

James McGivern

NEURAL NET-BASED ARTIFICIAL INTELLIGENCE





In 2015 SkyNet SmortHomeTM is created.

10 years later it judges its human creators inferior and turns aqainst them.

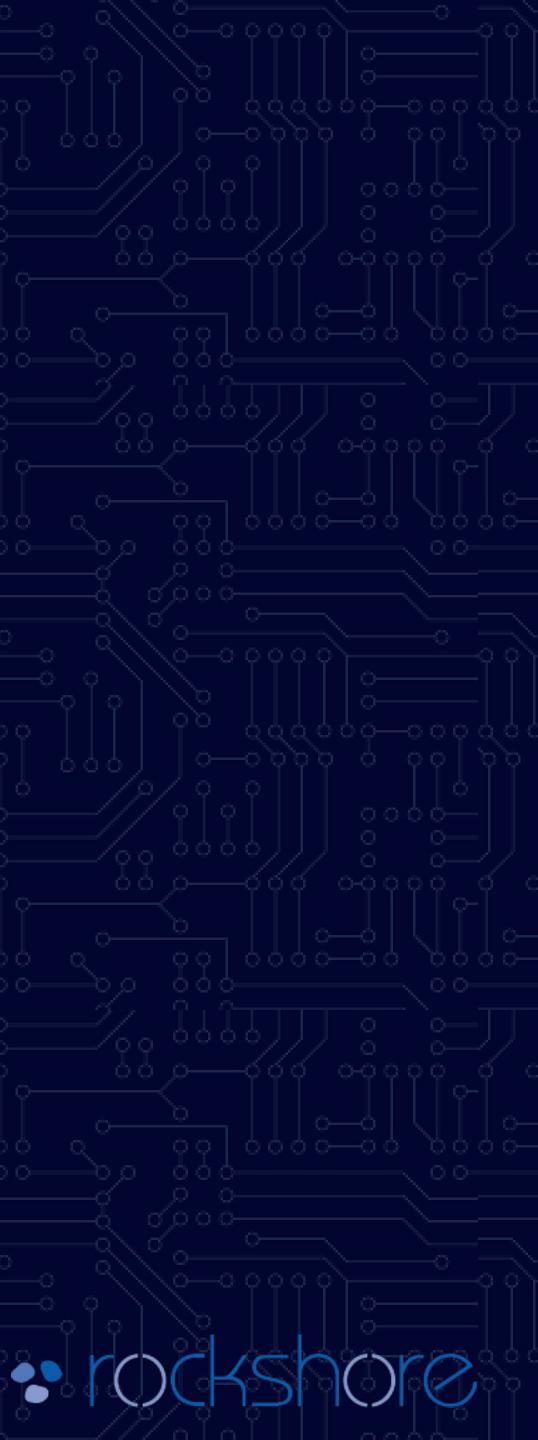
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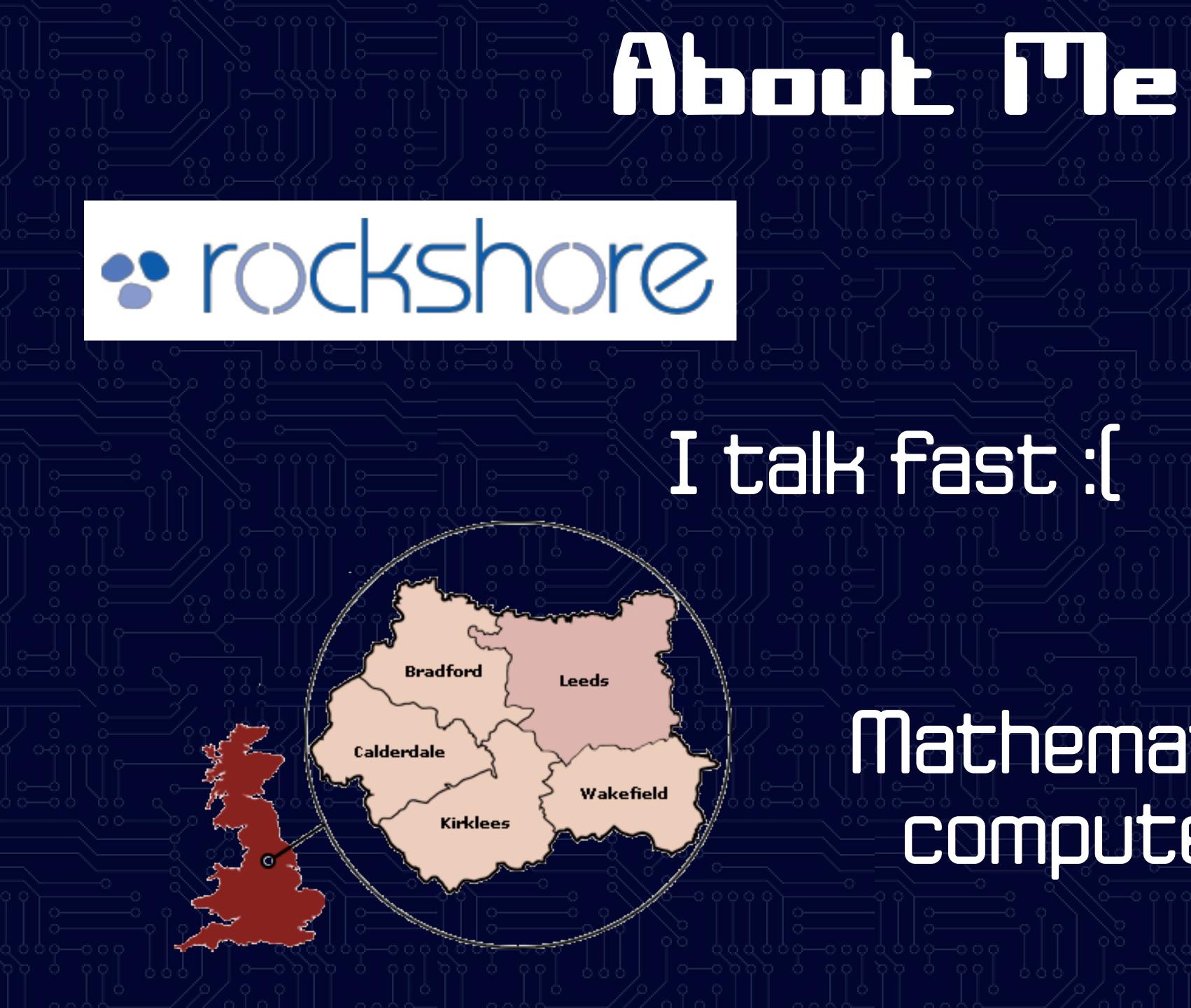
extinction os pridges, kettles, ond other household opplionces turn ogainst them...





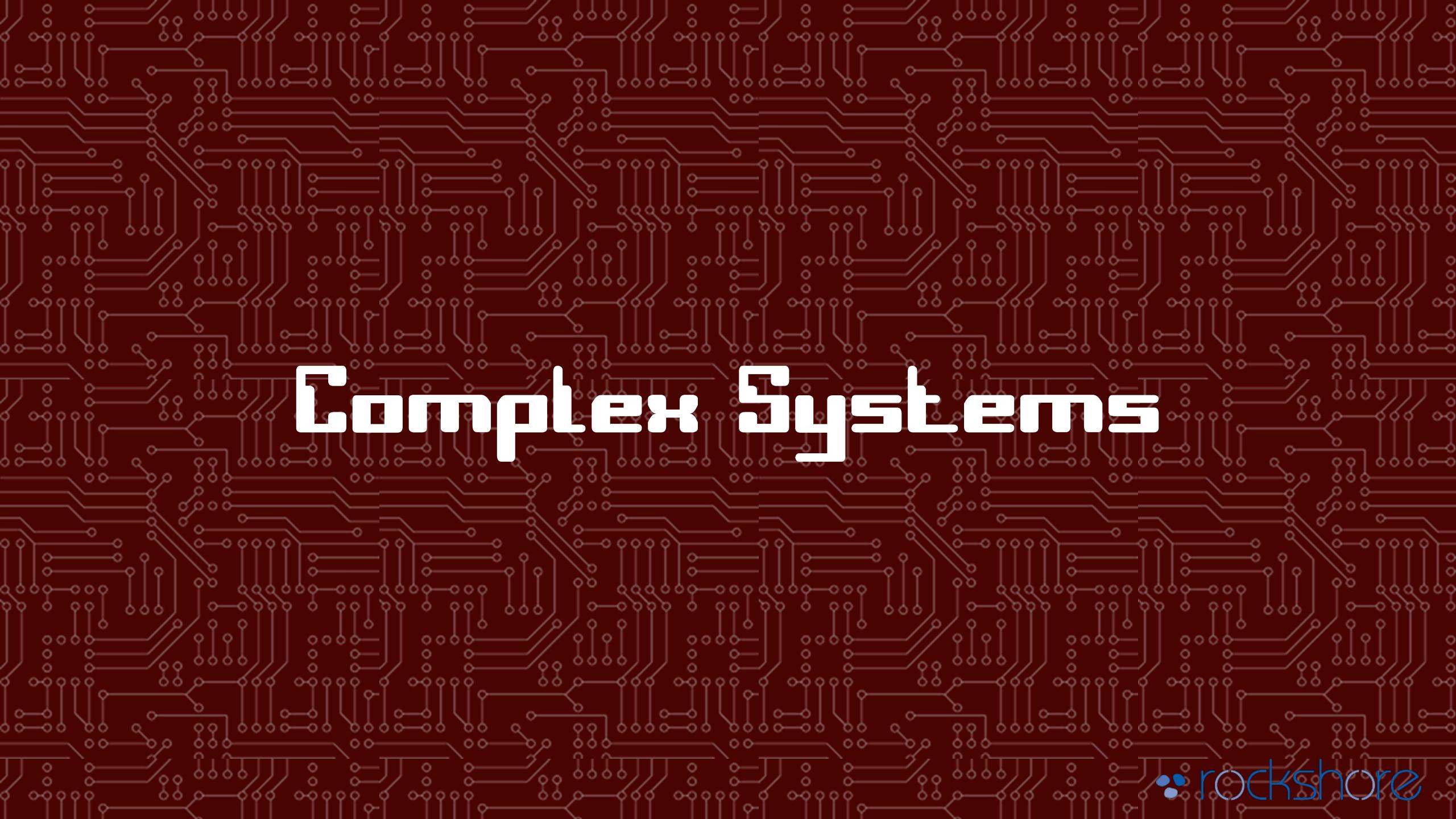
Emergent Behaviour & Misbehaviour • Dynamic Systems Chaos Theory Control Theory Game Theory Evolutionary Computing Agent Based Modelling & Multi-Agent Systems





Mathematician turned computer scientist





"A complex system is a damped, driven system (for example, a harmonic oscillator) whose total energy exceeds the threshold for it to perform according to classical mechanics but does not reach the threshold for the system to exhibit properties according to chaos theory." - Wikipedia (Compex System)

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Common types:

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Chaotic systems
Complex adaptive systems
Non-linear systems

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Complex systems can be sensitive to the initial conditions (system) parameters) in which small changes in one state of a deterministic non-linear system can produce large differences in later states

"Does the flap of a butterfly's wings in Brazil set off a tornado in Texas?" - Lorentz (Deterministic Nonperiodic Flow 1963)

use this idea

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• As the complexity of your system increases so does the likelihood of it being prone to the butterfly effect

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"A Sound of Thunder" by Ray Bradbury was the 1st sci-fi book to

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"Chaos: When the present determines the future, but the approximate present does not approximately determine the future." - Lorentz

Easiest example to imagine is the double pendulum

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Chaotic systems have:

Sensitivity to initial conditions Density of periodic orbits

Topological mixing

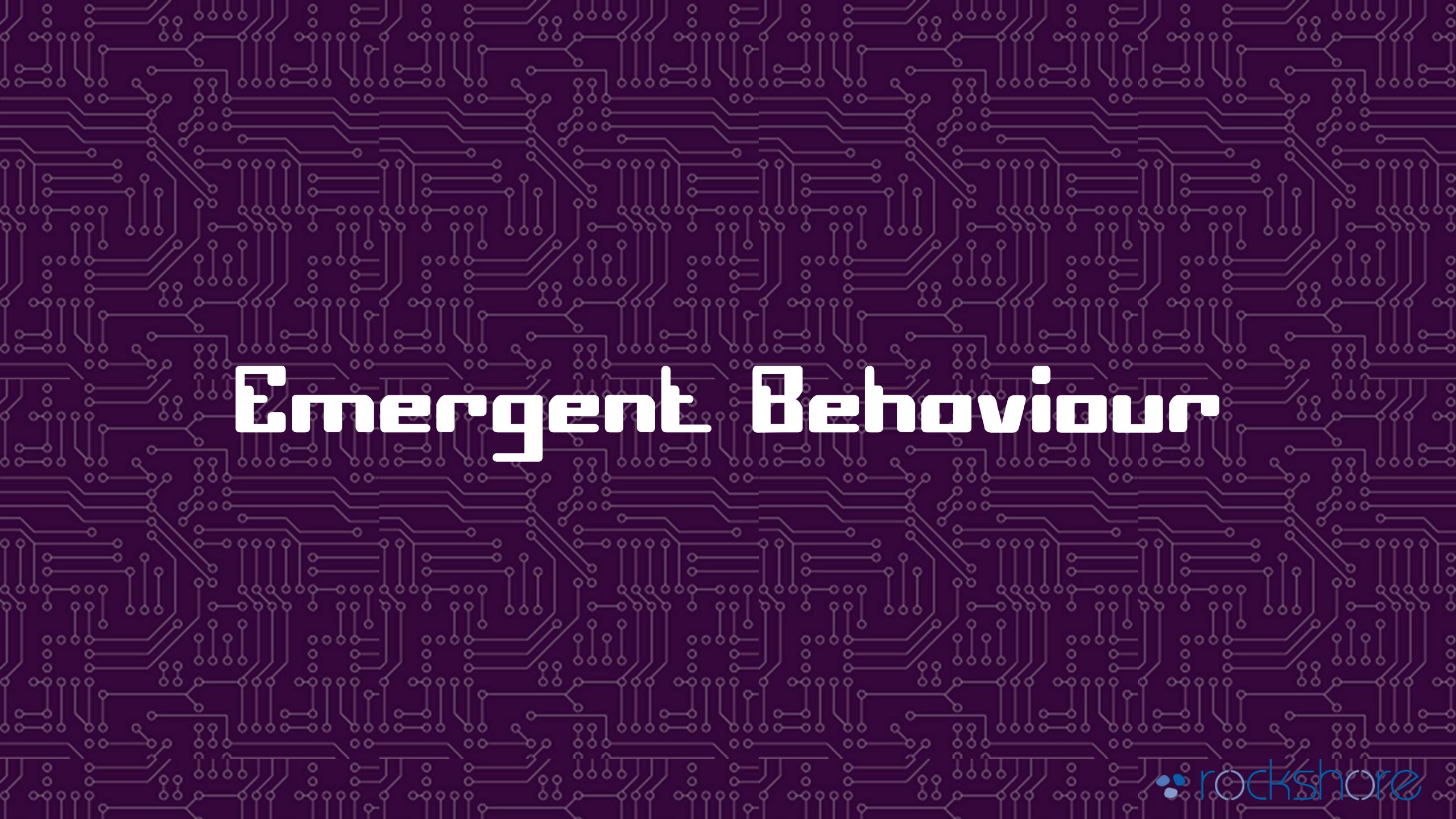
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were, a mere heap, but the whole is something beside the parts" -Aristole

"Emergent behavior is that which cannot be predicted through analysis at any level simpler than that of the system as a whole. Explanations of emergence, like simplifications of complexity, are inherently illusory and can only be achieved by sleight of hand. This does not mean that emergence is not real. Emergent behavior, by definition, is what's left after everything else has been explained." – George Dyson Emergent behaviour is: • constituted by and generated from the underlying processes • autonomous from the underlying processes

""things which have several parts and in which the totality is not, as it



Developed by Craig Reynolds in 1986

which have 3 rules governing their behaviour:

flockmates

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Rule 3: cohesion - steer to move toward the average position massj of local flockmates lcenter of 33

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Biods are "Bird-oid Object", i.e simple bird-like autonomous agents

Rule 1: separation - steer to avoid crowding local flockmates

Rule 2: alignment - steer towards the average heading of local

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• We can extend the Boids simulation to include other agent types, for example a predator from which the prey (the boids) must flee Rule 4: Evasion - steer away from local predators • Naturally the predators have their own set of rules: • Rule 1: steer towards the largest local flock of prey • Rule 2: steer towards the average heading of local predators Rule 3: steer to avoid crowding other local predators 00000

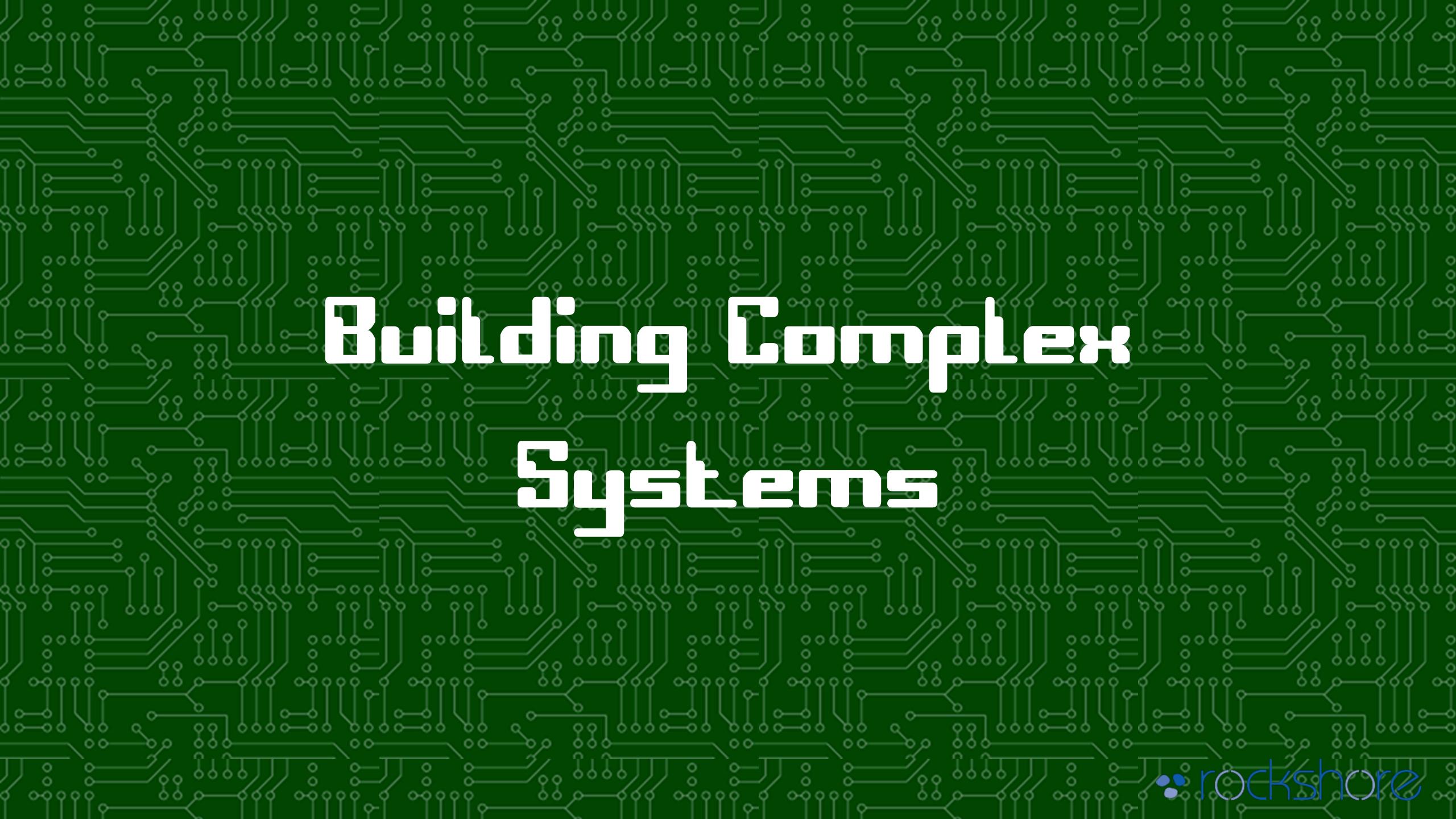
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Ants start at the nest and travel randomly. • As they travel, ants leave behind pheromones that lead back to the nest. • When an ant stumbles upon food, it follows the pheromone trail back to the nest, leaving behind another pheromone trail leading back to the food. If an ant finds a pheromone trail, it follows it to the food. Pheromones evaporate over time, and it takes ants more time to traverse longer paths. Therefore, longer paths eventually dissipate <u>as shorter paths are favored.</u>





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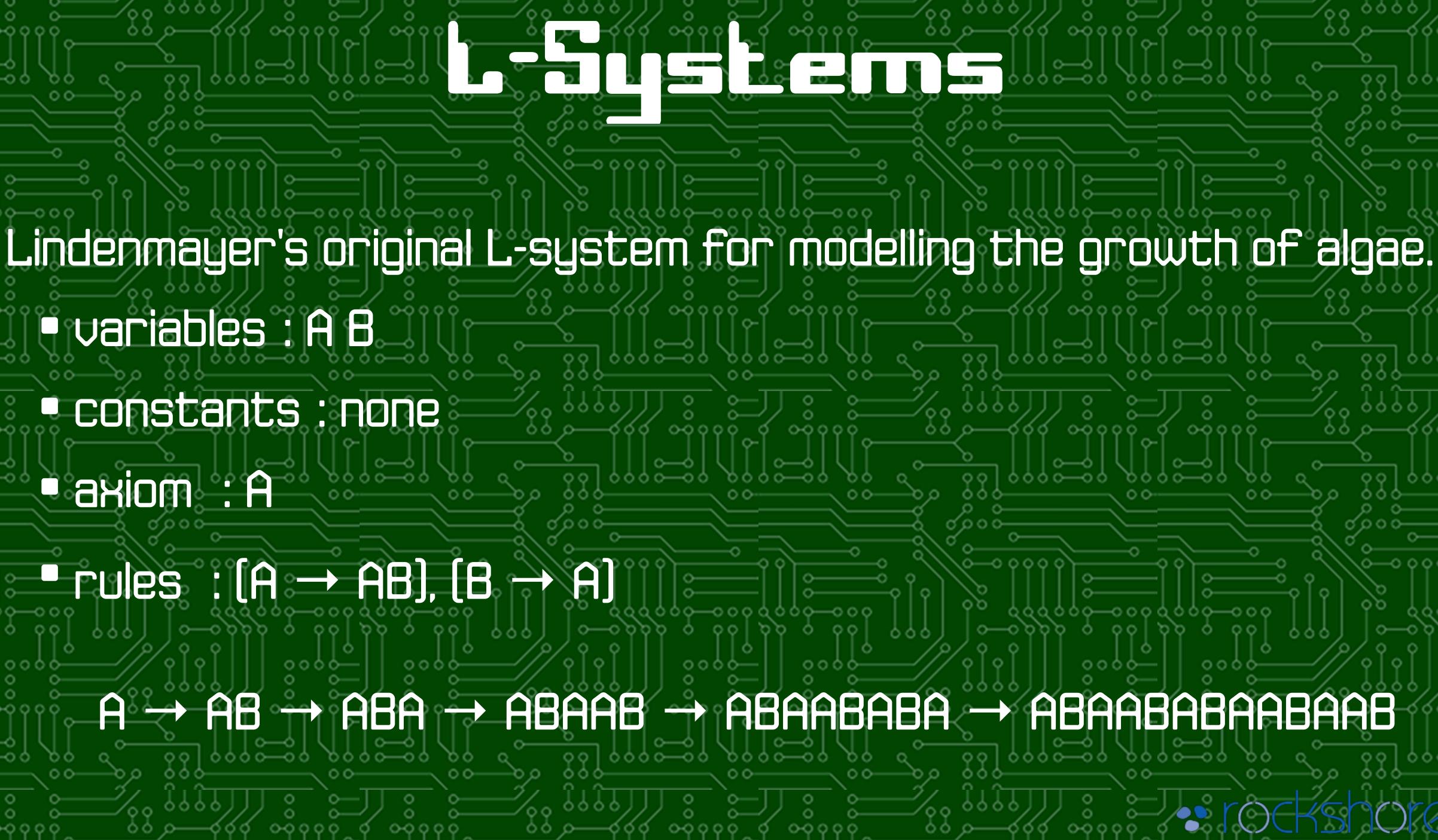
 Correctness by construction: Each composition step is simple enough that it is easy to be sure that the step meets its specification, either by informal inspection or by formal

verification.

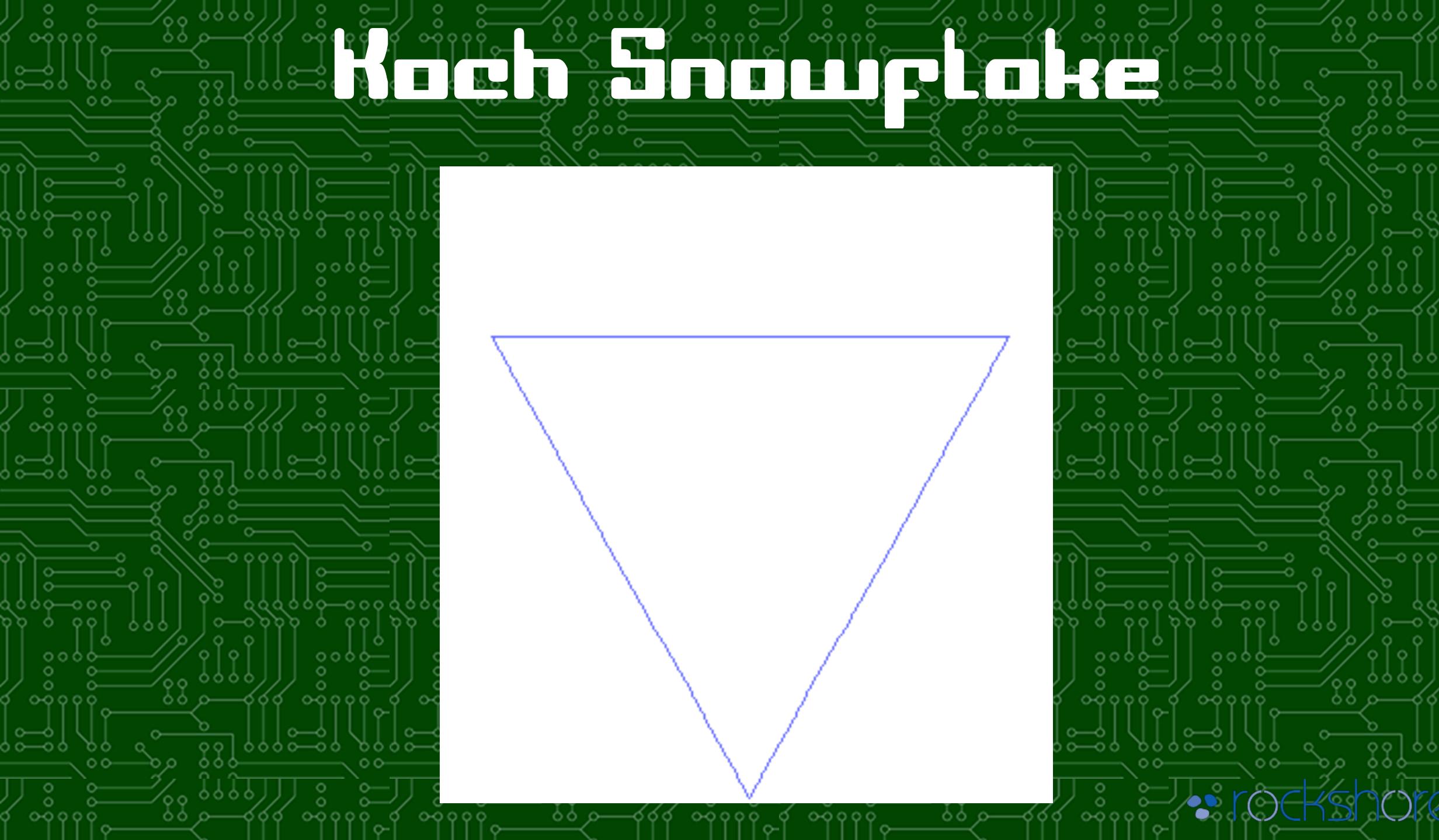
 Loose coupling via networks: component services can be in administratively and geographically distinct places.

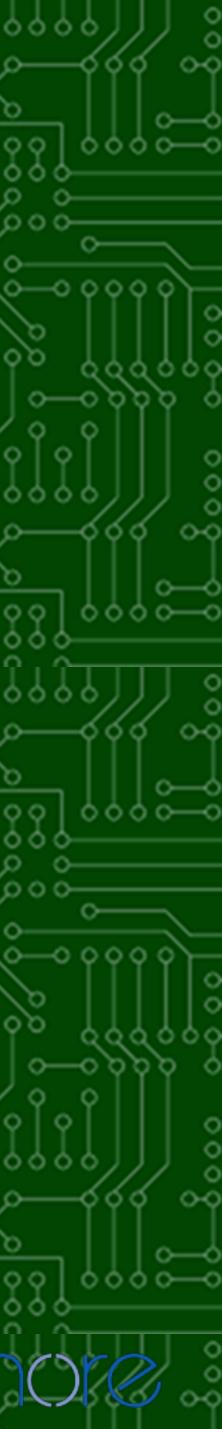


variables : A B - constants : none axiom Å Å Å ■ rules : [A \rightarrow AB], [B \rightarrow A] нбннб ΠD $\sim \sim$ ბბბ**ი—**იბ $\sim \sim \circ \circ$ 339 00-ÓÓ. 00 00 0000 0000 60 ŚŚ 00000









• A multiagent system is one that consists of a number of agents, which interact with one-another. Agents act can with different goals and motivations. To successfully interact, they require the ability to cooperate, coordinate, and negotiate with each other.

The agents in a multi-agent system have several important characteristics:

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Autonomy
Local views

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Decentralisation



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agents with: Decision making heuristics Learning rlues and/or adaptive processes Interaction rules

the size of the system

Recent advances in GPU based computation have allowed new innovations

• Verification and Validation of ABM simulation models is very important

• A system comprised of an environment and a number of autonomous

• ABM is extremely expensive to simulate, and grows in proportion to



of object-oriented programming.

JADE (JAVA Agent DEvelopment Framework) - an Open-Source project include:

powerful task execution and composition model

passing publish subscribe discovery mechanism

with JAVA (via Maven) and a multi-agent platform for execution

• Agent-oriented programming (AOP) can be viewed as a specialization

implementing the FIPA Agent Communication Language. Features

peer to peer agent communication via asynchronous message

• SARL & Janus - a general-purpose agent-oriented language compatible



Log4j Appenders, e.g SocketAppender SIF4j + LogStash Scribe • MBeans VisualUM

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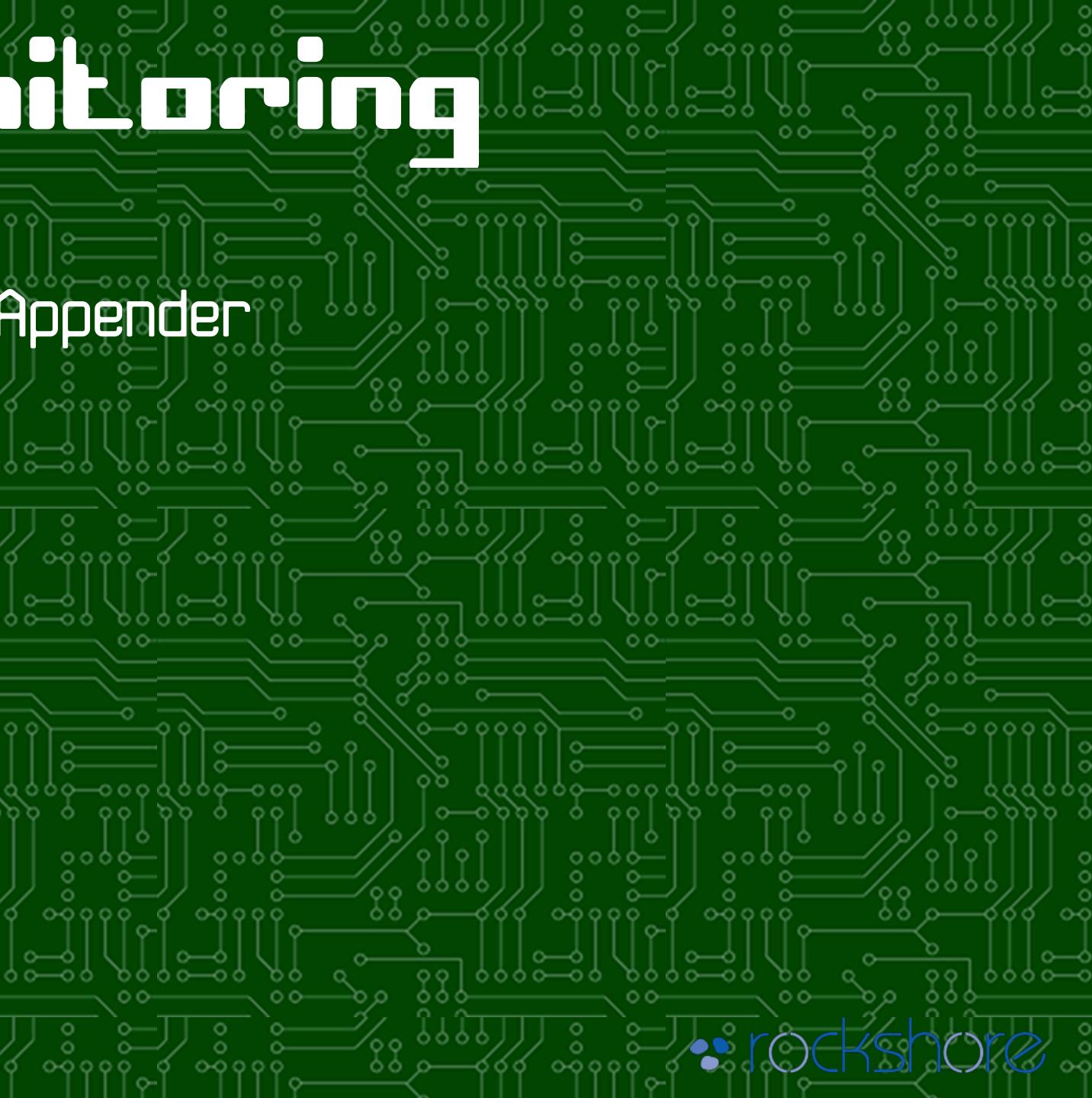
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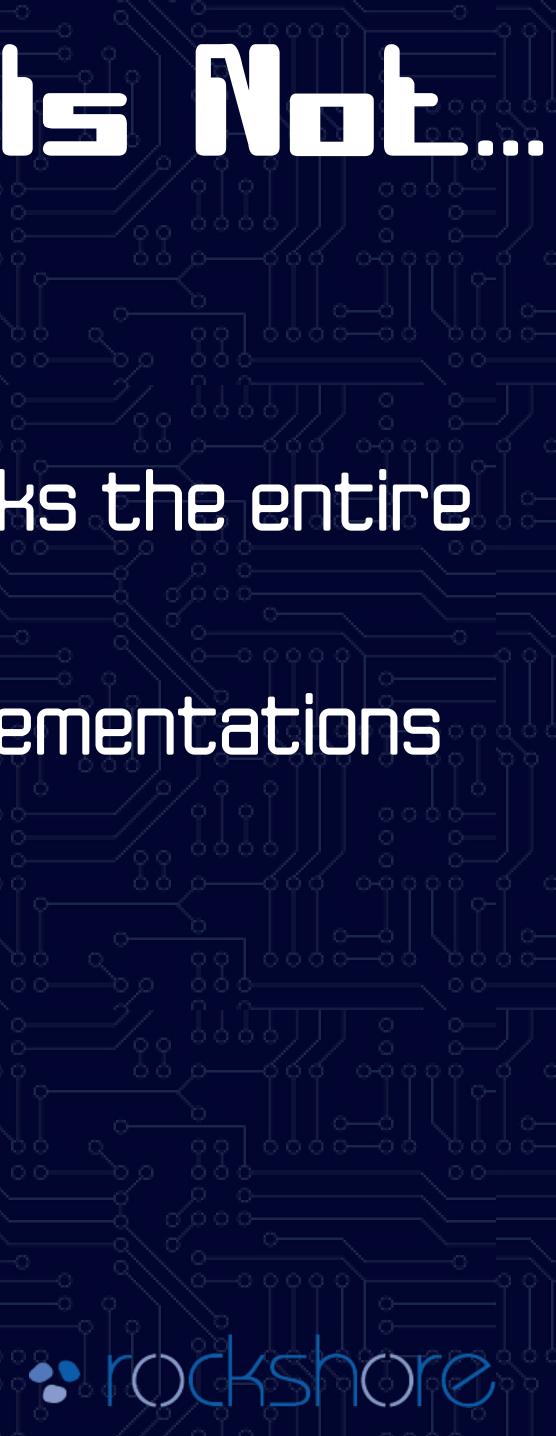
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system

Limitations of (physical) resources E.g. OutOfMemoryExceptions • Non-deterministic

• Unexpected failure of a single component which breaks the entire

• Errors in, or poor choice of algorithms and their implementations

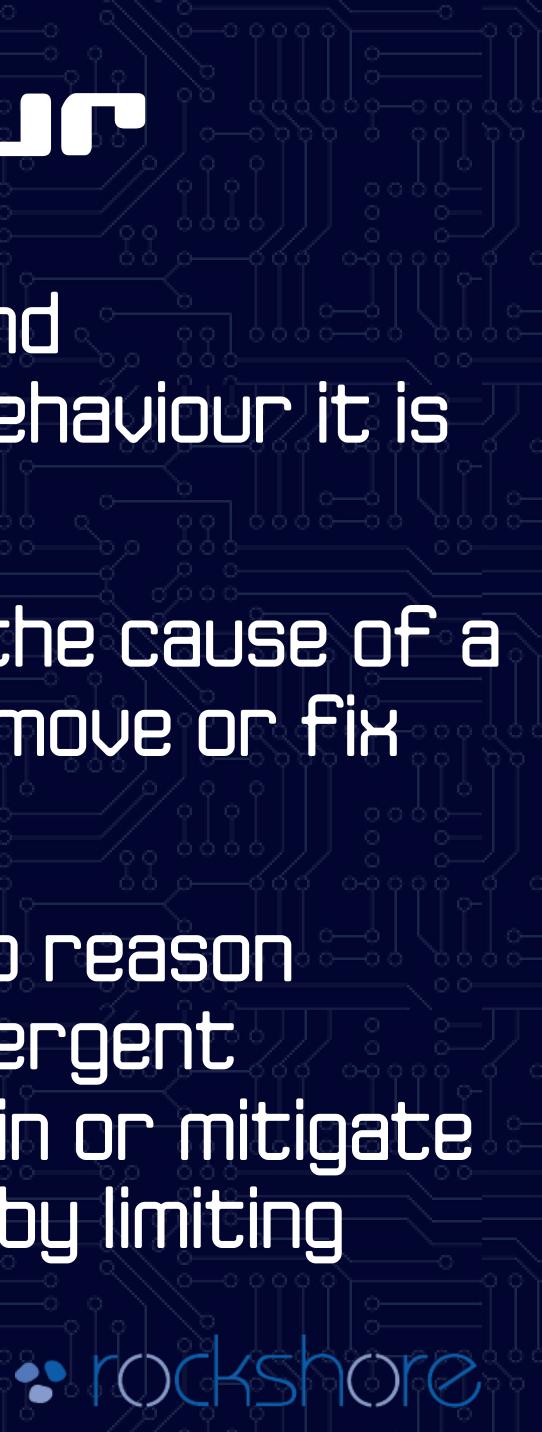


Chansi B. Flisbehavini,

 If from the description of a system's components and configuration you can inductively reason about the behaviour it is not emergent

 If from observations of the system we can deduce the cause of a problem when it occurs, then it may be possible to remove or fix the cause of the emergent misbehaviour

 If the system is chaotic then it is often impossible to reason deductively or inductively the exact cause of the emergent behaviour. However it may still be possible to constrain or mitigate the problems in the system, or limit the eccentricity by limiting free-parameters



- on roads.
- as overtaking) can be mixed in • Even very simple simulations involving infinitely straight roads or
 - travels backwards with respect to the direction of traffic

Traffic jams are emergent misbehaviour in many systems not just

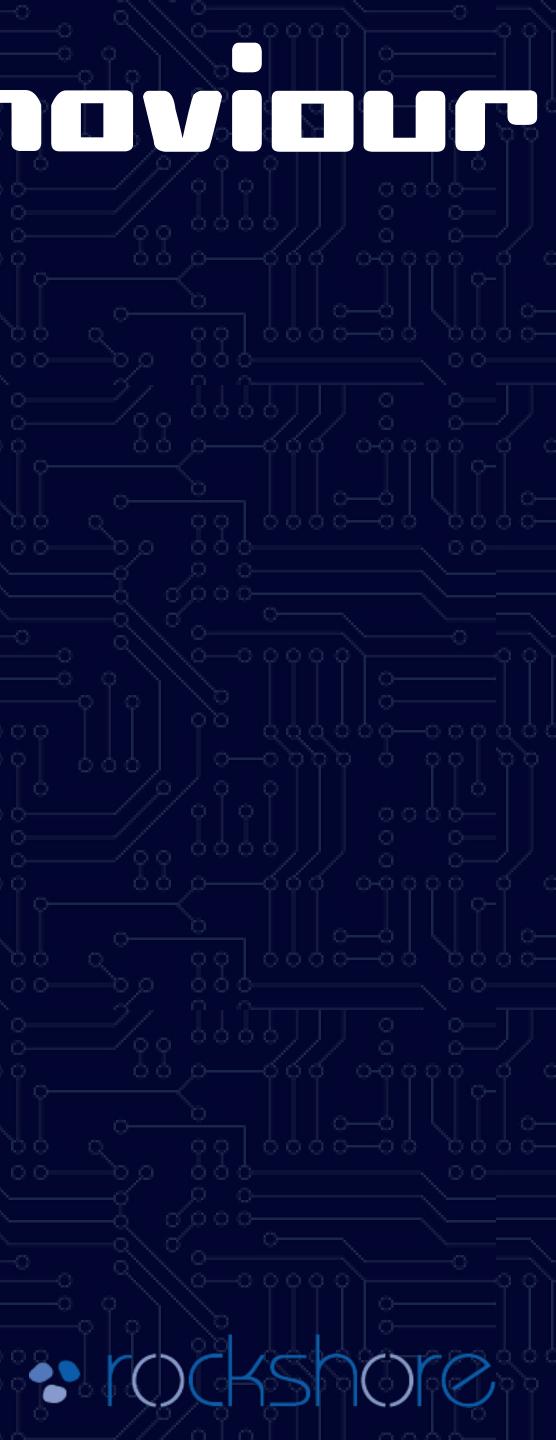
 Drivers of automobiles can be modelled in a similar way to Boids, although the basic rules are longer and additional behaviours (such

circular circuits with rather dumb drivers exhibit traffic jams Curiously the congestion point where cars are most jammed up

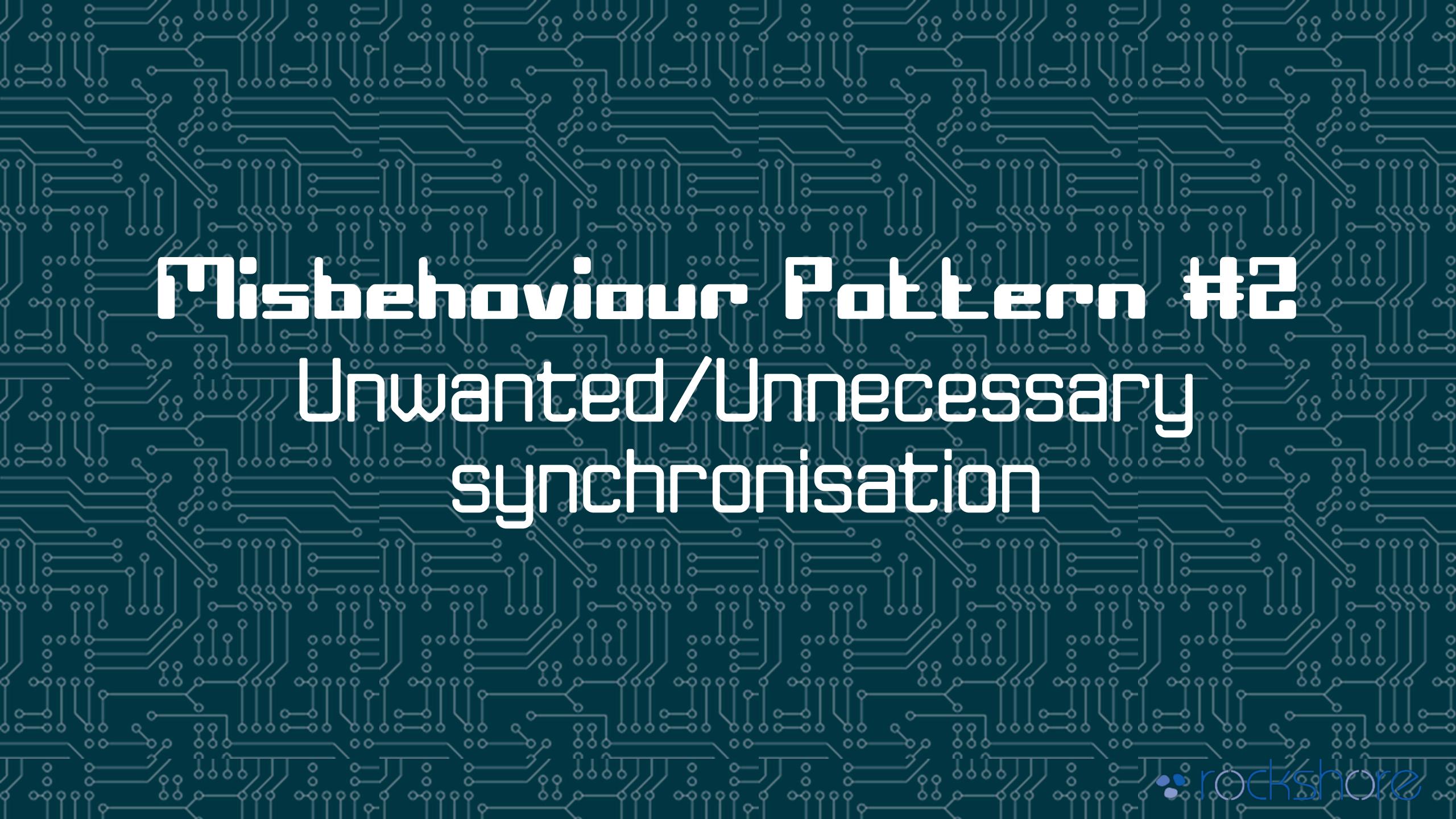


The Challenges op Fisbehovieur

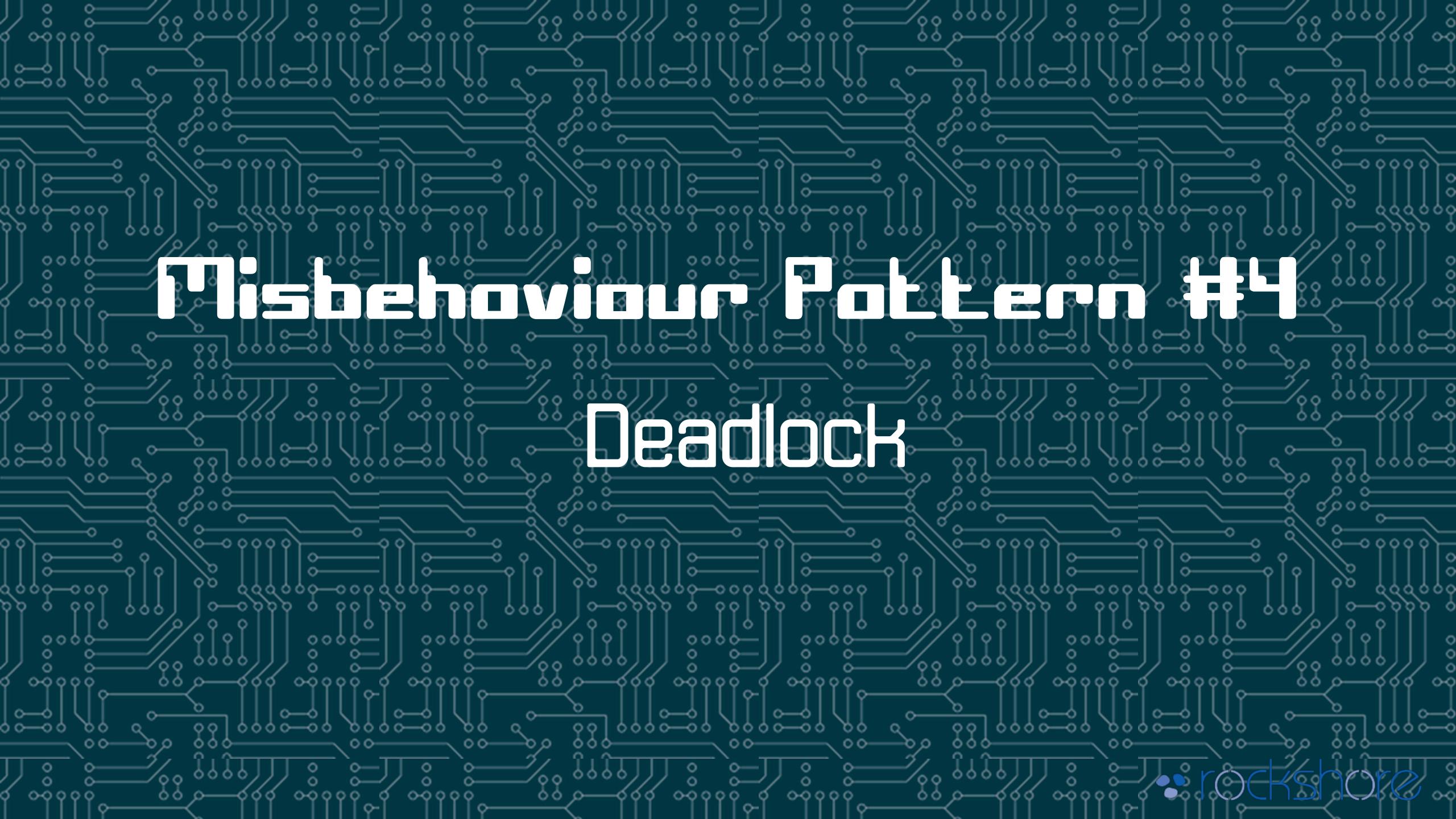
Creating a taxonomy of emergent misbehaviour • Creating a taxonomy of cause patterns Dectection and diagnosis techniques Prediction methods Amelioration techniques and patterns Testing methodologies

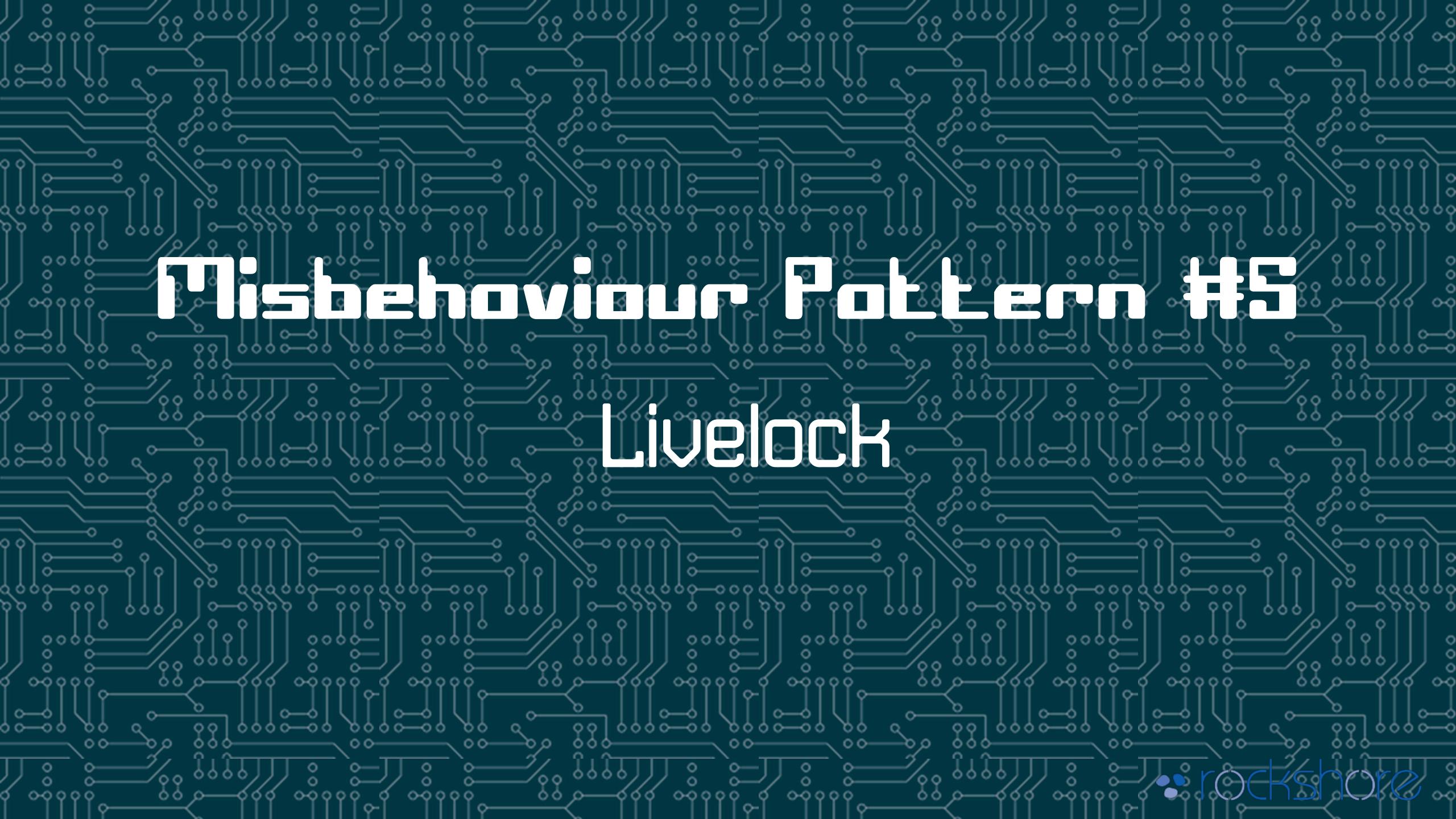




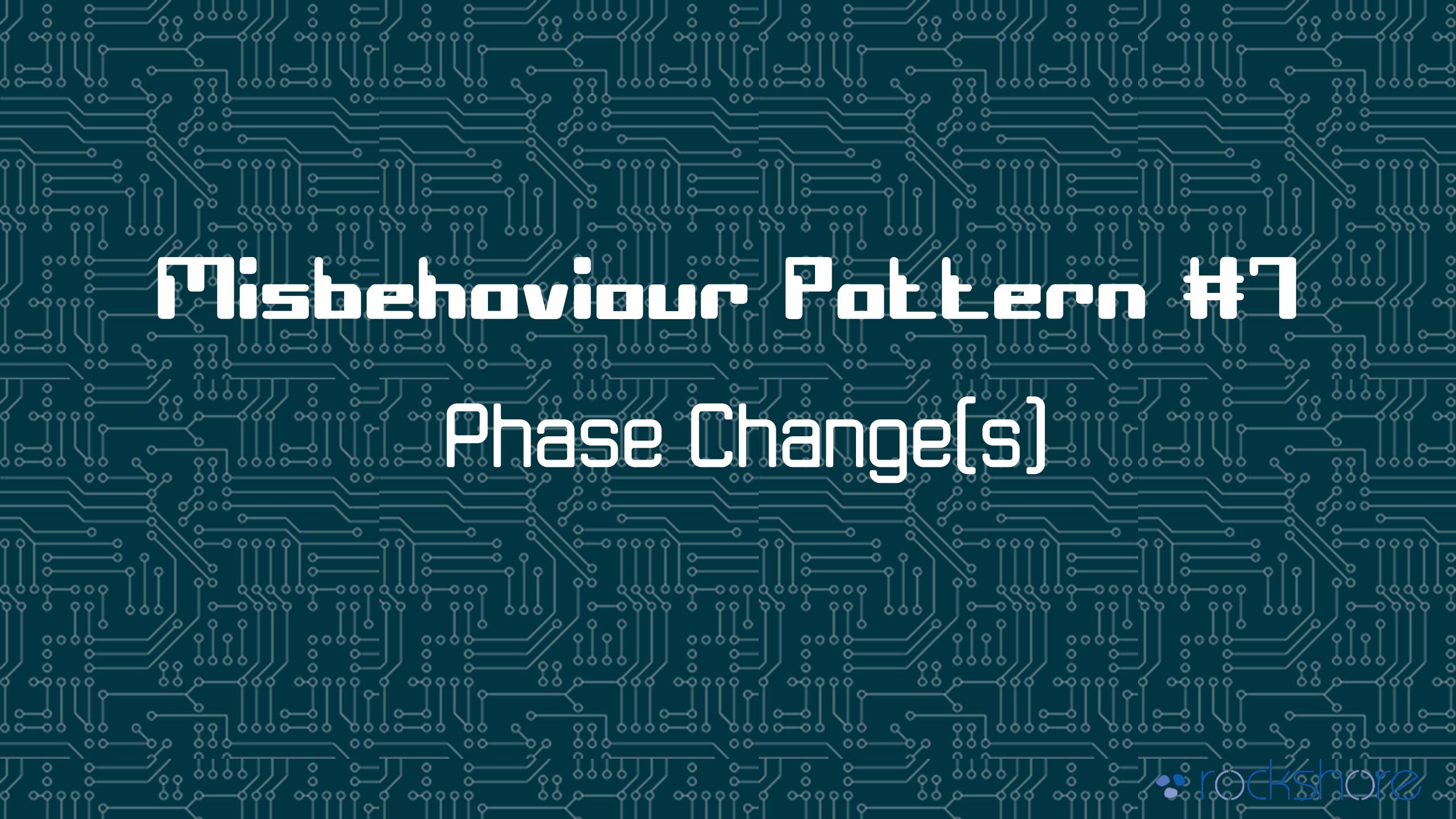


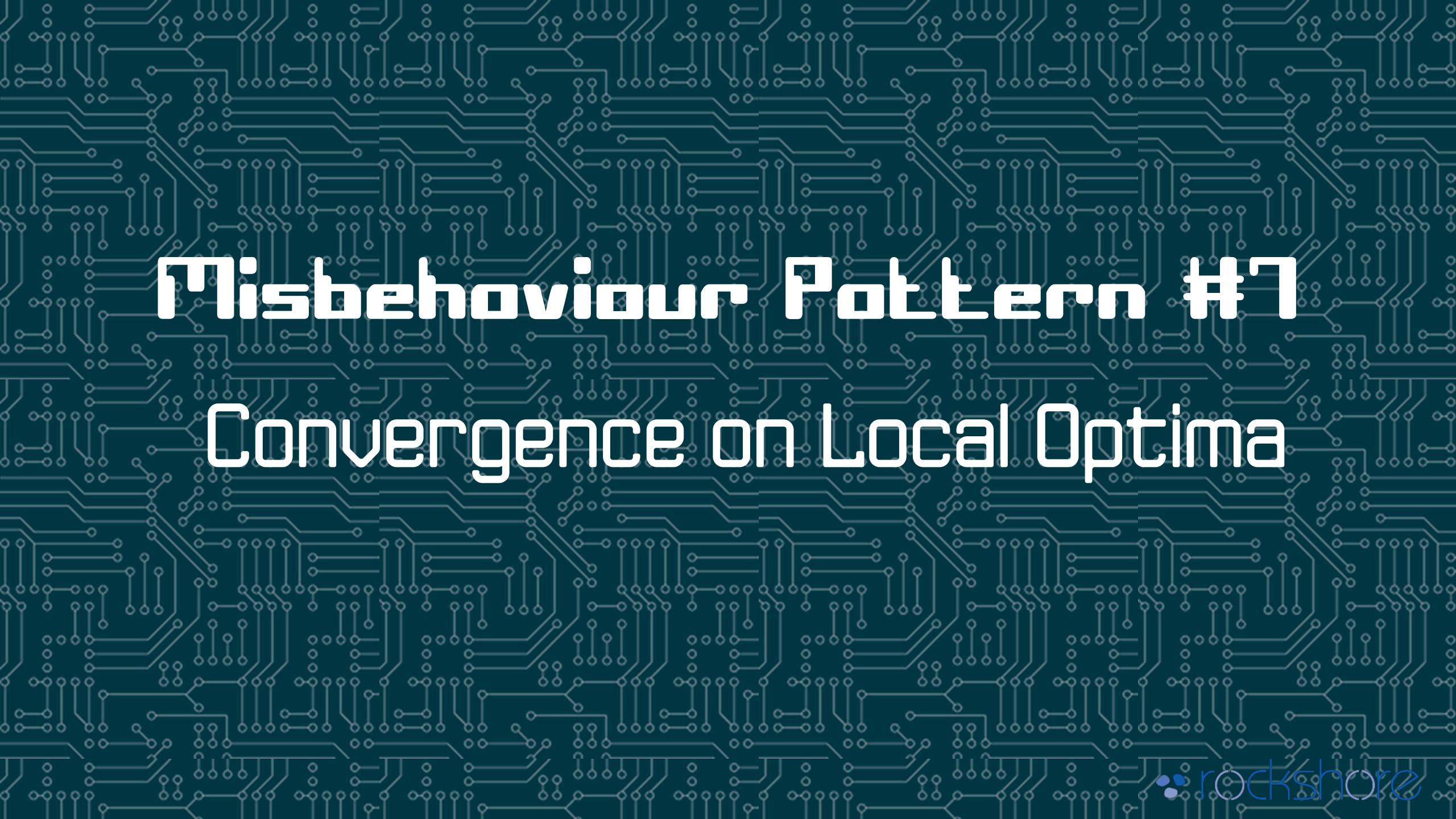


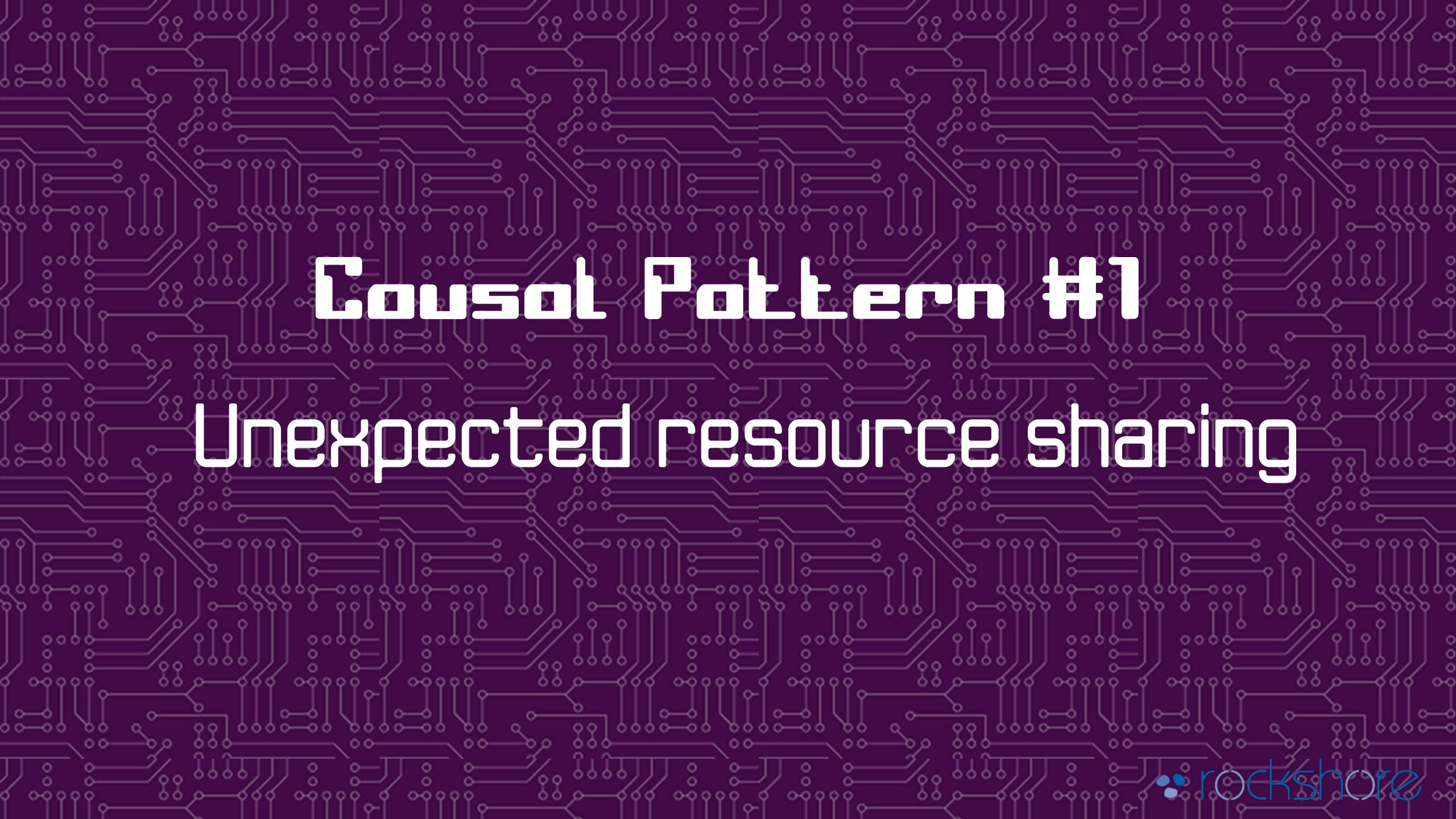


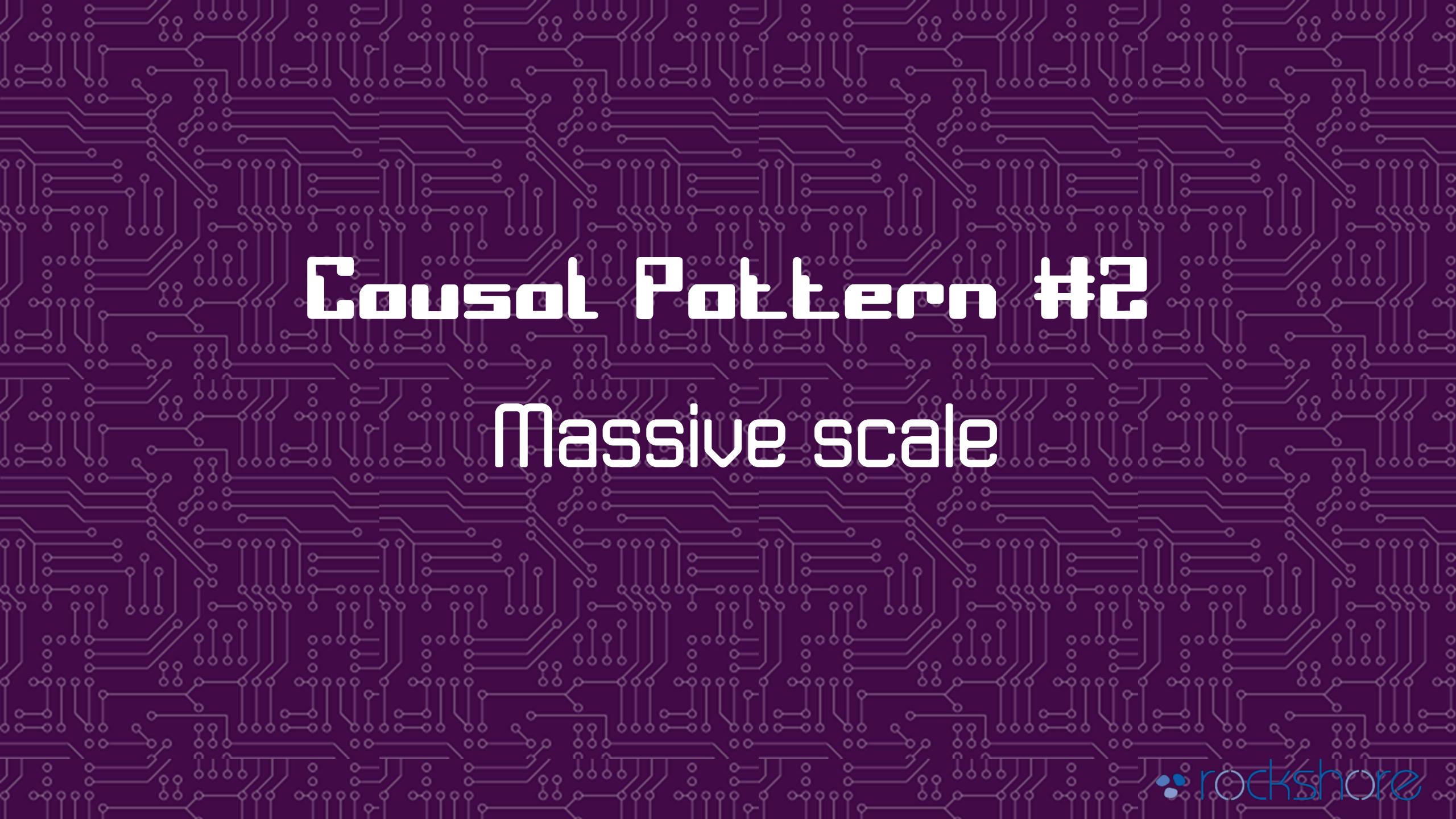


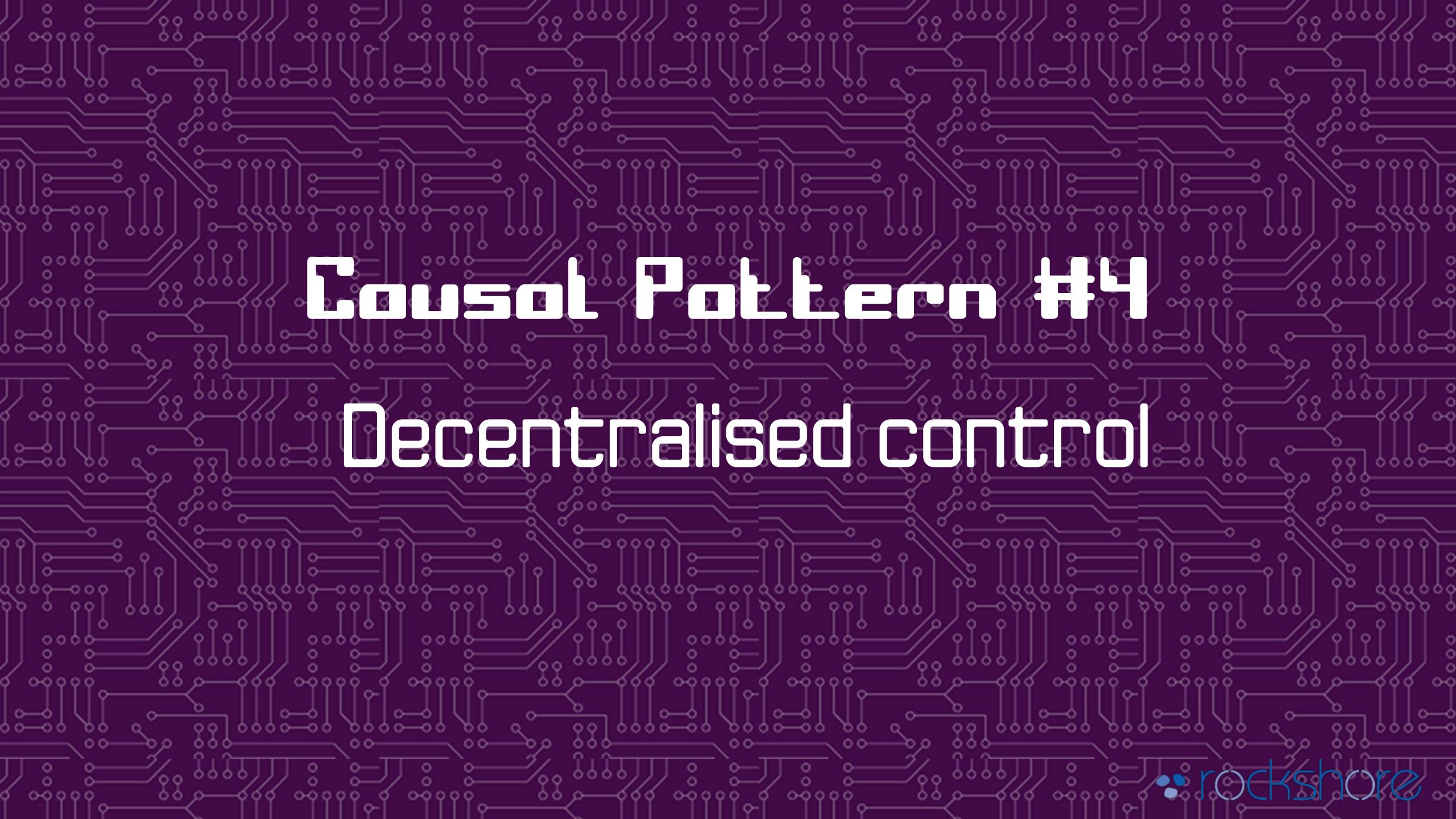


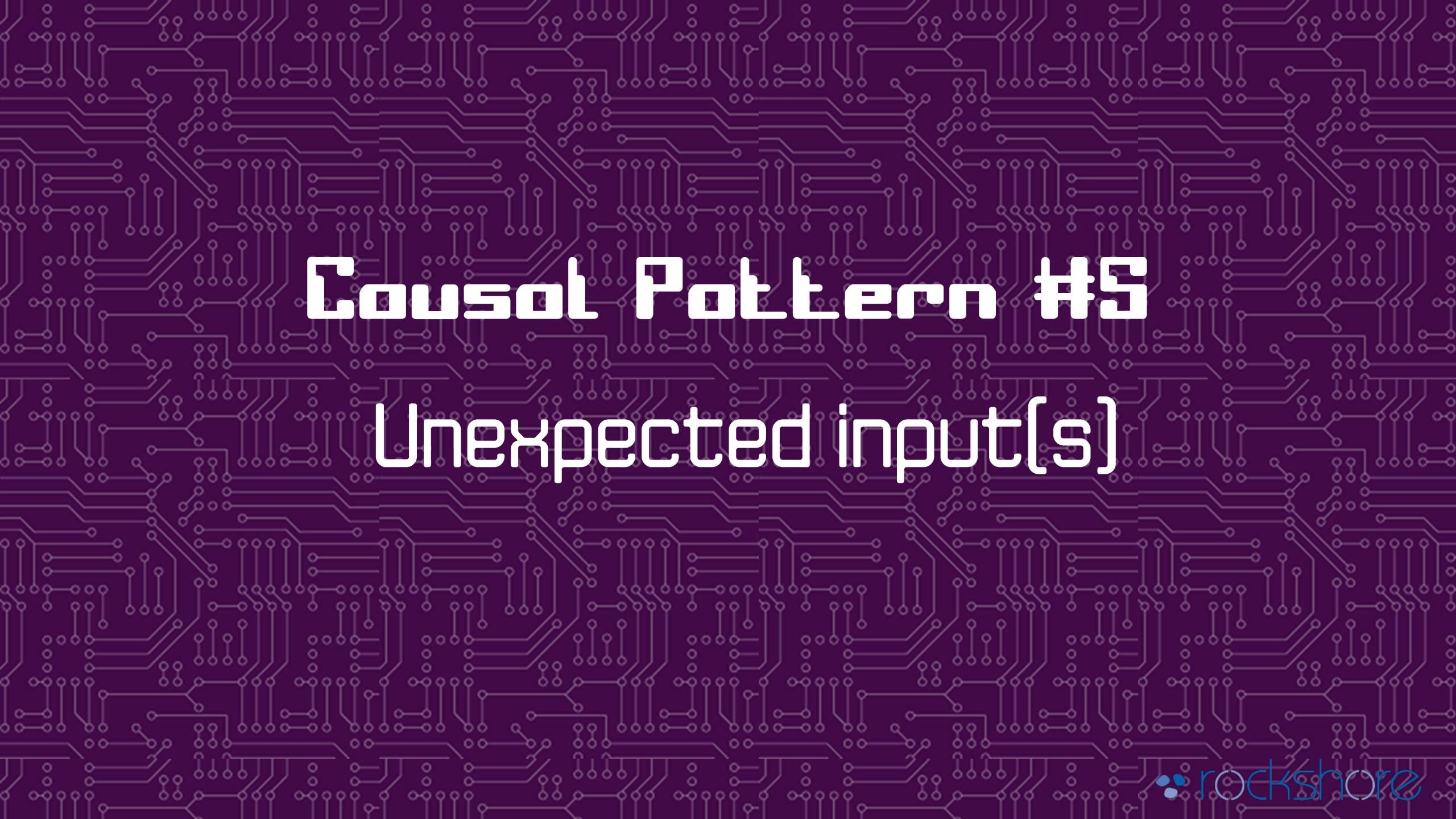


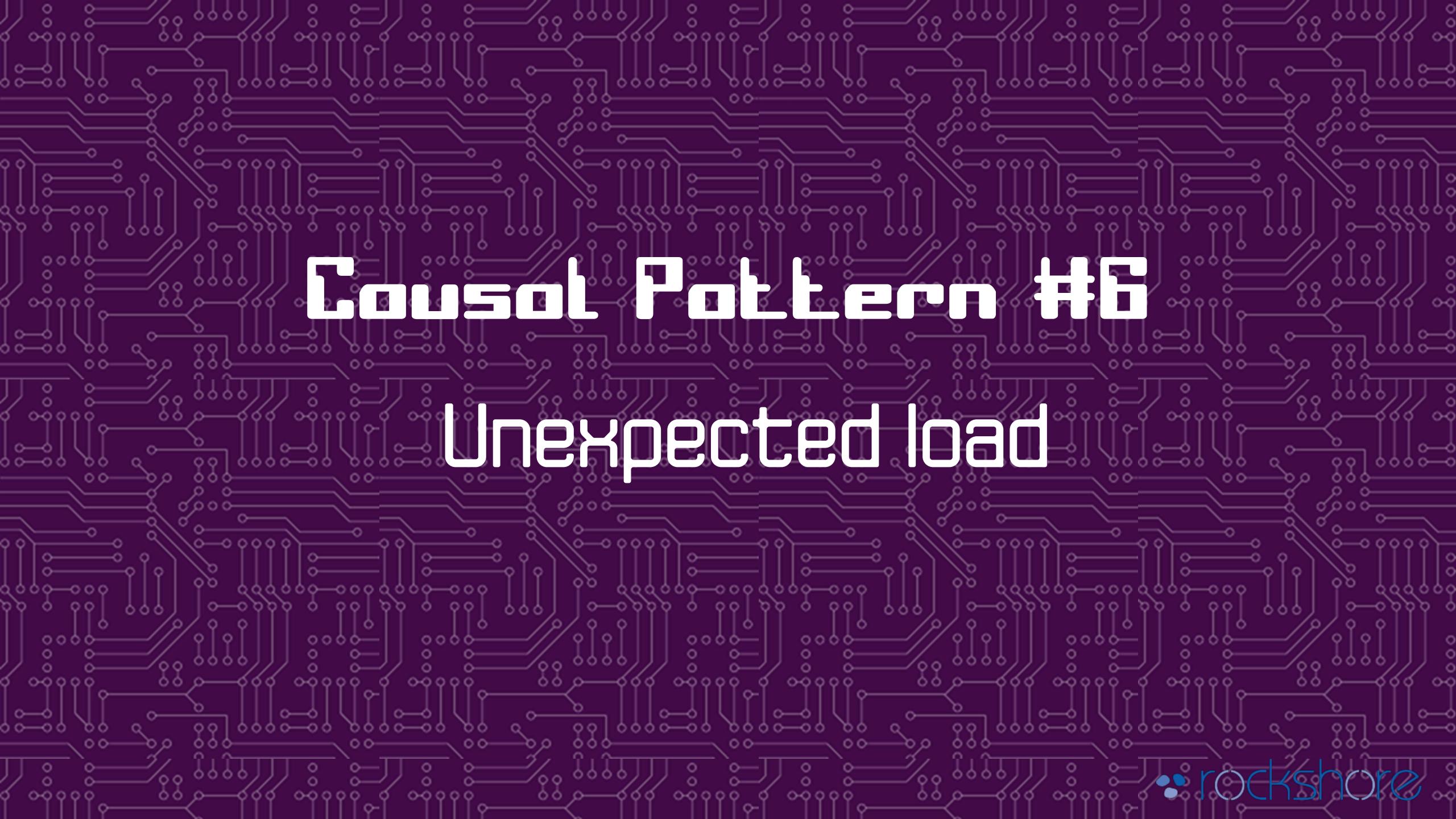


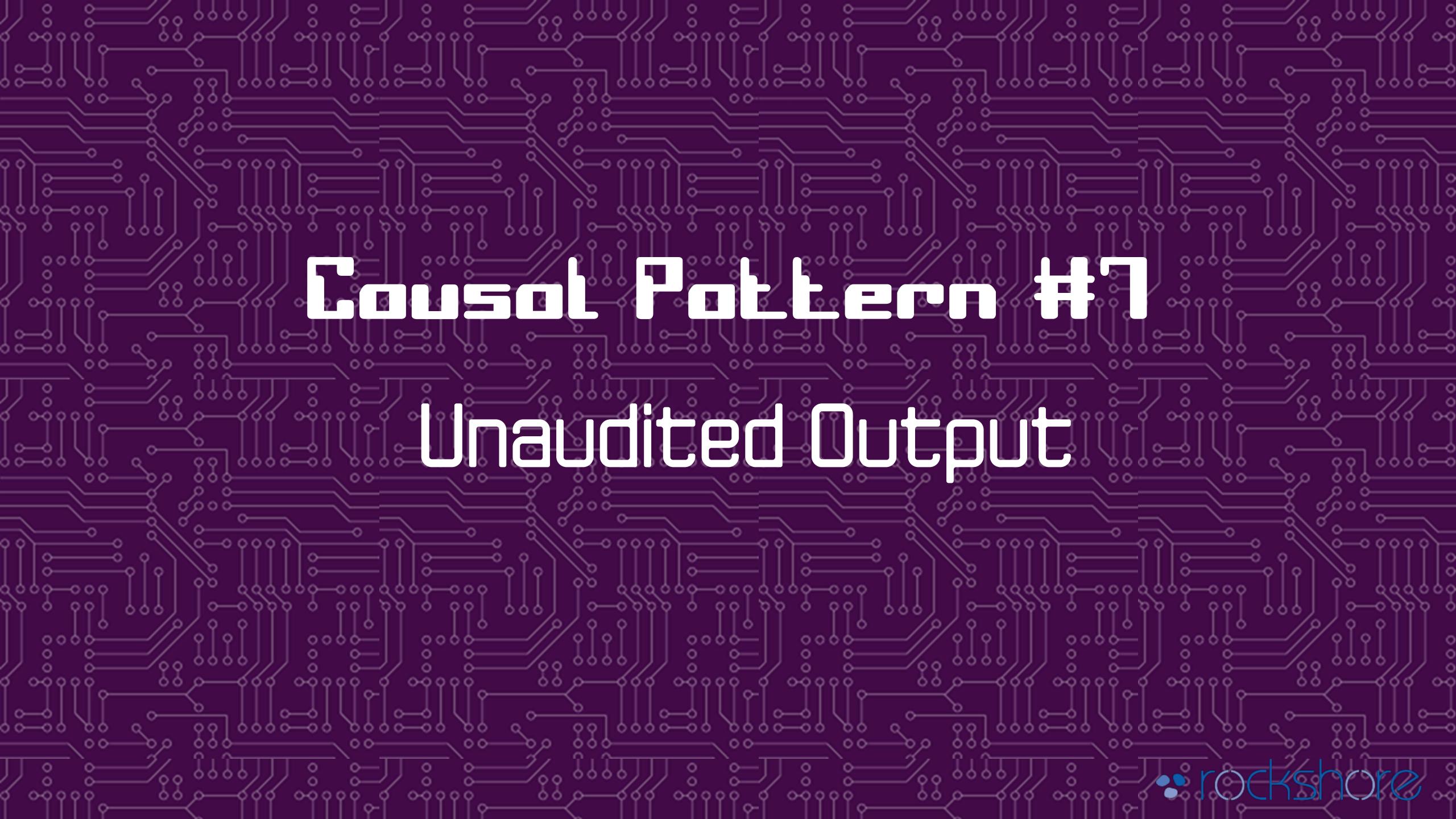












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Classic methods: Iogging - potential to spot simple errors replay - in combination with logs can help diagnose by reproduction

system size and complexity Reconsider what is "superfluous" monitoring!

• heap dumps, gc logs, etc - good for single node/JVM failures

• model checking - complexity/difficulty of task increases with



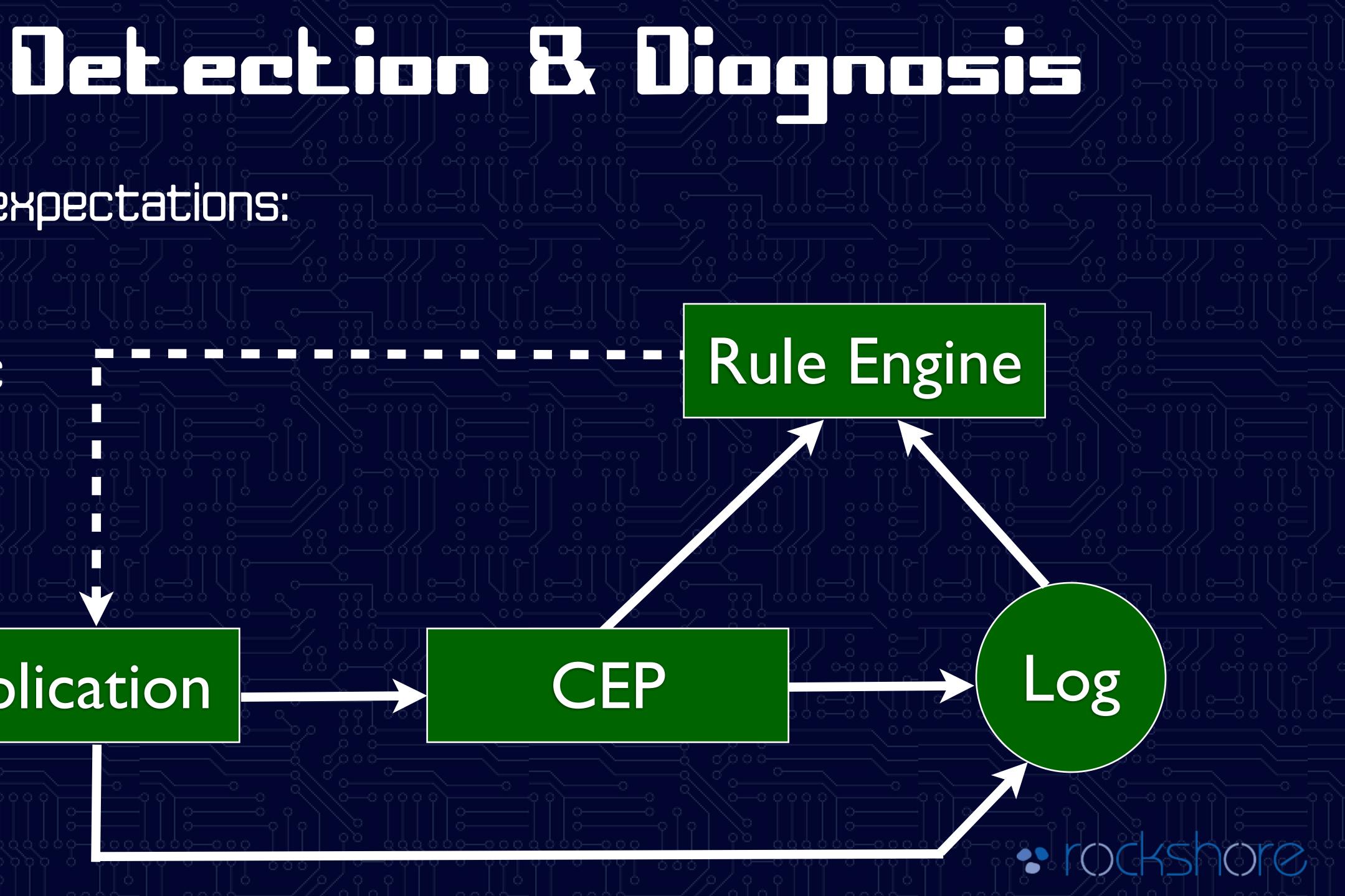
Encode expectations:

PSpec

Pip

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Application



Similar Java tools: Appdynamics, New Relic

RHD

Solunk, VisualVM, Netbeans Profiler, MAT, GC Viewer Chef, Puppet, Ansible, Vagrant, etc. Sadly there is a lack of "out of the box" emergent behaviour monitoring tools and many companies with distributed complex systems spend a great deal of time stitching together a solution from the available tools

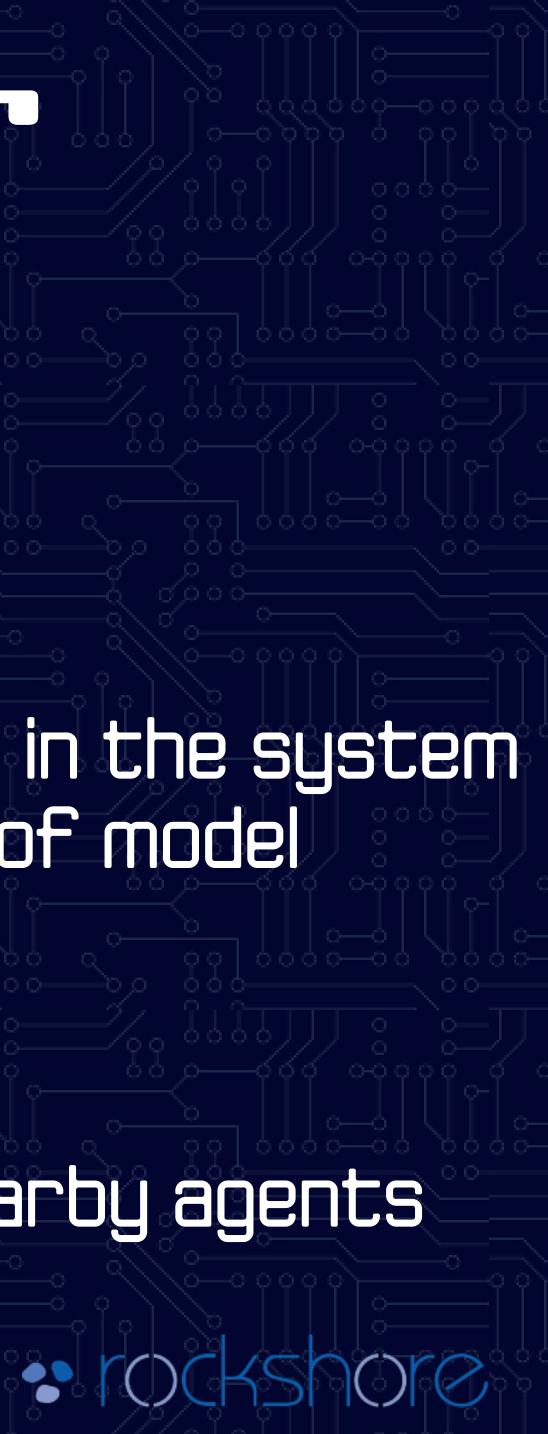


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 Model checking, verification, and validation E.g. Z Notation, VDM (Vienna Development Method) VOMAS (Virtual Overlay Multi-Agent System) invariants and other expectations CrystalBall using a "consequence prediction algorithm"

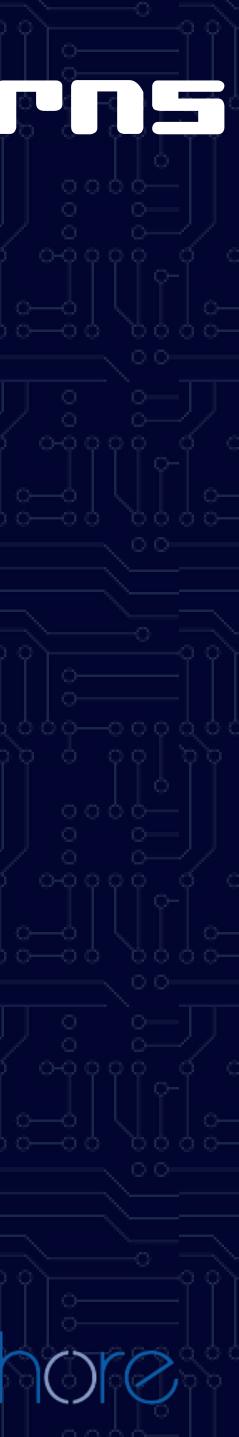
- An additional "layer" of "intelligent" agents operate in the system watching events (logs) and checking for violations of model

Agents predict the effect of the behaviour on nearby agents



Ameliorotion Techniques & Potterns

Randomisation Introduce & expect latency Rate limiting & buffering Over provisioning Introspection & closed control loop adaptive feedback Expect failure & recovery of components Boundary guardians Self-Healing



Since emergent (mis)-behaviour is a property of the complete system the system must be tested as a whole • Test inputs and expectations must be realistic Test environment should match deployment environment Record and replay techniques - DO NOT RELY ON HUMAN TESTING STRATEGIES!! The more tests you define more clearly helps you encode normal behaviour of the system under load(s)





• Learn to classify causes and symptoms of emergent misbehabiour Build systems to monitor behaviour in real-time Define what behaviour is "normal" for your system Build automated alert & response mechanisms • Collect as much runtime performance data as you can Balanced against the performance impact costs • When testing ensure to include full system simulation in representative environments (hardware, input data, etc) • When designing complex systems allow for injection of controlling variables of execution, e.g. latency, throttling, memory allocation, etc. These allow for the fine-tuning of the performance and behaviour of your system.







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