EMERGENT SOFTWARE SYSTEMS

Summer School

Barry Porter & Roberto Rodrigues Filho School of Computing and Communications Lancaster University Funded by The Royal Society Newton Fund





Distributed Emergent Systems

Applying our concept to complex distributed systems

 Starting from an objective and reward, how do we assemble and learn the best collection of behaviours?

- How do we convert all of the decision making in distributed systems into a simple action / reward model?
 - While guaranteeing that the system will always be "functional"

Distributed Emergent Systems

 We project our local component model directly into a distributed system, and we introduce the idea of "Class A" and "Class B" distributed interactions

- Given a set of available machines M, this allows us to decide on placement of all needed sub-systems; the replication factor of those systems; and the internal composition of each sub-system
 - All guided from one reward function

SELF-DISTRIBUTING SYSTEMS

Class A

 Let's imagine that we're building an entire datacentre ecosystem

 We have the necessary building blocks for a web server, database, and memcached system

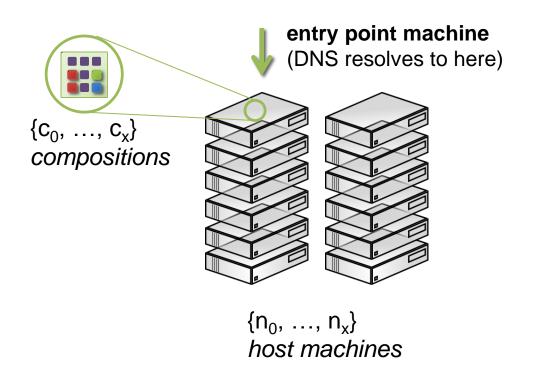
We'll start by considering only the web server system

 We have a set of available machines which can host components

 We have an entry point machine to which ingress traffic (HTTP requests from users) arrives to the datacentre

 We have a set of compositions which can form the functionality of a web server (receive HTTP request, respond with resource)

To illustrate:



Dana gives us two useful capabilities here

 We can hot-swap a component at runtime to a different implementation

 We can see which interfaces have state (declared as transfer state)

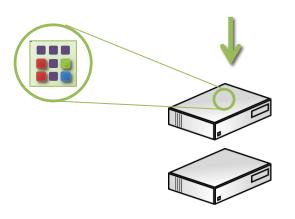
 This leads to the idea that we could relocate a component to a different host machine at runtime

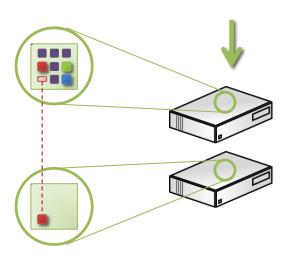
 To do this we would create a proxy version of the interface which forwards function calls to a remote version, marshalling / unmarshalling parameters / return values

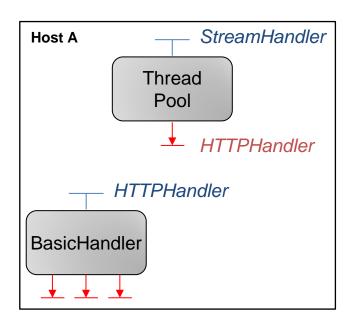
 By relocating a component we might move it closer to the source of data it uses, or we might benefit from extra computation power of another host machine

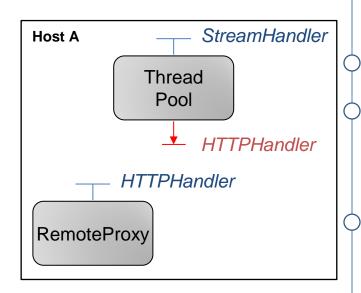
 But we can do more: if an interface has no associated state, we can replicate a component across multiple hosts

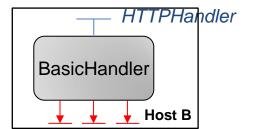
 Even if an interface does have state, we can still replicate it if we have a custom-built plug-in to manage the state

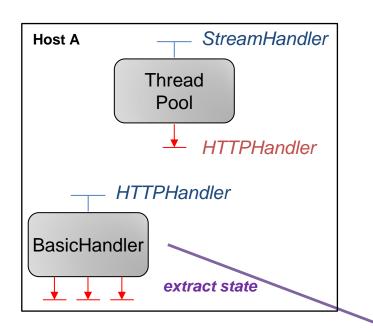


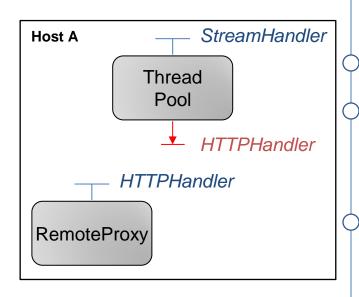


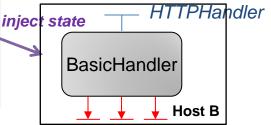


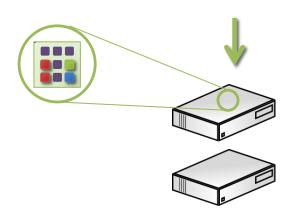


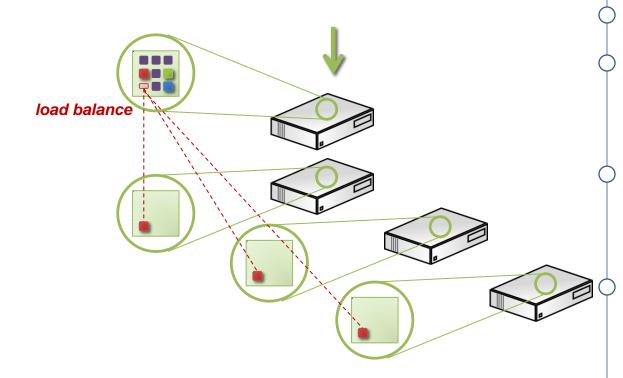


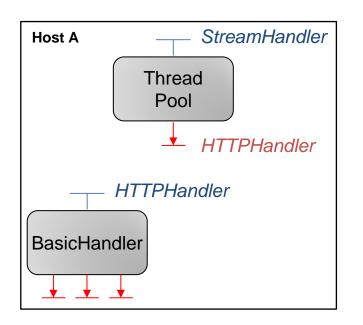


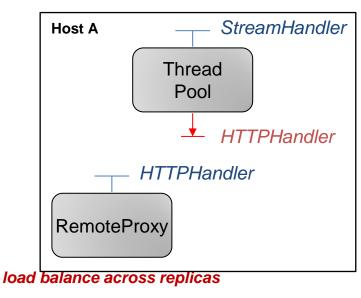


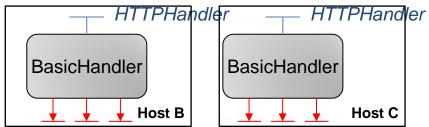


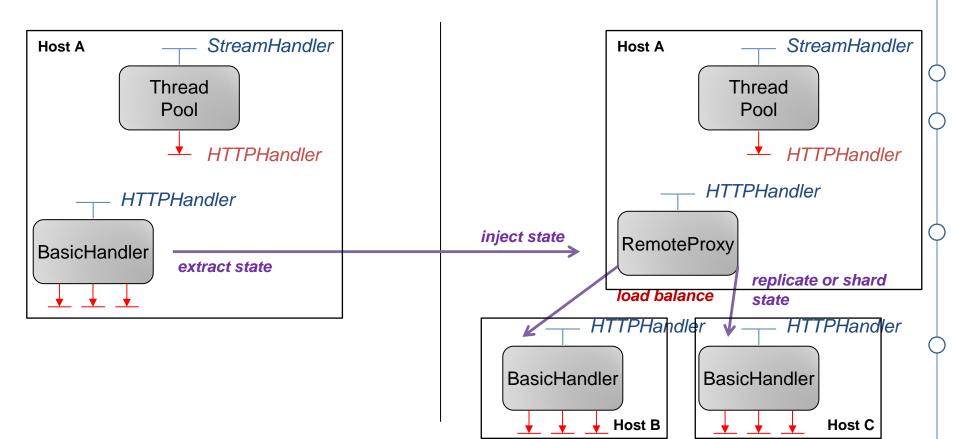




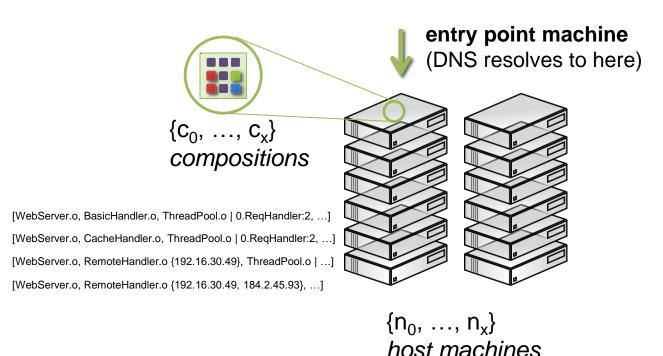




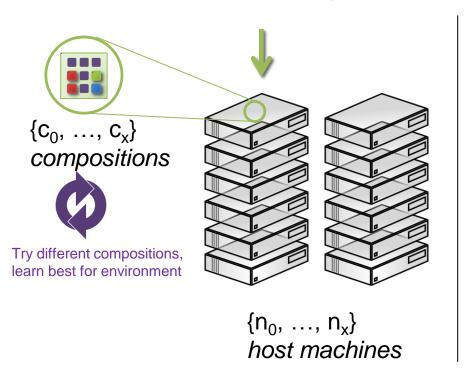


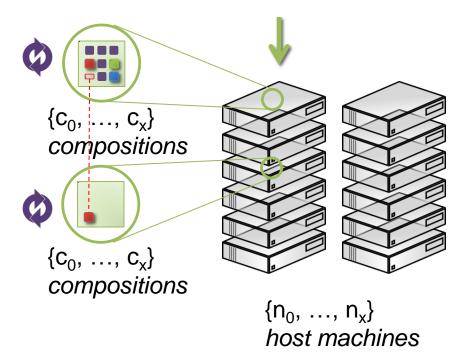


What does our action list look like?



How does learning work?





 What we're doing here is distributing code that was never designed for distribution

What if a remote host fails under this distribution model?

 A seminal paper by Waldo et al presents a concise analysis of the problems with distributed objects

Latency

 If distributing objects introduces unexpectedly high latency compared to a local object interaction, this can impact system behaviour in unexpected ways

Memory access

 Some programming paradigms have shared writable memory between objects, which is very hard to synchronise across a distributed system

Distribution causes partial failures

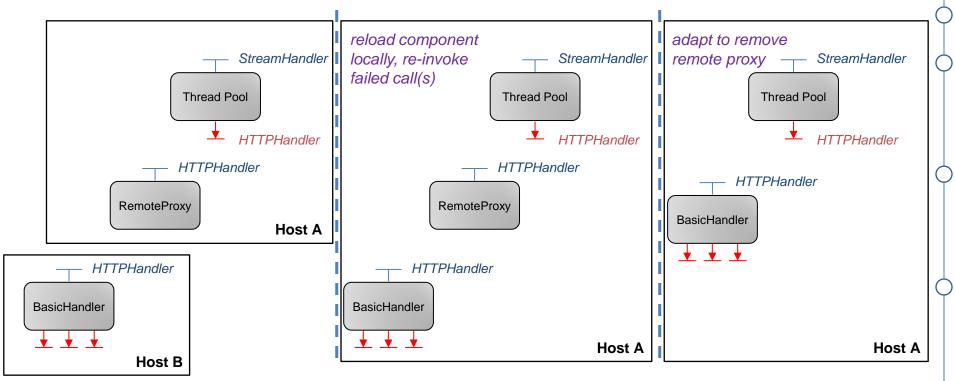
- In normal conditions, a local system is either fully working or not working at all, so there is no need for the programmer to deal with "partial failure" where one particular object stops functioning
- In distributed systems, partial failures are the norm, as network links experience dropouts and host machines fail
- Waldo et al propose two possible solutions to this situation for distributed object systems

Distribution causes partial failures

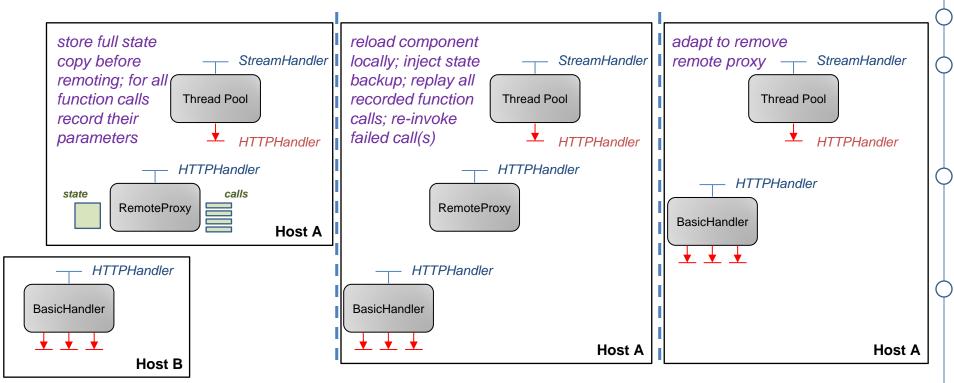
- Program everything as if all interactions are always local, and ignore remote errors caused by network or host problems; this can lead to catastrophic system conditions as remote errors are supressed without any available handling code
- Program everything as if all interactions are always distributed (even if they're local), so that there is explicit fault handling code for every function call; this introduces huge volumes of failure handling code in every object and also brings associated runtime costs

- Distribution causes partial failures
 - We take a third path: we program everything as if all interactions are always local, so there's no programmer overhead
 - **But** we only ever distribute an object if we can guarantee that we can *automatically* and *seamlessly* recover from remote errors
 - When distributing a component we therefore introduce failure handling infrastructure which can recover automatically from errors without the system ever noticing that an error occurred
 - This is **not** the same as "masking" remote errors, because we're making guarantees that system integrity is assured with auto-recovery

Mechanics of fault tolerance for automated distribution



Mechanics of fault tolerance for automated distribution



SELF-DISTRIBUTING SYSTEMS

Class B

 We've seen how automated distribution of apparently local code requires an automated and seamless approach to ensuring fault-tolerance of the overall system

 For interfaces with large amounts of state, and with frequent interactions, supporting this requirement this becomes very expensive

• To counter this issue, we introduce the idea of *explicitly* distributed interfaces, which we call "Class B" proxies

 These interfaces are designed to communicate with a remote service, and have explicit error handling built into the interface definition for remote failures

 Components using a Class B interface will be designed to deal with remote failures reported by the interface

 Because failure handling logic is explicit for Class B interfaces, we do not need to employ seamless recovery and rather can leave decision logic to the application

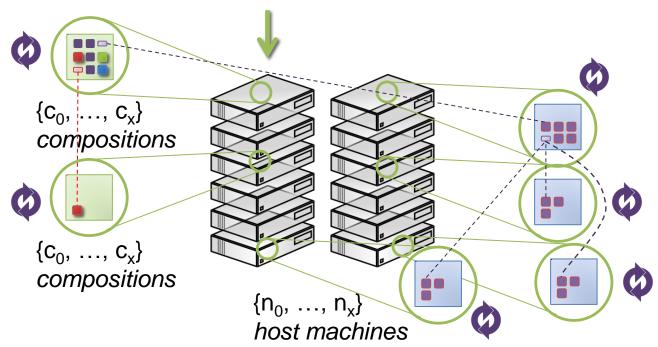
 It's much more common for things like databases to be represented in this way

 Besides this detail, the action / reward matrix looks identical to the Class A case

Each learning action is one composition, but a composition may include remoting or replication of particular components to particular hosts

This allows us to learn placement, co-location, or replication factors of all elements of a system, from a simple action list

When we move a component to a remote host, we start a new learning agent at that host to learn about the sub-composition



Summary

- We can take advantage of very strong encapsulation to allow local components to be distributed over a network
 - This applies generically to any component, including buttons on a user interface or machine learning implementations
- Doing this requires automated fault tolerance so that failures are seamlessly recovered
- The result is the ability to learn where to place each element of system logic and how much replication to apply, all from a simple action/reward matrix