

Prediction of portal and hepatic blood flow from intake level data in cattle

Article

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INTERPRETIVE SUMMARY:

2	Prediction of portal and hepatic blood flow in cattle. Ellis et al. Given the extent of variability in
3	post absorptive metabolism, there is growing interest in developing integrated post-absorptive metabolism models for cattle. An integral part of linking a multi-organ post absorptive model is
4 5	the prediction of nutrient flow between organs and thus blood flow. This paper applied a
6	multivariate meta-analysis technique to simultaneously predict incoming and outgoing blood flows
7	to the liver. Prediction equations based on DMI performed well, and division of DMI into forage
8	and concentrate DMI improved blood flow predictions.
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10	RUNNING HEAD: PREDICTION OF LIVER BLOOD FLOW IN CATTLE
11	
12	Prediction of portal and hepatic blood flow from intake level data in cattle
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ABSTRACT

29

30 There is growing interest in developing integrated post-absorptive metabolism models for dairy 31 cattle. An integral part of linking a multi-organ post-absorptive model is the prediction of nutrient fluxes between organs, and thus blood flow. It was the purpose of this paper to use a multivariate 32 33 meta-analysis approach to model portal blood flow (PORBF) and hepatic venous blood flow 34 (HEPBF) simultaneously, with evaluation of hepatic arterial blood flow (ARTBF; ARTBF = 35 HEPBF – PORBF) and PORBF/HEPBF (%) as calculated values. The database used to develop 36 equations consisted of 296 individual animal observations (lactating and dry dairy cows and beef cattle) and 55 treatments from 17 studies, and a separate evaluation database consisted of 34 37 treatment means (lactating dairy cows and beef cattle) from 9 studies obtained from the literature. 38 39 Both databases had information on DMI, MEI, body weight and a basic description of the diet including crude protein intake and forage proportion of the diet (FP; %). Blood flow (L/h or L/kg 40 BW^{0.75}/h) and either DMI or MEI (g or MJ/d or g or MJ/kg BW^{0.75}/d) with linear and quadratic 41 fits were examined. Equations were developed using cow within experiment and experiment as 42 random effects, and blood flow location as a repeated effect. Upon evaluation with the evaluation 43 database, equations based on DMI typically resulted in lower root mean square prediction errors, 44 expressed as a % of the observed mean (rMSPE%) and higher concordance correlation coefficient 45 (CCC) values than equations based on MEI. Quadratic equation terms were frequently non-46 47 significant, and the quadratic equations did not out-perform their linear counterparts. The best performing blood flow equations were: PORBF (L/h) = $202 (\pm 45.6) + 83.6 (\pm 3.11) \times DMI (kg/d)$ 48 49 and HEPBF (L/h) = $186 (\pm 45.4) + 103.8 (\pm 3.10) \times DMI (kg/d)$, with rMSPE% values of 17.5 and 50 16.6 and CCC values of 0.93 and 0.94, respectively. The residuals (predicted – observed) for PORBF/HEPBF were significantly related to the forage % of the diet, and thus equations for 51

52	PORBF and HEPBF based on forage and concentrate DMI were developed: PORBF $(L/h) = 210$
53	$(\pm 51.0) + 82.9 \ (\pm 6.43) \times$ Forage (kg DM/d) + 82.9 $(\pm 6.04) \times$ Concentrate (kg DM/d), and
54	HEPBF (L/h) = $184 (\pm 50.6) + 92.6 (\pm 6.28) \times$ Forage (kg DM/d) + $114.2 (\pm 5.88) \times$ Concentrate
55	(kg DM/d), where rMSPE% values were 17.5 and 17.6 and CCC values were 0.93 and 0.94,
56	respectively. Division of DMI into forage and concentrate fractions improved the joint Bayesian
57	Information Criterion (BIC) value for PORBF and HEPBF (BIC = 6512 vs. 7303), as well as
58	slightly improved the rMSPE and CCC for ARTBF and PORBF/HEPBF. This was despite
59	minimal changes in PORBF and HEPBF predictions. Developed equations predicted blood flow
60	well, and could easily be used within a post absorptive model of nutrient metabolism. Results also
61	suggest different sensitivity of PORBF and HEPBF to the composition of DMI, and accounting
62	for this difference resulted in improved ARTBF predictions.
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64	Key words: blood flow, portal, hepatic, cattle, meta-analysis, multivariate
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66	INTRODUCTION
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68	The ability of current feed ration systems to predict the effects of metabolizable protein
69	supply on milk protein production and nitrogen excretion to the environment by dairy cattle is
70	limited by an oversimplified representation of post-absorptive metabolism (Lapierre et al., 2006).
71	Given the variability in post-absorptive metabolism, there is interest in developing integrated post-
72	absorptive models of metabolism (portal-drained viscera, liver, mammary gland, and other organs
73	or tissues) to replace current empirical feeding systems for cattle. Integration of such organ-based
74	models requires prediction of nutrient flow between organs, including prediction of hepatic arterial
75	(ARTBF), portal (PORBF) and hepatic venous (HEPBF) blood flows (BF). Across the liver, the

relative contribution of ARTBF and PORBF can have a significant effect on nutrient fluxes 76 through the organ (e.g. Barnes et al., 1986), warranting reliable prediction of these blood flows. 77 Nutrient concentration in PORBF is modified by the net absorption of nutrients following digestion 78 79 of feeds (or the net utilization of nutrients from arterial blood), while ARTBF nutrient concentration is mainly the result of the residual balance between nutrient absorption, utilization, 80 81 endogenous synthesis, and mobilization from body tissues. Several attempts to model ARTBF, PORBF and/or HEPBF in ruminants are present in the literature, but 1.) were conducted on sheep 82 (e.g. Vernet et al., 2009), 2.) use older meta-analysis techniques which exclude random effects 83 84 (e.g. Lescoat et al., 1996), or 3.) examined only one of the 3 blood flows of interest (e.g. Huntington, 1984; Bermingham et al., 2008). Species differences in blood flow (e.g. between cattle 85 and sheep) have already been observed (Vernet et al., 2005; Bermingham et al., 2008), indicating 86 that cross-species application of blood flow equations may be poor. Equations developed using 87 older meta-analysis techniques may inherently contain prediction errors (St-Pierre, 2001; Sauvant 88 et al., 2008). A fully integrated post-absorptive model for cattle would require all 3 blood flows to 89 be estimated simultaneously. Therefore, a multivariate meta-analysis approach, simultaneously 90 fitting equations for ARTBF, PORBF and HEPBF, while accounting for the interrelationship 91 92 between BF, is warranted.

The purpose of this study was therefore to (1) investigate the simultaneous prediction of ARTBF, HEPBF and PORBF for cattle via a multivariate meta-analysis on published studies, considering DMI and MEI as driving variables, and (2) to compare these predictions to available extant prediction equations on an evaluation database, in order to identify the most appropriate prediction equations for use in future cattle metabolism models.

MATERIALS AND METHODS

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101 Developmental Database

The database used for equation development is summarized in Table 1. It consisted of 17 102 studies with 296 individual animal means and 55 treatment means. Published experiments 103 104 included: Reynolds et al. (1991; 1992a,b; 1993; 1994a,b; 1995a,b; 1998; 1999; 2001; 2003a,b), 105 Caton et al. (2001), Hanigan et al. (2004), Maltby et al. (2005) and Røjen et al. (2011). Experiments 106 covered both lactating and dry dairy cows and growing beef cattle (steers and heifers). Method of 107 BF measurement was downstream dilution of para-aminohippuric acid (PAH) (Katz and Bergman, 108 1969) for all studies. Within studies, BF results were means of (between) 5 to 12 hourly measurements. All reported BF values are on a whole blood basis. Criteria for inclusion in the 109 110 developmental database included availability of individual animal data and provision of information on both PORBF and HEPBF, DMI, metabolizable energy intake (MEI), BW and 111 112 forage % (FP) in the diet. Within study, any treatments which were not nutritional were removed in order to minimize non-nutritional variation in the database. 113

Within the database, the average SD within treatment across the database (indicator of within treatment animal variability) was 135 L/h, 210 L/h, 177 L/h and 0.852 kg/d for ARTBF, PORBF, HEPBF and DMI, respectively, and the average SD of treatment means (indicator of variation across treatment means) was 152 L/h, 548 L/h, 673 L/h and 6.35 kg/d for ARTBF, PORBF, HEPBF and DMI, respectively. Preliminary analysis (not shown) revealed that withintreatment BF variation was significantly related to within-treatment DMI variation (P < 0.01).

121 *Evaluation Database*

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The database used for equation evaluation is summarized in Table 2. It consisted of 9

123 studies with 34 treatment means extracted from the published literature (Wieghart et al., 1986; 124 Eiseman and Nienaber, 1990; Huntington et al., 1990; Guerino et al., 1991; Reynolds and Tyrrell, 1991; Casse et al., 1994; Eiseman and Huntington, 1994; Whitt et al., 1996; Alio et al., 2000), and 125 included both lactating dairy cows and beef cattle. Method of BF measurement for all studies was 126 downstream dilution of PAH (Katz and Bergman, 1969). Similar to the developmental database, 127 all reported BF values are whole blood. Criteria for inclusion in the database included published 128 studies with provision of information on PORBF, HEPBF, DMI, MEI, BW and FP. Having MEI 129 and simultaneous reporting of PORBF and HEPBF as inclusion criteria for the evaluation database 130 131 limited the number of potential studies which could be included, but ensured an equal comparison between DMI and MEI, and PORBF and HEPBF based equations. Similar to the developmental 132 database, within study, any treatments which were not nutritional were removed in order to 133 134 minimize non-nutritional variation in the database.

135The observed PORBF and HEPBF vs. DMI relationship for both the developmental and136evaluation databases are presented in Figure 1 and the distribution of FP across DMI in Figure 2.

137

138 Equation Development

To model the effect of DMI and MEI of cattle on ARTBF, PORBF and HEPBF, mixed model analysis was performed. Linear and quadratic multivariate mixed model analysis was conducted using the NLINMIX macro of SAS (NLMM 8.0 SAS; Moser, 2004; Littell et al., 2006), with simultaneous parameterization of the response variables (PORBF, HEPBF) and representation of the correlation between these variables via the repeated effects statement (Strathe et al., 2010). For a recent example of NLMM code, see the appendix of Strathe et al. (2009).

145 Due to the high degree of error and low sensitivity of ARTBF to the driving variables, it

was difficult to obtain convergence of the multivariate model when ARTBF was modelled directly
(not shown). This is likely because ARTBF is a comparatively small flow determined by difference
experimentally (*in vivo*, observed ARTBF = observed HEPBF – observed PORBF). As a result,
predicted ARTBF was determined by calculation of the difference between predicted PORBF and
HEPBF. Similarly, PORBF/HEPBF (%) was evaluated as the ratio of predicted blood flows, and
not modeled directly.

As the data were compiled from multiple studies, it was necessary to analyze not only the fixed effects of the dependent variables, but also the random effect of experiment as this accounts for differences between experiments such as physiological status of the animals, experimental design, measurement methods, techniques, and laboratory variation (St-Pierre, 2001; Sauvant et al, 2008). As it was desirable to examine the between animal variation in DMI and BW, the full model also included the random effect of cow nested within experiment.

158 The statistical model can be written as follows, where fixed and random effects are 159 incorporated directly into parameters:

160
$$Y_{ijk} = f(\mathcal{O}_{ij}, \text{ intake}_{ijk}) + e_{ijk},$$
 [1]

$$161 \qquad \emptyset_{ij} = \begin{bmatrix} \emptyset_{1ij} \\ \emptyset_{2ij} \\ \vdots \\ \vdots \\ \emptyset_{dij} \end{bmatrix} = \begin{bmatrix} B_{11} \cdot x_1 + B_{12} \cdot x_2 \\ B_{21} \cdot x_1 + B_{22} \cdot x_2 \\ \vdots \\ B_{d1} \cdot x_1 + B_{d2} \cdot x_2 \end{bmatrix} + \begin{bmatrix} b_i^{(1)} \\ b_i^{(2)} \\ \vdots \\ b_i^{(d)} \end{bmatrix} + \begin{bmatrix} b_{i,j}^{(1)} \\ b_{i,j}^{(2)} \\ \vdots \\ b_i^{(d)} \end{bmatrix}$$

In this equation, the function f is a linear or quadratic function of intake (DMI or MEI), with the parameter vector \emptyset_{ij} and model error e_{ijk} . The experiment and cow(experiment) random effects, $\{b_i\}$ and $\{b_{i,j}\}$, are assumed independent of each other and independent of within cow errors e_{ijk} . The *B*'s are the fixed effects influencing the curve parameters due to blood flow (PORBF, HEPBF), and are introduced via two dummy variables x₁ and x₂, respectively. Initial analysis revealed a potential 'fan' shape in the residuals, where residual variance increased with the predicted BF value. In addition, within-treatment and across treatment BF variation increased as BF and/or DMI increased (P < 0.05; data not shown). This may reflect the different type of animals used at low and high DMI (beef cattle vs. dairy cows), milk yield or body reserve mobilization, or the range of diets examined. To compensate, a variance weighting statement (**wt**) was added to the NLMM macro model, wt = 1/(predicted value)², which decreased variance weight with increasing predicted BF value (see Strathe et al., 2009 for discussion).

The joint distribution of random effects was assumed to be multivariate normal and the dual quasi-Newton technique was used for optimization with an adaptive Gaussian quadrature as the integration method.

177

178 Equation Evaluation

Goodness of fit of the statistical model (inclusion/exclusion of random effects, variance/covariance structure selection etc.) was evaluated using the Bayesian information criterion (**BIC**) fit statistic (SAS Inst. Inc., Cary, NC), where lower values indicate better model fit, and the value and significance of the fixed effect model parameters were tested against a *P* value of 0.05.

Evaluation of newly developed and extant equations against the evaluation database were performed via two methods. Firstly, root mean square prediction error (**rMSPE**) was performed, where the mean square prediction error (**MSPE**) is calculated as:

187
$$MSPE = \sum_{i=1}^{n} (O_i - P_i)^2 / n$$
 [2]

where *n* is the total number of observations, O_i is the observed value, and P_i is the predicted value. The rMSPE, expressed as a percentage of the observed mean, gives an estimate of the overall prediction error. The rMSPE can also be decomposed into error in central tendency or mean bias
(ECT), error due to deviation of the regression slope from unity (ER) and error due to the
disturbance (random error) (ED) (Bibby and Toutenburg, 1977).

- 193 Secondly, concordance correlation coefficient analysis (CCC) was performed (Lin, 1989),
 194 where CCC is calculated as:
- 195 $\operatorname{CCC} = r \times C_b$ [3]

where r is the Pearson correlation coefficient and C_b is a bias correction factor. The r 196 variable gives a measure of precision, while C_b is a measure of accuracy. Associated CCC variables 197 198 (used in calculation of C_b) are v, which provides a measure of scale shift, and u, which provides a 199 measure of location shift relative to the scale. The v value indicates the change in standard deviation, if any, between predicted and observed values. A v value greater than 1.0 indicates larger 200 201 variance in the predicted data compared to observed, while a v value less than 1.0 indicates a smaller variance in the predicted data compared to observed. A positive u value indicates over-202 prediction, while a negative *u* value indicates under-prediction. 203

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- 205

RESULTS AND DISCUSSION

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207 Low vs. High Intake

Visual inspection of the data revealed two potential clusters within the databases, representing a cluster of 'lower-intake' and 'higher-intake' data (Figure 1). These intake groups are confounded with animal type, and also represent clusters of studies, where the low intake group comprised all beef cattle data and the high intake group comprised all dairy cow data. As a result, analysis was initially performed by separating the data (by studies) into low and high intake groups 213 (or, alternatively, animal type) (Table 1), and analysing for statistical differences between intake-214 group parameter estimates. In sheep, Vernet et al. (2005; 2009) suggested that BF responses to DMI or MEI differed based on the level of intake. Additionally, physiological status may have an 215 216 effect on BF. A major difference between the data of Vernet et al. (2005; 2009) and the current data (aside from species) is, however, that in the current database level of intake did not fall far 217 below maintenance requirements (Table 1). In this study, the average multiple of maintenance 218 feeding level was 1.31 (\pm 0.378) for the low-intake and 2.65 (\pm 0.749) for the high-intake groups, 219 compared to 0.5 and 1.3 in the study of Vernet et al. (2009), respectively. Separation of studies 220 221 into two intake groups in the current dataset did not result in significantly different parameter 222 estimates between low- and high-intake groups (P > 0.09) (Table 3). As a result, separate equations for the low intake and high intake groups (or animal type) are not reported, and equations reported 223 224 were fit to the full database.

225

226 Linear and Quadratic Blood Flow Equations

227 Results of linear and quadratic curve fitting to the BF development database are presented in Table 3. Equations were fit to data with BF units of L/h combined with DMI or MEI units of 228 kg/d or MJ/d, or with BF units of L/kg BW^{0.75}/h combined with DMI or MEI units of kg/kg 229 BW^{0.75}/d or MJ/kg BW^{0.75}/d. Scaling relative to BW was also examined, but resulted in no 230 improvements over BW^{0.75}, and is not reported. Model structure (random effects, variance-231 232 covariance structures, variance weighting) was optimized to ensure convergence and to minimize the joint BIC value. Joint BIC values represent the BIC for PORBF and HEPBF combined, which 233 were fit simultaneously. The significance of parameter estimates (vs. zero) are reported, as well as 234 235 the P-value for testing the low vs. high intake parameter estimates against each other, via 236 CONTRAST statements in SAS (Table 3). This division into low and high intake groups was not performed for quadratic equations due mainly to lack of convergence, but also because a quadratic 237 fit should inherently capture changes in the slope of the relationship across intake level. In support 238 239 of the findings that parameter estimates did not differ significantly between low and high intake 240 groups, fitting quadratic equations to the database resulted in similar or marginally better joint BIC 241 values, and the quadratic parameter estimates were not always significant (Table 3). Lack of significance of the quadratic parameter indicates potential over-parametrization of the model or 242 that the relationship was linear within the range of data available. When BF was expressed in units 243 244 of L/h, the negative quadratic parameter was significant for HEPBF, but not for PORBF (driving variable of DMI or MEI). When BF was expressed in units of L/kg BW^{0.75}/h, the quadratic 245 parameter was only significant for PORBF with DMI as a driving variable. Linear equation 246 parameters (slope and intercept) were always significant (P < 0.01). 247

Equations based on MEI generally had lower BIC values compared to equations based on DMI (Table 3), indicating better model fit. Conclusions on BF units cannot be made based on BIC, as BIC values are scaled by the units.

251

252 DMI and MEI Based Equation Evaluation

Equations developed were tested on an independent evaluation database (described in Table 2) to compare prediction precision and accuracy. Although the evaluation database may be considered somewhat small relative to the size of the development database, it does represent a complete dataset, where all variables predicted and evaluated were reported in the publications. Results are presented in Table 4 for PORBF, HEPBF, ARTBF (predicted by difference) and PORBF/HEPBF (%, ratio of predicted blood flows) for each equation.

259 Comparing DMI to MEI as the driving variable, DMI typically resulted in slightly better 260 predictions based on rMSPE and CCC results, except for PORBF/HEPBF (Table 4). This could be the result of added variation or error due to MEI determination. However, Han et al. (2002) also 261 suggested that portal BF responded primarily to bulk fill rather than nutrient supply. Reynolds et 262 al. (1991) suggested that in addition to ME consumed, ME density of the diet affected PORBF and 263 HEPBF via effects of forage content on gut fill and subsequent effects on gut mass and the work 264 of digestion, which may also explain the better relationship for splanchnic blood flow and DMI. 265 Vernet et al. (2009) found a similar lack of improvement with MEI over DMI in predicting BF in 266 267 sheep. Therefore, it is likely that this observation has a physiological basis rather than being error related. 268

Comparing linear to quadratic equations, predictions were similar but slightly improved with the linear equations (Table 4). As many of the quadratic parameter estimates were not significant, this is not a surprising result.

Comparing L/h and L/kg BW^{0.75}/h as units for BF, CCC results were in general slightly 272 improved when L/h was used and rMSPE results were in general slightly improved when L/kg 273 BW^{0.75}/h was used (Table 4). Scaling with BW^{0.75} reduced the contribution of non-random error 274 275 sources (ECT, ER) to the rMSPE total, indicating improved predictions compared with scaling without BW^{0.75}. However for CCC, BW^{0.75} scaling reduced the total CCC via a decrease in C_b , 276 despite a slight increase in r. This difference in results is likely due to differences in division of 277 error within rMSPE and CCC calculations (for a discussion see Ellis et al., 2010). Scaling by 278 BW^{0.75} is presumed to extend the range of data the equations may be applicable on, and thus was 279 280 of interest when combining dairy and beef data, but it may also introduce additional variation due 281 to BW measurement (difficulty getting a precise scale number, variation in gut fill contribution to

BW, etc.). For whichever reason, these results indicate that scaling by BW^{0.75} may not improve
predictions of blood flow over units of L/h, as performance between the equations was similar.

Predictions for PORBF and HEPBF, as evaluated by rMSPE and CCC analysis, were 284 typically very good, with CCC values greater than 0.84 and rMSPE values less than 19% (Table 285 4). The best predictions of PORBF and HEPBF when blood flow was expressed in L/h, were the 286 linear equations with DMI as the driving variable (P1 and H1 equations; rMSPE = 17.5 and 16.6%, 287 CCC = 0.93 and 0.94, respectively). Similarly, when PORBF and HEPBF were expressed relative 288 to BW^{0.75}, the linear DMI equations resulted in slightly better predictions (P5 and H5 equations; 289 rMSPE = 15.4 and 14.9%, CCC = 0.87 and 0.90, respectively). However, in general predictions 290 were similar and good across all equations with only minor differences. 291

Residual analysis was conducted on the seemingly best performing equations (linear, DMI; 292 L/h and L/kg BW^{0.75}/h), and is displayed in Figure 3. Residuals plotted against predicted BF 293 (Figure 3) did not reveal any significant trends in the data (P > 0.05), nor for the most part when 294 plotted against the driving variable DMI (kg/d or g/kg BW^{0.75}/d; P > 0.05), with the exception of 295 296 residual ARTBF (L/h), where P = 0.04 (residual ARTBF (L/h) = 40.2 (± 29.2) - 6.4 (± 2.83) × DMI(kg/d); graphs not shown). The residuals were also plotted against the forage proportion (FP, 297 %) of the diet, and while the regression was not significant for ARTBF, PORBF or HEPBF (P >298 0.05), it was significant for PORBF/HEPBF (%) (P = 0.03 and 0.03, for L/h and L/kg BW^{0.75}/h 299 equations, respectively; Figure 4). As the result of the FP pattern in the residuals, the FP of the diet 300 301 was considered as an additional driving variable. The results of separating forage and concentrate DMI is outlined in the following section. 302

303

304 Separating Forage and Concentrate DMI

To further examine the potential effect of the FP of the diet, DMI was separated into forage and (starch-rich) concentrate components (kg/d) in the developmental database, and new equations were parameterized for PORBF and HEPBF, with ARTBF again calculated by difference. Equations developed are presented in Table 5.

309 When testing the PORBF forage and concentrate slopes against each other, the difference 310 between parameter estimates was non-significant (Table 5), indicating no difference in effect of type of DMI on PORBF. However, testing HEPBF forage and concentrate slope parameters against 311 each other revealed a significant difference, the slope for concentrate being higher (Table 5). This 312 313 result suggests a higher sensitivity of HEPBF to FP or energy intake compared to PORBF. In support of this, the slope of MEI based equations was also generally higher for HEPBF than for 314 PORBF (Table 3). This may reflect an increased absorption and liver metabolism of propionate 315 and other VFA with an increasing concentrate proportion in diet DM (Huntington, 1990). 316

Dividing DMI into forage and concentrate components resulted in improved joint BIC values (Table 3 vs. Table 5), slightly improved ARTBF and PORBF/HEPBF predictions, and similar PORBF and HEPBF predictions to equations based on total DMI (Table 4 vs. Table 6).

Interpretation of these FP equations is challenging. For PORBF, it appears forage and concentrate DMI do not differ in their magnitude of effect on BF (similar parameter estimates). This may, however, be the compound result of two opposing mechanisms: forage DMI may stimulate BF less than concentrate DMI due to lower energy content and digestibility, but this may be countered by a higher bulk fill value which is stimulatory to BF (Reynolds et al., 1991).

In contrast, it appears that HEPBF may be more sensitive to concentrate (or energy intake)
than to forage intake (significantly different parameter estimates), suggesting that total liver BF is

still more heavily regulated by energy status and absorption of VFA and other components of ME
than gut fill. Vernet et al. (2009) made similar observations in sheep.

While these differences did not greatly alter PORBF and HEPBF predictions, prediction of 329 330 the calculated ARTBF and PORBF/HEPBF were both improved. This suggests that while DMI alone may predict PORBF or HEPBF adequately, differences between them (ARTBF) may be 331 better predicted with consideration of the diet FP. While Vernet et al. (2009) did not examine 332 residuals of arterial/venous BF against FP, they did observe a significant relationship between the 333 residuals and OM digestibility, suggesting again that BF depend on both bulk and the nutrient 334 335 density of the diet. In order to better understand these effects, an examination of the regulation of liver BF is required. 336

337

338 Blood flow regulation through splanchnic tissues

Blood flow through the portal vein (PORBF), the main blood supply to the liver, is 339 regulated by the portal drained viscera (**PDV**) which is responsible for nutrient uptake and delivery 340 to the post-absorptive environment, as opposed to being controlled by the liver (Lautt, 2009). Bulk 341 fill as well as nutrient delivery to the animal impact this flow (e.g. see Reynolds et al., 1991) 342 343 through regulation by intrinsic and extrinsic mechanisms. Intrinsic mechanisms include local metabolic control (response to oxygen supply and demand), myogenic control (transmural 344 pressure), local reflexes (presence of lumen contents) and locally produced vasoactive substances 345 346 (e.g. gastrin, secretin, cholecystokinin) (Lautt, 1996; Lautt, 2009). The extrinsic factors include sympathetic innervation, circulating vasoactive substances and systemic haemodynamic changes 347 348 (Lautt, 1996; Lautt, 2009). Hepatic arterial blood flow (ARTBF), while regulated by local tissue 349 oxidation levels in other organs, is also not regulated by the liver (Lobley et al., 2000). Instead, it 350 appears that ARTBF regulation is linked to PORBF, ensuring the liver receives a constant total 351 blood flow relative to liver mass (Lautt, 1996; Lautt, 2009). This appears to be regulated via a continuous release of adenosine into the space of Mall, independent of oxygen supply or demand, 352 followed by removal through both ARTBF and PORBF. Adenosine itself is a powerful vasodilator 353 (Lautt, 2009). If PORBF is reduced, the local concentration of adenosine increases, stimulating 354 arterial vasodilation and increased ARTBF to remove the adenosine. On the other hand, when 355 PORBF is high, e.g. during peak absorption of nutrients from the rumen, this may cause a reduction 356 in ARTBF due to a decrease in local adenosine concentrations. This process is referred to as the 357 358 hepatic arterial buffer response. In this respect, the liver does not drive either of the incoming blood flows; PORBF is driven by the PDV, and ARTBF is driven, inversely, by PORBF. However, the 359 liver can have significant indirect regulatory effects on incoming BF, via mechanisms impacting 360 BF to splanchnic organs that drain into the PORBF. As well, longer-term effects on BF can be 361 mediated by changes in liver mass. For a full review of liver BF regulation, see Lautt (2009). 362

363 Based on the empirical blood flow prediction equations developed in the present work, it is possible that stimulation of PORBF by concentrate (energy) intake is countered by a depression 364 in PORBF by a lower forage intake (bulk fill), resulting in similar forage and concentrate 365 366 parameters for PORBF prediction across a range of FP. When FP was low and total DMI alone was the driving variable, PORBF/HEPBF was over predicted (P < 0.05) and as a result ARTBF 367 slightly under predicted (non-significant; Figure 4). This makes sense as ARTBF is calculated as 368 369 HEPBF – PORBF. At a low FP, over-prediction of PORBF/HEPBF could be due to over prediction of PORBF and/or under prediction of HEPBF. Examination of the (albeit non-significant) slope 370 371 terms in Figure 4, suggest that both are occurring to some extent.

372 Since parameterization with separate forage and concentrate DMI resulted in similar parameters for PORBF, if these results reflect *in vivo* observations, it suggests that for low FP 373 diets, HEPBF is under-represented by using total DMI because of an under-represented ARTBF 374 contribution. This suggests that while total blood flow through the liver is sensitive to energy 375 intake (and thus different forage and concentrate parameters for HEPBF), factors reducing PORBF 376 relative to the local adenosine concentration (in this case, FP or bulk fill) may drive an increase in 377 ARTBF to compensate. Thus, separating forage and concentrate DMI captures this effect of 378 ARTBF, without directly modeling ARTBF. 379

When interpreting these results, it should be noted that while DMI varied within all studies, FP did not. Although the equations were parameterized on kg/d of forage and concentrate, in the developmental database 5 of 17 studies specifically examined FP effects, and 4 of 9 studies in the evaluation database examined FP effects. The distribution of FP across DMI is illustrated in Figure 2. Therefore, the forage + concentrate equations require examination on an additional database with additional variation in FP to ensure it is not only an artifact of the data used.

386

387 Equations Based on Diet Chemical Composition

Although one of the main purposes of this paper was to compare DMI and MEI as the major drivers of PORBF and HEPBF in a multivariate analysis, CP and NDF content of the diet were also available in the development databases. Therefore, initially, development of equations based on CP or NDF intake (kg/d or g/kg BW^{0.75}/d) were also considered. However, while these equations had BIC values comparable to the forage + concentrate DMI equations (joint BIC values were: 6413 for CP (kg/d), 6534 for NDF (kg/d), 1391 for CP (g/kg BW^{0.75}/d), and 1491 for NDF (g/kg BW^{0.75}/d) based equations), their rMSPE and CCC values were worse than those of DMI 395 and MEI (for e.g., CP (kg/d) predicting PORBF (L/h) resulted in rMSPE% = 37.8, CCC = 0.63396 and HEPBF (L/h) rMSPE% = 37.9 and CCC = 0.69, on the evaluation database). As a result, these equations were not pursued further. However, equations developed considering multiple chemical 397 components of the diet may be considered in the future, in particular given the relationship 398 observed here with FP. 399

400

401

Comparison with Extant Blood Flow Equations

To compare predictions of the newly developed blood flow equations to extant equations, 402 403 several equations were selected from the literature and applied to the evaluation database. The equations of Lescoat et al. (1996) were not included, as the evaluation database used here shared 404 data with the developmental database used by Lescoat et al. (1996), resulting in unsurprisingly 405 good blood flow predictions by these equations (not shown). Although the equations of Vernet et 406 al. (2009) were developed on sheep, it represented an interesting challenge to include their 407 equations for comparison on cattle data. 408

409 Extant equation evaluations are presented in Table 7. Of the equations evaluated, the PORBF equation of Huntington et al. (1984) based on MEI performed comparably to the newly 410 411 developed PORBF equations in terms of rMSPE and CCC analysis. These equations (Huntington et al., 1984) were developed on beef and dairy heifer data. The linear PORBF equation of 412 Bermingham et al. (2008) performed adequately, with slightly more bias (over prediction) and 413 414 lower CCC values. However, similar to the results found in the current study, the quadratic equation for PORBF by Bermingham et al. (2008) did not improve predictions over their linear 415 416 equation. These equations were developed on a combination of sheep and cattle data.

417	The sheep equations of Vernet et al. (2009) also tended to over predict PORBF, HEPBF
418	and ARTBF, expressed relative to BW, likely illustrating a species difference. Of the 3 sets of
419	extant equations, only those of Vernet et al. (2009) allowed calculation and evaluation of ARTBF
420	and PORBF/HEPBF. Both the Vernet et al. (2009) above maintenance and above + below linear
421	equations tended to under predict the mean PORBF/HEPBF. Interestingly, the Vernet et al. (2009)
422	sheep equations also showed a relationship between the PORBF/HEPBF residual and the FP of
423	the diet (Figure 5) with a trend similar to that in the equations derived in the present study (Figure
424	4), and therefore seems to support the separation of forage and concentrate parameters.
425	
426	CONCLUSIONS
427	
428	Equations developed herein represent advancement over current PORBF, HEPBF, ARTBF
429	and PORBF/HEPBF prediction equations available in the literature for cattle. In the present
430	analysis, a more advanced meta-analysis technique was used, allowing simultaneous predictions
431	of multiple blood flows, as well as providing new equations which separate forage DMI from
432	concentrate DMI, resulting in improvements in ARTBF and PORBF/HEPBF predictions. All
433	PORBF and HEPBF equations performed well when evaluated on an evaluation database. These
434	equations can be applied within a post-absorptive model of cattle metabolism, in order to predict
435	nutrient fluxes to and from the liver, but should be further evaluated on additional data obtained
436	under a wider range of conditions.
437	
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439	

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Variable	<u>All data</u>	data Beef cattle (Low-intake group) Dairy cow						ow (High	v (High-intake group)			
	Mean	SD^2	Mean	SD^2	MIN ³	MAX ⁴	Mean	SD^2	MIN ³	MAX^4		
DMI (kg/d)	11.8	6.58	5.5	1.11	3.0	8.3	17.3	3.84	7.9	25.1		
MEI (MJ/d)	126.4	74.91	60.6	12.74	37.4	97.6	192.8	48.03	81.4	295.1		
CP (kg/d)	1.8	1.03	0.9	0.25	0.5	1.4	2.6	0.81	1.0	4.6		
NDF (kg/d)	3.7	2.44	1.3	0.58	0.6	3.3	5.9	1.13	3.3	9.0		
BW (kg)	510	140.3	412	84.4	251	598	637	85.2	487	878		
DMI (g/kg BW ^{0.75} /d)	96.3	46.93	61.0	13.94	42.6	104.1	142.2	32.85	57.0	202.6		
MEI (MJ/kg BW ^{0.75} /d)	1.01	0.515	0.68	0.175	0.49	1.13	1.55	0.423	0.56	2.26		
MEI (Multiple of MN ⁵)	1.81	0.850	1.31	0.378	0.86	2.27	2.65	0.749	0.90	3.90		
Forage Proportion (%)	44	18.0	41	23.3	25	75	47	10.3	35	66		
ARTBF (L/h)	234	206.4	91	64.2	3	437	359	207.1	18	1089		
PORBF (L/h)	1188	586.6	650	126.9	382	986	1655	398.6	762	2887		
HEPBF (L/h)	1409	708.7	736	138.3	428	1019	1992	431.0	929	3208		
ARTBF (L/kg BW ^{0.75} /h)	1.8	1.52	1.0	0.69	0.0	4.1	2.8	1.66	0.5	8.8		
PORBF (L/kg BW ^{0.75} /h)	10.0	3.88	7.2	1.57	4.6	14.4	13.5	3.00	5.6	19.4		
HEPBF (L/kg BW ^{0.75} /h)	11.7	4.79	8.2	1.69	5.5	14.7	16.2	3.57	6.9	21.6		
PORBF/HEPBF (%)	86	9.3	88	7.7	57	100	85	9.2	59	100		
n (data points)	296	-	137	-	-	-	159	-	-	-		
n (treatments)	55	-	22	-	-	-	33	-	-	-		
n (studies)	17	-	7	-	-	-	10	-	-	-		

Table 1. Summary of the blood flow (hepatic arterial - ARTBF; portal venous - PORBF; hepatic venous - HEPBF) developmental database¹.

 586

¹ Mean & SD reported are based on 'n (data points)'. ² Standard deviation. 587

588

³ Minimum value in database. 589

⁴ Maximum value in database. 590

⁵MN – maintenance energy requirement. 591

Variable	Mean	SD^2	MIN ³	MAX^4
DMI (kg/d)	8.4	4.71	4.3	21.8
MEI (MJ/d)	90.4	43.59	51.1	231.5
CP (kg/d)	1.3	0.89	0.7	4.0
BW (kg)	387	97.5	198	538
DMI (g/kg BW ^{0.75} /d)	94.0	36.91	57.1	204.6
MEI (MJ/kg BW ^{0.75} /d)	1.02	0.346	0.62	2.17
MEI (Multiple of MN ⁵)	1.99	0.613	1.16	3.94
Forage Proportion (%)	42	22.0	10	100
ARTBF (L/h)	165	137.9	26	563
PORBF (L/h)	832	369.3	336	1992
HEPBF (L/h)	996	495.9	400	2524
ARTBF (L/kg BW ^{0.75} /h)	1.8	1.24	0.3	5.3
PORBF (L/kg BW ^{0.75} /h)	9.4	2.95	6.4	18.7
HEPBF (L/kg BW ^{0.75} /h)	11.2	3.96	7.5	23.7
PORBF/HEPBF (%)	85	5.9	76	97
n (data points)	34			
n (treatments)	34			
n (studies)	9			

Table 2. Summary of the blood flow (hepatic arterial - ARTBF; portal venous - PORBF; hepatic venous - HEPBF) evaluation database¹.

¹ Mean & SD reported are based on 'n (data points)'.

595 ² Standard deviation.

³ Minimum value in database.

⁴ Maximum value in database.

 $598 \quad {}^{5}MN$ – maintenance energy requirement.

Response Variable	Driving Variable	Eqn	ID	Joint BIC	Int	SE	Р	P (Intake Level) ²	Slope (Lin)	SE	Р	P (Intake Level) ²	Slope (Quad)
<u>L/h</u>	kg/d or N	IJ/d											
PORBF	DMI	Linear	P1	7303	202	45.6	< 0.01	0.98	83.6	3.11	< 0.01	0.73	-
HEPBF	DMI		H1		186	45.4	< 0.01	0.64	103.8	3.10	< 0.01	0.90	-
PORBF	MEI	Linear	P2	6689	294	43.2	< 0.01	0.19	6.8	0.26	< 0.01	0.67	-
HEPBF	MEI		H2		264	42.8	< 0.01	0.09	8.9	0.26	< 0.01	0.96	-
PORBF	DMI	Quad	P3	7296	148	70.9	0.04	-	94.9	12.25	< 0.01	-	-0.44
HEPBF	DMI		H3		72	69.9	0.31	-	129.3	12.03	< 0.01	-	-1.03
PORBF	MEI	Quad	P4	6698	209	68.1	< 0.01	-	8.3	1.07	< 0.01	-	-0.01
HEPBF	MEI		H4		110	65.8	0.10	-	11.8	1.02	< 0.01	-	-0.01
L/kg BW ^{0.75} /h	g or MJ/	kg BW ^{0.75} /d											
PORBF	DMI	Linear	P5	1548	2.10	0.417	< 0.01	0.50	0.080	0.004	< 0.01	0.88	-
HEPBF	DMI		H5		1.91	0.421	< 0.01	0.17	0.100	0.004	< 0.01	0.15	-
PORBF	MEI	Linear	P6	1337	2.80	0.286	< 0.01	0.94	6.61	0.256	< 0.01	0.81	-
HEPBF	MEI		H6		2.41	0.286	< 0.01	0.43	8.71	0.258	< 0.01	0.09	-
PORBF	DMI	Quad	P7	1543	0.58	0.728	0.43	-	0.119	0.016	< 0.01	-	-0.0002
HEPBF	DMI		H8		0.84	0.769	0.27	-	0.128	0.018	< 0.01	-	-0.0002
PORBF	MEI	Quad	P8	1327	1.53	0.690	0.04	-	9.26	1.465	< 0.01	-	-1.09
HEPBF	MEI		H8		1.97	0.701	< 0.01	-	9.53	1.499	< 0.01	-	-0.30

Table 3. Summary of portal (PORBF) and hepatic venous blood flow (HEPBF) prediction equations based on DMI and MEI¹.

 1 Abbreviations: Eqn = equation form, ID = equation name, BIC = Bayesian information criterion, Int = intercept.

² Tested whether slope or intercept for data grouped into 'high' intake (dairy cow) differed from data grouped into 'low' intake (beef

604 cattle), via CONTRAST statements in SAS (data not shown).

Table 4. Summary of blood flow (hepatic arterial - ARTBF; portal venous - PORBF; hepatic venous - HEPBF) predictions on the evaluation database for ARTBF, PORBF and HEPBF locations, where ARTBF = predicted HEPBF – predicted PORBF, PORBF and HEPBF are according to equations presented in Table 3, and PORBF/HEPBF = predicted PORBF/predicted HEPBF \times 100.

Response Variable	Driving Variable	Eqn	ID	Pred Mean ¹	Pred SD ¹	rMSPE, % ²	ECT ,% ³	ER, % ⁴	ED, % ⁵	CCC ⁶	r^7	$C_b{}^8$	v ⁹	u^{10}
<u>L/h</u>	kg/d or M	J/d												
ARTBF	DMI	linear		154	93.8	42.4	2.2	14.0	83.8	0.82	0.93	0.88	0.69	-0.09
PORBF			P1	907	388.3	17.5	27.0	9.0	64.1	0.93	0.98	0.95	1.07	0.20
HEPBF			H1	1061	482.1	16.6	15.5	1.1	83.4	0.94	0.99	0.95	0.99	0.13
PORBF/ H	EPBF, %			86	1.7	6.6	5.7	0.2	94.1	0.18	0.52	0.34	0.30	0.42
ARTBF	MEI	linear		160	90.2	44.2	0.5	14.8	84.7	0.80	0.92	0.87	0.66	-0.05
PORBF			P2	909	292.0	17.6	27.7	13.8	58.5	0.90	0.95	0.95	0.80	0.24
HEPBF			H2	1068	382.2	18.7	14.8	19.5	65.7	0.91	0.96	0.95	0.78	0.17
PORBF/ H	EPBF, %			86	2.4	6.5	3.2	0.3	96.5	0.24	0.68	0.35	0.41	0.27
ARTBF	DMI	quad		160	93.6	44.3	0.5	10.4	89.1	0.80	0.93	0.86	0.69	-0.05
PORBF			P3	908	391.1	17.7	26.7	10.0	63.2	0.93	0.98	0.95	1.07	0.20
HEPBF			H3	1067	484.3	17.3	17.0	1.5	81.4	0.94	0.99	0.95	0.99	0.15
PORBF/ H	EPBF, %			86	2.2	6.9	2.8	3.3	93.9	0.13	0.65	0.20	0.38	0.27
ARTBF	MEI	quad		166	90.8	45.4	0.0	11.6	88.3	0.79	0.92	0.86	0.67	0.01
PORBF			P4	910	299.5	17.4	28.9	10.2	60.9	0.91	0.96	0.95	0.82	0.24
HEPBF			H4	1076	390.0	19.0	17.6	14.8	67.6	0.91	0.96	0.95	0.80	0.18
PORBF/ H	EPBF, %			85	2.8	6.7	0.9	4.2	95.0	0.22	0.78	0.28	0.48	0.13
$L/kg BW^{0.75}/h$ g or MJ/ kg $BW^{0.75}/d$		$^{5}/d$												
ARTBF	DMI	linear		1.7	0.73	43.3	2.0	10.2	87.8	0.70	0.87	0.80	0.60	-0.12
PORBF			P5	9.6	2.91	15.4	1.6	6.2	92.2	0.87	1.00	0.88	1.00	0.06
HEPBF			H5	11.3	3.64	14.9	0.2	0.5	99.3	0.90	1.00	0.90	0.93	0.02
PORBF/ H	EPBF, %			86	1.3	6.6	1.2	0.9	97.9	0.14	0.43	0.32	0.23	0.22

 ARTBF	MEI	linear		1.8	0.72	45.0	0.3	7.5	92.2	0.67	0.87	0.77	0.59	-0.05
PORBF			P6	9.6	2.25	14.4	0.7	6.2	93.1	0.86	0.97	0.89	0.78	0.05
HEPBF			H6	11.3	2.97	15.6	0.2	10.3	89.6	0.87	0.96	0.91	0.76	0.02
PORBF/ H	IEPBF, %			85	1.8	6.6	0.0	0.0	99.9	0.16	0.56	0.29	0.31	0.02
 ARTBF	DMI	quad		1.7	0.80	41.5	3.2	6.7	90.0	0.74	0.91	0.81	0.65	-0.14
PORBF			P7	9.8	2.62	15.9	5.7	0.5	93.8	0.85	0.99	0.87	0.90	0.13
HEPBF			H7	11.5	3.41	15.5	1.6	0.2	98.1	0.89	0.99	0.90	0.87	0.06
PORBF/ H	IEPBF, %			86	1.9	6.4	4.6	1.5	93.9	0.24	0.56	0.43	0.32	0.35
 ARTBF	MEI	quad		1.6	0.84	43.7	4.2	2.0	93.8	0.72	0.92	0.78	0.69	-0.16
PORBF			P8	9.7	2.14	15.4	3.7	8.4	87.9	0.84	0.95	0.88	0.74	0.11
HEPBF			H8	11.4	2.96	15.7	0.4	10.0	89.6	0.87	0.96	0.90	0.76	0.03
PORBF/ H	HEPBF, %			86	2.4	6.6	5.6	0.4	94.0	0.23	0.67	0.35	0.41	0.36

¹ Where: observed means \pm SD: ARTBF, PORBF, HEPBF (L/h): 165 \pm 137.9, 832 \pm 369.3, 996 \pm 495.9; ARTBF, PORBF, HEPBF

610 (L/kg BW^{0.75}/h): 1.8 ± 1.24 , 9.4 ± 2.95 , 11.2 ± 3.96 ; PORBF/HEPBF (%): 85 ± 5.9 , respectively.

 2 Root mean square prediction error, % of observed mean.

³Error due to mean bias, as a % of total MSPE.

 4 Error due to regression, as a % of total MSPE.

⁵Error due to disturbance, as a % of total MSPE.

615 ⁶Condordance correlation coefficient, where $CCC = r \times C_b$.

⁶¹⁶ ⁷Pearson correlation coefficient.

617 ⁸Bias correction factor.

618 ⁹Scale shift.

619 ¹⁰Location shift relative to the scale.

Table 5. Summary of portal (PORBF) and hepatic venous (HEPBF) blood flow prediction equations based on DMI divided into

Response Variable	Driving Variable	Eqn	ID	Joint BIC	Int	SE	Р	Slope (F)	SE	Р	Slope (C)	SE	Р	$P (F vs. C)^2$
L/h	<u>kg/d</u>													
PORBF	DMI^3	Linear	P9	6512	210	51.0	< 0.01	82.9	6.43	< 0.01	82.9	6.04	< 0.01	1.00
HEPBF			H9		184	50.6	< 0.01	92.6	6.28	< 0.01	114.2	5.88	< 0.01	0.03
<u>L/kg BW^{0.75}/h</u>	g/kg BW ^{0.75} /d	<u>d</u>												
PORBF	DMI ³	Linear	P10	1365	2.16	0.467	< 0.01	0.08	0.006	< 0.01	0.08	0.006	< 0.01	0.41
HEPBF			H10		1.91	0.468	< 0.01	0.09	0.006	< 0.01	0.11	0.006	< 0.01	0.01

622 forage (F) and concentrate (C) intake¹.

623

¹Abbreviations: Eqn = equation form, ID = equation name, BIC = Bayesian information criterion, Int = intercept.

² Tested whether the forage (F) and concentrate (C) slopes differed from each other, performed via CONTRAST statements in SAS.

 3 Separated into forage DMI (kg/d) + concentrate DMI (kg/d).

627 Table 6. Summary of blood flow (hepatic arterial - ARTBF; portal venous - PORBF; hepatic venous - HEPBF) predictions based on

628 separated forage + concentrate DMI, on the evaluation database for ARTBF, PORBF and HEPBF locations, where ARTBF =

629 predicted HEPBF – predicted PORBF, PORBF and HEPBF predictions are according to equations presented in Table 5, and

630 PORBF/HEPBF = predicted PORBF/predicted HEPBF \times 100.

Response Variable	Driving Variable	Eqn	ID	Pred Mean ¹	Pred SD ¹	rMSPE, % ²	ECT, % ³	ER, % ⁴	ED, % ⁵	CCC ⁶	r^7	$C_b{}^8$	<i>v</i> ⁹	u^{10}
v anacie	, and to to			1,10011	52	70	,,,	/0	/0					
<u>L/h</u>	<u>kg/d</u>													
ARTBF	\mathbf{DMI}^{11}	linear	-	160	105.1	41.3	0.6	3.9	95.5	0.84	0.97	0.87	0.77	-0.04
PORBF			P9	909	385.0	17.5	28.4	7.6	63.9	0.93	0.98	0.95	1.06	0.21
HEPBF			H9	1069	485.0	17.6	17.0	1.7	81.3	0.94	0.99	0.95	0.99	0.15
PORBF/ H	EPBF %			86	3.3	6.2	4.1	1.0	94.9	0.40	0.84	0.48	0.57	0.24
L/kg BW ^{0.75} /h	g or MJ/ k	kg BW ^{0.75} /	d											
ARTBF	DMI^{11}	linear	-	1.7	0.80	42.1	1.1	5.3	93.6	0.73	0.91	0.80	0.66	-0.08
PORBF			P10	9.7	2.91	15.7	3.2	6.2	90.6	0.87	1.00	0.87	1.00	0.09
HEPBF			H10	11.4	3.64	15.5	1.2	0.6	98.2	0.89	1.00	0.90	0.93	0.05
PORBF/ HEPBF %				85	2.9	6.2	0.8	0.5	98.7	0.36	0.80	0.44	0.51	0.11

631

¹ Where: observed means \pm SD: ARTBF, PORBF, HEPBF (L/h): 165 \pm 137.9, 832 \pm 369.3, 996 \pm 495.9; ARTBF, PORBF, HEPBF

633 (L/kg BW^{0.75}/h): 1.8 ± 1.24 , 9.4 ± 2.95 , 11.2 ± 3.96 ; PORBF/HEPBF (%): 85 ± 5.9 , respectively.

 2 Root mean square prediction error, % of observed mean.

 3 Error due to mean bias, as a % of total MSPE.

 4 Error due to regression, as a % of total MSPE.

⁵Error due to disturbance, as a % of total MSPE.

638 ⁶Condordance correlation coefficient, where $CCC = r \times C_b$.

639 ⁷Pearson correlation coefficient.

640 ⁸Bias correction factor.

641 ⁹Scale shift.

 10 Location shift relative to the scale.

¹¹Separated into forage DMI (kg/d) + concentrate DMI (kg/d).

Source	Response Variable	Driving Variable	Eqn	Pred Mean ¹	Pred SD ¹	rMSPE, % ²	ECT, % ³	ER, % ⁴	ED, % ⁵	CCC ⁶	r^7	$C_b{}^8$	<i>v</i> ⁹	u^{10}
Vernet et al.	<u>L/kg BW/h</u>	<u>g/kg BW/d</u>												
(2009)	ARTBF	DMI	Lin	0.6	0.17	65.1	57.4	1.0	41.6	0.48	0.64	0.75	0.65	0.96
(above MN)	PORBF	DMI	Lin	2.6	0.45	26.6	62.2	1.4	36.4	0.59	0.71	0.84	0.73	0.84
	HEPBF	DMI	Lin	3.2	0.62	29.9	72.4	1.5	26.1	0.61	0.69	0.88	0.76	0.91
	PORBF/ HEPBF	DMI	Lin ¹¹	81	1.2	7.7	29.6	1.9	68.5	0.11	0.30	0.37	0.22	-1.32
Vernet et al.	L/kg BW/h	<u>g/kg BW/d</u>												
(2009) (above +	ARTBF	DMI	Quad	0.6	0.15	66.5	60.1	3.5	36.4	0.45	0.58	0.77	0.58	1.07
below MN)	PORBF	DMI	Quad	2.7	0.64	32.4	70.6	3.1	26.3	0.58	0.70	0.82	1.02	0.92
	HEPBF	DMI	Quad	3.3	0.78	34.8	79.8	0.7	19.6	0.59	0.67	0.88	0.96	0.99
	PORBF/ HEPBF	DMI	Lin ¹¹	81	0.1	7.7	22.9	9.1	68.0	0.01	0.03	0.36	0.02	-4.03
Bermingham	L/kg BW/h	g/kg BW/d												
et al. (2008)	PORBF	DMI	Lin	2.4	0.51	20.5	38.6	0.0	61.4	0.74	0.88	0.84	0.82	0.48
	PORBF	DMI	Quad	2.7	1.21	45.4	35.5	51.2	13.4	0.57	0.69	0.82	1.94	0.67
Huntington (1984)	PORBF, L/h	MEI, MJ/d	Lin	876	249.1	18.6	8.3	39.6	52.1	0.88	0.92	0.95	0.68	0.15

Table 7. Blood flow (hepatic arterial - ARTBF; portal venous - PORBF; hepatic venous - HEPBF) predictions by extant equations on
 the evaluation database for ARTBF, PORBF and HEPBF locations.

¹ Where: observed means \pm SD: ARTBF, PORBF, HEPBF (L/h): 165 \pm 137.9, 832 \pm 369.3, 996 \pm 495.9; ARTBF, PORBF, HEPBF

648 (L/kg BW/h): 0.4 ± 0.26 , 2.1 ± 0.63 , 2.5 ± 0.83 ; ARTBF, PORBF, HEPBF (L/kg BW^{0.75}/h): 1.8 ± 1.24 , 9.4 ± 2.95 , 11.2 ± 3.96 ;

649 PORBF/HEPBF (%): 85 ± 5.9 , respectively.

 2 Root mean square prediction error, % of observed mean.

 3 Error due to mean bias, as a % of total MSPE.

 4 Error due to regression, as a % of total MSPE.

⁵Error due to disturbance, as a % of total MSPE.

⁶Condordance correlation coefficient, where $CCC = r \times C_b$.

⁷Pearson correlation coefficient.

- ⁸Bias correction factor.
- ⁹Scale shift.
- ¹⁰Location shift relative to the scale. ¹¹ PORBF/HEPBF % = (100 Arterial/venous % linear prediction equation from Vernet et al. (2009)).



Figure 1. Observed portal blood flow (PORBF; top) and hepatic blood flow (HEPBF; bottom)

679 vs. DMI (kg/d) for the developmental database (δ , y) and the evaluation database (\blacksquare , y').



Figure 2. Distribution of forage % across DMI (kg/d) for the developmental (◊) and evaluation
(■) databases.



- **Figure 3.** Residual (predicted observed value) vs. predicted blood flow values for the linear
- 711 DMI based equations (Table 3) based on blood flow in L/h (left) or L/kg BW^{0.75}/h (right),
- evaluated on the evaluation database for ARTBF (a), PORBF (b) HEPBF (c) and
- 713 PORBF/HEPBF % (d), and where ARTBF hepatic arterial, PORBF portal venous and HEPBF -
- 714 hepatic venous blood flows.



- **Figure 4.** Residual (predicted observed value) vs. the forage proportion (%) of the diet for the
- 748 DMI based equations (Table 3) based on blood flow in L/h (left) or L/kg BW^{0.75}/h (right),
- evaluated on the evaluation database for ARTBF (a), PORBF (b) HEPBF (c) and
- 750 PORBF/HEPBF (d), and where ARTBF hepatic arterial, PORBF portal venous and HEPBF -
- 751 hepatic venous blood flows.



Figure 5. Residual (predicted – observed value) PORBF/HEPBF (%) vs. the forage proportion
(%) in the diet for the DMI based sheep equations of Vernet et al. (2009), for their above
maintenance equation (linear) (Top), and above plus below maintenance equation (quadratic)
(Bottom), evaluated on the evaluation database.