

# SELF-DRIVING NETWORKS

## “LOOK, MA: NO HANDS”

Kireeti Kompella  
CTO, Engineering  
Juniper Networks

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**JUNIPER**  
NETWORKS

Engineering  
Simplicity

# VISION

# THE DARPA GRAND CHALLENGE

## BUILD A FULLY AUTONOMOUS GROUND VEHICLE

### GOAL

Drive a pre-defined 240km course in the Mojave Desert along freeway I-15

### PRIZE

\$1 Million

### RESULT

**2004: Fail** (best was less than 12km!)

**2005:** 5/23 completed it

### 2007: “URBAN CHALLENGE”

Drive a 96km urban course following traffic regulations & dealing with other cars  
6 cars completed this

## IMPACT

- **Programmers, not drivers**
- No cops, lawyers, witnesses
- Quadruple highway capacity
- Glitches, insurance?
- Ethical Self-Driving Cars?

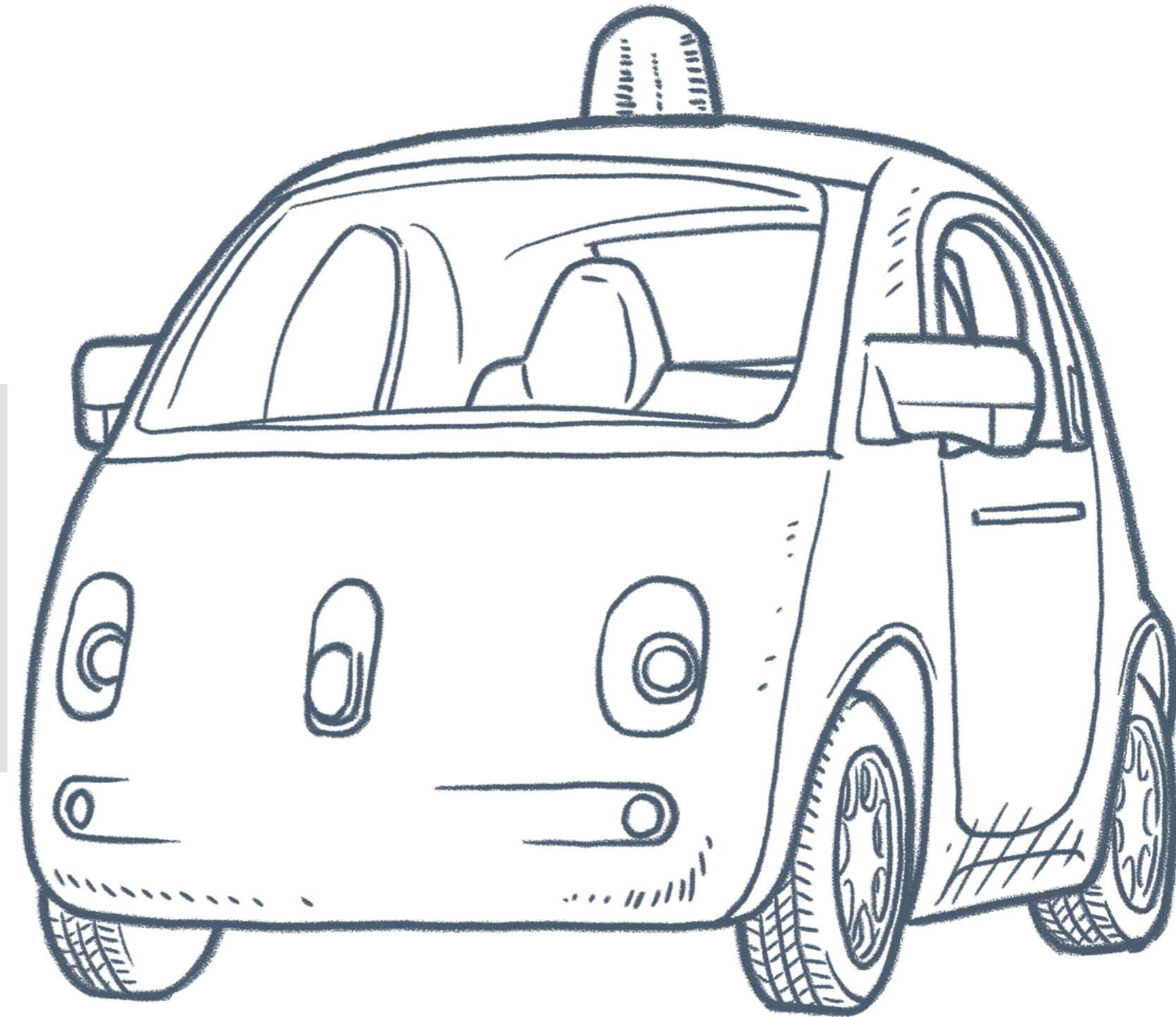
## POSSIBILITIES



## GRAND RESULT: THE SELF-DRIVING CAR: (2009, 2014)

No steering wheel, no pedals—  
a completely autonomous car  
Not just an incremental improvement

This is a **DISRUPTIVE** change  
in automotive technology!



# THE NETWORKING GRAND CHALLENGE

## BUILD A SELF-DRIVING NETWORK

### GOAL

Self-Discover—Self-Configure—Self-Monitor—Self-Correct—Auto-Detect  
Customers—Auto-Provision—Self-Diagnose—Self-Optimize—Self-Report

### RESULT

**Free up people** to work at a higher-level: new service design and “mash-ups”  
**Agile**, even **anticipatory** service creation  
**Fast, intelligent response** to security breaches

### CHALLENGE

Build and operate a self-driving **service network** that **greatly increases agility** and **vastly improves service quality** by **proactive maintenance**  
Autonomously run the end-to-end life-cycle of a service  
Learn user behavior and anticipate changing user requirements

## IMPACT

- **New skill sets required**
- **New focus**
  - BGP/IGP policies → AI policy
  - Service config → service design
  - Reactive → proactive
  - Firewall rules → anomaly detection

## POSSIBILITIES



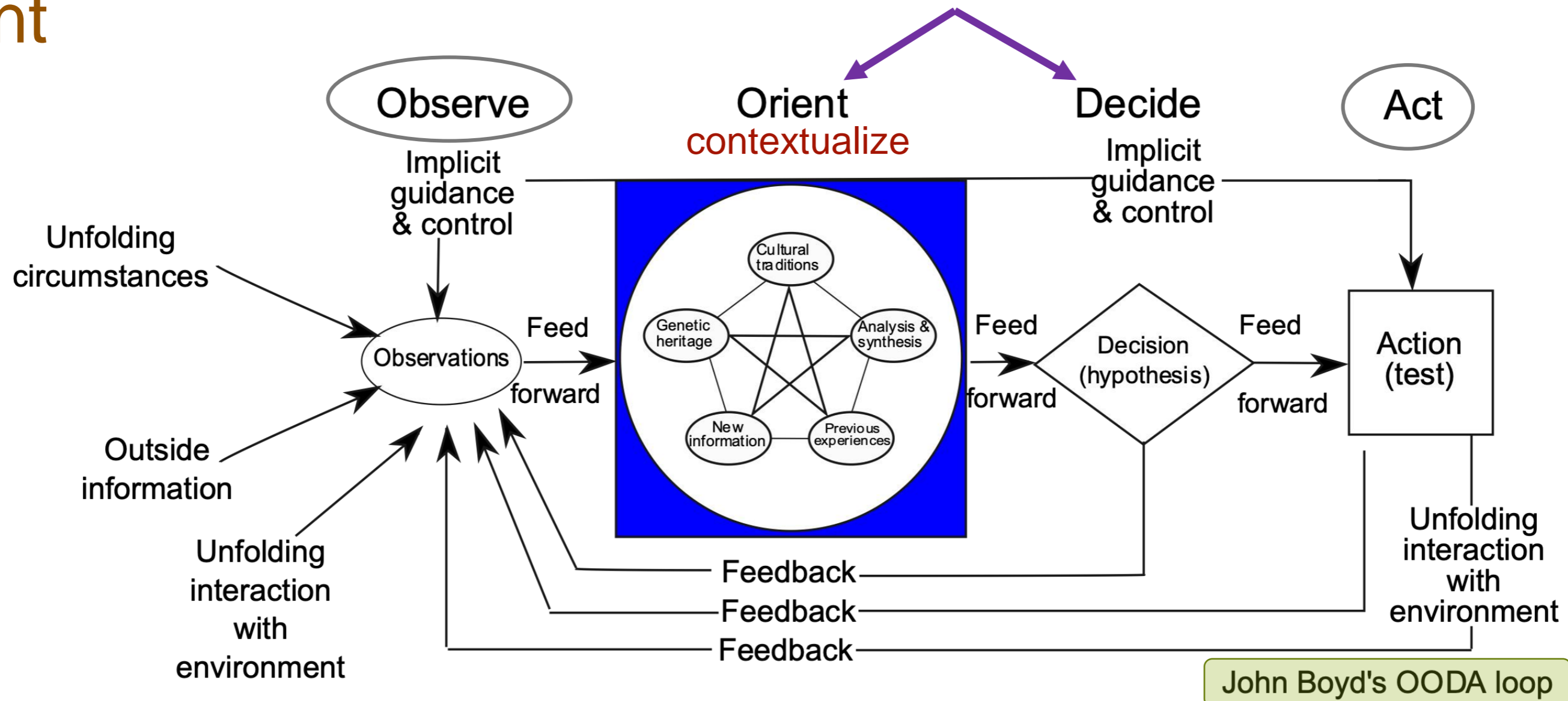
# CONCEPT

What would this look like?

Think **SON**, but applied to all aspects of operations, to all parts of the network

# CLOSED LOOP SYSTEMS (THE OODA FRAMEWORK)

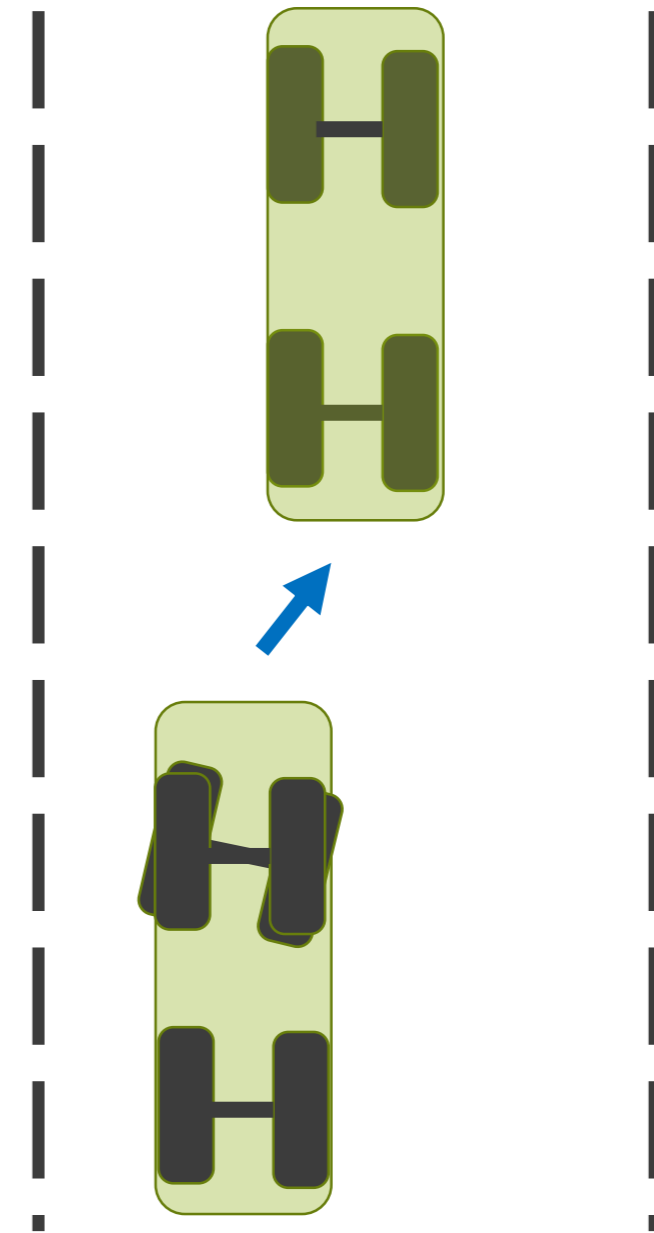
## Intent



BREAK DOWN THE PROBLEM TO  
MANAGEABLE CHUNKS



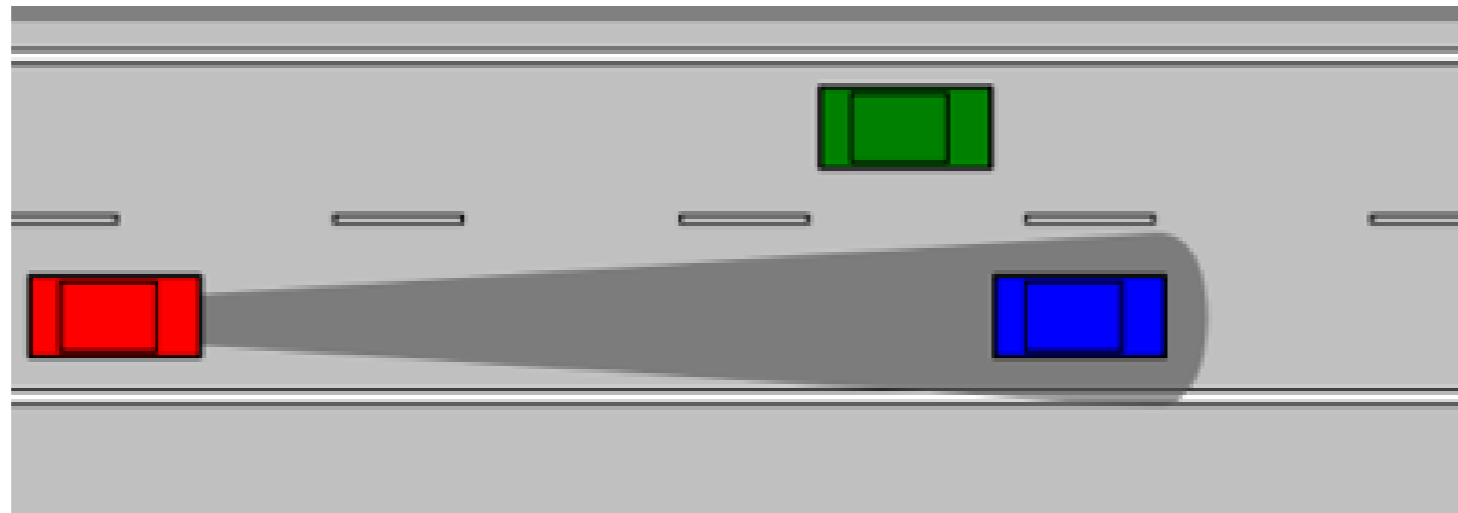
# ILLUSTRATION: AUTOMATIC LANE CENTERING



Intent

1. Observe
2. Orient
3. Decide
4. Act

# ILLUSTRATION: AUTOMATIC SPEED CONTROL



Act: while braking, must stay within lane  
Act: may decide to change lanes instead of (or even while) braking  
**As subsystems interact, you get more complex and more capable systems**

## Intent

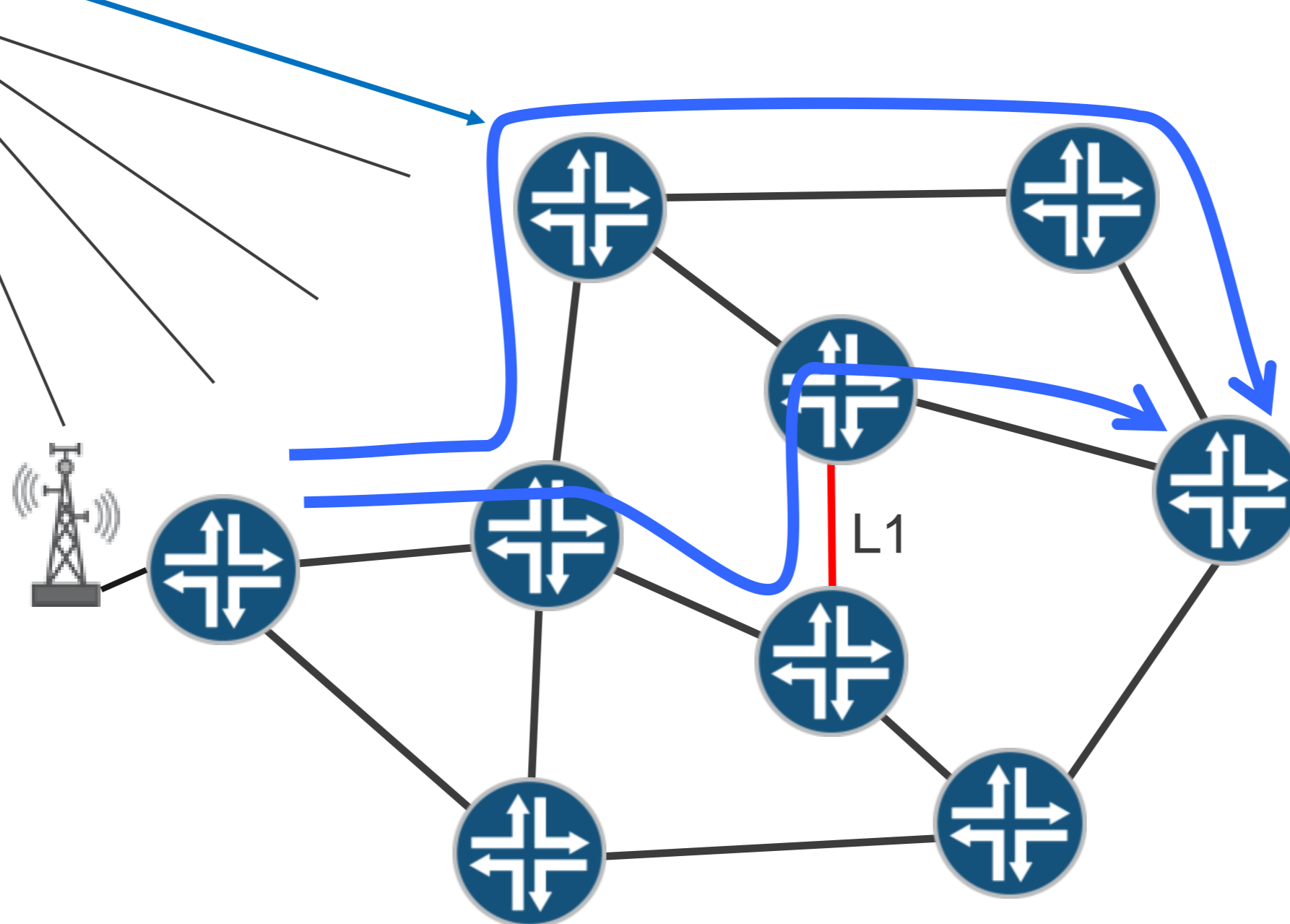
1. Observe
2. Orient
3. Decide
4. Act

# ILLUSTRATION: SELF-MANAGED SLAs



Intent: maintain SLAs

1. Streaming Telemetry from all network nodes
2. Are SLAs being met?
3. Decide that Link L1 is dropping traffic
4. Recompute all paths that pass through Link L1



Simple example, but consider the following:

1. Need this for real-time operation!
2. SLAs getting more complex, critical
3. Decisions must address root cause
4. Action must not violate other SLAs

# BREAKING THIS DOWN

## Architecture

1. Observe: build a robust data ingest engine

All kinds of data: stats, traps, syslogs, thresholds, .... and

expertise and experience

3. Decide: does the behavior match the intent?

Needs deep knowledge of the service, SLAs, deployment, expertise and experience

2. Orient: understand how the data fits together

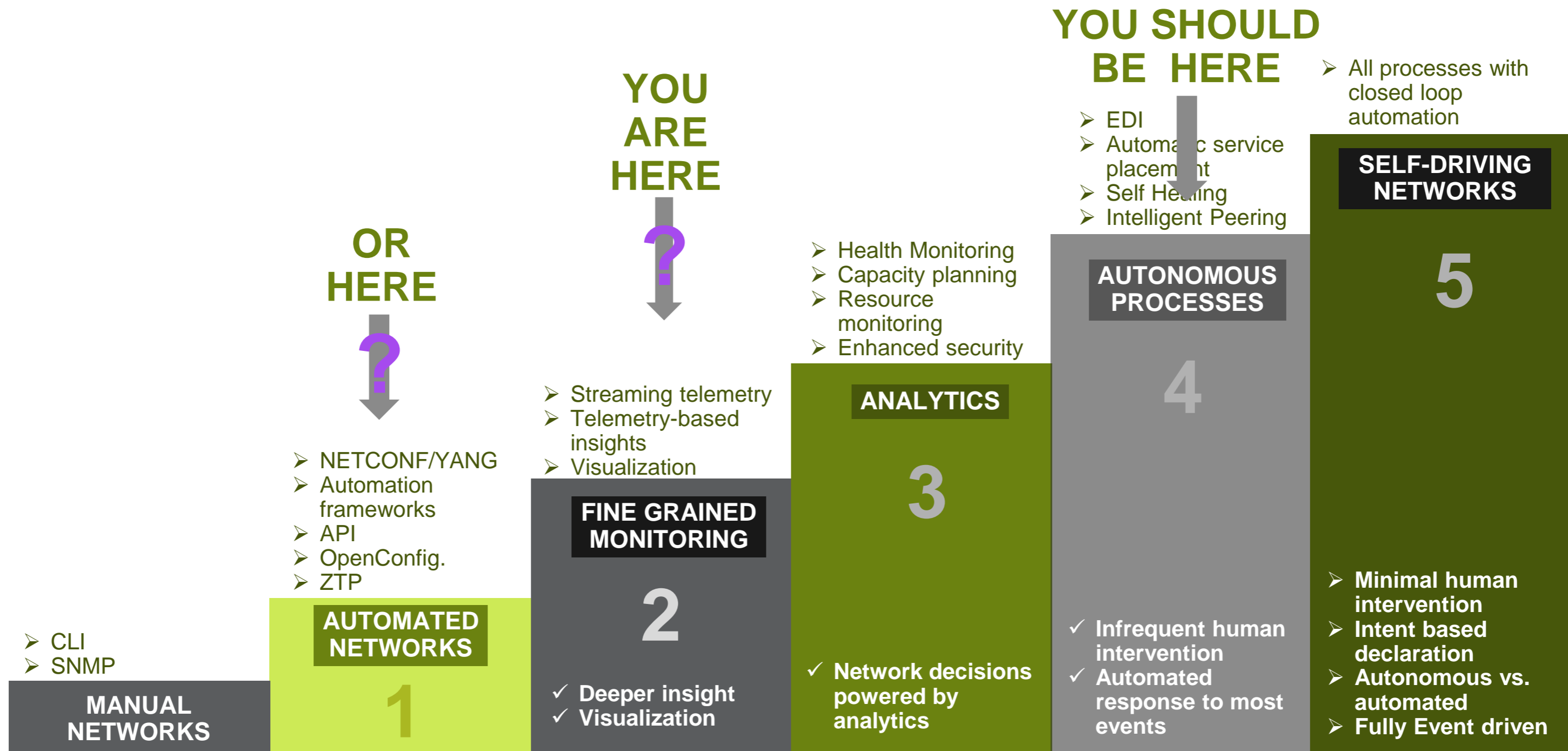
Needs deep knowledge of implementation, debugging, NOC expertise & experience

4. Act: return behavior to the intent (via netconf, etc.)

Needs clear understanding of consequences, careful sequencing of steps, expertise and experience

# SELF-DRIVING NETWORKS

## FIVE STAGES TOWARDS THE LONG-TERM VISION





# DEPLOYMENT

What does it take to put this in action?

# PERSPECTIVE ON CONTROL AND TRUST

MIT Technology Review,  
Will Knight, Oct 2013  
“Driverless Cars Are Further  
Away Than You Think”

Musk, Dec 2017:  
full self-driving by  
end of 2019

Musk, Dec 2015:  
full autonomy by  
end of 2018

# GIVING UP CONTROL WORRIES SOME, EXCITES OTHERS

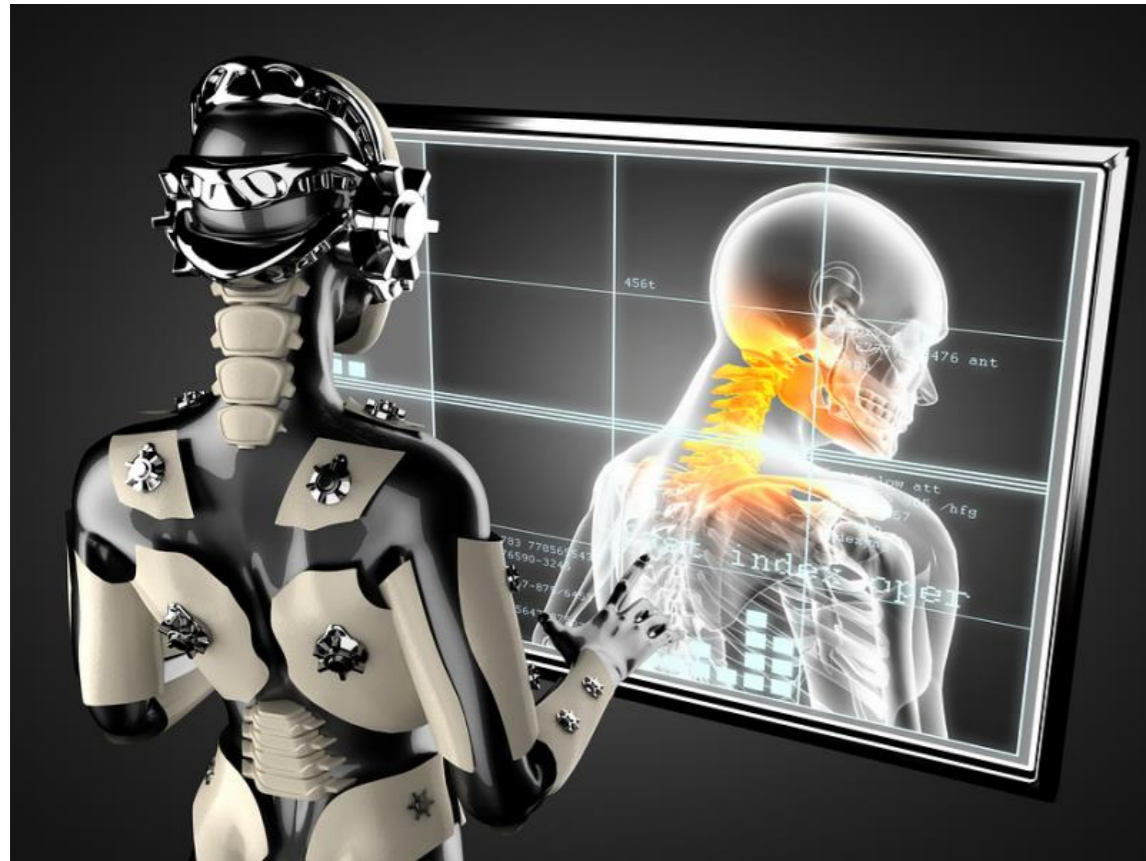


Photo: [www.dotmed.com](http://www.dotmed.com)



Photo: [www.cio.com.au](http://www.cio.com.au)

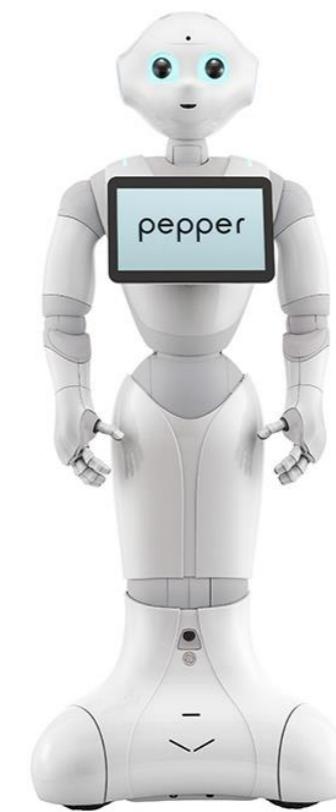


Photo: SoftBank/  
Aldebaran Robotics



Photo: [newyorker.com](http://newyorker.com)



Photo: [erobots.in](http://erobots.in)

- ML?**
- The Three Laws of Robotics
1. Don't hurt/kill humans
  2. Obey a human's orders
  3. Protect yourself



# IF CONTROL COMES FROM ML, UNDERSTAND THIS

Amazingly accurate in its domains

Limitations

- Opaque & un-understandable — Explainable AI?
- Blind, goal-seeking — Ethical AI?
- No “common sense” — but will that change?
- Easy to fool — adversarial techniques
- Three-year-olds can do much better in most contexts

Search: one pixel attack



Pragmatic approach:  
rule-based decision making

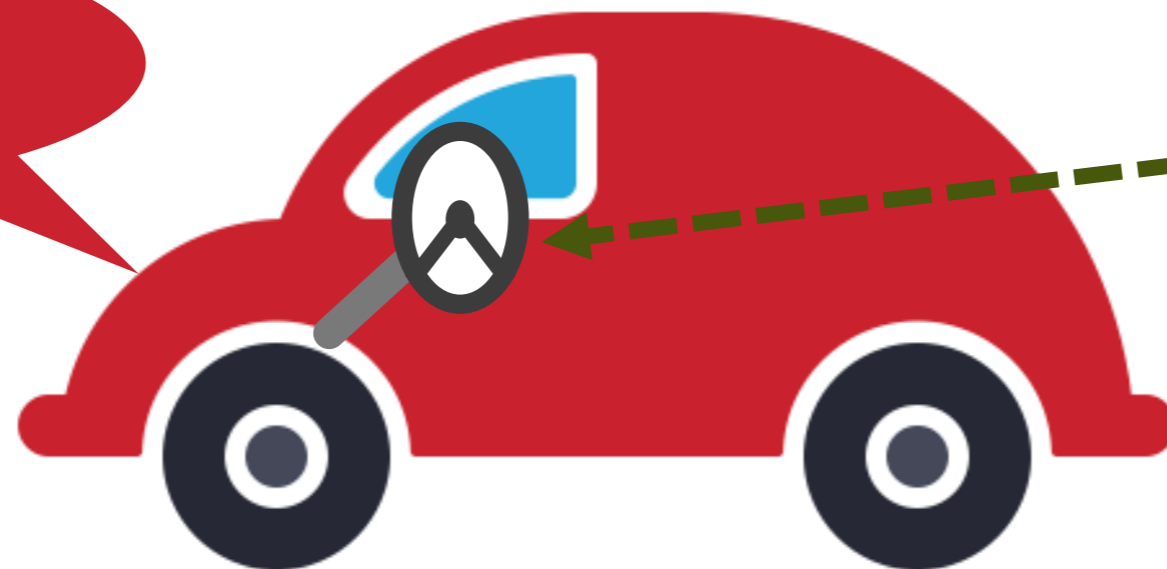
# SOFTWARE-DEFINED NETWORKS: “I WANT MORE CONTROL!”

SDN goes beyond this

Software

Hardware

SDN is about control



# MACHINE LEARNING: MAJOR PARADIGM SHIFT

Program for machine vision

look for feature X  
if this then that  
else the other

look for feature Y

custom feature definition;  
difficult to write;  
complex, fragile

Direct control



Machine Learning

image



label

cat

cat

dog

cat

new image



model



prediction

All you need is lots of labeled data

Indirect control

Photos from [Pexel](#)

# NO PEDALS, NO STEERING WHEEL: OK. NO WINDOWS: NOT OK!

Self-driving cars have raised some trust issues

- Why should I let it drive?
- How do I know it's driving well?
- What are its inner workings?
- My driving style is different

Autonomous, closed loop operation absolutely requires:

1. Customization
2. Feedback
3. Man-machine cooperation
4. Building confidence
5. Tangible benefits



# LESSONS TO LEARN

1. Give up control (but not *all* control)
2. ML is sexy, but may not be the place to start
3. Customization is crucial
4. Building trust is crucial



# IMPLEMENTATION

Make it customizable, controllable, safe

# ACHIEVING AUTONOMY

Closed loops + effective decision-making = self-driving

“Autonomous system” → transfer of control

- from **human to machine**
- what does the machine use for control?

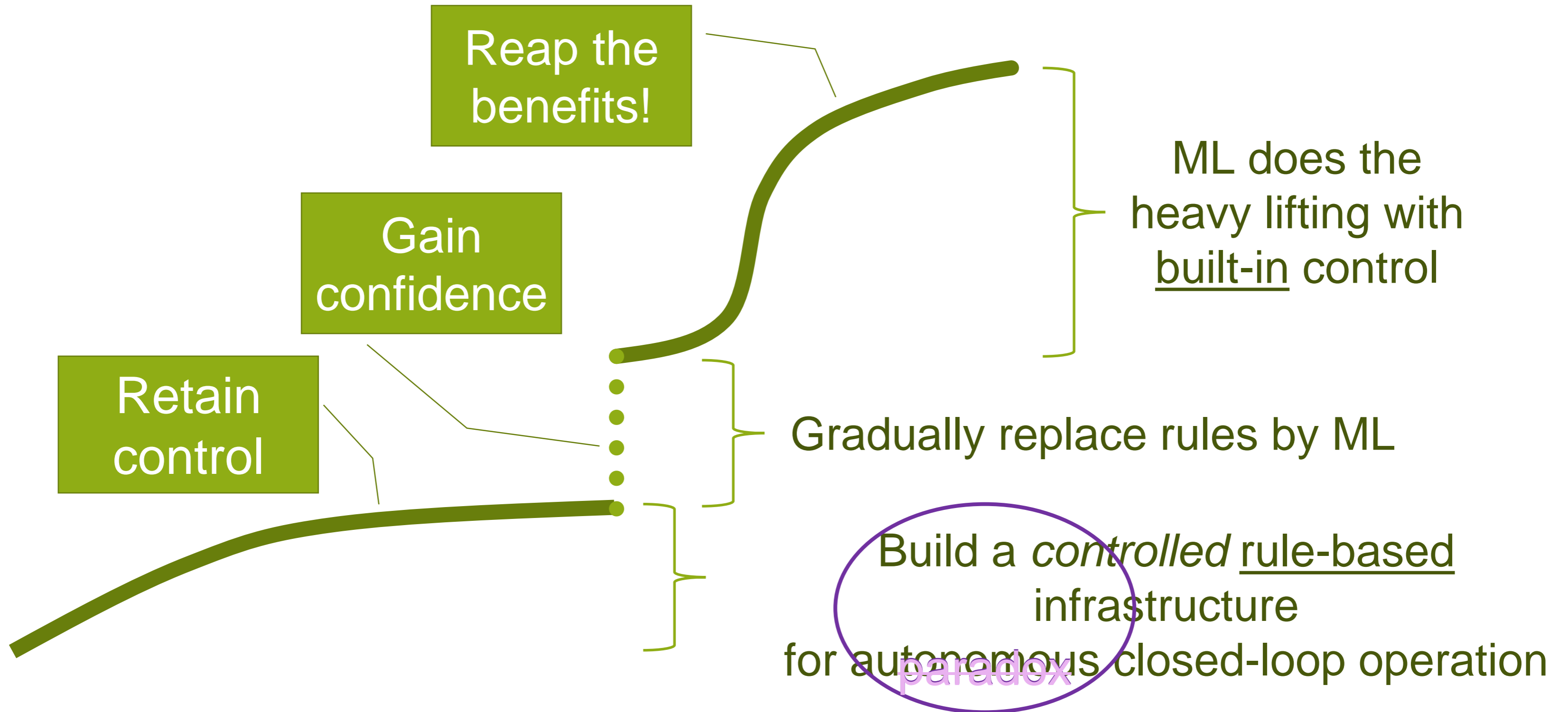
Partial autonomy can be dangerous

- Who's in charge?
- Can the transfer of control be done seamlessly?

Does this mean total abdication of control?  
Fortunately, no

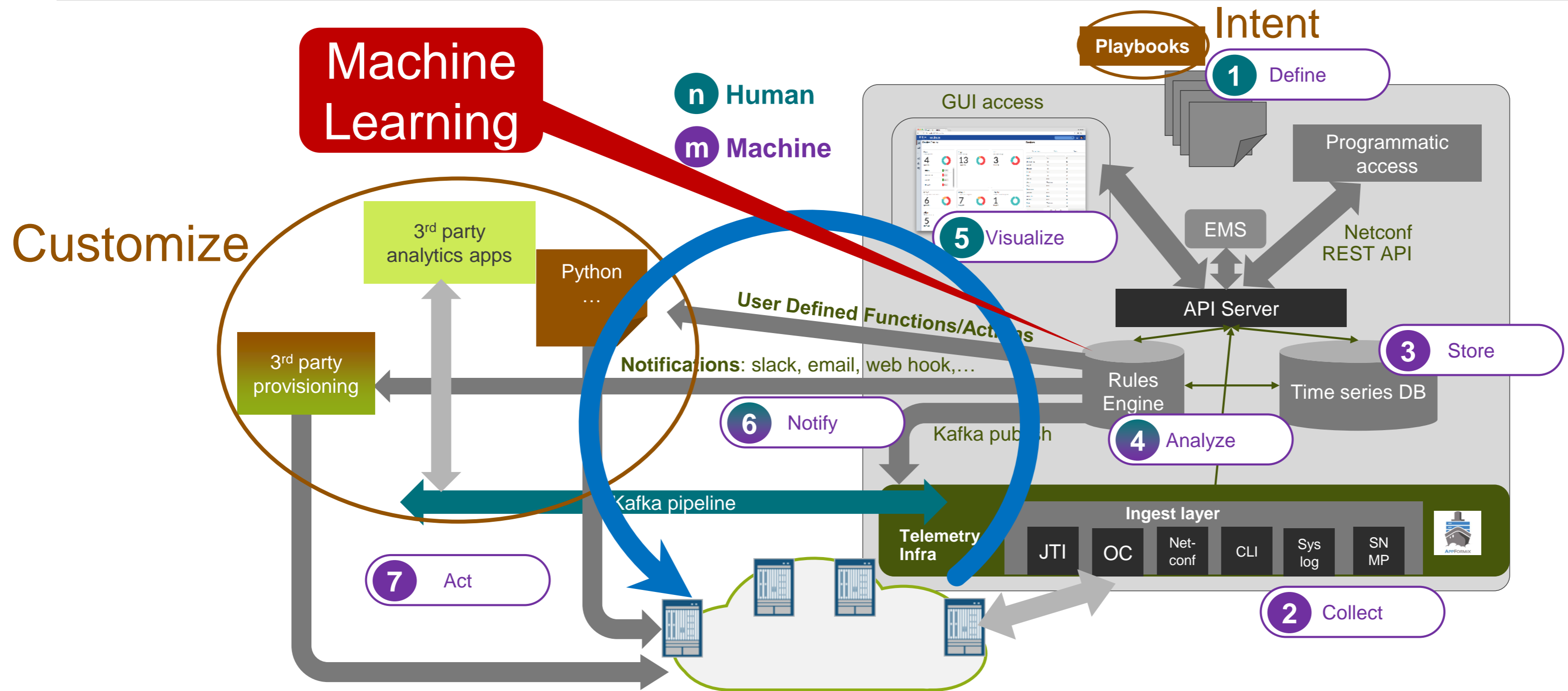
In networking, partial autonomy (Stage 4) is indeed practical

# ROADMAP FOR DEPLOYING SELF-DRIVING NETWORKS





# ENABLING CONTROLLABLE *AUTONOMOUS* CLOSED LOOPS



# GENERAL ARCHITECTURE FOR SELF-DRIVING

The same architecture can be used for various use cases:

- Device health
- Service health
- Underlay management
- ...

The key is the playbook, written in a DSL

- Playbooks are the result of collaboration of engineers, support and network operators
- Playbooks are in a github repo, and can be customized by each network operator (<https://github.com/Juniper/healthbot-rules>)

Things get really fun when playbooks interact

- For example, a “gray failure” of a link can lead to remapping the underlay
- Service (non-)health can be traced to an unhealthy device, which can then be fixed



This is where there is a need for more control until the interactions are better understood



# RELEVANCE TO NETWORK2030

Should this Focus Group take on this work?

# BENEFITS

The goals of Self-Driving Networks are:

1. Capture human expertise & experience
2. Reduce mundane work (and thus errors)
3. Effectively utilize all possible data (telemetry, logs, ...)
4. Improve efficiency on all fronts
5. Move to predictive operations (maintenance, service creation, ...)

Many OPS groups face loss of talent

To free up time for innovation, for paying down “technical debt”, etc.

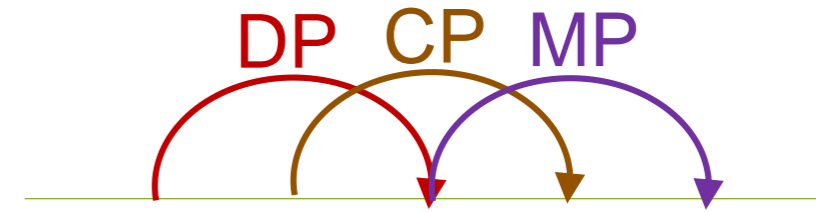
Humans can't possibly process all the data available, nor prioritize it well

Better utilization of resources, dynamic shifting of workloads, ...

This is the game-changer, but first, the infrastructure must be built

**Human-Machine cooperation!**

# START WITH MANAGEMENT PLANE!



As networks get bigger  
and more complex  
(densification, IoT) ...

As SLAs get tougher  
(e.g., latency → holographic)  
...

As the data plane evolves  
to meet new needs  
(e.g., in-band telemetry) ...

As control plane actions have  
more serious consequences  
(e.g., remote surgery) ...

Let's bring new tools to bear on the  
problem of managing networks!

# SUMMARY

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I've presented a new approach to managing networks, and a general architecture for **controllable** *autonomous* closed loops

- The starting point is rule-based, maintaining control via playbooks
  - Playbooks capture and distill human **expertise and experience**
  - This approach builds confidence and trust
- ***Closed loops can interact with each other, giving new use cases***
- Eventually, rules → ML, while retaining control and trust
- Check out the playbooks on GitHub!

THANK YOU!