

# Review: Impact of protein and energy supply on the fate of amino acids from absorption to milk protein in dairy cows

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Making dairy farming more cost-effective and reducing nitrogen environmental pollution could be reached through a reduced input of dietary protein, provided productivity is not compromised. This could be achieved through balancing dairy rations for essential amino acids (EAA) rather than their aggregate, the metabolizable protein (MP). This review revisits the estimations of the major true protein secretions in dairy cows, milk protein yield (MPY), metabolic fecal protein (MFP), endogenous urinary loss and scurf and associated AA composition. The combined efficiency with which MP (Eff<sub>MP</sub>) or EAA (Eff<sub>AA</sub>) is used to support protein secretions is calculated as the sum of true protein secretions (MPY + MFP + scurf) divided by the net supply (adjusted to remove the endogenous urinary excretion: MP<sub>adi</sub> and AA<sub>adi</sub>). Using the proposed protein and AA secretions, Eff<sub>MP</sub> and Eff<sub>AA</sub> were predicted through meta-analyses (807 treatment means) and validated using an independent database (129 treatment means). The effects of MP<sub>adi</sub> or AA<sub>adi</sub>, plus digestible energy intake (DEI), days in milk (DIM) and parity (primiparous v. multiparous), were significant in all models. Models using  $(MP_{adj}, MP_{adj} \times MP_{adj})$  DEI and DEI  $\times$  DEI) or  $(MP_{adj}/DEI and MP_{adj}/DEI \times MP_{adj}/DEI)$  had similar corrected Akaike's information criterion, but the model using MP<sub>adi</sub>/DEI performed better in the validation database. A model that also included this ratio was, therefore, used to fitting equations to predict  $Eff_{AA}$ . These equations predicted well  $Eff_{AA}$  in the validation database except for Arg which had a strong slope bias. Predictions of MPY from predicted Eff<sub>MP</sub> based on MP<sub>adi</sub>/DEI, MP<sub>adi</sub>/ DEI × MP<sub>adi</sub>/DEI, DIM and parity yielded a better fit than direct predictions of MPY based on MP<sub>adj</sub>, MP<sub>adj</sub>× MP<sub>adj</sub>, DEI, DIM and parity. Predictions of MPY based on each Eff<sub>AA</sub> yielded fairly similar results among AA. It is proposed to ponder the mean of MPY predictions obtained from each Eff<sub>AA</sub> by the lowest prediction to retain the potential limitation from AA with the shortest supply. Overall, the revisited estimations of endogenous urinary excretion and MFP, revised AA composition of protein secretions and inclusion of a variable combined  $Eff_{AA}$  (based on  $AA_{adi}/DEI$ ,  $AA_{adi}/DEI \times A_{adi}/DEI$ , DIM and parity) offer the potential to improve predictions of MPY, identify which AA are potentially in short supply and, therefore, improve the AA balance of dairy rations.

Keywords: efficiency, ration, formulation, requirement, nitrogen

#### Implications

To improve the formulation of dairy rations, allowing a reduction of crude protein intake, feeding costs and nitrogen excretion into the environment, the current review proposed revisited estimations of daily true protein secretions in dairy cows and associated amino acid composition. A good prediction of milk protein yield was obtained using the predicted combined variable efficiency of utilization of absorbed amino acids based on the ratio of absorbed amino acids/digestible energy intake, days in milk and parity. This approach could help to identify which amino acids are in short supply and, therefore, improve the amino acid balance of dairy rations.

#### Introduction

With an overall objective of increasing the sustainability of dairy farms, optimizing the efficiency of utilization of protein without compromising productivity becomes a must for dairy nutritionists. Emphasis is often put on the poor efficiency of utilization of N by dairy cows to produce milk protein (milk N/N intake) averaging, for example,  $24.7 \pm 4.1\%$  and  $27.7 \pm 3.6\%$  in 736 North American and 998 North

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European treatment means, respectively (Huhtanen and Hristov, 2009). However, human-edible feed conversion efficiency (**heFCE**), calculated as the ratio of human-edible output/human-edible input, has been proposed to better evaluate the contribution of animal production to the human food chain (e.g. Wilkinson, 2011; Ertl *et al.*, 2015). In this context and considering protein, Wilkinson (2011) concluded that dairy cows were offering the most efficient animal production system in the United Kingdom; Ertl *et al.* (2015) reported heFCE for protein varying from 0.5 to slightly more than 2.0 in commercial dairy farms in Austria, whereas Broderick (2018) calculated heFCE varying between 1.4 and 2.1, depending of the production context from different countries. Therefore, dairy cows can make a valuable contribution to the human food chain with a high heFCE for protein.

Improving the overall efficiency of N utilization still remains, however, a target due to its dual impact on reducing both feeding cost and environmental impact. Dijkstra *et al.* (2013a) suggested that focusing on an optimal supply of rumendegradable protein and optimizing the efficiency of utilization of absorbed amino acids (**AA**) for milk protein synthesis would be the potential strategies available for improving N efficiency. To improve AA recommendations for dairy rations, three major points need to be tackled: (1) quantify the net supply of AA; (2) assess the fate of absorbed AA (for which functions are they used for?) and (3) determine with which efficiency the absorbed AA are used to support the identified functions.

The current review focuses on the two latter points: revisiting true protein (**TP**) secretions and associated AA secretions and identifying the major factors affecting the efficiency of utilization of absorbed AA (**Eff**<sub>AA</sub>). We will also evaluate if predicted efficiency of utilization of metabolizable protein (**MP**) supply and Eff<sub>AA</sub> are sufficiently robust to predict milk true protein yield (**MPY**). To simplify the review, only secretions and not accretions are included in AA demand: the cows are, therefore, considered at constant body weight (**BW**) and not in gestation.

#### Updates of true protein and amino acid secretions

To follow the fate of absorbed AA, the quantification of AA secreted into milk protein but also on AA 'lost' by the cow in endogenous secretions found in scurf, urine and feces is required. These endogenous AA losses as MPY remove AA irreversibly from the free AA pool. On a net basis, these exported AA need to be replaced on a timely basis by a minimal equivalent flow of digested AA: this predicted 'replacement' constitutes the basis of MP and AA recommendations which are calculated as the sum of secretions divided by an efficiency of utilization of absorbed MP or AA to support different identified secretions.

#### Scurf

*True protein secretion.* In most of the formulation models, the equation from Swanson (1977) predicting net crude protein (**CP**) requirement for scurf is used:

$$CP \, scurf_{secretion}(g/day) = 0.2 \times BW^{0.60}$$
(1)

Swanson's prediction was retained but adjusted to take into account that not all CP is TP:

$$\begin{aligned} \text{TP scurf}_{\text{secretion}}(\text{g/day}) &= 0.2 \times \text{BW}^{0.60} \times 0.86 \\ &= 0.17 \times \text{BW}^{0.60} \end{aligned} \tag{2}$$

where 0.86 represents the TP/CP ratio of scurf, based on its AA composition, detailed below, and total N content; here and throughout the text, BW is in kg.

*Amino acids.* The secretion of AA into scurf is obtained by multiplying TP scurf<sub>secretion</sub> by its AA composition, estimated using the head, hide, feet and tail composition reported by Williams (1978) and van Amburgh *et al.* (2015). The mean from these studies, corrected for incomplete recovery of AA with 24-h hydrolyses (Lapierre *et al.*, 2019), is reported on a TP basis in Table 1.

$$\begin{split} \text{AA scurf}_{\text{secretion}}(\text{g}/\text{day}) &= 0.17 \times \text{BW}^{0.60} \\ &\times [\text{AA}_{\text{corr}-\text{Scurf}}]/100 \quad \ (3) \end{split}$$

where  $(AA_{corr-Scurf})$  is in g AA/100 g TP.

**Table 1** Amino acid (AA) composition of protein secretions used in the calculation of efficiency of utilization of AA in lactating dairy cows

	g AAcorr / 100 g CP <sup>1</sup>	<u>(</u>	g AAcalc/ 100 g TP <sup>2</sup>			
AA	Duodenal endogenous	Microbial	Scurf	Whole empty body	Metabolic fecal	Milk
Ala	4.69	7.38	9.17	8.59	6.32	3.59
Arg	4.61	5.47	9.60	8.20	5.90	3.74
Asx	4.75	13.39	8.39	9.61	7.56	8.14
Cys	2.58	2.09	2.70	1.74	3.31	0.93
Gİx	11.31	14.98	14.69	15.76	15.67	22.55
Gly	5.11	6.26	21.08	14.46	8.45	2.04
His	2.90	2.21	1.75	3.04	3.54	2.92
lle	4.09	6.99	2.96	3.69	5.39	6.18
Leu	7.67	9.23	6.93	8.27	9.19	10.56
Lys	6.23	9.44	5.64	7.90	7.61	8.82
Met	1.26	2.63	1.40	2.37	1.73	3.03
Phe	3.98	6.30	3.61	4.41	5.28	5.26
Pro	4.64	4.27	12.35	9.80	8.43	10.33
Ser	5.24	5.40	6.45	5.73	7.72	6.71
Thr	5.18	6.23	4.01	4.84	7.36	4.62
Trp	1.29	1.37	0.73	1.05	1.79	1.65
Tyr	3.62	5.94	2.62	3.08	4.65	5.83
Val	5.29	6.88	4.66	5.15	7.01	6.90

 $^{1}\mathrm{g}$  AA<sub>corr</sub>: AA composition corrected to account for incomplete recovery of AA with 24-h hydrolysis; TP = true protein.

<sup>2</sup>g AA<sub>calc</sub>: AA composition calculated from the primary structure of the reference protein of each family; see text for details.

#### Endogenous urinary

True protein secretion. Most formulation models predict endogenous urinary daily protein losses according to Swanson (1977), at 2.75 g/BW<sup>0.50</sup>. To better quantify the AA required to cover this loss, a literature review was conducted to quantify the composition of urinary N. Force is to admit that literature is scarce on that domain in dairy cattle (Dijkstra et al., 2013b). The major N metabolites in endogenous urinary N losses are: urea synthesized from endogenous sources, endogenous purine derivative (PD), creatinine and creatine, hippuric acid and 3-methyl-His. Therefore, endogenous N urinary loss is not a protein secretion per se. From studies with low CP dietary intake, daily excretion of endogenous urea has been guantified as 10 mg N/BW per day (Hutchinson and Morris, 1936; Biddle et al., 1975; Marini and Van Amburgh, 2005; Wickersham et al., 2008a and 2008b). To predict creatinine excretion, a database using exclusive dairy breeds, with growing and mature animals (111 treatment means from 24 publications from 1979 to 2015: Supplementary Material S1) was built. Urinary excretion of creatinine was regressed to 25.5 mg creatinine/BW per day, representing  $9.46 \pm 0.157$  mg N/BW per day. Creatine excretion was evaluated as 0.37 that of creatinine (Blaxter and Wood, 1951; Nehring et al., 1965; Bristow et al., 1992). Urinary excretion of endogenous PD was assumed to average 27.1 mg N/BW<sup>0.75</sup> per day (483  $\mu$ mol/ BW<sup>0.75</sup>: reviews from Tas and Susenbeth, 2007; Fujihara and Shem, 2011). Daily urinary excretion of 3-methyl-His ( $\mu$ mol) was evaluated at 50.4 + 3.54 × BW (Harris and Milne, 1981).

Using the database from Spek et al. (2013), the 'measured' endogenous urinary N excretion was calculated as non-urea urinary excretion plus endogenous urea (predicted as described above) minus estimation of PD derived from absorbed microbial protein (Chen and Gomes, 1992). The sum of predictions described above represented 54% of the 'measured' endogenous urinary N excretion. As previously mentioned, hippuric acid is another N metabolite excreted in urine. Hippuric acid is formed in the liver to detoxify benzoic acid originating from rumen fermentation of dietary phenolic compounds. Although this excretion cannot be purely defined as 'endogenous', it has probably been included in previous predictions of endogenous urinary N excretion. When determined, it averaged 25.7% of non-urea N urinary excretion (Nehring et al., 1965; Bristow et al., 1992; Kool et al., 2006). Including hippuric acid to the 'endogenous' urinary excretion, the calculated v. 'measured' values from Spek's database were not different (29.3  $\pm$  4.5 v.  $33.4 \pm 15.4$  g N/day), but there was a strong slope bias. The potential hippuric acid excretion was best related, in the database, to the proportion of urea N in urinary N excretion. Using this relationship to evaluate hippuric acid excretion, the calculated endogenous urinary N excretion averaged  $33.2 \pm 11.2$  g N/day compared with 33.4 g N/day for the 84 treatment means 'measured' as described above. Although smaller, there remained a slope bias that could not be corrected, indicating an important gap in our knowledge

#### Protein & energy affect efficiency of amino acids

on the composition of urinary N excretion. Besides hippuric acid, most of the estimations of urinary excretion were based on BW, and the sum of endogenous urinary N excretions was expressed relative to BW, averaging 53 mg N/BW (or 0.053 g N/BW) per day. Using a totally different approach, the prediction of daily endogenous urinary N loss averages 50 mg N/BW in the formulation model of the Institut National de Recherche Agronomique (INRA, 2018), very similar to our prediction and roughly twice as large as the Swanson (1977) prediction. As these compounds are expressed in g N/day, there is no need to convert from CP to TP. Therefore,

TP endogenous urinary<sub>secretion</sub>  $(g/day) = 0.33 \times BW$  (4)

where 0.33 is derived from  $0.053 \times 6.25$ .

Amino acid composition. The reason for revisiting endogenous urinary excretion was to identify which AA were upstream of these urinary excretions. After the examination of metabolic pathways yielding each of these urinary excreted compounds, only endogenous urea and 3-methyl-His excretions require a direct input of essential AA (EAA), if we exclude Arg from true EAA. Indeed, endogenous PD are synthesized from Asp, Gln and Gly; creatine and creatinine from Arg and Gly (it requires S-adenosyl Met, but as for other metabolic pathways, this does not represent a net Met requirement); and hippuric acid is synthesized from Glv. Endogenous urea excretion is assumed to have a revisited whole empty body AA composition (Williams, 1978; Rohr and Lebzien, 1991; Ainslie et al., 1993; Van Amburgh et al., 2015); the mean of these studies, corrected for incomplete recovery of AA with 24-h hydrolyses, is reported on a TP basis in Table 1. Therefore, to determine AA secretion in endogenous urinary N output, we need first to calculate endogenous urea excretion,

TP endogenous urinary-urea<sub>secretion</sub> (g/day)  
= 
$$0.0625 \times BW$$
, (5)

where 0.0625 was derived from 0.010 g N/day  $\times$  6.25; and multiply this secretion by the corresponding AA<sub>corr</sub> composition, assumed to be that of whole empty body (Table 1).

AA endogenous urinary<sub>secretion</sub>(g/day)

$$= 0.0625 \times \text{BW} \times \left[\text{AA}_{\text{corr}-\text{WholeEmptyBody}}\right]/100 \quad (6)$$

where  $(AA_{corr-WholeEmptyBody})$  is in g AA/100 g TP.

To complete the estimation of His excretion in endogenous urinary loss, 3-methyl His urinary excretion, as described above (mg His/day =  $7.82 + 0.55 \times BW$ ), needs to be added. And finally, to complete the estimation of contribution of Arg to urinary N excretion, we need to include its contribution to creatinine and creatine, that is,  $0.052 \times BW$  g Arg/day.

#### Metabolic fecal protein secretion

True protein secretion. This is certainly protein secretion with the largest discrepancy in its prediction varying, for example, between 337 and 621 g/day for a cow eating 23 kg/day of a 16.5% CP diet when predicted with five formulation models (Lapierre et al., 2018). Reasons for this high discrepancy are inherent to the difficulty in performing its measurements and to the ambiguous definition of MFP. First, the determination of MFP in ruminants cannot be done as simply as in monogastrics where MFP is measured in animals fed an N-free diet. Predictions of MFP in ruminants have been based on regressing intake of digestible CP on CP intake with the negative intercept estimated as MFP, averaging, for example, 34, 33 and 29 g CP/kg DM intake (DMI) in earlier studies (Holter and Reid, 1959; Waldo and Glenn, 1984) comparable to 32 and 27 g CP/kg DMI reported in more recent studies (Jonker et al., 1998; Kauffman and St-Pierre, 2001). In a meta-analysis using 65 growing-finishing cattle studies (291 treatment means) and 43 dairy cow studies (164 treatment means), Marini et al. (2008) obtained an intercept of 30 g CP/kg DMI when ignoring the multidimensionality of the relationship with other parameters. Predictions of MFP have also been calculated subtracting predicted undigested feed N from fecal N when animals were fed low CP diets (29.4 g CP/kg DMI; Swanson, 1977). However, it has been demonstrated that MFP losses are related more closely to feces output than to feed intake (Swanson, 1977): for this reason, some formulation models have based their estimation of MFP on indigestible DM (CNCPS; Fox et al., 2004) or the outflow of organic matter from the digestive tract (NorFor, 2011; INRA, 2018). However, because of the uncertainty related to the estimation of DM digestibility, the National Research Council (2001) predicted MFP based on DMI.

However, the values obtained from the methods described above are not strictly a measure of loss of true protein from endogenous origin. Indeed, it was already raised at the beginning of the 1980s that these estimations of MFP included bacteria and bacterial debris (Swanson, 1982) and then the question 'Is the source of bacteria primarily waste N rather than a metabolic cost to the animal?' was raised (question from Trenkle: Swanson, 1982). Indeed, the metabolic demand for MFP should be only derived from endogenous secretions originating directly from AA (either from arterial supply and small intestinal digestion) and not from urea recycled into microbial protein. Therefore, NRC (2001), recognizing that a part of this fecal material contains undigested ruminal microbial CP, assumed that 50% of indigestible microbial protein appears in the feces and should be excluded from initial MFP prediction.

To improve the estimation of endogenous secretions through the gut in dairy cows, Ouellet *et al.* (2002 and 2010) adapted an isotopic dilution approach used in pigs (e.g. Lien *et al.*, 1997) and sheep (Sandek *et al.*, 2001). This approach allowed the development of a model delineating the contribution of undigested rumen bacteria synthesized from endogenous secretions or from urea to fecal N, thus

allowing the exclusion of the latter from MFP. Because endogenous proteins have multiple origins (saliva, gastric juices, bile, pancreatic secretions, sloughed epithelial cells and mucin: Tamminga et al., 1995), it is a challenge to determine the isotopic enrichment of the precursor pool when using a dilution approach. Values obtained using the enrichment of the mucosa as representative of endogenous secretions have been retained for this revision. Furthermore, the metabolic cost of the loss of undigested endogenous secretion across the upper gut should be measured at the ileum, because the endogenous secretions flowing out of the small intestine and disappearing across the hindgut do not result in absorbed AA. Endogenous fecal loss was, therefore, adjusted using a factor of 1.13, representing the ratio of ileal endogenous flow divided by fecal endogenous flow, measured in dairy cows (Ouellet et al., 2007). Therefore, the prediction of endogenous ileal flow calculated using the enrichment of gut mucosa was retained as a basis for the estimation of MFP, averaging 14.9 g CP/kg DMI (Lapierre et al., 2007).

Although MFP is calculated relative to DMI, as discussed above, it has been recognized that the driving force of MFP should be indigestible DM (Swanson, 1977). However, because of the uncertainties associated with estimating DM digestibility, we propose to still use DMI as a basis to predict MFP, but to include the neutral detergent fiber of the ration (NDF, %DM) based on Marini et al. (2008) regression of total tract digestibility of N v. N content of the diet. This inclusion will partially account for diet DM digestibility. The Marini equation also included a carbohydrate fermentation rate (fast, medium or none), but for practical purposes and to remove subjectivity, the average of three values was used to derive the final equation. Therefore, the equation of Marini et al. (2008) was adjusted to yield the value mentioned above at 14.9 g CP/kg DMI for cows fed diets at 36% NDF (in the rations in Ouellet et al., 2002, 2007 and 2010). In addition, endogenous secretions occurring across the hindgut also create a demand on AA as demonstrated in pigs (Zhu et al., 2003). Based on observations in sheep (Sandek et al., 2001), this demand was estimated as 60% of N ileal flow of small intestinal endogenous secretion, the latter averaging 5.1 g CP/kg DMI in dairy cows (Ouellet et al., 2007). Due to the scarcity of data on the exact origin of this hindgut N, it is assumed that half of this input originated from endogenous proteins and the other half from urea. Therefore, the estimation of MFP excretion (q CP/day) =  $(11.62 + 0.134 \times$  $NDF_{\%DM}$  × DMI. Note that we keep the term metabolic fecal protein, although the small intestinal loss was truly predicted at the ileum. Based on its AA composition, detailed below, and N content, a ratio of 0.73 for TP/CP of MFP is calculated and

$$\mathsf{TP}\,\mathsf{MFP}_{\mathsf{secretion}}(\mathsf{g}/\mathsf{day}) = (8.5 + 0.1 \times \mathsf{NDF}_{\%\mathsf{DM}}) \times \mathsf{DMI} \tag{7}$$

where  $NDF_{\%DM}$  is the percentage of NDF in the ration; here and throughouot the text, DMI is in kg/day.

Amino acid composition. The AA composition of MFP is based on the AA composition of ruminal and abomasal isolates from Ørskov *et al.* (1986) and the endogenous flow at the ileum in pigs (Jansman *et al.*, 2002), assuming that 70% of MFP is from undigested endogenous duodenal flow and the remaining 30% from the intestine (Ouellet *et al.*, 2002 and 2010). The averaged composition corrected for incomplete recovery of AA with 24-h hydrolyses is reported on a TP basis in Table 1. Therefore, individual AA secretion in MFP is calculated as:

$$\begin{split} \text{AA MFP}_{\text{secretion}}(\text{g/day}) &= [(8.5 + 0.1 \times \text{NDF}_{\text{\%DM}}) \times \text{DMI}] \\ &\times [\text{AA}_{\text{corr}-\text{MFP}}]/100 \end{split}$$

where  $(AA_{corr-MFP})$  is in g AA/100 g TP.

#### Milk

*True protein secretion.* Milk true protein secretion is certainly the most accurate measurement of export protein to make. The factor used to convert the measured milk N concentration into CP varies between 6.34 and 6.39. Based on the AA composition of milk protein, 6.34 would be the best factor (Karman and van Boekel, 1986; authors' calculations), but using different factors only has a limited impact on the estimation of MPY, smaller than 1%. Similar to other protein secretions, it has to be expressed as TP. If the TP/CP ratio is not known, NPN content of milk is assumed to be 4.9% (DePeters and Cant, 1992).

Amino acid composition. Although critical in the definition of AA requirements, milk AA composition has not been recently investigated. Indeed, early studies reported (1) that the EAA composition of milk produced from cows fed urea and ammonium N as the sole source of N differed by <3% from the EAA composition of milk from control cows (Syvaöja and Virtanen, 1965), and (2) that a change in the forage-grain ratio of the ration did not alter the AA composition of milk (Featherston et al., 1964). From these observations, it has been assumed that the AA composition of milk protein is fairly constant, and this dogma has not been really challenged. Therefore, milk AA composition is still assumed to be constant, although this issue might need to be readdressed with improved techniques to measure AA concentration. The same approach as that used in Swaisgood (1995) has been adopted to determine the AA composition of milk TP. Milk AA composition has been calculated based on the primary structure of reference protein of each family as detailed by Farrell et al. (2004). Based on the distribution of milk proteins reported in 15 manuscripts published between 1980 and 2012 (Supplementary Material S2), protein fractions in milk TP were assumed to be 82.4% casein (as a percentage of total protein:  $35.2\% \alpha s1$ -casein; 7.6% $\alpha$ s2-casein; 30.9%  $\beta$ -casein; 8.7%  $\kappa$ -casein) and 17.6% whey (as a percentage of total protein: 3.7%  $\alpha$ -lactalbumin; 10.5% β-lactoglobulin; 1.04% albumin; 1.64% lgG1; 0.21%

IgG2; 0.04% IgA; 0.33% IgM; 0.21% lactoferrin). The AA composition of milk protein calculated using this procedure is presented in Table 1. Therefore, individual AA secretion in milk protein is calculated as:

$$\mathsf{AA}\,\mathsf{Milk}_{\mathsf{secretion}}(\mathsf{g}/\mathsf{day}) = \mathsf{MPY}(\mathsf{g}/\mathsf{day}) \times [\mathsf{AA}_{\mathsf{calc}-\mathsf{Milk}}]/100 \tag{9}$$

where  $AA_{calc-Milk}$  is in g/100 g TP.

Using this approach for milk, there is no need to do any correction for potential loss due to an incomplete recovery with hydrolyses. Due to the high diversity of proteins included in other types of secretions, an approach similar to milk protein cannot be used for these former proteins; therefore, the only way to obtain their AA composition is by hydrolysis. To correctly sum the AA in protein secretions, we used corrected AA concentrations for all protein secretions and calculated AA concentrations for milk.

### Efficiency of utilization of metabolizable protein and amino acids

#### Variable efficiency

It is recognized that the efficiency of utilization of MP (Eff<sub>MP</sub>) to support MPY is not fixed. Indeed, the marginal recovery of abomasal casein infusions averaged 21%, ranging from -5% to 45% in seven studies (Hanigan *et al.*, 1998), far below the traditional fixed efficiency of lactation of 65% to 67%. More recently, the marginal recovery of 81 comparisons of MPY response to post-rumen casein infusions averaged 24%, and was negatively related to the MP balance of control treatment (Martineau *et al.*, 2017). A similar trend is observed when variation in MP supply is achieved through a dietary change. For example, Metcalf *et al.* (2008) reported that the efficiency of lactation decreased from 77% to 50% when MP supply varied from 25% below to 25% above requirements, the efficiency for maintenance requirement assumed to be fixed.

In addition to MP supply per se, energy supply also has an impact on Eff<sub>MP</sub>. Increments in MPY have been observed in response to post-ruminal supply of energy, either as glucose (Vanhatalo *et al.*, 2003a; Nichols *et al.*, 2016), propionate (Raggio *et al.*, 2007) or dietary rumen-inert fat (Nichols *et al.*, 2018), although not always (e.g. Clark *et al.*, 1977; Vanhatalo *et al.*, 2003b). Obviously, MPY increment in response to increased post-rumen energy supply (no effect on MP supply) increased Eff<sub>MP</sub>. Using 825 treatment means, Daniel *et al.* (2016) concluded that both MP and net energy of lactation (**NE**<sub>L</sub>) supplies increased MPY, the effects being additive, as observed in most of the individual studies where the interaction was tested. Only the study of Brun-Lafleur *et al.* (2010) reported a protein × energy interaction with a very targeted experimental design.

Therefore, there was enough evidence of the good use of a variable  $Eff_{MP}$ , related to both MP and energy supplies. Currently, NorFor (2011) and the DVE/OEB

(Van Duinkerken *et al.*, 2011) system are using a fixed efficiency for maintenance and a variable efficiency for lactation. Although not estimating directly  $Eff_{MP}$ , INRA (2018) is using MP and NE<sub>L</sub> supplies to predict MPY, thereby introducing a variable  $Eff_{MP}$  for non-productive and lactation functions. Introducing a variable  $Eff_{MP}$  in the formulation models was yielding a better prediction of MPY in response to variations in MP and/or energy supply than the use of a fixed  $Eff_{MP}$ , still in use in most current North American models (Lapierre *et al.*, 2018).

#### Combined efficiency

Based on the observed metabolism of AA across tissues, it has been proposed to use a single Eff<sub>AA</sub>, different for each AA but identical for all the protein functions (Lapierre et al., 2007). The reason for this suggestion is that EAA catabolism does not occur at the site of protein synthesis or protein secretion but does occur in the organ(s) where appropriate enzymes are present (Lobley and Lapierre, 2003). For example, mammary uptake of Group 2 AA (Ile, Leu, Lys and Val) is in excess of MPY and the excess increases with MP supply; in contrast, Group 1 AA (His, Met, Phe+Tyr and Trp) net mammary uptake is almost equivalent to their secretion into milk protein (Lapierre et al., 2012). Therefore, it is proposed to use a single Eff<sub>MP</sub> and Eff<sub>AA</sub> (one for each EAA) for all the protein functions, except endogenous urinary loss. The latter represents end-products of metabolic pathways, and an efficiency of 1.0 should be used (Sauvant et al., 2015). Sauvant et al. (2015) reported that Eff<sub>MP</sub> was better predicted when the same efficiency was assigned to all protein functions rather than a fixed efficiency for the nonproductive functions and a variable efficiency for lactation.

With the objective of balancing dairy rations on an AA basis rather than MP basis, we developed equations to predict  $Eff_{MP}$  and  $Eff_{AA}$  in relation to MP or AA and energy supplies, which has not been done yet. Variable  $Eff_{AA}$  has already been proposed, but solely related to AA supply (Doepel *et al.*, 2004).

#### Calculation of efficiency

MP supply and AA net digestible flow were calculated as the sum of digestible flow from rumen-undegraded protein (**RUP**) flow and microbial protein; endogenous duodenal flow was not included. The model of White *et al.* (2017) was used to predict RUP flows, with the AA composition of RUP assumed to be the AA composition of feed ingredients; the equation from Roman-Garcia *et al.* (2016) was used to predict microbial N, with an adjustment proposed by Myers *et al.* (2018), converted to microbial true protein assuming 16% N, and using the TP/CP (82.4%) ratio and the AA composition from Sok *et al.* (2017).

Using the supplies and secretions described above, the combined  $\text{Eff}_{\text{MP}}$  or  $\text{Eff}_{\text{AA}}$  was calculated as follows:

$$Eff_{MP} = (TP scurf_{secretion} + TP MFP_{secretion} + MPY) / MP supply_{adj}$$
(10)

where MP  $supply_{adj} = MP supply - TP$  endogenous  $urinary_{secretion}$ 

#### and

$$\begin{split} \mathsf{Eff}_{\mathsf{A}\mathsf{A}} &= (\mathsf{A}\mathsf{A}\operatorname{scurf}_{\mathsf{secretion}} + \mathsf{A}\mathsf{A}\operatorname{\mathsf{MFP}}_{\mathsf{secretion}} \\ &+ \mathsf{A}\mathsf{A}\operatorname{\mathsf{Milk}}_{\mathsf{secretion}})/\mathsf{A}\mathsf{A} \text{ net digestible flow}_{\mathsf{adj}} \end{split} \tag{11}$$

where AA net digestible flow<sub>adj</sub> = AA net digestible flow – AA endogenous urinary<sub>secretion</sub>. In the text, MP supply<sub>adj</sub> will be referred to as MP<sub>adj</sub>, and AA net digestible flow<sub>adj</sub> as AA<sub>adj</sub>.

#### Prediction of efficiency

Databases. The calculations described above were applied to two databases, one used for the development of models, and the second for their validation. The developmental database included 208 publications (807 treatment means) and was an extension of the database used by Roman-Garcia et al. (2016) with studies added to offer a wider range of AA supply. An independent validation database was also built, including 32 publications (129 treatment means). The summary statistics of developmental and validation databases are presented in Tables 2 and 3, respectively. Publications included in the developmental database and in the validation database are listed in Supplementary Material S3. In both databases, studies have been coded to look specifically at the increment of MP supply. The relationship between MPY and MP supply depicts the overall meta-design (Figure 1).

Statistics. Models predicting Eff<sub>MP</sub> were developed using variables that were strong predictors of the sum of protein secretions when tested individually:  $MP_{adj}$ , digestible energy intake (**DEI**), days in milk (**DIM**) and parity (primiparous *v*. multiparous). Digestible energy was used because metabolizable energy requires the estimation of urinary N, unknown until the efficiency is predicted, and net energy requires, in addition, the quantification of MPY: both are unknown that we are trying to predict. Digestible energy was predicted based on nutrient digestibility (Daley *et al.*, 2018; de Souza *et al.*, 2018). Models tested the linear and quadratic effects of (1) MP<sub>adj</sub>, (2) MP<sub>adj</sub> and DEI and (3) MP<sub>adj</sub>/DEI; DIM and parity were tested in all models.

Models were developed using the *rma.mv* function from the metafor package in R. Potential outlying and influential observations and studies were detected using the *rstandard*, the *rstudent* and the *cook.distance* functions in the metafor package. All relationships were graphed and evaluated using the *ggplot* function of the ggplot2 package for R (Wickham, 2016). Using the *rma.mv* function of metafor, the hierarchy of studies, as a random effect, was taken into account. For example, two or more different studies could be reported in the same experiment; therefore, data were fitted to a three-level mixed-effect meta-regression model. To weigh data by  $\sqrt{N}$ , the V argument was set to zero and the R argument was used to specify a known matrix,

Variable	Mean	SD	Minimum	Maximum
Days in milk	135	56.4	28	344
BW (kg)	606	49.7	479	788
Year of publication	2002	8.5	1974	2019
DM intake (kg/day)	21.1	3.53	9.1	31.8
MP supply (g/day)	2072	415.9	997	3393
MP <sub>adi</sub> (kg/day)	1873	409.7	798	3157
DE intake (MJ/day)	275	47.9	120	421
MP <sub>adi</sub> /DE intake (g/MJ)	6.8	0.88	4.4	9.6
Digestible flow of Arg (g/day)	111	25.3	50	181
Digestible flow of His (g/day)	50	11.3	22	103
Digestible flow of Ile (g/day)	122	23.1	62	191
Digestible flow of Leu (g/day)	190	42.6	86	396
Digestible flow of Lys (g/day)	159	30.1	79	248
Digestible flow of Met (g/day)	48	9.7	23	83
Digestible flow of Phe (g/day)	120	24.8	59	203
Digestible flow of Thr (g/day)	110	20.9	56	176
Digestible flow of Trp (g/day)	27	5.6	13	44
Digestible flow of Val (g/day)	130	25.3	65	225
Sum of protein secretions (g/day)	1198	235.1	512	1803
Milk true protein yield (g/day)	945	207.1	389	1522
Metabolic fecal true protein (g/day)	245	38.5	113	382
Scurf true protein (g/day)	8.0	0.39	6.9	9.4
Endogenous urinary true protein loss (g/day)	201	16.4	159	261
Efficiency of utilization of MP <sub>adi</sub>	0.65	0.102	0.40	1.06
Efficiency of utilization of Arg	0.72	0.210	0.31	2.44
Efficiency of utilization of His	0.78	0.141	0.37	1.27
Efficiency of utilization of Ile	0.61	0.088	0.38	0.89
Efficiency of utilization of Leu	0.67	0.110	0.31	1.05
Efficiency of utilization of Lys	0.67	0.104	0.36	1.06
Efficiency of utilization of Met	0.71	0.113	0.37	1.10
Efficiency of utilization of Phe	0.54	0.084	0.31	0.81
Efficiency of utilization of Thr	0.58	0.079	0.36	0.85
Efficiency of utilization of Trp	0.77	0.125	0.42	1.26
Efficiency of utilization of Val	0.65	0.098	0.36	0.97

**Table 2** Summary statistics of the developmental database (n = 807) from studies conducted in lactating dairy cows

MP = metabolizable protein;  $MP_{adj}$  = metabolizable protein supply minus endogenous urinary loss; DE = digestible energy.

that is, diag(1/developmental database\$Nexp0.5), with an unknown multiplicative variance component which was then estimated by metafor (Viechtbauer, 2018). Unbiased estimates of fixed effects and valid estimates of SE were obtained using the robust function in the metafor package with publication as the clustering variable. The use of robust function does not change the weight matrix but only affects the way the variance-covariance matrix of the fixed effects and downstream SE and P values are computed (Viechtbauer, 2017a and 2017b). Model performance was evaluated using RMSE and root mean squared prediction error (RMSPE) for the development and validation databases as a percentage of the observed mean. Mean bias and slope bias, which are two MSE decomposition terms, were also computed and expressed as a percentage of MSE (Theil, 1966; Bibby and Toutenburg, 1978). Concordance correlation coefficients (CCC; Lin, 1989 and 1992) and the corrected Akaike's

information criterion (**AICc**; Hurvich and Tsai, 1993) are also reported. Ideal models are those with RMSE closest to 0, CCC closest to 1, mean and slope biases closest to 0, and smallest AICc.

Factors affecting efficiency. We initially tested the models using  $MP_{adj}$  as a surrogate of individual  $AA_{adj}$  to delineate which model(s) would yield the best goodness of fit. Three models are reported in Table 4: Eff<sub>MP</sub> as a function of  $MP_{adj}$  and its squared term (equation 12); the latter plus DEI and its squared term (equation 13); and the ratio of  $MP_{adj}/DEI$  and its squared term (equation 14); DIM and parity were significant in the three models. The inclusion of DEI, either as an independent term (equation 13) or as a ratio with  $MP_{adj}$  (equation 14), improved the goodness of fit compared with MP terms alone (Figure 2), as shown by a substantial reduction of AICc, a large decrease in slope bias

Variable	Mean	SD	Minimum	Maximum
Days in milk	124	58.2	11	273
BŴ (kg)	598	59.8	442	704
Year of publication	2002	7.8	1990	2018
DM intake (kg/day)	21.6	3.66	10.8	29.8
MP supply (g/day)	2105	393.2	1142	3252
MP <sub>adi</sub> (kg/day)	1908	384.1	943	3052
DE intake (MJ/day)	281	48.9	138	382
MP <sub>adi</sub> /DE intake (g/MJ)	6.8	0.69	5.3	8.8
Digestible flow of Arg (g/day)	110	22.7	58	173
Digestible flow of His (g/day)	49	10.0	26	77
Digestible flow of Ile (g/day)	124	22.6	69	186
Digestible flow of Leu (g/day)	195	38.5	97	303
Digestible flow of Lys (g/day)	159	28.8	88	236
Digestible flow of Met (g/day)	48	8.1	25	70
Digestible flow of Phe (g/day)	121	22.9	65	190
Digestible flow of Thr (g/day)	112	20.4	62	170
Digestible flow of Trp (g/day)	27	5.4	15	45
Digestible flow of Val (g/day)	131	24.5	73	203
Sum of protein secretions (g/day)	1206	256.4	655	1674
Milk true protein yield (g/day)	948	225.0	499	1390
Metabolic fecal true protein (g/day)	251	39.1	147	342
Scurf true protein (g/day)	7.9	0.49	6.6	8.7
Endogenous urinary true protein loss (g/day)	198	19.8	146	233
Efficiency of utilization of MP <sub>adj</sub>	0.64	0.079	0.41	0.82
Efficiency of utilization of Arg	0.70	0.144	0.39	1.18
Efficiency of utilization of His	0.77	0.100	0.50	1.03
Efficiency of utilization of Ile	0.60	0.074	0.37	0.78
Efficiency of utilization of Leu	0.65	0.090	0.42	0.84
Efficiency of utilization of Lys	0.66	0.081	0.41	0.89
Efficiency of utilization of Met	0.71	0.100	0.45	0.95
Efficiency of utilization of Phe	0.53	0.067	0.35	0.69
Efficiency of utilization of Thr	0.57	0.064	0.38	0.73
Efficiency of utilization of Trp	0.77	0.100	0.47	1.03
Efficiency of utilization of Val	0.65	0.080	0.41	0.84

**Table 3** Summary statistics of the validation database (n = 129) from studies conducted in lactating dairy cows

 $\label{eq:metabolizable} \begin{array}{lll} \mathsf{MP}=\mathsf{metabolizable} & \mathsf{protein}, & \mathsf{MP}_{\mathsf{adj}}=\mathsf{metabolizable} & \mathsf{protein}, & \mathsf{supply}, & \mathsf{minus}, & \mathsf{endogenous}, & \mathsf{urinary}, & \mathsf{loss}; \\ \mathsf{DE}=\mathsf{digestible}, & \mathsf{energy}. \end{array}$ 

in both the developmental and validation databases and an increased CCC (Table 4). The improvement of the relationship of Eff<sub>MP</sub> with MP<sub>adi</sub>/DEI compared with MP<sub>adi</sub> can also be visually appreciated in Figure 3. This agrees with a better prediction of MPY when energy supply is included in the model than based only on MP supply (Doepel et al., 2004) or with a final model which includes both protein and energy supplies (Daniel et al., 2016). Although equation (13) yielded a slightly lower AICc than equation (14), in the validation database, the slope bias was larger with the former equation and CCC was lower. Currently, two European systems, the DVE/OEB (Van Duinkerken et al., 2011) and NorFor (2011), are using the ratio of MP/NE<sub>L</sub> available for milk to predict the efficiency of lactation and MPY; Sauvant et al. (2015) estimated that Eff<sub>MP</sub> was better related to MP/DMI than MP/NE<sub>1</sub>. Another main advantage of using the ratio is the practicality of transferring results to 'cows of the future' eating more and

producing more than cows from the studies included in the current review. In theory, predictive equations should be used within the range of values of predictors used for their development. Even if cows in commercial farms are eating more than observations in the developmental database, the ratio MP<sub>adi</sub>/DEI remains within the limits of current observations, whereas MP<sub>adi</sub> and DEI of high-producing dairy cows are already higher than the maxima observed in the developmental database. Therefore, we decided to use MP<sub>adi</sub>/DEI as the driving force of Eff<sub>MP</sub> and apply the same concept to individual  $AA_{adj.}$  Results for Eff<sub>AA</sub>, equations (15) to (24), are presented in Table 5. All estimates were highly significant (P < 0.01) for all variables except parity (P < 0.05). Globally, the trends were very similar for the estimation of Eff<sub>AA</sub> compared with Eff<sub>MP</sub>, that is, very low mean and slope bias in the developmental database and a mean bias of  $\pm 5\%$ MSE in the validation database. In both databases, His and

#### Protein & energy affect efficiency of amino acids



Figure 1 Relationship between milk true protein yield and metabolizable protein (MP) supply in lactating dairy cows in the developmental and validation databases.

Table 4	Models of efficiency	of utilization o	of metabolizable	protein (MP)1	in lactating dai	ry cows
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	Ec	quation (12)	)	Eq	uation ( <mark>13</mark> )		Equation (14)		
ltem <sup>2</sup>	Estimate	SE	P value	Estimate	SE	P value	Estimate	SE	P value
Intercept	135	8.8	<0.001	83	8.9	<0.001	190	12.0	<0.001
MP <sub>adi</sub>	-44.5	8.31	<0.001	-62.7	7.84	<0.001			
$MP_{adj} \times MP_{adj}$	5.6	2.01	0.006	8.3	1.86	<0.001			
DEI				0.40	0.082	<0.001			
DEI × DEI				-0.00044	0.00014	0.002			
MP <sub>adi</sub> /DEI							-26.7	3.43	<0.001
$MP_{adj}$ /DEI $\times$ MP <sub>adj</sub> /DEI							1.31	0.246	<0.001
DIM	-0.051	0.0223	0.02	-0.037	0.0111	<0.001	-0.035	0.0086	<0.001
Parity	-10.8	1.65	<0.001	-5.3	1.28	<0.001	-2.5	1.20	0.04
AICc	4662			4407			4452		
Developmental database ( $n = 80$	)7)								
Observed mean	65.1			65.1			65.1		
Predicted mean	64.9			65.1			65.1		
RMSE (% observed mean)	14.9			11.1			10.7		
Mean bias (% MSE)	0.1			0.0			0.0		
Slope bias (% MSE)	21.5			2.1			0.4		
CCC		0.55			0.70			0.71	
Validation database ( $n = 129$ )									
Observed mean	63.5			63.5			63.5		
Predicted mean	65.3			65.7			65.4		
RMSPE (% observed mean)	16.0			11.5			10.4		
Mean bias (% MSE)	3.0			8.8			8.0		
Slope bias (% MSE)	41.1			9.1			3.4		
CCC	0.27			0.51			0.58		

AICc = corrected Akaike's information criterion; CCC = concordance correlation coefficient; RMSPE = root mean squared prediction error.

<sup>1</sup>Efficiency of utilization of MP = (true protein secretion in milk + scurf + metabolic fecal protein)/MP<sub>adj</sub>  $\times$  100.

 $^{2}$ MP<sub>adj</sub> (kg/d): metabolizable protein supply minus endogenous urinary loss; DEI (MJ/d): digestible energy intake; DIM: days in milk; parity (1 = primiparous; 0 = multiparous).



Figure 2 Observed (•) and residual ( $\blacktriangle$ ) values of efficiency of utilization of metabolizable protein in the function of efficiency predicted according to equations (12), (13) and (14) (see text for details of the equations) in lactating dairy cows in the validation database.



**Figure 3** Relationship between efficiency of utilization of metabolizable protein (MP) and MP adjusted (MP supply minus endogenous urinary loss) or the ratio of MP adjusted/digestible energy (DE) intake in lactating dairy cows in the developmental database. Efficiency is calculated as (true protein in milk + scurf + metabolic fecal)/MP adjusted.

Trp were used with the highest efficiency, and Phe and Thr with the lowest. Values are in the same range as previously reported for combined efficiency (Lapierre *et al.*, 2007; Van Amburgh *et al.*, 2015) except for Arg being much higher and Lys and Met being lower. The large difference for Arg is due to the change in the prediction of AA in endogenous urinary loss: the current proposition involves an important loss of Arg related to creatine and creatinine urinary excretion decreasing substantially Arg<sub>adi</sub>. Also, Arg displayed high

CCC but had a large slope bias in the validation database, probably related to the uncertainty of its true supply due to unknown and unaccounted supply from *de novo* synthesis. In fact, for this reason, although given for a comparison, the current estimates for Arg should not be used. It is also important to note that the current estimates could only be used to predict efficiency until the minimal predicted efficiency is reached according to the quadratic function: after that threshold ratio of AAadj/DEI, the function will not apply.

AA	Arg	His	lle	Leu	Lys	Met	Phe	Thr	Trp	Val
ltem <sup>2</sup>	Equation (15)	Equation (16)	Equation (17)	Equation (18)	Equation (19)	Equation (20)	Equation (21)	Equation (22)	Equation (23)	Equation (24)
Intercept	254***	207***	170***	172***	183***	178***	134***	167***	227***	181***
AA <sub>adi</sub> /DEI	-1030***	-1009***	-355***	-204***	-285***	-811***	-240***	-404***	-2258***	-53***
$AA_{adi}/DEI \times AA_{adi}/DEI$	1331***	1650***	265***	83***	155***	1256***	146***	346***	7754***	243***
DIM	-0.039***	-0.037***	-0.035***	-0.037***	-0.038***	-0.042***	-0.028***	-0.027***	-0.041***	-0.035***
Parity	-2.0\$	-2.6*	-2.4*	-2.6*	-2.7*	-3.4*	-2.0*	-2.2*	-2.9*	-2.7*
Developmental database ( $n = 8$	07)									
Observed mean	71.8	78.1	60.6	67.2	66.6	70.9	54.0	58.0	76.6	65.4
Predicted mean	72.1	78.1	60.6	67.3	66.6	70.9	53.9	58.0	76.5	65.2
RMSE (% observed mean)	12.3	10.3	11.0	10.9	10.9	11.4	10.5	9.7	10.7	10.6
Mean bias (% MSE)	0.1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Slope bias (% MSE)	1.4	0.2	0.6	0.2	0.8	1.3	0.6	0.6	0.8	0.4
CCC	0.91	0.81	0.62	0.72	0.70	0.69	0.71	0.68	0.75	0.68
Validation database ( $n = 129$ )										
Observed mean	70.1	77.3	59.6	64.8	66.0	70.7	53.2	57.0	77.2	64.5
Predicted mean	72.3	79.7	61.4	67.0	67.9	72.8	54.8	58.5	79.2	66.3
RMSPE (% observed mean)	11.4	10.0	10.4	10.6	10.7	11.4	10.2	9.1	10.1	10.2
Mean bias (% MSE)	7.4	9.6	7.8	9.6	8.0	7.2	8.3	7.8	6.8	7.5
Slope bias (% MSE)	27.9	4.6	2.2	0.9	2.4	0.0	1.8	3.4	6.0	2.2
CCC	0.87	0.67	0.55	0.64	0.51	0.55	0.60	0.60	0.66	0.58

Table 5 Models of efficiency of utilization of individual amino acids (AA)<sup>1</sup> in lactating dairy cows

CCC = concordance correlation coefficient; RMSPE = root mean squared prediction error. <sup>1</sup>Efficiency of utilization of AA = (AA secretion in milk + scurf + metabolic fecal protein)/AA<sub>adj</sub> × 100. <sup>2</sup>AA<sub>adj</sub> (g/d): net flow of digestible AA supply minus endogenous urinary loss; DEI (MJ/d): digestible energy intake; DIM: days in milk; parity (1 = primiparous; 0 = multiparous); all estimates were significant: P < 0.01 for all parameters except parity where P < 0.05. Degree of significance =  ${}^{S}P \le 0.10$ ;  $*P \le 0.05$ ;  $***P \le 0.001$ .

	Eff <sub>MP</sub> es	Eff <sub>MP</sub> estimation based on equation <sup>1</sup>								
	Equation (12)	Equation (13)	Equation (14)	equation (27) <sup>2</sup>						
Developmental database ( $n = 807$	7)									
Observed mean (g/day)	945	945	945	945						
Predicted mean (g/day)	925	943	947	943						
RMSE (% observed mean)	18.8	13.6	13.1	13.7						
Mean bias (% MSE)	1.3	0.0	0.0	0.0						
Slope bias (% MSE)	0.0	3.2	1.5	4.5						
CCC	0.43	0.74	0.80	0.73						
Validation database ( $n = 129$ )										
Observed mean (g/day)	948	948	948	948						
Predicted mean (g/day)	956	976	980	972						
RMSPE (% observed mean)	18.8	13.6	12.8	13.6						
Mean bias (% MSE)	0.2	4.8	6.9	3.5						
Slope bias (% MSE)	11.3	16.3	0.1	22.8						
CCC	0.45	0.76	0.83	0.76						

**Table 6** Predicted milk true protein yield (MPY, g/d) based on estimates of the efficiency of utilization of metabolizable protein (Eff<sub>MP</sub>) or predicted directly in lactating dairy cows

CCC = concordance correlation coefficient; RMSPE = root mean squared prediction error.

<sup>1</sup>Equations detailed in the text; MPY = Eff<sub>MP</sub>  $\times$  MP<sub>adj</sub> – (scurf true protein + metabolic fecal true protein); MP<sub>adj</sub> (kg/d): metabolicable protein supply minus endogenous urinary loss.

<sup>2</sup>Equation detailed in the text.

These threshold ratios are in  $AA_{adj}/DEI$  (g/MJ): 0.31, 0.67, 1.23, 0.92, 0.32, 0.82, 0.58, 0.15 and 0.73 for His, Ile, Leu, Lys, Met, Phe, Thr, Trp and Val, respectively. In the validation database, only one treatment mean was higher than the threshold ratio for His at 0.32, whereas the ratios were all lower than the threshold values in the validation database.

Efficiency of utilization and prediction of milk true protein yield. The ultimate goal in the estimation of  $Eff_{MP}$  and  $Eff_{AA}$  is the prediction of MPY which was calculated either as:

$$MPY = MP_{adj} \times Eff_{MP} - (TP scurf_{secretion} + TPMFP_{secretion})$$
(25)

or

$$\begin{split} \mathsf{MPY} &= (\mathsf{AA}_{\mathsf{adj}} \times \mathsf{Eff}_{\mathsf{AA}} - (\mathsf{AA}\,\mathsf{scurf}_{\mathsf{secretion}}) \\ &+ \mathsf{AA}\,\mathsf{MFP}_{\mathsf{secretion}})) / [\mathsf{AA}_{\mathsf{calc}-\mathsf{Milk}}] \times 100 \end{split} \eqno(26)$$

First it appeared clearly that the prediction of Eff<sub>MP</sub> based solely on MP<sub>adj</sub> (equation 12) predicts MPY with a low CCC in both databases (Table 6). As observed for the prediction of the efficiencies themselves, adding DEI to MP into the prediction equations greatly improved the predictions of MPY. Adding DEI as an independent variable (equation 13) improved CCC; slope bias contributed to a greater proportion of the total prediction error, but the total error decreased in the validation database (Table 6). Finally, the equation, including the ratio MP<sub>adj</sub>/DEI, provided the best goodness of fit in the two databases, with the highest CCC. For a comparison, a model was developed to predict the sum of protein secretions (MPY + MFP + scurf) directly from  $MP_{adj}$ ,  $MP_{adj} \times MP_{adj}$ , DEI, DIM and parity; MPY was then calculated as predicted protein secretions minus (MFP + scurf). The equation is:

$$\begin{aligned} \text{Protein secretions} &= 160(\pm 79) + 381(\pm 74) \times \text{MP}_{\text{adj}} \\ &- 74(\pm 19) \times \text{MP}_{\text{adj}} \times \text{MP}_{\text{adj}} \\ &+ 2.49(\pm 0.22) \times \text{DEI} - 0.6(\pm 0.1) \\ &\times \text{DIM} - 100(\pm 21) \\ &\times \text{parity}(1 = \text{primiparous}, \\ &0 = \text{multiparous}) \end{aligned}$$

The squared term of DEI was not significant (P = 0.93). Predictions from this model are detailed in Table 6 and are not as good as those obtained when Eff<sub>MP</sub> was predicted with equation (14). Moreover, a strong slope bias was observed in the validation database. Inclusion of the ratio MP<sub>adi</sub>/NE<sub>L</sub> supply in predictive models of Eff<sub>MP</sub> to predict MPY also yielded the best predictions when different formulation models were compared (Lapierre et al., 2018). Therefore, predicted Eff<sub>AA</sub> from equations (15) to (24), for each EAA, were used to predict MPY (Table 7). All AA are yielding fairly similar predictions of MPY except Arg. A comparison between observed and predicted MPY was also made using several combinations to explore the impact of individual Eff<sub>AA</sub>. Five combinations, all excluding predictions from Arg, are presented in Table 7, which are: the lowest of MPY predicted from Eff<sub>AA</sub> (MinAA), the mean of nine MPY predicted from Eff<sub>AA</sub> (MeanAA), the mean of MinAA and MeanAA (MinMeanAA) and the mean of estimations from His, Lys and Met (HLM). All predictions provided good fitness with

	, ,								. ,,	-	5 ,				
		Based on Eff <sub>AA</sub> of individual AA <sup>1</sup>									Combination <sup>2</sup>				
	Arg	His	lle	Leu	Lys	Met	Phe	Thr	Trp	Val	Min	Mean	Max	HLM	Average (min, mean)
Developmental database ( $n = 8$	807)														
Observed mean (g/day)	945	945	945	945	945	945	945	945	945	945	945	945	945	945	945
Predicted mean (g/day)	952	948	946	949	947	945	946	957	946	937	931	947	965	947	939
RMSE (% observed mean)	17.1	13.1	13.0	13.2	13.0	13.0	12.9	13.2	13.0	13.0	13.0	13.0	13.3	12.9	12.9
Mean bias (% MSE)	0.2	0.0	0.0	0.1	0.0	0.0	0.0	1.0	0.0	0.5	1.4	0.0	2.6	0.0	0.3
Slope bias (% MSE)	11.8	2.0	0.6	1.6	0.7	0.5	1.1	2.4	0.9	0.6	0.8	1.0	1.7	0.9	0.9
CCC	0.68	0.80	0.80	0.80	0.80	0.80	0.80	0.80	0.80	0.80	0.80	0.80	0.79	0.80	0.80
Validation database ( $n = 129$ )															
Observed mean (g/day)	948	948	948	948	948	948	948	948	948	948	948	948	948	948	948
Predicted mean (g/day)	976	983	978	984	981	978	980	990	976	968	966	980	995	981	973
RMSPE (% observed mean)	14.4	12.8	12.7	12.8	12.8	12.8	12.7	13.1	12.7	12.5	12.4	12.7	13.4	12.8	12.6
Mean bias (% MSE)	4.2	8.2	6.4	9.0	7.3	6.3	7.2	11.6	5.4	3.0	2.4	7.0	13.8	7.3	4.5
Slope bias (% MSE)	0.5	0.0	0.3	0.1	0.1	0.6	0.1	0.0	0.1	0.3	0.4	0.1	0.0	0.2	0.3
CCC	0.79	0.83	0.83	0.83	0.83	0.83	0.83	0.83	0.83	0.83	0.84	0.83	0.82	0.83	0.83

Table 7 Prediction of milk true protein yield (MPY, g/day) based on estimates of the efficiency of utilization of amino acids (Eff<sub>AA</sub>) in lactating dairy cows

CCC = concordance correlation coefficient; RMSPE = root mean squared prediction error.  $^{1}MPY = (Eff_{aa} \times AA_{adj} - (AA in scurf + AA in metabolic fecal))/concentration of AA in milk \times 100; AA_{adj} (g/d): net flow of digestible AA supply minus endogenous urinary loss.$   $^{2}Min: minimum predicted MPY; mean: average predicted MPY; max: maximum predicted MPY; HLM: average predicted MPY from His, Lys and Met; Arg predicted MPY excluded from all the combinations.$ 

observed MPY, and this is probably one of the limitations of the database and current work actually available. Although we extended the database from Roman-Garcia et al. (2016), trying to increase the number of studies where the supply of only one AA was changed at the time, in most of the studies variations of AA supply were achieved through a change in protein supply which affected simultaneously the supply of all AA. MinAA might be too severe, as the efficiency of a single AA might be maximized if only this one is in short supply (e.g. Lapierre and Ouellet, 2015). Until we have further development, MinMeanAA might be a prudent option as it ponders an average predicted MPY with the prediction from AA most probably in shortest supply. By doing so, it would certainly be fortuitous to verify which AA is yielding the lowest prediction as this might provide a tool to identify the AA with the shortest supply relative to estimated requirements.

#### Conclusion

The development of a factorial approach to balance dairy rations for individual EAA is moving forward with improvement of the quantification of proteins exported out of the animal and their respective AA composition. The major net utilization of EAA supports secretion into MFP and MPY, with a limited contribution to endogenous urinary and scurf secretions. To this net utilization, an inefficiency (100 – efficiency) needs to be added:  $Eff_{MP}$  and  $Eff_{AA}$  are positively related to energy supply and negatively related to MP<sub>adj</sub> or AA<sub>adj</sub>. And finally, the predictions of Eff<sub>MP</sub> and Eff<sub>AA</sub> can be used successfully to predict MPY. Although the concepts derived in the current study can probably be extended to most of the models used to balance dairy rations, it has to be noted that the current figures only apply when using the assumptions as presented. Also, although protein accretion from growth in cows from first parity is acknowledged because parity (primiparous v. multiparous) was included in the model, other changes in protein mass, either through gain or loss of BW or gestation, were not accounted for. Finally, studies where the supply of a single EAA is changed incrementally are currently lacking to really fine-tune our estimations of individual EAA recommendations in dairy rations.

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#### **Declaration of interest**

There is no potential conflicts of interest.

#### **Ethics statement**

None.

#### Software and data repository resources

None.

#### Supplementary material

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#### References

Ainslie SJ, Fox DG, Perry TC, Ketchen DJ and Barry MC 1993. Predicting amino acid adequacy of diets fed to Holstein steers. Journal of Animal Science 71, 1312–1319.

Bibby J and Toutenburg H 1978. Improved estimation and prediction. Journal of Applied Mathematics 58, 45–49.

Biddle GN, Evans JL and Trout JR 1975. Labile nitrogen reserves and plasma nitrogen fractions in growing cattle. Journal of Nutrition 105, 1584–1591.

Blaxter KL and Wood WA 1951. The nutrition of the young Ayrshire calf. I. The endogenous nitrogen and basal energy metabolism of the calf. British Journal of Nutrition 5, 11–25.

Bristow AW, Whitehead DC and Cockburn JE 1992. Nitrogenous constituents in the urine of cattle, sheep and goats. Journal of the Science of Food and Agriculture 59, 387–394.

Broderick GA 2018. Review: Optimizing ruminant conversion of feed protein to human food protein. Animal 12, 1722–1734.

Brun-Lafleur L, Delaby L, Husson F and Faverdin P 2010. Predicting energy  $\times$  protein interaction on milk yield and milk composition in dairy cows. Journal of Dairy Science 93, 4128–4143.

Chen XB and Gomes MJ 1992. Estimation of microbial protein supply to sheep and cattle based on urinary excretion of purine derivatives: an overview of the technical details. International Feed Resources Unit, Rowett Research Institute, Aberdeen, UK.

Clark H, Spires HR, Derrig RG and Bennink MR 1977. Milk production, nitrogen utilization and glucose synthesis in lactating cows infused postruminally with sodium caseinate and glucose. The Journal of Nutrition 107, 631–644.

Daley VL, Armentano LE, Kononoff PJ, Prestegaard JM and Hanigan MD 2018. Fatty acid digestion in dairy cows fed different fat sources: a meta-analytic approach. Journal of Dairy Science 101 (suppl. 2), 304.

Daniel JB, Friggens NC, Chapoutot P, Van Laar H and Sauvant D 2016. Milk yield and milk composition responses to change in predicted net energy and metabolizable protein: a meta-analysis. Animal 10, 1975–1985.

DePeters EJ and Cant JP 1992. Nutritional factors influencing the nitrogen composition of bovine milk: a review. Journal of Dairy Science 75, 2043–2070.

de Souza RA, Tempelman RJ, Allen MS, Weiss WP, Bernard JK and VandeHaar MJ 2018. Predicting nutrient digestibility in high-producing dairy cows. Journal of Dairy Science 101, 1123–1135.

Dijkstra J, Oenema O, van Groenigen JW, Spek JW, van Vuuren AM and Bannink A 2013b. Diet effects on urine composition of cattle and  $N_2O$  emissions. Animal 7, 292–302.

Dijkstra J, Reynolds CK, Kebreab E, Bannink A, Ellis JL, France J and Van Vuuren AM 2013a. Challenges in ruminant nutrition: towards minimal nitrogen losses in cattle. In Energy and protein metabolism and nutrition in sustainable animal production (ed. JW Oltjen, E Kebreab and H Lapierre), pp. 47–58. Wageningen Academic Publishers, Wageningen, The Netherlands.

Doepel L, Pacheco D, Kennelly JJ, Hanigan MD, López IF and Lapierre H 2004. Milk protein synthesis as a function of amino acid supply. Journal of Dairy Science 87, 1279–1297. Ertl P, Klocker H, Hörtenhuber S, Knaus W and Zollitsch W 2015. The net contribution of dairy production to human food supply: the case of Austrian dairy farms. Agricultural Systems 137, 119–125.

Farrell HMJ, Jimenez-Flores R, Bleck GT, Brown EM, Butler JE, Creamer LK, Hicks CL, Hollar CM, Ng-Kwai-Hang KF and Swaisgood HE 2004. Nomenclature of the proteins of cows' milk - Sixth revision. Journal of Dairy Science 87, 1641–1674.

Featherston WR, Frazeur DR, Hill DL, Noller CH and Parmelee CE 1964. Constancy of amino acid composition of cow's milk protein under changing ration. Journal of Dairy Science 47, 1417–1418.

Fox DG, Tedeschi LO, Tylutki TP, Russell JB, Van Amburgh ME, Chase LE, Pell AN and Overton TR 2004. The Cornell Net carbohydrate and protein system model for evaluating herd nutrition and nutrient excretion. Animal Feed Science and Technology 112, 29–78.

Fujihara T and Shem MN 2011. Metabolism of microbial nitrogen in ruminants with special reference to nucleic acids. Animal Science Journal 82, 198–208.

Hanigan MD, Cant JP, Weakley DC and Beckett JL 1998. An evaluation of postabsorptive protein and amino acid metabolism in the lactating dairy cow. Journal of Dairy Science 81, 3385–3401.

Harris CI and Milne G 1981. The urinary excretion of N-methyl histidine by cattle: validation as an index of muscle protein breakdown. British Journal of Nutrition 45, 411.

Holter JA and Reid JT 1959. Relationship between the concentrations of crude protein and apparently digestible protein in forages. Journal of Animal Science 18, 1339–1349.

Huhtanen P and Hristov AN 2009. A meta-analysis of the effects of dietary protein concentration and degradability on milk protein yield and milk N efficiency in dairy cows. Journal of Dairy Science 92, 3222–3232.

Hurvich CM and Tsai CL 1993. A corrected akaike information criterion for vector autoregressive model selection. Journal of Time Series Analysis 14, 271–279.

Hutchinson JC and Morris S 1936. The digestibility of dietary protein in the ruminant: the digestibility of protein following a prolonged fast, with a detailed study of the nitrogen metabolism. Biochemical Journal 30, 1695–1704.

Institut National de Recherche Agronomique (INRA) 2018. Feeding system for ruminants. Wageningen Academic Publishers, Wageningen, The Netherlands.

Jansman AJM, Smink W, Van Leeuwen P and Rademacher M 2002. Evaluation through literature data of the amount and amino acid composition of basal endogenous crude protein at the terminal ileum of pigs. Animal Feed Science and Technology 98, 49–60.

Jonker JS, Kohn RA and Erdman RA 1998. Using milk urea nitrogen to predict nitrogen excretion and utilization efficiency in lactating dairy cows. Journal of Dairy Science 81, 2681–2692.

Karman AH and van Boekel AJS 1986. Evaluation of the Kjeldahl factor for conversion of the nitrogen content of milk and milk products to protein content. Netherlands Milk and Dairy Journal 40, 315–336.

Kauffman AJ and St-Pierre NR 2001. The relationship of milk urea nitrogen to urine nitrogen excretion in Holstein and Jersey cows. Journal of Dairy Science 84, 2284–2294.

Kool DM, Hoffland E, Hummelink EWJ and van Groenigen JW 2006. Increased hippuric acid content of urine can reduce soil N2O fluxes. Soil Biology and Biochemistry 38, 1021–1027.

Lapierre H, Binggeli S, Sok M, Pellerin D and Ouellet DR 2019. Estimation of correction factors to determine the true amino acid concentration of protein after a 24-hour hydrolysis. Journal of Dairy Science 102, 1205–1212.

Lapierre H, Larsen M, Sauvant D, Van Amburgh ME and Van Duinkerken G 2018. Review: Converting nutritional knowledge into feeding practices: a case study comparing different protein feeding systems for dairy cows. Animal 12, S457–S466.

Lapierre H, Lobley GE, Doepel L, Raggio G, Rulquin H and Lemosquet S 2012. Triennial lactation symposium: mammary metabolism of amino acids in dairy cows. Journal of Animal Science 90, 1708–1721.

Lapierre H, Lobley GE, Ouellet DR, Doepel L and Pacheco D 2007. Amino acid requirements for lactating dairy cows: reconciling predictive models and biology. In Proceedings of the 69th Cornell Nutrition Conference for Feed Manufacturers, 23–25 October 2007, Syracuse, NY, USA, pp. 39–59.

Lapierre H and Ouellet DR 2015. How the efficiency of utilization of histidine varies with supply in dairy cows. Journal of Dairy Science 98 (suppl. 2), 609.

Lapierre H, Ouellet DR, Martineau R and Spek JW 2016. Key roles of amino acids in cow performance and metabolism – considerations for defining amino acid requirement. In Proceedings of the 78th Cornell Nutrition Conference for Feed Manufacturers, 18–20 October 2016, Syracuse, NY, USA, pp. 205–219.

Lien KA, Sauer WC, Mosenthin R, Souffrant WB and Dugan MER 1997. Evaluation of the <sup>15</sup>N-isotope dilution technique for determining the recovery of endogenous protein in ileal digestion of pigs: effect of dilution in the precursor pool for endogenous nitrogen secretion. Journal of Animal Science 75, 148–158.

Lin LI-K 1989. A concordance correlation coefficient to evaluate reproducibility. Biometrics 45, 255–268.

Lin LI-K 1992. Assay validation using the concordance correlation coefficient. Biometrics 48, 599–604.

Lobley GE and Lapierre H 2003. Post-absorptive metabolism of amino acids. In Progress in research on energy and protein metabolism (ed. WB Souffrant and CC Metges), pp. 737–756, Wageningen Academic Publishers, Wageningen, The Netherlands.

Marini JC, Fox DG and Murphy MR 2008. Nitrogen transactions along the gastrointestinal tract of cattle: a meta-analytical approach. Journal of Animal Science 86, 660–679.

Marini JC and Van Amburgh ME 2005. Partition of nitrogen excretion in urine and the feces of holstein replacement heifers. Journal of Dairy Science 88, 1778–1784.

Martineau R, Ouellet DR, Kebreab E, White RR and Lapierre H 2017. Relationships between postruminal casein infusion and milk production, and concentrations of plasma amino acids and blood urea in dairy cows: a multilevel mixed-effects meta-analysis. Journal of Dairy Science 100, 8053–8071.

Metcalf JA, Mansbridge RJ, Blake JS, Oldham JD and Newbold JR 2008. The efficiency of conversion of metabolisable protein into milk true protein over a range of metabolisable protein intakes. Animal 2, 1193–1202.

Myers AJ, Lapierre H, White RR, Tran H, Kononoff PJ, Martineau R, Weiss WP and Hanigan MD 2018. Predictions of rumen outflow of amino acids in dairy cattle. Journal of Dairy Science 101 (suppl. 2), 410.

National Research Council 2001. Nutrient requirements of dairy cattle. 7th revised edition. Natl. Acad. Sci., Washington, DC, USA.

Nehring K, Zelck U and Schiemann R 1965. Über die zusammensetzung des harns an organischen inhaltsstoffen bei rindern, schafen und schweinen. Archiv für Tierernaehrung 15, 45–52.

Nichols K, Bannink A, Pacheco S, van Valenberg HJ, Dijkstra J and van Laar H 2018. Feed and nitrogen efficiency are affected differently but milk lactose production is stimulated equally when isoenergetic protein and fat is supplemented in lactating dairy cow diets. Journal of Dairy Science 101, 7857–7870.

Nichols K, Kim JJM, Carson M, Metcalf JA, Cant JP and Doelman J 2016. Glucose supplementation stimulates peripheral branched-chain amino acid catabolism in lactating dairy cows during essential amino acid infusions. Journal of Dairy Science 99, 1145–1160.

NorFor 2011. The Nordic feed evaluation system. Wageningen Academic Publishers, Wageningen, The Netherlands.

Ørskov ER, McLeod NA and Kyle DJ 1986. Flow of nitrogen from the rumen and abomasum in cattle and sheep given protein-free nutrients by intragastric infusion. British Journal of Nutrition 56, 241–248.

Ouellet DR, Berthiaume R, Holtrop G, Lobley GE, Martineau R and Lapierre H 2010. Effect of method of conservation of timothy on endogenous nitrogen flows in lactating dairy cows. Journal of Dairy Science 93, 4252–4261.

Ouellet DR, Demers M, Zuur G, Lobley GE, Seoane JR, Nolan JV and Lapierre H 2002. Effect of dietary fiber on endogenous nitrogen flows in lactating dairy cows. Journal of Dairy Science 85, 3013–3025.

Ouellet DR, Valkeners D, Holtrop G, Lobley GE and Lapierre H 2007. Contribution of endogenous secretions and urea recycling to nitrogen metabolism. In Proceedings of the 69th Cornell Nutrition Conference for Feed Manufacturers, 23–25 October 2007, Syracuse, NY, USA, pp. 1–24.

Raggio G, Lobley GE, Berthiaume R, Pellerin D, Allard G, Dubreuil P and Lapierre H 2007. Effect of protein supply on hepatic synthesis of plasma and constitutive proteins in lactating dairy cows. Journal of Dairy Science 90, 352–359.

Rohr K and Lebzien P 1991. Present knowledge of amino acid requirements for maintenance and production. Paper pesented at the 6th International

Symposium on Protein Metabolism and Nutrition, 9–14 June 1991, Herning, Denmark, pp. 127–137.

Roman-Garcia Y, White RR and Firkins JL 2016. Meta-analysis of postruminal microbial nitrogen flows in dairy cattle. I. Derivation of equations. Journal of Dairy Science 99, 7918–7931.

Sandek A, Krawielitzki K, Kowalczyk J, Kreienbring F, Schoenhusen U, Gabel M, Zebrowska T, Hagemeister H and Voigt J 2001. Studies on N-metabolism in different gastrointestinal sections of sheep using the digesta exchange technique. 2. Passage of endogenous nitrogen. Journal of Animal & Feed Sciences 10, 605–618.

Sauvant D, Catalapiedra-Hijar G, Delaby L, Daniel J-B, Faverdin P and Nozière P 2015. Actualisation des besoins protéiques des ruminants et détermination des réponses des femelles laitières aux apports de protéines digestibles dans l'intestin (PDI). INRA Productions Animales 28, 347–368.

Sok M, Ouellet DR, Firkins JL, Pellerin D and Lapierre H 2017. Amino acid composition of rumen bacteria and protozoa in cattle. Journal of Dairy Science 100, 5241–5249.

Spek JW, Dijkstra J, van Duinkerken G, Hendriks WH and Bannink A 2013. Prediction of urinary nitrogen and urinary urea nitrogen excretion by lactating dairy cattle in northwestern Europe and North America: a meta-analysis. Journal of Dairy Science 96, 4310–4322.

Swaisgood HE 1995. Protein and amino acid composition of bovine milk. In Handbook of Milk Composition (ed. RG Jensen), pp. 464–472. Academic Press, London, UK.

Swanson EW 1977. Factors for computing requirements of protein for maintenance of cattle. Journal of Dairy Science 60, 1583–1593.

Swanson EW 1982. Estimation of metabolic protein requirements to cover unavoidable losses of endogenous nitrogen in maintenance of cattle. Paper presented at the Symposium Protein Requirements for Cattle (ed. FN Owens), 19–21 November 1980, Oklahoma State University, Stillwater, OK, USA, pp. 183–197.

Syvaöja E-L and Virtanen AI 1965. Fractionation of milk proteins, especially of milk produced on protein-free feed with urea and ammonium salt as the sole source of nitrogen. Enzymologia 29, 205–219.

Tamminga S, Schulze H, Van Bruchem J and Huisman J 1995. The nutritional significance of endogenous N-losses along the gastro-intestinal tract of farm animals. Archiv fur Tierernahrung 48, 9–22.

Tas BM and Susenbeth A 2007. Urinary purine derivates excretion as an indicator of in vivo microbial N flow in cattle: a review. Livestock Science 111, 181–192.

Theil H 1966. Measuring the accuracy of point predictions. In Applied economic forecasting, pp. 14–36. North Holland Publishing Company, Amsterdam, The Netherlands.

Van Amburgh ME, Collao-Saenz EA, Higgs RJ, Ross DA, Recktenwald EB, Raffrenato E, Chase LE, Overton TR, Mills JK and Foskolos A 2015. The Cornell Net carbohydrate and protein system: updates to the model and evaluation of version 6.5. Journal of Dairy Science 98, 6361–6380.

Van Duinkerken G, Blok MC, Bannink A, Cone JW, Dijkstra J, Van Vuuren AM and Tamminga S 2011. Update of the Dutch protein evaluation system for ruminants: the DVE/OEB2010 system. Journal of Agricultural Science 149, 351–367.

Vanhatalo A, Varvikko T and Huhtanen P 2003a. Effects of casein and glucose on responses of cows fed diets based on restrictively fermented grass silage. Journal of Dairy Science 86, 3260–3270.

Vanhatalo A, Varvikko T and Huhtanen P 2003b. Effects of various glucogenic sources on production and metabolic responses of dairy cows fed grass silagebased diets. Journal of Dairy Science 86, 3249–3259.

Viechtbauer W 2017a. [R-meta] Multilevel meta-analysis using metafor. Retrieved on 01 August 2018 from https://stat.ethz.ch/pipermail/r-sig-metaanalysis/2017-August/000174.html.

Viechtbauer W 2017b. [R-meta] Multilevel meta-analysis using metafor. Compute (cluster) robust test and confidence intervals for 'rma' objects. Retrieved on 30 January 2018 from https://www.rdocumentation.org/ packages/metafor/versions/1.9-9/topics/robust.

Viechtbauer W 2018. [R-meta] Weighing factors in rma.mv corresponding to those used in Imer. Retrieved on 01 August 2018 from https://stat.ethz.ch/pipermail/r-sig-meta-analysis/2018-March/000731.html.

Waldo DR and Glenn BP 1984. Comparison of new protein systems for lactating dairy cows. Journal of Dairy Science 67, 1115–1133.

White RR, Kononoff PJ and Firkins JL 2017. Technical note: methodological and feed factors affecting prediction of ruminal degradability and intestinal digestibility of essential amino acids. Journal of Dairy Science 100, 1946–1950.

Wickersham TA, Titgemeyer EC, Cochran RC, Wickersham EE and Gnad DP 2008a. Effect of rumen-degradable intake protein supplementation on urea kinetics and microbial use of recycled urea in steers consuming low-quality forage. Journal of Animal Science 86, 3079–3088.

Wickersham TA, Titgemeyer EC, Cochran RC, Wickersham EE and Mooref ES 2008b. Effect of frequency and amount of rumen-degradable intake protein supplementation on urea kinetics and microbial use of recycled urea in steers consuming low-quality forage. Journal of Animal Science 86, 3089–3099.

Wickham H 2016. ggplot2: elegant graphics for data analysis, 2nd edition. Springer International Publishing AG, Switzerland.

Wilkinson JM 2011. Re-defining efficiency of feed use by livestock. Animal 5, 1014–1022.

Williams AP 1978. The amino acids, collagen and mineral composition of preruminant calves. The Journal of Agricultural Science 90, 617–624.

Zhu CL, Lapierre H, Rademacher M and de Lange CFM 2003. Pectin infusion into the caecum reduces the utilization of threonine intake for body protein deposition and urea flux in growing pigs. In Proceedings of the 9th International Symposium on Digestive Physiology in Pigs, Volume 2 (ed. RO Ball), 14–17 May 2003, Banff, AB, Canada, pp. 346–348.