

Architectures, Technologies, Algorithms

Industry 4.0

Αλέξανδρος Μπουσδέκης

albous@mail.ntua.gr

Outline

- Introduction
- Architectures
- Technologies
- Algorithms
- Applications

Outline

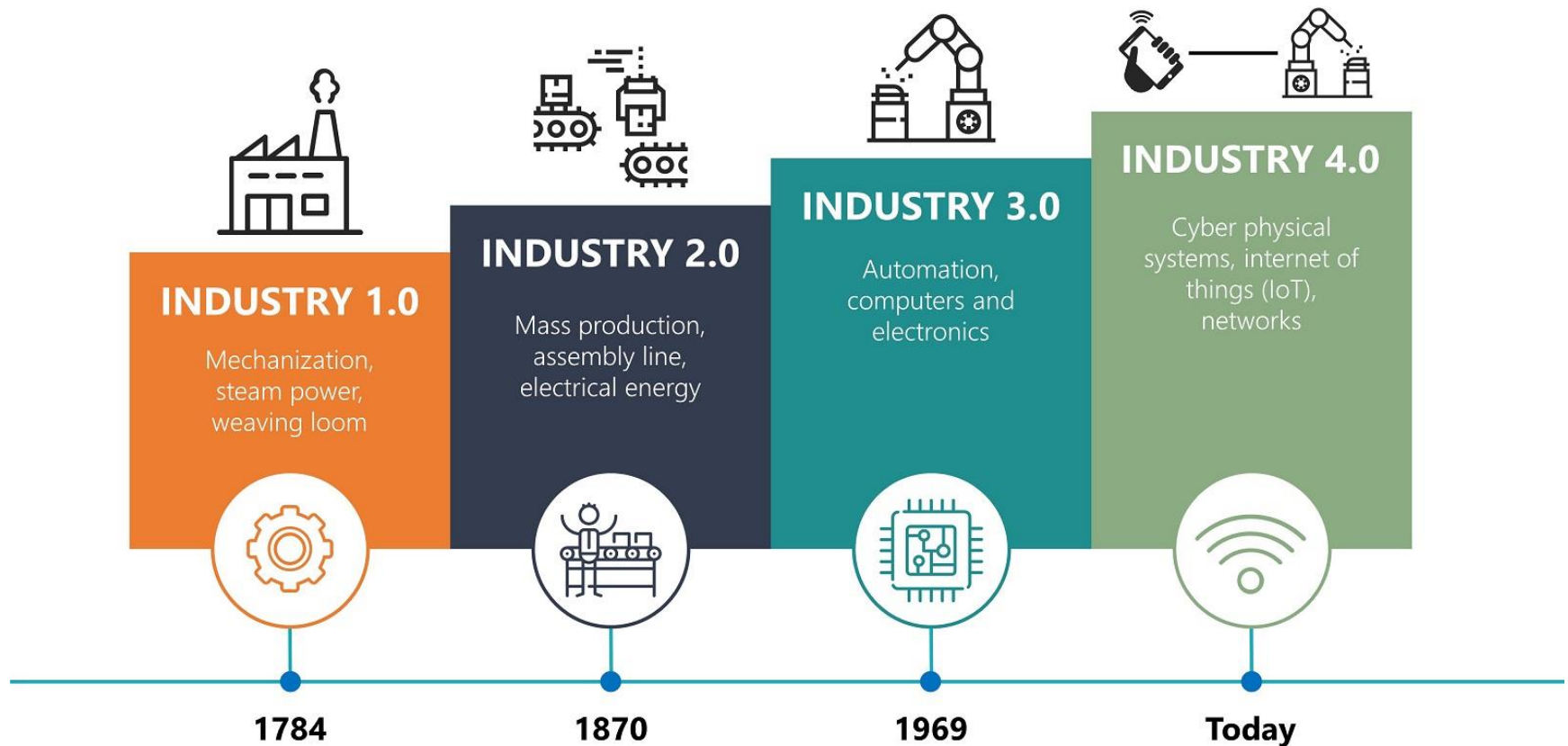
Introduction

- Architectures
- Technologies
- Algorithms
- Applications

The AUDI smart factory

**SMPTE
UNIVERSAL
LEADER**

The 4 industrial revolutions



Today's factories are not efficient

- If there is one decisive metric in the manufacturing environment, it is the OEE (Overall Equipment Effectiveness).
 - It measures how effectively a manufacturing operation is utilized compared to how effective it could be.
 - Average OEEs run at 60-70%. Today's world class OEEs are around 85%.
 - That means that even the most effective factories today lose 15% of time on non-value-adding tasks like machine changeovers, stoppages, maintenance, and production of faulty products.
 - The monetary equivalent to a one percentage point increase in OEE is gigantic for every company.

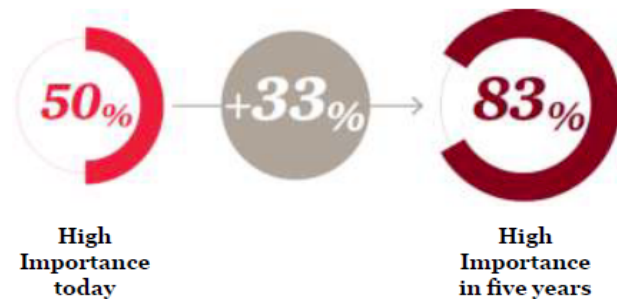
PwC survey (1/2)

The survey highlighted the importance of Data Analytics in our clients' decision-making processes over the next five years



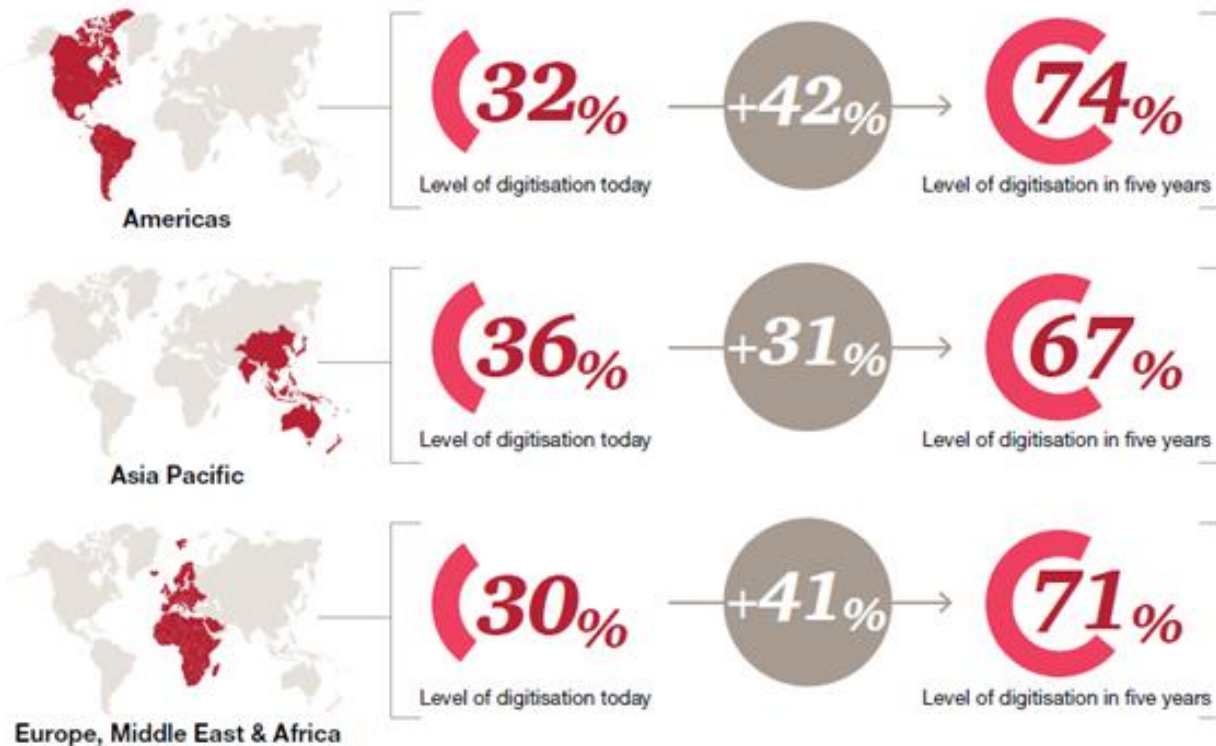
The Industry 4.0 Survey had a focus on Data Analytics, giving us a bird's eye view of this issue in Industry

- **83%** of respondents expect data analytics will have a significant influence on their decision-making processes in five years' time
- **33%** of respondents expect this change to happen in the next five years

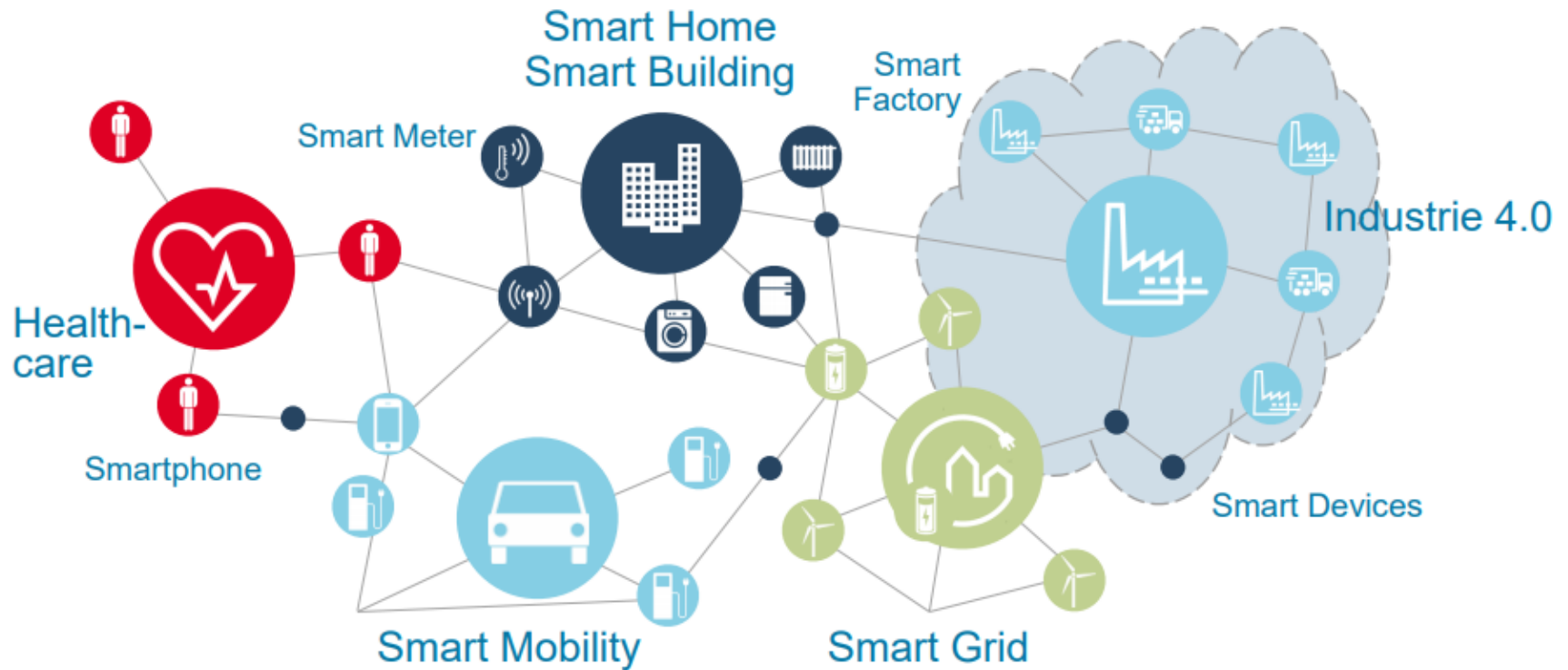


PwC survey (2/2)

Companies all over the world are expecting to dramatically increase digitisation over the next five years



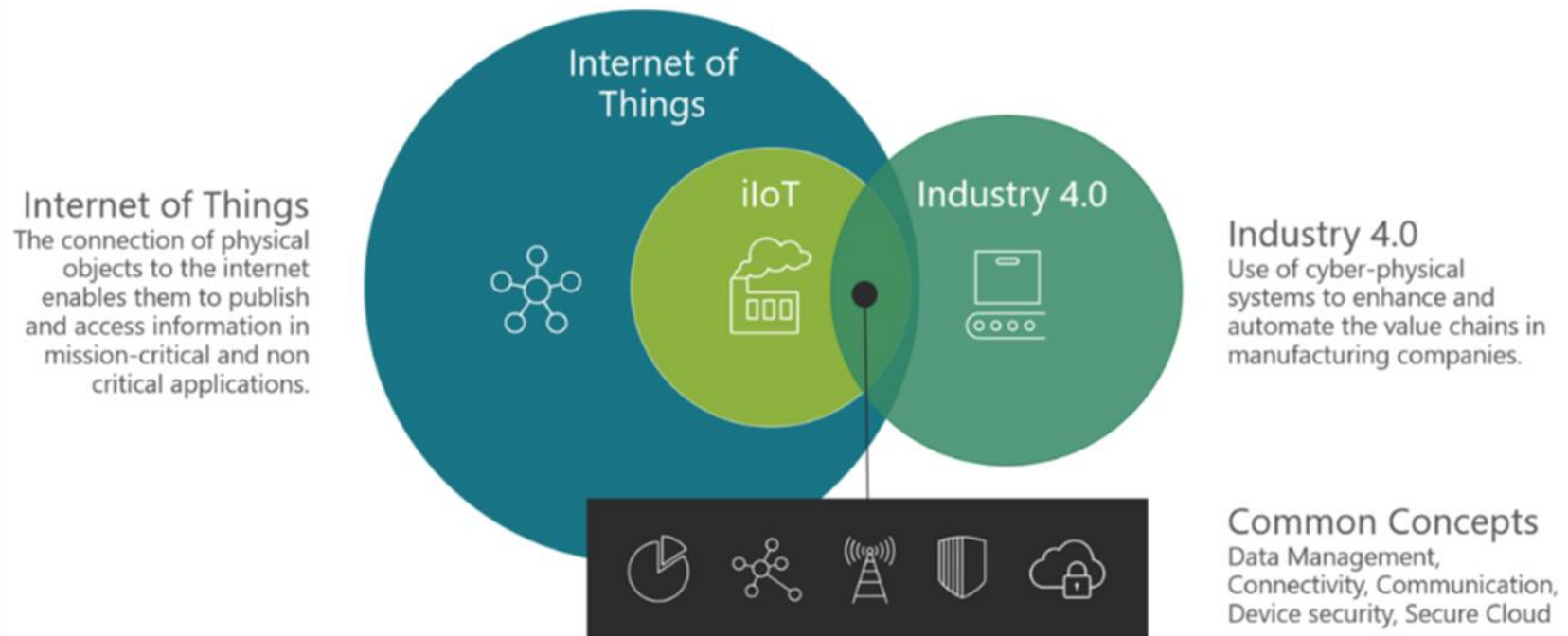
Internet of Things (IoT)



Industry 4.0 and Internet of Things

Making Sense of the Trends

The big picture of IoT and Industry 4.0



Source: Daniel Sontag

The future of Industrial IoT

- Many market researchers such as Gartner, Cisco and PwC consider the **industrial IoT** as the IoT concept with the **highest overall potential**.
- However, **it has not gathered yet the interest** that smart homes or wearables have gathered, due to the **high investments** required and the **long periods of implementation** needed.
- **The 2/3 of IoT benefits will deal with Industry** and the 1/3 with consumer benefits.

Industry 4.0: Revolution or Evolution?

- *"The electric light did not come from the continuous improvement of the candle."*
- *"Technological innovation is continuous and the concept of a "revolution" is based on a lack of knowledge of the details."*



Clash of two worlds

“Machine Guys“ meet “Internet Guys“.

5-year-thinking vs. continuous beta.

IoT needs both worlds.

How to build bridges?

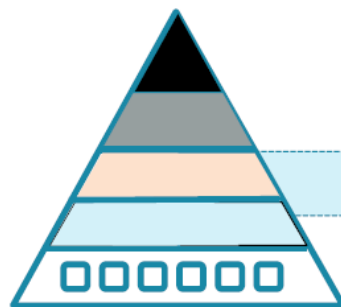


Source: www.enterprise-iot.org

Revolution or Evolution? – The answer

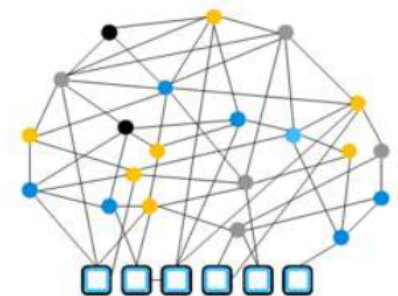
Evolution or Revolution ?

Revolution on Business Level



Automation
Pyramid

Interoperability/Standards
IT-Security/Industrial Data Space[®]
Dependability and Latency
Machine Learning/Data Analytics
Human-Machine Collaboration



Service
Network

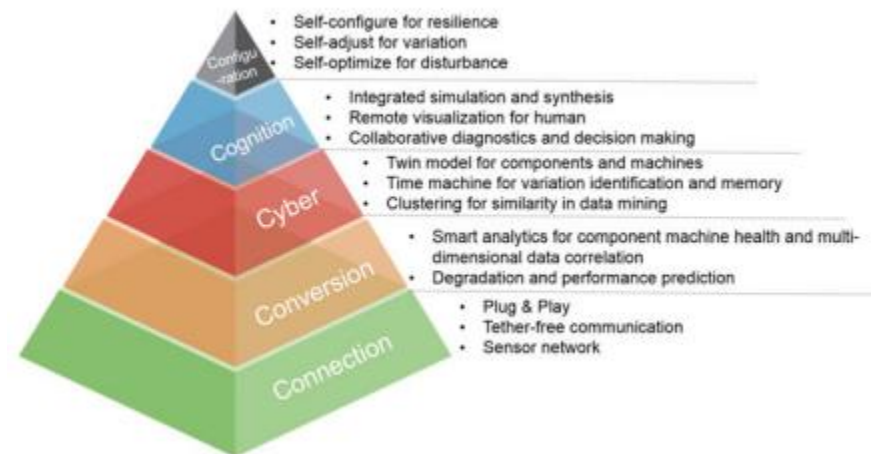
Evolution on Technological Level

Industry 4.0 pillars

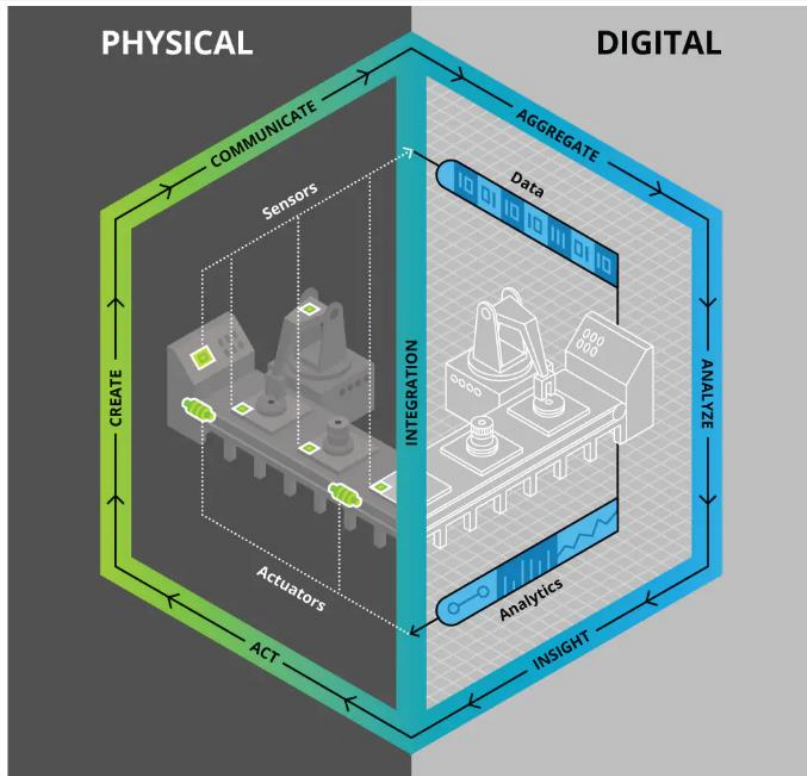


Cyber Physical System (CPS)

- CPS systems require seamless integration between computational models and physical components.
- Each physical component and machine will have a Digital Twin model in the cyber space composed of data generated from sensor networks and manual inputs.
- A CPS can be constructed by following the “5C” architecture, which serves as a guideline for the development of CPS for industrial applications

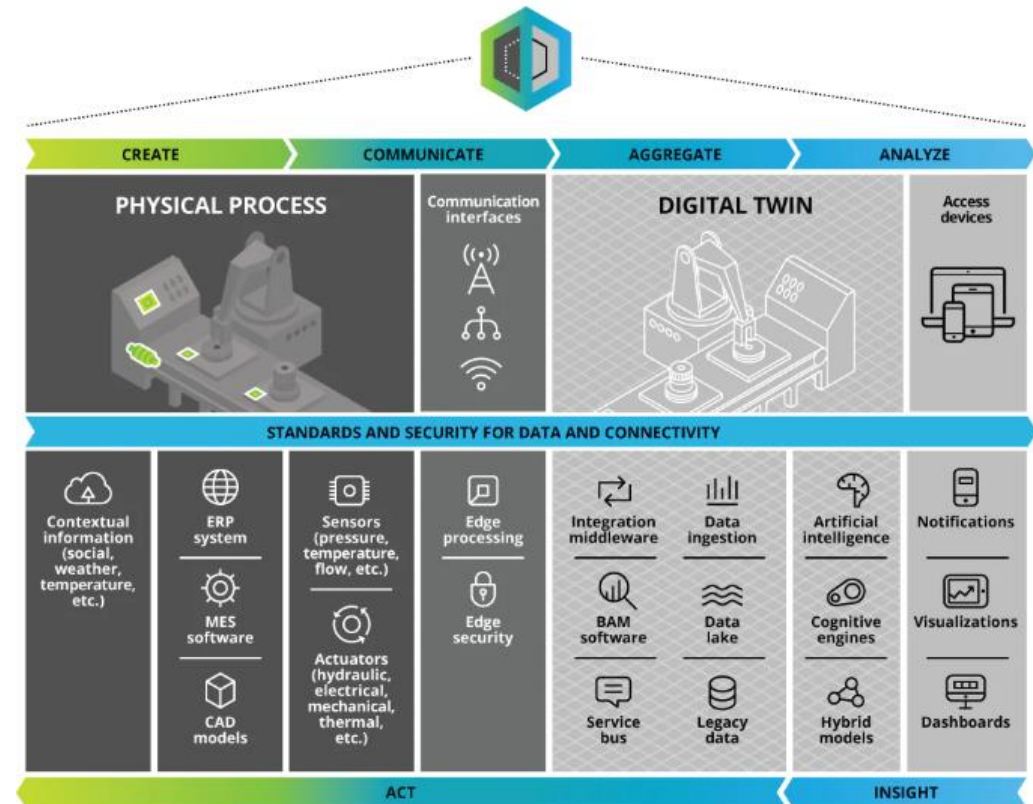


Digital twin



Source: Deloitte University Press.

Deloitte University Press | dupress.deloitte.com



Source: Deloitte University Press.

Deloitte University Press | dupress.deloitte.com

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- Introduction

Architectures

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Reference Architectural Model for Industry 4.0 (RAMI 4.0)

Platform Industrie 4.0

Platform Industrie 4.0



Working groups
Main topics

WG1	WG2	WG3
Reference Architectures, Standards and Norms Chair: Kai Garrels, <i>ABB STOTZ-KONTAKT GmbH</i>	Technology and Application Scenarios Chair: Johannes Kalhoff, <i>Phoenix Contact</i>	Security of Networked Systems Chair: Michael Jochem, <i>Robert Bosch GmbH</i>
Legal Framework Chair: Dr. Hans-Jürgen Schlinkert, <i>ThyssenKrupp</i>	Work, Education and Training Chair: Martin Kamp, <i>IG Metall</i>	Digital Business Models for Industrie 4.0 Chair: Prof. Dr. Svenja Falk, <i>accenture</i>
WG4	WG5	WG6

Industry partners involved



BOSCH



**Rockwell
Automation**

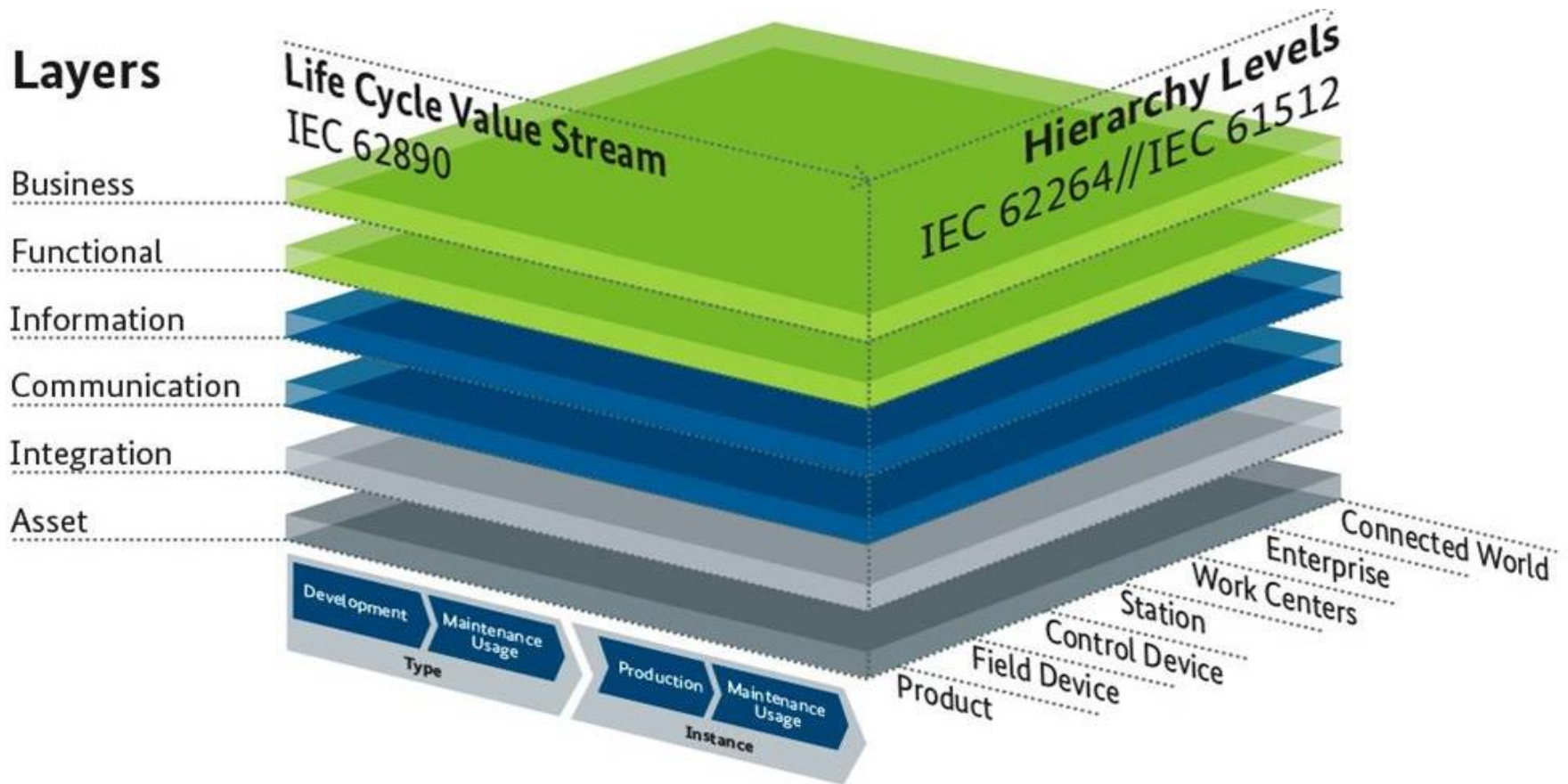
SIEMENS

FESTO

ABB

**Schneider
Electric**

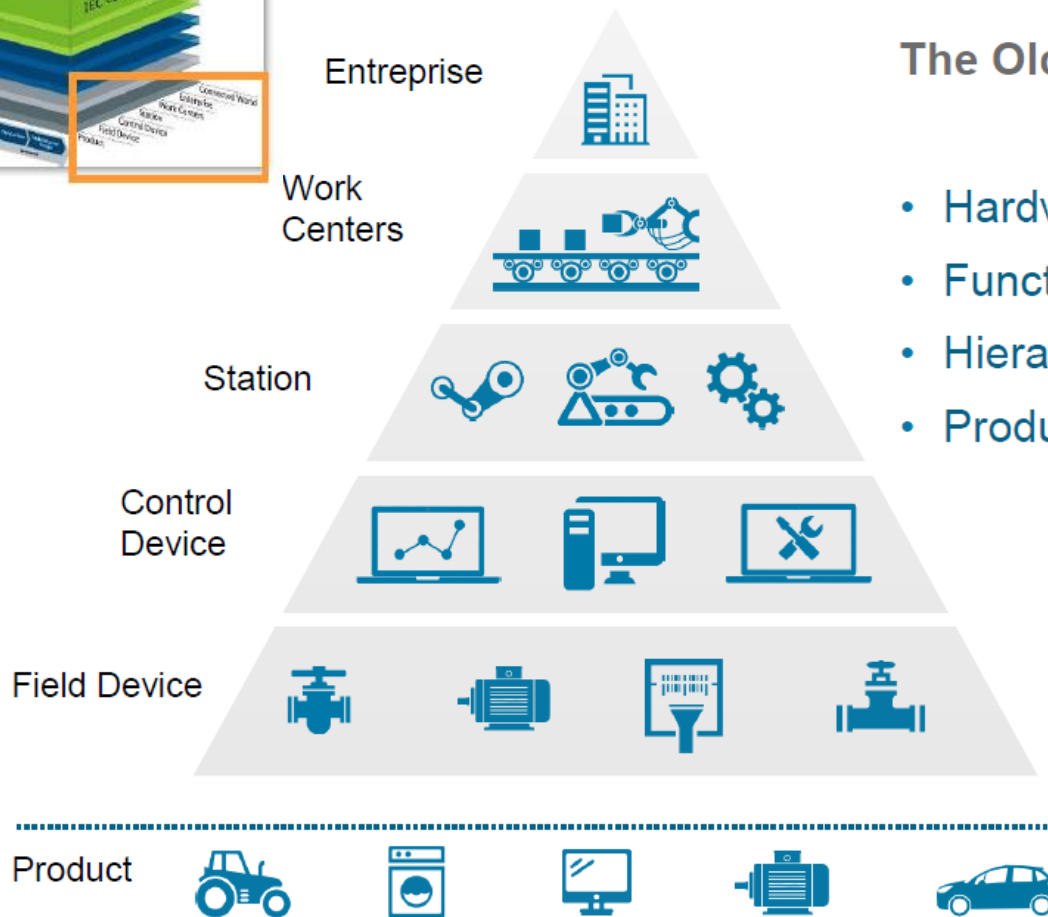
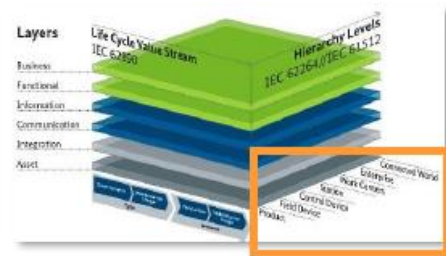
RAMI 4.0



RAMI 4.0

- RAMI 4.0 is a **three-dimensional map** showing how to approach the issue of Industrie 4.0 in a structured manner.
- RAMI 4.0 ensures that all participants involved in Industrie 4.0 discussions **understand each other**.
- RAMI 4.0 is a **SERVICE-ORIENTED ARCHITECTURE**.
- RAMI 4.0 **combines all elements and IT components** in a layer and life cycle model.
- RAMI 4.0 **breaks down complex processes** into easy-to-grasp packages, including data privacy and IT security.

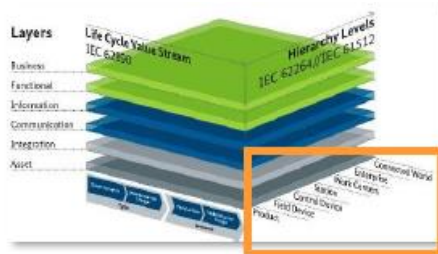
Axis 1 – Hierarchy: The Factory



The Old World: Industrie 3.0

- Hardware-based structure
- Functions are bound to hardware
- Hierarchy-based communication
- Product is isolated

Axis 1 – Hierarchy: The Factory



The New World: Industrie 4.0

- Flexible systems and machines
- Functions are distributed throughout the network
- Participants interact across hierarchy levels
- Communication among all participants
- Product is part of the network

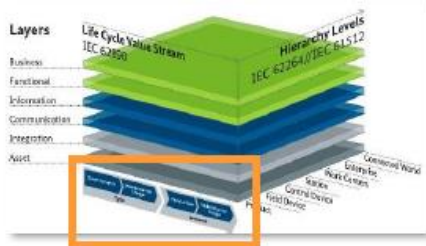
Connected World

Smart Factory

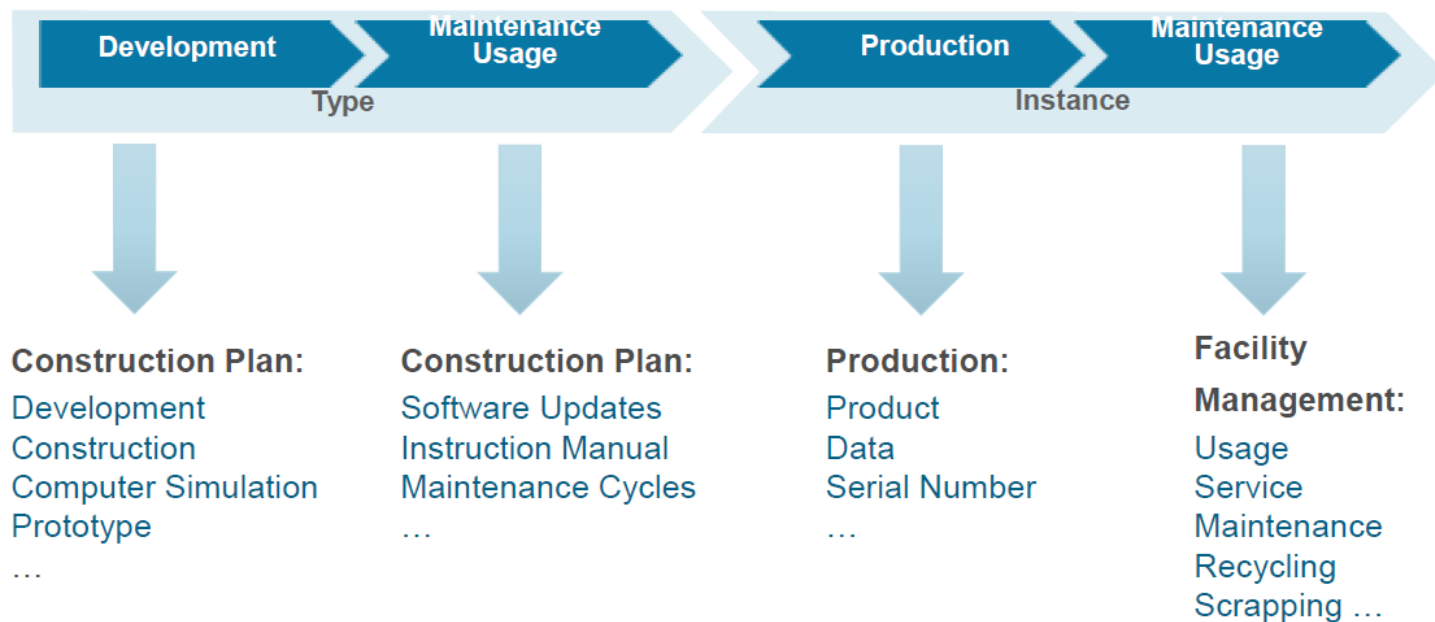
Smart Products



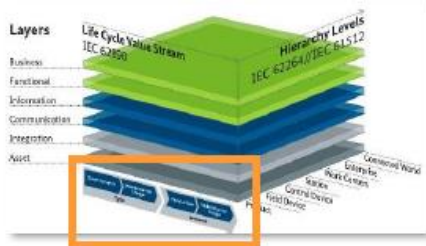
Axis 2 – Product Life Cycle



The Product: From the First Idea to the Scrapyard



Axis 2 – Examples



Manufacturer



Development

Maintenance Usage

Type

Production

Maintenance Usage

Instance

System integrator



Development

Maintenance Usage

Type

Production

Maintenance Usage

Instance

Operator / end customer



Development

Maintenance Usage

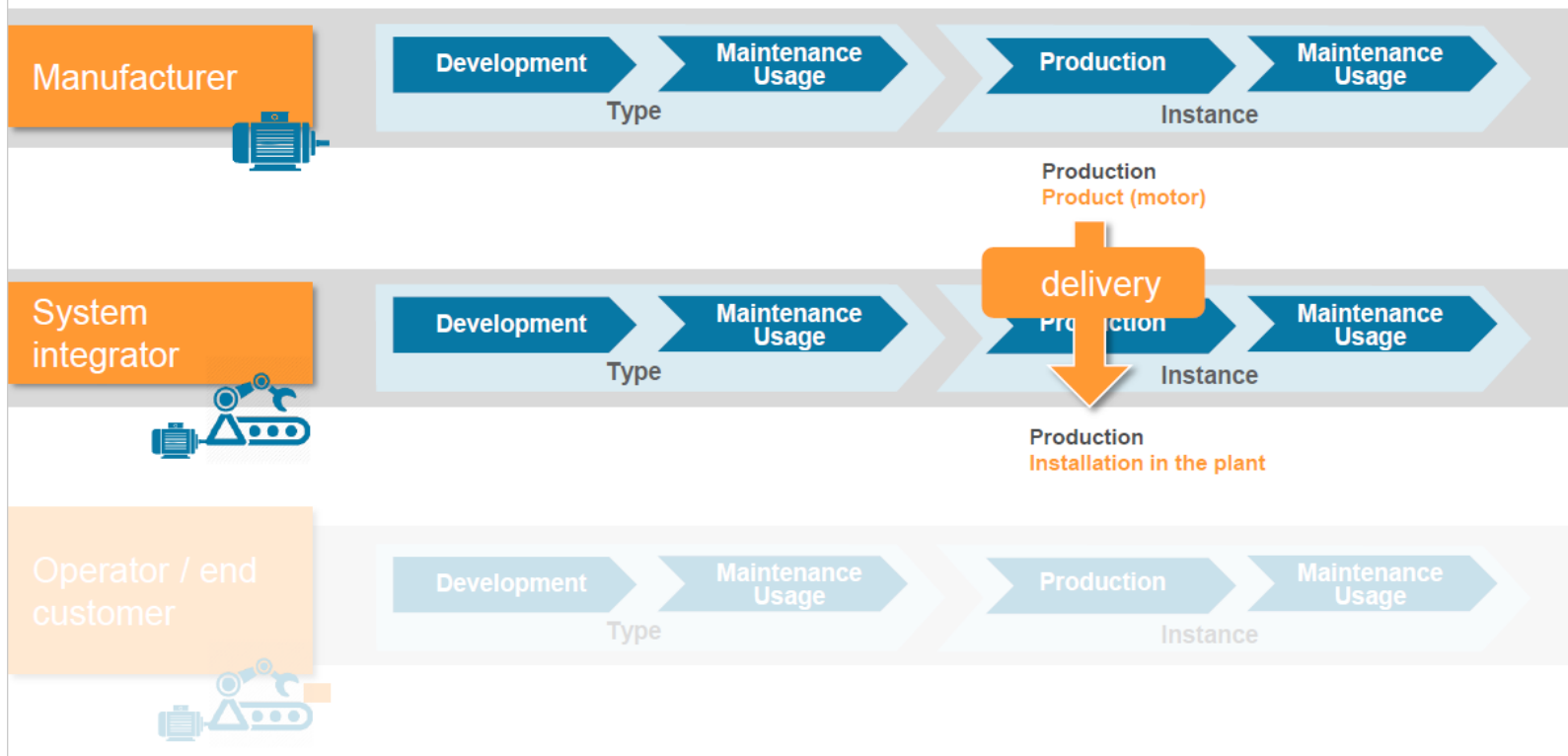
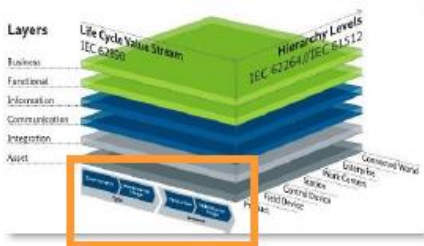
Type

Production

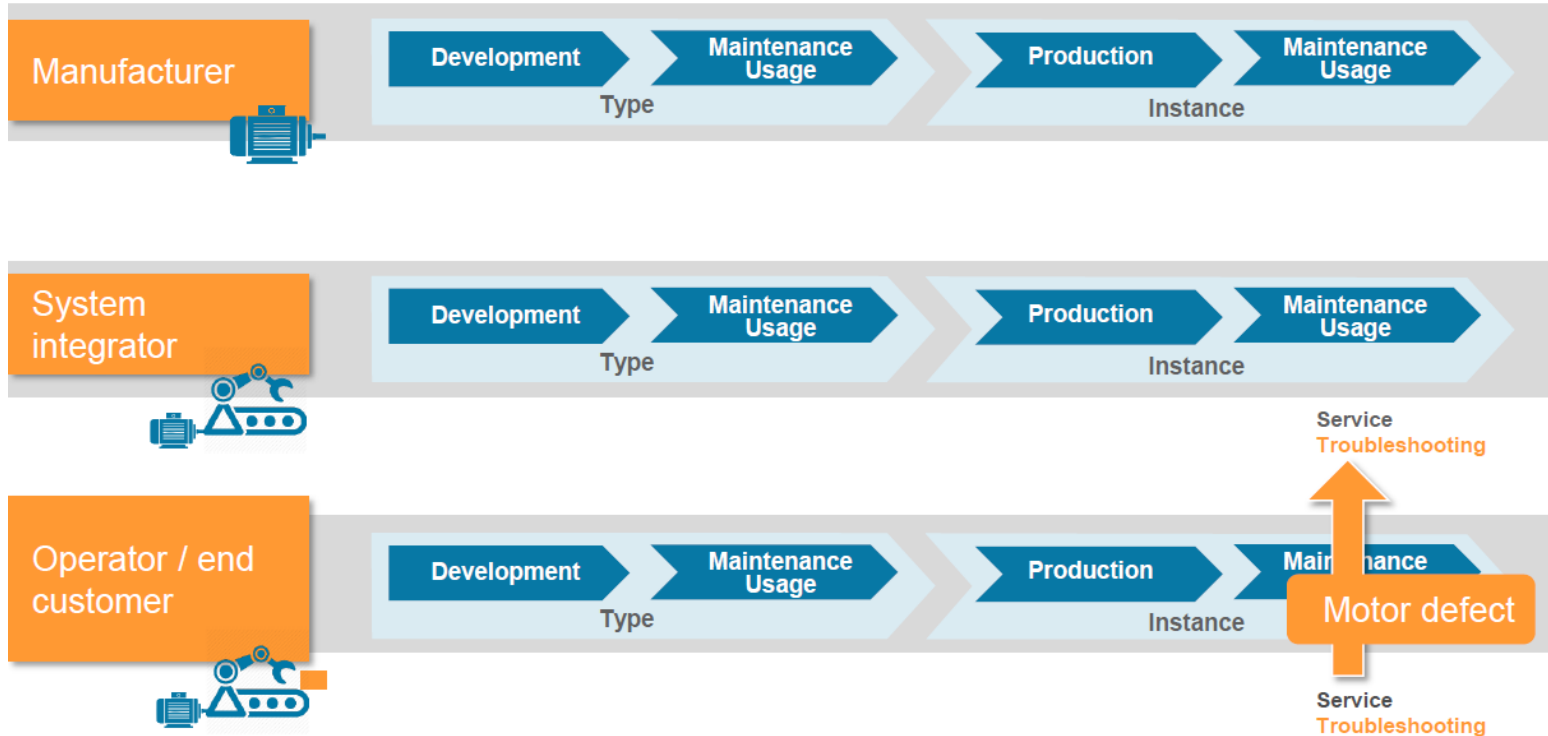
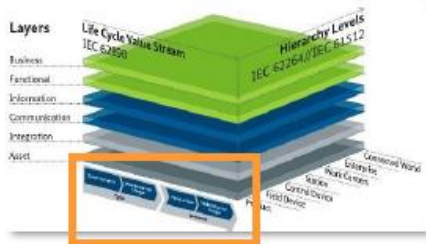
Maintenance Usage

Instance

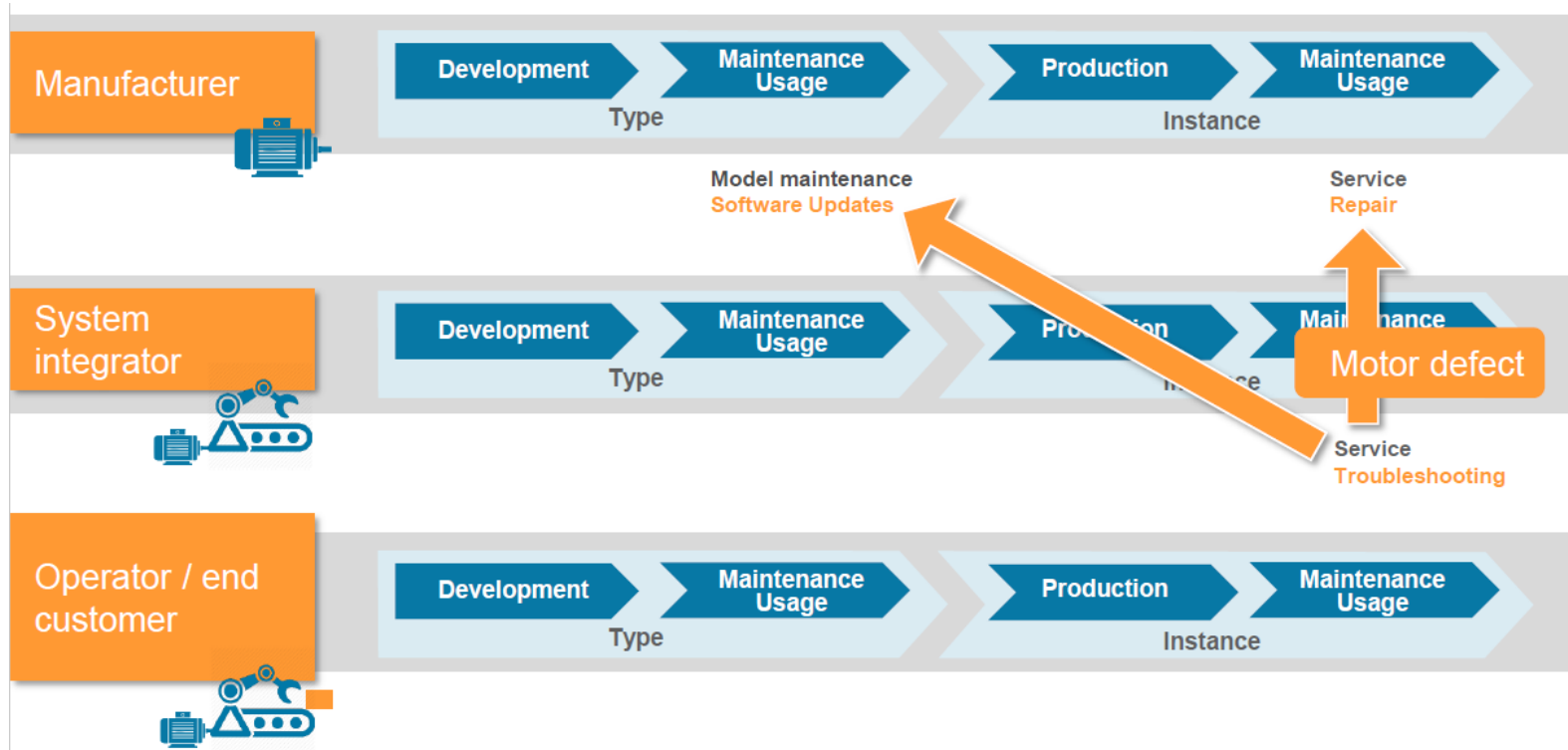
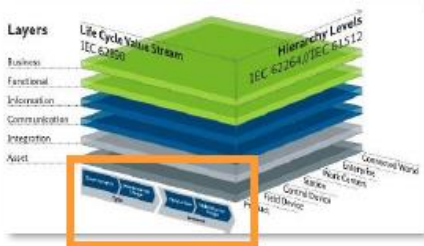
Axis 2 – Examples



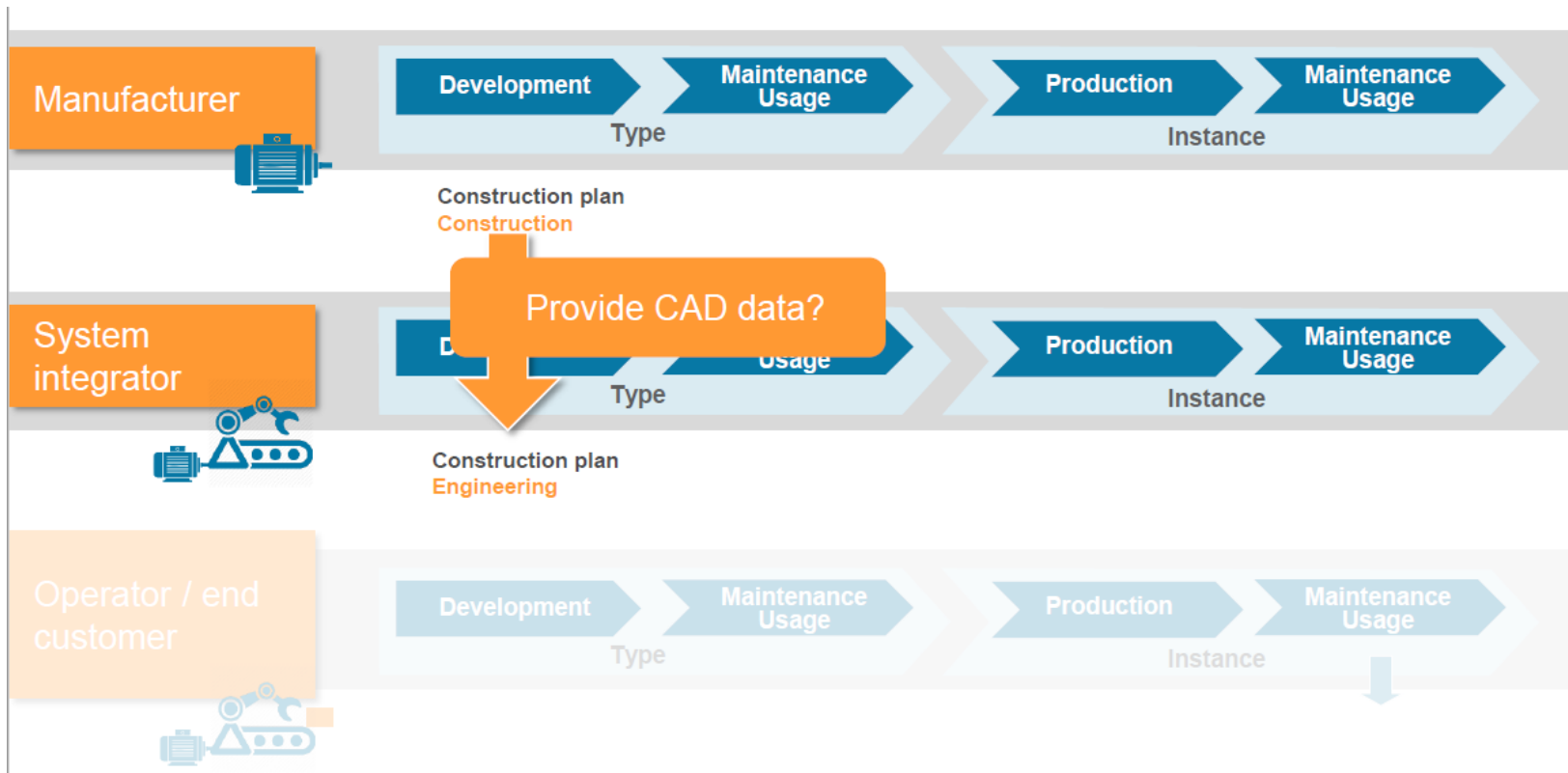
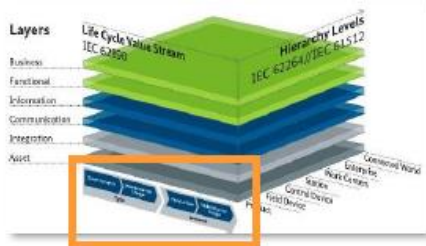
Axis 2 – Examples



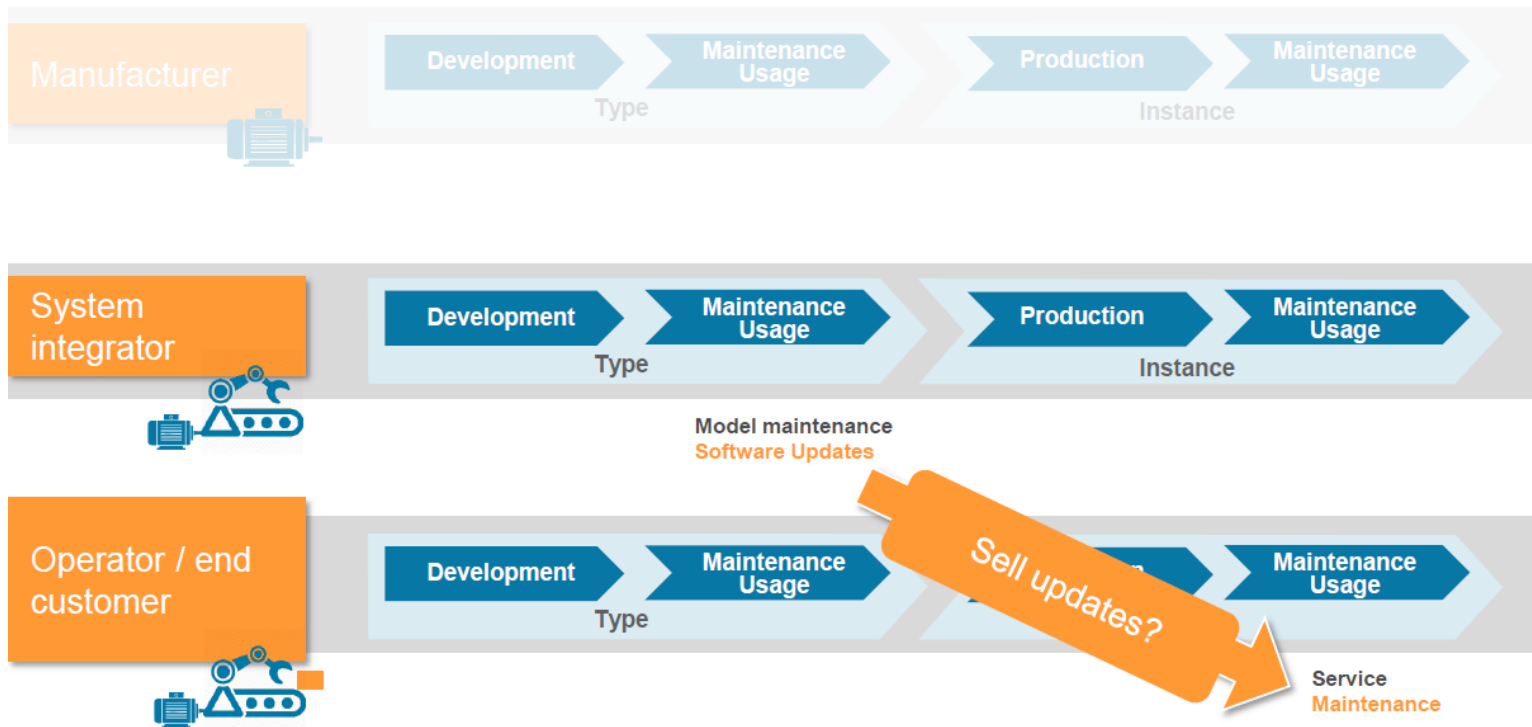
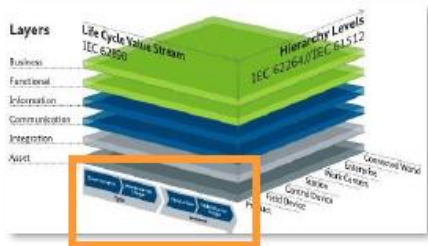
Axis 2 – Examples



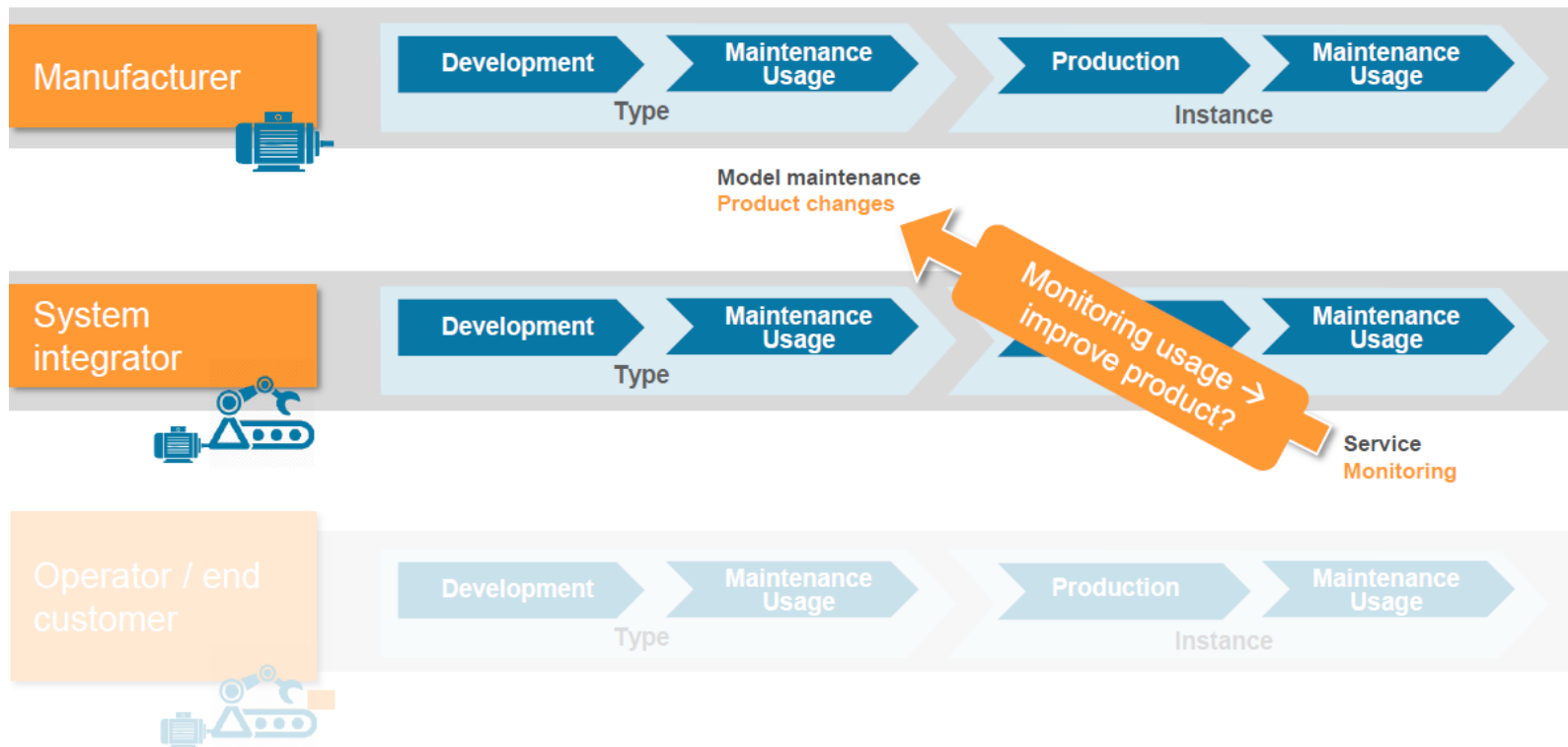
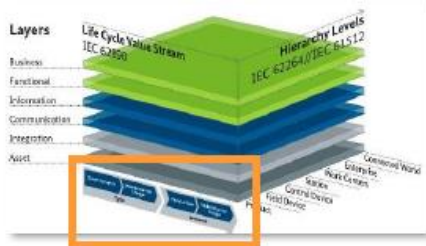
Axis 2 – Examples



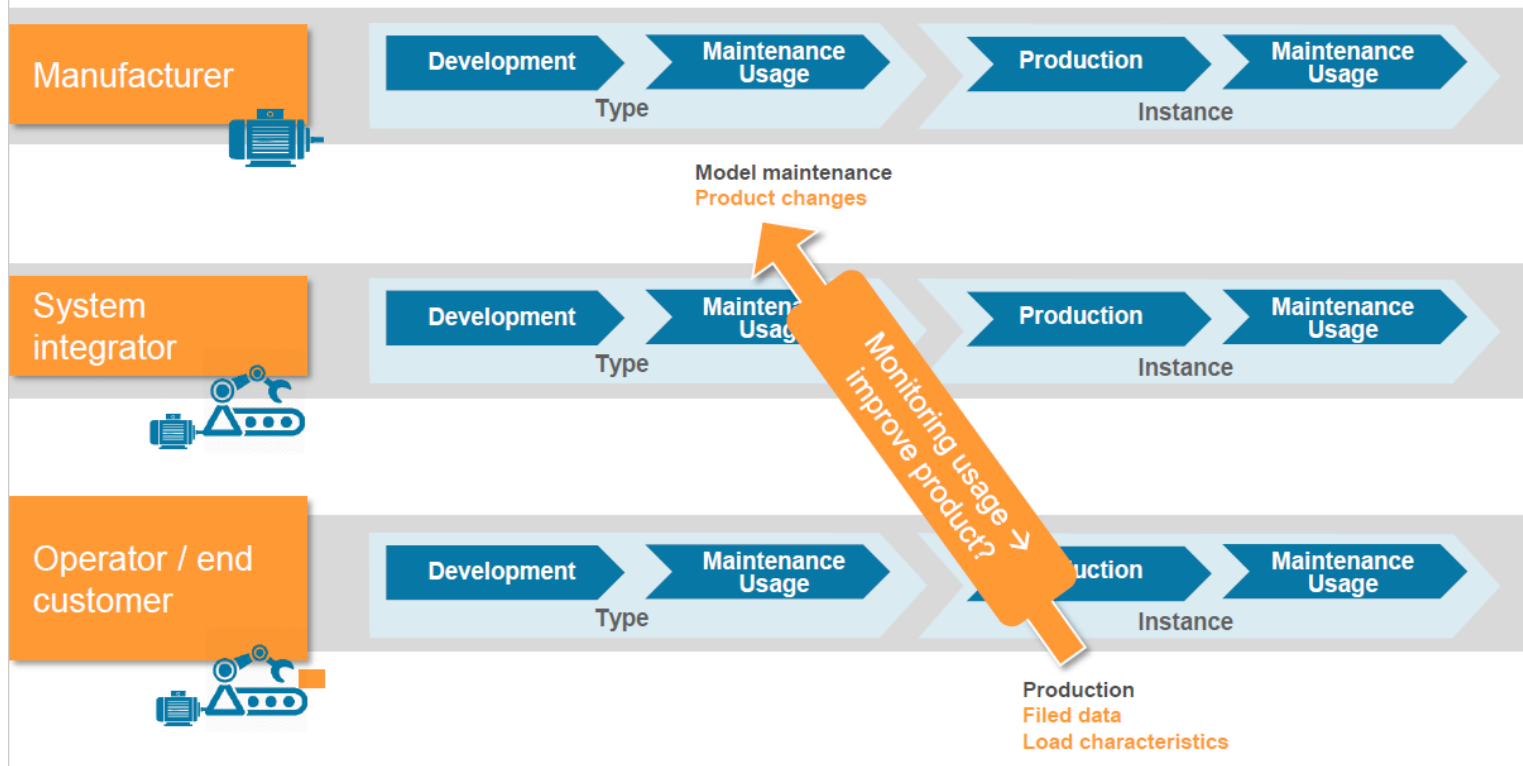
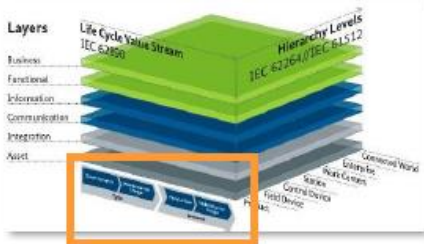
Axis 2 – Examples



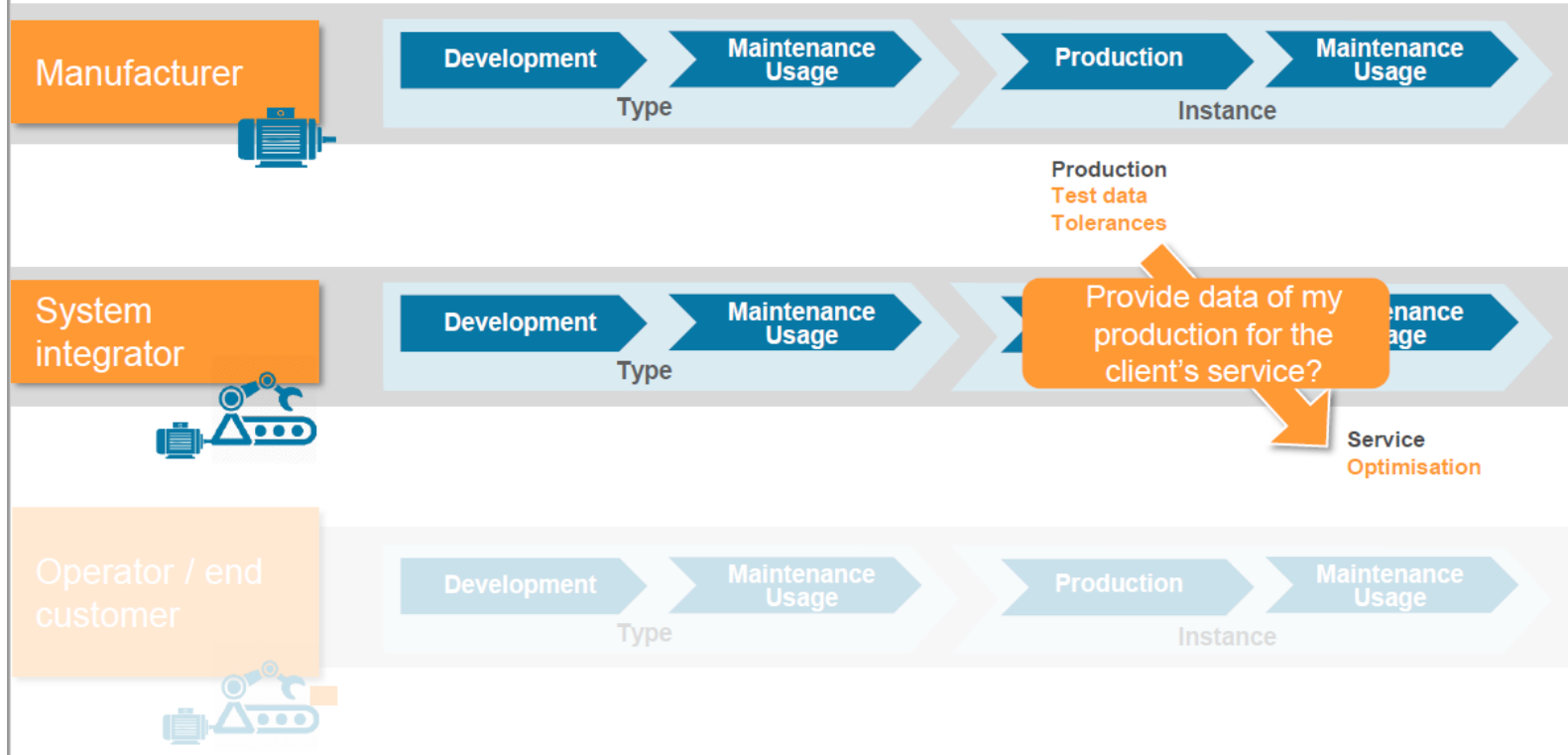
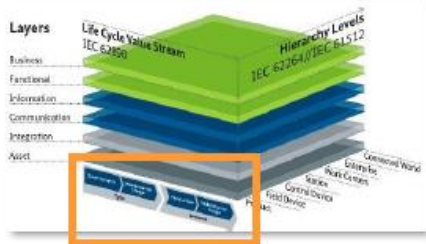
Axis 2 – Examples



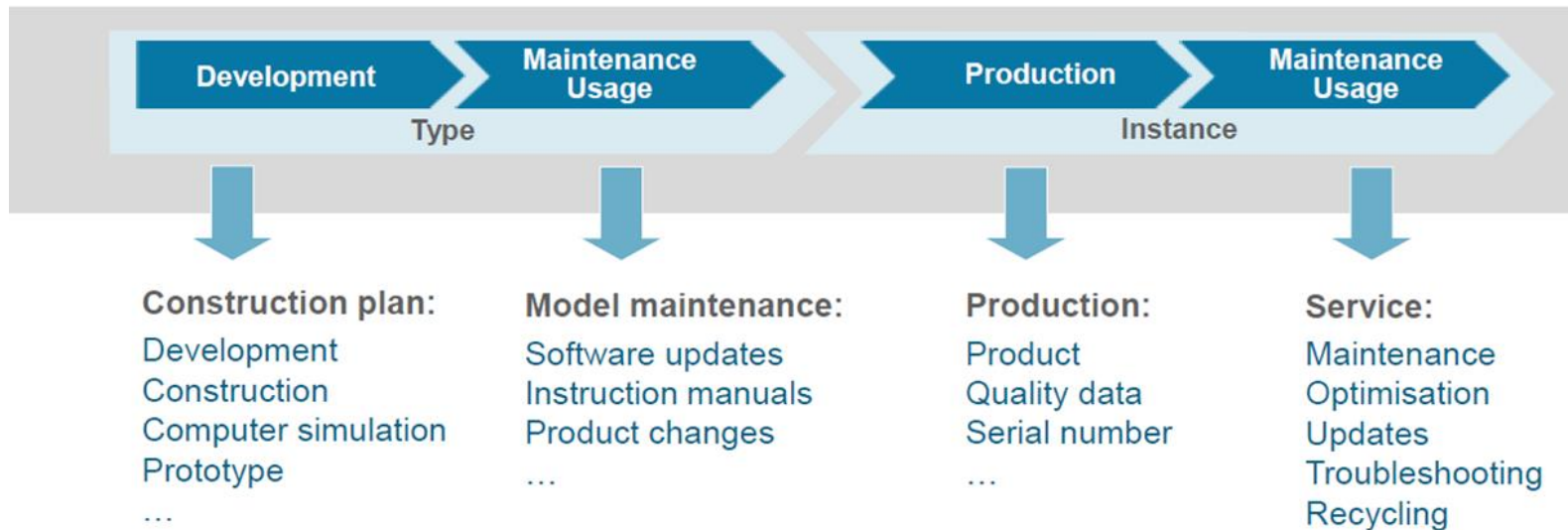
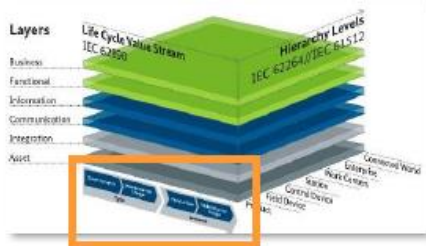
Axis 2 – Examples



Axis 2 – Examples



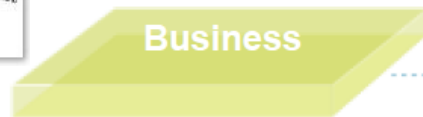
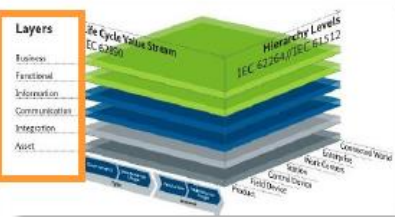
Axis 2 – Examples



Example: As a manufacturer of motors I can take two perspectives

- ▶ My product and service offers to my clients
- ▶ My internal processes and workflows

Axis 3 – Architecture



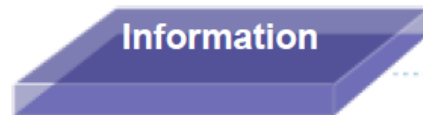
Business

Organisation and Business Processes



Functional

Functions of the Asset



Information

Necessary Data



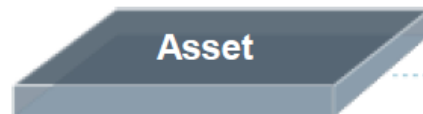
Communication

Access to Information



Integration

Transition from Real to Digital World

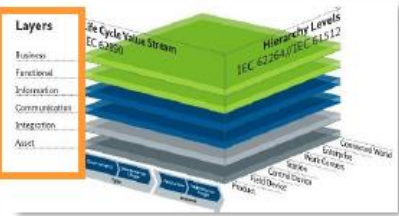


Asset

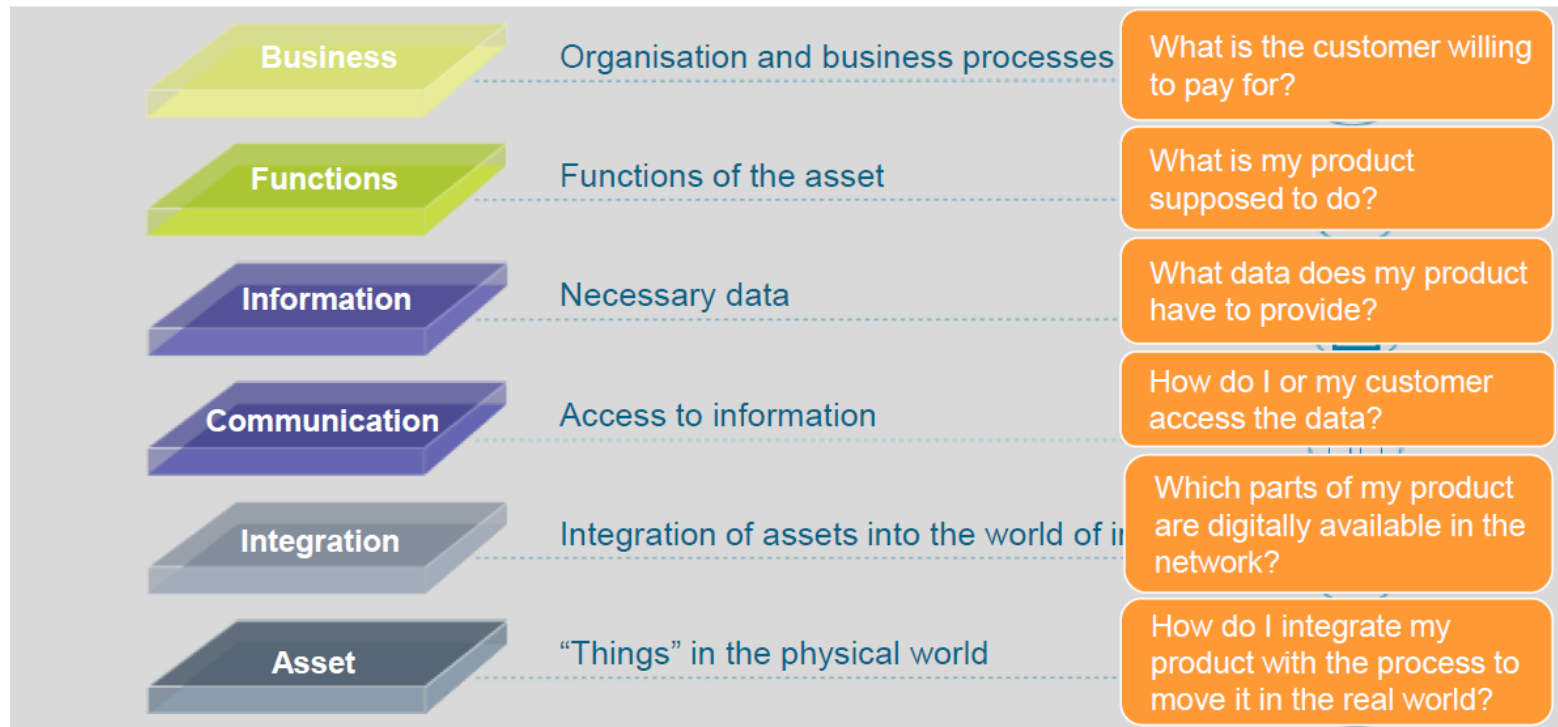
Physical Things in the Real World



Axis 3 – Architecture



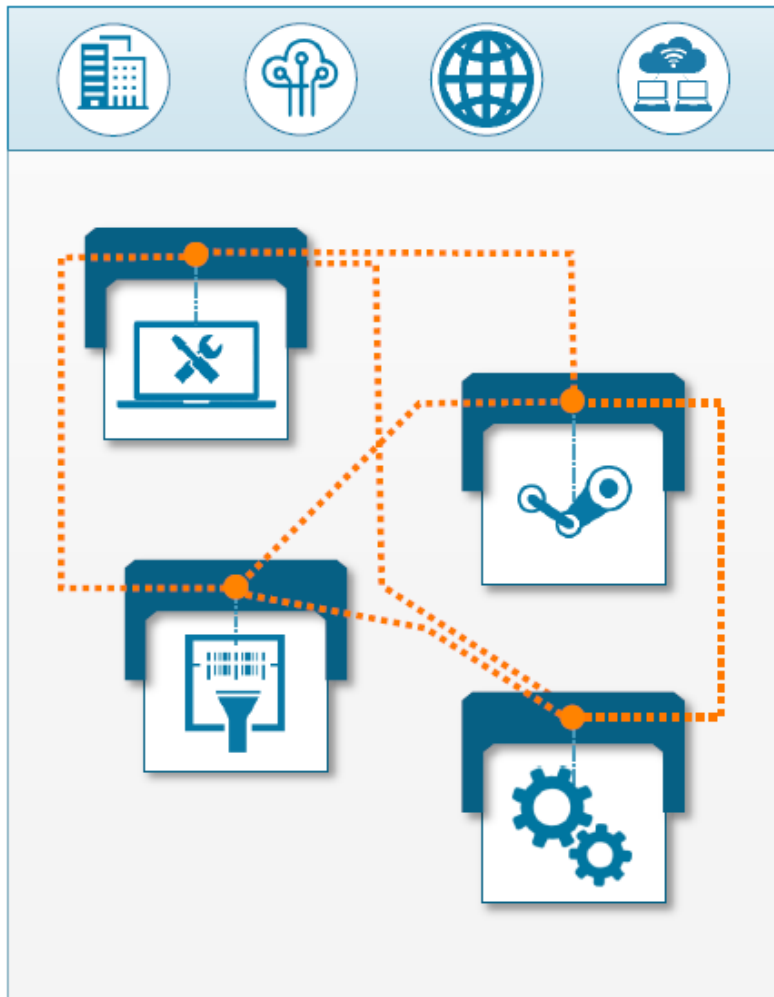
Basic questions about the business idea



Asset Administration Shell

Platform Industrie 4.0

Who provides interpretation?



- The Administration Shell...
 - ... is the interface connecting I4.0 to the physical Thing
 - ... stores all data and information about the asset
 - ... serves as the network's standardized communication interface
 - ... is also able to integrate passive assets

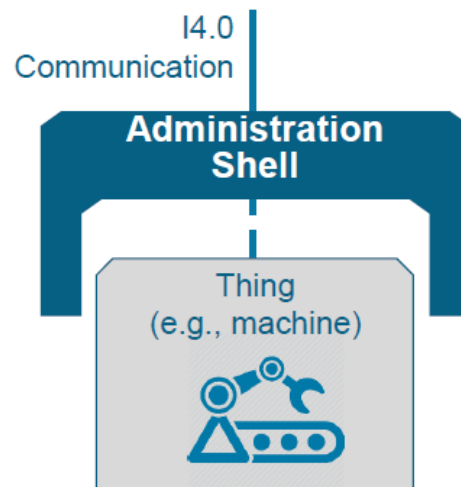
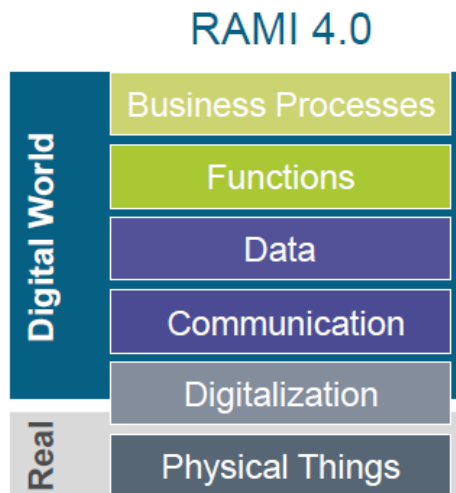
The Roles and Responsibilities of the Administration Shell

- Each physical thing has its own administration shell.
- Several assets can form a thematic unit with a common administration shell.



The Industrie 4.0 Component

- Each object needs its own administration shell that allows its integration into Industrie 4.0.



- The connection takes place over the I4.0 communication
- The administration shell forms the digital part
- The Thing forms the real part

Administration Shell implements the Digital Twin



Digital Twin

Definition 1: Digital representation of a physical asset

Definition 2: Simulation model



The Administration Shell is the implementation of the Digital Twin for Industrie 4.0.

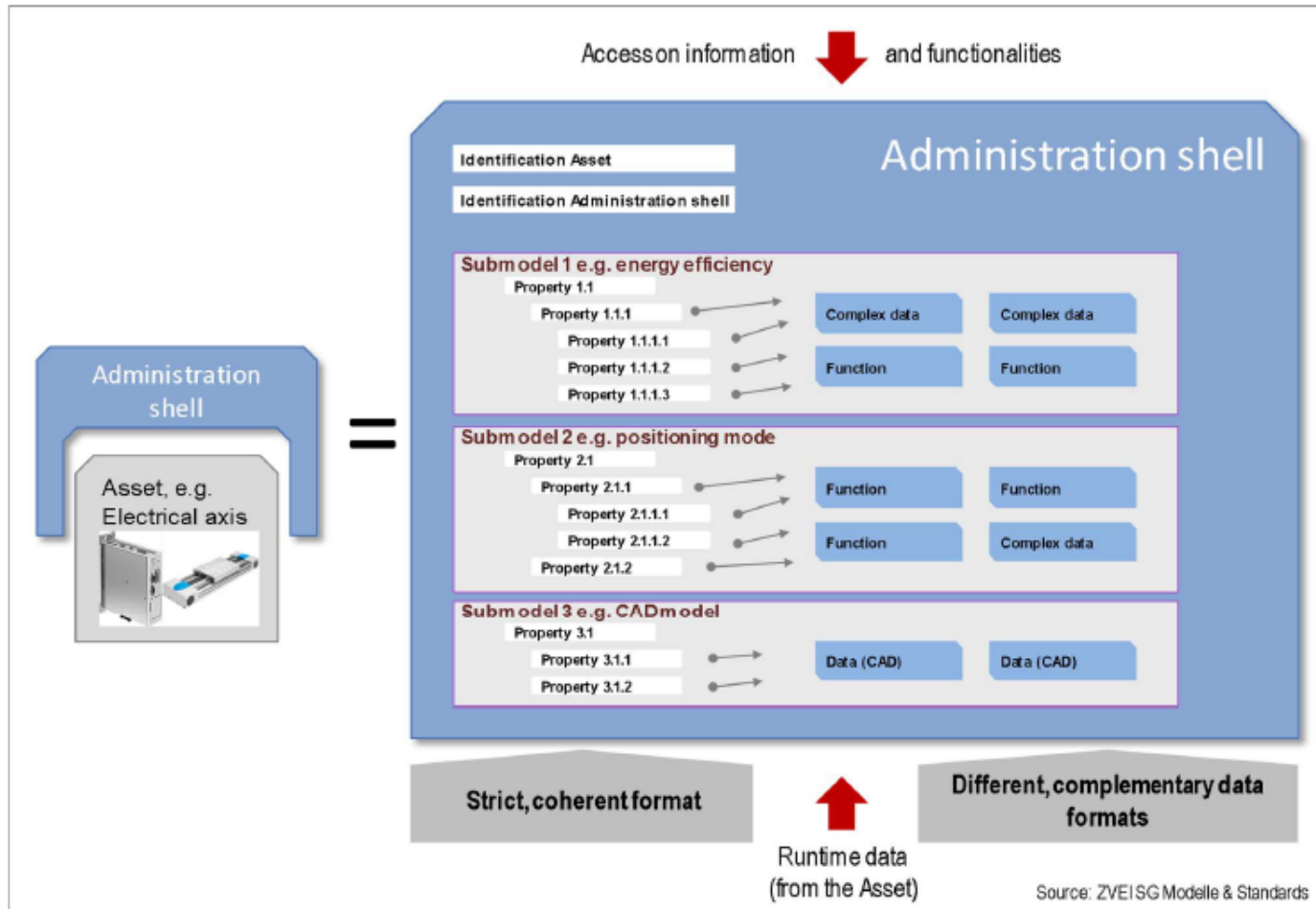
Administration Shell



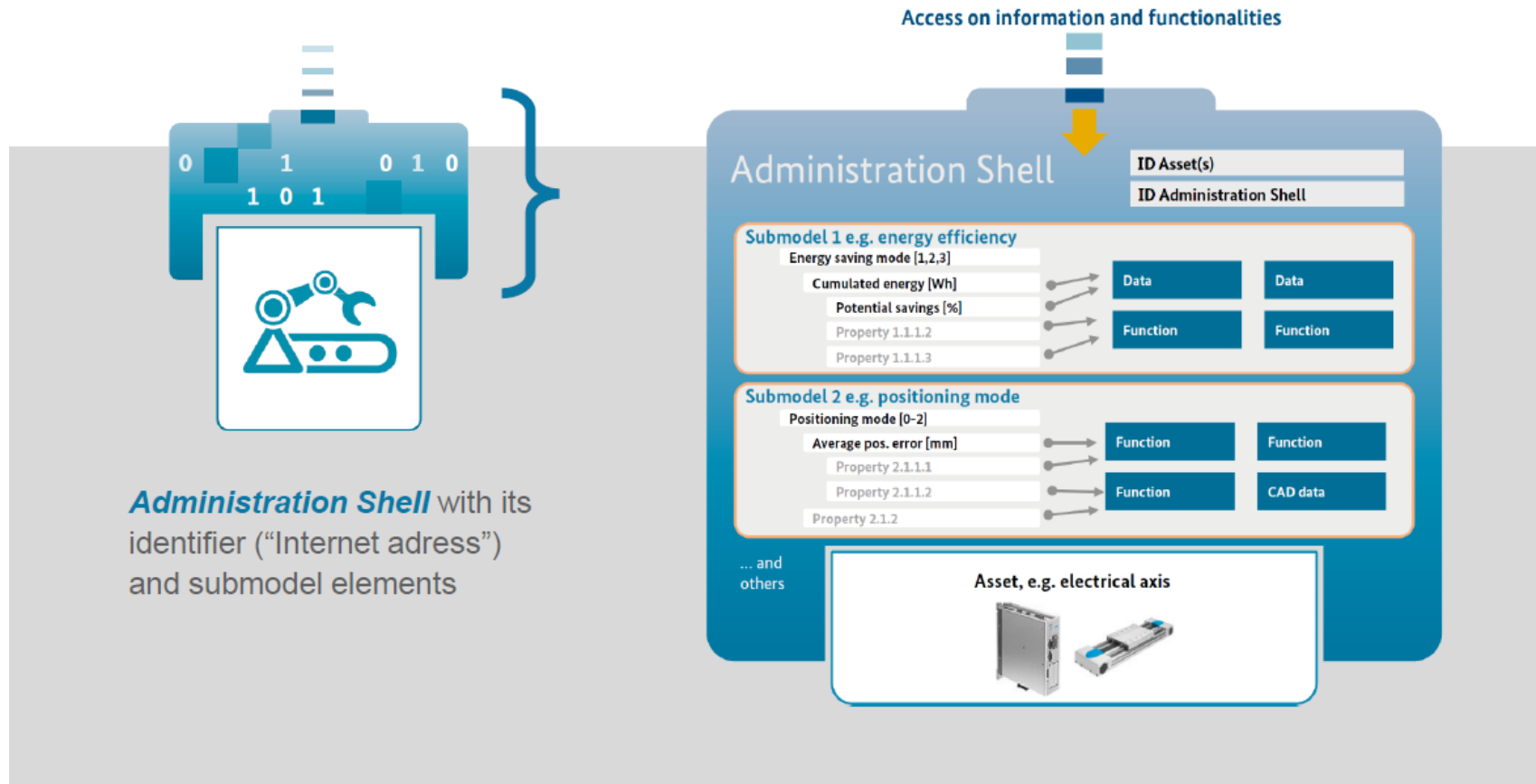
The Administration Shell...

- ▶ integrates the *asset into Industrie 4.0 communication*.
- ▶ is *addressable* in the network and *identifies the asset* unambiguously.
- ▶ provides a *controlled access to all information* of the asset.
- ▶ is the standardised and secure *communication interface*.
- ▶ can integrate *intelligent* and also *non-intelligent („passive“)* assets (without a communication interface), e.g. via bar codes or QR codes.

Basic structure



The generic structure of the meta information model

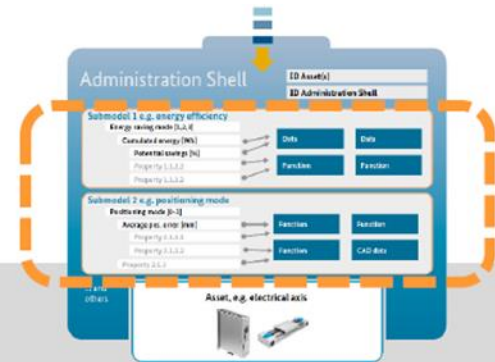


Source: Plattform Industrie 4.0 illustration based on ZVEI SG Modelle & Standards

Submodels

Submodel = aggregates information that belongs together

- ▶ *Submodels* combine different functional aspects of an Industrie 4.0 component
- ▶ *Basic submodels* (standardised): apply to many assets in the Industrie 4.0 world (e.g. catalogue data of products)
- ▶ *Free submodels*: agreed between partners in the value chain for a specific use case

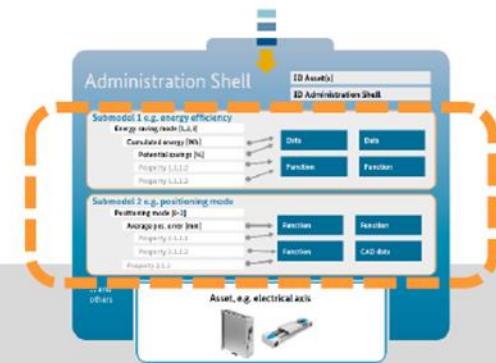


Submodels should always be linked to a **use case that creates value.**

An Administration Shell may contain many submodels.

Aim: To develop one submodel for each functional aspect.

Submodel elements / Properties



Submodels contain submodel elements

(Submodel elements include e.g. properties.)

- ▶ *Product properties* in terms of IEC61360-1 or ecl@ss
- ▶ *Process variables* and parameters, telemetry data
- ▶ *References* to external data sources or files
- ▶ *References* to other Administration Shells and their parts (submodels, properties), also from external partners in the value chain
- ▶ *Capabilities* of the asset, description of method calls
- ▶ *Sets* of properties, e.g. lists or arrays

Example

Submodel „energy efficiency“ contains e.g.

- ▶ Energy saving mode [1,2,3]
- ▶ Cumulated energy [Wh]
- ▶ Potential savings [%]

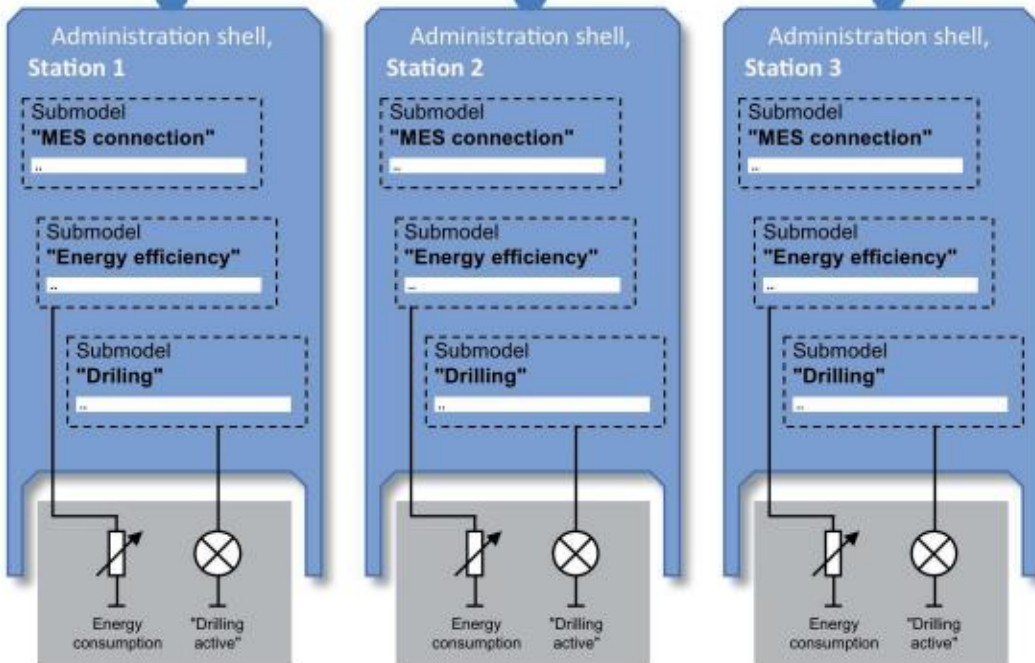
Submodel „positioning mode“ contains e.g.

- ▶ Positioning mode [0-2]
- ▶ Average pos. error [mm]

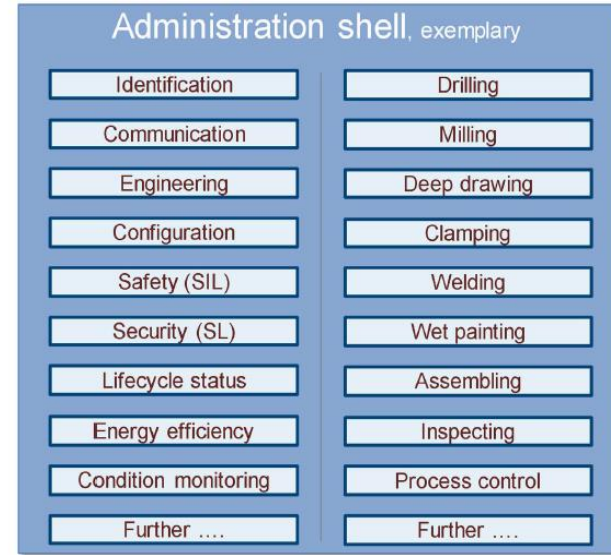
Examples

Hypothetical production planning and control system (MES)

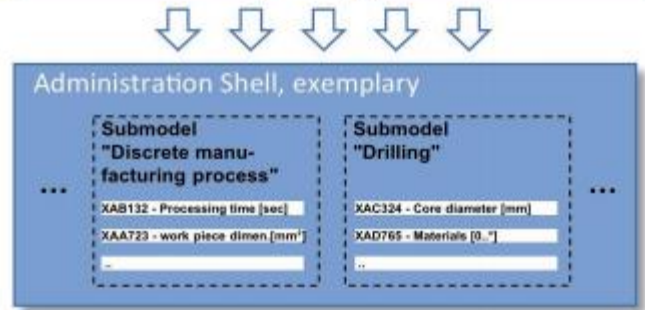
I4.0-compliant communication



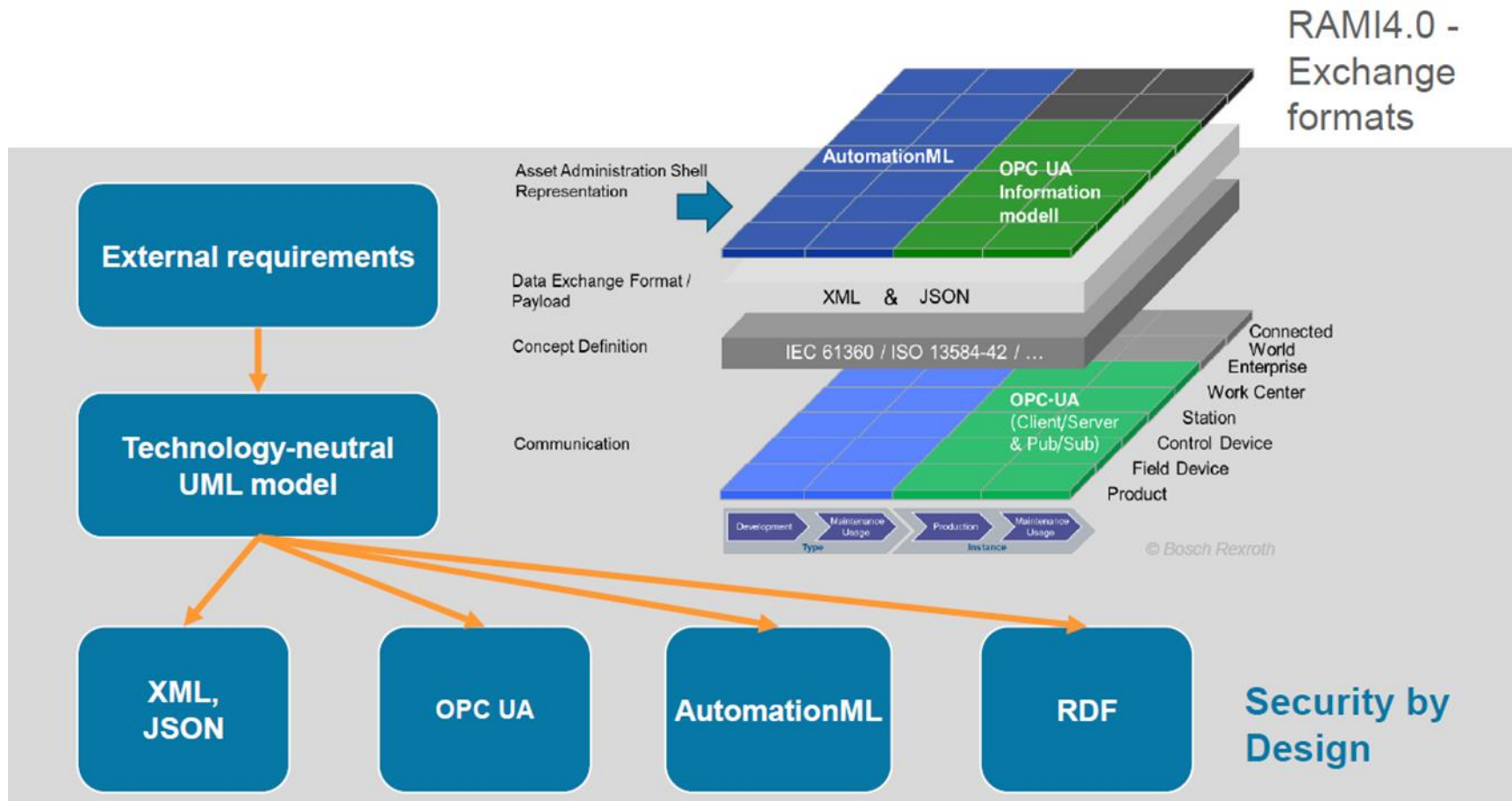
Architecture goal:
Separation of concern



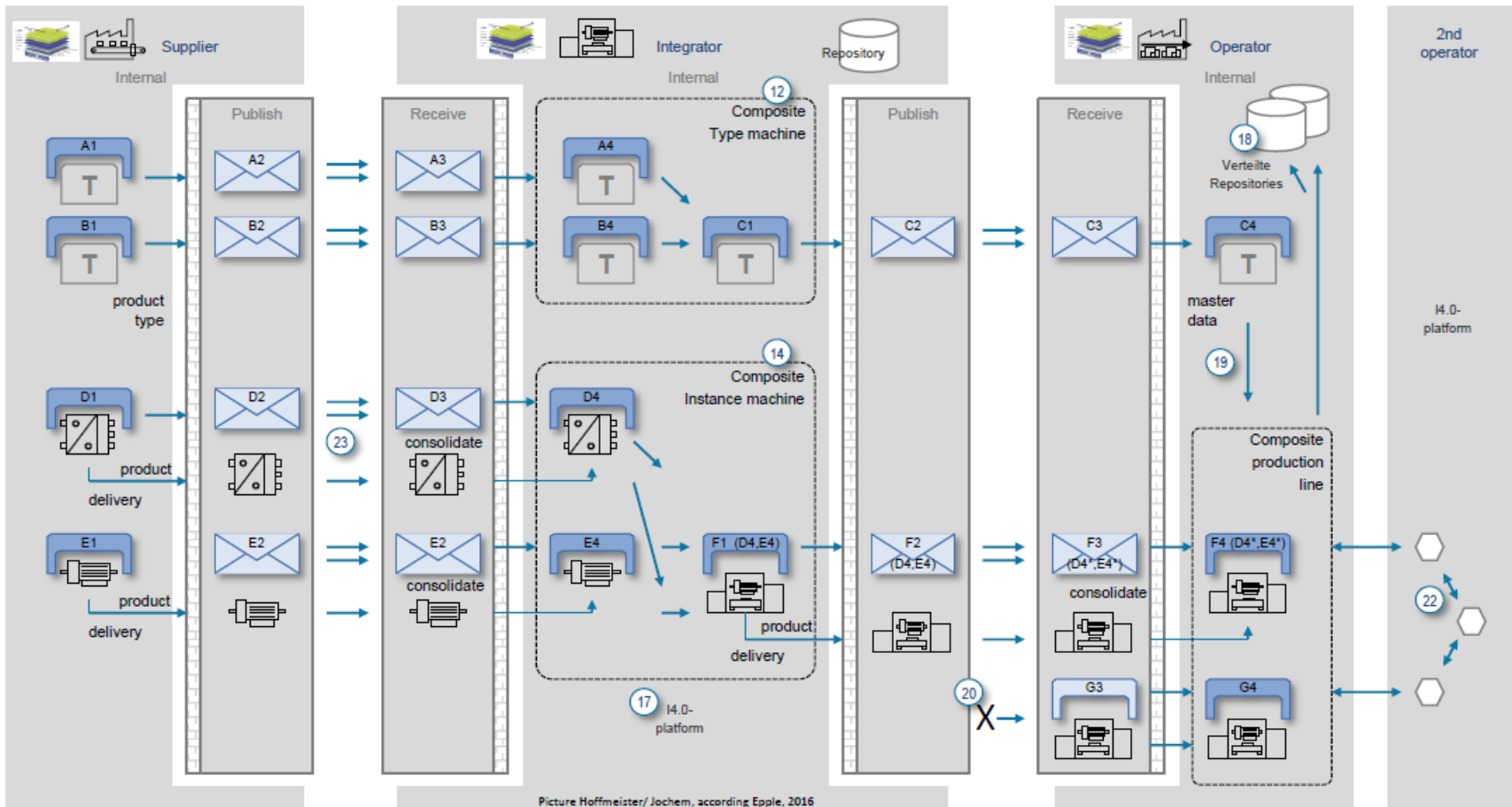
»Is a manufacturing process **Drilling** possible, having work piece dimension $\leq 500 \times 300 \text{mm}$, a core diameter of $\leq 12 \text{mm}$, a material V2A steel, taking processing time $\leq 3 \text{sec}$?«



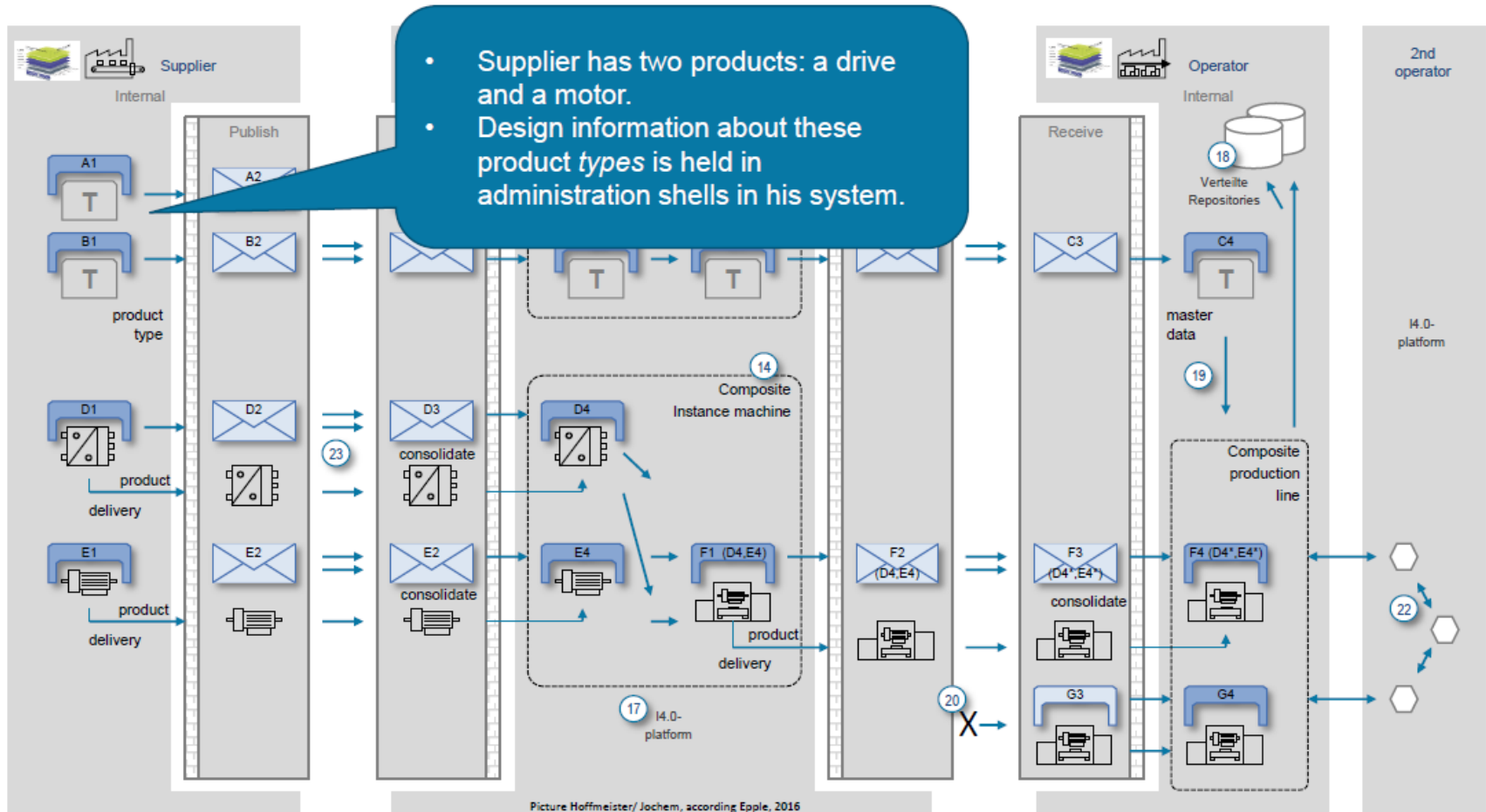
Overview of the Administration Shell



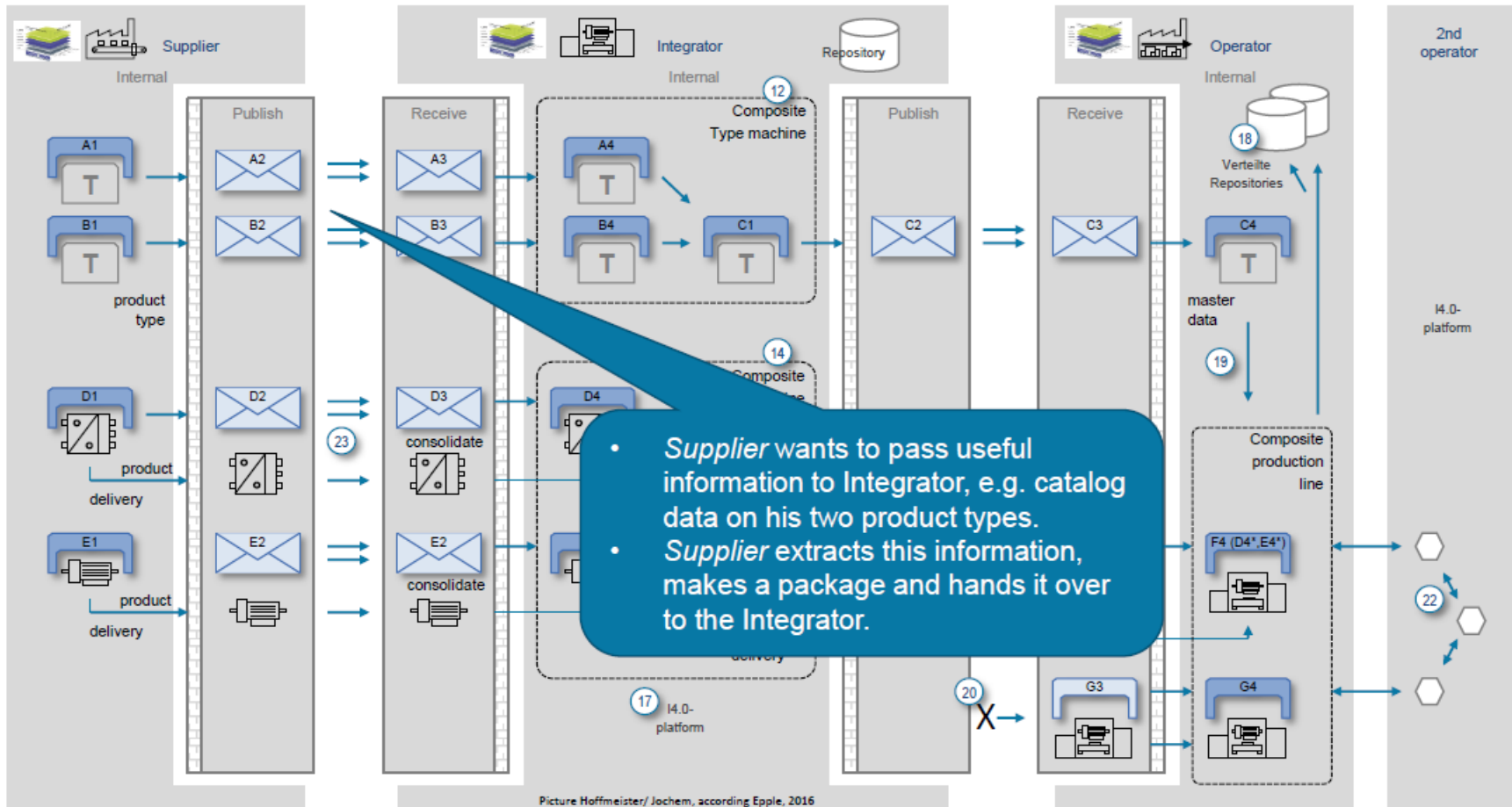
Leading picture for Use Cases: a three step value chain



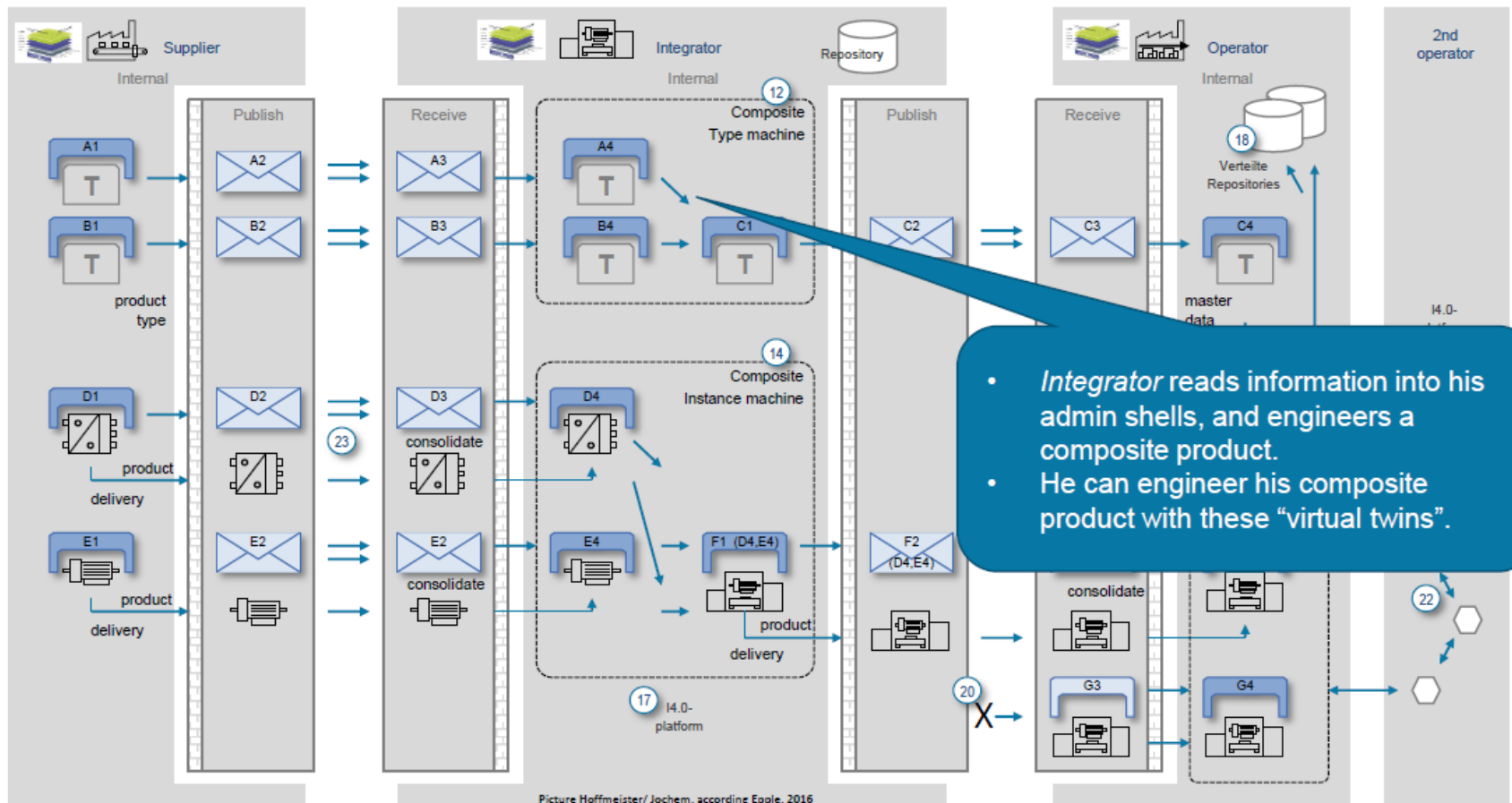
Use Case: Information about Products ("Types")



Use Case: Information about Products ("Types")

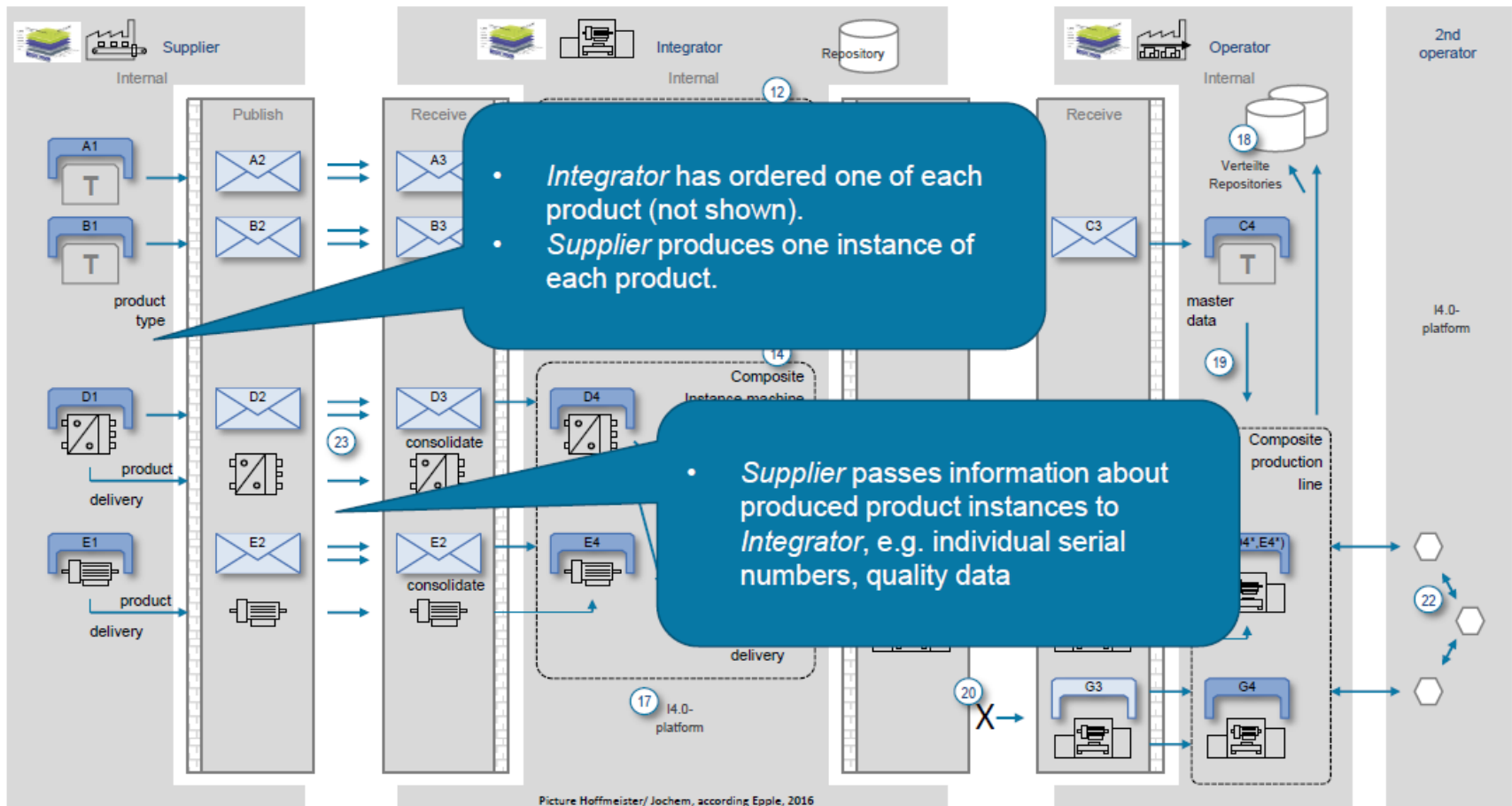


Use Case: Engineering with Product Types

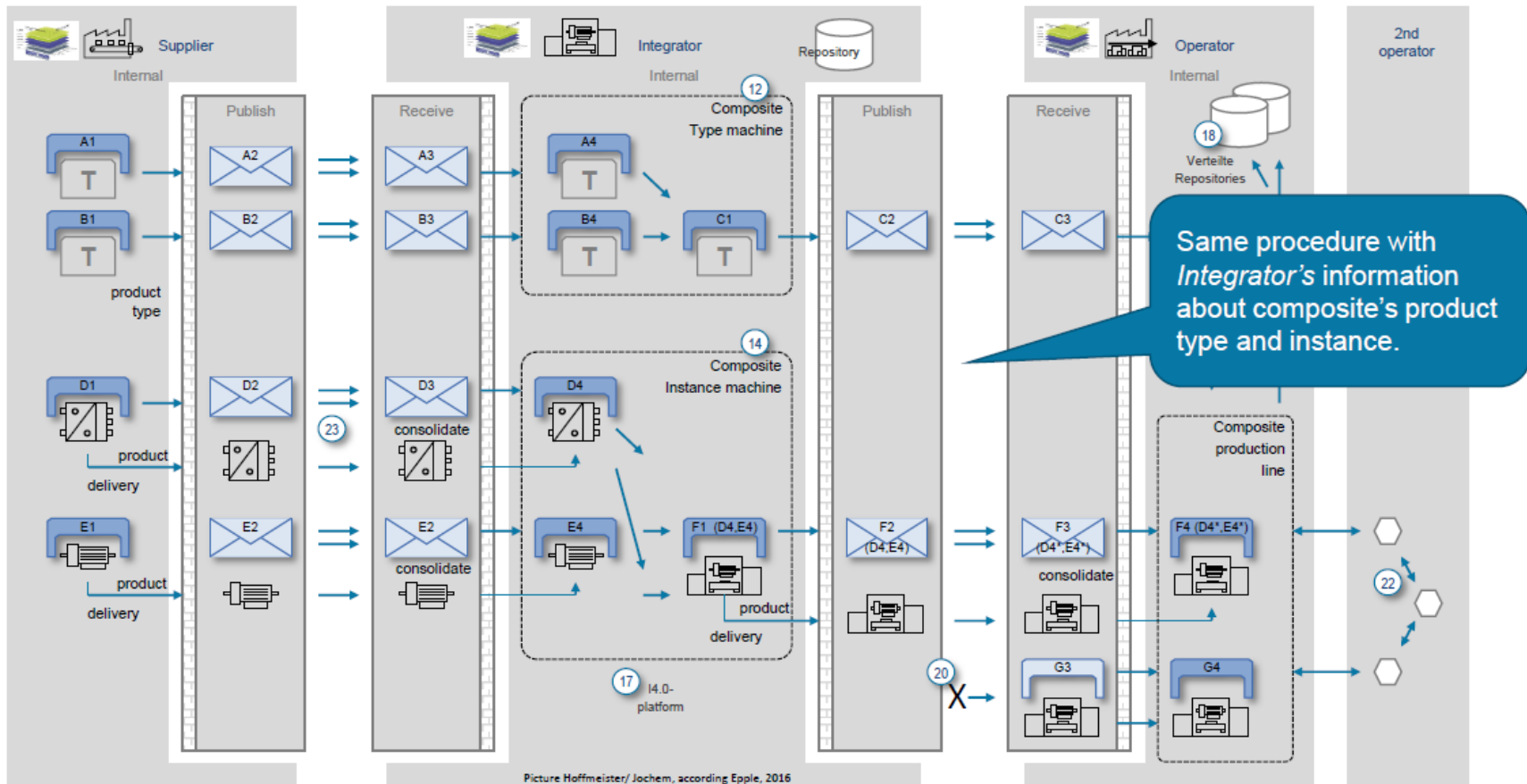


Picture Hoffmeister/ Jochem, according Epple, 2016

From Design to Reality: "Types" become "Instances"

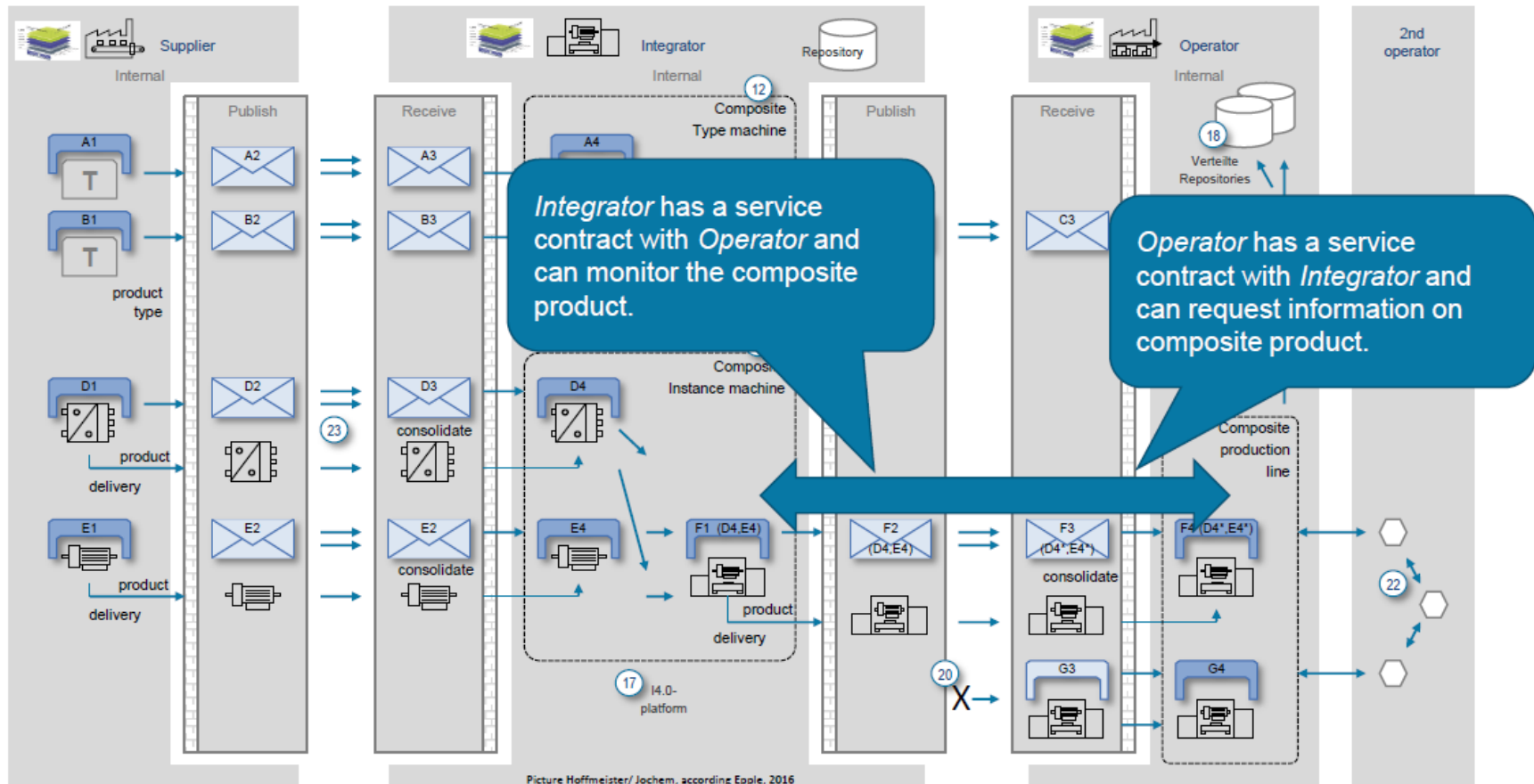


Leading picture for Use Cases: a three step value chain

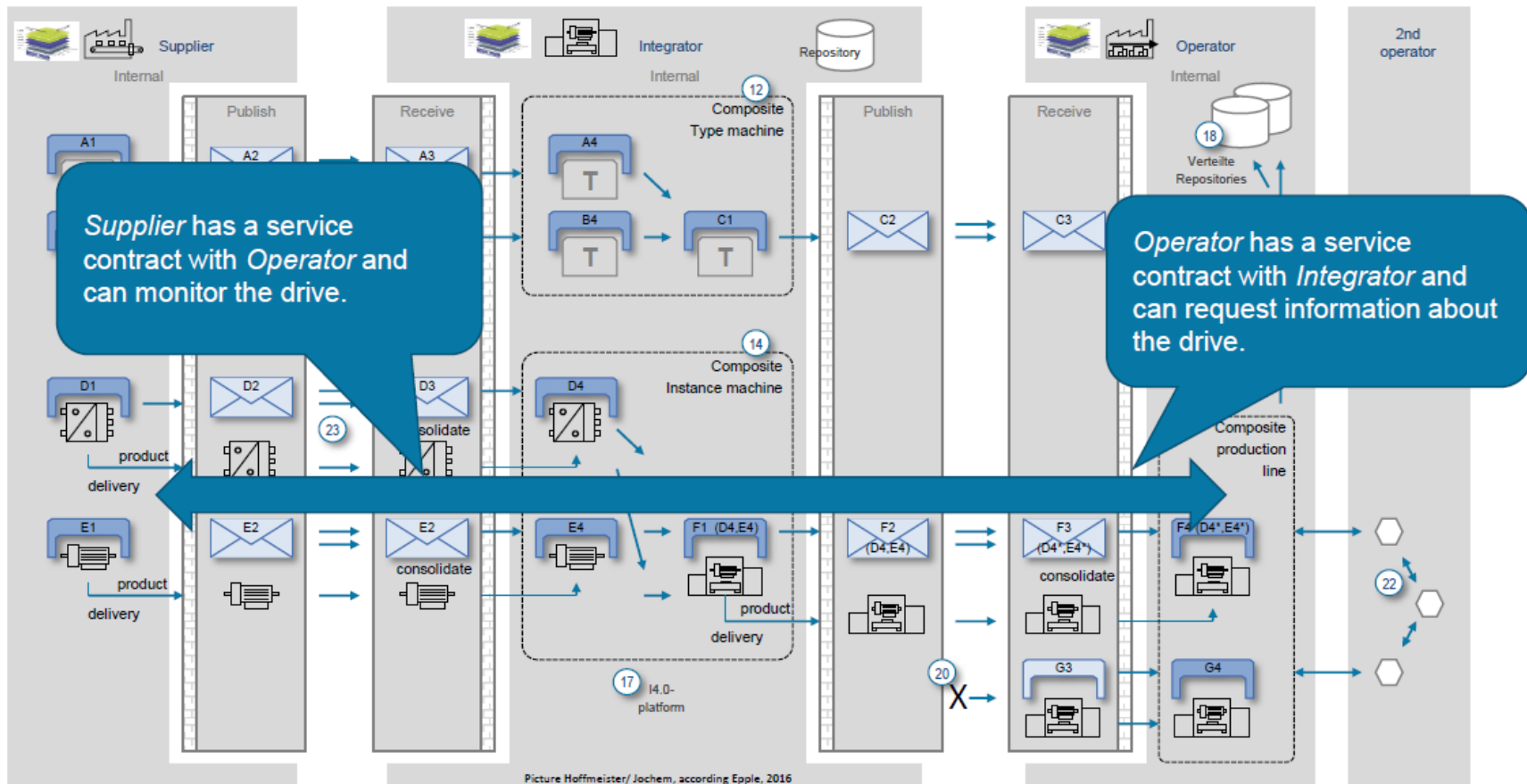


Picture Hoffmeister/ Jochem, according Epple, 2016

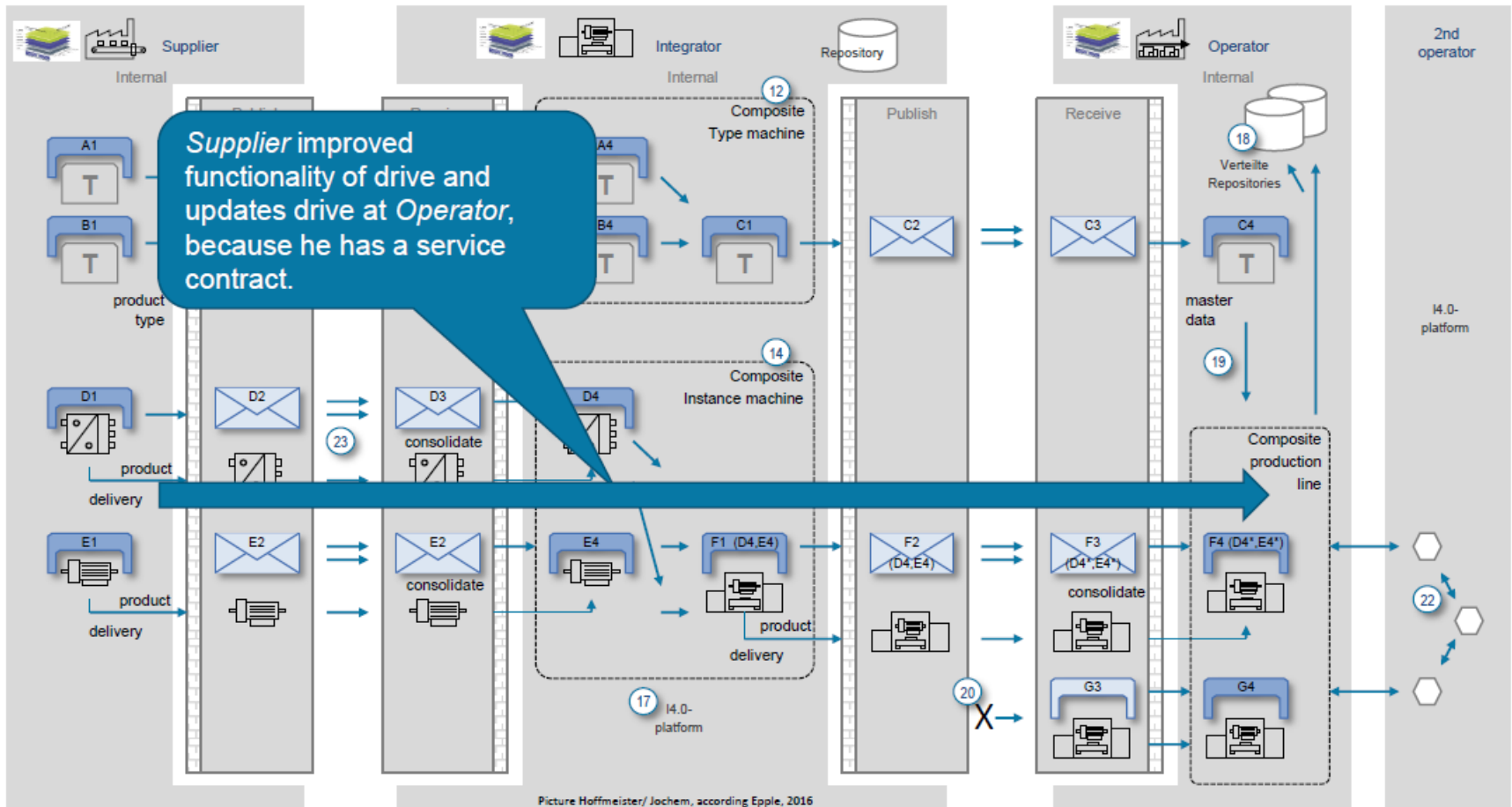
Use Case: Remote Monitoring & Asset Health



Use Case: Remote Monitoring & Asset Health

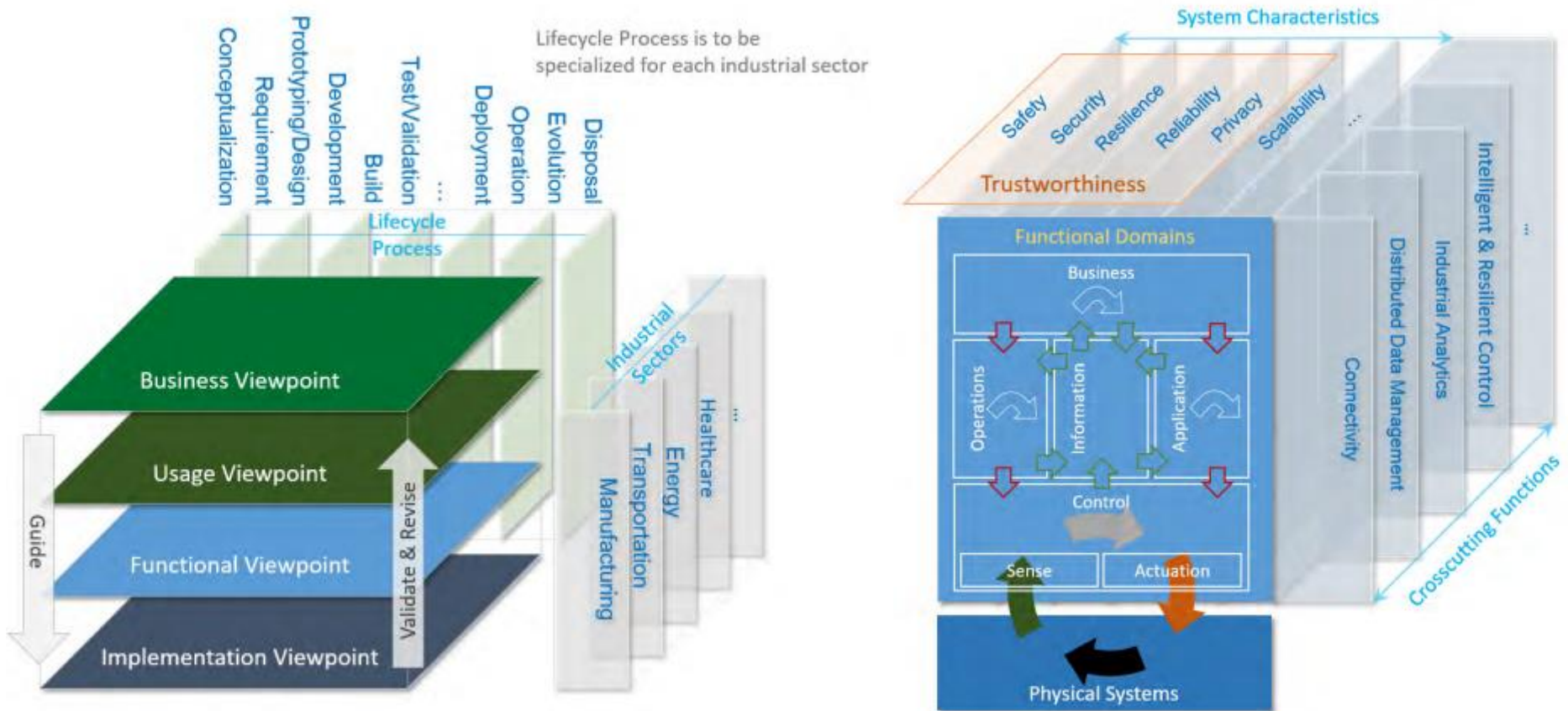


Use Case: Service Contracts



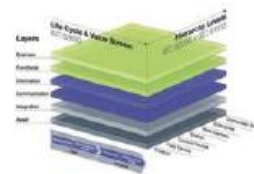
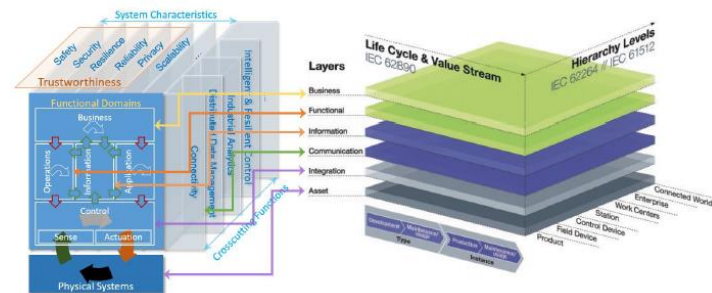
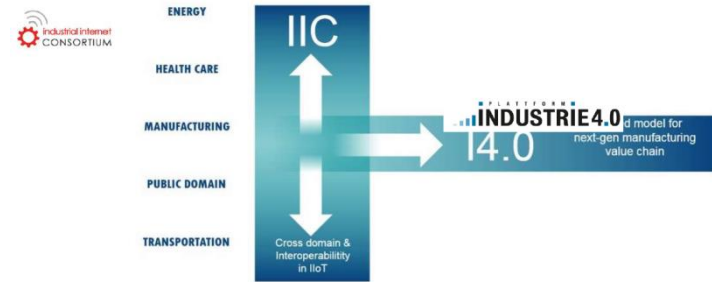
Other initiatives

Industrial Internet Reference Architecture (IIRA)

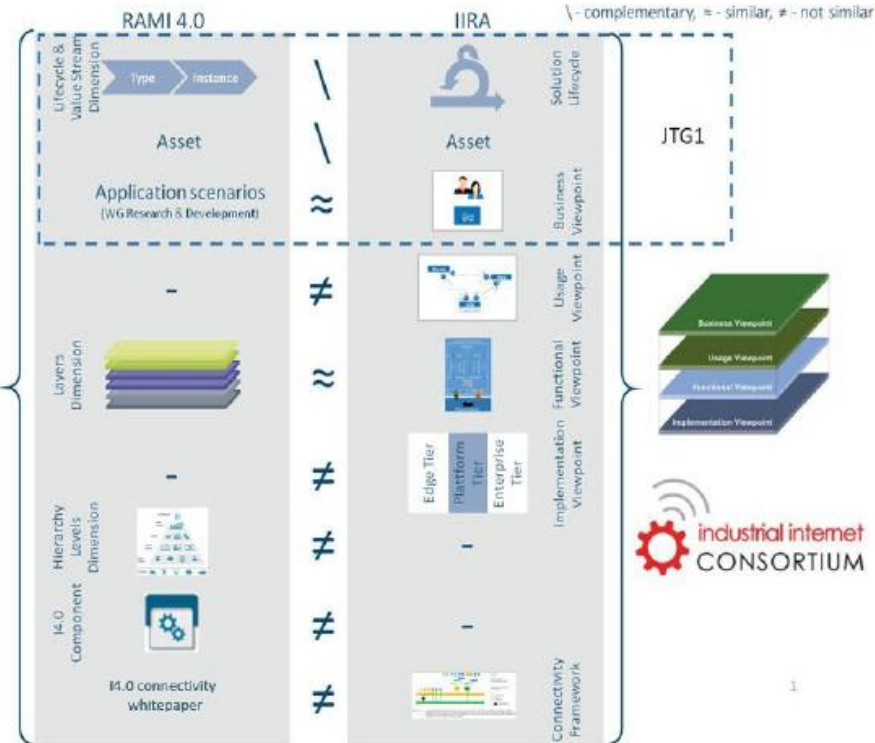


Mapping of IIRA and RAMI 4.0

Industrie 4.0 and IIC (Quelle: Plattform Industrie 4.0)

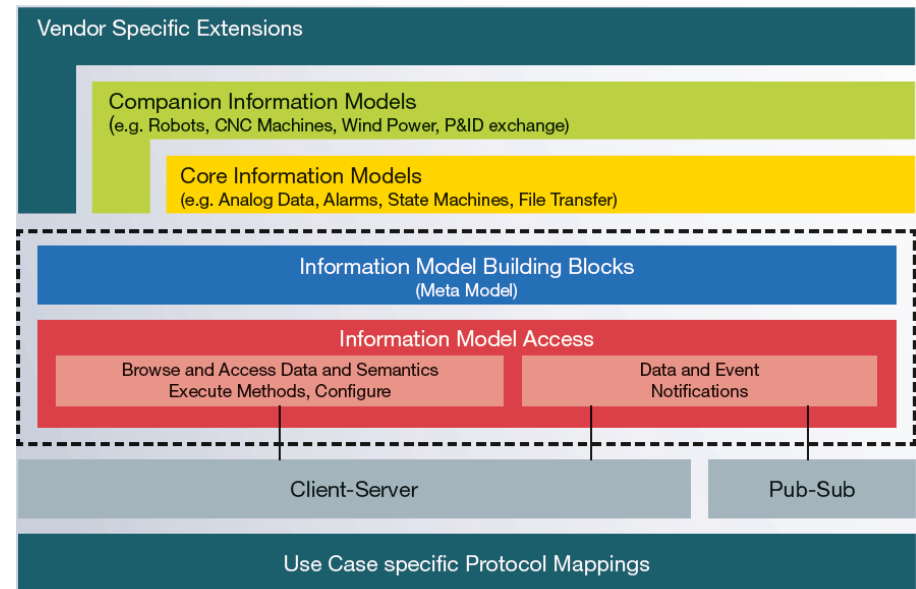


INDUSTRIE 4.0



OPC Unified Architecture (OPC UA)

- The OPC Unified Architecture (UA), released in 2008, is a **platform independent service-oriented architecture** that integrates all the functionality of the individual OPC Classic specifications into one extensible framework.
- This multi-layered approach accomplishes the original **design specification goals** of:
 - **Functional equivalence:** all COM OPC Classic specifications are mapped to UA
 - **Platform independence:** from an embedded micro-controller to cloud-based infrastructure
 - **Secure:** encryption, authentication, and auditing
 - **Extensible:** ability to add new features without affecting existing applications
 - **Comprehensive information modelling:** for defining complex information




Industrial Data Space (IDS)

Internet of Everything

Today

DATA PROVISIONING

- Interoperability
- Asset Digitisation
- Networks
- Processing




Data Economy

Tomorrow

DATA USAGE

- Reference Architecture
- Data Sovereignty
- Standard for CIM
- Data Models
- Data Monetization



Focus on concept and architecture
Focus on open source based implementation

Building Frameworks

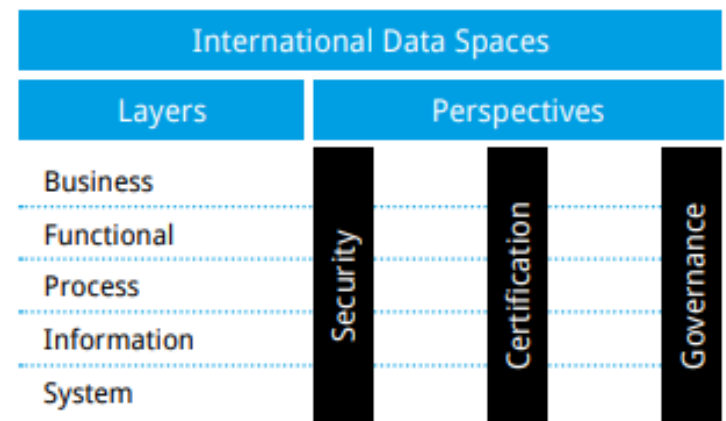
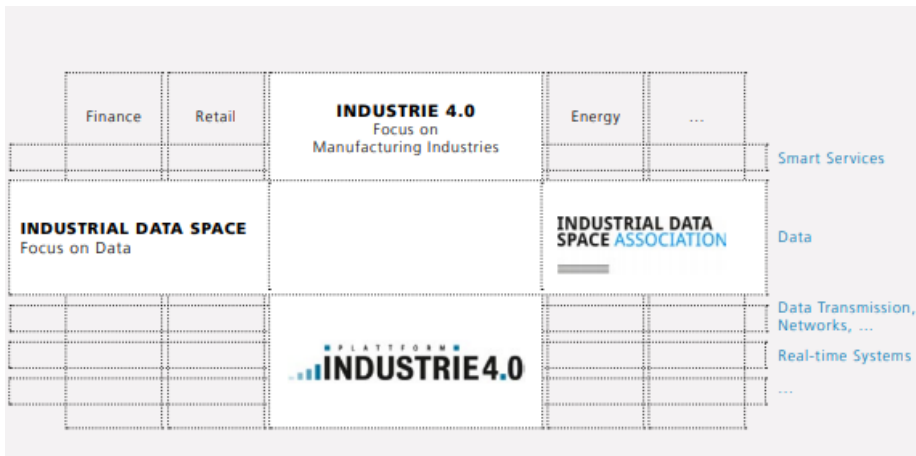
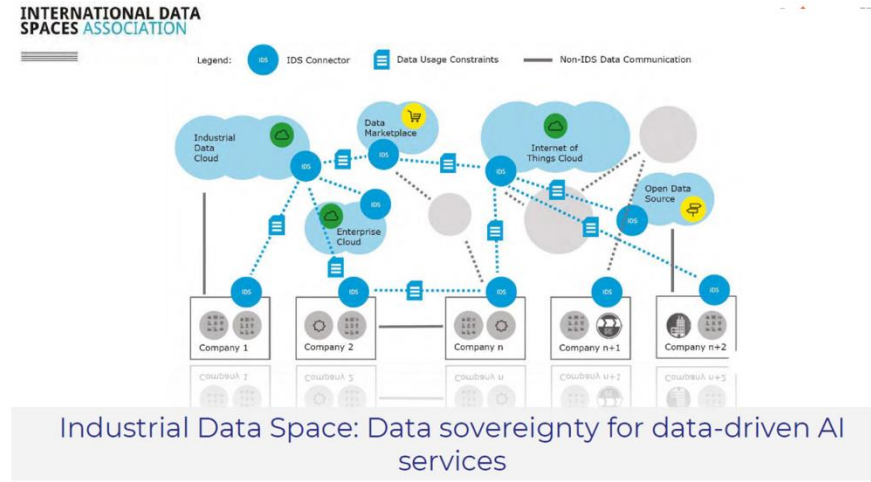
Frameworks of building blocks to assemble smart solutions

Commercial Solutions

Open Source Building Blocks

CEP Context Broker

INGSD



Outline

- Introduction
- Architectures

Technologies

- Algorithms
- Applications

(Big) Data is the new oil

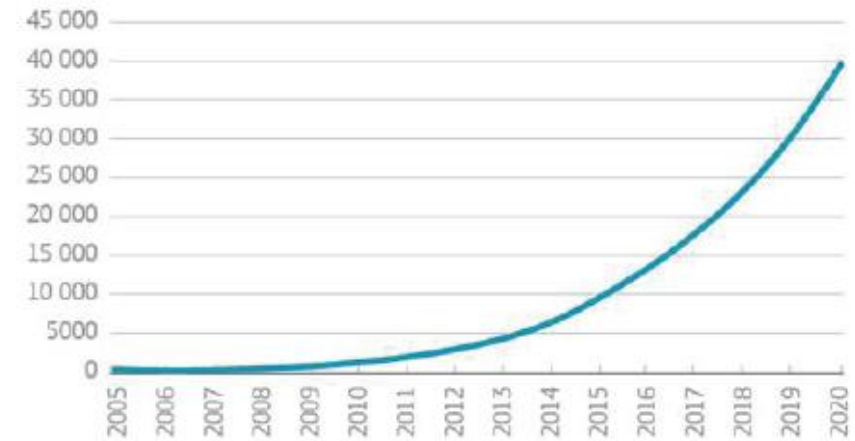
Emergence of Big Data

DATA AS THE NEW ECONOMIC ASSET

- Data is rapidly becoming the lifeblood of the global economy
- Gartner estimates there are currently about **4.9 billion connected devices** generating data
 - This is expected to reach 25 billion next years.
- The real value is no longer in the product, as such, but in the **opportunities** it can offer to users in terms of accessing information and experiences

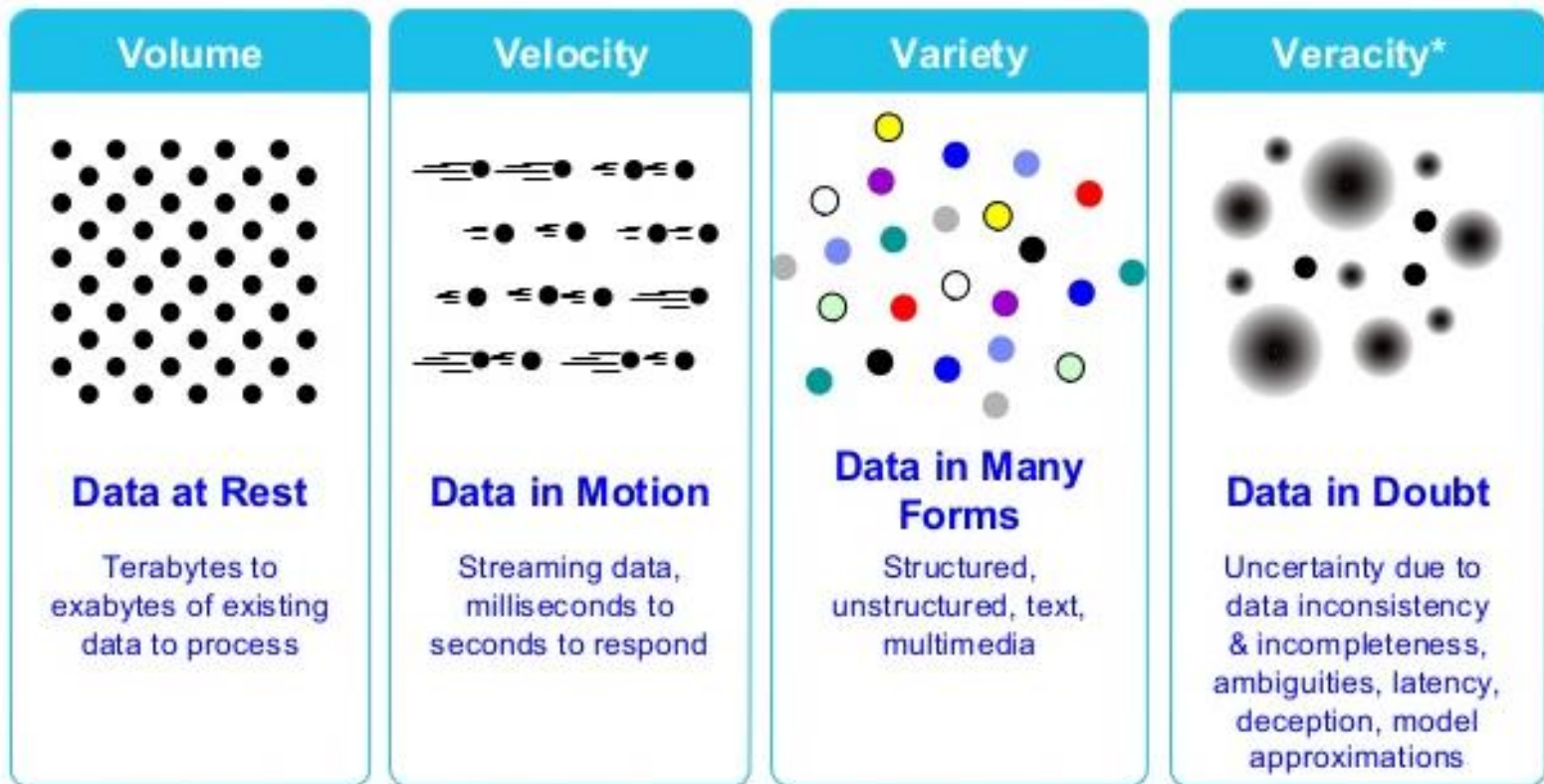
GLOBAL EXPLOSION OF DATA

Worldwide data storage in exabytes



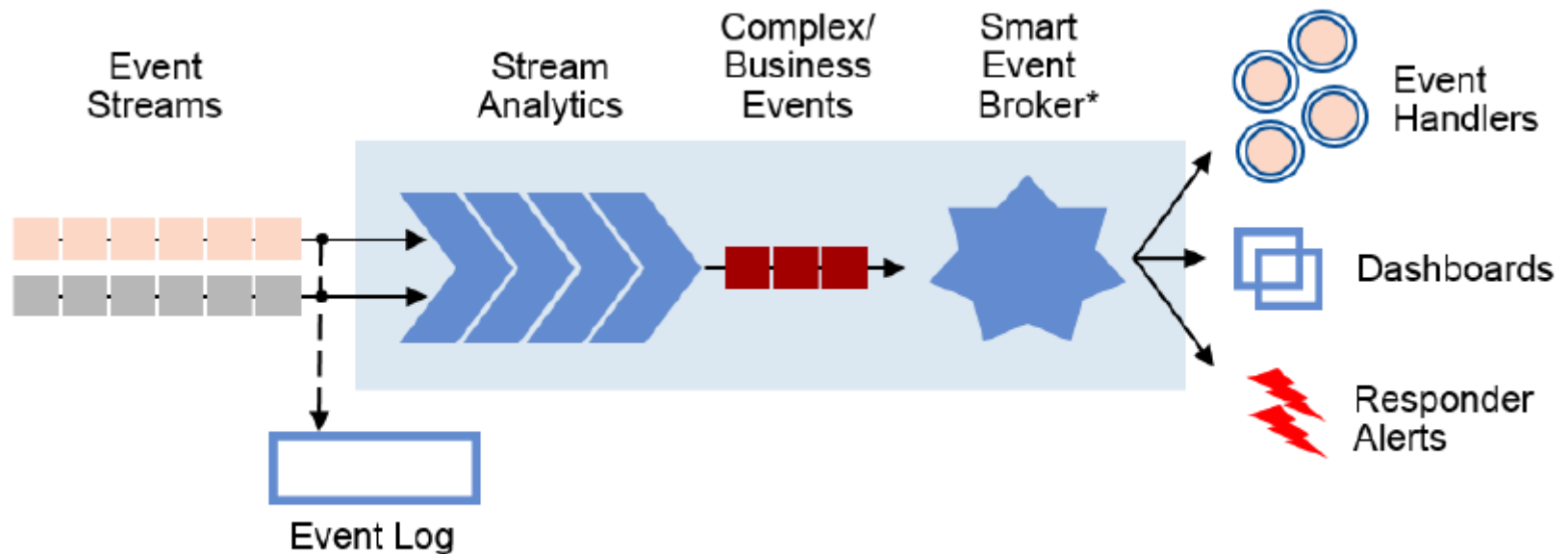
Source: International Data Corporation Digital University Study

The 4 Vs of Big Data



~~(Big) Data Events is the new oil~~

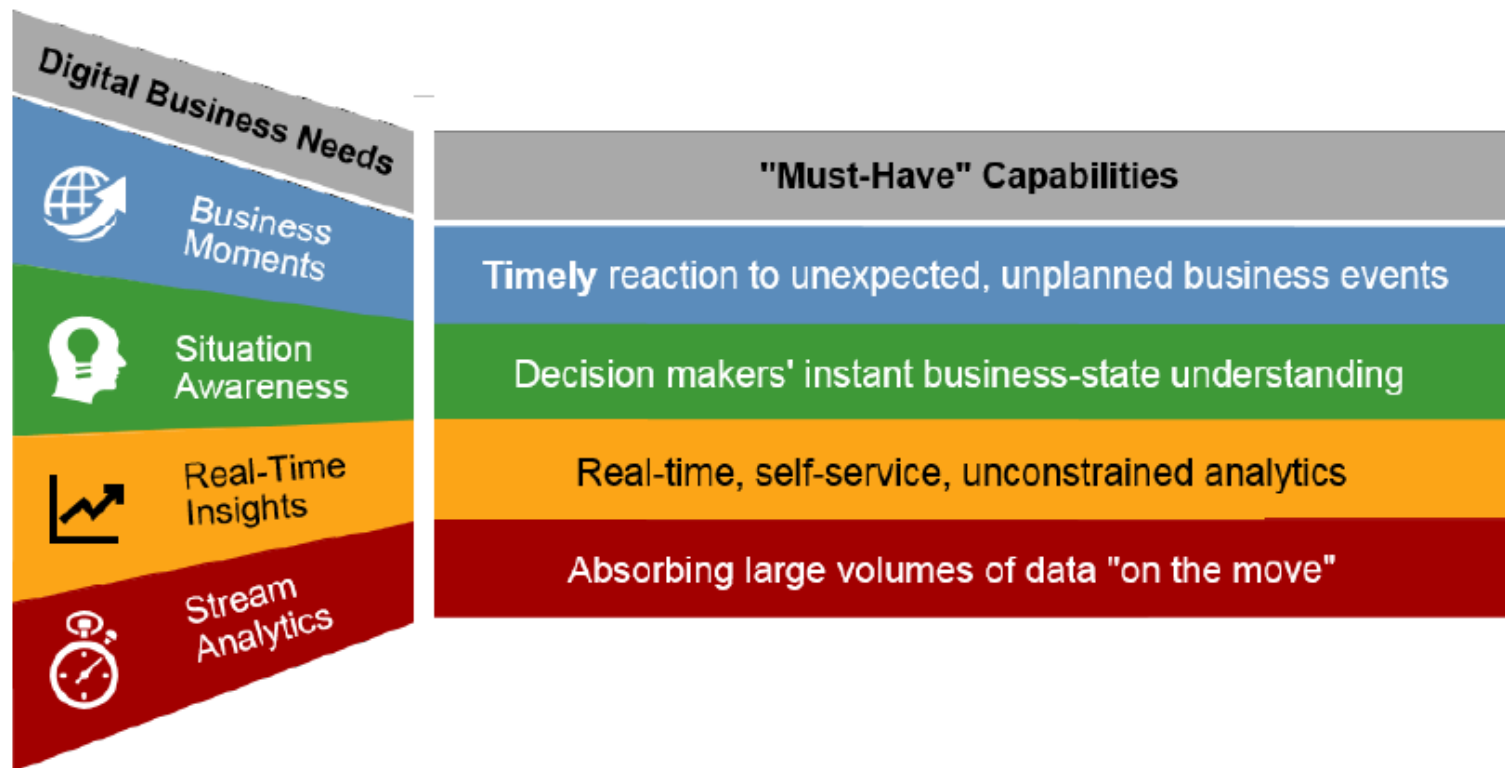
“Sense & Respond” Event-driven Architectures



* May include analytics, logging, integration, other mediation and routing of events to subscribers

**“Data-in-motion” (Big data + Events) is
the new oil**

Data and event stream processing



Plethora of platforms

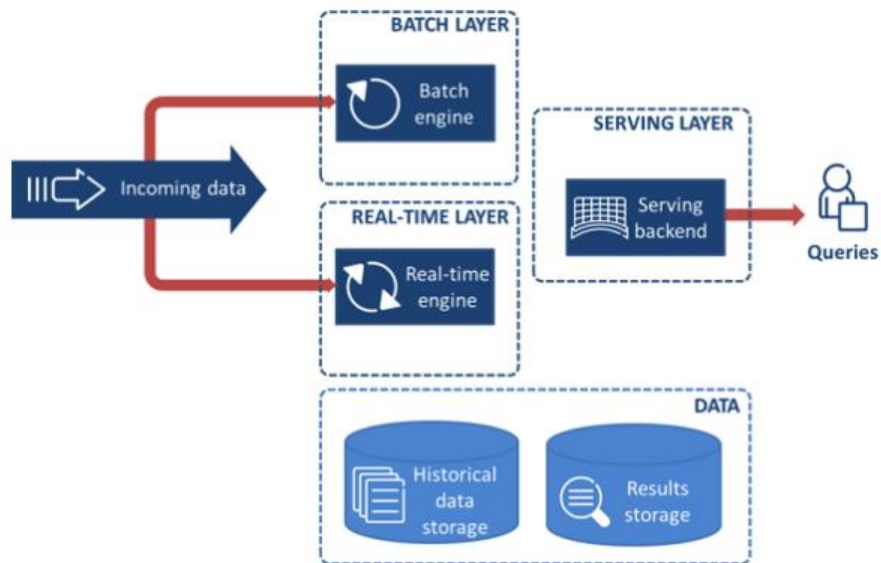
- Event processing
 - Esper (EsperTech)
 - Drools (Red Hat)
 - Apama (Software AG)
 - Business Events (Tibco)
 - IBM Streams ...
- Stream Computing
 - Apache Storm
 - Apache Flink
 - Apache Spark Streaming
 - Apache Samza
 - Twitter Heron ...
- Stream Data integration Platforms
 - Apache Kafka Streams
 - Apache Beam
 - Google Cloud Dataflow
 - Apache Gearpump
 - ...

The Technology Landscape is flourishing

- More than 35 vendors in this market segment
- Trends
 - Open source
 - E.g. Confluent (Apache Kafka), data Artisans (Apache Flink), Databricks (Apache Spark Streaming), DataTorrent (Apache Apex) etc.
 - Hybrid products
 - E.g. FICO Data Management Integration Platform (DMIP), Hortonworks DataFlow, Impetus Technologies StreamAnalytix, Rapidminer Streams, and Salesforce Thunder leverage Apache Storm
 - Cloud-enablement
 - Amazon Kinesis Stream Analytics, Microsoft Azure Stream Analytics and Salesforce now offer similar services

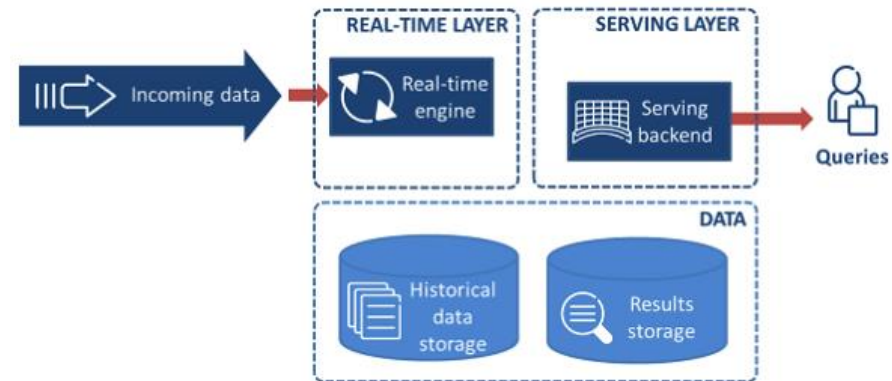
Lambda architecture

Lambda architecture



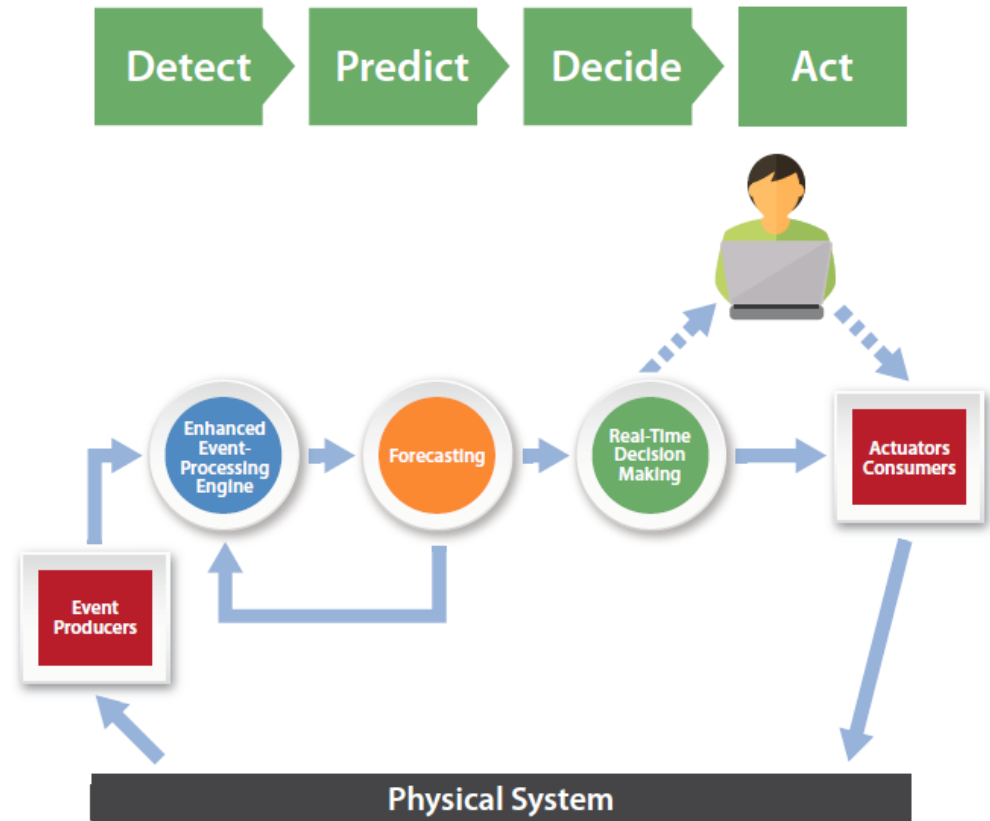
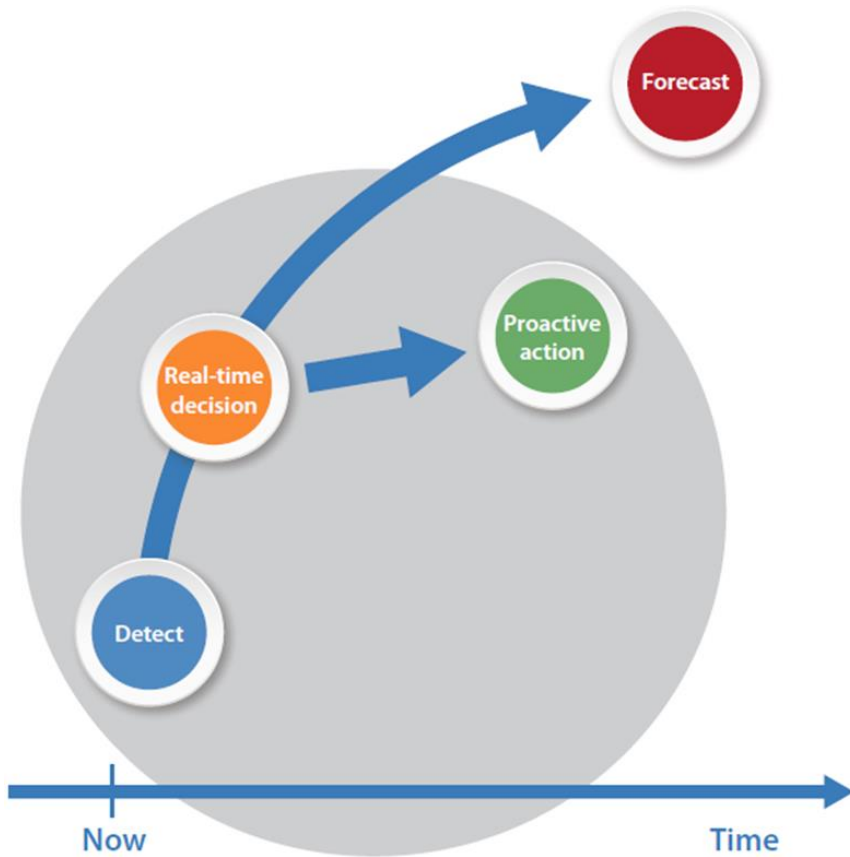
- Developed by Nathan Marz, creator of Apache Storm (2011)
- Used by Twitter and Spotify

Kappa architecture



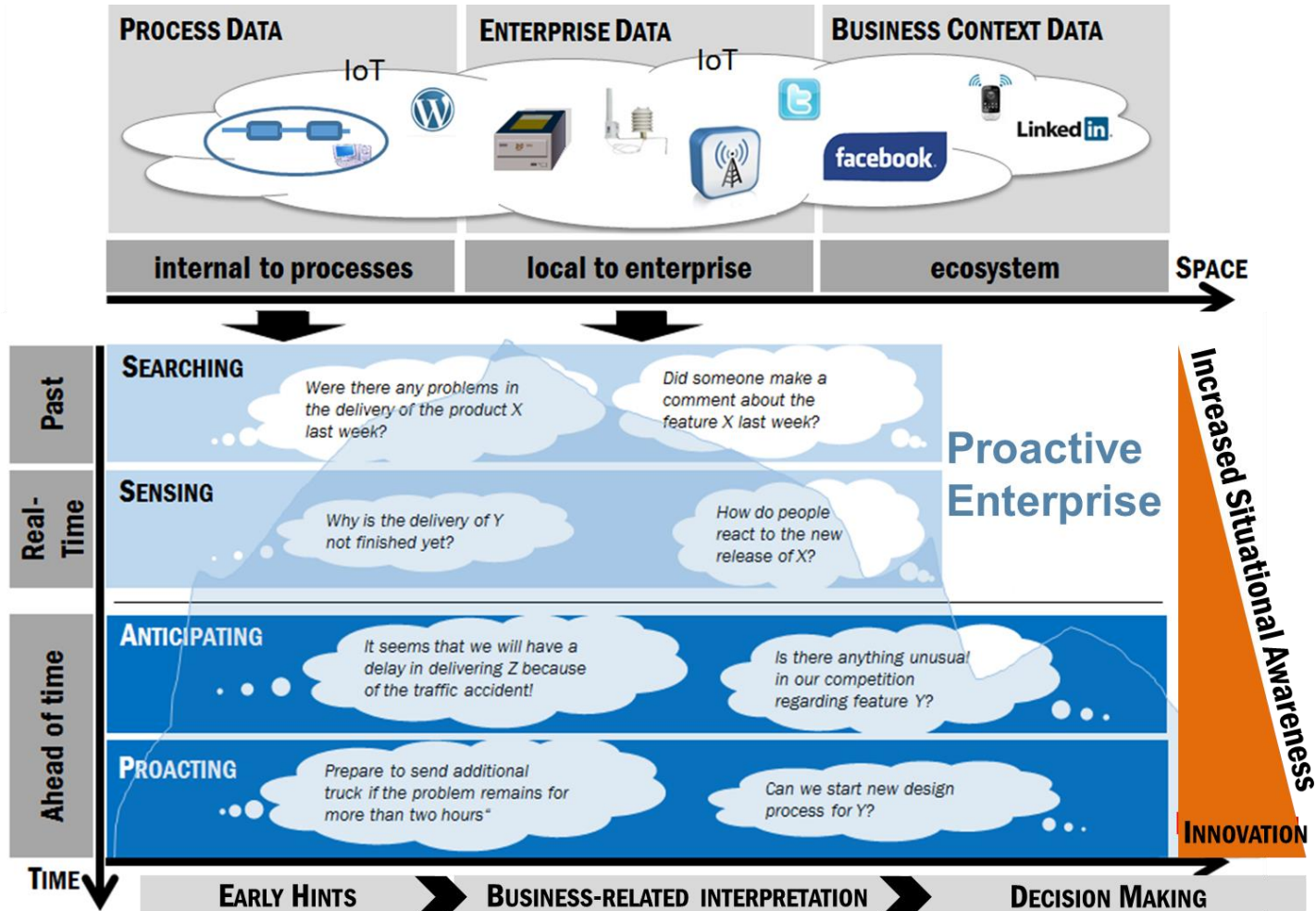
- Developed by Jay Kreps, creator of Apache Kafka (2014)
- Used by LinkedIn and Yahoo

Proactive computing



Source: Etzion O. (2016). Proactive Computing: Changing the Future. RTInsights.

The Proactive Enterprise



Outline

- Introduction
- Architectures
- Technologies

Algorithms

- Applications

Data analytics

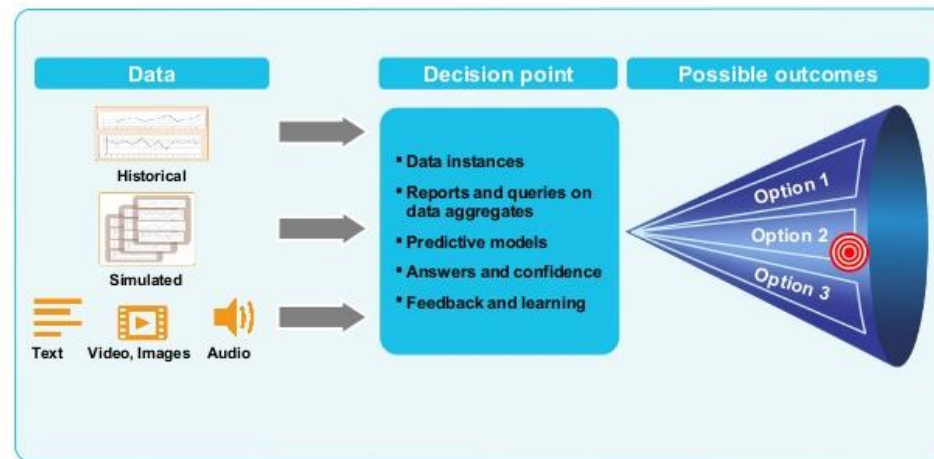
Data-in-Motion is the new oil, but ...



99%

Manufacturing Data Value is Lost

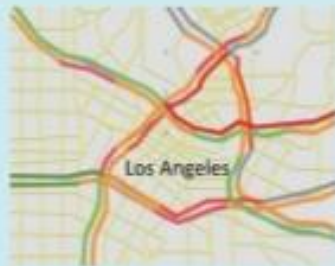
80% of all available data are uncertain



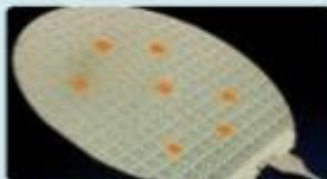
Uncertainty arises from many sources

Process Uncertainty

Processes contain
"randomness"



Uncertain travel times



Semiconductor yield

Data Uncertainty

Data input is uncertain



GPS Uncertainty



Testimony



{Paris Airport}

Ambiguity



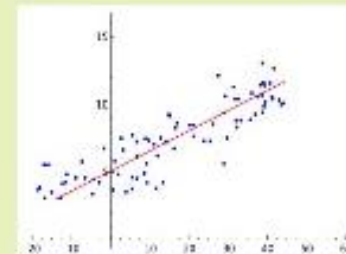
Contaminated?
Rumors

{John Smith, Dallas}
{John Smith, Kansas}

Conflicting Data

Model Uncertainty

All modeling is approximate



Fitting a curve to data

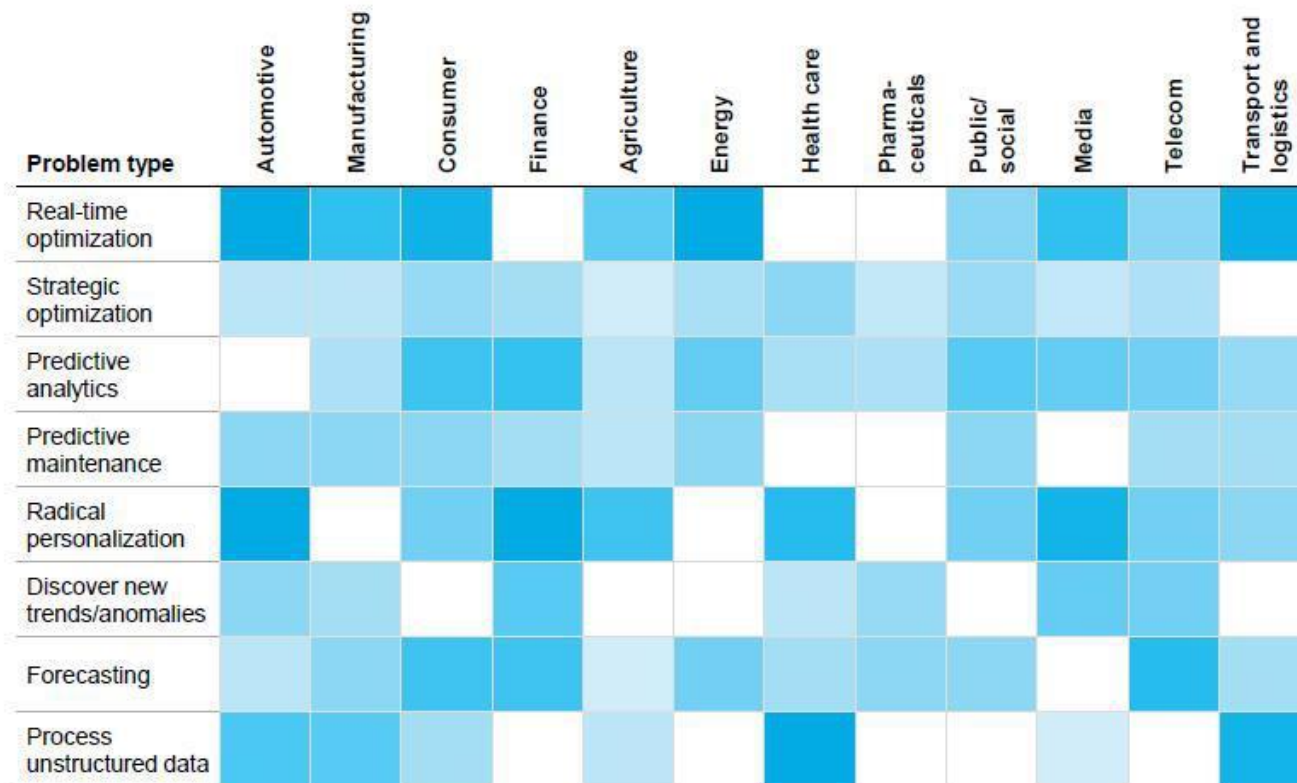


Forecasting a hurricane
(www.noaa.gov)

Potential in manufacturing

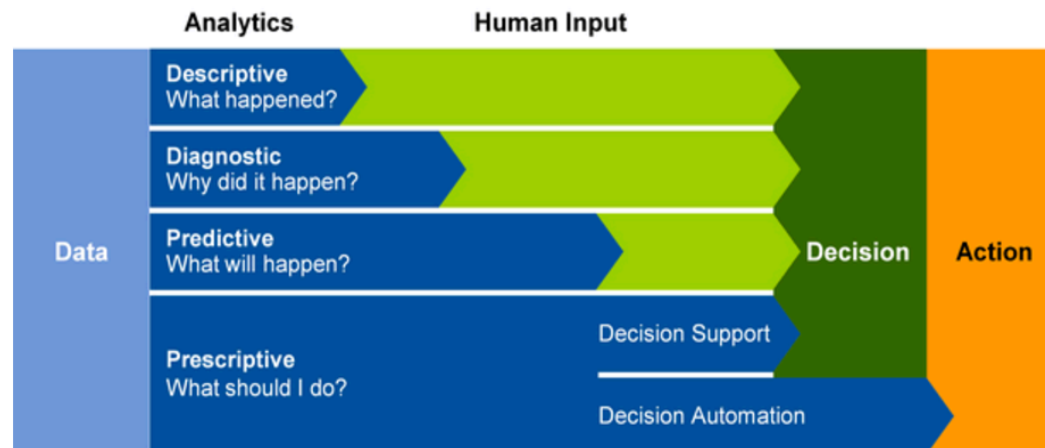
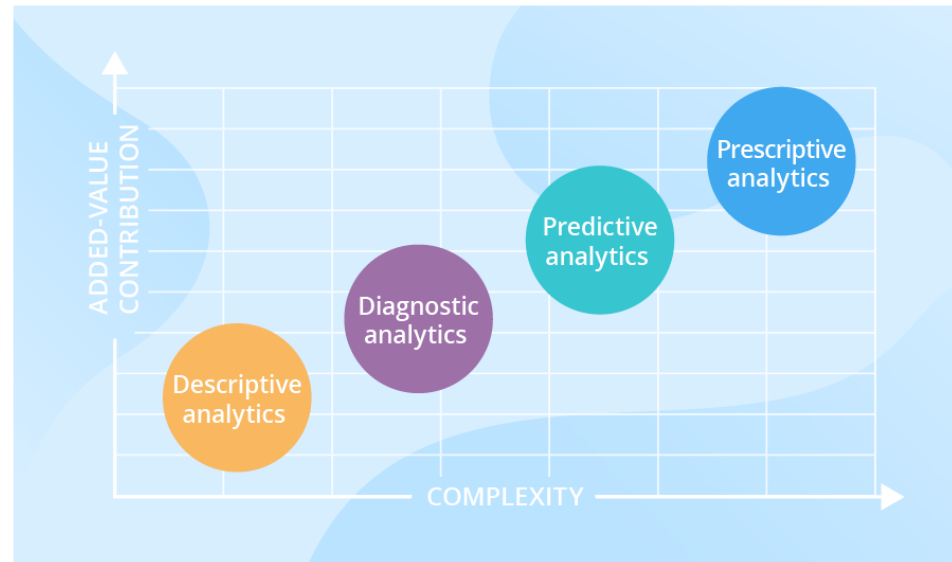
Rich data is an enabler in some use cases but the lack of it can be a barrier in others

Data richness
Low  High

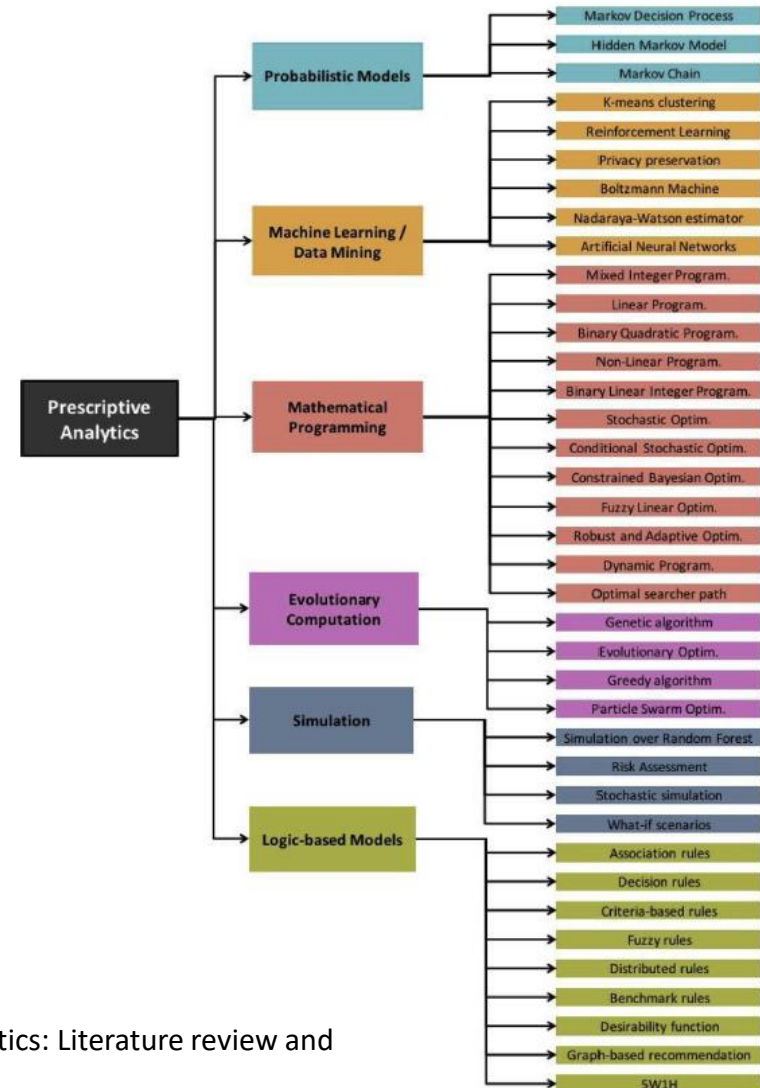
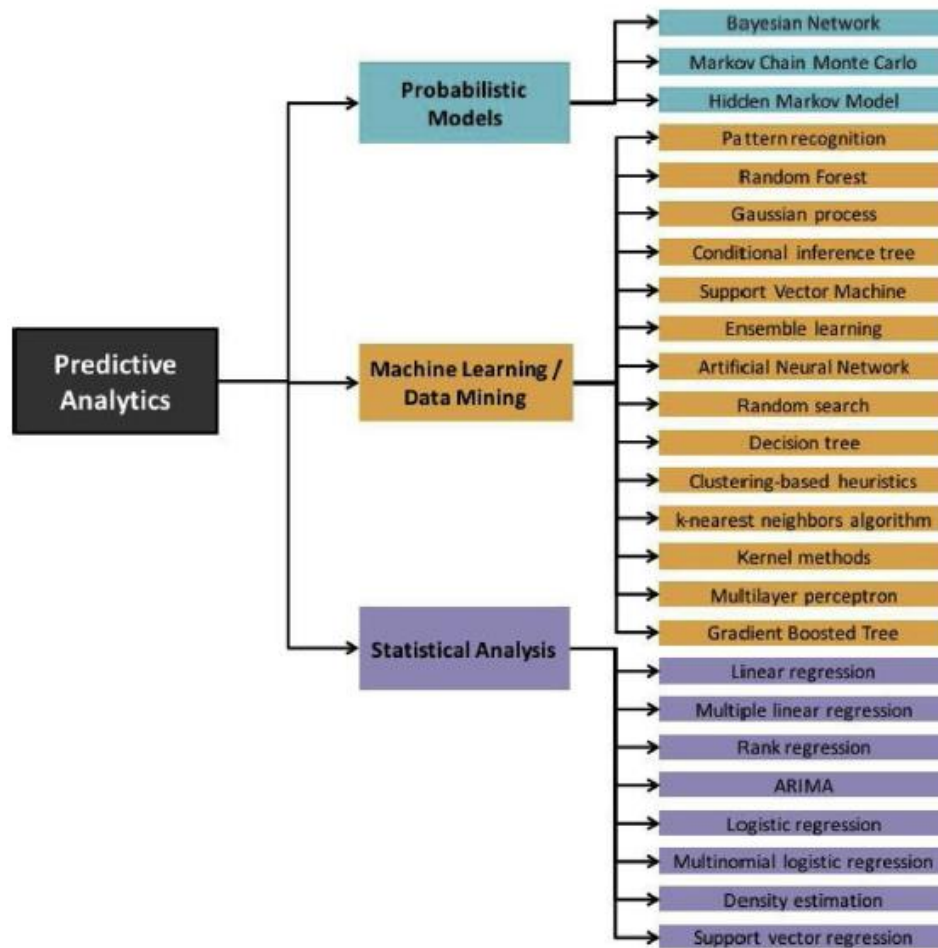


SOURCE: McKinsey Global Institute analysis

Data analytics categories



Classification of methods



Big data challenges in manufacturing



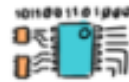
SMART FACTORY

SCENARIO RESEARCH
CHALLENGES V1.0



DATA MANAGEMENT & LIFECYCLE

CPS data sources integration
Systems semantic interoperability
Smart factory data annotation
Smart factory unstructured, semi-structured and missing data
Industrial IOT data availability



DATA PROCESSING ARCHITECTURES

On-premise / cloud architectures
Hybrid clouds and edge automation
Data in motion / data at rest
integration.



DATA ANALYTICS

Prescriptive analytics in industrial plants
Machine and deep learning
Analytics for data-human interaction
Analytics-based decision support
Embedded analytics
Analytics-oriented manufacturing




DATA VISUALISATION & USER INTERACTION

Context-aware visualization
Visual analytics for smart factory
Natural language interaction
interfaces
Cross-domain and data exploration
Simulation and training environments



DATA PROTECTION

Sensitive data privacy
Protection against cyber-attacks
Access control & data integrity
Selective anonymization  [RD.Antal](#)

(Deep) Machine learning and AI

Opportunities of machine learning

Machine learning opportunities in manufacturing

Highest-ranked use cases, based on survey responses	Use case type	Impact	Data richness
Predict failure and recommend proactive maintenance for production and moving equipment	Predictive maintenance	1.3	1.0
Optimize complex manufacturing process in real time—determine where to dedicate resources to reduce bottlenecks and cycle time	Operations/logistics optimization (real time)	1.1	1.0
Predict future demand trends and potential constraints in supply chain	Forecasting	0.8	0.7
Identify design problems in pre-production to reduce ramp-up time to maximum output (i.e., yield ramp)	Predictive analytics	0.6	0.3
Identify root causes for low product yield (e.g., tool-/die-specific issues) in manufacturing	Discover new trends/anomalies	0.5	0.7
Detect defects and quality issues during production using visual and other data	Process unstructured data	0.4	0.7

Machine Learning types

Supervised learning

- Teach the computer how to do something, then let it use its new found knowledge to do it
- Learning with teacher
- Inferring a function from labelled training data
- Training data includes desired outputs (labels)

Unsupervised learning

- Teach the computer how to do something, then let it use its new found knowledge to do it
- Learning without teacher
- Finding hidden structure in unlabelled data
- Training data does not include desired outputs (labels)

Semi-supervised learning

- Training data includes a few desired outputs (labels)

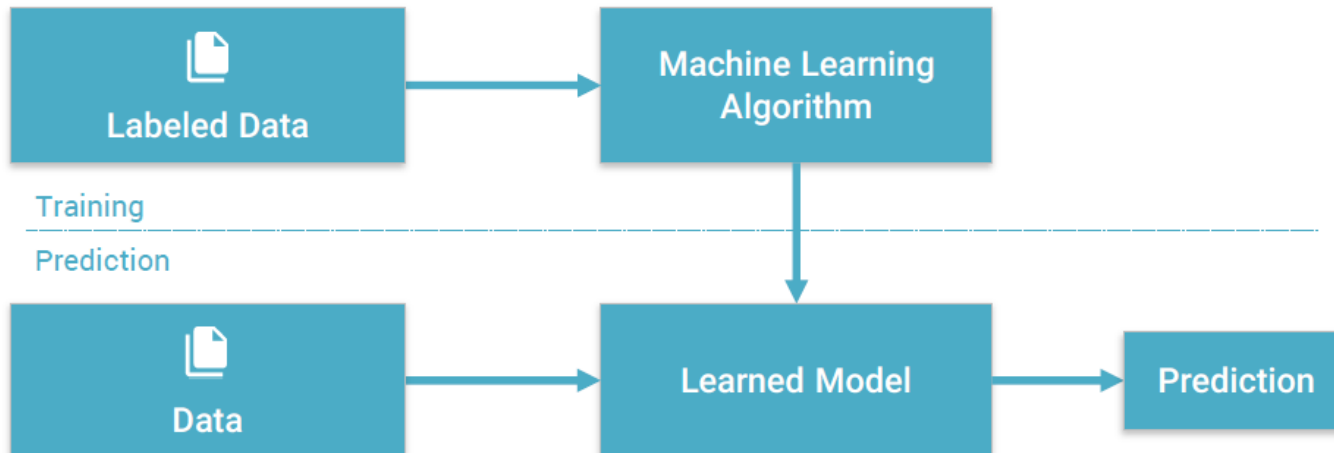
Reinforcement learning

- Rewards from sequence of actions

Machine learning methodology



Machine Learning is a type of Artificial Intelligence that provides computers with the ability to **learn without being explicitly programmed**.



Provides **various techniques** that can learn from and make predictions on data

Some well-known ML algorithms

Regression

- Least Squares Regression
- Linear regression
- Logistic regression

Classification

- Naïve Bayes
- Neural Networks
- Support Vector Machines (SVM)
- Kernel estimation (k-Nearest Neighbors)
- Random forests

Clustering

- K-Means
- Hierarchical clustering
- Spectral clustering
- Density-based algorithms (DBSCAN, OPTICS)

What is deep learning?



Part of the **machine learning** field of learning representations of data. Exceptional effective at learning patterns.

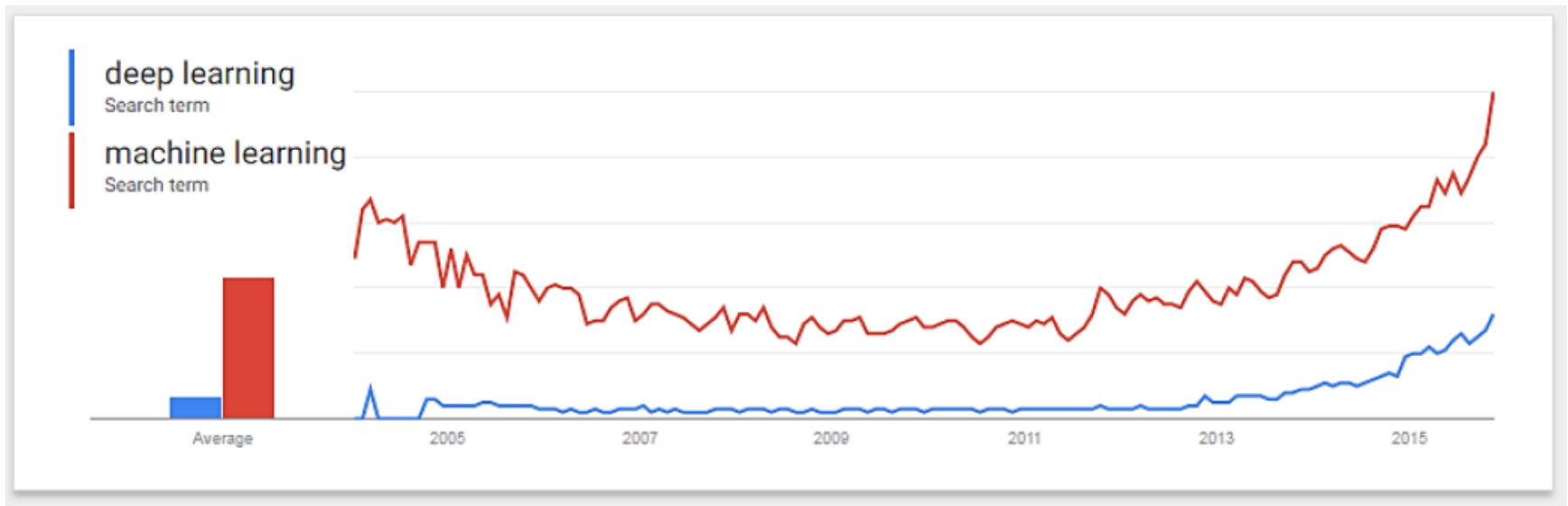


Utilizes learning algorithms that derive meaning out of data by using a **hierarchy** of multiple layers that **mimic the neural networks of our brain**.

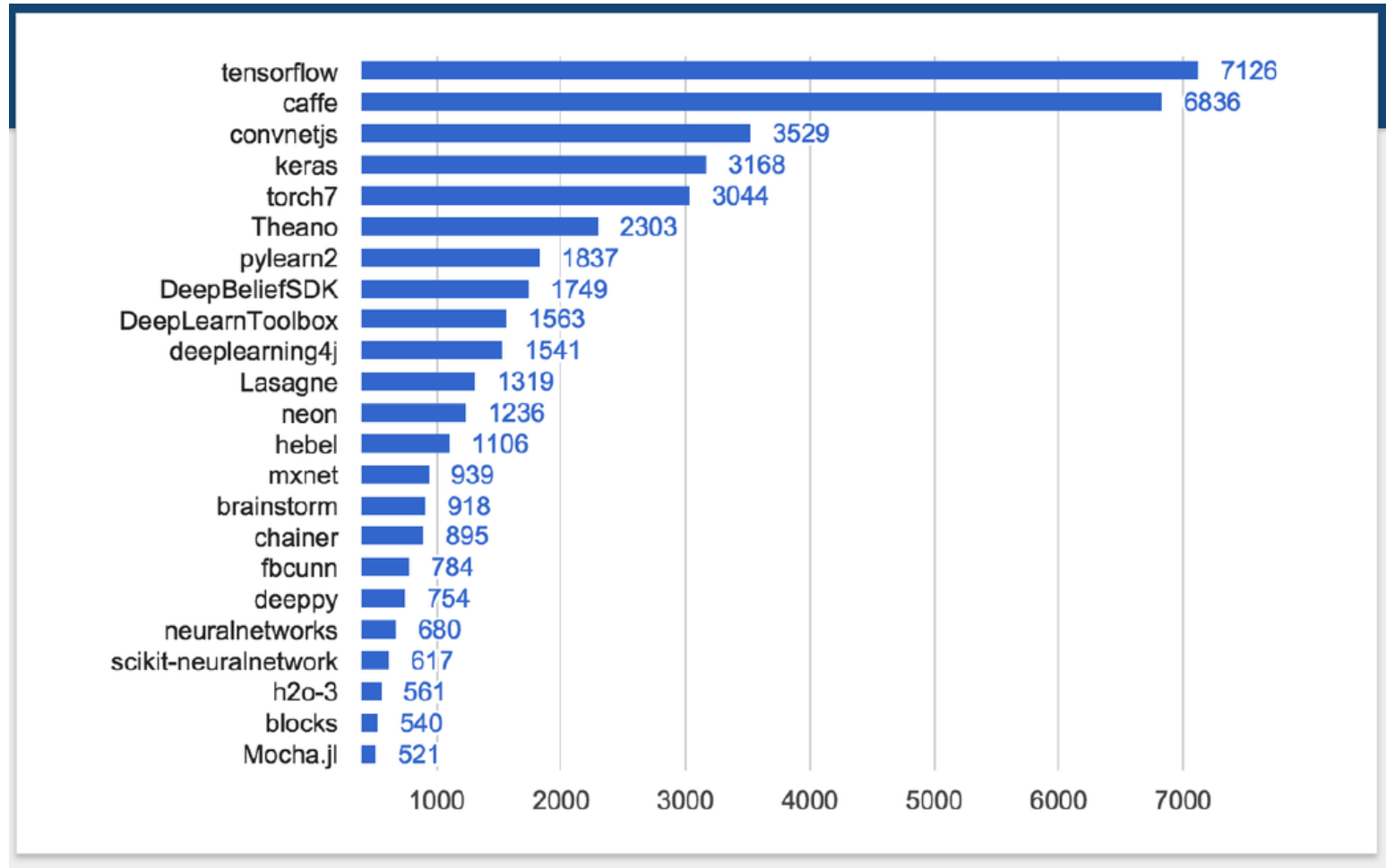


If you provide the system tons of information, it begins to understand it and respond in useful ways.

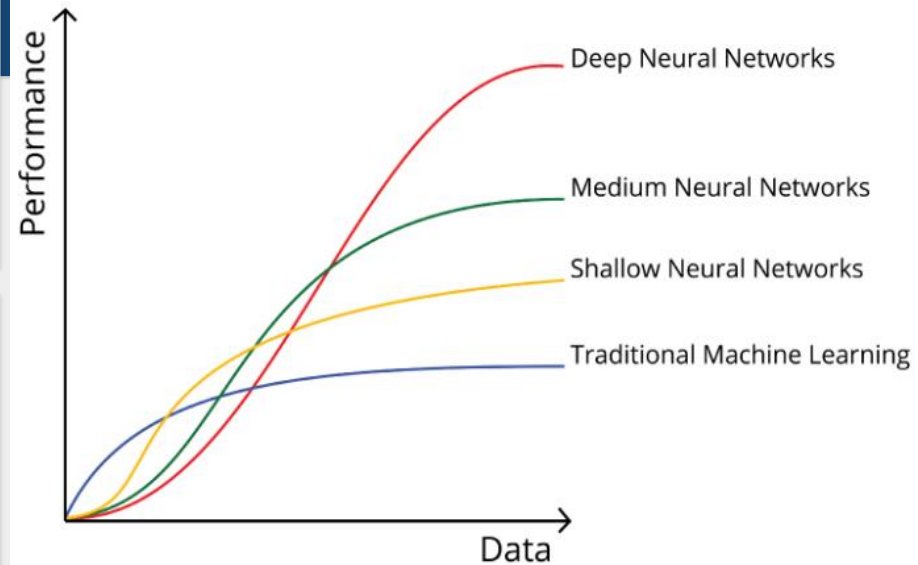
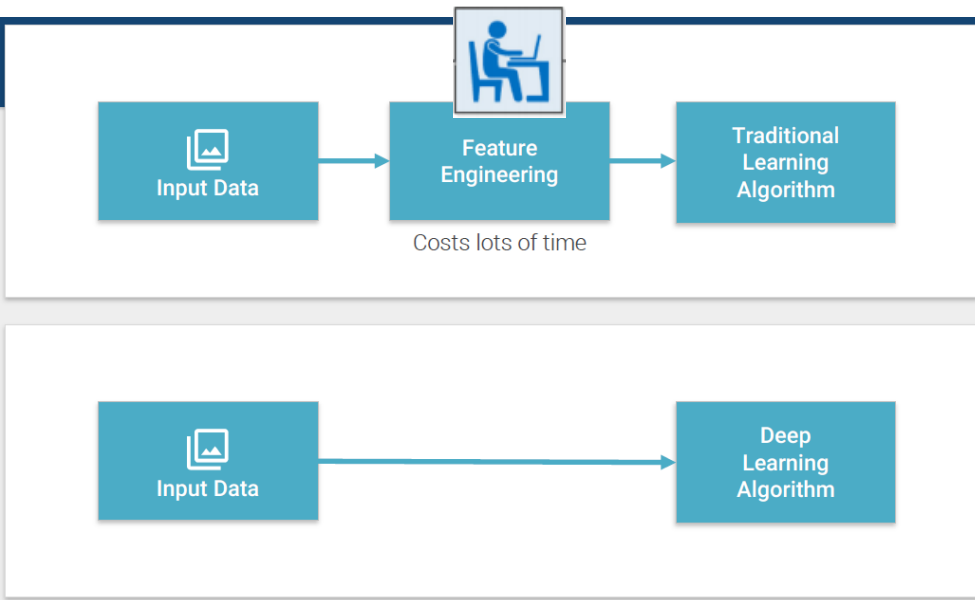
Deep learning trend and applications



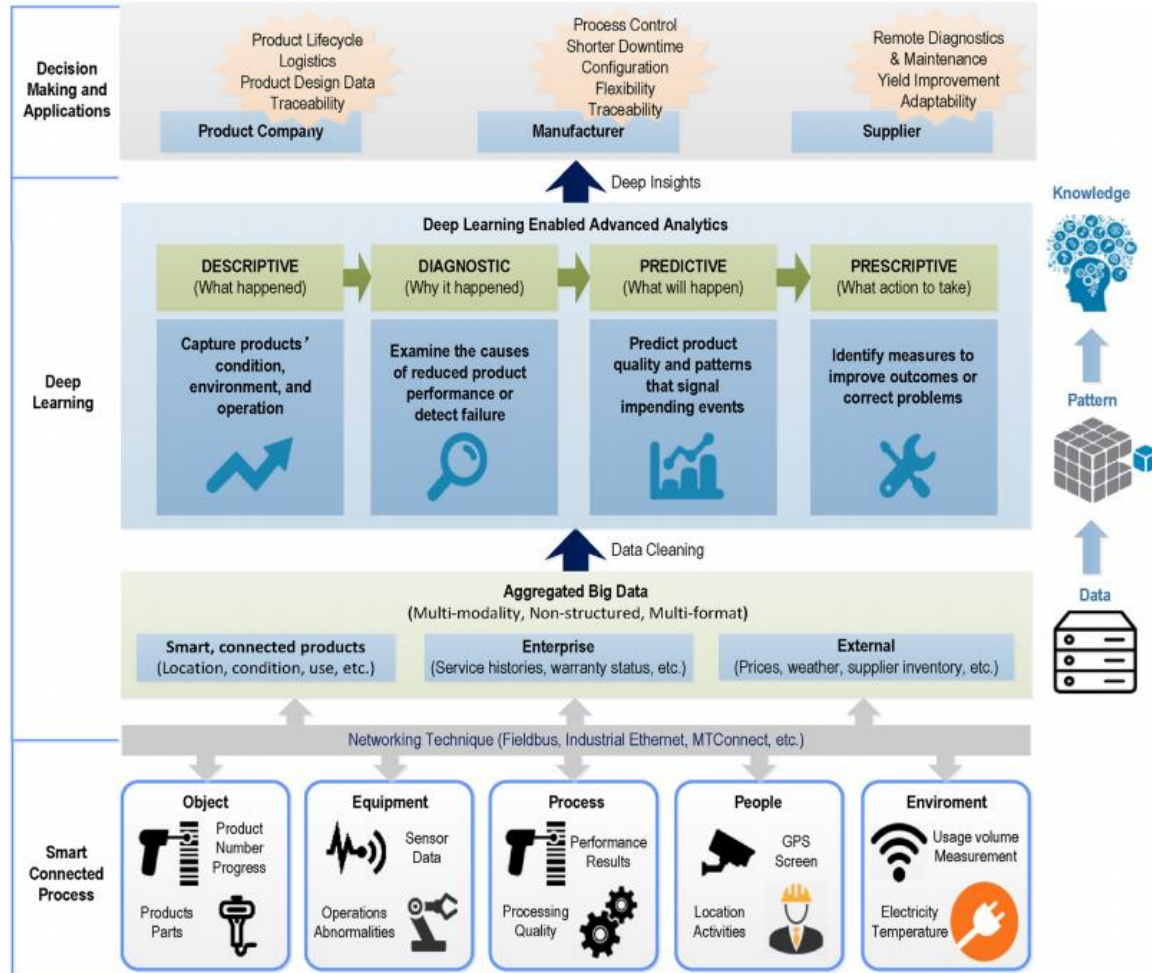
Deep learning tools



Machine learning vs deep learning



Deep learning enabled advanced analytics for smart manufacturing



Deep learning requirements



Large data set with good quality (*input-output mappings*)



Measurable and describable goals (*define the cost*)



Enough computing power (*AWS GPU Instance*)



Excels in tasks where the basic unit (*pixel, word*) has very little meaning in itself, but the **combination of such units has a useful meaning**

However, in manufacturing...

- ... data quality is usually poor.
- ... data quantity is usually low.
- ... computational infrastructures are usually weak.



Important to have high quality dataset

- *Valid, consistent, easily accessible, described and documented*



Important to have high quantity data

- *Stable process has less negative event to be correlated*
- *Stable process has less statistical deviation*



Process Experts and Data Experts must work side by side

Outline

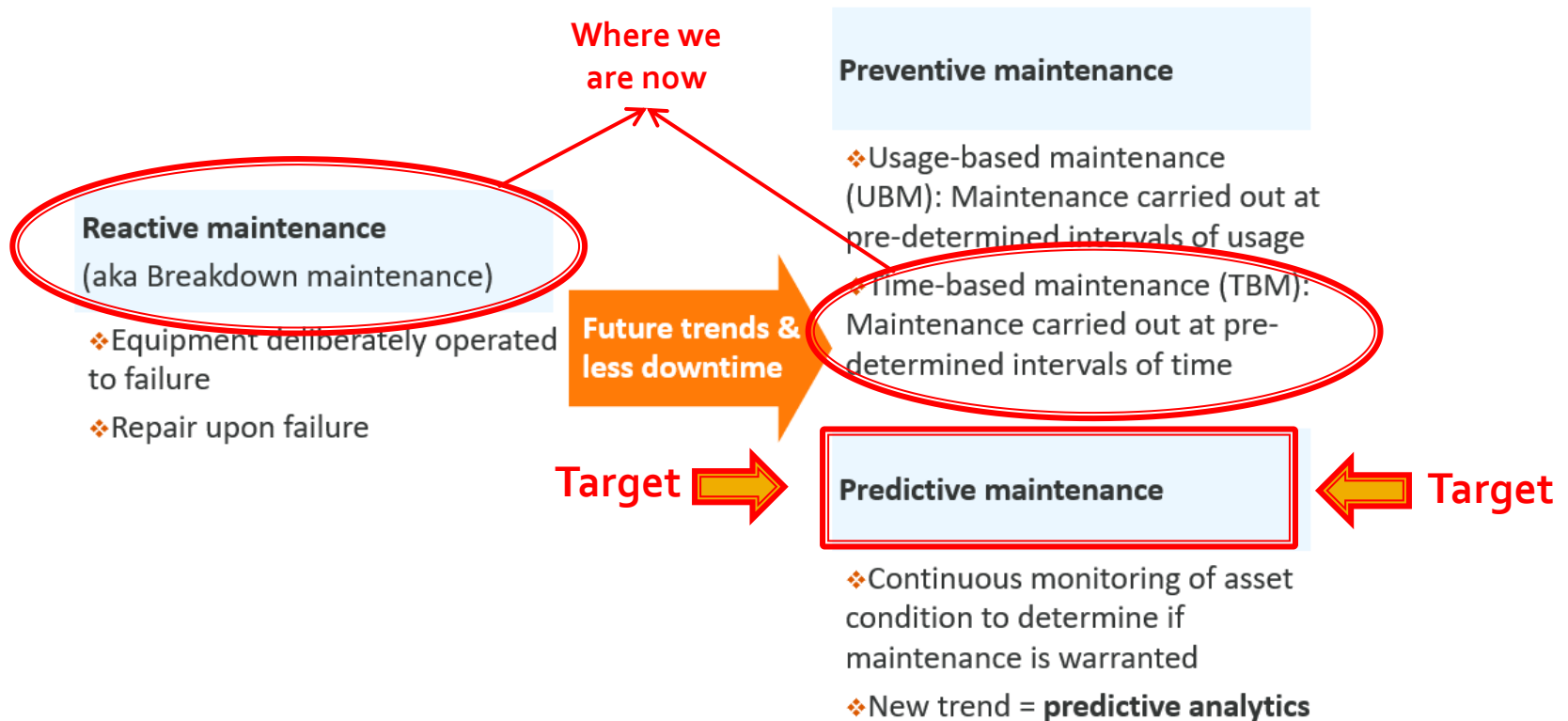
- Introduction
- Architectures
- Technologies
- Algorithms

Applications

Predictive maintenance

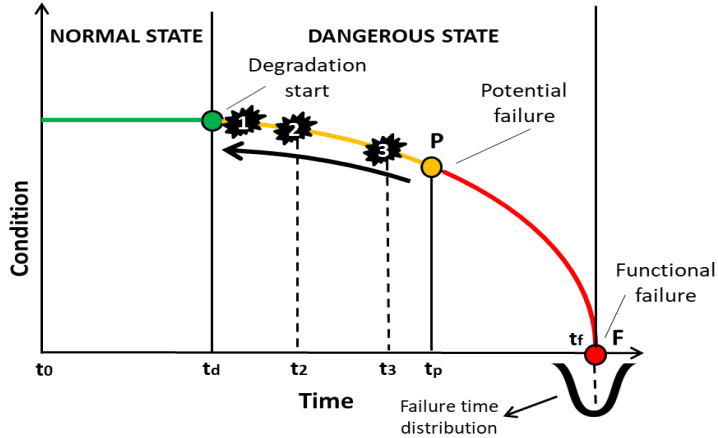
Maintenance strategies

The future trend is to move from reactive to preventive & predictive maintenance



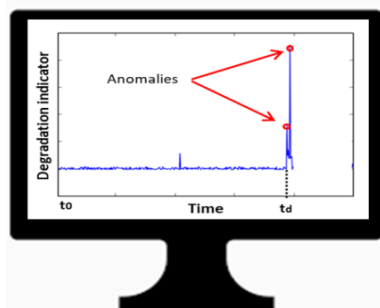
Predictive maintenance

- Predictive maintenance** uses condition monitoring equipment (e.g. sensors) in order to track the performance of equipment, to detect abnormal behaviour, to predict future failures and to support decision making about proactive actions.

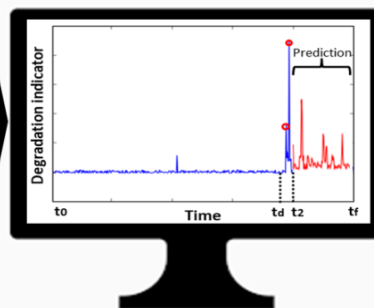


- The **P-F Curve** is a common way to represent the **behaviour of an asset** before actual functional failure has occurred.
- There is a sense in which **failure is a process**, not an event.
- The P-F interval can be seen as an **opportunity window** during which decision making ahead of time can eliminate the anticipated functional failure or mitigate its effect.

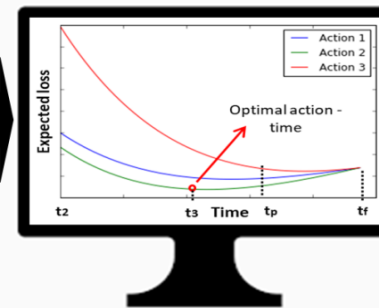
1 Anomaly detection



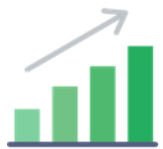
2 Failure prediction



3 Maintenance action



Impact of predictive maintenance



- OEE average improvement



- Reduction of MTTR

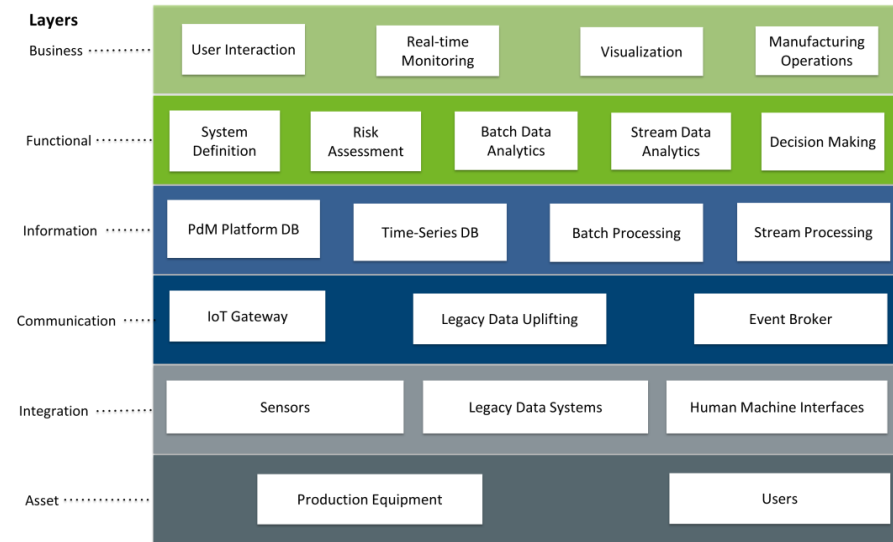


- Reduction of MTBF

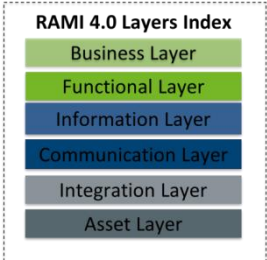
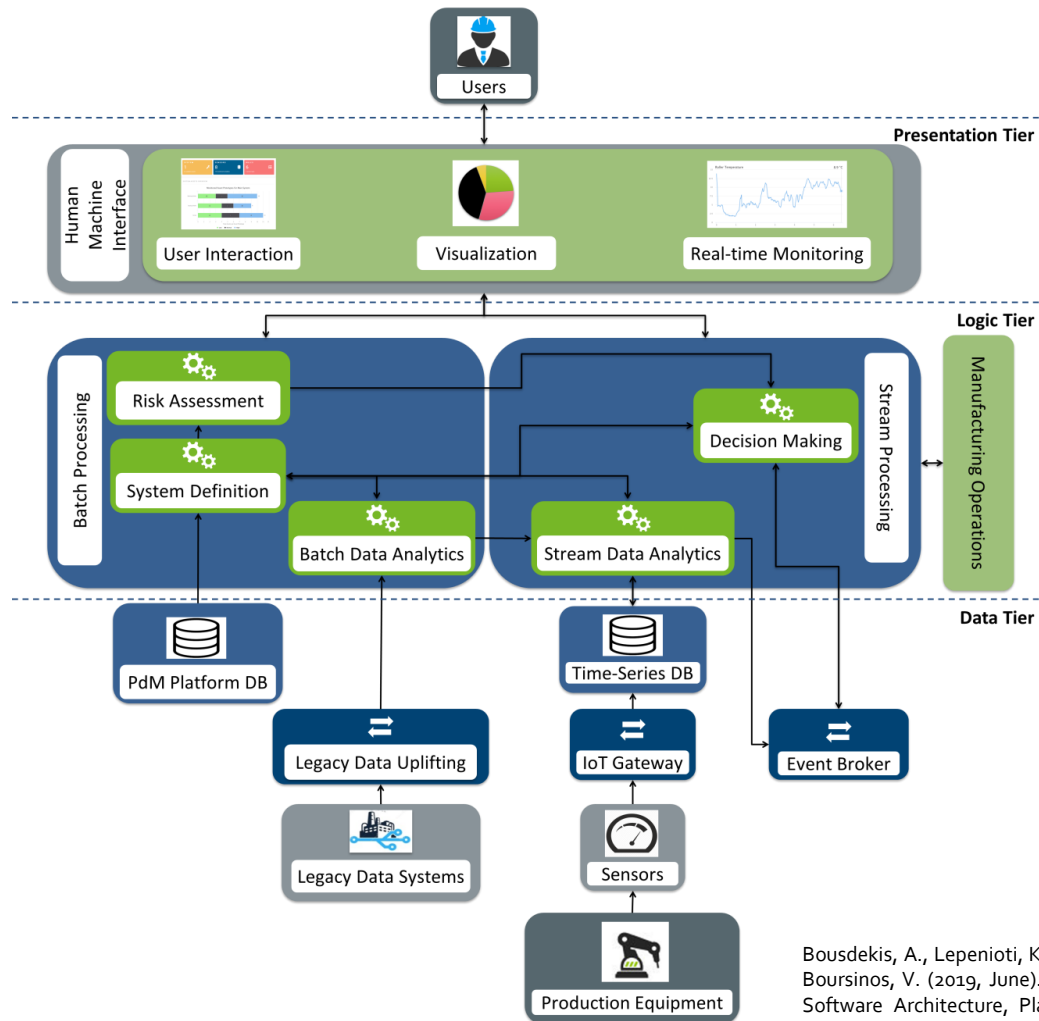


- Reduction of Total Cost of Maintenance

RAMI4.0-compliant predictive maintenance (1/2)



RAMI4.0-compliant predictive maintenance (2/2)

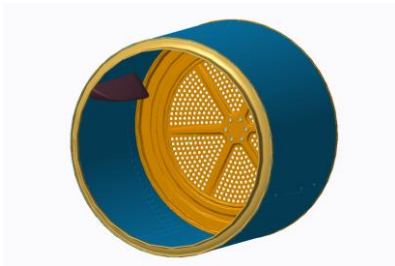


Bousdekis, A., Lepenioti, K., Ntalaperas, D., Vergeti, D., Apostolou, D., & Boursinos, V. (2019, June). A RAMI 4.0 View of Predictive Maintenance: Software Architecture, Platform and Case Study in Steel Industry. In International Conference on Advanced Information Systems Engineering (pp. 95-106). Springer, Cham.

Home appliances industry

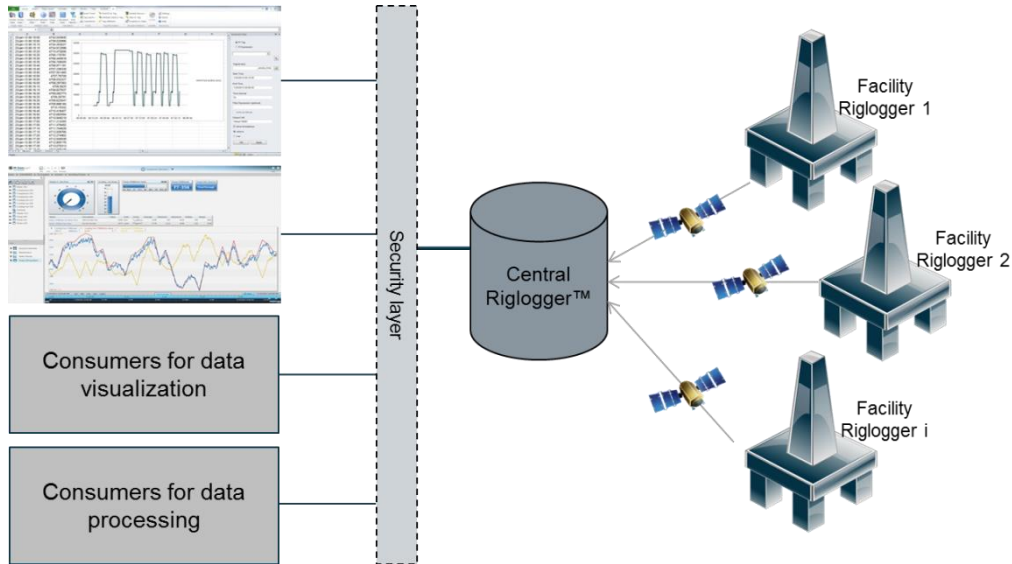
The Whirlpool case

Whirlpool
CORPORATION



Oil and gas industry

The MHWirth case



Steel industry

The M. J. Maillis case



Setup and process

M. J. MAILLIS



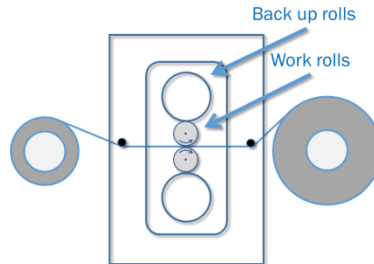
Raw material



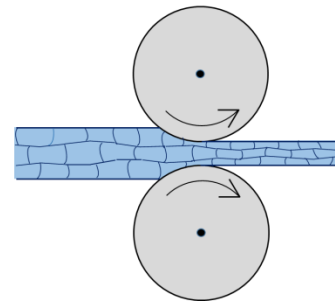
Cold rolling mill



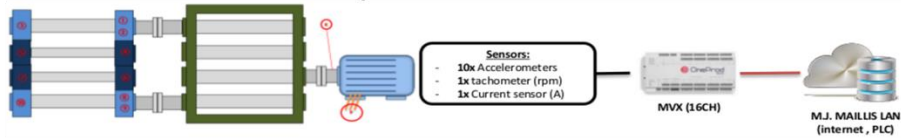
Roll Mill Stand



Deforming and Reducing the Grain Size



Infrastructure Setup for Sensor Data Collection



Front view of rollers



Rear view of rollers



Thank you