

Why EvoSysBio?

Quantify the 5 fundamental factors of evolution by
integrating rigorous models from across biology

Intra-Organism Biology

Rigorous models of
molecules ... in
organisms

*molecular systems
biology, biochemistry,
cell biology, ...*



**Organisms:
Abstracted
Modeled
Nested
Tested**

Trans-Organism Biology

Rigorous models of
organisms ... in
ecosystems

*evolutionary biology,
population genetics,
ecology, ...*



Better predictions of evolution
based on mechanistic insights

=

Evolutionary Systems Biology

Evolutionary Systems Biology *IS*

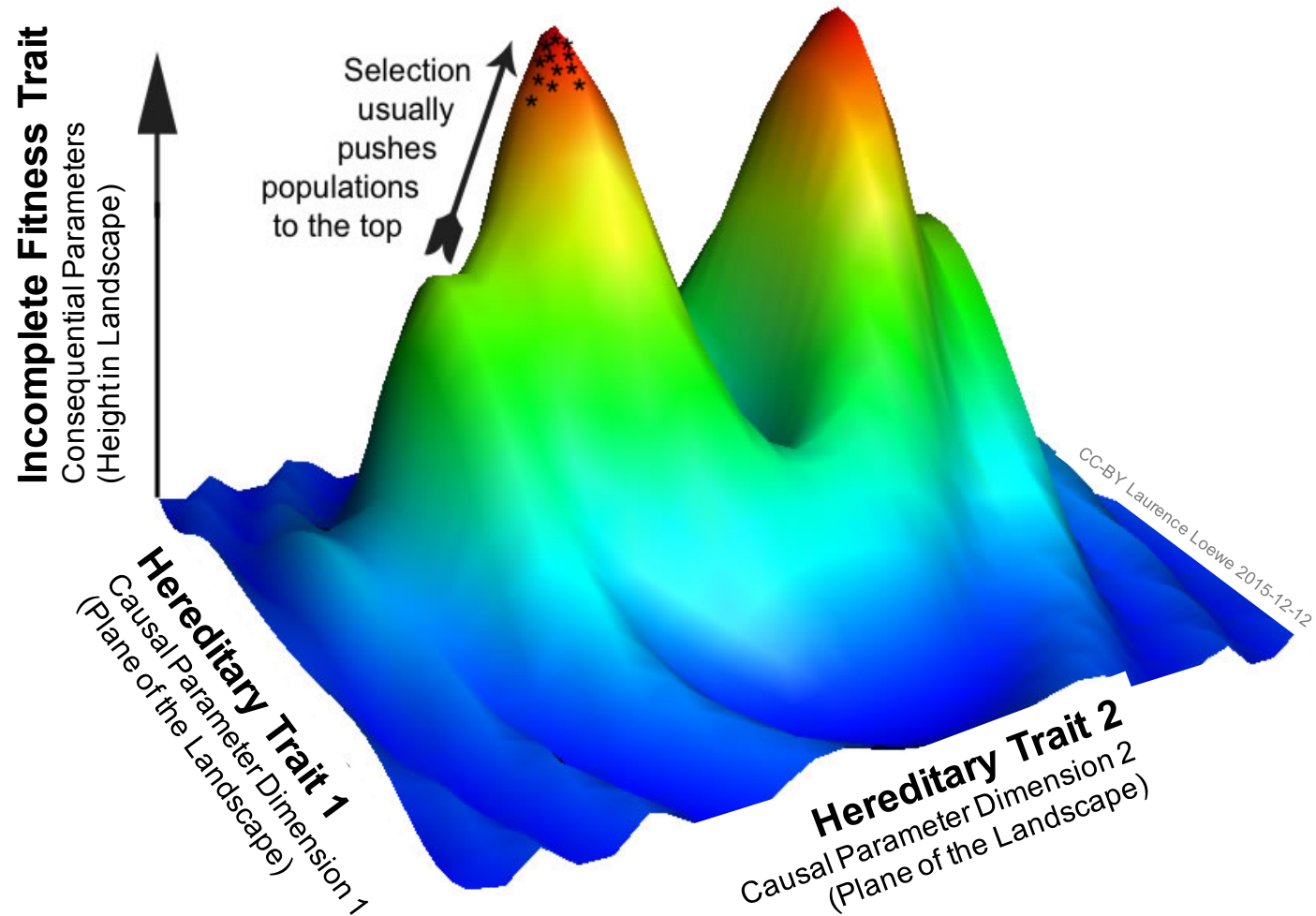
a trans-disciplinary **framework for constructing**
reliable, testable, interactive **overviews of**
nestable, dynamic, multi-dimensional

Fitness Landscapes

which mechanistically predict:

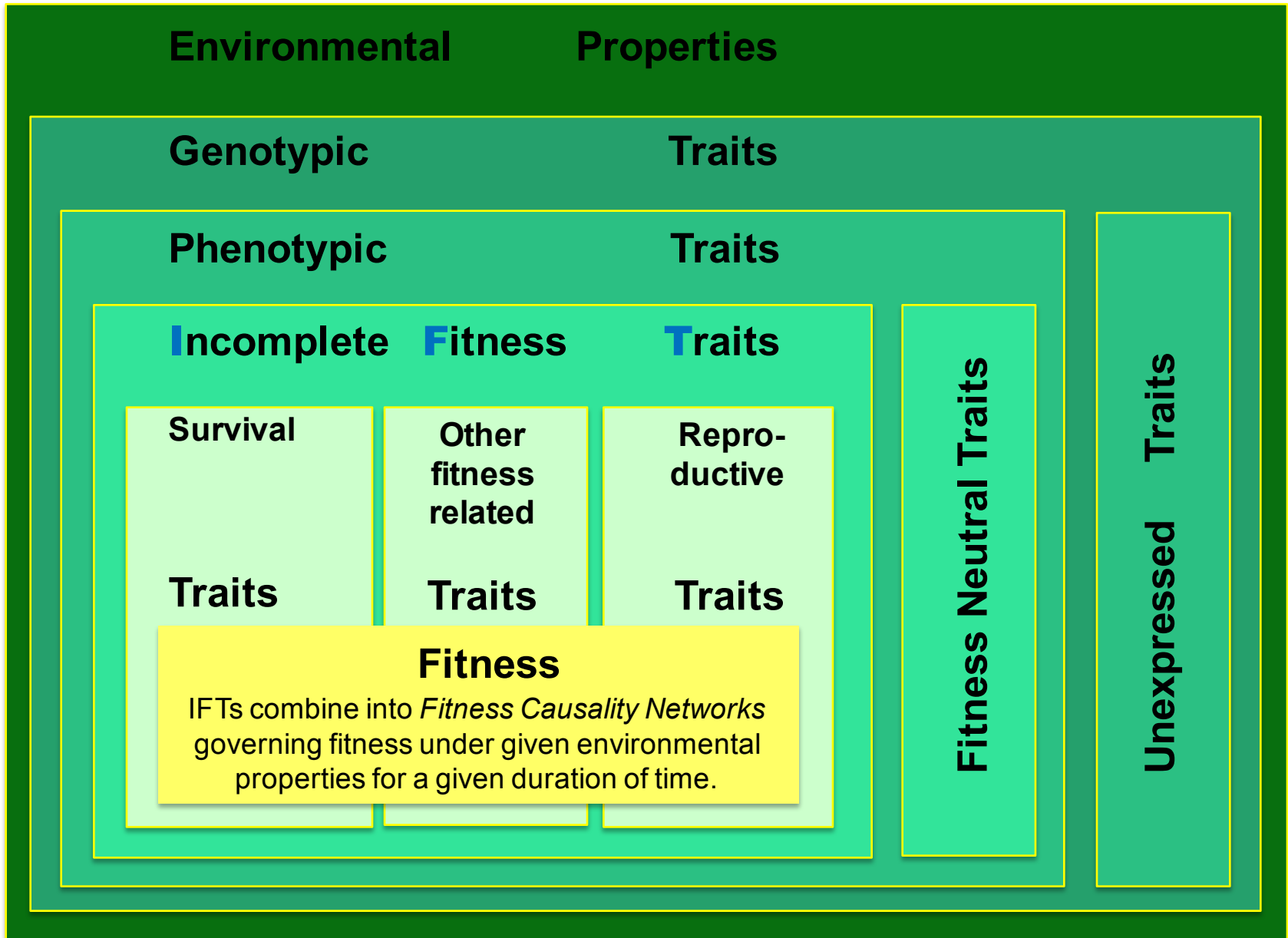
- ① **Changes in fitness of individual organisms**
when their states and environments change;
- ② **How populations evolve**
when organisms traverse fitness landscapes.

Massively Oversimplified Cartoonish Abstract Landscape of Incomplete Fitness Traits



Mostly MOCA-LIFTs **EvoSysBio's Journey** → Reliable LIFTs

Environments, Genes, Phenotypes, Chance → Fitness



IFTs of biological data in *Context*

Mean fitness of organisms in groups
(or other population statistics)

Organism survival, reproduction
(RIFTs network of tradeoffs)

Real-world Incomplete Fitness Traits
(RIFTs are reality calibrated SIFTs)

Simulated Incomplete Fitness Traits
(SIFTs are entirely computational)

Time series of phenotypic traits
(nested: ... in cells in tissues in ...)

Molecular functions network
(collection of interaction rates)

Molecular structures collection
(structures → govern functions)

Hereditary materials, genetic code, ...
(sequences → govern structures)

LIFTs are Methods for Mapping IFT points in *Planes* to *Heights*

7: **Summarizing** stats of relevant groups
in specified environments and times

6: **Balancing** trade-offs in IFT networks
of organism life history and physiology

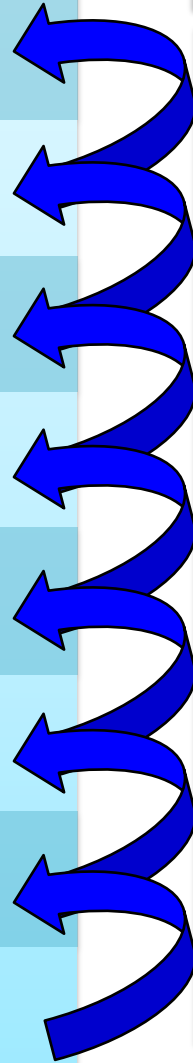
5: **Mapping** *in silico* IFTs to
relevant observed real-world IFTs

4: **Extracting** fitness-related traits (IFTs)
from analyses of time series

3: **Simulating** dynamic time series
of parts in systems biology networks

2: **Abstracting** the structural biology
of structure-function relationships

1: **Folding** of expressed sequences
into dynamic 3D structures



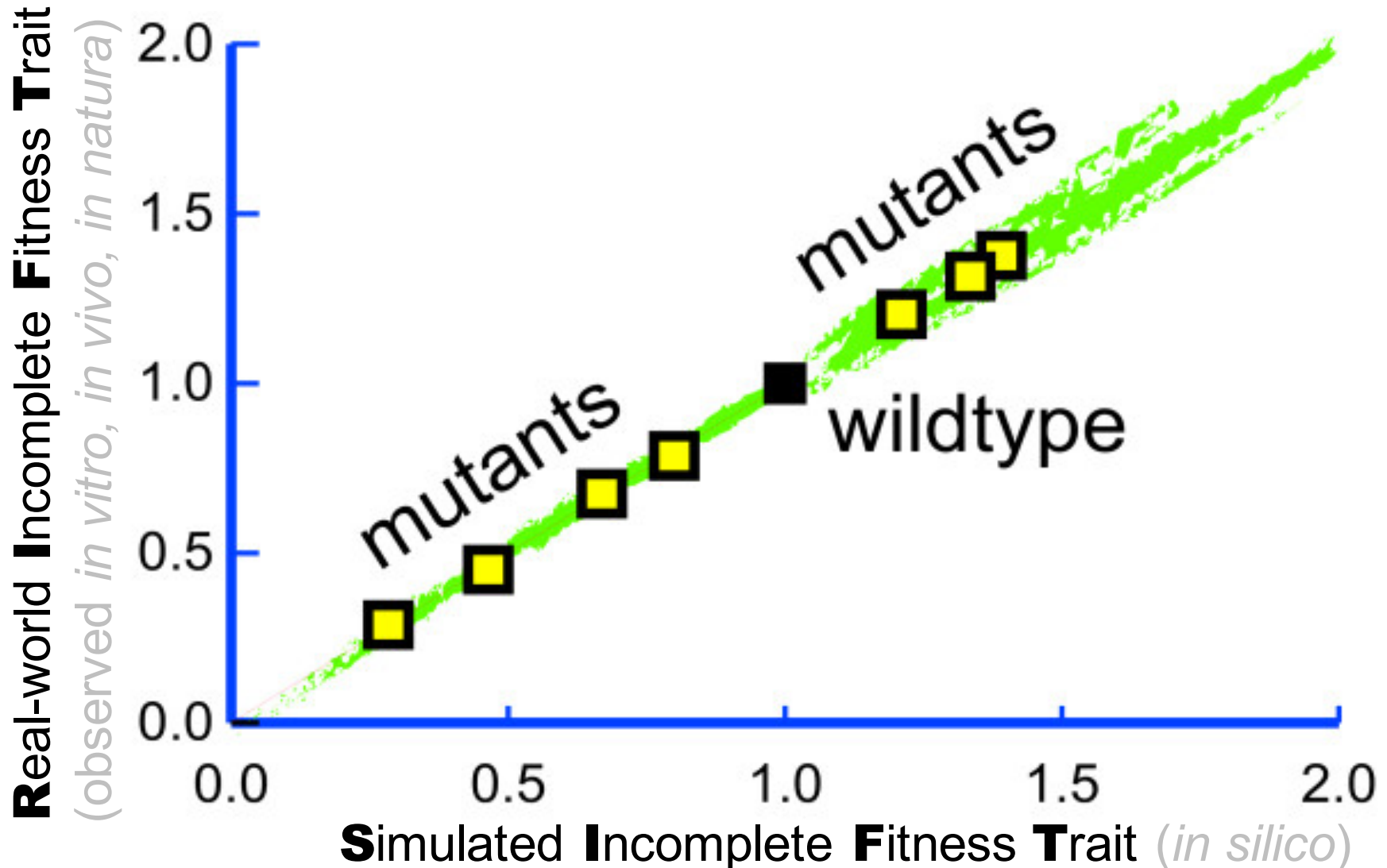
EvoSysBio in 10 Slides CC-BY Laurence Loewe 2015-12-12

Mechanistic Maps: Mutations to Phenotype Changes

compute from full Fitness Causality Networks of **L**andscapes of **I**ncomplete **F**itness **T**raits

Simulated Real-world Incomplete Fitness Trait Maps

Simulate SIFTs and measure RIFTs for the same wildtype. If linear extrapolations of SIFTs predict observed RIFTs, the SRIFT Map is usable. If not, statistical approximations may help with mapping.



Wildtypes instantiate **LIFTs** to **DMEs**

**Landscape of
Incomplete
Fitness
Traits**

Instantiate

Wildtype

Abstract

Distribution

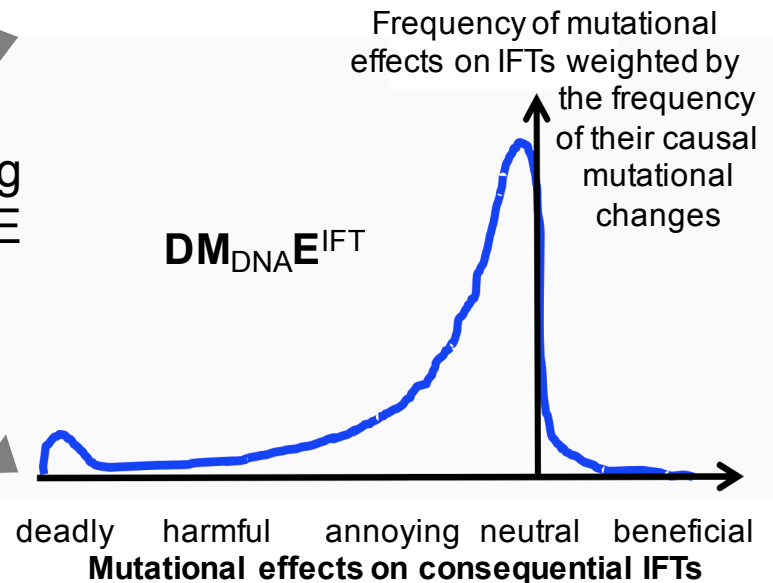
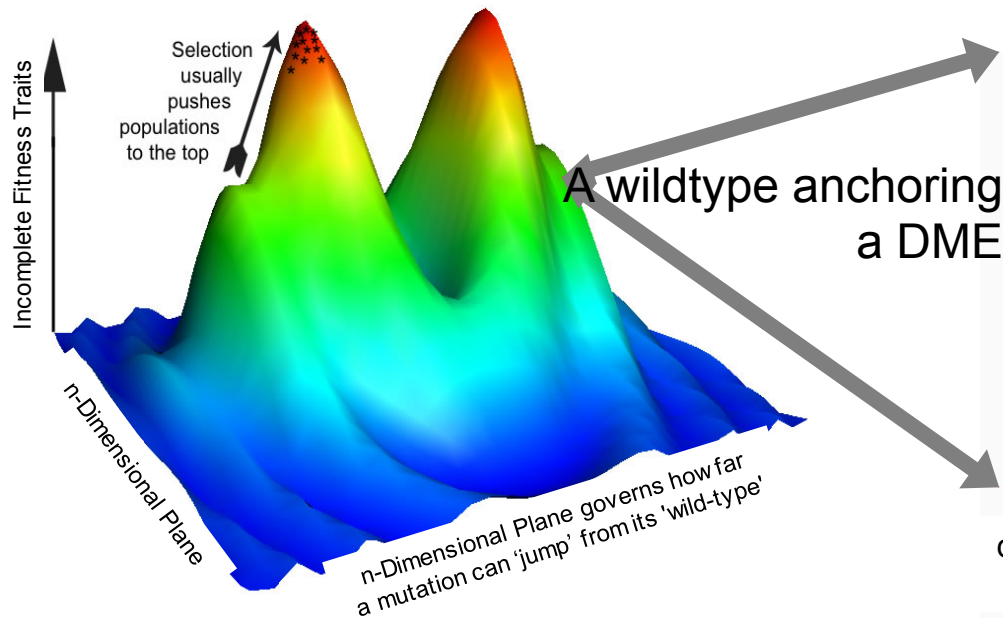
of increasing or decreasing

Mutational

changes in the causal
input parameter plane with

Effects

on consequential
Incomplete Fitness Traits



Estimation of Mutational Effects *in Silico*

Find: **D**istribution of **M**utational Effects of a given *MET* on a given *SIFT*, short:

$$\mathbf{DM}_{MET}\mathbf{E}^{SIFT}$$

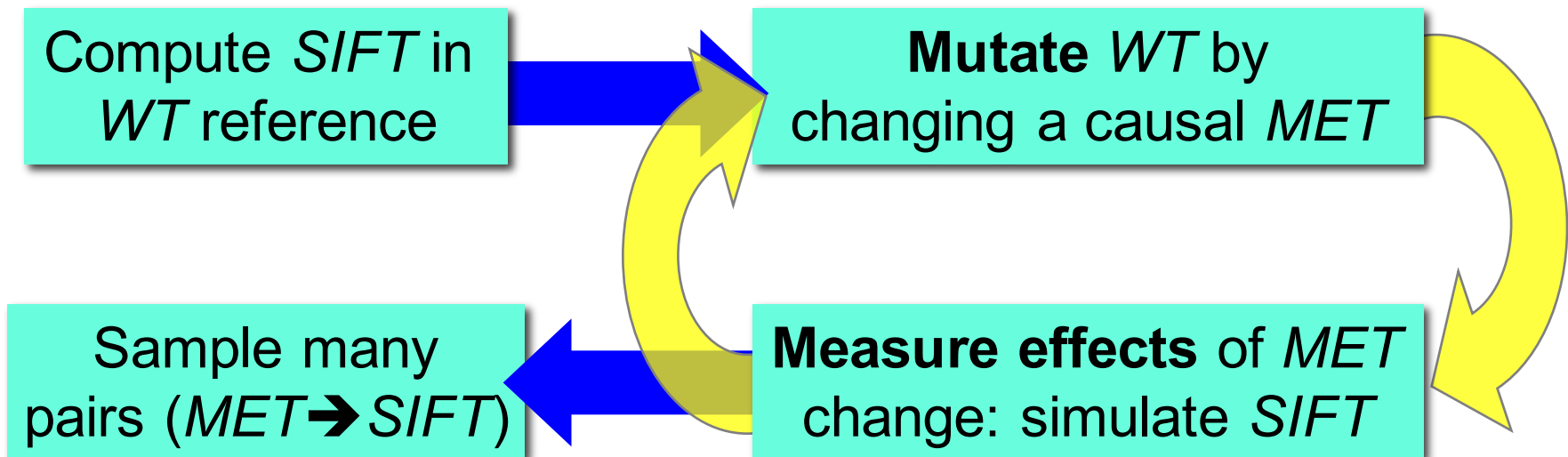
1. Define a *WT*: a reasonable systems biology model of a **WildType** system
2. Define a *SIFT*: a consequential **Simulated Incomplete Fitness Trait**

SIFTs are computable properties of a *WT* with a plausible effect on IFTs that are known to affect fitness in reality at least under some conditions. Use expert intuition to define candidates and expert judgment to test their quality. Examples: reproduction, survival, energy cost, gene regulation accuracy, speed, stability ...

3. Define a *MET*: a credible causal **Mutation Encoding Trait**

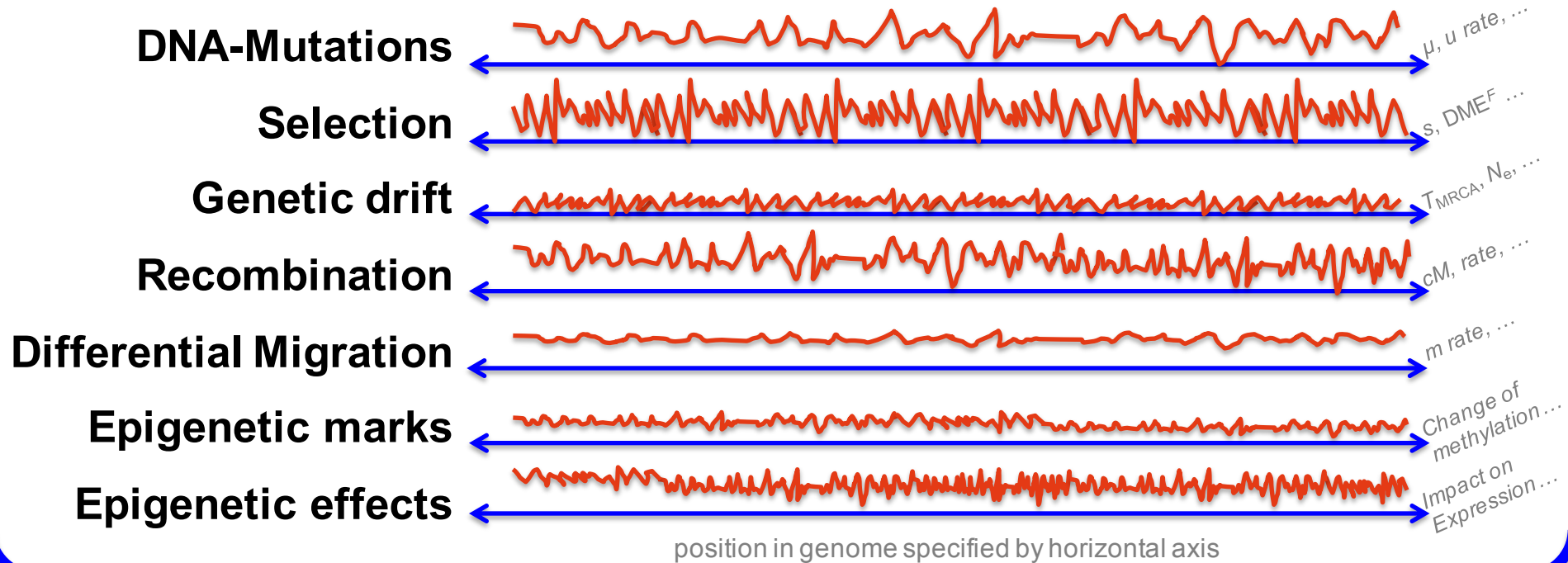
METs are causal traits of a *WT* with some stability ('heritable') that change independently from the SIFTs they impact. Examples: DNA sequence, promoter binding strength, kinetic parameter of enzyme, speed, ...

4. Define a causal $\mathbf{DM}_{MET}\mathbf{E}$, then sample its consequential $\mathbf{DM}_{MET}\mathbf{E}^{SIFT}$



Genomic Landscapes ...

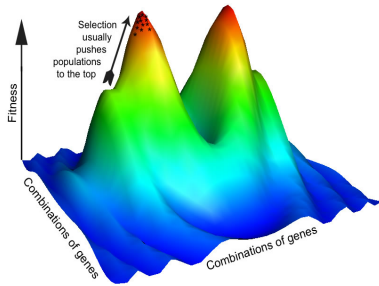
define genome-wide distributions of evolutionary factors that summarize Fitness Causality Networks incorporating effects from biochemistry to ecology



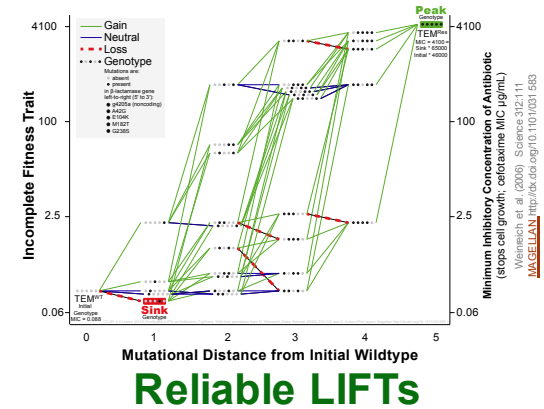
... Pose Huge Computing Challenges for Simulating Evolution:

- **Biological Species**: extremely diverse, often nested, different details matter
- **Population sizes**: can be huge (10^{14} bacteria @ 1KB / human gut = 100PB)
- **Genome sizes**: can be huge (*E.coli*: 4 million bp; *H.sapiens*: 6 billion bp)
- **Time heterogeneity**: can be huge (biochemistry: <ns; evolution: >1000My)

Where does **EvoSysBio** need to go from here?



EvoSysBio's Journey



Mostly MOCA-LIFTs

1. Embrace Trans-Disciplinarity

Develop a shared semantics and language to integrate existing mountains of diffuse physical, bio, and simulation data using the best math and stats methods.

2. Use Computational Data Science

Put computers to work. Reduce inefficient '*silicon digging*' and '*data shoveling*'. Automate!

3. Improve Models: reproducible, reliable, usable, documented, ...

Write models for efficiently extending them without the need for reimplementing.

4. Improve Tools: more accuracy, speed, usability, documentation, ...

Improve versatility and reliability of simulation tools for more diverse complex problems.

5. Allow Models to Inspire Biological Wet-Lab Experiments

Collect relevant raw time series data; use it to estimate better values for key parameters.