The inclusion of legumes such as red (*Trifolium pratense*) and white (*Trifolium*  
101 *repens*) clover on suitable soils was also highlighted as an important technique to reduce fertiliser  
102 requirements.

103 Examining the impacts of Canadian grazing management strategies on GHG intensities from  
104 beef herds, Alemu *et al.* (2017) modelled a typical herd structure of 120 cows, four bulls and their  
105 progeny over an eight-year period. A range of different grazing strategies were considered: light  
106 continuous grazing for all cattle; heavy continuous grazing for all cattle; light continuous grazing for  
107 cow-calf pairs and moderate rotational grazing for backgrounded cattle; and heavy continuous grazing  
108 for cow-calf pairs and moderate rotational grazing for backgrounded cattle. The system boundary was  
109 set as cradle-to-farmgate, and results based on two FUs (LW and CW) were reported side-by-side. The  
110 authors used Holos (a Canadian whole-farm model) to estimate farm-level emissions, and soil carbon  
111 changes were considered using the Introductory Carbon Balance Model, while farm management data  
112 were sourced from previous management studies. Emissions intensities were found to have narrow  
113 ranges (14.5–16.0 kg CO<sub>2</sub>-eq/kg LW and 24.1–26.6 kg CO<sub>2</sub>-eq/kg CW) across the grazing scenarios;  
114 however, GHG emissions tended to decrease as stocking density increased. Inclusion of soil as a carbon  
115 sink reduced impacts by up to 25%. The authors highlight the complexities in crediting a grassland  
116 system as a carbon sink due to the extremely dynamic nature of carbon flows.

117 Berton *et al.* (2017) applied the LCA method to examine the environmental footprint of the  
118 integrated French-Italian beef production system. The system boundary was set as cradle to farmgate;  
119 however, unlike many other studies, this boundary accounted for a cow-calf operation in one country,  
120 France, with animals ready for fattening transported to another country, Italy. All inputs and outputs  
121 (including transportation) associated with each stage were accounted for, and impacts were scaled to  
122 a FU of 1 kg LW (described as bodyweight). The authors considered a range of impact categories made  
123 up of GWP, acidification potential (AP), eutrophication potential (EP), cumulative energy demand  
124 (CED) and land use (LU) reported as land occupation. Regarding allocation of burdens to coproducts  
125 of the cow-calf operation, a mass approach was adopted along with a sensitivity analysis to consider  
126 the effect of this assumption. French farms (40) and Italian farms (14) were modelled based on best  
127 available data. The authors found that the burdens arising directly from the farms were greater than  
128 upstream processes in general; the only exception to this finding was CED, where energy demand was  
129 higher for off-farm processes for production of feed and agrochemicals. In terms of total impacts, the  
130 authors highlighted positive correlations between direct environmental burdens (GWP, AP and EP)

131 and resource requirements (CED and LU) and pointed out that agricultural policy design needs to  
132 account for multiple indicators rather than focusing on one.

133 In Italy once again, Buratti *et al.* (2017) compared the CF of conventional and organic beef  
134 production systems. Data were collected from two case study farms in the Umbria region of Italy, both  
135 of which operated as cow-calf systems rather than specialist fattening operations. The system  
136 boundary was from cradle to farmgate, and the FU was 1 kg LW of heifers and bullocks ready for  
137 slaughtering. Feed production primarily occurred on each of the farms, and burdens arising from  
138 fodder were modelled based on production data provided by the farms. The few imported products  
139 were treated as background processes and sourced from *ecoinvent V3*. Fertilising strategies differed  
140 between the enterprises. For example, the “organic” system solely used livestock manure to fertilise  
141 feed crops, while the “conventional” system used mineral N in addition to manure. Both systems  
142 transported excess manure to nearby but external cropland. GHG emissions were estimated using  
143 IPCC (2006) Tier 2 guidelines for all foreground sources, and, regarding enteric fermentation, the  
144 authors estimated  $Y_m$  values (CH<sub>4</sub> conversion factors) according to the digestible energy of the feed.  
145 The authors reported that lower GHG emissions were generated when producing organic feed due  
146 largely to lower mineral N requirements; however, interestingly, this did not translate to total CF  
147 rankings. The conventional system had a lower CF than the organic system, primarily driven by the  
148 shorter finishing times required.

149 de Figueiredo *et al.* (2017) examined the GHG balance and CF of three pasture-based beef  
150 finishing systems in Brazil. Three pasture systems all consisting of *Brachiaria spp.* were defined as: a  
151 degraded pasture receiving no external inputs; a managed pasture receiving annual fertiliser with  
152 animals receiving strategic supplementation consisting of maize (*Zea mays*) bran (82%), milled  
153 soybean (*Glycine max*) (14%), urea (3%) and mineral salt (1%) for a six-month period during dry season  
154 at a rate of 4 g/kg bodyweight; and a crop-livestock-forest integration system, a more complex system  
155 involving afforestation and rotational crop production and the same supplementation described  
156 under managed pasture. Both values were calculated using IPCC (2006) guidelines; the GHG balance  
157 was reported in terms of land area (1 ha), whereas the CF was reported as 1 kg LW leaving the  
158 farmgate. On an area basis, degraded pasture was found to have the lowest GHG balance, due to  
159 considerably lower stocking rates and no fertiliser requirement. Nevertheless, this finding was  
160 reversed in terms of 1 kg LW and degraded pasture was found to be the least efficient system due to  
161 low animal productivity. Between the two improved pastures, managed pasture was found to have  
162 considerably lower emissions (in terms of LW) than crop-livestock-forest, with livestock productivity  
163 again being a key factor. The crop-livestock-forest system brings its own merits in terms of other  
164 impact categories not assessed, such as improved biodiversity and utilising land to produce timber

165 and crops as coproducts from the system. Overall, the authors conclude that land designated as  
166 degraded pasture should be improved wherever feasible. This study further questions the use of area  
167 as a FU for system-level environmental evaluation.

168 Florindo *et al.* (2017) used LCA methodology in combination with life cycle costing (LCC) to  
169 evaluate both the CF and economic performance of beef cattle in the Brazilian Midwest. The authors  
170 point out that LCA studies often recommend mitigation strategies to reduce environmental footprints  
171 while failing to account for economic viability, a trade-off they explicitly consider. Primary data for the  
172 study, including machinery costs and management activity, were collected directly from a beef farm  
173 comprising 1,350 ha of grassland. The farm maintains 1,830 animals consisting of breeding stock as  
174 well as growing and finishing cattle. As part of the diversification strategy, the farm is split into four  
175 different production systems, differentiated by feeding regimes, stocking densities and slaughter  
176 weights. Feeding regimes were determined as with or without strategic supplementation which varied  
177 depending on the life stage of the cattle (e.g. creep feed). The protein mineral supplement included  
178 cornmeal (36%), soybean meal (12%) and urea (11%). Creep feed was made up of 30% cornmeal and  
179 51% soybean meal, while a 14% crude protein ration provided based on LW consisted of 72% cornmeal  
180 and 18% soybean meal. GHG emissions were calculated according to IPCC (2006) Tier 2 guidelines.  
181 Regarding LCC, the production system with the longest duration in terms of grazing was found to be  
182 the most cost-effective feed source, due to reduced supplementary feeding requirements. However,  
183 despite this positive aspect, it also resulted in the largest total financial cost due to lower stocking  
184 densities, and therefore, greater capital expenditure for land-use. The same finding was true for GHG  
185 emissions; higher stocking rates and lower grazing durations generated lower CFs, despite the  
186 subsequent lower finishing weights. This demonstrates the benefits of strategic supplementation,  
187 particularly in geographical regions affected by severe weather (extremely dry seasons in this  
188 instance). Care must be taken, however, at interpretation of these results, as strategic  
189 supplementation could potentially increase the level of food-feed competition (Wilkinson and Lee,  
190 2018).

191 Utilising interdisciplinary skills and expertise, Hesse *et al.* (2017) examined how Swedish beef  
192 and milk production systems could be environmentally and economically optimised under a range of  
193 different scenarios. Input was provided by experts in economics, LCA and supply chain management.  
194 The focus of this study was the environmental comparison of the reference situation (business-as-  
195 usual) with three hypothetical yet realistic scenarios. The three expertly-designed scenarios were  
196 based around Swedish environmental objectives and set as follows: an “ecosystem” scenario aimed  
197 at reducing impacts on biodiversity; a “nutrient” scenario which focused on optimising plant nutrient  
198 use and supply; and a “climate” scenario primarily concerned with reducing anthropogenic GHG

199 impacts. The overarching goal of each alternative scenario was to maintain or improve production  
200 efficiency, while simultaneously mitigating environmental impacts. Once the study panel had agreed  
201 upon the alternative systems, LCA models were constructed using a combination of literature and  
202 expert opinion. In most instances, the improved systems demonstrated reduced negative impacts.  
203 However, there were notable trade-offs; for example, the ecosystem scenario required more land  
204 being used as grassland to improve biodiversity, which in turn caused negative impacts on  
205 eutrophication (freshwater and marine) and cumulative energy demand across both beef and dairy  
206 systems. Despite this, the authors concluded that a common denominator in improving these livestock  
207 systems was more efficient use of resources such as energy and feed.

208 Tichenor *et al.* (2017) analysed differences in environmental performances between intensively  
209 managed grass-fed beef production and confinement dairy beef production systems in the Northeast  
210 US. The system boundary was from cradle to farmgate and the authors considered hot carcass weight  
211 as the FU to maximise comparative potential with previous North American studies. The impact  
212 categories considered were GWP, AP, EP, fossil fuel demand, water depletion and LU. For dairy beef,  
213 the authors adopted biophysical allocation at the ratio of 9.4%/0.4%/90.2% for beef/veal/milk,  
214 respectively. They also considered economic allocation in a sensitivity analysis, at the rate of  
215 7.8%/0.9%/91.3%. Across GWP, EP, AP and LU, grass fed was found to have higher burdens than dairy  
216 beef. On the other hand, dairy beef required more fossil fuel and water than grass fed. The authors  
217 also considered impacts on a per ha basis, which resulted in lower AP and EP burdens for grass fed. A  
218 sensitivity analysis to account for carbon sinks in grassland was also considered. While this inclusion  
219 substantially reduced the GWP of grass fed, it was not enough to offset the benefits of productivity  
220 from DB. The authors echoed the argument of Berton *et al.* (2017) that future research should  
221 consider multifaceted aspects of grass-fed systems that are socially important.

222 Wiedemann *et al.* (2017) examined resource use and GHG emissions associated with seven  
223 Australian feedlot beef systems. The authors adopted a gate to gate approach, with a primary focus  
224 on impacts arising from the grain-finishing stage. The FU for comparisons between the finishing stages  
225 was 1 kg LWG, while values for the entire system (including cow-calf enterprise) were reported as 1  
226 kg LW. Three classes of cattle were considered: short-fed (55–80 days) for domestic market; mid-fed  
227 (108–164 days) and long-fed (> 300 days) for alternative export markets. Similar to Hyland *et al.*  
228 (2016), Wiedemann *et al.* (2017) found that CH<sub>4</sub> emissions aggregated across enteric fermentation and  
229 manure management were the most significant contributors to emissions intensities. Across the three  
230 management strategies, long-fed generated more GHG emissions than mid-fed which in turn  
231 generated more emissions than short-fed, due largely to the length of production cycles. The same  
232 rankings were observed for fossil energy demand. However, the opposite rankings were noted for



233 water consumption, an impact category with high importance in the arid regions of Australia. While  
234 the differences were not significant between short- and mid-fed, long-fed cattle had considerably  
235 lower freshwater usage due to reduced irrigated water usage. In terms of cradle to gate analysis, the  
236 finishing systems were found to contribute 26–44% of the total emissions intensity, with higher  
237 maximum impacts (up to 72%) recorded for total energy demand. The authors note that switches from  
238 pasture based to grain-based systems have reduced Australia’s national emissions intensity from beef  
239 cattle, but these switches have been met with a trade-off of increased national energy demand. This  
240 signifies the complexities of drawing conclusions across multiple impact categories.

241 Willers *et al.* (2017) sought to identify environmental hotspots in semi-intensive beef  
242 production systems in Brazil’s Northeast. The study accounted for two farms: the cow-calf operation  
243 and a separate but nearby finishing system. Similar to most beef LCA studies, the authors adopted a  
244 cradle to farmgate system boundary and a FU of 1 kg LW leaving the finishing farm. Primary data were  
245 gathered from the managers of both farms, while background processes were sourced from *ecoinvent*  
246 V2. The authors considered five impact categories: GWP (reported as climate change); AP (reported  
247 as terrestrial acidification); EP (reported as freshwater eutrophication), LU (reported as agricultural  
248 land occupation) and fossil fuel depletion. Following Berton *et al.* (2017), Willers *et al.* (2017) used  
249 mass allocation to disentangle burdens arising from coproducts at the cow-calf stage. Regarding the  
250 identification of hotspots, the authors diverted from conventional approaches and considered pasture  
251 processes as separate entities to their modelled livestock. This resulted in an unusual attribution of  
252 the overall burdens, whereby “grassland production” has higher effects on all impact categories than  
253 “livestock burdens”, making inter-study comparison of the results (de Vries *et al.*, 2015) rather  
254 difficult.

255 Bragaglio *et al.* (2018) analysed the environmental footprints of a range of different beef  
256 production systems in Italy utilising data collected from 25 farms. The systems studied were:  
257 specialised extensive; high grain fattening; intensive cow-calf constantly kept in confinement; and  
258 native breed (Podolian) maintained on pasture and finished in housing. The authors considered GWP,  
259 water depletion, LU, AP and EP within a system boundary set as cradle to farmgate and a FU of 1 kg  
260 LW. In terms of GWP, the intensive systems (high grain fattening and cow-calf confinement) were  
261 found to have lower impacts due largely to improved growth rates. However, the authors found that  
262 the systems with durations of pasture grazing (specialised extensive and Podolian) had lower AP than  
263 cow-calf confinement. There was no significant difference noted for water depletion, while high grain  
264 fattening and Podolian demonstrated the lowest burdens in terms of water quality (EP). Significantly  
265 higher LU was required for specialised extensive and Podolian; however, the authors also  
266 acknowledged that competition with human edible feed was lower for the grazing systems,

267 particularly Podolian. A theme recurrent throughout grazing livestock LCA studies, namely the  
268 omission of ecosystem services and other societal benefits (e.g. improved animal welfare and meat  
269 quality) provided by grassland systems, is also highlighted by the authors. Bragaglio *et al.* (2018)  
270 conclude by acknowledging the importance of future LCA studies addressing these aspects of livestock  
271 systems that are more difficult to quantify.

272 Analytical approaches adopted by the above 14 studies are summarised in **Table 1**, with a  
273 particular attention to the treatment of major sources of uncertainty inherent in beef production  
274 systems. Overall, it demonstrates a considerable gap in knowledge regarding uncertainty within the  
275 existing literature. For example, none of the 14 studies used individual livestock data, meaning that  
276 intra-herd distributions of animal properties and performances could not be considered. Eight out of  
277 14 papers did use farm-level aggregated data; nonetheless, only one of them included primary  
278 information on forage quality, a parameter widely known to be affected by farm management and, in  
279 turn, contribute to the uncertainty surrounding CH<sub>4</sub> emissions through enteric fermentation. Only  
280 three studies conducted Monte Carlo analysis, reiterating the lack of attention bestowed upon  
281 uncertainty on the whole amongst LCA studies (Imbeault-Tétreault *et al.*, 2013).

282 Furthermore, none of the above studies adopted site-specific EFs for calculating GHG emissions.  
283 This finding is perhaps unsurprising given the considerable effort required to measure GHG fluxes at  
284 the farm scale; conducting field trials specifically for an LCA study is unlikely to be financially justified.  
285 Should an opportunity exist, however, introduction of site-specific EFs may provide an effective and  
286 computationally straightforward means to improve accuracy of CF estimation because, as already  
287 discussed in Section 1, uncertainty associated with these parameters is one of the important practical  
288 barriers to draw useful policy implications for farm management. An important operational question,  
289 then, is as follows: under practical constraints concerning staff and budgetary availability, which EFs  
290 should we prioritise to measure on the farm? The next section outlines the procedure of a virtual  
291 experiment designed to solve this problem. To the best of our knowledge, such an attempt has not  
292 previously been made in LCA literature.

### 293 **3. Case study: materials and methods**

294 The case study was carried out using data from the permanent pasture-based beef enterprise (farmlet)  
295 of the North Wyke Farm Platform National Capability (Orr *et al.*, 2016), an instrumented farm-scale  
296 grazing trial located in Devon, UK (50°46'10"N, 3°54'05"W). Under the attributional approach, the  
297 system boundary was defined as “cradle-to-gate” or from the production of raw materials to the  
298 departure of live animals for slaughter, and encompassed both cow-calf and finishing operations,

299 which are adjacent to each other but do not share pasture or other resources. FU was set as 1 kg LW  
300 of prime beef (calves) departing the farm. Environmental burdens attributable to the sale of culled  
301 cows were portioned out from the system-wide CF using economic allocation.

302 Farm management practices and the data collection strategy at the study site are detailed  
303 elsewhere (Takahashi *et al.*, 2018). Briefly, 30 Charolais × Hereford-Friesian calves and their dams  
304 constitute each year's herd, with the number of cows in the LCA model adjusted to account for extra  
305 heifers required to replace culled cows. As with the majority of beef farms in South West England,  
306 cattle graze in summer and are housed in winter, with both LWG of calves and the forage quality of  
307 pasture and silage evaluated at regular (2-4 week) intervals. All physical inputs into the system were  
308 appropriately recorded.

309 The present study utilised data associated with a generation of calves born in spring 2015 and  
310 slaughtered in winter 2016. On-farm GHG emissions from both livestock and pastures were calculated  
311 using the IPCC (2006) Tier 2 approach. Based on our earlier finding that ignoring the inter-animal  
312 difference in growth efficiency leads to a biased estimate of the farm-scale CF (McAuliffe *et al.*, 2018b),  
313 emissions for finishing animals were initially calculated for each individual animal separately and  
314 subsequently pooled together to create a whole-farm inventory. The resultant CF (kg CO<sub>2</sub>-eq/kg LW)  
315 was expressed as a 95% confidence interval, which was derived by a Monte Carlo analysis with 1000  
316 iterations. For each iteration, values for EFs were drawn from the distributions recommended by IPCC  
317 (2006); further details of this process are described in McAuliffe *et al.* (2018b).

318 Following the baseline CF estimation under which all EFs were assumed to be uncertain and  
319 follow predetermined probability distributions, eight groups of EFs (Table 2) were individually  
320 assumed to be certain at three levels defined by IPCC (2006): point estimate as well as lower and upper  
321 limits of the 95% confidence interval. By design, the CF derived under this setting was expected to  
322 have a narrower range than the baseline result, as one source of uncertainty had been eliminated  
323 from the model. These outputs represent hypothetical CFs when a particular group of site-specific EFs  
324 are perfectly quantified on the farm and therefore indicate the information value of pinpointing the  
325 corresponding EFs. Alternatively, as the true EFs are likely to lie somewhere between lower and upper  
326 limits of the IPCC 95% range, the difference in CF distributions derived under these two values can be  
327 seen as the level of uncertainty associated with the relevant EFs; a larger difference here suggests a  
328 higher priority for field measurements to obtain locally more accurate EF values. Overall, 25 CFs were  
329 derived, a baseline and 24 variants with distinct EFs (8 groups x 3 values).

330 Throughout the analysis, emissions pertaining to background processes were sourced from  
331 *Agri-footprint* v3 (Durlinger *et al.*, 2017) and *ecoinvent* v3 (Wernet *et al.*, 2016) databases. All CFs were

332 calculated under the IPCC (2013) 100-year average impact assessment method on *SimaPro* v8.2.3.

#### 333 4. Case study: results and discussion

334 The mean baseline CF was estimated to be 22.8 kg CO<sub>2</sub>-eq/kg LW, with a confidence interval of 21.3–  
335 25.2 kg CO<sub>2</sub>-eq/kg LW (**Figure 1**). These results are at the higher end of values and ranges reported by  
336 previous studies undertaken in comparative environments (reviewed in Section 1); while the reason  
337 behind this has not been completely elucidated, it is thought to be due to a combination of the low  
338 stocking rate and the high replacement of the breeding herd to ensure safe delivery of calves for  
339 fattening experiments. Assuming subsets of EFs to take point estimate values without uncertainty did  
340 not change the CF distribution to any noticeable extent, although knowledge in CH<sub>4</sub> emissions from  
341 manure management and N<sub>2</sub>O emissions from inorganic fertiliser application reduced the width of the  
342 confidence interval by 0.4 and 0.3 kg CO<sub>2</sub>-eq/kg LW, respectively (**Figure 1**). Interestingly, certainty  
343 regarding CH<sub>4</sub> emissions from enteric fermentation, widely considered to be the single largest source  
344 of ruminant-originated GHG emissions, contributed very little to the certainty regarding the resultant  
345 CF, likely because of the symmetric (triangular), rather than asymmetric (lognormal), nature of the  
346 probability distribution assumed under the baseline model (McAuliffe *et al.*, 2018b).

347 Comparisons of CFs derived under lower and upper limits of EF values revealed, however, that  
348 on-farm measurement of enteric CH<sub>4</sub> may still be one of the most effective approaches to reduce  
349 uncertainty (**Figure 2**). The difference in mean CFs between two scenarios was estimated to be 3.6 kg  
350 CO<sub>2</sub>-eq/kg LW, the second largest after N<sub>2</sub>O emissions from inorganic fertiliser application (4.4 kg CO<sub>2</sub>-  
351 eq/kg LW) and closely followed by N<sub>2</sub>O emissions from excreta deposited during grazing (3.5 kg CO<sub>2</sub>-  
352 eq/kg LW). The high information value of the latter two EFs is attributable to the high degree of  
353 uncertainty identified for these parameters by IPCC (2006), as discussed in Section 1 and quantitatively  
354 supported by **Figure 1**. On the other hand, improved knowledge of enteric CH<sub>4</sub> EF reduced uncertainty  
355 surrounding the overall CF because of its large contribution to the total environmental burdens, even  
356 though its own variation is confined to a relatively small range. The remaining five groups of EFs were  
357 shown to have considerably less impact on the CF.

#### 358 5. Conclusion

359 The above analysis revealed that uncertainty surrounding climate impacts of ruminant systems can  
360 potentially be reduced through on-farm measurements of GHG fluxes, but not all measurements carry  
361 the same degree of information value. In temperate grassland regions, the priority for measurements  
362 should be given to N<sub>2</sub>O from inorganic fertilisers applied and manure deposited to the soil, as well as

363 CH<sub>4</sub> from enteric fermentation. It is acknowledged that, strictly speaking, these site-specific EFs will  
364 still be accompanied by their own uncertainty, which stems from intra-farm variability in soil, weather,  
365 plant genetics, animal genetics, rumen microbial ecology and other confounding factors.  
366 Nevertheless, these local sources of uncertainty are likely to be considerably smaller than uncertainty  
367 about the farm location that needs to be embedded into generic EFs. In turn, CF analysis carried out  
368 under reduced uncertainty will likely offer more policy implications that are directly applicable to local  
369 production environments.

## 370 6. Acknowledgements

371 The case study component of this chapter was supported by the Biotechnology and Biological Sciences  
372 Research Council (BBS/E/C/00010320 and BBS/E/C/000J0100).

## 373 7. Where to look for further information

- 374 • The foundation theory behind the above discussion is summarised in: Heijungs R and Suh S  
375 (2002) *The Computational Structure of Life Cycle Assessment* ([https://doi.org/10.1007/978-](https://doi.org/10.1007/978-94-015-9900-9)  
376 [94-015-9900-9](https://doi.org/10.1007/978-94-015-9900-9)), Kluwer Academic Publishers.
- 377 • Oxford University Research Archive (<https://doi.org/10.5287/bodleian:0z9MYbMyZ>) stores a  
378 streamlined global database of CFs associated with food and beverage production. Also see  
379 [Poore and Nemecek \(2018\)](#).
- 380 • Global Farm Platform network (<http://www.globalfarmplatform.org>) is a worldwide initiative  
381 to compare sustainability of livestock production systems across different agroecological  
382 zones. Also see [Eisler et al. \(2014\)](#).
- 383 • North Wyke Farm Platform data portal ([https://www.rothamsted.ac.uk/north-wyke-farm-](https://www.rothamsted.ac.uk/north-wyke-farm-platform)  
384 [platform](https://www.rothamsted.ac.uk/north-wyke-farm-platform)) provides a wealth of raw data collected from farm-scale grazing trials, including  
385 those used in the present case study. Also see [Orr et al. \(2016\)](#) and [Takahashi et al. \(2018\)](#).

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**Table 1. Sources of uncertainty identified in recent LCA studies of beef production systems**

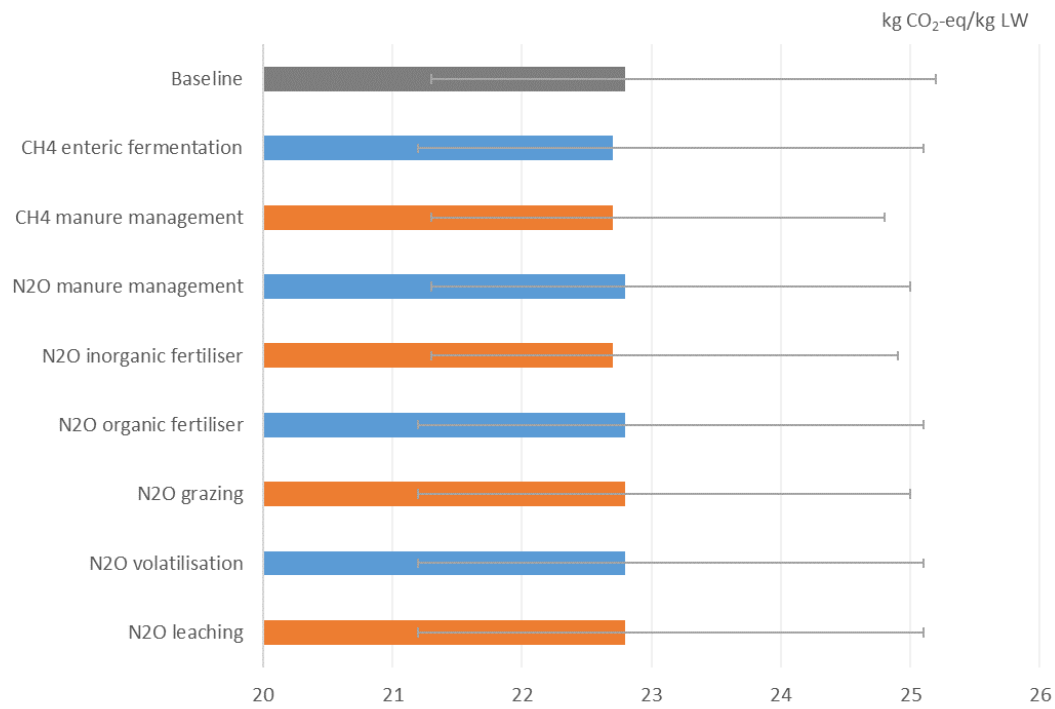
Study	Animal growth over time	Animal categories	Foreground data source	Feed quality	Emission factors for carbon footprints	Monte Carlo analysis
Dick <i>et al.</i> (2015)	Split between first year and subsequent years	9	National statistics	Literature	IPCC Tier 2	✘
Mogensen <i>et al.</i> (2015)	Fixed average daily gain (ADG) per system	6	Literature	Literature	Danish specific & IPCC Tier 2	✘
Wiedemann <i>et al.</i> (2015)	Fixed ADG	4	Farm	National statistics	Australian specific	✓
Hyland <i>et al.</i> (2016)	Monthly growth rates for growing stock	Unspecified	Farm	National statistics	UK specific & IPCC Tiers 1/2	✘
Alemu <i>et al.</i> (2017)	Fixed ADG per animal category	7	Literature and experimental data	Measured data	Canadian specific & IPCC Tier 2	✘
Berton <i>et al.</i> (2017)	Split into three growth stages on each farm	5	Farm	Literature	French specific and IPCC Tier 2	✘
Buratti <i>et al.</i> (2017)	Fixed ADG per animal category	7	Farm	Literature	IPCC Tier 2	✘
de Figueiredo <i>et al.</i> (2017)	Fixed ADG per system	Unspecified	Literature	Unspecified	Brazilian specific and IPCC Tier 1	✘
Florindo <i>et al.</i> (2017)	Varied ADG by age and scenario	4	Farm	Unspecified	IPCC Tier 2	✘
Hessle <i>et al.</i> (2017)	Unspecified	5	National statistics	Literature	IPCC (unspecified tier)	✘
Tichenor <i>et al.</i> (2017) <sup>d</sup>	Unspecified	6	Literature	Not applicable (only examines land use)	Not applicable (only examines land use)	✘
Wiedemann <i>et al.</i> (2017)	Varied ADG by farm and scenario	3	Farm	National statistics	Australian specific	✓
Willers <i>et al.</i> (2017)	Unspecified	4	Farm	Unspecified	IPCC (unspecified tier)	✓
Bragaglio <i>et al.</i> (2018)	Fixed ADG per system	3	Farm	Literature	IPCC Tier 2	✘



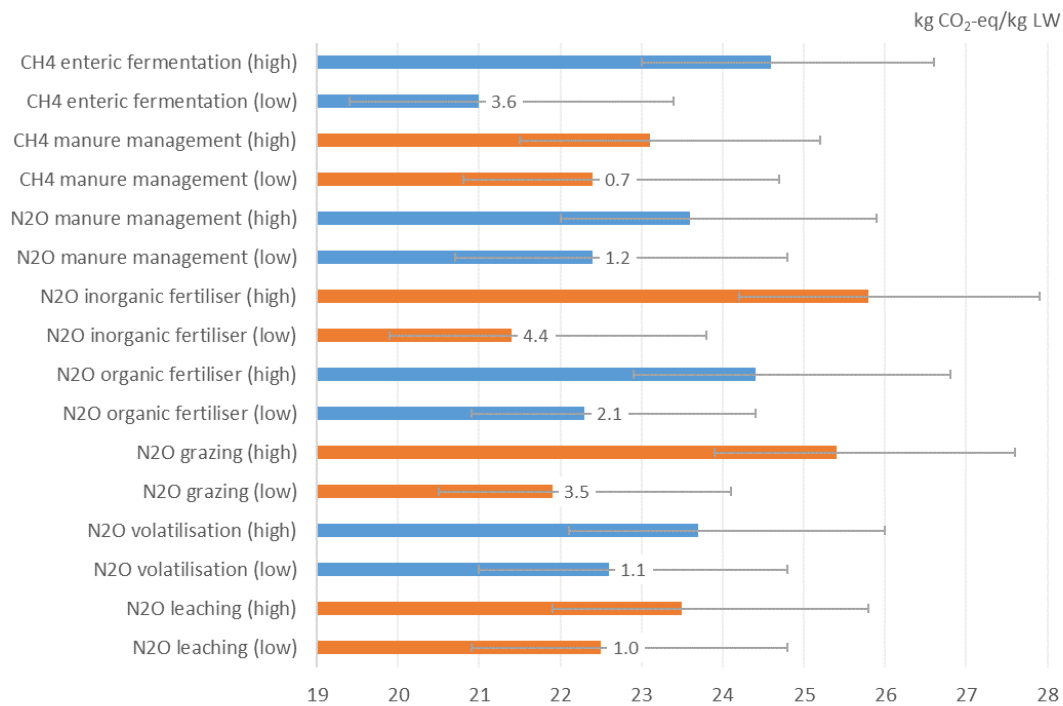
**Table 2. Emission factors considered in the case study**

	IPCC (2006) notation	IPCC (2006) reference
CH <sub>4</sub> enteric fermentation	$EF$	Equation 10.21
CH <sub>4</sub> manure management	$EF_{(T)}$	Equation 10.23
N <sub>2</sub> O manure management	$EF_3$	Table 10.21
N <sub>2</sub> O inorganic fertiliser (ammonium nitrate)	$EF_1$ (for $F_{SN}$ )	Table 11.1
N <sub>2</sub> O organic fertiliser (manure)	$EF_1$ (for $F_{ON}$ )	Table 11.1
N <sub>2</sub> O grazing	$EF_{3PRP}$	Table 11.1
N <sub>2</sub> O volatilisation	$EF_4$	Table 11.3
N <sub>2</sub> O leaching	$EF_5$	Table 11.3

Note: Emissions associated with other sources (including background processes) were assumed to be uncertain throughout the case study.



**Figure 1. Carbon footprints of beef production systems estimated with one group of emission factors fixed at the IPCC point estimate. Error bars show the 95% range derived from Monte Carlo simulation, where all but one group of emission factors were assumed to follow IPCC uncertainty distributions. The baseline result accounts for all sources of uncertainty.**



**Figure 2. Carbon footprints of beef production systems estimated with one group of emission factors fixed at lower and upper limits of the IPCC 95% confidence interval. The number under each group depicts the difference in means between the two scenarios, which represents the value of knowing that particular emission factor through site-specific field trials. Error bars show the 95% range derived from Monte Carlo simulation, where all but one group of emission factors were assumed to follow IPCC uncertainty distributions.**