

# Forage quality predicted by hyperspectral reflection measurements across climate zones

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# Grassland forage quality

**Metabolizable Energy [MJ kg<sup>-1</sup> DM<sup>-1</sup>]**  
(dep. on protein, fiber, fat, ash, ... content)

## Management

- Agriculture: milk and meat production
- AgroEcology: pasture carrying capacities

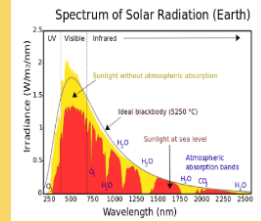
## Challenges

- Potential often unknown
- Expensive/labourous analysis

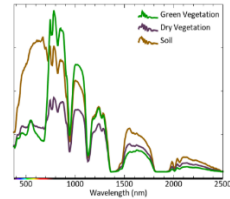
## REMOTE SENSING OF FORAGE QUALITY

1. Hyperspectral prediction of forage quality
2. Upscaling from ground to field & landscape level

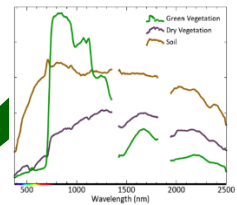
→ **Mapping of grasslands**



Radiance

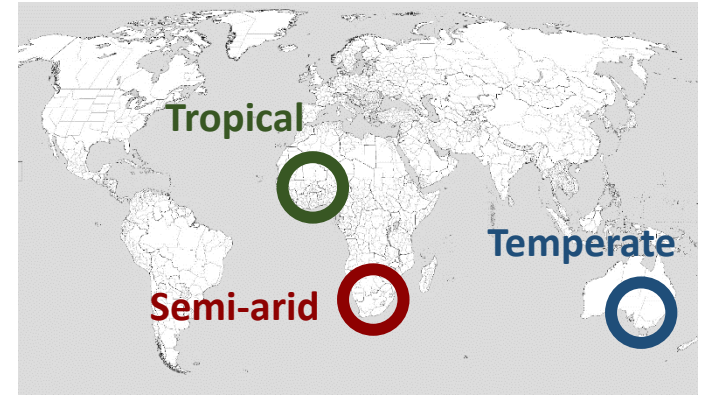


Surface Reflectance



# State of the art

- Ferner et al., 2015/2018:  
**Metabolizable energy** predictions in **tropical grasslands**
- Singh et al., 2017:  
**Nutritive chemical** prediction on **South-African savannahs**
- Smith et al., 2020:  
**Nutritive value** predictions on **temperate ryegrass** canopies



## Research Gap

Forage quality prediction modelling limited to regional scales



## Objectives

1. „Trans-climatic“ modelling of forage quality by hyperspectral measurements
2. Different predictor variables and model training/learning methods



# Study sites

Grasslands (meadows & rangelands) under different climatic conditions

- **Temperate pastures (n = 110)**

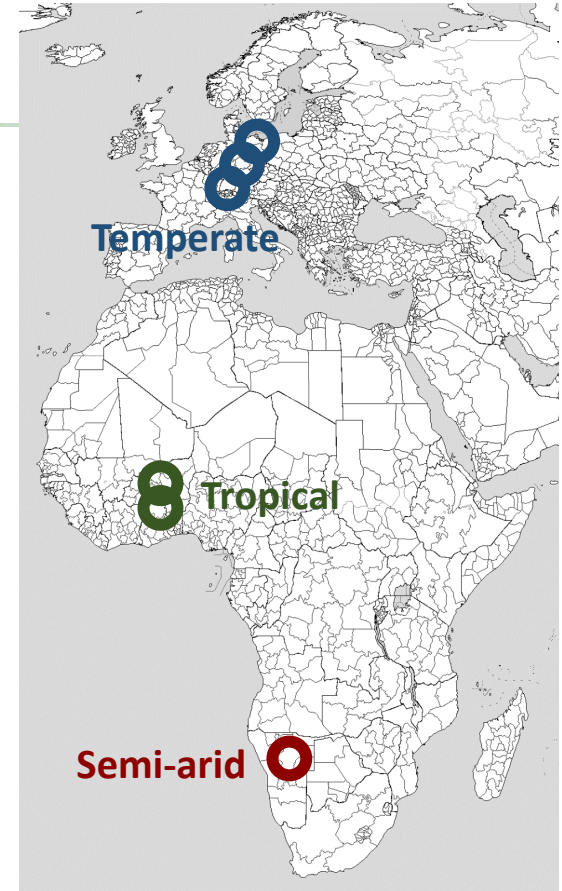
- Germany

- **Tropical grasslands (n = 101)**

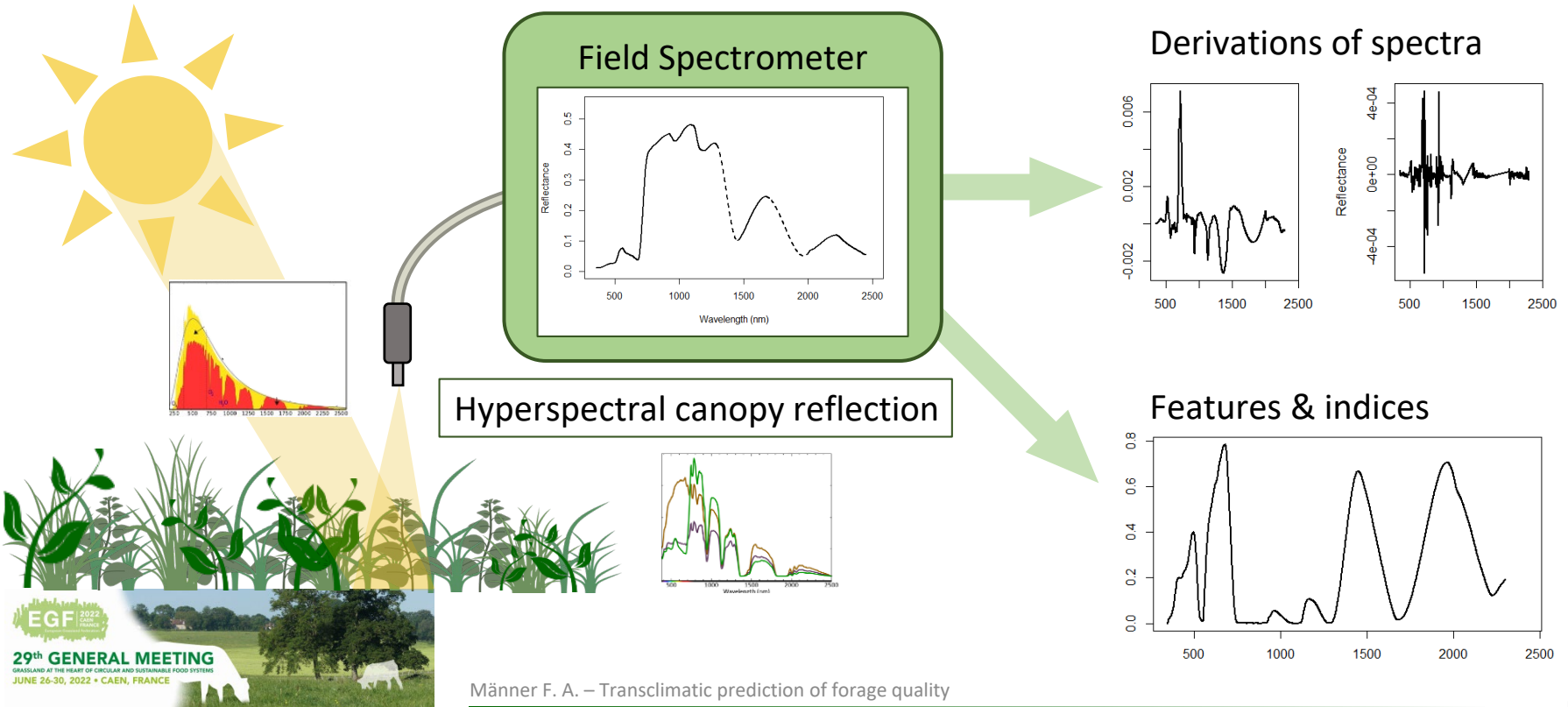
- Burkina-Faso (*Ferner et al. 2015*)
- Ghana (*Ferner et al. 2015*)

- **Semi-arid savannah (n = 127)**

- Namibia



# Hyperspectral measurements



# Forage quality measurements

1. Fermentation of biomass in rumen liquid (*Menke and Steingass, 1988*)  
→ Measurement of **gas production** ( $GP \left[ \frac{ml}{200mg TM} \right]$ )
2. Combustion of biomass in elemental analyzer  
→ Measurement of **crude protein** content ( $XP \left[ \frac{g}{kg TM} \right]$ )
3. Estimation of **metabolizable energy** ( $ME \left[ \frac{MJ}{kg TM} \right]$ )

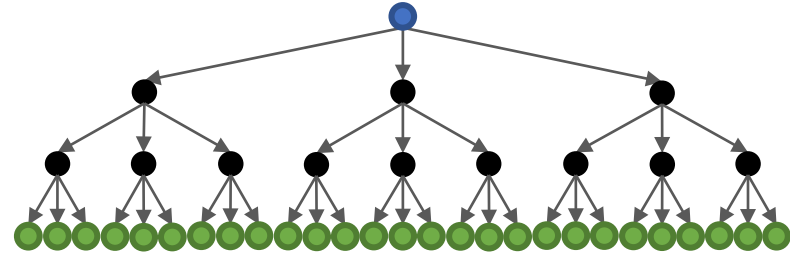
$$ME \approx 2.2 + 0.1357GP + 0.0057XP + 0.0002859XP^2$$



# Prediction Models

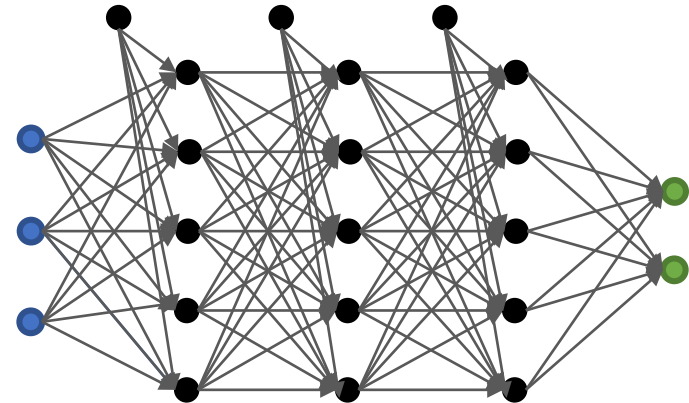
## Supervised machine learning

- Partial least squares regression (PLS)
- Random forest regression (RF)

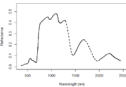
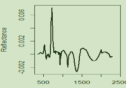
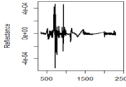
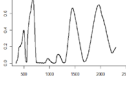


## Deep learning

- *Deep neural networks (DNN – in dev.)*



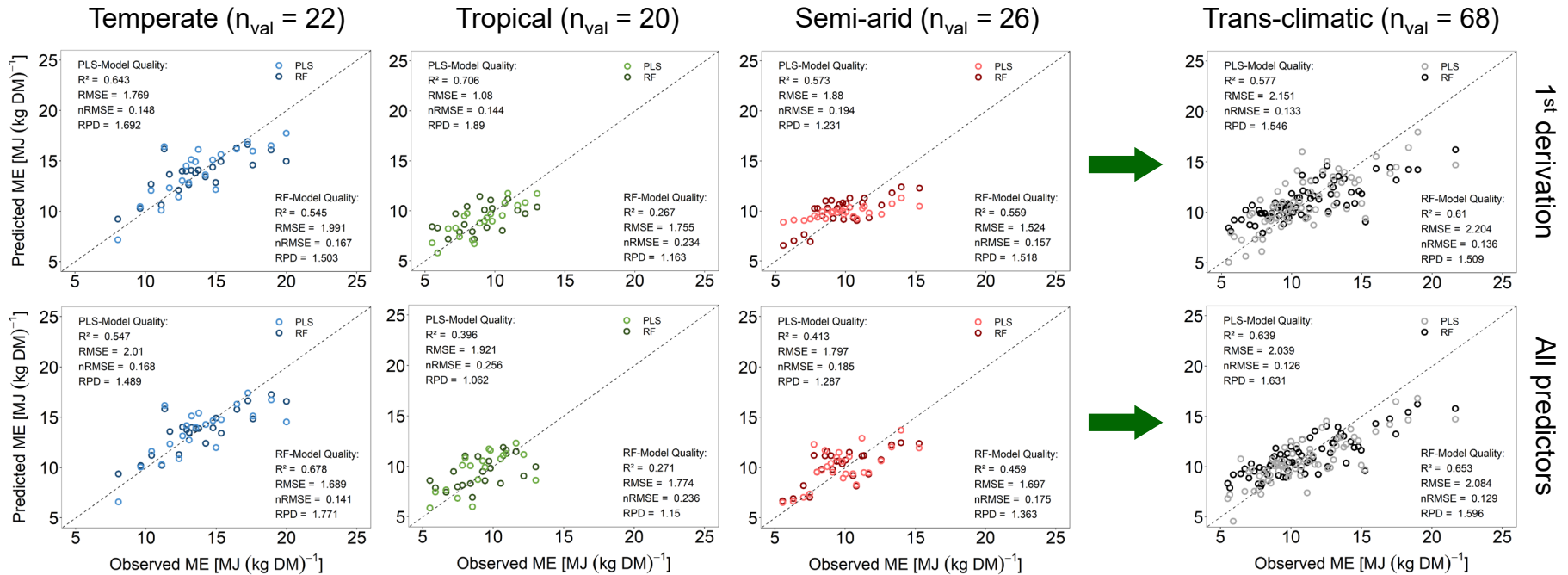
# Results & Discussion

		Random Forest Regression (RF)					Partial Least Squares Regression (PLS)			
		Temperate	Tropical	Semi-arid	Trans-Clim.	Temperate	Tropical	Semi-arid	Trans-Clim.	
 <b>Absolute spectra</b>	R <sup>2</sup>	0.310	0.290	0.275	0.438	0.712	0.500	0.430	0.700	
	RMSE [MJ (kg DM) <sup>-1</sup> ]	2.517	1.781	1.970	2.505	1.587	1.593	2.055	1.813	
	nRMSE	0.211	0.237	0.203	0.155	0.133	0.212	0.212	0.112	
	<b>RPD</b>	<b>1.189</b>	<b>1.146</b>	<b>1.174</b>	<b>1.327</b>	<b>1.209</b>	<b>1.886</b>	<b>1.281</b>	<b>1.126</b>	<b>1.834</b>
 <b>1st derivation</b>	R <sup>2</sup>	0.545	0.267	0.559	0.610	0.643	0.706	0.573	0.577	
	RMSE [MJ (kg DM) <sup>-1</sup> ]	1.991	1.755	1.524	2.204	1.769	1.080	1.880	2.151	
	nRMSE	0.167	0.234	0.157	0.136	0.148	0.144	0.194	0.133	
	<b>RPD</b>	<b>1.503</b>	<b>1.163</b>	<b>1.518</b>	<b>1.509</b>	<b>1.423</b>	<b>1.692</b>	<b>1.890</b>	<b>1.231</b>	<b>1.546</b>
 <b>2nd derivation</b>	R <sup>2</sup>	0.278	0.288	0.505	0.448	0.214	0.350	0.259	0.427	
	RMSE [MJ (kg DM) <sup>-1</sup> ]	2.502	1.711	1.701	2.488	2.760	1.756	2.044	2.510	
	nRMSE	0.209	0.228	0.175	0.154	0.231	0.234	0.211	0.155	
	<b>RPD</b>	<b>1.196</b>	<b>1.193</b>	<b>1.360</b>	<b>1.337</b>	<b>1.272</b>	<b>1.084</b>	<b>1.162</b>	<b>1.132</b>	<b>1.325</b>
 <b>Features/Indices</b>	R <sup>2</sup>	0.582	0.366	0.431	0.607	0.497	0.180	0.248	0.376	
	RMSE [MJ (kg DM) <sup>-1</sup> ]	1.903	1.586	1.744	2.105	2.091	5.799	2.083	2.726	
	nRMSE	0.159	0.211	0.180	0.130	0.175	0.772	0.215	0.169	
	<b>RPD</b>	<b>1.573</b>	<b>1.287</b>	<b>1.326</b>	<b>1.579</b>	<b>1.441</b>	<b>1.431</b>	<b>0.352</b>	<b>1.111</b>	<b>1.220</b>
<b>All predictors</b>	R <sup>2</sup>	0.678	0.271	0.459	0.653	0.547	0.396	0.413	0.639	
	RMSE [MJ (kg DM) <sup>-1</sup> ]	1.689	1.774	1.697	2.084	2.010	1.921	1.797	2.039	
	nRMSE	0.141	0.236	0.175	0.129	0.168	0.256	0.185	0.126	
	<b>RPD</b>	<b>1.771</b>	<b>1.150</b>	<b>1.363</b>	<b>1.596</b>	<b>1.470</b>	<b>1.489</b>	<b>1.062</b>	<b>1.287</b>	<b>1.631</b>





# Results & Discussion



# Outlook

## Current research

- Spatial/trans-climatic **cross validation**
- **Biomass prediction** with hyperspectral readings

- Upscaling to UAV/satellite imagery for **landscape forage mapping**



# References

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# Thank you for your attention!



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