

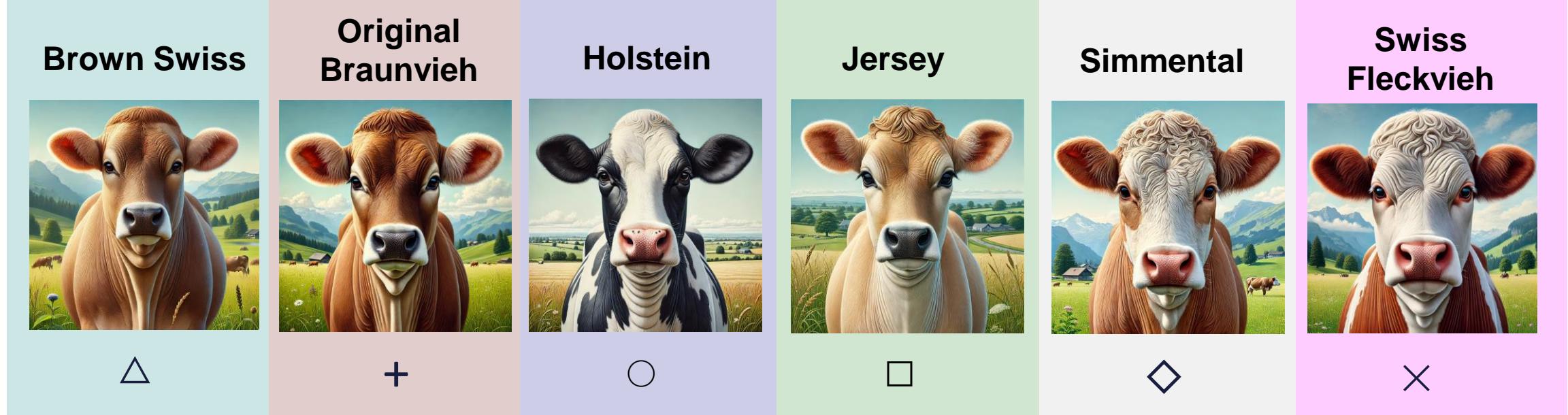
# Modelling the Effect of Heat Stress on the Performance of Swiss Dairy Cows

## A Big Data Statistical Analysis

Arno Schneuwly



# Poll: Which is the most heat-resilient breed in terms of ECM yield?



For which breed is the THI where the ECM yield starts to drop the highest?

# Dairy Farming and Climate in Switzerland

Im Tal haben die Kühe Hunger, auf der Alp haben sie Durst: Wie die Hitze das Vieh trifft

Das Wasser wird knapp. Das trifft die Bauernbetriebe besonders hart. Jetzt haben einige Kantone mehr Flächen freigegeben und wollen so dem Futtermangel vorbeugen. Aber nicht überall ist das Futter das Problem.



NZZ (2022)

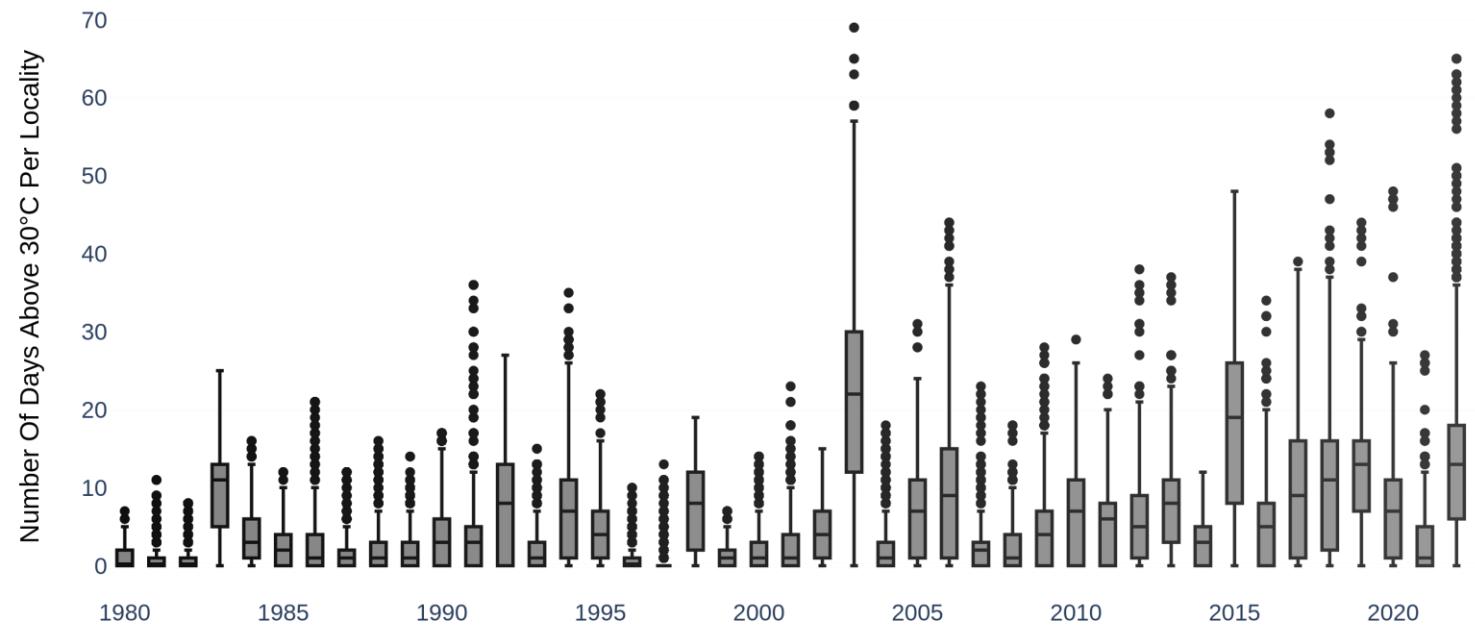


Agroscope (2023)

**Wasserdusche gegen Hitzestress für Kühe**



Milchbauernhof, Instagram (2022)



Zu hohe Temperaturen für Kühe  
Deshalb leiden Kühe unter Hitzestress

Kühe mögen es kühl und leiden im Sommer deshalb unter Hitzestress.  
Woran liegt das? Eine Studie bringt neue Erkenntnisse.

Samstag, 22.07.2023, 14:46 Uhr

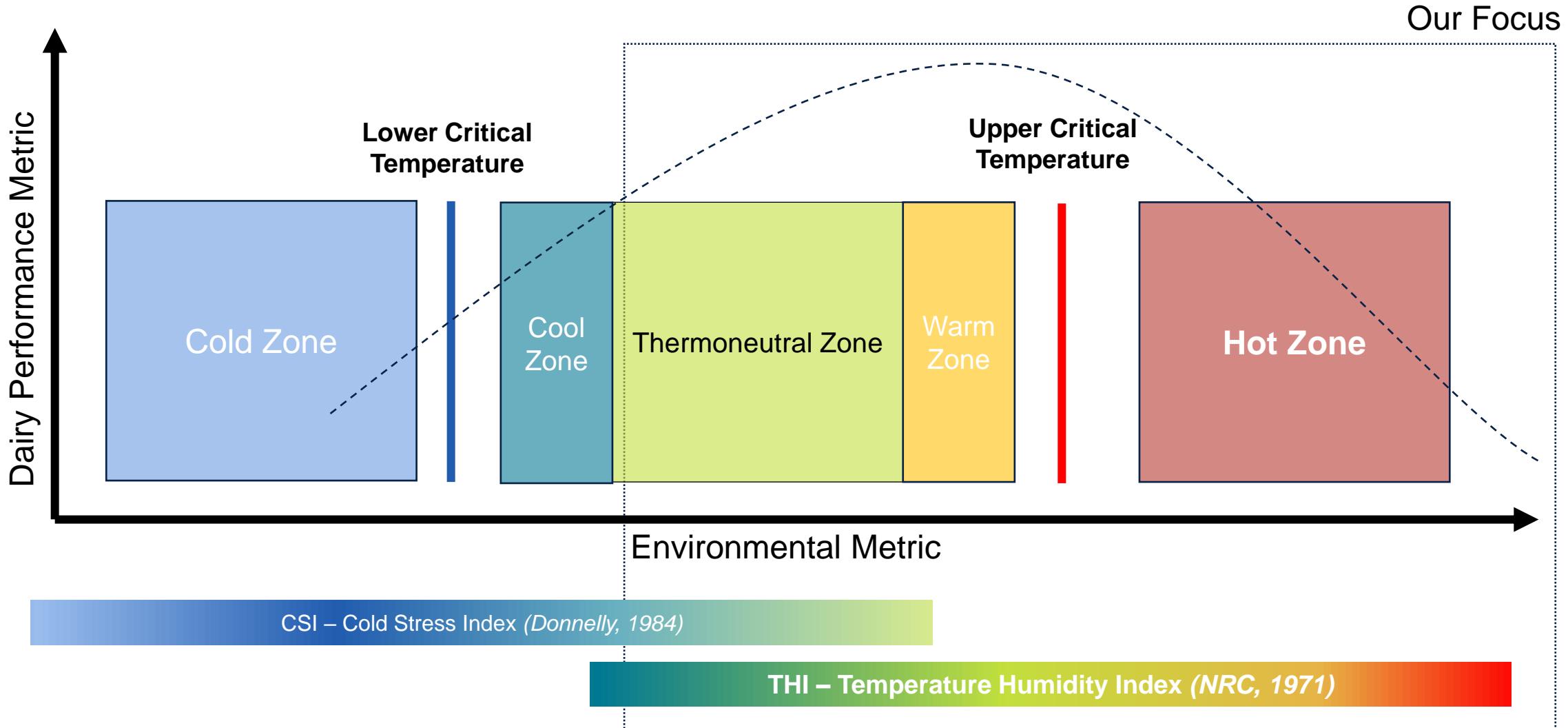
Auch jetzt im Sommer sind die Rinder und Kühe von Bauer Stefan Käser im

SRF (2023)

**Heat stress appears to be a problem. How big is it? Today, we try to quantify.**



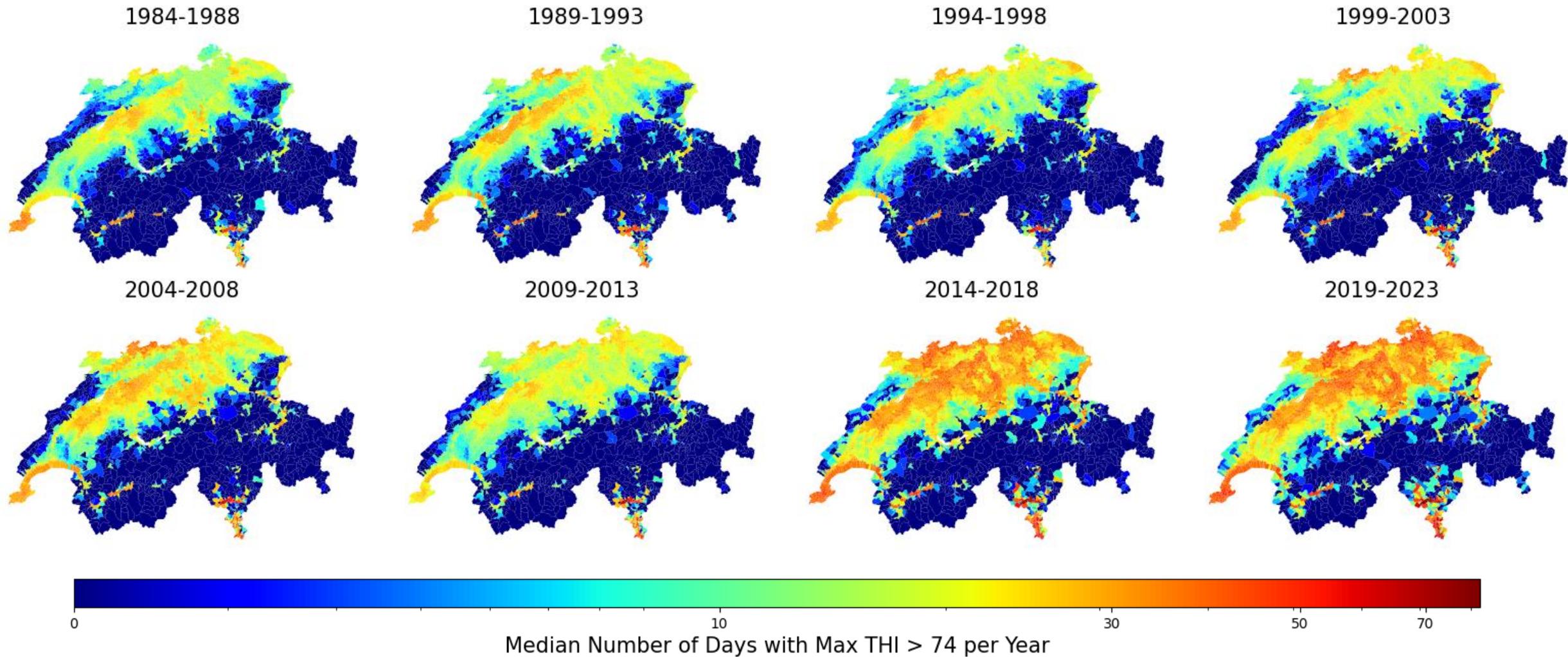
# Animal Physiology 101 - What is Heat Stress?



**THI is the agronomic metric for heat stress.**



# THI over Time in Switzerland



**High-THI exposure is rising. The effect of heat-incurred dairy performance losses is unknown for CH.**



# Heat Stress on Dairy Cow Performance Across Breeds

Study	Breeds	Records	Farms	Cows	Time	Location	Model
Bryant et al (2007)	3	~ 65 K	~ 0.5 K	> 19 K	1990 – 2002		Mixed
Gantner et al (2017)	2	~ 2.3 M	~ 1.5 K	> 156 K	2005 – 2012		Mixed
Ahmed et al (2022)	4	~ 5 M	~ 1.4 K	?	2016 – 2019		Linear / GAM
→ Our work	6	> 130 M	~ 46 K	~ 4.2 M	1982 – 2023		GAM

Heat stress studies of non-experimental production systems across breeds are very scarce.



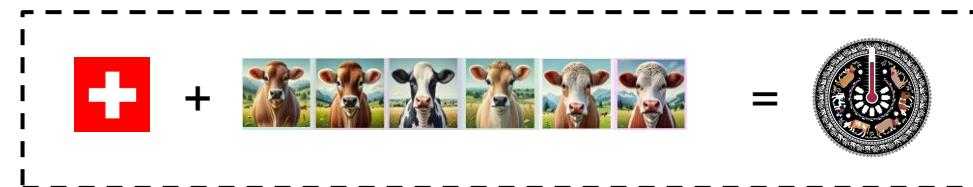
# Research Question & Hypothesis

Can we and at which THI value(s) do we observe a change (increase / decrease) in dairy performance variables for the different dairy cow breeds in Switzerland?

Variable	General Hypothesis	Breed Specific Hypotheses
Milk Yield [kg/day]	Decrease	Different THI thresholds for different breeds – Kadzere (2002) Jersey (□) more heat stress resistant than Holstein (○) – Bryant (2007)
ECM Yield [kg/day]	Decrease	Simmental (◇) more heat resistant than Holstein (○) – Gantner (2017)

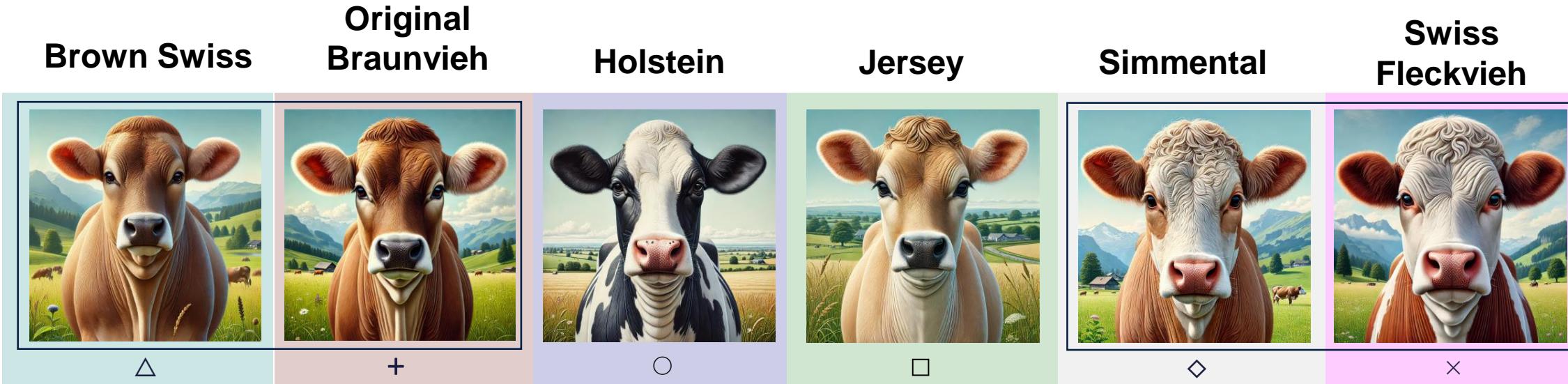
well-known

Gap & Contribution





# Popular Cow Breeds in Switzerland – 2023 Performance



Purpose	Milk	Milk & Meat	Milk	Milk	Milk & Meat	Milk & Meat
Milk [kg/day]	<b>23.04</b> ( $\pm 7.55$ )	<b>19.33</b> ( $\pm 6.48$ )	<b>27.20</b> ( $\pm 8.58$ )	<b>18.88</b> ( $\pm 6.10$ )	<b>19.39</b> ( $\pm 6.50$ )	<b>22.45</b> ( $\pm 7.51$ )
Protein [%]	3.56 ( $\pm 0.43$ )	3.42 ( $\pm 0.42$ )	3.42 ( $\pm 0.42$ )	3.97 ( $\pm 0.51$ )	3.44 ( $\pm 0.37$ )	3.46 ( $\pm 0.42$ )
Fat [%]	4.13 ( $\pm 0.61$ )	4.01 ( $\pm 0.57$ )	4.22 ( $\pm 0.67$ )	5.30 ( $\pm 0.94$ )	4.04 ( $\pm 0.60$ )	4.27 ( $\pm 0.69$ )
ECM [kg/day]	<b>25.56</b> ( $\pm 8.02$ )	<b>20.95</b> ( $\pm 6.77$ )	<b>30.13</b> ( $\pm 8.90$ )	<b>24.17</b> ( $\pm 7.15$ )	<b>21.14</b> ( $\pm 6.89$ )	<b>25.02</b> ( $\pm 7.88$ )

Breeding commonly optimizes for performance. Heat resilience is not explicitly included.



# Our Dataset Statistics

**Brown Swiss**



△

**Original Braunvieh**



+

**Holstein**



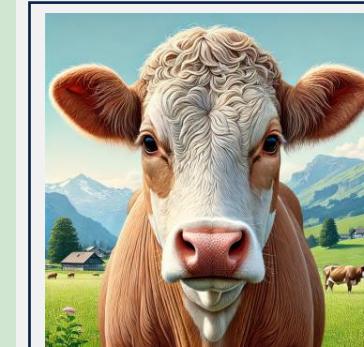
○

**Jersey**



□

**Simmental**



◇

**Swiss Fleckvieh**



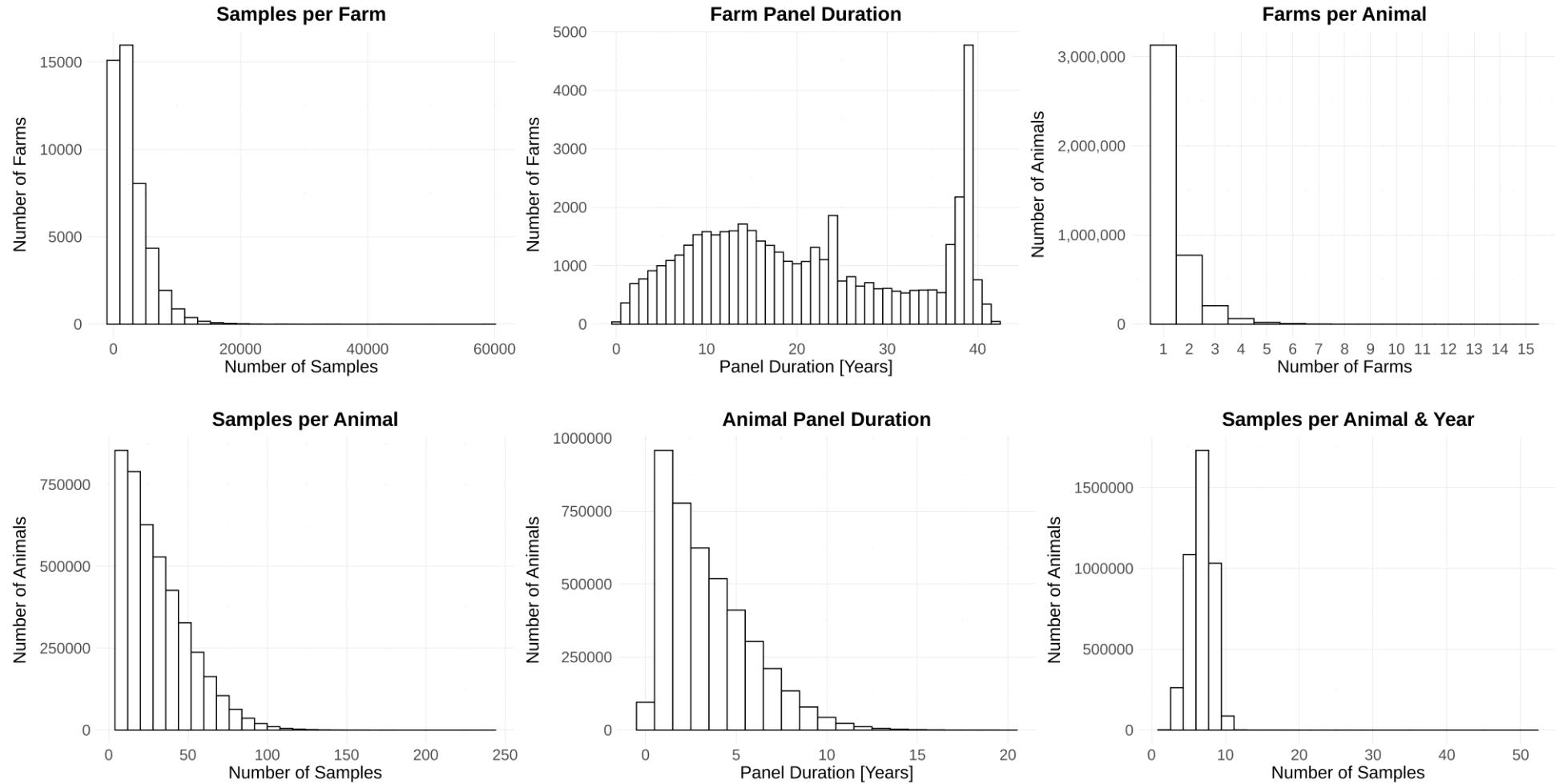
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<b># Samples</b>	56 M	5 M	28 M	700 K	9 M	30 M
<b># Farms</b>	26 K	18 K	25 K	4 K	19 K	27 K
<b># Animals</b>	1.7 M	150 K	900 K	24 K	300 K	1 M
<b>Timespan</b>	1982 - 2023	1982 - 2023	1985 - 2023	1998 - 2023	1984 - 2023	1984 - 2023

This is real-world data collected in commercial farming over 40 years. No experimental design.



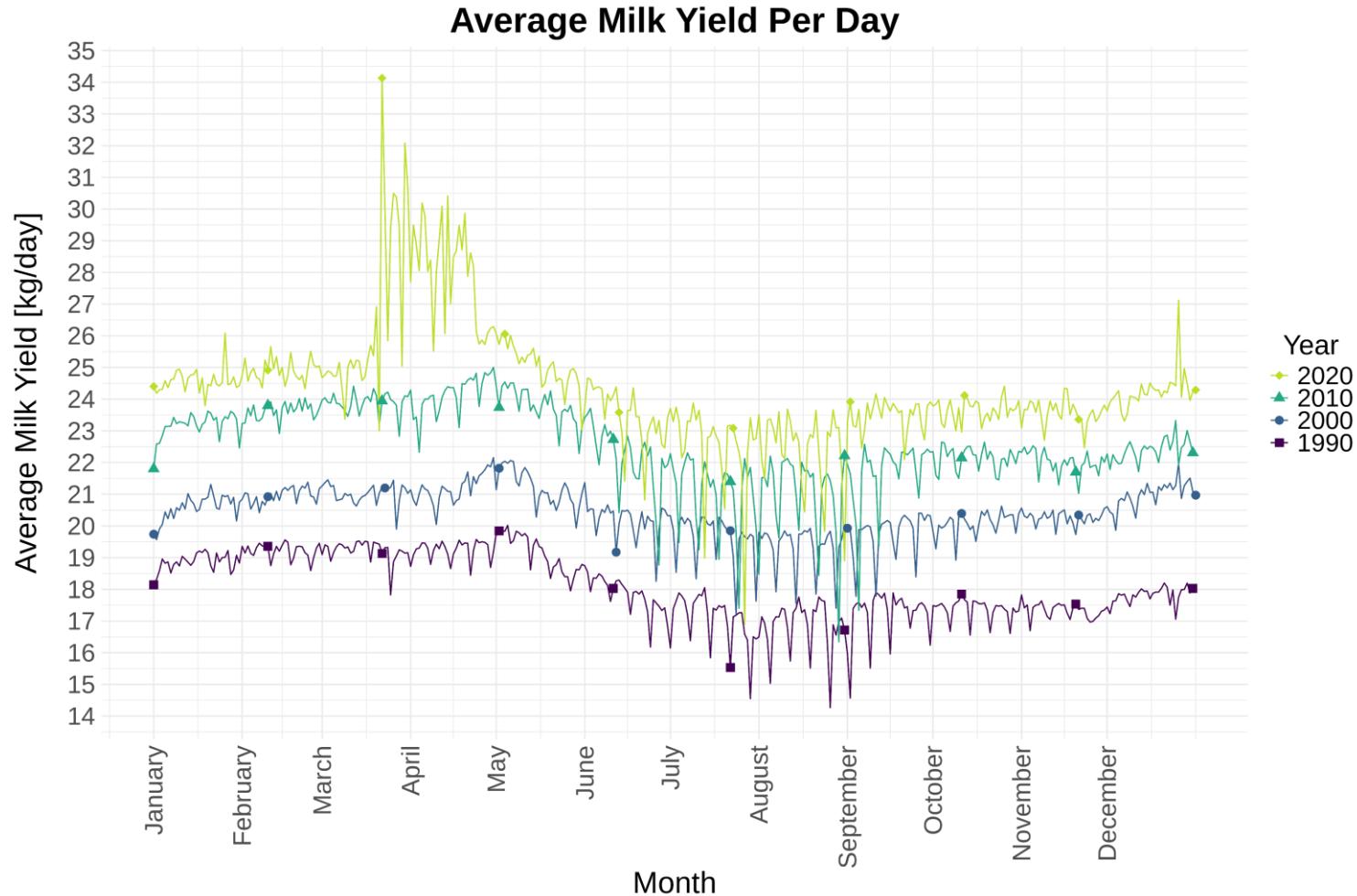
# Structure of Our Data – Sparsity & Dynamics



**Sparse and hierarchical data. Farms enter and exit. Animals enter, exit and change farms. Irregular sampling frequencies.**



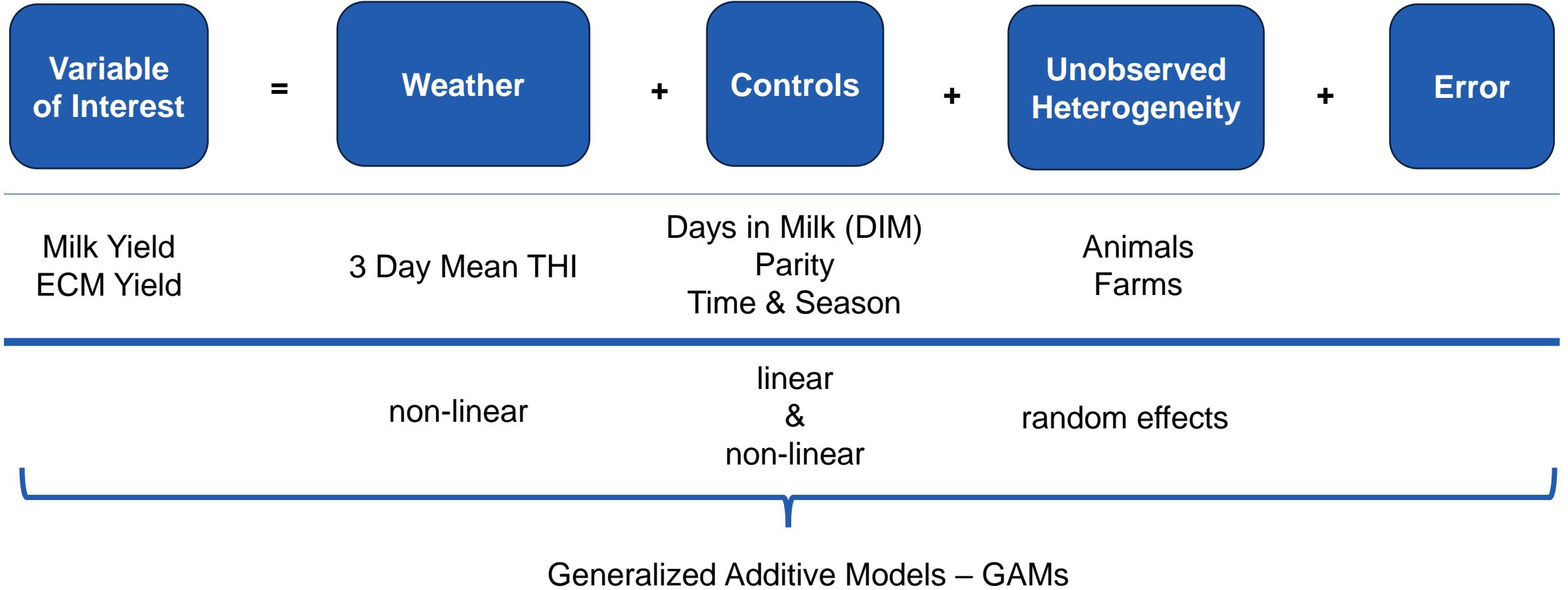
# Structure of Our Data – Time & Seasonality



Breeding Progress. Policy Changes. Sampling Patterns. Annual Seasonality.



# Empirical Strategy



**Model the non-linearities and derive the threshold values with numerical methods.**



# Single Breed Model

$$Y_{ijkt} = \beta_0$$

$$+ f_{1,1}(\text{THI}_{kt}) \cdot \mathbb{I}(\text{Primiparous}_i)$$

$$+ f_{1,2}(\text{THI}_{kt}) \cdot (1 - \mathbb{I}(\text{Primiparous}_i))$$

Intercept

$$+ f_{2,1}(\text{DIM}_{it}) \cdot \mathbb{I}(\text{Primiparous}_i)$$

$$+ f_{2,2}(\text{DIM}_{it}) \cdot (1 - \mathbb{I}(\text{Primiparous}_i))$$

$$+ \beta_1 \cdot \mathbb{I}(\text{Primiparous}_i)$$

Weather

Controls

$$+ \sum_{m=1}^{M-1} \beta_{2m} \cdot \mathbb{I}(\text{Year}_t = m)$$

Time

$$+ u_{kt}$$

Seasonality

$f_1(\text{THI}_{kt})$  is a smooth function of THI on farm  $k$  at time  $t$

$$+ v_k$$

Farms

$f_2(\text{THI}_{it})$  is a smooth function of DIM of sample  $i$  at time  $t$

$$+ w_i$$

Animals

primiparous vs multiparous

$$+ \epsilon_{ijkt}$$

Error

dummy for each year

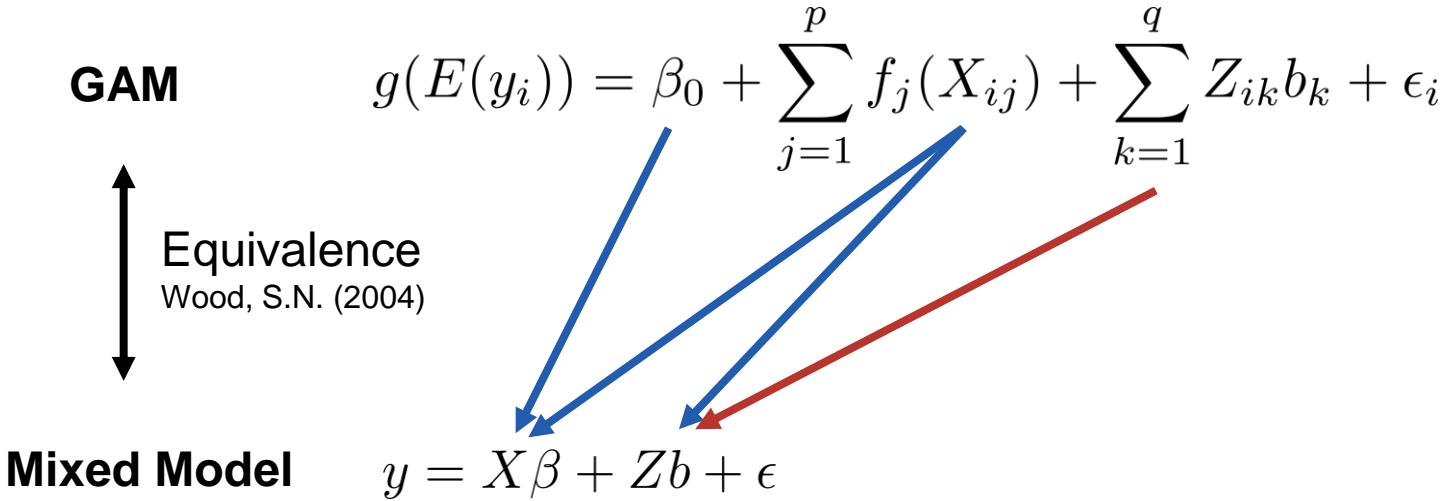
RE of zip code of farm  $k$  and month of  $t$

RE of farm  $k$

RE of for animal  $i$

GAMs are the state-of-the-art statistical framework to fit our model.

# An Unforeseen Challenge... The Sparsity of our Data...



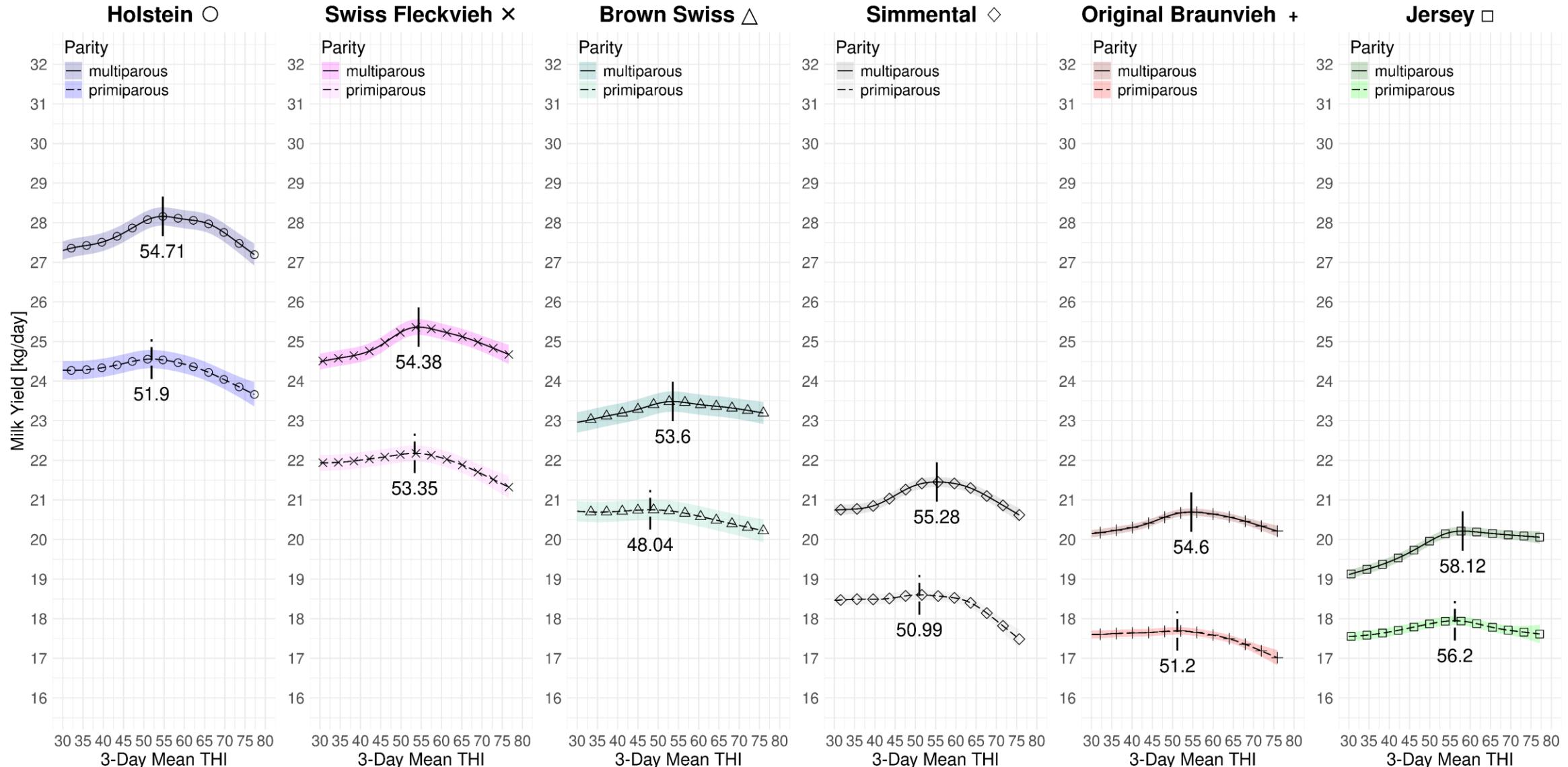
Our contribution →

Package	Factor Levels
<b>MGCV</b> Wood (2017)	hundreds
<b>gamm4</b> Wood, Scheipl (2020)	thousands
<b>gamm4b</b> <b>gammJ</b>	tens of thousands ~3000% speed-up

We extended libraries to accommodate GAMs with tens of thousands of factor levels.

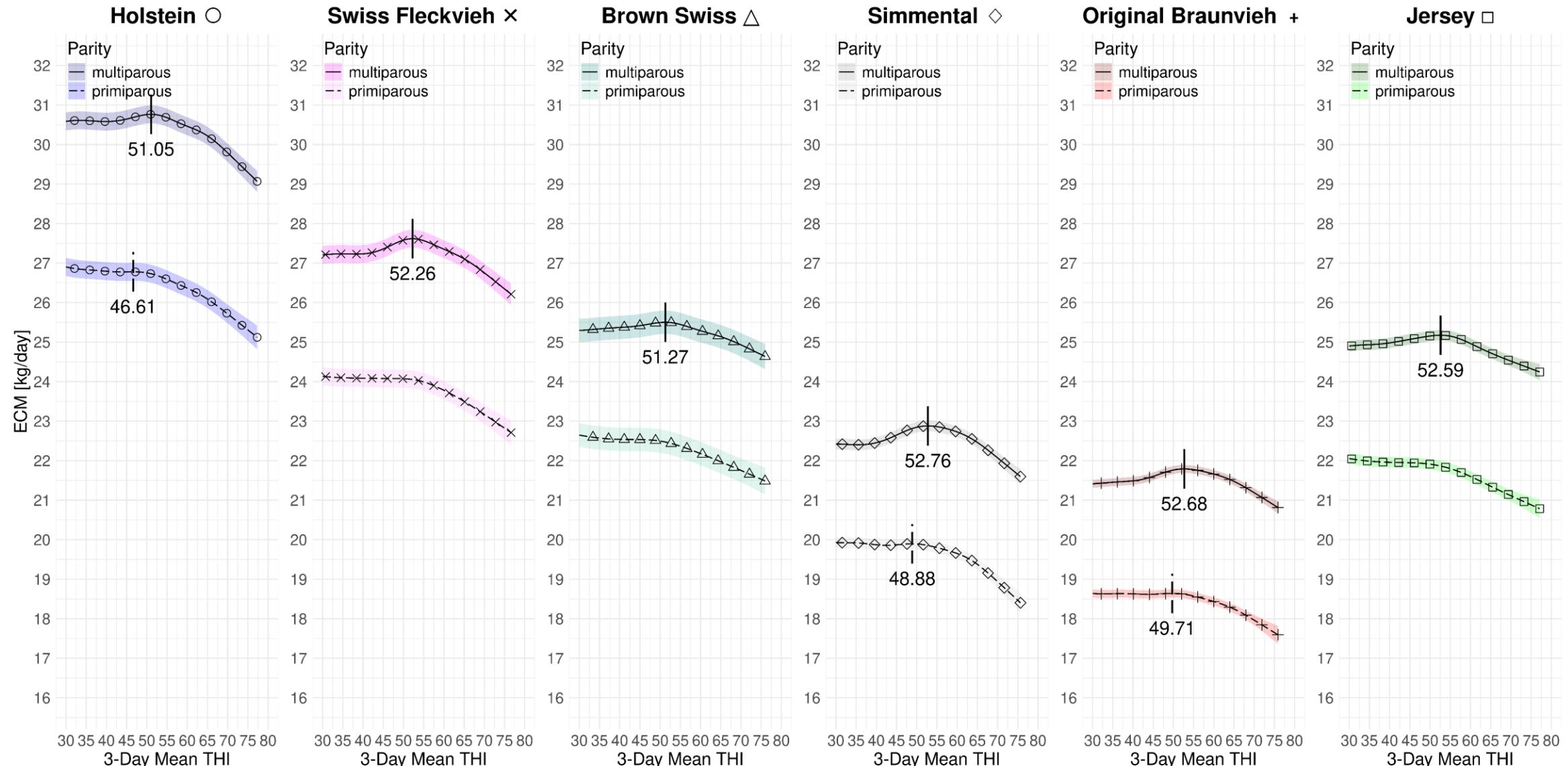


# Results: Milk Yield - THI Effect & Intercepts (Mean + Year 2023)





# Results: ECM Yield - THI Effect & Intercepts (Mean + Year 2023)



# Discussion

		THI Thresholds Milk Yield [kg/day]								THI Thresholds ECM Yield [kg/day]					
		HO	SF	BS	SI	OB	JE			HO	SF	BS	SI	OB	JE
Multiparous	HO	54.71	54.38	53.60	55.28	54.60	58.12	Multiparous	51.05	52.26	51.27	52.76	52.68	52.59	
	SF	51.90	53.35	48.04	50.99	51.20	56.20		46.61	-	-	48.88	49.71	-	
Primiparous	○	✗	△	◇	+	□	Primiparous	○	✗	△	◇	+	□		
	BS	54.71	54.38	53.60	55.28	54.60	58.12	51.05	52.26	51.27	52.76	52.68	52.59		
Multiparous	SI	51.90	53.35	48.04	50.99	51.20	56.20	Primiparous	46.61	-	-	48.88	49.71	-	
	OB	54.71	54.38	53.60	55.28	54.60	58.12		51.05	52.26	51.27	52.76	52.68	52.59	
Primiparous	JE	51.90	53.35	48.04	50.99	51.20	56.20		46.61	-	-	48.88	49.71	-	

**Low THI Thresholds:** not unseen, but not common - *Ahmed et al (2022), Hill et al (2015), Vinet et al (2023)*

**Component drop:** Bryant et al (2007), Chen et al (2024)

**Lower thresholds for primiparous:** similar findings by *Maggiolino et al (2022) , J. Castro-Montoy (2019)*

contradicts *Bernabucci et al (2014), West (2004), Aguilar et al (2010)*

**Early component drop weighs out differences in volume drops across breeds!**



# Limitations & Potential Next Steps

## 6 Single Breed Models

- Computational limits → multi-breed model
- Subsampling strategies
- Stacking / ensemble techniques



Limited interpretability of p-values (non-experimental design) **p\*\*\***

## Spatial Autocorrelation & Confounding & Seasonality

- Better modelling than with the *zip X month RE* → longitude, latitude, day of year smooths 
- Spatial+ Dupont et al (2020)
- Model selection

Farm as fixed effects (MixedModels.jl such an experiment – sparse FE matrix support) 

Other weather effects: precipitation, sunshine duration, radiation, exact lactation numbers 

Other performance variables: Somatic Cell Count, lactose, (rotein, fat separately - included in ECM)

# Contributions

Agronomy / Animal Science



performance critical **THI thresholds** for 6 breeds



**non-linear marginal effects** of THI for 6 breeds

unprecedented **granularity** and **scale** of data, even if we subsampled

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Statistics / Computational Science  $\frac{-x}{+ =}$

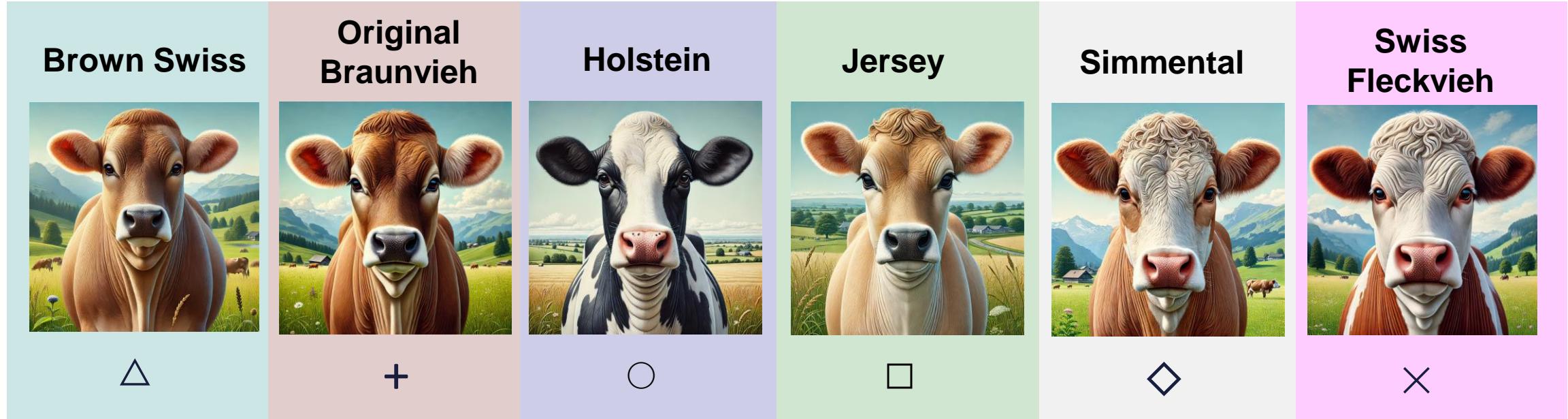
Identified a **bottleneck** in MGCV for random effects with a high number of factor levels

Fixed gamm4 bugs, introduced an improved version **gamm4b**

**gammJ**: modified gamm4 with a bridge to MixedModels.jl to support GAMs



# Poll: Which is the most heat-resilient breed in terms of ECM yield?



5

2

6

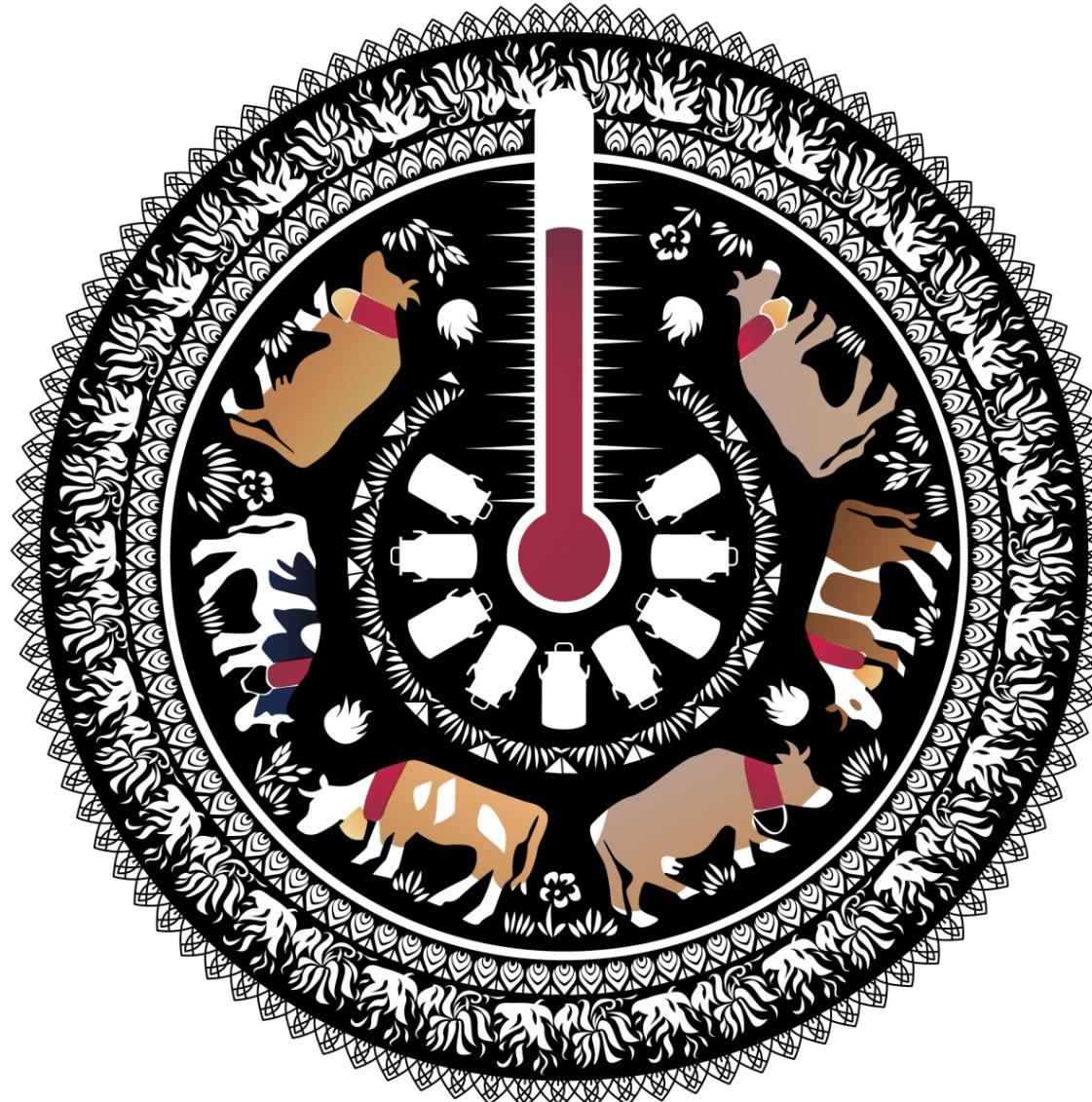
3

1

4

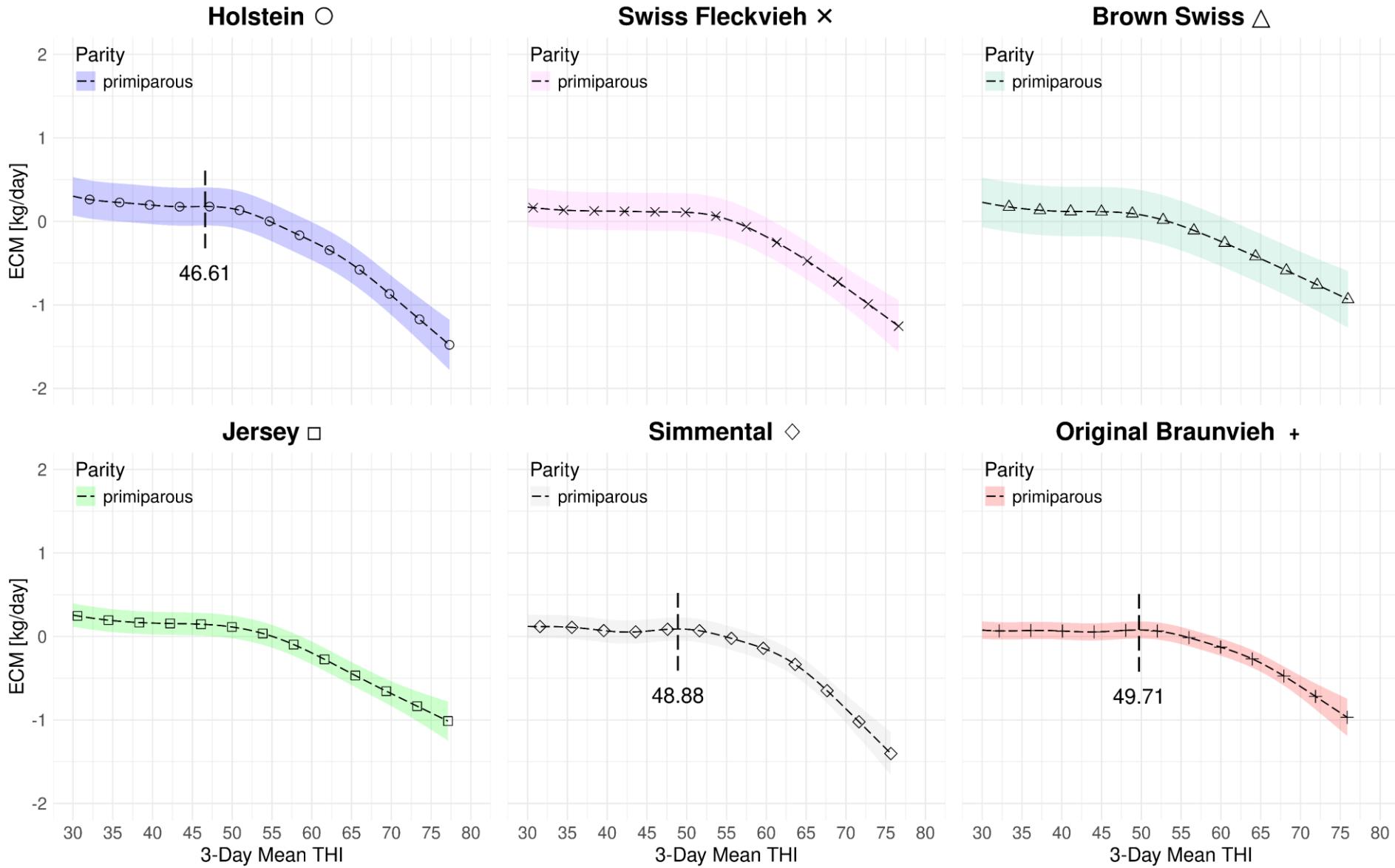
The threshold differences across breeds are minimal for ECM yield! Similar resilience for all breeds.

# Appendix

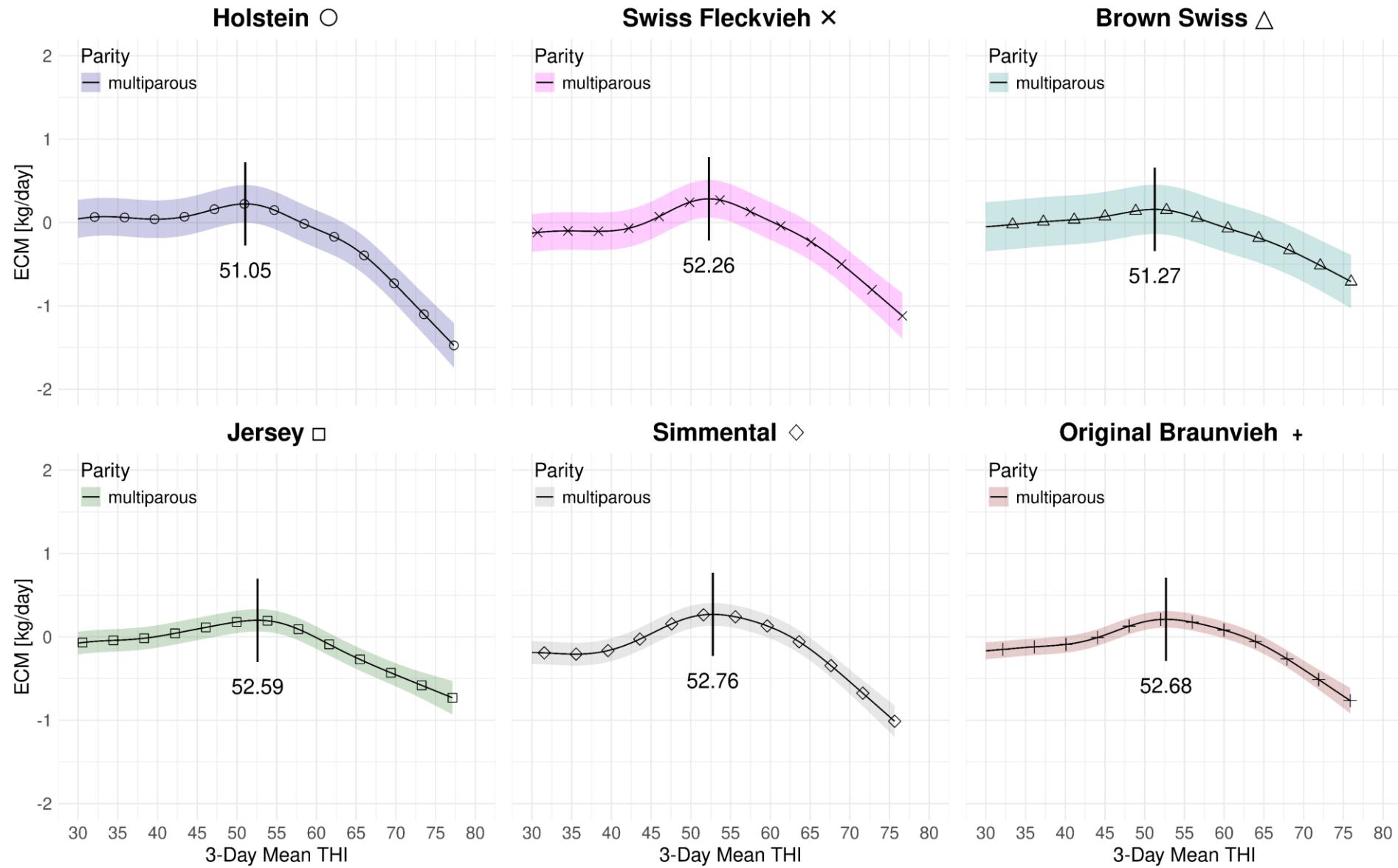


# ECM Full Time Range

# Results: ECM Yield – Marginal Effects for Primiparous Cows

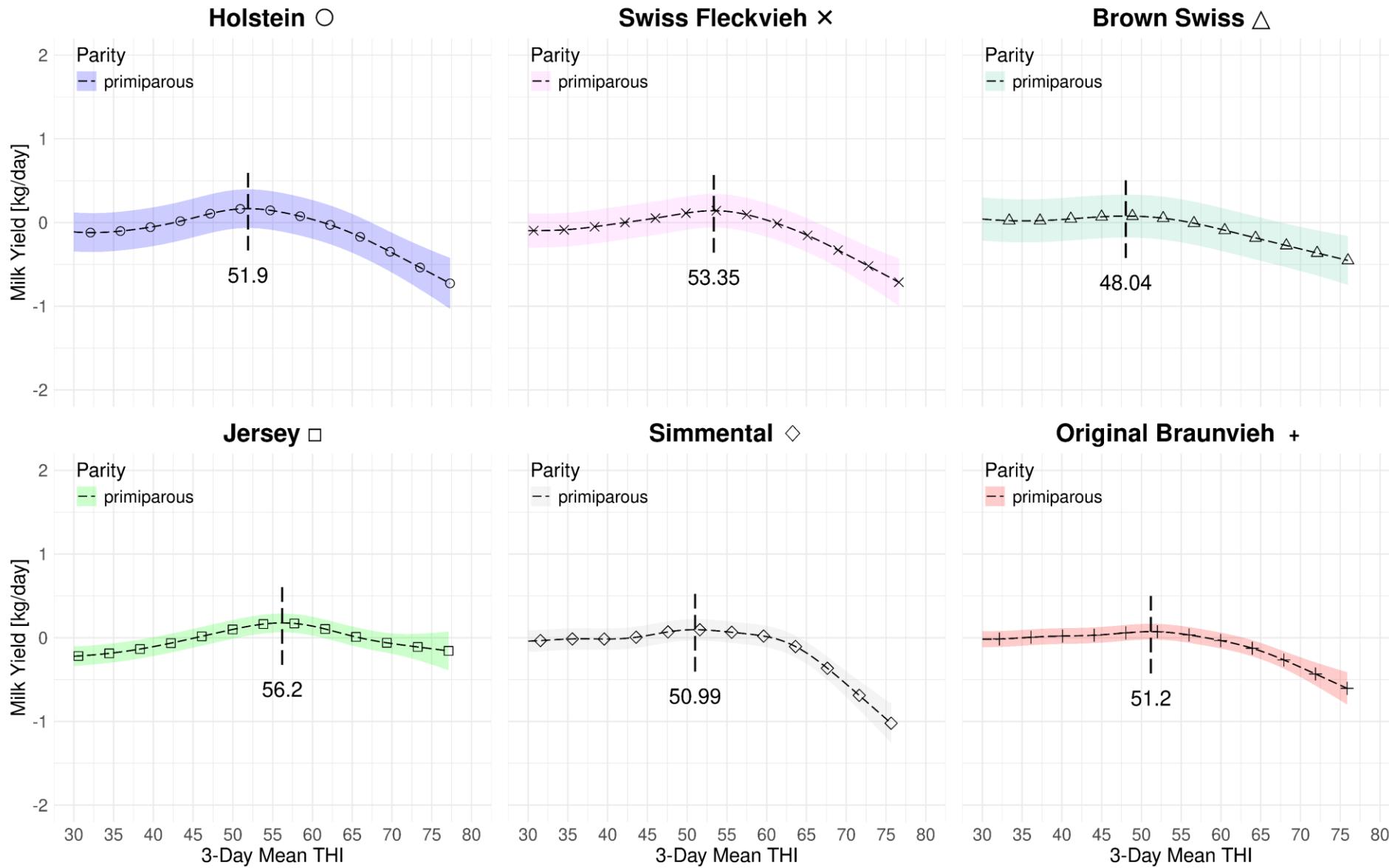


# Results: ECM Yield – Marginal Effects for Multiparous Cows

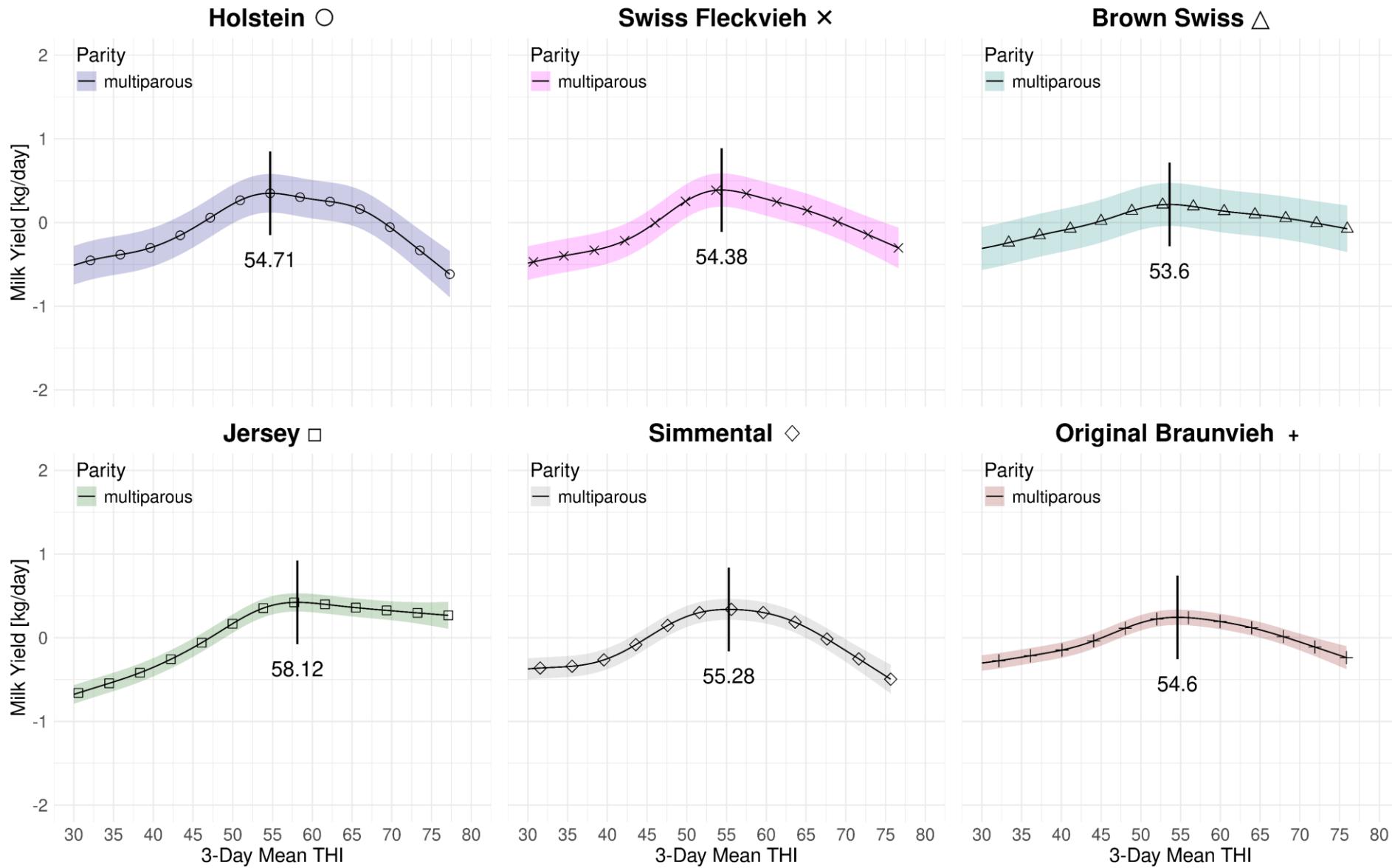


# Milk Yield Full Time Range

# Results: Milk Yield – Marginal Effects for Primiparous Cows

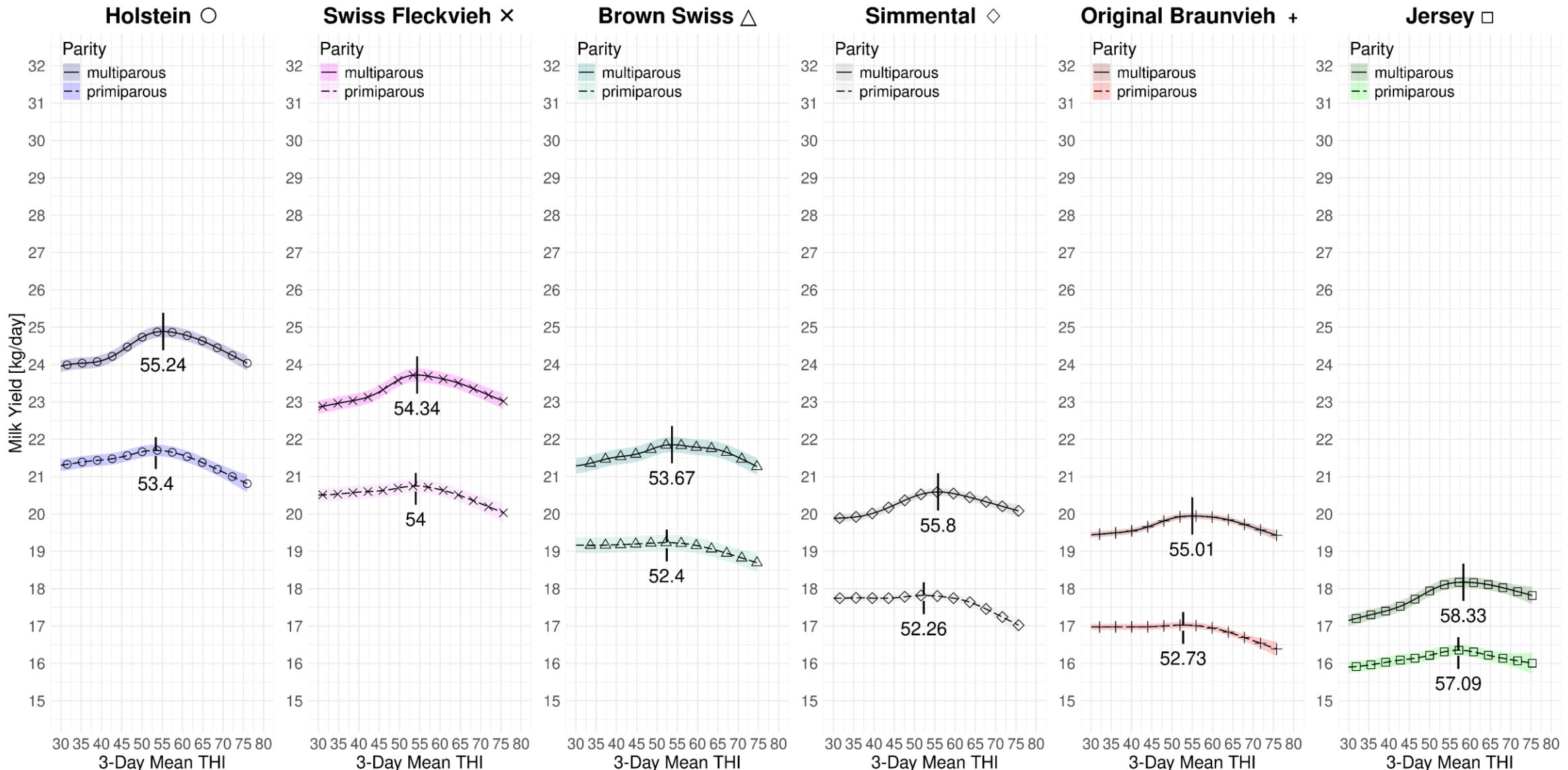


# Results: Milk Yield – Marginal Effects for Multiparous Cows

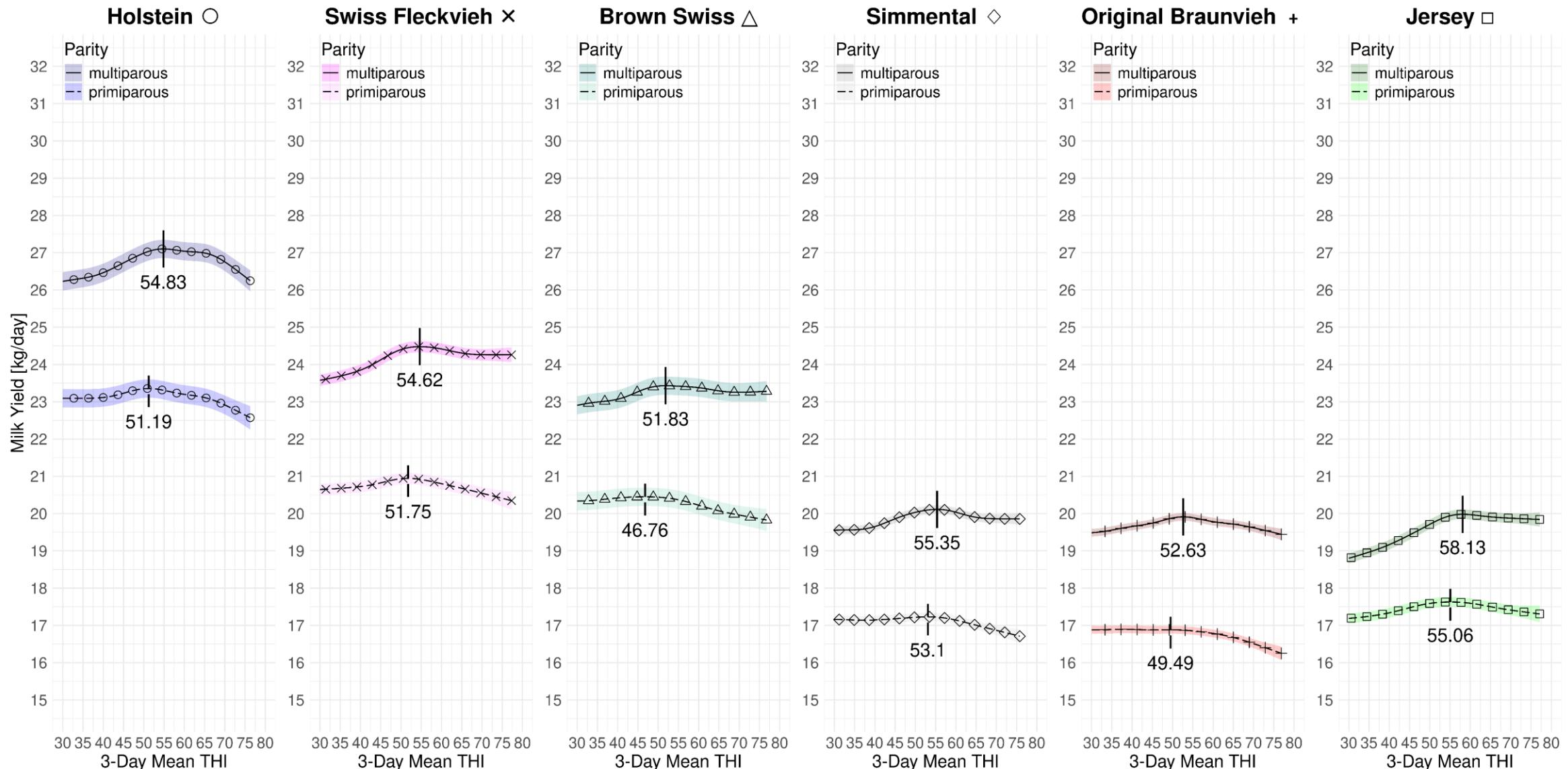


# Milk Yield Before & After 2010

# Results: Milk Yield – Marginal Effects for Primiparous Cows before 2010

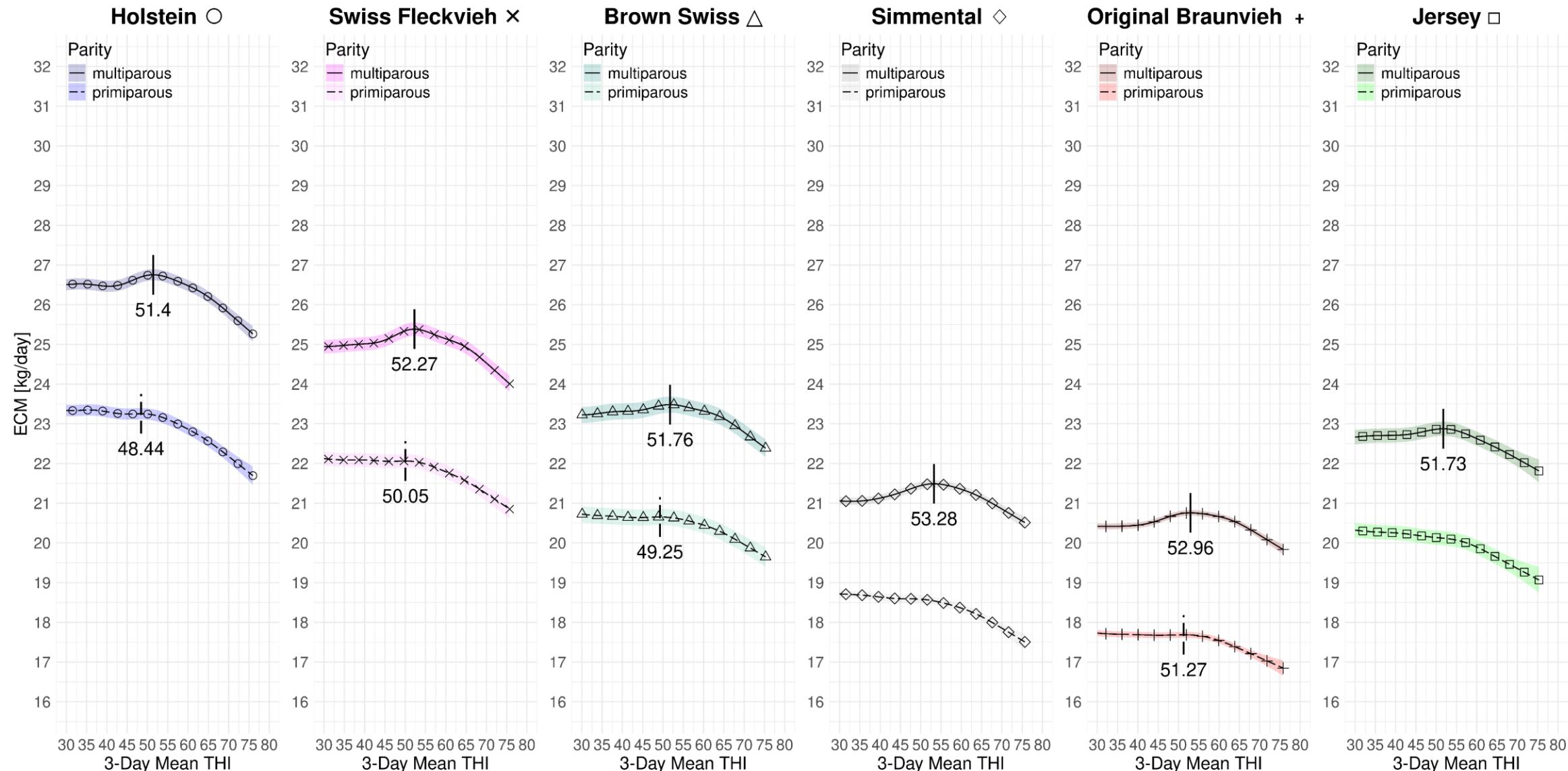


# Results: Milk Yield – Marginal Effects for Primiparous Cows after 2010

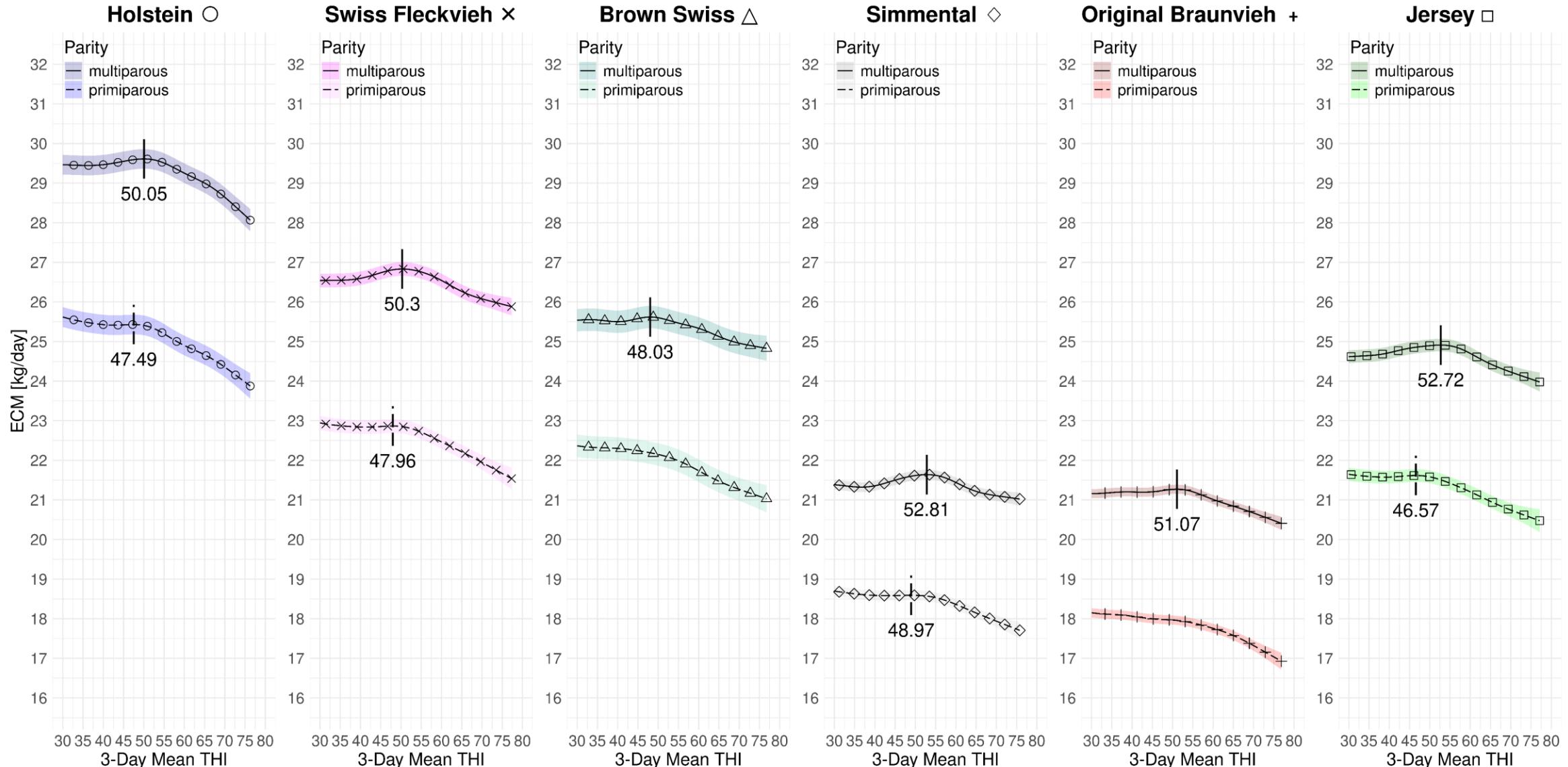


# ECM Yield Before & After 2010

# Results: ECM Yield – Marginal Effects for Primiparous Cows before 2010

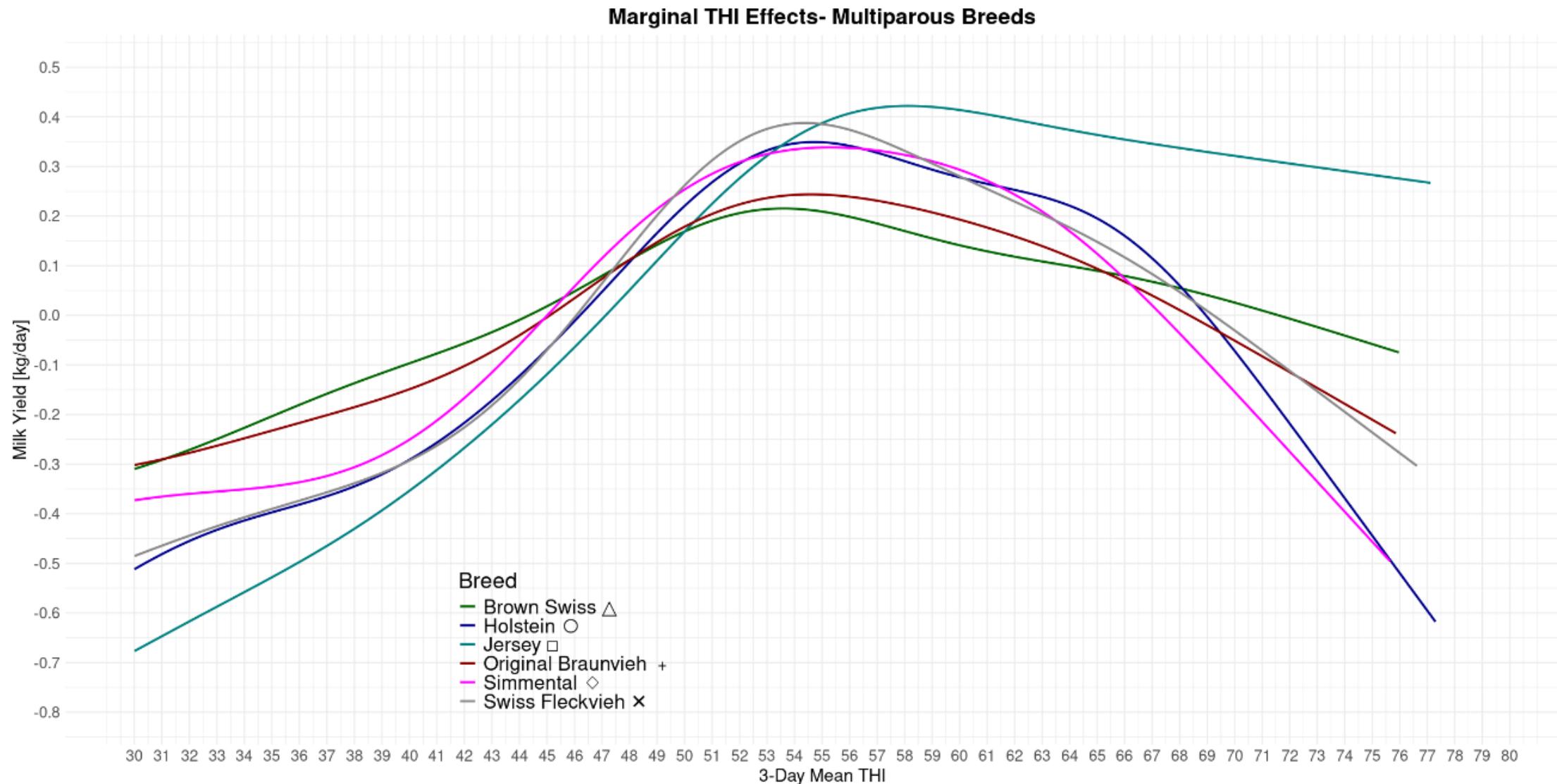


# Results: ECM Yield – Marginal Effects for Primiparous Cows after 2010

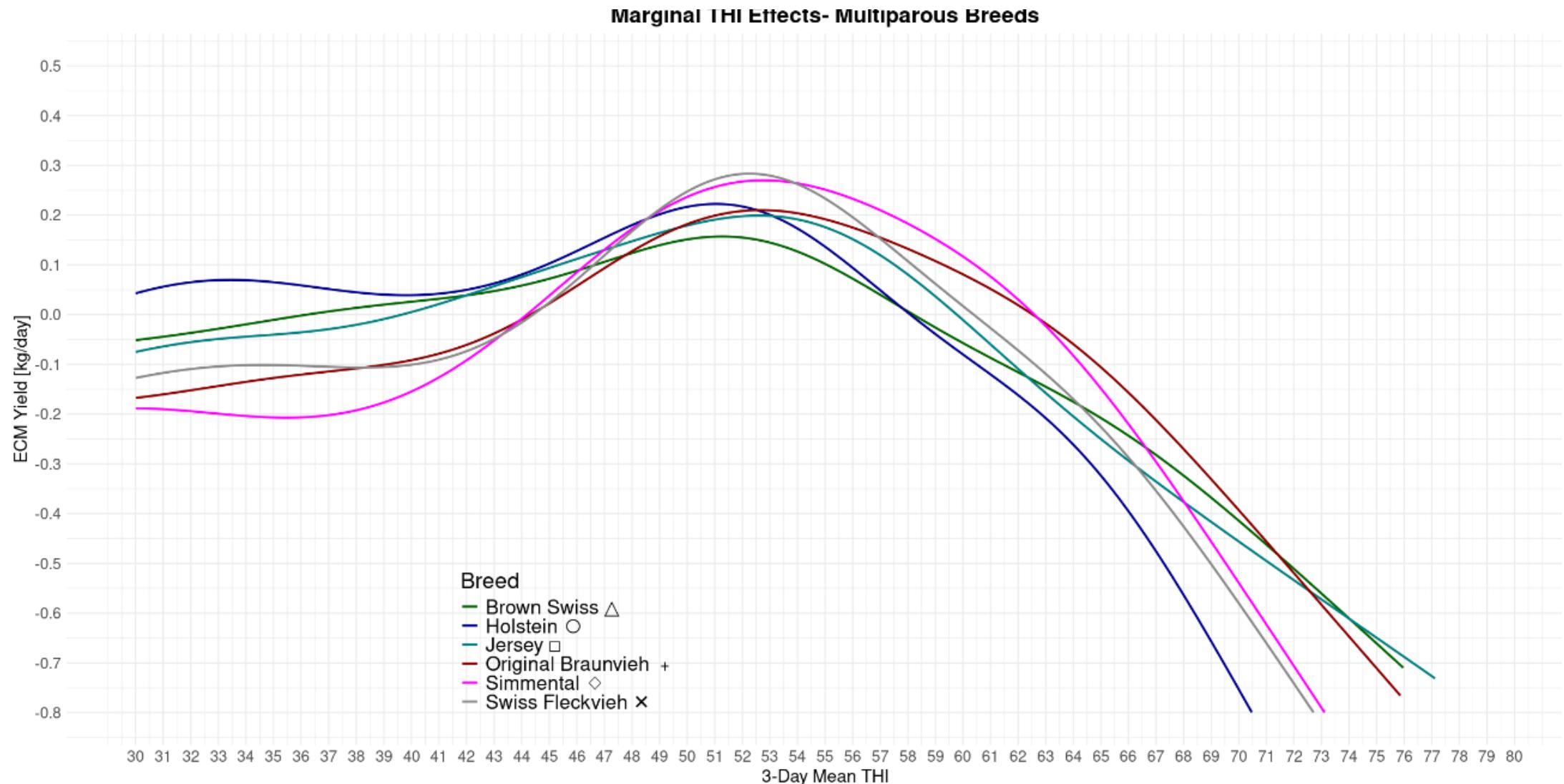


# Marginal Effects – Full Period

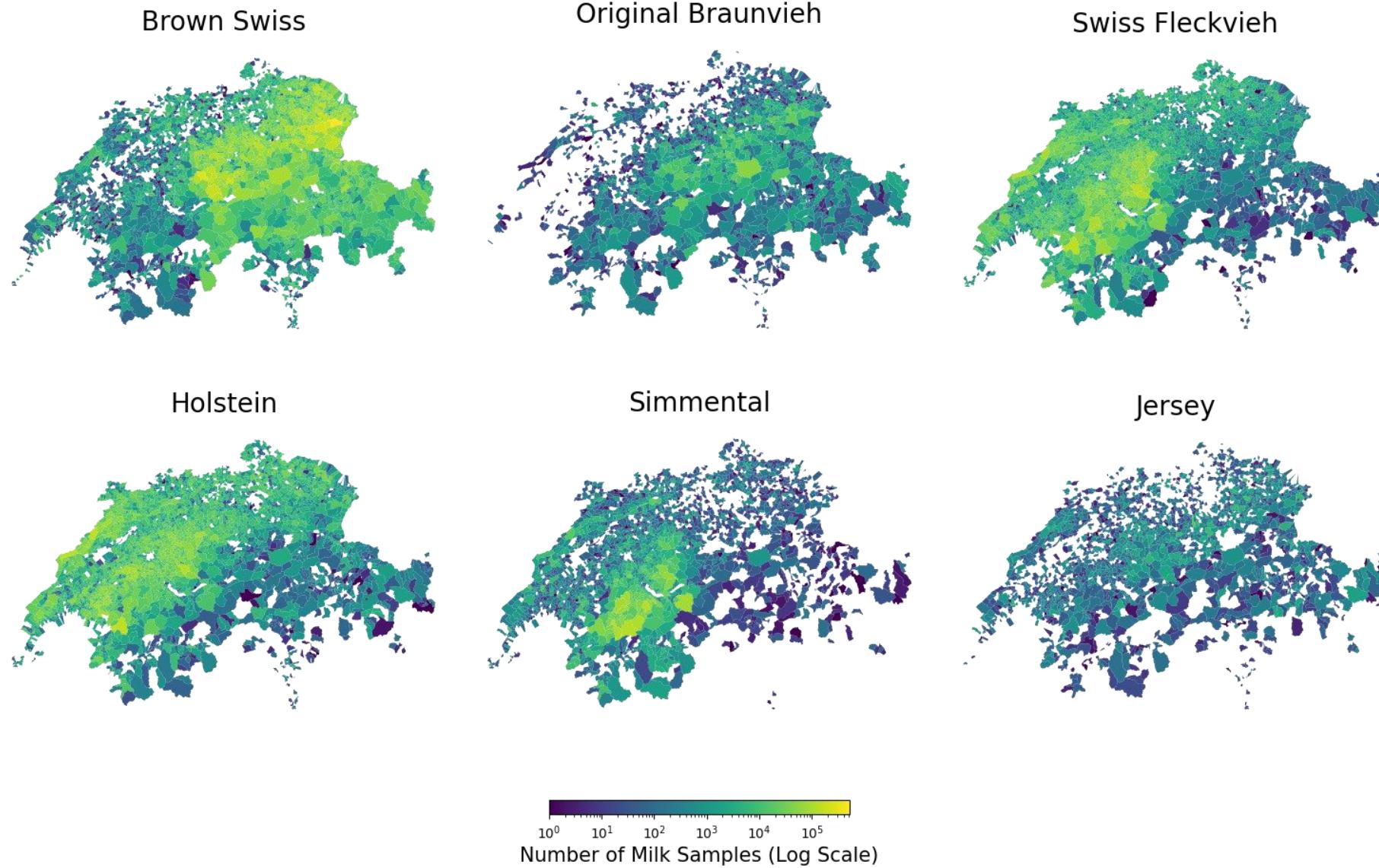
# Marginal Effects



# Marginal Effects



# Sample Distribution by Breed ( > 1000 samples)



# Multi-Stage Data Cleaning

IQA Filtering – Drop unrealistic values

E.g. 90 kg milk, 90% Protein

Only take samples where all target variables of interest are simultaneously available

Milk, Protein, Fat

Drop Research Farms, Farm Schools, Breeding Associations, Research Organizations

ETH, Agroscope, Qualitas

Only Farms in Switzerland – Drop foreign farms

Keep cows with international ID

**Conservative data cleaning approach.**

# Agricultural Policies in Switzerland

Agricultural Policies	Enactment
RAUS	1993
BTS	1996
Milk price supplement	1999
Milk quota abolition	2009
Grassland-based feeding	2014
Commercial milk	2019
Pasture payment	2023