



Impact of high throughput phenotyping on estimation of genetic parameters

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Impact of high throughput phenotyping on estimation of genetic parameters

Katri Kärkkäinen, Luke
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Impact of high throughput phenotyping on estimation of genetic parameters

- High throughput phenotyping?
- Why?
- What?
- How?
- When?

High throughput phenotyping?

Forest Research:
phenology,
Sex reproduction,
Frost and drought resistance
Pest resistance
Annual radial growth
Root characteristics
Stem architecture
Wood properties
(some examples from bolded ones)





WHY?

Aim:

- Tree breeding
- Assessing optimal deployment of forest regeneration material
- Assessing evolution of tree populations

- Estimation of genetic parameters
- Estimation of genetic basis of phenotypes

Estimation of genetic parameters

Phenotypic variance,

$$V_P = V_A + V_D + V_I + V_E$$

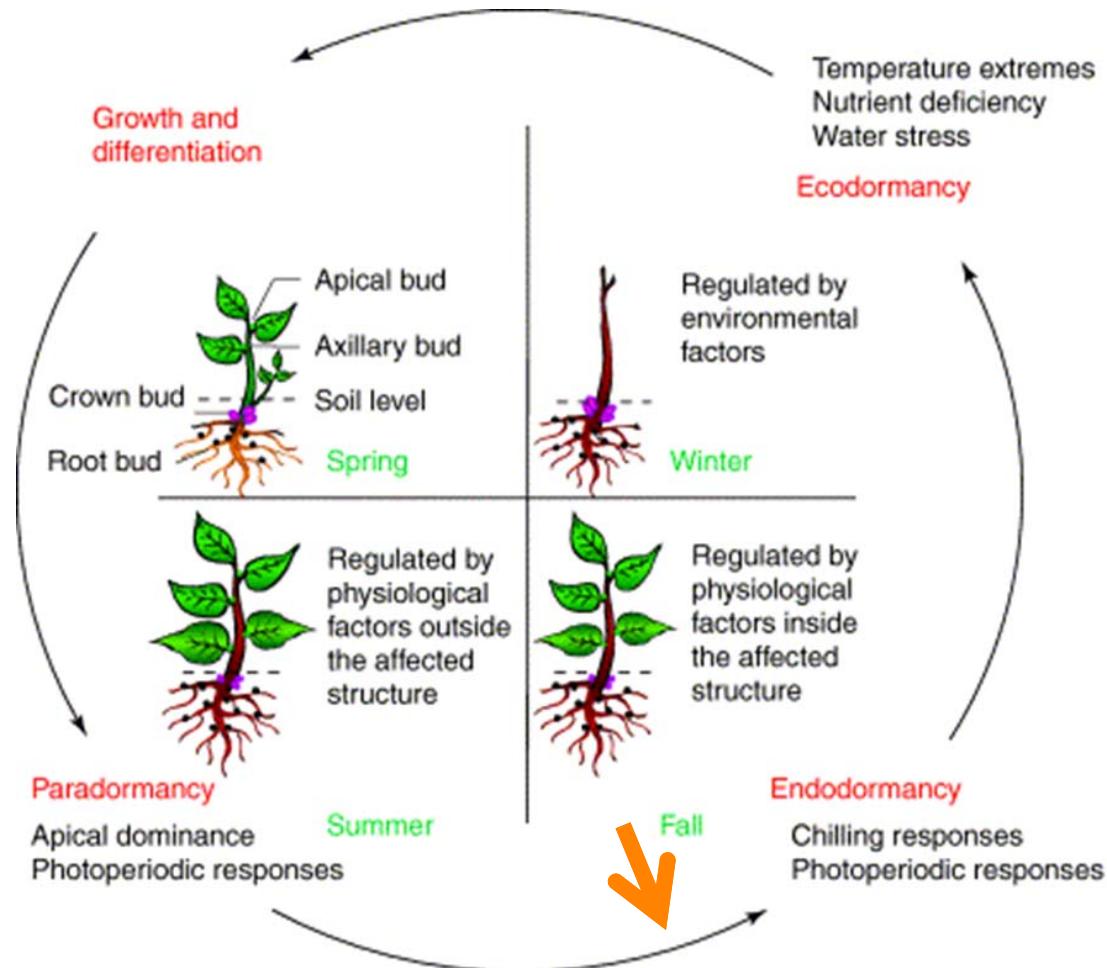
Additive value

Environmental deviation

"true" environmental variation
+ measurement errors

$$h^2 = V_A / V_P$$

WHAT? Aim: How to identify the most suitable trait? Case 1: Phenology – selection on growth rhythm



TRENDS in Plant Science

Aim: How to identify the most suitable trait?

Case 1: Phenology – selection on growth rhythm:
Cessation of growth (95% of final height) vs. bud-set,
greenhouse study



Cessation of growth (when 95% of final height reached) and timing of budset

Variance components and genetic parameters for cessation of growth

Population	Seedlings	Mean	Variance components			VA	Genetic parameters				
			$\widehat{\sigma_f}^2$	$\widehat{\sigma_i}^2$	$\widehat{\sigma_e}^2$		CV _A	V _P	CV _P	h^2	SD h^2
Punkaharju	11919/12000	96.04	2.71	0.00	66.27	10.83	0.03	68.98	0.09	0.16	0.02

Variance components and genetic parameters for budset

Population	Seedlings	Mean	Variance components			VA	Genetic parameters				
			$\widehat{\sigma_f}^2$	$\widehat{\sigma_i}^2$	$\widehat{\sigma_e}^2$		CV _A	V _P	CV _P	h^2	SD h^2
Punkaharju	11884/12000	102.99	10.00	0.71	73.33	39.99	0.06	84.03	0.09	0.48	0.04

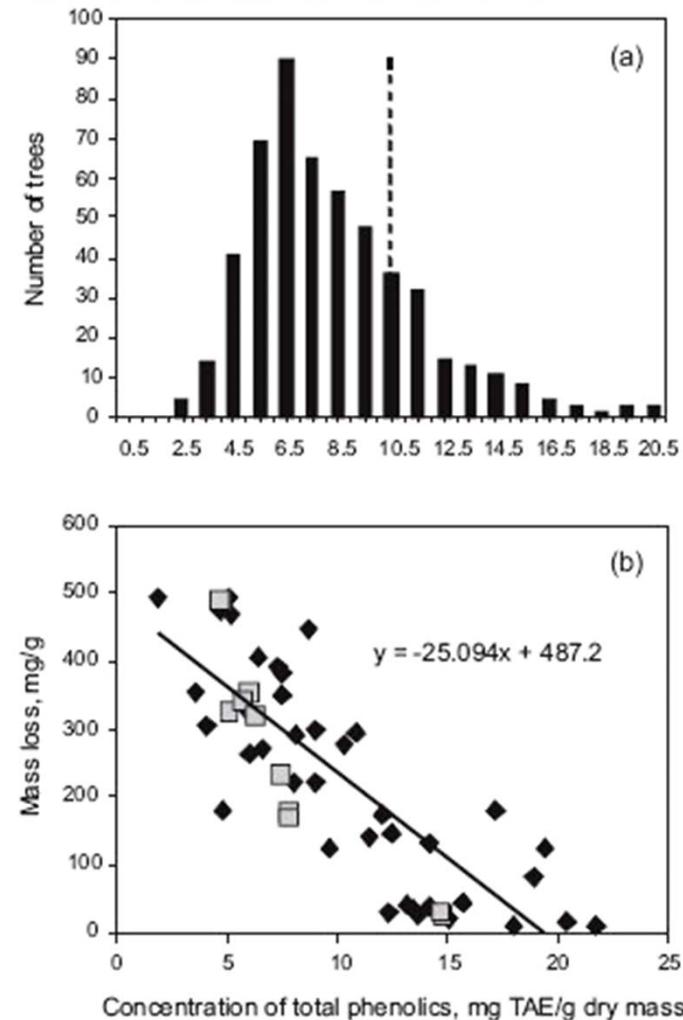
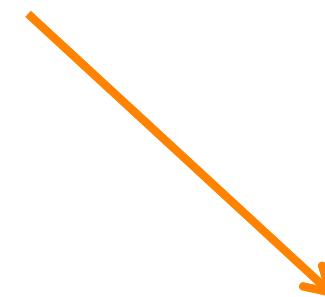
WHAT? Aim: How to identify the most suitable trait?

Case 2: Decay resistance of timber

Decay tests: correlation between decay resistance and concentration of phenolics

Conclusions:

Decay tests necessary,
but phenotyping may be
more efficient
using chemical traits





How?

Aim: High-throughput phenotyping

- Non-destructive
- Low cost, fast
- Optical analyses

How? Aim: High-throughput phenotyping

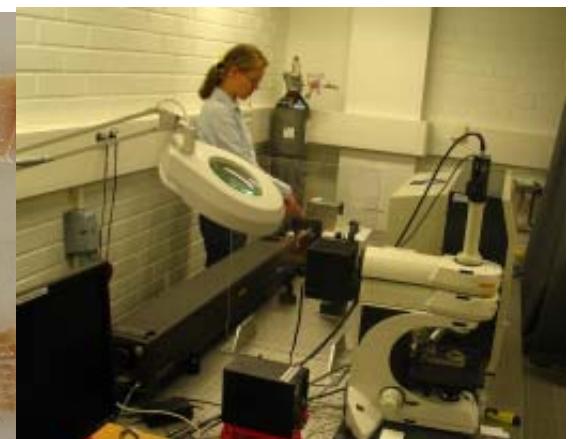
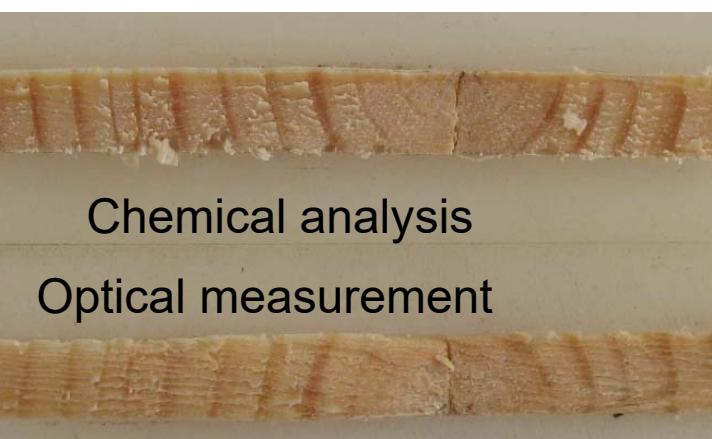
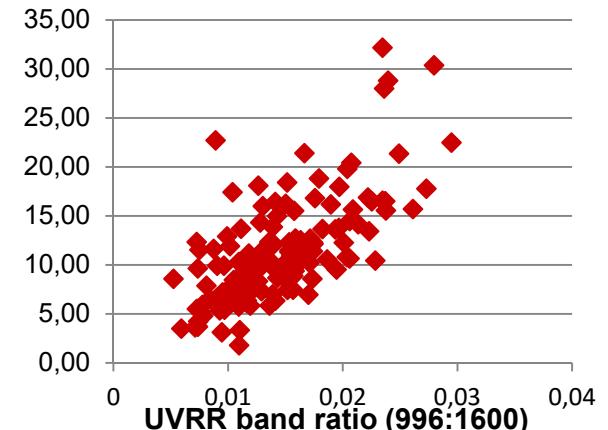
Case 1: Heartwood extractives in Scots pine: wet chemistry vs. two optical methods

Parallel samples for

GC-MS (wet chemistry; Luke)

UVRRaman (Aalto Univ)

NIRS (INRA, Innventia)



Aim: High-throughput phenotyping

Case 1: Heartwood extractives in Scots pine: wet chemistry vs. two optical methods

Method	V _a	h ²	Mean	C _{Va}	SD(h ²)
GC	14,15	0,85	11,30	0,33	0,30
UVRaman	6,94	0,33	10,14	0,26	0,21
NIRS	12,93	0,84	11,39	0,32	0,29

Aim: High-throughput phenotyping

Case 2: Heartwood extractives in larch:

Objective:

Evaluate variability of extractives content in relation to natural durability

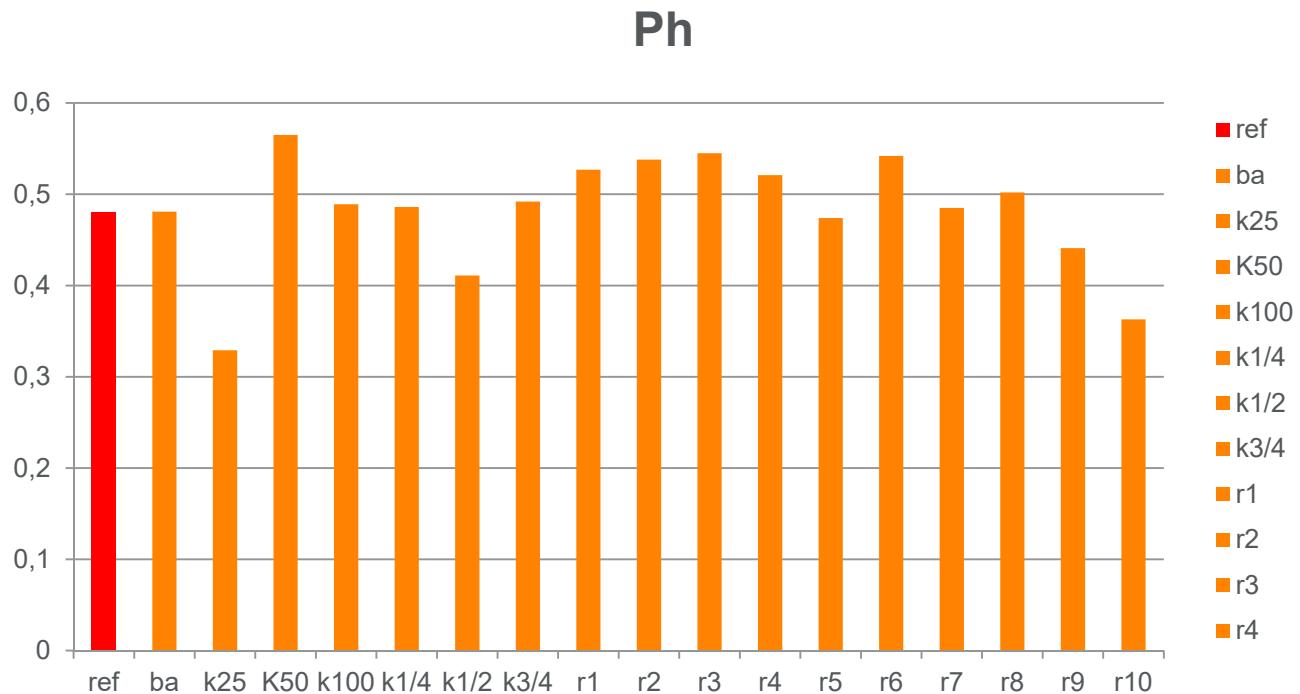
Develop a NIRS calibration model for both extractives and natural durability

Material: 42 FS progenies from 12 EL x 8 JL

355 individuals (out 1302 increment cores collected), +/- 8.4 sibs/family

NIRS spectra and calibration models

Family heritability variation: wet chemistry (red) vs. optical analyses: impact of different calibration models



Ref: Reference HPLC (wet chemistry)

- 1) batch: $\frac{1}{4}$ used to build up the model, prediction of the next $\frac{3}{4}$.
- 2-7) sets of calibration made of 25, 50, 100, $\frac{1}{4}$, $\frac{1}{2}$ and $\frac{3}{4}$ of individuals
- 8-17) 10 random samples of 100 individuals in the training set



How? Aim: High-throughput phenotyping

Case 3: Physical wood characteristics in larch: towards high-throughput phenotyping

Material:

- 25 FS hybrid larch progenies (~10 trees/family)

Measurements:

- Pilodyn: 2 directions; in 1993 (age 11); mm
- Density (from microdensitometry): ring 1989-1995 (age 13); kg/m³
- MOE (from rigidimeter on standing trees): in 1998 (age 16); MPa

Case 3: Physical wood characteristics in larch: towards high-throughput phenotyping

Level of throughput (trees around 20-25 yrs)

Technique	Operation	Timing	Throughput Nber of samples per M/D	'efficiency' towards MOE
MOE (rigidimeter)	Pruning & measure	20-25 trees/D Need 2 persons	12	1
Pilodyn	Measure	250 trees/M/D	250	21
Infradensity (cores)	Collect Measure	100 cores/M/D 200 cores/M/D	70	6
Microdensitometry	Collect Sawing RX Windendro	100 core/M/D 50 cores/M/D 150 /M/D 50/M/D	18	1.5

M = person

D = day

Estimation of heritability (family level) and of direct and indirect genetic gains from family selection

Heritability, absolute and relative gains for direct FS family selection (i=1)

	Mean	σ_{FS}	H^2_{FS}	ΔG_{abs}	$\Delta G_{rel} (\%)$
MOE (MPa)	7169.827	819.378	0.706 (0.500-0.851)	578.48	8.1
Density (Kg/m ³)	411.764	19.658	0.726 (0.534-0.861)	14.27	3.5
Pilodyn (mm)	14.312	1.325	0.678 (0.452-0.837)	0.898	6.3

Case 3: Physical wood characteristics in larch: towards high-throughput phenotyping

Correlated response

Target trait: MOE

	Correlated gain with MOE (MPa) (i=1)	Selection efficiency	i Due to throughput level	Correlated gain with MOE (MPa)	Selection efficiency
Pilodyn	394.612	0.682	2.459	970.351	1.677
Infradensity	375.786*		1.96	736.541	1.273
Density	375.786	0.650	1.271	477.624	0.826

* Supposed gain equal to that from microdensity

WHEN?

→ Phenotypic variation in different life stages

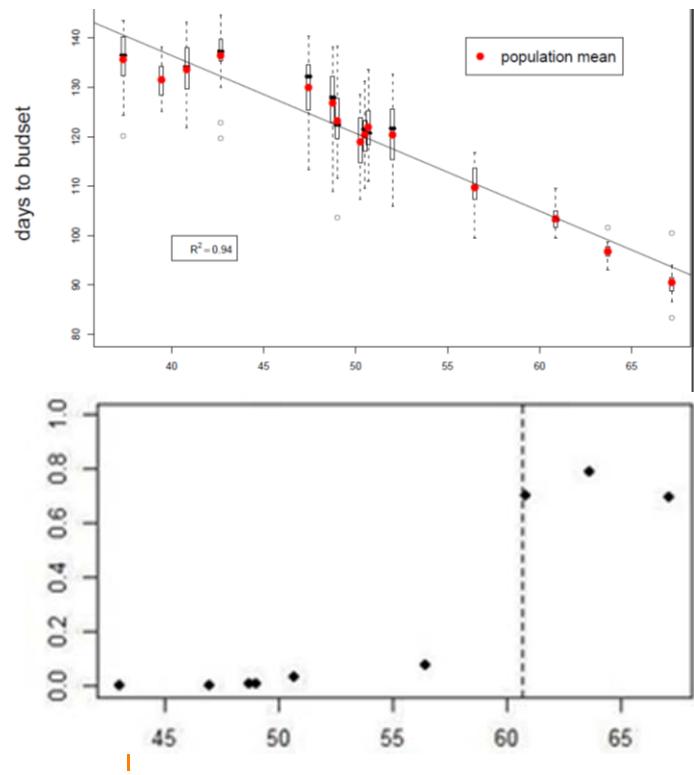


→ Phenotyping when it matters

When? Correlation between phenotypic variation in different life stages

Case1: Growth rhythm in different life stages

Bud-set in greenhouse of 1st yr seedlings v. survival (10 yrs) in field



Apuli et al, MS

When? Correlation between phenotypic variation in different life stages

Case 2: Can induced defense of seedlings used to assess constitutive heartwood extractives?



Extractive (mg g^{-1} d.w.)	h^2_{OP}
PS	0.31
PSM	0.09
Sum of stilbenes	0.20
Sum of RAC	0.13

Harju et al 2008. Tree Physiol.



Conclusions

High-throughput phenotyping

Why?

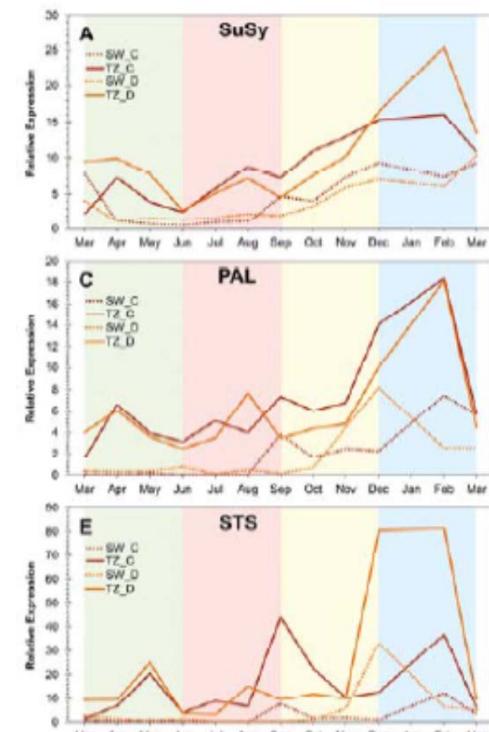
What?

How?

When?

Important to plan well beforehand

Future demands: more detailed phenotypes
for genomic selection



A photograph of a forest with many tall, thin trees. In the foreground, there is a fallen log or a pile of branches on the ground. The scene is somewhat dim, suggesting it might be early morning or late afternoon.

Thank you