

# Transcriptomes as phenotypes

Bringing Genetics to Genomics

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*Online Slides Available at [dangeles.github.io](https://dangeles.github.io)*

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# The geneticist's arsenal

Null mutants (**Epistasis**)

Allelic series (**dominance**)

Crosses (**maternal effects**)

# Genetics orders genes along pathways

Genes

Phenotypes

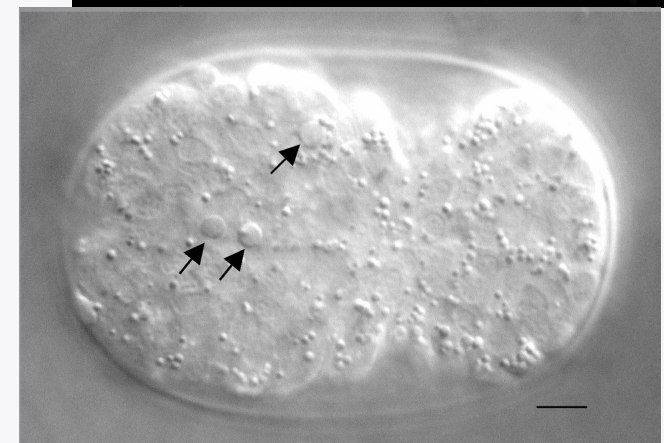
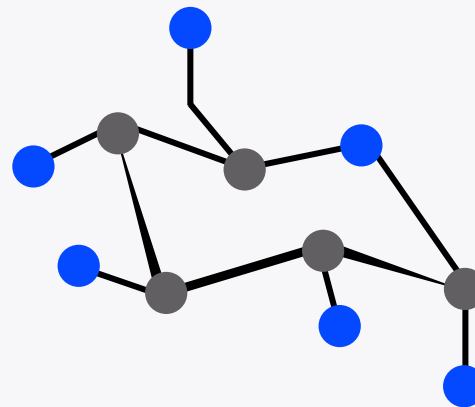
*a*

*b*

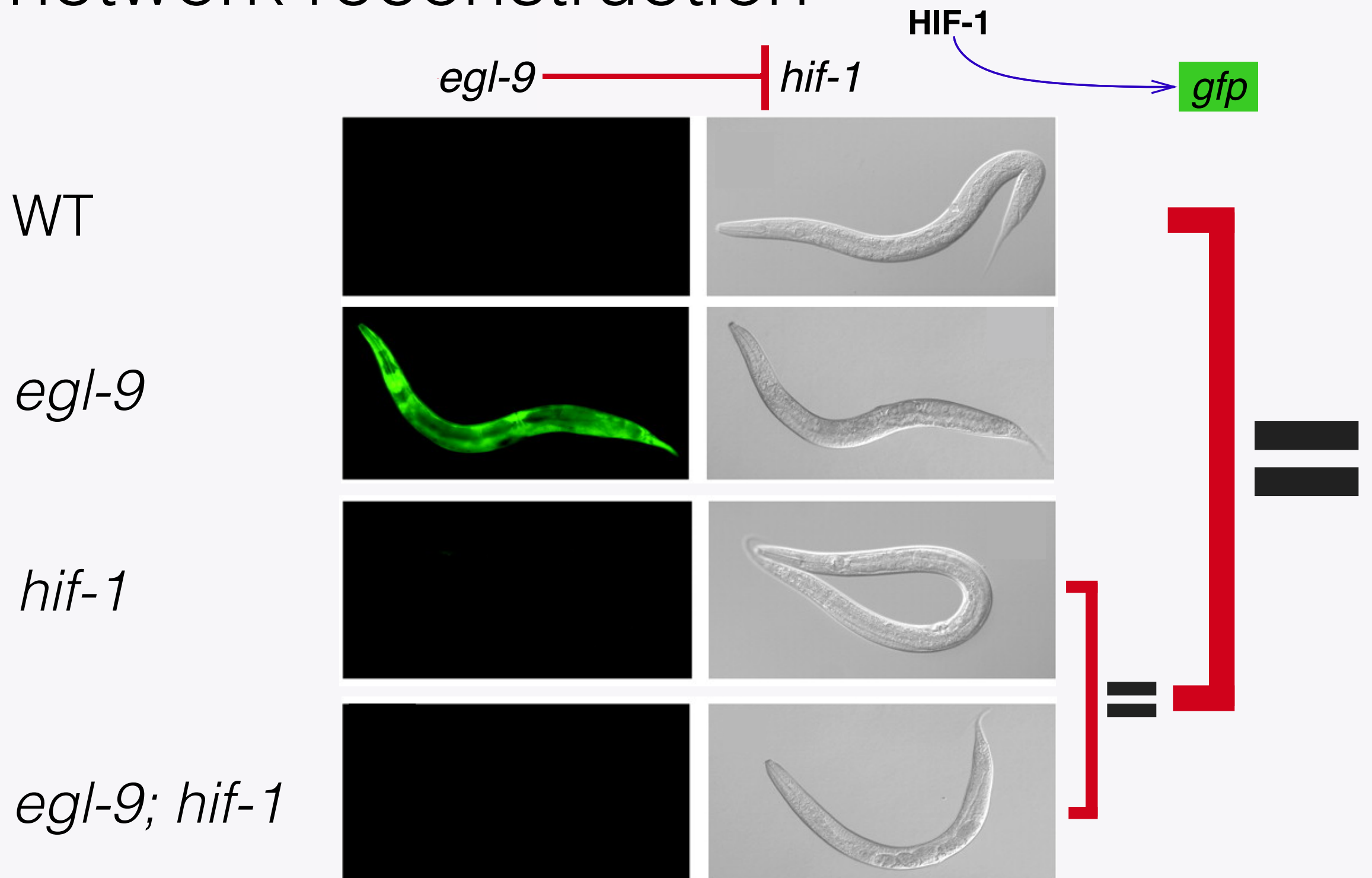
*c*

*d*

*e*



# Batesonian epistasis is a powerful method for network reconstruction



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# 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

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

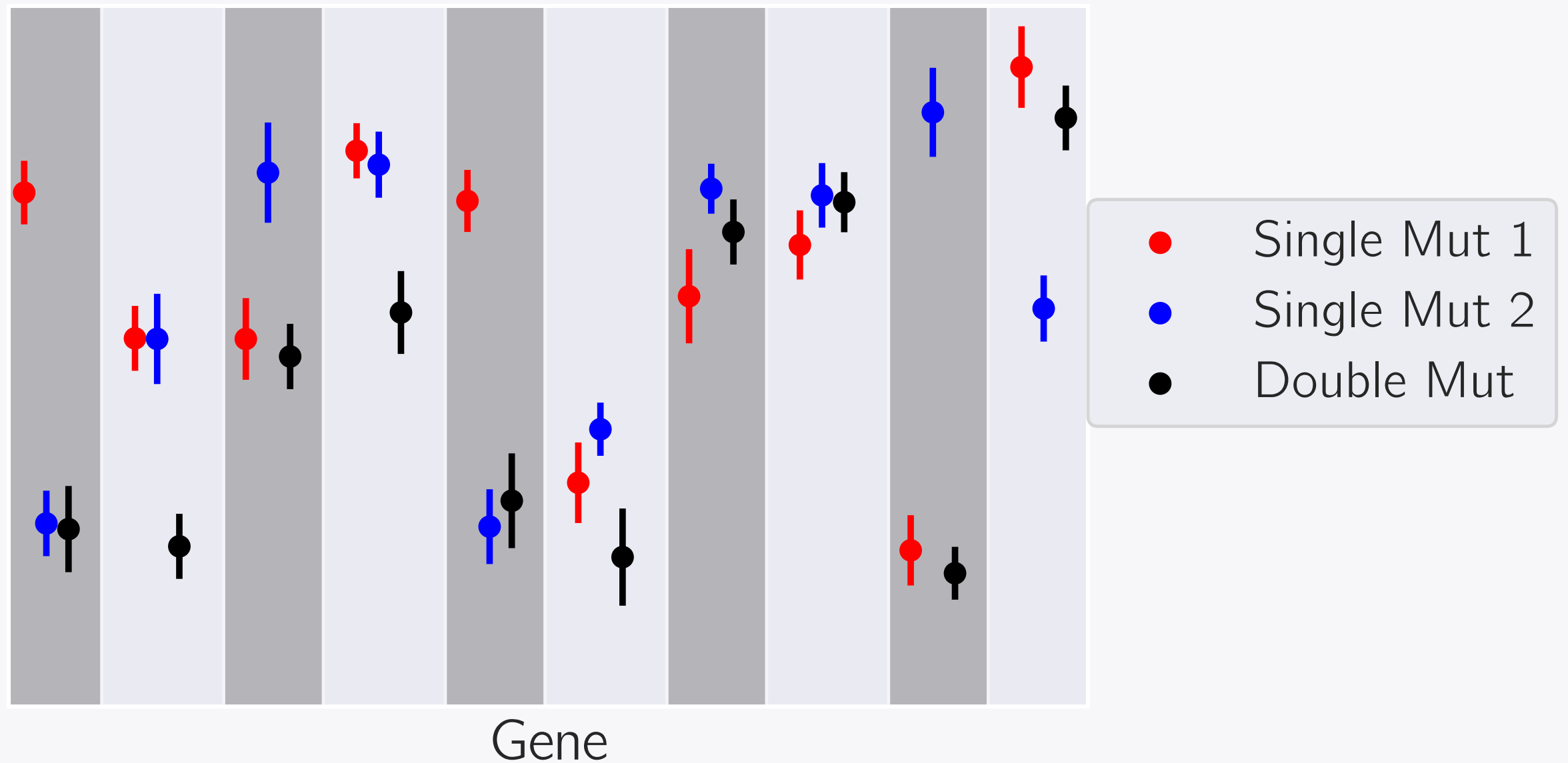
Genome-wide

Quantitative

Unbiased

# Transcriptomes are powerful, but complicated

log Fold Change of Expression



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

**For details, see:**

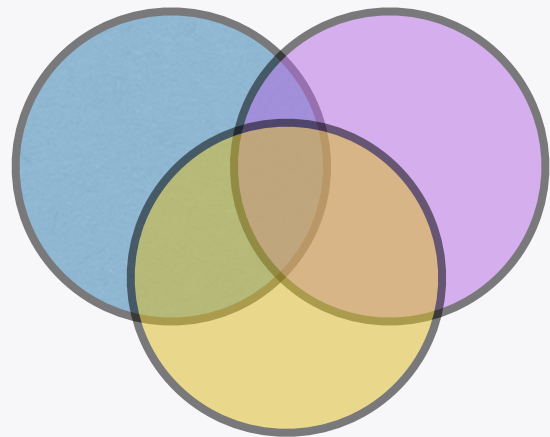
- **Epistasis:** Angeles-Albores *et al*, *PNAS*, 2018;  
Angeles-Albores *et al*, *G3*, 2017
- **Dominance:** Angeles-Albores, *Genetics*, 2018



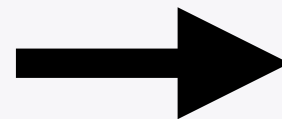
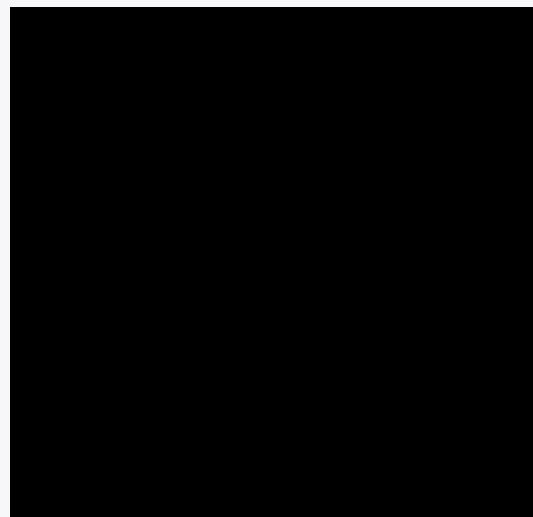
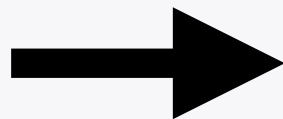
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# Transcriptome-wide epistasis analysis in a nutshell:

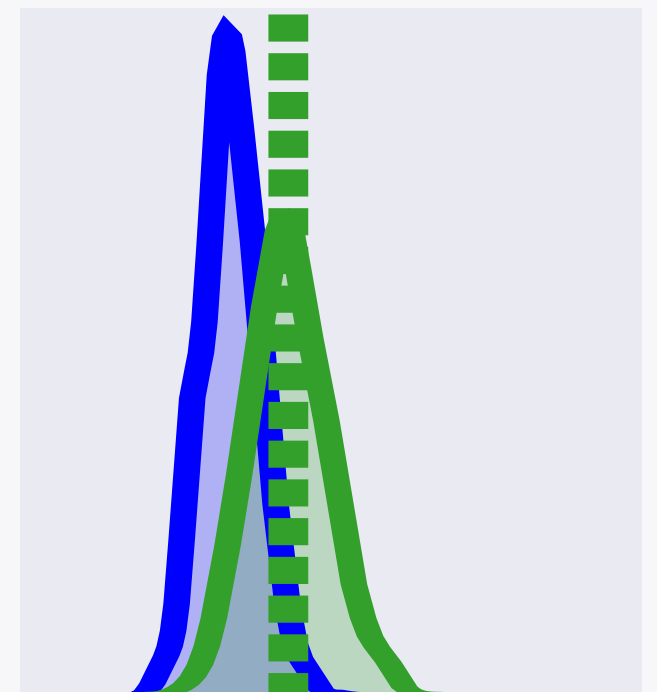
Choose  
phenotype



Compute a statistic for  
all genes in phenotype

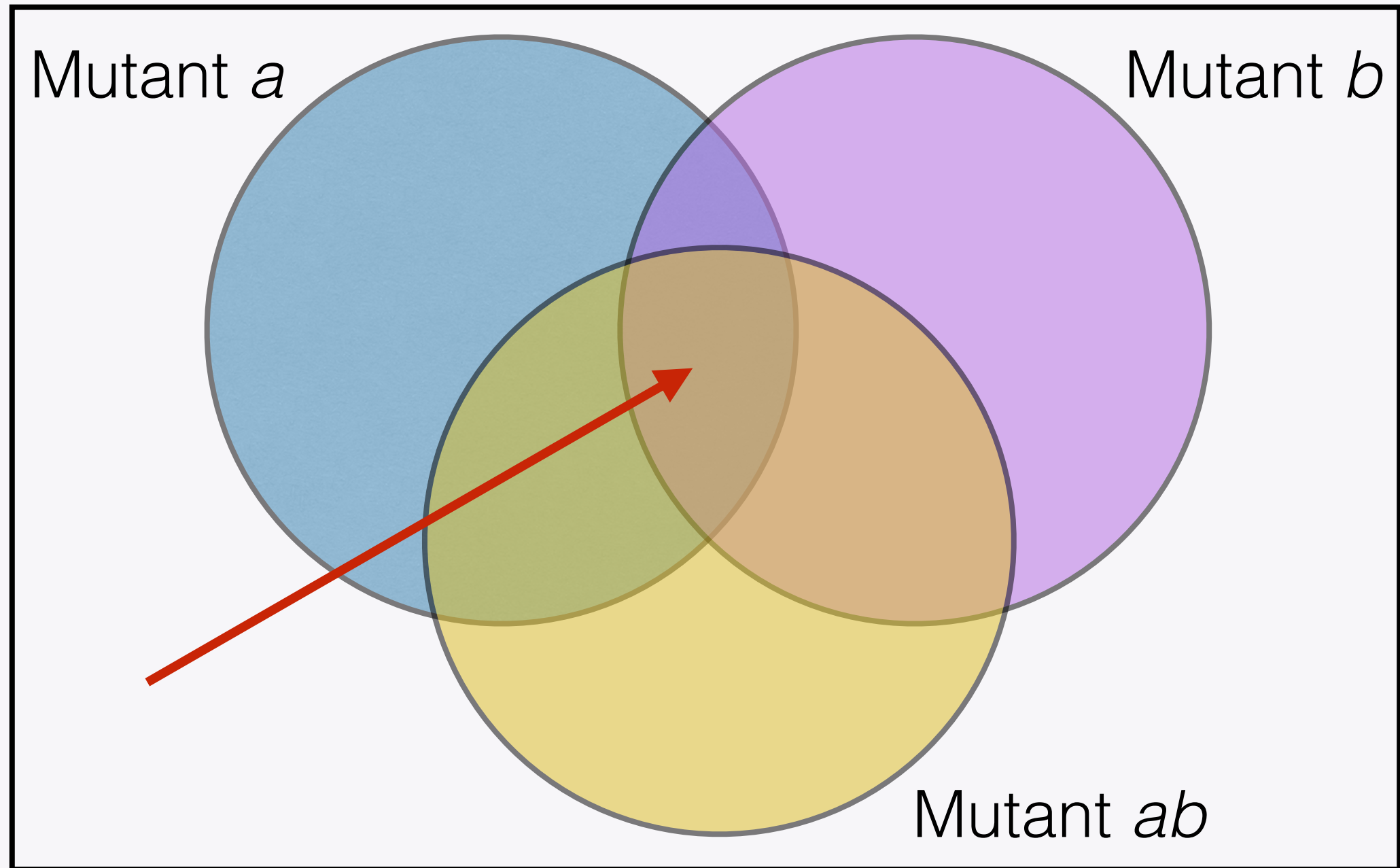


Check if statistic  
is Batesonian



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# Transcriptome-wide epistasis: Defining a phenotype

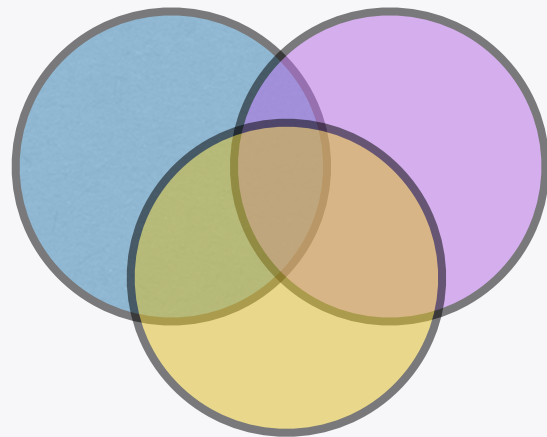


Diff. Exp. Genes relative to WT

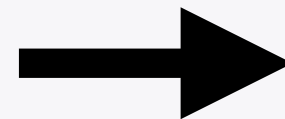
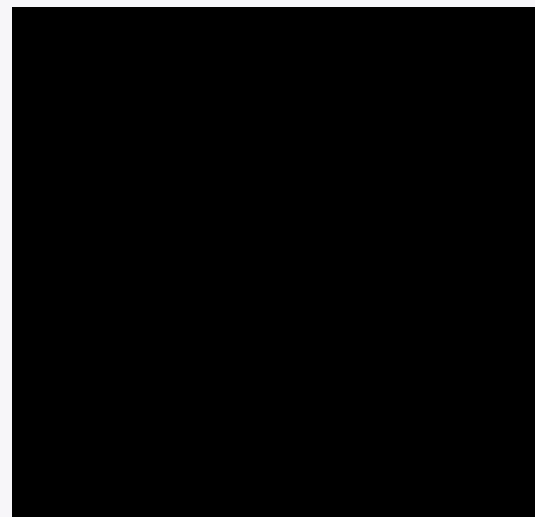
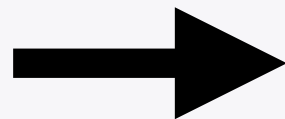
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# Transcriptome-wide epistasis analysis in a nutshell:

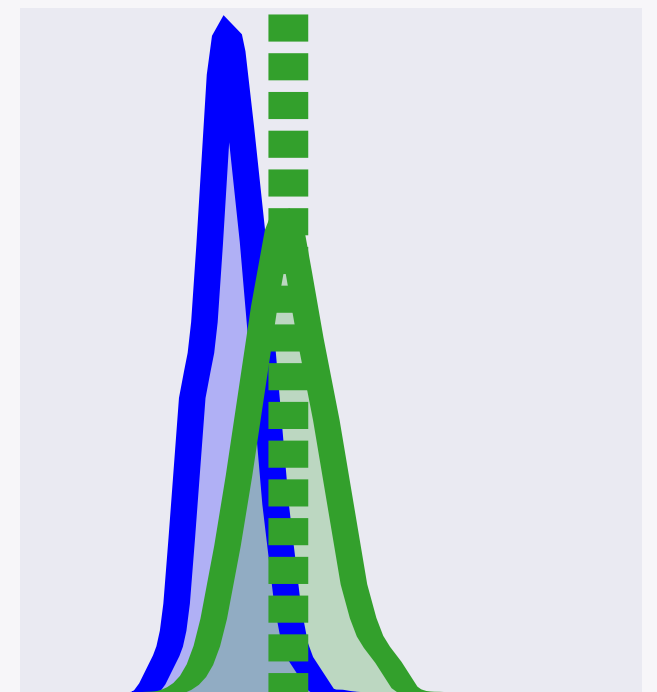
Choose  
phenotype



Compute a statistic for  
all genes in phenotype



Check if statistic  
is Batesonian



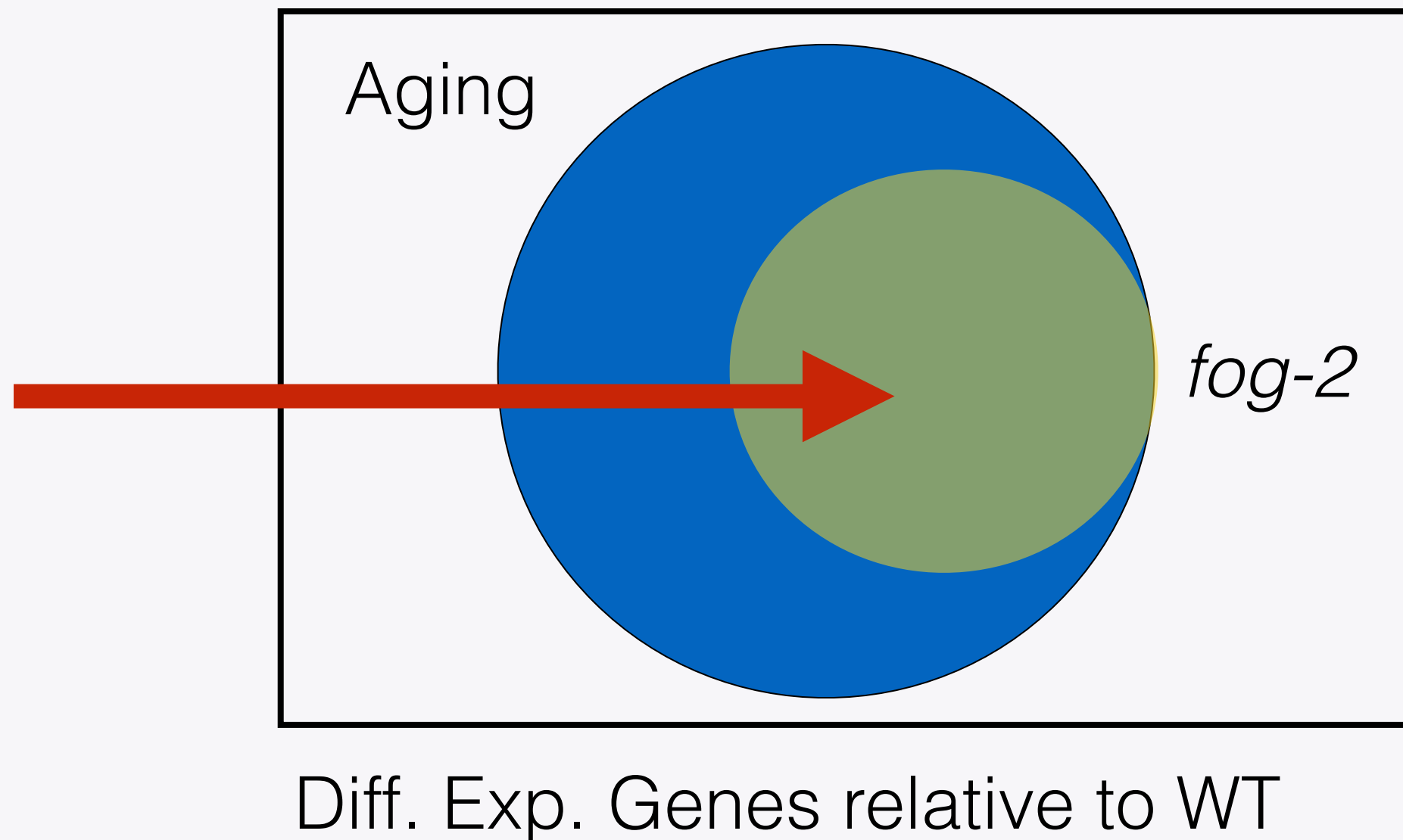
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An example: Does sperm status have effects independent of aging?

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

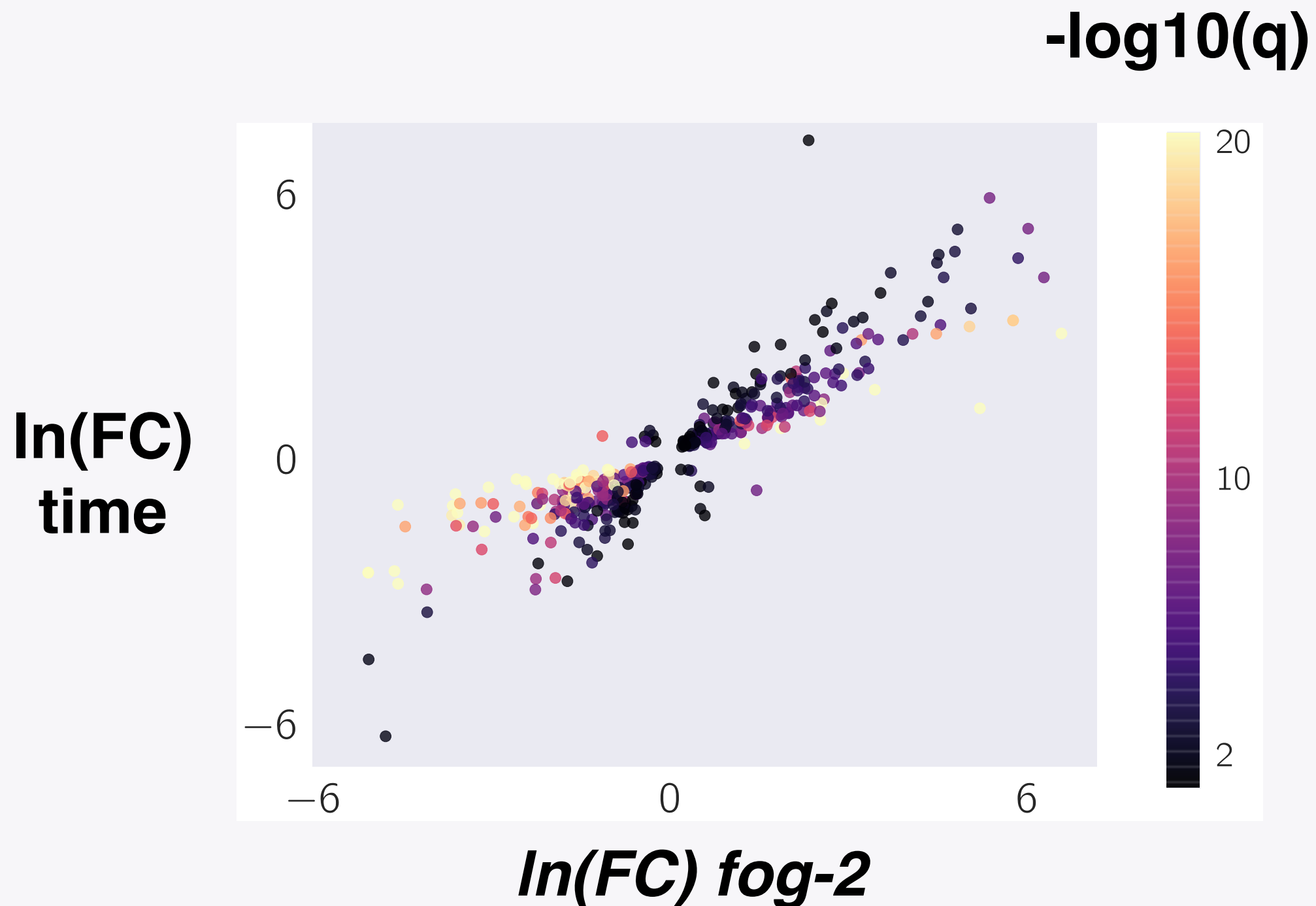
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Age affects more genes than *fog-2*, so we find the commonly affected subset



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First, show that both perturbations  
have equivalent effects



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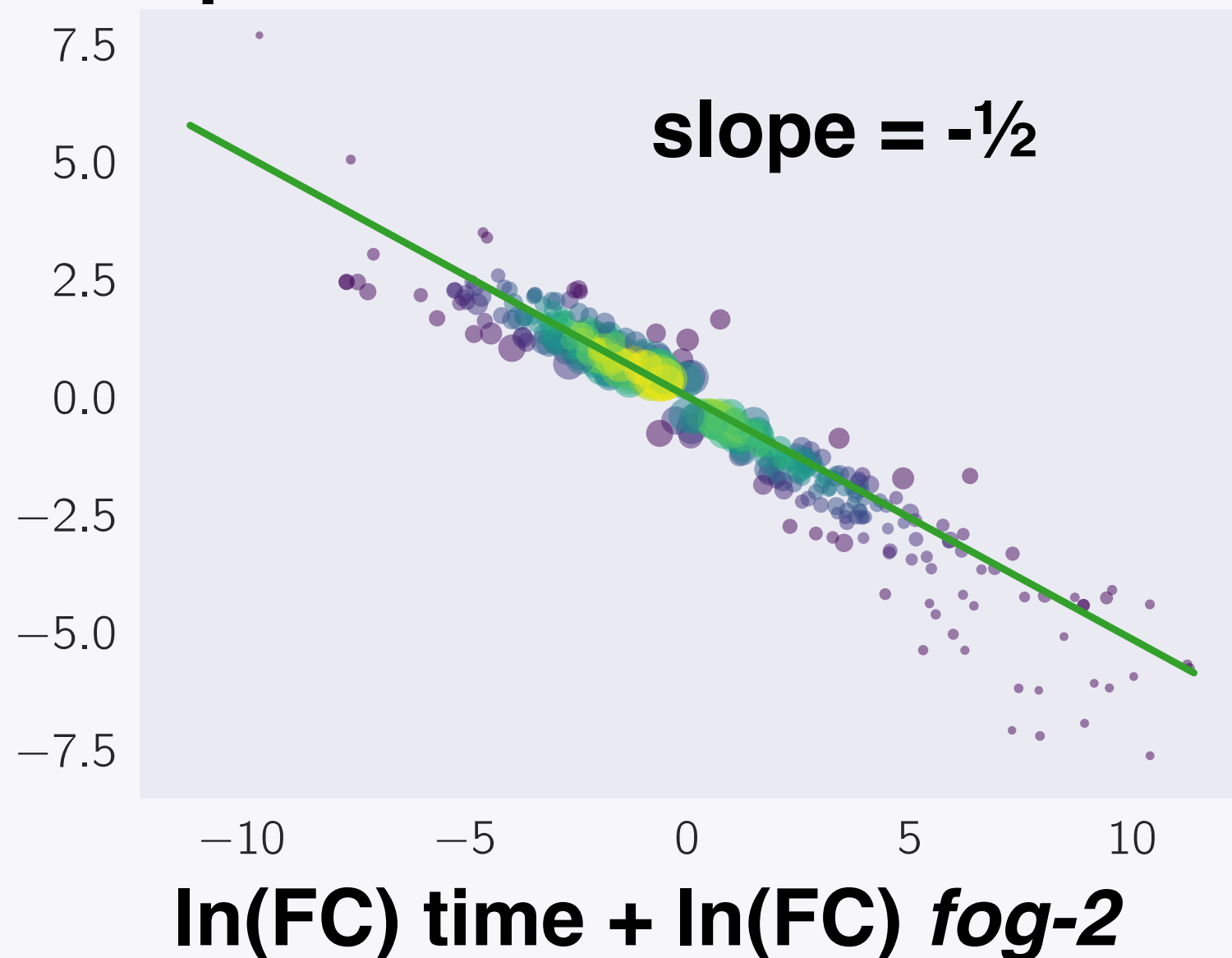
# What does our black box do?

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

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

**Observed - Expected**



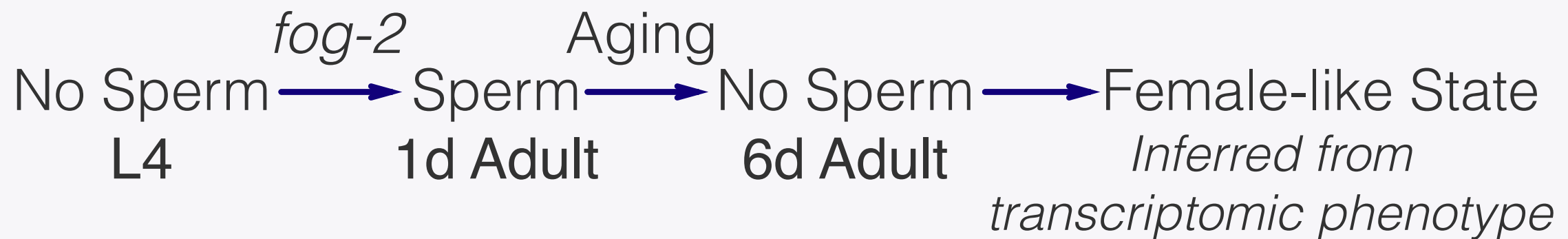
Behind the math:

**Observed**  
=  
**ln(FC) time**  
=  
**ln(FC) fog-2**



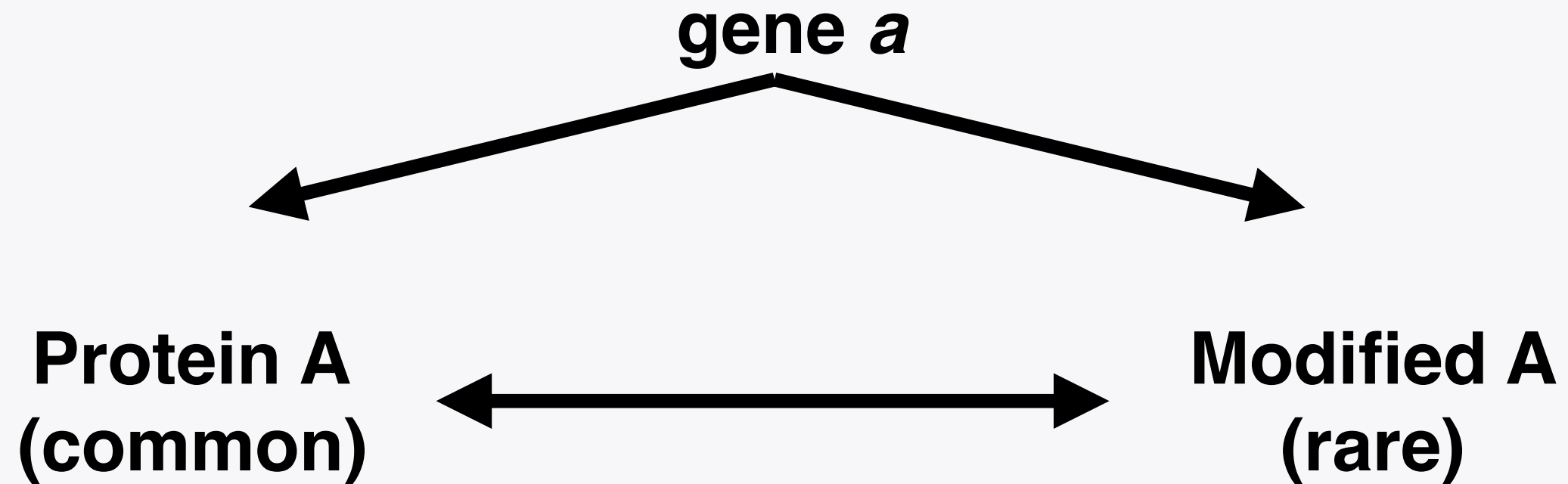
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The *C. elegans* female state was  
inferred from transcriptome profiling



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# Transcriptomes can be used to think about biochemistry

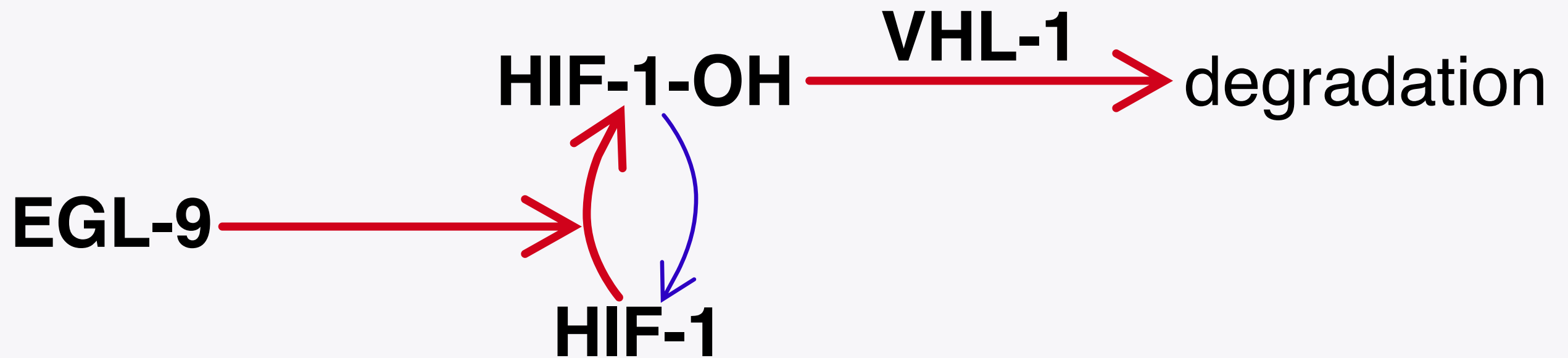


Accounts for most effects  
of knocking out ***a***

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

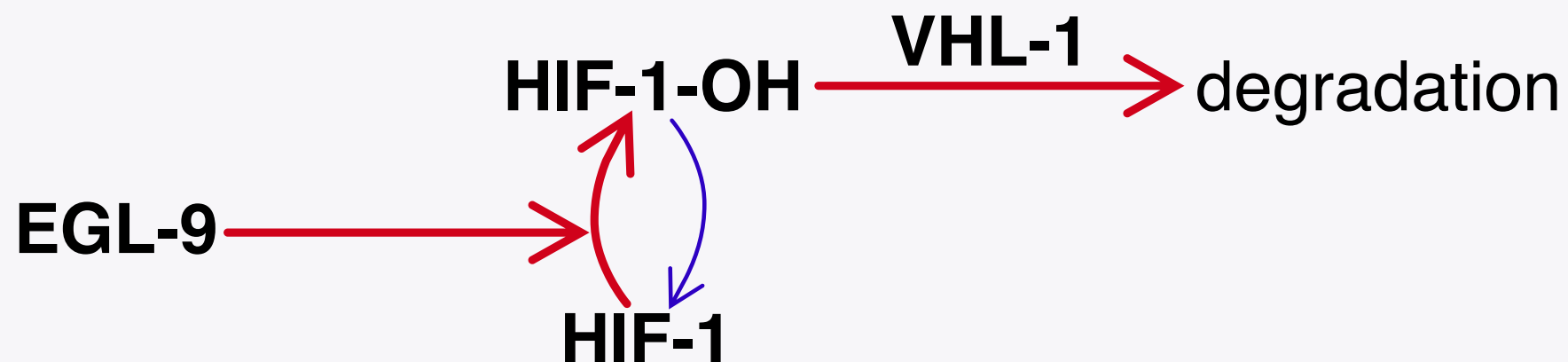
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Hypoxia factor 1 (*hif-1*) is degraded by VHL-1 in an EGL-9 dependent manner



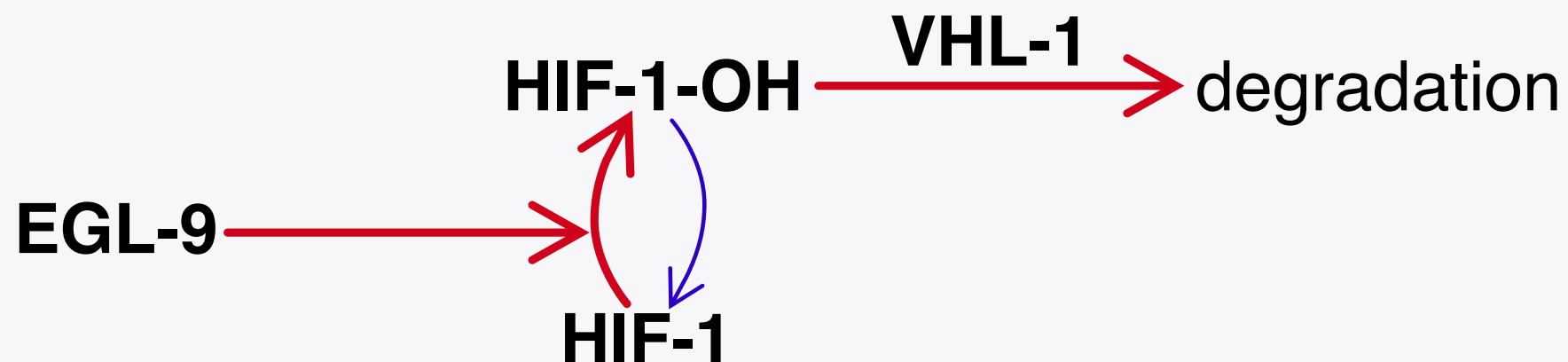
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Using HIF-1 abundance as phenotype leads to the canonical genetic diagram:



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



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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*

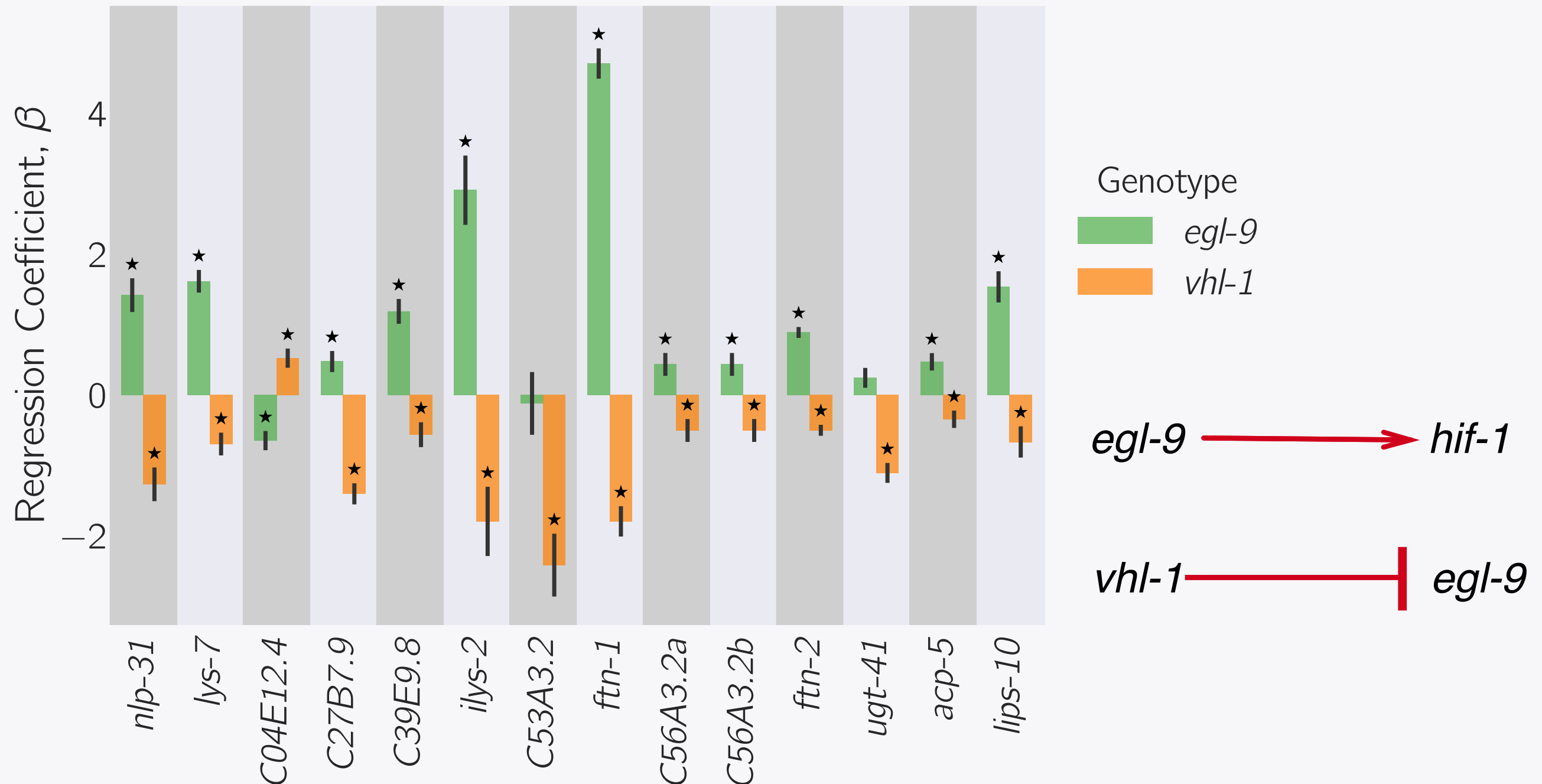
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However, both pathways obey the same set of epistatic rules!

***egl-9 = egl-9;vhl-1***

***hif-1 = egl-9;hif-1***

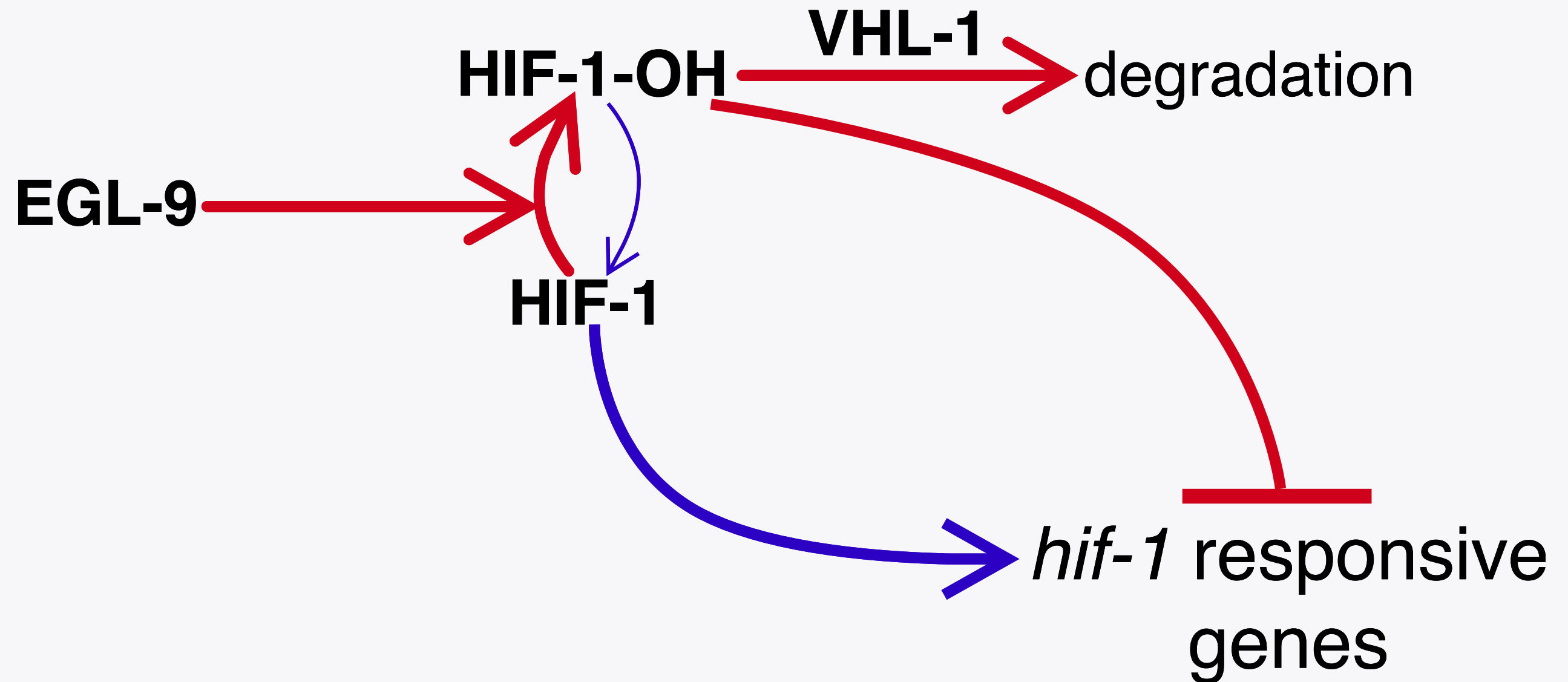
# Sequencing hypoxia pathway mutants reveal ~50 genes that behave as if controlled by HIF-1-OH





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



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

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Transcriptomes are phenotypes in other organisms, such as bacteria!

Fuqing's Question:

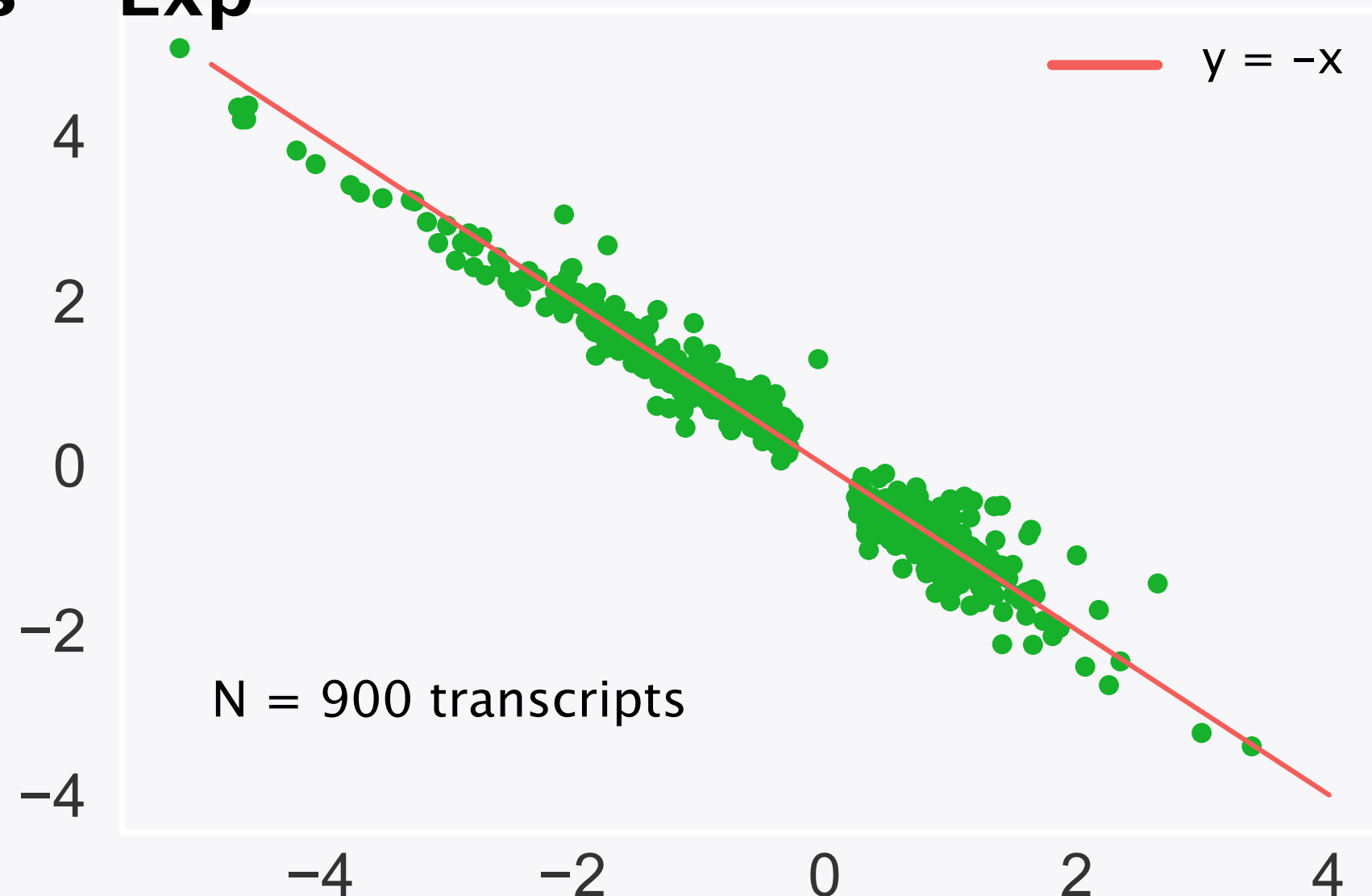
Do prebiotics affect antibiotic response in a gut bacterium?

**+/- Prebiotic**  
**+/- Antibiotic**

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

**Obs - Exp**



**$\log FC (\text{antibiotic}) + \log FC (\text{Prebiotic})$**

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Transcriptomes are phenotypes in other organisms, such as bacteria!

**Prebiotic** ————— **Antibiotic**

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# Transcriptomes as phenotypes: The geneticist's new arsenal

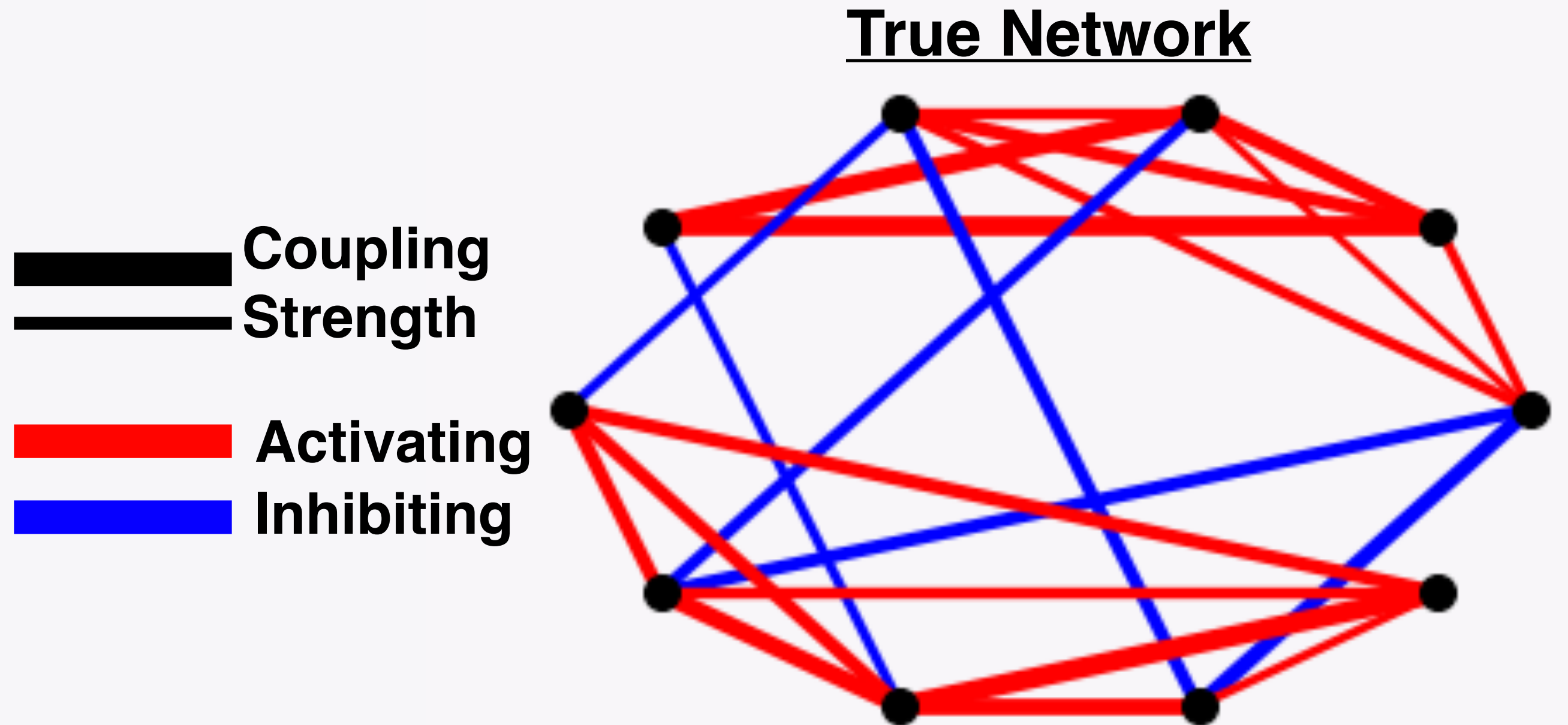
Null mutants (**Transcriptome-wide Epistasis**)

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

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

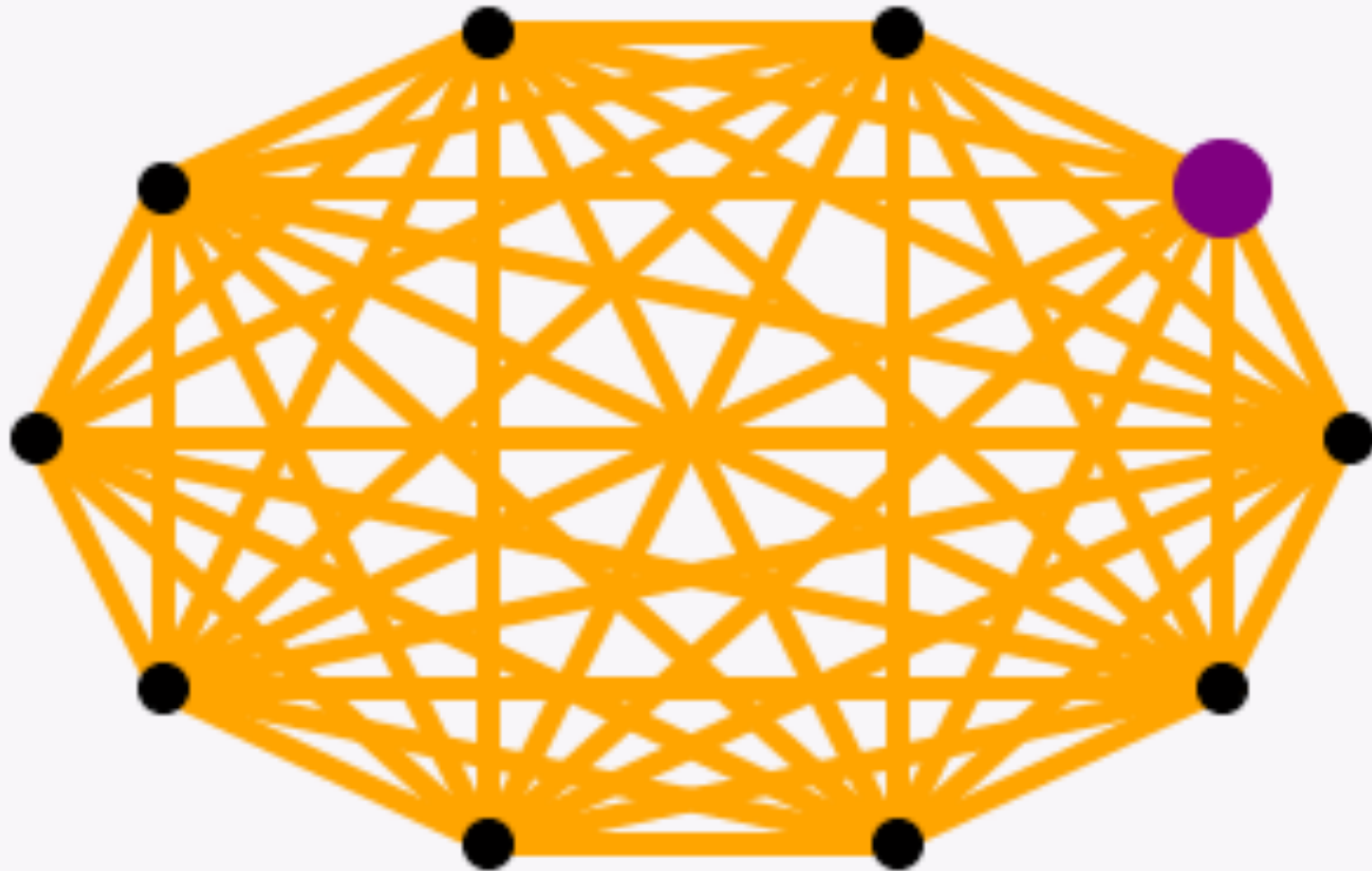
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# Epistasis analyses can be automated



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# An example of automated reconstruction



**Mutated Gene (Null)**



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# Reconstructed network structure (no valences!)



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

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# Transcriptomes are phenotypes

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

We now have both **statistical and conceptual machinery** to use transcriptomes productively

# Transcriptomes are Phenotypes!

**Paul Sternberg**



## **Sternberg Lab**

Carmie Puckett Robinson

Daniel Leighton

Tiffany Khaw

Tiffany Tsou

Hillel Schwartz

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

Igor Antoshechkin

Vijaya Kumar

**Erich Schwarz**

## **Barbara Wold**

Brian Williams

## **Matt Thomson**

