Review of quantitative genetics

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Something you need to carefully look at, or that may impair your GWP



Something to do, or that optimizes your GWP



Don't. Discourage to use this.



Smart tip. Something that makes the trick.



Advanced. Something to dive in.

Challenges

What you need to know from this lecture

Basic concepts

How genes and environment modulate the phenotype

What is genomic heritability and how it affects GWP

Interpret what a GWP implies

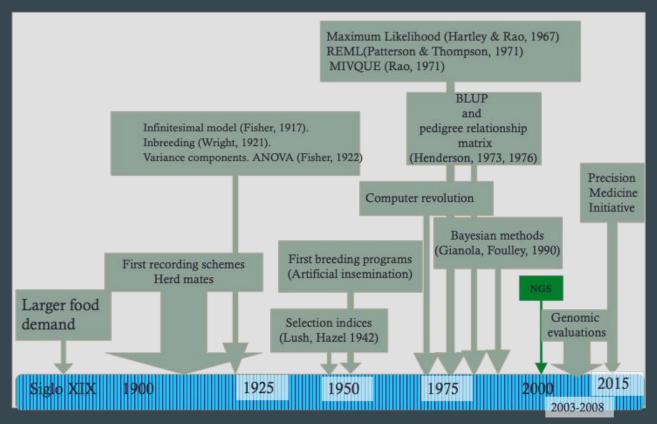
Genotyping strategies

Phenotyping strategies



A bit of history

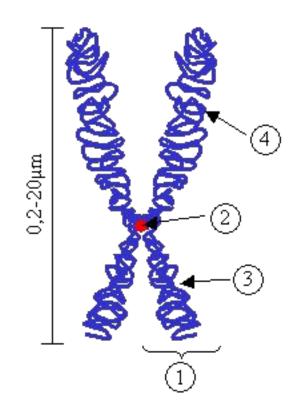






Locus, loci

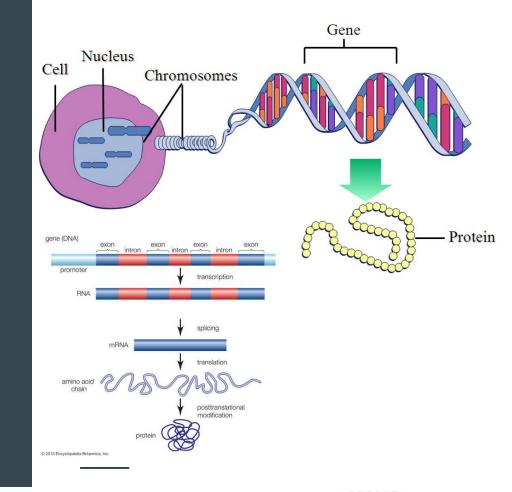
A specific physical location of a gene, DNA sequence or genetic marker on a chromosome; like a genetic street address





Gene

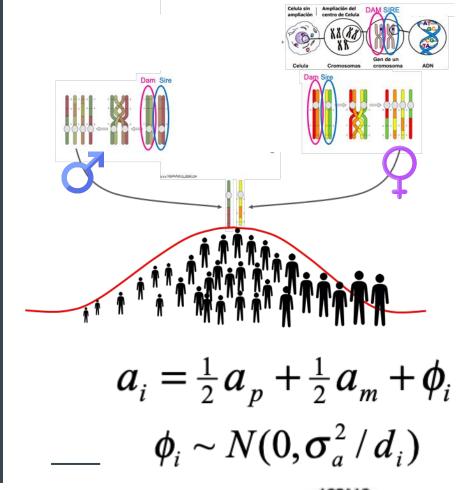
Gene, unit of hereditary information that occupies a fixed position (locus) on a chromosome. Genes achieve their effects by directing the synthesis of proteins.





Mendelian effect

Deviation from the expected parent average





Pedigree index

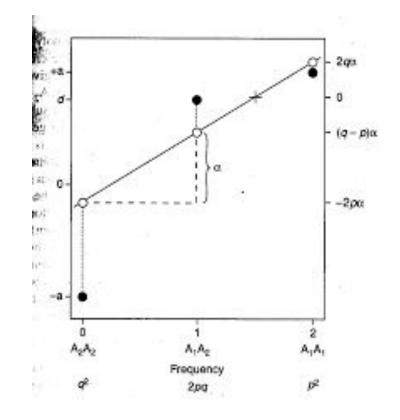
Parent average

1/2 EBV sire +1/2 EBV dam



Allele substitution effect

The effect that the presence of a copy of an allele has on the phenotype (regarding the reference allele).



f(A) = mean(Aa) - mean(aa)



Pleiotropy

the phenomenon in which a single locus affects two or more apparently unrelated phenotypic traits

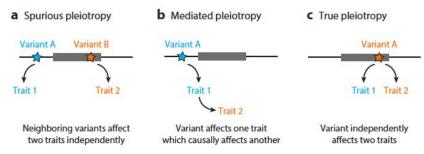
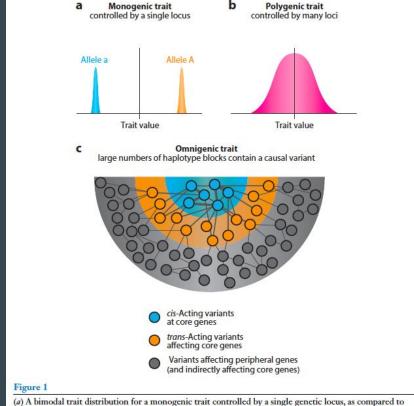


Figure 4

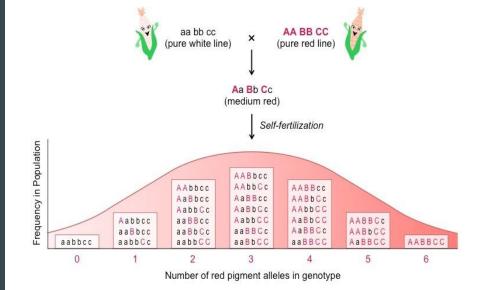
Diagrams illustrating (a) spurious pleiotropy, in which two neighboring, separately causal variants (blue and orange stars) are mistakenly inferred to be pleiotropic because they cannot be statistically distinguished; (b) mediated pleiotropy, in which a variant is statistically associated with two traits because it has a causal effect on one trait that in turn causally impacts another; and (c) true pleiotropy, in which a single unambiguous causal variant is separately biologically causal for two independent traits.



(a) A bimodal trait distribution for a monogenic trait controlled by a single genetic locus, as compared to (b) a continuous trait distribution for a polygenic trait controlled by many genetic loci. (c) Schematic of one possible architecture for an omnigenic trait, in which several large-effect cis-acting and many smaller-effect trans-acting variants modulate a set of core genes, as does a much larger ensemble of cis- and trans-acting variants impacting peripheral genes that only indirectly modulate the phenotype.

Infinitesimal model

A quantitative trait is influenced by an infinitely large number of genes, each of which makes an infinitely small (infinitesimal) effect, as well as by environmental factors. Random sampling of alleles at each gene produces a continuous, normally distributed phenotype in the population (at least around the average of that of the individual's parents).



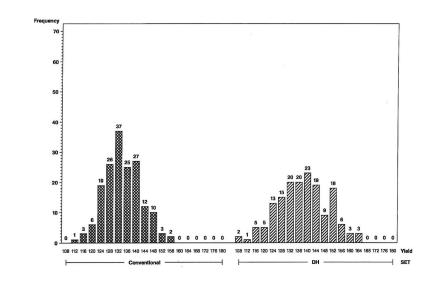
Probability of j major alleles in k biallelic loci \Rightarrow

$$\Rightarrow \binom{2k}{j} = \left(\frac{1}{2}\right)^{2k} \frac{2k!}{j! (2k-j)!}$$

Dogenome-wide prediction

Genetic variance

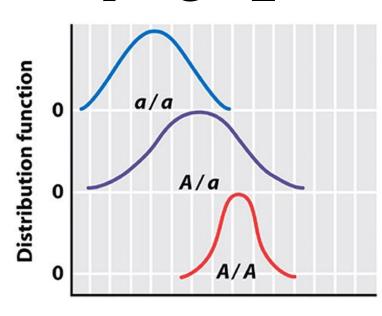
Phenotype deviation from the mean phenotype caused by the combination of alleles inherited from parentals and these alleles independent effects on the specific phenotype



Phenotype decomposition

Phenotype is affected by genetic (additive +dominance +epistasis), environment and their interactions.

$$P = G + E$$



Height (h)



Heritability

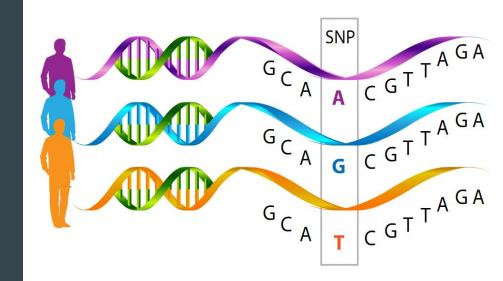
The amount of phenotypic (observable) variation in a population that is attributable to individual genetic differences

"Narrow and broad sense"

$$H^2 = \frac{V_g}{V_g + V_e}$$

Genetic marker

DNA sequence with a known location on a chromosome that can be used to identify individuals or species. It can be described as a variation (which may arise due to mutation or alteration in the genomic loci) that can be observed





Marker variance

Phenotype deviation from the mean phenotype caused by the inheritance of a particular allele from parentals and this allele's independent effect on the phenotype

Notation	Variance component	Genotype coding		
V_A	$2pq[a+d(p-q)]^2$	$x_A \in \{0, 1, 2\}$		
V_D	$(2pqd)^2$	$x_D \in \{0, 2p, 2(p-q)\}$		
V_D'	$\frac{4pq^2}{1+q}(a+dq)^2$	$x_D' \in \{0, 2, 2\}$		
V_A'	$\frac{2p^2q}{1+q}(a-d)^2$	$x_A' \in \{0, \frac{1-q}{1+q'}, \frac{-2q}{1+q}\}$		
$V_{AA}^{\prime\prime}$	computed numerically	$x_{AA}^{\prime\prime} \in (x_{A,1} - 1)(x_{A,2} - 1)$		

Genomic variance

The amount of variance explained by marker effects (<genetic variance, because of incomplete LD with QTLs or missingness)

$$Var(\beta'x_i) = \beta'Cov(x_i, x_i')\beta$$

$$= \beta'\Sigma_x\beta$$

$$= \alpha'\Sigma_{xx}\Sigma_x^{-1}\Sigma_x\Sigma_x^{-1}\Sigma_{xx}\alpha$$

$$= \alpha'\Sigma_{xx}\Sigma_x^{-1}\Sigma_{xx}\alpha$$

Genomic heritability

The proportion of variance of a trait that can be explained (in the population) by a linear regression on a set of markers

$$h_g^2 = \frac{\sigma_g^2}{\sigma_y^2} = \frac{\sigma_a^2}{\sigma_y^2} \frac{\sigma_g^2}{\sigma_a^2} = h^2 \frac{\sigma_g^2}{\sigma_a^2}$$

$$h_{g}^{2} < = h^{2}$$

Missing heritability

The problem of missing heritability, that is to say the gap between heritability estimates from genotype data and heritability estimates from twin data WENNE PERFORM PERSONAL GENOMES.



The case of the missing heritability

When scientists opened up the human genome, they expected to find the genetic components of common traits and diseases. But they were nowhere to be seen, **Erendan Maher** shines a light on six places where the missing lost could be steaked every.



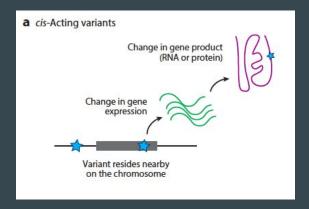
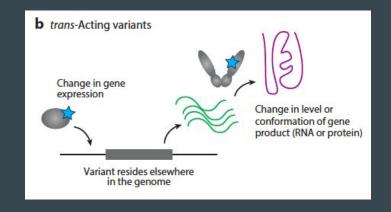
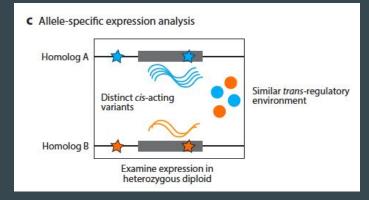


Figure 2

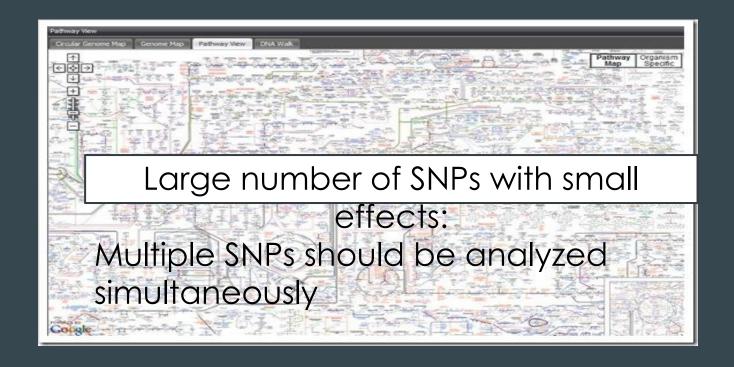
(a) cis-Acting variants that impact the expression of a gene immediately proximal on the chromosome.
(b) trans-Acting variants that impact a gene product originating from a distal genetic locus. (c) Schematic of an experimental design to measure allele-specific mRNA levels. Due to the presence of both parental alleles in the F₀ heterozygote, cis-acting regulatory activity is inferred from differential expression of the messenger RNA attributable to one of the two homologous loci. This is because both homologs exist in an essentially equivalent trans-regulatory environment; any difference in abundance must therefore be due to a nearby cis-acting variant.





Jakobson and Jarosz, 2020



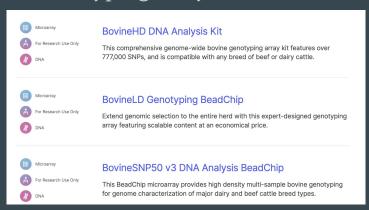


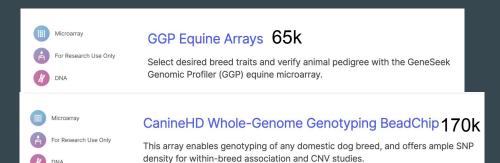
- →Whole Genome Sequencing (WGS)
- →Genotyping chips (SNPchips)
- →Restriction site-associated DNA sequencing (RADseq)
- →Genotype by low-pass sequencing (skim-Seq)

- →Whole Genome Sequencing (WGS)
 - Next generation sequencing
 - Wide range of genetic variants (SNP, Indels, CNV, Structural Variants)
 - High cost

- →Genotyping arrays
 - Most used / Widely implemented
 - Large variety (species and densities)
 - Most available from sequencing services /labs
 - SNP + short indels
 - Biallelic markers

→Genotyping arrays





OvineSNP50 DNA Analysis Kit

This sheep microarray features over 54,241 evenly spaced SNP probes for genome-wide association studies, genome-wide selection, and genetic merit determination.Read More...

The BeadChip was developed in collaboration with leading ovine researchers from AgResearch, Baylor UCSC, CSIRO, and the USDA as part of the International Sheep Genomics Consortium. It features over 54,241 evenly spaced probes that target single nucleotide polymorphisms (SNPs).

GGP Porcine HD Array

This genome-wide porcine genotyping array is ideal for markerassisted selection and prediction applications. Array content includes:

- . 70,000 SNPs for all major porcine breeds
- · Average marker spacing of ~42 kb
- · 20 key causative mutations

GGP Porcine LD Array

This array is designed for marker-assisted selection, Illumina PorcineSNP60 imputation, GGP Porcine HD imputation, and prediction applications. Array content includes:

- · More than 10,000 SNPs for all major porcine breeds
- . Average marker spacing of ~250 kb
- ~20 important causative mutations



m Asian Screening Array-24 v1.0 BeadChip

ful, cost-effective genotyping array for large-scale genetic studies and eogenomics in East Asian populations.

m Core-24 Kit

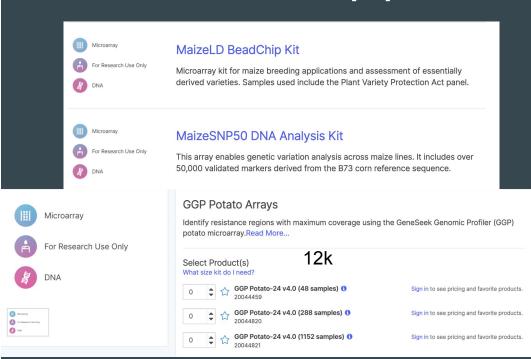
rays support economical large-scale human genotyping studies, with bughput capabilities and the option to add up to 300K semi-custom

m CoreExome-24 Kit

A microarray kit delivers genome-wide SNP and genetic variant ion for genetic studies, especially large-scale human genotyping studies.

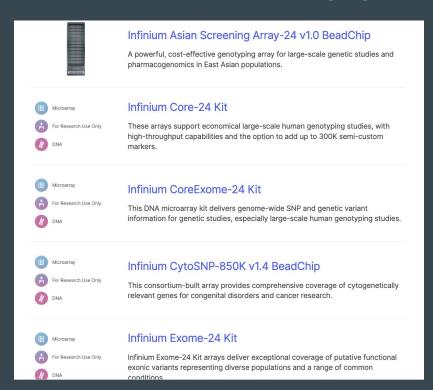
m CytoSNP-850K v1.4 BeadChip

sortium-built array provides comprehensive coverage of cytogenetically genes for congenital disorders and cancer research.



e-wide prediction

→Genotyping arrays

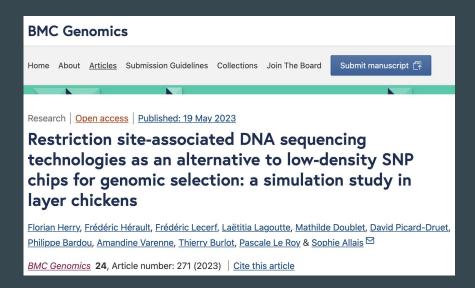




- \rightarrow Restriction site-associated DNA sequencing (RADseq)
 - Uses NGS
 - Requires specific library preparation with specific restriction enzymes
 - Most effective in organisms with well-characterized reference genomes
 - Cost-effective and medium-throughput

→Restriction site-associated DNA sequencing (RADseq)





→Restriction site-associated DNA sequencing (RADseq)

Highly dependent on the restriction enzyme

GEBV correlation <0.50 with HD SNP arrays

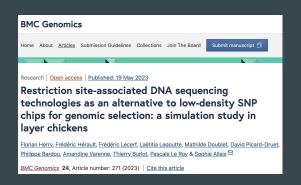


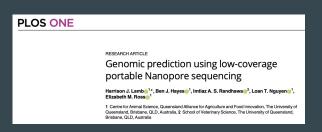
Table 6 Pearson correlations between true "Full_HD" GEBVs and imputed HD GEBVs based on ancestry for the 67 G1 breeders, according to each enzyme used for egg weight (EW), eggshell colour (ESC), eggshell strength (ESS) and albumen height (AH).

	Number of SNPs	EW	ESC	ESS	АН
EcoRI	1,797	0.3774	0.2962	0.3420	0.4261
Taql	4,126	0.4476	0.2453	0.3906	0.4478
Taql_Pstl	11,193	0.4740	0.2442	0.3869	0.4684
Avall	12,453	0.4681	0.2430	0.3859	0.4794
Pstl	14,390	0.4664	0.2450	0.3953	0.4689
HD SNP chip	300,028	0.4713	0.2460	0.3940	0.4802

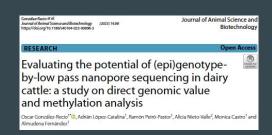
The line HD SNP chip corresponds to the Pearson correlation between true "Full_HD" GEBVs and true HD GEBVs based on ancestry for the 67 G1 breeders.



- →Genotype by low-pass sequencing (Skim-Seq)
 - Low cost
 - Non targeted sequencing (needs imputation)
 - NGS (Illumina or ONT)
 - Minimum coverage ranges between 0.5x and 4x depending on population and sequencing method









→Genotype by low-pass sequencing (Skim-Seq)

Imputation accuracy improves with

sequencing depth





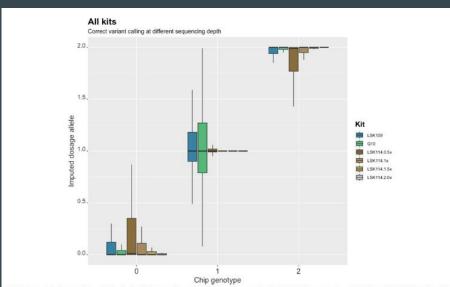
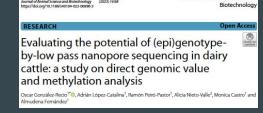


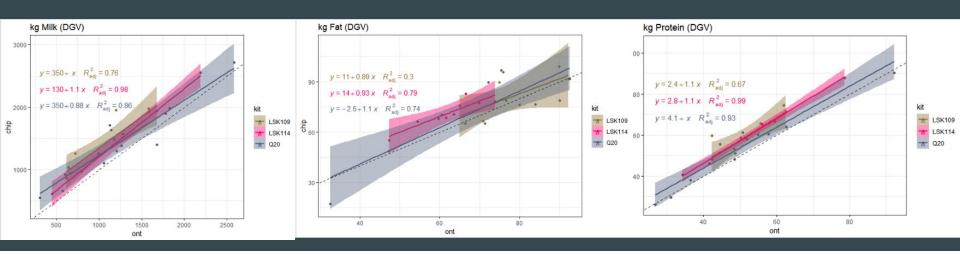
Fig. 3 Imputed dosage allele obtained after LPS according to the SNP genotype code. Each kit and sequencing depth (from LSK114) is depicted in different color. Samples from LSK109 (Q20) had average sequencing depth of 0.6 × (0.4 ×)



→Genotype by low-pass sequencing (Skim-Seq)

High correlation (>0.98) with latest technology



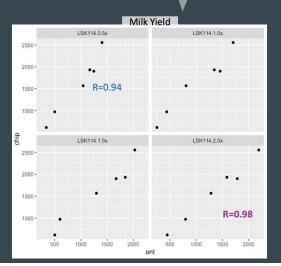




→Genotype by low-pass sequencing (Skim-Seq)

High correlation (>0.98) with latest technology

and seq-depth>2x.





Reference population

A Reference population is needed in Genome-wide prediction to train the statistical models

- Genotypes and phenotypes
- Statistical association between genotypes and phenotypes
 - Covariates
 - Genomic relationship
- Genomic predictions may be achieved in individuals without phenotype (but w/genotypes)

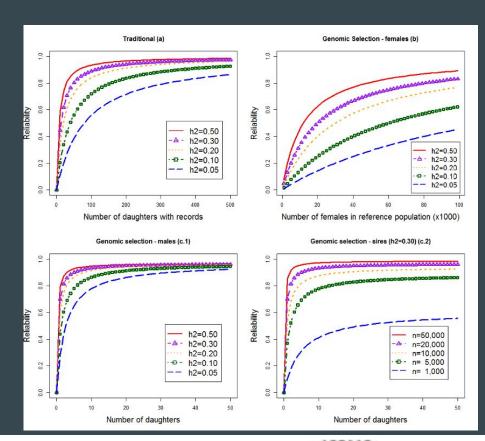
Strategies for phenotyping

Cost-benefit function

 How much phenotyping cost vs how much prediction accuracy is gained (trait dependent)









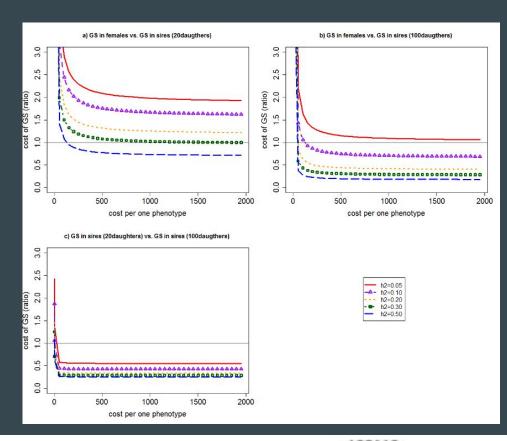
Strategies for phenotyping

Cost-benefit function

 Cost of phenotype influences the genotyping strategy (parents vs individuals)









Strategies for phenotyping

Cost-benefit function

- Traits easy to measure
 - Use progeny tests and genotype parents
 - Establish a routine phenotype recording
- Traits difficult to measure and expensive
 - Use individual phenotype and record
 - Experimental conditions
- (This rule of thumb may not work in human medicine, depending on the importance of the trait)



RECAP

Assume gaussian distribution on phenotypes (... subsequently residuals)

Why variance is important

Inference is different from prediction.

Genetic architecture challenge

Genotyping strategies

Phenotyping strategies

