

# Derivation of the metabolic status in dairy cows - prediction of ruminal daily pH mean

A. Mensching<sup>1</sup>, M. Zschiesche<sup>2</sup>, J. Hummel<sup>2</sup>, A. R. Sharifi<sup>1</sup>

- <sup>1</sup> University of Goettingen, Animal Breeding and Genetics Group, Center for Integrated Breeding Research, Goettingen, Germany
- <sup>2</sup> University of Goettingen, Ruminant Nutrition Group, Goettingen, Germany





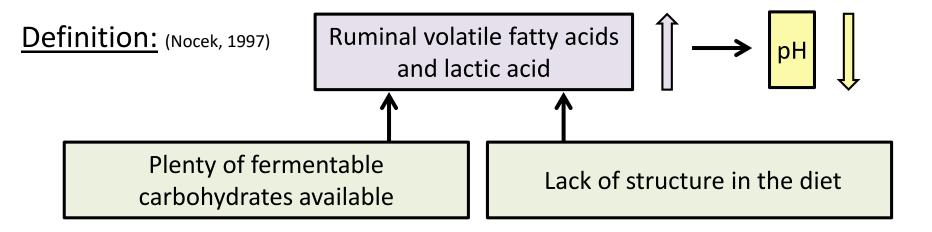




#### Introduction



Metabolism disorder in dairy cows: Subacute ruminal acidosis (SARA) "It may [...] be appropriate to suggest that SARA is the most important nutritional disease of dairy cattle." (Enemark, 2008)



Adequate ruminant nutrition (!)

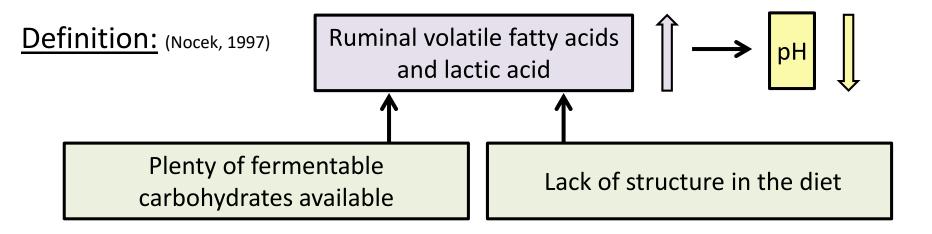
#### Present indicators for SARA

- Daily mean pH <6.2 and time pH <5.8 >5.2 h per day (Zebeli et al., 2008)
- Fat-to-Protein ratio <1.0 (Enemark et al., 2002; KTBL, 2016)</p>
- pH <5.5 using rumenocentesis (Nordlund et al., 1995; Garett et al., 1999)</li>

#### Introduction



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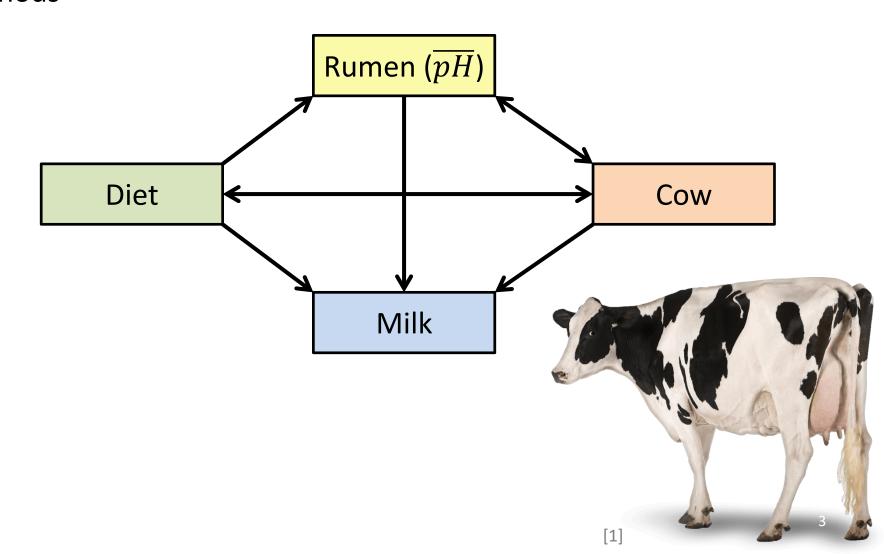
The phenotype of SARA is <u>not</u> clearly defined yet

→ Main reason: Subacute = No clinical signs

# Objective



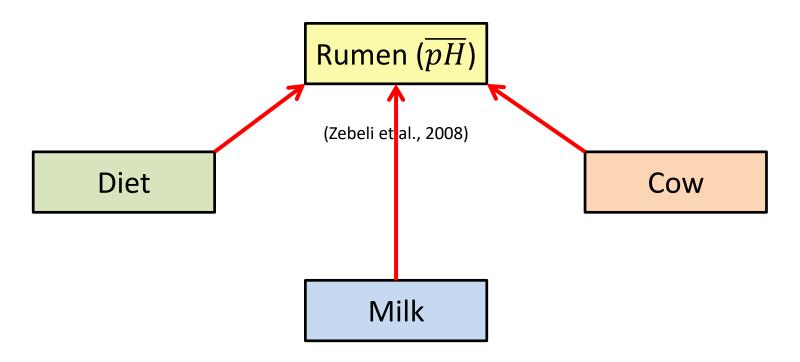
Modeling of the daily mean ruminal pH  $(\overline{pH})$  using meta-analytical methods



# Objective



Modeling of the daily mean ruminal pH  $(\overline{pH})$  using meta-analytical methods



- Predictors: Milk- and diet-specific parameters
  - Focus on parameters that are available in agricultural practice
- Analysis and validation of present indicators



#### Workflow

Step I

Literature research and building a data set

→ Problem: ≈21 % missing values

Step II

Statistical imputation

Method comparison via cross-validation: Liner Models (LM) vs. linear mixed Models (LMM)

Step III

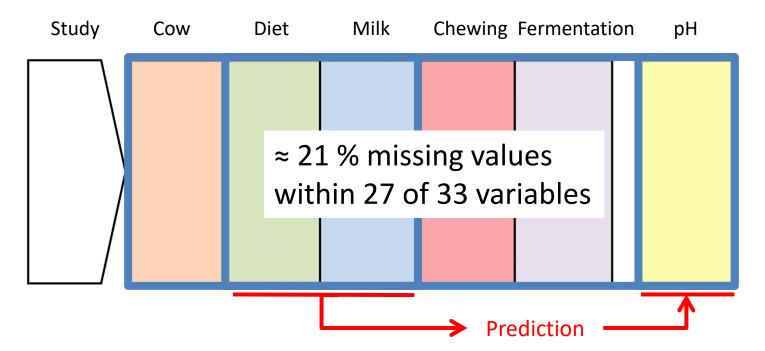
Variable selection for  $\overline{pH}$ 

Method comparison via cross-validation: LM vs. LMM



#### Step I: Data

- Selection criteria for studies
  - Continuous pH measurement in the ventral sac of the rumen
  - Measurement >24 h in cannulated lactating Holstein cows
- Hierarchical data structure with group level "study"
  - n = 39 studies, k = 141 treatment means ( $\Sigma$  = 279 cows)





#### Meta-regression for Step II and III:

LM: linear and quadratic effects

$$y_i = \beta_0 + \beta_1 x_{i,1} + \beta_2 x_{i,2} + \beta_2 x_{i,2}^2 + \dots + e_i$$

LMM: linear and quadratic effects, study as random effect

$$y_{ij} = \beta_0 + \beta_1 x_{ij,1} + \beta_2 x_{ij,2} + \beta_2 x_{ij,2}^2 + \dots + study_j + e_{ij}$$

- Weighted least squares estimation
  - Weighting by the reciprocal standard error
     (Inverse variance method) (Borenstein et al., 2009; Schwarzer et al., 2015)
- Model comparison via leave-one-study-out cross-validation on the correlation between observed and predicted values  $r(y, \hat{y})$  (Prediction with LMM only on the fixed effect part)

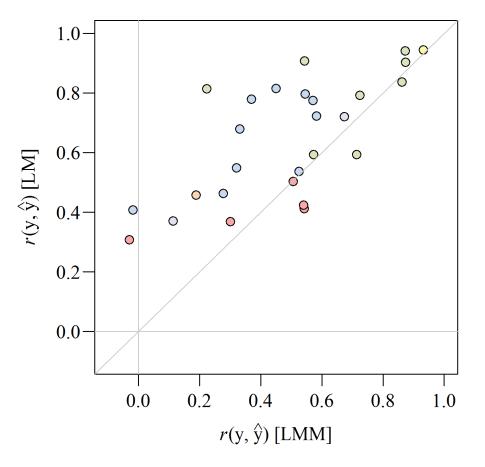


#### Step II: Statistical imputation

- Multiple Imputation by Chained Equations (MICE) (Azur et al., 2011)
  - Each variable with missing values is modeled by other variables of the same dataset
    - Developed algorithm for variable selection with a stepwise forward selection using the Bayesian information criterion (BIC)
    - Considering linear effects
  - Series of regression models in an iterative process
- Within statistical imputation all variable groups were used



#### Step II: Statistical imputation, Method comparison via $r(y, \hat{y})$



Parameter group

Milk
Diet
Cow
pH
Fermentation
Chewing

- Imputation performance of LM better than of LMM
  - → Further analysis on the data base which was imputed by LM



#### Step III: Variable selection for $\overline{pH}$

- Developed algorithm for variable selection with a stepwise forward selection using the BIC
- Considering linear and quadratic effects
- Separate variable selection for  $\overline{pH}$  on parameters of

Milk

Diet

Milk & Diet

## Results and Discussion



#### Variable selection for $\overline{pH}$

	Liı	Linear models			Linear mixed models		
Predictors	LM1	LM2	LM3	LMM1	LMM2	LMM3	
Protein							
FPR							
Lactose							
Lactose <sup>2</sup>							
NEL							
EE							
ADF							
peNDF>8							
(peNDF>8)							
NFC							

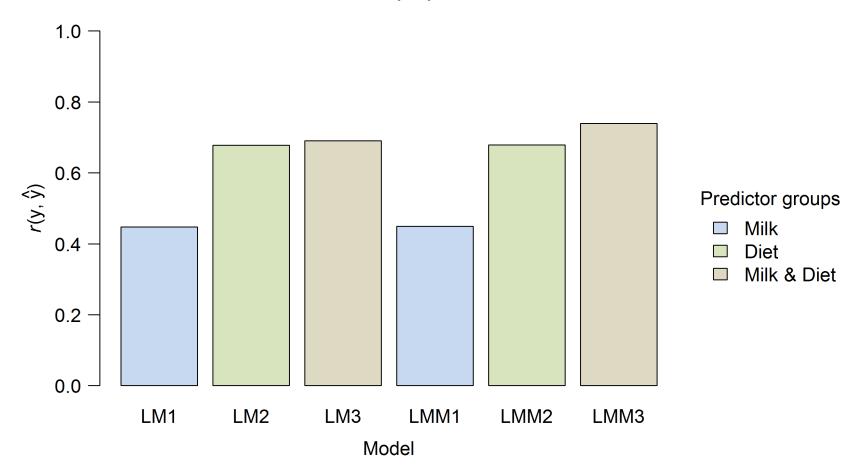
**NEL** = Net energy lactation, **EE** = Ether extract, **ADF** = Acid detergent fiber, **peNDF>8** = physically effective neutral detergent fiber >8mm, **NFC** = Non fiber carbohydrates

## Results and Discussion



#### Variable selection for $\overline{pH}$

• Results of cross-validation  $r(y, \hat{y})$ 



## Conclusion



- Statistical imputation is a useful tool for meta-analytical data
- The explorative variable selection detected known associations
- Moderate predictability of ruminal  $\overline{pH}$
- Additional consideration of milk parameters improves the modeling of  $\overline{pH}$
- Clear signs for protein and lactose as indicators

#### <u>Outlook</u>

- Investigation of the influence of location for pH measurement
  - ventral sac of the rumen vs. reticulum
- Validation on data which were collected on 100 cows equipped with boluses on practical farms



## Thank you for your attention!

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#### Gefördert durch:



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# **Appendix**



#### Variable selection for $\overline{pH}$

	Linear mixed models					
Predictors	LM1	LM2	LM3	LMM1	LMM2	LMM3
Intercept	10.172	2.353	43.936	9.993	3.873	10.123
Protein	-0.389		-0.431	-0.457		-0.460
FPR				0.453		
Lactose	-0.615		-15.253	-0.639		-0.529
Lactose <sup>2</sup>			1.565			
NEL		0.290			0.249	
EE		0.086			0.078	
ADF		0.027	0.027			
peNDF>8		0.077			0.063	0.049
(peNDF>8) <sup>2</sup>		-0.002			-0.001	-0.001
NFC					-0.015	-0.015

Predictor groups

- Milk
- Diet
- Milk & diet