

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

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10 Highlights
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- Environment, animal and feed characteristics influence nutrient utilization in pigs.
- Mathematical models can be used to estimate real-time daily nutrient requirements.
- Thanks to technological advances, each pig can receive its daily nutrient requirements.
- Precision feeding may also reduce feed cost and environment load.
- 15
- 16
- 17 Abstract

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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

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

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43 Keywords: farming conditions, feed conversion ratio, nutritional models, pig production,

44 precision feeding

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47 Introduction

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

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71 **1. Feed and nutrient efficiency**

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1.1. Definition and measure of feed efficiency

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

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

For fattening pigs, FCR is usually calculated from 10 weeks of age (around 30 kg of body weight, BW) to slaughter (around 115 kg of BW). FCR can also be calculated from weaning 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, 2015). Ultimately, FCR is not determined by growth rate and feed intake, but by factors that 105 106 influence them, such as genetics, feeding practices, environmental conditions and health status. For example, data obtained from a French pig farm survey (Table 1) indicate that FCR 107 108 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). FCR is greater for the fattening period than for the post-weaning period, and these values vary 110 111 among farms.

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1.2. Factors that influence feed efficiency

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

119 [Eq. 1] FCR = (feed intake + spillage) / pig growth

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

121 [Eq. 3] FCR = (indigestible + maintenance + growth) / (protein + water + lipids + minerals)

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

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

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

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

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

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

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

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

as amino acids (AA) and minerals, is also affected by nutrient digestibility and maintenance 197 198 requirements Above maintenance, the apparent efficiency of use of digestible minerals and AA for tissue deposition depends on their metabolic efficiency of retention, and on their 199 200 possible oversupply. In case of nutrient undersupply, the marginal efficiency of retention is at highest, but due to insufficient supplies, growth rate decreases resulting in an increased FCR. 201 Conversely, when nutrients are supplied in excess to the requirement they contribute to 202 increasing excretion, which results in reduced efficiency of retention. For instance, in growing 203 204 pigs, metabolic efficiency of digestible lysine retention is about 72%, whereas on average in practice, over the fattening period, its apparent efficiency with conventional two-phase 205 206 feeding programs is only about 45% (van Milgen et al., 2008). Thus, improving efficiency of use of nutrients requires (i) to improve their digestibility and (ii) to provide them, over time, 207 208 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 produce enough phytase enzyme to degrade phytate, which encapsulates P, protein, and AA in 210 211 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). Adequate supplies of minerals and AA are needed to respectively maximize growth and bone 213 214 mineralization (Letourneau et al, 2016) and protein retention, and therefore minimize FCR. The supply of an AA such as lysine influences FCR in different ways (Fig. 2). The lack of one 215 or more AA limits protein deposition as protein synthesis decreases. This results in an 216 increase in lipid deposition and consequently a decrease in growth rate and an increase in 217 218 FCR (van Milgen, 2008). When one AA is limiting, the other AA in excess are catabolized and excreted as urea, reducing energy efficiency and increasing cost and waste. Similarly, an 219 220 excessive supply of nutrients increases nutrient excretion, fat deposition and FCR. Nutrient requirements also depend on pig characteristics such as sex, age and genotype. For example, 221

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

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

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

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

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

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

have additive effects. Structural changes in buildings like the addition of cooling systems, 271 floor type reduce climate stress (Black et al., 2001). During a short-term exposure to hot 272 conditions (32°C as opposed to 21°C), pigs ate about 60 to 100 g less feed each day per °C of 273 274 heat stress (Heitman and Hughes, 1949; Heitman et al., 1958). In their meta-analysis, Renaudeau et al. (2011) reported that feed intake and average daily weight gain of growing-275 finishing pigs are decreasing with increasing temperature starting from 20°C. Moreover, these 276 effects were more pronounced with increasing pig body weight. Ambient temperature clearly 277 impact feed intake and consequently FCR. Structural changes in the building also help 278 improving cleanliness. Reducing the microbial load by ensuring building hygiene has 279 increased production and decreased disease incidence (Le Floc'h et al., 2006). In a dirty 280 environment measured through air quality (amount of ammonia, CO2 and dust), pigs' feed 281 intake has been found to decrease of 100 g/kg compared to a cleaner environment, especially 282 283 for individually housed pigs (Currie et al., 1997; Lee et al., 1997). The type of feeder (design, access, location) also has an effect on FCR (Rantanen et al., 1994). 284

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

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

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2. Individual variability, the key point of precision feeding

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

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

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338 **2.2.Variation in nutrient requirements**

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

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

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

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2.3.Feeding technologies and real-time data

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

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

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

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

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

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3. Models and applications of precision feeding

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

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406 **3.1.** Fattening pigs

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

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

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

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

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3.2. Sows in gestation and lactation

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

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)

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

On most farms, sows are fed different diets for gestation and lactation instead of the same diet for both, reducing N and P excretion by 20-25% (Dourmad et al., 1999). Currently, few studies have focused on improving feeding strategies for lactating sows, even though sow requirements vary greatly. A precision feeding strategy might be useful for lactating sows because nutrient requirements per kg of diet vary greatly as a function of milk production and feed intake. For example, sows of parity 1 have greater requirements due to their lower feed intake. Nutrient requirements also vary by season due to the influence of temperature on feed intake and milk production. Gauthier et al. (2019) developed a decision support system based
on InraPorc that could be incorporated in automated feeding equipment. Simulations
compared a conventional feeding strategy to a precision feeding strategy; the latter could
reduce mean lysine intake by 6.8%, P intake, and the number of under- or over-fed lactating
sows.

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3.3. Modeling mineral requirements

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

consider that mineral requirements change during each physiological stage (e.g. an increase incalcium requirements at the end of gestation).

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

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534 Conclusion

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

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

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553

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System Breeder, sale at weaning Traditional breeder Breeder-Fattener Fattener Weaner-Fattener Number of farms 1579 82 80 15 330 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 1.78 1.68 1.68 FCR, kg/kg Fattening Feed intake, kg/d 2.23 2.31 2.29 2.69 2.88 2.74 FCR, kg/kg

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

809 FCR: Feed Conversion Ratio

Genotype	Piétrain		Large-White		
Sex	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

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

812 Figure captions

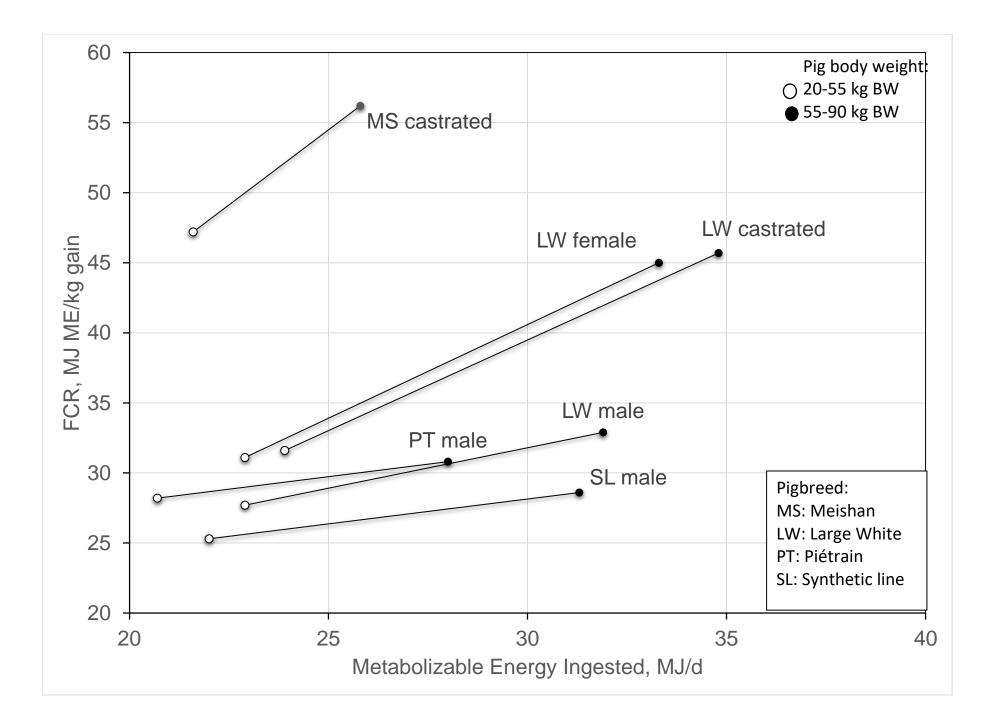
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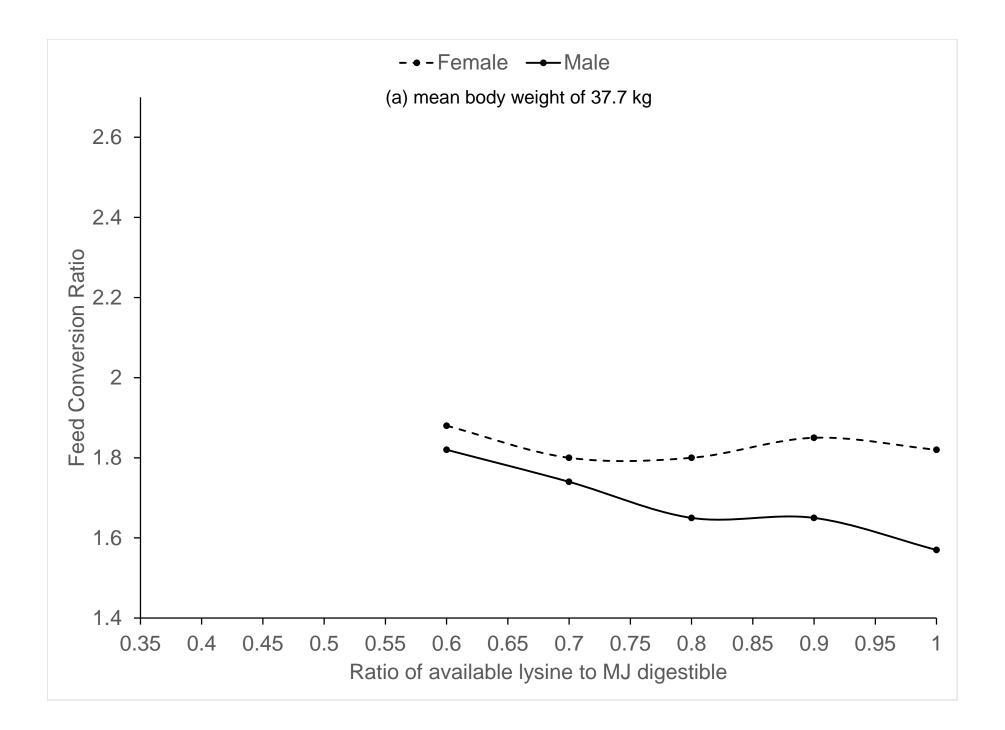
Figure 1. Relationship between feed conversion ratio (FCR) (MJ ME/kg gain) and ME 814 ingested (MJ/d) for different breeds (MS = Meishan, LW = Large-White, PT = Piétrain, SL = 815 Synthetic line of animals selected for their low adiposity) and sexes of pigs from 20-55 kg 816 BW (open circles) and 55-90 kg BW (solid circles) (data from Noblet et al, 1994). ME = 817 818 metabolizable energy. 819 Figure 2. Effect of the ratio of available lysine to MJ digestible energy on mean feed conversion ratio (± 1 SEM) for (Large-White x Landrace) x Duroc female and male pigs with 820 a mean body weight of (a) 37.7 kg and (b) 76.4 kg (Mullan et al., 2011) 821 Figure 3. Principles of precision feeding (adapted from Allain et al., 2014) 822 Figure 4. General outline of the Hauschild et al. (2012) model, with empirical and 823

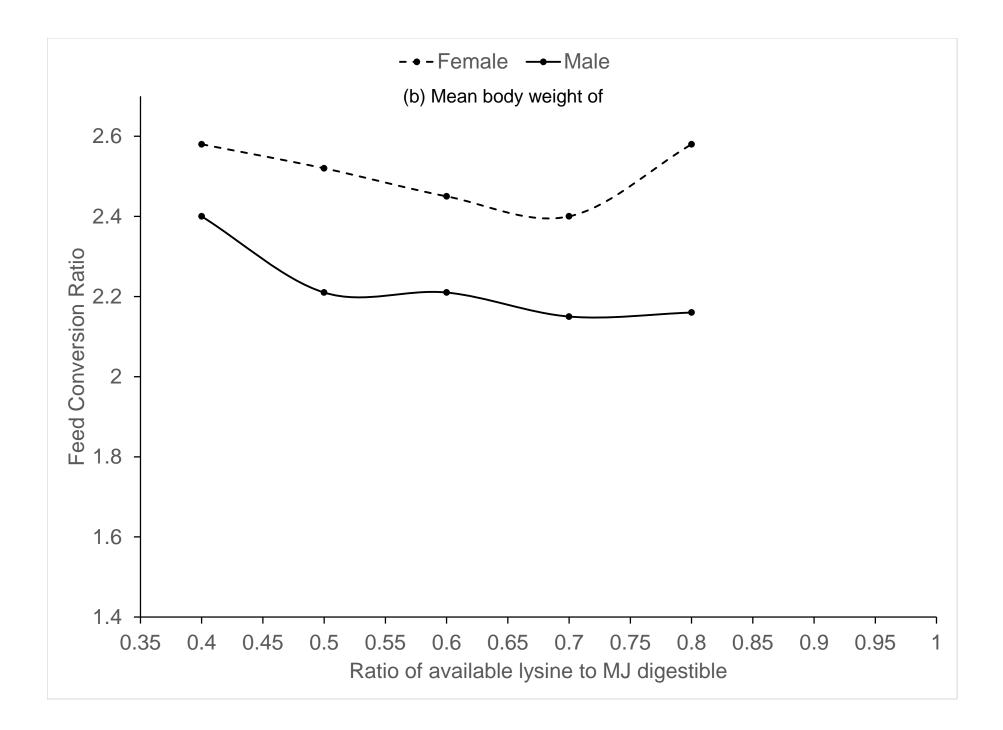
mechanistic model components used to estimate daily nutrient requirements for each individual in a pig population according to its measured growth and feed intake patterns (adapted from Hauschild et al., 2012). BW = body weight

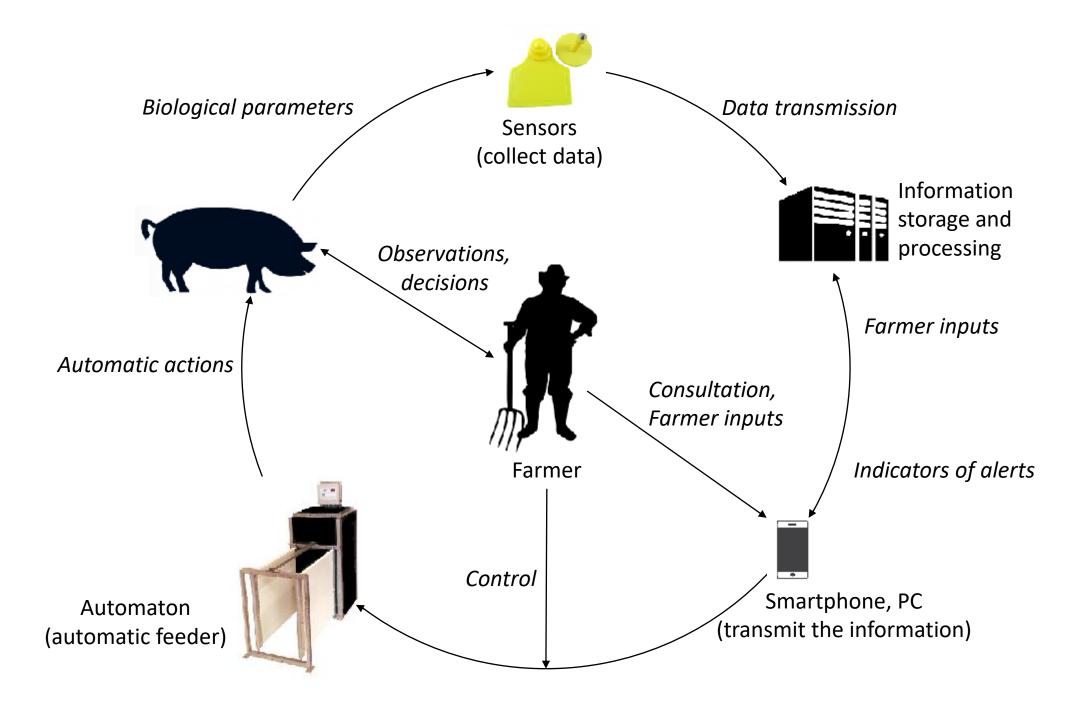
Figure 5. Configuration of the InraPorc decision making tool for sow nutrition (from
Dourmad et al., 2008). BW = body weight

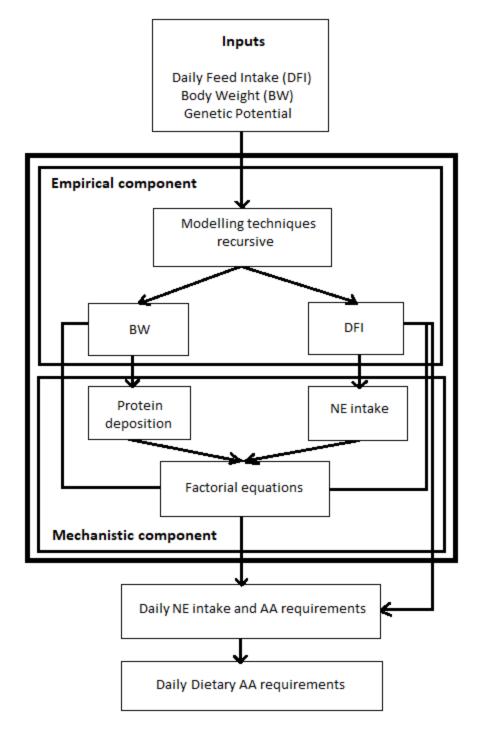
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NE: Net Energy AA: Amino Acid

