

Transcriptomes as phenotypes

Bringing Genetics to Genomics

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Online Slides Available at dangeles.github.io

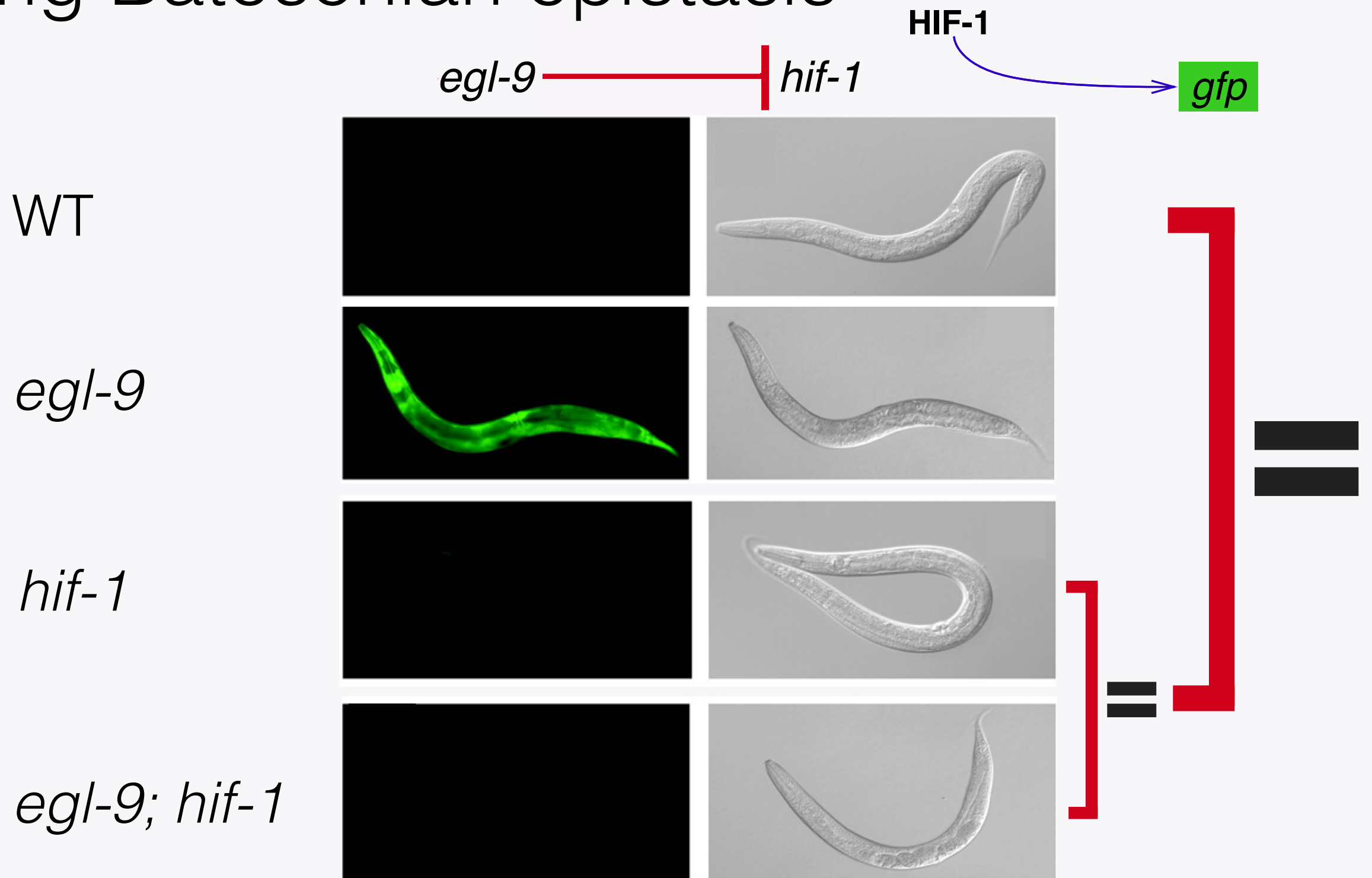
The geneticist's arsenal

Null mutants (**Epistasis**)

Allelic series (**dominance**)

Crosses (**maternal effects**)

Genetics orders genes along pathways using Batesonian epistasis



Epistasis analysis in a nutshell:

- (A) Choose phenotype (based on expertise)
- (B) Phenotype single, double NULL mutants
- (C) Check if double mutant = a single mutant

Yes?

Infer pathway

No?

Genetic interaction
is 'complex', need
more information

RNA-seq offers the possibility of a new
kind of phenotypes

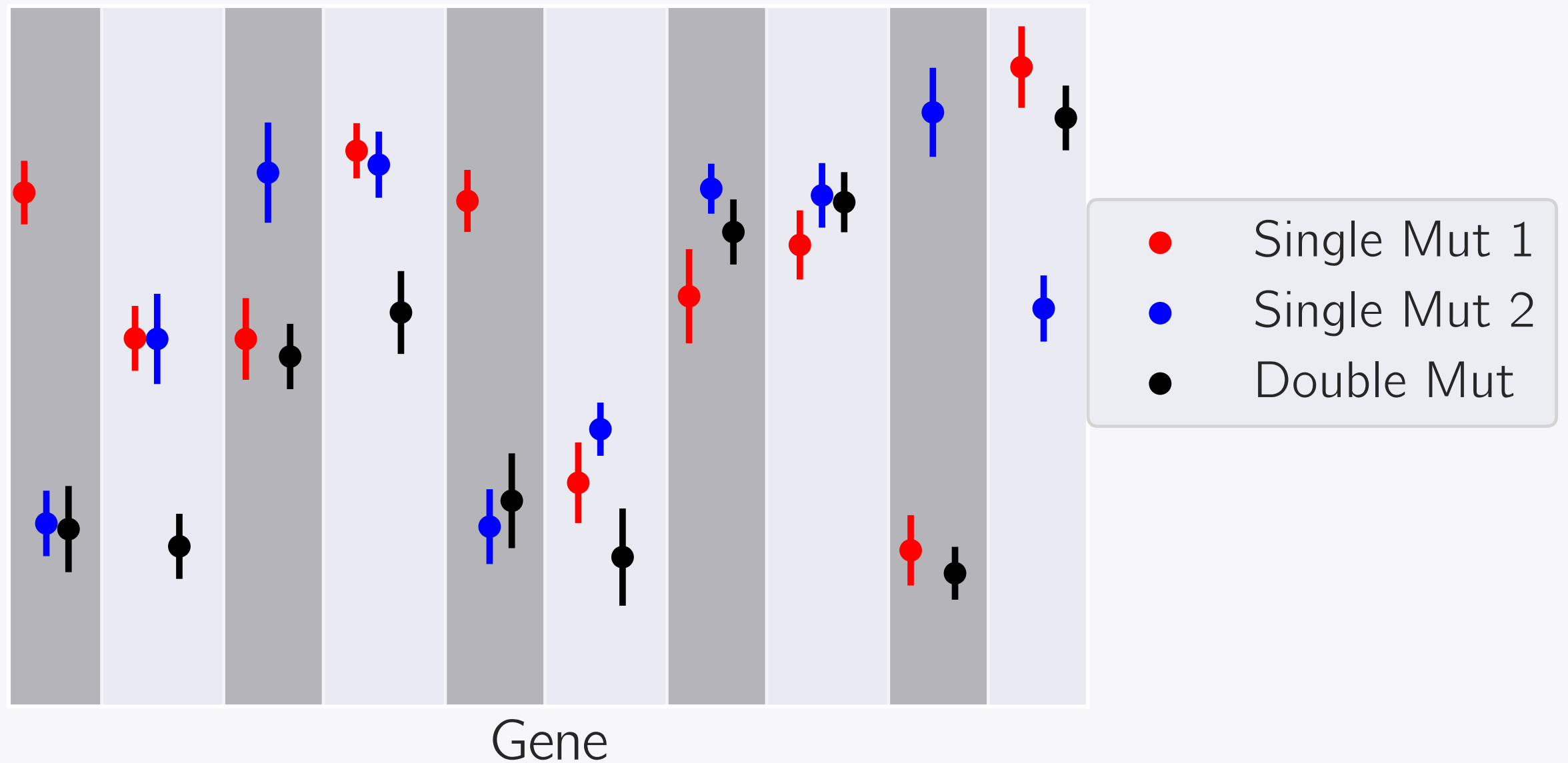
Genome-wide

Quantitative

Unbiased

Transcriptomes are powerful, but complicated

log Fold Change of Expression



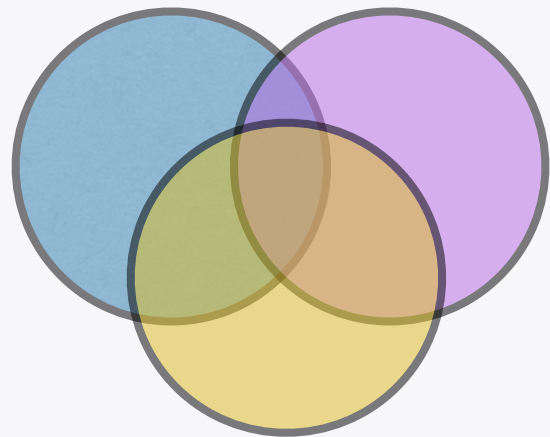
To use genetic methods in a genomic context, we need **specialized statistical machinery**

For details, see:

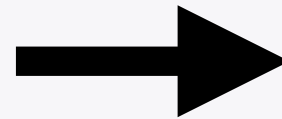
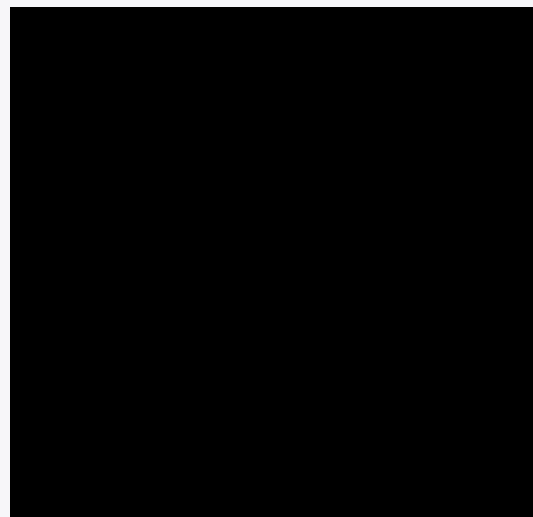
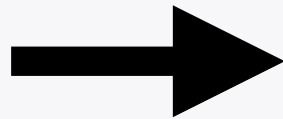
- **Epistasis:** Angeles-Albores *et al*, *PNAS*, 2018;
Angeles-Albores *et al*, *G3*, 2017
- **Dominance:** Angeles-Albores, *Genetics*, 2018

Transcriptome-wide epistasis analysis in a nutshell:

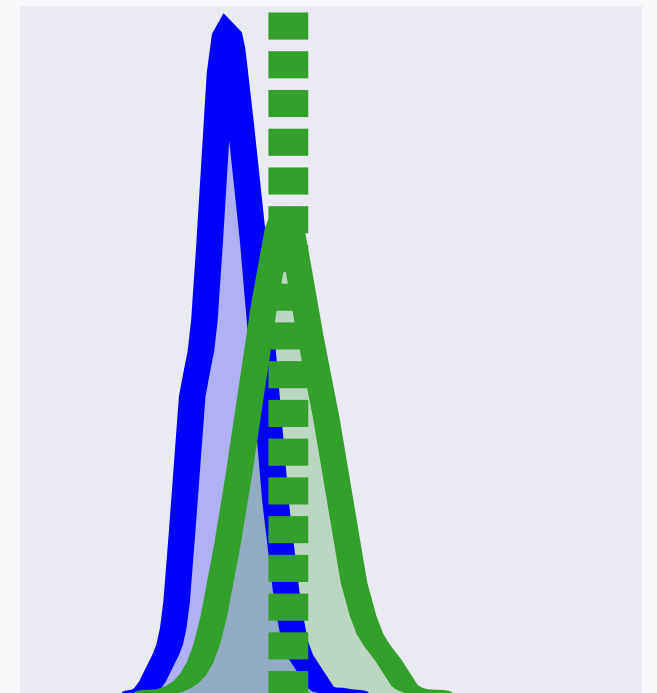
Choose
phenotype



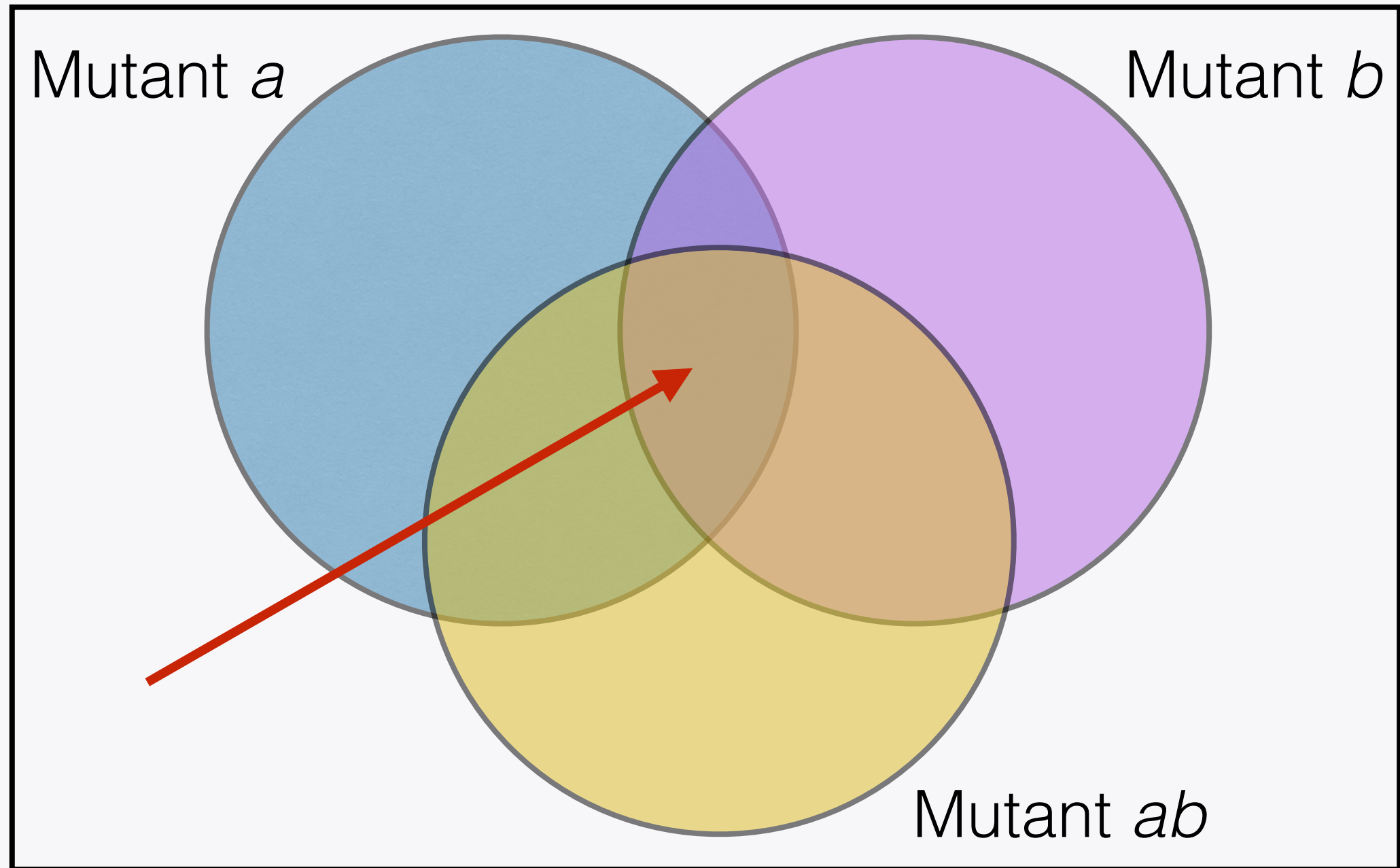
Compute a statistic for
all genes in phenotype



Check if statistic
is Batesonian



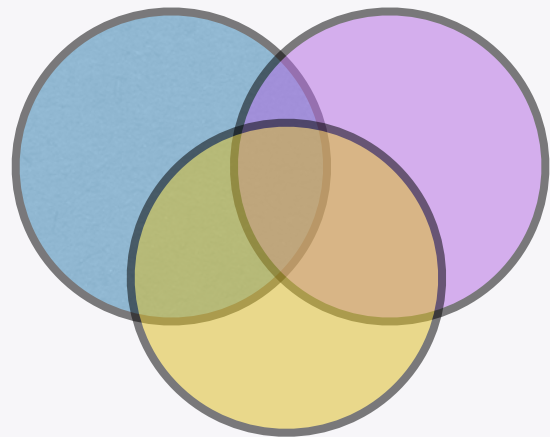
Transcriptome-wide epistasis: Defining a phenotype



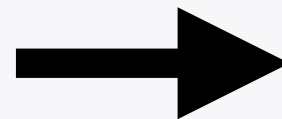
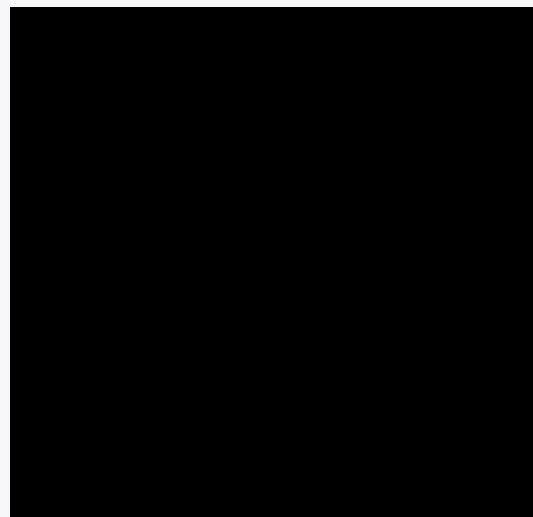
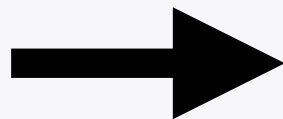
Diff. Exp. Genes relative to WT

Transcriptome-wide epistasis analysis in a nutshell:

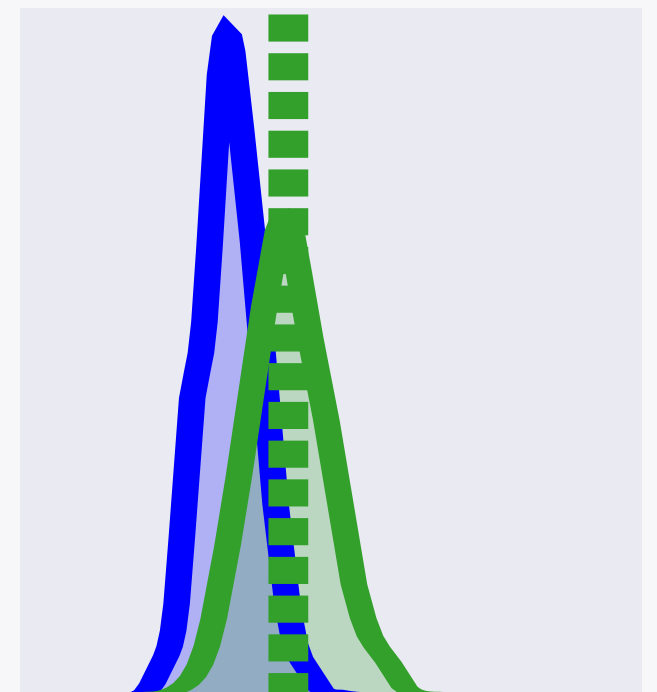
Choose
phenotype



Compute a statistic for
all genes in phenotype



Check if statistic
is Batesonian



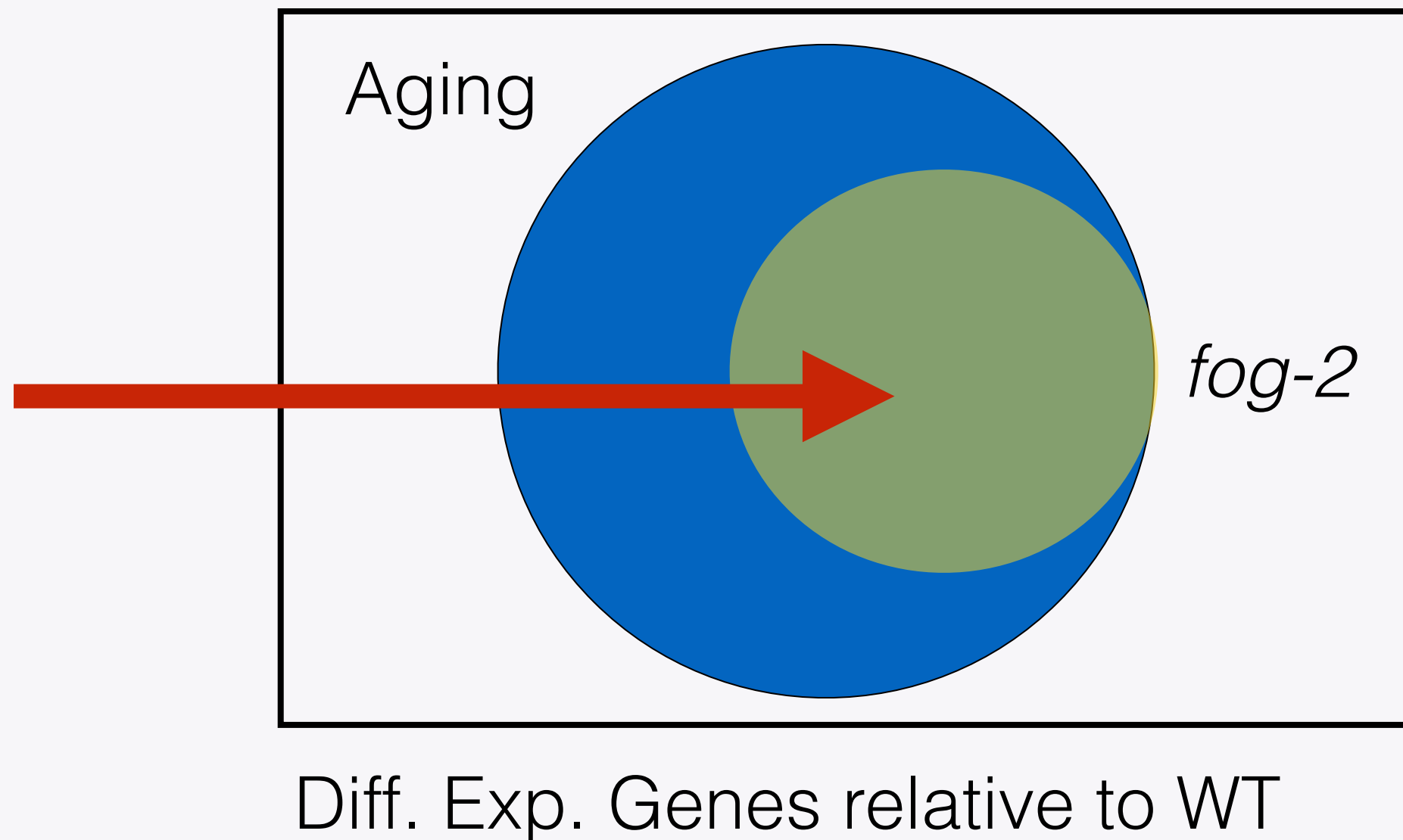
What does our black box do?

- (1) Calculate: **expected** double mutant value
(Add the single mutant log Fold Changes)
- (2) Compute: **difference = observed - expected**
- (3) Plot difference vs. expected for all transcripts
and **determine line of best fit**

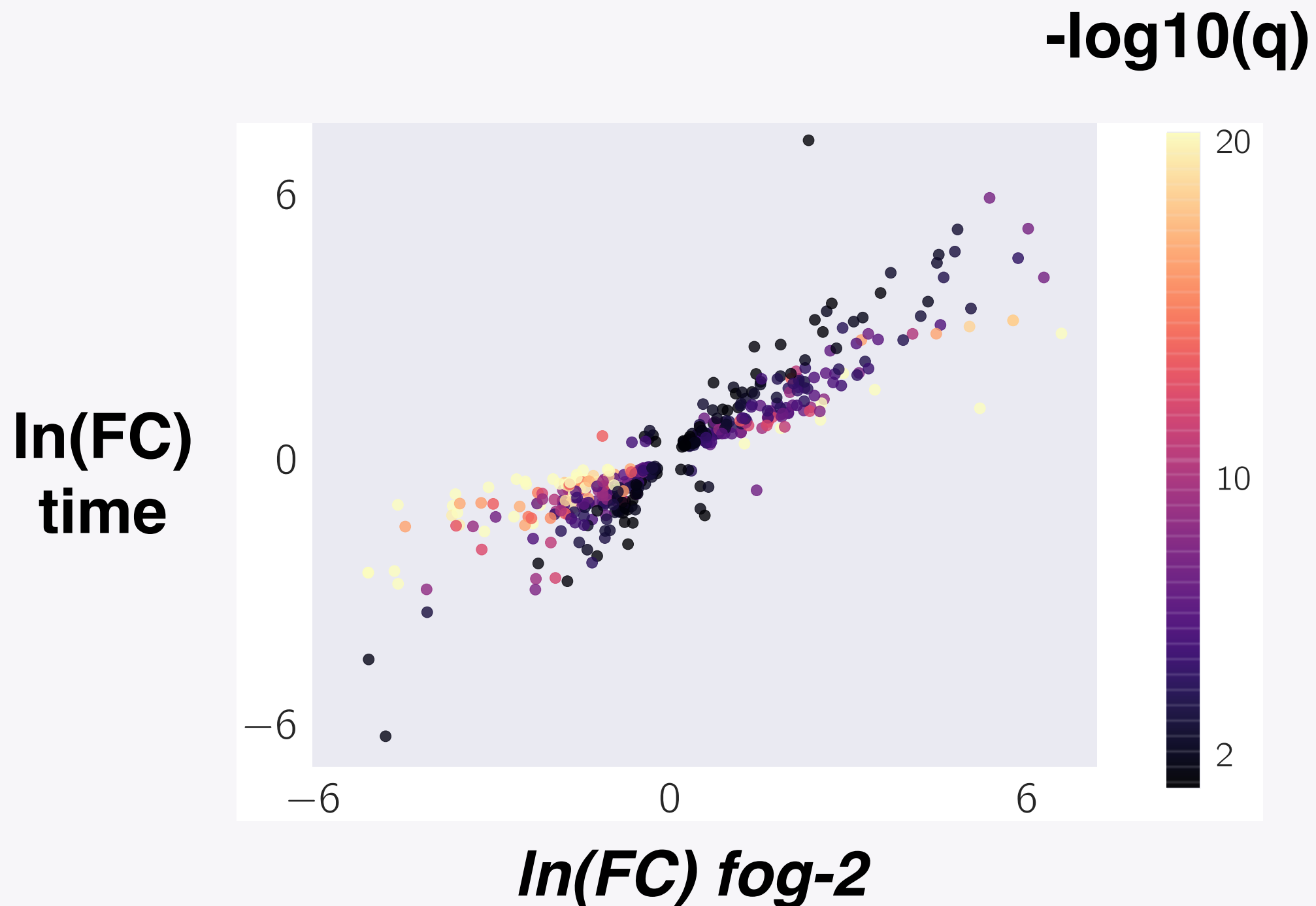
An example: Does sperm status have effects independent of aging?

	WT	<i>fog-2</i>
Young adult	Young, Sperm	Young, NO Sperm,
‘Middle-aged’ adult	Aged, NO Sperm,	Aged, NO Sperm,

Age affects more genes than *fog-2*, so we find the commonly affected subset

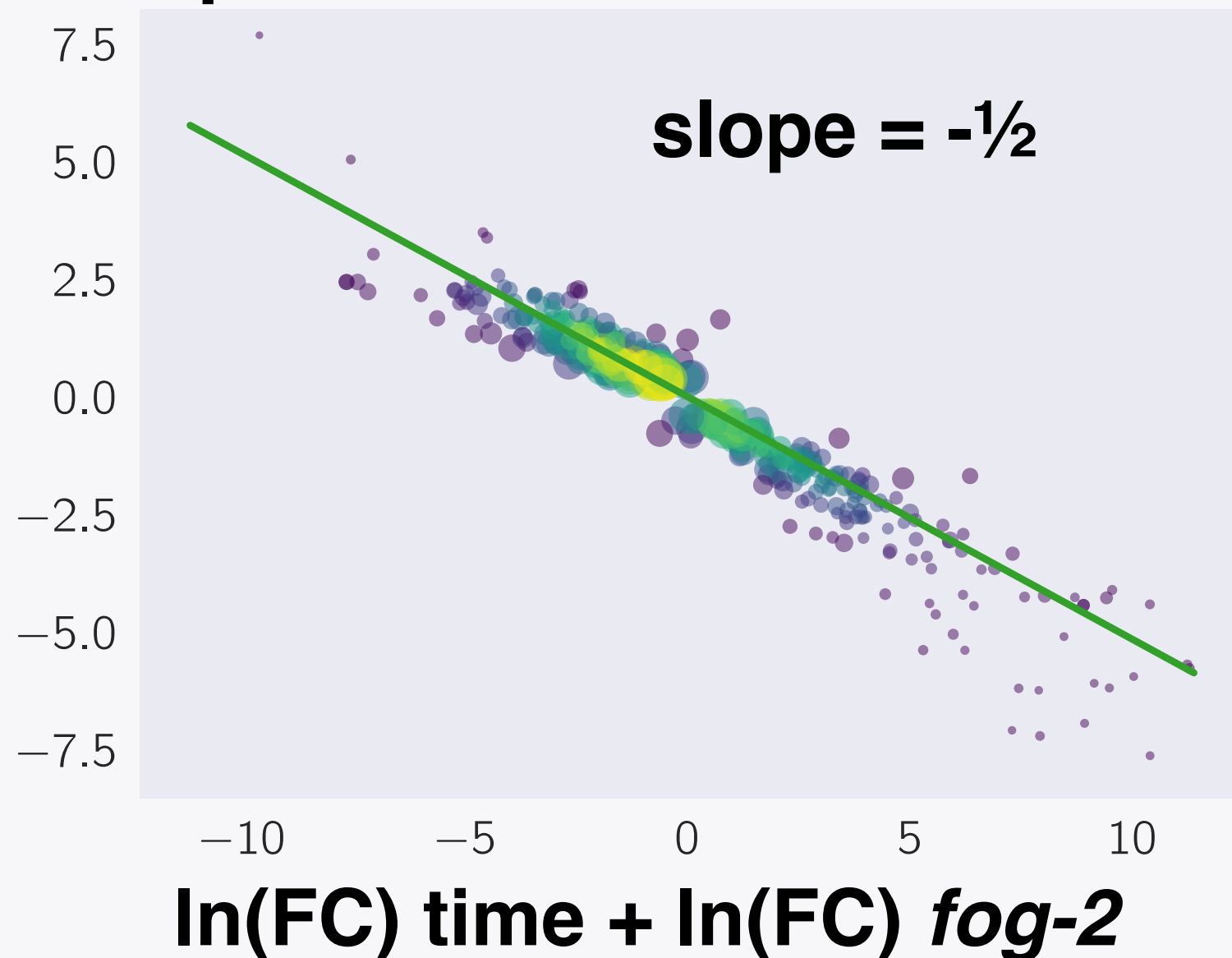


First, show that both perturbations
have equivalent effects



A slope of $-\frac{1}{2}$ indicates that sperm loss through aging is the same as never having sperm

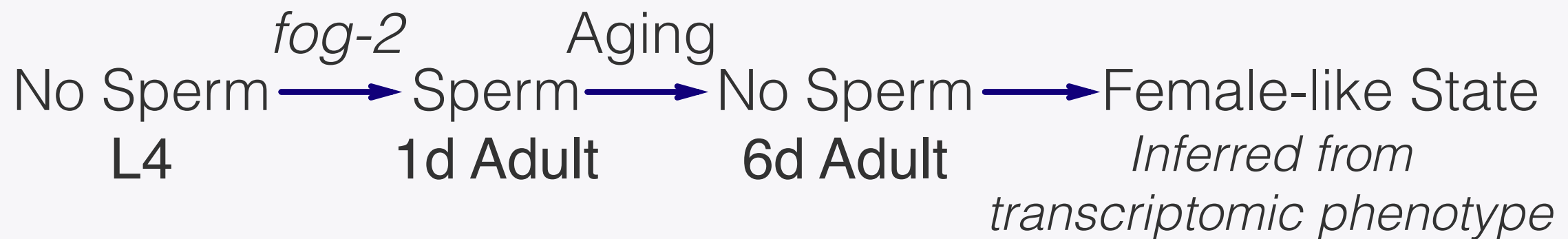
Observed - Expected



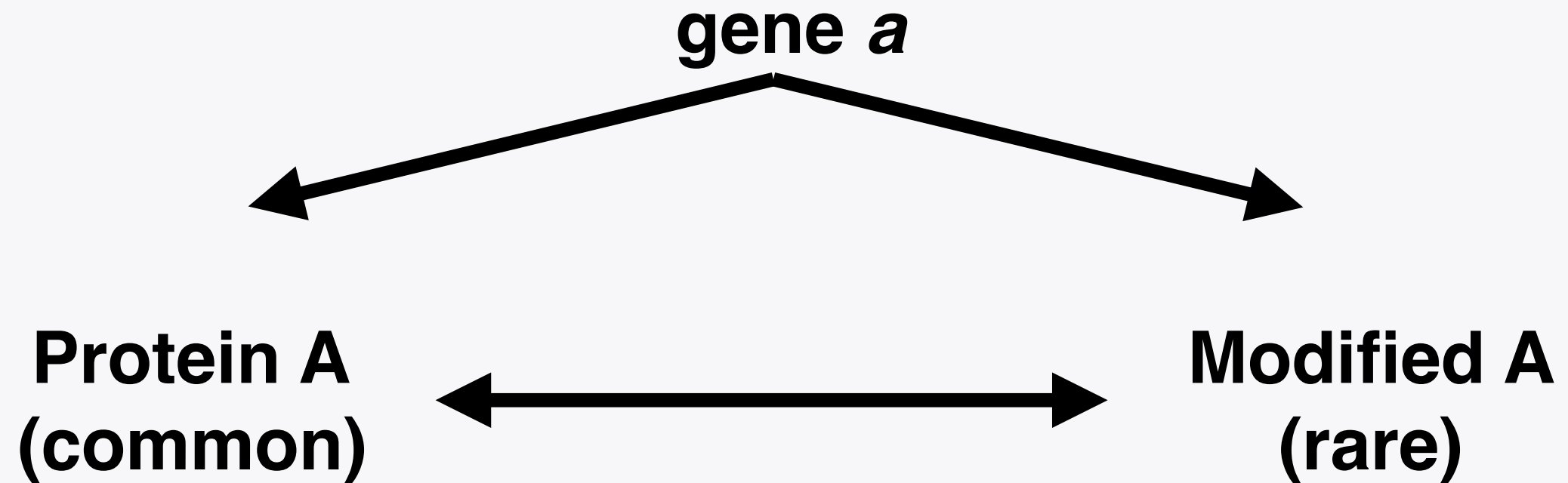
Behind the math:

$$\begin{aligned} \text{Observed} &= \\ &= \frac{1}{2} \text{ Expected} \\ &= \ln(\text{FC}) \text{ time} \\ &= \ln(\text{FC}) \text{ fog-2} \end{aligned}$$

The *C. elegans* female state was
inferred from transcriptome profiling



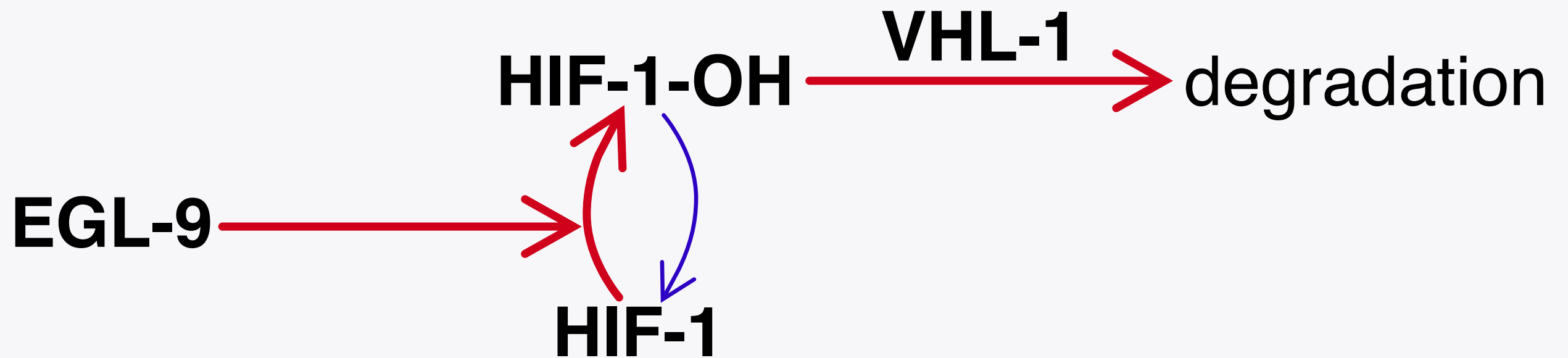
Transcriptomes can be used to think about biochemistry



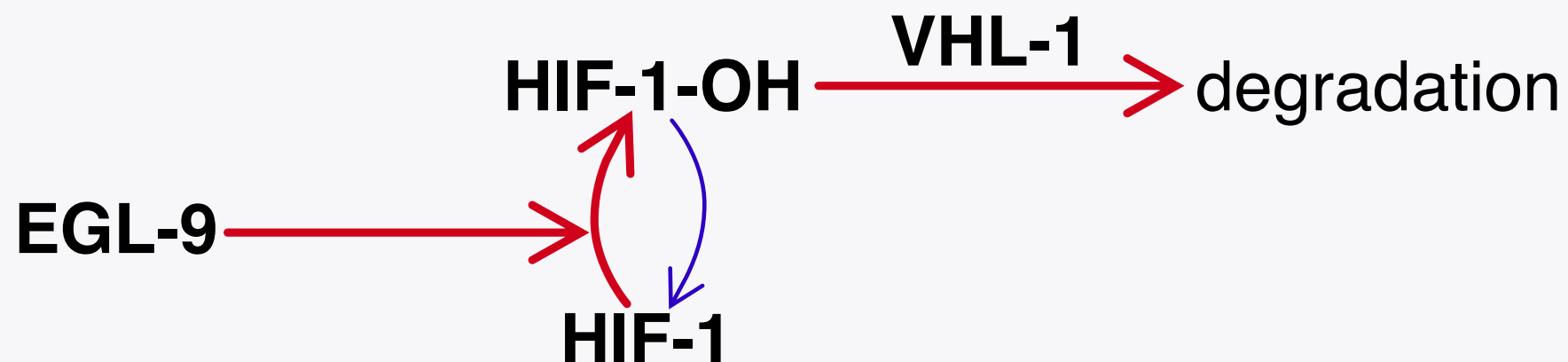
Accounts for most effects
of knocking out ***a***

Accounts for a few effects
of knocking out ***a***

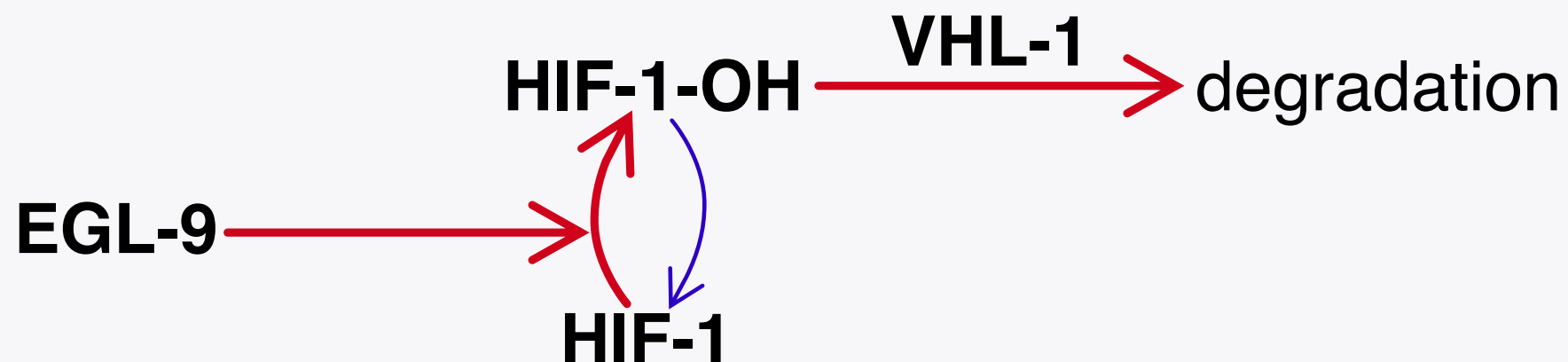
Hypoxia factor 1 (*hif-1*) is degraded by VHL-1 in an EGL-9 dependent manner



Using HIF-1 abundance as phenotype leads to the canonical genetic diagram:



If we could measure HIF-1-OH abundance,
we would write the genetic pathway as:



Choosing a phenotype affects the outcome of the genetic reconstruction:

**HIF-1 abundance
as phenotype**

egl-9 ———| *hif-1*

vhl-1 —————→ *egl-9*

**HIF-1-OH abundance
as phenotype**

egl-9 —————→ *hif-1*

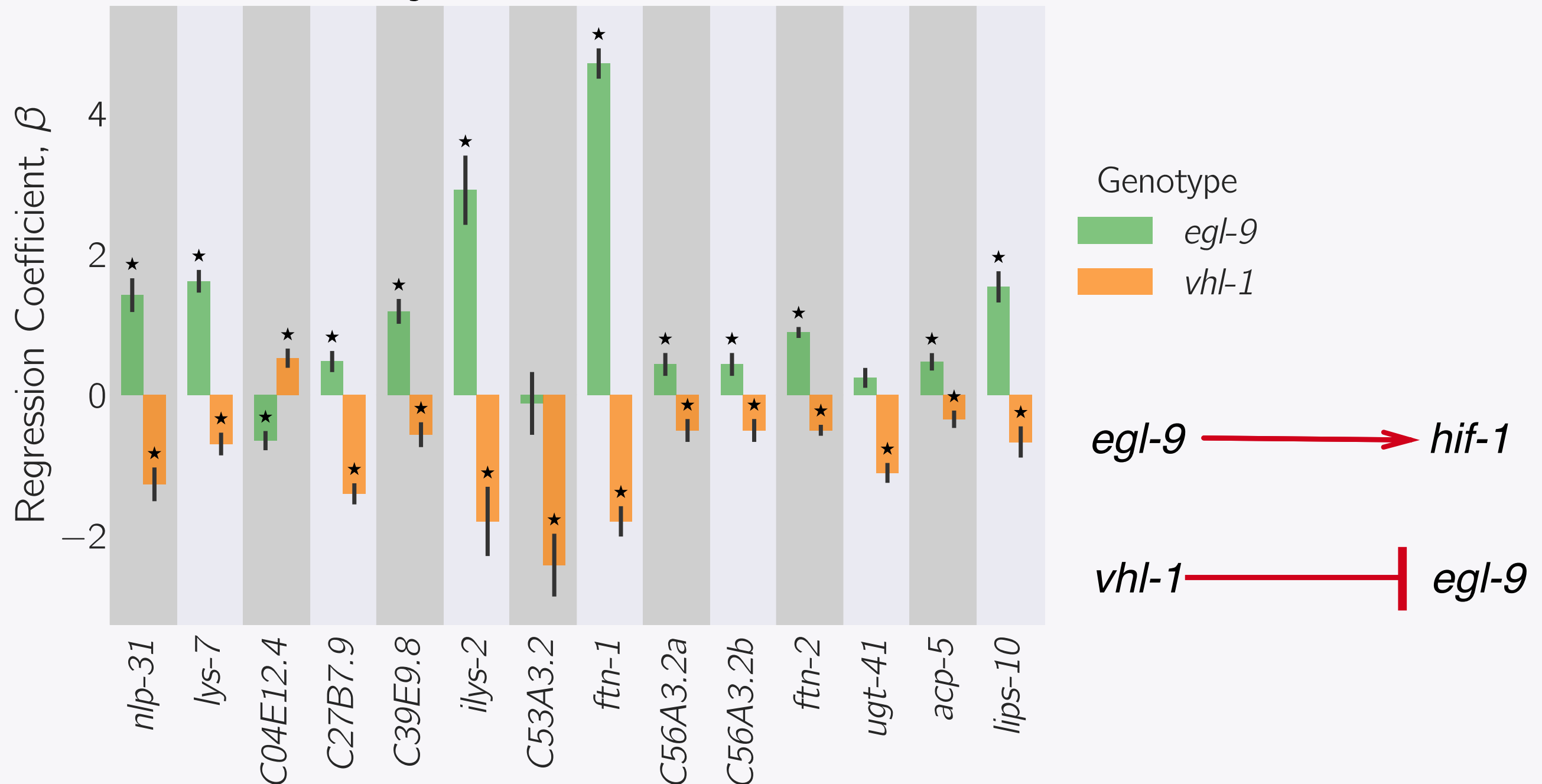
vhl-1 ———| *egl-9*

However, both pathways obey the same set of epistatic rules!

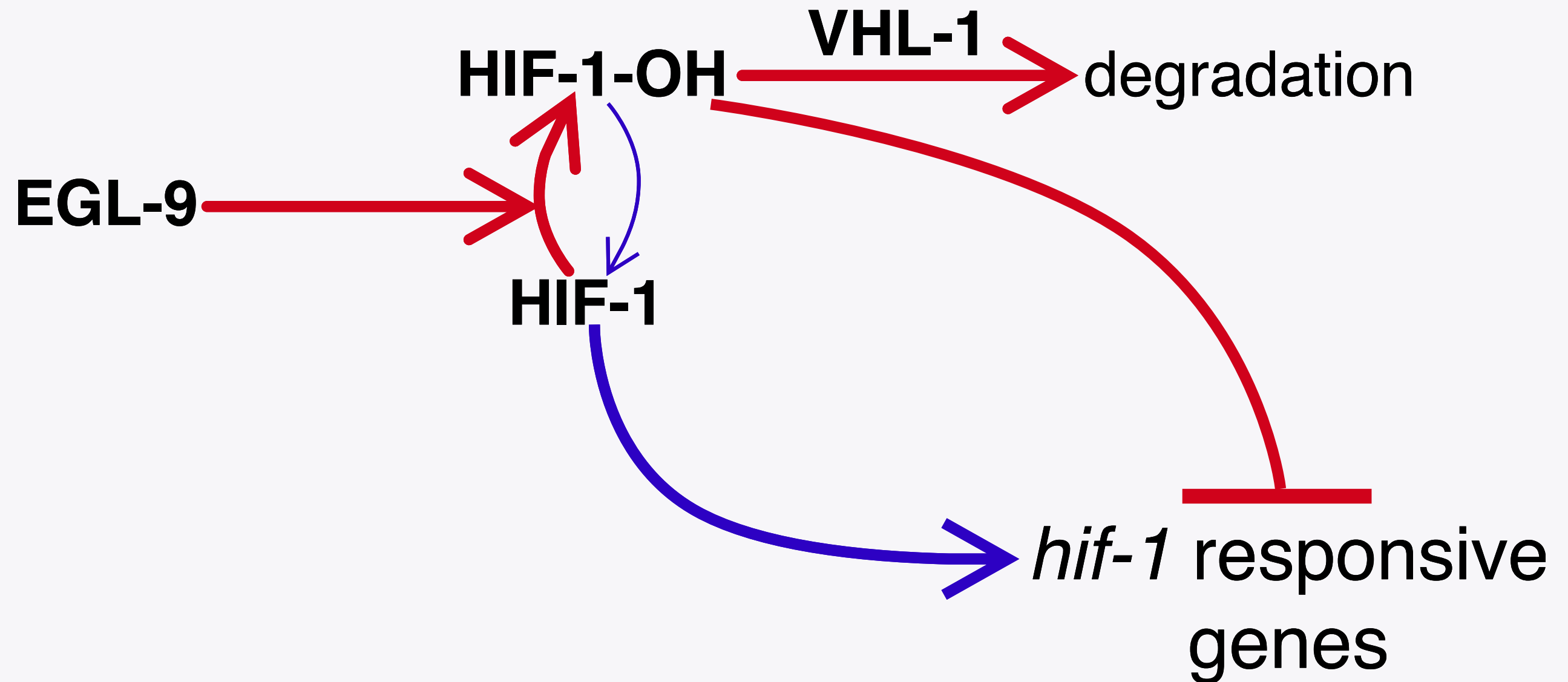
egl-9 = egl-9;vhl-1

hif-1 = egl-9;hif-1

Sequencing hypoxia pathway mutants reveals ~50 genes that behave as if controlled by HIF-1-OH



Hypothesis: A subset of genes is strongly responsive to HIF-1-OH levels



Transcriptomes + Biochemical Models can lead to testable hypotheses about molecular functions.

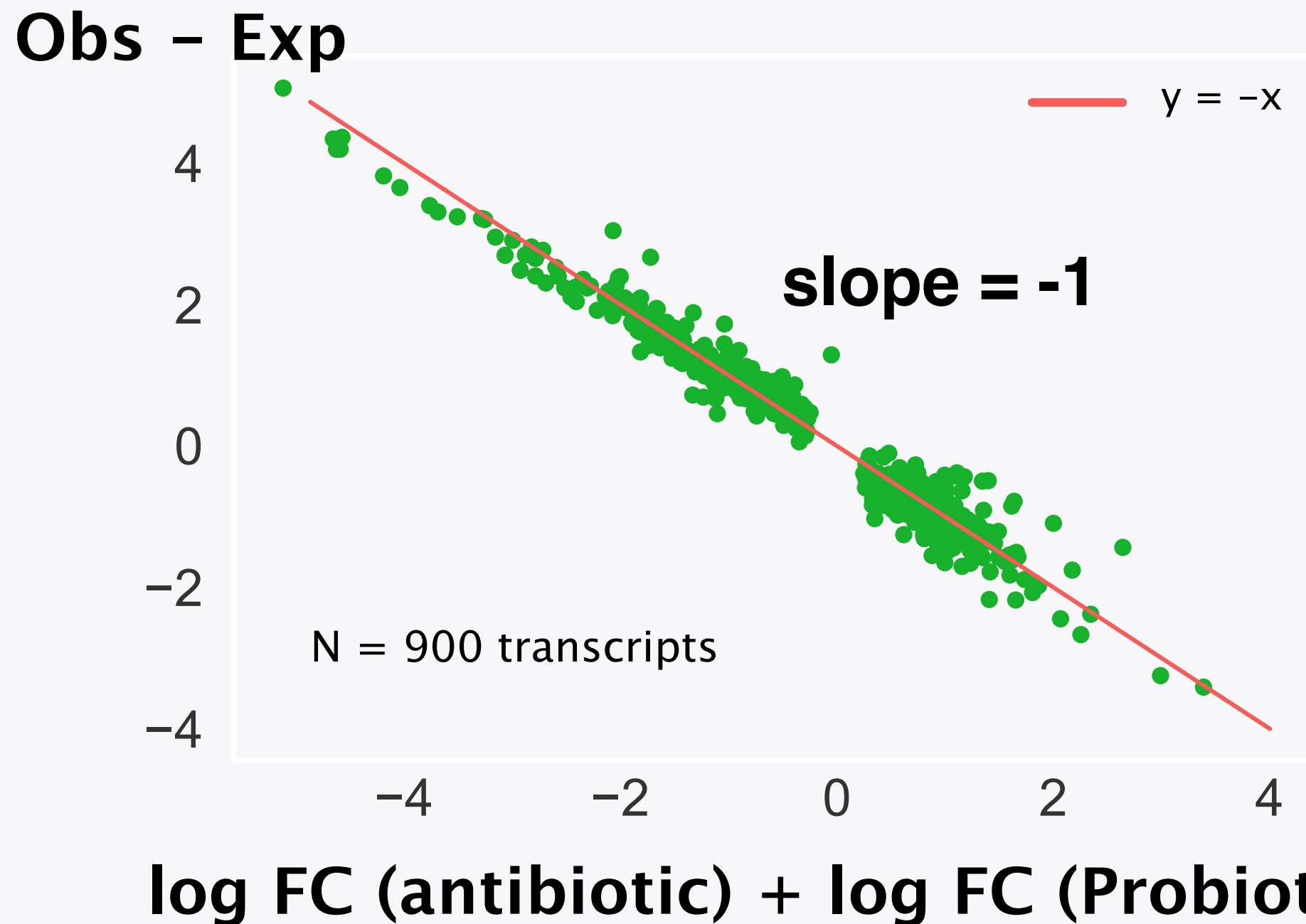
Transcriptomes are phenotypes in other organisms, such as bacteria!

Fuqing's Question:

Do probiotics affect antibiotic response in a gut bacterium?

+/- Probiotic
+/- Antibiotic

A slope of -1 indicates complete inhibition of the effect of antibiotics by probiotics for a subset of genes



Transcriptomes are phenotypes in other organisms, such as bacteria!

Probiotic  **Antibiotic**

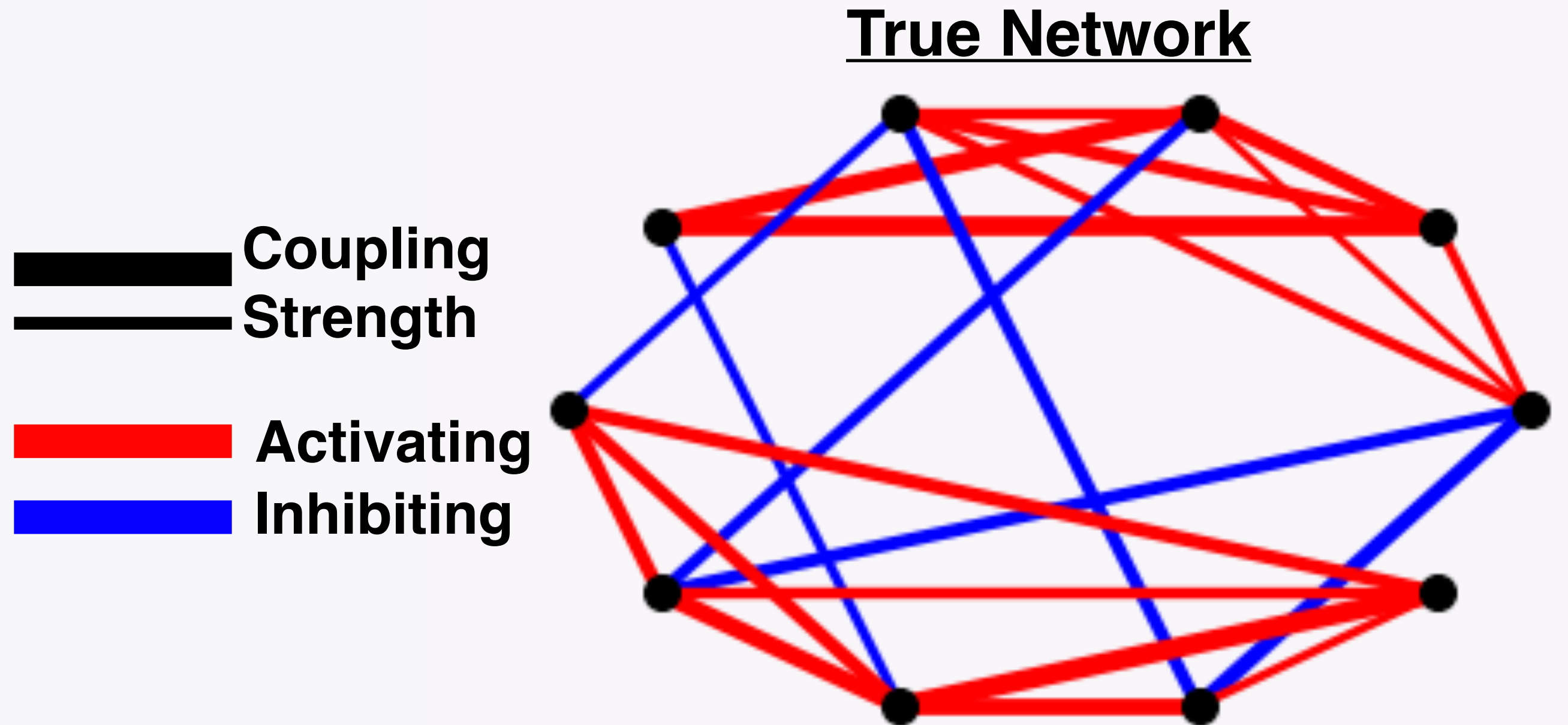
Transcriptomes as phenotypes: The geneticist's new arsenal

Null mutants (**Transcriptome-wide Epistasis**)

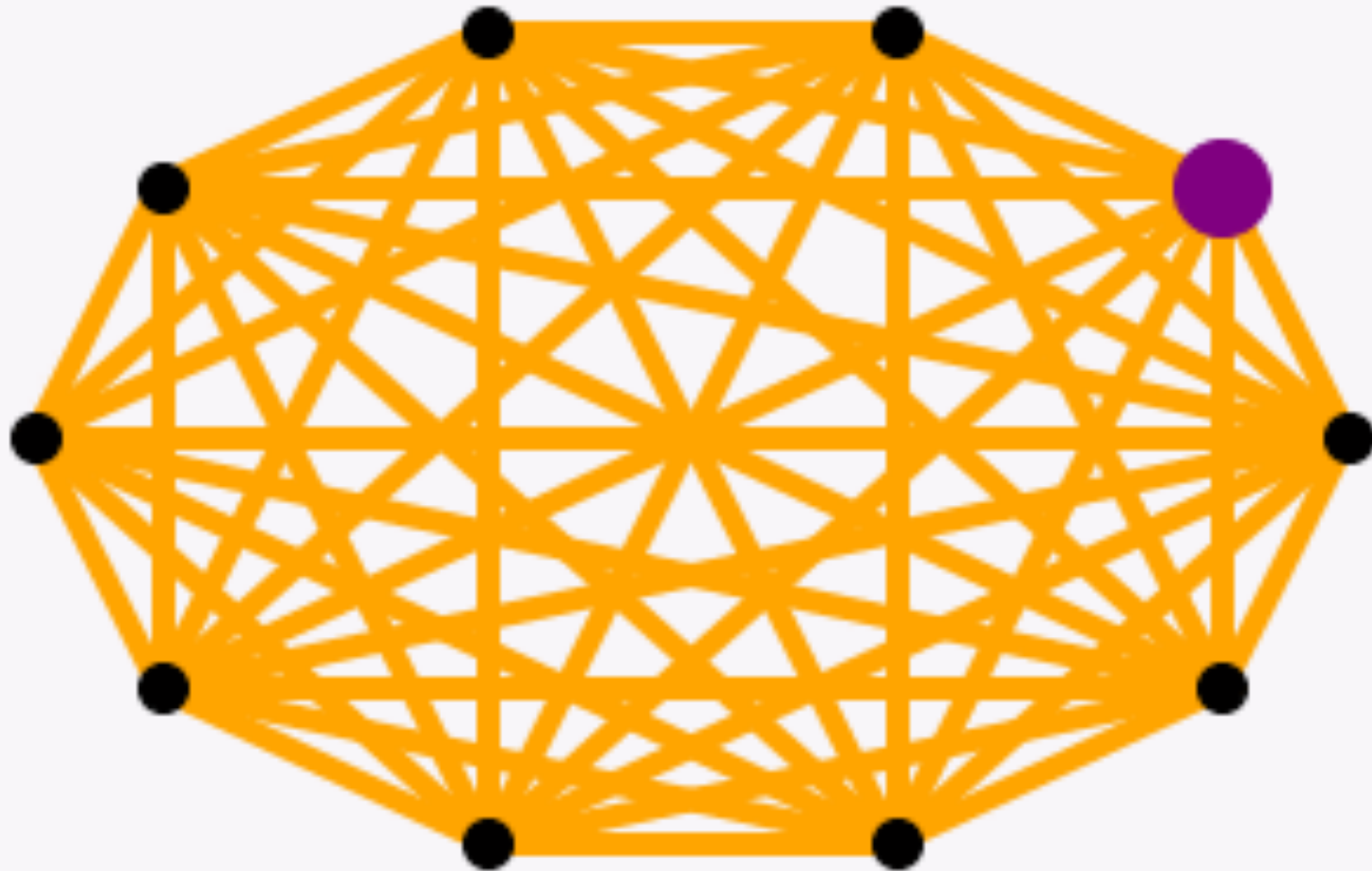
Allelic series (**Transcriptome-wide dominance**)

Crosses (**Transcriptome-wide maternal effects**)

Epistasis analyses can be automated

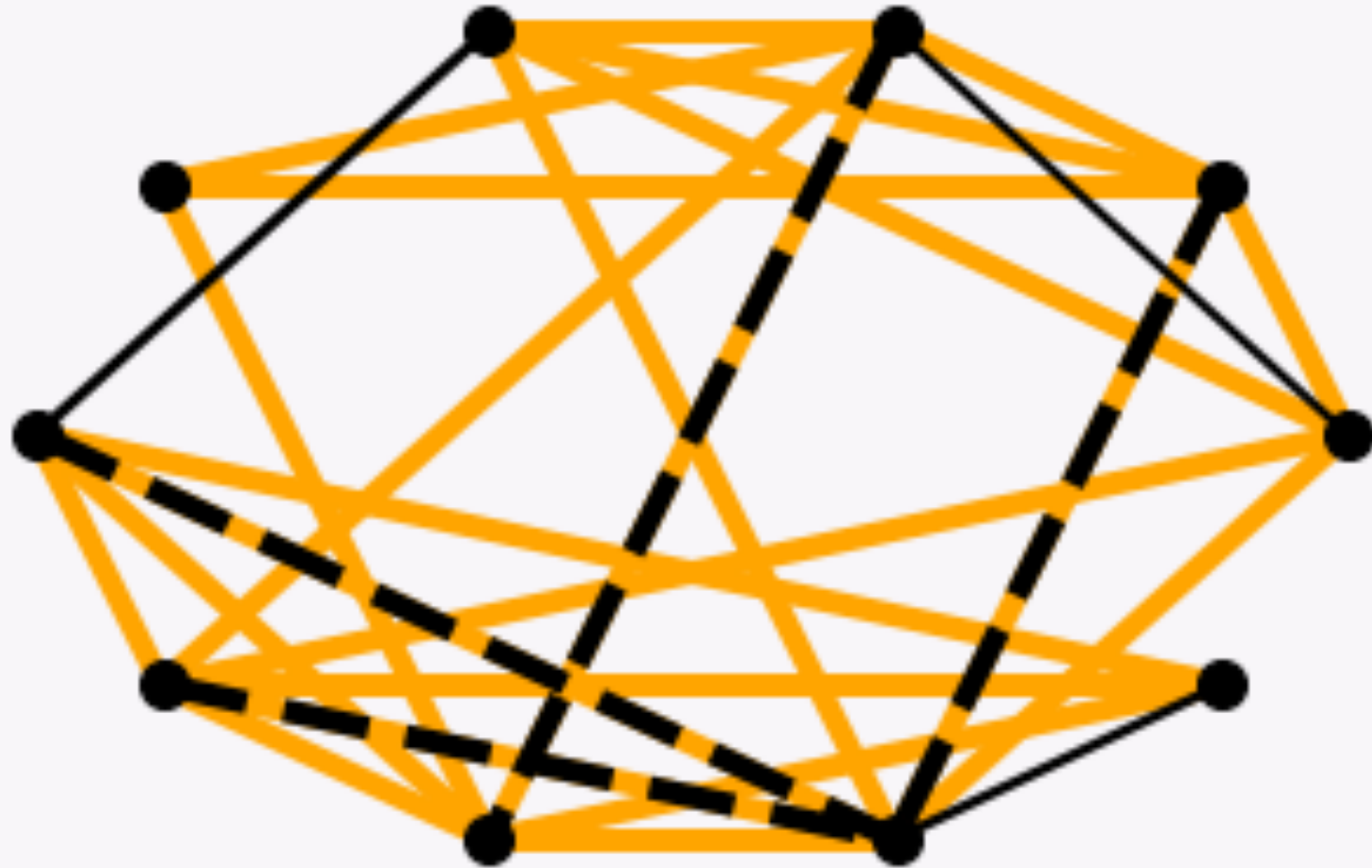





An example of automated reconstruction



Mutated Gene (Null)

Reconstructed network structure (no valences!)



-  **Real edges**
-  **Missing edges (smaller = weaker)**
-  **Extra edges (should not be there)**

Transcriptomes are phenotypes

Deploying transcriptomes in a **rich experimental context**
makes them powerful

We developed **statistical and conceptual machinery**
to use transcriptomes productively

Transcriptomes are Phenotypes!

Paul Sternberg



Howard Hughes
Medical Institute

Sternberg Lab

Carmie Puckett Robinson

Daniel Leighton

Tiffany Khaw

Tiffany Tsou

Hillel Schwartz

Barbara Wold

Brian Williams

Matt Thomson



**Millard and Muriel Jacobs
Genetics and Genetics Lab**

Igor Antoshechkin

Vijaya Kumar

Erich Schwarz

