

# UIC Railway Asset Management Global Conference 2017

4 - 6 OCTOBER 2017  
UIC HEADQUARTERS, PARIS

## Lessons learned in Heavy Assets Predictive Maintenance



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5 Octobre 2017

# Challenges

- Demand is increasing **while budgets are shrinking**
- Infrastructure and Rolling Stock are **under stress**
- Regulatory bodies are more demanding on **Quality of Service**
- Trains have to be on-time and mission ready
- MRO costs need to be drastically reduced
- Sacrificing Customer Service for Cost reduction is not an option

## Doing more with Less



# Maturity driven by safety, availability and TCO



	Safety	Availability	Costs	Maturity
Air Forces	●	●	●	●
Airlines	●	●	●	●
Aircraft Manufacturer	●	●	●	●
Aircraft Engines Manufacturer	●	●	●	●
<b>Railway Company</b>	●	●	●	●
Rolling Stock Manufacturer	●	●	●	●
Turbine Manufacturer	●	●	●	●
Truck / Tractor manufacturer	●	●	●	●
Manufacturing Process	●	●	●	●
Electrical Grid	●	●	●	●
Telco	●	●	●	●
Mining Industry	●	●	●	●
Facility Management	●	●	●	●
Oil & Gas	●	●	●	●
Car manufacturer	●	●	●	●

Even though Railway pays lot of attention to availability and TCO, it's still lagging behind on predictive maintenance

- Hard constraint
- Average constraint
- Soft Constraint

# Use Cases



## US Army

- Predictive Maintenance program has produced for a fleet of 120 helicopters:
  - **11 % increase in fleet availability**
  - **41 000 Maintenance Man Hours saved in a year**



## Navistar

- Telematics data is enabling
  - Failures prediction
  - Fleet operator notification
  - Proactive routing of spare parts
- **Logistic response enhanced from 7 to 2 days**



## Caterpillar

- Monitoring and analysis of Truck engines sensor data has led to **100 M\$ savings in executing Performance Based Contracts**



## Union Pacific

- 3350 trains/day on a 32,000 miles network
- Infra red sensors installed every 20 miles
  - 20 millions readings / day
  - Monitoring of mechanical wear, overheating and derailments
  - 1500 problems qualified every day
- **Reduction of damages, delays and customers claims**

# Siemens Vision 2020



Teradata wins a Stevie® Award

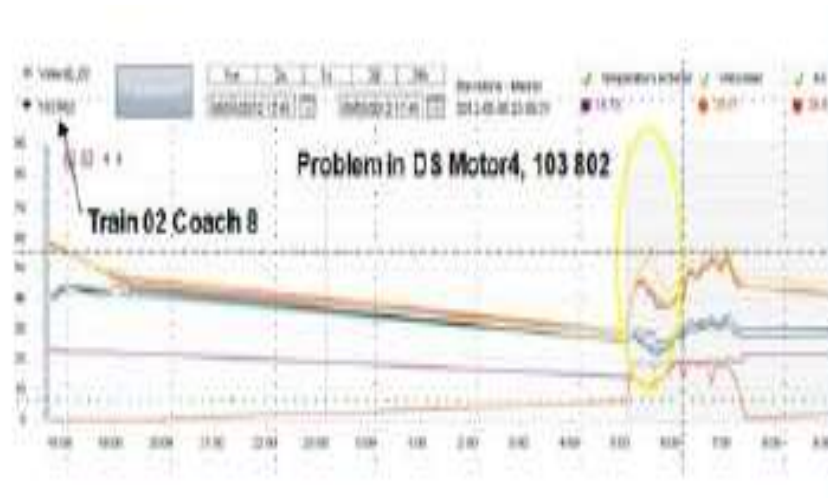
<http://www.teradata.com/Resources/Videos/Siemens-Capitalizing-on-Digitization,-the-In>

Teradata was named a Bronze Stevie Winner for the [video](#) "Siemens: Speeding Down the Path of a Successful Future with Big Data and Analytics" in the "sales & technology video" category.

# Identifying risks and anticipating breakdowns

## Predictive Maintenance

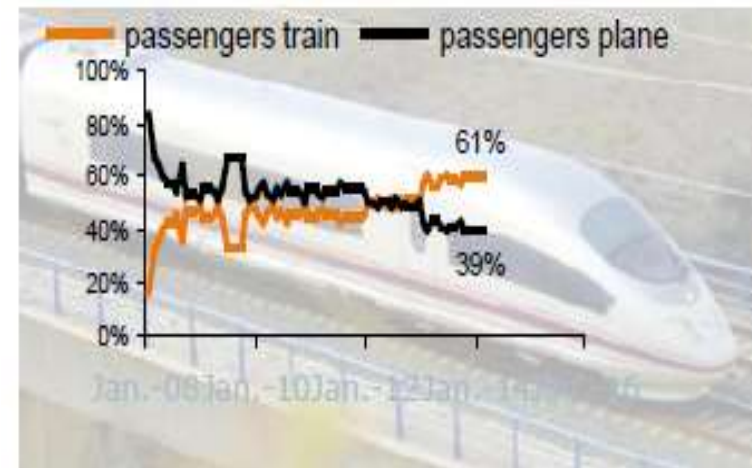
- Sensors continuously gather a breadth of data
- Most of the data is analysed on the fly
- Drafts are identified
- Alerts are triggered on abnormal situations



Identifying risky situations

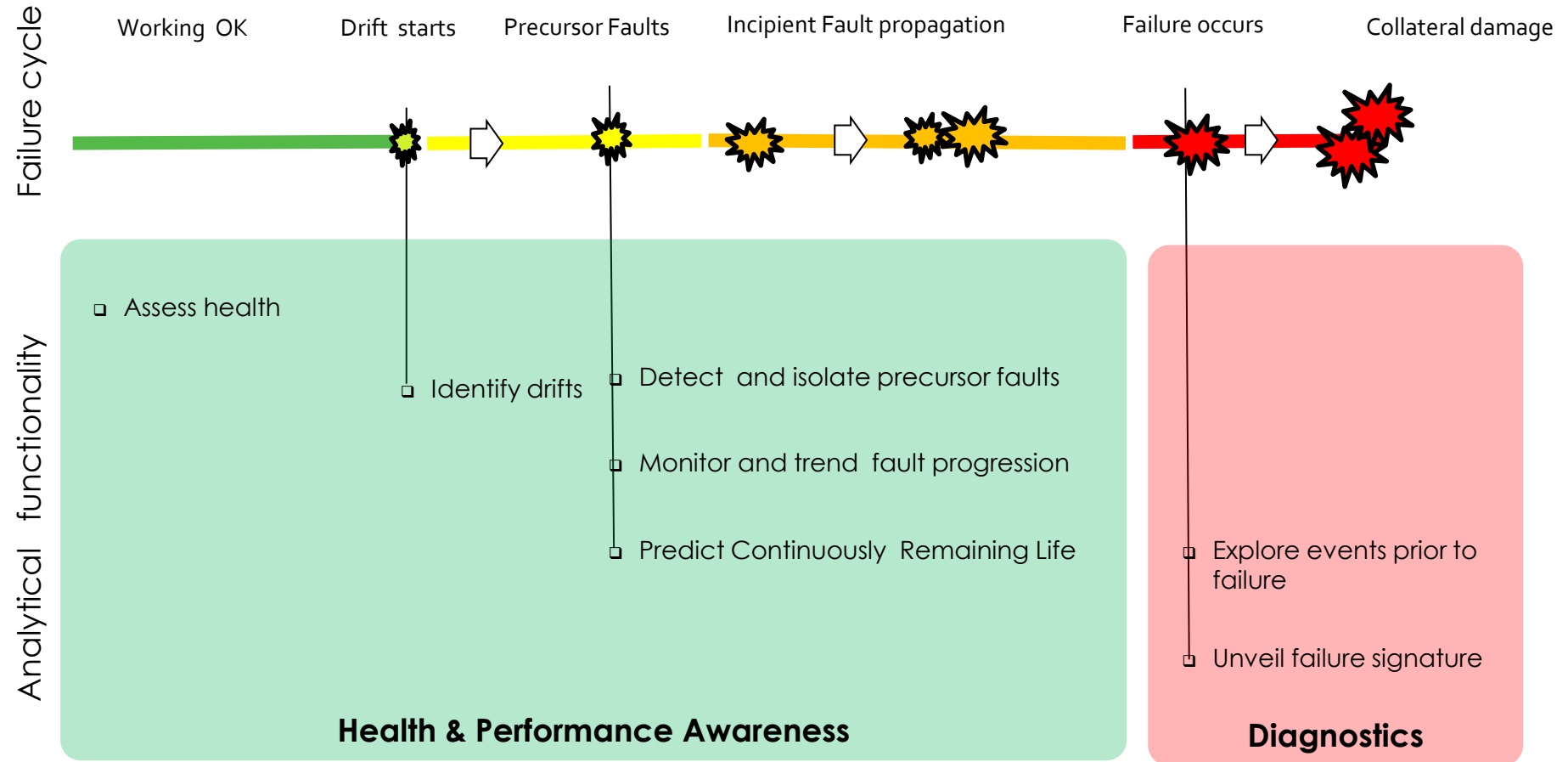
## Benefits

- «Performance-based-contracting
- One delay out of 2300 trips
- Valero E est the reliable RENFE High Speed Train
- Big market share on Madrid – Barcelone leg

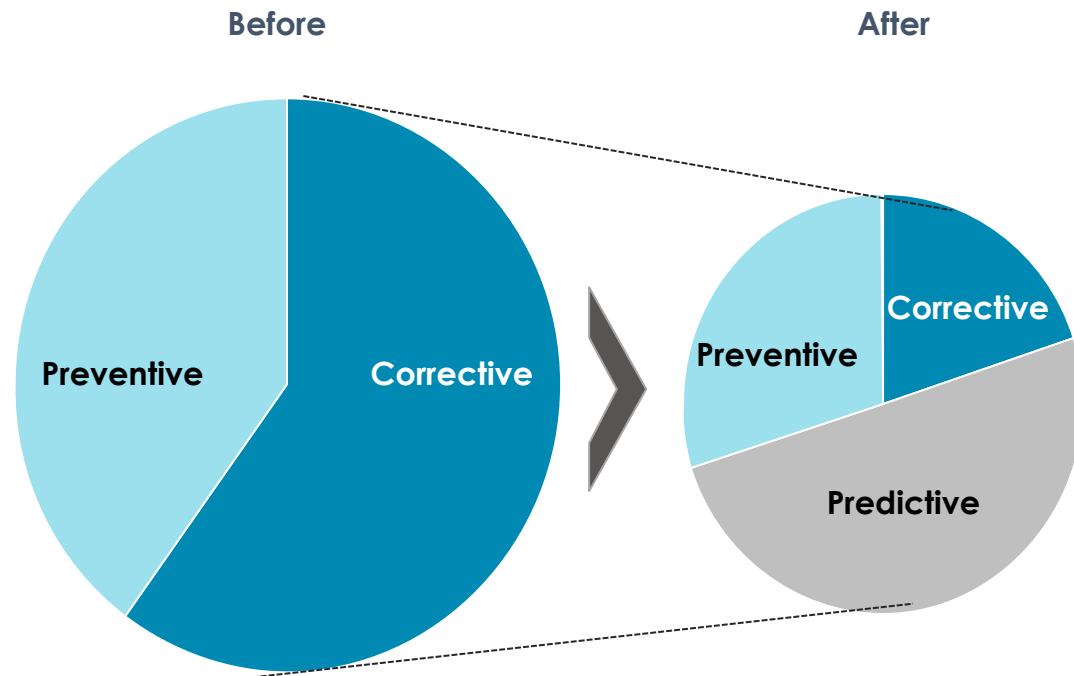


Market share Railway vs Airlines

# Prognostics & Health Management



# Impact of Predictive Maintenance



- Develop a deeper knowledge on equipment health and aging effects
- Improve Equipment reliability and maintainability
- Optimise Maintenance Planning and Execution
- Make better trade off between Repair vs. Leave in field with Partial Mission readiness
- Increase the sharing of equipment life cycle data
- Process Reengineering

**The amount and mix of Maintenance workload will change**



# Focus on Predictive maintenance

Track

Plan

Execute



## Intelligent Asset Tracking

*Analyse asset demography*



## Health and performance Surveillance

*Scrutinise asset operations & pathologies*



## Maintenance Planning

*Fine tune the maintenance ailor maintenance pro*



## Material Management

*Predict spare parts demand and align supply & inventories*



## Maintenance & Repair Efficiency

*Accelerate MRO tasks completion*



## LifeCycle Costing

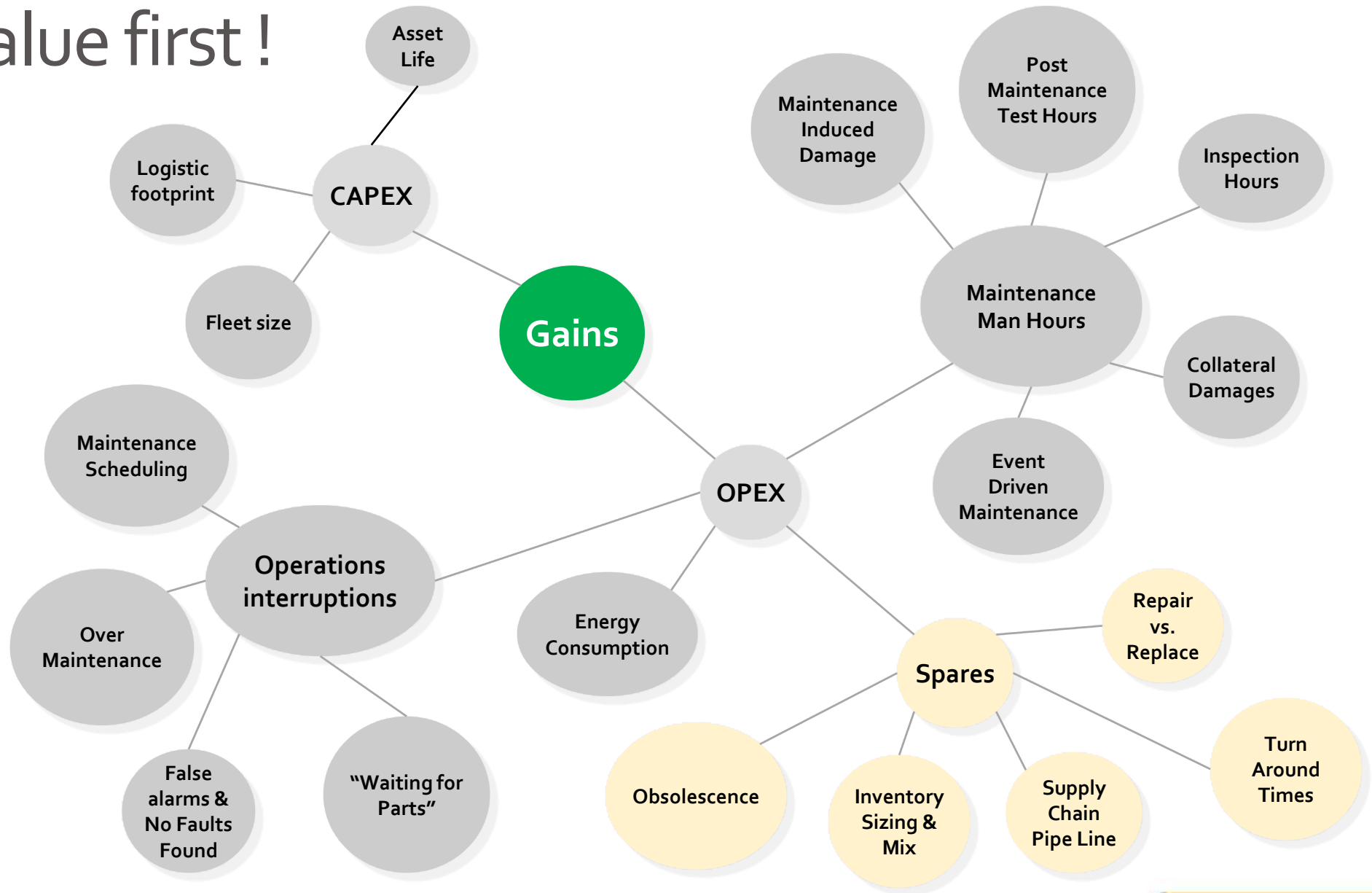
*Planning and controlling MRO costs over the asset life*



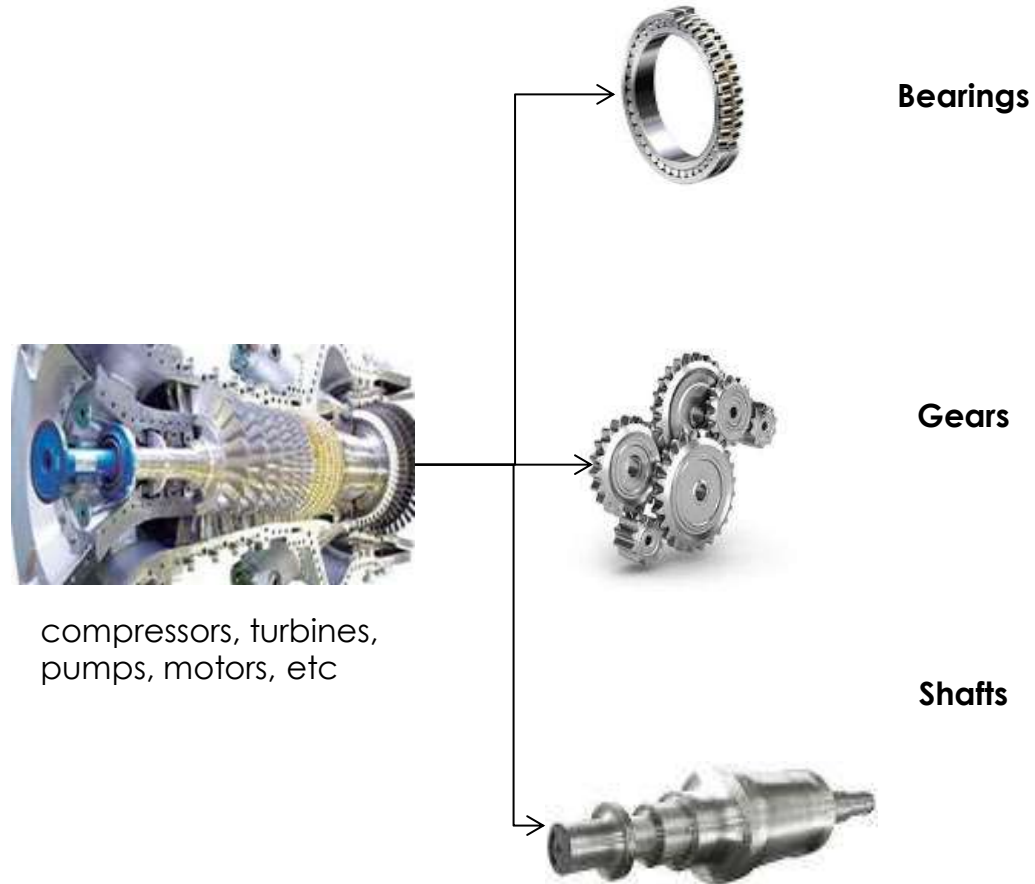
## Asset Improvement & retrofit

*Upgrading and extending asset capabilities*

# Value first!



# Focus first on organs that actually drive failures



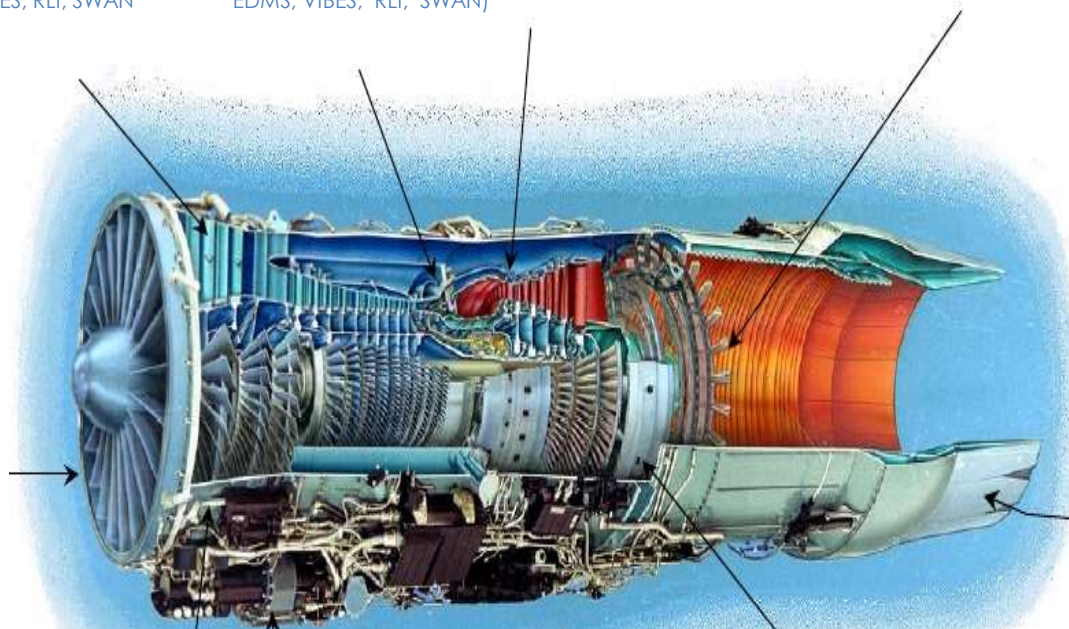
# Leverage any kind of data

**FAN 1 ROTOR TIP/ CRACKS**  
 (ECS, BVM, AFD, IDMS,  
 EDMS, VIBES, RLI, SWAN)

**OVERFLOWING / BLOCKED  
 FUEL NOZZLE**  
 EDMS, VIBES, RLI, SWAN)

**COMBUSTOR/HPT EROSION**  
 EDMS

**LPT BLADE RUB**  
 EDMS, VIBES, RLI, SWAN



**BEARING FAILURE**  
 ODM, EODM, EBM, VIBES, RLI,  
 SWAN

**BULK OIL  
 DEGRADATION**  
 OCM, ODM, EBM

**VAPOR PHASE COKING**  
 OCM, ODM, EODM, EBM




**GEARBOX FAILURE**  
 VIBES, RLI, SWAN

**BEARING FAILURE**  
 ODM, EBM, VIBES, RLI, SWAN

**BULK OIL CONTAMINATION**  
 OCM, EBM, ODM, EODM

AFD	Acoustic FOD Detector
BVMX	Blade Vibration Meter
EBM	Electrostatic Bearing Monitor
ECS	Eddy Current Blade Sensor
EODM	Electrostatic Oil Debris Monitor
EDMS	Engine Distress Monitoring
IDMS	Ingested Debris Monitoring
ODM	Oil Debris Monitor
PZT	Piezoceramic Patch Crack Detection
RLI	Robust LASER Interferometer
SWAN	Stress Wave Analysis
VIBES	Vibrations sensor

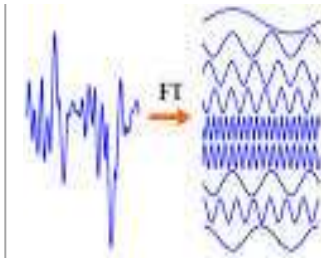
# Understand the dynamic of the machinery

	Failure modes	Sensor Data	Data characteristic
 <p>Bearings</p>	<ul style="list-style-type: none"> <li>• Failure of outer/inner races, rollers or cage</li> </ul>	<ul style="list-style-type: none"> <li>• Vibration</li> <li>• oil debris</li> <li>• acoustic</li> <li>• emission</li> </ul>	<ul style="list-style-type: none"> <li>• high noise in raw data</li> </ul>
 <p>Gears</p>	<ul style="list-style-type: none"> <li>• Tooth pitting</li> <li>• Gear crack</li> <li>• Wear</li> </ul>	<ul style="list-style-type: none"> <li>• Vibration</li> <li>• oil debris</li> <li>• Acoustic emission</li> </ul>	<ul style="list-style-type: none"> <li>• High noise</li> <li>• Signal modulation (coupling with bearing, shaft transmission path)</li> </ul>
 <p>Shafts</p>	<ul style="list-style-type: none"> <li>• Misalignment</li> <li>• Unbalance</li> <li>• Bend</li> <li>• Crack</li> <li>• Rubbing</li> </ul>	<ul style="list-style-type: none"> <li>• Vibrations</li> </ul>	<ul style="list-style-type: none"> <li>• Vibration signal is relatively clean and harmonic</li> <li>• frequency components of rotating speed can indicate the defects</li> </ul>

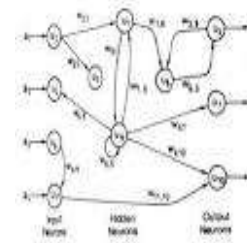
# Pick and train relevant analytical technique

## ▲ Feature extraction

- Fourier Transform Analysis
- STFT Analysis
- Wavelet Packet Decomposition
- Autoregressive Modeling



## ▲ Diagnosis



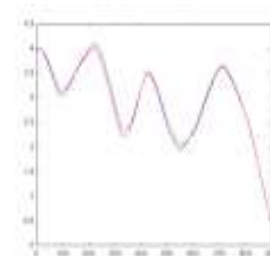
- ▲ Feature Map
- ▲ Hidden markov Model
- ▲ **Bayesian Network**
- ▲ Neural network

## ▲ Pattern Recognition

- ▲ Logistic Regression
- ▲ Feature Map
- ▲ Statistical Pattern Recognition
- ▲ Hidden Markov Model



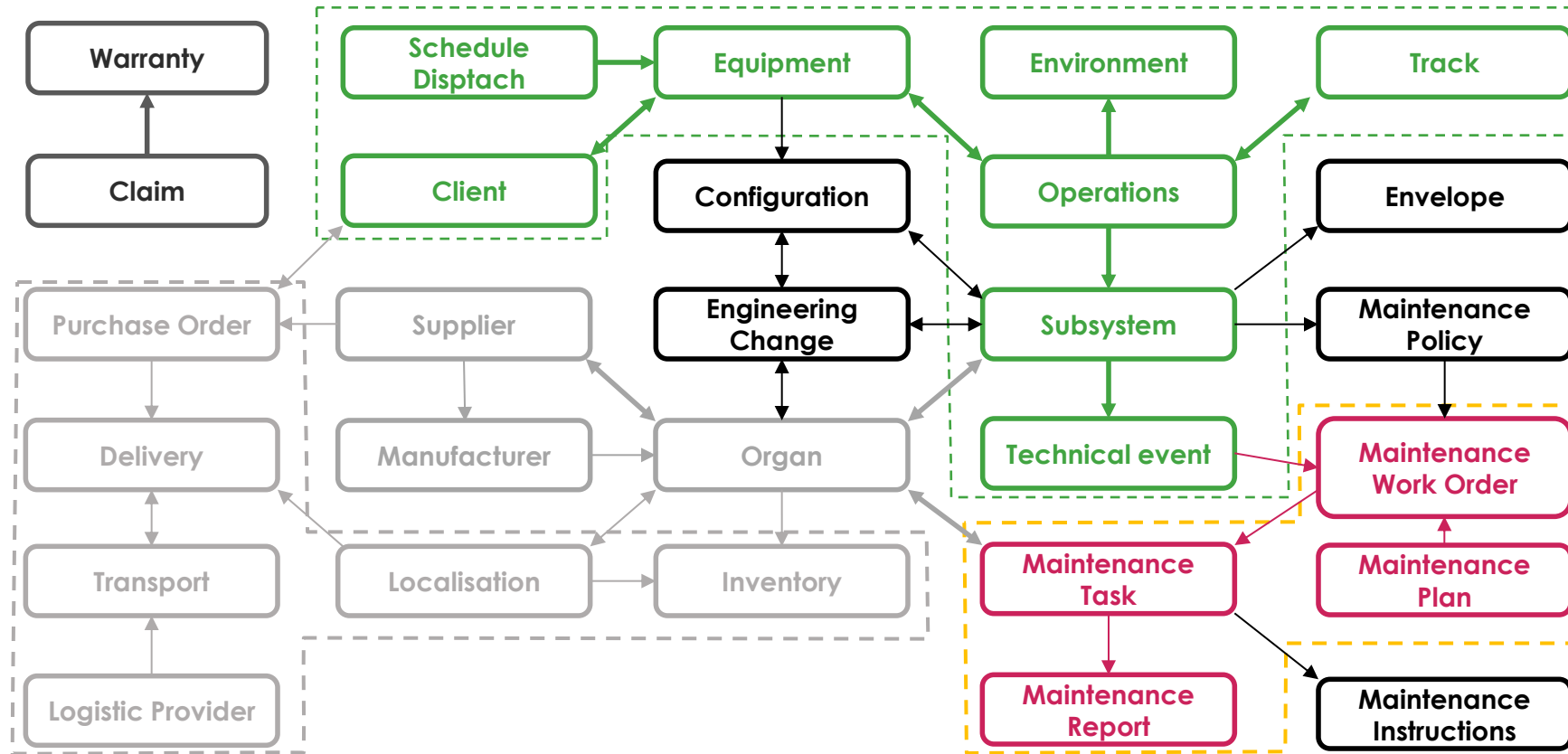
## ▲ Forecasting (*Remaining Life*)



- ▲ Autoregressive Moving Average
- ▲ Match Matrix Prediction
- ▲ Fuzzy Logic Prediction
- ▲ Neural Network Prediction



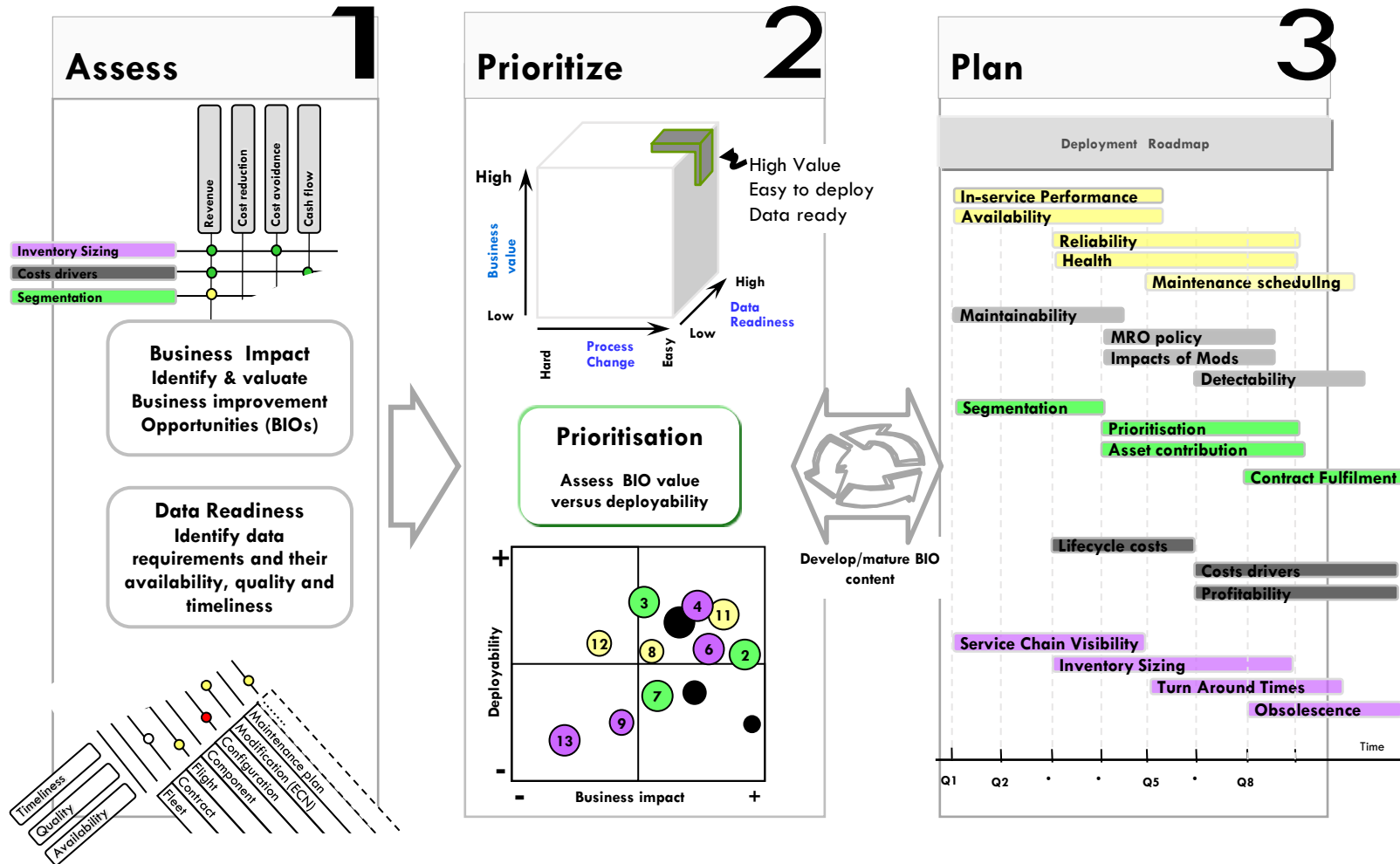
# Enlarge progressively your data scope



<span style="border: 1px solid green; display: inline-block; width: 15px; height: 10px;"></span> Operational data	<span style="border: 1px solid grey; display: inline-block; width: 15px; height: 10px;"></span> Engineering Data	<span style="border: 1px solid pink; display: inline-block; width: 15px; height: 10px;"></span> Repair Shop Data	<span style="border: 1px solid lightgrey; display: inline-block; width: 15px; height: 10px;"></span> Supply Chain Data
<span style="border: 1px solid lightgrey; display: inline-block; width: 15px; height: 10px;"></span> PDM data	<span style="border: 1px solid grey; display: inline-block; width: 15px; height: 10px;"></span> CRM Data	<span style="border: 1px dashed black; display: inline-block; width: 15px; height: 10px;"></span> Key Functional Scope	



# Adopt a step by step approach



Thank you!

Q & A

