

SmartFarm: IoT Systems That Simplify & Automate Agriculture Analytics



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UC Santa Barbara



UCSB

International Conference on the Internet of Things (IoT), Oct. 17 2018



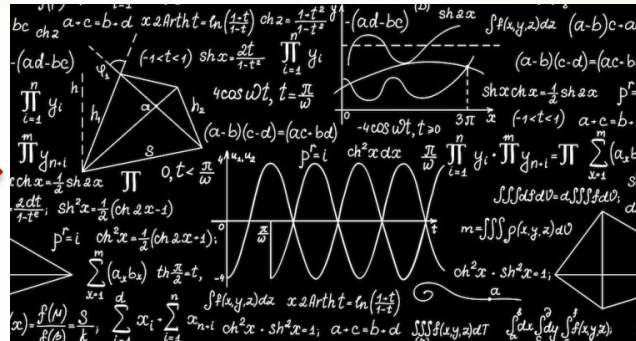


Major life insurance holder

Cloud + Data Analytics: Revolutionizing Commerce

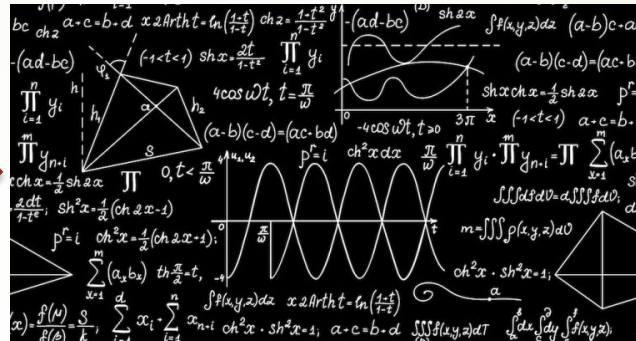


Internet Activity



Analytics: Inference and Prediction

Cloud + Data Analytics: Revolutionizing Commerce

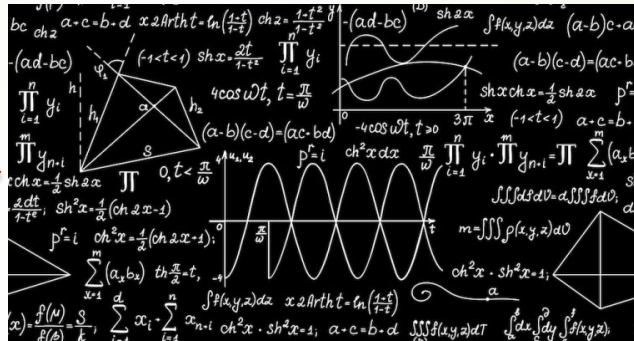


Internet Activity

Analytics: Inference and Prediction



Cloud + Data Analytics: Revolutionizing Commerce



Internet Activity

Analytics: Inference and Prediction



Cloud + Data Analytics: Revolutionizing Commerce



Internet Activity



Math and Statistics (Code!)

Analytics: Inference and Prediction

- *What will you buy?*
- *When will you buy it?*
- *What will you pay?*



What Else Can We Revolutionize With It?

- To solve a *very hard, impending* problem: **feeding the planet**
 - A complex system: Food-Energy-Water nexus
 - Global: *821M people today are undernourished
 - *6.5M children live in food-insecure households in US
 - *9.8B people to feed by 2050
- Just how complex is it?



* <http://www.cdfa.ca.gov/statistics/>
<https://www.ers.usda.gov/topics/food-nutrition-assistance/food-security-in-the-us/key-statistics-graphics.aspx>
<http://www.fao.org/state-of-food-security-nutrition/en/> FAO 2018
<https://www.un.org/development/desa/en/news/population/world-population-prospects-2017.html>

Critical Needs & Complex Challenges Surrounding Food Production



The world needs more **food** for a growing population.
We use 70% of fresh **water** for growing food.



We lose ~40% of the food we produce to **spoilage**



30% of **global energy** is used to produce food
22% of **greenhouse gases** come from agriculture



Worker shortages and high **labor costs**

Invasive pests and disease threaten production



Sources: USDA, UN FAO

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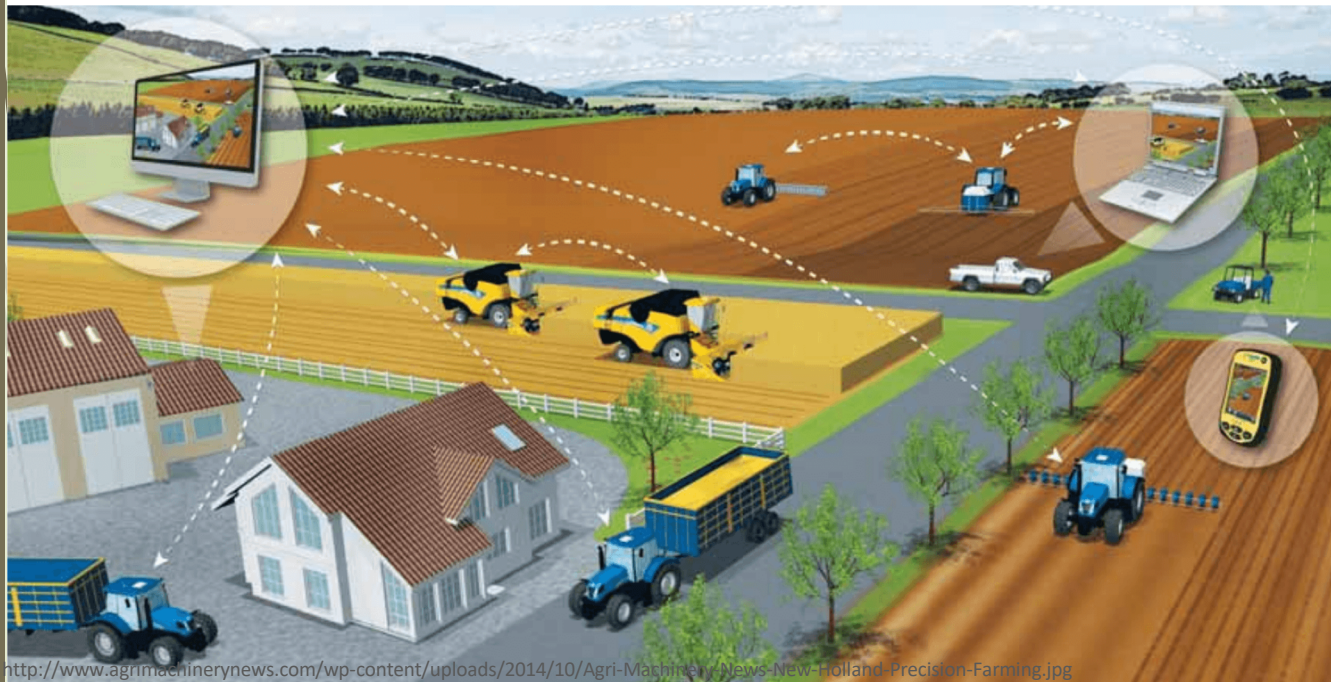
What Else Can We Revolutionize With Cloud+Analytics?

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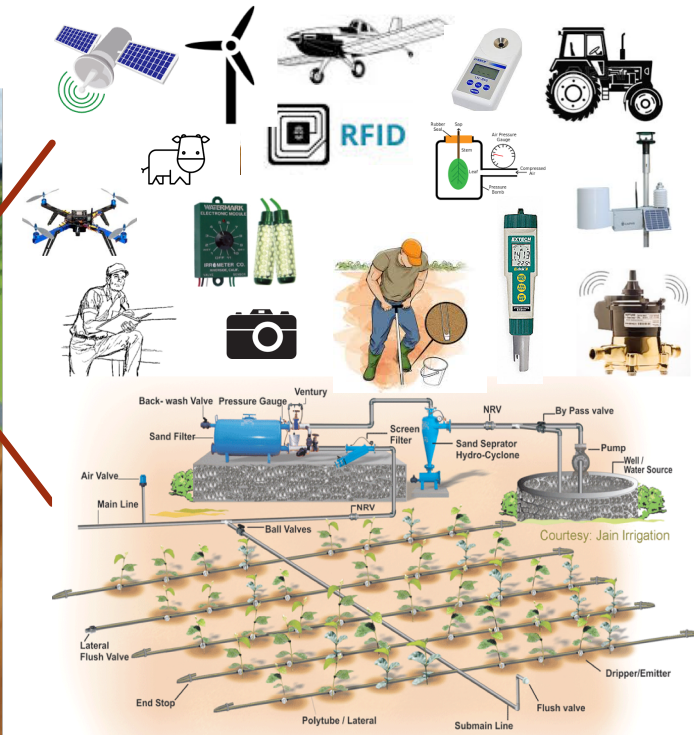
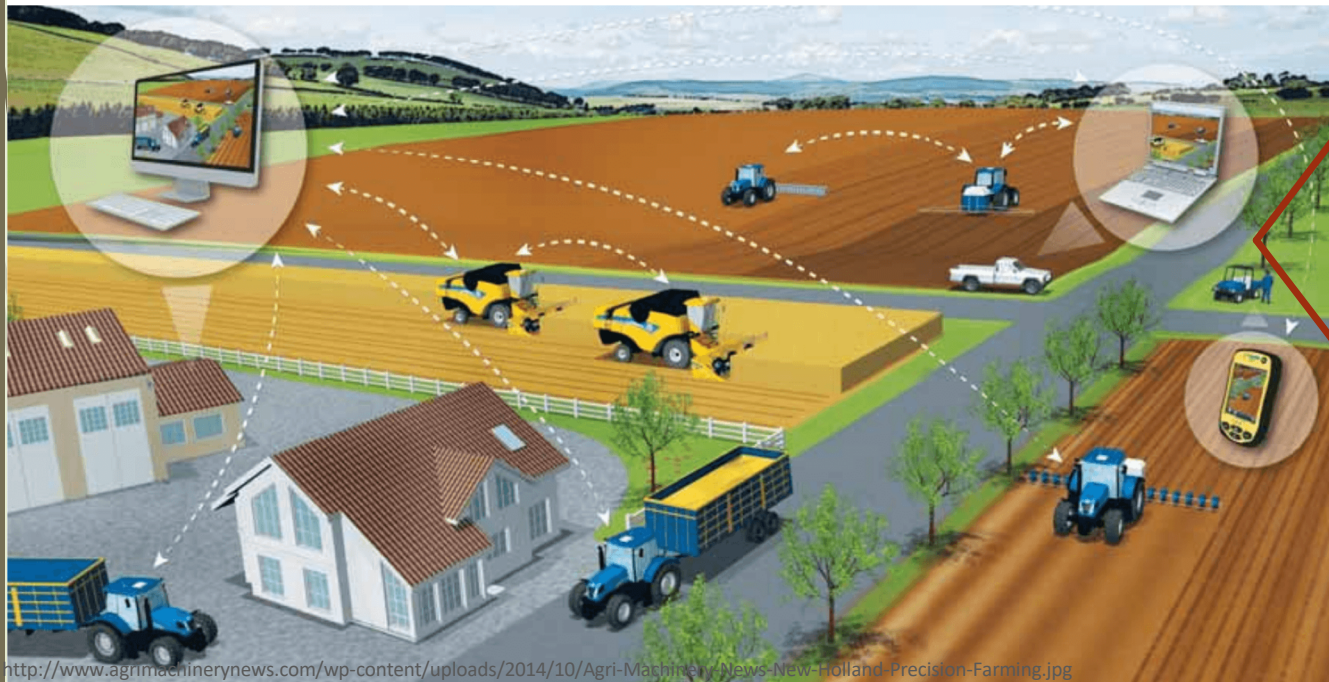


- **The Cloud+Data Analytics boom has not come to Ag, despite**
 - Maturity of big data and machine learning technologies available
 - Vast amounts of data surrounding food production, processing & the crop lifecycle
 - Weather, historical records, sensors, images (NDVI/thermal), ...
 - At the farm, community, county, region, state, nation, global level

IoT to Bridge the Gap

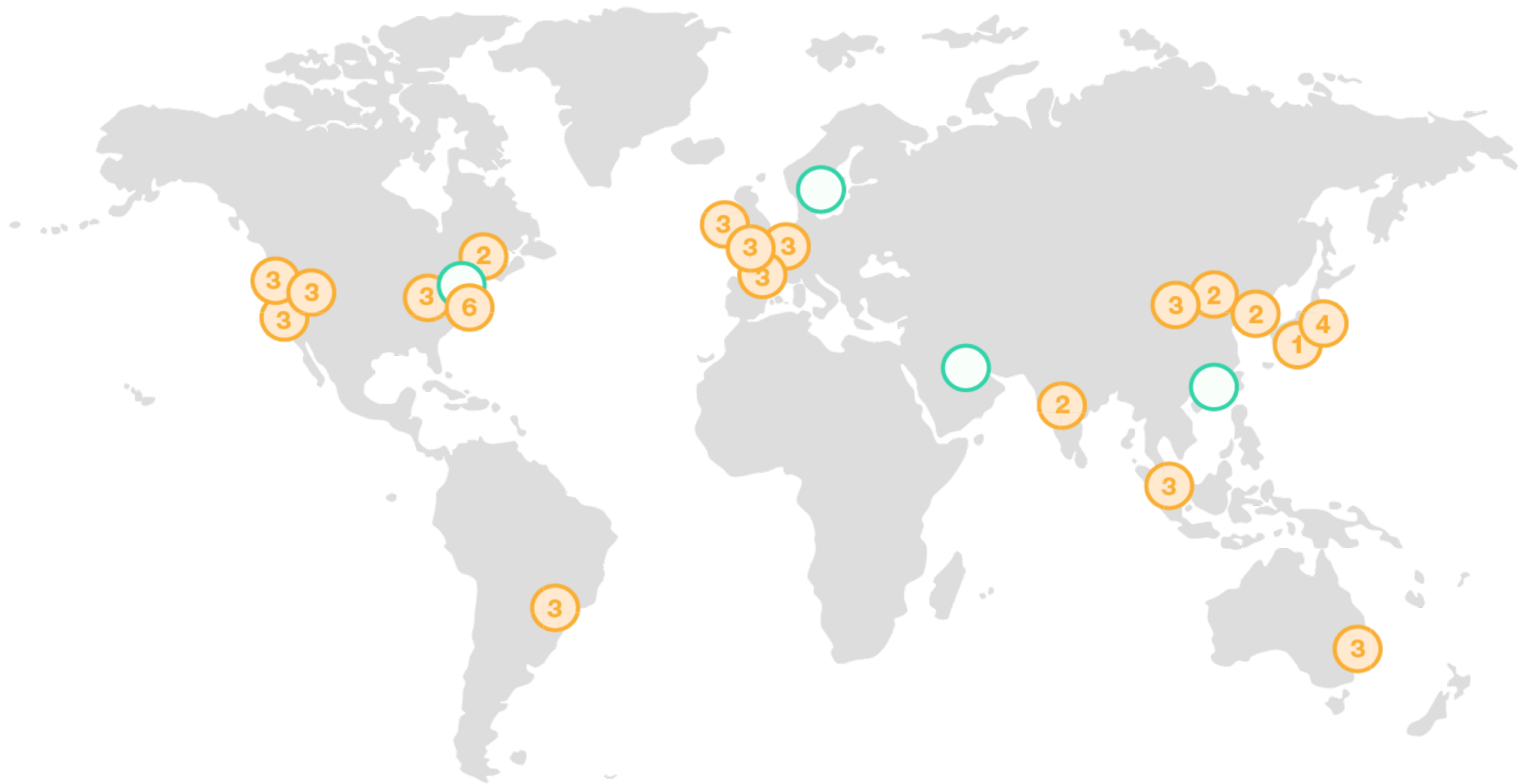


IoT to Bridge the Gap



How Do We Get There?

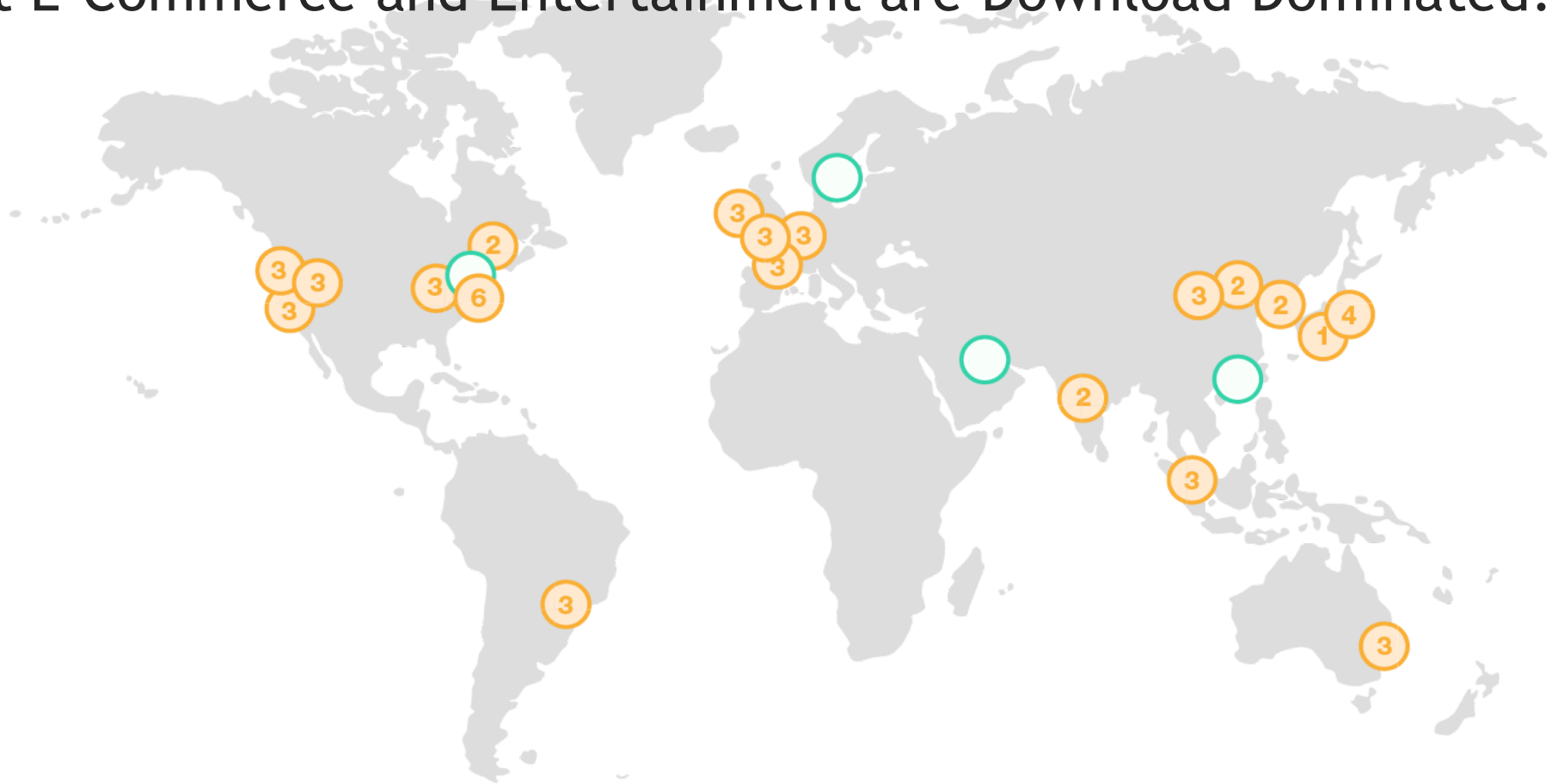
How Do We Get There? aka What Would Amazon Do?



Amazon Cloud Infrastructure

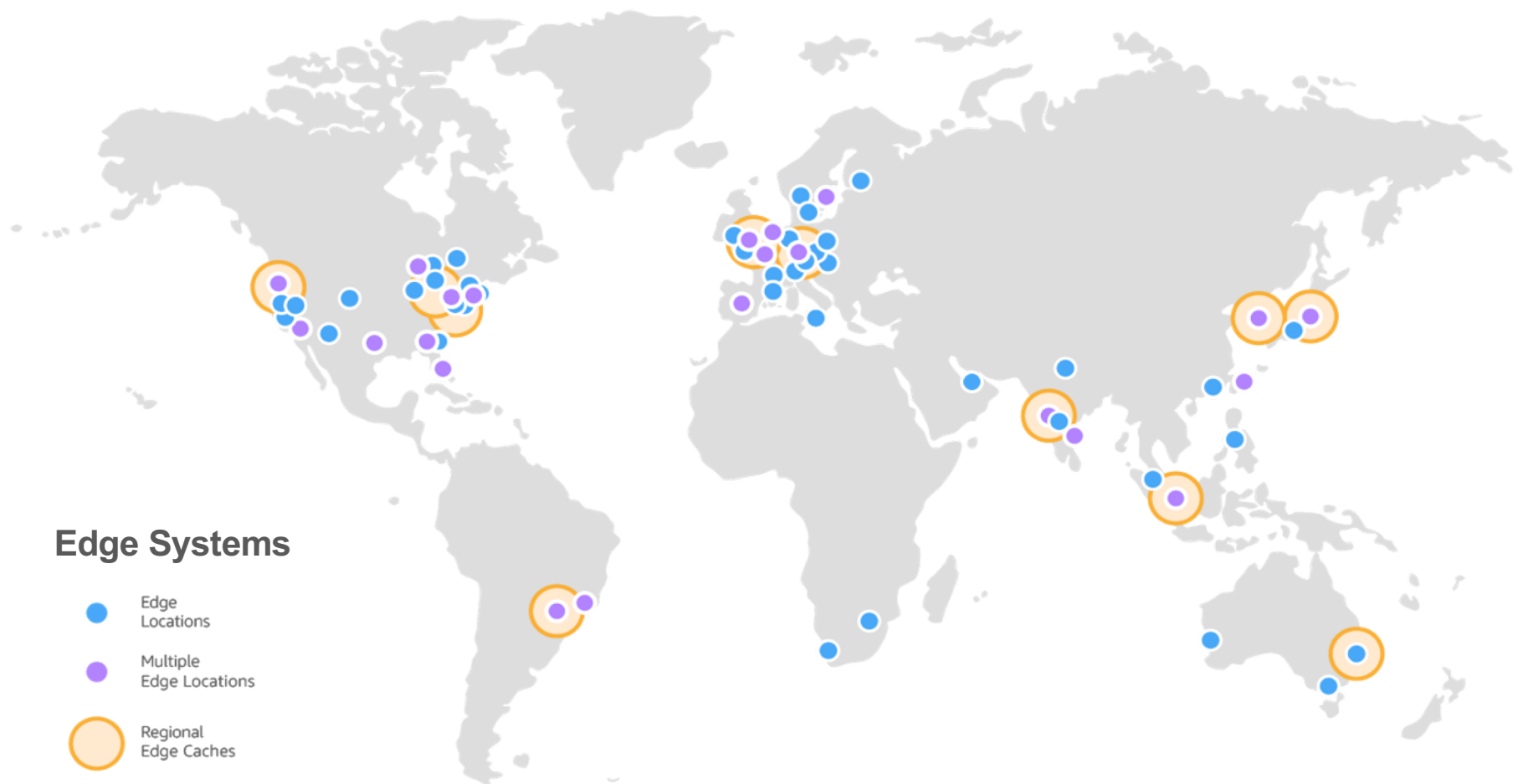
How Do We Get There? aka What Would Amazon Do?

But E-Commerce and Entertainment are Download Dominated!



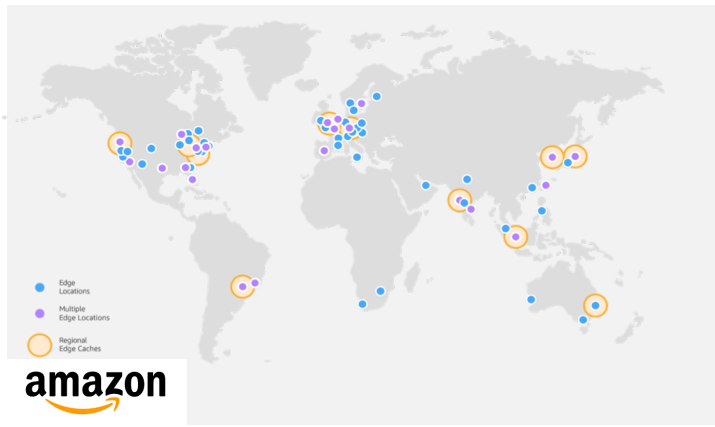
Amazon Cloud Infrastructure

The "Cloud" is More Distributed Than You Think!

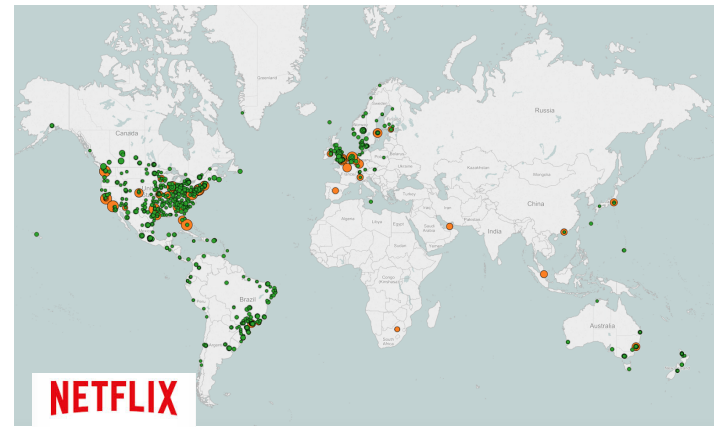


Amazon Content Delivery Infrastructure

Content (Download) Delivery -- is the (Edge) Cloud



AWS 2018



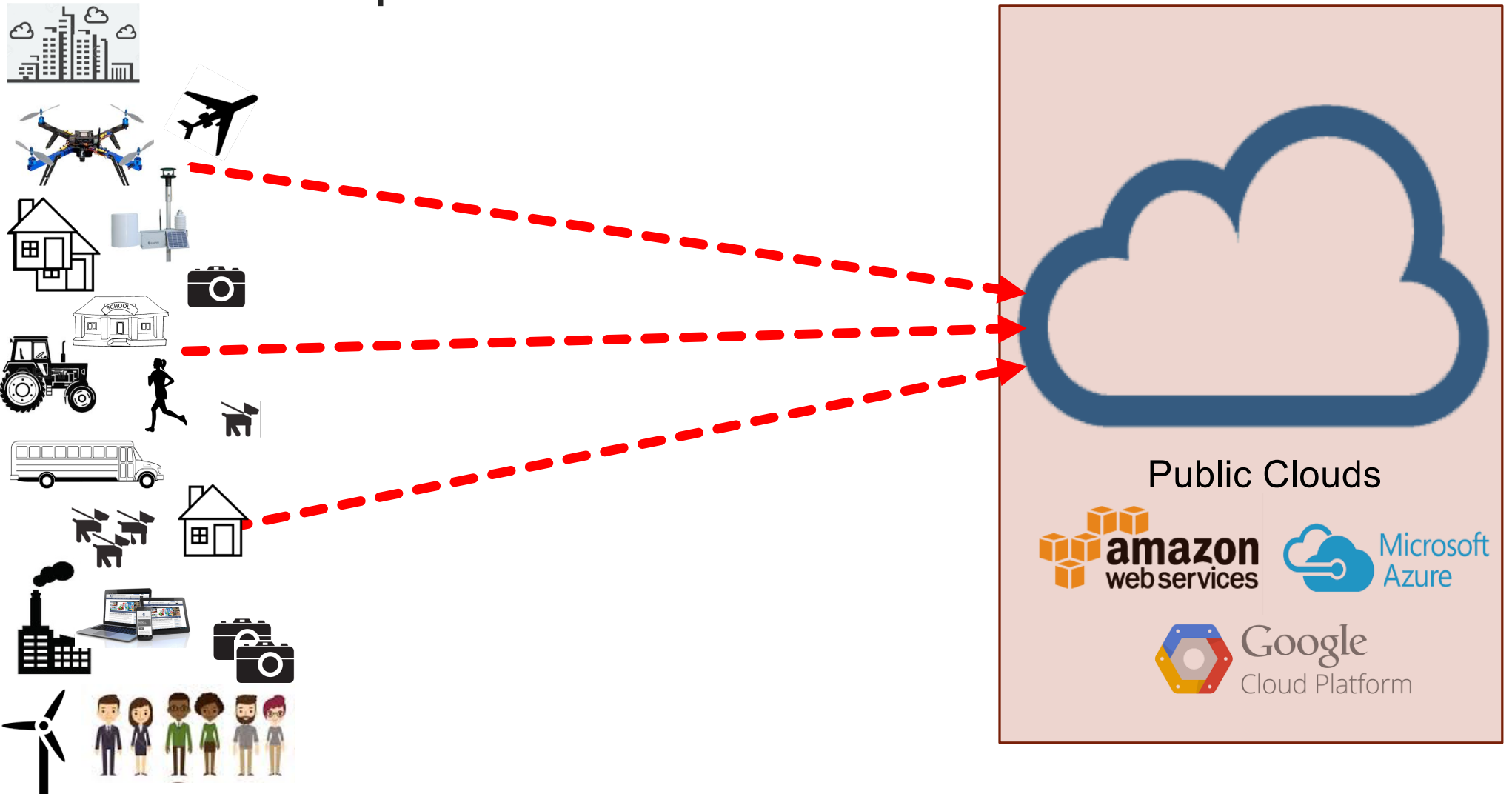
Netflix 2018



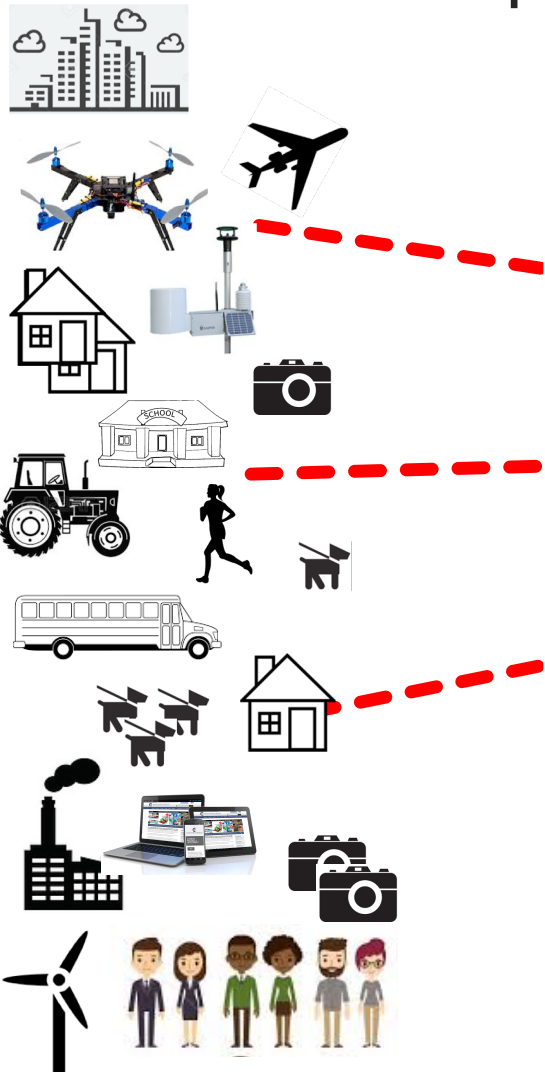
Adhikari et al, HotMD'11



IoT is Write/Upload Dominated

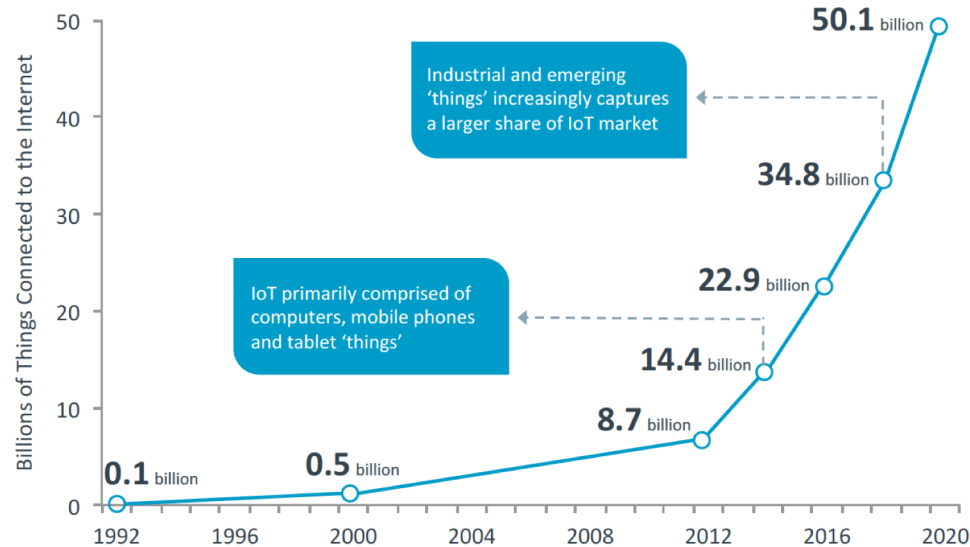


IoT is Write/Upload Dominated



Projecting the 'Things' Behind the Internet of Things

From 2014-2020, IoT grows at an annual compound rate of 23.1% CAGR



CompTIA

Sources: Group SJR | Cisco | CompTIA

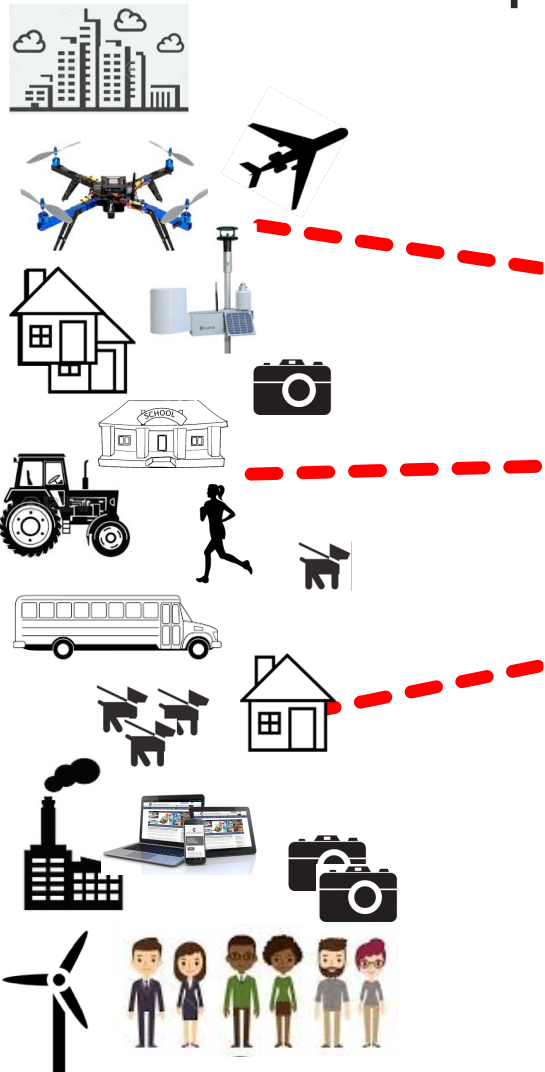
Public Clouds

Amazon
Services

Microsoft
Azure

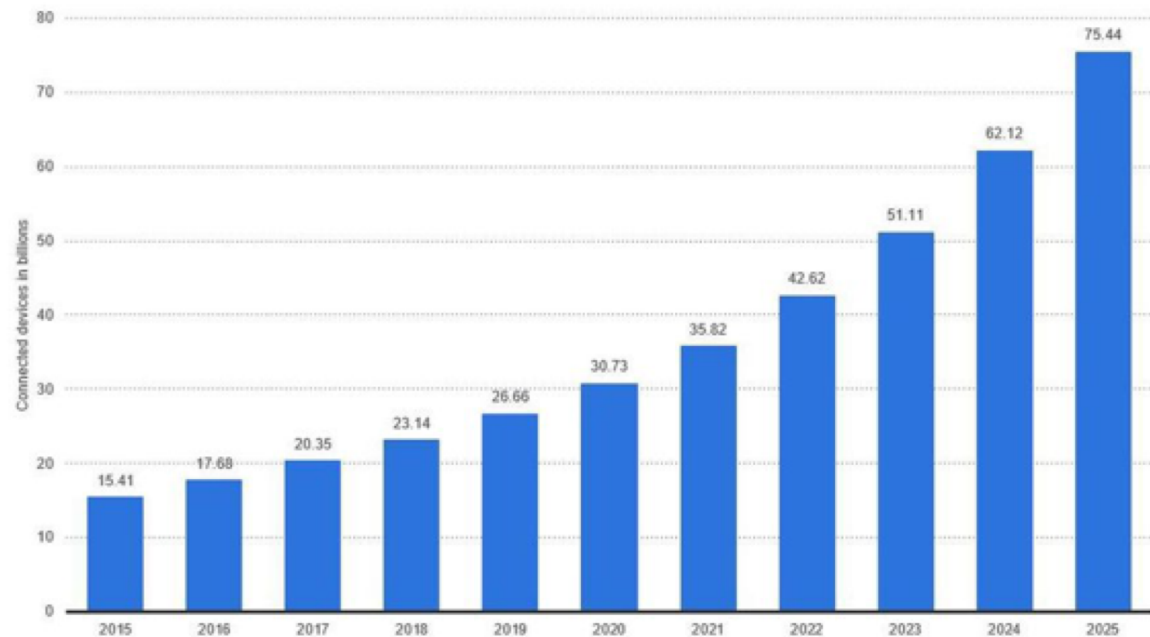
Google
Cloud Platform

IoT is Write/Upload Dominated



Internet of Things - number of connected devices worldwide 2015-2025

Internet of Things (IoT) connected devices installed base worldwide from 2015 to 2025 (in billions)



statista

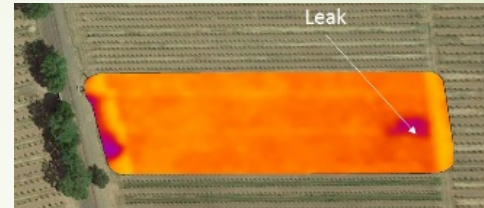
Microsoft
Azure

Key Differences Between Farm & E-Commerce

- Many more "things" (1T) than people (3.5B) connected
- Write not *read* dominated
- Decision-making, actuation, & control is **local**
- Farmers must maintain ownership and control over their data
- Most farms are **not** well connected

Farm vs E-Commerce/Entertainment

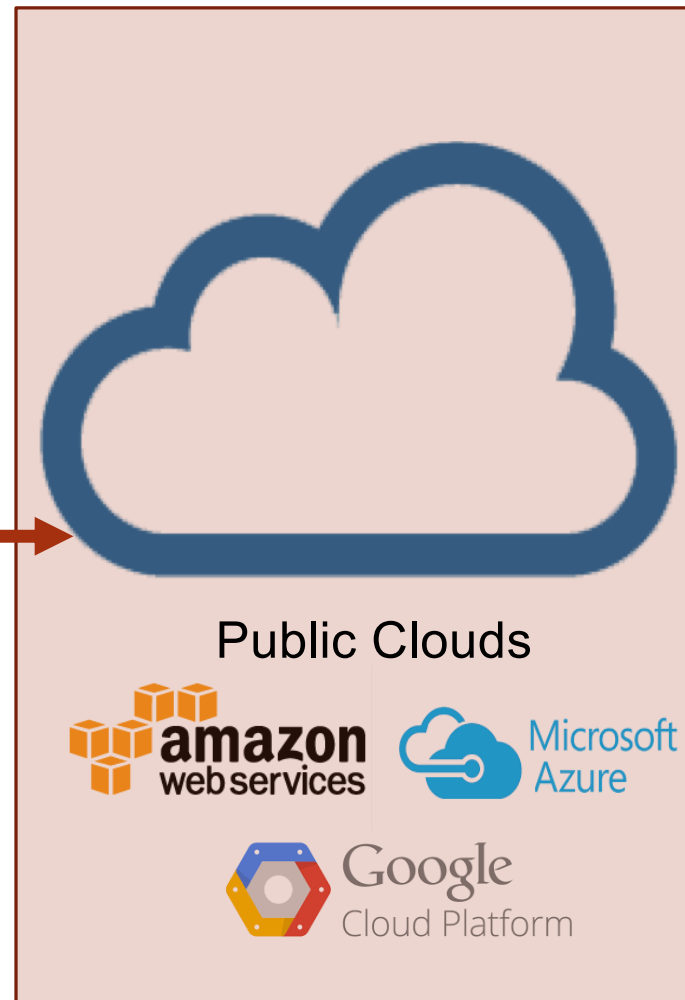
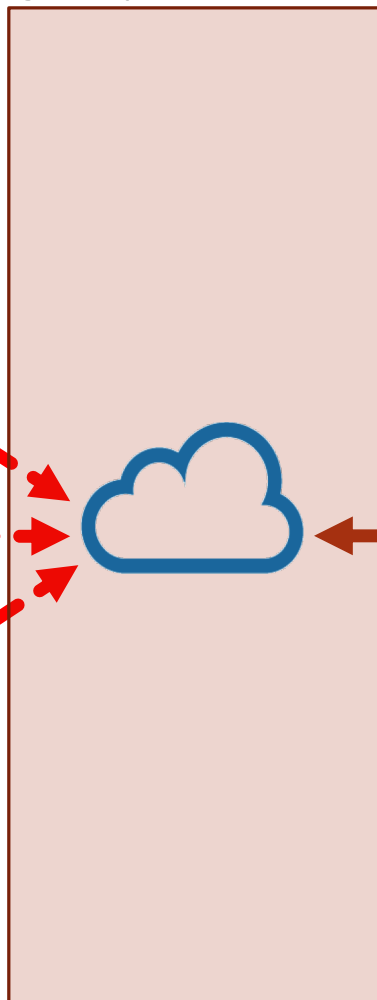
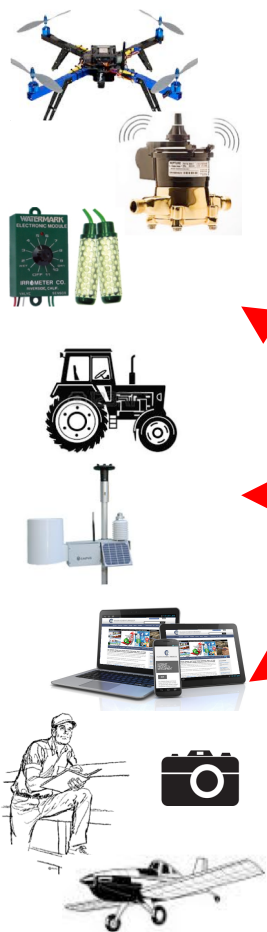
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- **Similarity:** Latency matters



Devices

Edge Systems?

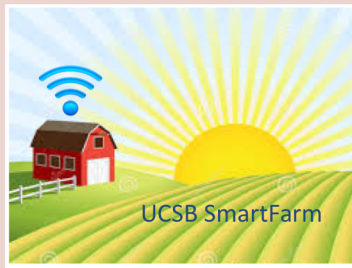
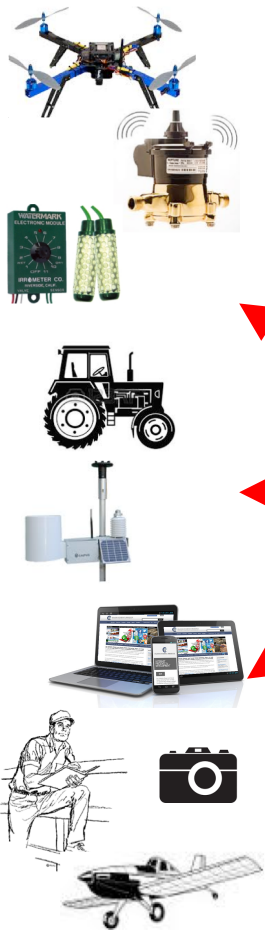
Public Cloud Tier



Devices

Edge Tier

Public Cloud Tier



Edge Clouds

Service Distribution
Network For Data
Analytics



Public Clouds



Devices

Edge Tier

Public Cloud Tier



Edge Clouds

Service Distribution
Network For Data
Analytics



Public Clouds



amazon
web services



Microsoft
Azure



Google
Cloud Platform

Data appliance

- Self managing, fault tolerant
- Hardened for hostile environments
- API-compatible with public clouds
- Variety of analytics (multi-analytics)

Devices

Edge Tier

Public Cloud Tier



Edge Clouds

Service Distribution
Network For Data
Analytics

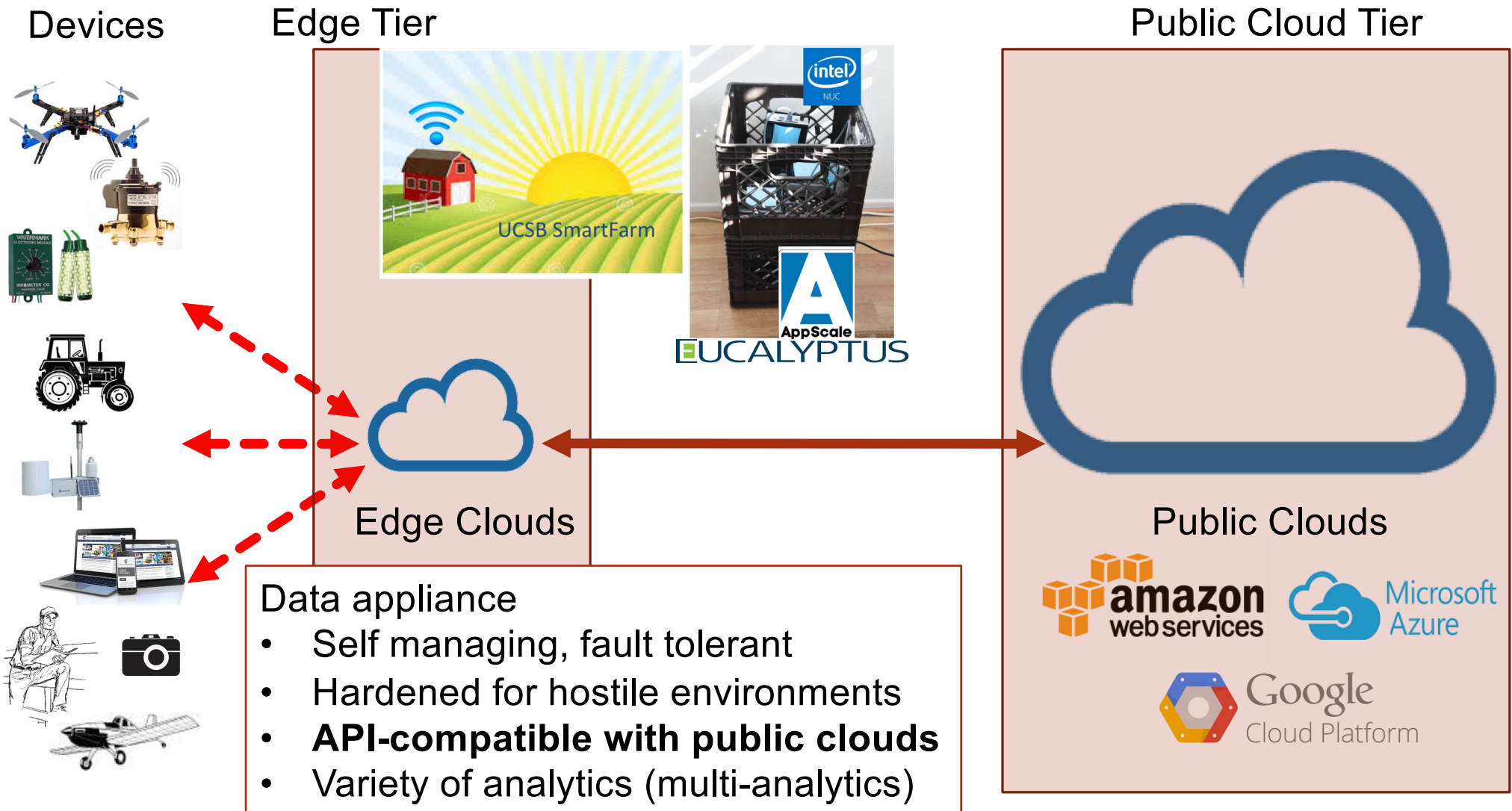


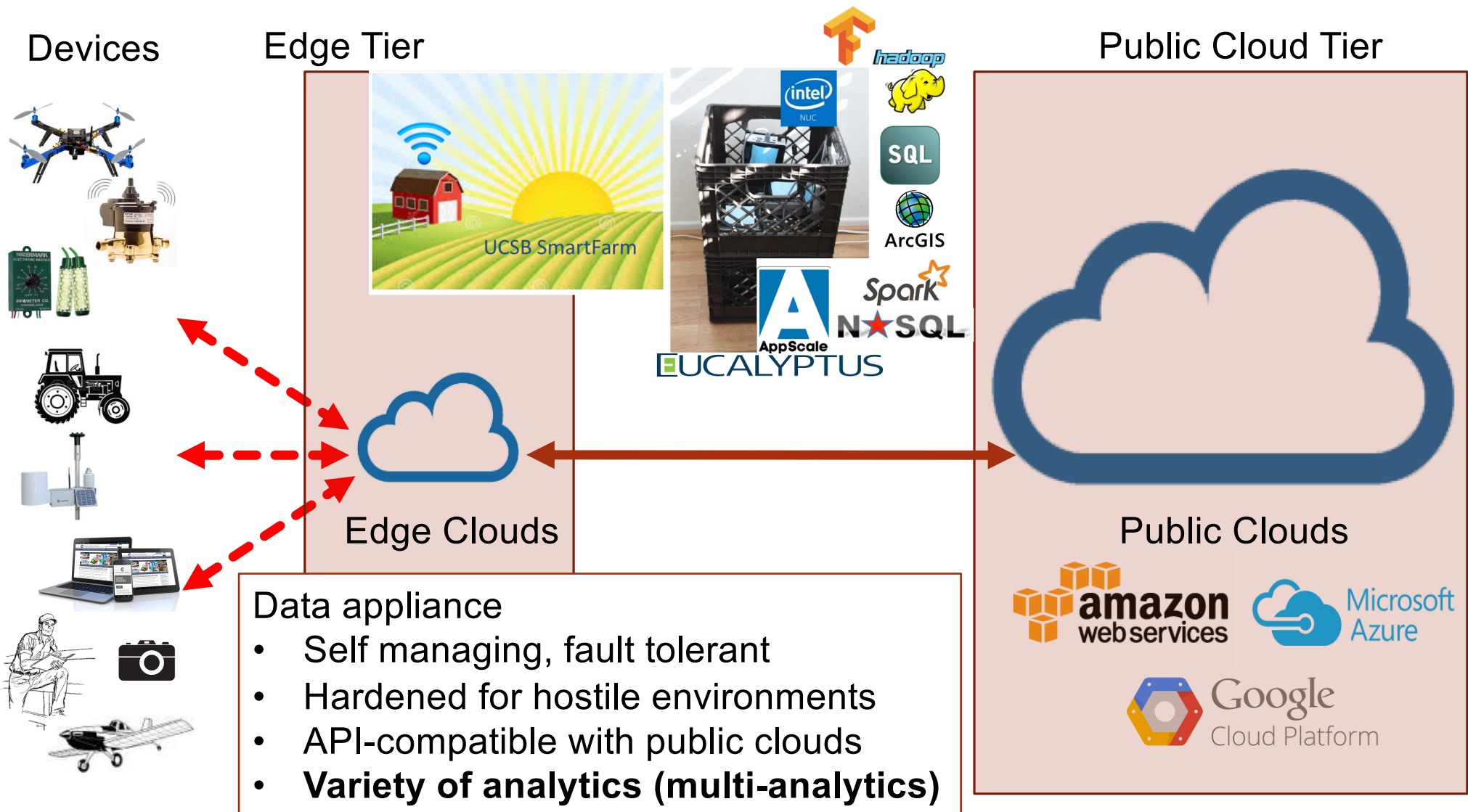
Public Clouds

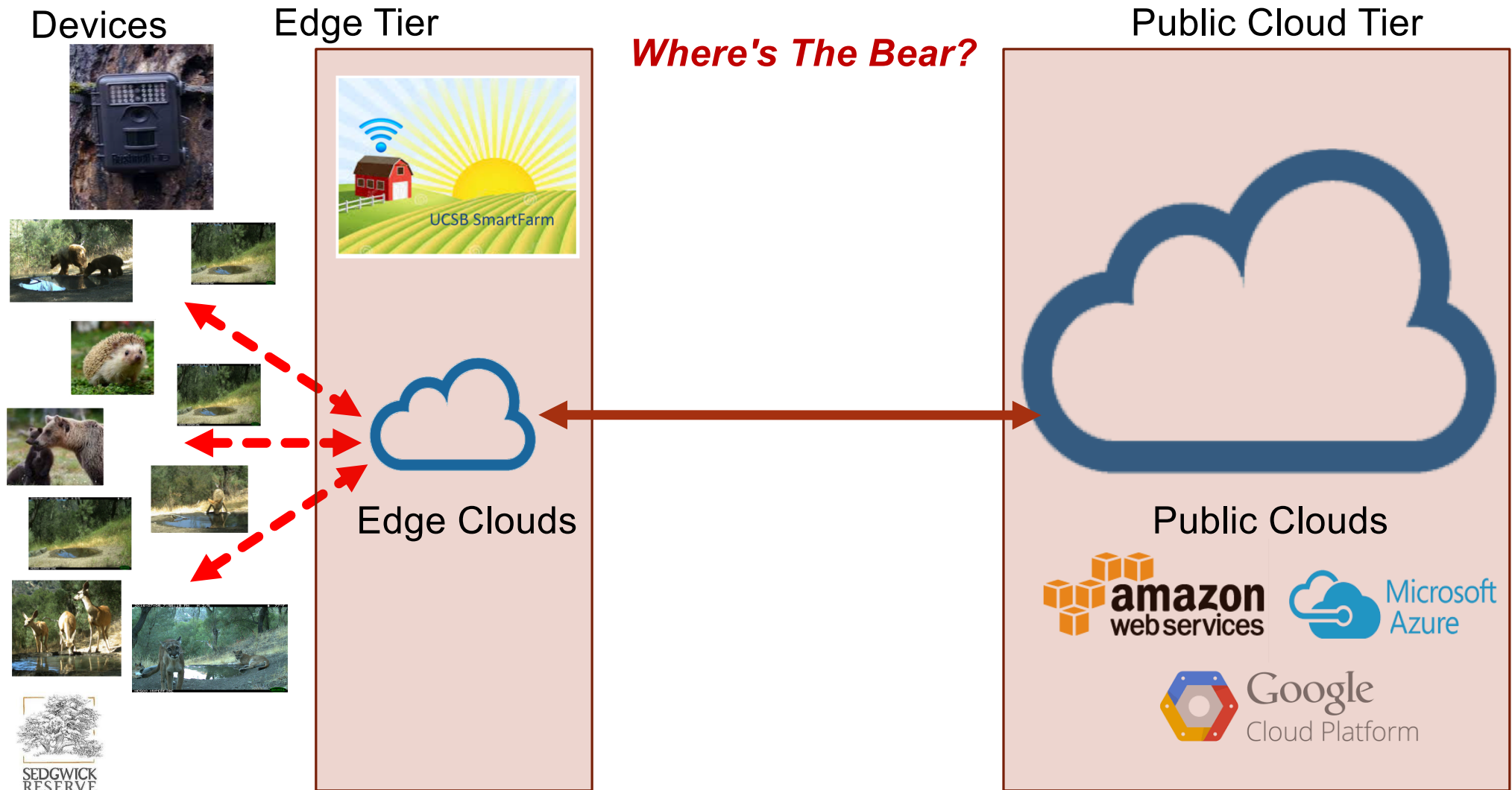


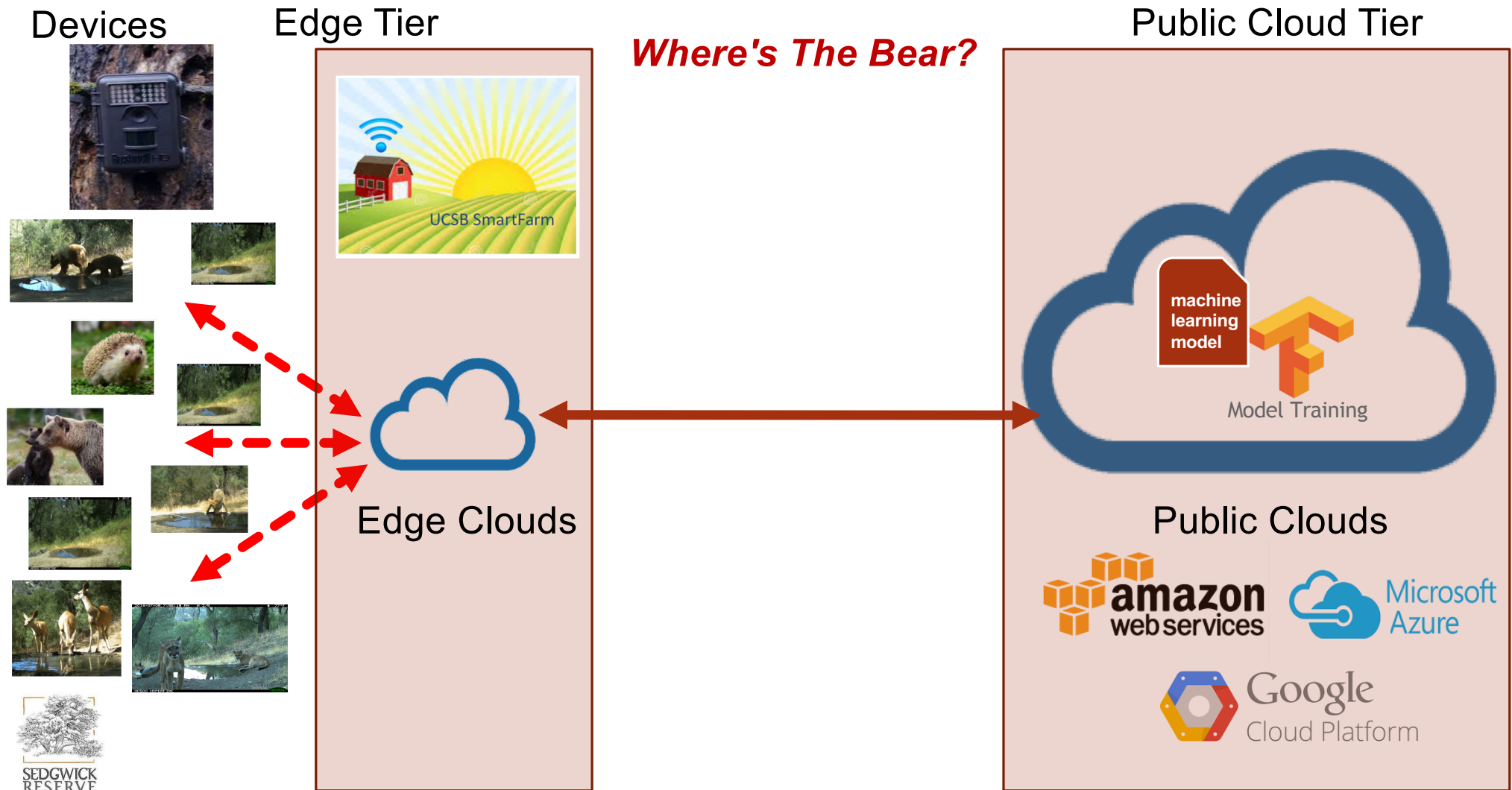
Data appliance

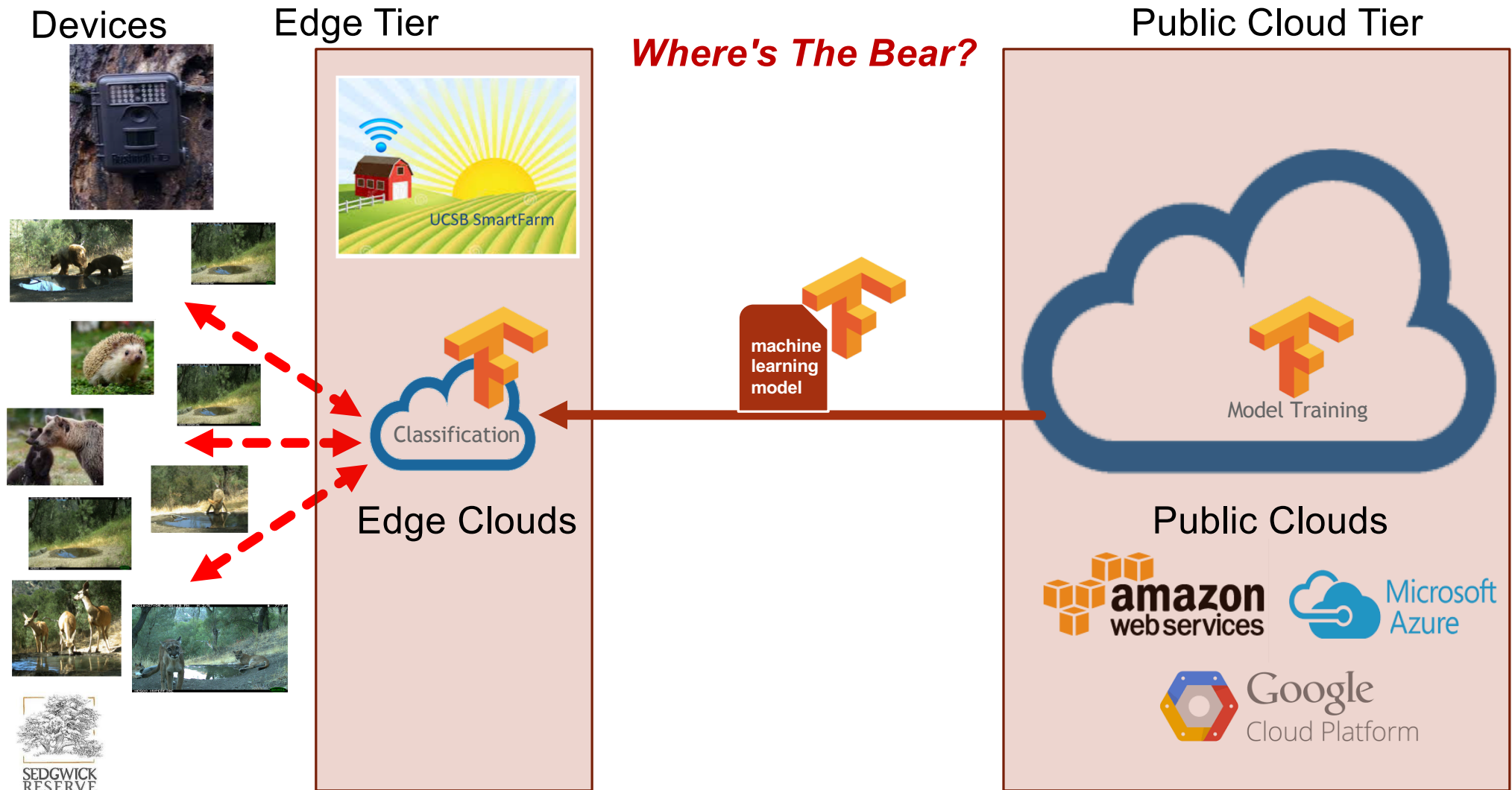
- Self managing, fault tolerant
- Hardened for hostile environments
- API-compatible with public clouds
- Variety of analytics (multi-analytics)

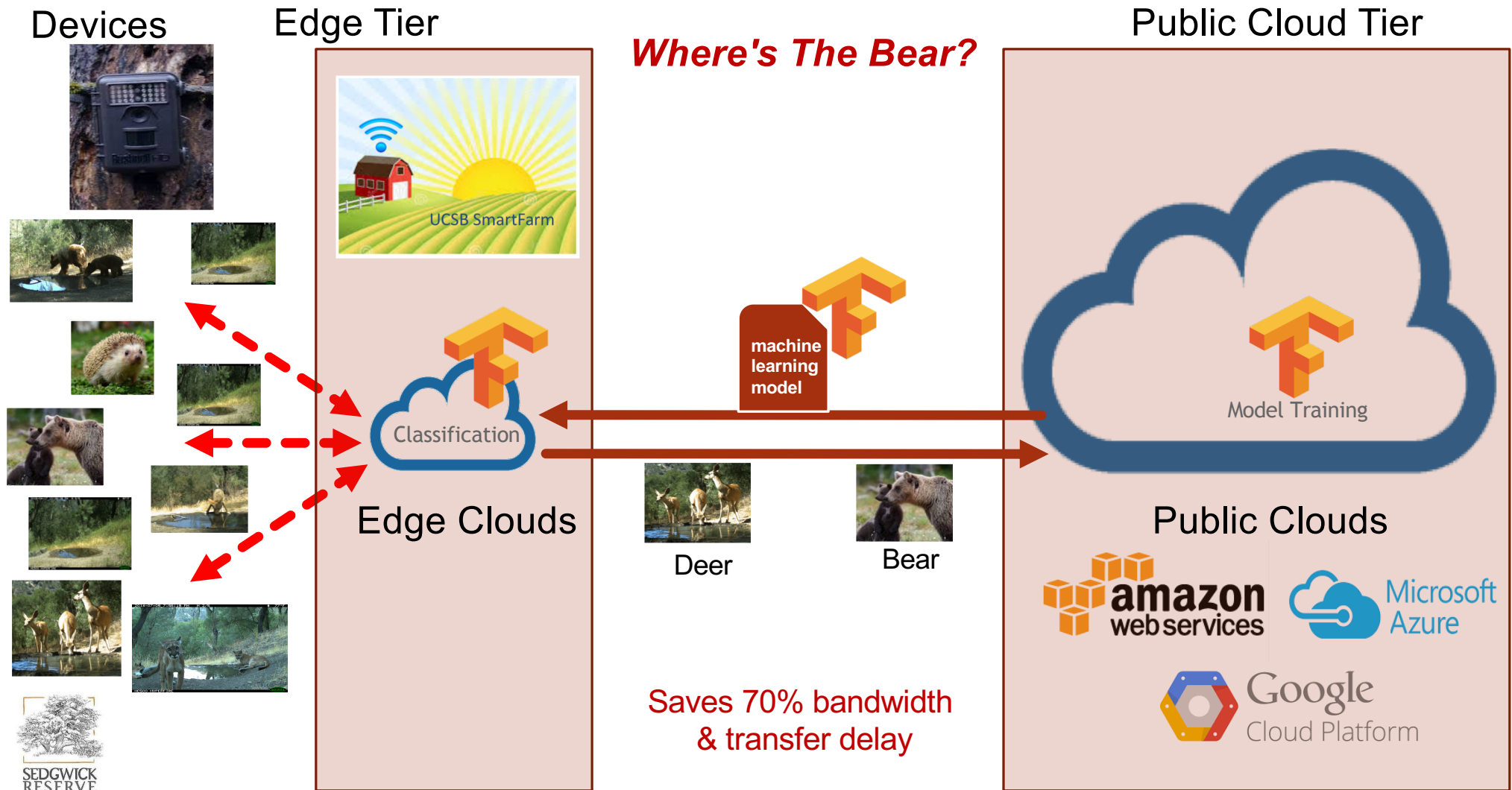


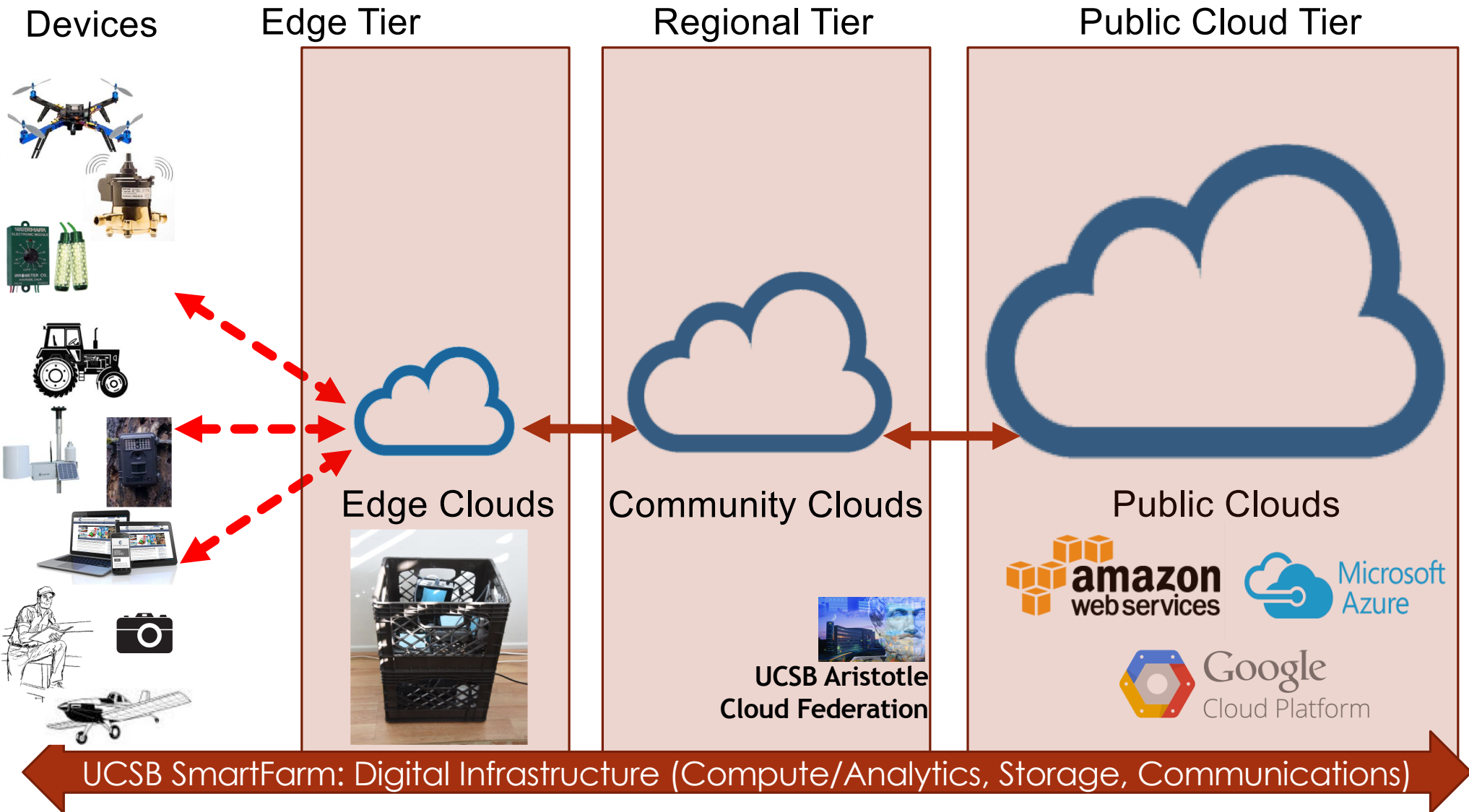


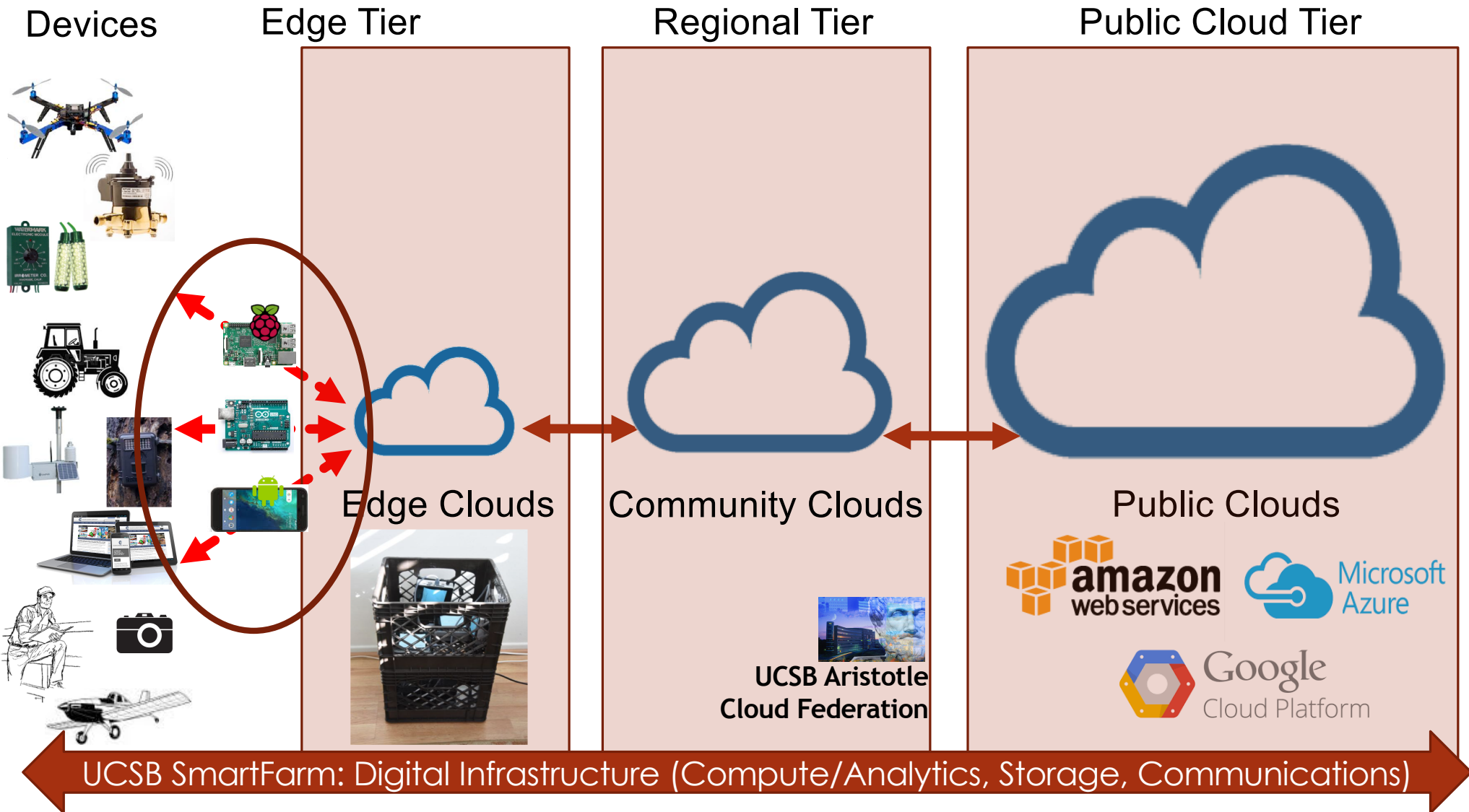






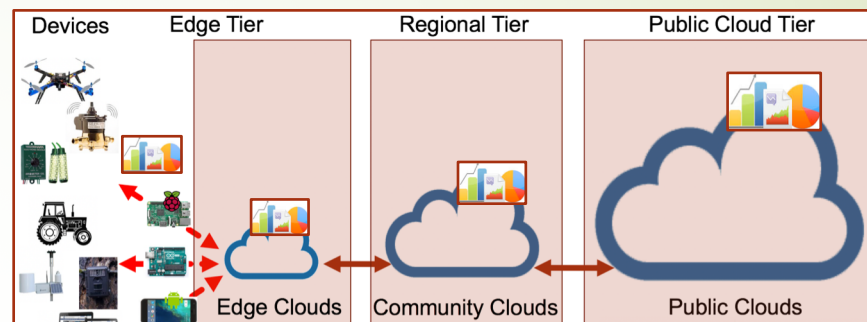






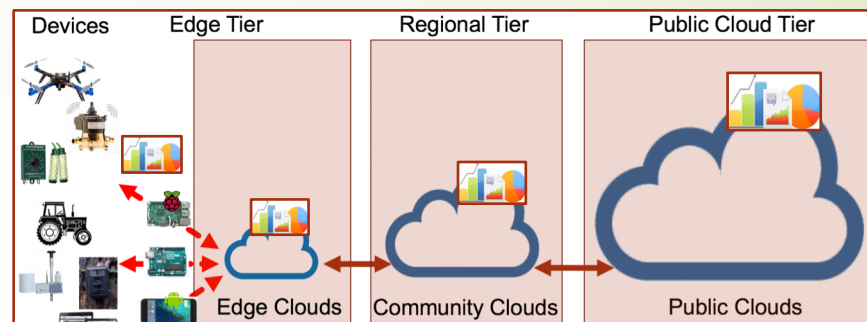
Write-Once Run-Anywhere for Multi-Tier IoT Apps

- Portable across
 - Low-level, small footprint
 - Low latency response
- Easy to develop/deploy distributed applications (novices & experts alike)
- Tailored IoT application use cases
 - Data driven: *event-based*
 - Multi-analytics
 - Executable on resource constrained or resource rich systems
 - Updatable, debuggable, maintainable



Write-Once Run-Anywhere for Multi-Tier IoT Apps

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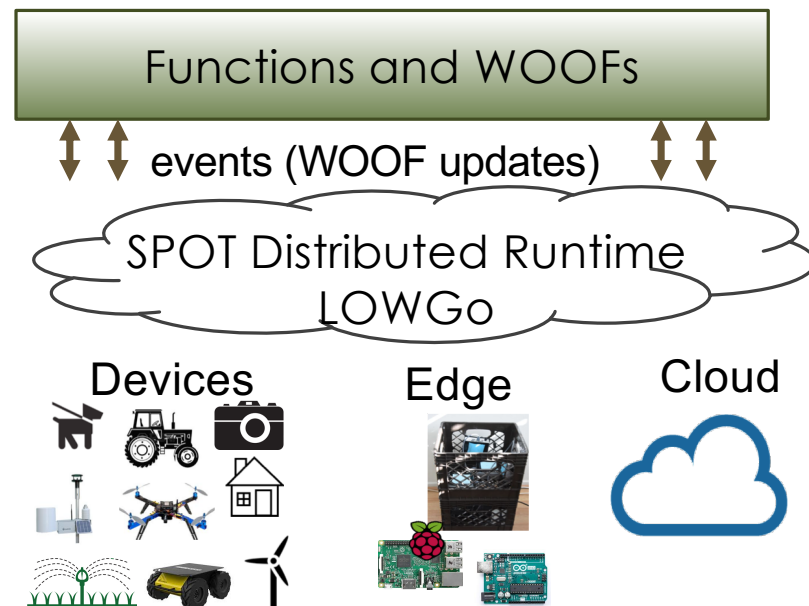
Serverless / Functions-as-a-Service (FaaS)
programming model

Platform = Language + Infrastructure

Write-Once Run-Anywhere for Multi-Tier IoT Apps

➤ Serverless Platform of Things (SPOT)

- Program = Simple, *event-triggered functions*
 - Portable across IoT systems (all tiers)
- Open source, self-service, easy to use

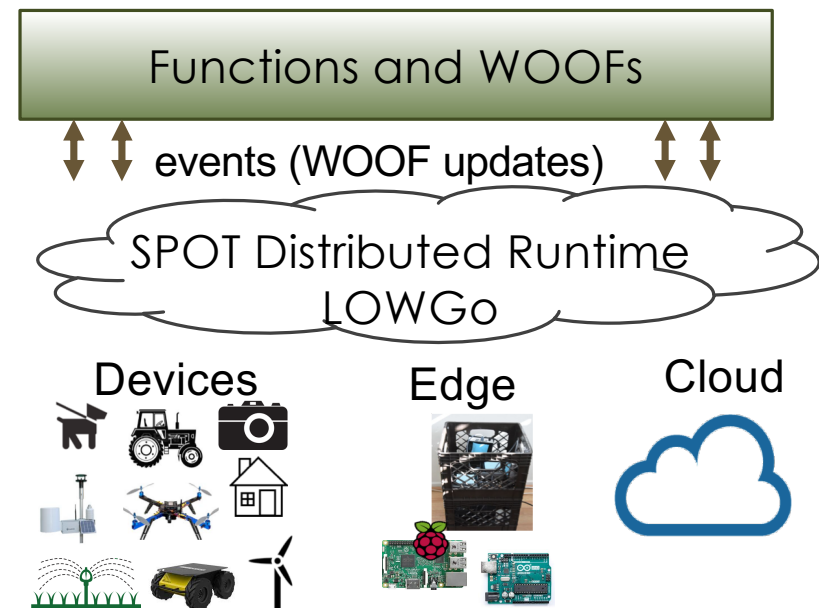


Write-Once Run-Anywhere for Multi-Tier IoT Apps



Serverless Platform of Things (SPOT)

- Program = Simple, *event-triggered functions*
 - Portable across IoT systems (all tiers)
 - Leverages Linux containers for deployment
- Open source, self-service, easy to use
- Wide-area Objects of Functions (WOOFs)
 - Append-only, persistent data structure
 - Event = object update
 - Tracked via causally ordered, distributed log (LOWGo)
- Tools for debugging, replay, pause/resume, root cause analysis, ...

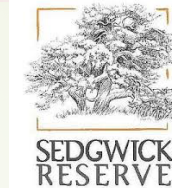


SmartFarm Applications

➤ Sensor synthesis

- Use one sensor to estimate the data of another
- Example: Estimating outdoor air temperature from IoT CPU temperature sensor
 - Micro-climate monitoring
 - Estimating evapo-transpiration (ET)

SmartFarm
Deployments
& Testbeds



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SmartFarm Applications

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Estimating Outdoor Temperature from CPU Temperature for IoT Applications in Agriculture



Chandra Krintz
Rich Wolski

Nevena Golubovic

Fatih Bakir

Computer Science Department
University of California, Santa Barbara

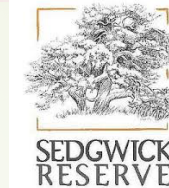


IoT-Conference



3:20pm
Today!

SmartFarm
Deployments
& Testbeds



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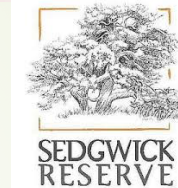
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Research and Extension Center System

SmartFarm Applications

- Sensor synthesis
 - Use one sensor to estimate the data of another
 - Example: Estimating outdoor air temperature from IoT CPU temperature sensor
 - Micro-climate monitoring, ET
- Soil electrical conductivity (EC) analysis for management zone delineation
 - Estimates water holding capacity, soil health, compaction, ...
 - Multi-scale *scored* clustering (@Edge)

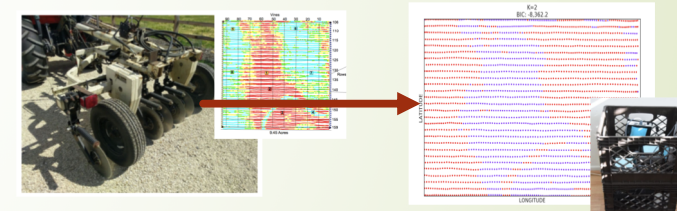
SmartFarm
Deployments
& Testbeds





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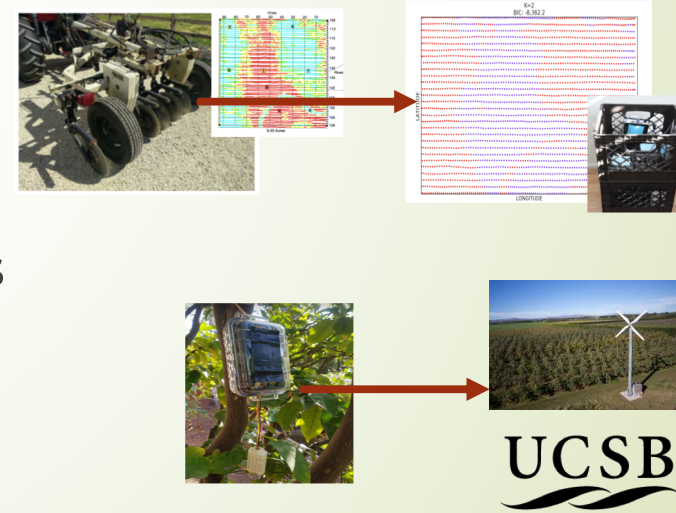
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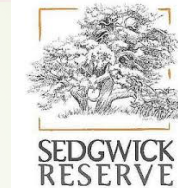


SmartFarm Applications

- **Sensor synthesis**
 - Use one sensor to estimate the data of another
 - Example: Estimating outdoor air temperature from IoT CPU temperature sensor
 - Micro-climate monitoring, ET
 - Soil electrical conductivity (EC) analysis for management zone delineation
 - Automated frost protection (@Edge) triggered by *predicted* micro-climate temperature inversions
 - Autonomous sensing (ground robots)
 - Precision weeding & input application
- 
- 



SmartFarm Deployments & Testbeds



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The logo for the RECS (Recreation, Education, and Community Services) department at Cal Poly San Luis Obispo. It features the word "RECS" in a bold, blue, sans-serif font. To the right of the text is a stylized graphic consisting of several curved lines in shades of blue and green, suggesting a landscape or a dynamic movement.

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A New Kind of IoT Research



- Problem driven and empirical
 - Food-Energy-Water nexus
- Societal and regional impact
- Multidisciplinary collaboration
- Repeatable, demonstrable, applied (tech-transfer ready)
- Engaging students & the community

Thanks!

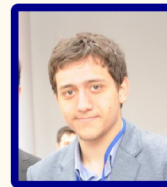
UCSB RACELab

The Lab for Research on Adaptive Computing Environments
Computer Science Department, Harold Frank Hall (E-5), Santa Barbara, CA



- Co-Lead: Dr. Rich Wolski
- Collaborators: UCSB, LREC, Dr. Bo Liu - CalPoly, Dr. Balaji Seth - Fresno State, Powwow Energy, Sedgwick Reserve, Private Growers
- Support: Google, Huawei, IBM Research, Microsoft Research, NSF, NIH, California Energy Commission

Students:



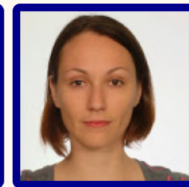
Fatih Bakir



Kyle Carson



Gareth George



Nevena Golubovic



Carly Larsson



Wei-Tsung Lin



Andy Rosales Elias



Nazmus Saquib



John Thomason



Michael Zhang



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