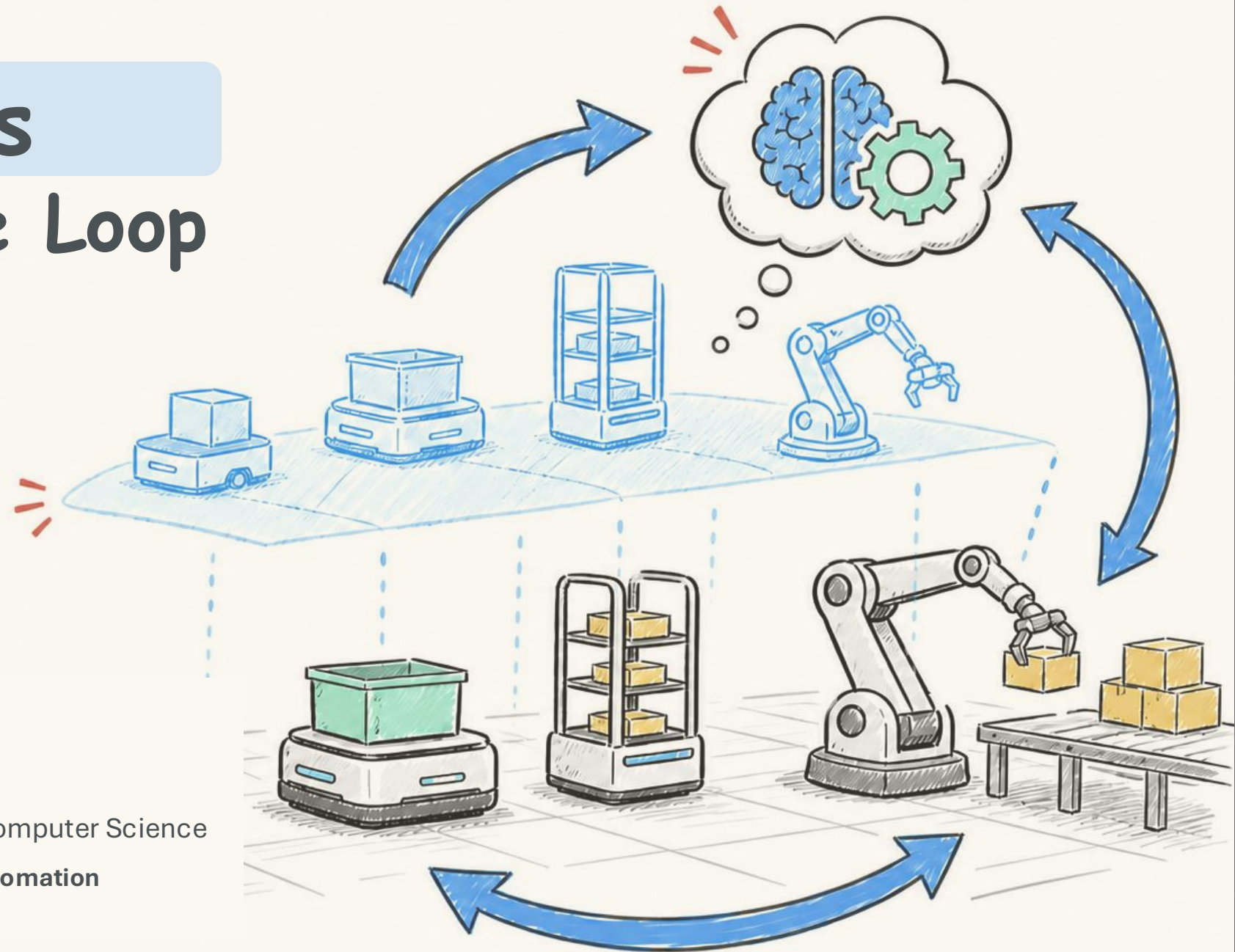


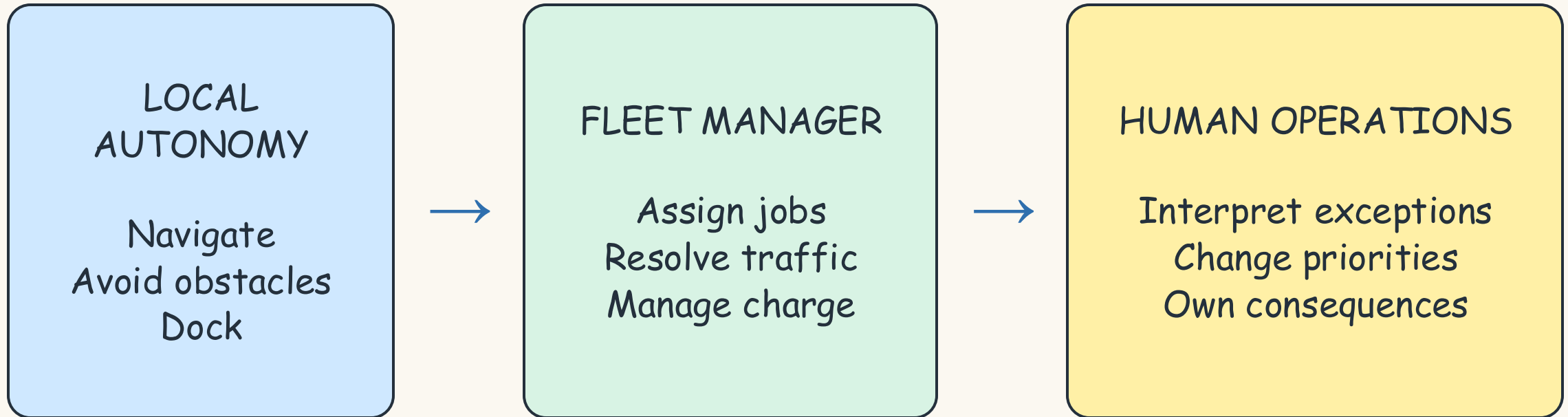
Digital Twins That Close the Loop Agentic AI for Robot Fleets

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Workshop: Agentic AI and Digital Twin for Automation
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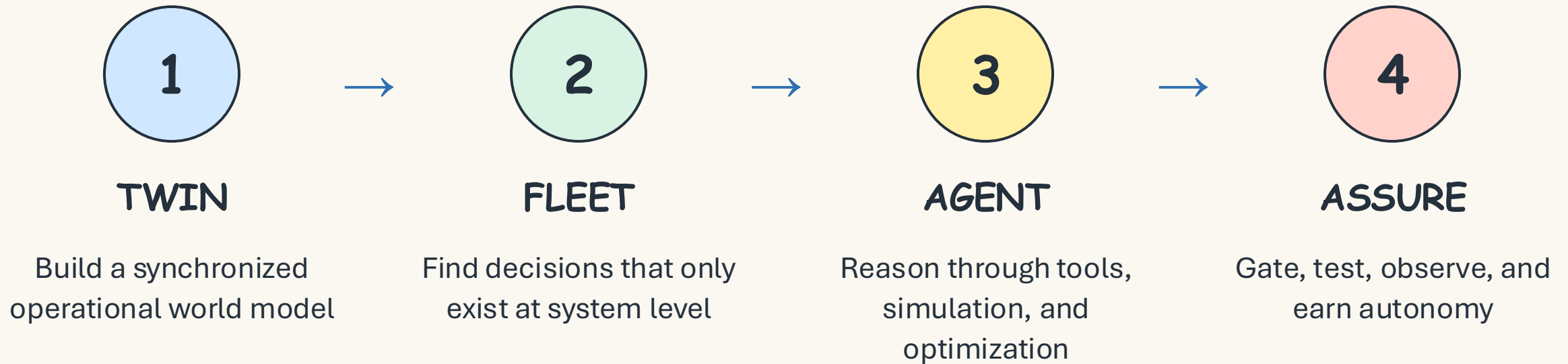


Robot fleets already act. Their intelligence is fragmented.



The missing layer is shared operational reasoning across the whole fleet.

Four steps from mirror to bounded action

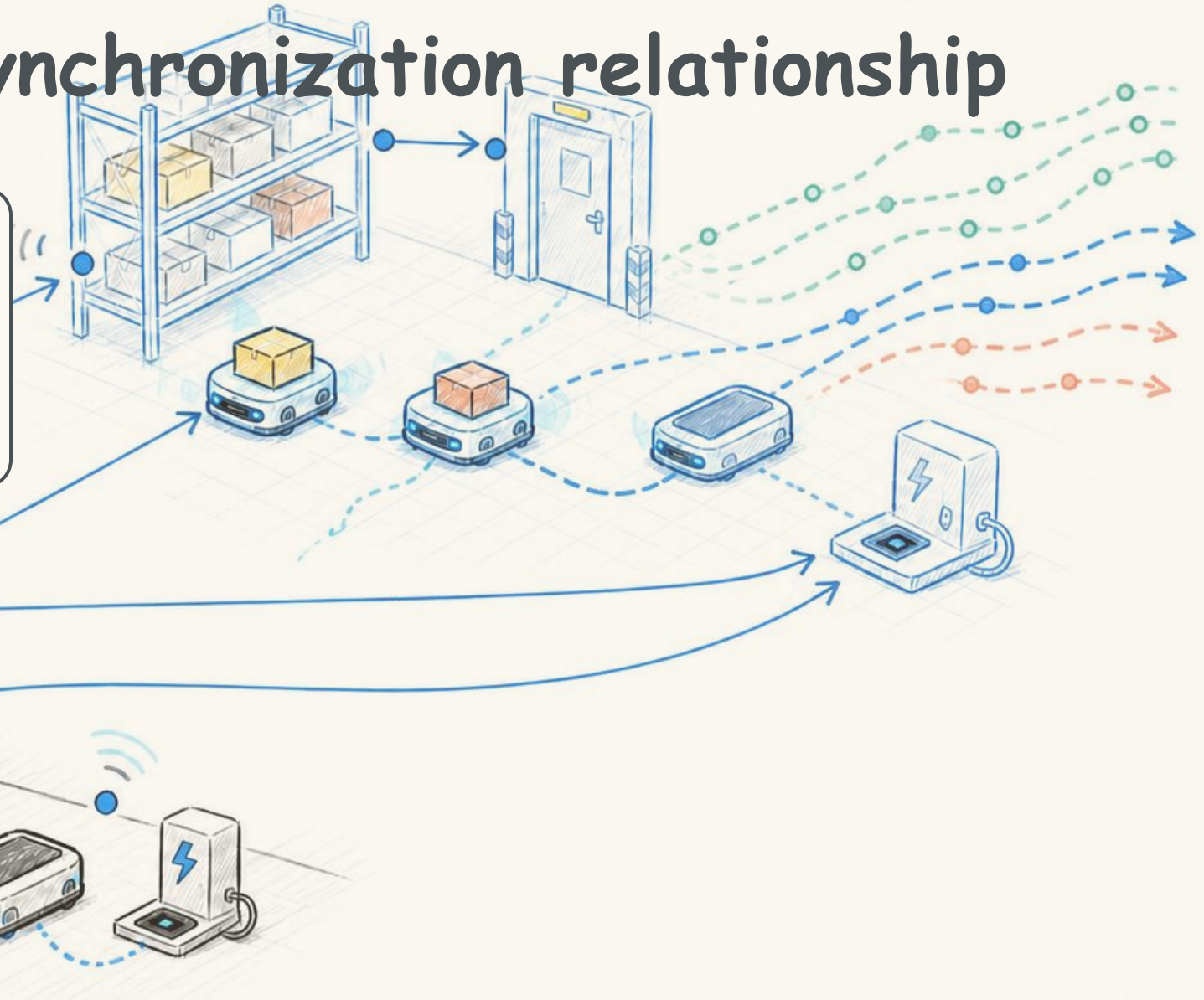


The goal is not an LLM that drives robots. It is an accountable system that closes a useful loop.

A digital twin is a synchronization relationship

An integrated, data-driven virtual representation of real entities and processes, with synchronized interaction at a specified frequency and fidelity.

A twin is defined by the decision it must support.



Model, shadow, or twin?

MODEL

Offline representation

No live state

No operational action

DIGITAL SHADOW

Physical $\rightarrow$ virtual

Live observation

No direct intervention

DIGITAL TWIN

Physical $\leftrightarrow$ virtual

Observation and
intervention

Explicit authority



If the virtual side cannot be trusted to observe or influence operations, the agent has no stable world to reason about.

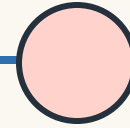
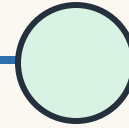
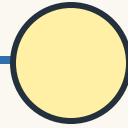
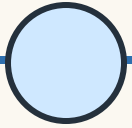
Every twin makes a synchronization contract

STATE

FREQUENCY

FIDELITY

AUTHORITY



What is represented?
Pose, battery, task,
health, map, queue

How fresh must it be?
Milliseconds, events,
minutes, shifts

How accurate is
enough?
Routing, energy, wear,
human flow

What may flow back?
Advice, dispatch,
parameters, stop

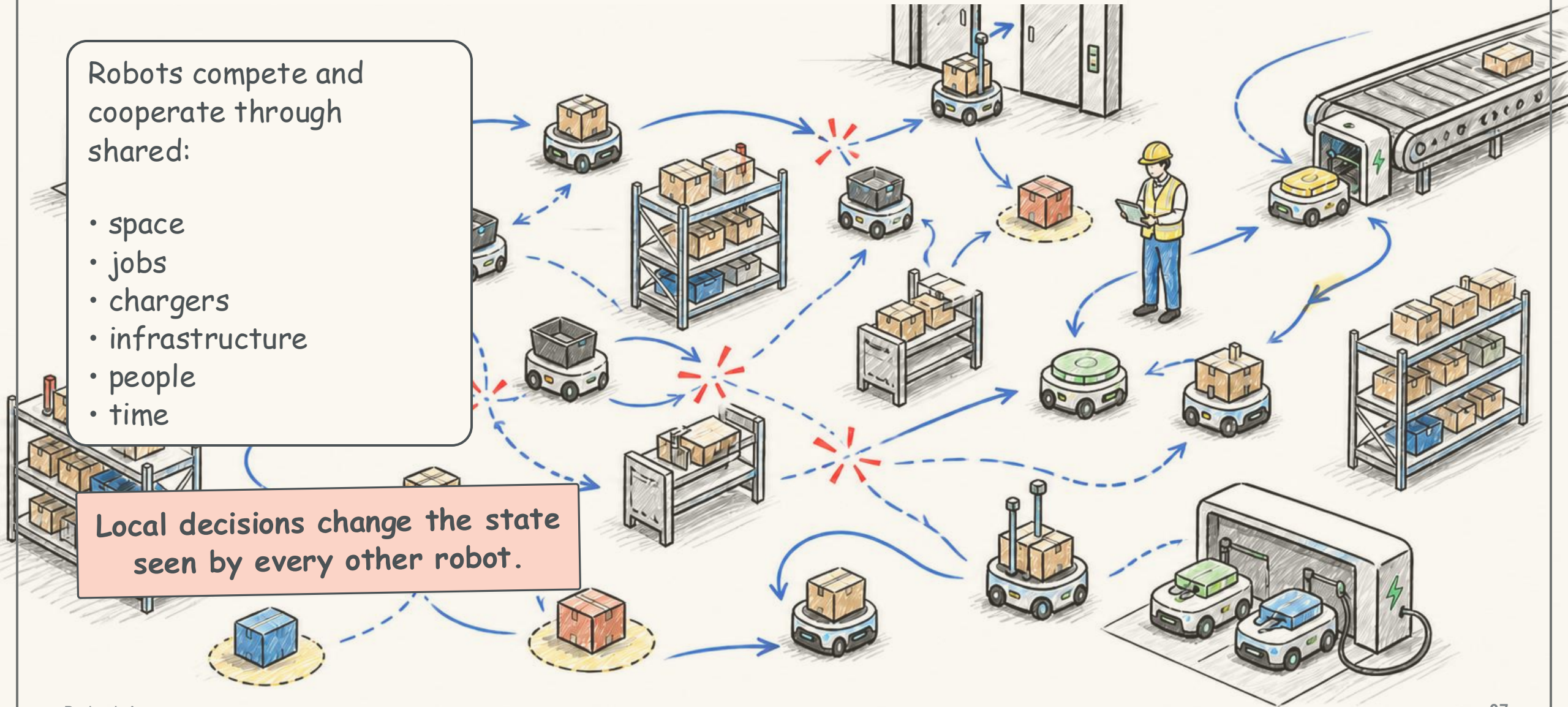
Staleness is not only a data problem. In a closed loop, it becomes a control risk.

A fleet is a system of systems

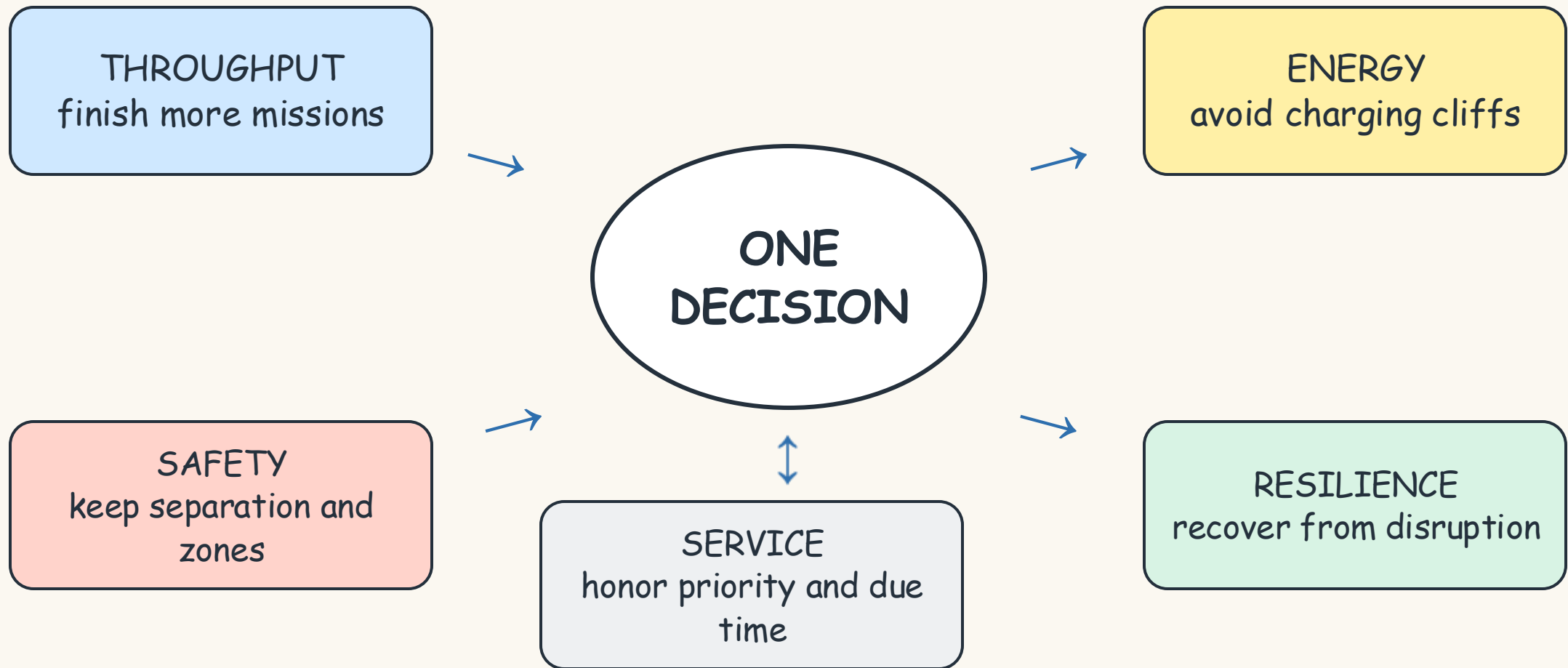
Robots compete and cooperate through shared:

- space
- jobs
- chargers
- infrastructure
- people
- time

Local decisions change the state seen by every other robot.

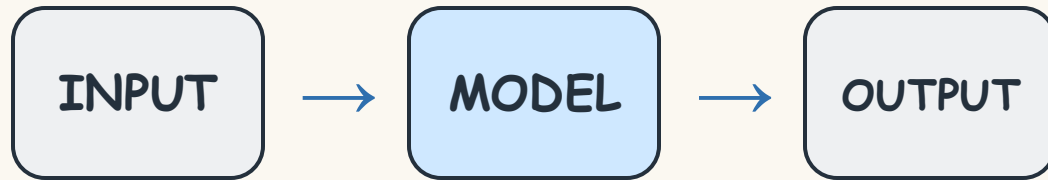


Fleet objectives are coupled



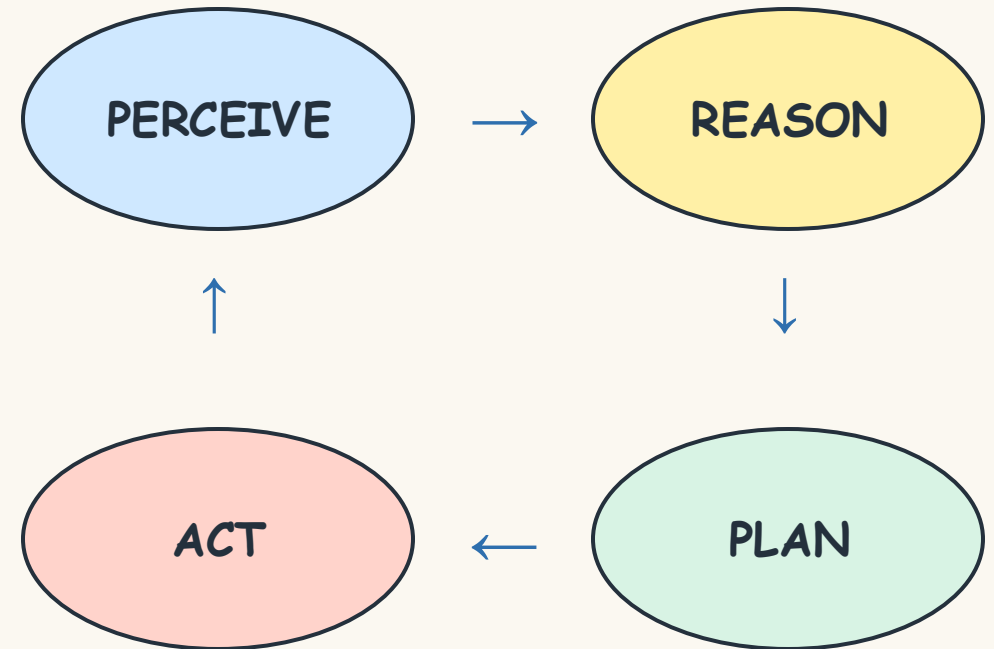
Agentic AI adds goal-directed tool use

TRADITIONAL PIPELINE



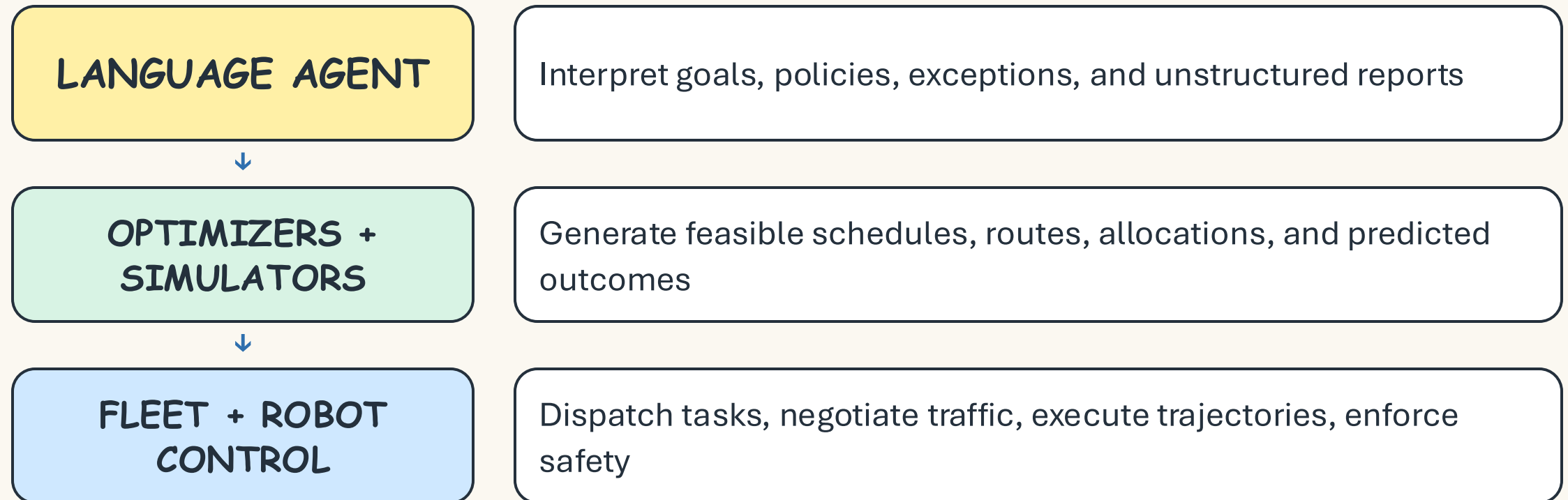
one pass

AGENTIC LOOP



Agency appears at the system level: memory + tools + feedback + policy.

Keep the language model at the semantic layer



The LLM should not close the millisecond control loop.

Twin = operational world model

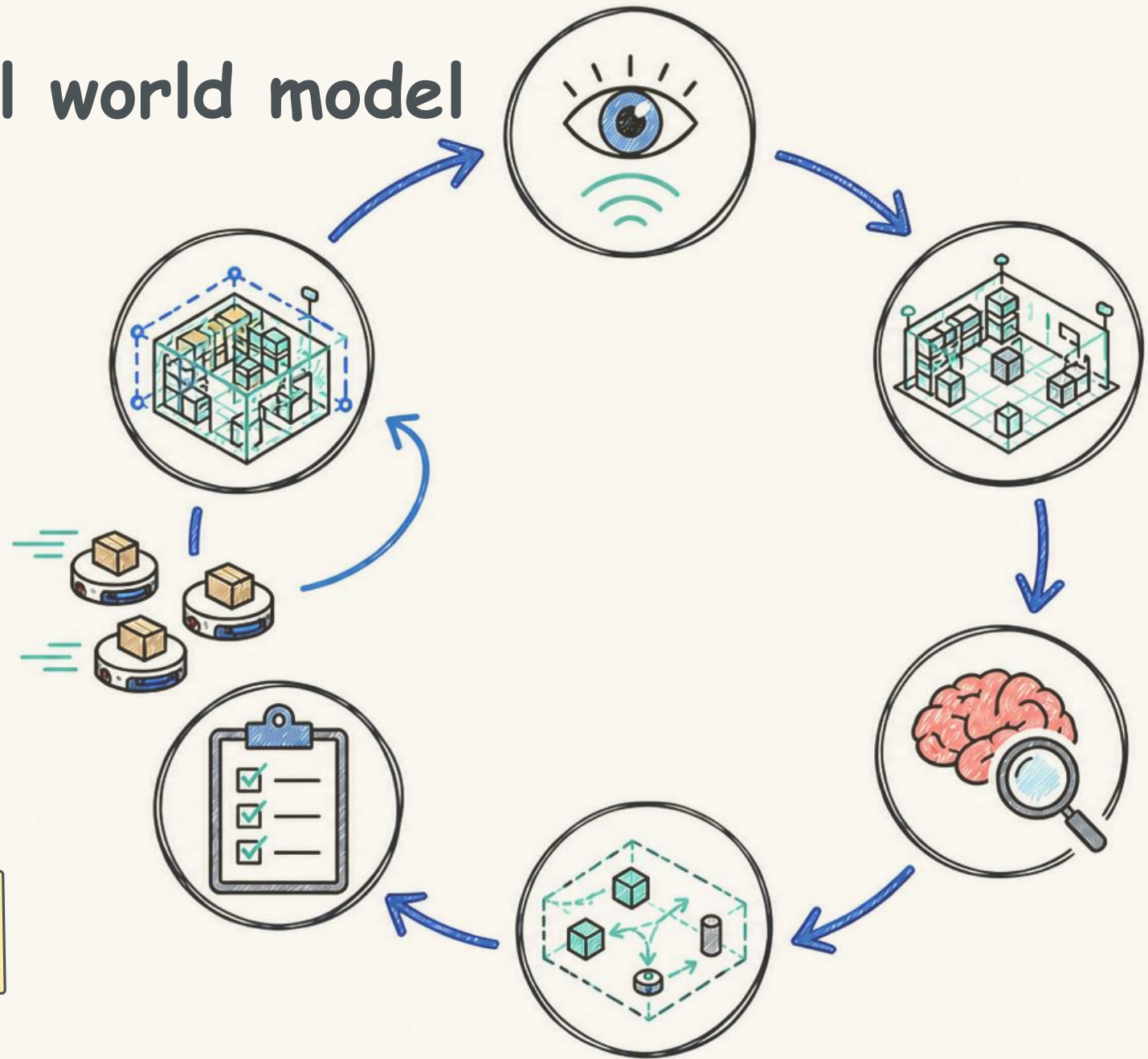
OBSERVE
Twin state

DIAGNOSE
Evidence + context

SIMULATE
Candidate futures

ACT + LEARN
Bounded intervention

The twin is both memory and rehearsal space.



Reference architecture: agent + twin + fleet



Give the agent tools, not raw authority

READ

query_fleet_state

get_incident_context

retrieve_policy

ANALYZE

simulate_candidate

solve_allocation

check_constraints

ACT

propose_dispatch

request_zone_change

start_canary

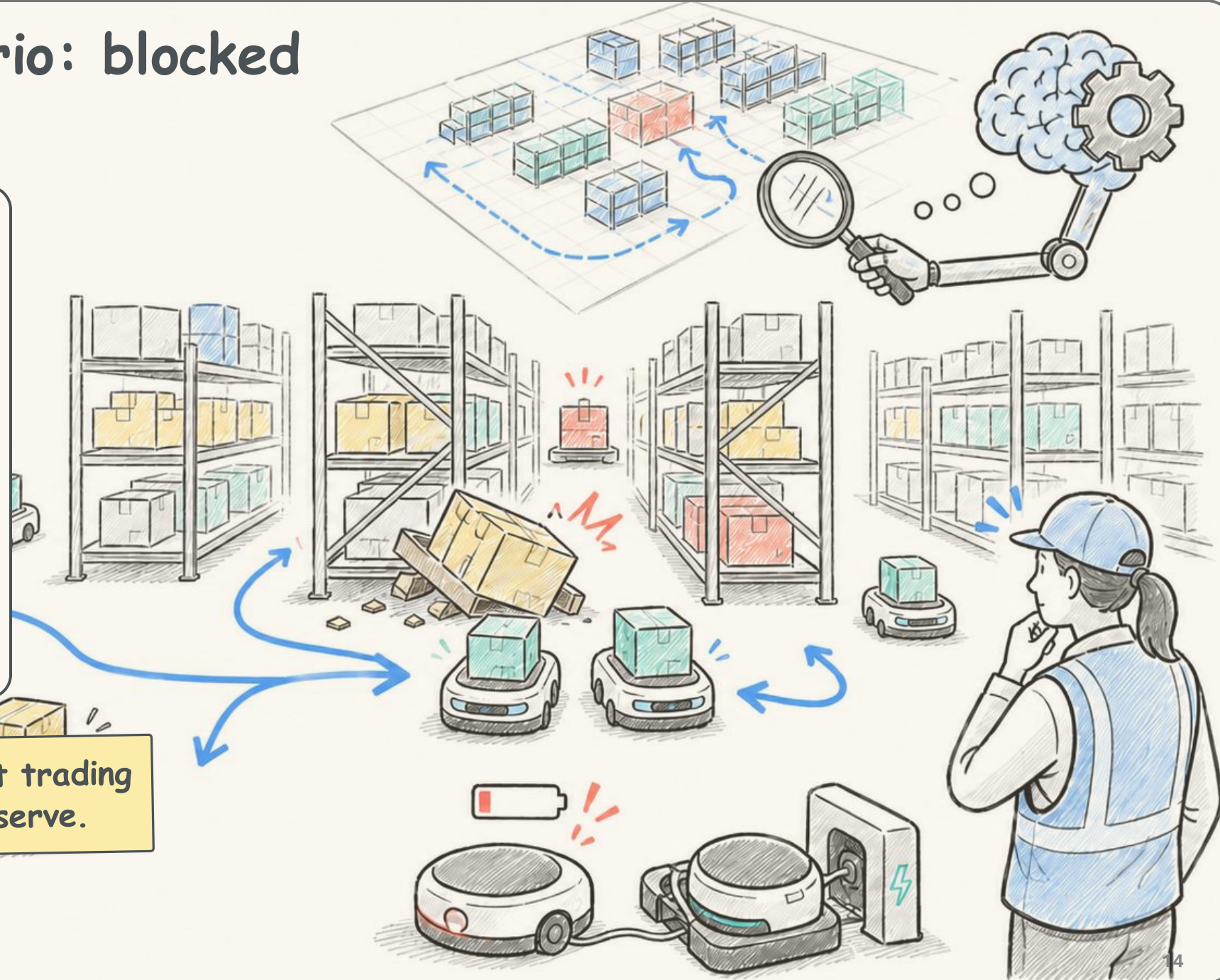
Typed • scoped • authorized • reversible •
logged

Illustrative scenario: blocked aisle

40 AMRs

Blocked aisle
Urgent order at risk
Two robots queued
One charger occupied
One battery near reserve

Goal: recover service without trading away safety or charge reserve.



Step 1: observe and diagnose from evidence

TWIN OBSERVATIONS

Aisle B-7 blocked

Order O-184 due in 11 min

Robot R-12 at 14%

Charger C-2 free in 6 min



DIAGNOSIS

The delay is not only a path problem.
It is a coupled traffic, task, and
charging problem.

Evidence confidence: high

Twin freshness: 1.2 s

Ground every conclusion in observable
state.

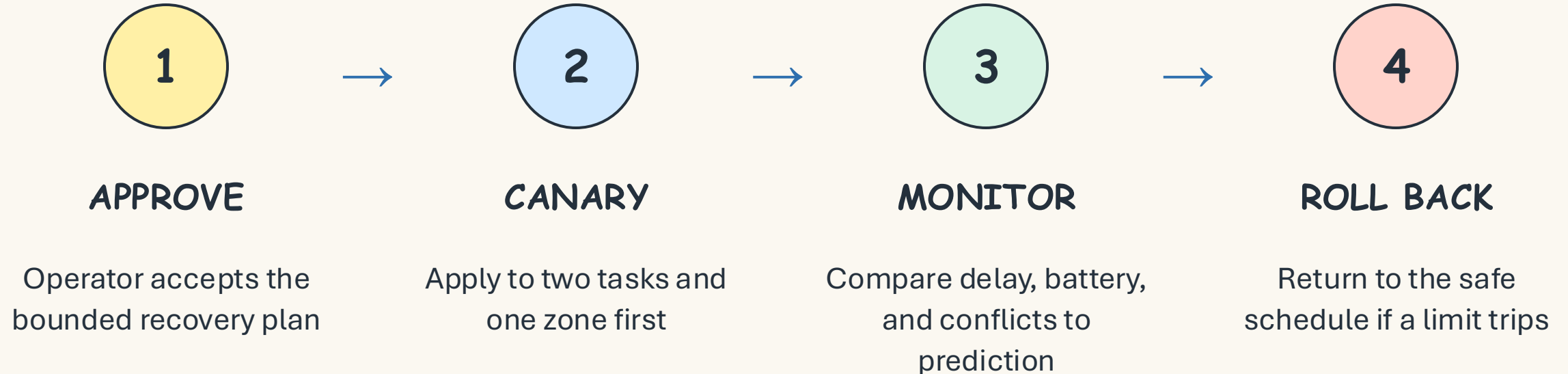
Step 2: compare candidate futures

Candidate	Predicted service	Energy	Risk	
A Reroute all	Order on time	High detour	Congestion	
B Reassign + reroute	Order on time	Reserve held	Low	CHOOSE
C Wait for clearance	Order late	Lowest	Low	

Language proposes. Simulation and optimization provide the evidence.

Illustrative predicted outcomes

Step 3: act, verify, and roll back



Exit conditions: safety event • KPI deviation • stale twin • tool failure • human override

A decision needs a record, not a rationale

Trigger

Blocked aisle B-7 at 18:22:14

Evidence

State snapshot + freshness + confidence

Candidates

A / B / C with predicted KPIs

Constraints

Safety zones, battery floor, order SLA

Authorization

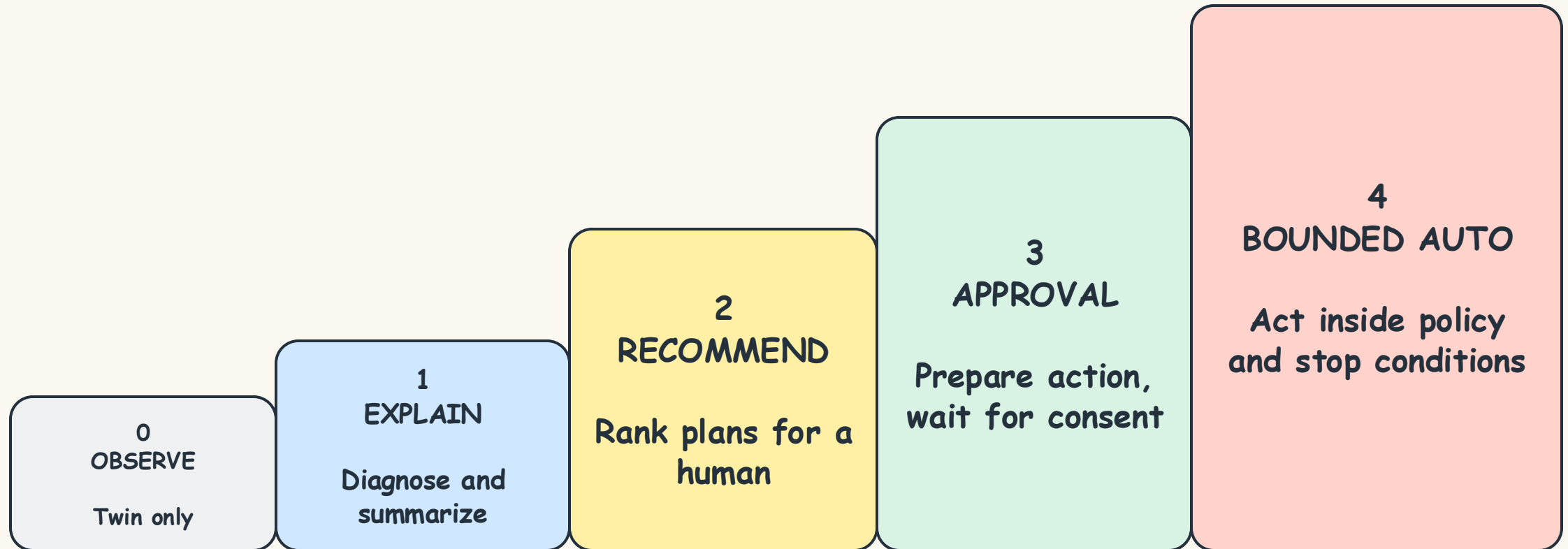
Approved by operator 23; scope: zone B

Outcome

Delay -7 min; no safety event; twin error 4%

Store state and consequences, not hidden chain-of-thought.

Autonomy is a policy, not a model feature



More authority requires more evidence, narrower scope, and stronger recovery.

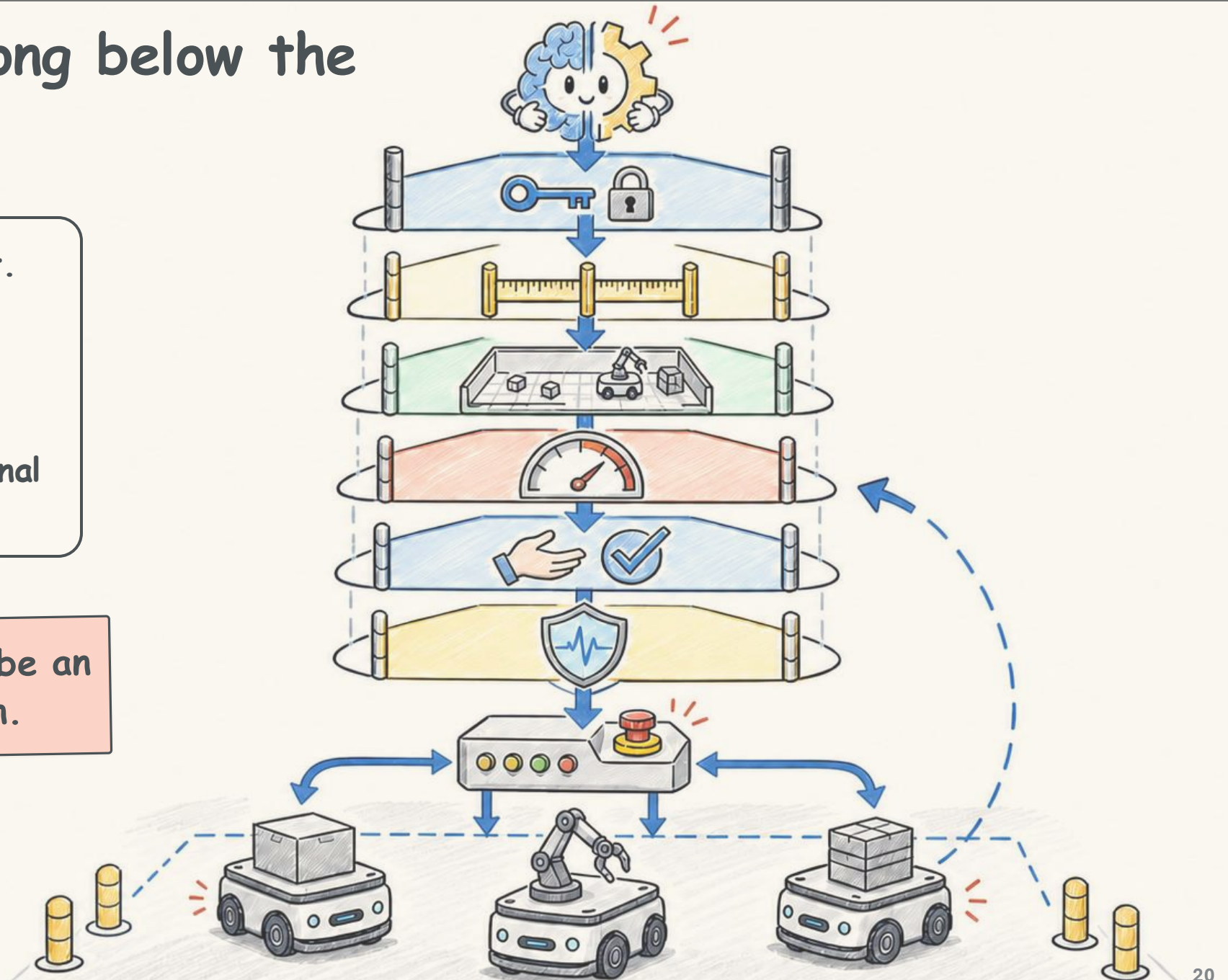
Safety gates belong below the agent

The agent proposes intent.

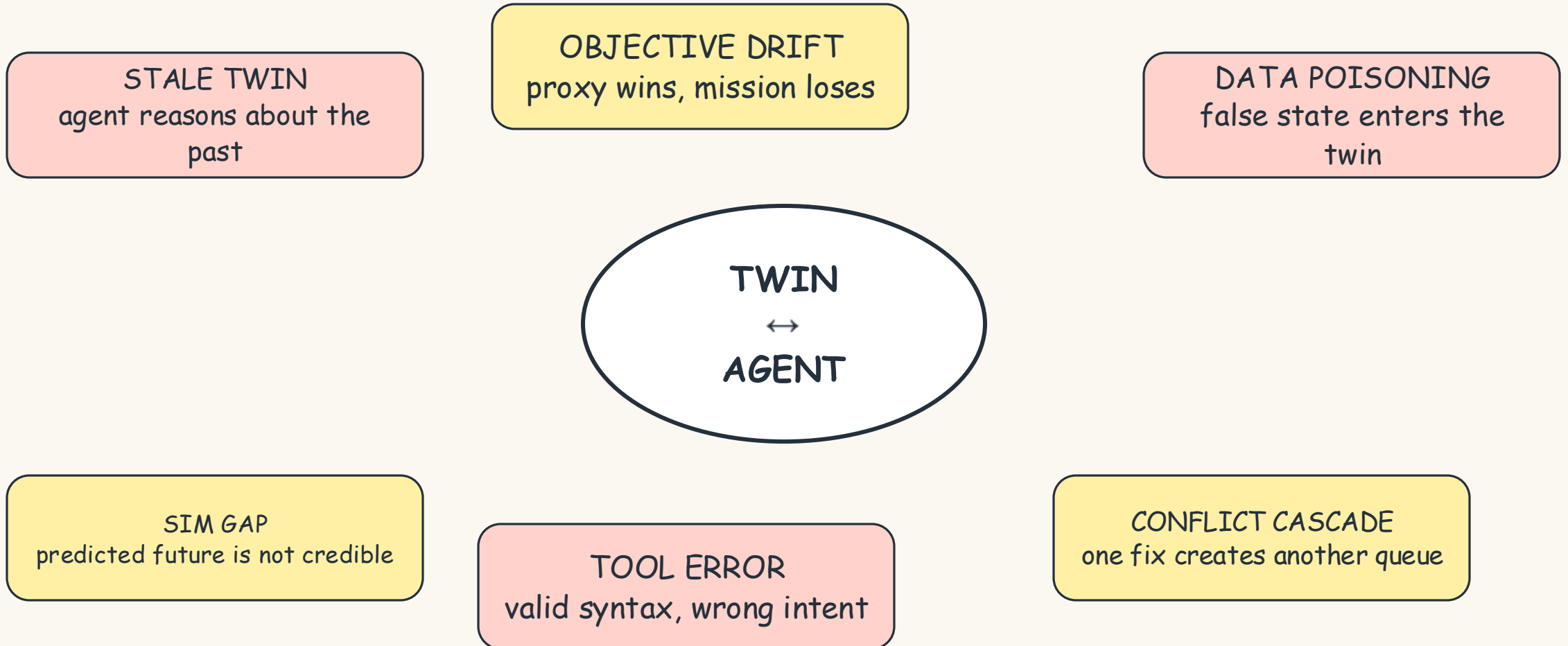
Hard constraints decide feasibility.

Safety controllers retain final authority.

A valid tool call can still be an unsafe physical action.



Two coupled loops can amplify errors



Test the policy before the plant floor

1 **REPLAY** Historical incidents and edge cases

2 **SHADOW** Live recommendations with no authority

3 **COUNTERFACTUAL** Twin simulations against alternative actions

4 **STRESS** Latency, packet loss, stale data, adversarial events

5 **CANARY** Small scope, tight limits, immediate rollback

6 **RUNTIME** Continuous monitors and human override

Interoperability creates safe handles

ISO 23247

Manufacturing twin
architecture and
information exchange

VDA 5050 v3

Command and status
interface for mixed
mobile-robot fleets

Open-RMF

Task allocation, traffic
negotiation, and shared
infrastructure

ISO 21423

Emerging AMR
communication and
interoperability standard

Interoperability defines what can be exchanged. Safety still needs its own standard, control logic, and verification.

For driverless industrial trucks and AMRs, see ISO 3691-4:2023.

Earn autonomy in stages

INSTRUMENT

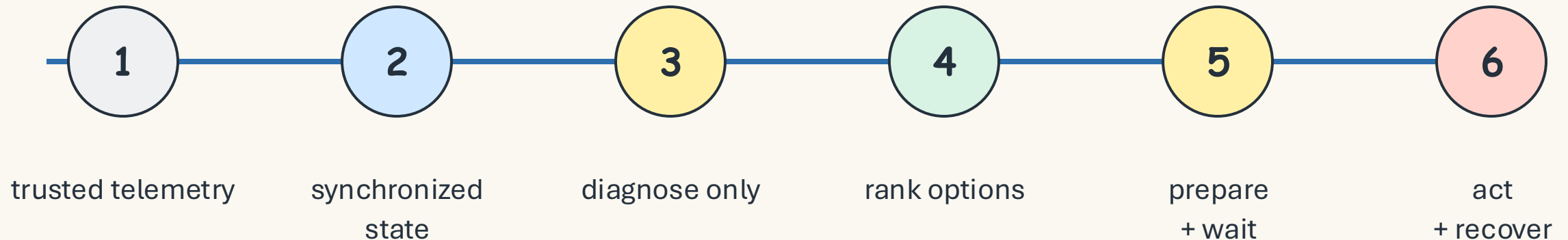
TWIN

SHADOW AGENT

RECOMMEND

APPROVAL

BOUNDED AUTO



Do not automate a decision you cannot observe, simulate, constrain, and reverse.

A practical first 90 days

DAYS 0-30

Choose one decision

Map state, authority,
and failure cost

Instrument freshness
and data quality

DAYS 31-60

Build the twin slice

Add replay and scenario
simulation

Expose read-only tools

DAYS 61-90

Run a shadow agent

Compare
recommendations to
operators

Define gates for a
canary

Best first target: frequent, costly, bounded, and
currently manual.

Research questions that matter now

1. How should twin fidelity adapt to the decision and risk?

2. How do we measure agent value beyond task success?

3. How can fleet memory remain useful without becoming stale?

4. How should humans supervise many interacting agents and robots?

5. What evidence is sufficient to earn more action authority?

Three ideas to carry into your next system

1

THE TWIN

is a synchronized decision model, not a dashboard

2

THE AGENT

reasons through tools, models, and constraints

3

AUTONOMY

is earned through policy, evidence, and recovery

Discussion: What decision in your operation is frequent, costly, and still manual?

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