

# Genomic predictions for dry matter intake using the international reference population of **gDMI**

Yvette de Haas and **gDMI** consortium



# Breeding for feed efficiency is gaining importance!

**Farmers  
Guardian**

HOME YOUR SPACE SHOWS & SALES DIGITAL EDITIONS CL  
Business Arable Machinery Livestock Equestrian Young farmers Rural Li

[Home](#) | [Livestock](#) |

## Research will be key in improving dairy sector efficiency

13 May 2014 | By Olivia Midgley

MORE research must be targeted at feed efficiency in ruminants in order to 'catch up' with advancements in the pig and poultry sector.

**FARMERS WEEKLY**

FWI Classifieds Suppliers Jobs Property Academy Awards

HOME BUSINESS MACHINERY ARABLE LIVESTOCK POULTRY FARM LIFE WEATHER

Dairy Beef Sheep Pigs Farmer Focus Taking Stock Videos

Livestock

Invest dairy profits wisely to boost business efficiency

Tuesday 15 April 2014 06:00

**Alberta Farmer**  
EXPRESS

NEWS OPINION LIVESTOCK CROPS MARKETS WEATHER

## New study aims to improve protein efficiency in dairy cows

Researcher to study how to lower protein content without compromising milk production



Posted May. 6, 2014 by Alma Release in [Livestock](#), [News](#)



# What is missing to breed on feed efficiency?

- **Feed intake** records on daughters of sires

- Expensive measurements
- Labour intensive
- Not practical for daily practice -> only research herds (small impact)



# Pilot international collaboration (AUS-NL-UK)

## Conclusion of pilot:

- Accuracy of genomic breeding values for dry matter intake can be increased by:
  - combining datasets across countries, and
  - using a multitrait approach
- Average accuracy  $r(GEBV, TBV)$ : 0.35



J. Dairy Sci. 95:6103–6112  
<http://dx.doi.org/10.3168/jds.2011-5280>  
© American Dairy Science Association®, 2012.

**Improved accuracy of genomic prediction for dry matter intake of dairy cattle from combined European and Australian data sets**

Y. de Haas,<sup>\*1</sup> M. P. L. Calus,\* R. F. Veerkamp,\* E. Wall,<sup>†</sup> M. P. Coffey,<sup>†</sup> H. D. Daetwyler,<sup>‡</sup> B. J. Hayes,<sup>‡\$#</sup> and J. E. Pryce<sup>‡\$</sup>

<sup>\*</sup>Animal Breeding and Genomics Centre, Wageningen UR Livestock Research, PO Box 65, NL-8200 AB Lelystad, the Netherlands

<sup>†</sup>Sustainable Livestock Systems Group, Scottish Agricultural College, Easter Bush, Midlothian, EH25 9RG, United Kingdom

<sup>‡</sup>Biosciences Research Division, Department of Primary Industries Victoria, 1 Park Drive, Bundoora, Victoria 3083, Australia

<sup>\$</sup>Dairy Futures Cooperative Research Centre, Victoria 3083, Australia

<sup>#</sup>La Trobe University, Bundoora, Victoria 3086, Australia

# global Dry Matter Initiative: gDMI

- 15 parties in consortium (science + industry)
- 9 countries, 10 groups
- ~10,000 phenotyped animals
- ~6,000 genotyped animals
- ~12,000 parities
- 591,621 SNPs HD-imputed



# global Dry Matter Initiative: gDMI

- Key research questions:
  - How to combine, homogenise and standardise phenotypes? (Berry et al., 2014)
  - Genomic similarity between population? (Pryce et al., 2014)
  - Can we predict DGV for DMI for different partners? (De Haas et al., 2014 in prep.)
- Aim of today is to present:
  - Estimated accuracies of genomic breeding values (GEBV's) across countries for dry matter intake with international dataset of gDMI

# Our data (Berry et al., 2014)

Country	N	Mean
<b>Cows</b>		
All	10,008	19.7
Canada	411	22.2
Denmark	668	22.1
Germany	1,141	20.2
Iowa	398	23.5
Ireland	1,677	16.7
Netherlands	2,956	21.4
UK	2,840	17.4
Wisconsin	447	24.9
Australia	103	15.6
<b>Heifers</b>		
Australia	843	8.3
New Zealand	941	7.6

# Phenotype for genetic parameters

“Weighted average” phenotype per animal for DMI on day 70 in parity 2

- Predicted from estimated quadratic DMI curve for each animal by 5 parities => 120 (co)variances
  - Fixed effects: parity-DIM, HYS, experimental treatment
  - Random effects: cow-parity-DIM difference in residual variances across HYS and herd curve
- Predicted DMI with low reliabilities were discarded => ~8,500 animals remained
- Parameters estimated with H-matrix

# Heritability for each country

DMI on 70d in par2	
CAN	<b>0.21</b>
DNK	<b>0.46</b>
AU_h	<b>0.32</b>
NZ_h	<b>0.24</b>
GER	<b>0.17</b>
US_I	<b>0.53</b>
IRL	<b>0.26</b>
NLD	<b>0.38</b>
UK	<b>0.26</b>
US_W	<b>0.12</b>

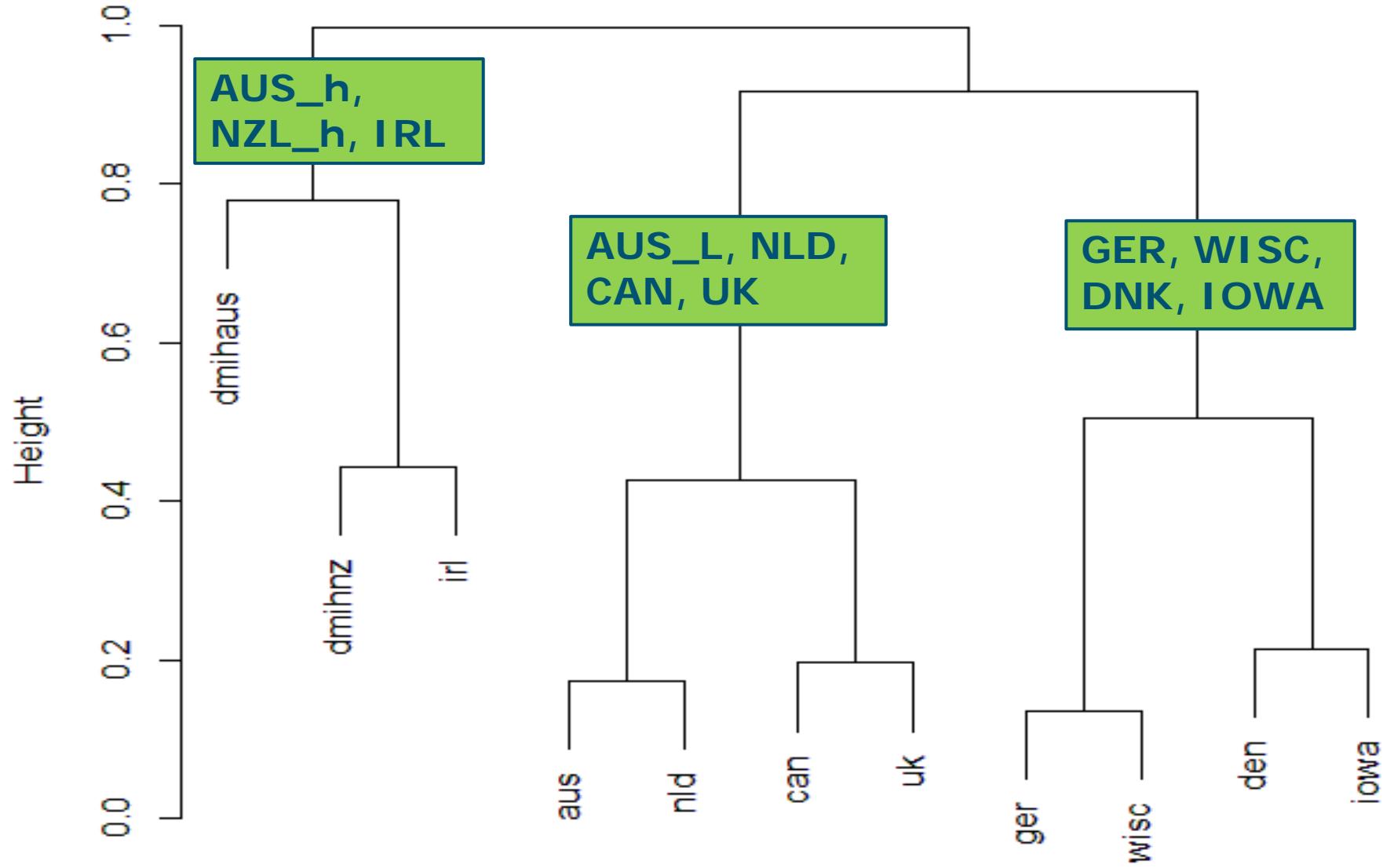
# Genetic correlations between countries

	AUS	CAN	DNK	AU_h	NZ_h	GER	US_I	IRL	NLD	UK
CAN	<b>0.66</b>									
DNK	0.56	0.32								
AU_h	0.27	0.30	0.05							
NZ_h	-0.26	0.09	-0.19	0.22						
GER	0.32	0.13	0.85	0.17	0.17					
US_I	0.36	0.14	0.79	-0.14	-0.06	0.68				
IRL	0.00	0.04	0.16	0.39	0.56	0.45	-0.15			
NLD	0.83	0.77	0.82	0.20	-0.14	0.62	0.63	0.02		
UK	0.57	0.80	0.37	0.61	0.40	0.46	0.08	0.50	0.68	
US_W	0.53	0.52	0.75	0.29	0.15	0.86	0.50	0.35	0.80	0.76

# Genetic correlations between countries

	AUS	CAN	DNK	AU_h	NZ_h	GER	US_I	IRL	NLD	UK
CAN	<b>0.66</b>									
DNK	0.56	0.32								
AU_h	0.27	0.30	0.05							
NZ_h	-0.26	0.09	-0.19	0.22						
GER	0.32	0.13	<b>0.85</b>	0.17	0.17					
US_I	0.36	0.14	0.79	-0.14	-0.06	0.68				
IRL	0.00	0.04	0.16	0.39	0.56	0.45	-0.15			
NLD	<b>0.83</b>	0.77	<b>0.82</b>	0.20	-0.14	0.62	0.63	0.02		
UK	0.57	<b>0.80</b>	0.37	0.61	0.40	0.46	0.08	0.50	0.68	
US_W	0.53	0.52	0.75	0.29	0.15	<b>0.86</b>	0.50	0.35	<b>0.80</b>	0.76

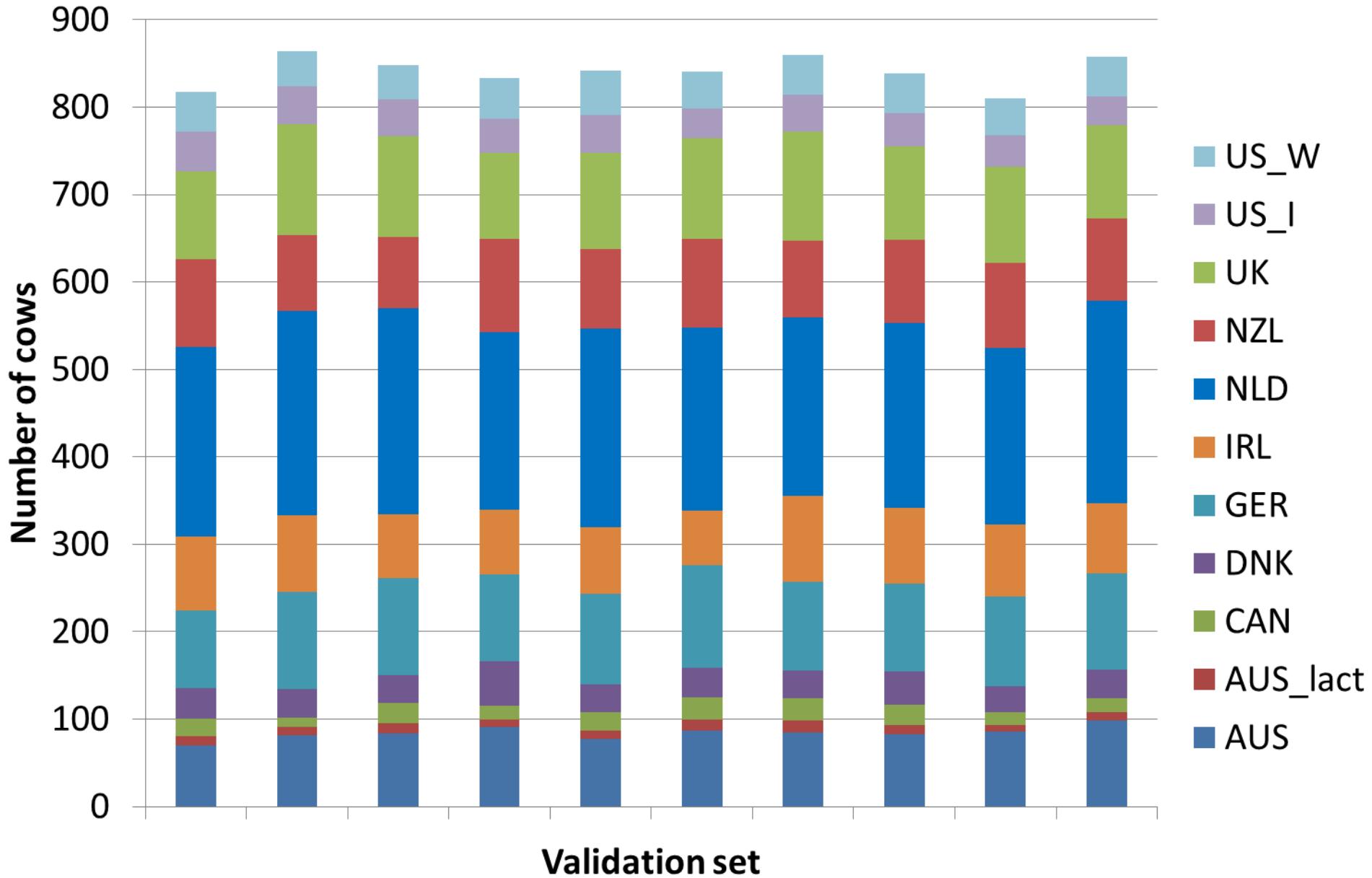
# Dendrogram



# Validation strategy

- Validation sets based on progeny groups of sires in the different countries
  - *Can we increase the accuracy of bull GEBVs by using multi-country reference populations?*
  - Risk that by having all the progeny of particular sires only within one validation population might mean lower accuracies than if the progeny were spread across the validation populations.

# Validation sets



# Accuracy of genomic selection $r(\text{GEBV}, \text{TBV})$

	CAN	DNK	AU_h	NZ_h	GER	IOWA	IRL	NLD	UK	WISC
$h^2$	0.21	0.46	0.32	0.24	0.17	0.53	0.26	0.38	0.26	0.12
Acc	0.40	0.37	0.39	0.44	0.45	0.46	0.48	0.54	0.49	0.38
SE	0.19	0.08	0.04	0.05	0.07	0.06	0.06	0.04	0.08	0.15

Average	
Accuracy	0.44
Standard error	0.08

AUS\_lact: 0.48 (0.08)

# Accuracy of genomic selection $r(\text{GEBV}, \text{TBV})$

Correlations between countries: 0.0

	CAN	DNK	AU_h	NZ_h	GER	IOWA	IRL	NLD	UK	WISC
$h^2$	0.21	0.46	0.32	0.24	0.17	0.53	0.26	0.38	0.26	0.12
Acc	0.17	0.33	0.39	0.42	0.39	0.43	0.46	0.51	0.43	0.20
SE	0.19	0.10	0.05	0.05	0.07	0.05	0.06	0.04	0.08	0.13

# Accuracy of genomic selection $r(\text{GEBV}, \text{TBV})$

Correlations between countries: 0.95

	CAN	DNK	AU_h	NZ_h	GER	IOWA	IRL	NLD	UK	WISC
$h^2$	0.21	0.46	0.32	0.24	0.17	0.53	0.26	0.38	0.26	0.12
Acc	0.33	0.36	0.30	0.38	0.43	0.43	0.44	0.51	0.48	0.35
SE	0.20	0.09	0.04	0.05	0.07	0.07	0.07	0.04	0.08	0.14

# Accuracy of genomic selection $r(\text{GEBV}, \text{TBV})$

## Overview

	CAN	DNK	AU_h	NZ_h	GER	IOWA	IRL	NLD	UK	WISC
Rg	0.40	0.37	0.39	0.44	0.45	0.46	0.48	0.54	0.49	0.38
Rg0	0.17	0.33	0.39	0.42	0.39	0.43	0.46	0.51	0.43	0.20
Rg1	0.33	0.36	0.30	0.38	0.43	0.43	0.44	0.51	0.48	0.35

Avg Acc	
Rg	0.44
Rg0	0.37
Rg1	0.40

# Conclusions

- Average accuracy of genomic breeding values with **gDMI** dataset: 0.44
  - Average accuracy in pilot (AUS-UK-NL): 0.35

Further research questions:

- Sensitivity for chosen phenotype (70DIM; par 2)
- Sensitivity for validation strategies
- Sensitivity for half sib groups across countries
- Sensitivity for genetic correlations (bending procedure)

# Acknowledgements

## gDMI consortium:

Roel Veerkamp  
Mario Calus  
Donagh Berry  
Mike Coffey  
Nina Krattenmacher  
Georg Thaller  
Peter Lovendahl  
Jennie Pryce  
Ben Hayes  
Kevin MacDonald  
Jeremy Bryant  
Filippo Miglior  
Kent Weigel  
Diane Spurlock

And many others



*Thank you for your attention*



Open meeting of gDMI on  
Thursday, 13.30h, room 30610

