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Getting prepared for Metabolomics studies: deliberate selection of samples from large animal studies

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Context

Large sample sizes allow for studying complex traits and rare incidents (e.g. fertility and diseases)



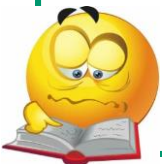
BUT: Sample numbers for OMICs studies **have to be limited to fewer samples**



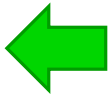
Need for selecting meaningful subsets of animals/samples representing the desired traits



OPTION 1
You can select animals based on specific and fixed labels
→ ***“human selection”***



How?



OPTION 2
You can select animals based on Artificial Intelligence techniques
→ ***“Machine Learning selection”***





Aims

**THE AIM OF THIS PROJECT WAS TO USE MACHINE LEARNING AS A
TOOL TO SELECT MEANINGFUL SAMPLES THAT ARE REPRESENTATIVE
OF A VARIATION IN THE POPULATION BY TESTING DIFFERENT
HYPOTHESIS**





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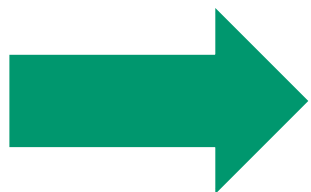
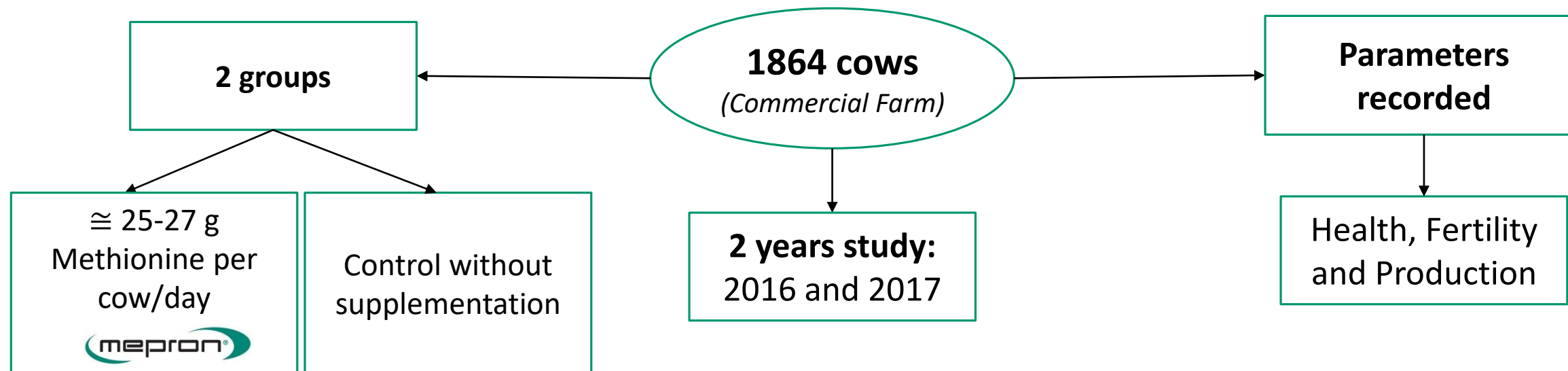
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Starting Database





Animal Trial



All these data were collected in Excel files + all the records relative to the cows were taken with the management software **DairyComp DC 305** (Valley Agricultural Software, USA) (e.g. cows changing groups, mastitis, sick pen...)



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Screening & Animal Selection





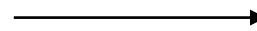
Screening of the Database



Starting from **1864 cows**

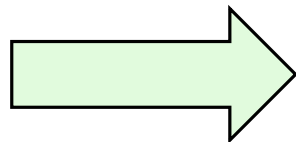


Checked: **blood samples** and **availability of results** (*BCS, BFT, Production results...*)



Checked Met supplementation:
20-70 DIM + no more 5 DIM out of a group

Cows who entered the **trial twice** receiving Met in the 1st one were excluded



From **1864 cows** to **1038** (*Masterfile*)



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Using Machine Learning Techniques

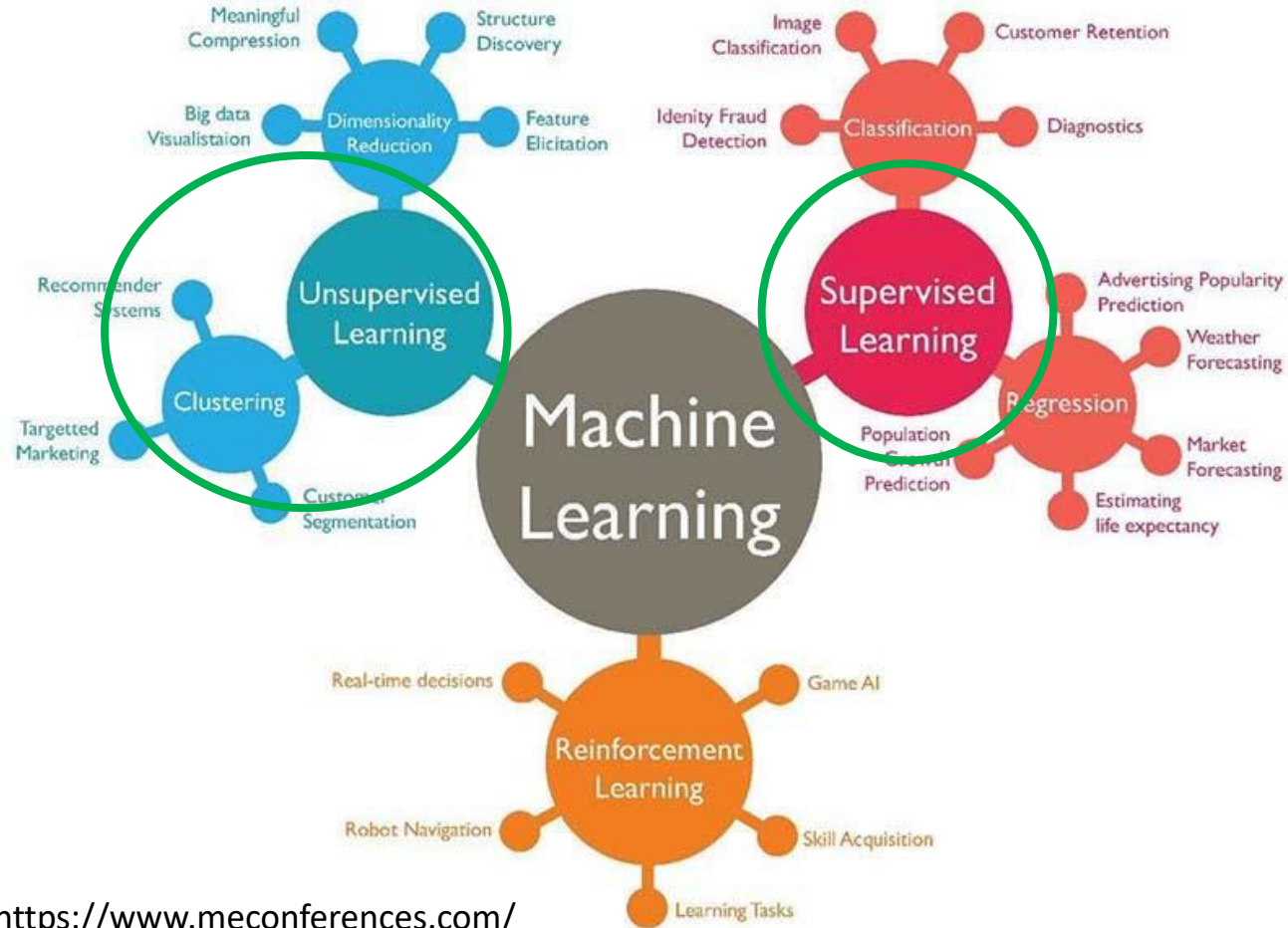




What is Machine Learning (ML) ?

Is a subfield of **art**

on and inference



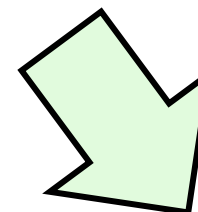
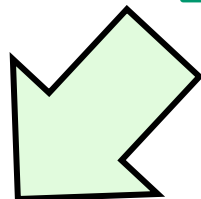
company of the
data preparation,
 dictive analytics

<https://www.meconferences.com/>



Why using ML techniques?

Having a **large Database**, in terms of **number of animals** and **availability of results**



Permits to validly use
ML techniques

→ high number of data means
higher statistical power

Permits to analyse
different variables and
have a complete picture
of the situation

**RESULTING IN A BETTER
OVERVIEW OF THE POSSIBLE
VARIATION IN THE POPULATION**



How did we use ML?

SUPERVISED ML – PREDICTION

- Permits to **identify patterns** in data
- Uses different models (*Native Bayes, Generalized Linear Model, Deep Learning...*)
- Importance of **precision, accuracy** and **sensitivity**

→ We aimed to find interactions between treatments (Control vs. Met) and disease by this method

→ → we identified **Endometritis x Met being most significant**

UNSUPERVISED ML - CLUSTERING

- Permits to **identify groups** of similar data
- **K-means** and **X-means** are the models that were used
- We used both of them → trying to identify the best possible grouping

→ Via this method we identified different grouping based on the **BFT-BHB** interaction & **Pattern of Body Condition Loss** during transition period



Results from ML

We were able to select cows based on 4 different biological aspect

Supervised Machine Learning

Unsupervised Machine Learning

**Endom or Healthy
vs
Methionine
or Control**

**Success or not in
the 1st AI
considering health
status**

**Clustering based
on BHB and BFT**

**Body Condition
Loss around
Parturition**

**GROUPINGS WERE MADE DEPENDING ON WHAT WE WERE INTERESTED IN!!
EACH GROUPING IS DIFFERENT FROM THE OTHER → THE POWER OF ML**



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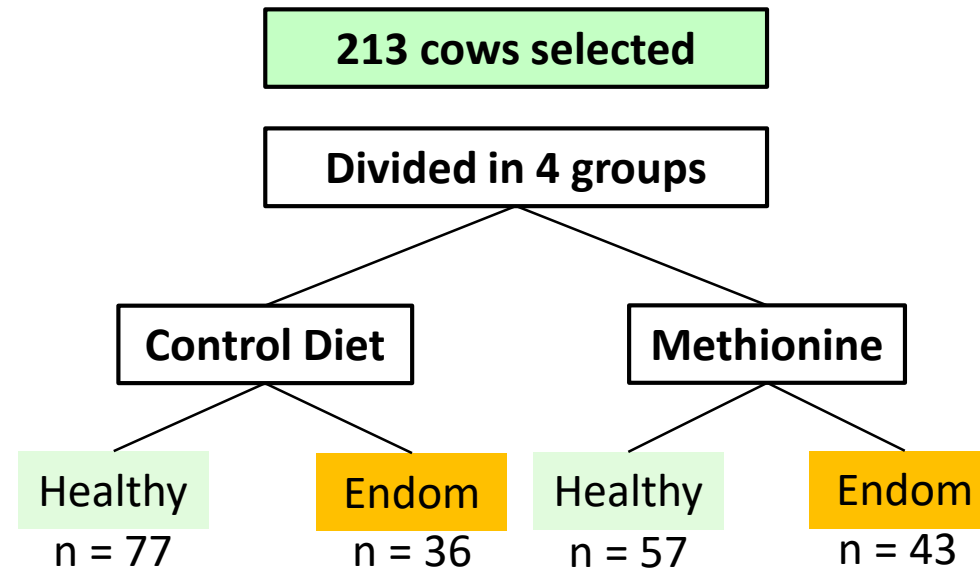
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Validation of Results





Diet Treatment vs Health Condition



Data analyzed using the *MIXED procedure of SAS* (SAS Institute Inc., Cary, NC)

Model included: random effect of cow and the fixed effect of the main factors M, CE, and their interaction (M × CE)

Days in milk was considered a covariate for the dependent variables



Diet Treatment vs Health Condition

Disease Effect

	P-value
Calving to conception interval (d)	0.0053
Milk at 30 DIM (Kg/d)	0.013
ECM at 30 DIM	0.079
Urea at 30 DIM (mg/100 mL)	0.003
Fat at 60 DIM (kg)	0.05
Protein at 30 DIM (kg)	0.04

Disease x Diet Effect

	P-value
BFT at 50 DIM (mm)	0.065
BCS at 50 DIM (score)	0.04
SCC at 70 DIM (10 ³ /ml)	0.097
ECM at 30 DIM	0.065
F:P Ratio at 30 DIM	0.008
Fat at 30 DIM (%)	0.015

Diet Effect

	P-value
BFT at 30 DIM (mm)	0.06
BCS at 30 DIM (score)	0.04
SCC at 70 DIM (10 ³ /ml)	0.039
Milk at 60 DIM (kg/d)	0.055
Protein at 30 DIM (kg)	0.094
Protein at 60 DIM (kg)	0.025



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Conclusion & Future Objectives





Conclusion & Future Objectives

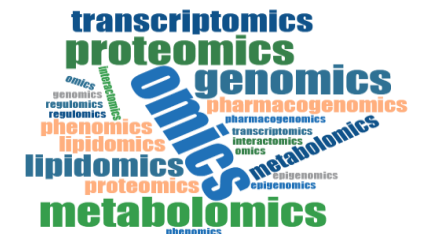
Having large datasets is desirable, but a considerable amount of time is necessary to prepare the data for Machine Learning

Machine Learning is an important tool that enables to find patterns and thus select meaningful samples

Results from the *Treatment vs Disease* study showed that the grouping was effective as statistical differences and interactions were found between the groups

Different phenotypes were detected
→ future **OMICs analyses** will permit us to have further insights and obtain a complete picture of the mechanisms involved

Machine Learning will be used again to further analyse the OMICs results with the **aim to identify markers** which can then be **targeted on all cows from the starting database**





“Data is the new science. Big Data holds the answers. Are you asking the right questions?”

Patrick P. Gelsinger





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THANK YOU FOR YOUR ATTENTION





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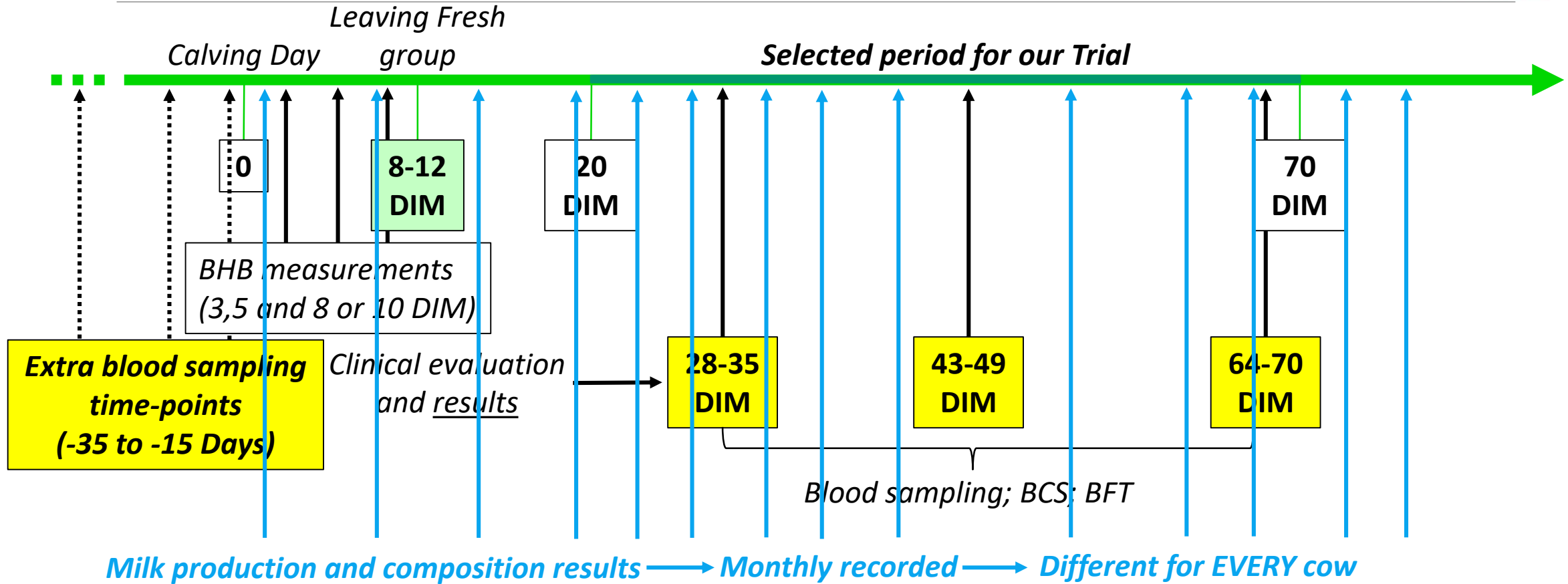
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Supplementary Material



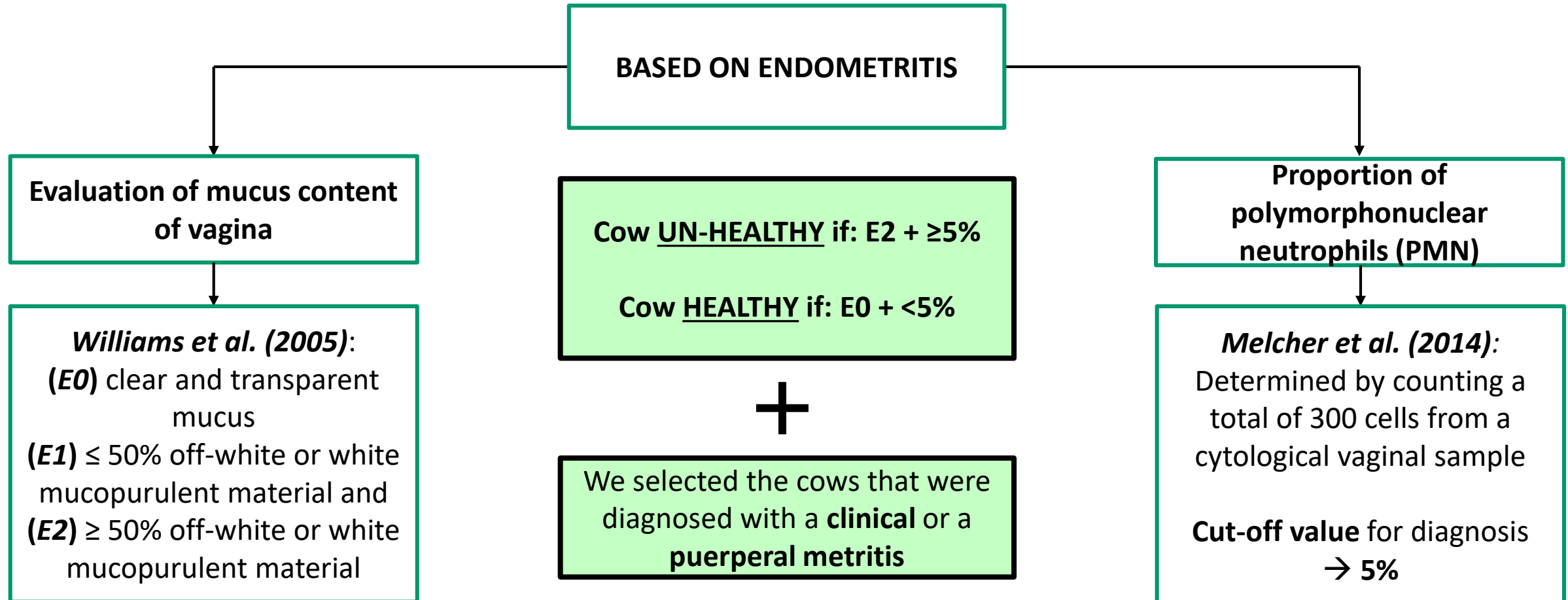


Overview - Animal Trial





Endometritis vs Healthy Selection





Statistical Analysis

	Control		Methionine		SEM	P-Value		
	CE	HEL	CE	HEL		TREAT	DISEASE	T x D
BFT30DIM (mm)	10.17	11.04	10.14	9.24	0.238	0.06	0.79	0.11
BFT50DIM	8.81	10.25	9.23	8.69	0.249	0.16	0.40	0.065
BFT70DIM	8.61	9.91	9.07	8.82	0.265	0.86	0.23	0.14
BCS30DIM (score)	2.69	2.75	2.64	2.58	0.027	0.04	0.91	0.26
BCS50DIM	2.56	2.74	2.63	2.56	0.028	0.29	0.33	0.04
BCS70DIM	2.58	2.70	2.56	2.57	0.031	0.33	0.31	0.43

	Control		Methionine		SEM	P-Value		
	CE	HEL	CE	HEL		TREAT	DISEASE	T x D
Calving to Conception Interval (d)	114.43	95.47	121.67	101.56	3.299	0.47	0.0053	0.92

EA Euro	Control		Methionine		SEM	P-Value					
	CE	HEL	CE	HEL		TREAT	DISEASE	T x D	PARITY	DIM	
1st Time Point	Milk (kg/d)	35.32	38.07	34.70	38.49	0.594	0.50	0.004	0.52	0.10	<.0001
	ECM	33.55	36.55	32.94	36.94	0.585	0.62	0.004	0.67	0.006	<.0001
	SCC (10 ³ /ml)	157.59	115.72	148.52	128.93	15.307	0.79	0.27	0.39	0.76	<.0001
	Prot (%)	3.17	3.26	3.24	3.23	0.026	0.72	0.26	0.25	0.25	<.0001
	Fat (%)	3.73	3.75	3.58	3.81	0.065	0.95	0.42	0.43	0.002	<.0001
	Urea mg/100ml	21.78	24.12	21.13	25.24	0.549	0.63	0.003	0.51	0.55	0.099
	2nd Time Point	Milk (kg/d)	42.13	42.45	43.14	43.90	0.383	0.12	0.50	0.78	0.009
ECM		37.03	38.72	38.10	39.77	0.449	0.24	0.07	0.99	<.0001	0.14
SCC (10 ³ /ml)		205.87	73.00	76.14	100.73	18.293	0.12	0.27	0.63	0.25	0.86
Prot (%)		3.01	3.07	3.04	3.03	0.017	0.91	0.57	0.24	0.53	0.20
Fat (%)		3.30	3.44	3.17	3.39	0.051	0.43	0.05	0.65	0.0003	0.014
Urea mg/100ml		24.05	26.03	25.22	25.55	0.564	0.70	0.41	0.76	0.057	0.62
3rd Time Point	Milk (kg/d)	40.59	40.70	41.66	40.28	0.404	0.69	0.45	0.37	0.017	0.054
	ECM	36.57	37.82	36.73	36.78	0.420	0.60	0.44	0.47	<0.0001	0.054
	SCC (10 ³ /ml)	246.62	108.72	85.42	135.13	18.815	0.039	0.71	0.097	0.40	0.75
	Prot (%)	3.09	3.14	3.13	3.12	0.018	0.84	0.49	0.40	0.90	0.75
	Fat (%)	3.35	3.56	3.18	3.42	0.057	0.17	0.54	0.87	0.0002	0.84
	Urea mg/100ml	24.79	25.43	25.84	25.83	0.438	0.44	0.74	0.73	0.33	0.76



Statistical Analysis

Too many variations in DIM for Milk production



Further selection based on DIM (Analyses with 54 cows)

	Control		Methionine		SEM	P-Value				
	CE	HEL	CE	HEL		TREAT	DISEASE	T x D	PARITY	
Around 30 DIM	Milk (kg/d)	40.13	44.51	41.99	46.27	0.695	0.30	0.013	0.94	0.78
	ECM	40.22	40.23	38.10	44.67	0.853	0.63	0.079	0.065	0.30
Around 60 DIM	Milk (kg/d)	40.05	43.19	42.92	47.24	0.810	0.055	0.046	0.73	0.35
	ECM	37.08	39.19	37.65	43.78	0.930	0.16	0.034	0.26	0.038

	Control		Methionine		SEM	P-Value			
	CE	HEL	CE	HEL		TREAT	DISEASE	T x D	PARITY
F:P 30DIM	1.36	1.10	1.08	1.23	0.037	0.32	0.49	0.008	0.075
F:P 60DIM	1.12	1.13	1.01	1.13	0.025	0.28	0.27	0.31	0.06



Statistical Analysis

	Control		Methionine		SEM	P-Value				
	CE	HEL	CE	HEL		TREAT	DISEASE	T x D	PARITY	
Around 30 DIM	Prot %	3.01	3.06	3.11	3.11	0.025	0.19	0.65	0.75	0.30
	Prot kg	1.23	1.36	1.31	1.44	0.021	0.094	0.005	0.93	0.66
Around 60 DIM	Prot %	3.12	3.01	3.10	3.14	0.034	0.42	0.78	0.30	0.046
	Prot kg	1.25	1.30	1.32	1.48	0.029	0.025	0.075	0.31	0.014

	Control		Methionine		SEM	P-Value				
	CE	HEL	CE	HEL		TREAT	DISEASE	T x D	PARITY	
Around 30 DIM	Fat %	4.09	3.36	3.36	3.80	0.112	0.53	0.60	0.015	0.051
	Fat kg	1.66	1.50	1.41	1.77	0.054	0.92	0.40	0.016	0.09
Around 60 DIM	Fat %	3.49	3.42	3.11	3.53	0.085	0.44	0.34	0.16	0.009
	Fat kg	1.40	1.48	1.34	1.66	0.048	0.58	0.04	0.24	0.003