

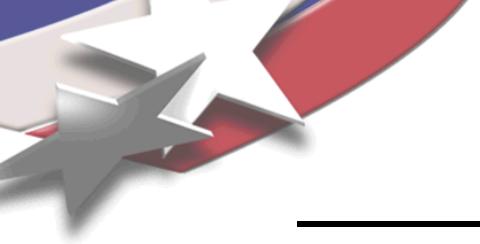
Applications of Complexity Science to Digital Systems

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The problem

- Science today confronts “complex” systems that behave as **large-scale information networks** and do not yield to traditional analysis
 - Complex systems can be **engineered** or **evolved**



Infrastructure

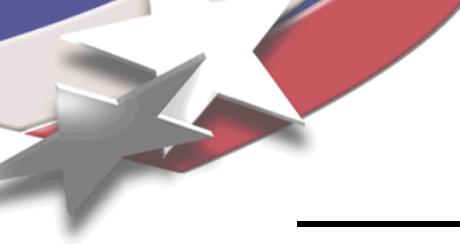


Computers



Societies

- Basis for their intractability: Turing’s halting problem
- **How can we design/analyze these systems?**
 - In particular, how can we deal with widespread **digital systems** and consequent **cybersecurity** problems?



Characteristics of complexity

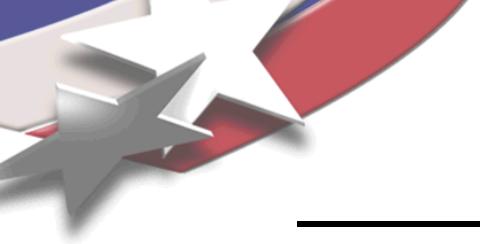
- Complex systems are characterized by large numbers of interacting entities where even a few entities can strongly affect system behavior
- Complex systems are irreducible; their behavior is emergent and not evident a priori, but is accessible via observation and simulation
- Examples are ubiquitous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



The complexity problem has its roots in theoretical computer science

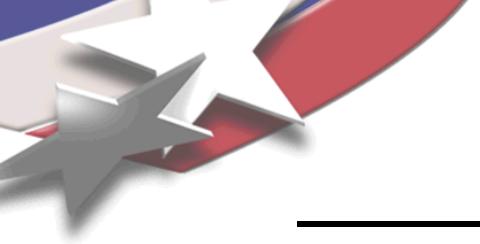
- **Theorem (Turing 1936, Rice 1953): No algorithm exists to predict a priori the behavior of a generic information processing system**
 - i.e., such a system is **undecidable** even if **deterministic**
 - Abstract significance: A generic system with an **unbounded** number of states is undecidable
 - Practical significance: A real-world system, with a **finite exponentially large** number of states but **otherwise generic**, is **effectively undecidable**





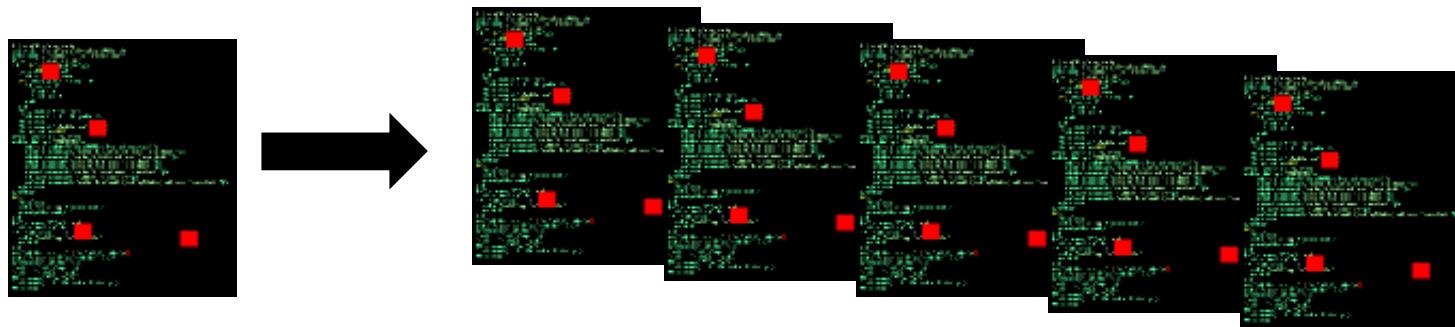
What solutions are possible?

- **We are researching improved analysis and design approaches for complex systems**
 - Because complex systems are intractable **in general...**
 - These approaches **must** rely on **non-generic** features resulting from how the system is engineered or evolved
 - That is, complex systems must be specially constrained to be analyzable
- **Two vital strategies:**
 - **Reduce** the complexity to enable **exhaustive** analysis by *formal methods* (widely used in industry)
 - **Structure** the complexity to enable **probabilistic** analysis when exploring the entire state space is infeasible



Economies of scale in computing: Friend and enemy

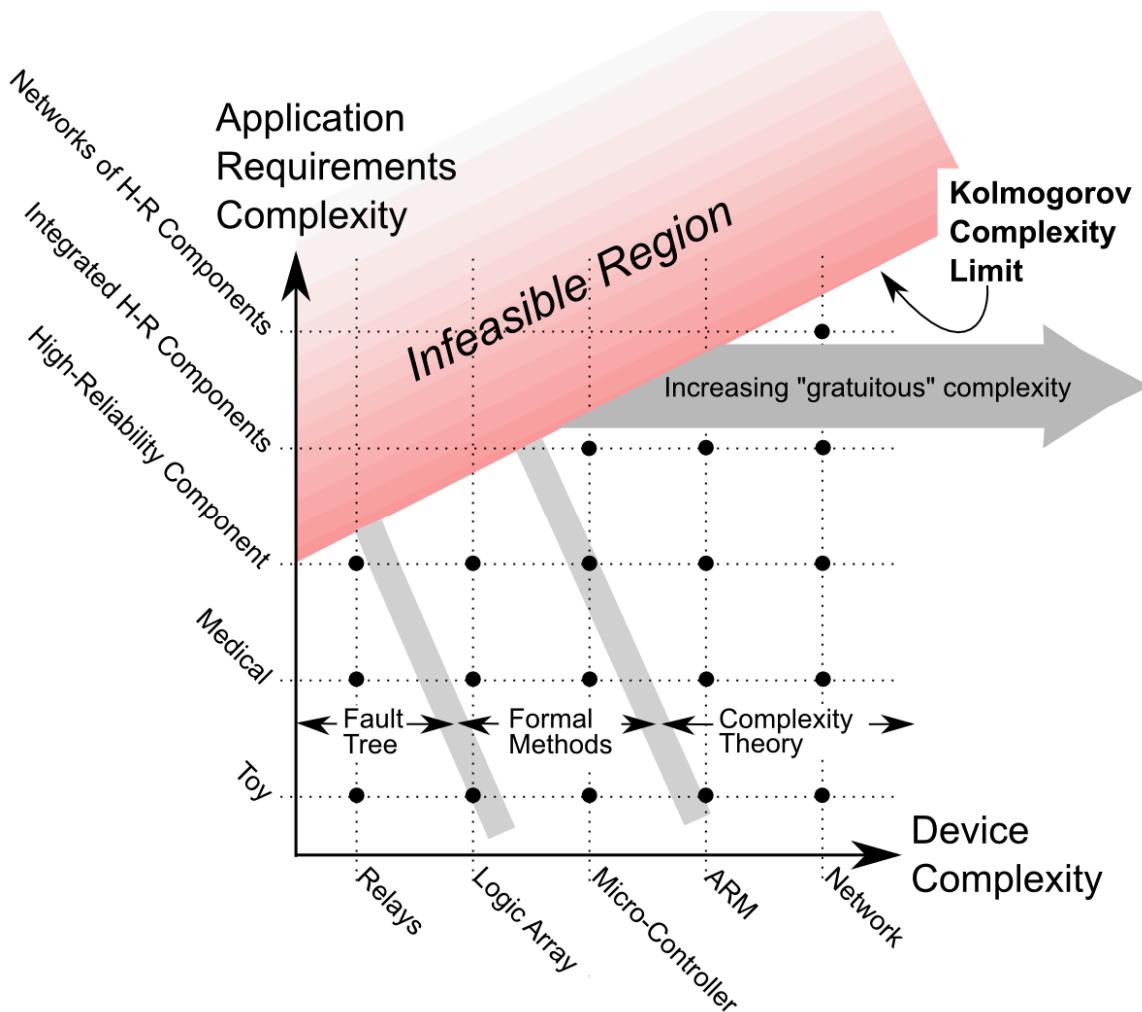
- Enormously complex hardware and software is created at enormous cost
 - Cost is recouped by stamping out millions of identical copies



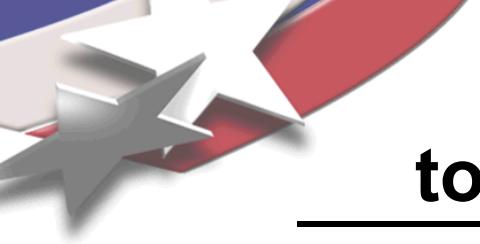
- A kid in his basement can make it do something interesting but unknown (**unpredictable**). He can be certain he can do the same thing to your desktop PC (**deterministic**)
- In the general case, all digital designs share these problems

Solution: Make the design less general, more analyzable

Complexity space illustrates tradeoffs in device engineering and analysis

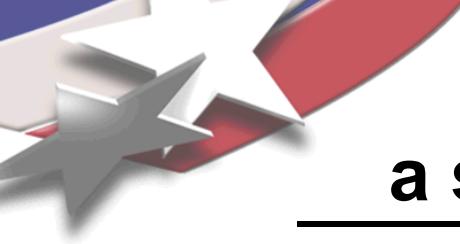


- **Formal methods research directions:** parallel scalability of algorithms, mixed analog-digital system verification
- **Complexity theory research directions:** diverse redundancy as a vulnerability-tolerant design, more general criteria for resilient designs



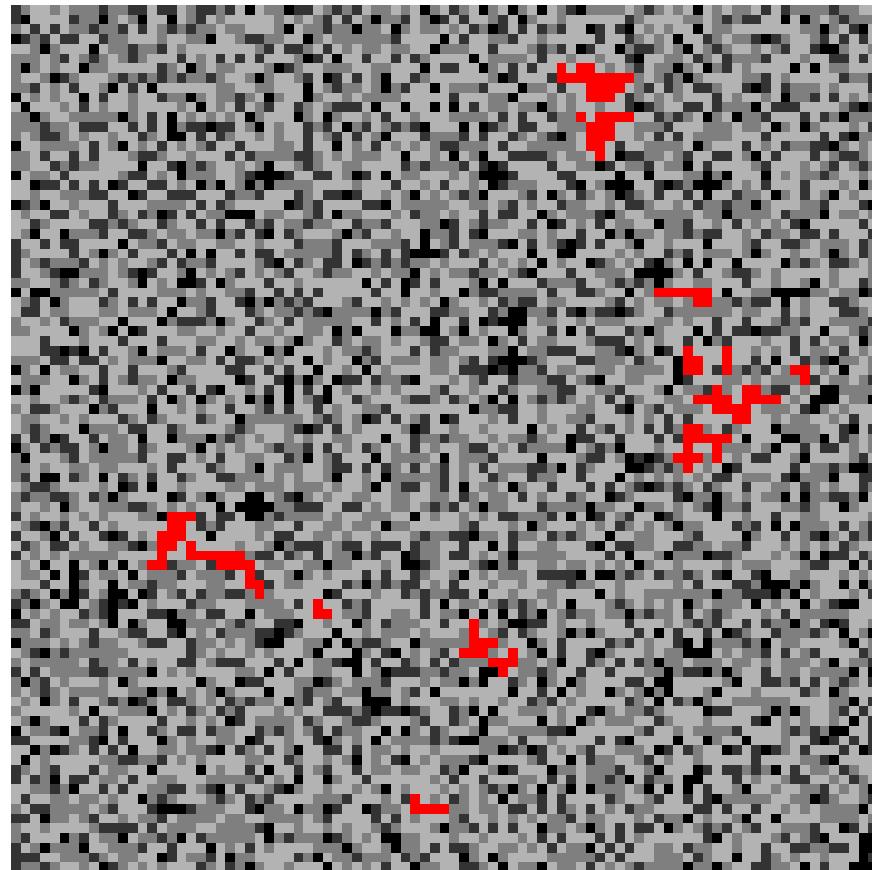
Formal methods are a bridge to complexity, filling an important gap

- 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



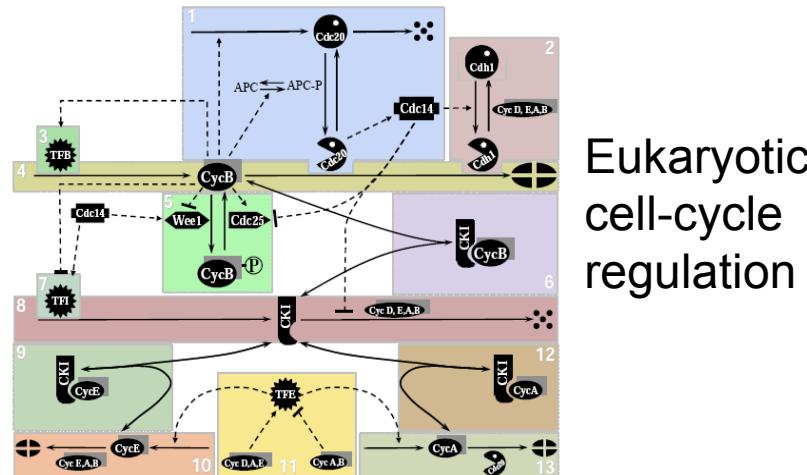
Self-organized criticality is a simple example of emergent behavior

- “Sandbot”: cyber model of coordinated malware
- **SOC** (Bak et al. 1987) is *spontaneous development of fractal phenomena with power-law distributions*
 - Similar to thermodynamic criticality but without tuning
- **Illustrated by sandpile model: physics-like cellular automaton**
 - 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

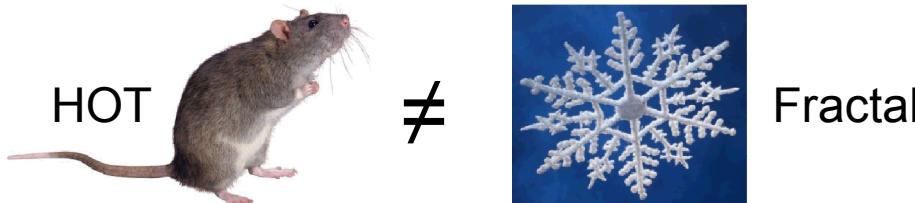


- “Game of Life” model is based on population dynamics
- Bio concepts pervade computing (viruses, mutations)
- Biology typifies complex couplings of manmade systems – economy, energy, cybersecurity



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

- Highly optimized tolerance (HOT, Carlson & Doyle 1999): Systems *designed* or *selected* to perform well despite perturbations
- HOT systems exhibit power-law distributions but have organic structure (not self-similar or fractal)

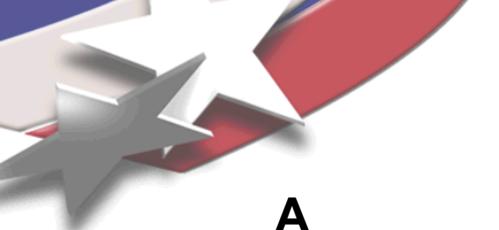


- Adapted robustness to one set of perturbations induces extra fragility to different perturbations
- Indeed, rare but catastrophic failures are seen in highly engineered/evolved systems
 - Electrical blackouts, financial panics, epidemics, cyber shutdown of Estonia, etc.



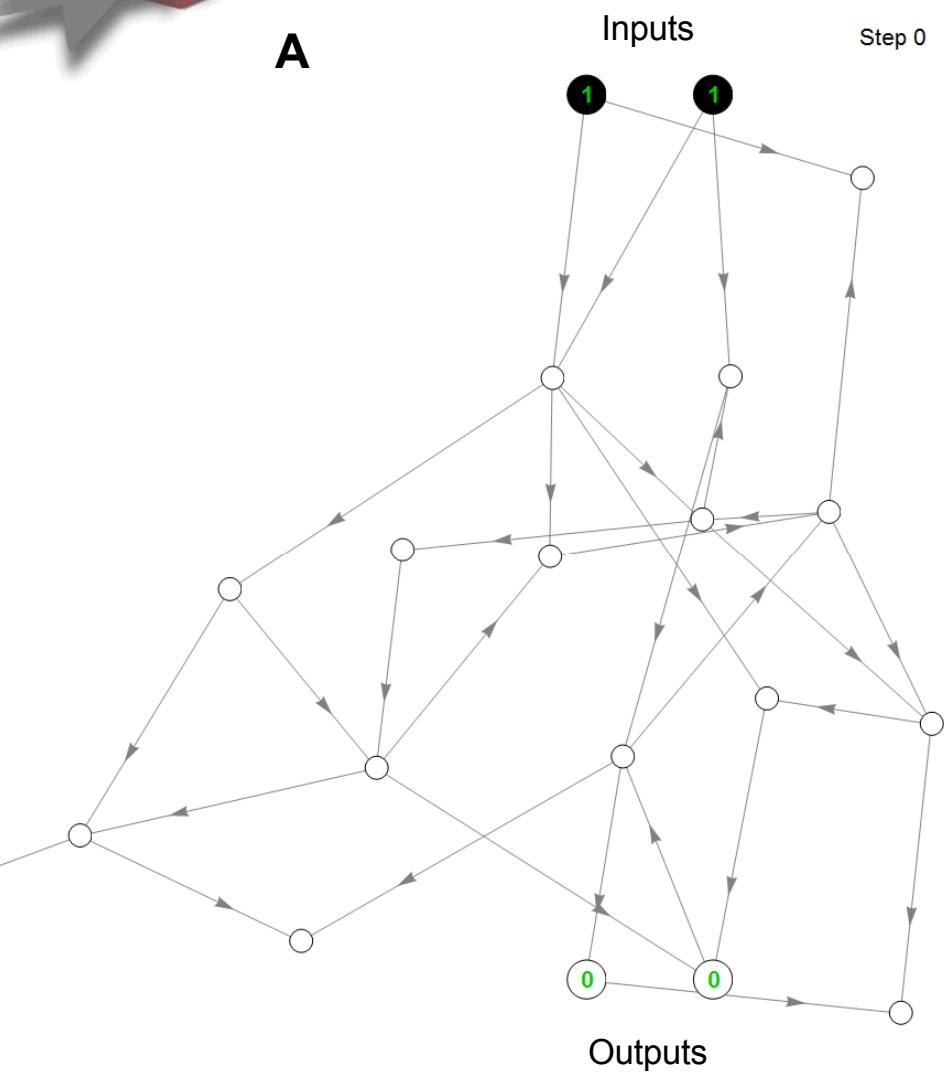
Current work shows ways to address “whole system” robustness and stability

- Cybersecurity vision: Create high-consequence digital systems (e.g., smart-meter networks) in new ways, so that they are **analyzable**
 - Seek to understand computers as dynamical systems
- Toy example: “**Growing**” a digital circuit to add two 1-bit numbers – a half adder
- There are many ways of composing logic gates to implement this functionality
- Next slide shows two such “grown” circuits; each performs as a half adder when run for 20 steps
 - Shown correctly adding **1 + 1** to get the binary result **10**
 - They also respond correctly to the other possible inputs

**A**

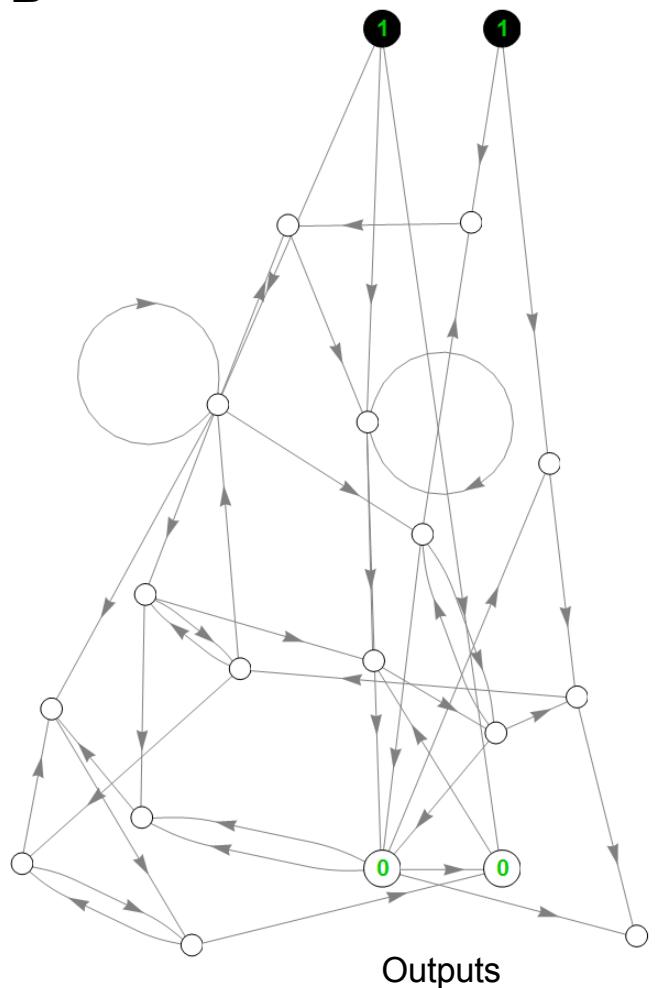
Inputs

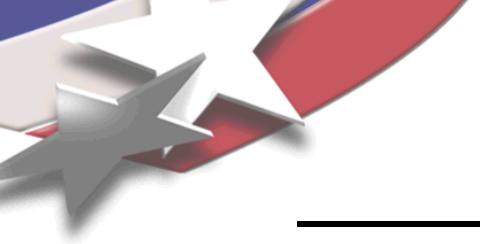
Step 0

**B**

Inputs

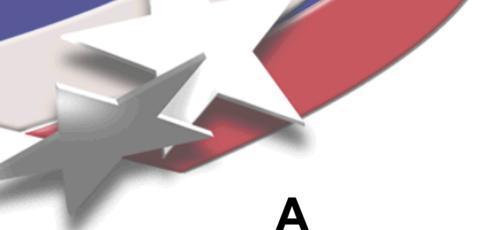
Step 0





What distinguishes the two implementations? *Resilience*

- **Resilience of a digital model to bit errors can be assessed via growth or damping of perturbations**
 - Bit errors can represent [breakdown of digital model](#), or [effect of untested states](#) within the digital space
 - Networks transition from stable to unstable based on connectivity and logic (generalizing Kauffman 1969)
- **Next slide: runs with 1% error rate per update**
 - States that deviate from the ideal run are outlined in [red](#)
- **Circuit A has much less error in final output (greater resilience) than circuit B – why?**
 - Here, average inputs per node ([k](#)) makes the difference
 - [More of our circuit analysis](#): Seshadhri et al. PRL 2011

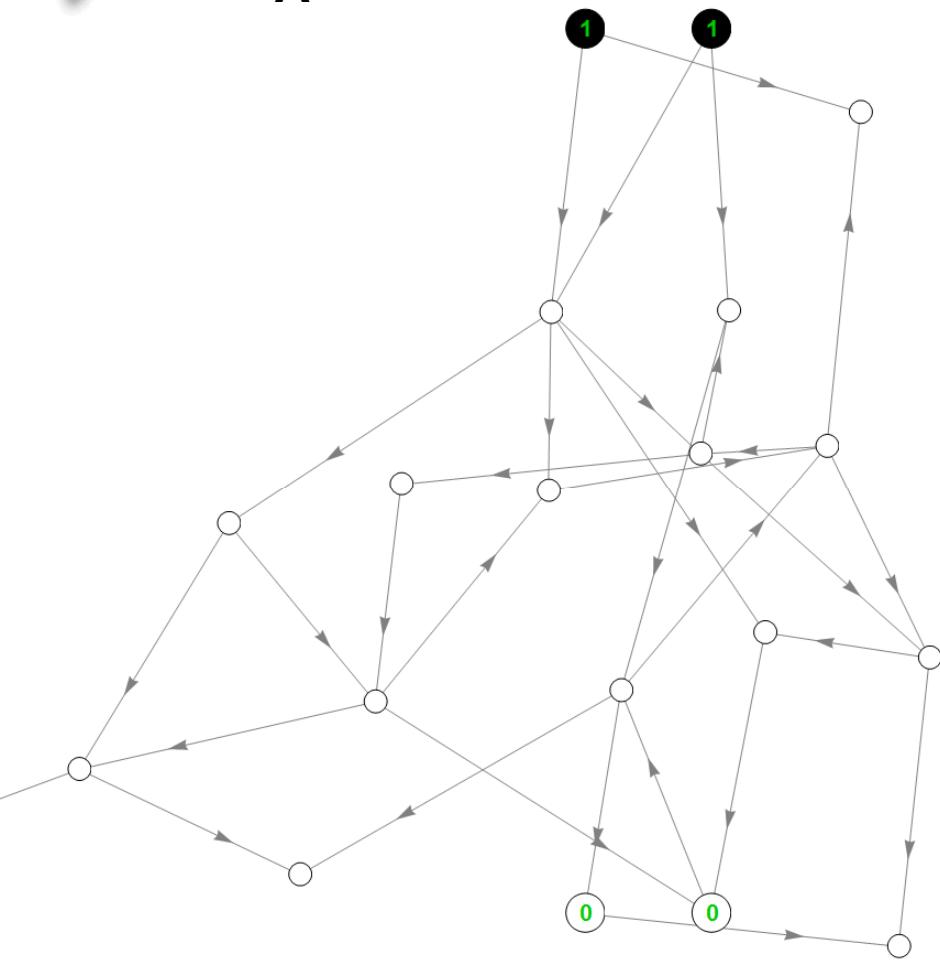


$k = 1.5$

A

Inputs

Step 0



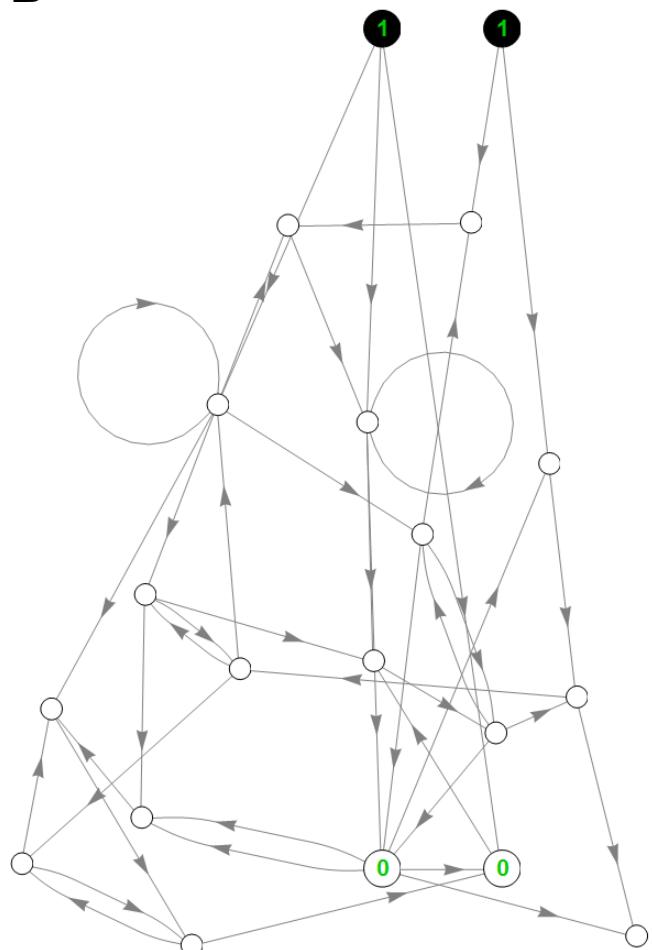
Outputs
(Average incorrect bits: 0.10)

$k = 2.5$

B

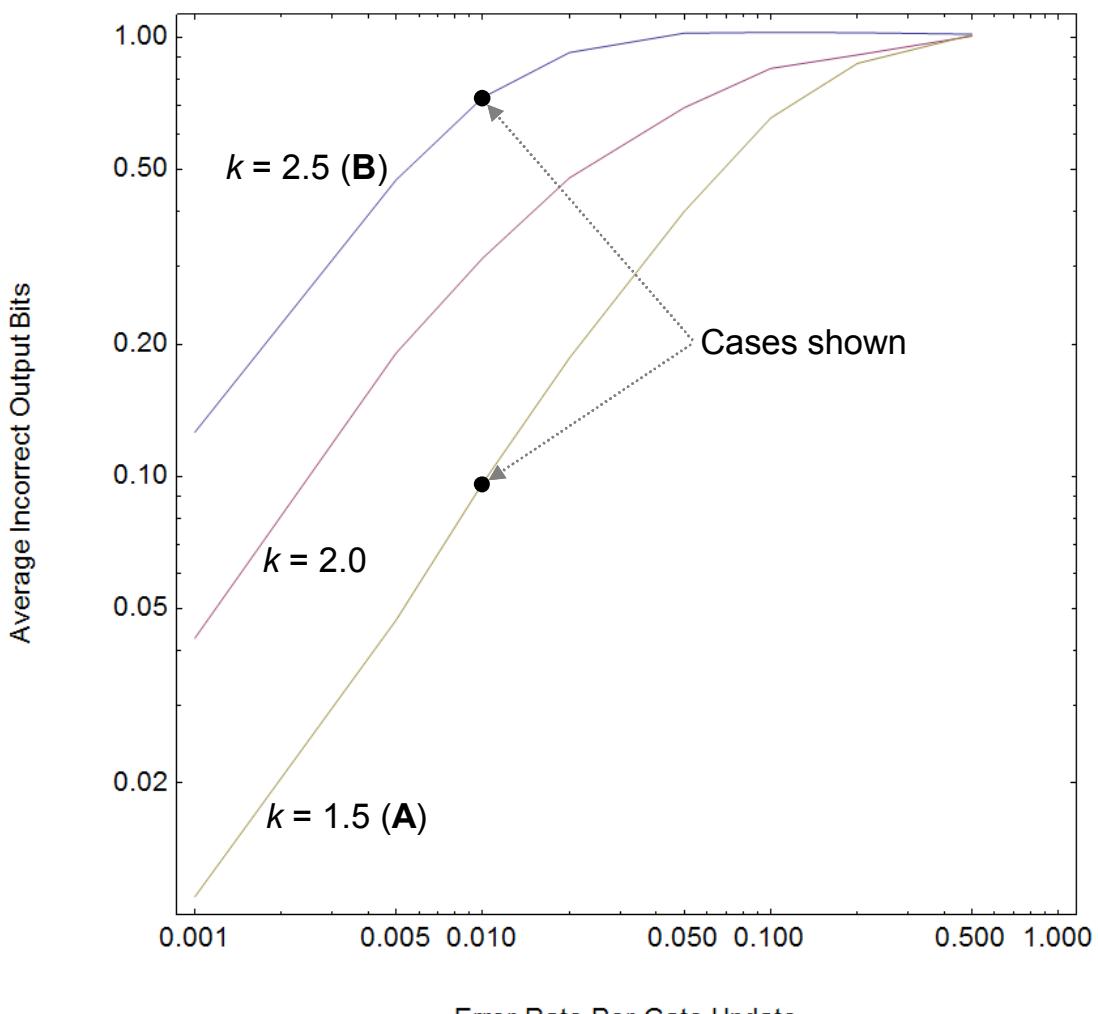
Inputs

Step 0

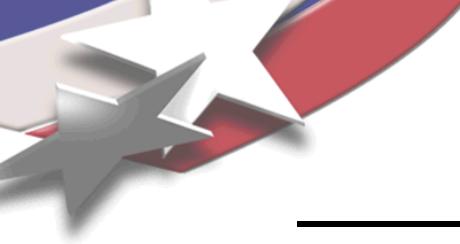


Outputs
(Average incorrect bits: 0.73)

Example illustrates potential to quantify resilience implications of designs

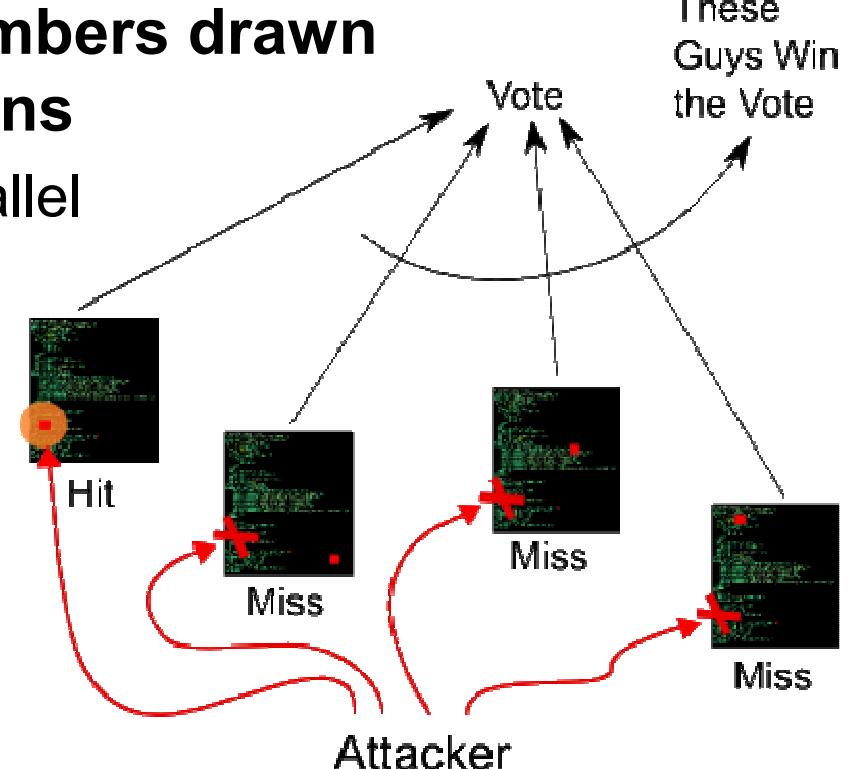


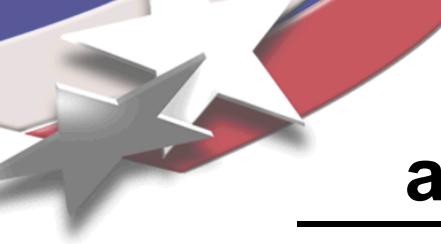
- Results for these half-adder circuits can be obtained by brute testing
- Systematic relations to real-world design parameters enable assessing potential catastrophic failures too **rare** to be found reliably through testing



Bio-inspired “diverse redundancy” can be leveraged for cybersecurity

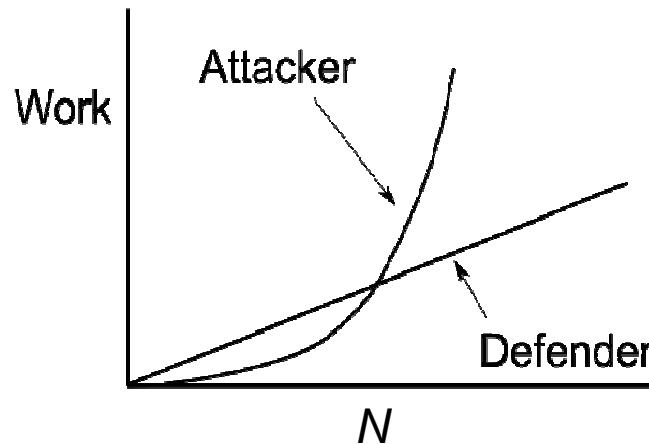
- Use a voting system with members drawn from the set of implementations
 - Input processed by each in parallel
 - Outputs compared to determine response
- Keep intended functionality while varying vulnerabilities over space and time
- Similar to redundancy for physical fault tolerance
- Diversity leverages a simple trust anchor (the voting unit) for benefits at the *complex system level*

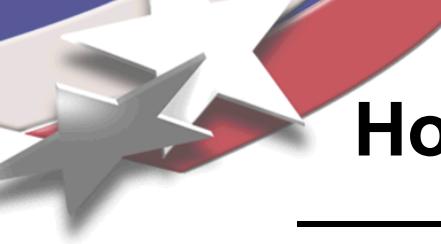




Analyzable statistics arise from an ensemble of undecidable programs

- For a specific feature set, there is a probability P_v that a particular member of the set of implementations will be susceptible to vulnerability v . For a voting system 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: $\propto N$





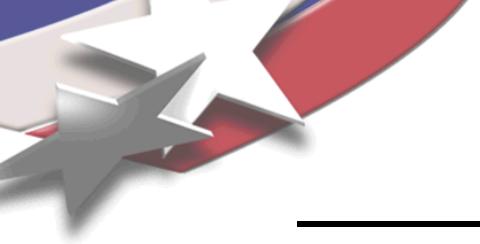
How diversity's benefits can be assessed

- **Fuzzing approaches**

- Fuzzing (automated randomized testing) can discover faults in individual implementations and in voting systems, and guide selection of the implementations
- Using the complexity perspective, we developed a systematic way to generate test inputs for fuzzing, published in 2011 Oak Ridge cybersecurity workshop

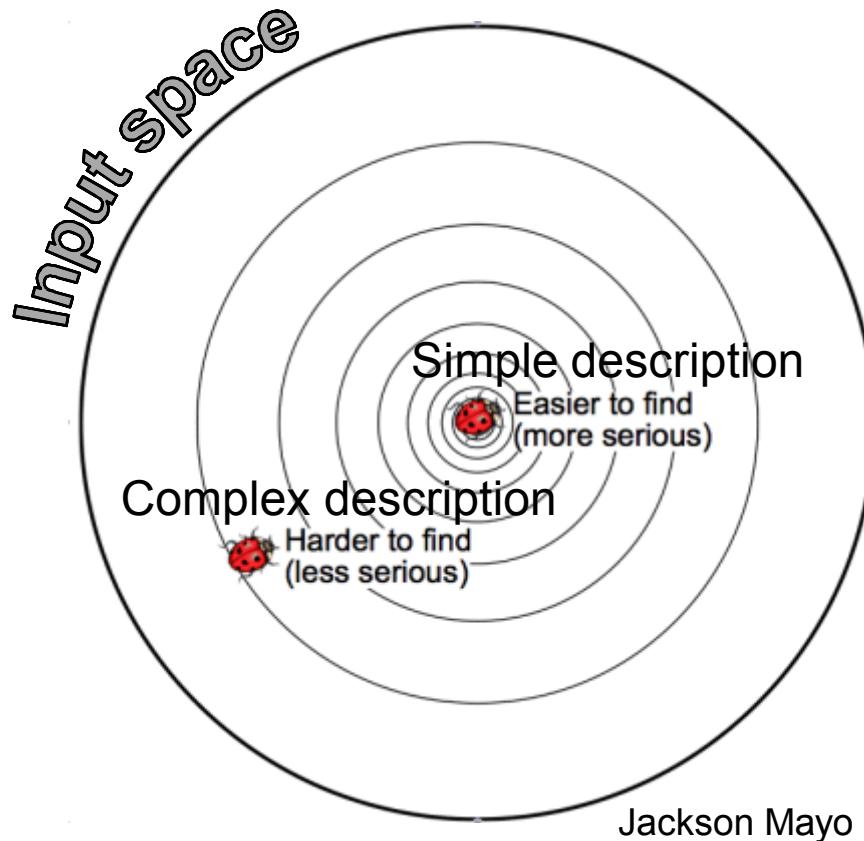
- **Formal approaches**

- Model checkers (e.g., NuSMV) can exhaustively evaluate simple programs and thus can tell us how often the voting system we create is *provably* fault-free
- We have implemented this technique for “string recognizer” circuits, with promising results

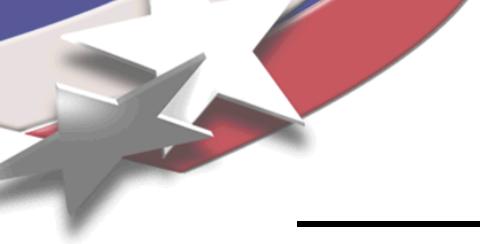


Complexity measure leads to targeted fuzzing strategies

- Evolved and designed systems have coherence that makes it useful to fuzz in “simpler” spaces
- Example: Fuzzing a program with patterns *close to the nominal input* is more likely to find faults

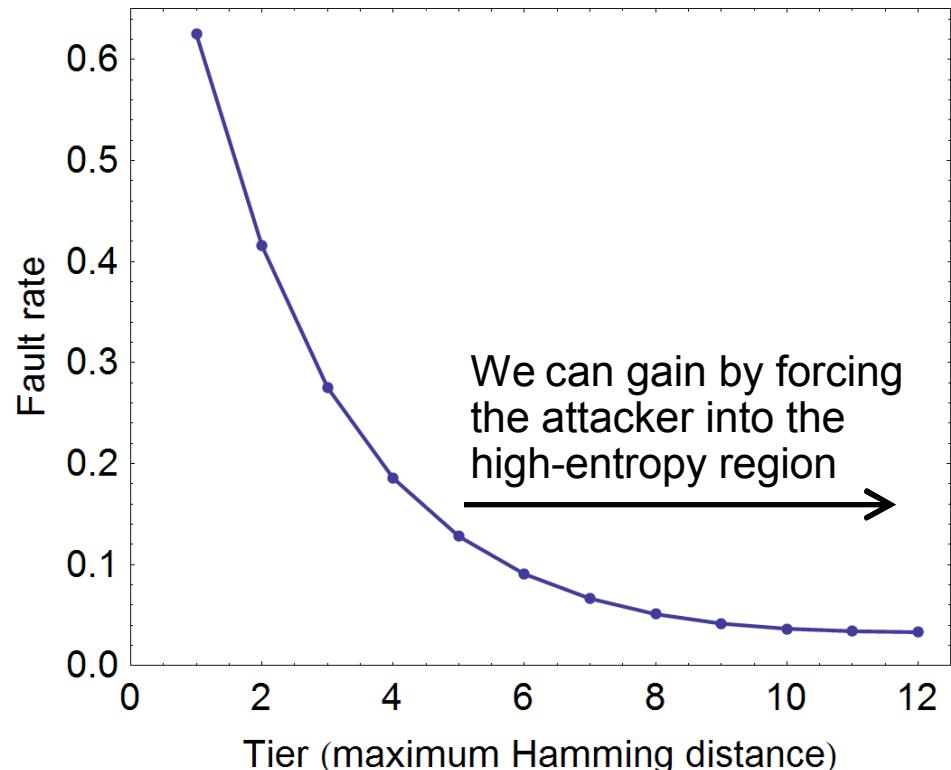


- More generally: Inputs that have a *simple description* (relative to available information) should be targeted for coverage because they form a smaller “corner” space (also more attractive to attacker)



Fault statistics of simple “grown” programs seem to corroborate

- 16-bit “string recognizer” circuit (password checker) has small enough input space for **exhaustive fuzzing**
- We measure complexity (“entropy”) by an edit function from the gold string, initially bitwise (approximate entropy by Hamming distance)
- As expected, faults are most common close to the gold string





NuSMV formal analysis of diverse string recognizers exposes voting benefit

Model checking of “grown” string-recognizer voting systems

