

1 Running head: Genetic parameters of CT traits

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3 **Genetic parameters of *in vivo* primal cuts and body composition (PigAtlas) in pigs**
4 **measured by computed tomography (CT)¹**

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8 **ABSTRACT:**

9 Genetic parameters of *in vivo* primal cuts in breeding pigs using computed tomography were
10 estimated. 2439 Duroc and 1998 Landrace boars from the Topigs Norsvin boar testing station
11 in Norway were CT scanned as part of the genetic program. *In vivo* primal cuts were derived
12 from the CT images using atlas segmentation; the method called the Pig Atlas. The
13 (co)variance estimates were obtained from univariate (heritabilities) and multivariate
14 (correlations) animal genetic models using DMU software. The heritabilities for all primal
15 cuts proportions (%) were intermediate to high for both breeds, h^2 ranging from 0.15 to 0.50.
16 Negative genetic correlations were found between most of the other primal cuts, and the
17 strongest correlation was between belly and ham. Carcass lean meat percentage showed a
18 positive correlation to shoulder and ham, but was negatively correlated to belly. In this study,
19 *in vivo* primal cuts from atlas segmentation are used for genetic parameter calculations for the
20 first time. Computed Tomography (CT) makes it possible to measure *in vivo* body or carcass
21 composition. This will aid the selection response by measuring on the candidates themselves
22 instead of using relatives. Primal cut proportion and composition measured *in vivo* by

¹ The Norwegian Research Council is acknowledged for funding this project with grant #256316

23 computed tomography and atlas segmentation show heritable variation comparable to
24 previous post mortem studies.

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26 **Key words:** pigs, body composition, genetic parameters, primal cuts, computed tomography,
27 PigAtlas

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INTRODUCTION

30 The most important traits in pig breeding have for decades been related to efficient lean meat
31 production, i.e. selection for increased growth rate, feed efficiency and low fat carcass
32 composition (Cameron, 1990). However, breeding goals have been changing from focus on
33 cost reduction toward retail carcass yield and meat quality (Miar et al., 2014), and recently
34 including sustainability (Rydhmer et al., 2014), social breeding values (Reimert et al., 2014)
35 and robustness (Herrero-Medrano et al., 2015). Carcass yield is still highly valid in breeding
36 goals, but in order to give more space to the more recent traits, we need to find more efficient
37 ways of measuring carcass traits *in vivo*, bypass the use of sibs or half-sibs in order to
38 measure these traits on the selection candidates themselves. *In vivo* or non-invasive measures
39 of carcass composition can be derived either indirectly by ultrasound and other
40 anthropometric measures like body conformation score, or more directly by Computed
41 Tomography (CT) , Magnetic Resonance Imaging (MRI) and dual-energy X-ray
42 absorptiometry (DXA or DEXA) (Scholz et al., 2015). Heritability, h^2 , is defined as the
43 genetic contributions to a population's phenotypic variance. Heritabilities of the primal cuts
44 ham, loin, belly and shoulder, based on dissected carcasses, are found to be medium to high.
45 For example, the ham, belly and loin have been reported to range from 0.29 to 0.57 in
46 previous studies, and for deboned cuts, the heritabilities were even higher ($h^2 > 0.70$)
47 (Newcom et al., 2002; Van Wijk et al., 2005). Previous studies have shown that primal cut

48 traits are highly heritable on the carcass level (post mortem), however the genetic parameters
49 for *in vivo* primal cuts and carcass composition remain elusive. The first *in vivo* trials for
50 measuring body composition in pigs using CT was done in Norway already in early 1980s
51 (Skjervold et al., 1981). Several trials were performed during the 1980s, and the method of
52 using *in vivo* CT to describe body composition was shown to be very accurate in terms of
53 protein (meat) and fat, summarized by Horn (1995). Although these early trials were limited
54 by the number of images obtained from each animal, a small set of images could still provide
55 accurate results with respect to total body composition of protein and fat. In order to get more
56 accurate results on weights of primal cuts and measure more details in bone, body
57 composition and eg. internal organs, more images are needed per animal. Recent advances in
58 computing power have made it possible to analyze larger sets of images, and in 2008, Norsvin
59 introduced CT scanning in pig breeding based on full body scan with slice thickness of 1.25
60 mm, 1.100 images per animal (Gjerlaug-Enger et al., 2012). The resolution of such data
61 makes us able to estimate volumes of different parts of the pig body. In order to handle large
62 amounts of data and animals, an automation was needed in the process of virtually dissect the
63 pig based on CT. In order to aid the automated process, the authors have recently looked at
64 developing a pig atlas based on CT (Gangsei et al., 2016a; Gangsei and Kongsro, 2016). The
65 Pig Atlas is a digital map based on computed tomography body scans of a subset of ~500 pigs
66 from the population of purebred Landrace and Duroc pigs in Norway. The Pig Atlas is an
67 average pig based on that particular subset. Volumes of CT scanned pigs are continuously
68 fitted onto the pig atlas and deformed in order to transfer information from the atlas to the
69 new animal. The primal cuts are identified using the already established “landmarks”, and the
70 pig is brought back to its original shape. This has proven to be an effective way of recording
71 primal cut data from a large number of animals in an automated way (Gangsei et al., 2016a).
72 This is, to our knowledge, the first attempt to study *in vivo* carcass and primal cut

73 composition using CT in a commercial breeding program. Study of genetic parameters for *in*
74 *vivo* carcass and primal cut composition is necessary in order to implement these traits in a
75 commercial breeding program for pigs. Thus, the aim of this study was to estimate genetic
76 parameters of *in vivo* primal cuts and carcass composition in breeding pigs measured by atlas
77 segmented images from computed tomography (CT).

78

79

MATERIALS AND METHODS

80 All animals were cared for according to the laws and regulations for keeping pigs in Norway
81 (Regulation for the keeping of pigs in Norway 2003-02-18-175, 2003; Animal welfare Act
82 2009-06-19-97, 2009).

Animals

84 2439 purebred Duroc and 1998 Landrace boars from the boar testing station, Norsvin
85 Delta in Norway, were CT scanned as part of the Topigs Norsvin genetic program as
86 described more in detail by Aasmundstad et al., 2013. The purebred boars were born and
87 raised to 25-30 kg in different nucleus herds located in Norway. As part of the test program,
88 the pigs are sent to the boar testing station, Norsvin Delta, when reaching 25-30 kg. The test
89 includes feed and weight recordings, score of exterior traits and CT scanning at the end of test
90 when they reach 120 kg as described more in detail by Gjerlaug-Enger et al., 2012.

91

CT scanning

93 The pigs were CT scanned using a GE Healthcare VCT 32 scanner at 120 kg body
94 weight. The protocol used was optimized for soft tissue using a slice thickness of 1.25 mm,
95 120 kV and dynamic mA based on the thickness of the slice. Prior to CT scanning, the boars
96 were sedated using Azaperone, 8 mg/kg live weight (Stresnil Vet® , Janssen-Cilag Ltd,

97 Buckinghamshire, UK) administered intramuscularly. Boars were scanned approximately 45
98 min after injection, as the sedation was given to keep the boars calm during scanning.

99

100 ***Image analysis and atlas segmentation***

101 Atlas segmentation (Gangsei et al., 2016a; Gangsei et al., 2016b) was the impetus for
102 obtaining new and better phenotypes *in vivo* to improve accuracy and potentially increase
103 heritabilities for the traits of interest. By atlas segmentation every voxel was assigned to one
104 of the primal cuts, i.e. ham, belly, loin (including tenderloin) or shoulder, or as non-
105 commercial parts, i.e. internal organs, testicles etc. The labelled atlas was based on expert
106 segmentation of CT images (Gangsei et al., 2016a). After atlas segmentation all voxels in the
107 primal cuts were classified into the tissue classes; meat, fat or bone. Voxels with HU > 200
108 were classified as bone. Thereafter, in order to reduce noise, a simple filter was applied. HU
109 values used for tissue classification of non-bone voxels were set to the minimum HU value of
110 a 2×2 sliding frame within each 2D image slice. Non-bone voxels with filtered HU > 0 were
111 classified as meat and the rest were classified as fat. The weight of each voxel in the primal
112 cuts was calculated as the product of its volume and density, where the density was set based
113 on a simple linear relationship ($1.0062 + 0.0016 \times \text{HU}$) (Campbell et al., 2003). Thereafter, the
114 total weights of meat, fat and bone within each primal cut were easily calculated as the sum of
115 weights of voxels were classified to the actual cut and tissue. These weights constitute the
116 basis for all genetic parameter calculations in this study.

117

118 ***Traits***

119 Carcass weight was defined as the sum of primal cuts (ham, belly, loin and shoulder),
120 excluding the head. The primal cuts ham, belly, loin and shoulder were obtained as described
121 above, and were expressed as primal cut weights (absolute weights in kg) or as relative primal

122 yield (relative proportions of carcass weight in %). Lean meat in each of the primal cuts was
 123 also expressed as a percentage of lean meat (LMP) in the specific primal cut. Carcass lean
 124 meat was expressed as carcass lean meat percentage (CLMP). Carcass yield was defined as a
 125 ratio of carcass weight (sum of all primal cuts) over live weight estimated from the entire CT
 126 volume of the pig. Loin depth was defined as muscle (*m. longissimus*) depth in cm at the last
 127 rib measured on CT images. Backfat thickness was defined as subcutaneous fat layer over
 128 loin depth at the last rib measured on CT.

129

130 *Statistical analysis*

131 The difference in phenotypic means between Duroc and Landrace was analyzed by a student
 132 t-test in SAS (SAS, 2011). Duroc and Landrace were analyzed separately as there is no
 133 common pedigree. The (co)variance estimates were obtained from univariate (heritabilities)
 134 and multivariate (correlations) animal genetic models. The DMU software (Madsen P and
 135 Jensen J, 2013) was used. The model for each trait was defined as:

$$y_{ijklmno} = PARITY_i + HY_j + MONTH_k + SECTION_l + \beta_1 * LW_{ijklmno} \\ + \beta_2 * LW_{ijklmno}^2 + pen_m + litter_n + animal_o + error_{ijklmno}$$

136 Where Y is the studied trait(s). For all y, PARITY (parity of dam) included four levels (i = 1,
 137 2, 3 or a higher parity, unknown parity), HY (herd of origin and year of birth) included j = 35
 138 levels for Duroc and 108 levels for Landrace. MONTH (month born in) included k = 12
 139 levels. SECTION (section in the barn at the testing station) included l = 74 levels for Duroc
 140 and 77 for Landrace. LW of each animal (live body weight in kg from CT scanning) was
 141 fitted as a fixed regression coefficient of first (β_1) and second (β_2) degree. PARITY, HY,
 142 MONTH and SECTION were treated as fixed effects. Pen (at the testing station), litter
 143 (common environmental effect for full sibs), animal and error were treated as random effects.
 144 The additive genetic effects were expected $\sim N(0, G \otimes A)$ and both the pen, litter and residual

145 effect were expected where A is the additive relationship matrix, G is the additive genetic (co)
146 variance matrix, I is an identity matrix of dimension equal to the number of animals with
147 phenotypic records and R is the residual (co)variance matrix. Heritability estimates were
148 obtained from univariate analysis, and heritability was defined as additive genetic variance
149 divided by the phenotypic variance (sum of additive genetic-, pen-, litter and residual
150 variance). Genetic and residual correlations were obtained from multivariate analysis using
151 the same model as for univariate analysis. For the random effects of the pen and litter no co-
152 variance was assumed between traits. Groups of traits analyzed together corresponds to the
153 traits presented in table 3, table 4, and table 6 (tables with genetic correlations). Due to
154 convergence problems, the traits in table 5 (Landrace) had to be split in four blocs with the
155 following traits: 1) percentage primal cuts, 2) percent primal cuts with LMP and carcass yield,
156 3) percent primal cuts with muscle and backfat, 4) LMP, carcass yield, muscle and backfat.

157

158

RESULTS

159 Descriptive Statistics of Primal Cuts and Carcass Traits

160 Descriptive statistics are shown in Table 1 and 2. Ham was the largest primal cut for both
161 breeds, and loin was the smallest. Loin and ham were also the primal cuts with the largest
162 differences between breeds where Duroc had the largest ham, and Landrace the largest loin.
163 Duroc had also a slightly smaller belly and a larger shoulder compared to Landrace. Landrace
164 showed the largest variance in primal cut proportions. Duroc had a lower CLMP compared to
165 Landrace, and this was reflected in the LMP within each primal cut where the largest
166 differences were seen for LMP within loin and ham. Duroc had larger variation in lean
167 content (%) in loin, shoulder and belly compared to Landrace. Further, Duroc had a lower
168 measure of loin depth and a higher measure of backfat compared to Landrace. Duroc gave a
169 lower carcass yield compared to Landrace. The mean live weight was 120 kg deviating with

170 2.14 kg and 2.79 for Duroc and Landrace, respectively. The average number of days from 40
171 to 120 kg was 76.54 and 72.96, for Duroc and Landrace, respectively.

172

173 **Heritability estimates**

174 Heritability estimates (+/- standard errors) are presented in table 3-6 (diagonal elements). The
175 heritabilities for all primal cuts were intermediate to high for both breeds, h^2 ranging from
176 0.15 to 0.50 (table 3 and table 5). The lowest and highest heritabilities were found in Landrace
177 for shoulder and belly, respectively. The standard errors were relatively small (0.05-0.07) and
178 similar across cuts and breeds. Lean meat percentage had a high-intermediate heritability,
179 whereas carcass yield showed a lower heritability. Loin depth and backfat was also
180 intermediate heritable (table 3 and table 5). The heritability of LMP within each primal cut was
181 similar to the CLMP, although somewhat smaller for shoulder in Landrace (table 4 and table
182 6).

183

184 **Genetic correlations**

185 Genetic correlations close to zero were found for shoulder and ham (both breeds) and
186 shoulder and loin (Duroc) (table 3 and table 5). Negative genetic correlations were found
187 between most of the other primal cuts, and the strongest correlation was between belly and
188 ham. However, a weak, positive genetic correlation was found between ham and loin in
189 Landrace (table 5). CLMP showed a positive correlation to shoulder and ham, but was
190 negatively correlated to belly. Loin showed no genetic correlation to CLMP. Carcass yield
191 was positively correlated to ham and shoulder, but the correlation to loin was close to zero.
192 Correlation between carcass yield and belly was close to zero in Duroc and negative in
193 Landrace. Loin depth and loin was positively correlated in both breeds, but there was a
194 negative correlation between loin depth and shoulder. Belly and ham did not correlate to loin

195 depth in Landrace, but Duroc showed weak negative (belly) and positive (ham) correlations.
196 Backfat showed a positive correlation with belly, but a negative correlation with shoulder and
197 ham. For Duroc, there was a positive correlation between backfat and loin but this was not
198 seen in Landrace (table 3 and table 5).

199

200 LMP within each primal cut were strongly correlated to each other (0.83-0.91) and even more
201 so to CLMP (table 4 and table 6). Carcass yield showed intermediate genetic correlations to
202 LMP within all primal cuts for Landrace, but the same correlations were weak for Duroc.
203 Loin depth showed intermediate-strong genetic correlations to the LMP within each primal
204 for Landrace, but again, weaker correlations for Duroc. Backfat was consistently positively
205 correlated to LMP within each primal cut in both breeds (table 3-6).

206

207

DISCUSSION

208 The objective of the study was to identify and measure genetic parameters of primal cuts
209 and carcass composition measured *in vivo* by CT, in order to implement these traits in a
210 commercial breeding program for pigs. The focus of this study is the new phenotypes
211 obtained from atlas segmentation, and results for the other traits will only be discussed when
212 relevant for the new atlas based traits. The results show that the traits measured *in vivo* using
213 CT and atlas segmentation are intermediate to highly heritable.

214 To the best of our knowledge the use of atlas segmentation based on CT scans of boars
215 *in vivo* on an industrial scale is exclusive to the Topigs Norsvin genetic programs. In this
216 study, *in vivo* primal cuts from atlas segmentation are used for genetic parameter calculations
217 for the first time. The primal cuts have different value to the market and the value varies also
218 considerably between markets (European vs. Asian and North American). Primal cuts from *in*
219 *vivo* measurements implemented in the breeding goal could more efficiently breed for or sort

220 lines with the most profitable carcasses to different markets. *In vivo* measurements mean that
221 selection candidates will have own phenotype which is more efficient than post mortem
222 measurements where data is available only to relatives of the AI boars.

223 Moreover, the total phenotypic variance is for the calculation of heritability
224 decomposed into two parts; genetic and environmental variation. Variation coursed by
225 erroneous segmentation, atlas or manual, will add to the environmental part. Consequently,
226 precise segmentation is important in order to improve heritability estimates. As reported by
227 Nissen et al. (2006) substantial errors might also occur by manual segmentation. Recently,
228 Olsen et al. (2017) points out that the precision of CLMP obtained by CT has the potential of
229 outperforming the precision obtained by manual dissection. Under the assumption of flawless
230 atlas segmentation, these arguments should also be valid for primal cuts.

231 The results of the present study provide indirect support for the validity of atlas
232 segmentation. If the atlas segmentation offered no information, the heritability estimates
233 should be results of random noise. As we get heritability results in line with, or slightly
234 inferior to, heritability results obtained by manual segmentation, the precision of atlas
235 segmentation are indicated to be in line with, or at least close to the precision of manual
236 segmentation. The only comparison with manual primal cuts from other studies have been
237 possible so far, however, in another study results will be validated by comparing results from
238 *in vivo* segmented boars with results from manual segmentation.

239 In this paper, it was decided to use primal cuts as the relative proportions (%) of
240 carcass weight, in order to avoid confounding between carcass weight and primal cut yields.
241 This contrasts with previous studies on traits related to primal cuts, where the focus has been
242 on primal cut weights, with carcass weight included as a covariate in the models. The authors
243 find it more relevant to use proportions directly as a response in the models, as emphasis on

244 the proportion of valuable primal cuts in breeding goals will facilitate more valuable carcasses
245 and not just bigger carcasses (Moore et al., 2017).

246 The results for both breeds strongly indicate that selection for increased LMP will
247 decrease the percent belly, which is natural as the belly is the fattest primal cut, and per
248 definition the pig gets leaner when LMP increase. The loin is minimally affected by selection
249 for LMP, and the percent of shoulder and ham will increase when selecting for increased
250 LMP.

251 Nevertheless, a high CLMP means a carcass with high muscle to fat ratio. Belly is a
252 primal cut with high fat content, and the negative correlation between belly and CLMP is
253 therefore not surprising. Backfat and CLMP were negatively correlated which is natural as the
254 backfat measure is expected to be lower in pigs with high CLMP (lean pigs). The positive
255 correlation of backfat with the belly is therefor expected. The positive correlation of loin and
256 backfat in Duroc is a bit unexpected, and is not seen in Landrace. Correlations for backfat
257 with shoulder and ham showed the same pattern as CLMP indicating that pigs with high
258 backfat and/or high CMLP (lean pigs) have larger proportions of shoulder and ham compared
259 to fatter pigs. Carcass yield was the least heritable of the studied traits, and show that
260 environmental factors (e.g. measuring technique, trait definition) plays a major role for this
261 trait. Ham was the trait with the strongest correlation to carcass yield in both breeds
262 suggesting that selection for increased yield primarily leads to increased proportion of ham.
263 The negative correlation of carcass yield and belly reported for Landrace suggests that
264 increased yield will lead to a smaller proportion of belly in Landrace. The correlations for
265 CLMP and carcass yield with the primal cuts were generally intermediate to low, and this
266 indicates that these two measures does not accurately capture the variation in primal cut
267 composition. Loin depth was measured at one point of the loin, and the positive genetic
268 correlation to loin as a primal cut is therefor expected. The genetic and residual correlations

269 are however only intermediate (0.21-0.40) showing that loin depth may not be an optimal
270 indicator of the proportion loin in a carcass. Loin depth is an important predictor of carcass
271 value and composition in the industry today, and the results in this study show the need for a
272 revision of the use of loin depth to predict carcass value and composition. The CLMP seems
273 to be a good indicator of lean meat percentage within each primal as the genetic and residual
274 correlations were very high. The genetic correlations between the primal cuts with respect to
275 LMP were around 0.9 which is very high, but it will still be possible to change the lean
276 content more in some parts than in others.

277

278

IMPLICATIONS

279 Computed Tomography (CT) makes it possible to measure *in vivo* body or carcass
280 composition. This will aid the selection response by measuring on the candidates themselves
281 instead of using relatives. Primal cut proportion and composition measured *in vivo* by
282 computed tomography and atlas segmentation show heritable variation comparable to
283 previous post mortem studies. The detailed information on carcass composition can be used to
284 improve carcass value through selection using *in vivo* information. It can also be used to sort
285 i.e. boars used for artificial insemination, based on different markets worldwide. Future
286 studies should include the value of cuts in relation to different markets and regions, and *in*
287 *vivo* primal cut yields in relation to growth, feed intake and meat quality. Hopefully, the
288 PigAtlas will give us new insight on the relationship between *in vivo* body composition and
289 other traits in the breeding goal for pigs in the future.

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291

LITERATURE CITED

292 Cameron, N. D. 1990. Genetic and phenotypic parameters for carcass traits, meat and eating
293 quality traits in pigs. *Livest. Prod. Sci.* 26:119–135.

- 294 Campbell, A., W. Bain, A. McRae, T. Broad, P. Johnstone, K. Dodds, B. Veenvliet, G. Greer,
295 B. Glass, A. Beattie, and others. 2003. Bone density in sheep: genetic variation and
296 quantitative trait loci localisation. *Bone*. 33:540–548.
- 297 Gangsei, L. E., and J. Kongsro. 2016. Automatic segmentation of Computed Tomography
298 (CT) images of domestic pig skeleton using a 3D expansion of Dijkstra’s algorithm. *Comput.*
299 *Electron. Agric.* 121:191–194.
- 300 Gangsei, L. E., J. Kongsro, K. Olstad, E. Grindflek, and S. Sæbø. 2016a. Building an *in vivo*
301 anatomical atlas to close the phenomic gap in animal breeding. *Comput. Electron. Agric.*
302 127:739–743.
- 303 Gangsei, L. E., J. Kongsro, K. Olstad, E. Grindflek, and S. Sæbø. 2016b. Building an *in vivo*
304 anatomical atlas to close the phenomic gap in animal breeding. *Comput. Electron. Agric.*
305 127:739–743.
- 306 Gjerlaug-Enger, E., J. Kongsro, J. Odegård, L. Aass, and O. Vangen. 2012. Genetic
307 parameters between slaughter pig efficiency and growth rate of different body tissues
308 estimated by computed tomography in live boars of Landrace and Duroc. *Animal*. 6:9–18.
- 309 Herrero-Medrano, J. M., P. K. Mathur, J. ten Napel, H. Rashidi, P. Alexandri, E. F. Knol, and
310 H. A. Mulder. 2015. Estimation of genetic parameters and breeding values across challenged
311 environments to select for robust pigs¹. *J. Anim. Sci.* 93:1494–1502.
- 312 Horn, P. 1995. Using X-ray Computed Tomography to Predict Carcass Leanness in Pigs. In:
313 National Swine Improvement Federation Annual Conference. National Swine Improvement
314 Federation.
- 315 Madsen P and Jensen J. 2013. DMU A Package for Analysing Multivariate Mixed Models.
316 Available from: http://dmu.agrsci.dk/DMU/Doc/Current/dmuv6_guide.5.2.pdf
- 317 Miar, Y., G. S. Plastow, S. S. Moore, G. Manafiazar, P. Charagu, R. A. Kemp, B. Van
318 Haandel, A. E. Huisman, C. Y. Zhang, R. M. McKay, H. L. Bruce, and Z. Wang. 2014.

- 319 Genetic and phenotypic parameters for carcass and meat quality traits in commercial
320 crossbred pigs. *J. Anim. Sci.* 92:2869–2884.
- 321 Moore, K. L., R. Mrode, and M. P. Coffey. 2017. Genetic parameters of Visual Image
322 Analysis primal cut carcass traits of commercial prime beef slaughter animals. *Animal*. 1–7.
- 323 Newcom, D. W., T. J. Baas, J. W. Mabry, and R. N. Goodwin. 2002. Genetic parameters for
324 pork carcass components. *J. Anim. Sci.* 80:3099–3106.
- 325 Nissen, P. M. 2006. The estimated accuracy of the EU reference dissection method for pig
326 carcass classification. *Meat Sci.* 73:22–28.
- 327 Olsen, E. V., L. B. Christensen, and D. B. Nielsen. 2017. A review of computed tomography
328 and manual dissection for calibration of devices for pig carcass classification - Evaluation of
329 uncertainty. *Meat Sci.* 123:35–44.
- 330 Reimert, I., T. B. Rodenburg, W. W. Ursinus, B. Kemp, and J. E. Bolhuis. 2014. Responses to
331 novel situations of female and castrated male pigs with divergent social breeding values and
332 different backtest classifications in barren and straw-enriched housing. *Appl. Anim. Behav.*
- 333 Rydhmer, L., J. L. Gourdine, K. de Greef, and M. Bonneau. 2014. Evaluation of the
334 sustainability of contrasted pig farming systems: breeding programmes. *animal*. 8:2016–2026.
- 335 SAS. 2011. SAS/STAT 9.3 User's Guide. User's Guid. SAS Inst. Inc., Cary, NC. 8640.
- 336 Scholz, A. M., L. Bünger, J. Kongsro, U. Baulain, and a. D. Mitchell. 2015. Non-invasive
337 methods for the determination of body and carcass composition in livestock: dual-energy X-
338 ray absorptiometry, computed tomography, magnetic resonance imaging and ultrasound:
339 invited review. *Animal*. 9:1–15.
- 340 Skjervold, H., K. Grønseth, O. Vangen, and A. Evensen. 1981. *In vivo* estimation of body
341 composition by computerized tomography. *Z. Tierzuchtgsbiol.* 98:77–79.
- 342 Van Wijk, H. J., D. J. G. Arts, J. O. Matthews, M. Webster, B. J. Ducro, and E. F. Knol. 2005.
343 Genetic parameters for carcass composition and pork quality estimated in a commercial

344 production chain. *J. Anim. Sci.* 83:324–333.

345 Aasmundstad, T., J. Kongsro, M. Wetten, N. I. Dolvik, and O. Vangen. 2013.

346 Osteochondrosis in pigs diagnosed with computed tomography: heritabilities and genetic

347 correlations to weight gain in specific age intervals. *Animal.* 7:1576–82.

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350

351 **Tables and Figures**

352

353 Table 1. Summary statistics for traits. Duroc, n=2439.

Trait	Mean	SD	Min	Max	CV ¹
Shoulder (%)	31.05	0.85	28.18	34.33	2.74 %
Belly (%)	20.09	1.26	13.72	27.05	6.26 %
Loin (%)	13.40	0.69	10.27	22.42	5.17 %
Ham (%)	35.00	1.32	22.31	46.47	3.77 %
Shoulder lean (%)	59.55	3.32	44.57	71.48	5.57 %
Belly lean (%)	50.37	4.20	35.66	64.78	8.34 %
Loin lean (%)	61.99	3.03	50.85	71.96	4.89 %
Ham lean (%)	63.96	2.53	53.99	73.52	3.95 %
Carcass lean (%)	59.66	2.99	48.25	69.68	5.02 %
Carcass yield ² (%)	74.29	1.10	69.33	80.55	1.49 %
Backfat ³ (mm)	6.23	1.68	3	13	27.23 %
Loin ² (mm)	61.20	4.24	49	78	6.77 %

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¹ Coefficient of variation (CV) = SD/mean

² Carcass (kg) / Live animal (kg) x 100%

³ Backfat and loin depth over last rib

356 Table 2. Summary statistics for traits. Landrace, n=1998.

Trait	Mean	SD	Min	Max	CV ¹
Shoulder (%)	30.46	1.02	26.83	34.54	3.36 %
Belly (%)	20.50	1.34	15.88	26.27	6.56 %
Loin (%)	14.47	0.79	7.90	20.65	5.43 %
Ham (%)	34.07	1.35	26.84	45.63	3.97 %
Shoulder lean (%)	61.93	3.25	47.09	73.88	5.25 %
Belly lean (%)	55.04	4.46	38.46	69.98	8.11 %
Loin lean (%)	68.16	2.62	55.03	77.19	3.85 %
Ham lean (%)	69.27	2.74	50.42	77.97	3.95 %
Carcass lean (%)	63.32	2.95	47.04	73.53	4.66 %
Carcass yield ⁴ (%)	75.41	1.14	71.00	79.56	1.51 %
Backfat ⁵ (mm)	5.25	1.79	2	20	34.23 %
Loin ² (mm)	65.34	4.38	49	79	6.70 %

357

358

¹ Coefficient of variation (CV) = SD/mean

⁴ Carcass (kg) / Live animal (kg) x 100%

⁵ Backfat and loin depth over last rib

359 Table 3. Heritabilities (diagonal, bold) and genetic (below diagonal) and residual (above
 360 diagonal) correlations (with SE). Primal cuts, lean meat percentage carcass (CLMP) and
 361 carcass yield in (%), backfat and loin depth in millimeters. Duroc

Trait	Shoulder	Belly	Loin	Ham	CLMP	YIELD	BF	LD
Shoulder	.26 (.06)	-.28 (.04)	.08 (.05)	-.45 (.04)	-.07 (.07)	.24 (.04)	.10 (.05)	.00 (.06)
Belly	-.50 (.10)	.39 (.06)	-.15 (.05)	-.58 (.03)	-.12 (.07)	-.45 (.04)	.25 (.05)	.01 (.07)
Loin	-.02 (.13)	-.24 (.11)	.34 (.06)	-.44 (.04)	-.02 (.07)	-.10 (.05)	.03 (.05)	.21 (.06)
Ham	-.04 (.14)	-.68 (.07)	-.32 (.11)	.27 (.05)	.15 (.06)	.28 (.04)	-.28 (.04)	-.10 (.06)
CMLP	.38 (.11)	-.46 (.09)	-.05 (.10)	.31 (.11)	.58 (.07)	-.03 (.06)	-.23 (.07)	.17 (.09)
YIELD	.11 (.17)	.02 (.13)	.02 (.15)	.34 (.15)	.05 (.14)	.14 (.04)	.08 (.05)	.27 (.05)
BF	-.18 (.14)	.38 (.11)	.26 (.12)	-.47 (.11)	-.67 (.08)	.10 (.16)	.33 (.06)	.09 (.06)
LD	-.18 (.14)	-.23 (.11)	.26 (.11)	.22 (.13)	.21 (.10)	.50 (.13)	.05 (.12)	.49 (.07)

362

363 Table 4. Heritabilities (diagonal, bold) and genetic (below diagonal) and residual (above
 364 diagonal) correlations (with SE). Primal cut lean meat composition (%), CLMP and carcass
 365 yield in (%), backfat and loin depth in millimeters. Duroc

Trait	Shoulder	Belly	Loin	Ham	CLMP	YIELD	BF	LD
Shoulder LMP	.53 (.05)	.74 (.04)	.66 (.04)	.66 (.04)	.91 (.01)	-.09 (.01)	-.23 (.07)	.09 (.08)
Belly LMP	.89 (.02)	.55 (.07)	.79 (.03)	.67 (.04)	.91 (.01)	-.12 (.06)	-.20 (.07)	.08 (.09)
Loin LMP	.90 (.02)	.90 (.02)	.53 (.07)	.55 (.05)	.82 (.03)	.08 (.06)	-.27 (.07)	.29 (.08)
Ham LMP	.91 (.02)	.89 (.02)	.91 (.02)	.51 (.07)	.82 (.02)	-.07 (.06)	-.04 (.07)	.24 (.08)
CMLP	.97 (.01)	.96 (.01)	.95 (.01)	.96 (.01)	.58 (.07)	-.03 (.06)	-.23 (.07)	.17 (.09)
YIELD	.03 (.13)	-.03 (.13)	.19 (.13)	.05 (.13)	.06 (.14)	.14 (.04)	.08 (.05)	.27 (.05)
BF	-.63 (.08)	-.70 (.07)	-.62 (.08)	-.57 (.09)	-.66 (.08)	.11 (.16)	.33 (.06)	.07 (.07)
LD	.13 (.10)	.18 (.10)	.32 (.09)	.17 (.10)	.19 (.10)	.48 (.15)	.05 (.12)	.49 (.07)

366

367 Table 5. Heritabilities (diagonal, bold) and genetic (below diagonal) and residual (above
 368 diagonal) correlations (with SE). Primal cuts, lean meat percentage carcass (CLMP) and
 369 carcass yield in (%), backfat and loin depth in millimeters. Landrace

Trait	Shoulder	Belly	Loin	Ham	CLMP	YIELD	BF	LD
Shoulder	.15 (.05)	-.39 (.05)	-.16 (.04)	-.41 (.04)	.06 (.07)	.34 (.05)	-.08 (.07)	-.12 (.06)
Belly	-.32 (.15)	.51 (.07)	-.07 (.06)	-.43 (.05)	-.15 (.10)	-.40 (.07)	.25 (.09)	.07 (.08)
Loin	-.29 (.19)	-.35 (.14)	.21 (.06)	-.46 (.05)	.05 (.07)	-.26 (.05)	.15 (.07)	.23 (.06)
Ham	-.02 (.19)	-.87 (.05)	.12 (.17)	.32 (.07)	.04 (.08)	.22 (.06)	-.23 (.07)	-.10 (.07)
CMLP	.30 (.16)	-.42 (.10)	.05 (.15)	.32 (.13)	.50 (.07)	-.07 (.09)	-.10 (.11)	.05 (.10)
YIELD	.14 (.19)	-.25 (.13)	-.05 (.17)	.26 (.15)	.43 (.12)	.27 (.06)	.21 (.07)	.25 (.05)
BF	-.35 (.16)	.66 (.08)	.04 (.16)	-.57 (.11)	-.73 (.07)	-.30 (.14)	.41 (.07)	.35 (.10)
LD	-.21 (.19)	-.03 (.14)	.40 (.15)	-.02 (.15)	.61 (.11)	.61 (.11)	-.27 (.14)	.33 (.07)

370

371 Table 6. Heritabilities (diagonal, bold) and genetic (below diagonal) and residual (above
 372 diagonal) correlations (with SE). Primal cut lean meat composition (%) , CLMP and carcass
 373 yield in (%), backfat and loin depth in millimeters. Landrace

Trait	Shoulder	Belly	Loin	Ham	CLMP	YIELD	BF	LD
Shoulder LMP	.38 (.07)	.68 (.04)	.60 (.05)	.60 (.04)	.88 (.02)	-.06 (.07)	-.09 (.08)	-.04 (.08)
Belly LMP	.89 (.03)	.54 (.07)	.81 (.03)	.67 (.04)	.91 (.02)	-.04 (.08)	-.17 (.10)	-.04 (.10)
Loin LMP	.89 (.04)	.91 (.02)	.58 (.08)	.59 (.05)	.82 (.03)	.05 (.09)	-.19 (.10)	.13 (.10)
Ham LMP	.89 (.04)	.83 (.04)	.84 (.04)	.41 (.07)	.83 (.02)	-.15 (.07)	-.07 (.09)	.10 (.08)
CMLP	.97 (.01)	.96 (.01)	.94 (.02)	.93 (.02)	.58 (.07)	-.05 (.08)	-.10 (.10)	.02 (.09)
YIELD	.36 (.13)	.40 (.11)	.47 (.11)	.40 (.13)	.42 (.12)	.14 (.04)	0.24 (.07)	.18 (.08)
BF	-.63 (.09)	-.75 (.07)	-.71 (.07)	-.56 (.10)	-.72 (.07)	-.29 (.14)	0.33 (0.06)	.31 (.09)
LD	.59 (.12)	.55 (.11)	.71 (.09)	.69 (.10)	.64 (.10)	.59 (0.12)	-.25 (.14)	.49 (.07)

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1 Running head: Genetic parameters of CT traits

2

3 **Genetic parameters of *in vivo* primal cuts and body composition (PigAtlas) in pigs**
4 **measured by computed tomography (CT)¹**

5

6 **J. Kongsro, L. E. Gangsei, T. M. Karlsson-Drangsholt and E. Grindflek**

7

8 **ABSTRACT:**

9 Genetic parameters of *in vivo* primal cuts in breeding pigs using computed tomography were
10 estimated. 2439 Duroc and 1998 Landrace boars from the Topigs Norsvin boar testing station
11 in Norway were CT scanned as part of the genetic program. *In vivo* primal cuts were derived
12 from the CT images using atlas segmentation; the method called the Pig Atlas. The
13 (co)variance estimates were obtained from univariate (heritabilities) and multivariate
14 (correlations) animal genetic models using DMU software. The heritabilities for all primal
15 cuts proportions (%) were intermediate to large for both breeds, h^2 ranging from 0.15 to 0.50.
16 Negative genetic correlations were found between most of the other primal cuts, and the
17 strongest correlation was between belly and ham. Carcass lean meat percentage showed a
18 positive correlation to shoulder and ham, but was negatively correlated to belly. In this study,
19 *in vivo* primal cuts from atlas segmentation are used for genetic parameter calculations for the
20 first time. Computed Tomography (CT) makes it possible to measure *in vivo* body or carcass
21 composition. This will aid the selection response by measuring on the candidates themselves
22 instead of using relatives. Primal cut proportion and composition measured *in vivo* by

¹ The Norwegian Research Council is acknowledged for funding this project with grant #256316

23 computed tomography and atlas segmentation show heritable variation comparable to
24 previous post mortem studies.

25

26 **Key words:** pigs, body composition, genetic parameters, primal cuts, computed tomography,
27 PigAtlas

28

29

INTRODUCTION

30 The most important traits in pig breeding have for decades been related to efficient lean meat
31 production, i.e. selection for increased growth rate, feed efficiency and low fat carcass
32 composition (Cameron, 1990). However, breeding goals have been changing from focus on
33 cost reduction toward retail carcass yield and meat quality (Miar et al., 2014), and recently
34 including sustainability (Rydhmer et al., 2014), social breeding values (Reimert et al., 2014)
35 and robustness (Herrero-Medrano et al., 2015). Carcass yield is still highly valid in breeding
36 goals, but in order to give more space to the more recent traits, we need to find more efficient
37 ways of measuring carcass traits *in vivo*, bypass the use of sibs or half-sibs in order to
38 measure these traits on the selection candidates themselves. *In vivo* or non-invasive measures
39 of carcass composition can be derived either indirectly by ultrasound and other
40 anthropometric measures like body conformation score, or more directly by Computed
41 Tomography (CT), Magnetic Resonance Imaging (MRI) and dual-energy X-ray
42 absorptiometry (DXA or DEXA) (Scholz et al., 2015). Heritability, h^2 , is defined as the
43 genetic contributions to a population's phenotypic variance. Heritabilities of the primal cuts
44 ham, loin, belly and shoulder, based on dissected carcasses, are found to be medium to large.
45 For example, the ham, belly and loin have been reported to range from 0.29 to 0.57 in
46 previous studies, and for deboned cuts, the heritabilities were even greater ($h^2 > 0.70$)
47 (Newcom et al., 2002; Van Wijk et al., 2005). Previous studies have shown that primal cut

48 traits are highly heritable on the carcass level (post mortem), however the genetic parameters
49 for *in vivo* primal cuts and carcass composition remain elusive. Today, CT is used in pig
50 breeding, and in order to handle large amounts of data and animals, more automation is
51 needed in the process of virtually dissect the pig based on CT. In order to aid the automated
52 process, the authors have recently looked at developing a pig atlas based on CT (Gangsei et
53 al., 2016a; Gangsei and Kongsro, 2016). The Pig Atlas is a digital map based on computed
54 tomography body scans of a subset of ~500 pigs from the population of purebred Landrace
55 and Duroc pigs in Norway. The Pig Atlas is an average pig based on that particular subset.
56 Volumes of CT scanned pigs are continuously fitted onto the pig atlas and deformed in order
57 to transfer information from the atlas to the new animal. The primal cuts are identified using
58 the already established “landmarks”, and the pig is brought back to its original shape. This has
59 proven to be an effective way of recording primal cut data from a large number of animals in
60 an automated way (Gangsei et al., 2016a). This is, to our knowledge, the first attempt to study
61 *in vivo* carcass and primal cut composition using CT in a commercial breeding program.
62 Study of genetic parameters for *in vivo* carcass and primal cut composition is necessary in
63 order to implement these traits in a commercial breeding program for pigs. Thus, the aim of
64 this study was to estimate genetic parameters of *in vivo* primal cuts and carcass composition
65 in breeding pigs measured by atlas segmented images from computed tomography (CT).

66

67

MATERIALS AND METHODS

68 All animals were cared for according to the laws and regulations for keeping pigs in Norway
69 (Regulation for the keeping of pigs in Norway 2003-02-18-175, 2003; Animal welfare Act
70 2009-06-19-97, 2009).

71 *Animals*

72 Two thousand, four hundred thirty-nine (2439) purebred Duroc and 1998 Landrace
73 boars from the boar testing station, Norsvin Delta in Norway, were CT scanned as part of the
74 Topigs Norsvin genetic program as described more in detail by Aasmundstad et al., 2013. The
75 purebred boars were born and raised to 25-30 kg in different nucleus herds located in Norway.
76 As part of the test program, the pigs are sent to the boar testing station, Norsvin Delta, when
77 reaching 25-30 kg. The test includes feed and weight recordings, score of exterior traits and
78 CT scanning at the end of test when they reach 120 kg as described more in detail by
79 Gjerlaug-Enger et al., 2012.

80

81 ***CT scanning***

82 The pigs were CT scanned using a GE Healthcare VCT 32 scanner at 120 kg body
83 weight. The protocol used was optimized for soft tissue using a slice thickness of 1.25 mm,
84 120 kV and dynamic mA based on the thickness of the slice. Prior to CT scanning, the boars
85 were sedated using Azaperone, 8 mg/kg live weight (Stresnil Vet[®], Janssen-Cilag Ltd,
86 Buckinghamshire, UK) administered intramuscularly. Boars were scanned approximately 45
87 minutes after injection, as the sedation was given to keep the boars calm during scanning.

88

89 ***Image analysis and atlas segmentation***

90 Atlas segmentation (Gangsei et al., 2016a; Gangsei et al., 2016b) was the means for
91 obtaining new and better phenotypes *in vivo* to improve accuracy and potentially increase
92 heritabilities for the traits of interest. By atlas segmentation every voxel was assigned to one
93 of the primal cuts, i.e. ham, belly, loin (including tenderloin) or shoulder, or as non-
94 commercial parts, i.e. internal organs, testicles etc. The labelled atlas was based on expert
95 segmentation of CT images (Gangsei et al., 2016a). After atlas segmentation all voxels in the
96 primal cuts were classified into the tissue classes; lean meat, fat or bone. Voxels with HU >

97 200 were classified as bone. Thereafter, in order to reduce noise, a simple filter was applied to
98 the soft tissue ($HU < 200$). HU values used for tissue classification of soft tissue voxels were
99 set to the minimum HU value of a 2×2 sliding frame within each 2D image slice. Soft tissue
100 voxels with filtered $HU > 0$ were classified as lean meat and the rest were classified as fat.
101 The weight of each voxel in the primal cuts was calculated as the product of its volume and
102 density, where the density was set based on a simple linear relationship ($1.0062 + 0.0016 \times$
103 HU) (Campbell et al., 2003). Thereafter, the total weights of meat, fat and bone within each
104 primal cut were easily calculated as the sum of weights of voxels were classified to the actual
105 cut and tissue. These weights constitute the basis for all genetic parameter calculations in this
106 study.

107

108 ***Traits***

109 Carcass weight was defined as the sum of primal cuts (ham, belly, loin and shoulder),
110 excluding the head. The primal cuts ham, belly, loin and shoulder were obtained as described
111 above, and were expressed as primal cut weights (absolute weights in kg) or as relative primal
112 yield (relative proportions of carcass weight in %). Lean meat in each of the primal cuts was
113 also expressed as a percentage of lean meat (LMP) in the specific primal cut. Carcass lean
114 meat was expressed as carcass lean meat percentage (CLMP). Carcass yield was defined as a
115 ratio of carcass weight (sum of all primal cuts) over live weight estimated from the entire CT
116 volume of the pig. Loin depth was defined as muscle (*m. longissimus*) depth in cm at the last
117 rib measured on CT images. Backfat depth was defined as subcutaneous fat layer over loin
118 depth at the last rib measured on CT. Both backfat and loin depth was measured by ruler on
119 the images across the widest part of the *m. longissimus* muscle approximately 7 cm off the
120 midline of the animal.

121

122 ***Statistical analysis***

123 The difference in phenotypic means between Duroc and Landrace was analyzed by a student
 124 t-test in SAS (SAS, 2011). Duroc and Landrace were analyzed separately as there is no
 125 common pedigree. The (co)variance estimates were obtained from univariate (heritabilities)
 126 and multivariate (correlations) animal genetic models. The DMU software (Madsen P and
 127 Jensen J, 2013) was used. The model for each trait was defined as:

$$y_{ijklmno} = PARITY_i + HY_j + MONTH_k + SECTION_l + \beta_1 * LW_{ijklmno} \\ + \beta_2 * LW_{ijklmno}^2 + pen_m + litter_n + animal_o + error_{ijklmno}$$

128 Where Y is the studied trait(s). For all y, PARITY (parity of dam) included four levels (i = 1,
 129 2, 3 or a later parity, unknown parity), HY (herd of origin and year of birth) included j = 35
 130 levels for Duroc and 108 levels for Landrace. Mmonth born in (MONTH) included k = 12
 131 levels. The section in the barn at the testing station (SECTION) included l= 74 levels for
 132 Duroc and 77 for Landrace. The live body weight in kg from CT scanning of each animal
 133 (LW) was fitted as a fixed linear (β_1) and quadratic (β_2) regression coefficients. The effects
 134 of PARITY, HY, MONTH and SECTION were treated as fixed effects. Pen (at the testing
 135 station), litter (common environmental effect for full sibs), animal and error were treated as
 136 random effects. The additive genetic effects were expected $\sim N(0, G \otimes A)$ and both the pen,
 137 litter and residual effect were expected where A is the additive relationship matrix, G is the
 138 additive genetic (co) variance matrix, I is an identity matrix of dimension equal to the number
 139 of animals with phenotypic records and R is the residual (co)variance matrix. Heritability
 140 estimates were obtained from univariate analysis, and heritability was defined as additive
 141 genetic variance divided by the phenotypic variance (sum of additive genetic-, pen-, litter and
 142 residual variance). Genetic and residual correlations were obtained from multivariate analysis
 143 using the same model as for univariate analysis. For the random effects of the pen and litter
 144 no co-variance was assumed between traits. Groups of traits analyzed together corresponds to

145 the traits presented in table 3, table 4, and table 6 (tables with genetic correlations). Due to
146 convergence problems, the traits in table 5 (Landrace) had to be split in four blocs with the
147 following traits: 1) percentage primal cuts, 2) percent primal cuts with LMP and carcass yield,
148 3) percent primal cuts with muscle and backfat, 4) LMP, carcass yield, muscle and backfat.

149

150

RESULTS

151 Descriptive Statistics of Primal Cuts and Carcass Traits

152 Descriptive statistics are shown in Table 1 and 2. Ham was the greatest primal cut for both
153 breeds, and loin was the smallest. Loin and ham were also the primal cuts with the greatest
154 differences between breeds where Duroc had the greatest ham, and Landrace the greatest loin.

155 Duroc had also a slightly smaller belly and a greater shoulder compared to Landrace.

156 Landrace showed the greatest variance in primal cut proportions. Duroc had a smaller CLMP

157 compared to Landrace, and this was reflected in the LMP within each primal cut where the

158 greatest differences were seen for LMP within loin and ham. Duroc had greater variation in

159 lean content (%) in loin, shoulder and belly compared to Landrace. Further, Duroc had a

160 smaller measure of loin depth and a greater measure of backfat compared to Landrace. Duroc

161 gave a smaller carcass yield compared to Landrace. The mean live weight was 120 kg

162 deviating with 2.14 kg and 2.79 for Duroc and Landrace, respectively. The average number of

163 days from 40 to 120 kg was 76.54 and 72.96, for Duroc and Landrace, respectively.

164

165 Heritability estimates

166 The heritability estimates (+/- standard errors) are presented in table 3-6 (diagonal elements).

167 The heritabilities for all primal cuts were intermediate to large for both breeds, h^2 ranging

168 from 0.15 to 0.50 (table 3 and table 5). The smallest and largest heritabilities were found in

169 Landrace for shoulder and belly, respectively. The standard errors were relatively small (0.05-

170 0.07) and similar across cuts and breeds. Lean meat percentage had a large-intermediate
171 heritability, whereas carcass yield showed a smaller heritability. Loin depth and backfat was
172 also intermediate heritable (table 3 and table 5). The heritability of LMP within each primal cut
173 was similar to the CLMP, although somewhat smaller for shoulder in Landrace (table 4 and
174 table 6).

175

176 **Genetic correlations**

177 Genetic correlations close to zero were found for shoulder and ham (both breeds) and
178 shoulder and loin (Duroc) (table 3 and table 5). Negative genetic correlations were found
179 between most of the other primal cuts, and the strongest correlation was between belly and
180 ham. However, a weak, positive genetic correlation was found between ham and loin in
181 Landrace (table 5). The carcass lean meat percentage showed a positive correlation to
182 shoulder and ham, but was negatively correlated to belly. Loin showed no genetic correlation
183 to CLMP. Carcass yield was positively correlated to ham and shoulder, but the correlation to
184 loin was close to zero. Correlation between carcass yield and belly was close to zero in Duroc
185 and negative in Landrace. Loin depth and loin was positively correlated in both breeds, but
186 there was a negative correlation between loin depth and shoulder. Belly and ham did not
187 correlate to loin depth in Landrace, but Duroc showed weak negative (belly) and positive
188 (ham) correlations. Backfat showed a positive correlation with belly, but a negative
189 correlation with shoulder and ham. For Duroc, there was a positive correlation between
190 backfat and loin but this was not seen in Landrace (table 3 and table 5).

191

192 The lean meat percentage within each primal cut were strongly correlated to each other (0.83-
193 0.91) and even more so to CLMP (table 4 and table 6). Carcass yield showed intermediate
194 genetic correlations to LMP within all primal cuts for Landrace, but the same correlations

195 were weak for Duroc. Loin depth showed intermediate-strong genetic correlations to the LMP
196 within each primal for Landrace, but again, weaker correlations for Duroc. Backfat was
197 negatively correlated to LMP in each primal cut in both breeds (table 3-6).

198

199

DISCUSSION

200 The objective of the study was to identify and measure genetic parameters of primal cuts
201 and carcass composition measured *in vivo* by CT, in order to implement these traits in a
202 commercial breeding program for pigs. The focus of this study is the new phenotypes
203 obtained from atlas segmentation, and results for the other traits will only be discussed when
204 relevant for the new atlas based traits. The results show that the traits measured *in vivo* using
205 CT and atlas segmentation are intermediate to highly heritable.

206 To the best of our knowledge, this is the first attempt to apply the use of atlas
207 segmentation based on CT scans of boars *in vivo* in a commercial genetic program. In this
208 study, *in vivo* primal cuts from atlas segmentation are used for genetic parameter calculations
209 for the first time. The primal cuts have different market values and their value varies also
210 considerably between markets (European vs. Asian and North American). The relative value
211 change; i.e. belly becoming increasingly more valuable relative to loin and ham, particularly
212 in the US markets. Primal cuts from *in vivo* measurements implemented in the breeding goal
213 could more efficiently breed for or sort lines with the most profitable carcasses to different
214 markets. *In vivo* measurements mean that selection candidates will have own phenotype
215 which is more efficient than post mortem measurements where data is available only to
216 relatives of the AI boars.

217 Moreover, the total phenotypic variance is for the calculation of heritability
218 decomposed into two parts; genetic and environmental variation. Variation coursed by
219 erroneous segmentation, atlas or manual, will add to the environmental part. Consequently,

220 precise segmentation is important in order to improve heritability estimates. Correlation
221 within the residuals like side effects (left / right) are not regarded in this study as a source of
222 error since we use both sides or whole *in vivo* carcasses, not half carcasses like in other post
223 mortem studies. As reported by Nissen et al. (2006) substantial errors might also occur by
224 manual segmentation. Recently, Olsen et al. (2017) points out that the precision of CLMP
225 obtained by CT has the potential of outperforming the precision obtained by manual
226 dissection. Under the assumption of flawless atlas segmentation, these arguments should also
227 be valid for primal cuts.

228 Furthermore, when calculating the weight of each voxel as a product of its volume and
229 density, a formula by Campbell et al. (2003) was used. This formula was tested for sheep, and
230 there is no such formula for specifically for pigs. There might be unexplained variation,
231 especially in pig fat tissue, due to variable lipid content, saturation and water in fat. A
232 validation of the equation might be needed in order to improve the accuracy. However, the
233 authors made an approximation that this formula was valid also for pigs, since the variation is
234 picked up by the variation in HU. However, further studies is recommended in order to
235 validate the relationship between HU and density for a wide range of species, in particular
236 farmed animals, and HU levels.

237 The results of the present study provide indirect support for the validity of atlas
238 segmentation. If the atlas segmentation offered no information, the heritability estimates
239 should be results of random noise. Heritability for ham was 0.27 for Duroc and 0.32 for
240 Landrace which is a bit lower than reported by other studies based on manual segmentation
241 Newcom et al. (2002) and Van Wijik et al. (2002) ($h^2=0.57$ and 0.40 , respectively). Further,
242 Newcom et al. (2002) and Van Wijik et al. (2002) found heritability of 0.29 and 0.51 for loin,
243 and our results are in between for Landrace ($h^2=0.34$) and a bit smaller for Duroc ($h^2=0.21$).
244 For belly, Newcom et al. (2002) reported heritability of 0.51 which is same as we found for

Comment [TMK1]: Jeg har forsøkt meg på litt mer detaljer. Kan forsøke å inkludere Hermesch når jeg får tilgang til den.

245 Landrace, but again the heritability for Duroc was a bit smaller ($h^2=0.39$). With respect to the
246 composition of cuts, Hermesch et al. found a heritability of 0.34 on predicted fat percentage
247 of the belly by image analysis of a belly image of the anterior side (Hermesch and O'Shea,
248 2005; Hermesch, 2008). This is slightly lower than the heritability of Belly LMP found in this
249 paper (0.55), which might be related to the accuracy of methods, since Hermesch et al. stated
250 that it might difficult to take quality images of bellies from lean carcasses. As we get
251 heritability results in line with, or slightly inferior to, heritability results obtained by manual
252 segmentation, the precision of atlas segmentation are indicated to be in line with, or at least
253 close to the precision of manual segmentation. The only comparison with manual primal cuts
254 from other studies have been possible so far, however, in another study results will be
255 validated by comparing results from *in vivo* segmented boars with results from manual
256 segmentation.

257 In this paper, it was decided to use primal cuts as the relative proportions (%) of
258 carcass weight, in order to avoid confounding between carcass weight and primal cut yields.
259 This contrasts with previous studies on traits related to primal cuts, where the focus has been
260 on primal cut weights, with carcass weight included as a covariate in the models. The authors
261 find it more relevant to use proportions directly as a response in the models, as emphasis on
262 the proportion of valuable primal cuts in breeding goals will facilitate more valuable carcasses
263 and not just bigger carcasses (Moore et al., 2017).

264 The results for both breeds strongly indicate that selection for increased LMP will
265 decrease the percent belly, which is natural as the belly is the fattest primal cut, and per
266 definition the pig gets leaner when LMP increase. The loin is minimally affected by selection
267 for LMP, and the percent of shoulder and ham will increase when selecting for increased
268 LMP.

269 Nevertheless, a large CLMP means a carcass with high muscle to fat ratio. Belly is a
270 primal cut with large fat content, and the negative correlation between belly and CLMP is
271 therefore not surprising. Backfat and CLMP were negatively correlated which is natural as the
272 backfat measure is expected to be smaller in pigs with high CLMP (lean pigs). The positive
273 correlation of backfat with the belly is therefor expected. The positive correlation of loin and
274 backfat in Duroc is a bit unexpected, and is not seen in Landrace. Correlations for backfat
275 with shoulder and ham showed the same pattern as CLMP indicating that pigs with large
276 backfat and/or large CMLP (lean pigs) have greater proportions of shoulder and ham
277 compared to fatter pigs. Carcass yield was the least heritable of the studied traits, and show
278 that environmental factors (e.g. measuring technique, trait definition) plays a major role for
279 this trait. Ham was the trait with the strongest correlation to carcass yield in both breeds
280 suggesting that selection for increased yield primarily leads to increased proportion of ham.
281 The negative correlation of carcass yield and belly reported for Landrace suggests that
282 increased yield will lead to a smaller proportion of belly in Landrace. The correlations for
283 CLMP and carcass yield with the primal cuts were generally intermediate to small, and this
284 indicates that these two measures does not accurately capture the variation in primal cut
285 composition. Loin depth was measured at one point of the loin, and the positive genetic
286 correlation to loin as a primal cut is therefor expected. The genetic and residual correlations
287 are however only intermediate (0.21-0.40) showing that loin depth may not be an optimal
288 indicator of the proportion loin in a carcass. Loin depth is an important predictor of carcass
289 value and composition in the industry today, and the results in this study show the need for a
290 revision of the use of loin depth to predict carcass value and composition. The carcass lean
291 meat percentage seems to be a good indicator of lean meat percentage within each primal as
292 the genetic and residual correlations were very strong. The genetic correlations between the
293 primal cuts with respect to LMP were around 0.9 which is very strong, but it will still be

294 possible to change the lean content more in some parts than in others. Some primal cuts like
295 shoulder and belly might also be used as predictors of carcass lean meat if a whole carcass is
296 difficult to obtain or to reduce costs of dissection.

297 On the other hand, the percentages of primal cuts change as the animals mature, i.e.
298 the weights of the cuts have different allometric growths. Less mature animal will
299 consequently have more ham and leaner hams, and lower percent and leaner bellies. The
300 genetic correlations may to some extent reflect variation in the degree of maturity of pigs
301 within the genetic population. However, since we do not have performed this study over the
302 entire growth period of the pig, only using a fixed target weight of 120 kg, the effect of
303 maturity remains elusive.

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IMPLICATIONS

306 Computed Tomography (CT) makes it possible to measure *in vivo* body or carcass
307 composition. This will aid the selection response by measuring on the candidates themselves
308 instead of using relatives. Primal cut proportion and composition measured *in vivo* by
309 computed tomography and atlas segmentation show heritable variation comparable to
310 previous post mortem studies. The detailed information on carcass composition can be used to
311 improve carcass value through selection using *in vivo* information. It can also be used to sort
312 i.e. boars used for artificial insemination, based on different markets worldwide. Future
313 studies should include the value of cuts in relation to different markets and regions, and *in*
314 *vivo* primal cut yields in relation to growth, feed intake and meat quality. Hopefully, the
315 PigAtlas will give us new insight on the relationship between *in vivo* body composition and
316 other traits in the breeding goal for pigs in the future.

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LITERATURE CITED

- 319 Cameron, N. D. 1990. Genetic and phenotypic parameters for carcass traits, meat and eating
320 quality traits in pigs. *Livest. Prod. Sci.* 26:119–135..
- 321 Campbell, A., W. Bain, A. McRae, T. Broad, P. Johnstone, K. Dodds, B. Veenvliet, G. Greer,
322 B. Glass, A. Beattie, and others. 2003. Bone density in sheep: genetic variation and
323 quantitative trait loci localisation. *Bone.* 33:540–548.
- 324 Gangsei, L. E., and J. Kongsro. 2016. Automatic segmentation of Computed Tomography
325 (CT) images of domestic pig skeleton using a 3D expansion of Dijkstra’s algorithm. *Comput.*
326 *Electron. Agric.* 121:191–194.
- 327 Gangsei, L. E., J. Kongsro, K. Olstad, E. Grindflek, and S. Sæbø. 2016a. Building an in vivo
328 anatomical atlas to close the phenomic gap in animal breeding. *Comput. Electron. Agric.*
329 127:739–743.
- 330 Gangsei, L. E., J. Kongsro, K. Olstad, E. Grindflek, and S. Sæbø. 2016b. Building an in vivo
331 anatomical atlas to close the phenomic gap in animal breeding. *Comput. Electron. Agric.*
332 127:739–743.
- 333 Gjerlaug-Enger, E., J. Kongsro, J. Odegård, L. Aass, and O. Vangen. 2012. Genetic
334 parameters between slaughter pig efficiency and growth rate of different body tissues
335 estimated by computed tomography in live boars of Landrace and Duroc. *Animal.* 6:9–18.
- 336 Hermesch, S. 2008. Genetic relationships between composition of pork bellies and
337 performance, carcass and meat quality traits. *Animal.* 2:1178–1185.
- 338 Hermesch, S., and J. M. O’Shea. 2005. Genetic Parameters for Characteristics of Pork Bellies.
339 In: *Proc. Assoc. Advmt. Anim. Breed. Genet.* Vol. 16. p. 137–140.
- 340 Herrero-Medrano, J. M., P. K. Mathur, J. ten Napel, H. Rashidi, P. Alexandri, E. F. Knol, and
341 H. A. Mulder. 2015. Estimation of genetic parameters and breeding values across challenged
342 environments to select for robust pigs1. *J. Anim. Sci.* 93:1494–1502.
- 343 Madsen P and Jensen J. 2013. DMU A Package for Analysing Multivariate Mixed Models.

344 Available from: http://dmu.agrsci.dk/DMU/Doc/Current/dmuv6_guide.5.2.pdf

345 Miar, Y., G. S. Plastow, S. S. Moore, G. Manafiazar, P. Charagu, R. A. Kemp, B. Van
346 Haandel, A. E. Huisman, C. Y. Zhang, R. M. McKay, H. L. Bruce, and Z. Wang. 2014.
347 Genetic and phenotypic parameters for carcass and meat quality traits in commercial
348 crossbred pigs. *J. Anim. Sci.* 92:2869–2884.

349 Moore, K. L., R. Mrode, and M. P. Coffey. 2017. Genetic parameters of Visual Image
350 Analysis primal cut carcass traits of commercial prime beef slaughter animals. *Animal*. 1–7.

351 Newcom, D. W., T. J. Baas, J. W. Mabry, and R. N. Goodwin. 2002. Genetic parameters for
352 pork carcass components. *J. Anim. Sci.* 80:3099–3106.

353 Olsen, E. V., L. B. Christensen, and D. B. Nielsen. 2017. A review of computed tomography
354 and manual dissection for calibration of devices for pig carcass classification - Evaluation of
355 uncertainty. *Meat Sci.* 123:35–44.

356 Reimert, I., T. B. Rodenburg, W. W. Ursinus, B. Kemp, and J. E. Bolhuis. 2014. Responses to
357 novel situations of female and castrated male pigs with divergent social breeding values and
358 different backtest classifications in barren and straw-enriched housing. *Appl. Anim. Behav.*
359 *Sci.* 151:24–35.

360 Rydhmer, L., J. L. Gourdine, K. de Greef, and M. Bonneau. 2014. Evaluation of the
361 sustainability of contrasted pig farming systems: breeding programmes. *animal*. 8:2016–2026.

362 SAS. 2011. SAS/STAT 9.3 User's Guide. User's Guid. SAS Inst. Inc., Cary, NC. 8640.

363 Scholz, A. M., L. Bünger, J. Kongsro, U. Baulain, and a. D. Mitchell. 2015. Non-invasive
364 methods for the determination of body and carcass composition in livestock: dual-energy X-
365 ray absorptiometry, computed tomography, magnetic resonance imaging and ultrasound:
366 invited review. *Animal*. 9:1–15.

367 Van Wijk, H. J., D. J. G. Arts, J. O. Matthews, M. Webster, B. J. Ducro, and E. F. Knol. 2005.
368 Genetic parameters for carcass composition and pork quality estimated in a commercial

369 production chain. *J. Anim. Sci.* 83:324–333.

370 Aasmundstad, T., J. Kongsro, M. Wetten, N. I. Dolvik, and O. Vangen. 2013.

371 Osteochondrosis in pigs diagnosed with computed tomography: heritabilities and genetic

372 correlations to weight gain in specific age intervals. *Animal.* 7:1576–82.

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376 **Tables and Figures**

377

378 Table 1. Summary statistics for traits. Duroc, n=2439.

Trait	Mean	SD	Min	Max	CV ¹
Shoulder (%)	31.05	0.85	28.18	34.33	2.74 %
Belly (%)	20.09	1.26	13.72	27.05	6.26 %
Loin (%)	13.40	0.69	10.27	22.42	5.17 %
Ham (%)	35.00	1.32	22.31	46.47	3.77 %
Shoulder lean (%)	59.55	3.32	44.57	71.48	5.57 %
Belly lean (%)	50.37	4.20	35.66	64.78	8.34 %
Loin lean (%)	61.99	3.03	50.85	71.96	4.89 %
Ham lean (%)	63.96	2.53	53.99	73.52	3.95 %
Carcass lean (%)	59.66	2.99	48.25	69.68	5.02 %
Carcass yield ² (%)	74.29	1.10	69.33	80.55	1.49 %
Backfat ³ (mm)	6.23	1.68	3	13	27.23 %
Loin ² (mm)	61.20	4.24	49	78	6.77 %

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¹ Coefficient of variation (CV) = SD/mean² Carcass (kg) / Live animal (kg) x 100%³ Backfat and loin depth over last rib

381 Table 2. Summary statistics for traits. Landrace, n=1998.

Trait	Mean	SD	Min	Max	CV ¹
Shoulder (%)	30.46	1.02	26.83	34.54	3.36 %
Belly (%)	20.50	1.34	15.88	26.27	6.56 %
Loin (%)	14.47	0.79	7.90	20.65	5.43 %
Ham (%)	34.07	1.35	26.84	45.63	3.97 %
Shoulder lean (%)	61.93	3.25	47.09	73.88	5.25 %
Belly lean (%)	55.04	4.46	38.46	69.98	8.11 %
Loin lean (%)	68.16	2.62	55.03	77.19	3.85 %
Ham lean (%)	69.27	2.74	50.42	77.97	3.95 %
Carcass lean (%)	63.32	2.95	47.04	73.53	4.66 %
Carcass yield ⁴ (%)	75.41	1.14	71.00	79.56	1.51 %
Backfat ⁵ (mm)	5.25	1.79	2	20	34.23 %
Loin ² (mm)	65.34	4.38	49	79	6.70 %

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¹ Coefficient of variation (CV) = SD/mean

⁴ Carcass (kg) / Live animal (kg) x 100%

⁵ Backfat and loin depth over last rib

384 Table 3. Heritabilities (diagonal, bold) and genetic (below diagonal) and residual (above
 385 diagonal) correlations (with SE). Primal cuts, lean meat percentage carcass (CLMP) and
 386 carcass yield in (%), backfat and loin depth in millimeters. Duroc

Trait	Shoulder	Belly	Loin	Ham	CLMP	YIELD	BF	LD
Shoulder	.26 (.06)	-.28 (.04)	.08 (.05)	-.45 (.04)	-.07 (.07)	.24 (.04)	.10 (.05)	.00 (.06)
Belly	-.50 (.10)	.39 (.06)	-.15 (.05)	-.58 (.03)	-.12 (.07)	-.45 (.04)	.25 (.05)	.01 (.07)
Loin	-.02 (.13)	-.24 (.11)	.34 (.06)	-.44 (.04)	-.02 (.07)	-.10 (.05)	.03 (.05)	.21 (.06)
Ham	-.04 (.14)	-.68 (.07)	-.32 (.11)	.27 (.05)	.15 (.06)	.28 (.04)	-.28 (.04)	-.10 (.06)
CMLP	.38 (.11)	-.46 (.09)	-.05 (.10)	.31 (.11)	.58 (.07)	-.03 (.06)	-.23 (.07)	.17 (.09)
YIELD	.11 (.17)	.02 (.13)	.02 (.15)	.34 (.15)	.05 (.14)	.14 (.04)	.08 (.05)	.27 (.05)
BF	-.18 (.14)	.38 (.11)	.26 (.12)	-.47 (.11)	-.67 (.08)	.10 (.16)	.33 (.06)	.09 (.06)
LD	-.18 (.14)	-.23 (.11)	.26 (.11)	.22 (.13)	.21 (.10)	.50 (.13)	.05 (.12)	.49 (.07)

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388 Table 4. Heritabilities (diagonal, bold) and genetic (below diagonal) and residual (above
 389 diagonal) correlations (with SE). Primal cut lean meat composition (%), CLMP and carcass
 390 yield in (%), backfat and loin depth in millimeters. Duroc

Trait	Shoulder LMP	Belly LMP	Loin LMP	Ham LMP	CLMP	YIELD	BF	LD
Shoulder LMP	.53 (.05)	.74 (.04)	.66 (.04)	.66 (.04)	.91 (.01)	-.09 (.01)	-.23 (.07)	.09 (.08)
Belly LMP	.89 (.02)	.55 (.07)	.79 (.03)	.67 (.04)	.91 (.01)	-.12 (.06)	-.20 (.07)	.08 (.09)
Loin LMP	.90 (.02)	.90 (.02)	.53 (.07)	.55 (.05)	.82 (.03)	.08 (.06)	-.27 (.07)	.29 (.08)
Ham LMP	.91 (.02)	.89 (.02)	.91 (.02)	.51 (.07)	.82 (.02)	-.07 (.06)	-.04 (.07)	.24 (.08)
CMLP	.97 (.01)	.96 (.01)	.95 (.01)	.96 (.01)	.58 (.07)	-.03 (.06)	-.23 (.07)	.17 (.09)
YIELD	.03 (.13)	-.03 (.13)	.19 (.13)	.05 (.13)	.06 (.14)	.14 (.04)	.08 (.05)	.27 (.05)
BF	-.63 (.08)	-.70 (.07)	-.62 (.08)	-.57 (.09)	-.66 (.08)	.11 (.16)	.33 (.06)	.07 (.07)
LD	.13 (.10)	.18 (.10)	.32 (.09)	.17 (.10)	.19 (.10)	.48 (.15)	.05 (.12)	.49 (.07)

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392 Table 5. Heritabilities (diagonal, bold) and genetic (below diagonal) and residual (above
 393 diagonal) correlations (with SE). Primal cuts, lean meat percentage carcass (CLMP) and
 394 carcass yield in (%), backfat and loin depth in millimeters. Landrace

Trait	Shoulder	Belly	Loin	Ham	CLMP	YIELD	BF	LD
Shoulder	.15 (.05)	-.39 (.05)	-.16 (.04)	-.41 (.04)	.06 (.07)	.34 (.05)	-.08 (.07)	-.12 (.06)
Belly	-.32 (.15)	.51 (.07)	-.07 (.06)	-.43 (.05)	-.15 (.10)	-.40 (.07)	.25 (.09)	.07 (.08)
Loin	-.29 (.19)	-.35 (.14)	.21 (.06)	-.46 (.05)	.05 (.07)	-.26 (.05)	.15 (.07)	.23 (.06)
Ham	-.02 (.19)	-.87 (.05)	.12 (.17)	.32 (.07)	.04 (.08)	.22 (.06)	-.23 (.07)	-.10 (.07)
CMLP	.30 (.16)	-.42 (.10)	.05 (.15)	.32 (.13)	.50 (.07)	-.07 (.09)	-.10 (.11)	.05 (.10)
YIELD	.14 (.19)	-.25 (.13)	-.05 (.17)	.26 (.15)	.43 (.12)	.27 (.06)	.21 (.07)	.25 (.05)
BF	-.35 (.16)	.66 (.08)	.04 (.16)	-.57 (.11)	-.73 (.07)	-.30 (.14)	.41 (.07)	.35 (.10)
LD	-.21 (.19)	-.03 (.14)	.40 (.15)	-.02 (.15)	.61 (.11)	.61 (.11)	-.27 (.14)	.33 (.07)

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396 Table 6. Heritabilities (diagonal, bold) and genetic (below diagonal) and residual (above
 397 diagonal) correlations (with SE). Primal cut lean meat composition (%), CLMP and carcass
 398 yield in (%), backfat and loin depth in millimeters. Landrace

Trait	Shoulder LMP	Belly LMP	Loin LMP	Ham LMP	CLMP	YIELD	BF	LD
Shoulder LMP	.38 (.07)	.68 (.04)	.60 (.05)	.60 (.04)	.88 (.02)	-.06 (.07)	-.09 (.08)	-.04 (.08)
Belly LMP	.89 (.03)	.54 (.07)	.81 (.03)	.67 (.04)	.91 (.02)	-.04 (.08)	-.17 (.10)	-.04 (.10)
Loin LMP	.89 (.04)	.91 (.02)	.58 (.08)	.59 (.05)	.82 (.03)	.05 (.09)	-.19 (.10)	.13 (.10)
Ham LMP	.89 (.04)	.83 (.04)	.84 (.04)	.41 (.07)	.83 (.02)	-.15 (.07)	-.07 (.09)	.10 (.08)
CMLP	.97 (.01)	.96 (.01)	.94 (.02)	.93 (.02)	.58 (.07)	-.05 (.08)	-.10 (.10)	.02 (.09)
YIELD	.36 (.13)	.40 (.11)	.47 (.11)	.40 (.13)	.42 (.12)	.14 (.04)	0.24 (.07)	.18 (.08)
BF	-.63 (.09)	-.75 (.07)	-.71 (.07)	-.56 (.10)	-.72 (.07)	-.29 (.14)	0.33 (0.06)	.31 (.09)
LD	.59 (.12)	.55 (.11)	.71 (.09)	.69 (.10)	.64 (.10)	.59 (0.12)	-.25 (.14)	.49 (.07)

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