



Whole-Cell Modeling

PREDICTING PHENOTYPE FROM GENOTYPE
FOR SCIENCE, MEDICINE & ENGINEERING

Toward WC models for predicting cellular phenotypes

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July 17, 2019



**Mount
Sinai**



Join us at Mount Sinai!

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Acknowledgements



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Arthur Goldberg



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Outline

Genotype to cellular phenotype

- What is a WC model?
- Why do we need WC models?
- Challenges & feasibility
- Foundational principles and state of the art
- Progress toward comprehensive models

Tips for modeling complex systems



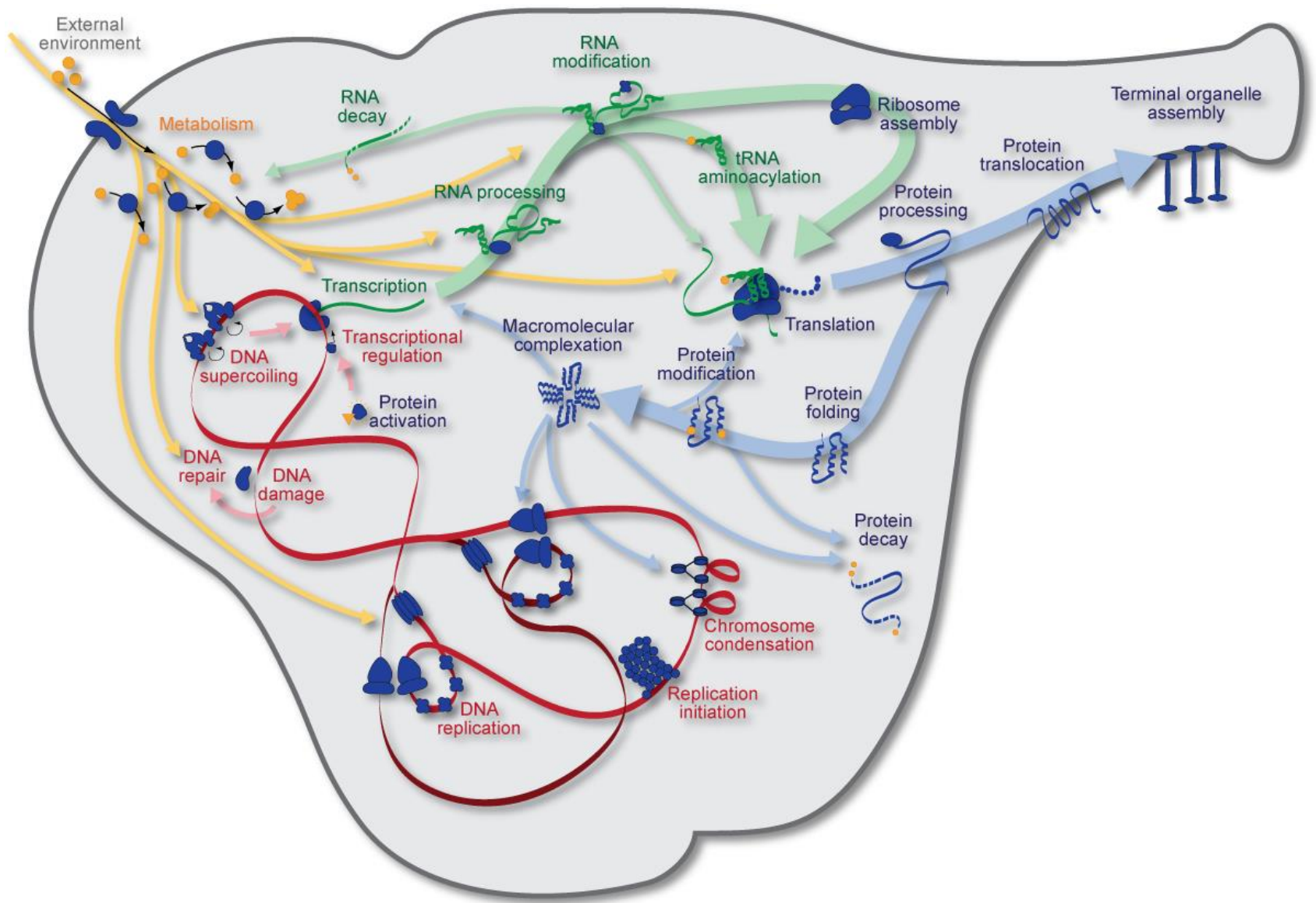
What is a WC model?



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Goals of WC modeling



Species-specific



Mechanistic



Whole cell



Dynamic



Whole genome



Stochastic



Whole cell cycle

Motivation

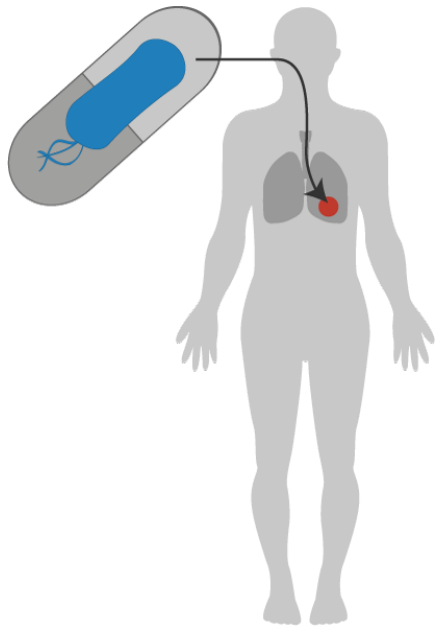


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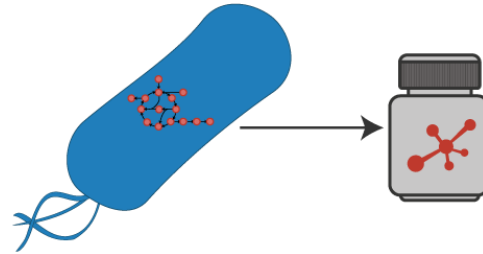


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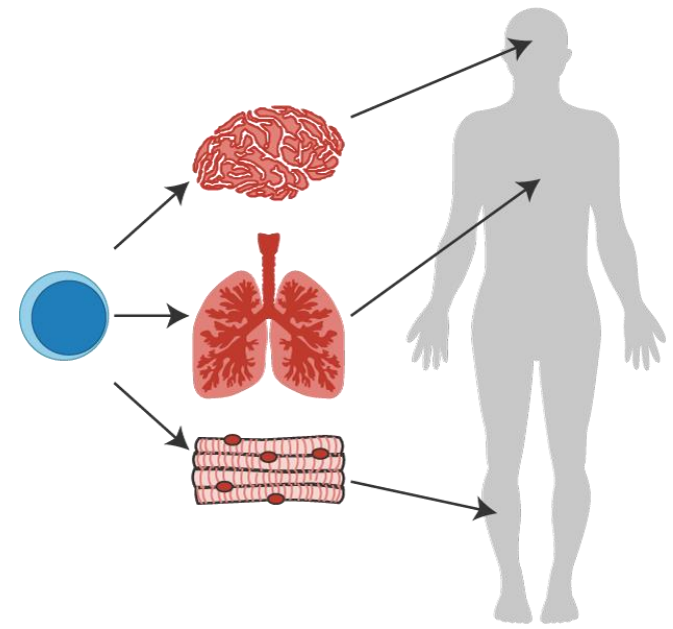
Synthetic biology requires WC models



Biosensors



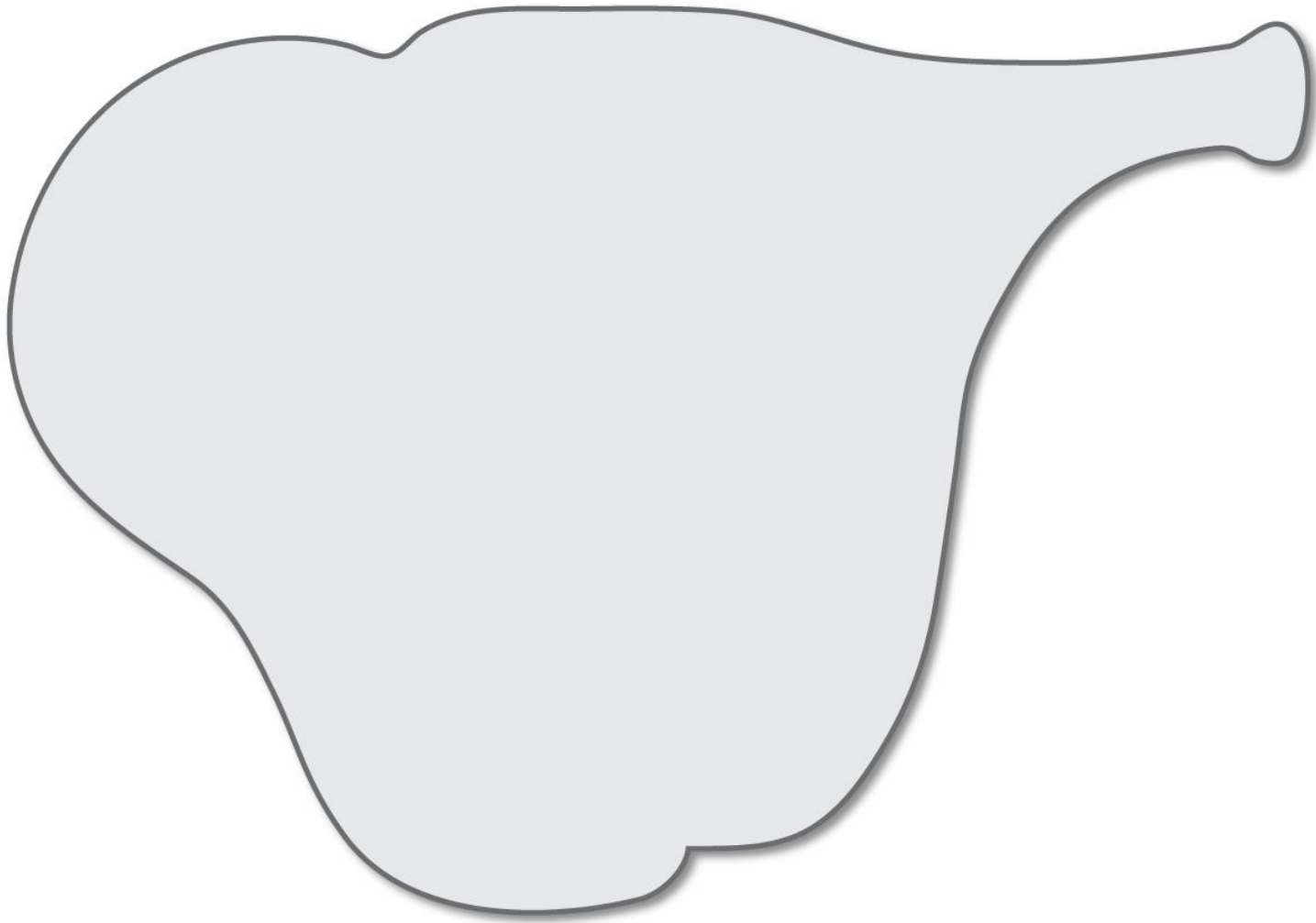
Biofactories



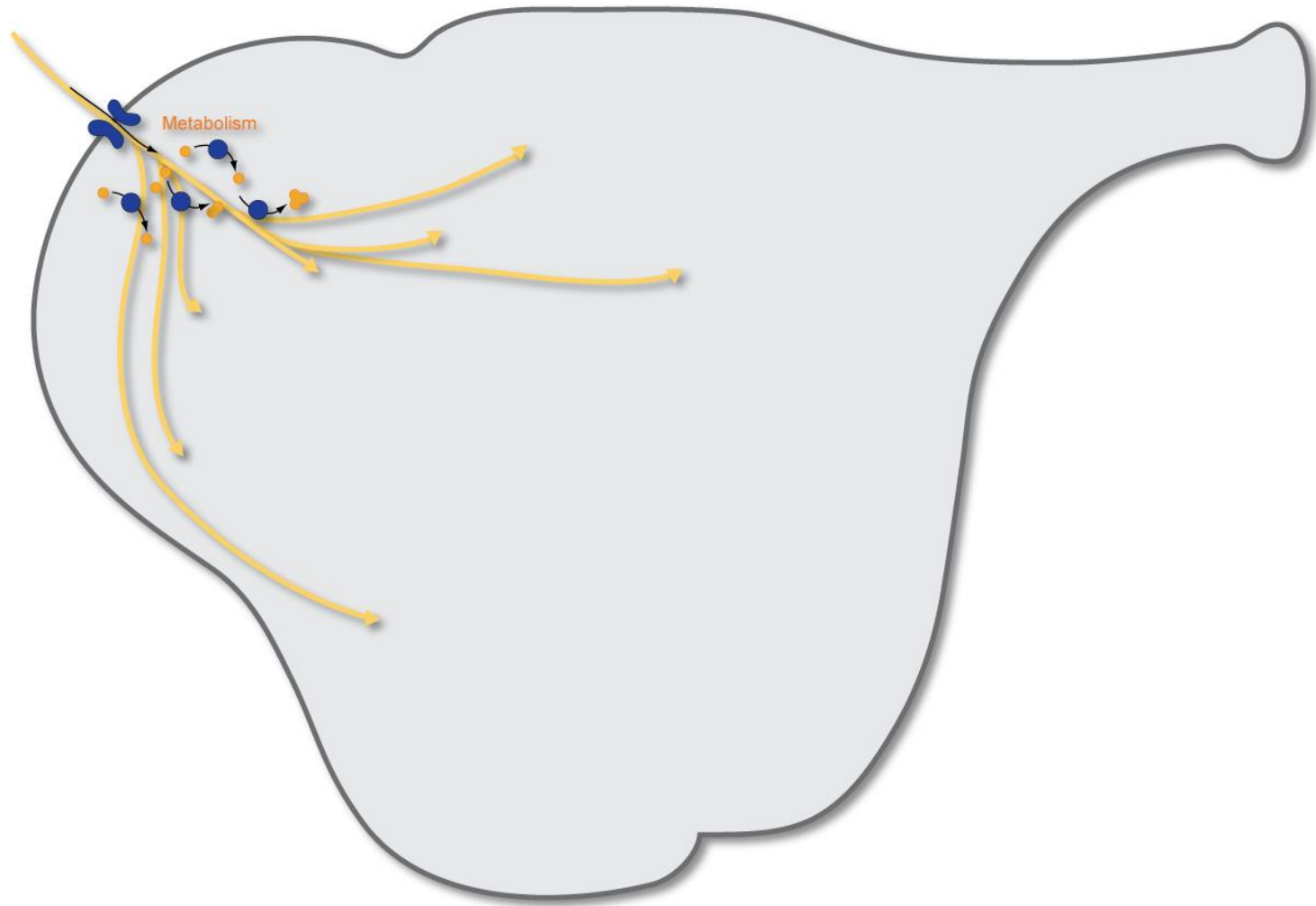
Tissue engineering



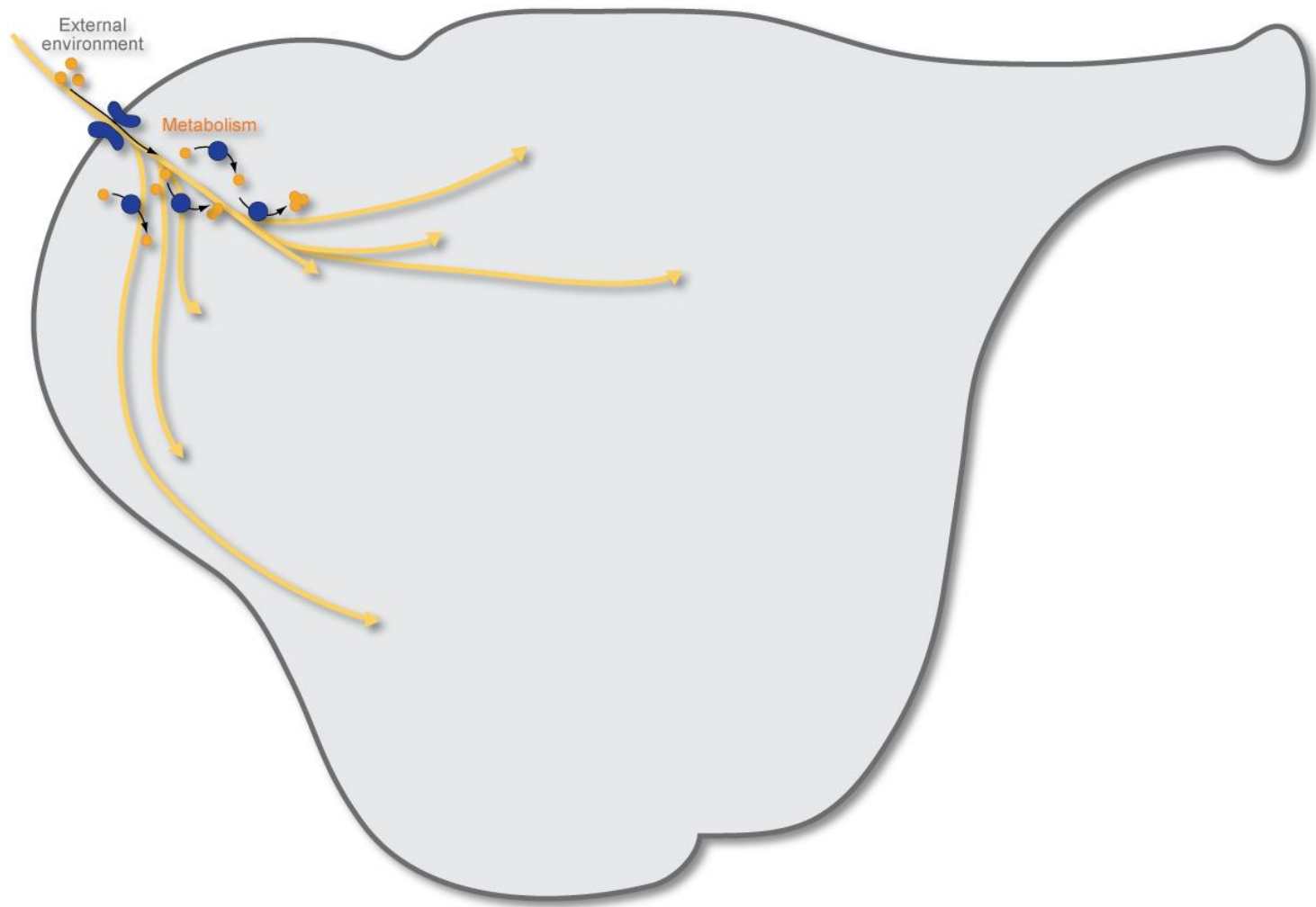
Example: drug biosynthesis



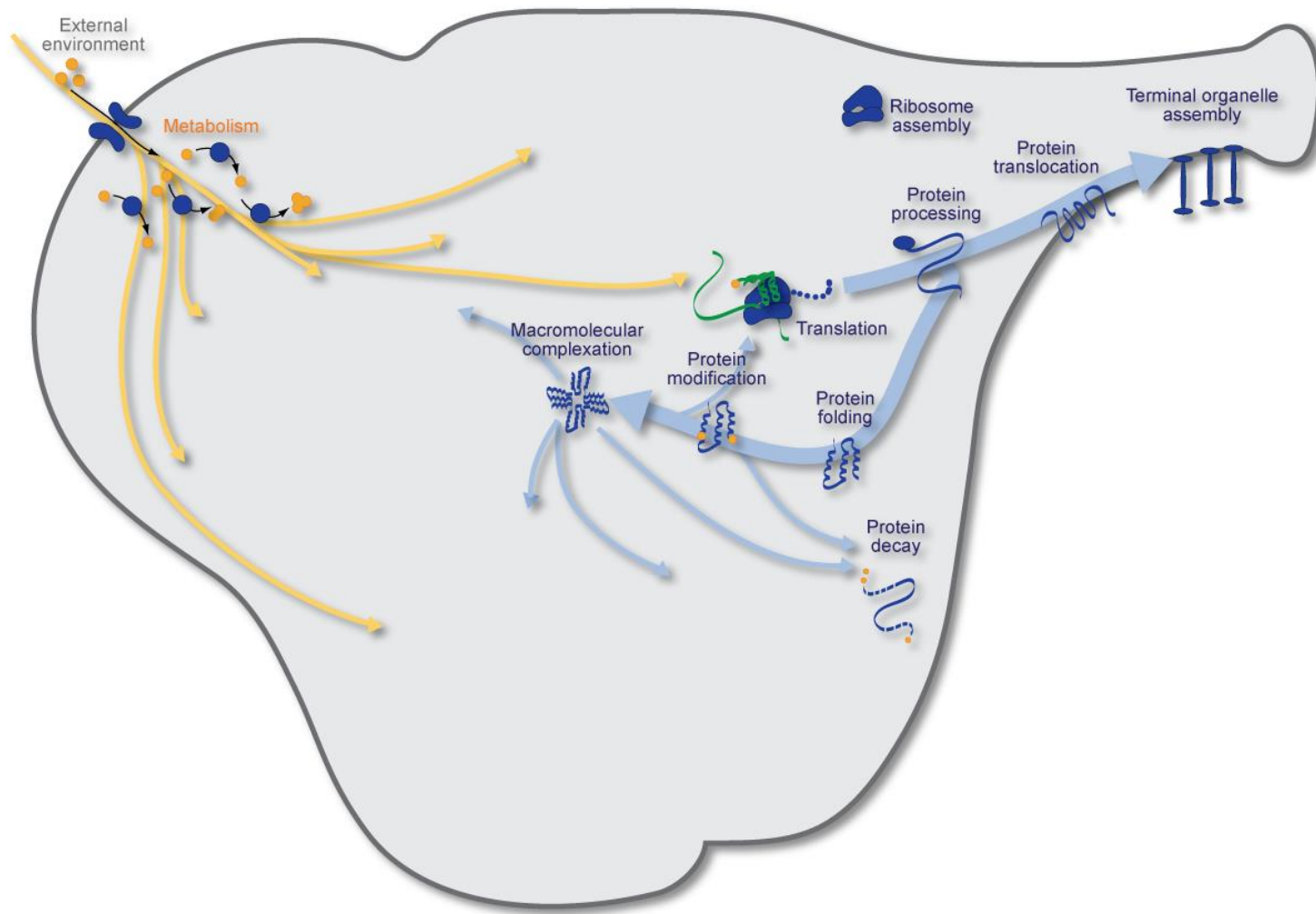
Example: drug biosynthesis



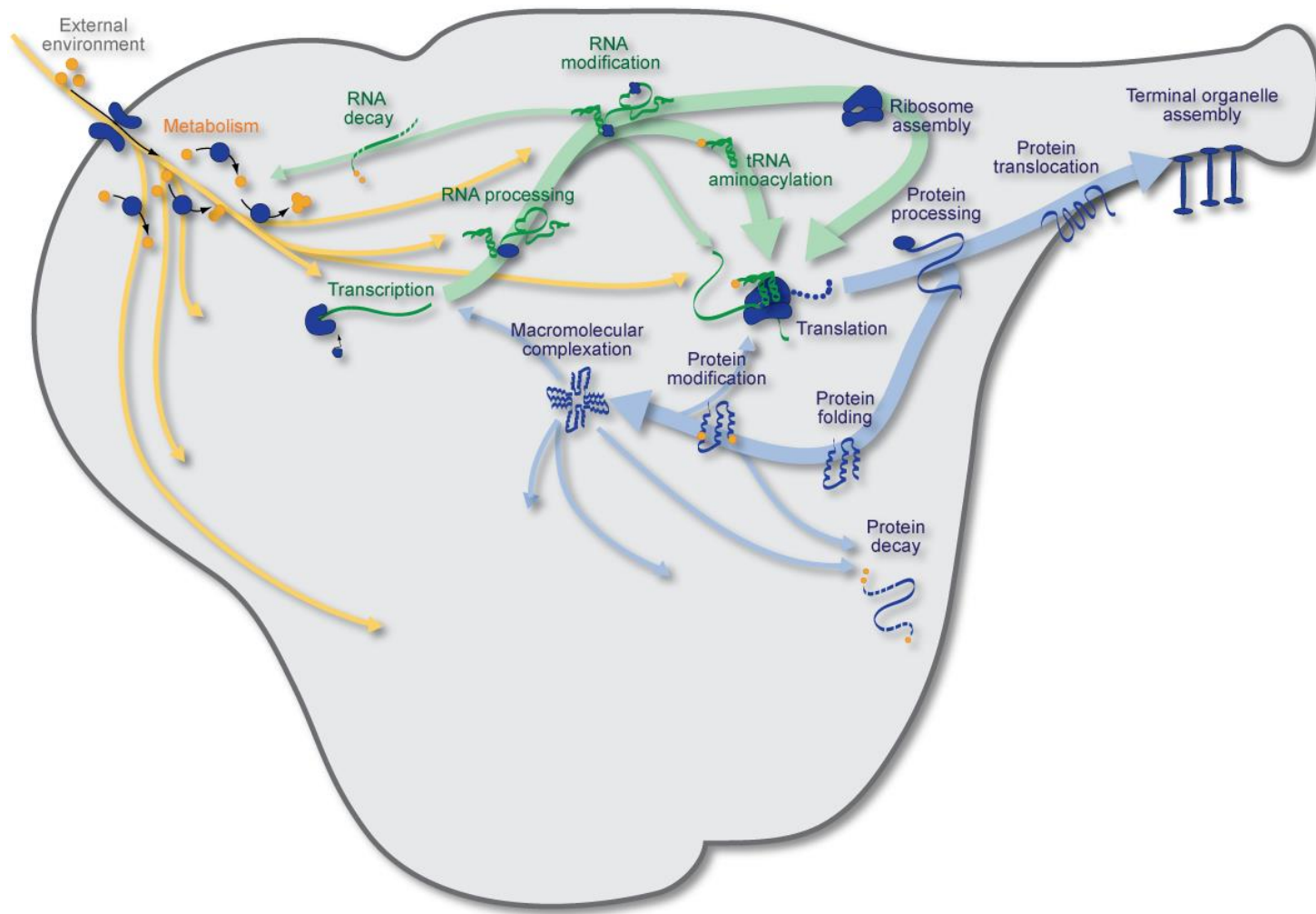
Example: drug biosynthesis



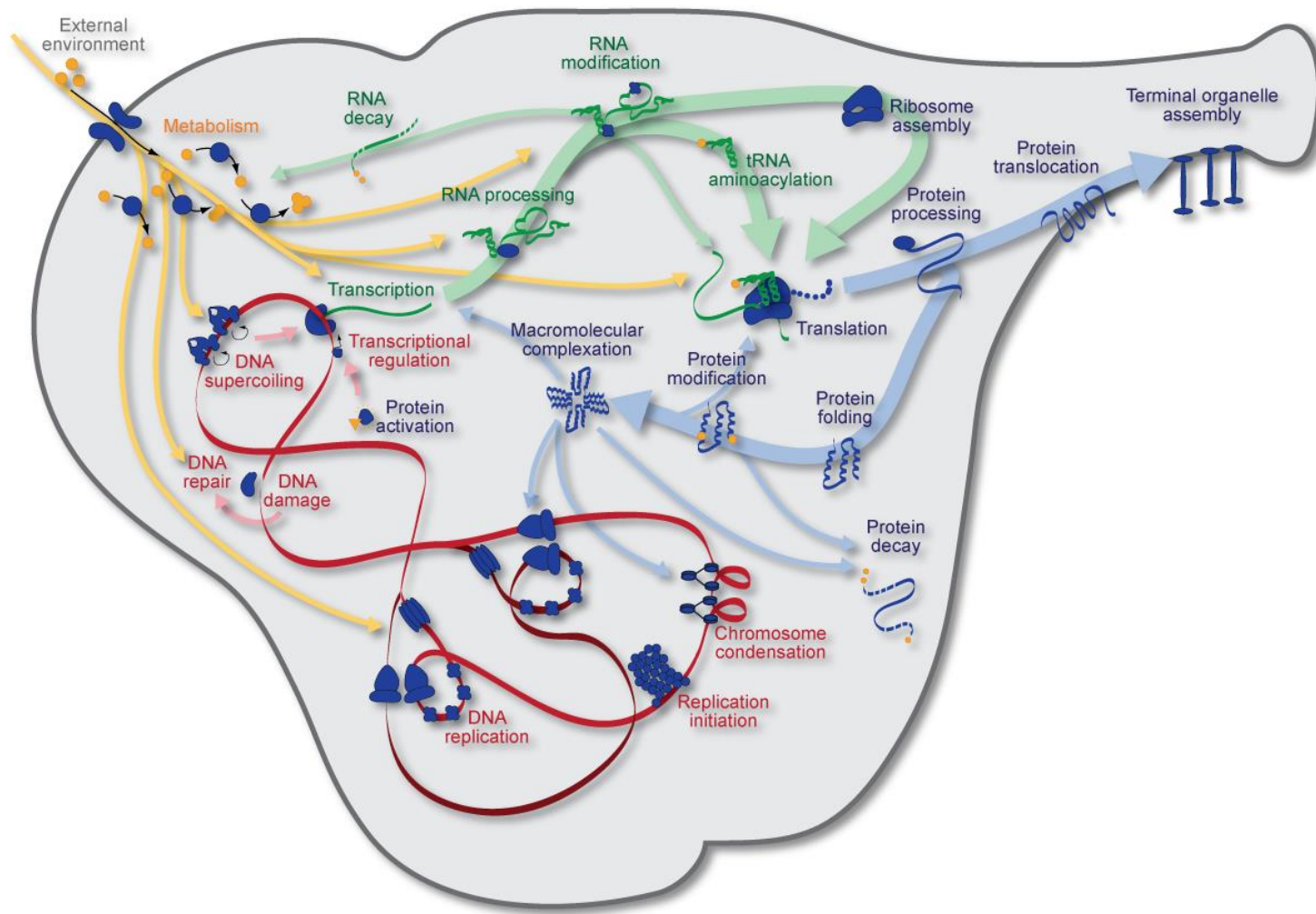
Example: drug biosynthesis



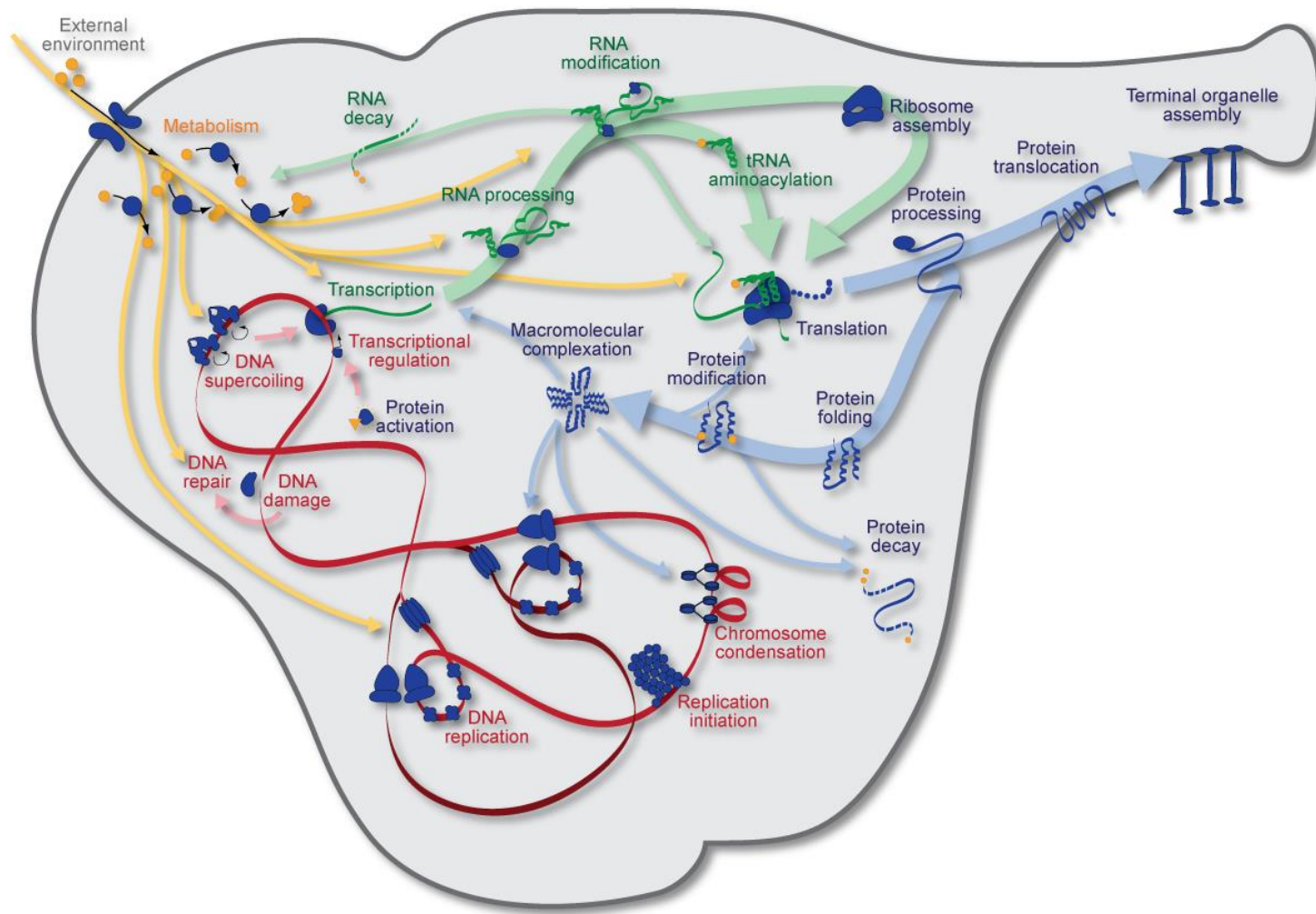
Example: drug biosynthesis



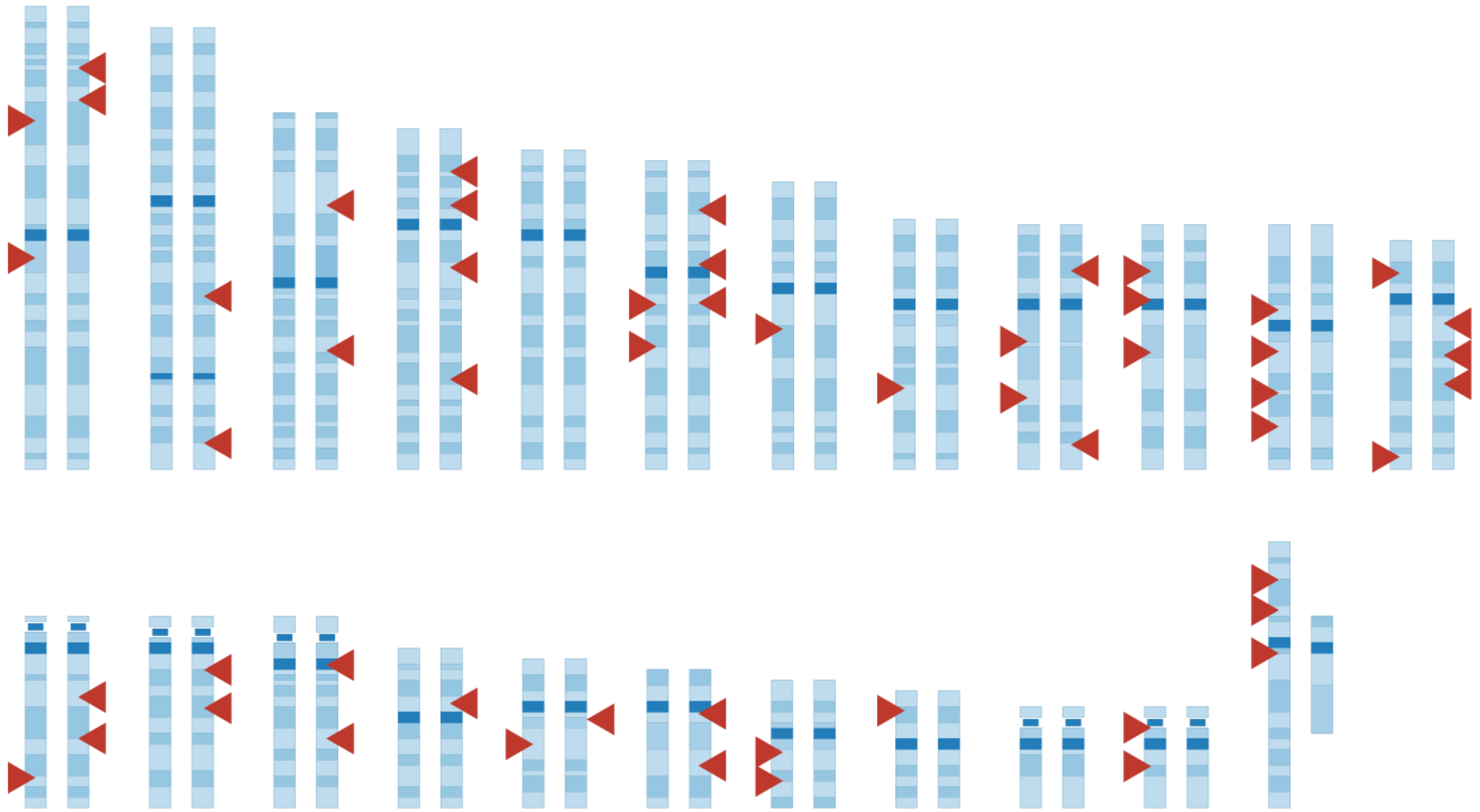
Example: drug biosynthesis



Example: drug biosynthesis



Precision medicine requires WC models



Challenges

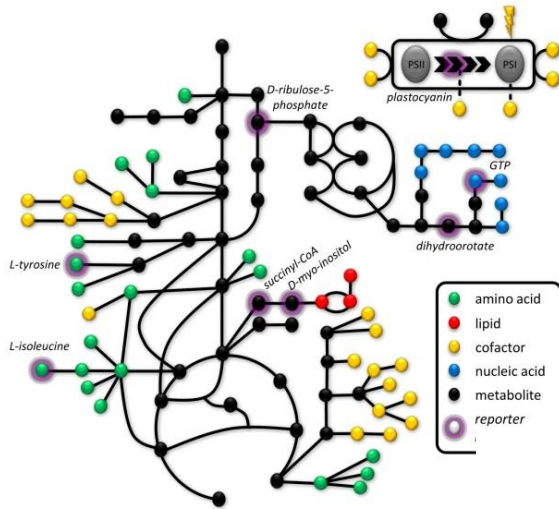


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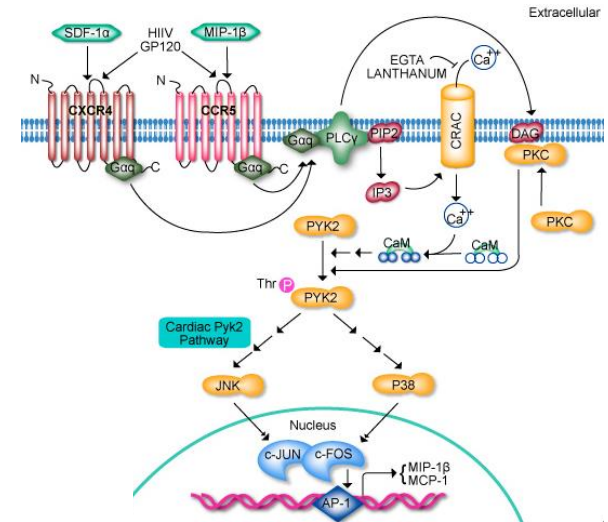
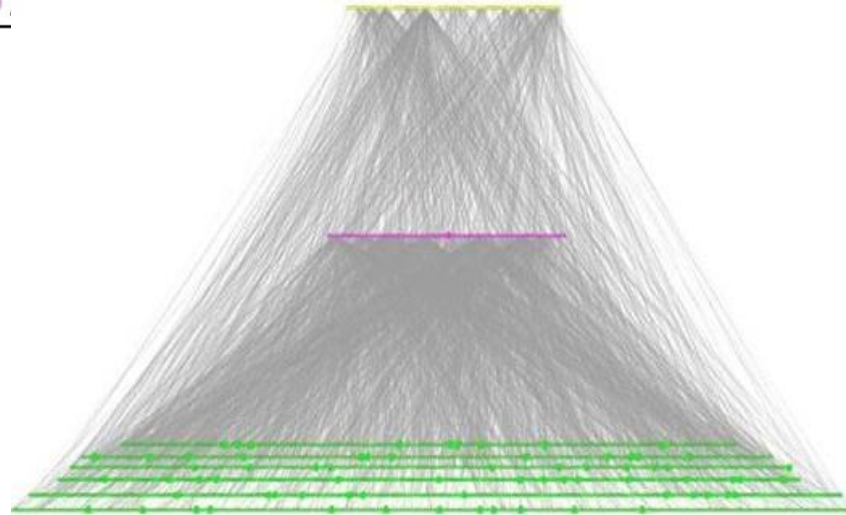
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Challenge: explain diverse chemistry



Metabolism
FBA

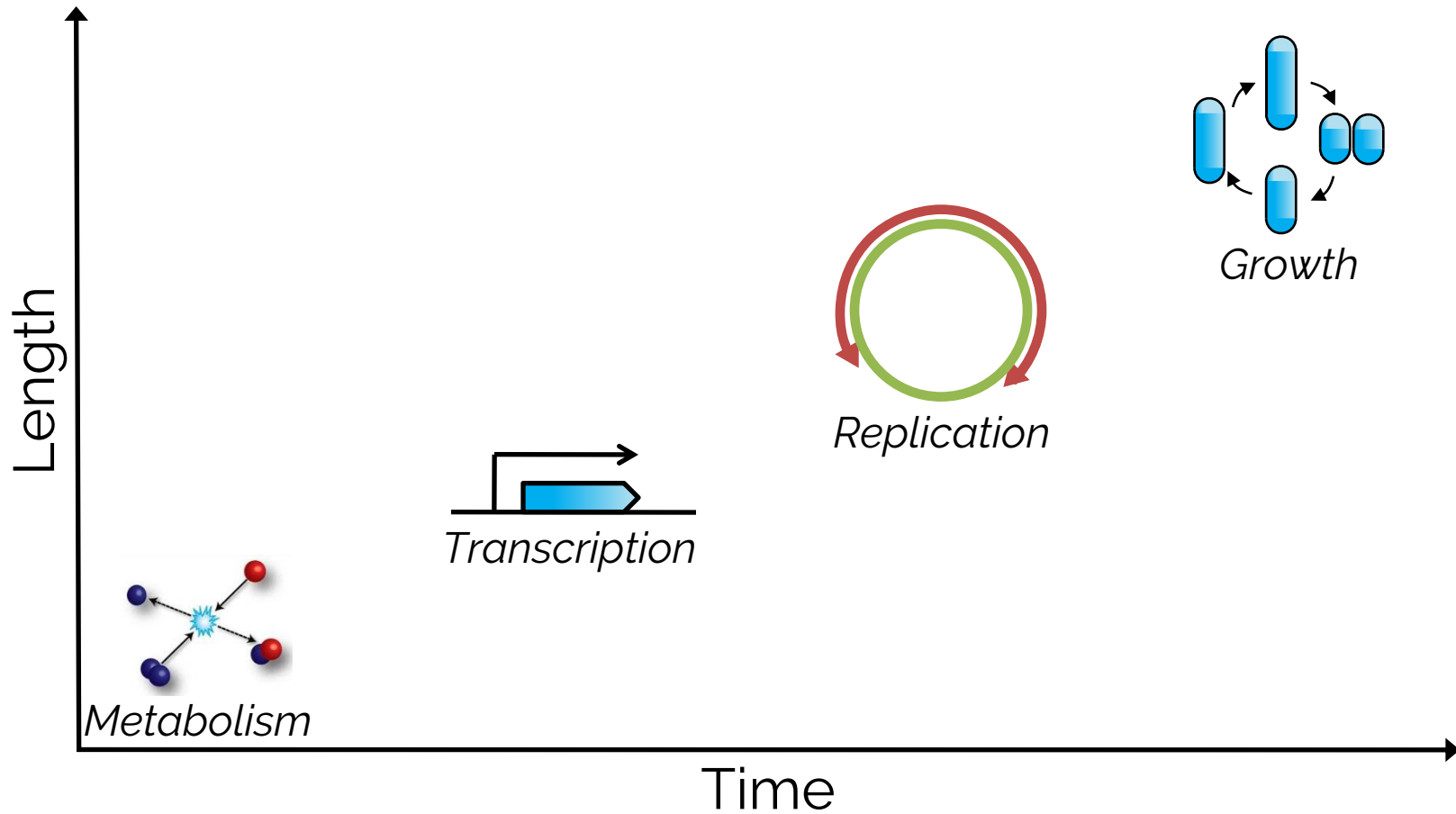
Transcriptional
regulation
Logical



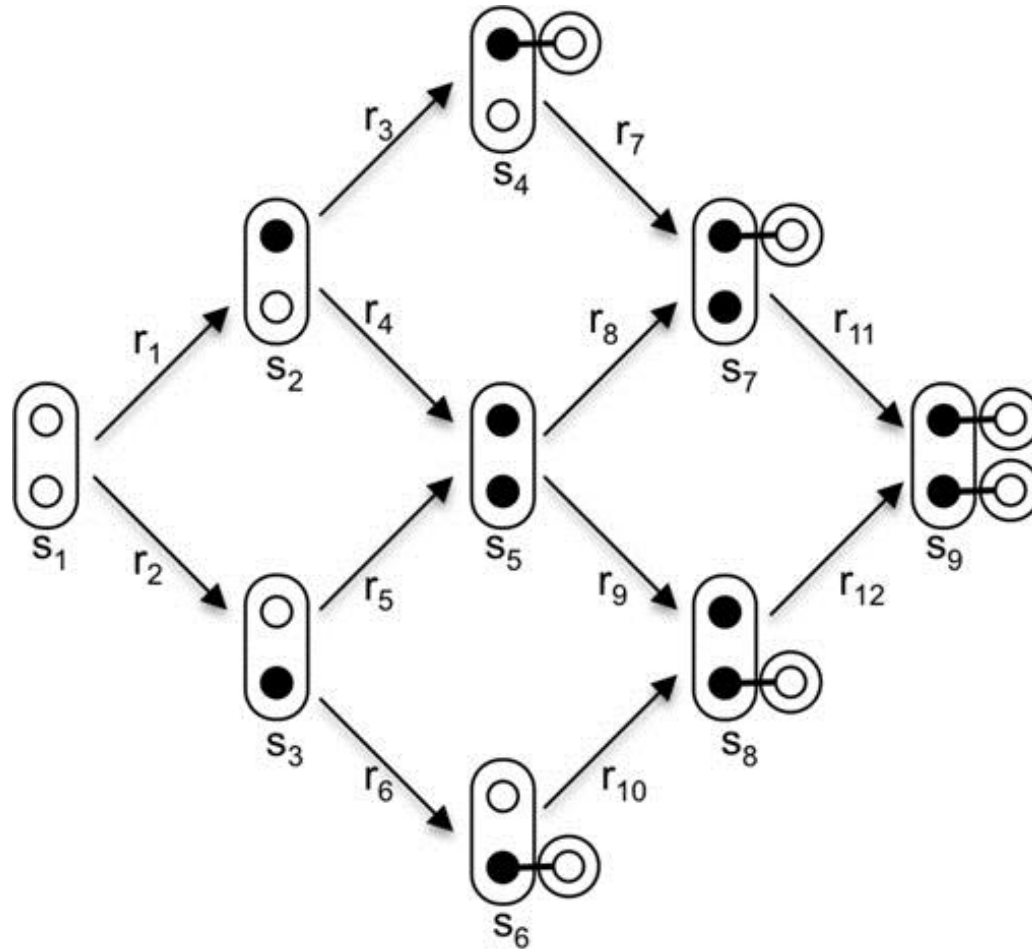
Signaling
ODE, SSA



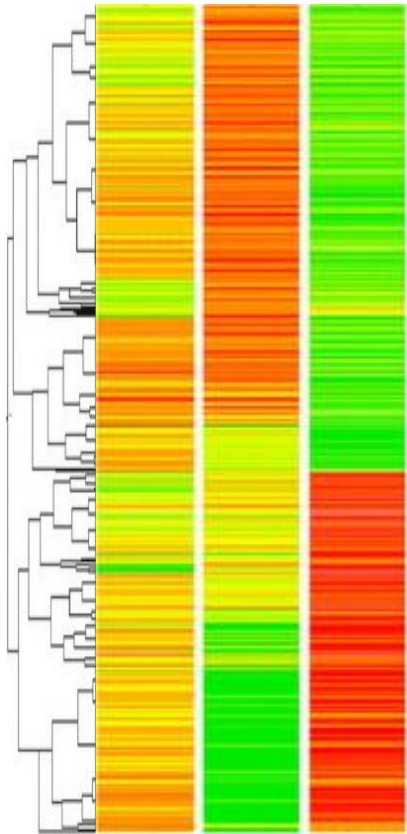
Challenge: explain multiple scales



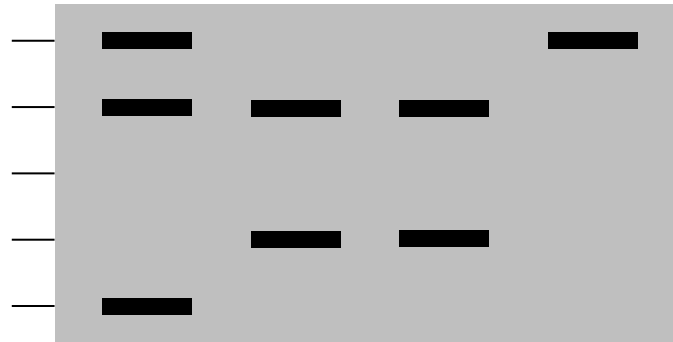
Challenge: capture chemical complexity



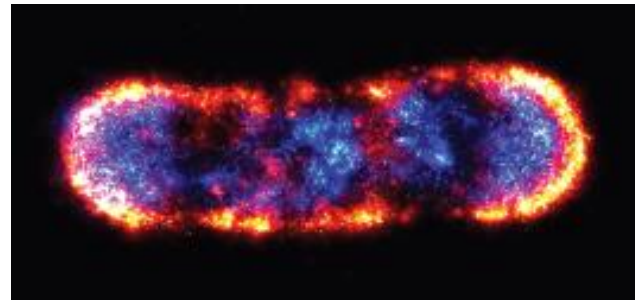
Challenge: heterogeneous data



Transcription
RNA-seq

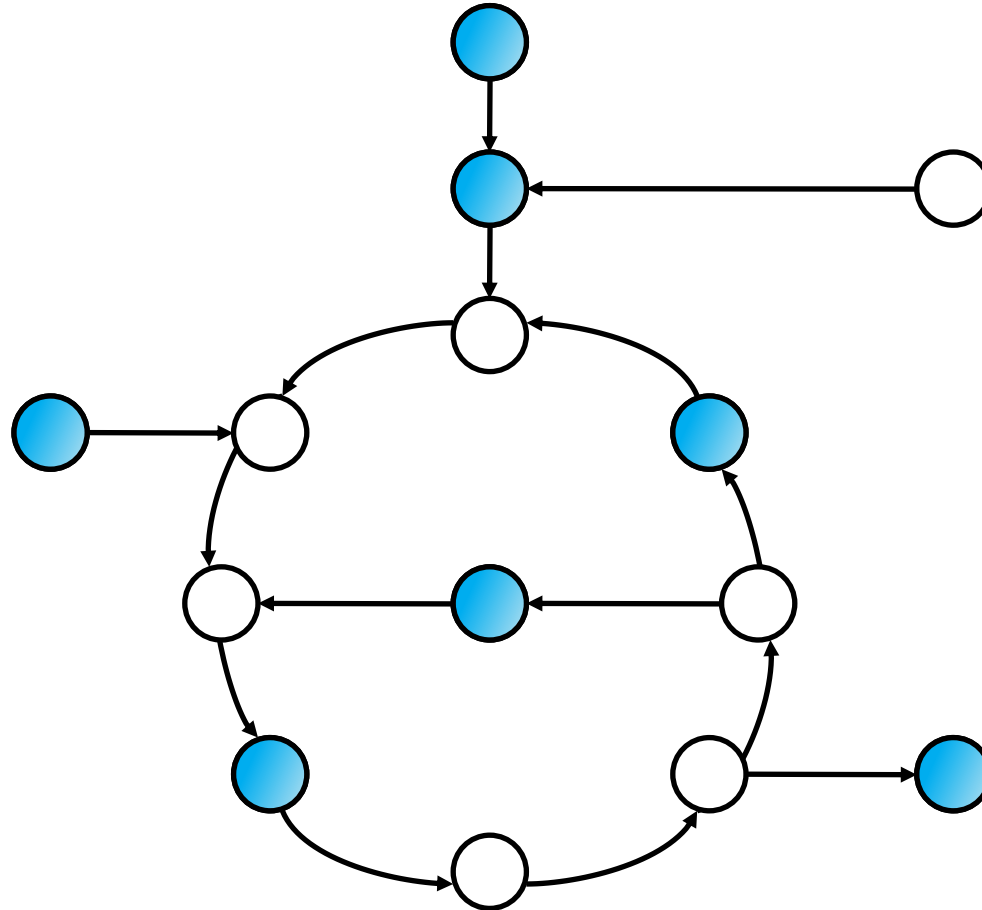


Protein expression
Mass-spec, Western blot



Single-cell variation
Microscopy

Challenge: incomplete data



Feasibility

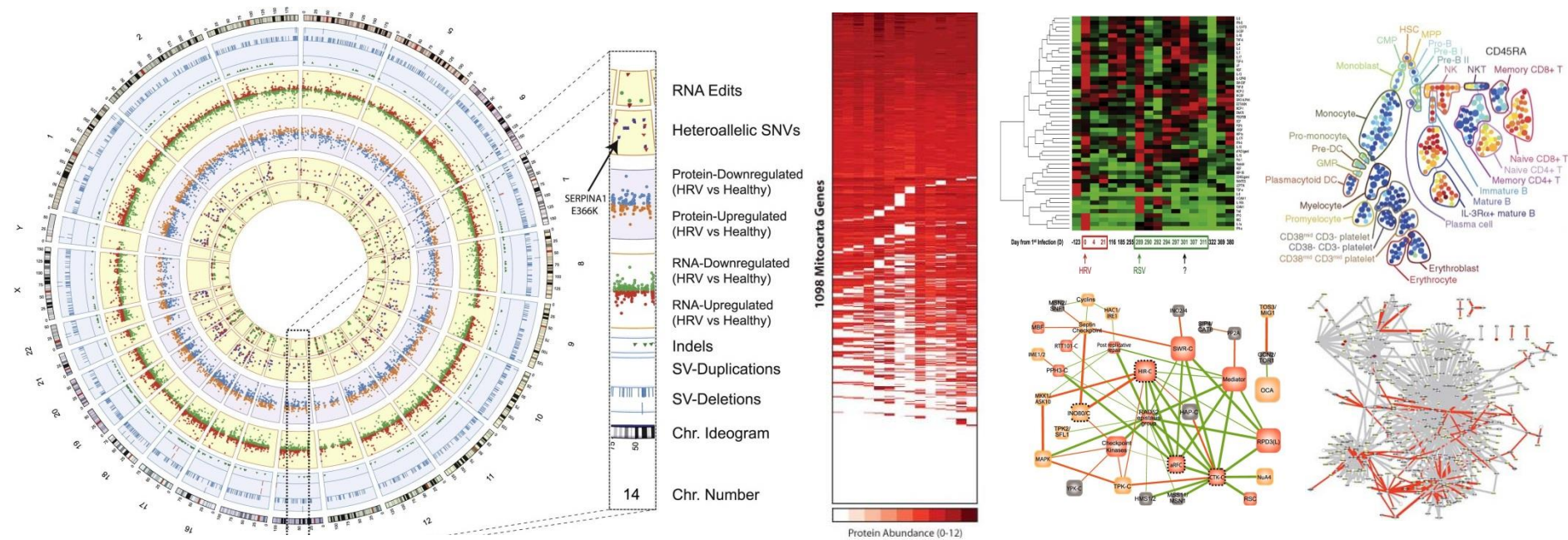


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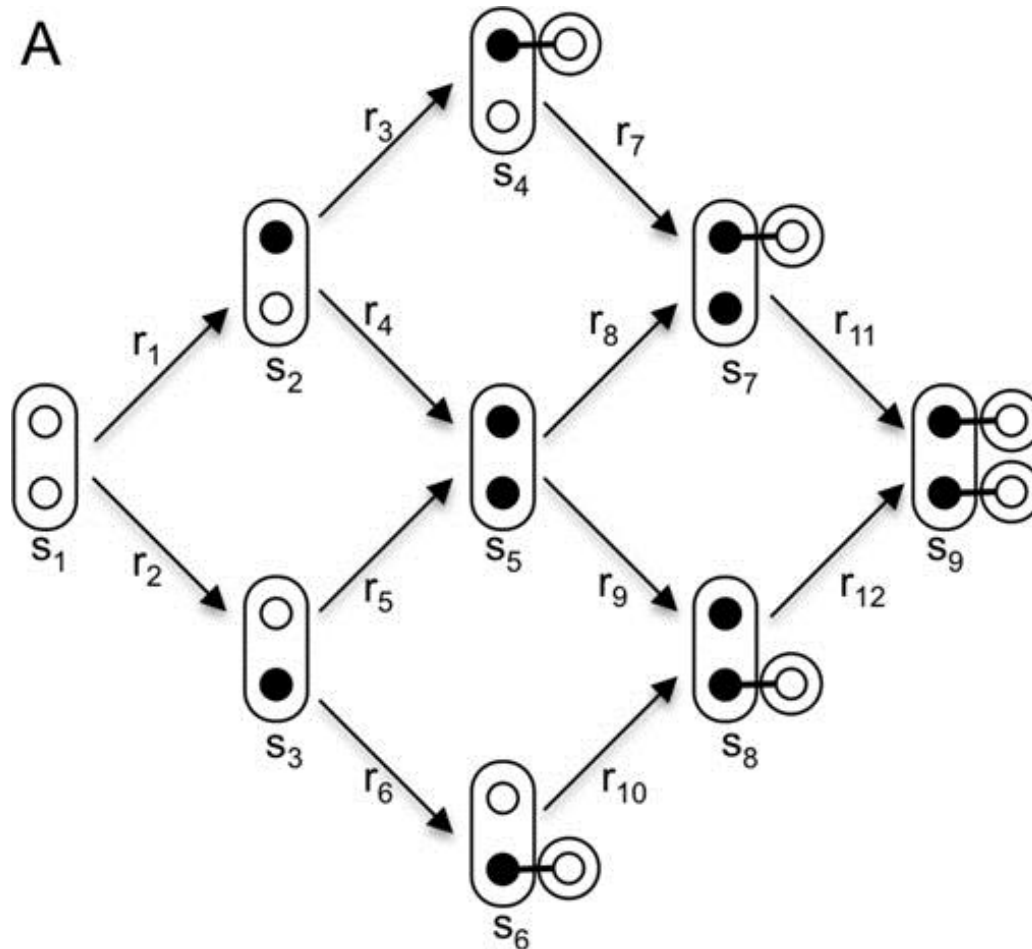


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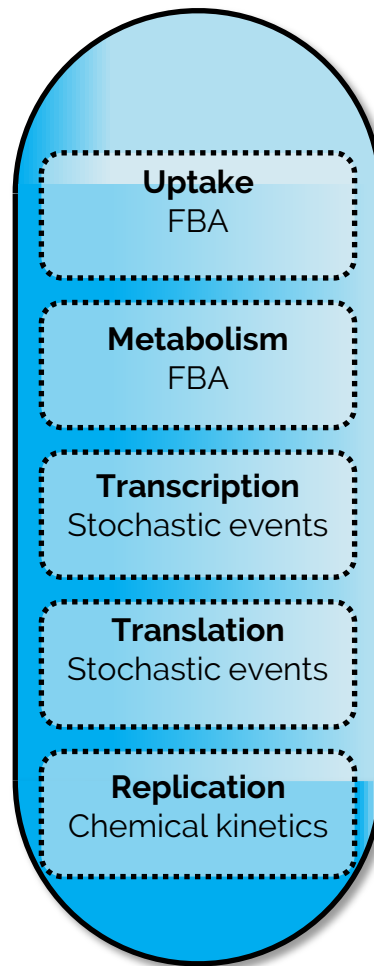
Feasibility: Extensive data



Feasibility: Rule-based modeling



Feasibility: Multi-algorithm simulation



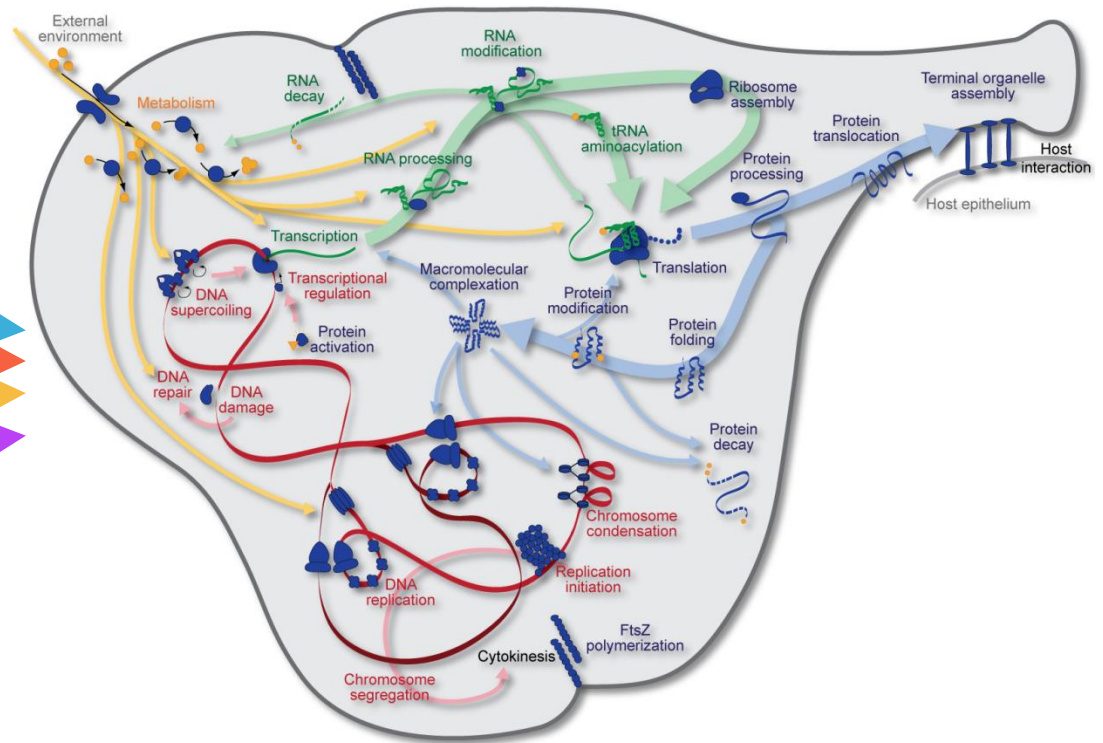
WC modeling is becoming feasible

Genomic and
biochemical data

Pathway
submodels

Rule-based
modeling

Multi-algorithmic
simulation



Workflow

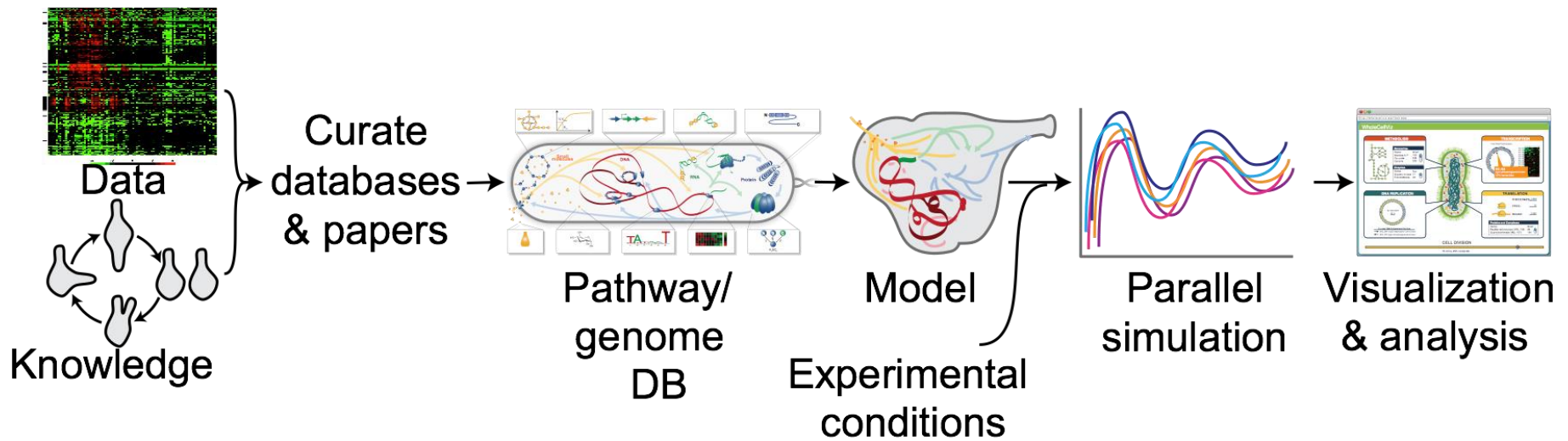


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Modeling

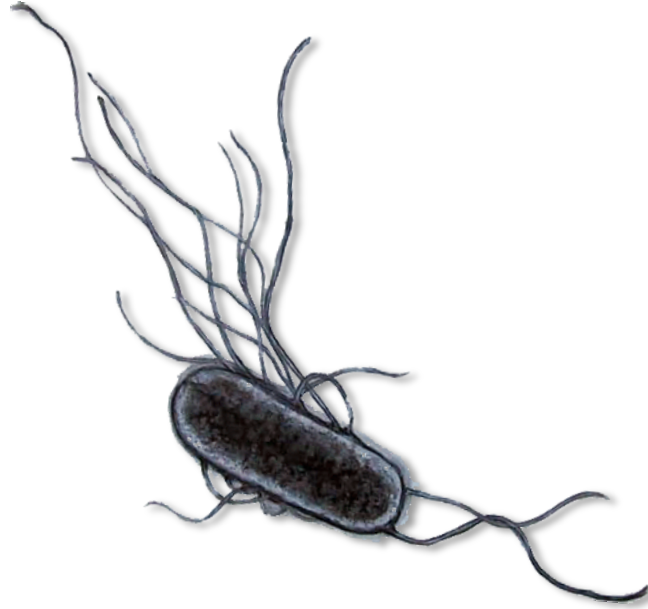


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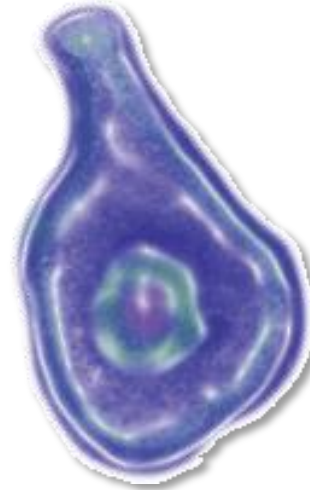
Workflow



1. Focus on simple cells



E. coli



M. genitalium

Genome

4700 kb

580 kb

Genes

4461

525

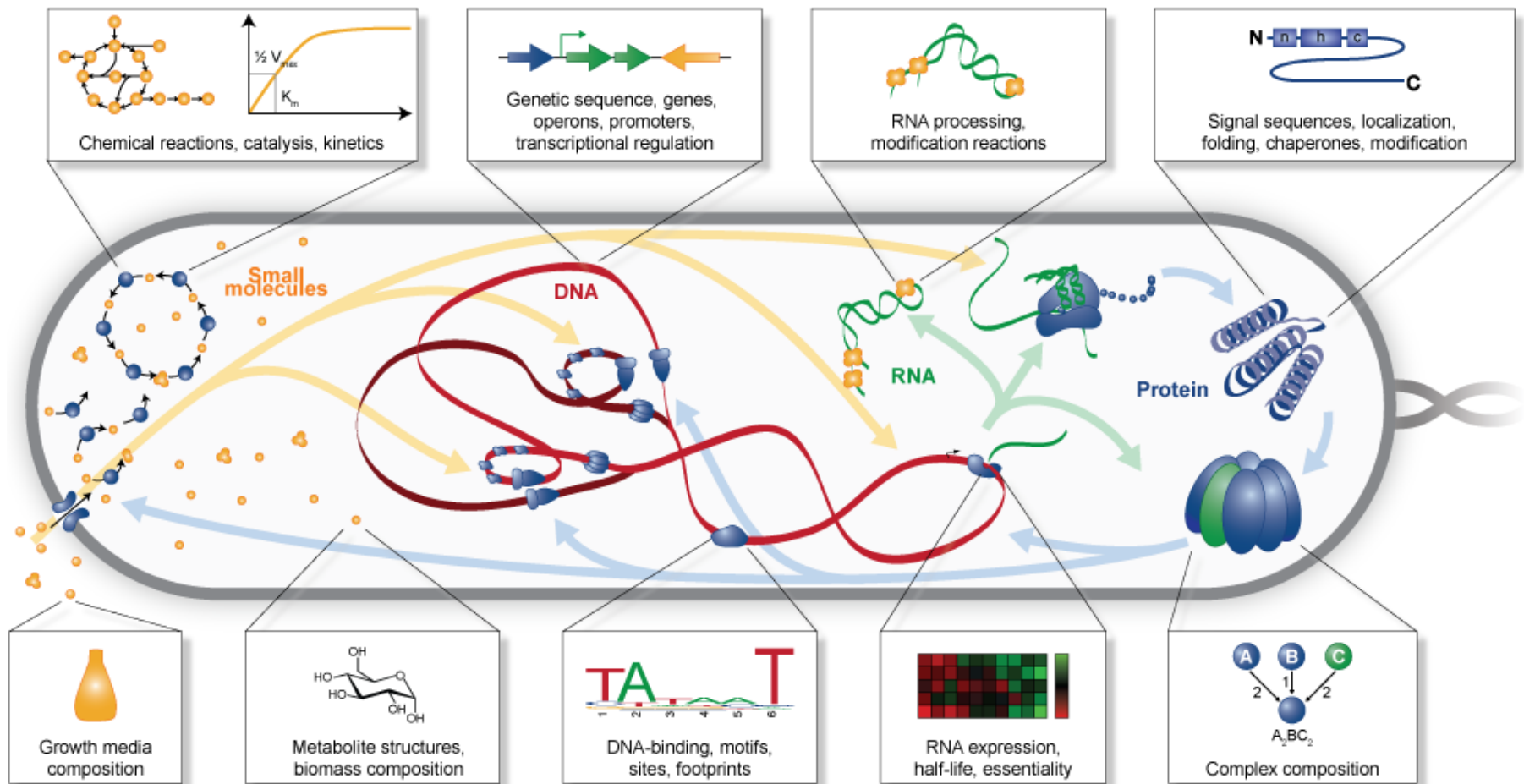
Size

2 μm \times 0.5 μm

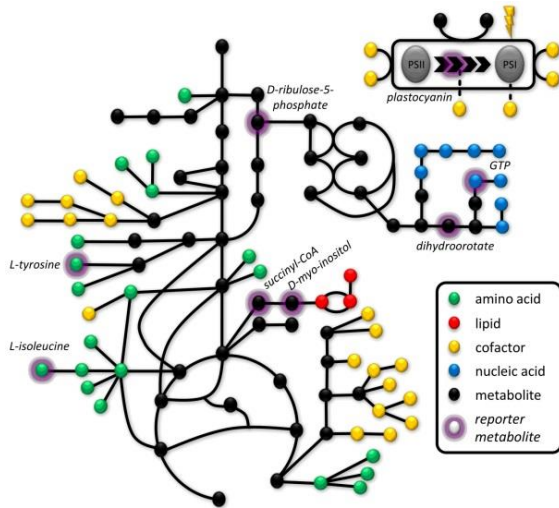
0.2-0.3 μm



2. Aggregate and integrate data

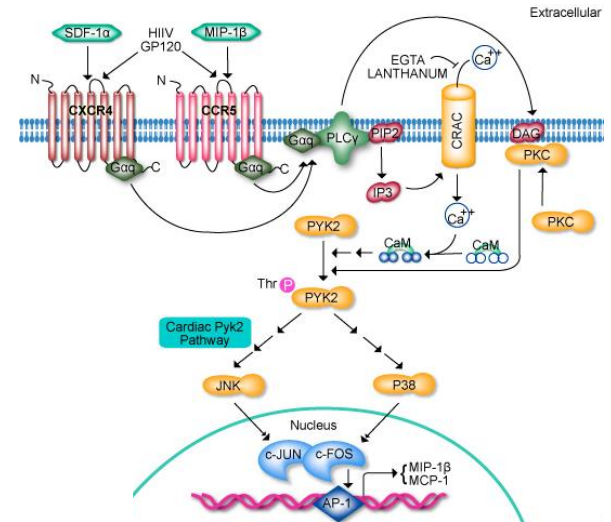
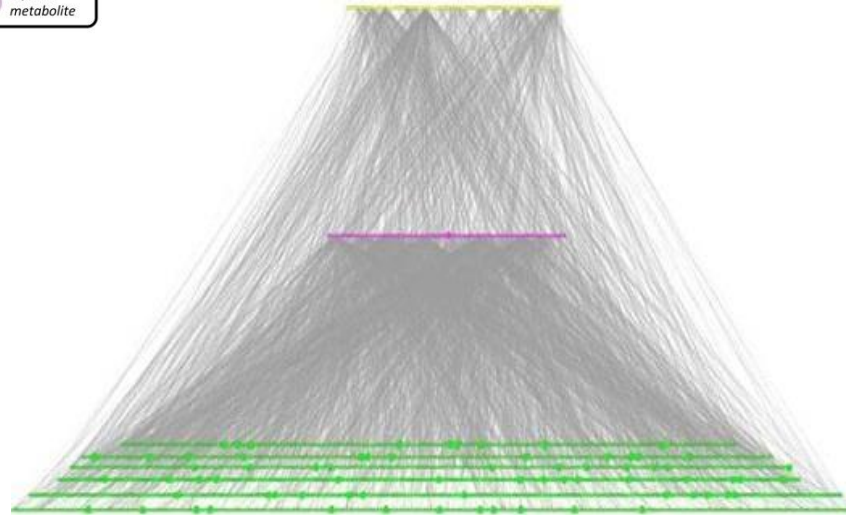


3. Model each process



Metabolism

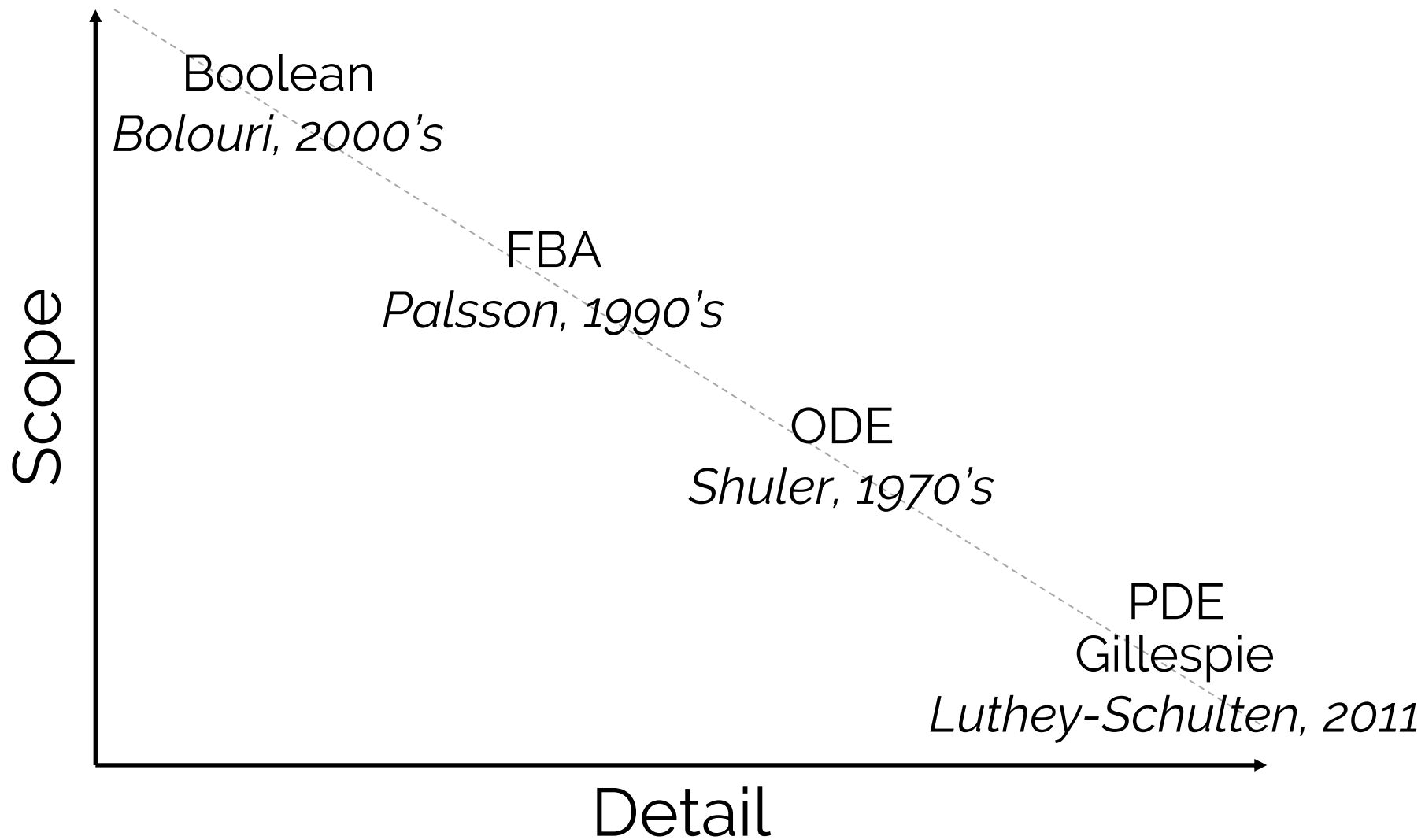
Transcriptional regulation



Signaling

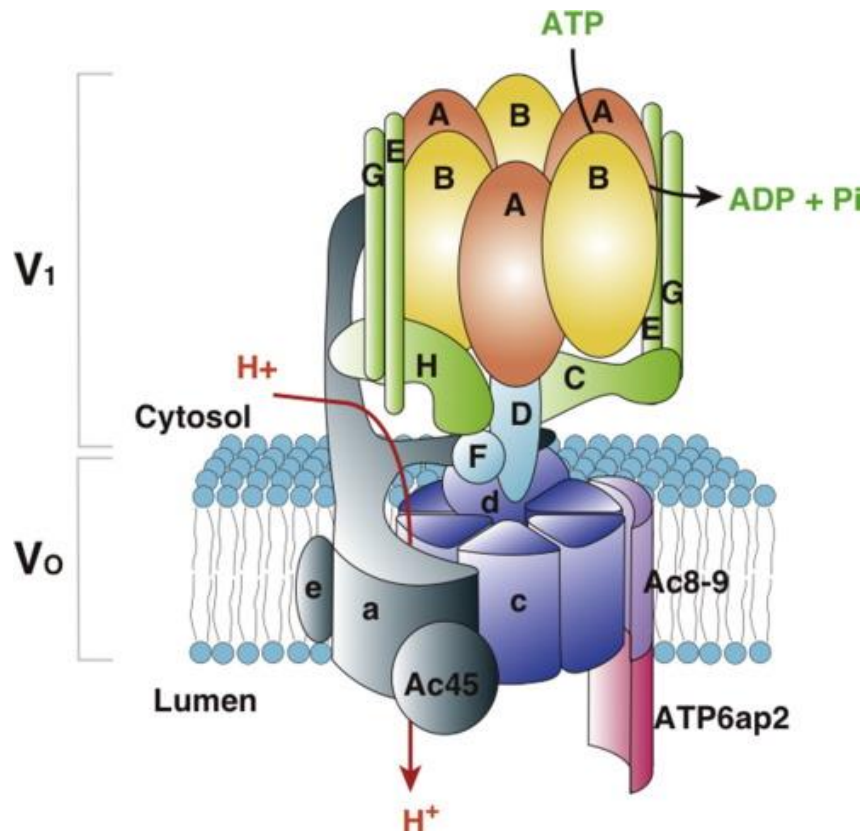


3. Model each process

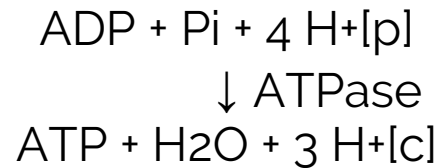


3. Model each process

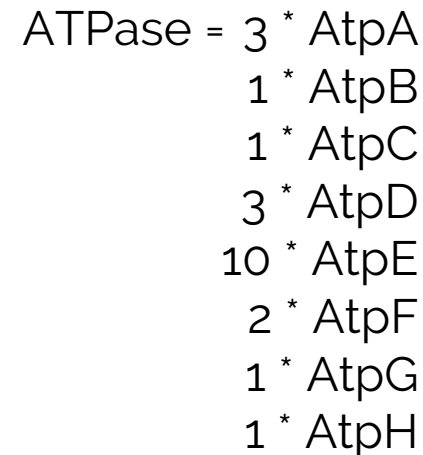
Metabolism



Species and reactions



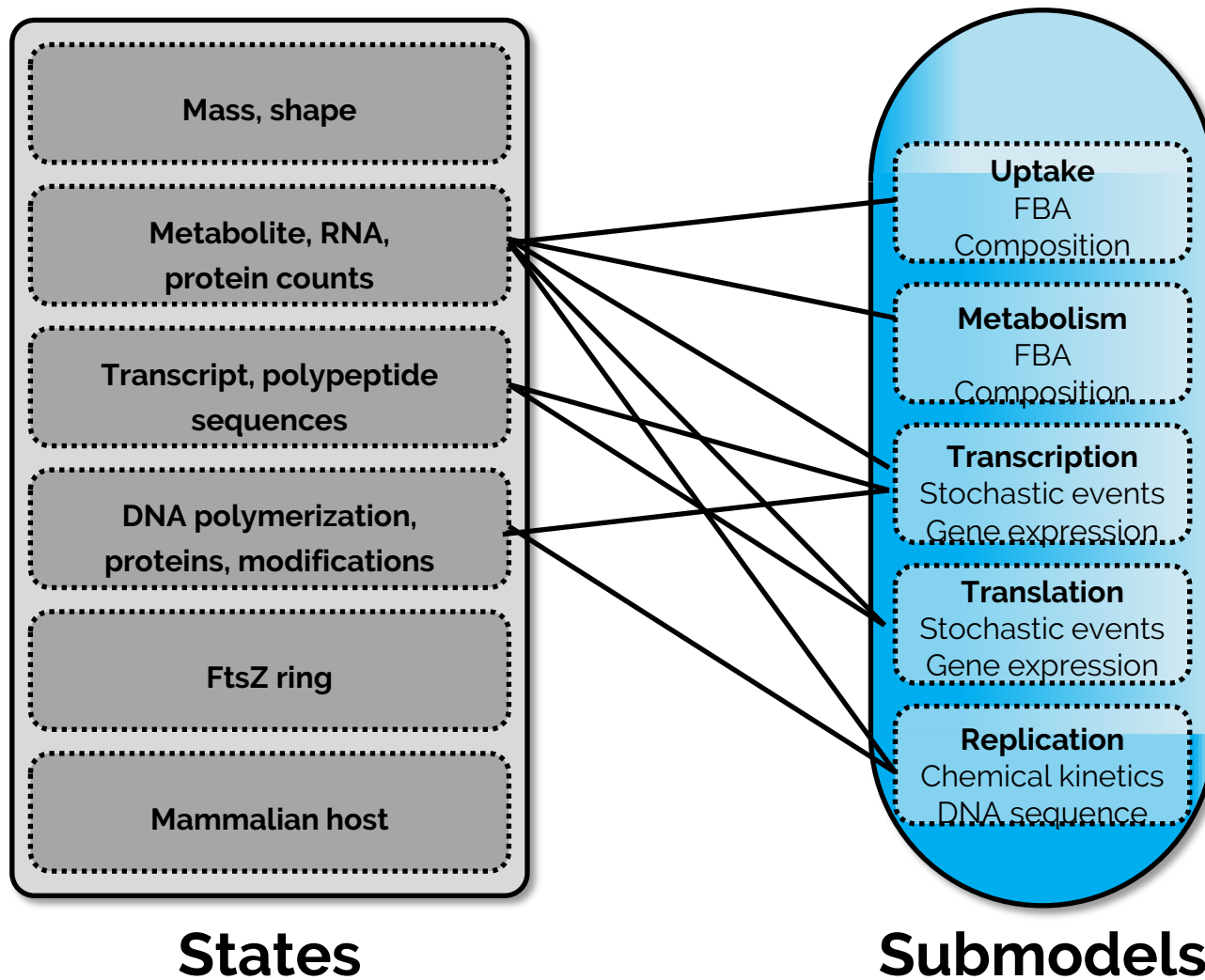
Catalysis



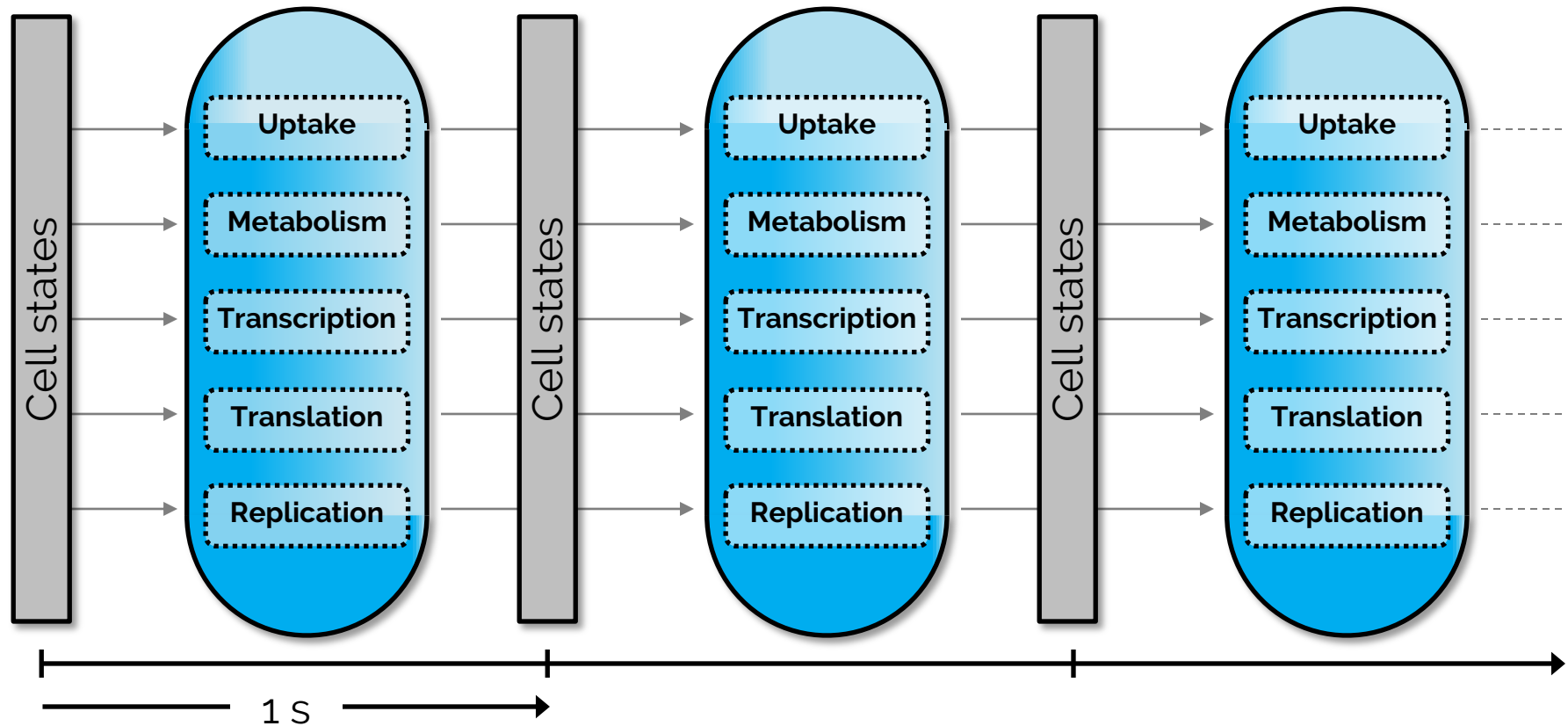
Kinetics

$$v = k_{\text{cat}}[\text{ATPase}] \frac{[\text{ADP}]}{K_m + [\text{ADP}]}$$

4. Merge models into a single model



5. Co-simulate models



6. Verify model

☑ Matches training data

- ☑ Cell mass, volume
- ☑ Biomass composition
- ☑ RNA, protein expression, half-lives
- ☑ Superhelicity

☑ Matches published data

- ☑ Metabolite concentrations
- ☑ DNA-bound protein density
- ☑ Gene essentiality

☑ Matches theory

- ☑ Mass conservation
- ☑ Central dogma
- ☑ Cell theory
- ☑ Evolution

☑ No obvious errors

- ☑ Plot model predictions
- ☑ Manually inspect data
- ☑ Compare to known biology

State of the art

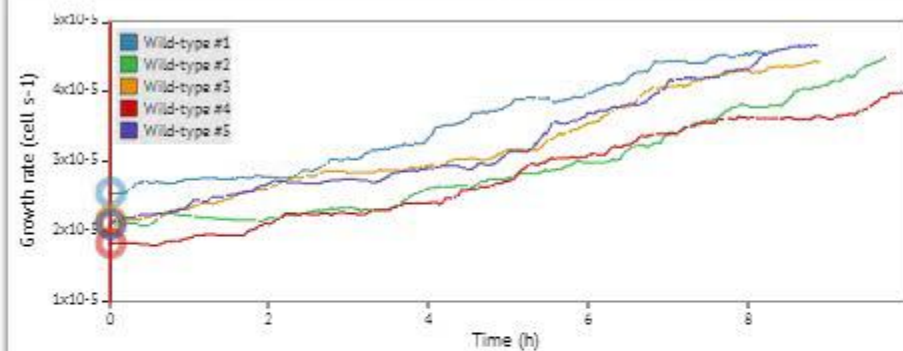


Whole-Cell
Modeling

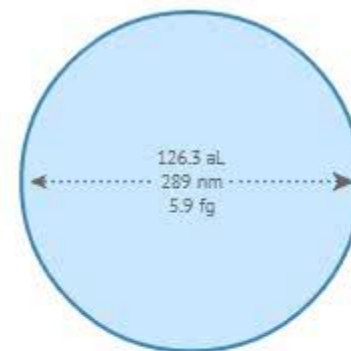


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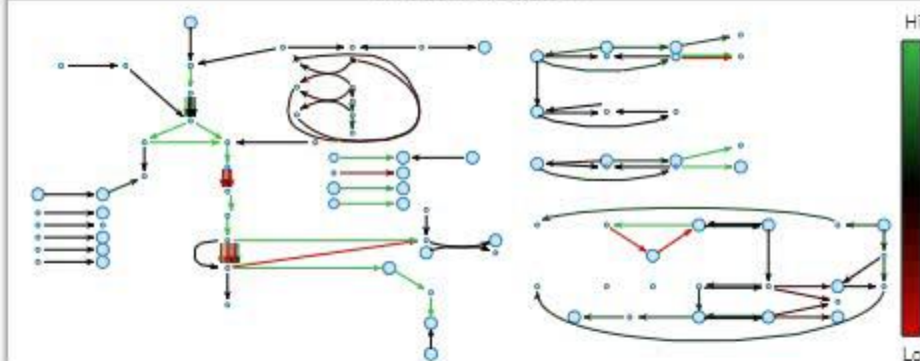
Growth rate
Simulation: multiple selected



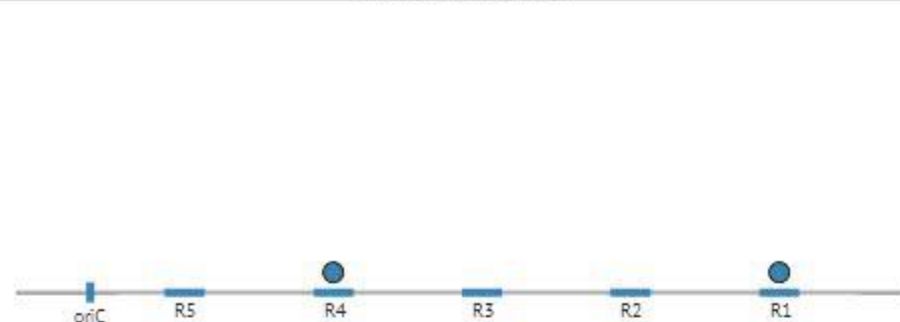
Cell shape
Simulation: Wild-type #1



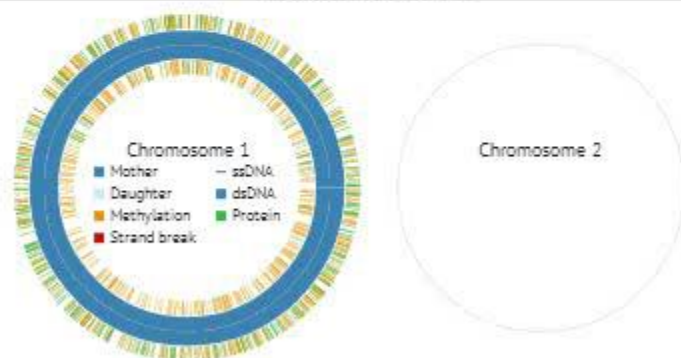
Metabolism
Simulation: Wild-type #1



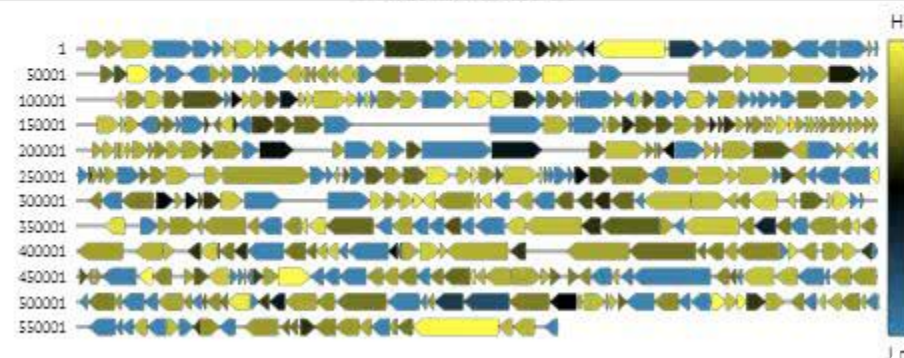
Replication initiation – oriC DnaA Boxes
Simulation: Wild-type #1



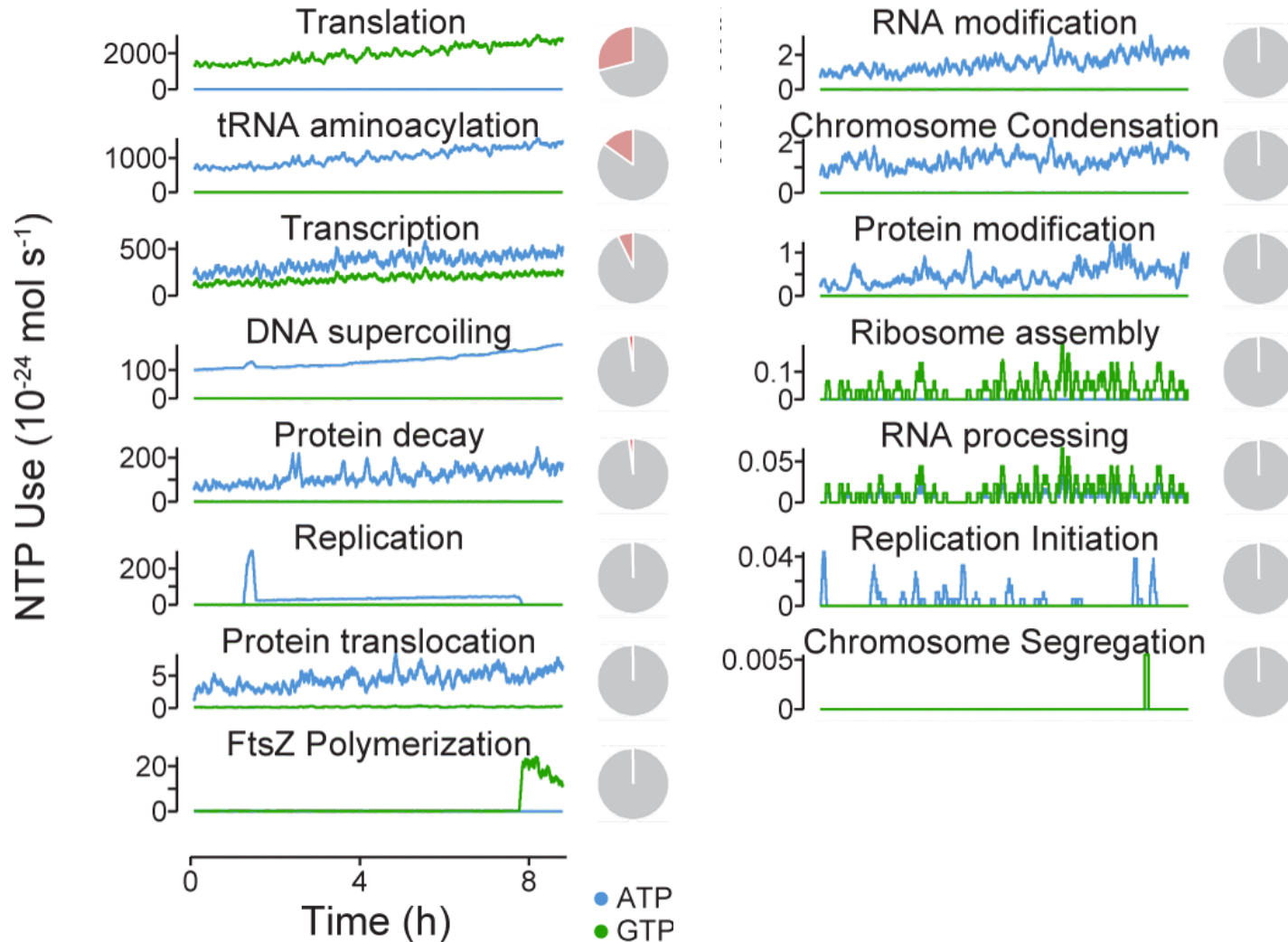
DNA replication, protein occupancy, methylation, & damage
Simulation: Wild-type #1



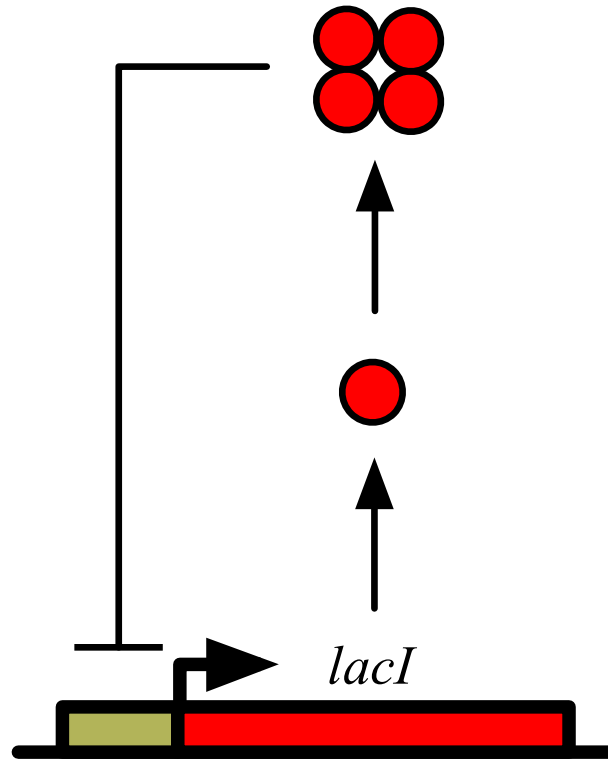
Mature protein monomer expression
Simulation: Wild-type #1



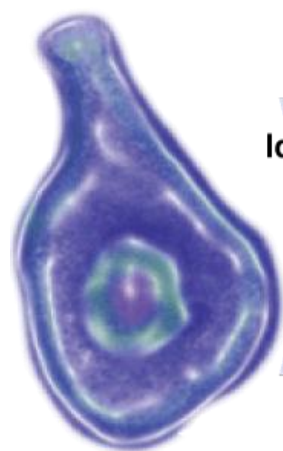
WC models provide novel insights



WC models help design cells

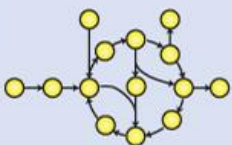


WC models help purpose drugs

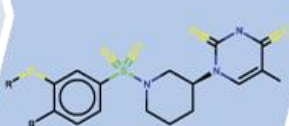


M. genitalium
525 genes

Identify fragile genes
Systems modeling



Identify inhibitors
Bioinformatics

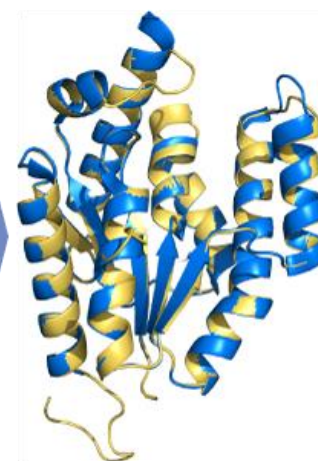
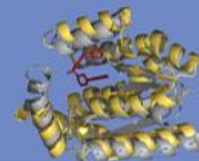


**Assess active
& binding sites
homology**

Bioinformatics

```
--TMEPLTEAYLFFAAARTEHIS  
--DMDIRTEAMLFFAASRREHLV  
DEVITDKAEVLMFYAARVQLVE  
--DVEDHSVHLLESANRWEQVP
```

Assess affinity
Structural modeling



Tmk
1 gene



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Limitations of the *Mycoplasma* model

- Represents one of the smallest bacteria
- Ignores several processes
- Mispredicts several phenotypes
- Methods were ad hoc
- Hard to understand, reuse, and expand
- Time-consuming to build

Toward more comprehensive and more accurate models

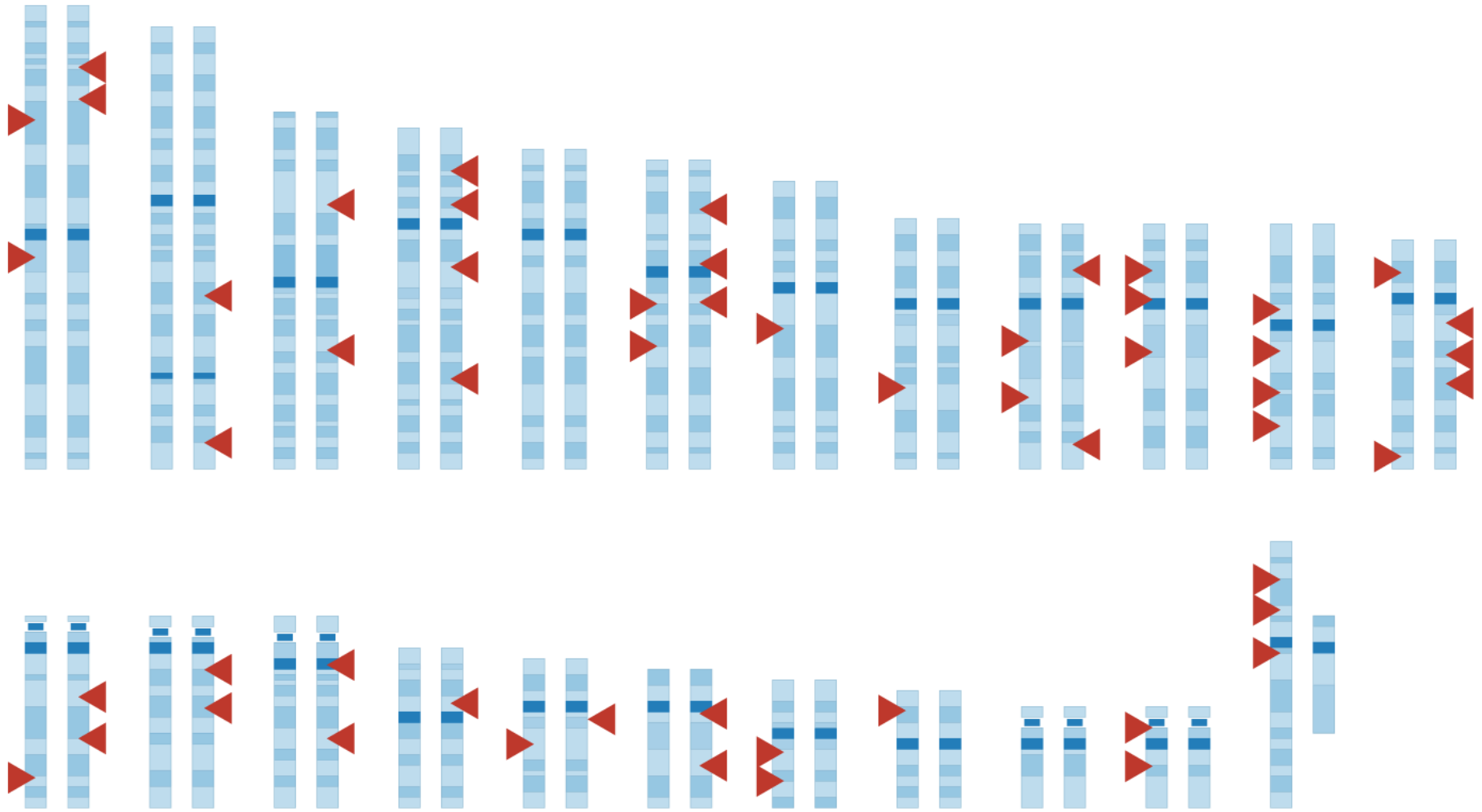


Whole-Cell
Modeling



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Goal: design precise therapy



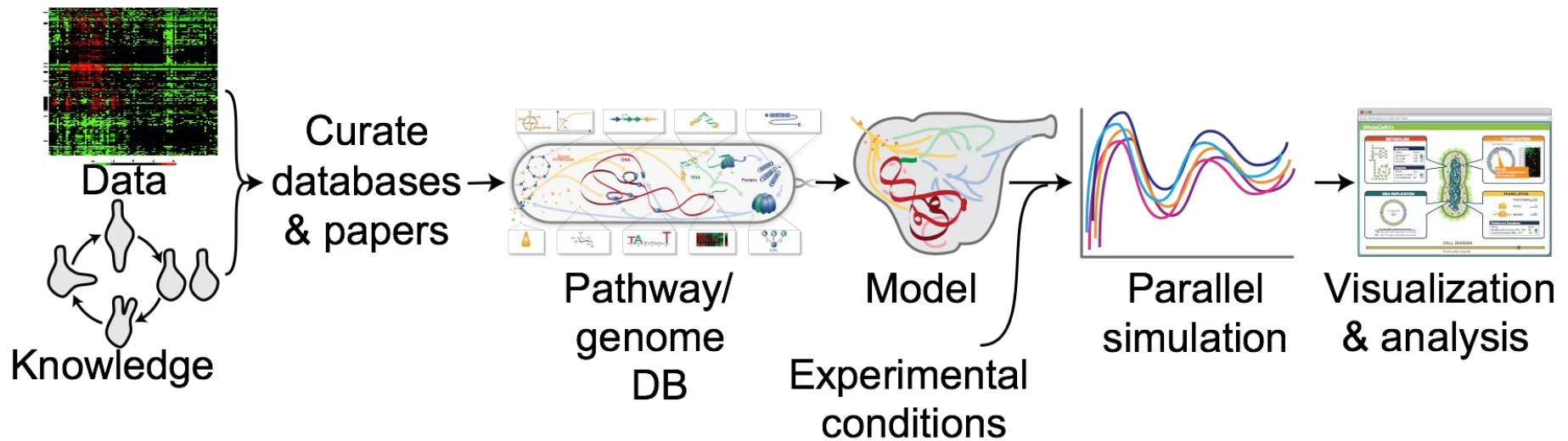
Challenge: H1 hESCs



- Karyotypically normal
- Autonomous
- Well-characterized



Bottlenecks



Bottlenecks

- **Data aggregation:** Hard to find relevant data
 - Data is incomplete, scattered, and insufficient annotated
- **Model design:** Hard to capture multiple scales and describe models modularly
 - Insufficient abstraction and metadata
- **Simulation:** Hard to simulate multiple scales
 - Simulators are only support individual formalisms and are slow
- **Verification:** Little formalism or standardization
- **Collaboration:** Difficult to describe the data, assumptions, and decisions that underlie modeling



Data needed for WC modeling

$$v = k_{\text{cat}} [\text{enzyme}] \frac{[\text{substrate}]}{[\text{substrate}] + K_m}$$

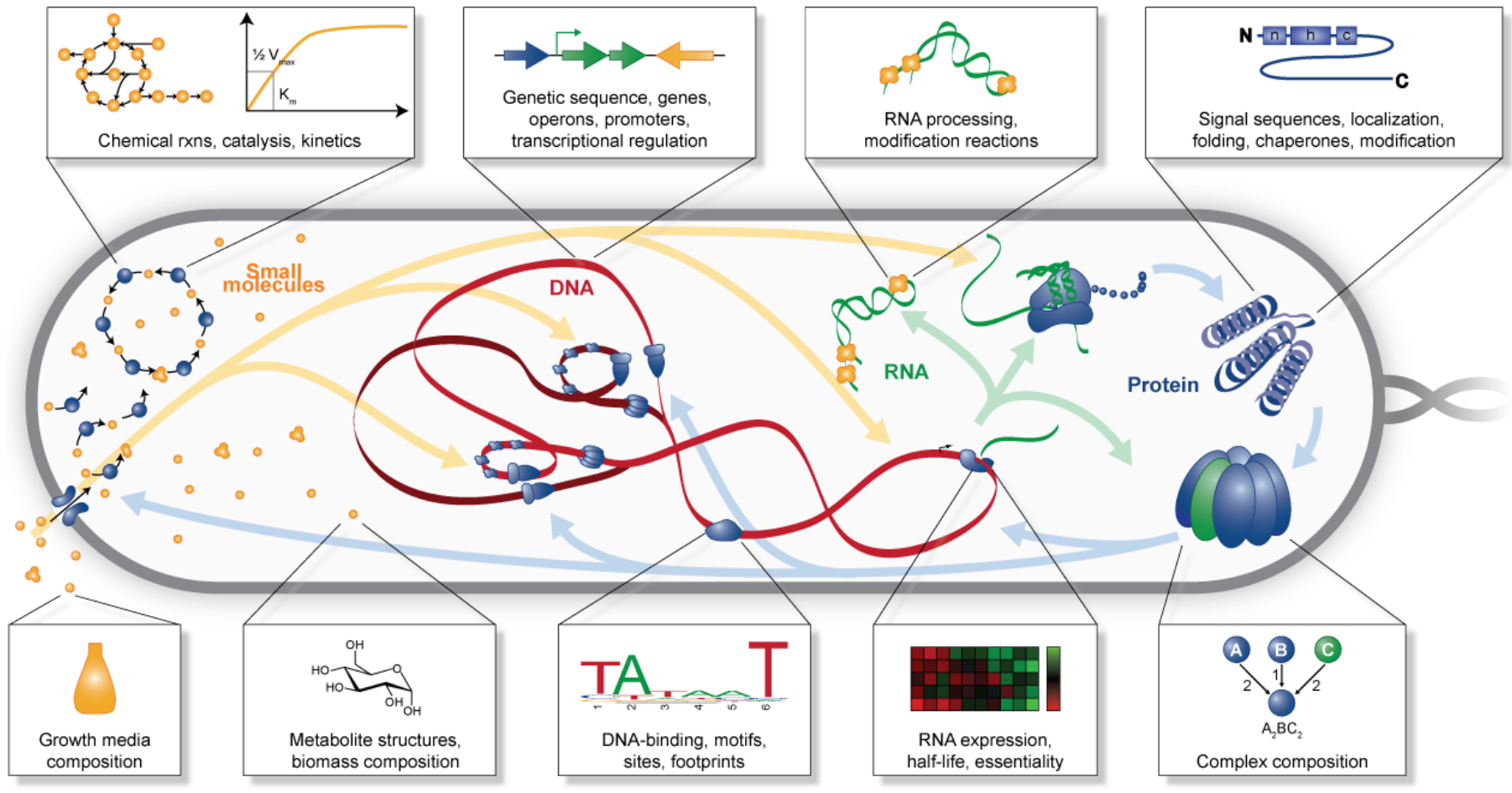
Metabolite
concentrations

Enzyme
concentrations

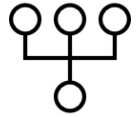
Reaction kinetics



Data needed for WC modeling



Datanator: data integration & discovery



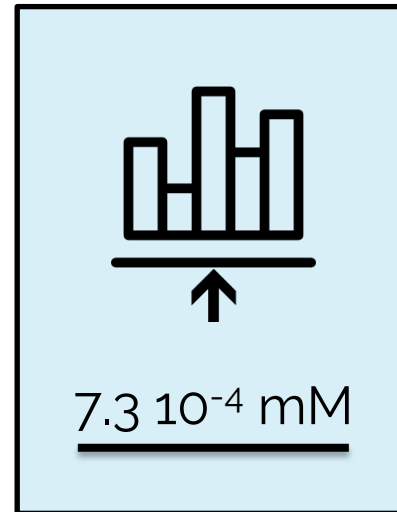
Aggregate



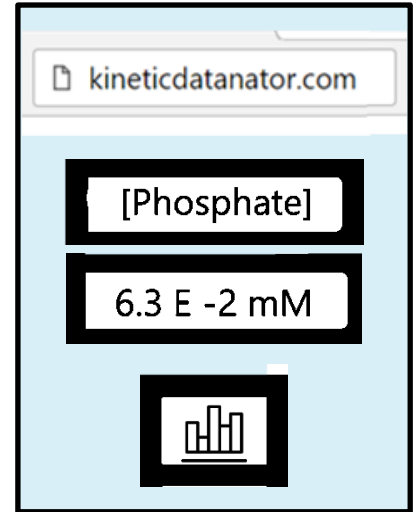
Find



Reduce



Review



Datanator: data aggregation

Metabolites

- ChEBI
- [ECMDB](#), [YMDB](#)
- [PubChem](#)

DNA

- GenBank

RNA

- [Array Express](#)
- [MODOMICS](#)
- [RNALocate](#)
- [RNA MOD](#)

Protein

- COMPARTMENTS
- [CORUM](#)
- Human Protein Ref. DB
- [Pax-DB](#)
- PDB
- PSORTdb
- [RESID](#)
- [UniProt](#)

Interactions

- BioCyc
- DBTBS
- DrugBank
- [JASPAR](#)
- KEGG
- SuperTarget

Taxonomy

- [NCBI](#)

Pathways

- KEGG
- Pathway Commons
- Reactome
- WikiPathways

Rates

- BRENDA
- [SABIO-RK](#)

Datanator: actionable metadata

Measured entity/property

Measured value, uncertainty, units

Genotype

- Taxon
- Genetic variant
- Cell, tissue type

Environment

- Temperature
- pH
- Growth media

Data generation process

- Experimental design
- Measurement method

Data analysis process

- Software
- Version

Metadata

- Authors
- Curator
- Date
- Citation



Datanator: Finding relevant data

Chemical similarity

- Tanimoto index
- Sequence similarity

Genetic similarity

- Whole-genome similarity
- Taxonomic distance

Environmental similarity

- Temperature
- pH

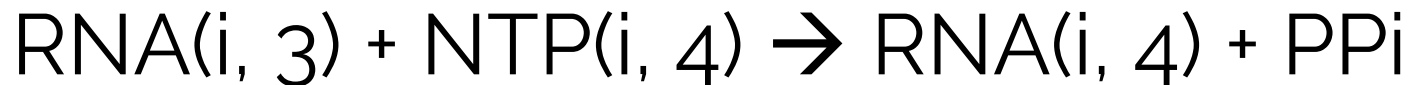
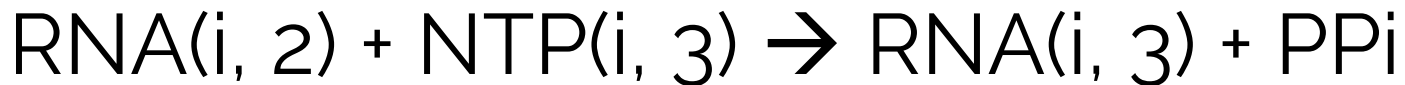
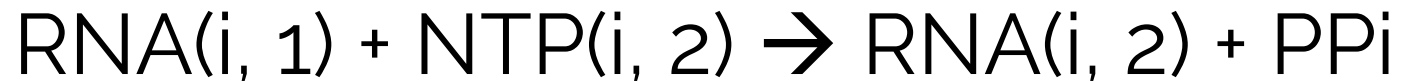
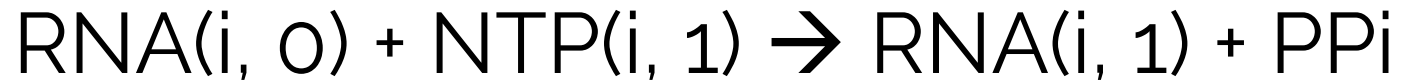


WC-Lang: scalable model descriptions

- Concretely describe composite multi-algorithmic models
- Concrete descriptions of every model element
- Capture data and assumptions underlying models
- Explicit descriptions of mixed granularity / lumping
- Structured description of initial conditions
- User interfaces suited to large models



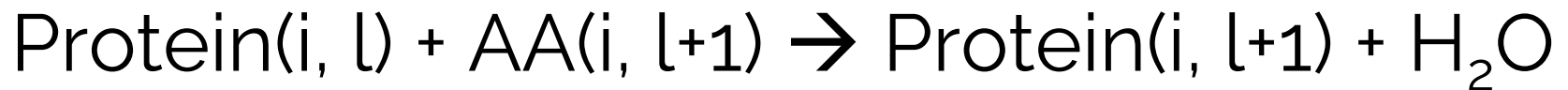
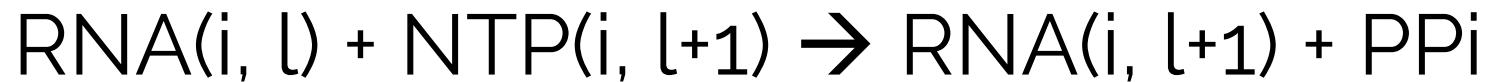
WC-Lang: scalable model descriptions



...



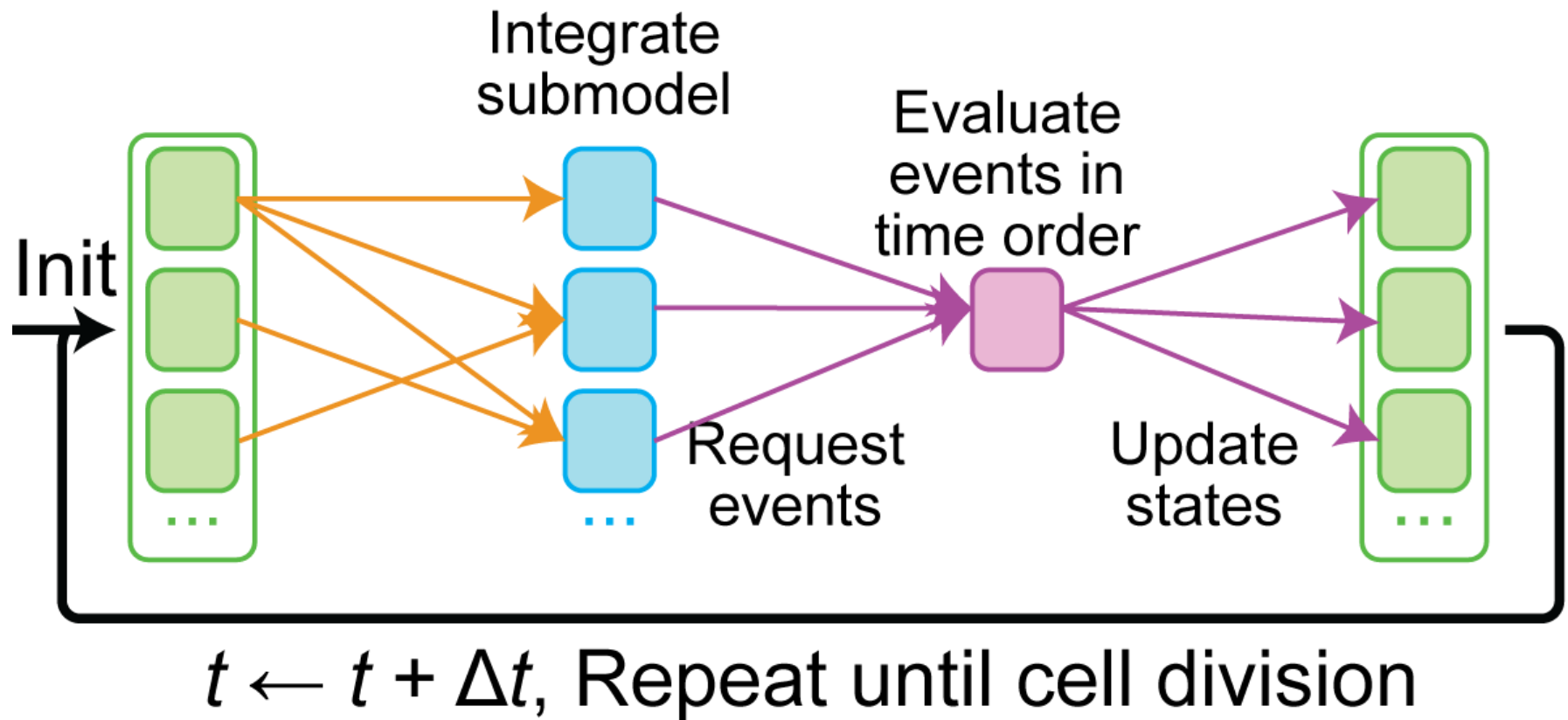
WC-Lang: scalable model descriptions



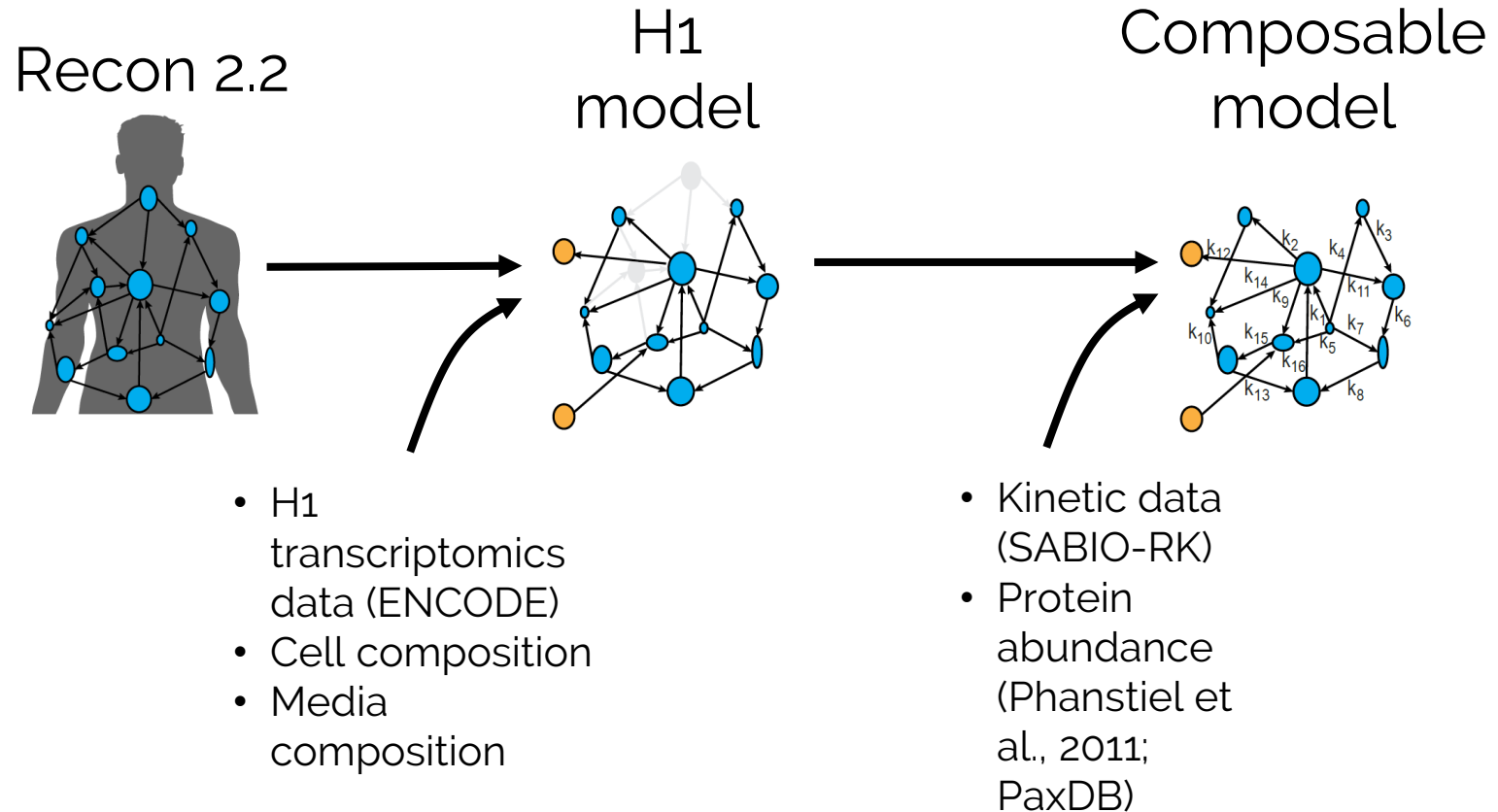
WC-Lang: scalable model descriptions

	Initiation		Elongation		Termination	
SBML	1 per RNA	335	1 per base	~500k	1 per RNA	335
Rules	1	1	1	1	1	1

WC-Sim: scalable co-simulation



H1-hESC model



Summary



Whole-Cell
Modeling



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Availability

- **Code:** code.karrlab.org (GitHub, PyPI)
- **Data:** data.karrlab.org (Quilt)
- **Images:** DockerHub
- **Primer and docs:** docs.karrlab.org
- **Tutorials:** sandbox.karrlab.org



Summary

Bioengineering and medicine needs WC models

WC modeling is becoming feasible

New technologies will enable WC modeling

Pilot models will show the feasibility of bacteria and human models



Tips & tricks



Whole-Cell
Modeling



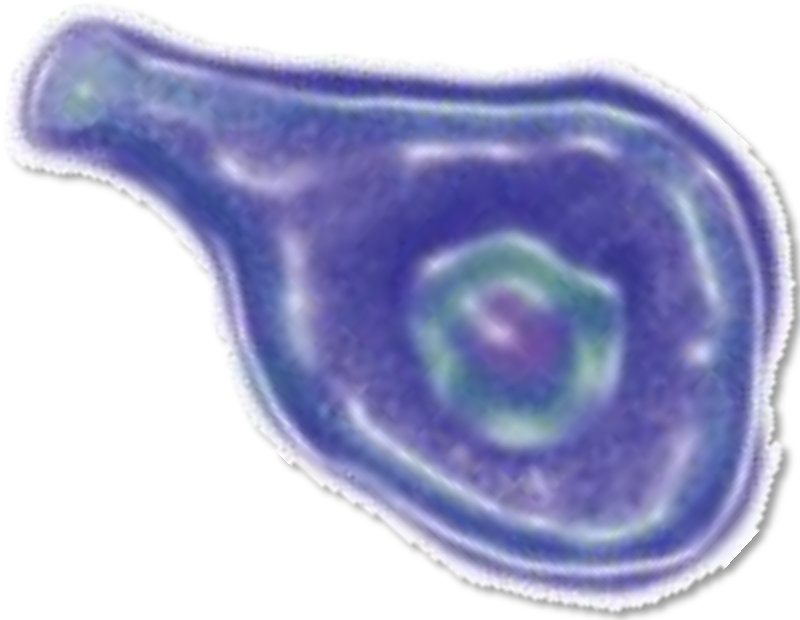
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Challenges to g2p2pop

- Build models from imperfect data
- Capture complexity within and between scales
- Systematically link scales
- Scalably simulate multiple scales
- Collaborate



Stretch goals inspire innovation

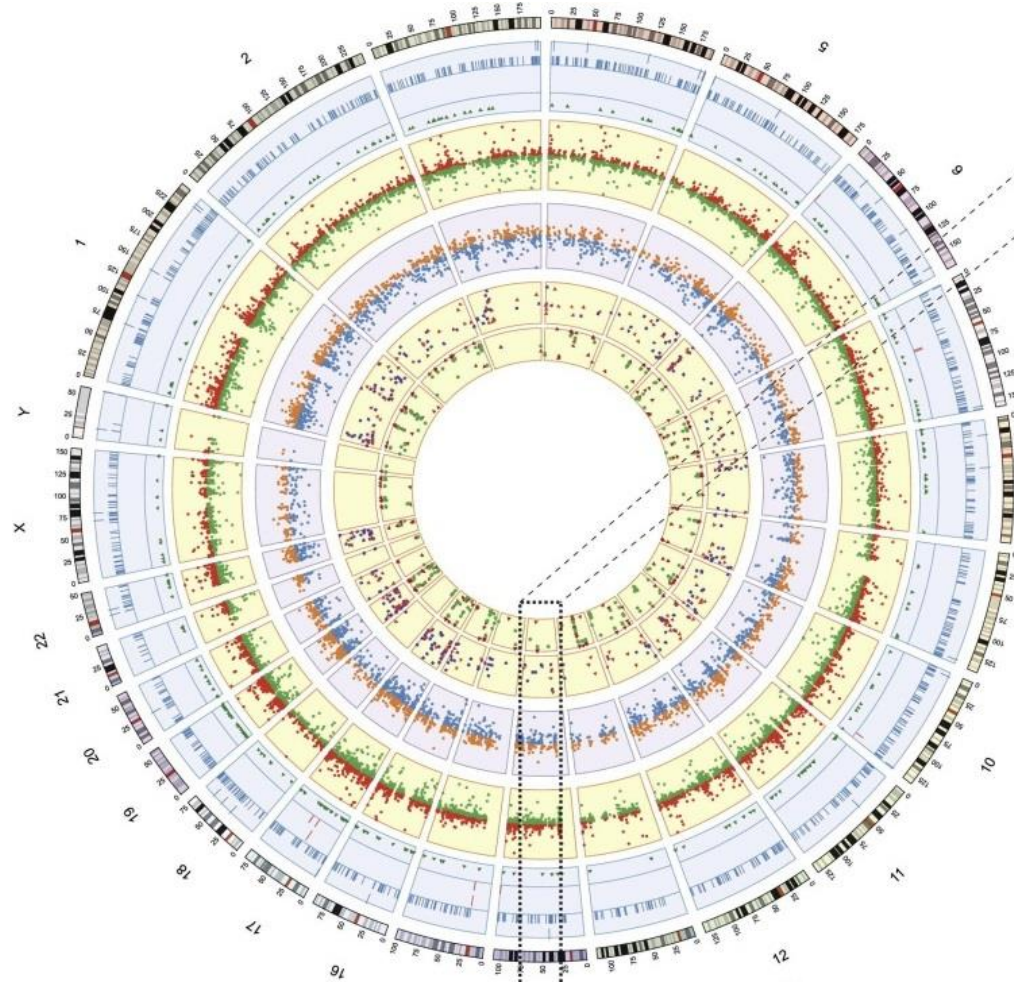


Integration enables great scope and depth

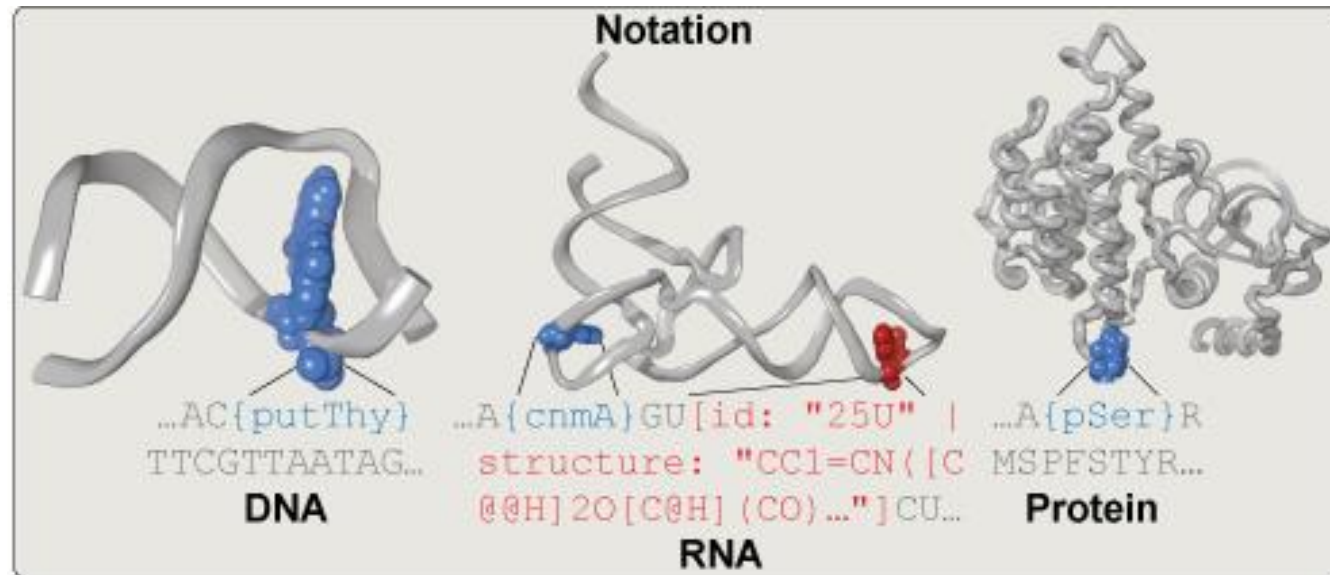
- Data aggregation
- Model composition
- Multi-algorithmic co-simulation
- Modular methods and software
- Interdisciplinary collaboration



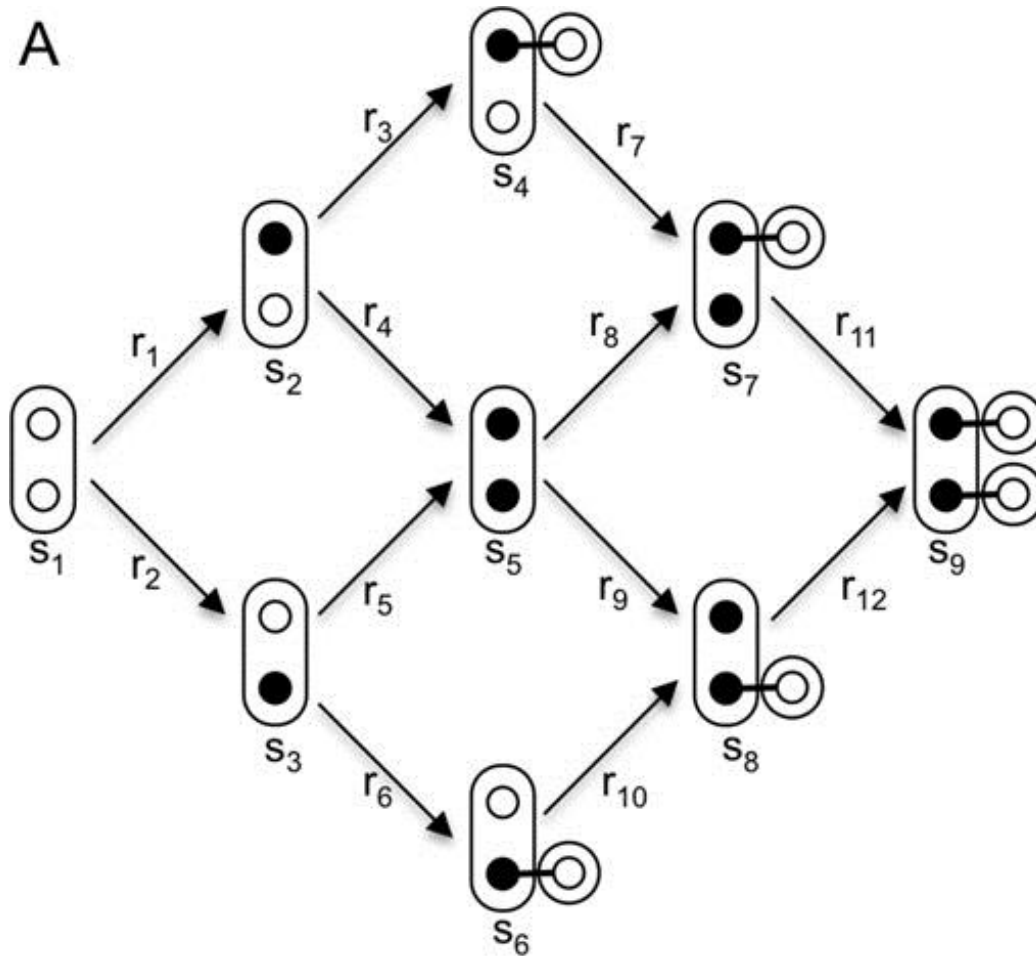
Frameworks enable scalable integration



Common languages enable frameworks



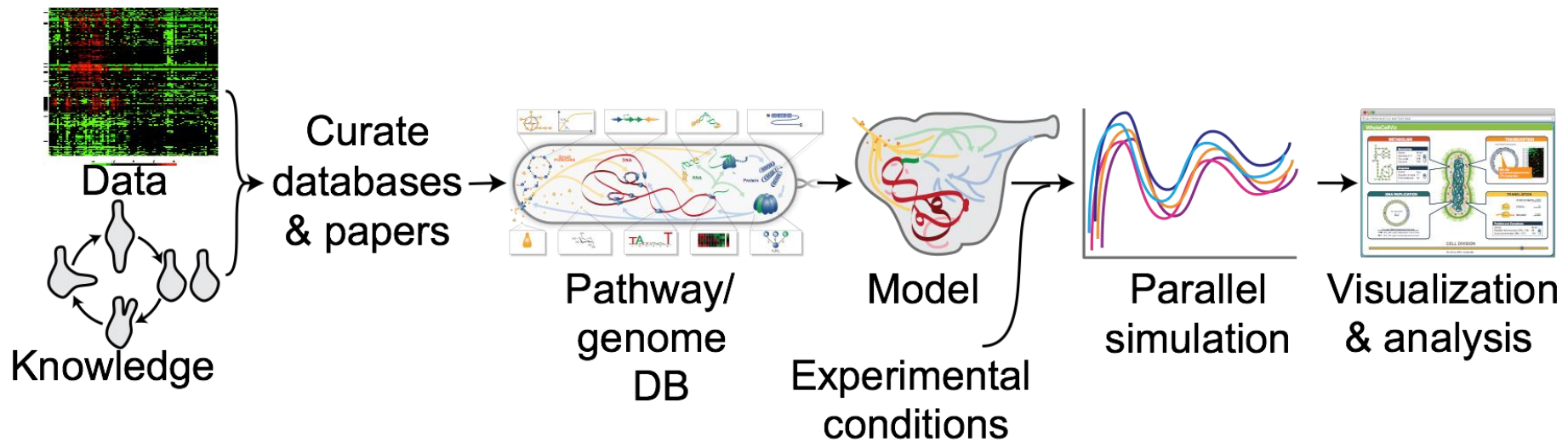
Agent-based modeling can capture complexity



Collaboration enables solutions



Modularity enables collaboration



Sharing promotes collaboration

- **Quilt:** data
- **GitHub:** code
- **PyPI:** packaged code
- **Docker:** computing environments
- **Google Docs, Overleaf:** written documents
- **Google Drive:** other files
- **GitHub issues:** tasks



Common practices ease collaboration

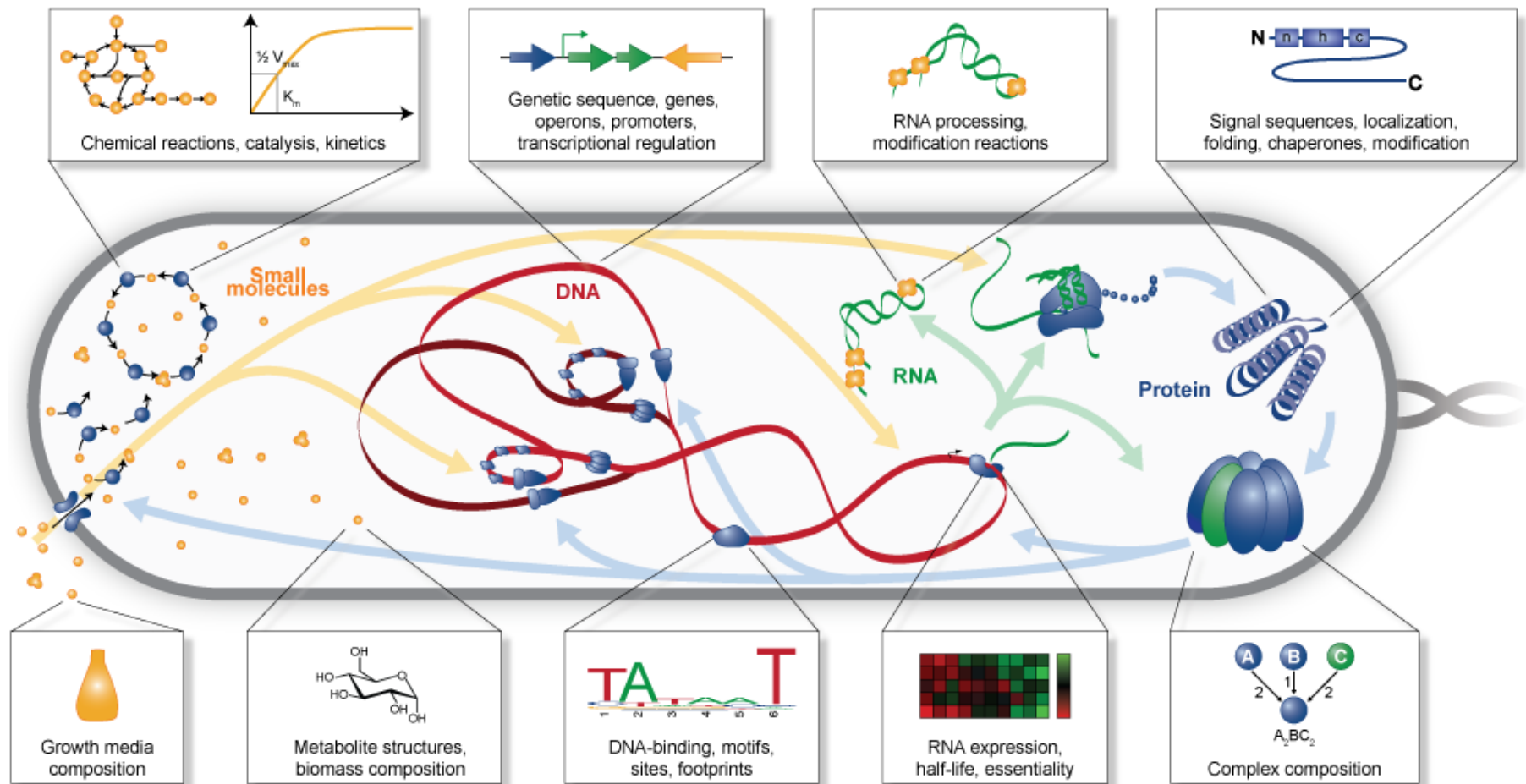
- Interfaces between modules
- Coarse-graining
- Package organization
- Coding, documentation styles
- Software libraries

QC inspires trust among collaborators

KarrLab



Data integration enables modelers to drive science



Summary



Whole-Cell
Modeling



Mount
Sinai

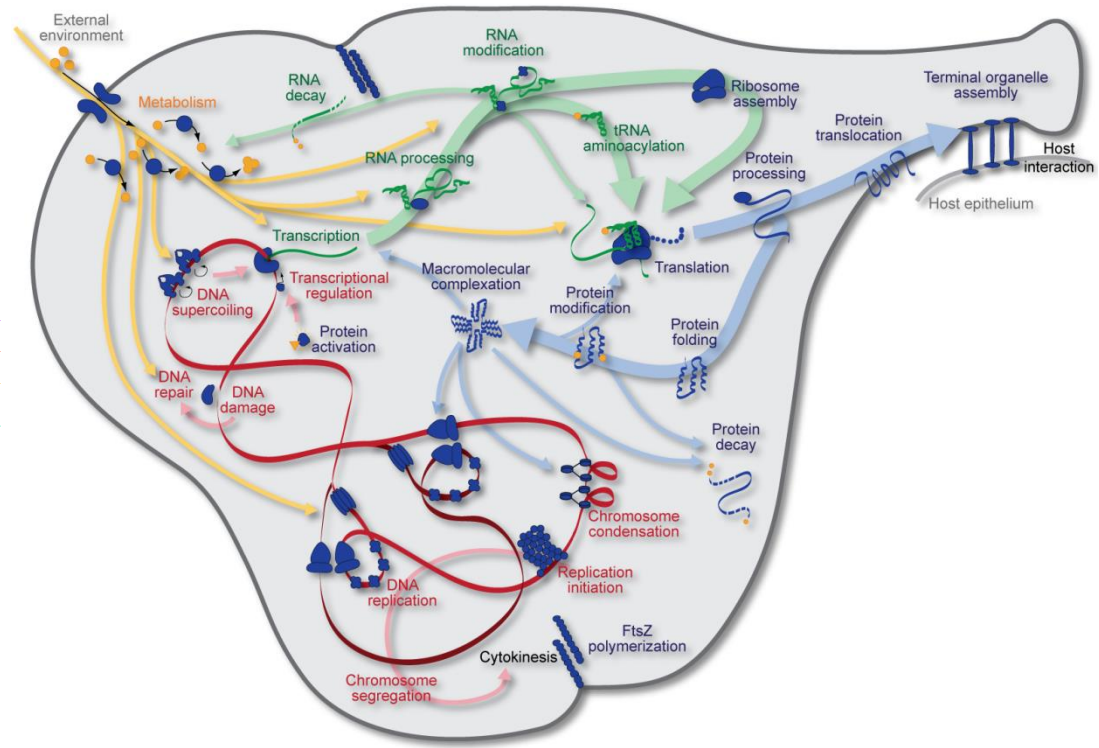
Integration is enabling WC modeling

Genomic and
biochemical data

Pathway
models

Rule-based
modeling

Multi-algorithmic
simulation



Whole-Cell
Modeling



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