# Monitoring Global Soybean Production Using Elementomics to Combat Rainforest Destruction

**Brian Quinn** 





# Soybean Farming





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# Soybean Production



In 2018 China slapped a 25% tariff on US soybeans in response to US taxes being added to all imported Chinese goods

➤ In the aftermath, China now imports 70% of Brazil's soybean production

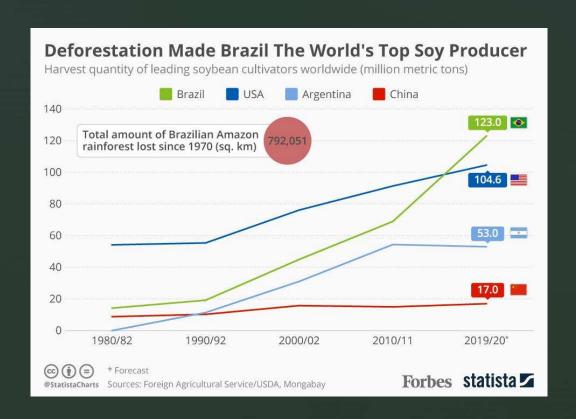
#### **Amazon Deforestation**



Brazil's response to the massive demand for soybeans has been to accelerate the systematic clearance of the Amazon Rainforest to create agricultural land

Over 11 million acres of the Amazon Rainforest was destroyed from 2019-2022

# Global Soybean Production



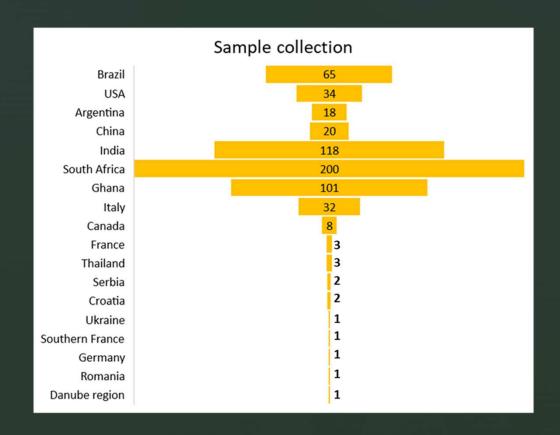
#### **Amazon Deforestation**



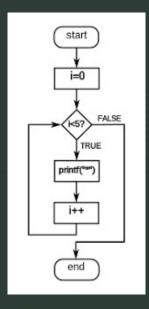
➤ In the EU and UK, consumers are applying significant pressure on grocery chains, livestock producers and feed companies to make sure the soya they sell is not from a recently deforested area

No robust analytical tool exists to help protect businesses from selling soybeans grown in impacted areas

## Global Sample Collection



## General Analytical Workflow



Sourced samples (10-15%) were analysed by Agilent 6546 LC/QToF, Agilent 5975C GC/MS/MS, and an Agilent 7850 ICP/MS to determine the best platform to take forward

> The ICP/MS proved to produce the best elementomic models, so it was the instrument of choice to continue this research

# ICP/MS Analytical Workflow



- Sample size is 100 mg with 10 replicates per sample
- Samples are allowed to digest in 2 mL of hydrogen peroxide and 2 mL of concentrated nitric acid before microwaving for 1.5 hours and diluting to 20 mL before analysis
- A soybean CRM was purchased from Merck and used in every analytical run to evaluate the instrument performance over time
- Samples are randomised on the autosampler prior to analysis

# Modelling Analytical Workflow



- Results from the ICP/MS are quantitated (45 elements)
- Quantitated results are log 10 transformed

➤ Models are created using SIMCA – unsupervised PCA first and supervised models (OPLS-DA, PLS-DA, RF, SVM, etc.) next

#### **Model Validations**

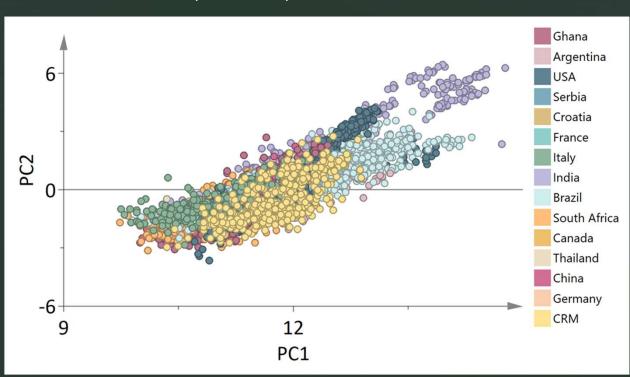
Permutation tests (n=50) were conducted to ensure the model was not overfit

External validation – 70% of samples were used as the training set and 30% were used as a test set to again test for overfitting

Internal validation – Leave 20% out testing was used

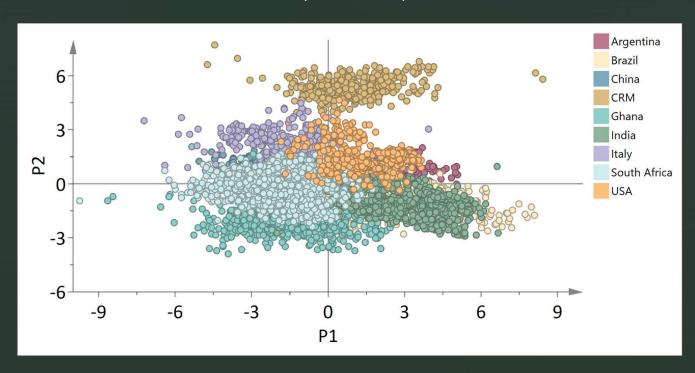
# PCA Model

R<sup>2</sup>X=0.971; Q<sup>2</sup>=0.919; LOG10 normalization



# OPLS-DA Model

• R<sup>2</sup>X=0.995; R<sup>2</sup>Y=0.693; Q<sup>2</sup>=0.69

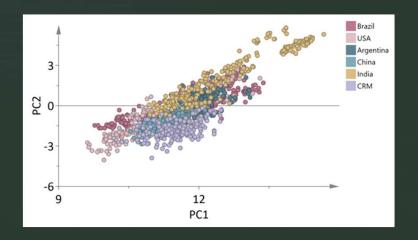


# OPLS-DA Model

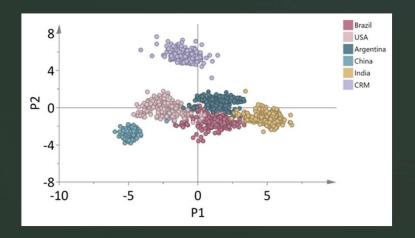
	Members	Correct	Argentina	Brazil	China	CRM	Ghana	India	Italy	South Africa	USA
Argentina	160	50.63%	81	76	0	0	0	0	0	0	3
Brazil	630	99.84%	0	629	0	0	1	0	0	0	0
China	200	100%	0	0	200	0	0	0	0	0	0
CRM	410	100%	0	0	0	410	0	0	0	0	0
Ghana	940	98.09%	0	11	0	1	922	0	0	6	0
India	760	98.56%	0	10	0	0	1	749	0	0	0
Italy	270	95.56%	0	0	0	0	0	3	258	6	3
South Africa	1760	99.94%	0	0	0	0	0	0	1	1759	0
USA	430	96.51%	0	0	0	0	0	1	4	10	415
Total	5560	97.5%	81	726	200	411	924	753	263	1781	421

## Elementomic Models – Top 5 Producers

**PCA:**  $R^2X = 0.976$ ;  $Q^2 = 0.916$ 



**OPLS-DA:** R<sup>2</sup>X=0.927; R<sup>2</sup>Y=0.869; Q<sup>2</sup>=0.836



# Elementomic Models – Top 5 Producers

#### Classification

		Brazil	USA	Argentina	China	India	CRM
	Brazil	249 99.6%	0	1 0.40%	0	0	0
Real Groups	USA	0	250 100%	0	0	0	0
	Argentina	11 6.87%	0	149 93.13%	0	0	0
	China	0	0	0	200 100%	0	0
	India 0		0	0	0	0 250 100%	
	CRM	0	0	0	0	0	200 100%

Group	# of samples	# of passes	# of failures	# of outliers	Classification rate
Total	1310	1298	12	0	99.1%

### Future Work

- Obtain samples from impacted areas in Brazil
- Keep collecting samples to improve models
- > Explore bean versus ground sample disparities
- > ISO accreditation

## Acknowledgements



Dr Nick Birse

Prof Chris Elliott







#### Soya Project Summary – Coalition of the Good





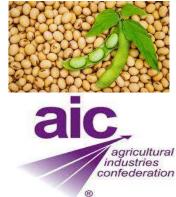


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