

Using lactation curves as a tool for breeding, nutrition and health management decisions in pasture-based dairy systems.

S.A. Adediran¹, A.E.O. Malau-Aduli¹, J. R. Roche^{1,2}, and D.J. Donaghy^{1,2}

¹ *School of Agricultural Science, University of Tasmania, Private Bag 54 Hobart, TAS 7001 Australia.*

² *Tasmanian Institute of Agricultural Research, P.O. Box 3523, Burnie, TAS 7320.*

samuel.adediran@utas.edu.au

Introduction

Milk yield and reproductive efficiency are crucial to profitable dairying.

Although, genetic improvement in the last few decades has led to substantial increases in milk yield/cow, fertility and reproductive health have declined (Dematawewa and Berger, 1998).

In a pasture-based system, a 365 day calving interval is crucial for optimum profit. Hence the need to increase milk yield by improving persistence of lactation rather than peak lactation which puts increased stress on the cows at the time when they should be rebreeding. Peak milk yield, persistency and lactation length are the key components of the lactation profile. Dairy cows with high peak yields are more prone to metabolic and physiological disorders (Terkeli *et al* 1999). Although estimated breeding values (EBV) in dairy cows in Australia incorporates indices of economic

value, such as survival and milking speed, the impact of the current

breeding approach and management on the shape of the lactation profile is not clear.

Mathematical functions such as those previously used to describe a series of milk test day records (Wood, 1967, Wilmink, 1987), have the advantage of minimizing random variation while simultaneously summarising the lactation profile into biologically interpretable parameters.

The shape of the lactation curve provides valuable information about the biological and economic efficiency of the animal or herd and is useful for genetic evaluation, health monitoring, feed management decisions and planning purposes (Sherchand *et al.*, 1995). Cow's genetic merit, breed, parity, calving season, nutrition, and pregnancy affect the shape of her lactation curve (Tozer and Huffaker, 1999, Roche *et al.*, 2006). A robust model should adequately mimic the biological process of lactation and adjust for factors affecting it.

The functions available to model lactation profiles are many (Beever *et al.* 1991) but their suitability varies across systems. For instance, Olori *et al.* (1999) reported that the polynomial model gave the best fit in a farm-based study, while Garcia and Holmes (2001) found no difference in average lactation predicted by both diphasic and linear-based split-plot models.

The objectives of this study was to identify suitable lactation models and determine the factors affecting the lactation profile, in pasture-based dairy systems.

Materials and Methods

Test-day milk yield data from 428 herds in Tasmania, consisting of 65,000 milk yield records from 2002-2005, were edited to exclude incomplete lactations. The five empirical and two mechanistic functions used to evaluate average daily milk (kg/d) were:

Incomplete Gamma (IG) $Y(t) = a t b e^{-ct}$
(Wood, 1967)

Modified Gamma (MG) $Y(t) = a t e^{-ct}$
(Jenkins and Ferrel, 1984)

Mixed Log (ML) $Y(t) = a + bt^{1/2} + c \log t$ (Guo and Swalve, 1995)

Exponential (EXP) $Y(t) = a + b e^{-kt} + ct$
(Wilmink, 1987)

Polynomial (PR) $Y(t) = a + bt + ct^2 + d \log t + \epsilon (\log t)^2$ (Ali & Schaeffer, 1987).

Bicompartmental (BC) $Y(t) = a e^{-bt} + de^{-ct}$
(Ferguson and Boston, 1993)

Dijkstra (DJ) $Y(t) = a \exp[b(1-e^{-ct})/c-dt]$
(Dijkstra *et al.*, 1997)

where Y is milk yield (L/kg/cow/day), at time t (weeks), and a, b, c d and ϵ are parameters that define the scale and shape of the curve. Each model was fitted using the NLIN function of SAS; model accuracy was evaluated based on residual mean square error (RMSE), the magnitude and distribution of residuals, and the correlation between the observed and predicted values.

Results and Discussion

All the models, except MG, equally well portrayed the lactation profile (Table 1). Parameter estimates were significant ($P < 0.05$), with large serial correlations indicating biased predictions at various lactation stages. Lactation curves of individual cow milk yields were more varied and exhibited the tendency for a second peak which was not accurately modeled. Lactation profiles differ between dairy regions in Tasmania (Figure 1) and herd size presumably due to differences in management and pasture growth rate. Production seems to be in favour of medium to large herds although the impact of higher yields on fertility remains to be seen. However it is interesting to note the lack of a distinct peak milk yield in early lactation. The results underscore the importance of

feed intervention strategies during early lactation and suitable body condition score at calving to mitigate the impact of negative energy balance on peak milk yield.

Conclusion

Mechanistic models performed best with herd data and offer insights into

the physiological basis of lactation, while the PR models fitted better overall. Future studies will focus on the genetic and phenotypic relationship between lactation curve parameters and reproductive traits as well as the comparative advantages of crossbred cows under the prevailing production systems.

Table 1. Initial, peak, mid-lactation and final milk yields (L/cow/day) of pasture-based Holstein-Friesian cows with residual mean standard errors (RMSE) obtained by fitting average milk yield data to seven lactation functions.

| Milk Yield (L/cow/day) | | | | | | | | |
|------------------------|----------------------------|------|------|------|------|------|------|------|
| Actual | Predicted Lactation models | | | | | | | |
| Item | EXP | IG | MG | ML | PR | BC | DJ | |
| Initial | 12.0 | 12.7 | 13.2 | 2.3 | 12.8 | 12.1 | 12.0 | 12.0 |
| Peak | 14.6 | 13.9 | 13.8 | 15.1 | 13.9 | 14.2 | 14.3 | 14.5 |
| Mid | 12.4 | 12.0 | 12.0 | 13.7 | 11.9 | 11.8 | 11.9 | 11.9 |
| Final | 10.5 | 9.4 | 9.38 | 6.3 | 9.4 | 9.8 | 9.5 | 9.5 |
| Peak week | 4 | 5 | 5 | 17 | 5 | 5 | 4 | 4 |
| RMSE | 0.24 | 0.26 | 10.4 | 0.23 | 0.19 | 0.19 | 0.19 | 0.19 |

IG – incomplete Gamma, MG – modified Gamma, ML – mixed log, Exp – exponential, PR – Polynomial, BC – Biocompartmental, DJ - Dijkstra

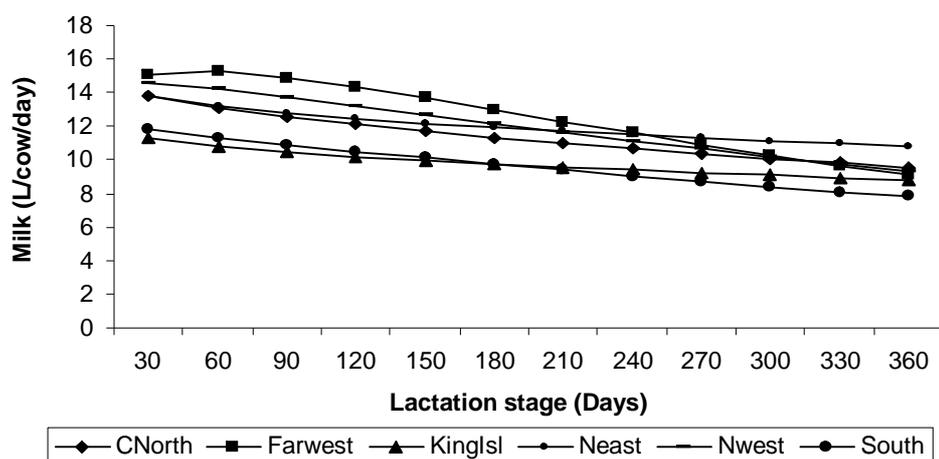


Figure 1. Predicted daily milk yield (L/cow/day) of Holstein-Friesian cows in different dairy regions of Tasmania

Acknowledgements

We thank Dairy Australia and University of Tasmania for funding and Tasherd for access to data.

References

Ali, T. E., and Schaeffer, L. R. (1987) Accounting for covariances among test day milk yields in dairy cows. *Can. J. An Sci.* **67**: 637-644.

Beever D.E, Rook, A.J, France, J, Danoa, M.S and Gill, M. (1991). A review of empirical and mechanistic models of lactation performance by the dairy cow. *Livest Prod. Sci.*, **29**: 115-130.

Dematawewa, C.M.B., Berger, P.J. (1998). Genetic and phenotypic parameter for 305-day yield, fertility and survival in Hosteins. *J. Dairy Sci.* **81**, 2700-2709.

Dijkstra, J, France, J, Dhanoa, M.S, Maas, J.A, Hanigan, M.D, Rook, A.S and Beever, D.E. (1997). A model to describe growth patterns of the mammary gland during pregnancy and lactation. *J. Dairy Sci.* **80**:2340–2354.

Ferguson J. D., and Boston, R. (1993) Lactation curves analysis: Comparison of gamma function, polynomial and exponential methods towards a mechanistic model of milk production.

J. Dairy Sci. **76** (Suppl. 1): 268. (Abstr.)

Garcia, S.C and Holmes, C.W. (2001). Lactation curves of autumn- and spring-calved cows in pasture-based dairy systems. *Livest. Prod. Sci.* **68**: 189-203.

Guo, Z., and Swalve, H. H. (1995) Modelling of the lactation curve as a sub-model in the evaluation of test day records. Paper presented at the INTERBULL open meeting, 7-8 September, 1995, Prague, Czech Republic.

Jenkins, T. G., and Ferrell, C. L. (1984) A note on lactation curves of crossbred cows. *Anim. Prod.* **39**: 470-482.

Olori, V. E., S. Brotherstone, *et al.* (1999). "Fit of standard models of the lactation curve to weekly records of milk production of cows in a single herd." *Livest. Prod. Sci.* **58**(1).

Roche, J. R., Berry, D.P. and Kolver, E.S. (2006). Holstein-Friesian Strain and Feed Effects on Milk Production, Bodyweight and Body Condition Score Profiles in Grazing Dairy Cows *J. Dairy Sci.* **89**: 3532–3543.

SAS Institute Inc. (2003) SAS User's Guide: Statistics. Version 9.1 Edition. SAS Institute Inc.: Cary, NC.

Sherchand, L., R. W. McNew, D. W. Kellog, and Z. B. Johnson. (1995). Selection of a mathematical model to generate lactation curves using daily milk yields of Holstein cows. *J. Dairy Sci.* **78**:2507–2513

Terkeli, M., Akinchi, Z. Dogan I., and Akcan A. (2000). Factors affecting the shape of Lactation Curves of Holstein Cows from the Balikesir Province of Turkey. *J. Dairy Sci* **83**:1381–1386
1381

Tozer, P.R; Huffaker; R.G; 1999: Mathematical equations to describe lactation curves for Holstein-Friesian cows in New South Wales. *Aust.J. Agric. Res.* **50**:431-440

Wilmink, J. B. M. (1987) Adjustment of test-day milk, fat and protein yield for age, season and stage of lactation. *Livestock Prod. Sci.* **16**: 335-348.

Wood, P. D. P. (1967). Algebraic model of the lactation curve in cattle. *Nature* 216, 164.5.

In: Bill Fulkerson (Editor) Current Topics in Dairy Production, Volume 12', *Proceedings of the Dairy Research Foundation Symposium*, The University of Sydney, Australia, 74-78 (2007)