

# Situational Awareness for Industrial Operations

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Research School of Computer Science

# Problem Context

## Factory Floor

- Are the operations carried out according to the schedule?

## Food Supply Chain

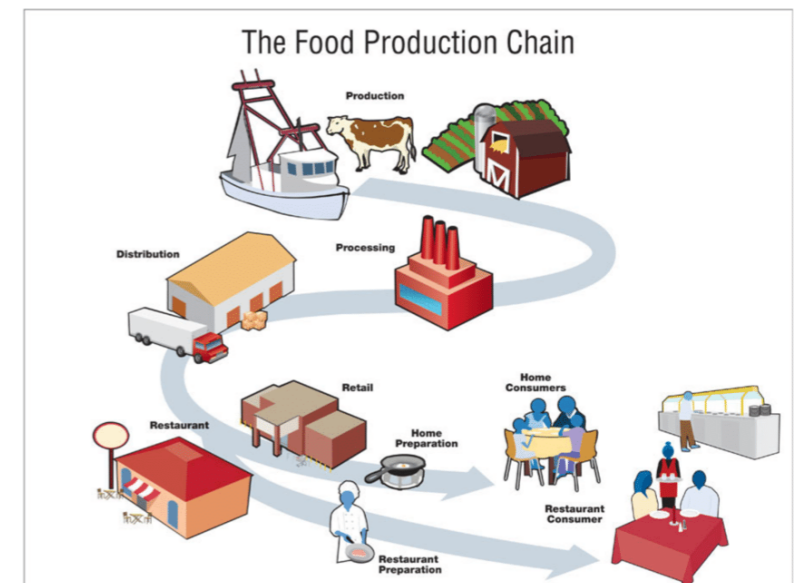
- Are the tomatoes delivered within 3 hours and stored below 25°C?
- Is “sold milk quantity”  $\leq$  “produced milk quantity”?

## Data Cleansing

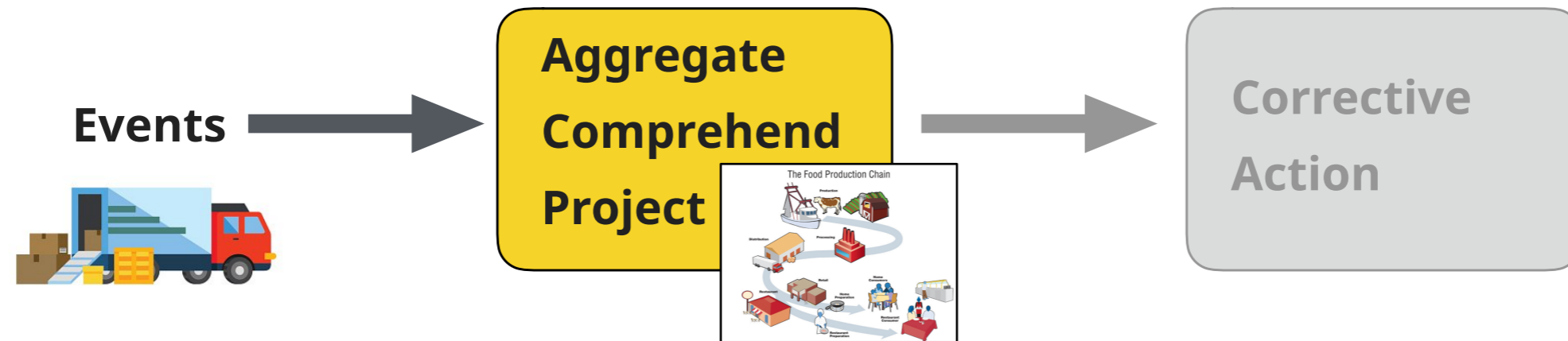
- Does the database have complete, correct, accurate and relevant data?

## This Work

**A situational awareness approach to answer such questions**

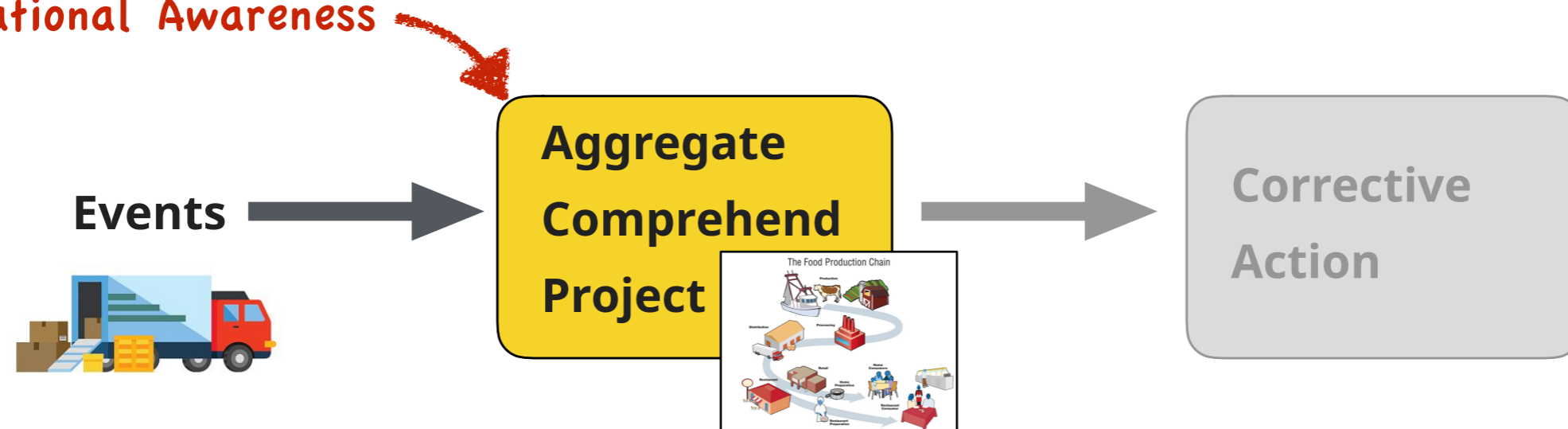


# Situational Awareness



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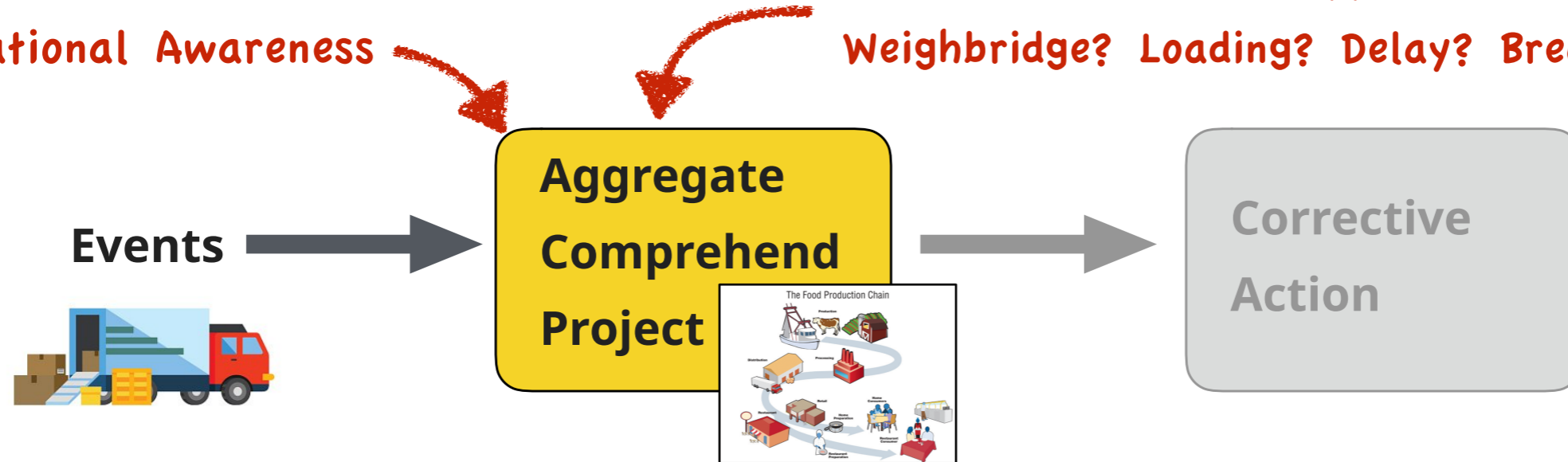
Situational Awareness



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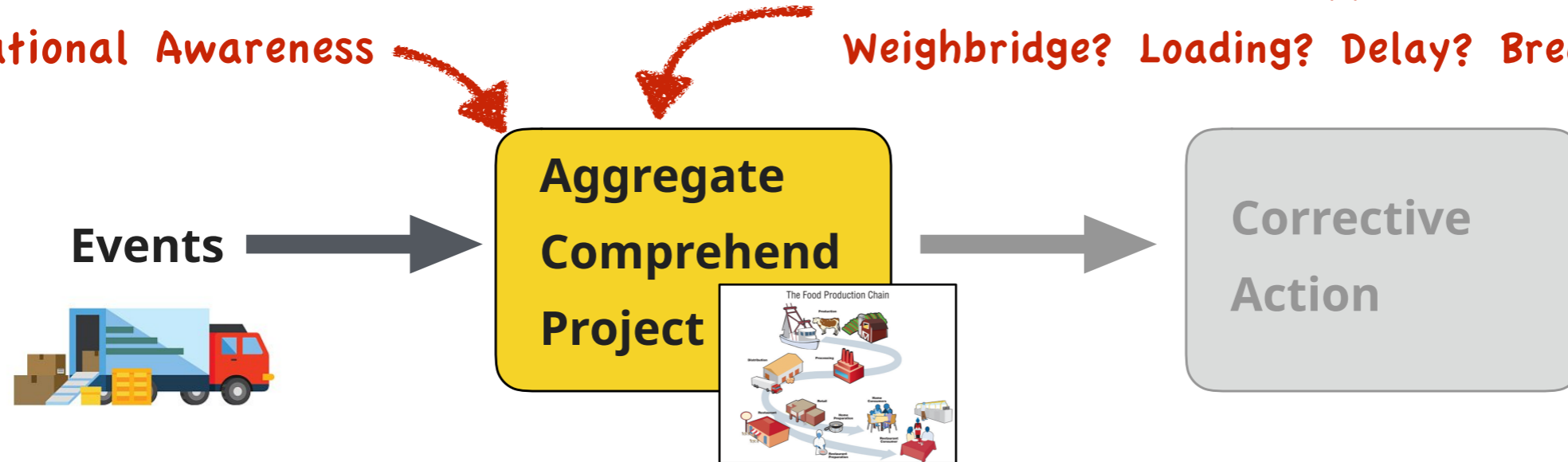
E.g. Given a truck's GPS trace,  
why has the truck stopped?  
Weighbridge? Loading? Delay? Break?



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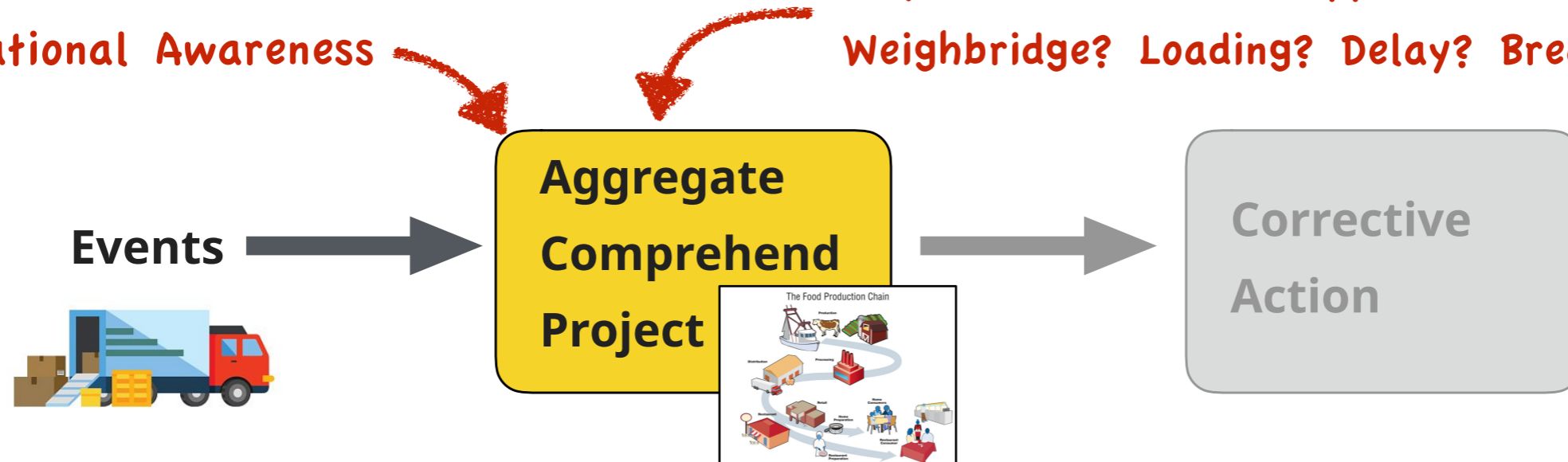
## Situational Awareness System Design Issues

- Want re-usability across domains
- What is an appropriate general system modelling language?
- How to derive situational awareness from events + model?
- How to deal with incomplete/noisy/erroneous/absent events

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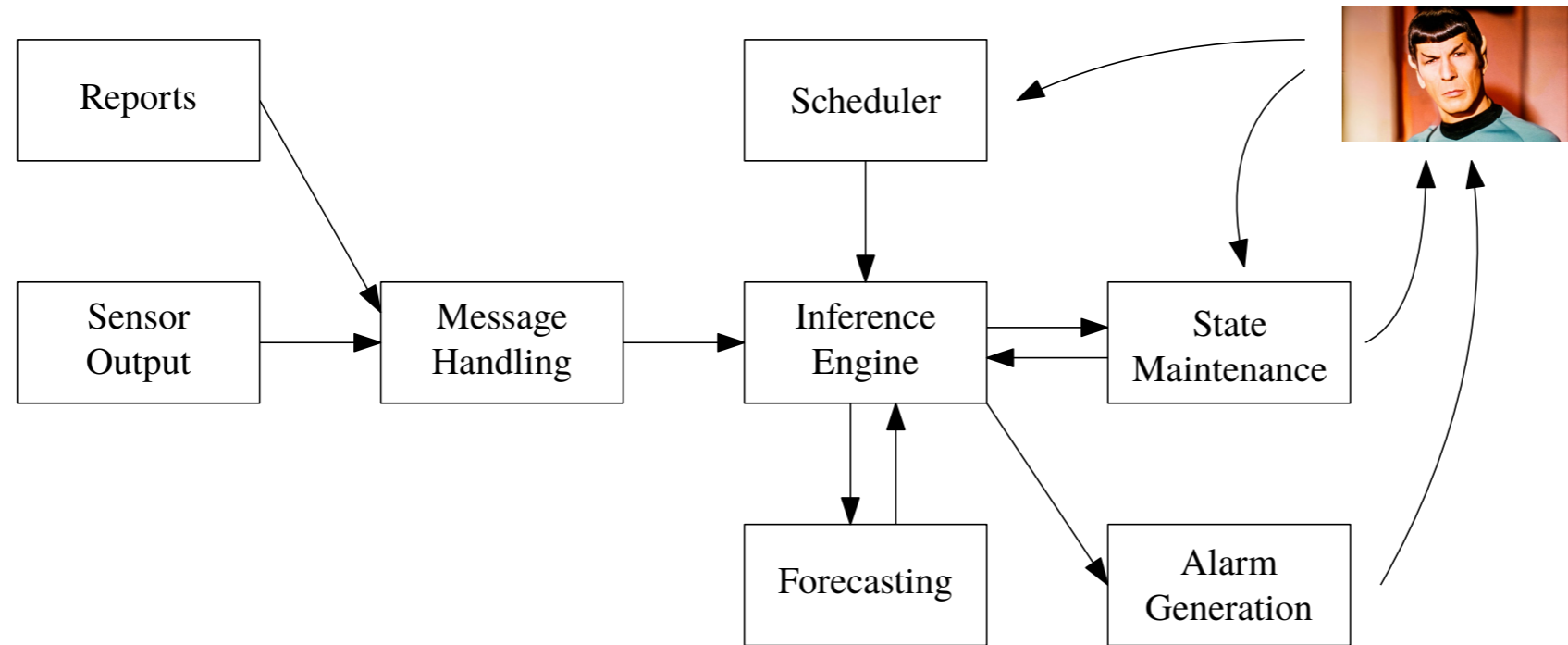
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## This Talk

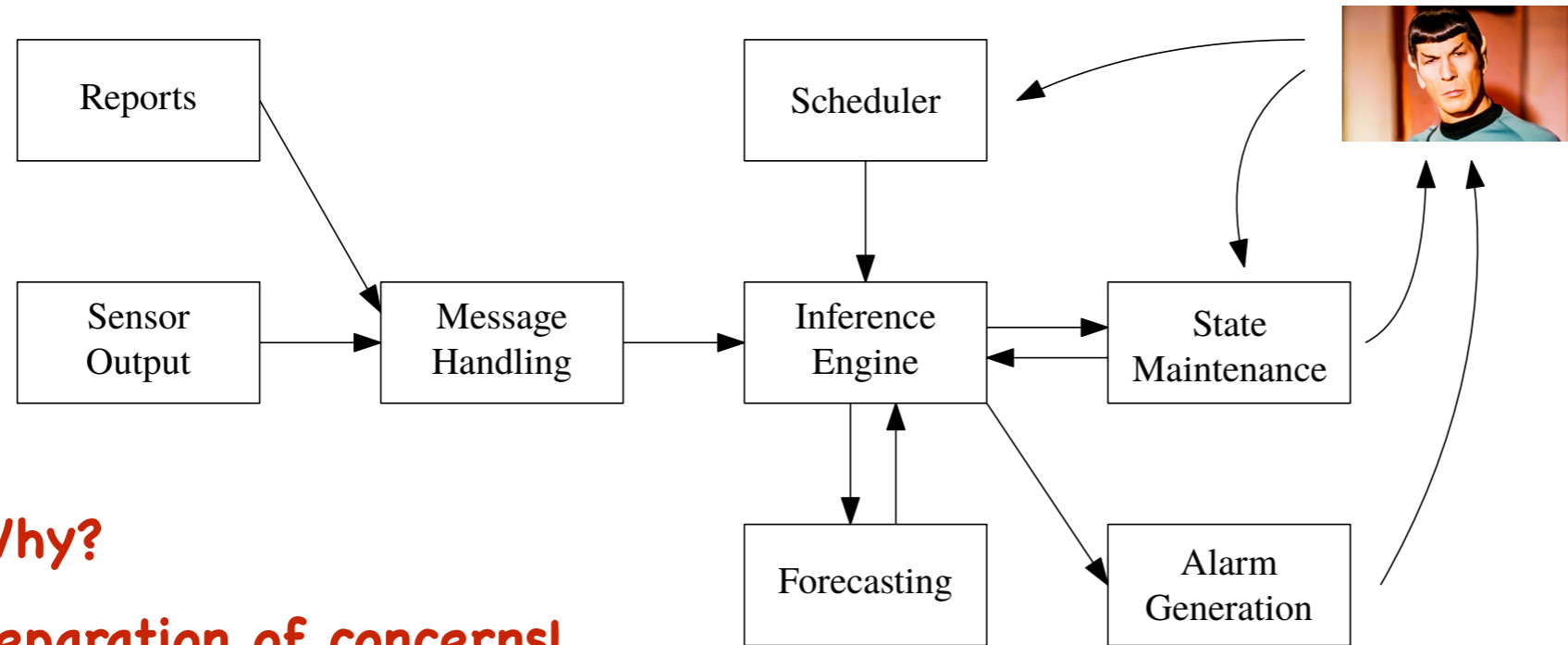
- **System Architecture**
- **Modelling Language**
- **Inference Engine and State Maintenance**
- **Implementation on top of Scala**

# System Architecture





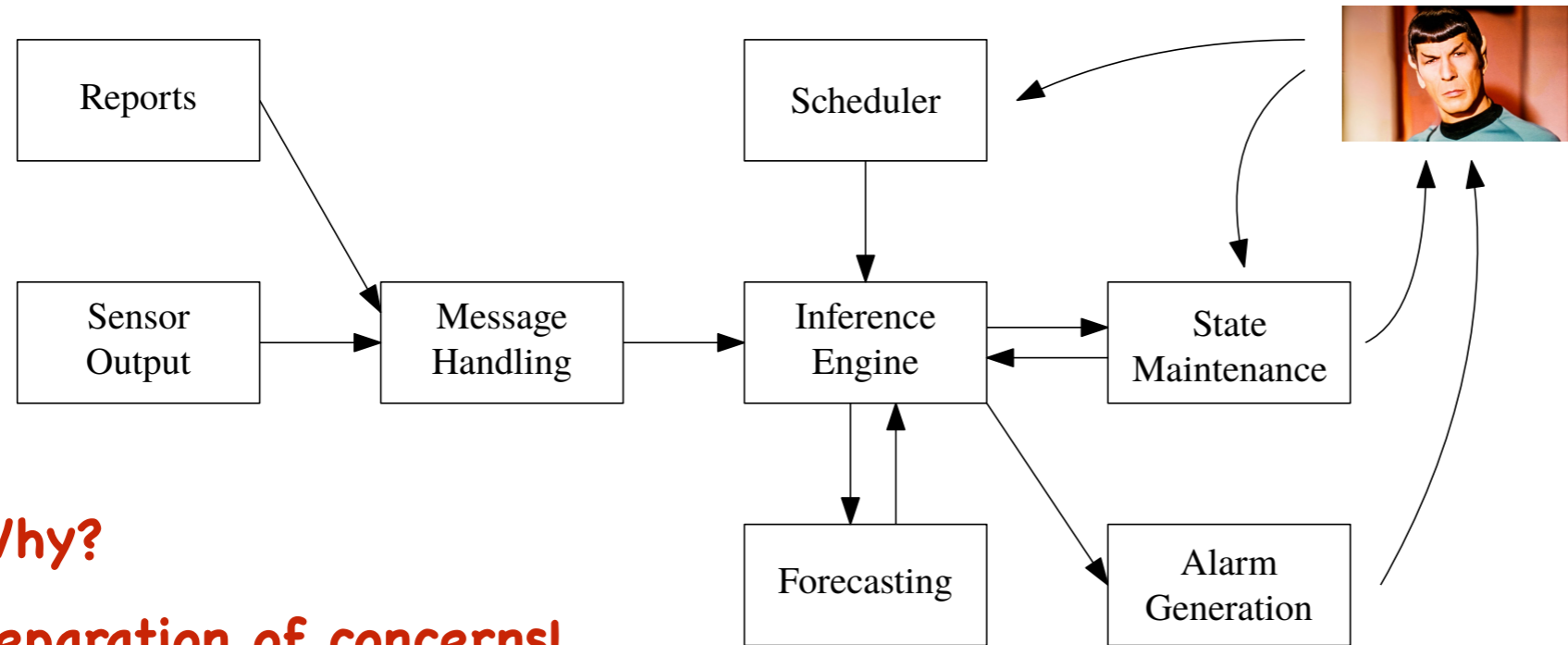
# System Architecture



**Why?**

**Separation of concerns!**

# System Architecture

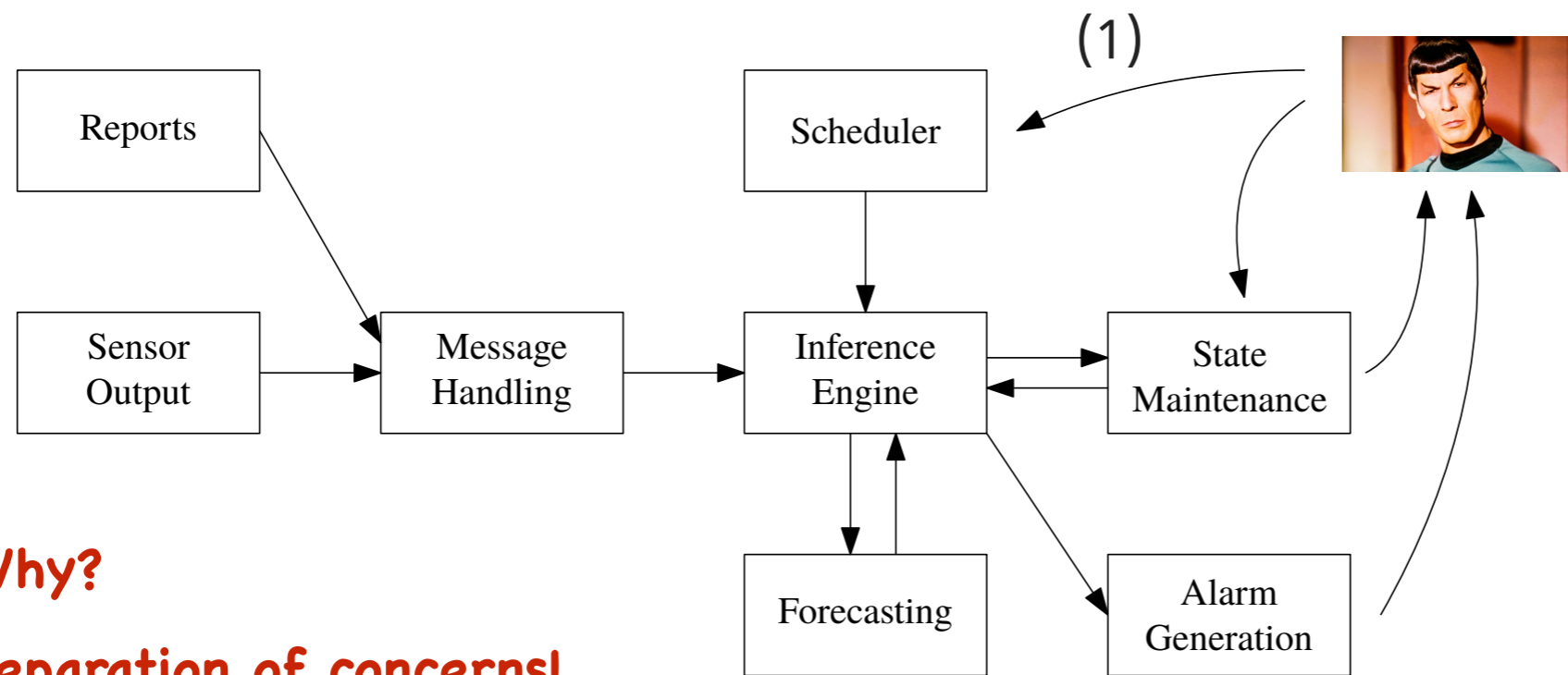


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**Execution**

# System Architecture



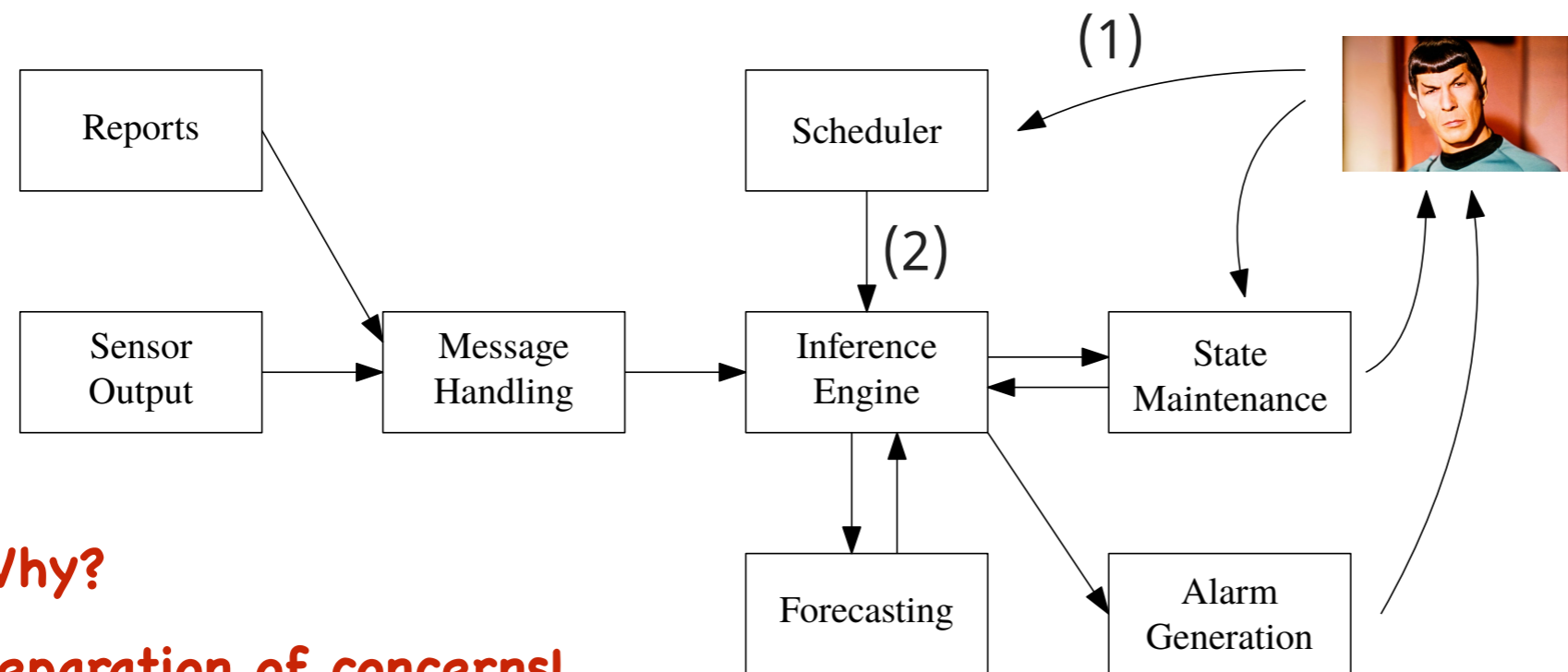
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## Execution

(1) Human Operator invokes Scheduler

# System Architecture



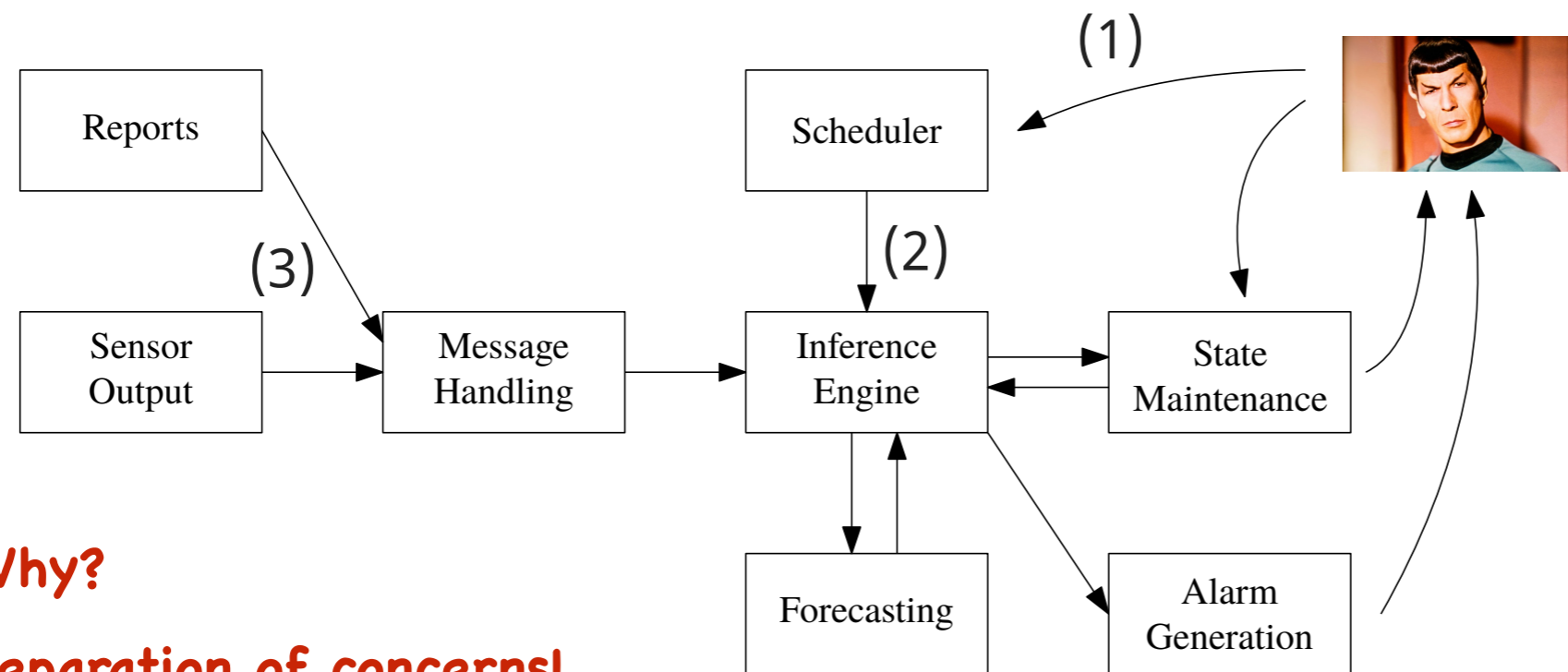
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## Execution

- (1) Human Operator invokes Scheduler
- (2) Schedule given to Inference Engine defining *expected* course of events

# System Architecture



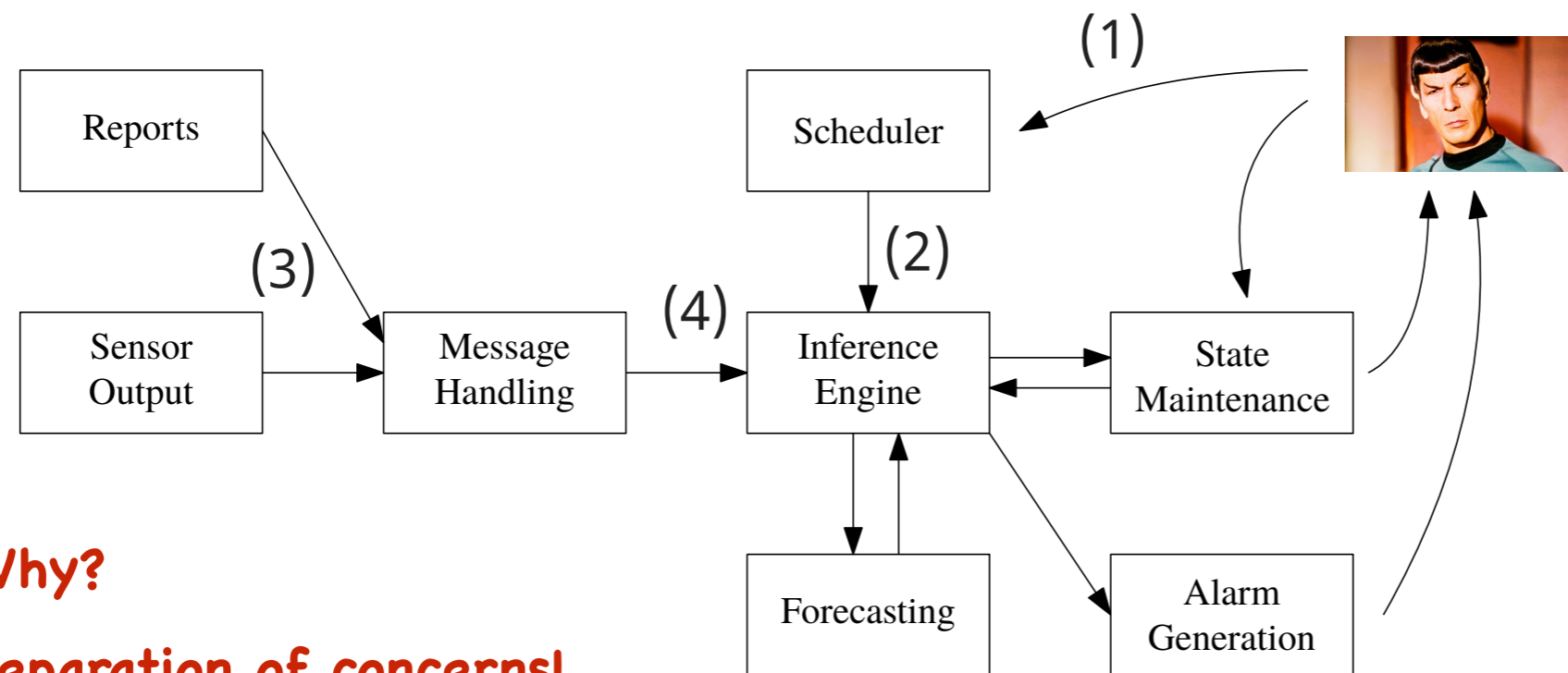
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## Execution

- (1) Human Operator invokes Scheduler
- (2) Schedule given to Inference Engine defining *expected* course of events
- (3) Collect messages from external world e.g. event reports and sensor output

# System Architecture



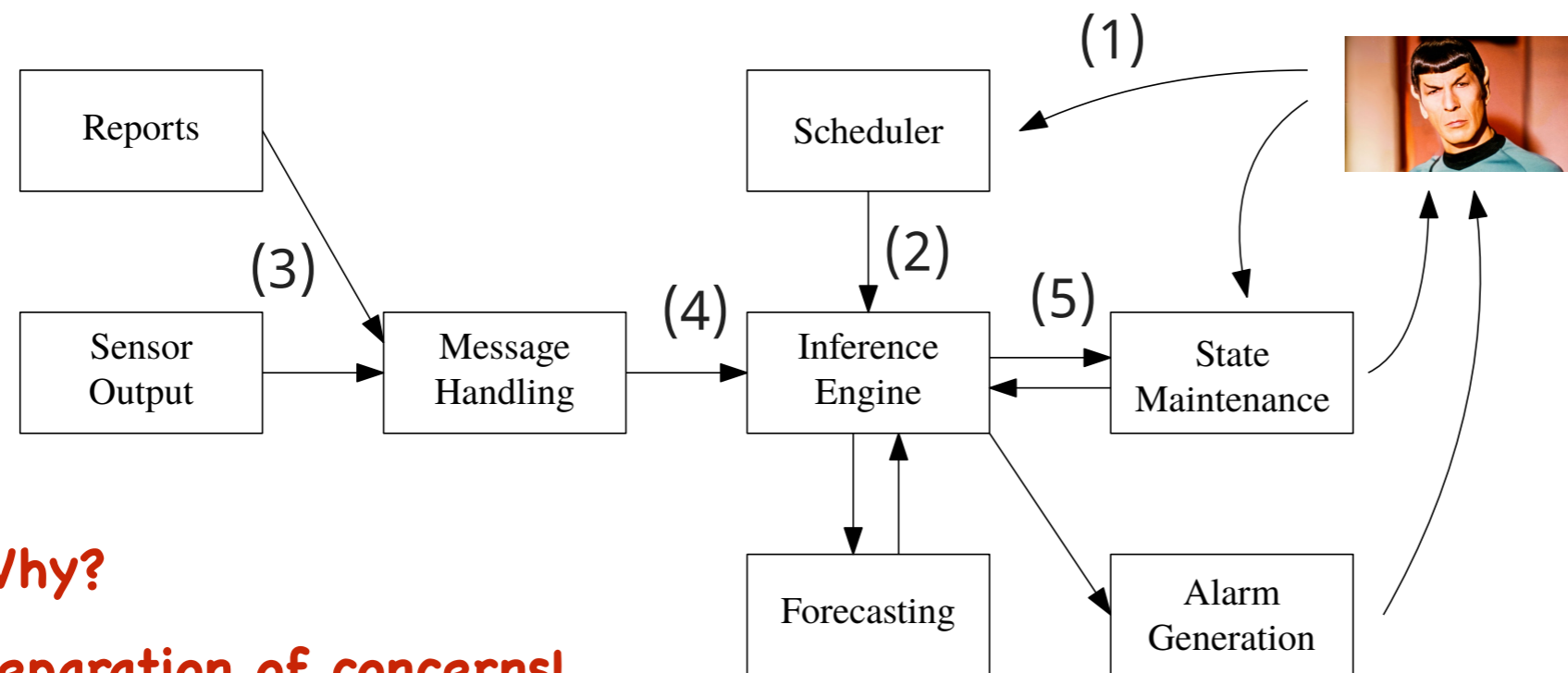
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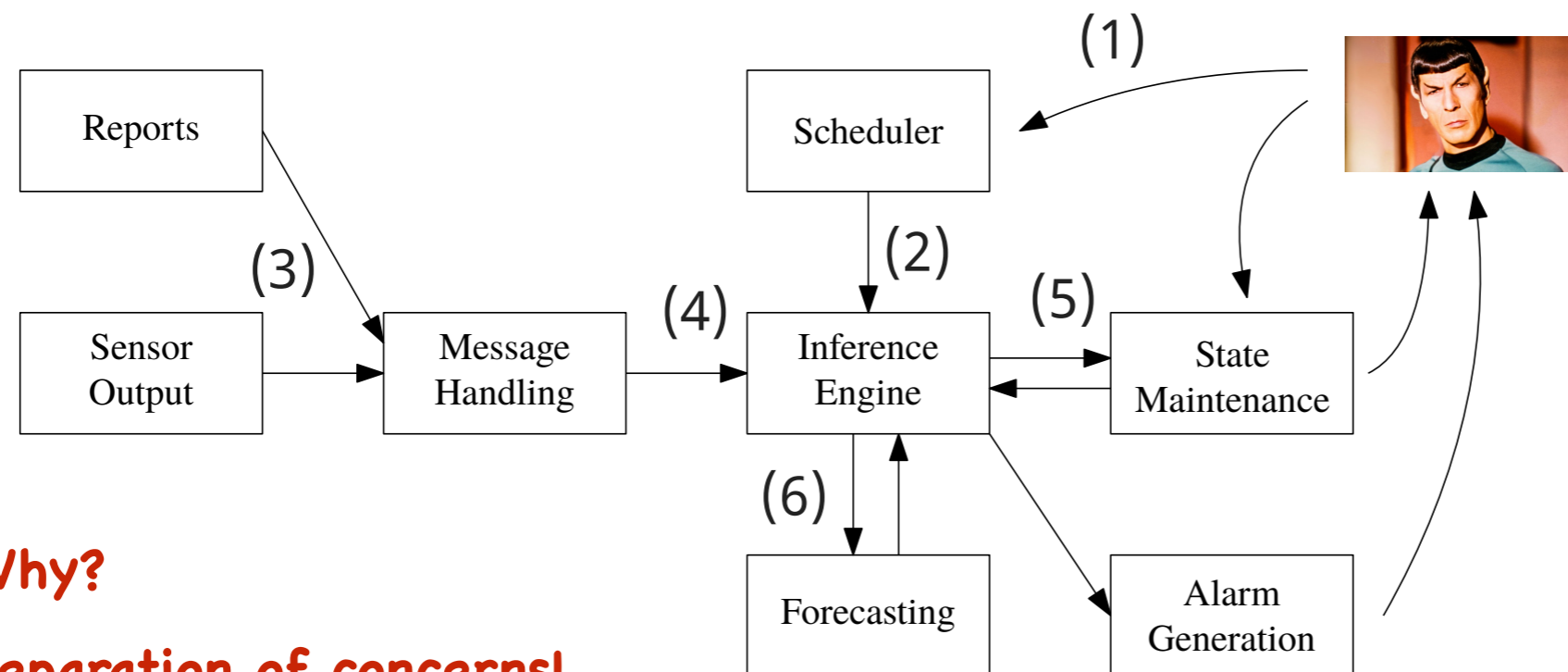
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# System Architecture



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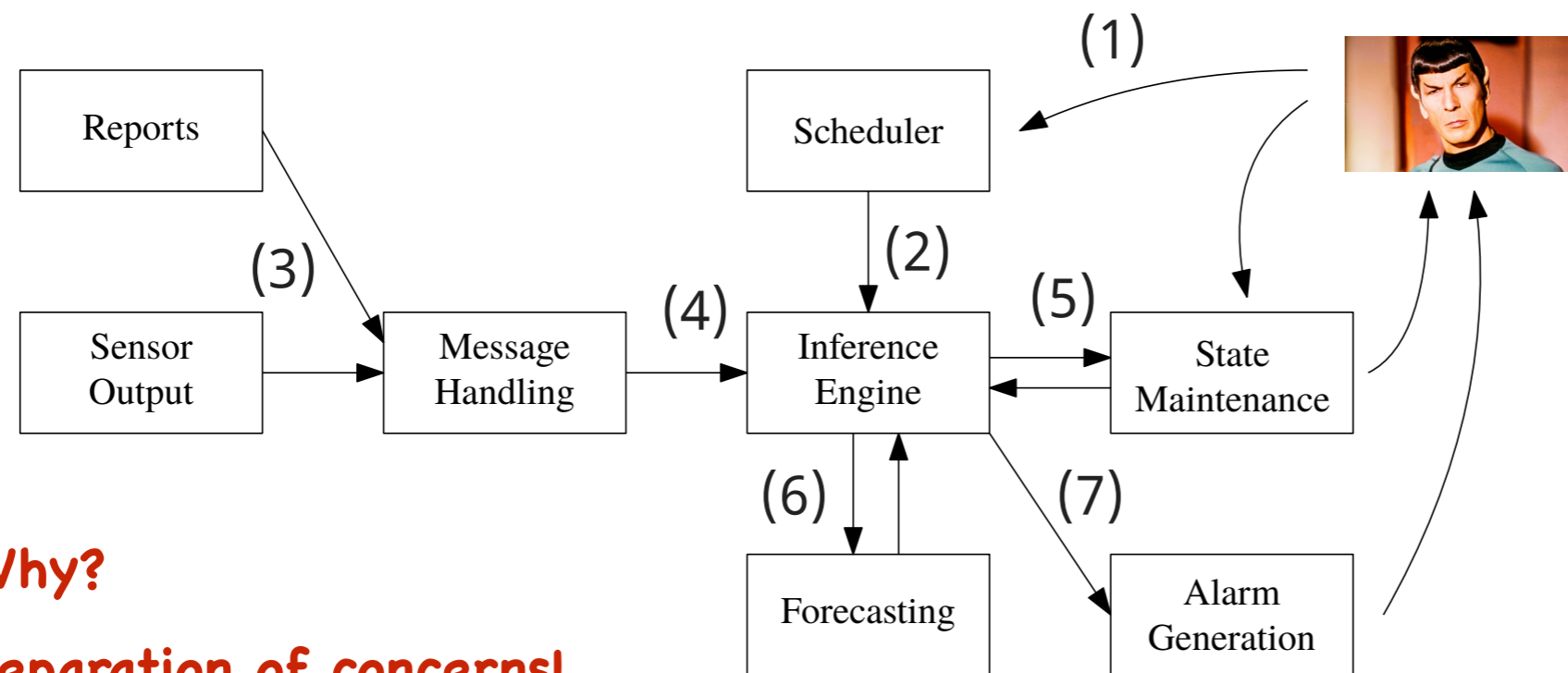
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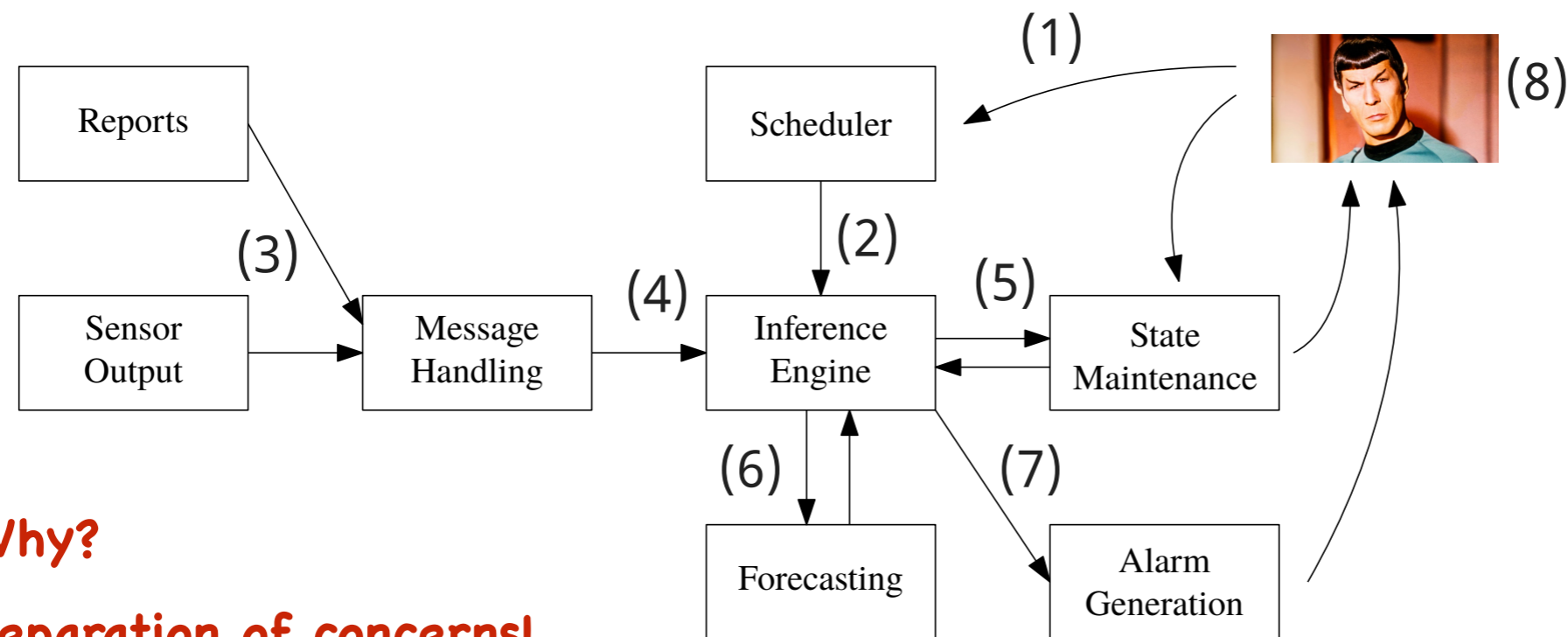
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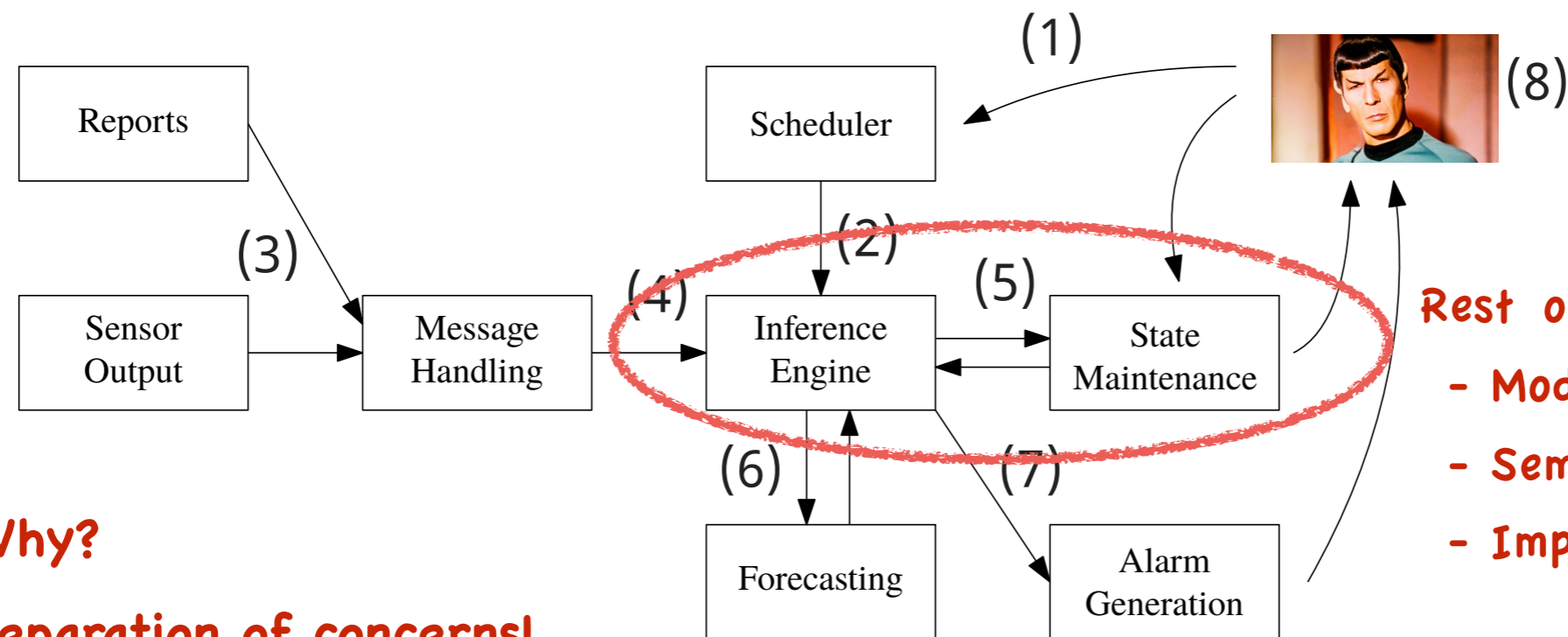
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- (8) Human Operator may reconcile/correct *current state*

# System Architecture



Rest of this talk:

- Modelling Language
- Semantics
- Implementation

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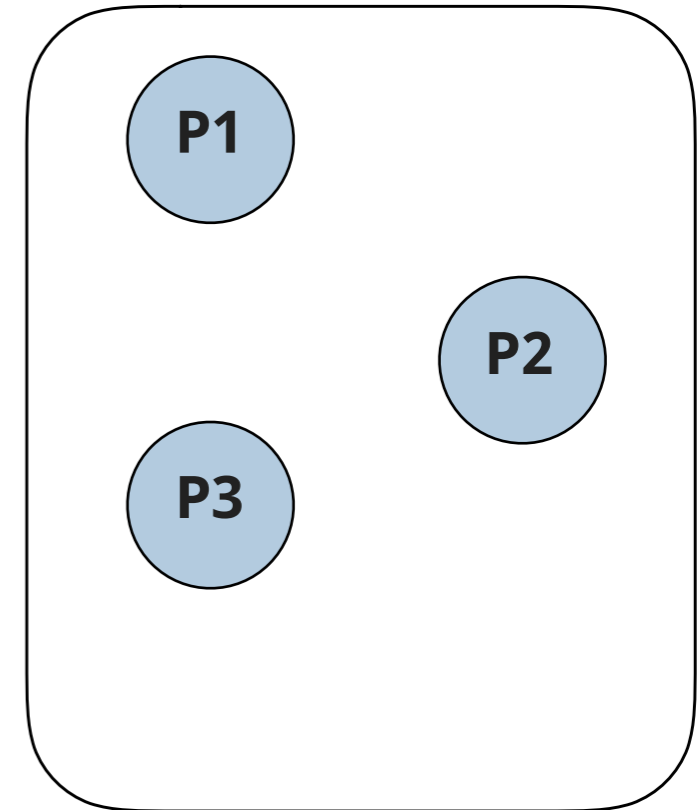
# System Model

Model = processes + channels + state maintenance

## Processes

- Codify logic of an actor of the model
- Process state = set of pairs (variable, value)
- System state = set of its process states
- Run in parallel and act on messages


Truck and  
its load

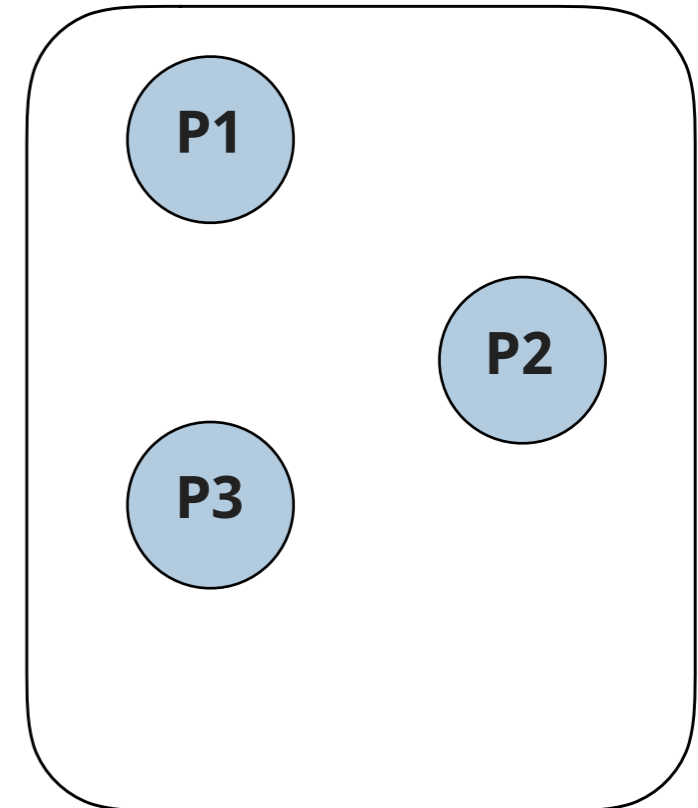


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- Why?  
Supports dynamically adding/removing actors



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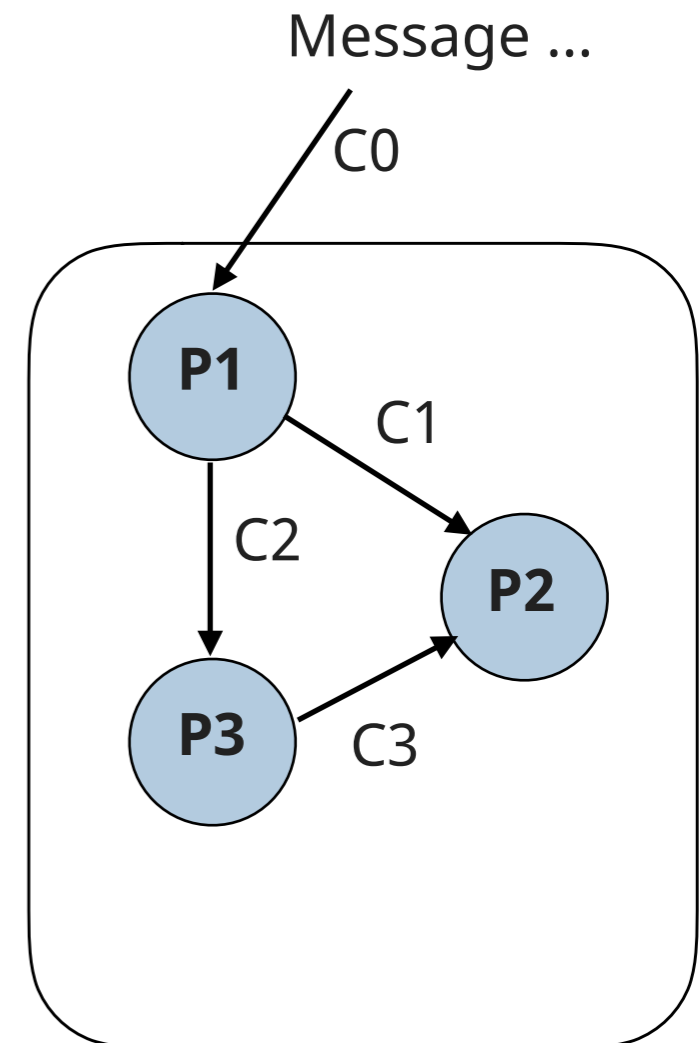
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## Channels and Messages

- Internal: m-to-n inter-process comms
- External: channels for input/output
- Interface to state maintenance module



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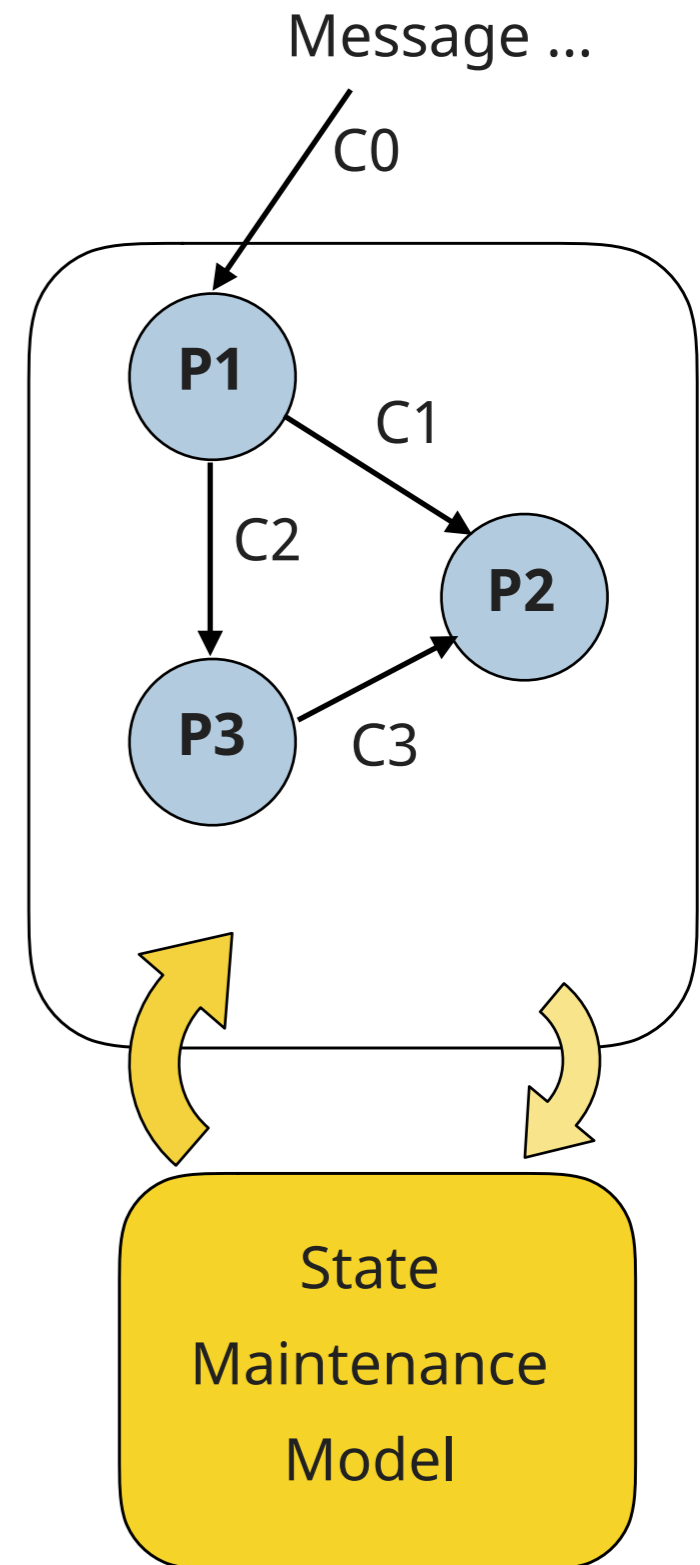
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- Internal: m-to-n inter-process comms
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## State Maintenance Model

- Analysis and amends system states
- “Global view” - has system state history, e.g. RTV based on (probabilistic) temporal logic, Conflict-directed diagnosis

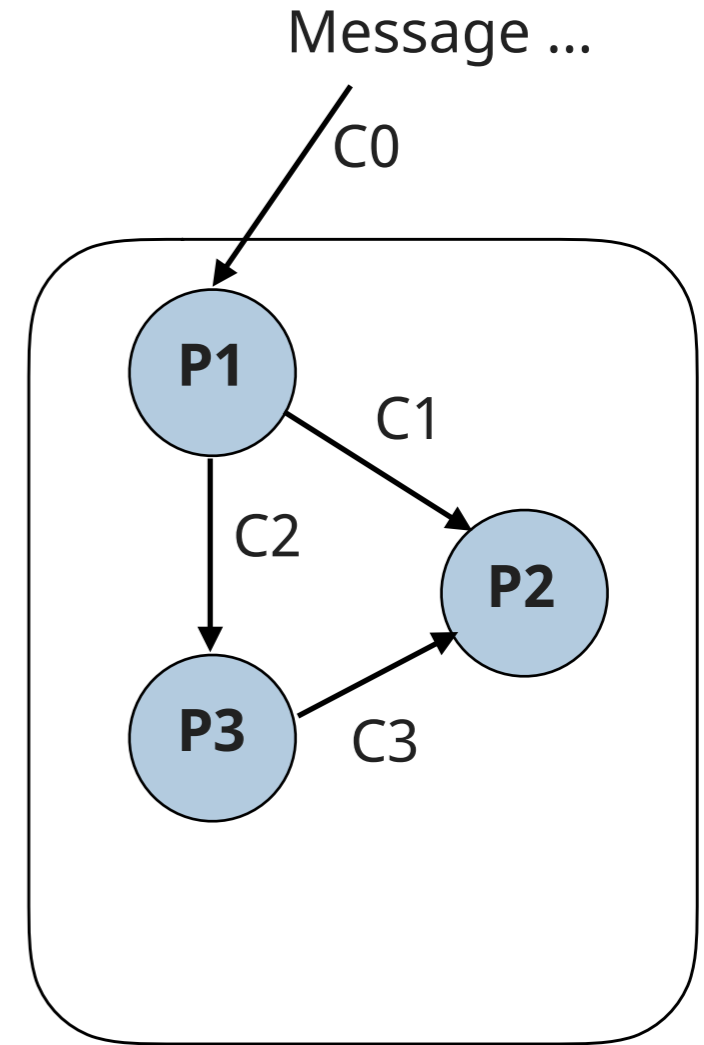


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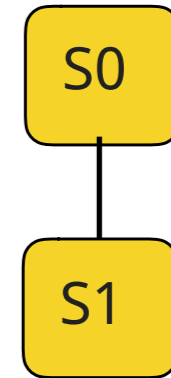
The main loop

Inference Engine

Inference Engine



State Maintenance





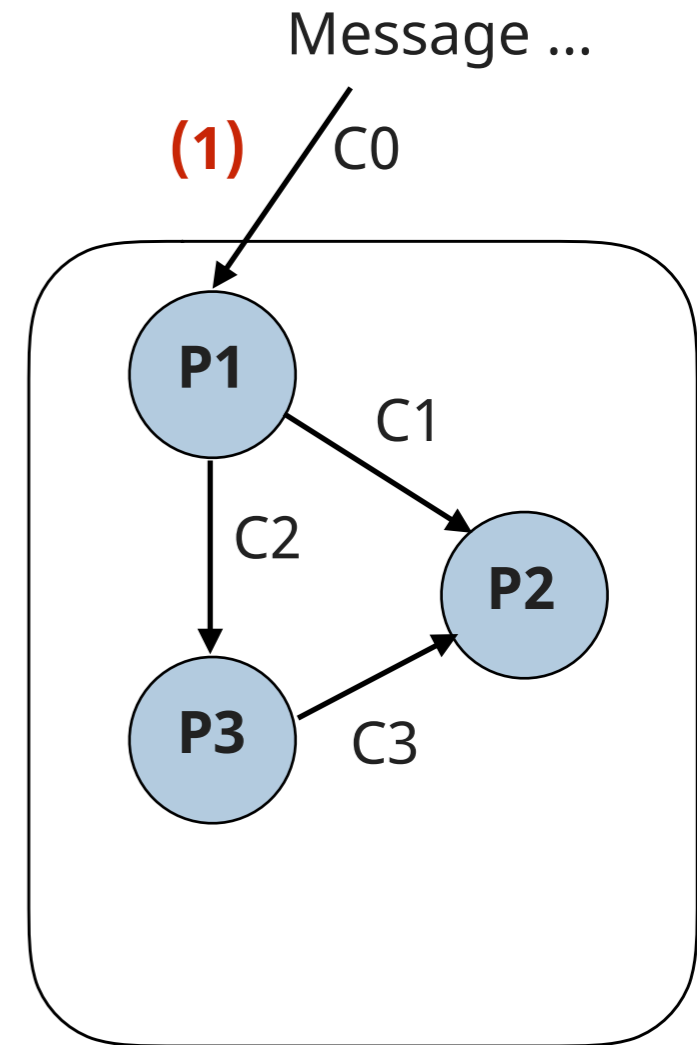
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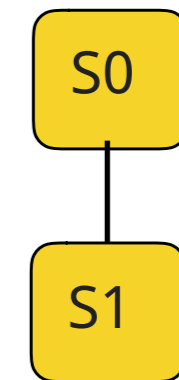
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(1) Next external message comes in

Inference Engine



State Maintenance



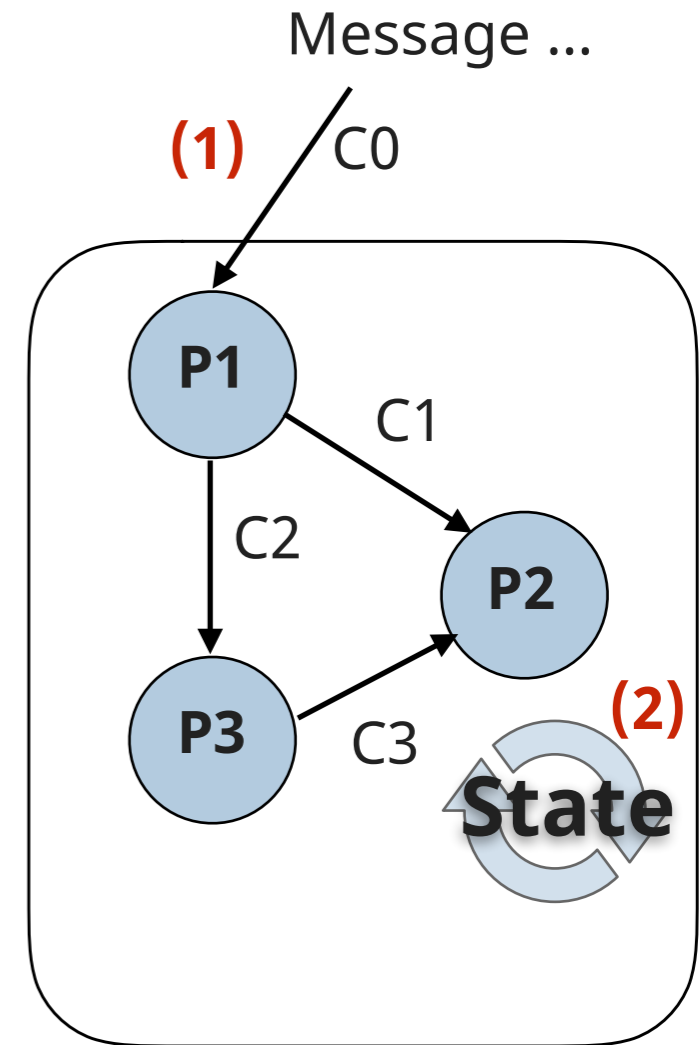
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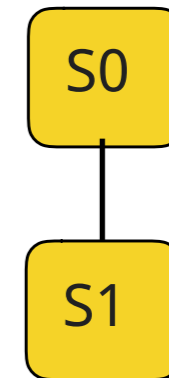
## Inference Engine

- (1) Next external message comes in
- (2) Processes run until exhaustion  
(No more messages sent around)

Inference Engine



State Maintenance



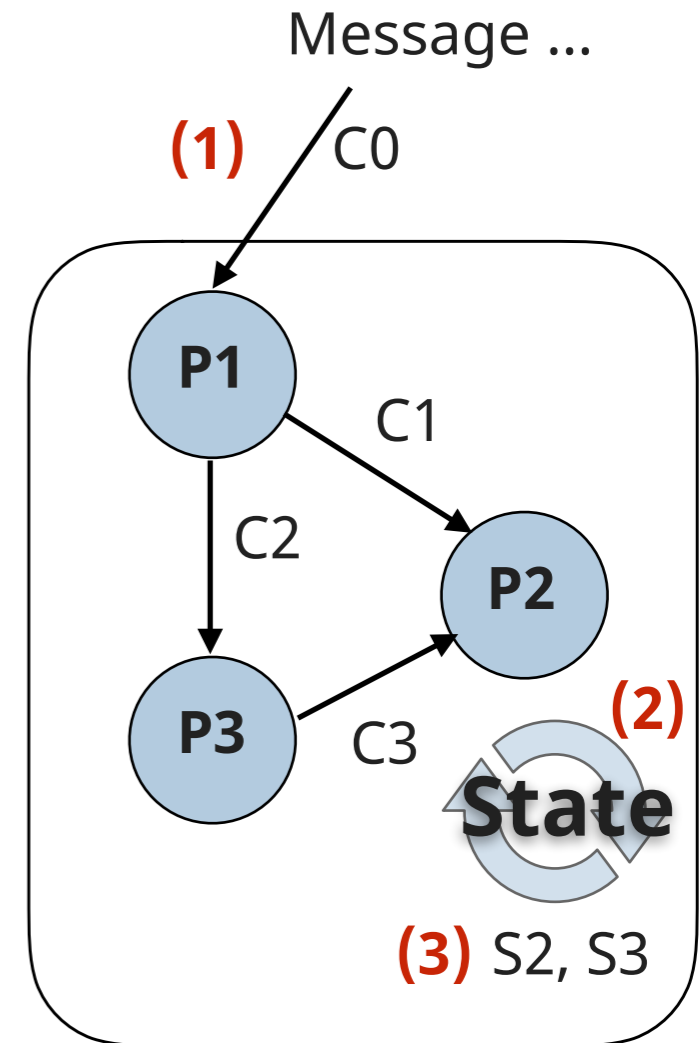
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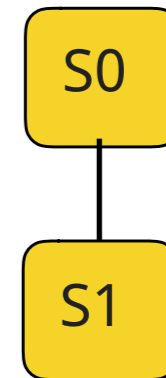
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State Maintenance



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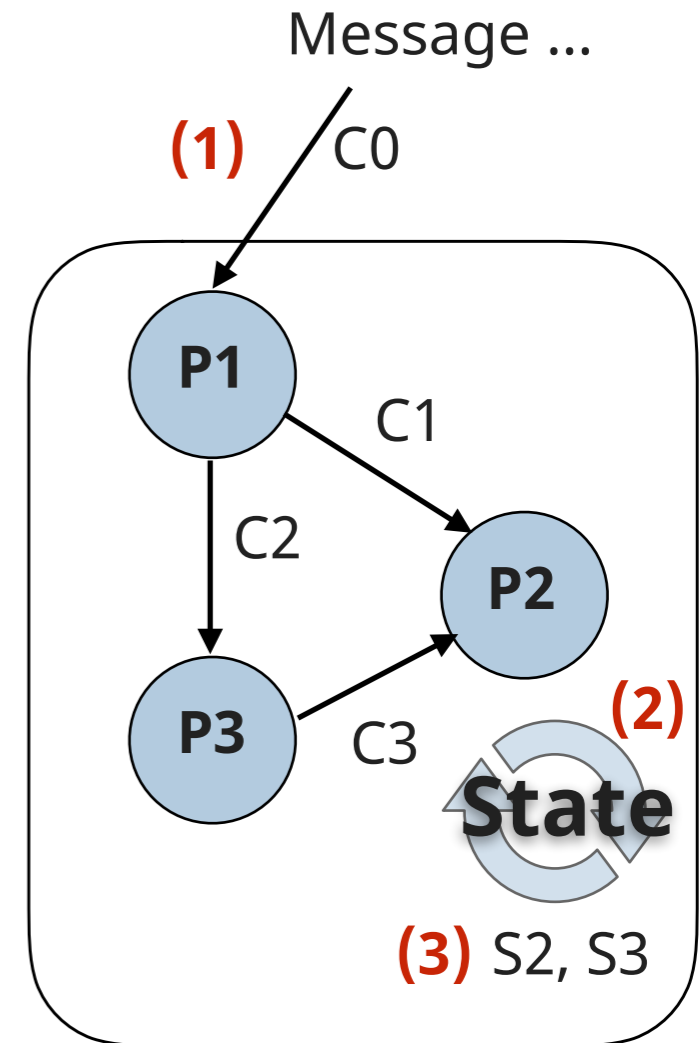
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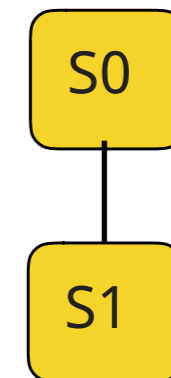
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State Maintenance



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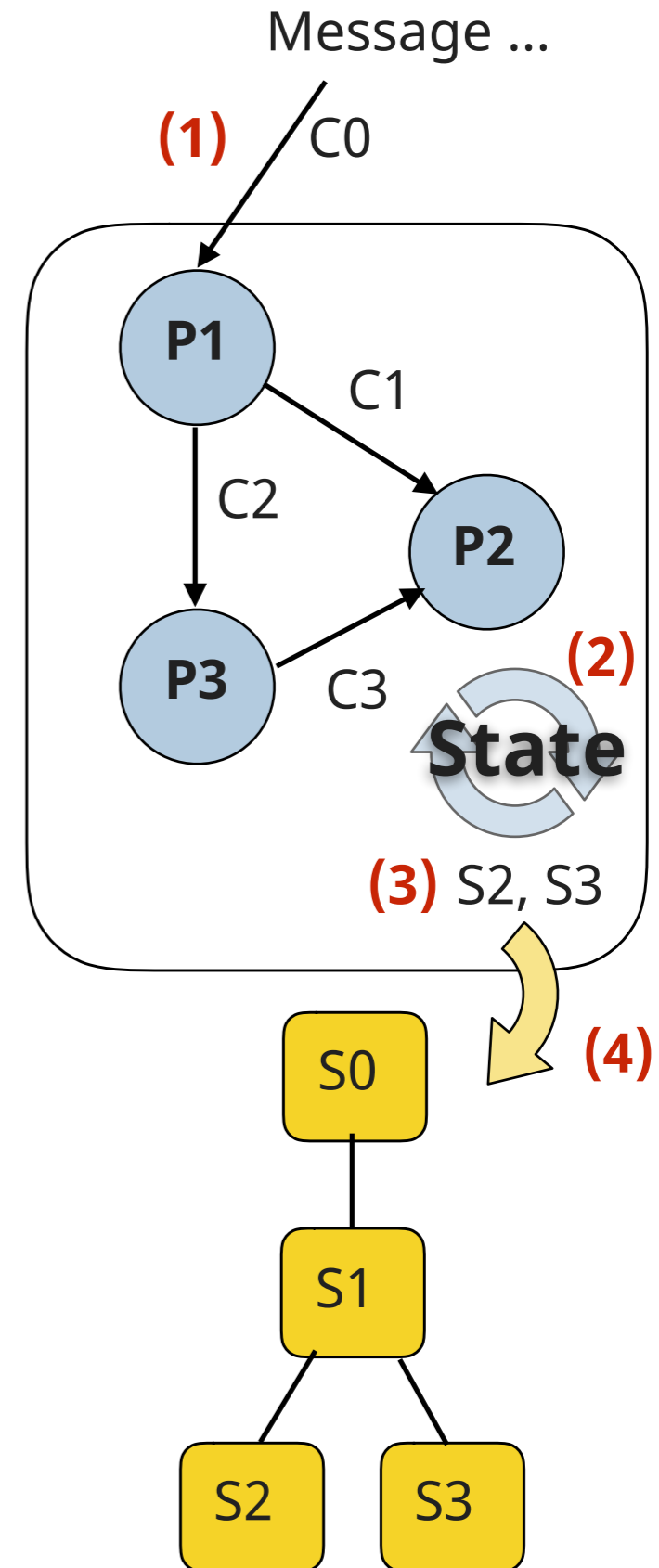
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- (4) Derived possible system states are sent to state maintenance

Inference Engine



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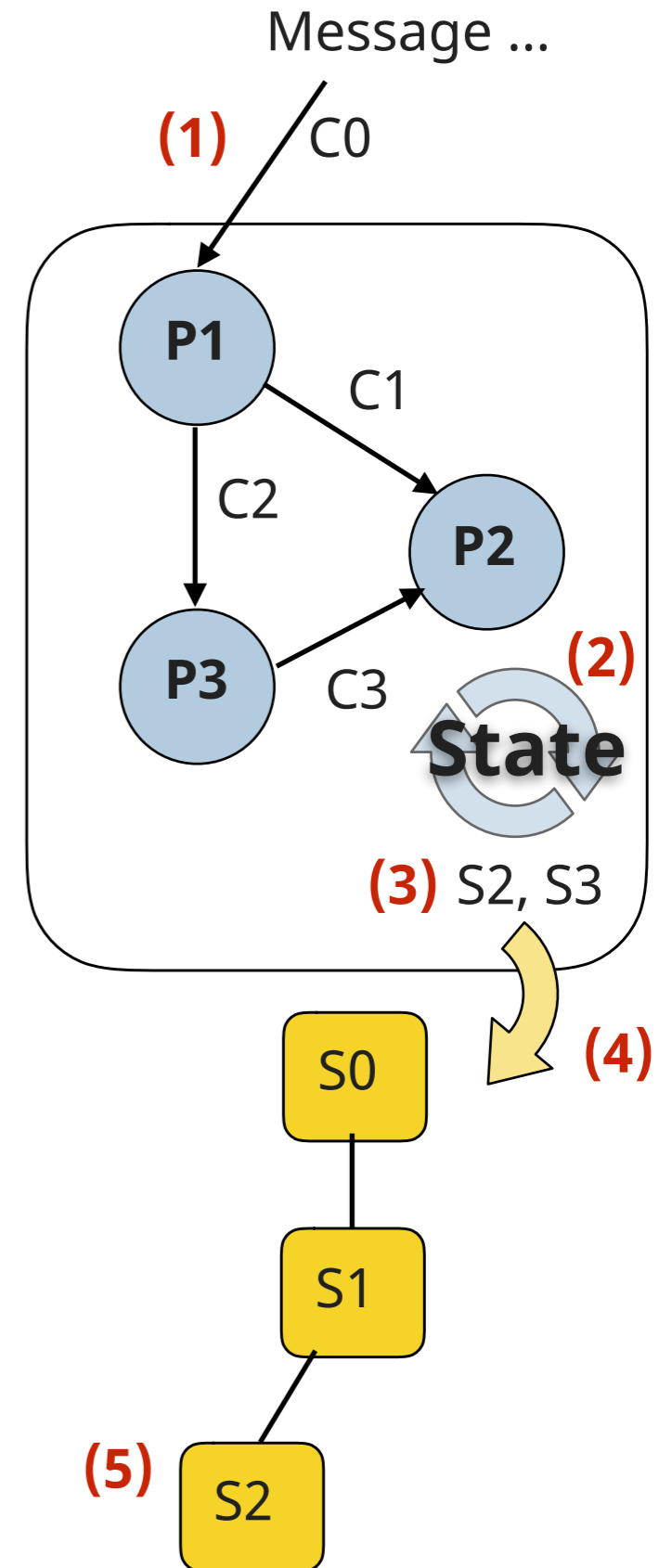
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Inference Engine



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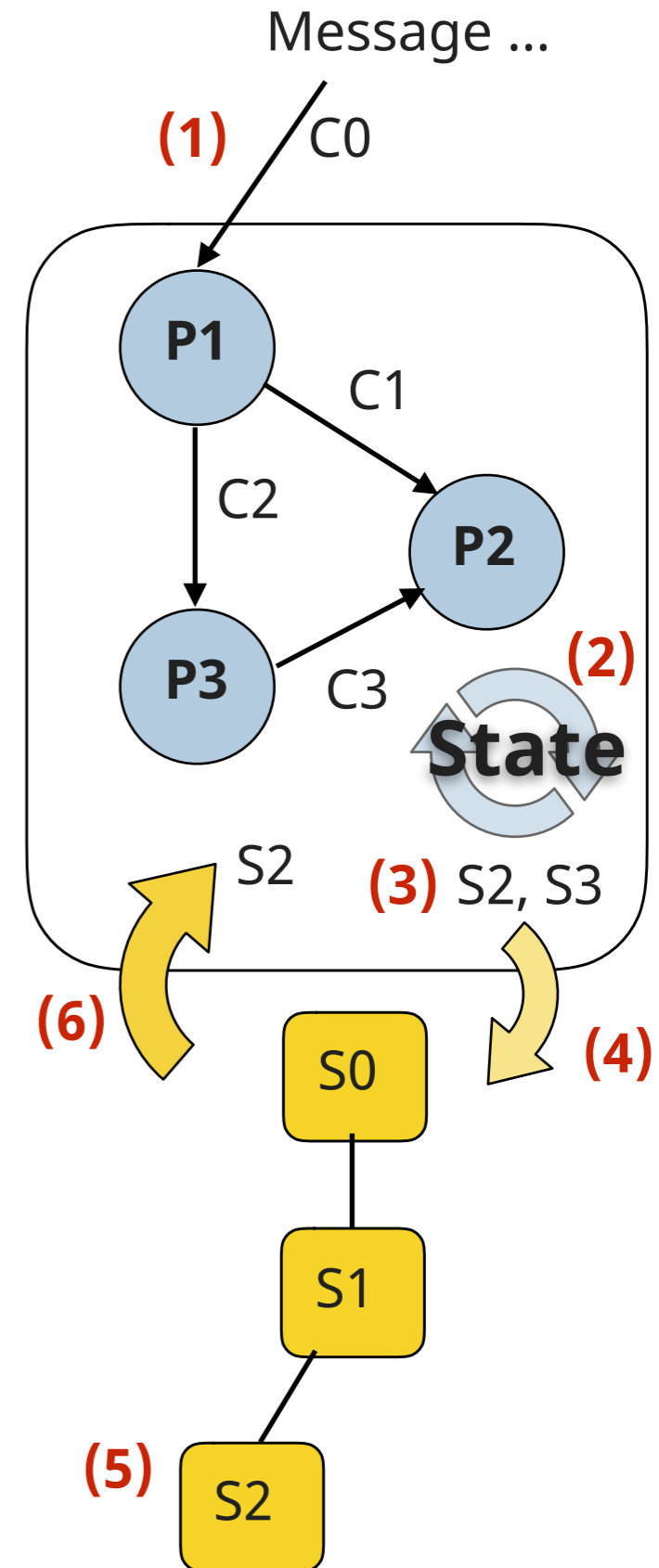
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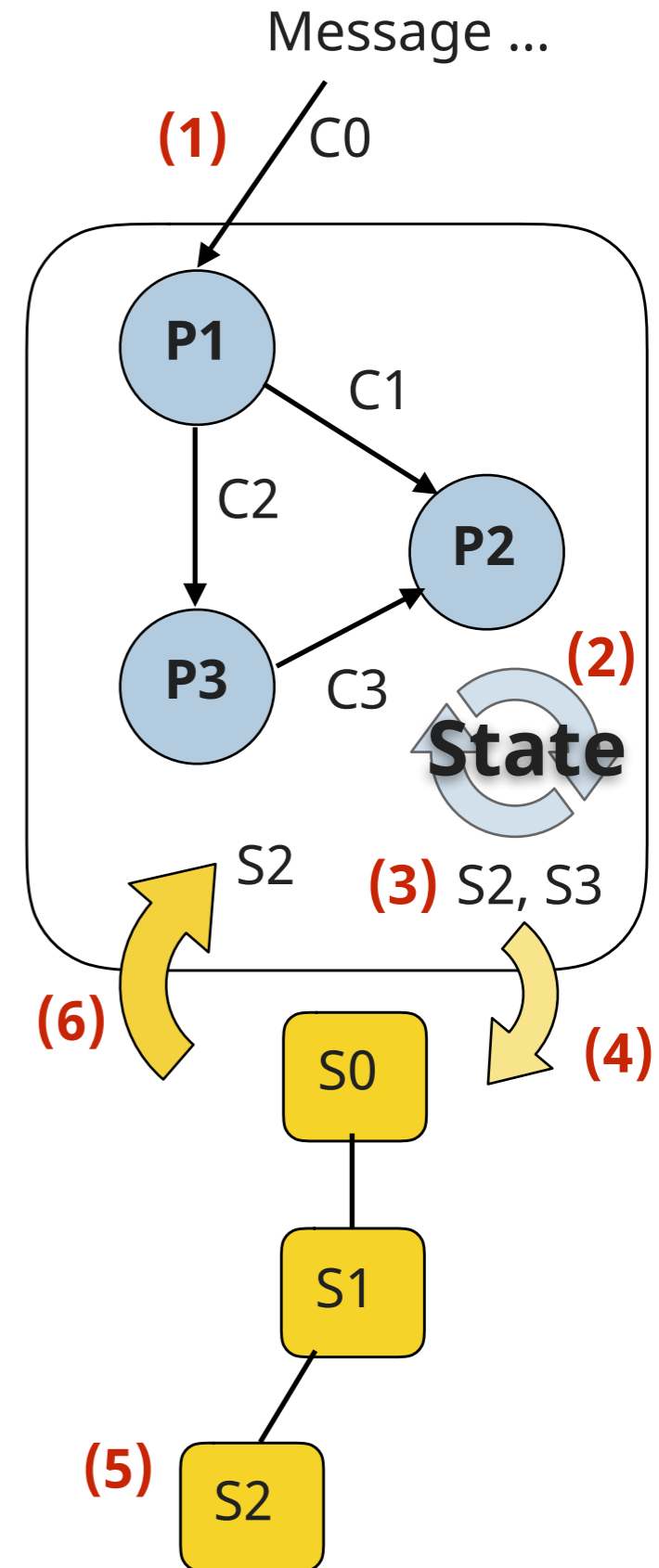
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- (7) Continue with (1)

## Inference Engine





# Example - Food Supply Chain

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## Model

- *Goods* (apples and oranges) of specific *origin* (Riverina and Batlow)
- *Warehouses* (Sydney, Goulburn, Canberra)
- *Trucks* (TruckA, TruckB and TruckC)
- *Waypoints* for trucks in terms of time and location (GPS)
- *Dockets* for loading goods on trucks at warehouses (EPCIS)

Processes

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- What trucks are where at what time?
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Recover from other messages?
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Informed guess?

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Certain trucks are loaded with certain goods that are not from the origin

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- What trucks are where at what time?
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Informed guess?

```
...
{ "time": "2018-02-18T02:00",
  "type": "Waypoint",
  "truck": "TruckB",
  "location": "Goulburn" }
...
{ "time": "2018-02-18T06:45",
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```

houses

# Example - Food Supply Chain

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**Infer Waypoint message  
from Loading message  
and broadcast it**

# Example - Food Supply Chain

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```

Infer Waypoint message  
from Loading message  
and broadcast it

Guess  
origin = "Riverina" OR  
origin = "Batlow"



# Implementation - Shallow Embedding in Scala

## Scala

- *Scala combines object-oriented and functional programming in one concise, high-level language* ([www.scala-lang.org](http://www.scala-lang.org))
- Runs on JVM
- Static type system, type inference, pattern matching, call-by-name/call-by-value, libraries
- Syntactic sugar  
`unless (x == 0) { println("One over x is " + 1/x) }`

## Shallow Embedding

- Modelling language =  
Scala + syntactic sugar + class definitions for "Process" and "Channel"
- Scheduler is library function
- Why?
  - Full power of host language
  - Existing libraries for DB connectivity, RabbitMQ, JSon, ...
  - Easy to implement

# Food Supply Chain Example

I\_am\_a\_modelling\_language\_expression

## Processes

```
object Dispatch extends Process("Dispatch") { ... }  
class Truck(Id: String) extends Process("Truck") { ... }  
class Warehouse(Location: String) extends Process("Warehouse") { ... }
```

## Message Data Structures

```
abstract class Message  
case class Waypoint(time: DateTime, truck: String, location: String) extends Message  
case class Loading(time: DateTime, truck: String, location: String,  
                  goods: String, origin: String) extends Message
```

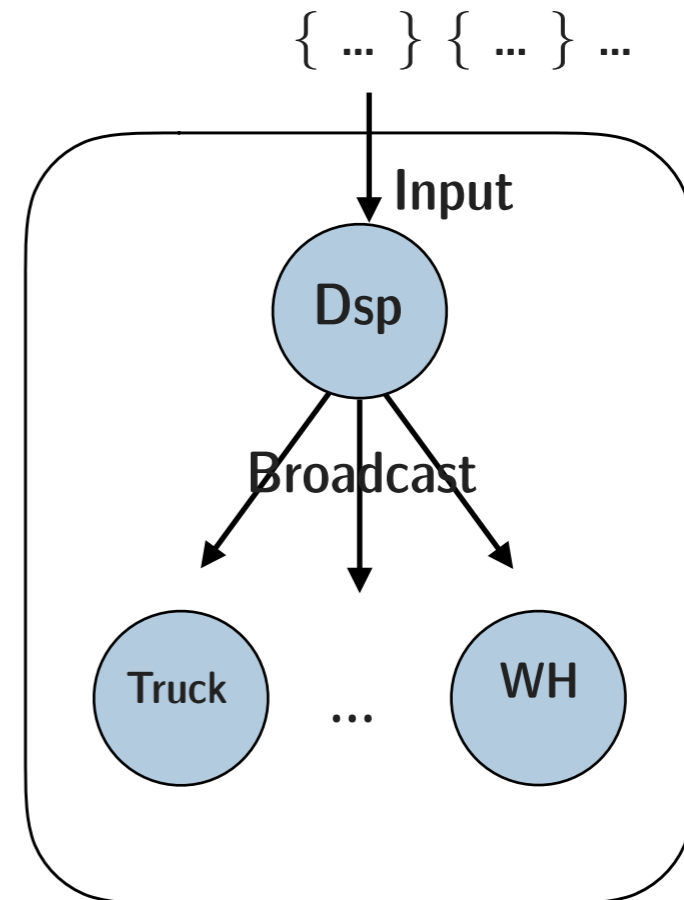
## Channels

```
object Input extends Channel[JsonObject]("Input", withInputPort = 5554, window = 1)  
// For receiving messages from the external world  
object Broadcast extends Channel[Message]("Broadcast")  
// For inter-process communication
```

# Food Supply Chain Example

## Dispatch Process

```
object Dispatch extends Process("Dispatcher") {  
  import collection.mutable.Set  
  val trucks = Set.empty[String]  
  rules (  
    Input --> { msg =>  
      (msg \ "type").as[String] match {  
        case "Waypoint" =>  
          val wp = msg.toWaypoint  
          if (! (trucks contains wp.truck)) {  
            Scheduler.schedule(new Truck(wp.truck))  
            trucks += wp.truck  
          }  
          Broadcast <- wp  
        case "Loading" => ... //similar  
        case _ => error(s"Dispatch: cannot handle message $msg")  
      }  
    }  
  )  
}
```



### Dispatcher process:

- Receive external messages
- Broadcasts messages to all processes
- Creates new Truck processes on the fly

# Food Supply Chain Example

## Truck Process

```
class Truck(id: Int) extends Process("Truck") {  
  
    val broadcast = Broadcast.subscribe()  
  
    var location = "unknown"  
    var load = Set.empty[(String, String)]  
  
    // Externally visible state variables  
    stateVar("id", ..., ...)  
    stateVar("location", ..., ...)  
    stateVar("load", ..., ...)  
  
    rules(  
        ... // next slide  
    )  
}
```

```
{ id = "TruckA",  
  location = "Sydney"  
  load = [ ("Oranges", "Batlow") ]  
}
```

⇒ Automatic mapping to/from Json

# Food Supply Chain Example

## Truck process (cont'd)

```
var location = "unknown"
var load = Set.empty[(String, String)]

// This rule infers a missing Waypoint message from a Loading message
rules (
  broadcast --> {
    case msg @ Loading(time, id, loc, goods, origin) if location != loc =>
      Broadcast <-- Waypoint(time, id, loc) // Infer Waypoint
      Broadcast <-- msg
    ... // (cont'd)
  }
}
```

# Food Supply Chain Example

## Truck process (cont'd)

```
var location = "unknown"
```

```
var load = Set.empty[(String, String)]
```

```
// This rule updates the load on this truck
```

```
rules (
```

```
  broadcast --> {
```

```
    case msg @ Loading(time, Id, loc, goods, origin) if origin != "unknown" =>
```

```
      load += (goods, origin)
```

```
// Disjunctive rule to resolve "unknown"
```

```
case msg @ Loading(time, Id, loc, goods, origin) if origin == "unknown" =>
```

```
  or( { Broadcast <-- Loading(time, Id, loc, goods, "Riverina") },
```

```
       { Broadcast <-- Loading(time, Id, loc, goods, "Batlow") } )
```

```
}
```

```
)
```

# Food Supply Chain Example

## Truck process (cont'd)

```
var location = "unknown"
```

```
var load = Set.empty[(String, String)]
```

```
// This rule updates the load on this truck
```

```
rules (
```

```
  broadcast --> {
```

```
    case msg @ Loading(time, Id, loc, goods, origin) if origin != "unknown" =>
```

```
      load += (goods, origin)
```



Rules are allowed to "fail"

```
// Disjunctive rule to resolve "unknown"
```

```
case msg @ Loading(time, Id, loc, goods, origin) if origin == "unknown" =>
```

```
  or( { Broadcast <-- Loading(time, Id, loc, goods, "Riverina") },
```

```
       { Broadcast <-- Loading(time, Id, loc, goods, "Batlow") } )
```

```
}
```

```
)
```

# Food Supply Chain Example

## Truck process (cont'd)

```
var location = "unknown"
```

```
var load = Set.empty[(String, String)]
```

```
// This rule updates the load on this truck
```

```
rules (
```

```
  broadcast --> {
```

```
    case msg @ Loading(time, Id, loc, goods, origin) if origin != "unknown" =>
```

```
      load += (goods, origin)
```



Rules are allowed to "fail"

```
// Disjunctive rule to resolve "unknown"
```

```
case msg @ Loading(time, Id, loc, goods, origin) if origin == "unknown" =>
```

```
  or( { Broadcast <-- Loading(time, Id, loc, goods, "Riverina") },
```

```
       { Broadcast <-- Loading(time, Id, loc, goods, "Batlow") } )
```

```
  }
```

```
)
```

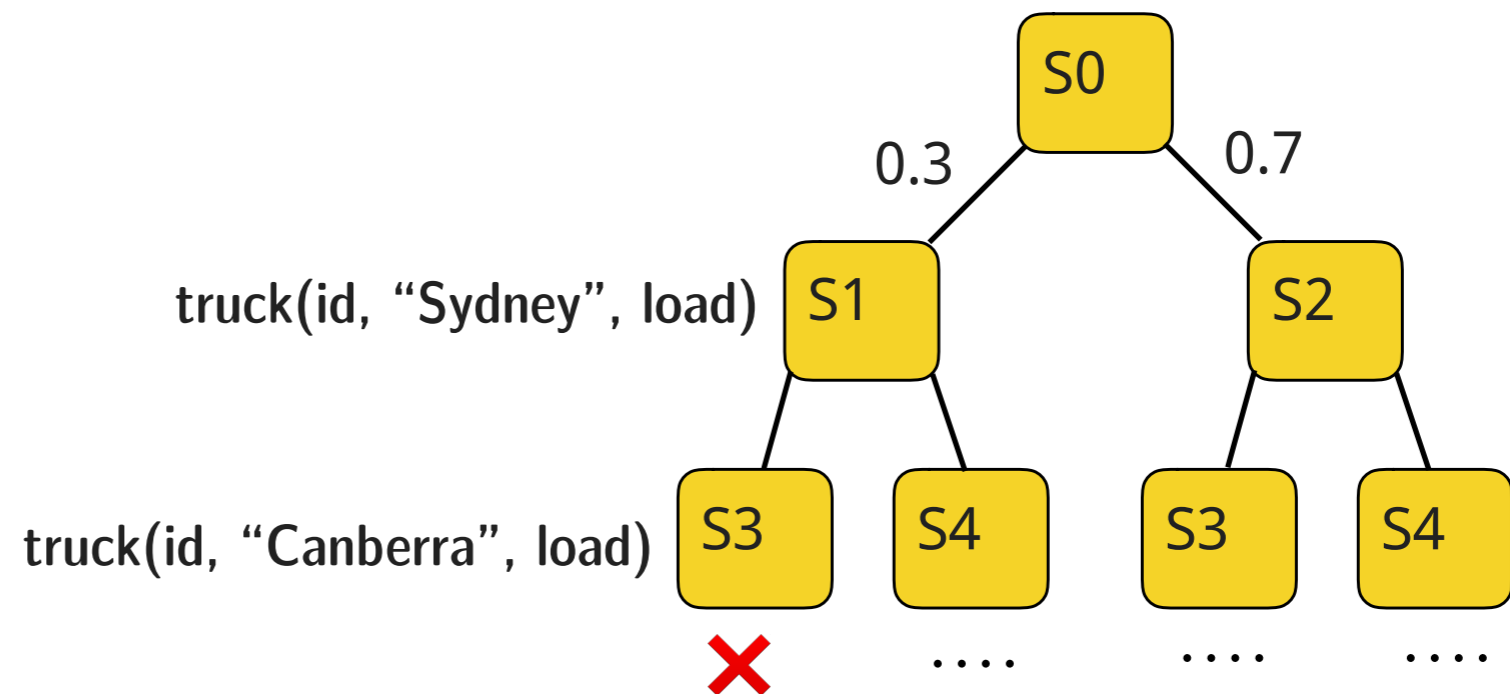
## What else?

- State Maintenance:  
"no oranges from Batlow"
- Inference engine algorithm  
and implementation
- Scheduler algorithm  
and implementation



## Next Steps

- Improve and complete implementation
- Realistic food supply chain based on EPCIS events
- Probabilistic state transitions
- State maintenance *trees*
- State maintenance declaratively specified by temporal logic constraints
  - $\mathbf{G} (\text{truck}(\text{id}, \text{"Sydney"}, \text{load}) \rightarrow \neg \mathbf{F} (\text{truck}(\text{id}, \text{"Canberra"}, \text{load})))$



## Next Steps

- Improve and complete implementation
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