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## Improvement of feed and nutrient efficiency in pig production through precision feeding

Charlotte Gaillard, Ludovic Brossard, Jean-Yves Dourmad

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1 **Review: Improvement of feed and nutrient efficiency in pig production through**  
2 **precision feeding**

3

4 Charlotte Gaillard\*, Ludovic Brossard, Jean-Yves Dourmad

5 PEGASE, INRAE, Institut Agro, 35590, Saint Gilles, France

6 \*Corresponding author: Charlotte Gaillard, PEGASE, INRAE, Institut Agro, 35590, Saint

7 Gilles, FranceSaint-Gilles, France

8 E-mail: [charlotte.gaillard@inrae.fr](mailto:charlotte.gaillard@inrae.fr)

9

10 **Highlights**

- 11 • Environment, animal and feed characteristics influence nutrient utilization in pigs.  
12 • Mathematical models can be used to estimate real-time daily nutrient requirements.  
13 • Thanks to technological advances, each pig can receive its daily nutrient requirements.  
14 • Precision feeding may also reduce feed cost and environment load.

15

16

17 **Abstract**

18

19 Nutrient requirement change over time and individual variability in pigs influences the  
20 efficiency of nutrient utilization. These variabilities should be considered to predict nutrient  
21 requirements more accurately. The goal of precision feeding is to develop systems able to  
22 estimate and deliver, at the right time, a ration with a quantity and composition adapted to the  
23 daily requirements of each animal. It would improve feed and nutrient efficiency, which is a  
24 major issue for the sustainability of all pig production systems. The objectives of this review

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Abbreviations: Feed conversion ratio (FCR), residual feed intake (RFI), metabolizable energy (ME), crude protein (CP), amino acid (AA), body weight (BW), phosphorus (P), nitrogen (N)

25 were: 1) to define feed efficiency and present the factors that affect it, as well as challenges to  
26 and strategies for improving it; 2) to define precision feeding and the sources of variability in  
27 nutrient requirements and show the need for new technology to obtain real-time data; and 3)  
28 to present current models and applications of precision feeding for fattening pigs and sows.  
29 Feed efficiency is expressed as the ratio of mean daily weight gain to mean daily feed  
30 consumption over a given period. In practice, the inverse of this ratio is generally used for  
31 breeding animals and represents the efficiency of converting feed into weight gain (feed  
32 conversion ratio, FCR). Several factors influence FCR, such as spillage, feed digestibility,  
33 composition of weight gain, feed intake and nutrient utilization. Selecting the appropriate  
34 form of feed and the appropriate nutrient density and supply, as well as reducing negative  
35 effects of environmental factors should improve FCR. New feeding technologies (e.g. sensors,  
36 feeders) allow group-housed animals to be fed based on their individual requirements, which  
37 improves group efficiency. Predictive models of nutrient requirements and excretion, such as  
38 InraPorc, have been developed and used to select the best feeding strategies. For growing  
39 pigs, precision feeding strategies are a promising solution to reduce nutrient excretion by  
40 adjusting the nutrient supply to each individual at different points in time. Recent simulations  
41 indicate that precision feeding might also be a relevant strategy for sows.

42

43 **Keywords:** farming conditions, feed conversion ratio, nutritional models, pig production,  
44 precision feeding

45

46

## 47 **Introduction**

48

49 Animal feed, human food, and bio-industries compete for crop resources, which places  
50 societal pressure on farming. Moreover, feed cost represents around two thirds of the  
51 production costs for fattening pigs (Pomar et al., 2009), and 15-17% of the production costs  
52 for sows and their litters until weaning (Solà-Oriol and Gasa, 2017). Consequently, nutrition  
53 is a major mechanism for improving the sustainability of pig production. Reducing the use of  
54 feed would reduce feed cost, and consequently nutrient excretion. It can also influence  
55 product quality: lean-to-fat ratio, fat quality, and the homogeneity of products. Currently,  
56 most of fattening pigs are group-housed and fed based on the average pig requirements of the  
57 room or the pen (Whittemore, 2006). Consequently, some pigs are overfed and others are  
58 underfed. Sows are usually fed two diets, one restrictively during gestation and the other  
59 nearly ad libitum during lactation (Solà-Oriol and Gasa, 2017); both based on an average  
60 sow's requirements. The goal of precision feeding is to develop systems that estimate and  
61 deliver, at the right time, a ration with a quantity and composition adapted to the requirements  
62 of each animal. The challenges in these systems reside in estimating individual requirements  
63 and distributing different diets to animals in the same group. These systems would improve  
64 feed and nutrient efficiency, major issues for the sustainability of all pig production systems  
65 (conventional and alternative). The objectives of this review were 1) to define feed efficiency  
66 and present the factors that influence it, as well as challenges to and strategies for improving  
67 it; 2) to define precision feeding and the sources of variability in nutrient requirements and  
68 show the need for new technology to obtain real-time data; and 3) to present the current  
69 models and applications of precision feeding for fattening pigs and sows.

70

### 71 **1. Feed and nutrient efficiency**

72

73 **1.1. Definition and measure of feed efficiency**

74 Feed efficiency (FE) is the ratio of mean daily weight gain to mean daily feed consumption  
75 over a given period (Gilbert, 2015). It can also be expressed as the ratio of growth to energy  
76 intake, which depends less on energy density in the diet. However, in practice, the inverse of  
77 FE is generally used for pigs, representing the efficiency of converting feed into weight gain  
78 (Gilbert, 2015). This feed conversion ratio (FCR) is similar to an economic measure of feed  
79 cost, whereas FE is similar to biological efficiency. The animals with the lowest FCR tend to  
80 be the most efficient (Bouquet, 2013).

81 Genetic improvement has reduced the FCR for most of the conventional pig breeds (Bouquet,  
82 2013). Nevertheless, producers still seek to minimize FCR. Selecting animals for increased  
83 growth may lead to increased ingestion, whereas selecting for decreased ingestion may lead to  
84 decreased growth. More recently, to avoid this selection difficulty, residual feed intake (RFI)  
85 was used as another measure of efficiency. The RFI is calculated as an animal's daily feed  
86 consumption minus the quantity of feed required to meet its theoretical energy requirements  
87 (Bouquet, 2013). Pigs with high RFI are less energy efficient because they produce more heat,  
88 mainly due to increased physical activity and basal metabolic rate (Barea, 2010). RFI has high  
89 variability, but genetic selection can decrease it. Since RFI is not correlated with growth but is  
90 positively correlated with feed intake and FCR, selecting animals for lower RFI should have  
91 no influence on growth but would reduce ingestion. However, recent studies at IFIP (French  
92 Pork and Pig Institute) showed that selecting for FCR results is still more economic than  
93 selecting for RFI (Bouquet, 2013).

94 For fattening pigs, FCR is usually calculated from 10 weeks of age (around 30 kg of body  
95 weight, BW) to slaughter (around 115 kg of BW). FCR can also be calculated from weaning  
96 to around 10 weeks of age, or from weaning to slaughter (Gilbert, 2015). FCR for

97 reproductive sows is more difficult to assess. It can be expressed, as for fattening pigs, as the  
98 amount of feed consumed over a given period divided by the BW gain of sows and piglets  
99 (total BW of piglets at weaning plus net increase in sow BW) over the same period. Another  
100 way to express FCR for sows would be as the amount of feed consumed per weaned piglet  
101 produced.

102 FCR can be expressed in different units; kg feed/kg gain is the most common unit, but MJ  
103 energy/kg gain is often used to consider variations in feed energy content. Cost of feed/kg  
104 gain is another way to express FCR that is more similar to the economic efficiency (Gilbert,  
105 2015). Ultimately, FCR is not determined by growth rate and feed intake, but by factors that  
106 influence them, such as genetics, feeding practices, environmental conditions and health  
107 status. For example, data obtained from a French pig farm survey (Table 1) indicate that FCR  
108 depends on the breeding system and production period: herd FCR is greatest for the breeder-  
109 sale-at-weaning system (5.45 kg) and lowest for the weaner-fattener system (2.64 kg/kg).  
110 FCR is greater for the fattening period than for the post-weaning period, and these values vary  
111 among farms.

112

## 113 **1.2. Factors that influence feed efficiency**

114 Three equations can be used to express the FCR of growing pigs in different ways to identify  
115 the factors that influence FCR and how they do so. FCR depends on feed intake, feed spillage  
116 and animal growth (Eq. 1). It also depends on feed digestibility, digestive efficiency of pigs,  
117 the relative importance of maintenance and growth (Eq. 2 and 3), and the tissue (i.e. lean-to-  
118 fat ratio, Eq. 2) or chemical (Eq. 3) composition of weight gain.

119 [Eq. 1]  $FCR = (\text{feed intake} + \text{spillage}) / \text{pig growth}$

120 [Eq. 2]  $FCR = (\text{indigestible} + \text{maintenance} + \text{growth}) / (\text{lean} + \text{fat} + \text{bone} + \text{skins} + \text{organs})$

121 [Eq. 3]  $FCR = (\text{indigestible} + \text{maintenance} + \text{growth}) / (\text{protein} + \text{water} + \text{lipids} + \text{minerals})$

122           **Spillage.** In Eq. 1, increased spillage increases the FCR. Spillage can be reduced by  
123 selecting feeder type. In the study of Pierozan et al. (2016), a linear dump feeder had less feed  
124 waste and lower FCR than other types of feeders, such as conical semiautomatic feeders (2.41  
125 and 2.44, respectively,  $P = 0.04$ ). Comparisons between feeder types are limited due to their  
126 diversity and the difficulty in developing trials to test the effect of feeder type on FCR, as it  
127 requires modifying the feeders (Pierozan et al., 2016). Feeder characteristics (individual  
128 shoulder protection, number of places per pig, and water supply) and localization in the pen  
129 are essential when selecting a feeder to reduce feed waste (Averós et al., 2012). Averós et al.  
130 (2012) reported that feeders with shoulder protection resulted in lower FCR than unprotected  
131 feeders due to a reduction of pig aggressions at the feeder. Feed restriction may also help  
132 reduce feed waste and improve FE (Patience et al., 2015). Another way to reduce spillage is to  
133 feed pellets instead of mash diets. Indeed, the review of Vukmirović et al. (2017) reported a  
134 general agreement on the fact that feeding pellets to pigs improved FCR, by improving DM  
135 digestibility and reducing feed wastage, compared to feeding mash diets. However, size  
136 particle is reduced during pelleting process, which can have negative effects on the  
137 gastrointestinal tract health.

138           **Feed intake, growth and maintenance.** Genotype determines growth rate because  
139 feed intake and growth rate differ among breeds. For example, Piétrain males have lower feed  
140 intake, growth rate, and fat deposition than Large White, Landrace and Duroc males (Edwards  
141 et al., 2006)(Fig. 1). FCR is a function of BW, and as pigs grow toward market weight, they  
142 become less efficient at converting feed into BW gain (Patience et al., 2015). This increase in  
143 FCR results from the increase in maintenance requirements, which depend on BW, and the  
144 decrease in the muscle lean-to-fat ratio. There is an increase in ingestion far above the protein  
145 deposition capacity; therefore, nutrients are deposited as fat, which decreases feed efficiency.  
146 However, this increase in FCR with BW varies with sex. For entire males, FCR increases

147 slightly as BW increases, whereas for female and castrated pigs, FCR increases more rapidly  
148 with BW due to differences in feed intake and growth rate between sexes (Noblet et al., 1994)  
149 (Fig. 1).

150 During lactation, parity affects sow's feed intake. Sows of parity 1 or 2 consume about 15%  
151 less feed than older sows (Koketsu et al., 1996). This gradual increase in feed intake with  
152 advancing parity is consistent with the increase in maintenance energy requirements  
153 associated with the age-related increase in BW (O'Grady et al., 1985). Sows' mean daily feed  
154 intake also increases as litter size increases from small litters of 3-6 piglets up to 11 piglets,  
155 whereas it remains relatively constant for more than 11 piglets. Sows with small litters (< 7  
156 piglets) have a mean daily feed intake of 4 kg lower than that of sows with larger litters  
157 (Koketsu et al., 1996). This increase in feed intake can be related to the increase in milk  
158 production as litter size increases (Auldist et al., 1998; Ngo et al., 2012).

159 Environmental factors such as space allowance, group size, number of feeders, flooring  
160 conditions, enrichment, temperature, ventilation rate, relative humidity, pathogens and  
161 stressors influence FCR (Averós et al., 2012; Averós et al. 2010) due to variation in feed  
162 intake.

163 **Digestibility of feed and digestive efficiency of pigs.** In Eq. 2 and 3, FCR increases  
164 as the indigestible portion of the feed increases. The digestibility of feed depends greatly on  
165 its composition (Ponter et al., 2004) and on animal digestive capacity, which has been shown  
166 to vary among pigs in interaction with feed composition (Noblet et al., 2013). For example, a  
167 greater proportion of fiber in the diet is a challenge for digestion and results in high variability  
168 in pig performance. Noblet et al. (2013) found that the digestibility of energy is influenced by  
169 sire, which suggests that digestibility depends on heritable genetic variability. Kyriazakis  
170 (2011) also reports that selecting pigs for digestive efficiency would improve nutrient  
171 efficiency. Further research is needed to understand underlying mechanisms.

172           **Composition of weight gain (lean-to-fat ratio).** Eq. 2 and 3 indicate that the  
173 chemical and tissue compositions of BW gain influence FCR. The chemical composition of  
174 empty BW gain in minerals and protein (around 3% and 16-17% of empty BW, respectively)  
175 (Table 2) is similar among breeds and sexes, except for certain traditional breeds whose  
176 protein content is lower (Noblet et al., 1994). Conversely, the percentages of water and lipids  
177 are largely related to genotype and sex, which influence the energy content of gain. Adipose  
178 tissue yields around four times as much energy as muscle tissue (protein and lipid deposition  
179 cost 9.1 and 29.2 MJ of ME per kg, respectively) because it contains less water. This results in  
180 large differences in the energy content of BW gain among genotypes (e.g. 11.2 MJ/kg in  
181 Piétrain males vs. 21.1 MJ/kg in Meishan castrates) and between sexes (12.3, 13.8, and 15.6  
182 MJ/kg for Large-White males, females, and castrated males, respectively) (Noblet et al.,  
183 1994). Over the past 40 years, most genetic improvement in FCR has been obtained by  
184 reducing the proportion of lipids in BW gain. Selection for fast growing and lean animals has  
185 increased the potential for protein deposition and reduced the amount of energy required to  
186 achieve this potential, usually with little influence on feed intake (Gilbert, 2015).

187           In the meta-analysis of Averós et al. (2012), pigs fed ad libitum had higher FCR than  
188 pigs fed restrictively. The effect of feed restriction on FCR varied with the type of animal. For  
189 a lean animal, a restriction in energy supply reduced similarly the deposition of proteins and  
190 lipids while for a fat animal, fat deposition is reduced in priority without increasing protein  
191 deposition (Gilbert, 2015). Therefore, in fat animals, FCR decreases, whereas in lean animals  
192 FCR changes little or even increases due to increased maintenance requirements (Bikker et  
193 al., 1996).

194           **Nutrient utilization.** As previously mentioned, FCR is affected by the efficiency of  
195 energy utilization, which depends on the energy content of BW gain and the effect of  
196 maintenance requirements. Similarly, to the energy, the efficiency of utilizing nutrients, such

197 as amino acids (AA) and minerals, is also affected by nutrient digestibility and maintenance  
198 requirements Above maintenance, the apparent efficiency of use of digestible minerals and  
199 AA for tissue deposition depends on their metabolic efficiency of retention, and on their  
200 possible oversupply. In case of nutrient undersupply, the marginal efficiency of retention is at  
201 highest, but due to insufficient supplies, growth rate decreases resulting in an increased FCR.  
202 Conversely, when nutrients are supplied in excess to the requirement they contribute to  
203 increasing excretion, which results in reduced efficiency of retention. For instance, in growing  
204 pigs, metabolic efficiency of digestible lysine retention is about 72%, whereas on average in  
205 practice, over the fattening period, its apparent efficiency with conventional two-phase  
206 feeding programs is only about 45% (van Milgen et al., 2008). Thus, improving efficiency of  
207 use of nutrients requires (i) to improve their digestibility and (ii) to provide them, over time,  
208 as close as possible to individual animals' requirements in order to limit their oversupply.

209 Up to 80% of P in feedstuff (cereals and seeds) is tightly bound in phytate, but pigs do not  
210 produce enough phytase enzyme to degrade phytate, which encapsulates P, protein, and AA in  
211 feedstuffs. Adding exogenous phytase is thus an effective way to improve the digestibility and  
212 efficiency of utilizing P and, to some extent, other nutrients (Jondreville and Dourmad, 2005).

213 Adequate supplies of minerals and AA are needed to respectively maximize growth and bone  
214 mineralization (Letourneau et al, 2016) and protein retention, and therefore minimize FCR.  
215 The supply of an AA such as lysine influences FCR in different ways (Fig. 2). The lack of one  
216 or more AA limits protein deposition as protein synthesis decreases. This results in an  
217 increase in lipid deposition and consequently a decrease in growth rate and an increase in  
218 FCR (van Milgen, 2008). When one AA is limiting, the other AA in excess are catabolized  
219 and excreted as urea, reducing energy efficiency and increasing cost and waste. Similarly, an  
220 excessive supply of nutrients increases nutrient excretion, fat deposition and FCR. Nutrient  
221 requirements also depend on pig characteristics such as sex, age and genotype. For example,

222 entire males require more lysine than females (Fig. 2) and castrated males. In growing pigs,  
223 the AA requirement relative to energy decreases as BW increases (Van Milgen et al., 2008,  
224 Noblet et al., 2016). In sows, the AA requirement increases with the stage of pregnancy and is  
225 greatest during lactation (Dourmad et al., 2008). Similar trends are observed for mineral  
226 requirements of growing pigs and sows. This indicates the need to optimize each individual's  
227 nutrient ingestion to maximize individual efficiency and reduce excretion. Taking into  
228 account individual characteristics (age, breed, sex, BW) can be used to better feed to animals  
229 up to their individual requirement. Nutrient ingestion can be optimized by using predictive  
230 models of nutrient requirements (Brossard et al., 2017) (see section 3).

231

### 232 **1.3.Strategies to improve feed efficiency**

233 **Genetics and microbiota.** Feed efficiency is one of the most important selection criteria  
234 in breeding programs, as it affects total cost of pig production and environmental footprint  
235 (Kanis et al., 2005; Reckmann et al., 2016). Thus, FE and related traits are a major target for  
236 genomic selection, a growing and promising method. Until recently, FCR and RFI were the  
237 two main traits used to evaluate FE, as described in the first part of this review. With the  
238 development of new technology and automatic data recording, recent studies have been  
239 looking for new traits related to FE like, for example, feed intake and feeding behavior (daily  
240 occupation time, daily feeder visit, and daily feeding rate) recorded daily and individually.  
241 Major quantitative trait loci for feed intake and for feeding behavior traits have been  
242 identified on different chromosomes as well as the positional and functional candidate genes  
243 (Reyer et al., 2017). This is a first step toward the understanding of the genetic connection  
244 between distinct feeding behavior traits and FE that can be used to select the most efficient  
245 animals.

246 Genetic factors are also influencing the abundance of distinct bacterial species (Benson et  
247 al., 2010). Several studies found a link between the porcine intestinal microbiome and FE  
248 (McCormack et al., 2019a; Tan et al., 2017). While the effect of intestinal microbiota in  
249 expressing FE has been confirmed, methods to phenotype the microbiota should be developed  
250 to use this information on farms. McCormack et al. (2019b) reported the effect of fecal  
251 microbiota transplantation (FMT), using fecal extracts from highly feed-efficient pigs, in  
252 pregnant sows on the offsprings performance. The FE of the offsprings from a mother under  
253 FMT was improved compared to the one of offsprings from a mother not under FMT (similar  
254 feed intake but growth differences). However, the negative effect of FMT on the offsprings  
255 growth limits the application of this procedure in commercial farms. Manipulations of  
256 bacterial populations can also be used to improve digestibility and FE (Le Sciellour et al.,  
257 2018) that would allow more flexibility regarding ration composition especially on the  
258 amount of fibers. Niu et al. (2015) reported that several bacteria were correlated with apparent  
259 crude fiber digestibility; of these, *Clostridium* is associated with dietary fiber metabolism.  
260 They also found that the abundance and diversity of the gut microbiota in pigs increased and  
261 changed with increasing age. Intestinal microbiota facilitates digestion of fiber, but its effects  
262 on the variability in FE needs to be assessed and broken down into the fraction that depends  
263 on animal genetics, the fraction that depends on breeding conditions and the fraction that sows  
264 transmit to piglets at birth and during lactation.

265 Finally, robust indicators need to be developed to quantify the sensitivity of animals to  
266 variations in the environment. For example, it seems that feed intake and growth rate of pigs  
267 selected for lower RFI are less affected by the quality of the environment than those of pigs  
268 selected for higher RFI (Gilbert et al., 2017).

269 **Removing external stressors.** Black et al. (2001) reported that removing one or more  
270 stressors, or reducing their influence, improves pig performance because the stressors may

271 have additive effects. Structural changes in buildings like the addition of cooling systems,  
272 floor type reduce climate stress (Black et al., 2001). During a short-term exposure to hot  
273 conditions (32°C as opposed to 21°C), pigs ate about 60 to 100 g less feed each day per °C of  
274 heat stress (Heitman and Hughes, 1949; Heitman et al., 1958). In their meta-analysis,  
275 Renaudeau et al. (2011) reported that feed intake and average daily weight gain of growing-  
276 finishing pigs are decreasing with increasing temperature starting from 20°C. Moreover, these  
277 effects were more pronounced with increasing pig body weight. Ambient temperature clearly  
278 impact feed intake and consequently FCR. Structural changes in the building also help  
279 improving cleanliness. Reducing the microbial load by ensuring building hygiene has  
280 increased production and decreased disease incidence (Le Floc'h et al., 2006). In a dirty  
281 environment measured through air quality (amount of ammonia, CO<sub>2</sub> and dust), pigs' feed  
282 intake has been found to decrease of 100 g/kg compared to a cleaner environment, especially  
283 for individually housed pigs (Currie et al., 1997; Lee et al., 1997). The type of feeder (design,  
284 access, location) also has an effect on FCR (Rantanen et al., 1994).

285 **Adequate nutrient supply.** As previously indicated, feed and nutrient efficiency depend  
286 partly on nutrient utilization, which is based on adjusting nutrient supply to requirements.  
287 Therefore, one way to improve FE is to refine this adjustment by estimating animal  
288 requirements more accurately. Two methods are generally used to estimate nutrient  
289 requirements for pigs: empirical and factorial. Briefly, requirements in the empirical method  
290 correspond to those of a population for a given performance target and time interval.  
291 However, the estimated requirements cannot be extrapolated to other situations because they  
292 vary as a function of animal characteristics and environments (Pomar et al., 2003). In contrast,  
293 nutrient requirements in the factorial method are estimated for an average animal at a given  
294 stage. However, pig performance depends on pig characteristics (genetic, age, weight, sex,  
295 social status and health), feed characteristics (feed allowance, nutrient composition and

296 digestibility), and housing conditions (ambient temperature and space allowance) (Noblet and  
297 Quiniou, 1999). Models based on the factorial approach have been developed to simulate  
298 performance of a single animal and can predict nutrient requirements and appropriate feeding  
299 strategies (e.g., van Milgen et al., 2008). However, because they are based on an average  
300 animal, feeding strategies based on this approach means that many animals are inevitably  
301 underfed or overfed (Pomar et al., 2003). Individual variability influences the efficiency of  
302 nutrient utilization (Pomar et al., 2003, Brossard et al., 2009). In addition, the fact that  
303 nutrient requirements change over time needs to be considered to predict them more  
304 accurately. Precision feeding requires developing new feeding strategies to refine the  
305 adjustment of nutrient supply to requirements.

306

## 307 **2. Individual variability, the key point of precision feeding**

308

### 309 **2.1.Precision feeding: definition, objectives, advantages**

310 Precision feeding is one way to better consider individual variability in nutrient requirements  
311 within a group. It involves using technology to provide the right amount of feed, with the  
312 right composition and at the right time, to a group of animals or to individuals (Pomar et al.,  
313 2009). Precision feeding aims to improve characterization of individuals (feed intake, growth  
314 potential, body condition, physical activity, health, etc.) or small groups to better adapt the  
315 quantity, quality and timing of feed supplied to them. It also aims to improve efficiency by  
316 reducing farm costs, reducing excretion, and monitoring quality (Fig. 3). Applying precision  
317 feeding and doing so accurately requires assessing the nutritional potential of feed ingredients  
318 and nutrient requirements of each animal to formulate balanced diets accurately to minimize  
319 nutrient deficiency or excess (Pomar et al., 2009).

320 The results of previous precision feeding assessments are promising (Andretta et al., 2014,  
321 Pomar et al., 2014, Andretta et al., 2016). For growing-finishing pigs, compared to a classic  
322 three-phase group-feeding strategy, adjusting feed composition daily based on the  
323 performance of an average animal in the group decreased N excretion by 12% without  
324 influencing growth (Pomar et al., 2007). This continual adjustment also has an economic  
325 advantage because it can be based on a mixture of two feeds, one with a high nutrient content  
326 and one with a lower nutrient content. At the individual scale, precision feeding of growing-  
327 finishing pigs further reduces N and P excretions compared to a multiphase group-feeding  
328 strategy (respectively, 38 vs. 42 g/d for N, and 5 vs. 6 g/d for P) (Andretta et al., 2014).  
329 Simulations indicate that precision feeding could also be beneficial for sows: using a  
330 multiphase feeding strategy (a mixture of two feeds) during gestation reduced the quantity of  
331 lysine ingested (-17%), N excretion (-19%), and feed cost (-8%) (Dourmad et al., 2015).  
332 These results for sows need to be confirmed with trials in experimental farms. “On-farm”  
333 application of precision feeding requires designing and developing measuring devices (for  
334 intake, BW), calculation methods and a feeding system that provides the required amount of  
335 feed with a composition that optimizes animal performance while minimizing the use of farm  
336 resources (Pomar et al., 2009).

337

## 338 **2.2.Variation in nutrient requirements**

339 Two main sources of variation in nutrient requirements must be considered: variations over  
340 time and differences between animals. Nutrient requirements vary over time (Andretta et al.,  
341 2014) and among growing pigs in a group receiving the same feed (Pomar et al., 2007,  
342 Brossard et al., 2009), due to sex (castrated or entire males, females), age (different nutritional  
343 requirements), weight and individual potential. Nutritional requirements of sows also vary  
344 with individual characteristics such as physiological status, age, weight and prolificacy

345 (Dourmad et al., 2017). Sows of parity 1, 2, or 3 continue to grow while gestating, whereas  
346 sows of parity 4 or more have already reached their mature weight, which means that their  
347 requirements are limited to maintenance, gestation and rebuilding body reserves used up  
348 during gestation. These between-animal variations influence the population response, the  
349 efficiency of nutrient utilization, and consequently the optimal nutrient supply for the  
350 population (Pomar et al., 2003, Brossard et al., 2009). Therefore, stochasticity has been  
351 introduced into models to address variability, simulate responses of groups of pigs (Pomar et  
352 al., 2003) and define strategies to improve on-farm nutrient efficiency (Brossard et al., 2017).  
353 These variations among animals and over time show the relevance of developing more  
354 individualized feeding strategies (see section 3).

355 In growing pigs, Cloutier et al. (2015) reported that the factorial method used to estimate  
356 individual daily lysine requirements was able to accommodate the small genetic differences in  
357 feed intake without a specific correction for genetic differences. This method can be used in  
358 precision feeding systems without adjustments for small genetic differences but should be  
359 studied further for larger genetic differences in feed intake and protein deposition patterns.

360

### 361 **2.3. Feeding technologies and real-time data**

362 Improving FE requires considering individual feeding requirements. Precision feeding uses  
363 feeding technologies to adjust animals' diets. Development of precision feeding systems  
364 requires automatic data collection, data processing, and system monitoring.

365 **Technology and individual data collection.** Farm animals can be identified by  
366 radiofrequency (RFID), which makes data collection reliable and simplifies management of  
367 individuals (Cornou and Kristensen, 2013). Using RFID and sensors automates farm  
368 equipment, which can transfer real-time data to a farmer or an automated decision support  
369 system that can make rapid management decisions (Fig. 3). Cornou and Kristensen (2013)

370 listed several sensors used in pig production and how their data can be used to support  
371 decisions. Automatic identification of an animal is the first step in monitoring production  
372 efficiency and is performed on pig farms usually by placing an ear-tag containing an RFID  
373 chip on each animal to recognize the animal, for example at the feeder. Electronic feeding  
374 stations can record the number, time and duration of the visits, and the quantity of feed  
375 ingested by each pig. In commercial herds, only sows are individually identified at electronic  
376 feeding stations. Individual identification is uncommon for fattening pigs due to its cost,  
377 although the technology is available for selection herds. Several technologies exist that  
378 automatically record BW: foreleg weighing systems (Ramaekers et al., 1995), image analysis  
379 (Parsons et al., 2007), the walk-through using machine vision (Banhazi et al., 2011) and  
380 photogrammetry to determine pigs' three-dimensional shapes (Wu et al., 2004). Knowing the  
381 body composition might also be required to individualize the diet, especially in gestating  
382 sows. Body composition can be determined by analyzing images or videos or measuring  
383 backfat thickness with ultrasound; however, this last technique is performed manually and  
384 would need to be automated. Pig activity, which may also be of interest, can be automatically  
385 recorded using photocells, force sensors for sows housed in crates (Oliviero et al., 2008), and  
386 accelerometers for sows in loose housing or crates (Cornou and Lundbye-Christensen, 2012).  
387 Finally, pig temperature influences FE and can be automatically recorded using an ear-based  
388 temperature sensor or estimated using an image-analysis procedure based on the pig's  
389 thermoregulatory behavior (Wouters et al., 1990).

390 **Data processing.** The sensors described above provide large amounts of data on a daily  
391 basis. The biological characteristic of interest needs to be extracted from each measurement.  
392 First, the data are cleaned by removing abnormal values, which requires defining thresholds.  
393 Then, the characteristic of interest is generally extracted by smoothing the data; the amount of

394 smoothing is based on the objective (Friggens and Robert, 2016). These data will then serve  
395 as inputs to models to predict animals' nutrient requirements.

396

### 397 **3. Models and applications of precision feeding**

398

399 Applying precision feeding requires developing models that predict nutrient requirements and  
400 using the models to test and select new feeding strategies. These models are of interest to  
401 compare alternative production systems to existing ones, from a time and economical point of  
402 view, and also to gain confidence in the success of a new strategy before testing it in real-life.  
403 Until now, most of the production models were based on an average animal, but individual  
404 variability need to be considered to gain in precision (Knap et al., 1995; Kyriazakis, 1999).

405

#### 406 **3.1. Fattening pigs**

407 Hauschild et al. (2012) developed a model that predicts real-time individual AA requirements  
408 of growing-finishing pigs. The model consists of two components (Fig. 4). The empirical  
409 component uses individual pig information (intake and BW) in real time to estimate daily feed  
410 intake and daily gain for the current day. Based on these estimates, the mechanistic model  
411 uses factorial equations to predict net energy intake and AA requirements (expressed through  
412 standardized ileal digestible lysine, SID Lys). The optimal AA concentration needed to meet  
413 each pig's requirements is predicted daily. To do so, the model requires at least seven  
414 consecutive feed intake measurements and two BW measurements to begin predicting feed  
415 intake, BW and the nutrient requirements.

416 The Hauschild et al. (2012) model was evaluated using data from a previous trial (Pomar et  
417 al., 2007) that tested the influence of a daily 3-phase or multiphase feeding strategy on pig  
418 efficiency. Daily feed intake and BW trajectories of an animal could be predicted 1 day or 7

419 days in advance, respectively, with an average mean absolute error of 12% and 1.8%,  
420 respectively. The mechanistic component of the Hauschild model has been used in two animal  
421 trials (Zhang et al., 2012, Cloutier et al., 2015). In the Zhang et al. (2012) trial, the model  
422 accurately predicted SID Lys requirements of pigs of 25-55 kg BW, but underpredicted the  
423 requirements of heavier animals. In the Cloutier et al. (2015) trial, the model was used to  
424 predict individual daily SID Lys requirements and to consider the influence of small genetic  
425 differences on them. Three trials evaluated the overall approach of estimating real-time AA  
426 requirements and the effect of switching from conventional to precision feeding systems in  
427 growing-finishing pig operations on productive performance, nutrient utilization, body  
428 composition and environmental costs (Andretta et al., 2014, Pomar et al., 2014, Andretta et  
429 al., 2016). Pomar et al. (2014) found that a daily phase-feeding strategy (mixing two feeds)  
430 reduced N intake by 7.3%, P intake by 3.3%, N excretion by 11.7%, P excretion by 1.9% and  
431 feed cost by 1.3% compared to those of a 3-phase feeding strategy. Andretta et al. (2014)  
432 found that a multiphase individual feeding strategy reduced SID AA intake by 27%, P  
433 excretion by 27% and N excretion by 20% compared to those of a 3-phase feeding strategy.  
434 Andretta et al. (2016) found that an individual feeding strategy (in which the mixing  
435 proportions of two feeds were updated daily to meet 100% of the lysine requirement) reduced  
436 SID Lys intake by 26%, N excretion by 30% and feeding cost by 10% compared to those of a  
437 group-feeding strategy. These three trials show that using precision feeding techniques to feed  
438 growing-finishing pigs with diets that are tailored daily is an effective approach to reduce  
439 nutrient excretion without compromising performance. It confirms that combining precision  
440 feeding with real-time modeling of requirements can improve the efficiency of use of feed  
441 and nutrient, and to some extent the economic result. However, this requires more  
442 sophisticated equipment (e.g. equipment for feed storage and distribution, smart feeders,  
443 weighing scale), with more supervision, inducing additional costs that were generally not

444 considered in the economic evaluation. Predictive models require further improvements, such  
445 as including health factors (environmental stressors, pathogen levels), and prediction of  
446 technical, economic and environmental effects of precision feeding on commercial farms. For  
447 example, Monteiro et al. (2016) used the InraPorc decision support tool  
448 ([https://inraporc.inra.fr/inraporc/index\\_en.html](https://inraporc.inra.fr/inraporc/index_en.html)) to predict production data that they then used  
449 as input data for life cycle assessment to compare environmental effects of four pig feeding  
450 strategies. They predicted that an individual feeding strategy yielded the lowest life cycle  
451 effects for pig fattening in all situations (in France and Brazil). Finally, this technology needs  
452 to be implemented and validated on commercial farms. Current technology can feed pig  
453 groups based on their weight, but automatic feeders with a decision support tool are not yet  
454 commercially available.

455

### 456 **3.2. Sows in gestation and lactation**

457 Mechanistic models such as InraPorc (Dourmad et al., 2008) and the model of Hansen et al.  
458 (2014) were developed to simulate energy and nutrient partitioning of reproductive sows on a  
459 daily time step. These models represent sows as the sum of multiple compartments: body  
460 protein, body lipids, body minerals and the uterus (Dourmad et al., 2008) (Fig. 5). Equations  
461 describing nutrient utilization by sows were used to build InraPorc, which predicts daily  
462 nutrient and energy flows from feed to storage in the body and then excretion. InraPorc  
463 simulates daily utilization of key nutrient pools by a sow. InraPorc also predicts energy and  
464 AA requirements of sows based on production objectives, as well as changes in body  
465 composition due to a given feeding strategy or housing condition (Gaillard et al., 2019a).

466 Dourmad et al. (2015) used InraPorc to simulate and evaluate two-phase and multiphase  
467 feeding strategies during gestation. Simulations results indicate that compared to one-phase  
468 feeding, the two-phase and multiphase strategies could respectively reduce crude protein (CP)

469 intake by 10% and 14%, SID Lys intake by 11% and 17%, P intake by 5% and 7 %, N  
470 excretion by 15% and 20%, and P excretion by 9% and 12%. Dourmad et al. (2017) and  
471 Gaillard et al. (2019a) developed a decision support tool for gestating sows based on InraPorc.  
472 Optimal supply for a given sow was determined each day by a factorial approach that  
473 considered all the information available about the sows (genotype, parity, gestation stage,  
474 etc.). Energy supply was calculated for each sow to reach a target BW at farrowing. Precision  
475 feeding with the mixing of two feeds was then simulated and compared to conventional  
476 feeding (single feed). Simulations indicated that compared to conventional feeding, precision  
477 feeding could reduce total SID Lys supply by 27%, total CP supply by 28%, and the number  
478 of under- or over-fed sows (Gaillard et al., 2019b). Adapting the feeding strategy during  
479 gestation to capture changes in nutrient requirements more adequately appears a promising  
480 approach to reduce N and P excretion without increasing feed cost, but this remains to be  
481 validated on experimental farms. During gestation, sows are housed in groups, offering the  
482 potential to use automatic feeders and apply these new feeding plans that consider sow  
483 characteristics, such as parity, weight, and backfat thickness at the start of gestation.  
484 However, although this approach is possible, the use of models and the potential to improve  
485 FE remains limited in practice, mainly due to insufficient data collection and the lack of  
486 decision support systems.

487 On most farms, sows are fed different diets for gestation and lactation instead of the same diet  
488 for both, reducing N and P excretion by 20-25% (Dourmad et al., 1999). Currently, few  
489 studies have focused on improving feeding strategies for lactating sows, even though sow  
490 requirements vary greatly. A precision feeding strategy might be useful for lactating sows  
491 because nutrient requirements per kg of diet vary greatly as a function of milk production and  
492 feed intake. For example, sows of parity 1 have greater requirements due to their lower feed  
493 intake. Nutrient requirements also vary by season due to the influence of temperature on feed

494 intake and milk production. Gauthier et al. (2019) developed a decision support system based  
495 on InraPorc that could be incorporated in automated feeding equipment. Simulations  
496 compared a conventional feeding strategy to a precision feeding strategy; the latter could  
497 reduce mean lysine intake by 6.8%, P intake, and the number of under- or over-fed lactating  
498 sows.

499

### 500 **3.3. Modeling mineral requirements**

501 Minerals are a major component of pig nutrition. Because P is necessary for bone  
502 development and the metabolism of growing pigs, it must be supplied in pig diets. Dietary P  
503 of plant origin has low digestibility for pigs, but addition of P and/or phytase increases feed  
504 cost. The oversupply and low digestibility of P also results in high P excretion, which affects  
505 the environment. Therefore, models that predict mineral requirements are required to optimize  
506 mineral supply and minimize excretion (Brossard et al., 2017). Minerals have received little  
507 modeling attention because most models have focused on AA. Jondreville and Dourmad  
508 (2005) used a factorial approach to estimate P requirements for maintenance and production  
509 in different physiological stages, and it was later added to InraPorc for growing pigs (van  
510 Milgen et al., 2008) and sows (Dourmad et al., 2008). This approach considers the influence  
511 of the type of diet (pellets or mash) and the addition of phytase on digestibility. The model  
512 allows dietary P supply to be adjusted to pig performance and physiological status and  
513 predicts the influence of performance level on apparent digestible P requirements. However, P  
514 requirements for growth are estimated from animal BW gain, which has certain limitations.  
515 More mechanistic models have therefore been developed in which mineral content (P and  
516 calcium) can vary independently of protein and lipid mass (Letourneau-Montminy et al.,  
517 2015). These deterministic and mechanistic research models can be used to improve decision  
518 support tools to develop feeding strategies that minimize P excretion. These models must also

519 consider that mineral requirements change during each physiological stage (e.g. an increase in  
520 calcium requirements at the end of gestation).  
521 For gestating sows, recent simulations of a precision feeding strategy based on lysine  
522 requirements still report an important excess in phosphorus (Gaillard et al., submitted). This is  
523 partly because the implemented strategy was based on lysine requirements with only two  
524 diets. Hence, in such condition it was not possible to deal with the different dynamic of lysine  
525 and phosphorus requirements over the whole gestation. To modulate lysine and phosphorus  
526 supplies independently, one solution would be to calculate the proportions of the two mixed  
527 diets (High Lysine and Low Lysine) based on lysine and phosphorus requirements  
528 simultaneously, and therefore propose 3 different diets to combine for precision feeding  
529 instead of two (High Lysine + High P, Low Lysine + High P, Low Lysine + Low P).  
530 However, the strategy might be less efficient than for lysine and will need to be evaluated and  
531 compared with the present feeding strategy, based on lysine only, in terms of production,  
532 excretion, and costs.

533

## 534 **Conclusion**

535 Feed efficiency can be expressed as the ratio of mean daily weight gain to mean daily feed  
536 consumption during a given period. In practice, the inverse of this ratio is generally used for  
537 breeding animals and represents the efficiency in converting feed into weight gain (feed  
538 conversion ratio, FCR). Several factors influence FCR, such as feed spillage, feed  
539 digestibility, composition of weight gain, feed intake and nutrient utilization. The FCR can be  
540 decreased by selecting the appropriate form of feed and nutrient density and supply, and by  
541 reducing negative effects of environmental factors. Precision feeding is based on managing  
542 individual variability within a group and uses feeding technologies (e.g. sensors, feeders) to  
543 provide the right amount of feed, with the right composition, and at the right time, to a group

544 of animals or to individuals. Predictive models of nutrient requirements and excretion, such as  
545 InraPorc, have been developed to select optimized feeding strategies. For growing pigs,  
546 precision feeding is a promising solution to reduce nutrient excretion by daily adjusting the  
547 supply of nutrients to individuals. Recent simulations results indicate that this might also be  
548 an appropriate feeding strategy for sows. Decision support models could be enhanced by  
549 improving sensors or considering factors such as ambient temperature and animal physical  
550 activity, which also influence energy utilization and consequently the FCR.

551

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557

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807

808 **Table 1.** Mean of FCR of French pig farms in 2016 by production system. Source: IFIP (French Pork and Pig Institute, <http://en.ifip.asso.fr/>)

System	Breeder, sale at weaning	Traditional breeder	Breeder-Fattener	Fattener	Weaner-Fattener
Number of farms	80	15	1579	82	330
Number of sows	525	569	228	1831	2848
Overall performance					
Feed per sow, kg/year	1221	1251	1218		
Overall FCR, kg/kg	5.45	3.07	2.82	2.96	2.64
Feed cost, €/kg	1.456	0.904	0.680	0.660	0.635
Post weaning					
Feed intake, kg/piglet		33	42		42
FCR, kg/kg		1.78	1.68		1.68
Fattening					
Feed intake, kg/d			2.23	2.31	2.29
FCR, kg/kg			2.69	2.88	2.74

809 FCR: Feed Conversion Ratio

810 **Table 2.** Chemical and tissue composition of the empty BW gain (from Noblet et al., 1994)

Genotype Sex	Piétrain		Large-White		Meishan
	Male	Male	Female	Castrated	Castrated
Growth, g/d	804	881	726	751	458
Tissues, g/kg					
Muscles	580	472	450	420	242
Adipose tissues	181	206	253	309	430
Composition, %					
Water	61.6	58.5	55.0	51.0	39.2
Minerals	2.6	3.0	3.0	3.0	2.1
Proteins	17.4	16.7	15.9	16.0	11.1
Lipids	18.2	21.1	25.0	30.4	48.8
Energy, MJ/kg	11.2	12.3	13.8	15.6	22.1

811

812 **Figure captions**

813

814 **Figure 1.** Relationship between feed conversion ratio (FCR) (MJ ME/kg gain) and ME  
815 ingested (MJ/d) for different breeds (MS = Meishan, LW = Large-White, PT = Piétrain, SL =  
816 Synthetic line of animals selected for their low adiposity) and sexes of pigs from 20-55 kg  
817 BW (open circles) and 55-90 kg BW (solid circles) (data from Noblet et al, 1994). ME =  
818 metabolizable energy.

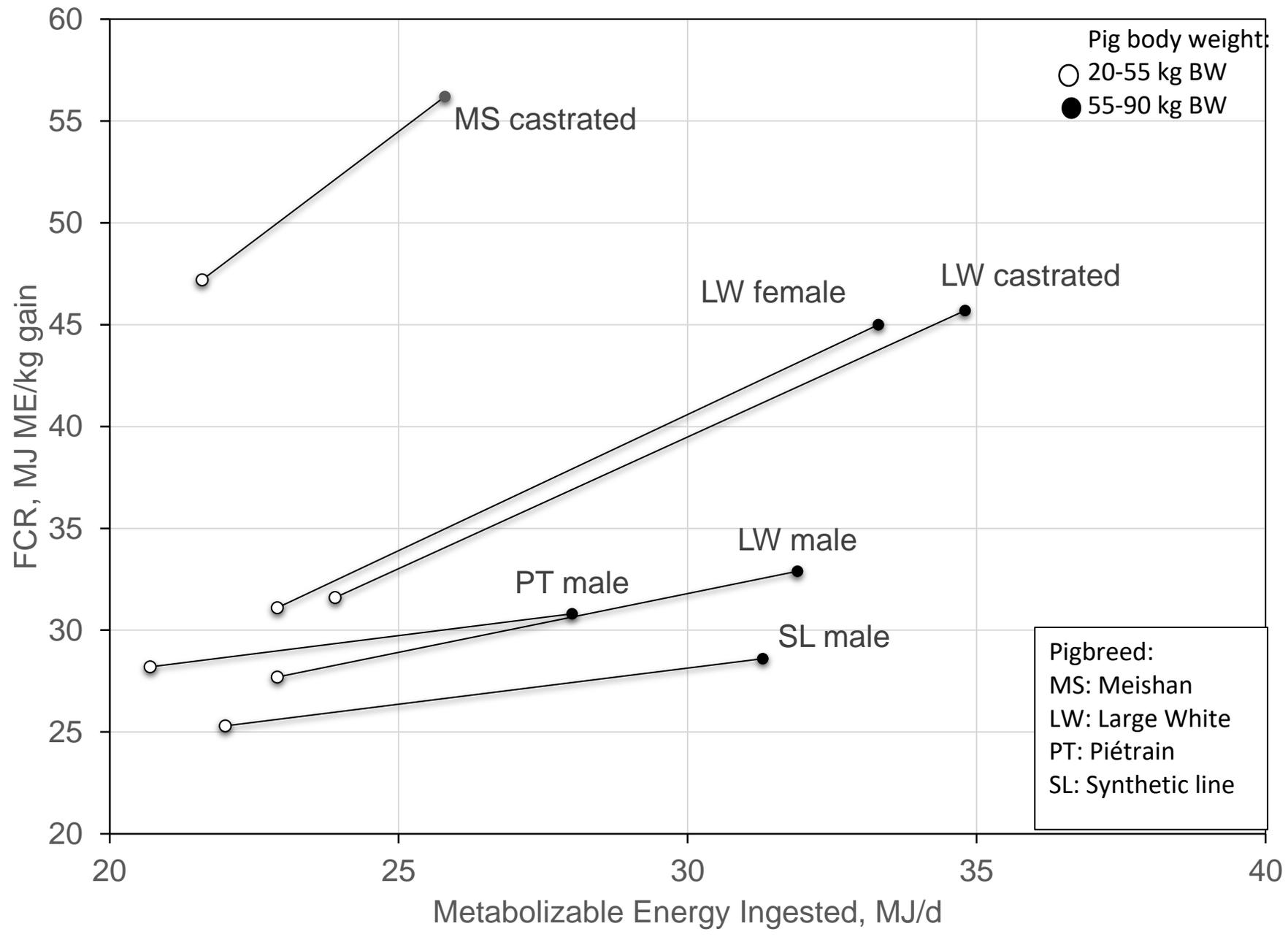
819 **Figure 2.** Effect of the ratio of available lysine to MJ digestible energy on mean feed  
820 conversion ratio ( $\pm 1$  SEM) for (Large-White x Landrace) x Duroc female and male pigs with  
821 a mean body weight of (a) 37.7 kg and (b) 76.4 kg (Mullan et al., 2011)

822 **Figure 3.** Principles of precision feeding (adapted from Allain et al., 2014)

823 **Figure 4.** General outline of the Hauschild et al. (2012) model, with empirical and  
824 mechanistic model components used to estimate daily nutrient requirements for each  
825 individual in a pig population according to its measured growth and feed intake patterns  
826 (adapted from Hauschild et al., 2012). BW = body weight

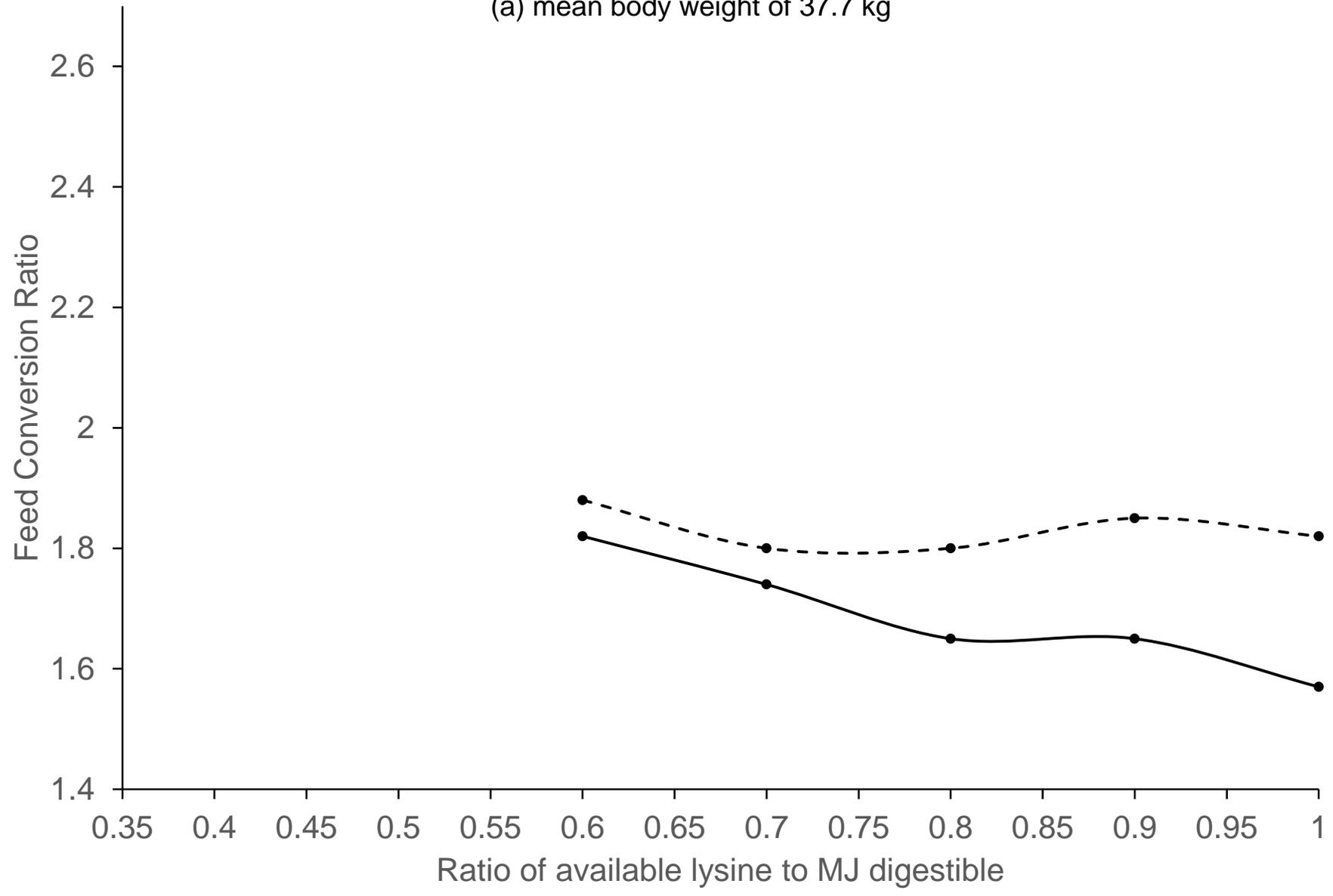
827 **Figure 5.** Configuration of the InraPorc decision making tool for sow nutrition (from  
828 Dourmad et al., 2008). BW = body weight

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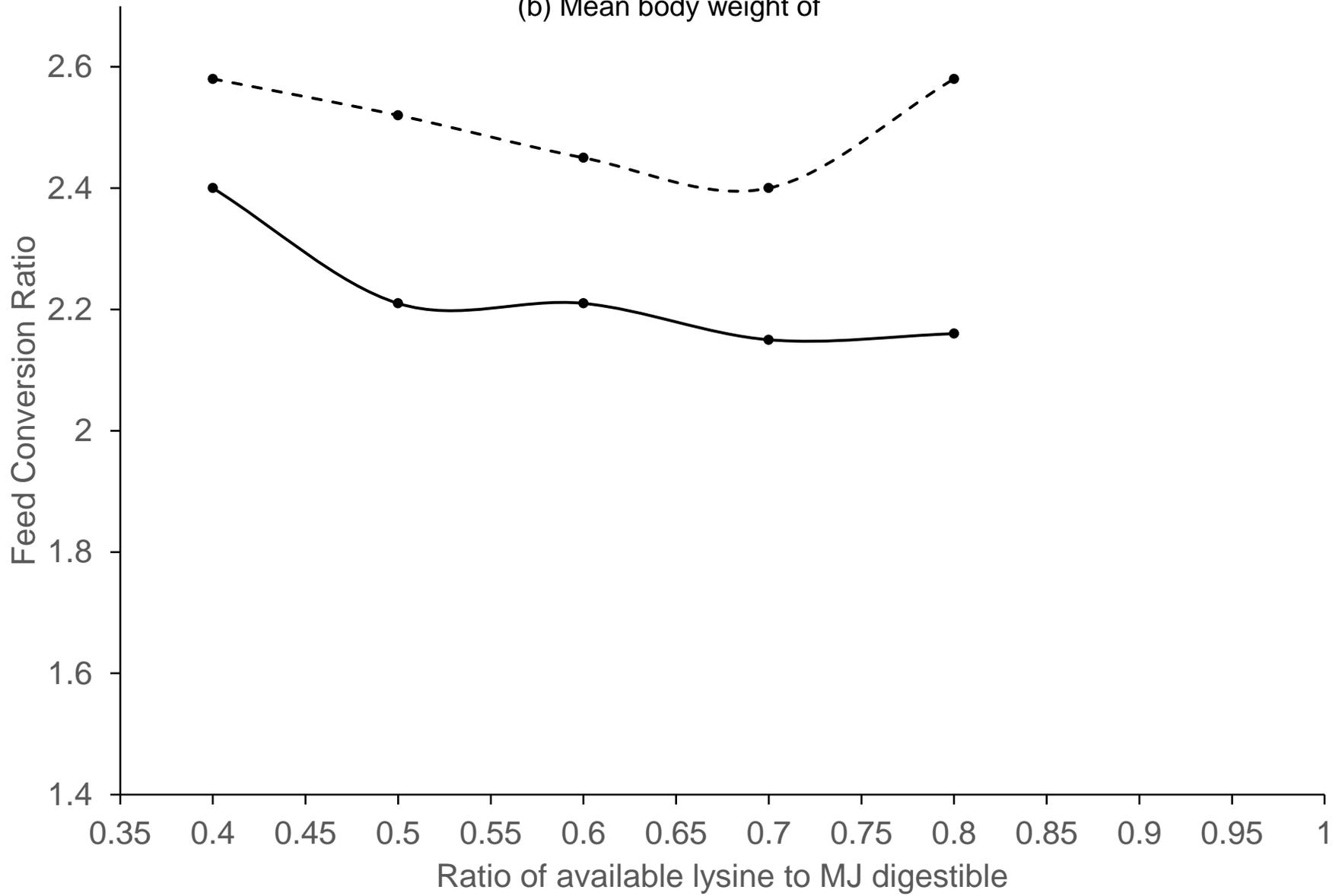
-●- Female    -●- Male

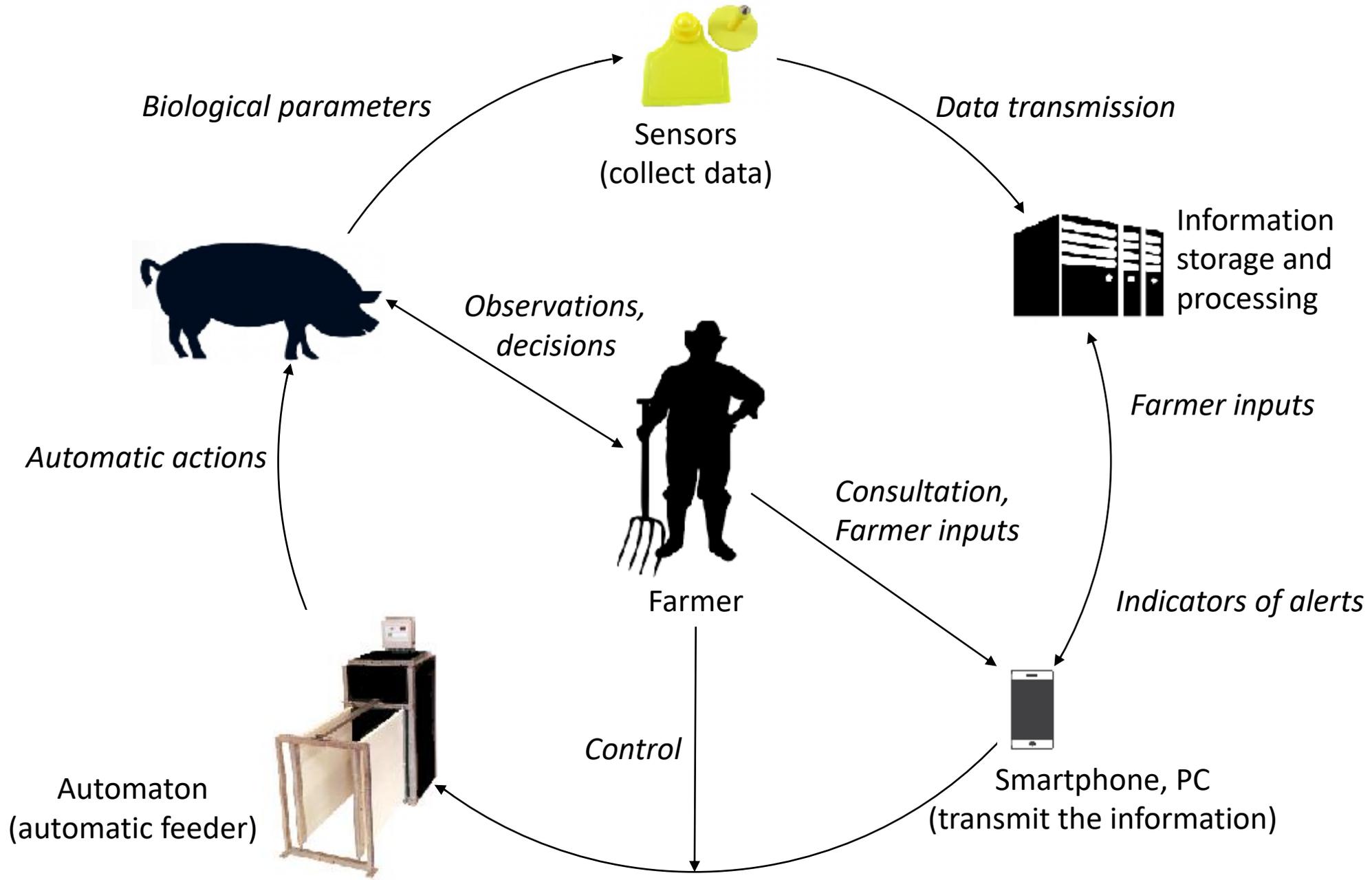
(a) mean body weight of 37.7 kg

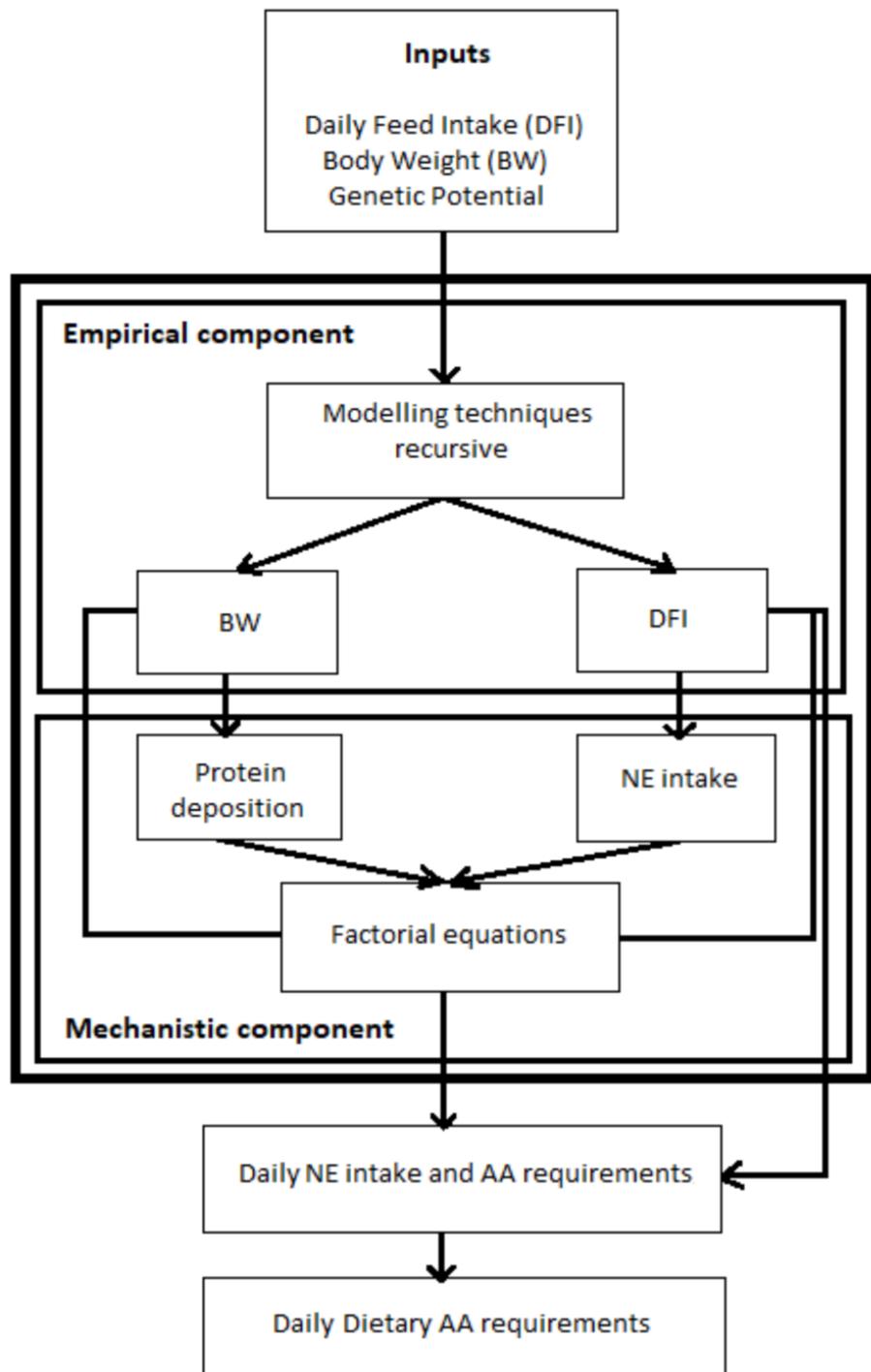


-•- Female —•— Male

(b) Mean body weight of







NE: Net Energy  
AA: Amino Acid

