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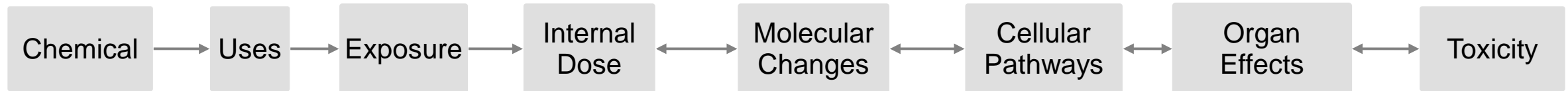
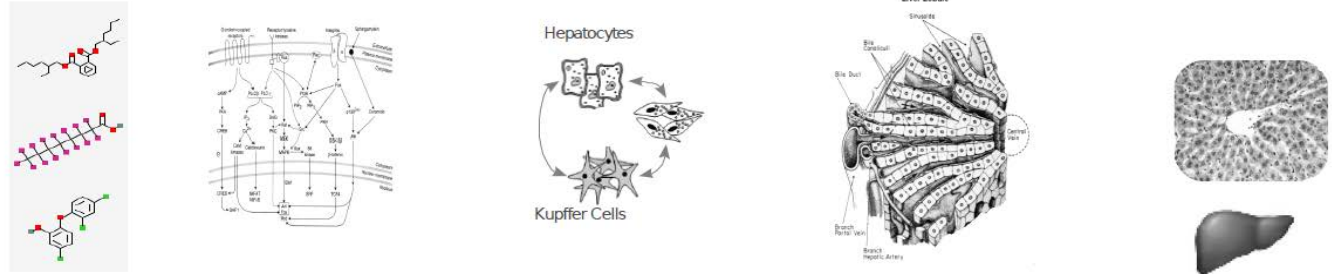
National Center for Computational Toxicology

US EPA

Modelling The Biological Complexity Of Our Environment

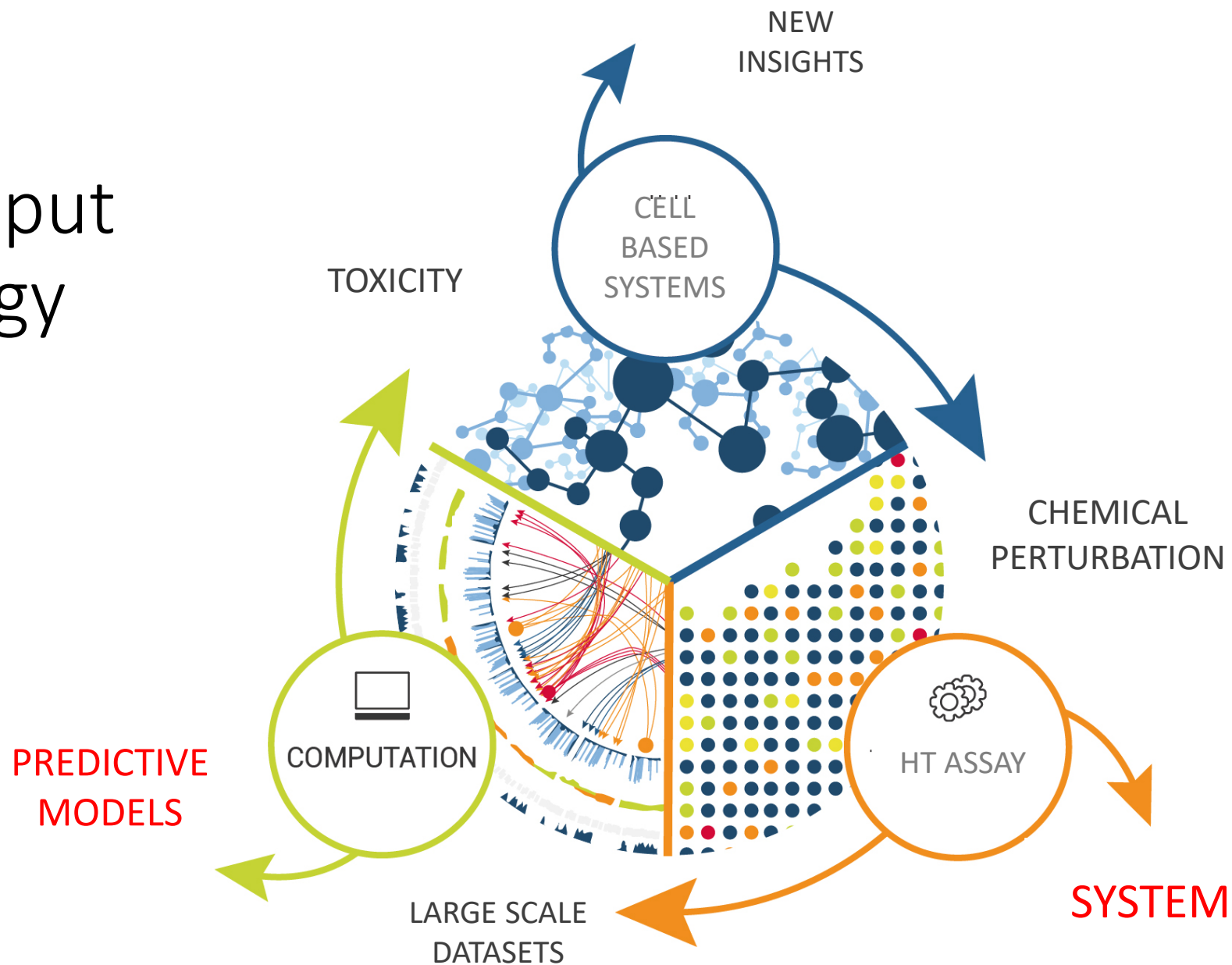
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Complex Pathways to Toxicity



- There are ~ 80,000 chemicals in commerce
- Chemicals are evaluated based on adverse outcomes (derived from animal testing)
- Only ~1000 chemicals have been evaluated systematically via animal testing
- How do we effectively determine the health risks of the remaining thousands of chemicals?

High Throughput Toxicology

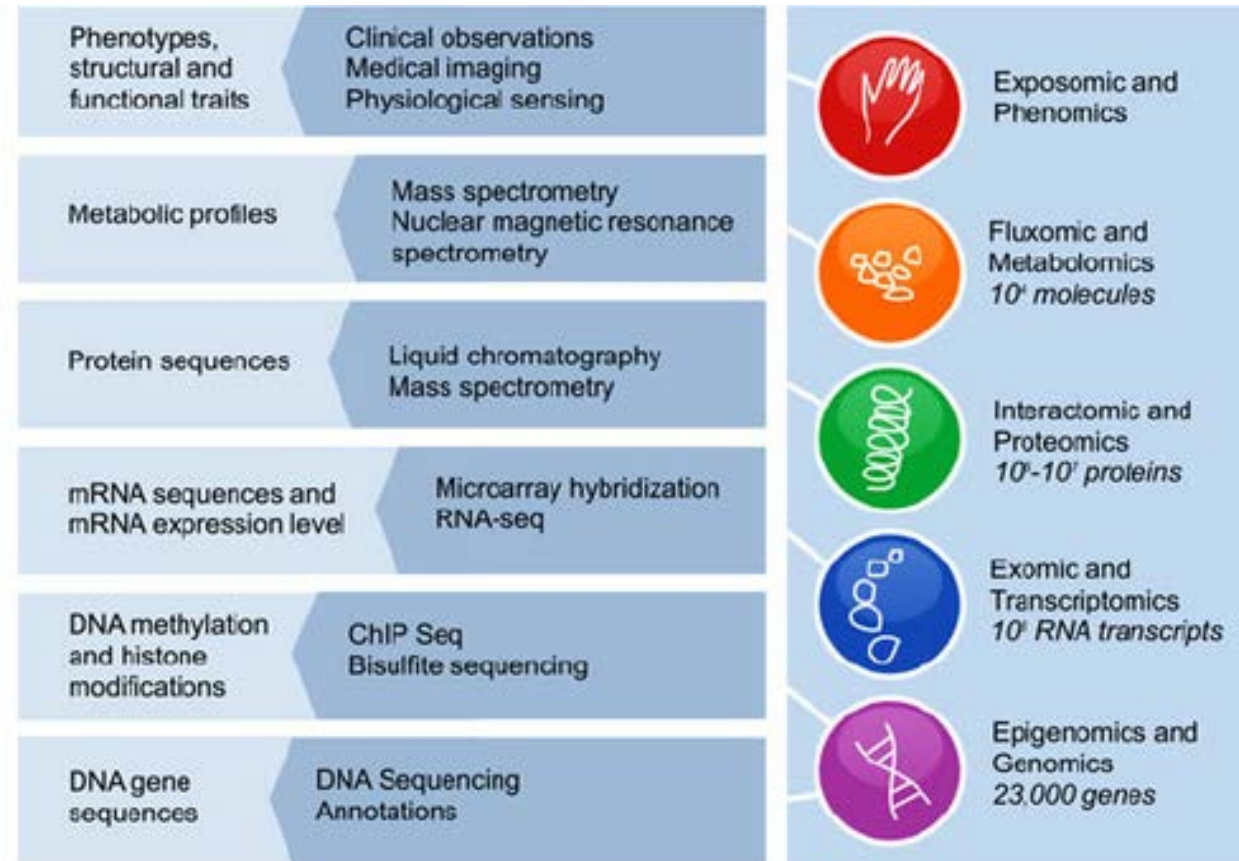


Multiple *-omic* technologies

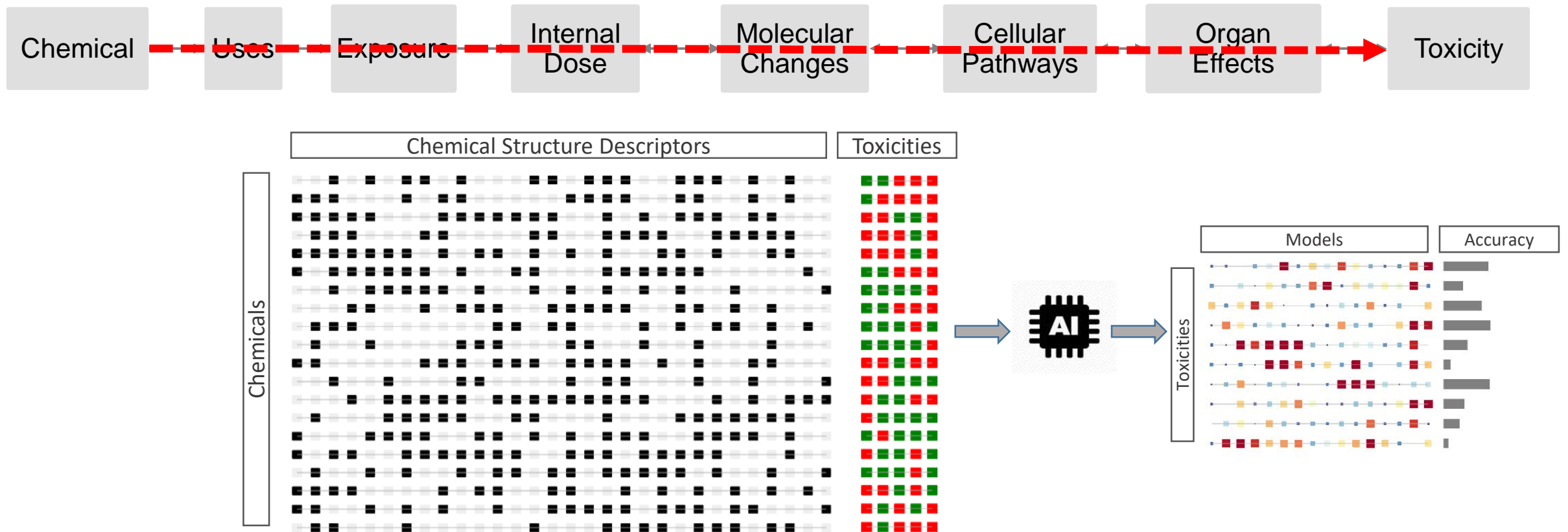
Varying

- levels of maturity/acceptance
- biological coverage
- mechanistic value
- translational utility
- cost \$\$\$

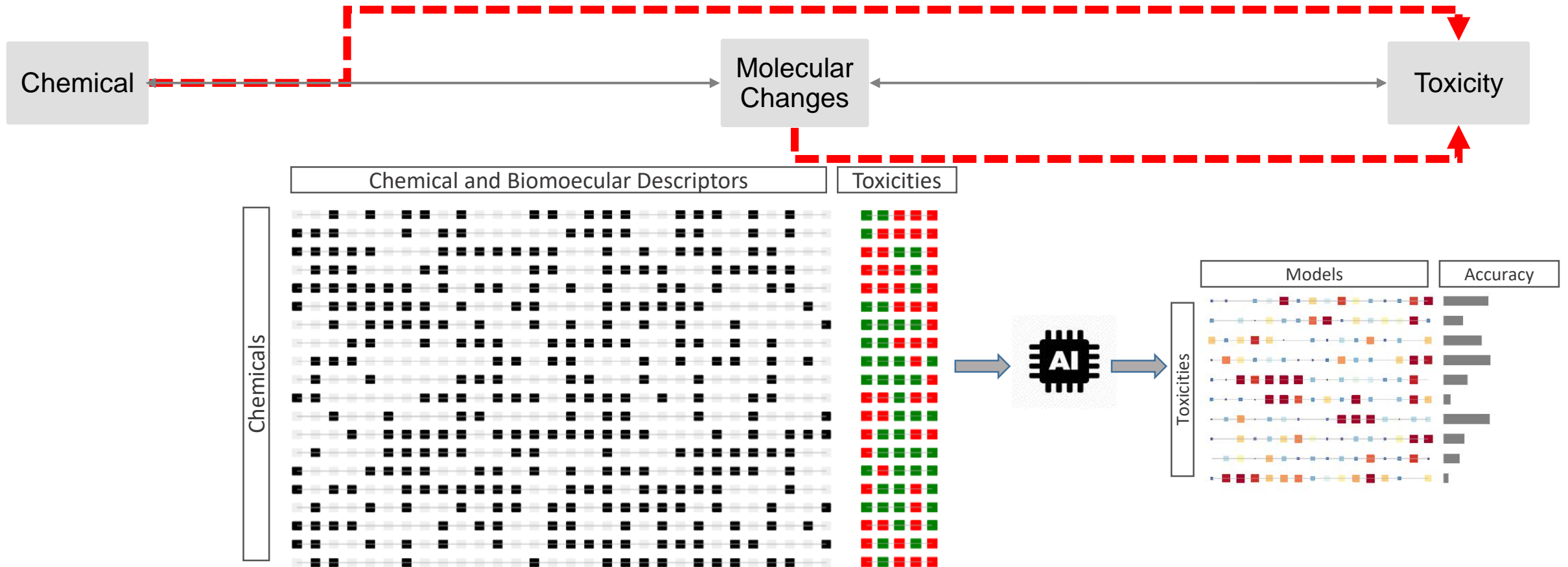
High-throughput transcriptomics



Data-Driven Toxicity Prediction

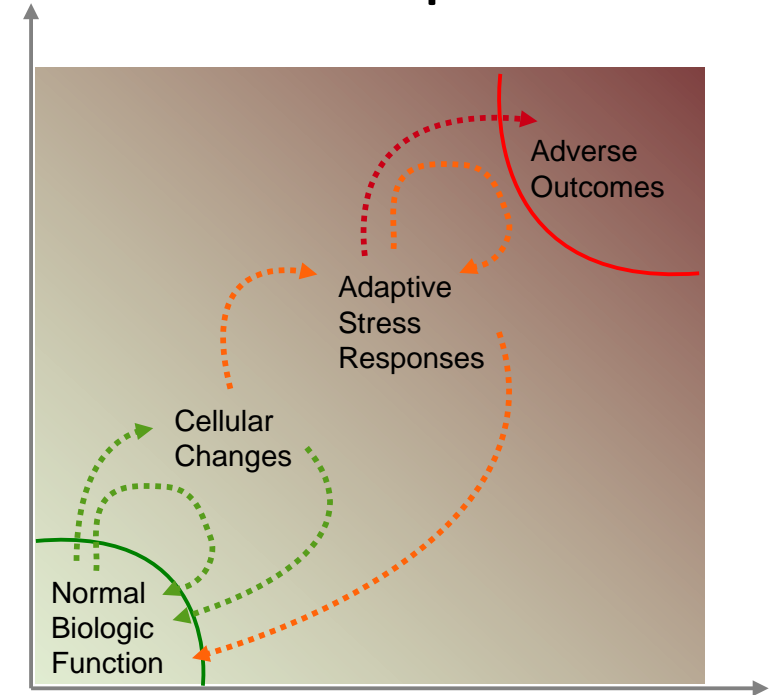


Data-Driven Toxicity Prediction



CAUTION: Large-Scale Data are Complex!

- HTT assays provide a snapshot of system state
- System state (trajectory) can adapt over time and recover or undergo injury
- Analyzing system trajectories reveals “tipping points”
- Tipping points: critical dose-dependent thresholds between adaptation and injury



System Trajectories:

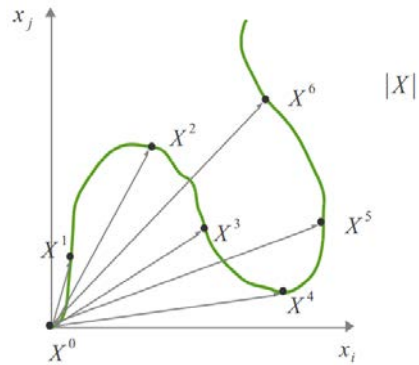
-➡ Some perturbation/
Recovery
-➡ Adaptive stress response/
Recovery
-➡ Adaptive capacity
exceeded/
Cell injury/
No recovery

Using ToxCast™ Data to Reconstruct Dynamic Cell State Trajectories and Estimate Toxicological Points of Departure

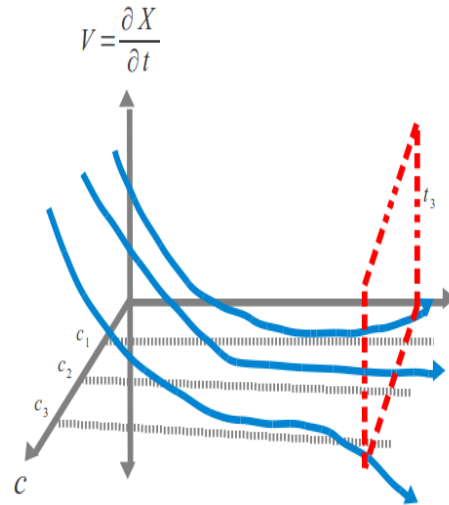
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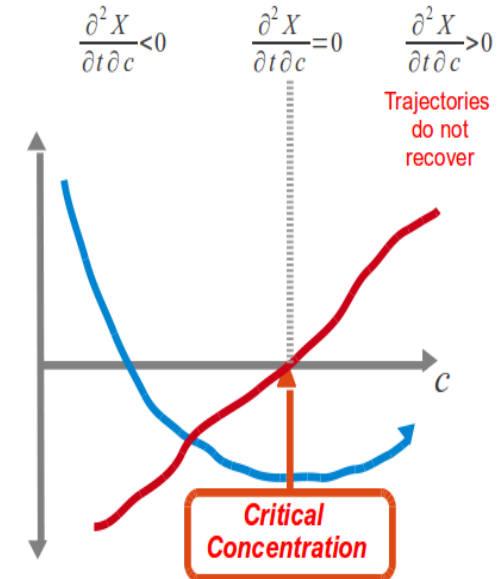
Tipping Point in System Recovery



1 Scalar
Perturbation

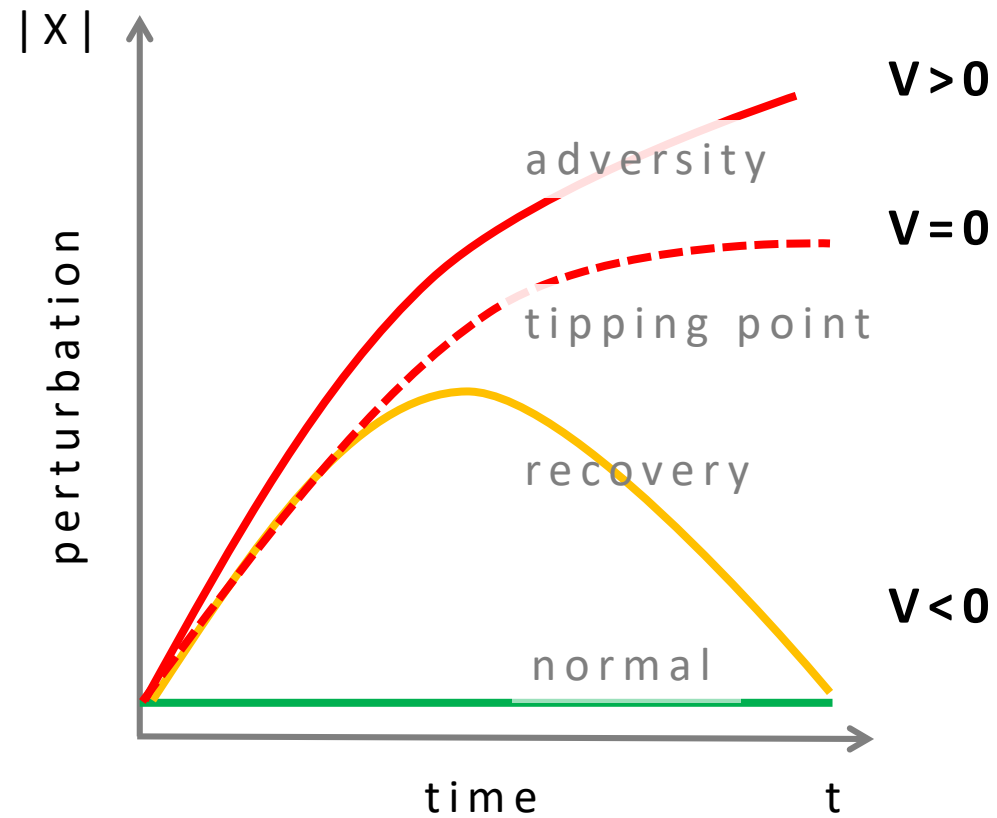


2 Velocity



3 Tipping
Point

System Trajectories & Tipping Points

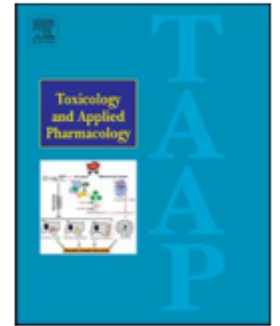




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Defining toxicological tipping points in neuronal network development[☆]

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Summary

- ❑ High-throughput data-driven approaches provide alternative toxicity testing strategy
- ❑ A broad array of computational predictive approaches can be used to predict chemical-induced effects
- ❑ Linking complex disparate data-streams is challenging
- ❑ Key issue: How do we differentiate adaptation from adversity *in vitro*?
- ❑ Tipping point analysis rigorously identifies critical dose-dependent thresholds in system recovery
- ❑ Further experiments underway to evaluate utility

Acknowledgements and Questions



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