

Large-Scale Phenotyping Inferences: From Trees to Forest through Machine Learning

Maxime Bombrun, PhD – Data Scientist, Data Analytics Team Jonathan Dash, Heidi Dungey, David Pont, Michael Watt February 2019



Context

New Zealand's Top 3 Exports:

- Dairy, eggs, honey: \$15 billion (27.6%)
- Meat: \$7 billion (12.7%)
- Wood: \$4.9 billion (9%)

Radiata Pine accounts for 89% of the total planted resource in New Zealand.

The Grow Confidence in Future Forestry programme goals are:

- increase returns from existing forests through mid-rotation interventions
- improve returns from existing forests through better knowledge of wood quality



SCION

Objectives

The genomics revolution

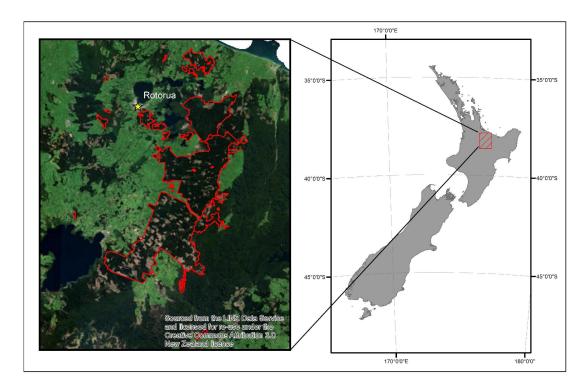
- Shorten the breeding cycle,
- Increase crop productivity.

However,

- Accurate phenotyping is challenging,
- Phenotyping at a forest scale.

Currently,

- Large dataset for the Kaingaroa forest,
- Stand productivity quantified using the Site Index,
- Machine learning approach to model productivity based on these variables.

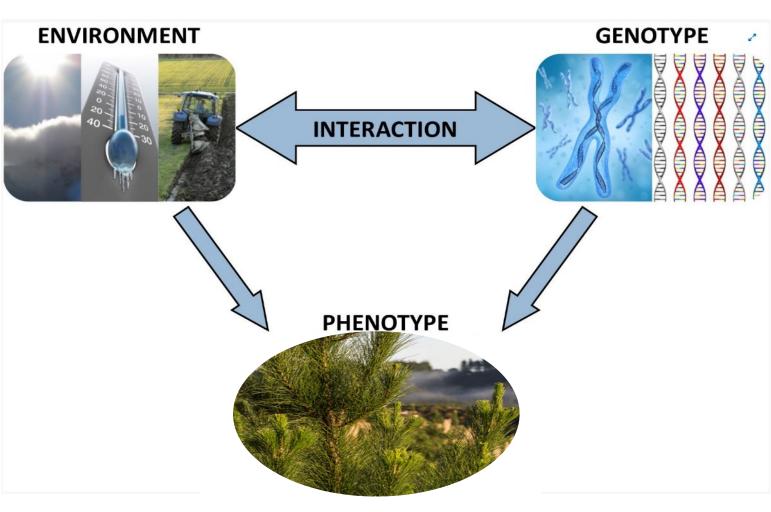


SCIOR

In-forest phenotyping – a new concept

 Enables quantification of plant traits and environmental conditions to identify and assess the performance on germplasm of interest

 Allows better matching of seedlot to site, based on past and current performance



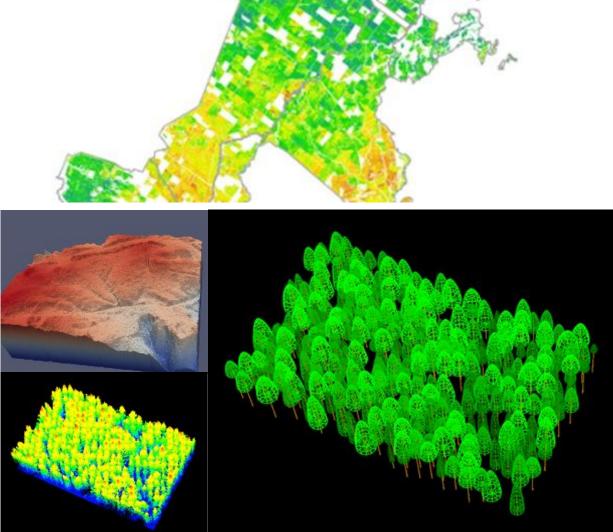


Phenotyping methodology

Many sensors and platforms have been developed for plant phenotyping

Two levels of phenotyping

- tree-based, applied to stands and eventually the forest estate
- area-based (25 x 25 m grid), applied to the forest estate





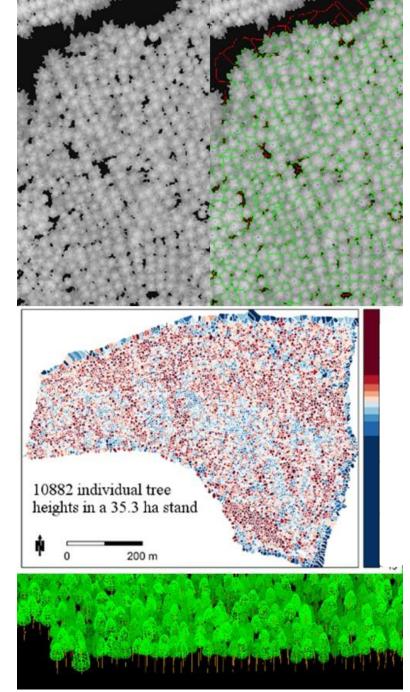
Application of tree-based phenotyping

• Airborne Laser Scanning used to carry out tree detection and crown delineation

• We have identified crown metrics that accurately estimate tree height and volume

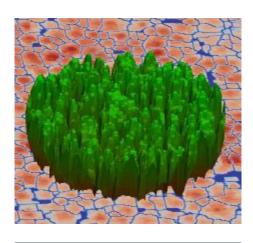
• Currently detecting and characterising every tree in operational stands (30-40 ha, 10,000+ trees)

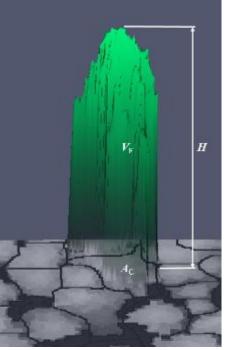
 Developing methods to select individual trees exhibiting exceptional height or volume growth

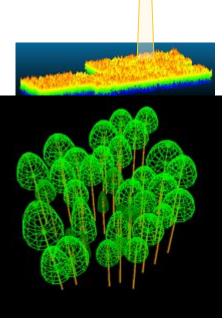


Benefits of tree-based phenotyping

- Quantifying GxExS interaction to determine tree growth
- Identification of superior tree performance
- Accelerated tree breeding
- Optimising the matching of breeds to sites
- Precision stand mapping and management
- Supports higher productivity higher returns from stands









Application of area-based phenotyping

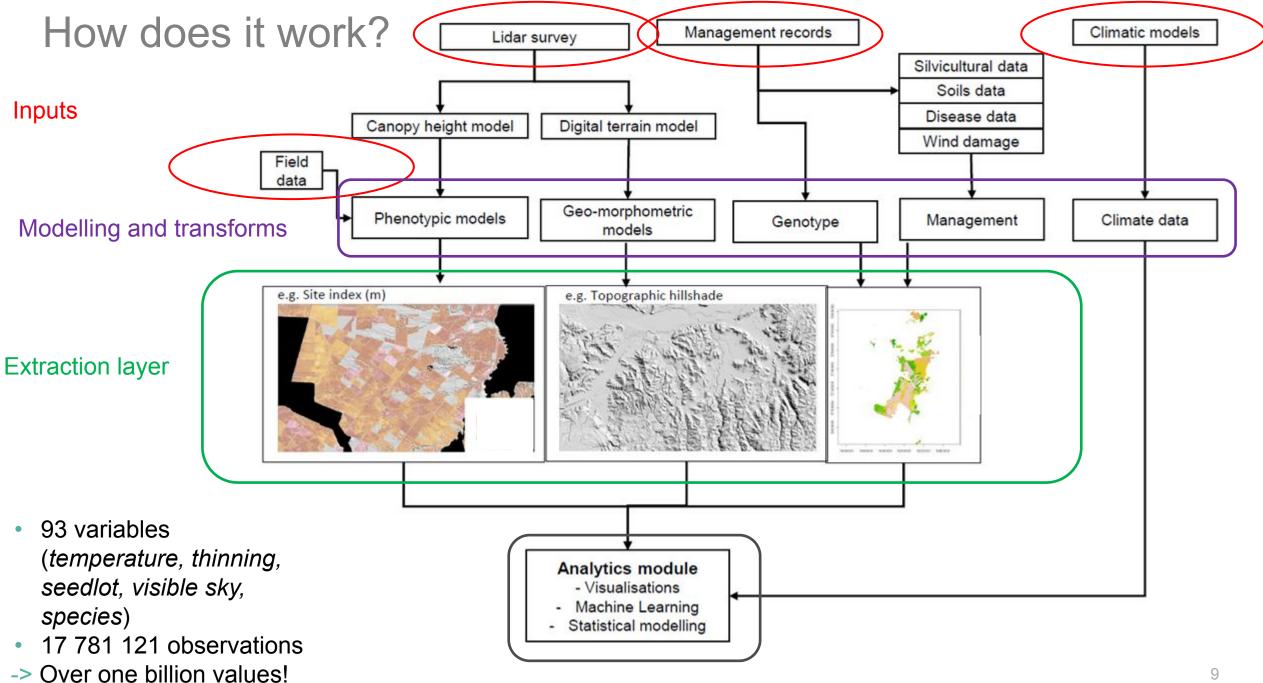
 Remote sensing to model forest phenotype across landscapes on an area basis in Kaingaroa

 Seedlot information provides us with a useful starting point

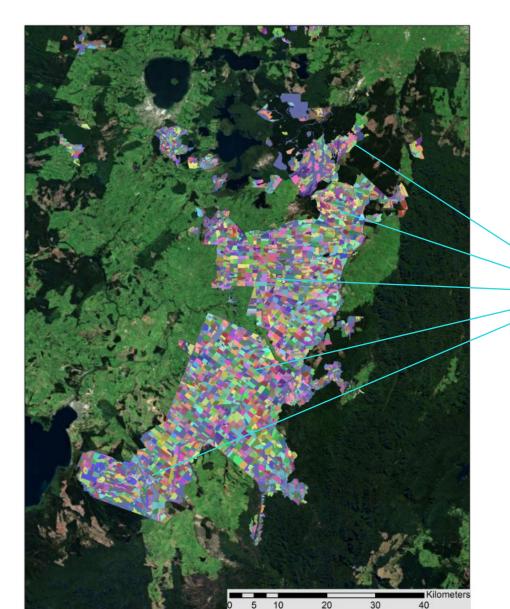
 Climatic data is integrated to identify key sites properties across the landscape

• The forest phenotyping platform allows us to assess genetic effects independently of site



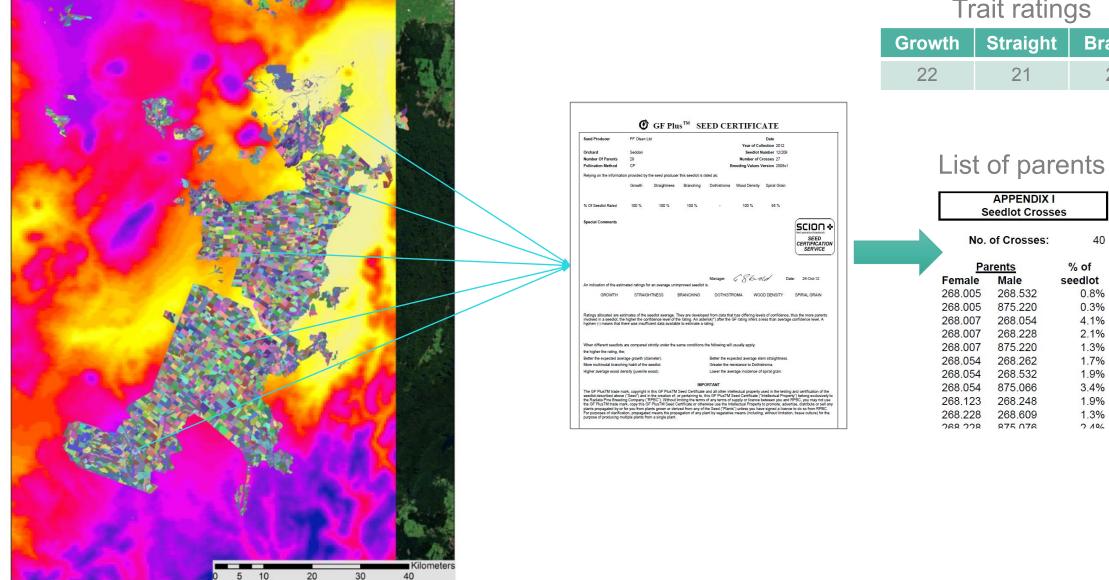


Seedlot distribution



	Trait ratings		
	Growth	Straight	Branch
	22	21	23
Content of Determine Sector Se	Lis	t of par	ents
Growth Straightness Branching Dothistroma Wood Density Sprai Grain % Of Seedlor Rated 100 % 100 % 100 % - 100 % 05 % Special Comments Sp	APPENDIX I Seedlot Crosses No. of Crosses: 40		
Manager: CEC-0CF Date: 24-06-12 An indication of the estimated ratings for an average unimproved seedlot is: GROWTH STRAGHTNESS BRANCHING DOTHISTROMA WOOD DENSITY SPIRAL GRAIN	Femal 268.00 268.00	5 268.532	% of seedlot 0.8% 0.3%
Ratings allocated are estimates of the seedort average. They are developed from data that hos offering levels of confidence, thus the more parents involved in a seedid, the higher the confidence level of the rating. An asterisk ¹¹ plate the GF rating inverse a less than average confidence level. A hyphier in mean that there was insufficient data available to estimate a rating. When different seedings are compared strictly under the same conditions the following will usually apply;	268.00 268.00 268.00 268.00	07 268.054 07 268.228	0.3% 4.1% 2.1% 1.3%
the higher the rating, the; Better the expected average soft (standard). Better the expected average soft endstands. Better the expected average	268.05 268.05 268.05 268.05 268.12	4 268.262 4 268.532 4 875.066	1.7% 1.9% 3.4% 1.9%
the GP Plant M take mark, copy this GP Plant M Seed Coefficient or otherwise use the Intellicitual Property to provide, advectes, calabitude or self any plants prograded by or froy using plant grown or derived from any of the Seed (Plant) unless puls taxe splants alicence to dis so the PSEC. Por purpose of admictance, propagated marks the propagation of any plant by wapitative marks (including, without limitation, taske outure) for the purpose of problem; multiple plants from a single plant.	268.22	8 268.609	1.3%

Climate overlay



Trait ratings **Branch** 23

40

XGBoost model

eXtreme Gradient Boosting (XGBoost):

- Machine learning combining
 - Boosting algorithm
 - Base learner tree method
- High performing solutions in large and complex competitions (e.g., Kaggle)
- + Fast (GBM: 24hr, Xgboost: ~6hr) and robust, automatic feature selection, even correlated,

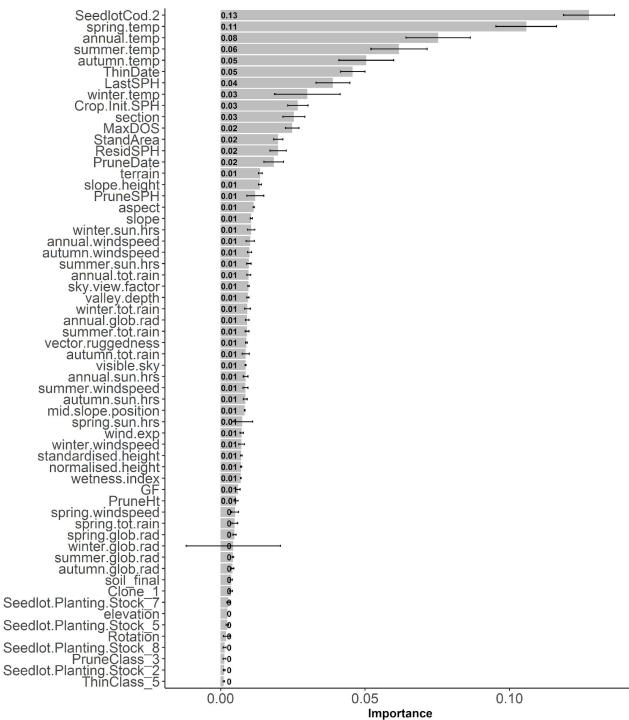
+ Understandable, feature of importance, thresholds of decision, constructed tree(s)



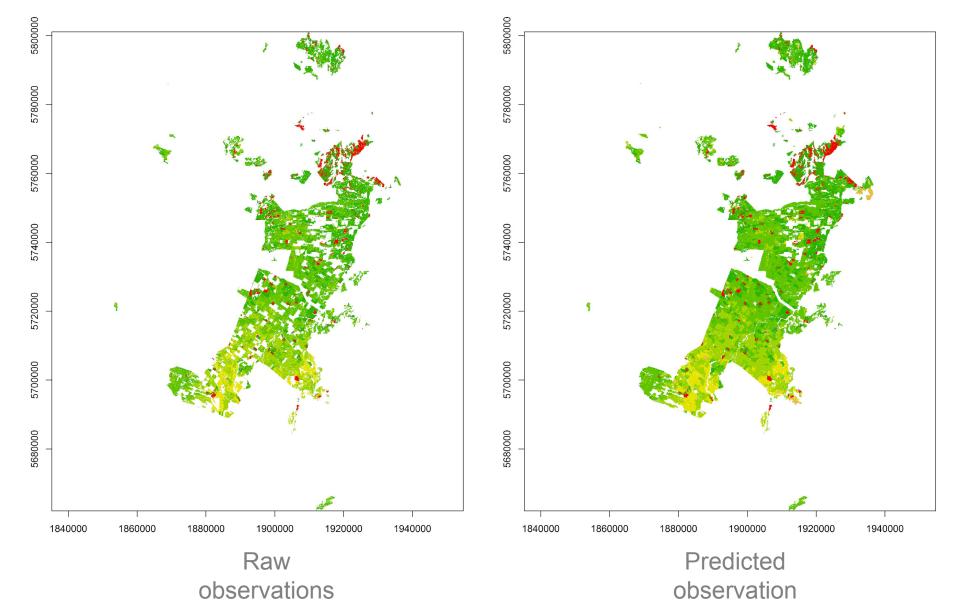
Early results

Initial models account for ~91 % of the variance in forest productivity across the forest

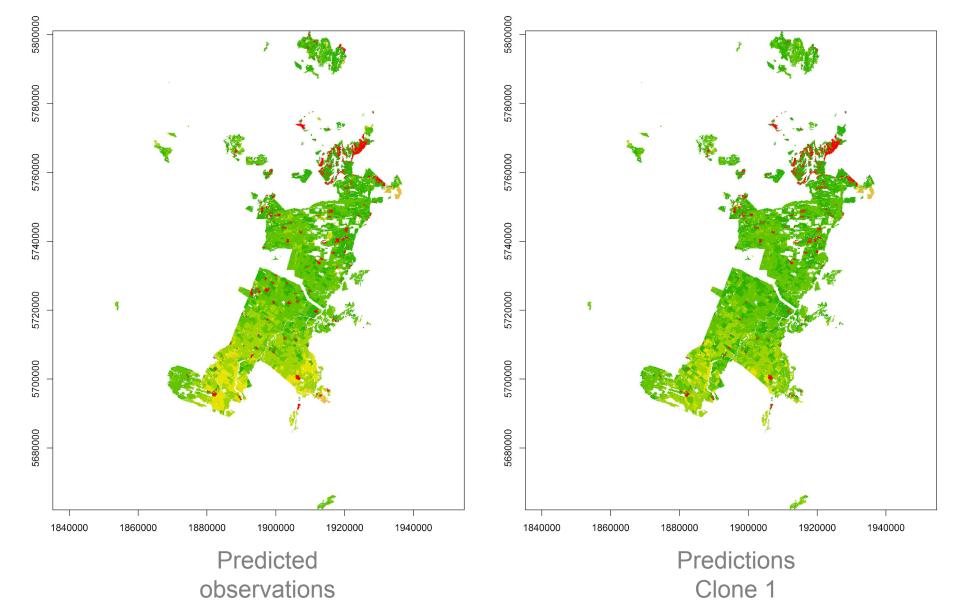
- Seedlot and site attributes are consistently the most important factors, highlighting the importance of matching seedlot to site
- We can use the XGBoost outputs to examine the interaction between genetic and site factors
- Future analysis will indicate where best to locate specific seedlots



Predictions of Site Index values



Predictions of Site Index values for Clone 1



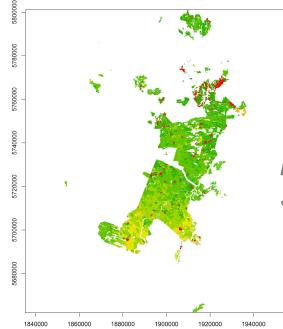
Productivity from genetics

Investigation on seedlot with the highest productivity:

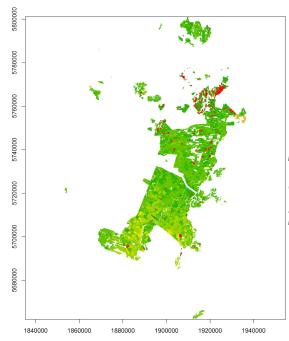
- Selective tree stocks based on environmental conditions,
- Predictions of productivity based on the XGBoost model, thus, enhance the genetic potential,
- Estimations in missing area from known parameters,
- Superior genotype across the forest.

Indicators of interplay between site and genetic factors, possibilities of higher productivity

ID	Percentage _Tot	Percentage_ noZero	Improvement
Pos_Clone1	60.77	61.45	2 813 899
Pos_Optimised	16.75	57.8	184 712







Predicted Site Index for clone1

Conclusions

- The Kaingaroa dataset
 - Vast, informative and revealing of the productivity at a forest-scale level.
 - Phenotyping at broad spatial scale was achieved through combining the analytical power of advanced machine learning methods with spatial layers acquired from remotely sensed data, management records and climatic surfaces.
- This methodology can be used to map variation in productivity between seedlots at fine spatial scale under varying environments to identify superior genotypes across the forest and thus, the continual optimisation of deployed genetically improved tree stock.



Maxime Bombrun Data Scientist Maxime.Bombrun@scionresearch.com

www.scionresearch.com

www.gcff.nz www.fgr.nz Date: 18/02/2019





www.scionresearch.com



Prosperity from trees *Mai i te ngahere oranga*

Scion is the trading name of the New Zealand Forest Research Institute Limited