

Streaming Data Analytics

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The vision

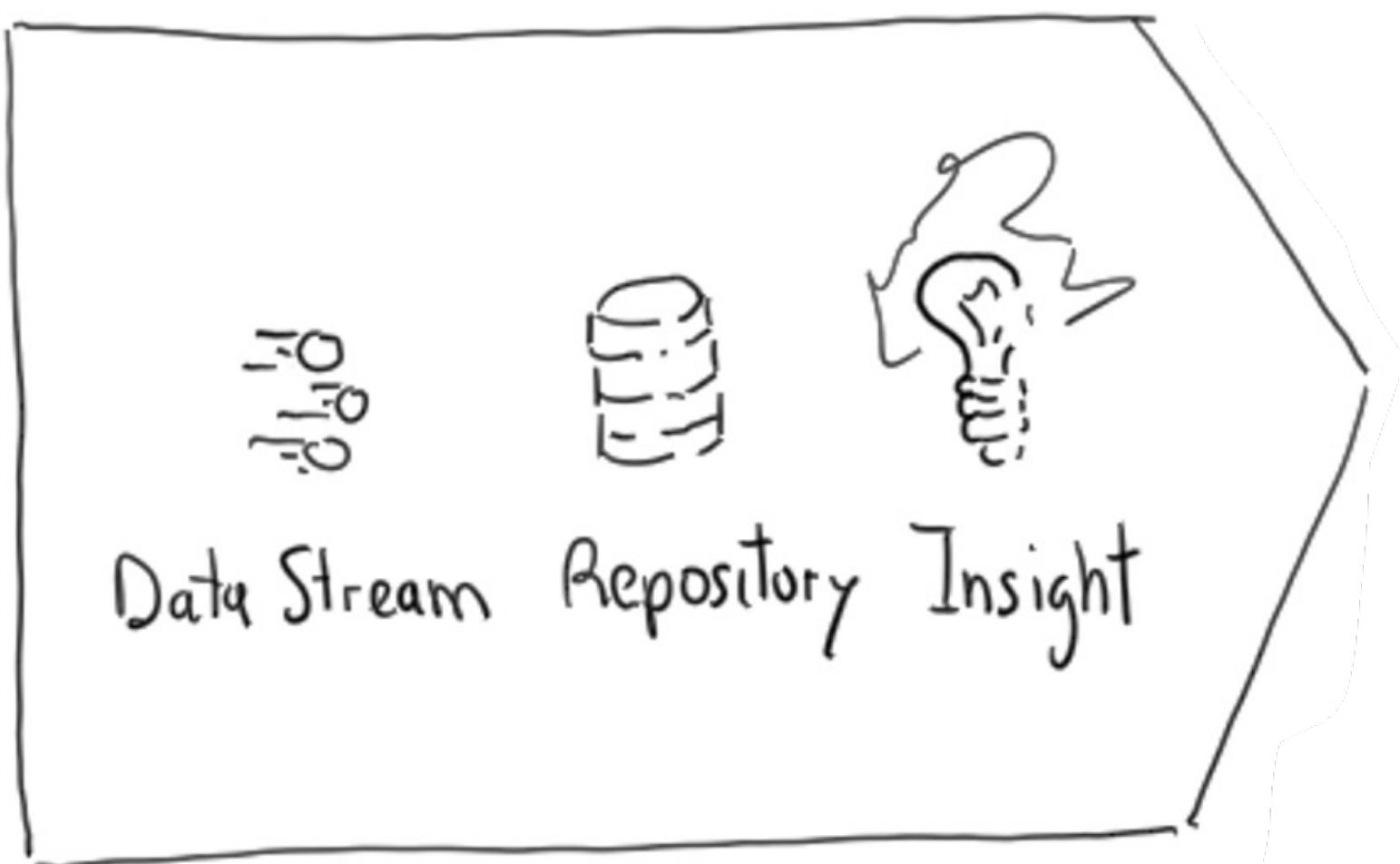
Continuity matters!



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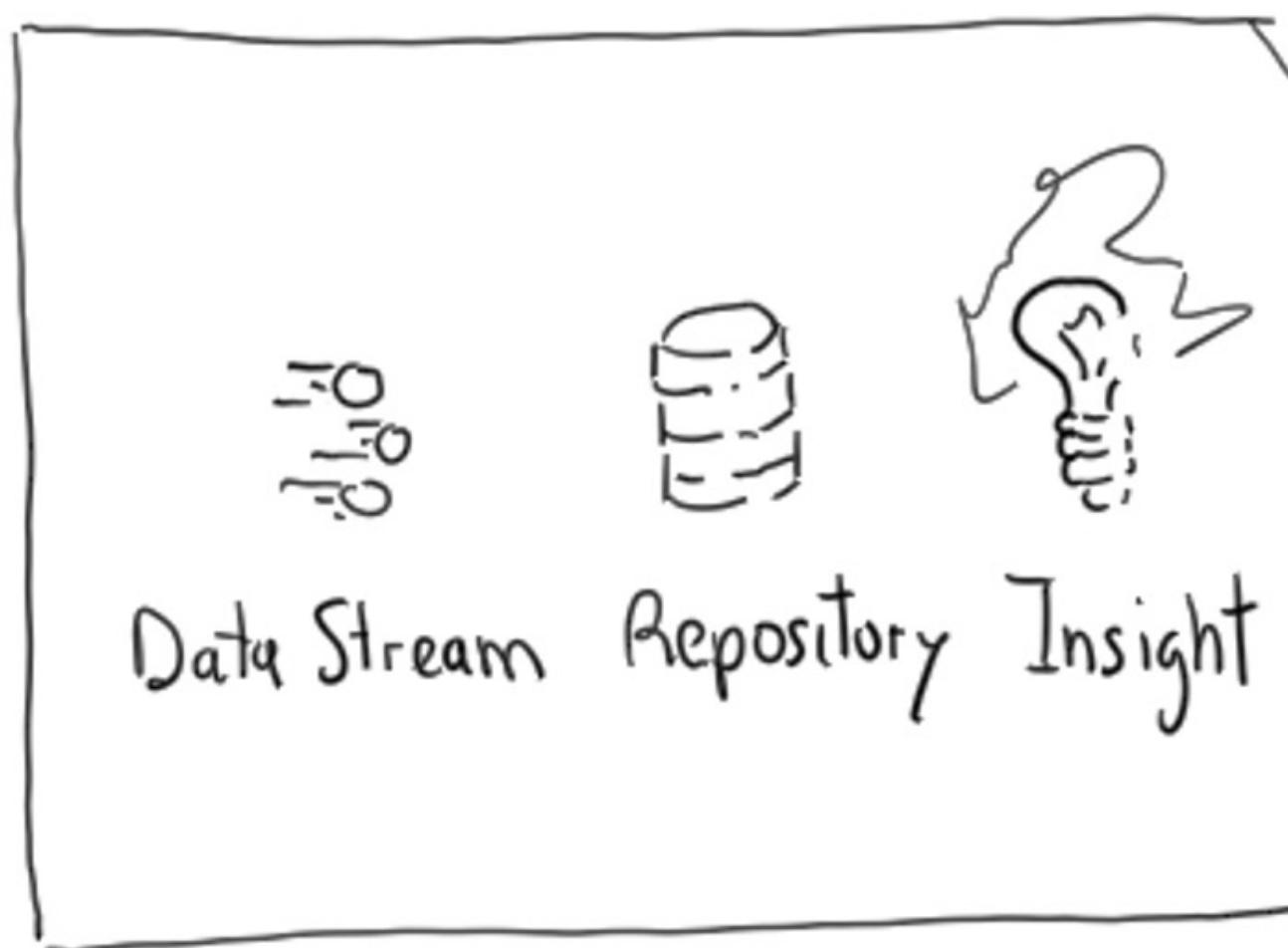


Traditional approach



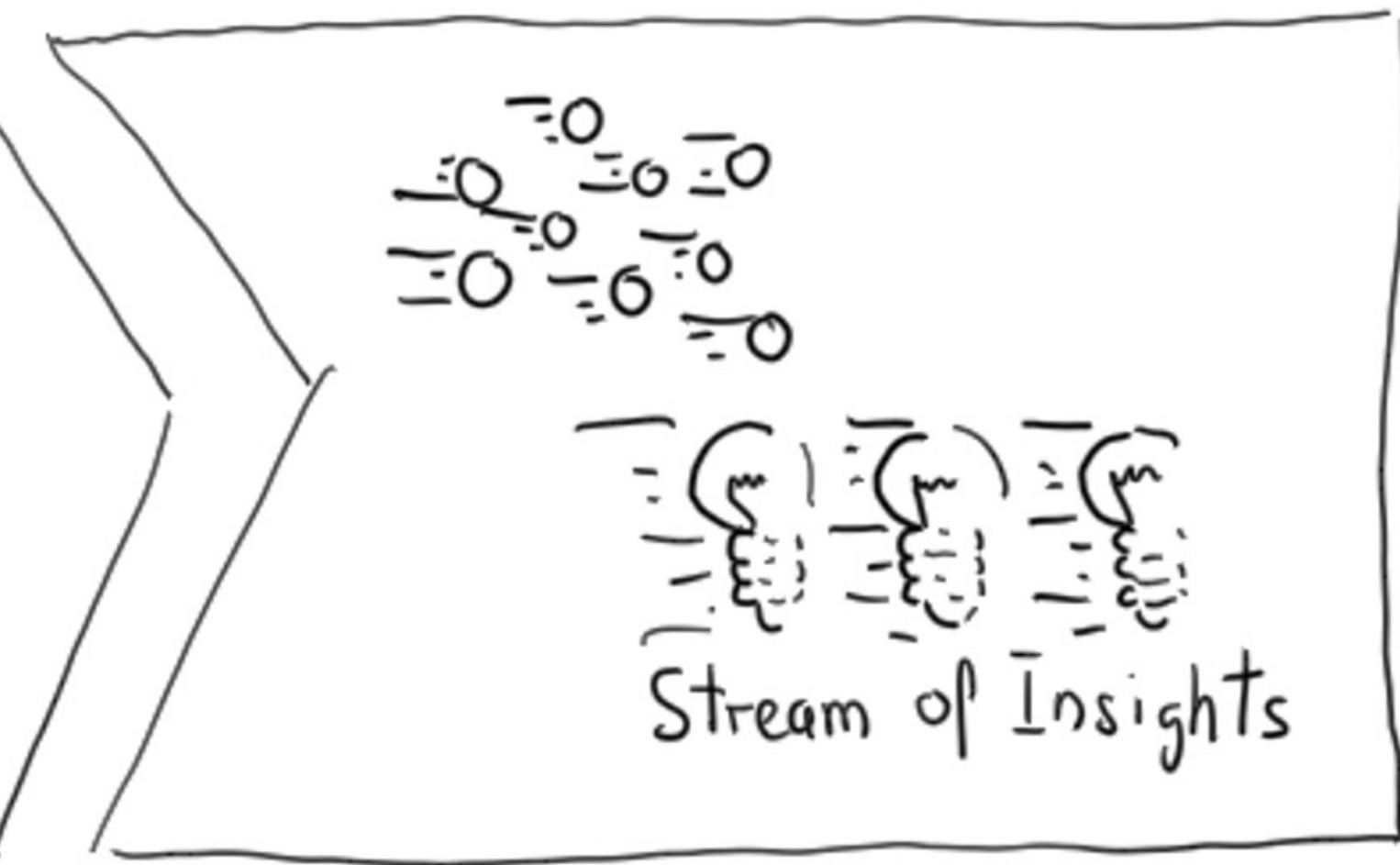
Stop data to analyse

Traditional approach



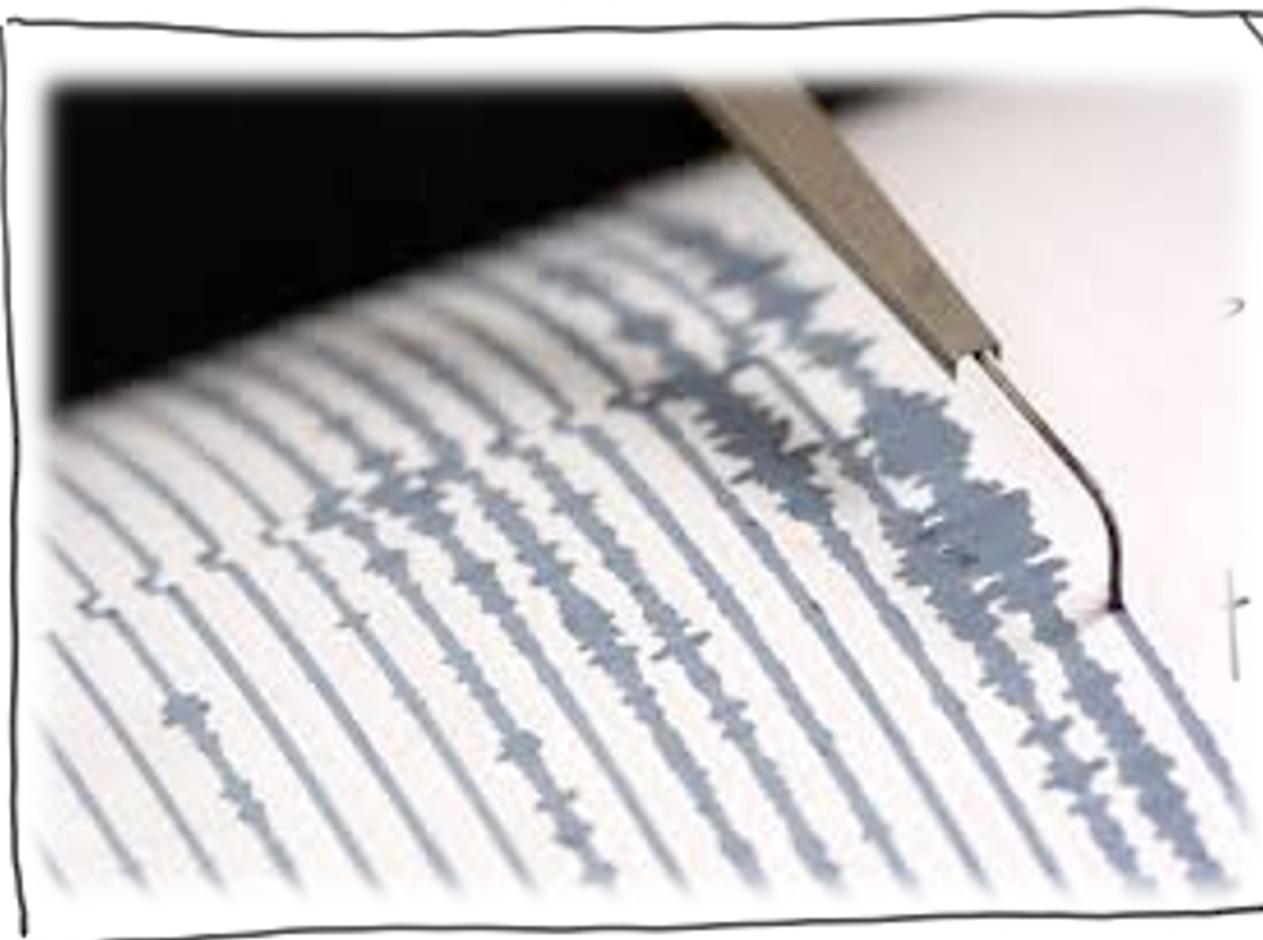
Stop data to analyse

Velocity approach



Analyse data in motion

Traditional approach



Stop data to analyse

Velocity approach



Analyse data in motion

The key ingredients

The key ingredients

1. Sensor and actuators
2. Connectivity
3. Streaming Data Engineering
4. Streaming Data Science
5. People & Processes

The key ingredients

1. **Sensor and actuators** (*as sources and sinks, not covered in the course*)
2. Connectivity
3. Streaming Data Engineering
4. Streaming Data Science
5. People & Processes

Sensors

Eyes using cameras

Color Cameras



Fixed focus



auto focus



optical image stabilization

Color+Depth Cameras



Short range

Long range

Actuators

Lights to display

Screens



Projections



Holograms

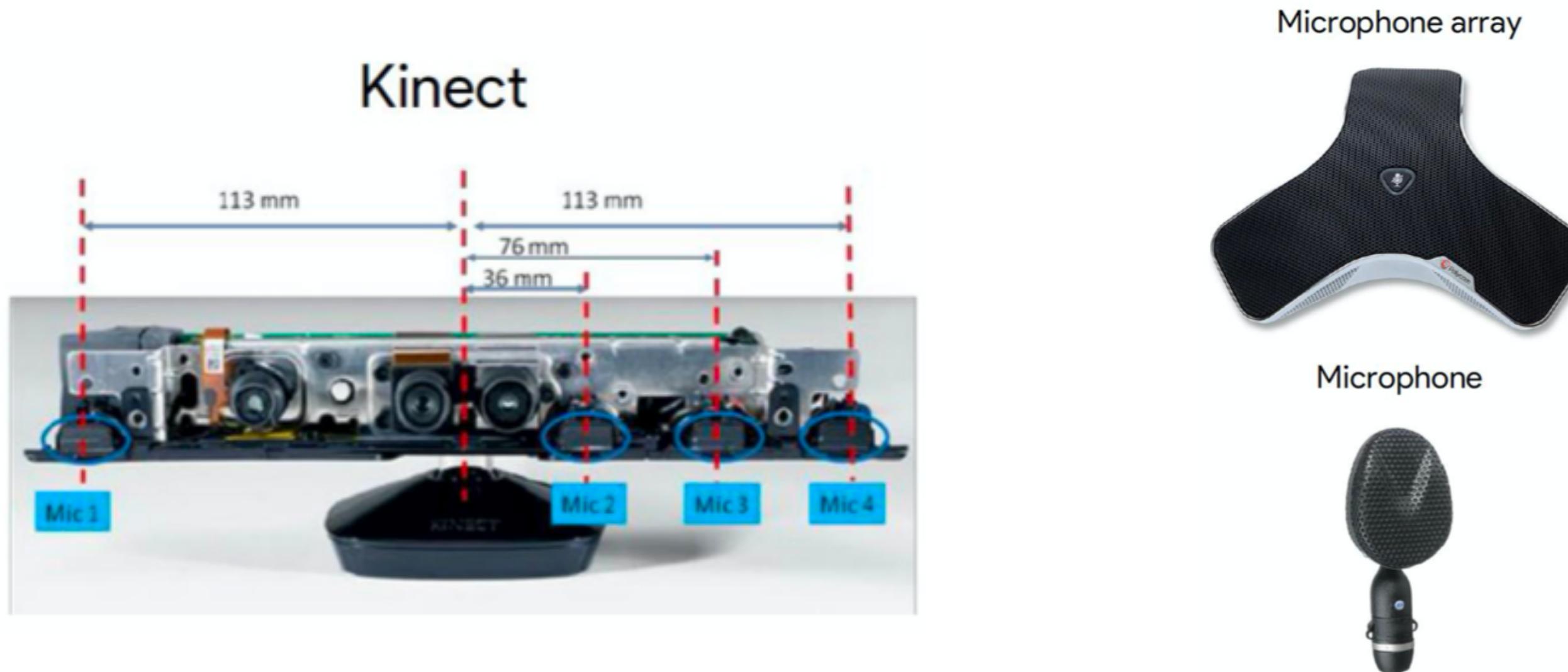


Lights



Sensors

Ears using microphones



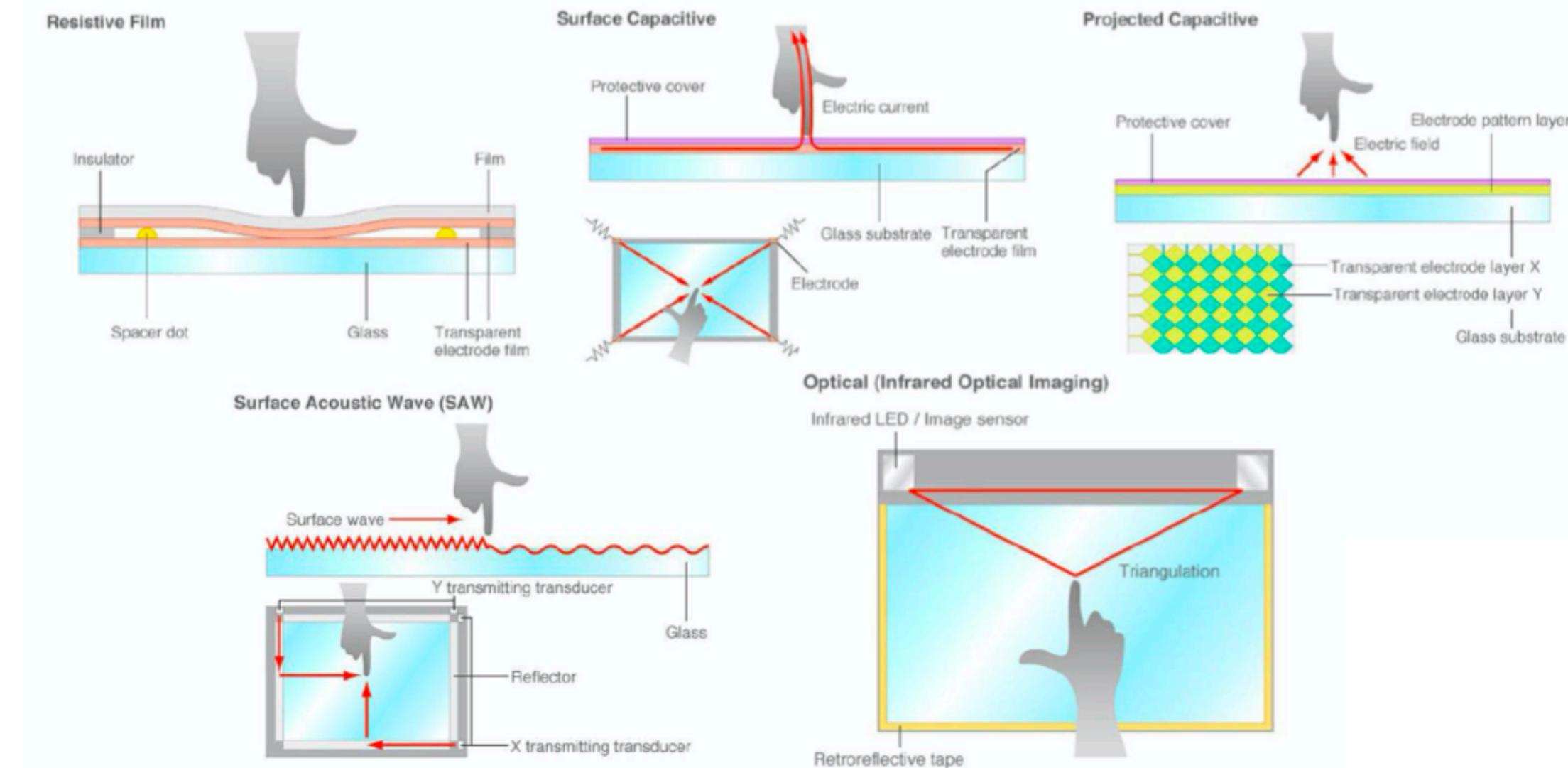
Actuators

Speakers to speak



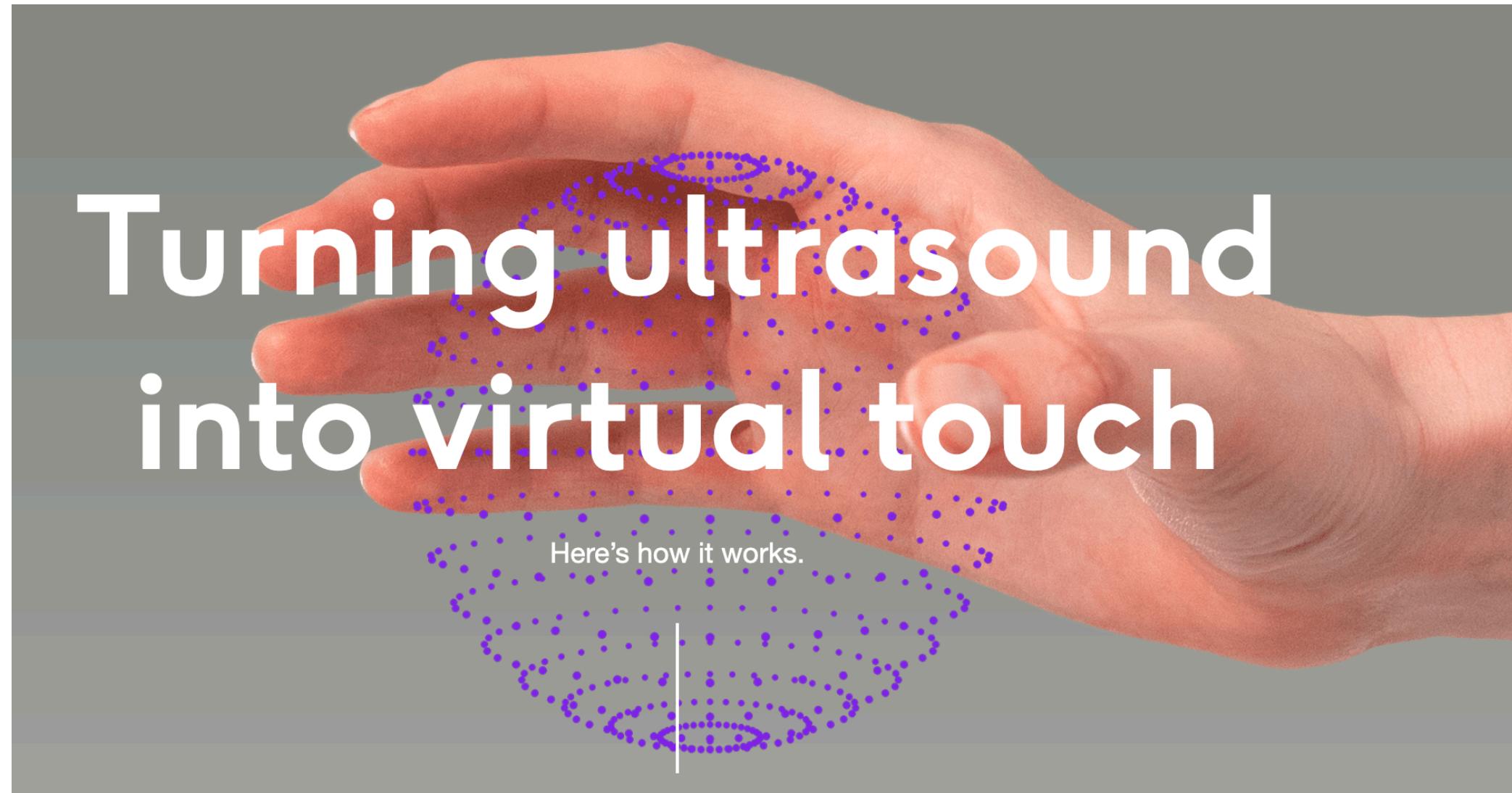
Sensors

Skin to feel touches



Actuators

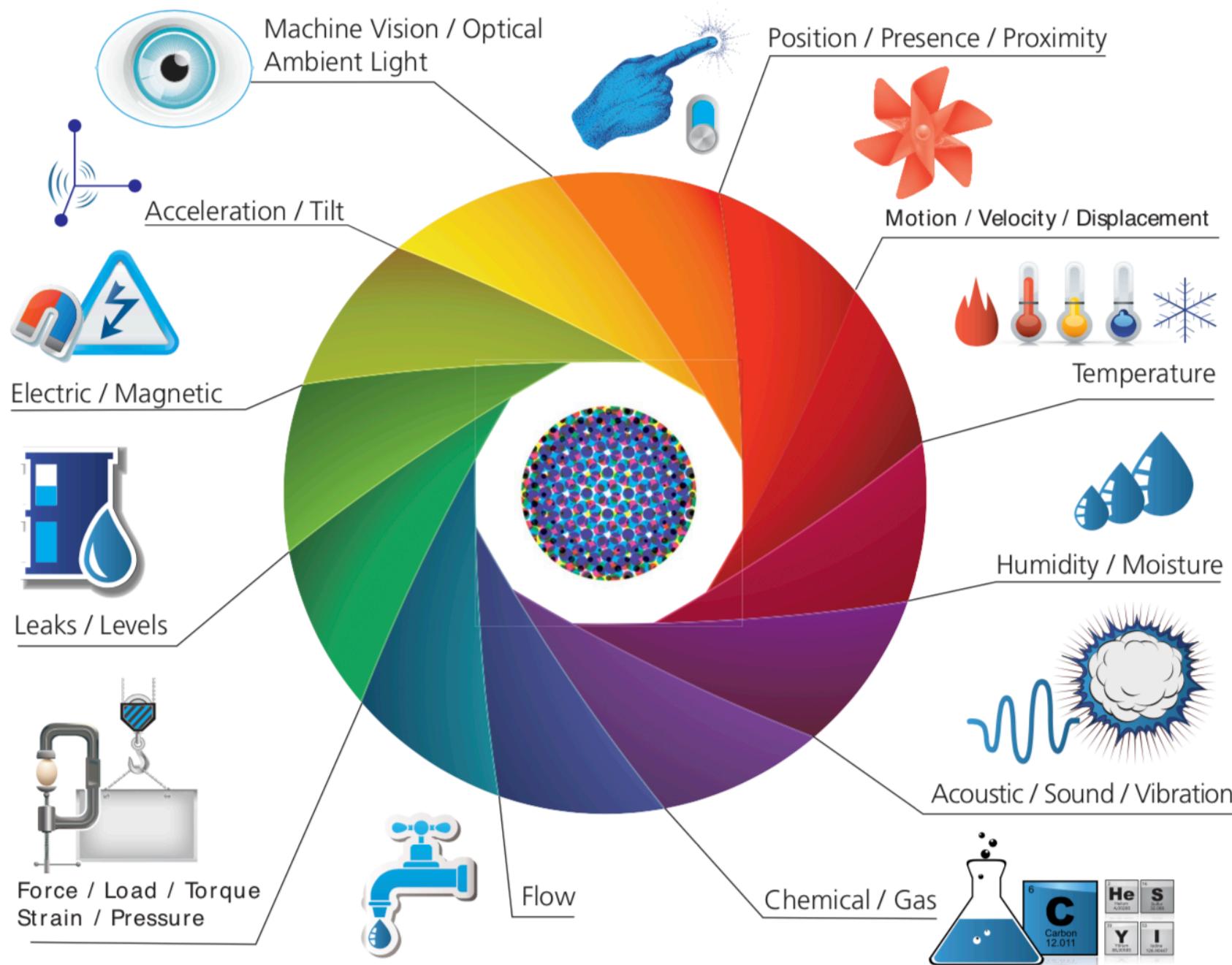
Ultrasounds to generate tactile sensations



Sensors and actuators

... and more!!!

- A **digital nervous system**
 - Orienteering capabilities using **GPS, compass, ...**
 - Sensory organs that can measure **temperature, pressure, ...**
- ... and an increasing number of **actuators**
 - **hands**
 - **a locomotor system**
 - ...

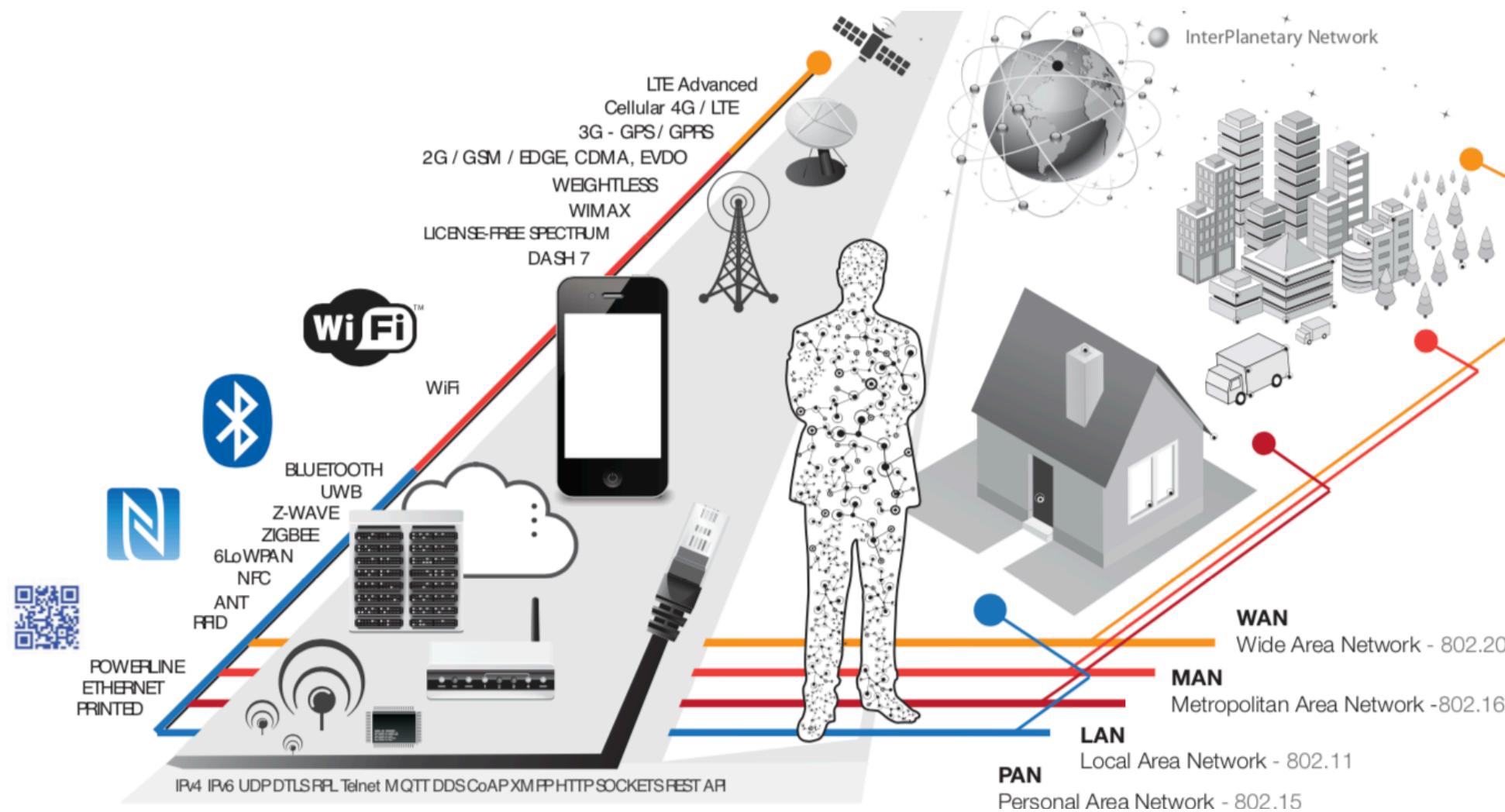


The key ingredients

1. Sensor and actuators
2. **Connectivity** (*as enabling and constraining factor, not covered in the course*)
3. Streaming Data Engineering
4. Streaming Data Science
5. People & Processes

Connectivity

To transport sensory observations and actuation signals



GPRS

1997

50 Kbps

EDGE

1998

250 Kbps

3G

2001

384 Kbps

4G

2009

150 Mbps

5G

2020

6400 Mbps



**WHAT SIZE SHOULD
YOUR WEBPAGE BE?***

* Based upon studies that have shown the maximum allowed waiting time for a loading webpage is **4 seconds**

**MAX
25
KB**

**MAX
125
KB**

**MAX
192
KB**

**MAX
75
MB**

**MAX
3,2
GB**



**HOW LONG WOULD IT TAKE TO
DOWNLOAD A 800MB MOVIE?**

**1 days
12 hours
24 minutes
32 seconds**

**0 days
7 hours
16 minutes
54 seconds**

**0 days
4 hours
44 minutes
27 seconds**

**0 days
0 hours
0 minutes
43 seconds**

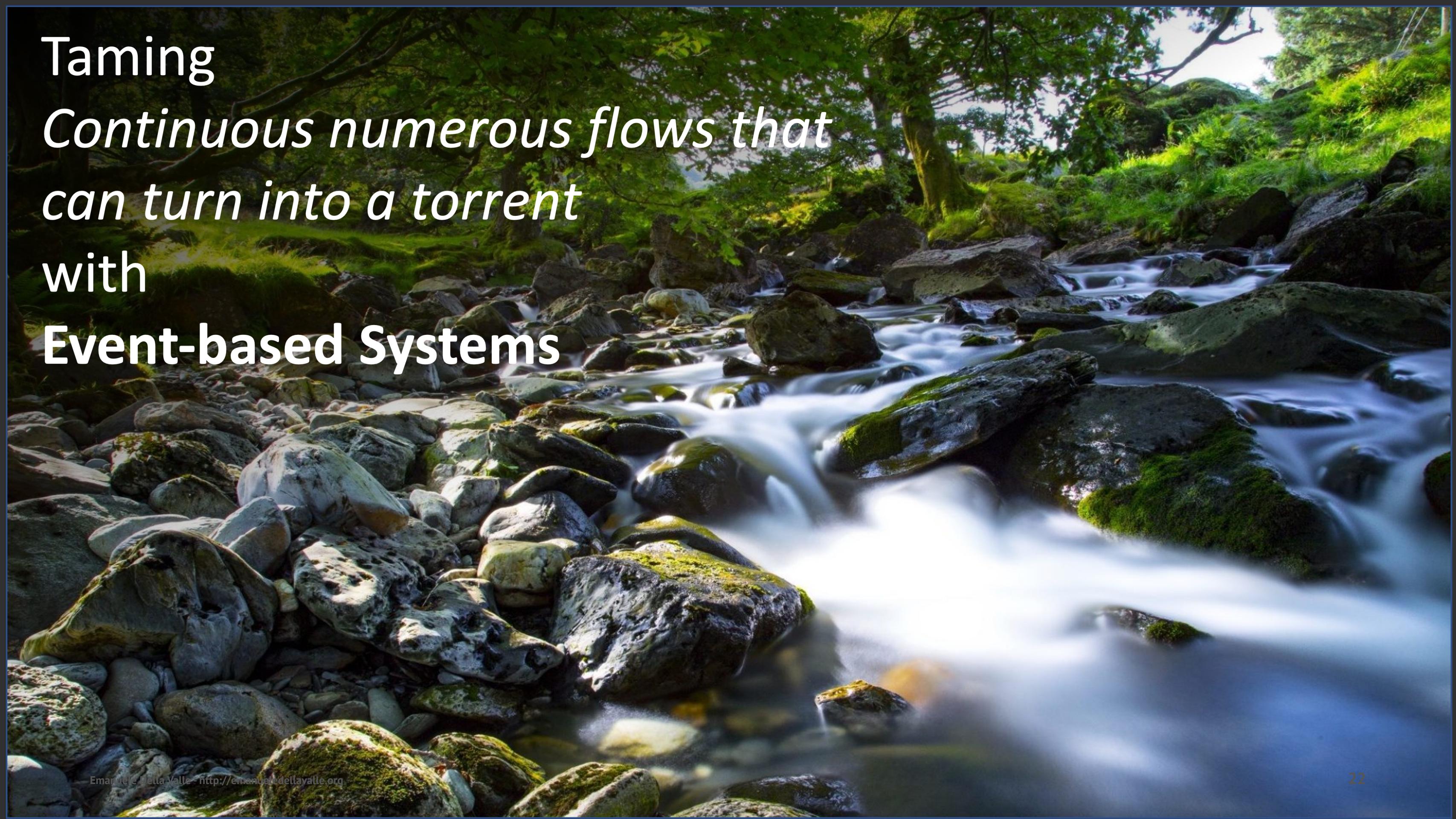
**0 days
0 hours
0 minutes
1 seconds¹⁹**

The key ingredients

1. Sensor and actuators
2. Connectivity
3. **Streaming Data Engineering** (*core topic of the course*)
4. Streaming Data Science
5. People & Processes

An aerial photograph of a river flowing through a green landscape. The river has a distinct bend, with one arm appearing dark blue/green and the other yellowish-brown. A small wooden bridge spans the river, and a small circular observation deck with people is visible on the bank. In the background, there are buildings and more greenery.

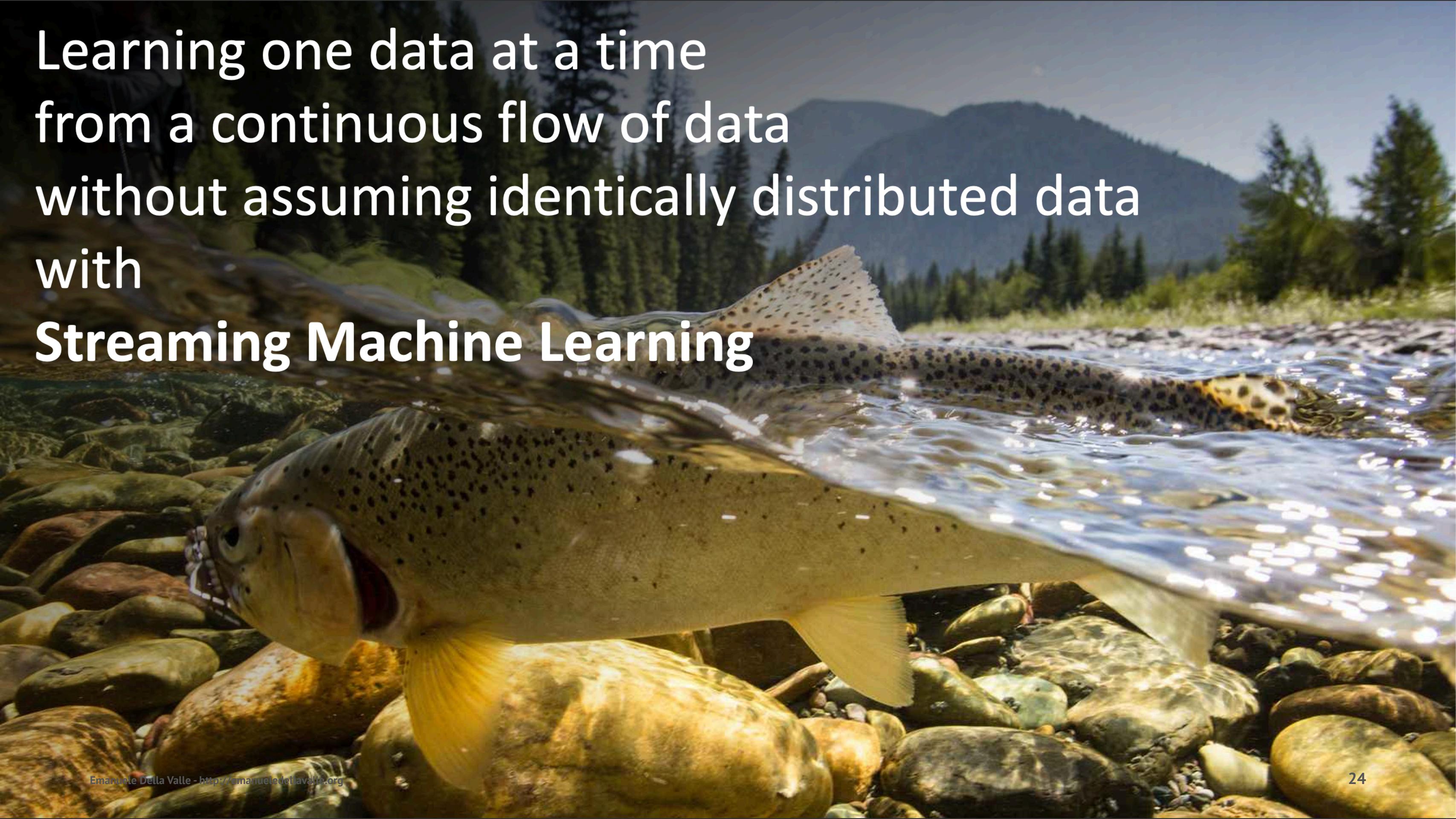
Taming
Continuous massive flows
than you cannot stop
with
Data Stream Mng. Systems



Taming
*Continuous numerous flows that
can turn into a torrent*
with
Event-based Systems

The key ingredients

1. Sensor and actuators
2. Connectivity
3. Streaming Data Engineering
4. **Streaming Data Science** (*core topic of the course*)
5. People & Processes

A close-up photograph of a large brown trout swimming in a river. The fish is oriented horizontally, facing left. It has a mottled brown and tan pattern on its body and a prominent yellowish-orange lateral stripe. The water is clear, reflecting sunlight, and the bottom consists of smooth, light-colored stones. In the background, a dense forest of green coniferous trees is visible, with a range of mountains under a clear blue sky.

Learning one data at a time
from a continuous flow of data
without assuming identically distributed data
with
Streaming Machine Learning

A scenic view of a park with a pond, a bridge, and city buildings in the background.

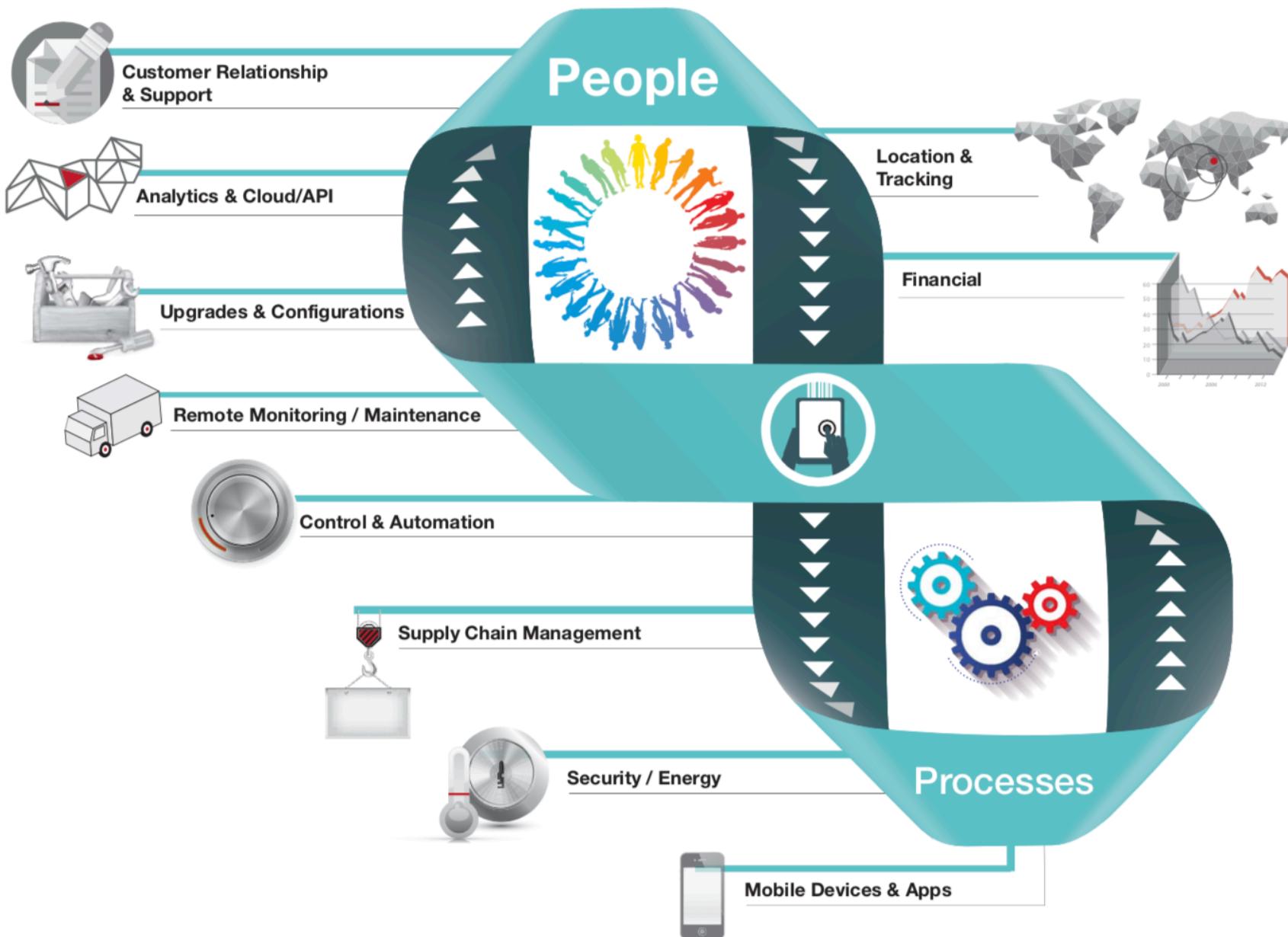
Explaining the past and forecast the future
of a continuous flow of data
without assuming data independence
with

Time Series Analytics

The key ingredients

1. Sensor and actuators
2. Connectivity
3. Streaming Data Engineering
4. Streaming Data Science
5. **People & Processes** (*to stress real-world applicability*)

People & processes



Case studies

Learning Thermostats

Save resources and money on your heating bills by adapting to your usage patterns and turning the temperature down when you're away from home.

E.g., Nest



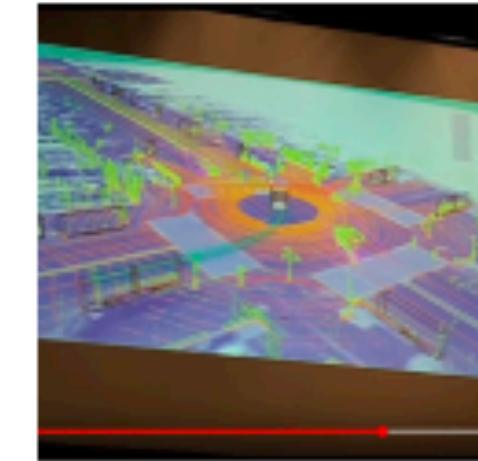
Self-driving cars

From Google Car research to Tesla autopilot and Vitibot

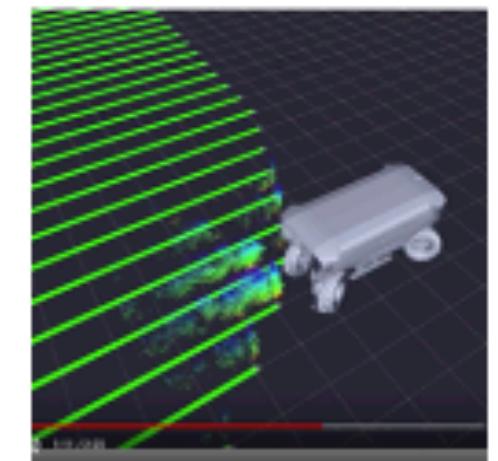
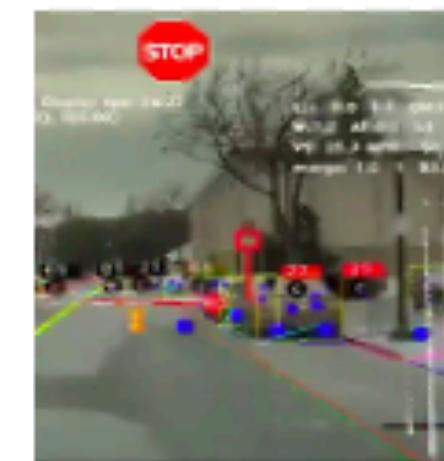
E.g., A movie I assembled

IROS 2011

International Conference on
Intelligent Robots and Systems



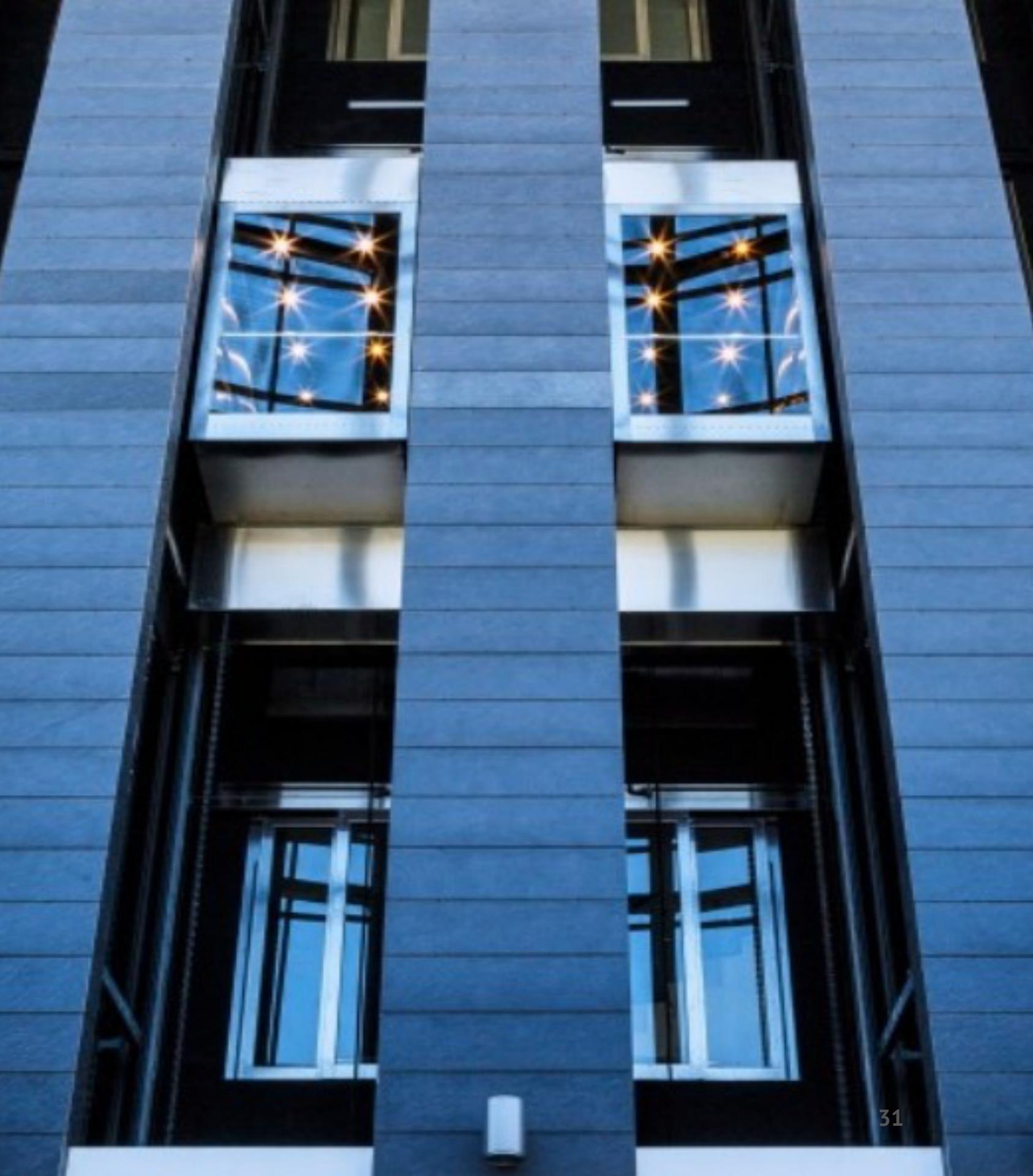
Carscoops
38.700 iscritti



Smart connected industrial products

The data collected from these products are analyzed to inform decision-making, enable operational efficiencies and continuously improve the product's performance.

E.g., thyssenkrupp's Max elevator



Equipment as a service

Pioneered in 1997 by Rolls-Royce, which lets airlines pay for their engines based on the number of flight hours. It is now a standard business model in many industrial settings.

E.g., Heller's insurance for its machines



Smart logistics

They are employing innovative technology to increase efficiency, effectiveness, and security by making logistics more environmentally sustainable, economically efficient, and capable of handling increasing traffic.

E.g., Hamburg Port Authority's Smart Port



The ultimate goal

Better decision making

Better decision making



Better decision making

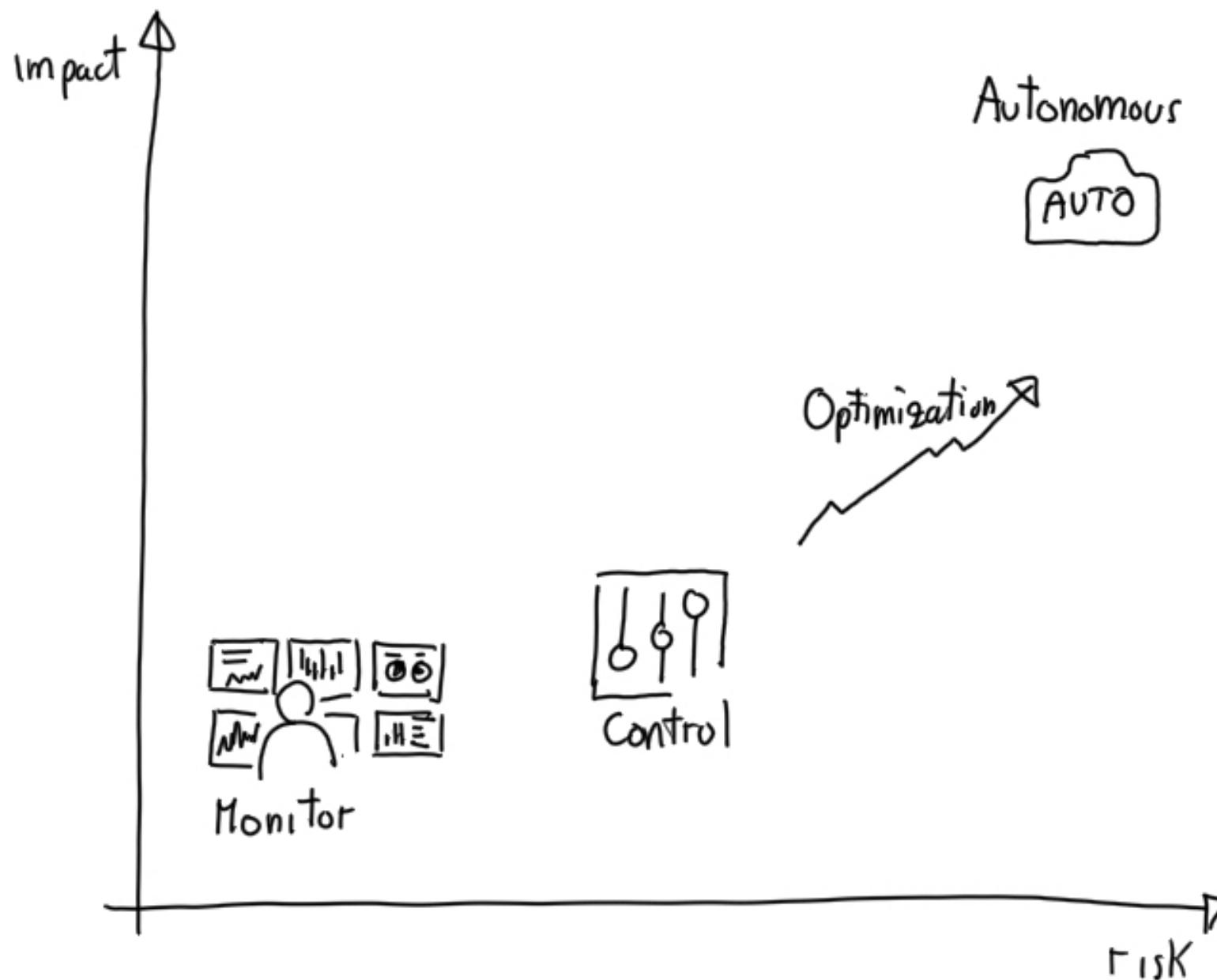


Better decision making

E.g.:

- reduce costs
- improve efficiency
- create innovative products
- new revenue streams

Streaming Data Analysis Maturity Model



Conclusions

It's about time

- **Continuous ingestion**
 - Events instead of tables of data
 - Publish / Subscribe middlewares instead of blob storage solutions
- **Continuous data preparation**
 - Stream- and Event-first data management
 - Continuous processing, no blocking operations
 - Incremental algorithms are a must
- **Reactive answers**
 - minimize latency while keeping high throughput
 - stream of answers

It's about change and temporal dependencies

- **The identically distributed assumption does not hold**
 - Traditional ML hardly applies
 - Change detection plays a key role
 - Streaming Machine Learning is a must
- **The data independence assumption does not hold**
 - Traditional ML hardly applies
 - Time Series Analytics is a must

It's at scale

- Trillions of concurrent messages
 - Billions of sensors/actuators
 - Often disconnected
-
- Whenever possible
 - **stay decentralized**
 - **do it incrementally**
 - **minimize resource footprint**
 - **do it at the edge**
 - Bring data together only if merging provides a real benefit

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