

Complex System Modeling and Science-Based Cybersecurity

**Computer Sciences and Information Systems
Sandia National Laboratories, Livermore, California**

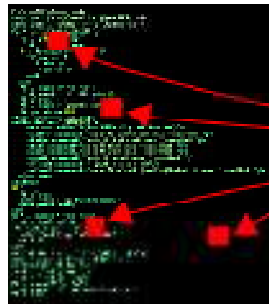
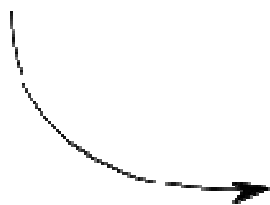
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Securing an arbitrary code is not just hard; it's impossible

- **Restated: Generic code has vulnerabilities that are unprovable and unknowable**
 - *Not* statistical, even in principle
 - Turing completeness demands that a generic code is undecidable

Program



vulnerabilities

- **So now what?**

Complexity makes cyber threats *asymmetric*

Bad Guy needs
to find one



You have to
find them all

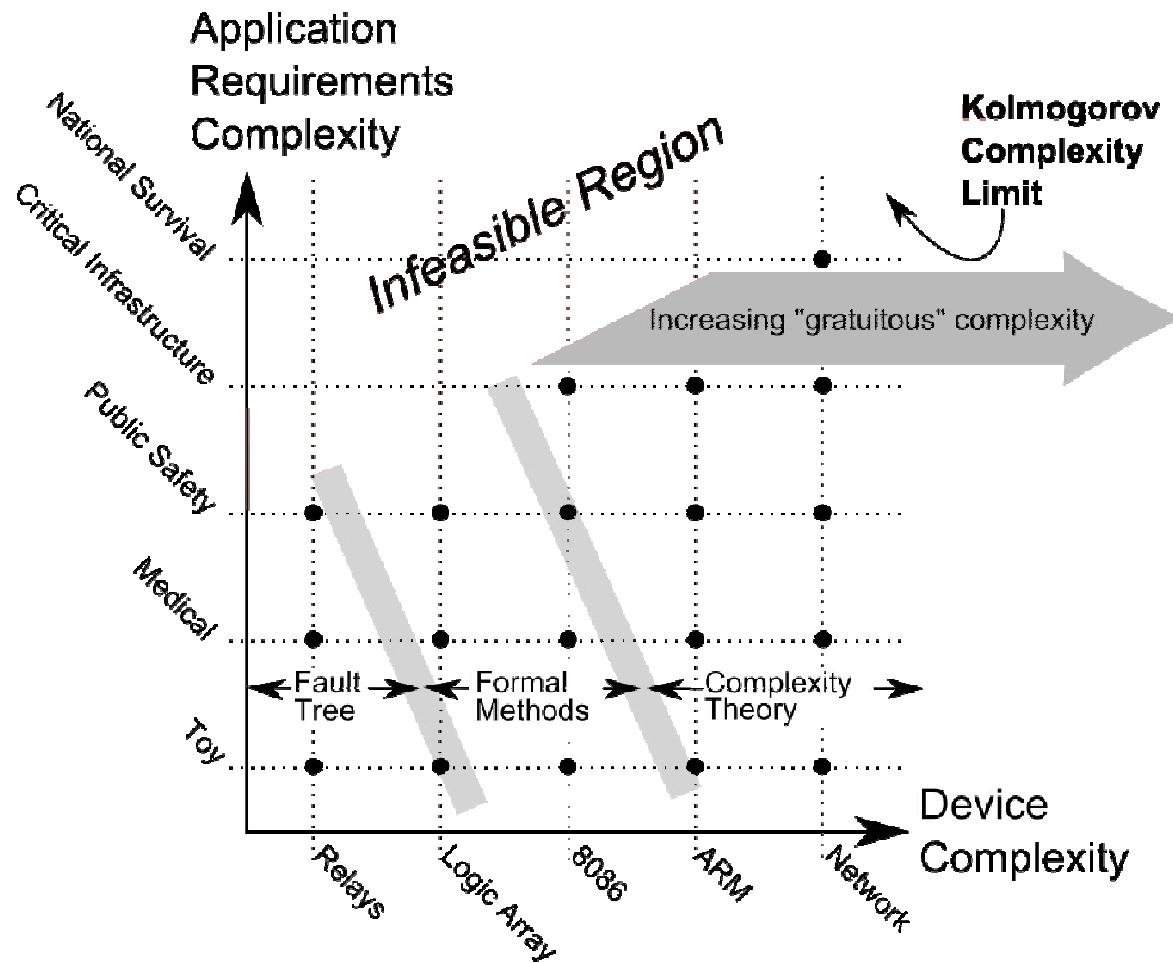
- Developer, user, *and attacker* all don't know where the vulnerabilities are (*undecidable*)
- Worse, attacker may have planted a vulnerability
- **Asymmetry: One vulnerability compromises the whole code**
 - Developer has to find all of them (impossible in general)
- **No one can guarantee “this code is clean” or even quantify improvement**



What is complexity?

- Complex systems are characterized by **large** numbers of interacting entities where even a **few** entities can **strongly** affect system behavior
- Complex systems are **irreducible**; their behavior is **emergent** and not evident *a priori*, but is accessible via **observation** and **simulation**
- **Examples are ubiquitous**
 - Living things and ecosystems
 - Human societies, economies, and institutions
 - Highly engineered artifacts – e.g., airplanes, NWs
 - Large-scale infrastructure – e.g., power grids
 - Computer software, hardware, and networks

Complexity space illustrates tradeoffs in device engineering and analysis





How formal methods work

- **Feed a formal methods tool with a mathematical model of the system (formal specification) and the claimed properties of the system**
 - Automated theorem proving
 - **Uses logical truths and inference to prove the properties of the system**
 - Model checking
 - **Partial order reduction, symbolic manipulation, etc. to reduce the state space**
 - **Exhaustive checking of the reduced states**
- **If a property can be proved false, a counter-example is also provided**



Formal methods are a bridge to complexity, filling an important gap

- **Formal methods use computer analysis to verify digital systems rigorously and exhaustively**
 - Applicable to less complex systems that are still beyond the reach of manual analysis
 - Widely used in high-consequence industrial applications such as aviation and medical devices
- **Verification of components does not generally translate to verification of whole system**
- **Irreducible complexity enters when exploring entire state space is infeasible**
 - Reliability and security assertions become probabilistic
- **Both formal verification and complexity science are vital for gaining confidence in digital systems**



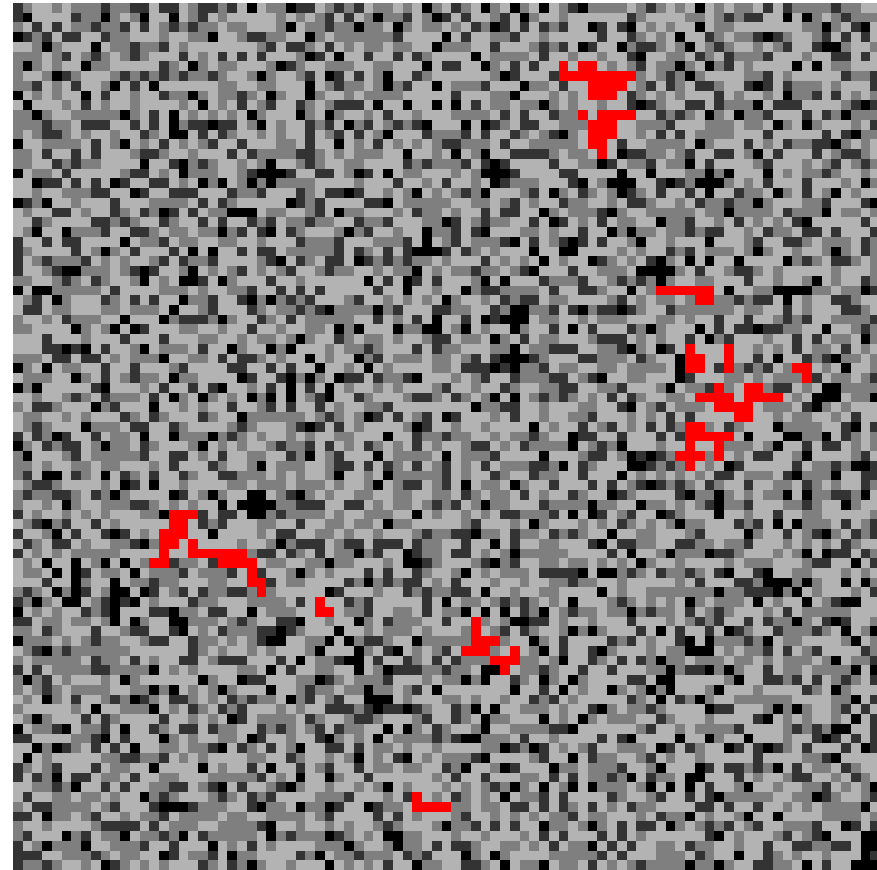
Complexity science offers a new perspective on modeling and design

- **Most real-world systems are too intricate to analyze directly; they are irreducible**
- **Reductionism requires “bottom-up” understanding**
 - Use expert knowledge to model component entities
 - Validate system model vs. observations
 - Make each component entity as reliable as possible
 - Formal methods are the pinnacle of this approach
- **Complexity science provides “top-down” insight relating system structure to emergent behavior**
 - New modeling paradigm: Identify entities by abstraction from idealized models with known emergent behavior
 - New design paradigm: Build real systems based on models with desired emergent behavior



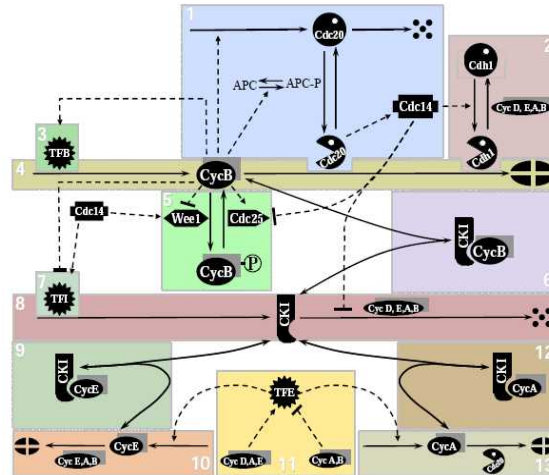
Self-organized criticality is a simple example of emergent behavior

- “Sandbot”: cyber model of coordinated malware
- **SOC** is *spontaneous* development of multi-scale phenomena with power-law distributions
 - Similar to thermodynamic criticality but without tuning
- **Illustrated by sandpile model: physics-like CA**
 - Sand is sprinkled randomly
 - Avalanches occur at all scales



Complexity is a fact of “life”

- **Biological phenomena are a prototype and inspiration for many complex domains**
 - Life involves a large chemical regulatory network



Eukaryotic
cell-cycle
regulation

- “Game of Life” model is based on population dynamics
- Bio concepts pervade computing (viruses, mutations)
- **Biology typifies complex couplings of manmade systems – economy, energy, cybersecurity**



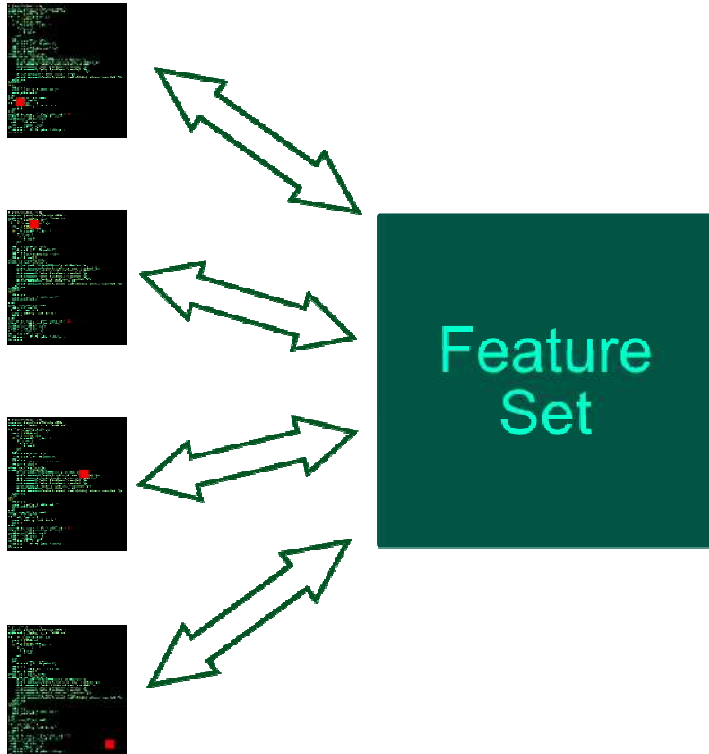
Robustness is key to understanding real-world systems with “organic” behavior

- **Highly optimized tolerance (HOT):** Systems *designed or selected* to perform well despite perturbations
- **Robustness is necessary for biological evolution and for effective engineering**
- **HOT systems exhibit power-law distributions like SOC but have organic structure (not self-similar)**
- **Adapted robustness to one set of perturbations induces **extra fragility** to different perturbations**
- **Indeed, rare but catastrophic failures are seen in highly engineered/evolved systems**
 - Electrical blackouts, cyber shutdown of Estonia, financial panics, hacker penetration of bank database, etc.

Observation #1: A program's feature set has many implementations

Implementations

Input/Output

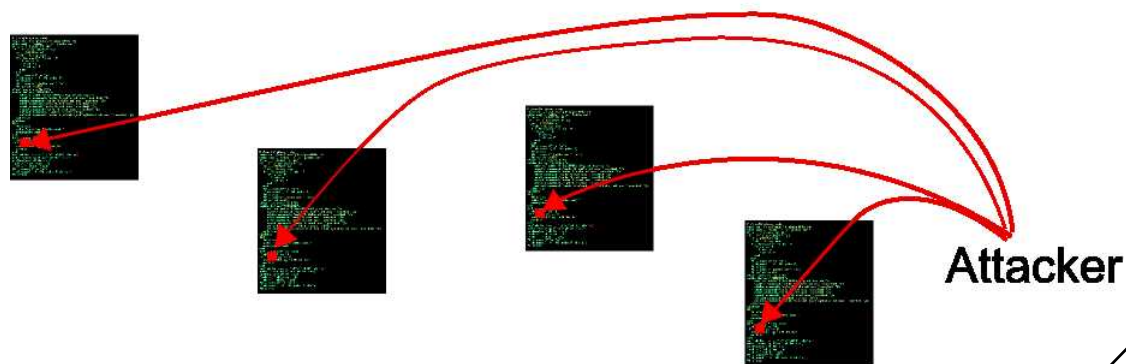


- Feature set is defined by a test suite
- Test suite verifies that an implementation conforms to desired functionality
- Test suite is a sample; cannot realistically cover all possible input/outputs
- Vulnerabilities arise from untested input/outputs
- Any feature set has infinitely many implementations
 - Finite large number if size is bounded

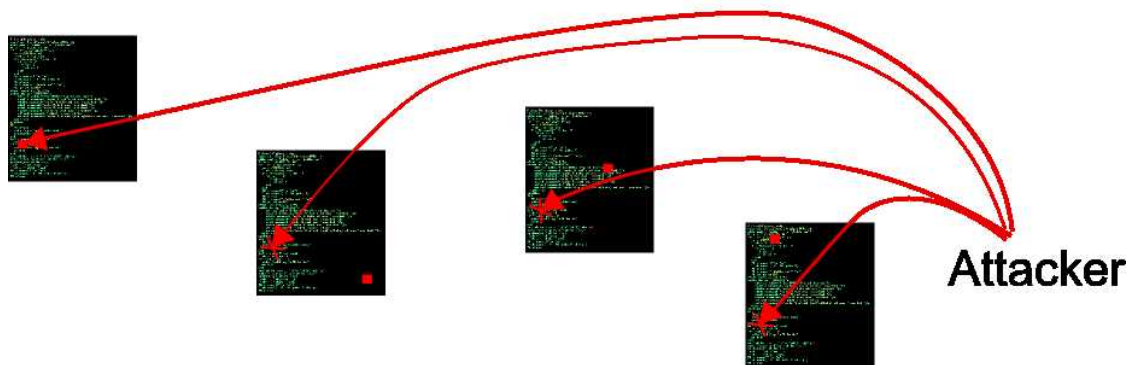
Observation #2: Ensemble of instances permits the formulation of statistics

- Assume: Multiple implementations randomize security holes
- Ensemble of multiple-version, “randomized” undecidable codes allows formation of security *improvement* statistics

Monoclonal



Diverse



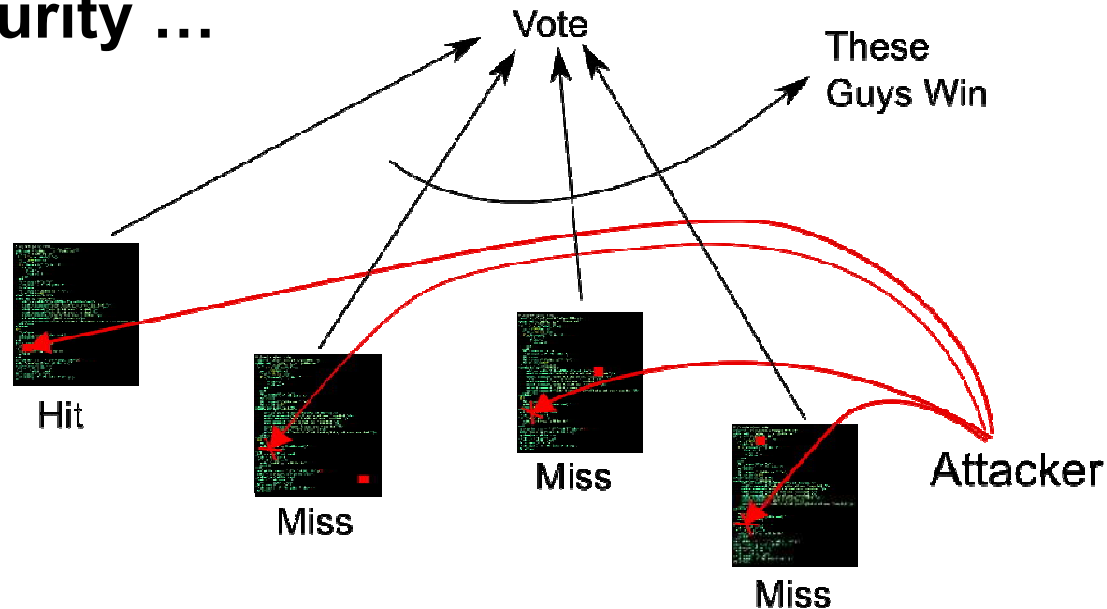
High-reliability systems can be constructed from “*N*-version software”

- **Space Shuttle: 4 computers, identical software, different hardware, same design**
 - Focus is on hardware faults
- **Similarly, software redundancy used mostly for control systems up to now**
 - *N*-version software: Multiple versions implemented to the same feature set by different developers
- **Models of *N*-version software view the control system as a stochastic process that walks the code graph of the software**
 - Control system takes the place of a “fuzzer”



Similarly, *N*-version software can quantifiably improve cybersecurity

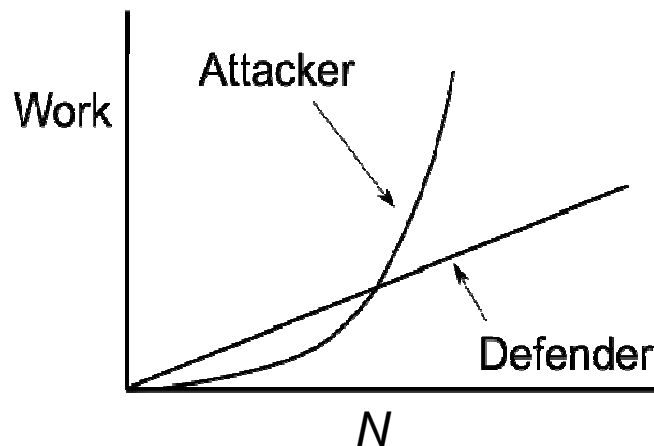
- Clear generalization of *N*-version reliability to cybersecurity ...



- ... but there are important differences requiring enabling technology
 - Compromised versions must be removed and replaced
 - Hand-made new versions are time-consuming and expensive
 - May repeat previous mistakes

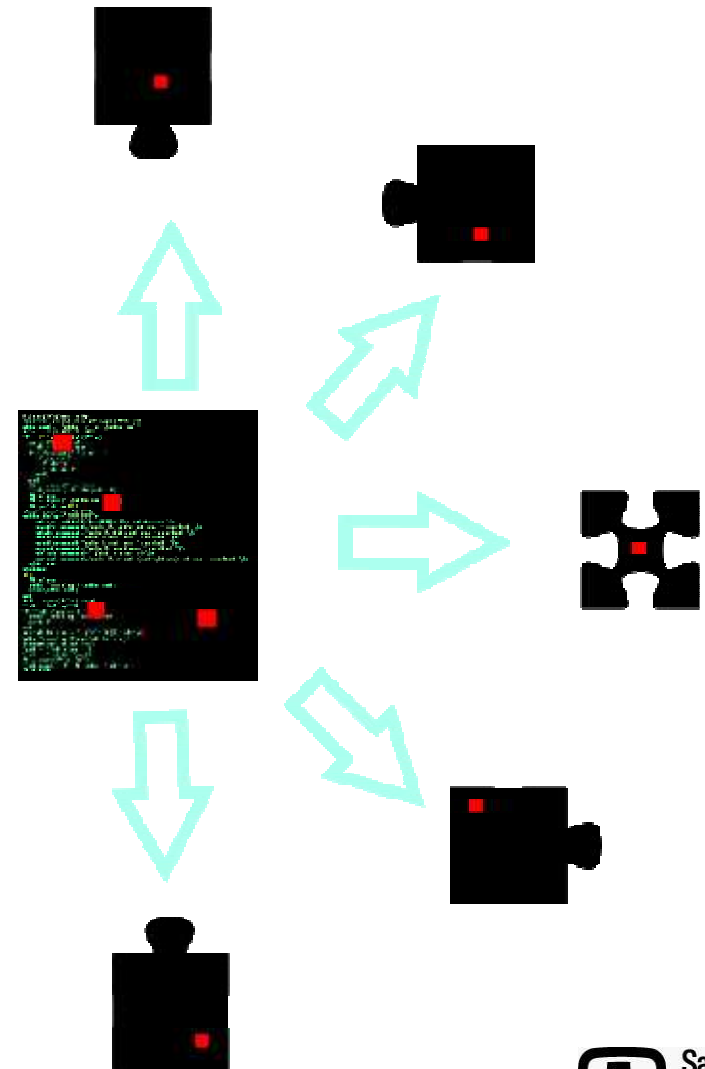
Simple statistics arise from an ensemble of undecidable programs

- On a specific feature set F there is a probability P_v that a particular sample from the set of implementations of F will be susceptible to vulnerability v . For a voting ensemble of size N :
 - The probability of success for the attacker is $(P_v)^{N/2}$
 - The attacker “work” is the expected number of tries: $(P_v)^{-N/2}$
 - The work for defender is the cost of producing N implementations: $\sim N$



A simple example: Diverse software can be constructed from components

- **Component-based codes automatically conform to a feature set if the constituent components conform to their individual feature sets (semantic interfaces)**
 - Multiple implementations of the code amount to multiple versions of components
 - Components can be mixed and matched to form a combinatorial number of code implementations

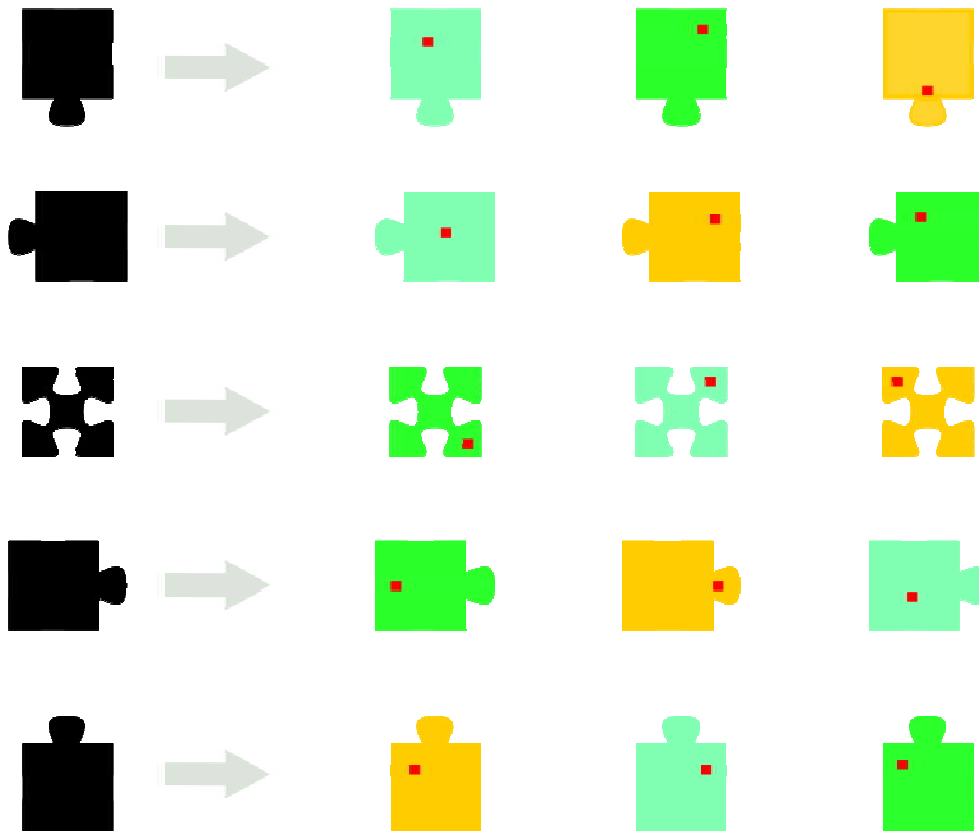


Living systems adapt to cope with unknowable attacks

Genome

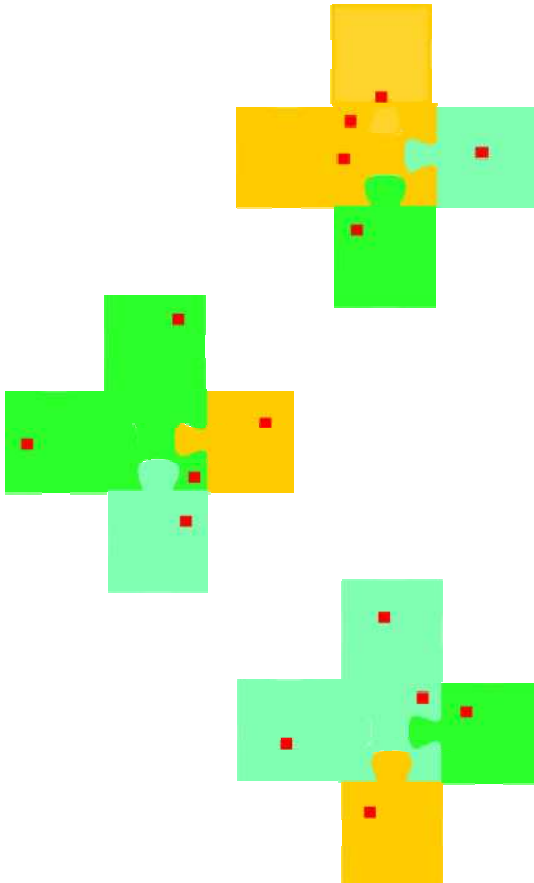
Alleles

- A component type is similar to a gene; component implementations are similar to alleles of a gene



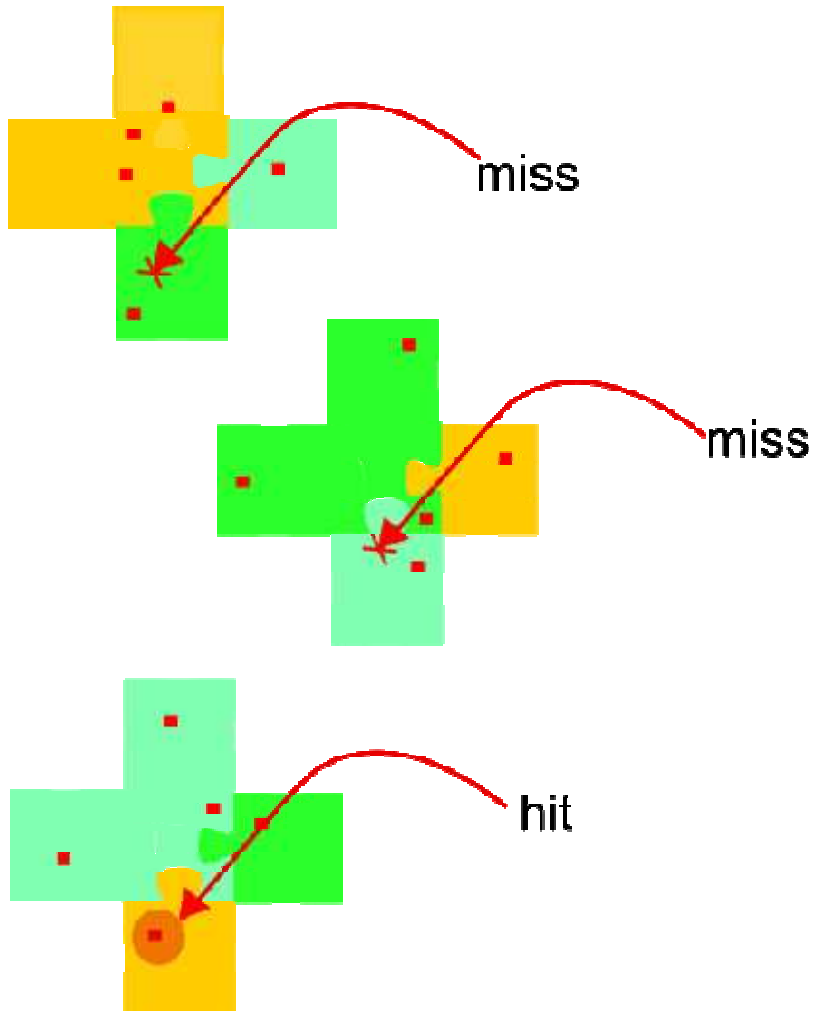


Reassemble alleles into individuals



- Different alleles can be assembled into new individuals that have “randomized” security holes
- New individuals are differently vulnerable and potentially adaptive
- Excess functionality and planted vulnerabilities can be “annealed” away

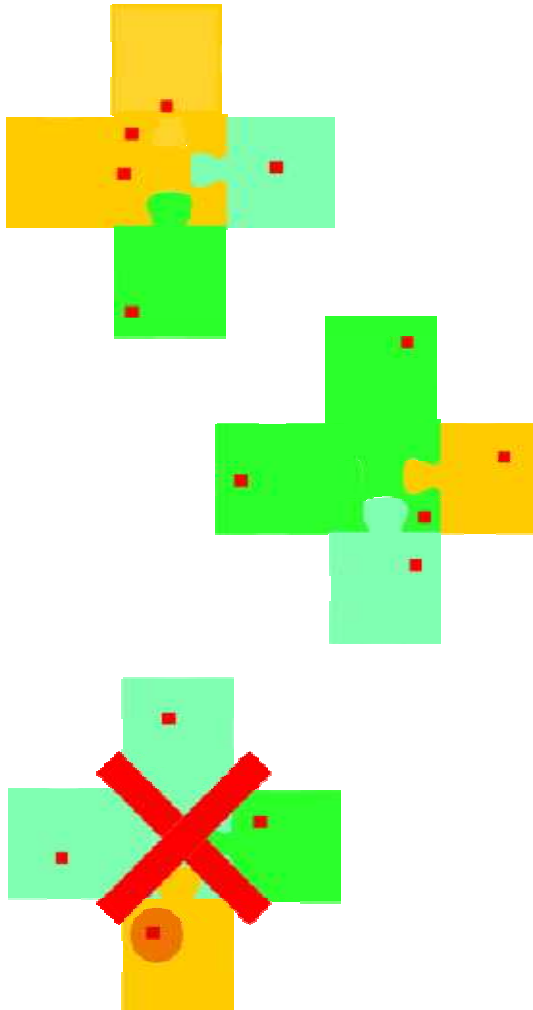
Compare responses from individuals



- Now different individuals will produce the same feature set but react differently to attacks



Evolve new and more robust individuals



- Eliminate the one with the differentiated response

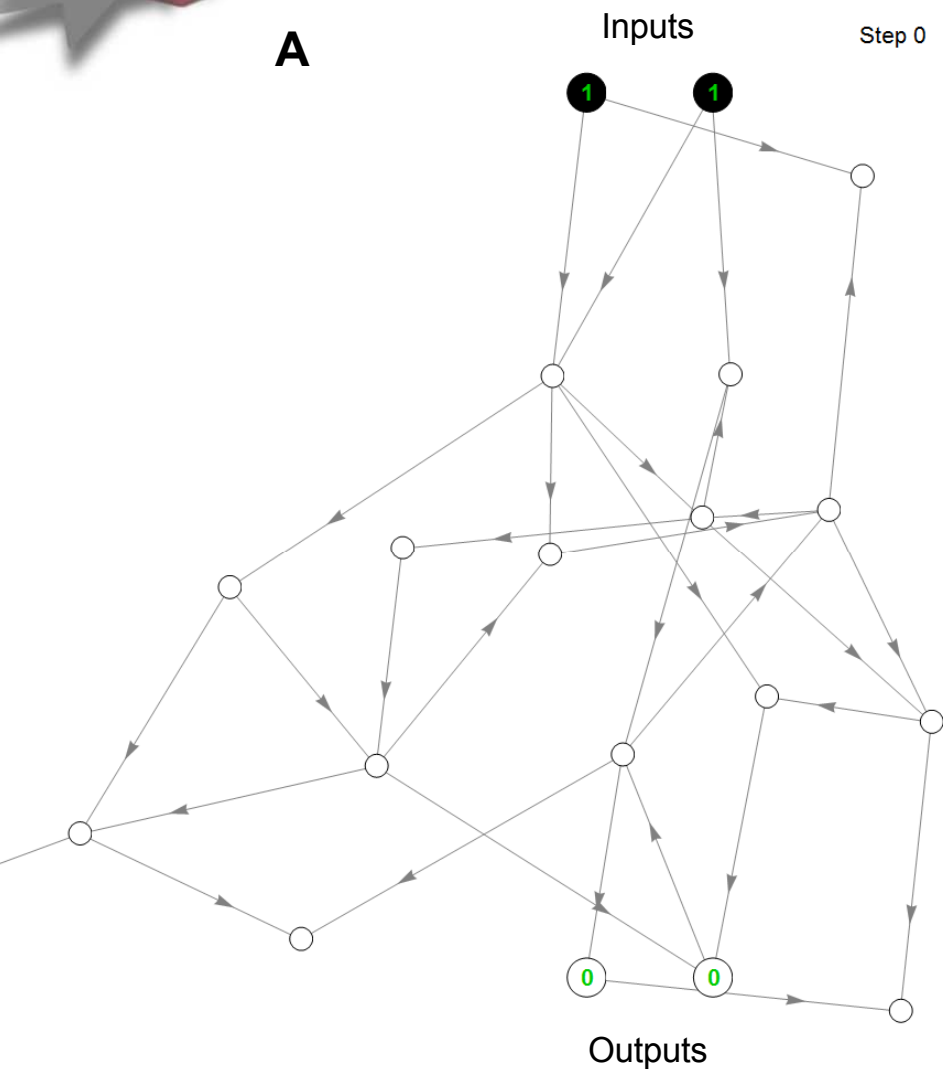


Complexity can address “whole system” robustness and stability

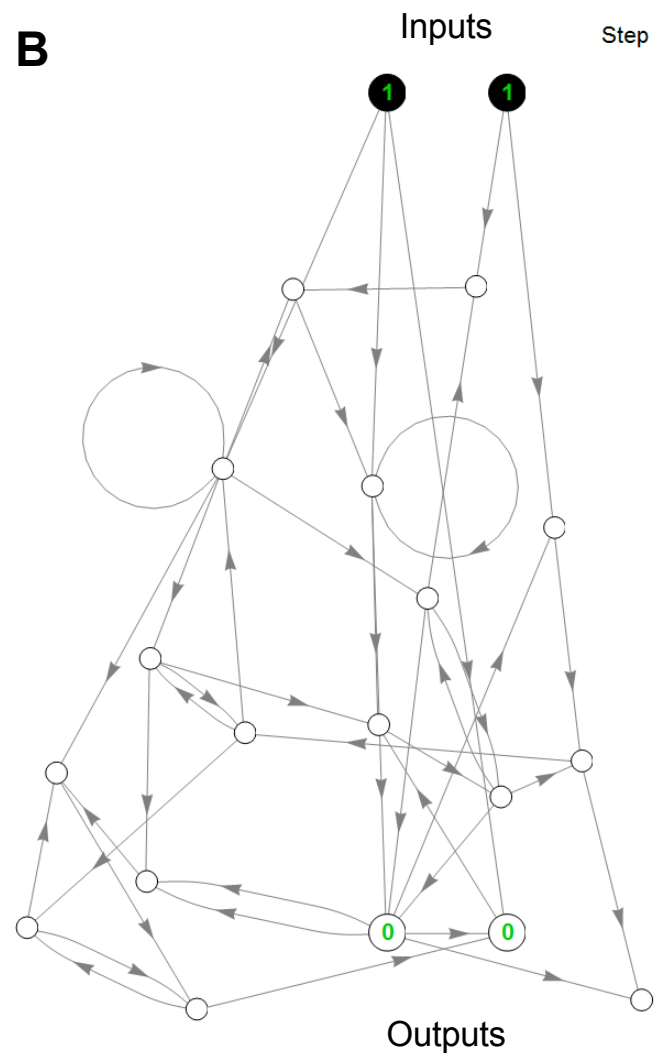
- **Consider designing a digital circuit to add two 1-bit numbers (a “half adder”)**
 - This is among the most basic functions appearing in microelectronics
- **There are many ways of composing logic gates to implement this functionality**
- **The next slide shows two such circuits; each performs as a half adder when run for twenty steps**
 - Shown correctly adding $1 + 1$ to get the binary result 10
 - They also give correct answers for the other possible inputs



A



B





What distinguishes the two implementations? *Resilience*

- For this very simple functionality, both circuits can be verified by exhaustive testing
- More realistic circuits cannot be tested exhaustively, so we need to understand the effect of untested states
- In this example, we introduce occasional gate errors to represent unanticipated behavior
- The next slide shows a typical run of each circuit with a 1% error rate per gate update
 - States that deviate from the ideal run are outlined in red
- Circuit A has much less error in the final output (greater resilience) than circuit B – why?
 - In this case, average inputs per node (k) makes the difference

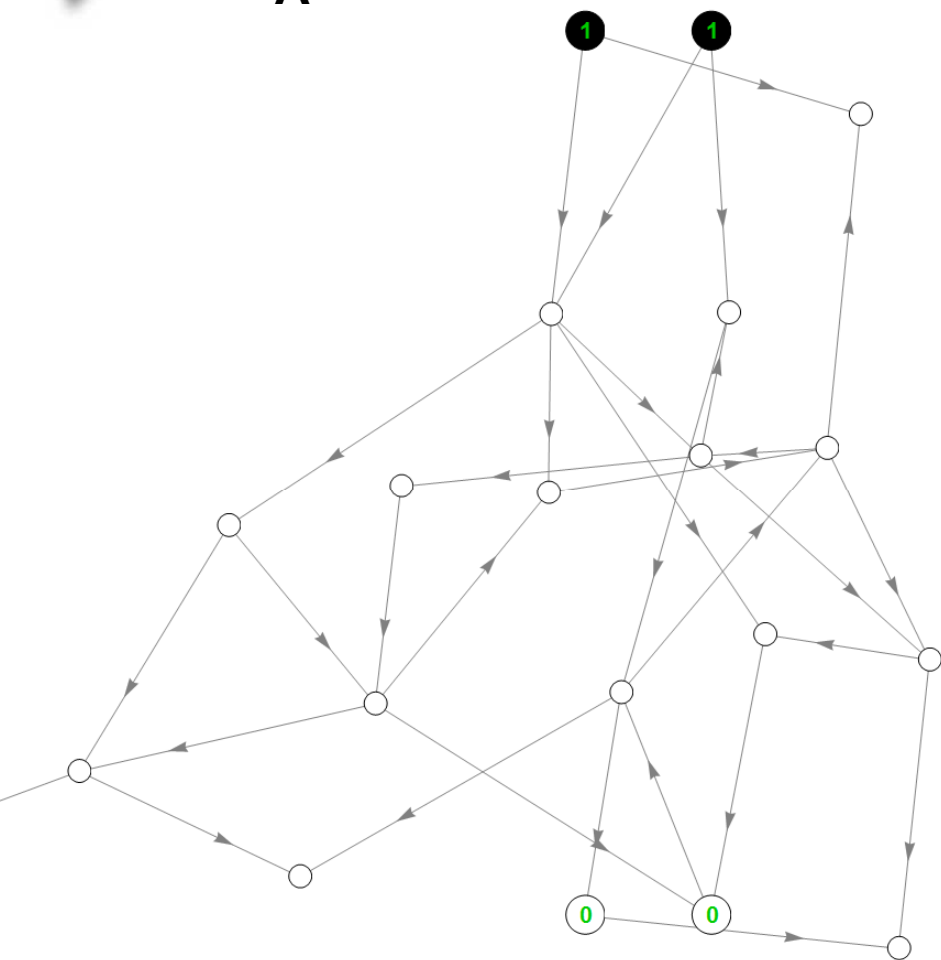


A

$k = 1.5$

Inputs

Step 0



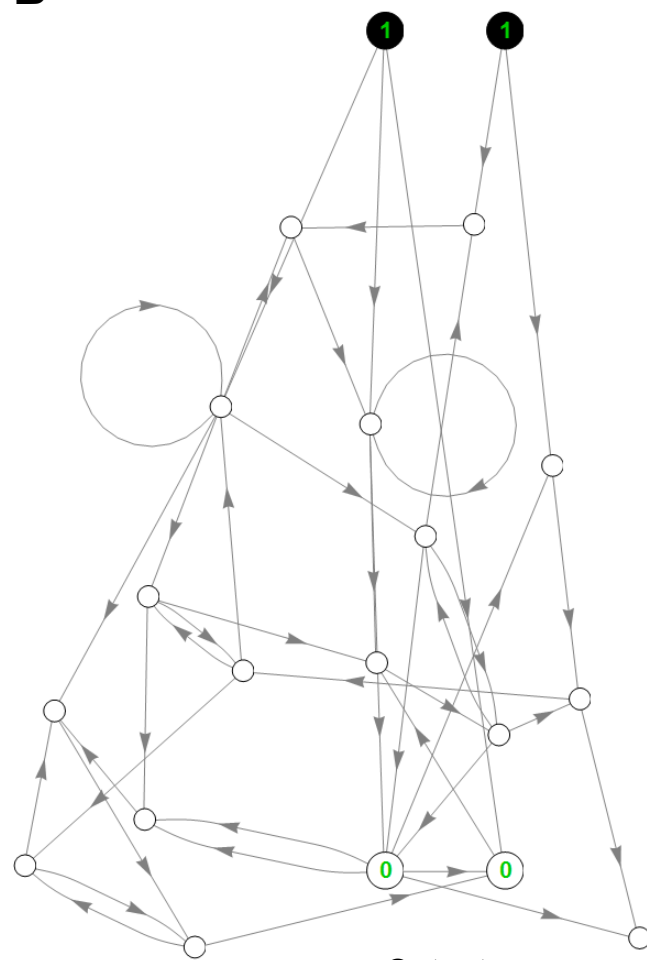
Outputs
(Average incorrect bits: 0.10)

B

$k = 2.5$

Inputs

Step 0



Outputs
(Average incorrect bits: 0.73)