# **EXAMPLE RIALS**

MATERIALS WITH INTELLIGENCE

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Award Most Innovative Portuguese SME 2015 NOS Award Award Investment Success Case 2017 European Commission

#### J.C. Viana, CTO

Data Science in Structural Health Management of critical systems

2019 March



#### AGENDA

#### 1. SHM – STRUCTURAL HEALTH MONITORING

#### 2. APPLICATION OF DATA SCIENCE

3. SOME CHALLENGES

#### **HAZARDS AND DISASTERS** MATERIALS

bird strikes!

ice formation!



hail!



lightning strike!



blade failure



tower collapse





tool drop







in-flight structural failure of a F-15C



## **CRITICAL** ASSETS/EQUIPMENT



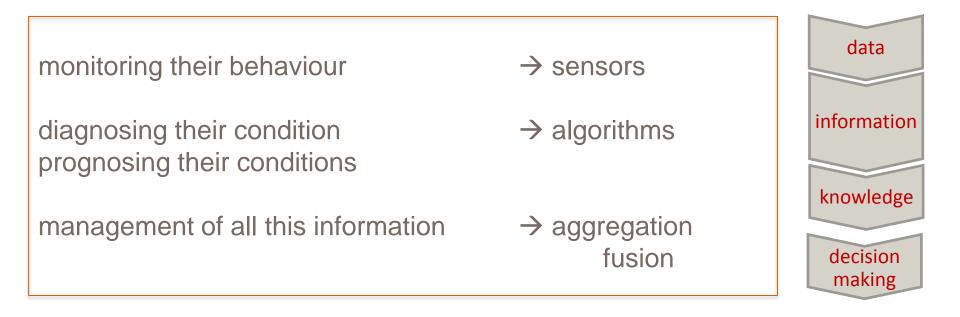
- 15% to 20% of equipment failures are age related
- 80% to 85% failures are totally time-random events, not predictable, but their start can be detectable

## Failure is not an option



## **CRITICAL** SHM - STRUCTURAL HEALTH MONITORING

#### **KNOWING THE CONDITION OF CRITICAL SYSTEMS**





### **CRITICAL** SHM - STRUCTURAL HEALTH MONITORING





monitoring

	monitoring	Embedded sensing system		
	Symptoms: temperature hacke tiredess	diagpagaia	prognosis	prescription
		diagnosis		
	Analysis:	Guess:	time to get good	exercise
C) Decers	Analysis: blood pressure	Guess: system		
		Guess: system disease	time to get good	exercise
	blood pressure blood analysis	Guess: system disease organ	time to get good next query	exercise medicine
	blood pressure	Guess: system disease	time to get good next query	exercise medicine
	blood pressure blood analysis	Guess: system disease organ	time to get good next query	exercise medicine

prognosis

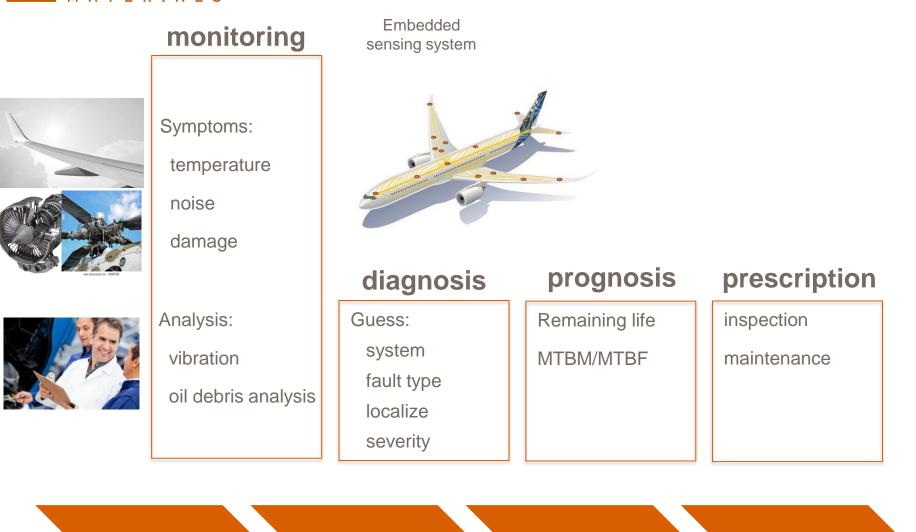
diagnosis

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prescription

### **CRITICAL** SHM - STRUCTURAL HEALTH MONITORING



prognosis

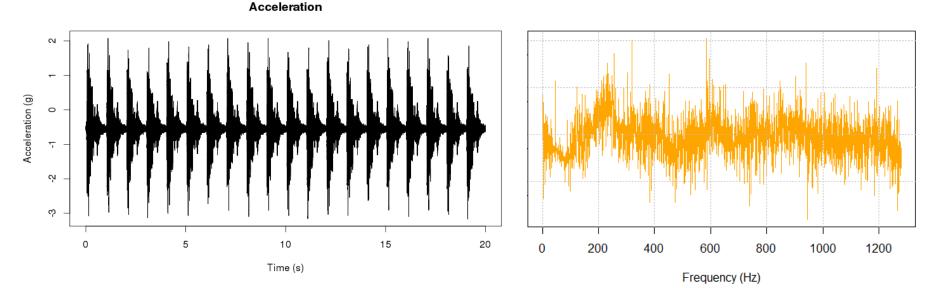
diagnosis

monitoring

prescription

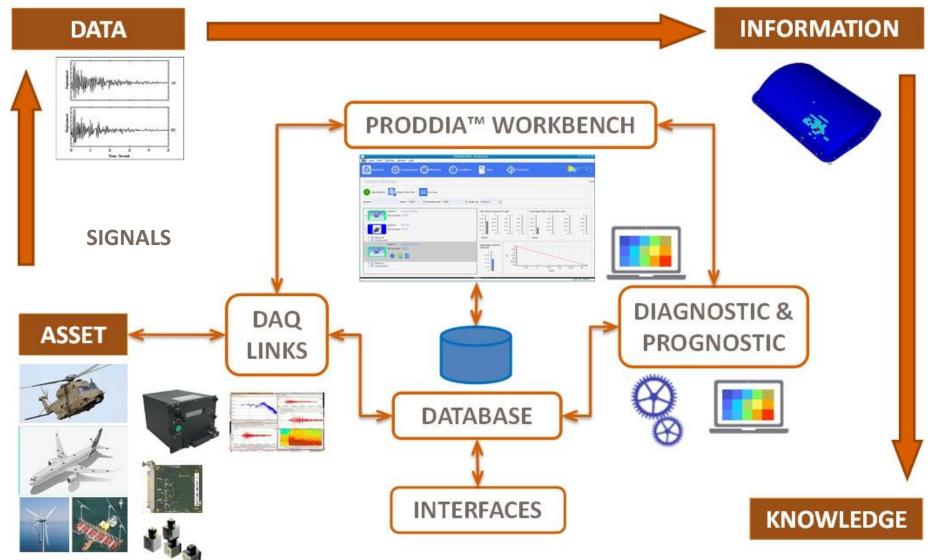


## CRITICAL CHALLENGES I

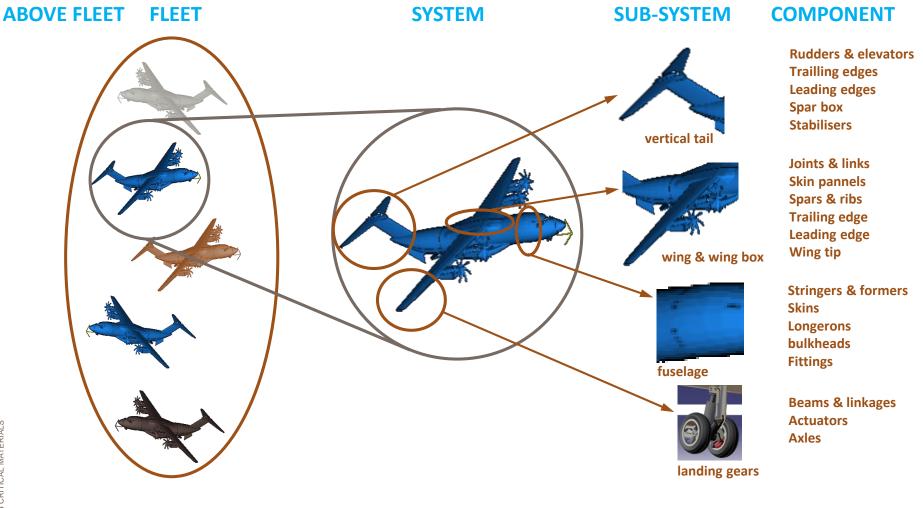


- How to transform sensor signals into valuable data?
- How to extract useful information from sensor data?
- How to detect failure / evaluate damage from sensor data?
- How to fuse data from various sensors types?

## **CRITICAL** SHM - STRUCTURAL HEALTH MANAGEMENT



## **CRITICAL** SHM - STRUCTURAL HEALTH MANAGEMENT



#### to IVHM – INTEGRATED VEHICLE HEALTH MANAGEMENT

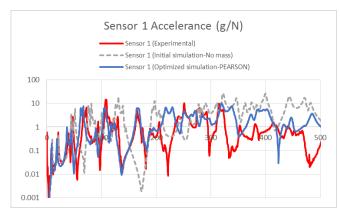
## CRITICAL CHALLENGES II

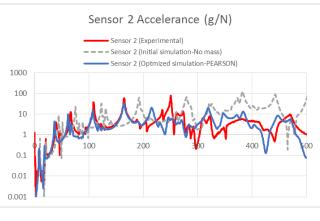


- How to integrate physical and digital data?
- How to deal with huge amounts of data (quality, latency)?
- How to fuse data from different assets?

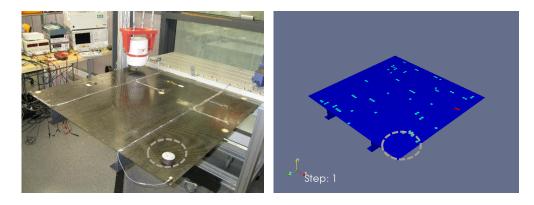
## CRITICAL SHM AND DIGITAL TWINS

#### Model adjustment - Data matching technology





## Damage Cases – Detected by digital twin operation



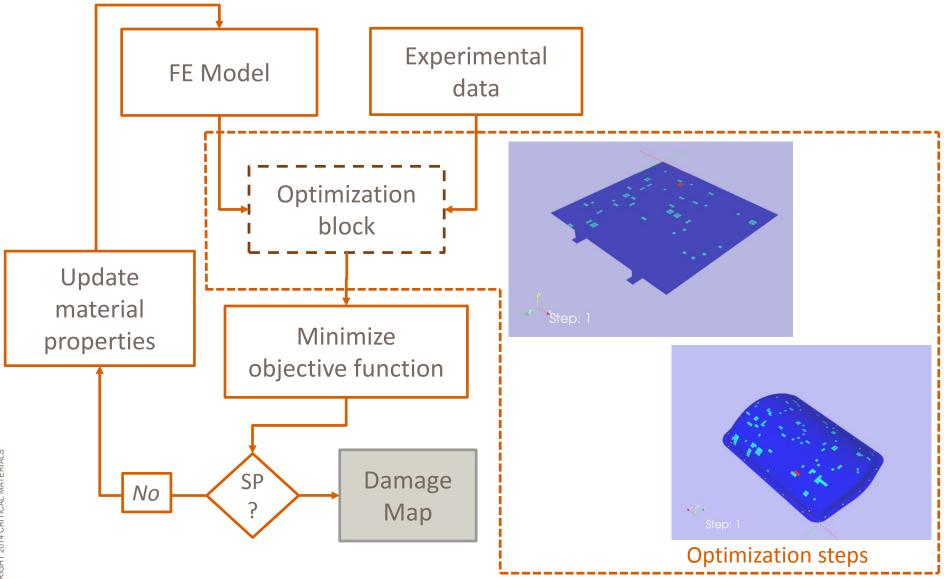
- The Digital Twin enables the localization and estimates the damage severity and size.
- Combines data from sensors (Acc) with a FEM model and a Global nonlinear Optimization method (*Particle Swarm Class*)

55 SIMULIA



PostgreSQL

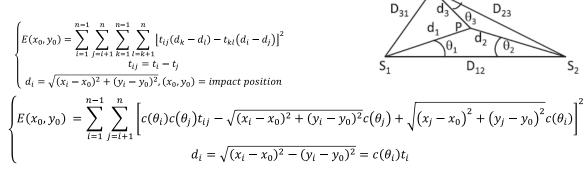
## CRITICAL SHM AND DIGITAL TWINS

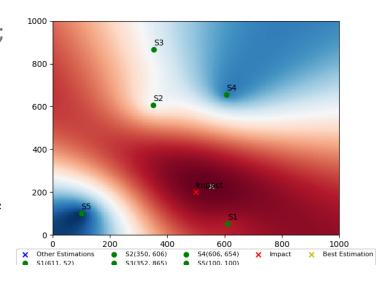


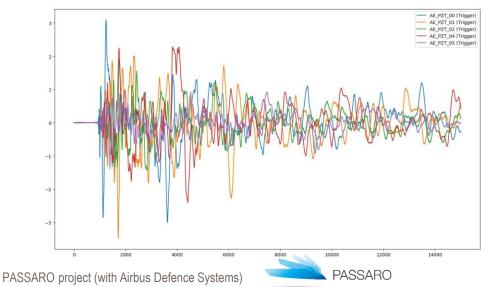
#### SHM AND OPTIMIZATION PASSARO TE

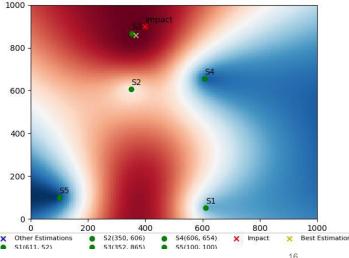
#### **Impact location based on Acoustic Emissions**

- PZT sensors
- Algorithm base on DToA (diference in time of arrival);
- Isotropic and anisotropic materials
- **Optimization procedure**





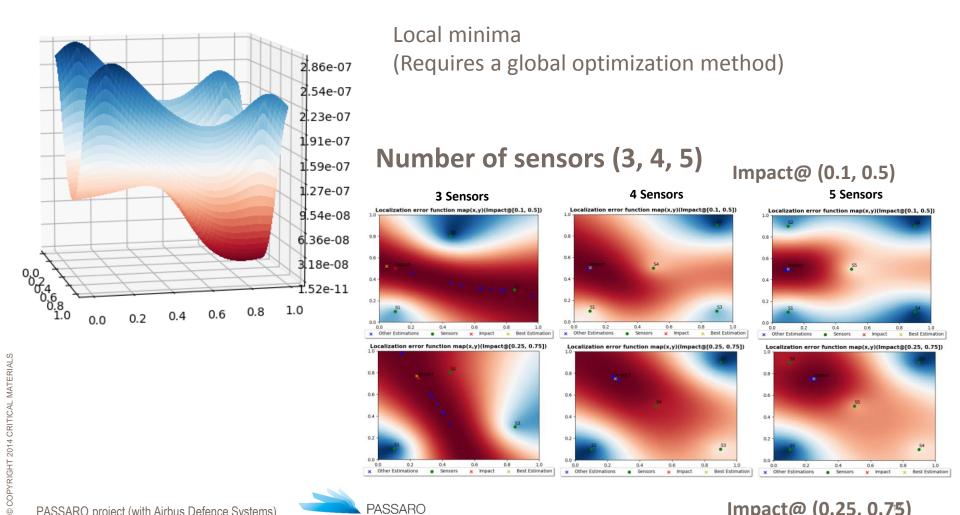




#### CRITICAL SHM AND OPTIMIZATION MATERIALS

#### **Optimization of sensor location**

Error maps for 5 sensors with impact @ (0.5, 0.9)



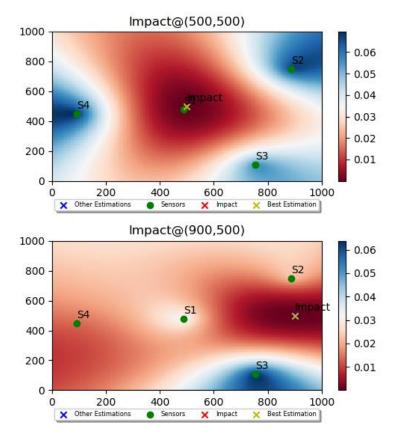
PASSARO project (with Airbus Defence Systems)

PASSARO

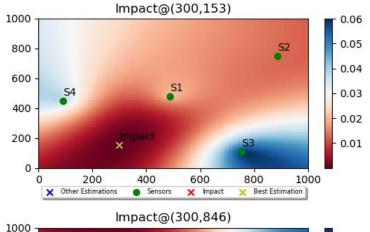


#### **Optimization of sensor location**

#### **Position of sensors**



#### Best sensors position | Mean Error = 0.83



0.06 × 800 0.05 0.04 600 51 **S**4 0.03 400 0.02 200 53 0.01 0 -0 200 400 600 800 1000 X Other Estimations Sensors × Impact X Best Estimation

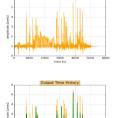


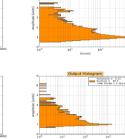
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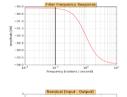
#### **SHM AND MACHINE LEARNING** ATERIALS

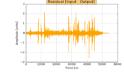
#### **Flight profiles**





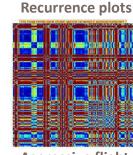




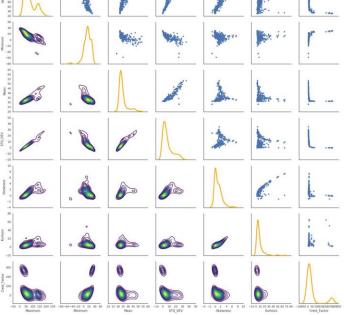


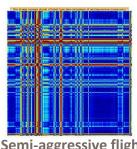
- **1. Regression with Deep Neural Networks**
- 2. Classification with Deep Neural Networks **2D Texture Image Classification with CNN** 
  - **Time Series Classification with RNN**
- 3. Natural language descriptions with Deep Neural Networks 2D Texture Image caption description with LSTM, CNN











**Aggressive flight** 

Semi-aggressive flight

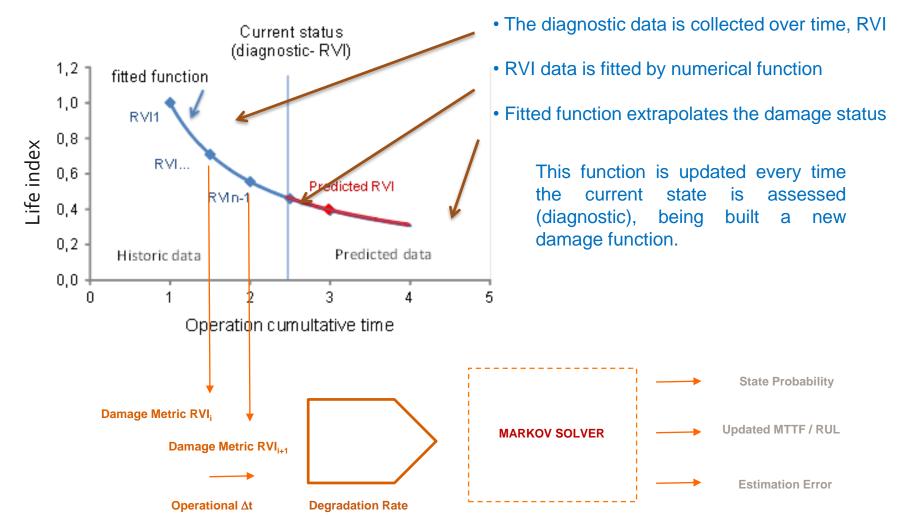
**3HT 2014 CRITICAL MATERIALS** 



A.M. Moura, P.J. Antunes, J.C. Viana, G.R. Dias, B. Serrano, D. Duarte, J. Cardoso, R. Basto, F. Machado, V. Infante, C. Mills, Application of Machine Learning Techniques for the Fatigue Life Estimation of Critical, Components, 9th European Workshop on Structural Health Monitoring, July 10-13, 2018, Manchester, UK

## CRITICAL SHM AND MACHINE LEARNING

#### **Health Prognosis**

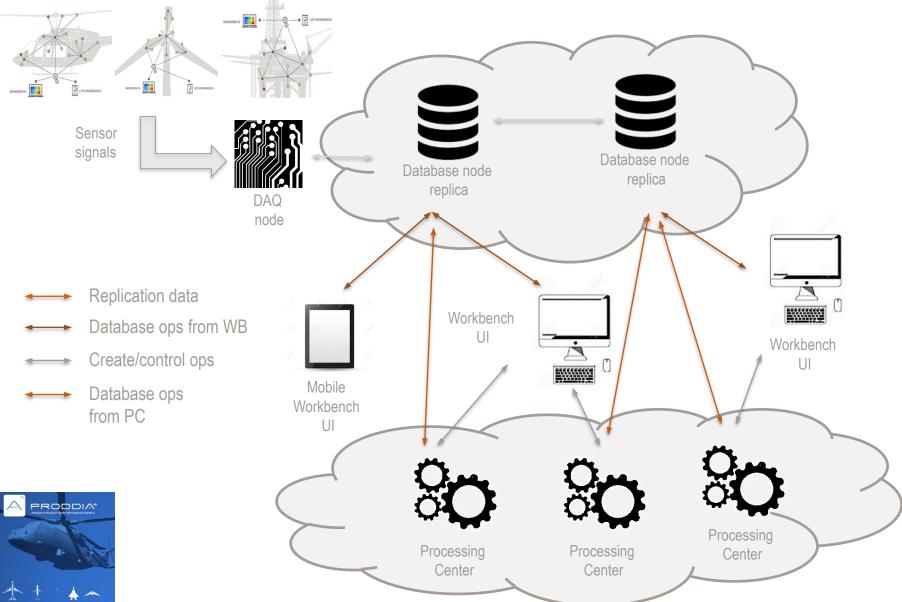


## SHM AS A CYBER-PHYSICAL SYSTEM

TTTT

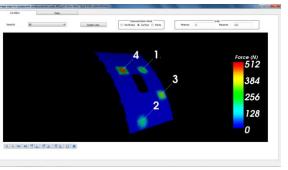


#### **CRITICAL** MATERIALS SHM - STRUCTURAL HEALTH MANAGEMENT



## CRITICAL SHM as a CPS

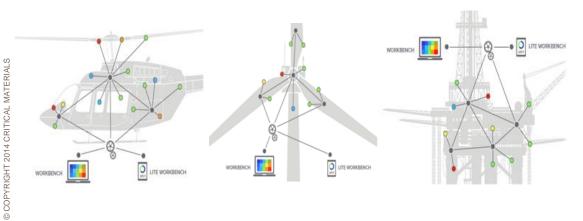




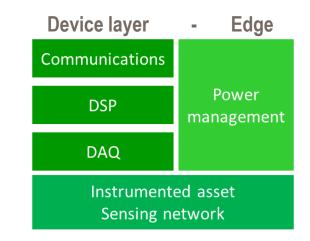






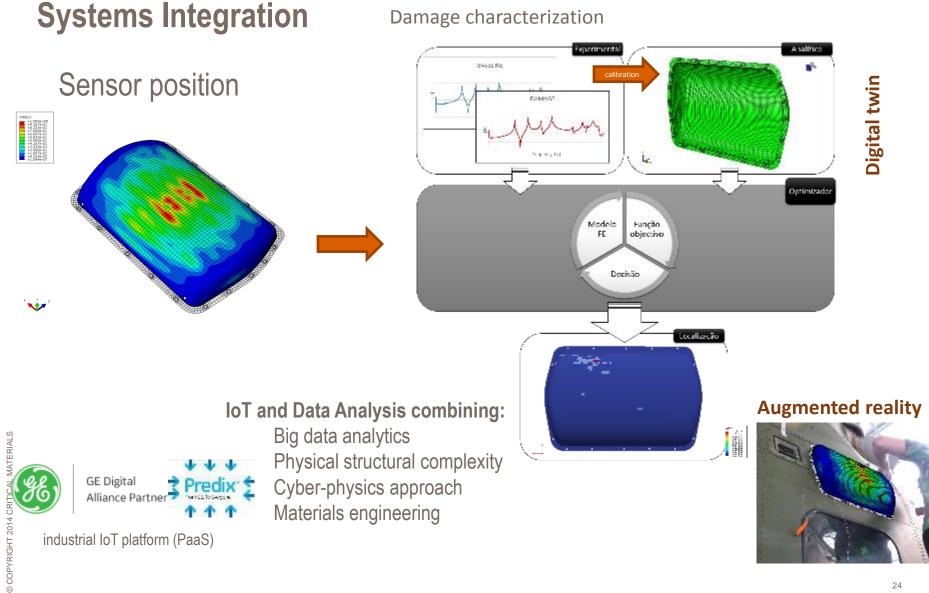


Processing layer - IT platform			
HMI Presentation Layer			
Reasoner: Condition monitoring, Health assessment, Decision support			
Data manipulation	Database		
Communication layer			





#### Damage characterization



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## CRITICAL CHALLENGES III



- Interoperability
- Big data
- IoT

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- Reasoning tools
- Operational intelligence
  - Business model

## CRITICAL UBIQUITOUS SHM SYSTEMS

DATA

components sub-system vehicle fleet company country everywhere

real/experimental data virtual/simulated data

time data

DATA ANALYTICS DATA CENTERS DATA COMPUTATION SENSING

CONDITION MONITORING

**DIAGNOSTIC & PROGNOSTIC** 

STRUCTURAL HEALTH MANAGEMENT

DATA ANALYTICS

#### PHYSICAL MODELS

**DIGITAL TWINS** 

**DECISON-MAKING** 

CYBER-PHYSICAL SYSTEMS

AUGMENTED REALITY

# **EXAMPLE RIALS**

MATERIALS WITH INTELLIGENCE

100



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