

eQ Technologic Inc,

# Navy Demonstration DaaS (Data as a Service) / IDE

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GLOBAL PRODUCT DATA  
INTEROPERABILITY  
**S U M M I T**  
**2019**



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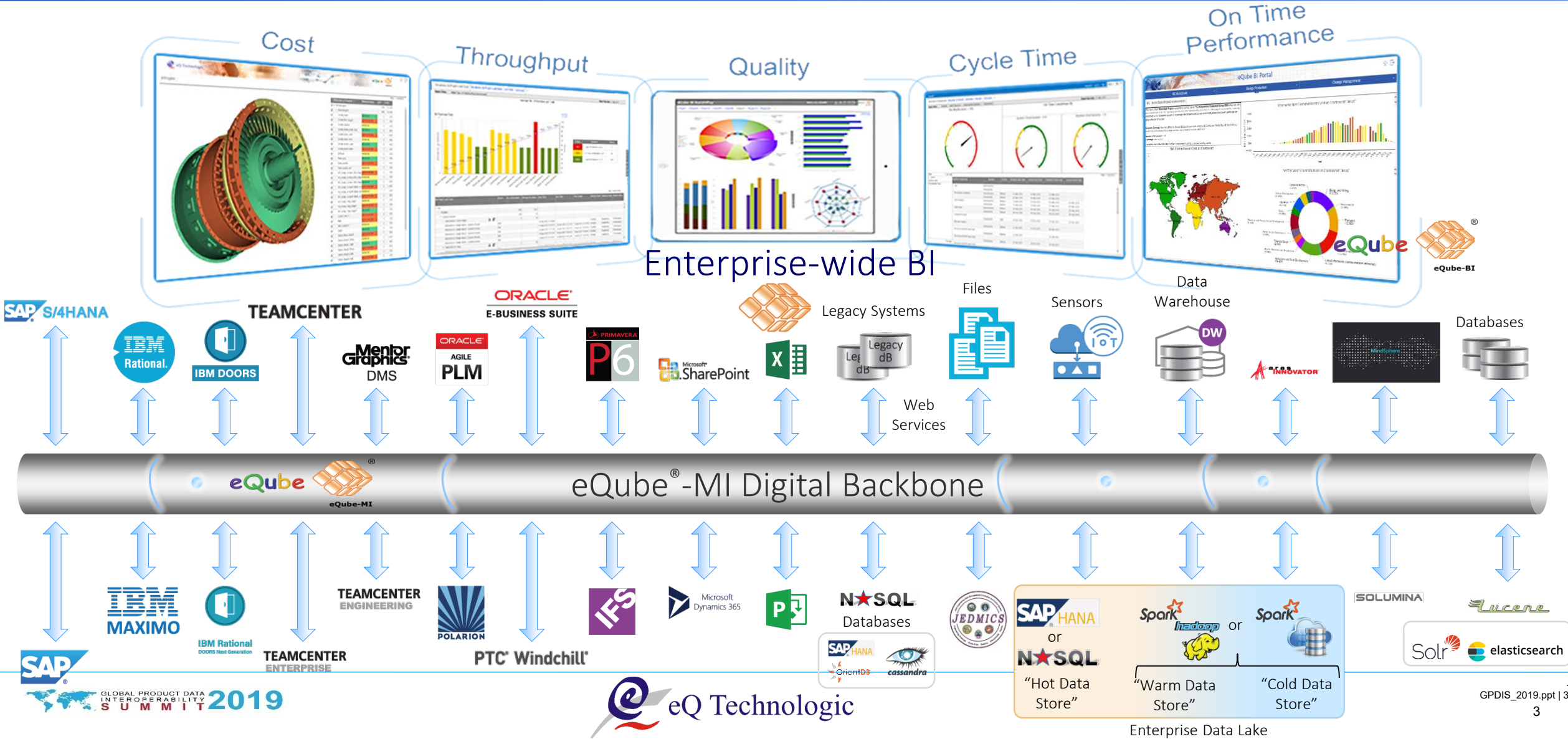


a modern platform for Data as a Service (DaaS) / Integrated  
Data Environment (IDE)

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eQube® platform forms a Connected Network of integrated data,  
applications, and devices that puts the power of analytics in the hands of  
end-users.

any data – at any speed, any application, any format, any device ....



# Navy Demonstrator For Purpose App with Predictive Maintenance

An intuitive, easy to use web application built using eQube-MI API Gateway

Seamless user experience across several underlying applications

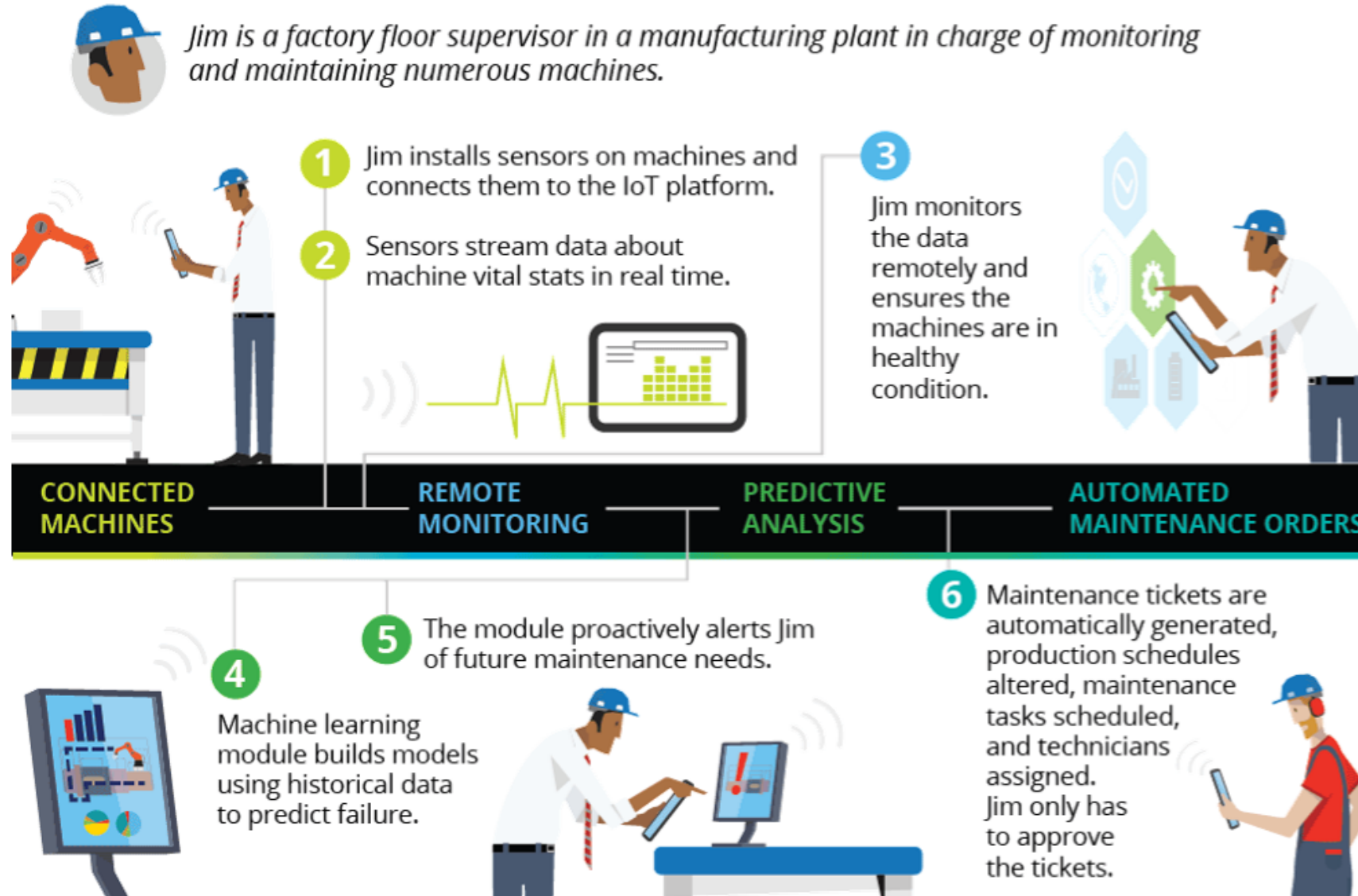
This 'For Purpose' application:

- Interacts & updates multiple applications behind the scene
- Has built-in business intelligence using eQube-BI
- Uses eQube-ADA that harnesses Machine Learning and Augmented Analytics to
  - Proactively analyze sensor data and detect unseen problems
  - Create a PR automatically and initiate a workflow



# Predictive Maintenance in a nutshell

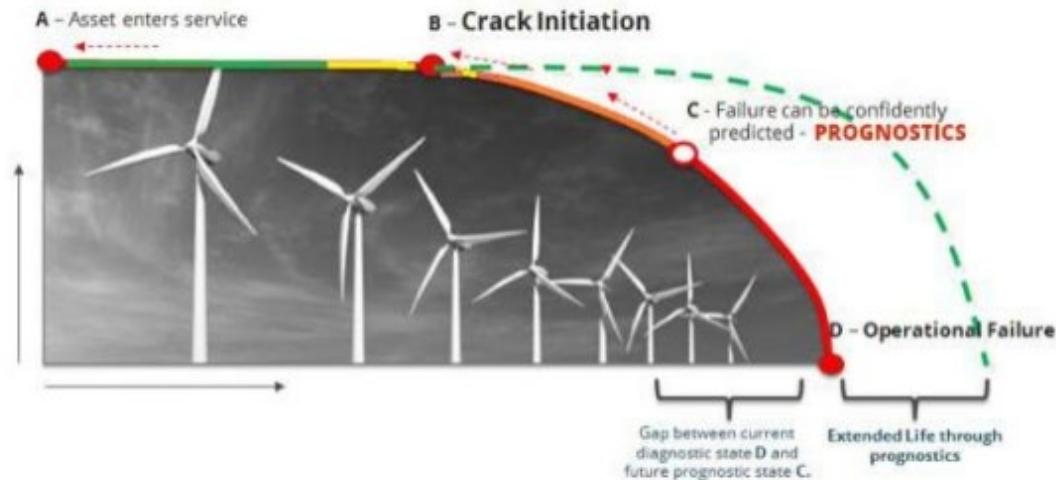
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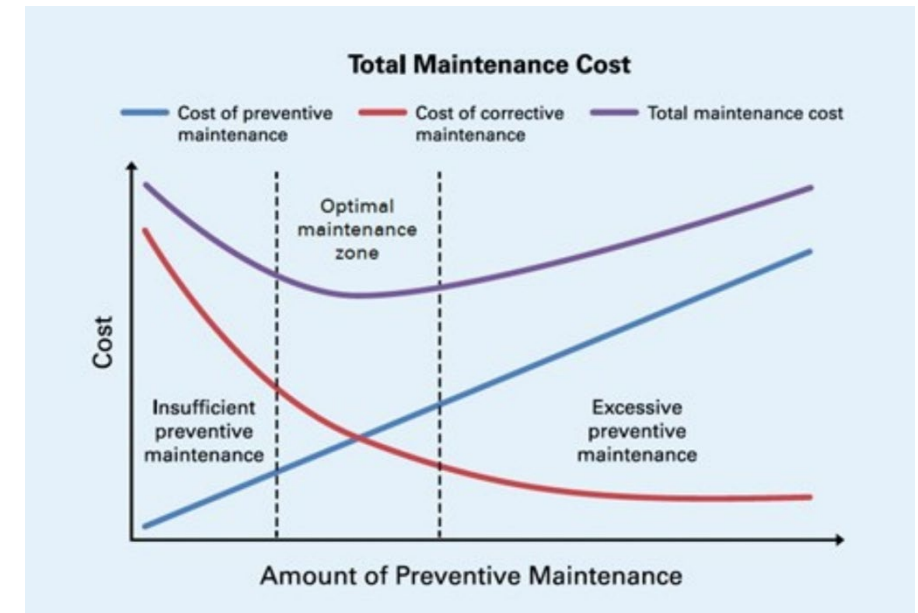
# Where can Predictive Maintenance help?

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- Helps extend remaining useful life



- Aid in planning maintenance to optimize downtime cost



# Demo Use Case

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Upload engine log in  
Navy demonstrator  
App

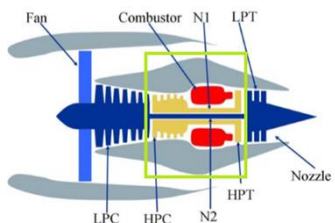
Analyse and identify  
problems

Automatically raise PR in  
Teamcenter  
using eQ's product suite

Review auto created  
PR Analysis Report

- Dataset

- Turbofan engine run to failover simulation dataset
- 4 sets of approx. 700 engine runs data available for training models
- Each engine containing 200-300 cycles and data for 21 sensors + 3

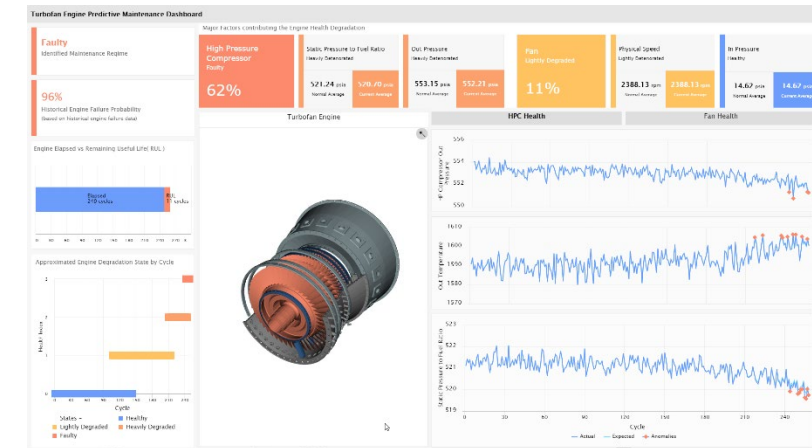


- Analysis objectives

- Identify remaining useful life of engine
- Anomalous behavior in sensor readings
- Degradation status of engine and parts
- Main causes (sensors) for engine degradation

- Machine Learning and Statistical Models used,

- XGBoost
- Deep learning based auto encoders
- Spectral Clustering
- Hidden Markov Models
- Predefined Semantic Rules

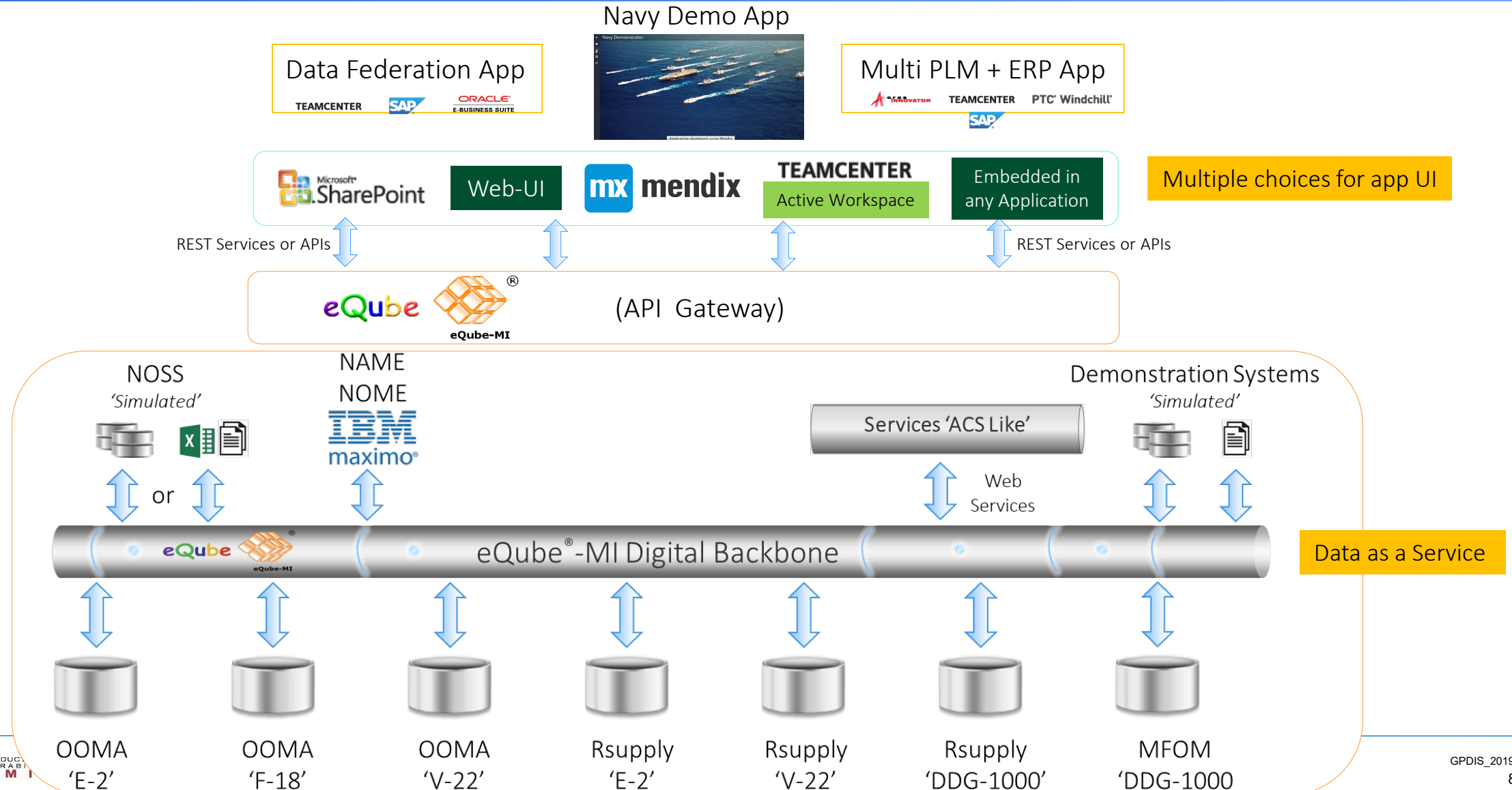


User can approve or reject the PR after reviewing the report

Figure 1. Simplified diagram of engine simulated in C-MAPSS [11].

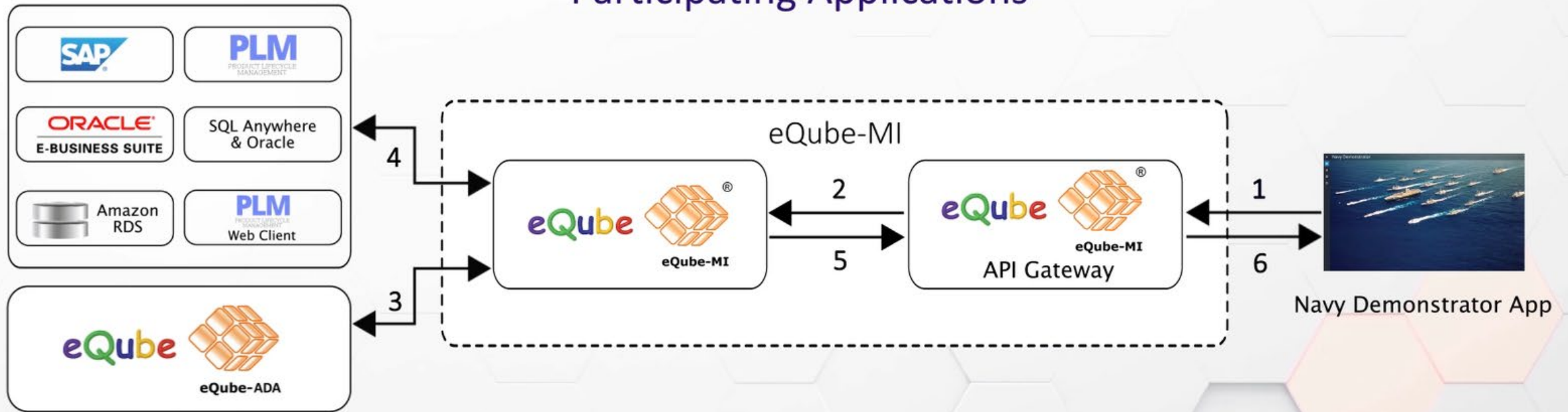
# 'For-purpose' apps leveraging eQube

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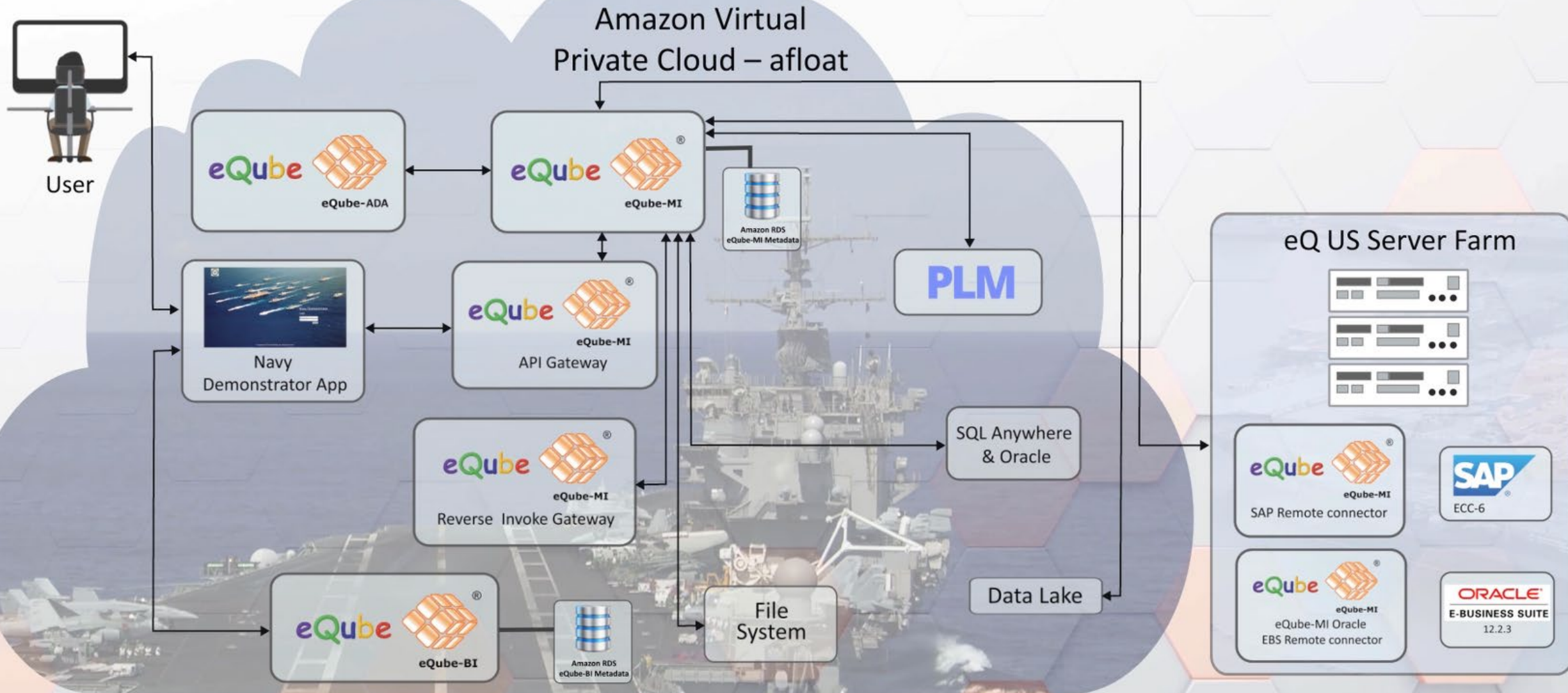
## Participating Applications



1. Navy Demonstrator invokes REST service exposed on eQube-MI API Gateway
2. eQube-MI API Gateway invokes underlying eQube-MI Business process
3. eQube-MI invokes eQube-ADA which analyzes sensor data and identifies anomalies / remaining life of parts
4. eQube-MI Business process fetches/updates data in all the other applications
5. JSON response sent to eQube-MI API Gateway from eQube-MI Business process
6. JSON response sent to Navy Demonstrator from eQube-MI API Gateway

Web application invokes  
eQube-MI processes exposed as REST  
or SOAP Web service

# About this demonstration





# About this demonstration

Web application invokes  
eQube-MI processes exposed as REST  
or SOAP Web service



Amazon Virtual  
Private Cloud – afloat

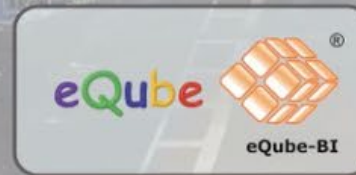
User fetches master information such as Aircraft type,  
Tail #, Point of Arrival(POA) and Point of Departure(POD)



PLM



SQL Anywhere  
& Oracle



File  
System

Data Lake

eQ US Server Farm



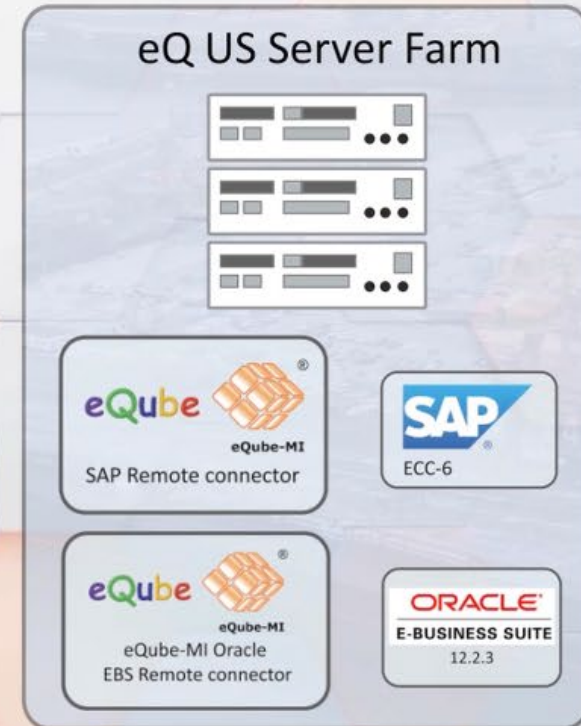
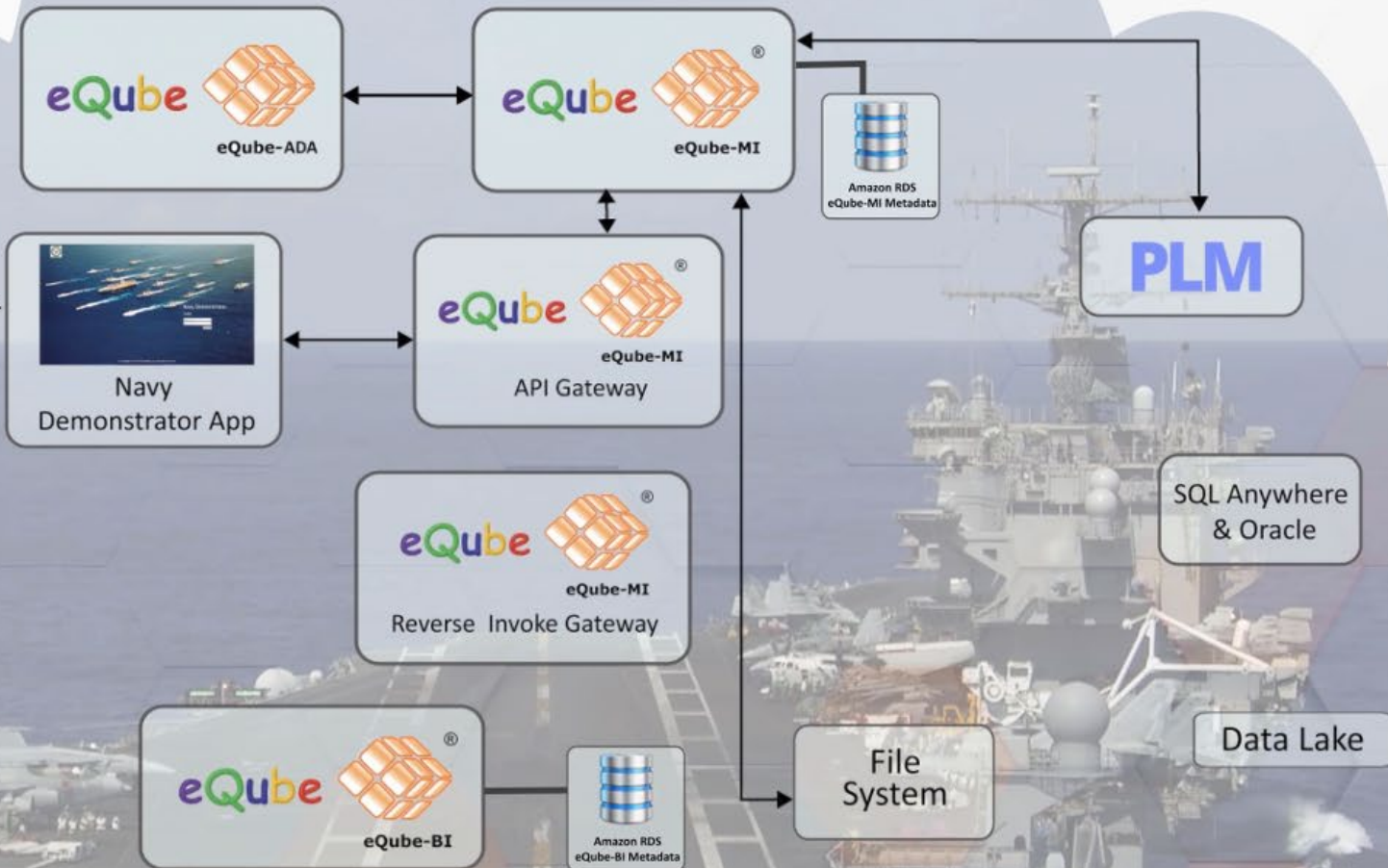
# About this demonstration

Web application invokes  
eQube-MI processes exposed as REST  
or SOAP Web service



Amazon Virtual  
Private Cloud – afloat

Flight information saved in Amazon RDS,  
flight data saved on file system and PR  
created on PLM and submitted to workflow





# About this demonstration

Web application invokes  
eQube-MI processes exposed as REST  
or SOAP Web service

Amazon Virtual  
Private Cloud – afloat

Flight information saved in Amazon RDS,  
flight data saved on file system and PR  
created on PLM and submitted to workflow



Even if user does not detect/report a problem,  
eQube-ADA automatically analyzes sensor data,  
identifies faulty part and causes a PR to be created

Demonstrator App



File System

SQL Anywhere  
& Oracle

Data Lake

eQ US Server Farm



# About this demonstration

Web application invokes  
eQube-MI processes exposed as REST  
or SOAP Web service

Advances  
Workflow



User

Amazon Virtual  
Private Cloud – afloat

Workflow advanced by user



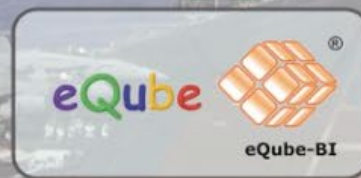
Navy  
Demonstrator App



PLM



SQL Anywhere  
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File  
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Data Lake

eQ US Server Farm

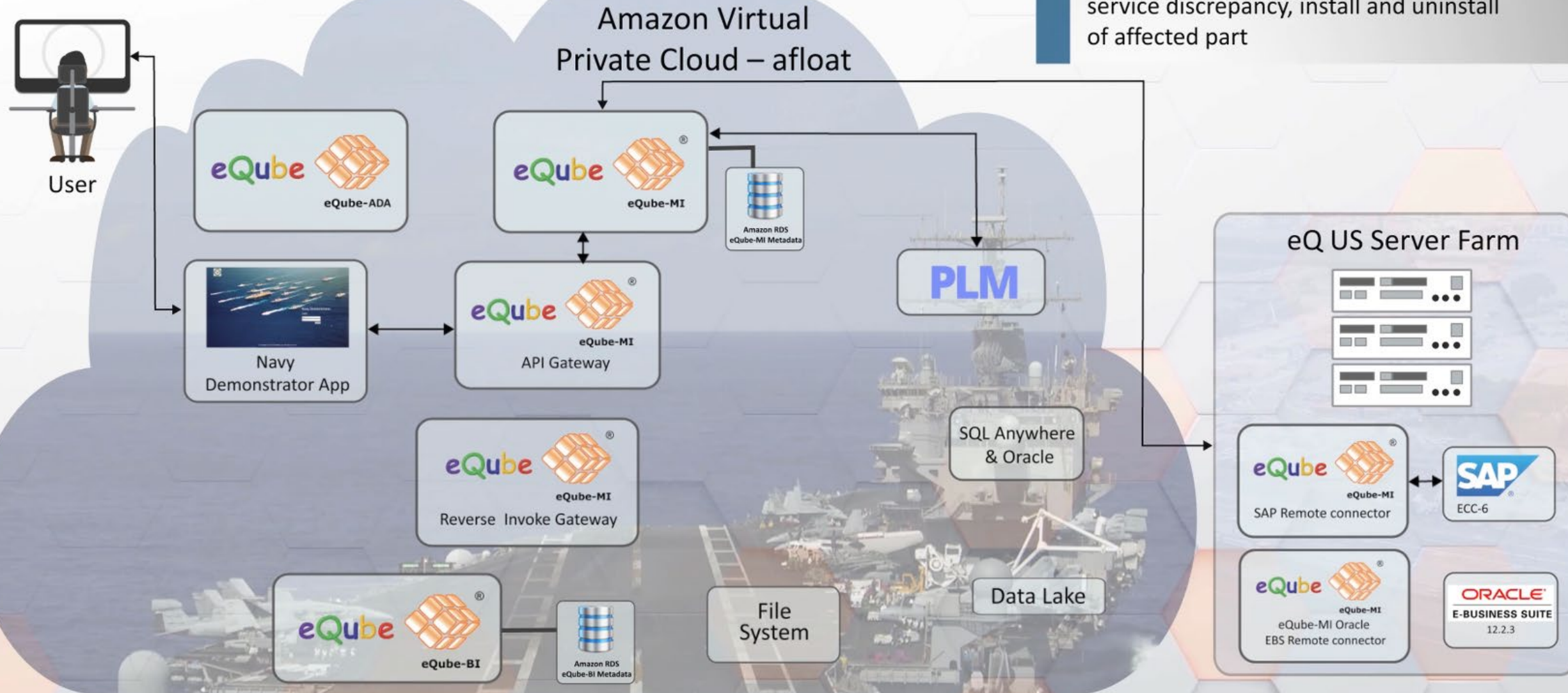




# About this demonstration

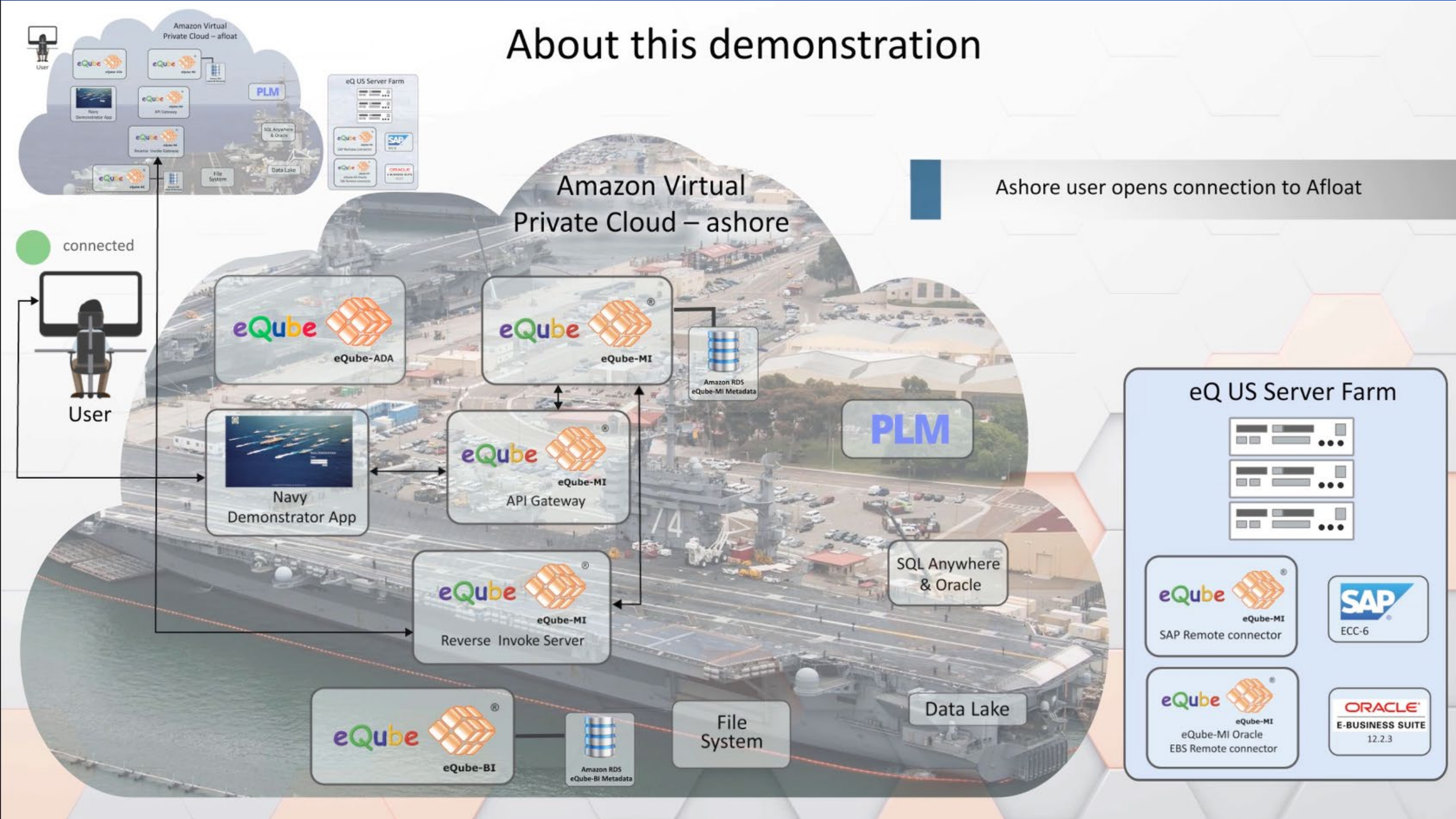
Web application invokes  
eQube-MI processes exposed as REST  
or SOAP Web service

Through Navy Demonstrator App, user  
performs action on worklist resulting in  
service discrepancy, install and uninstall  
of affected part





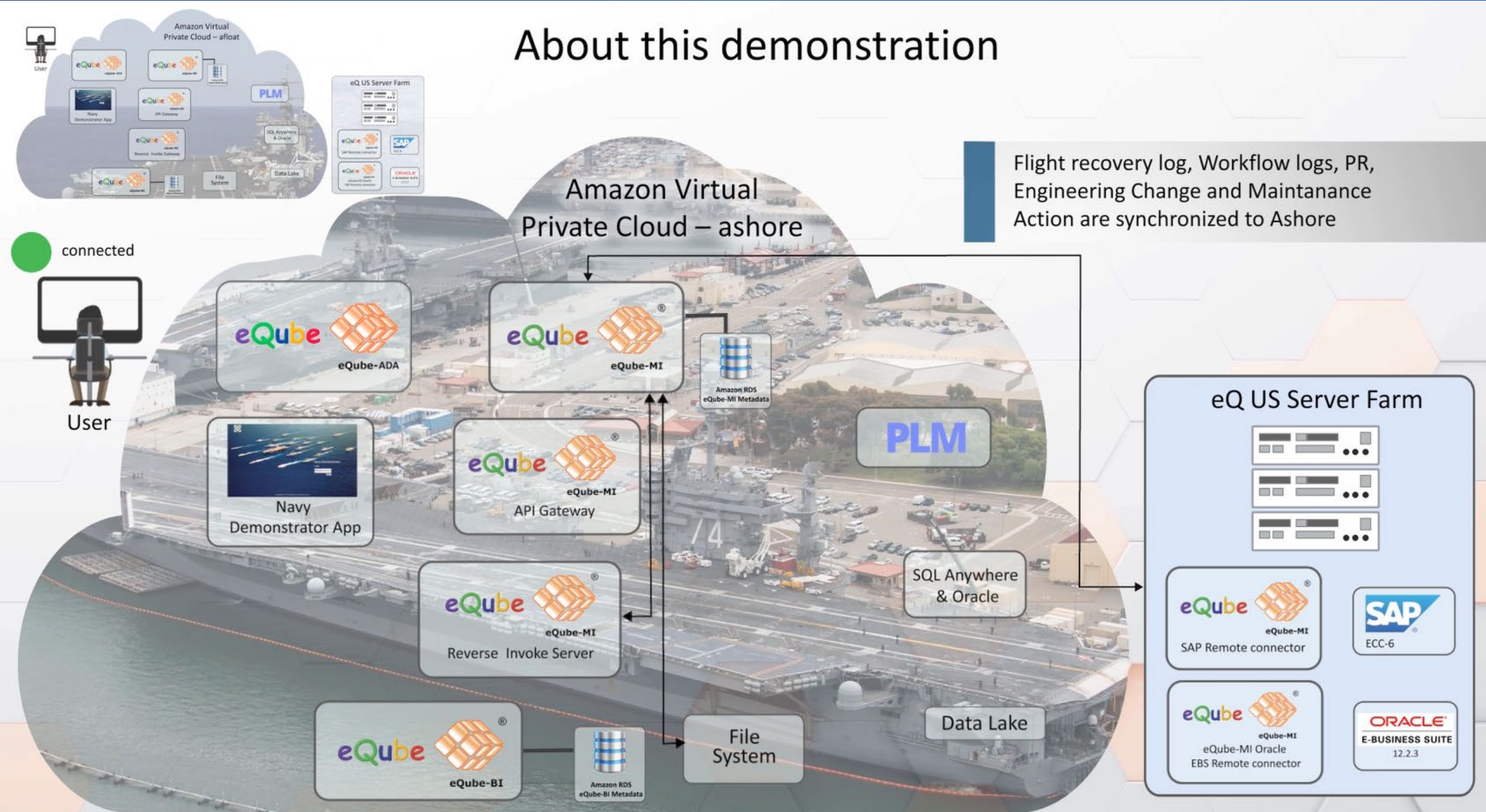
# About this demonstration





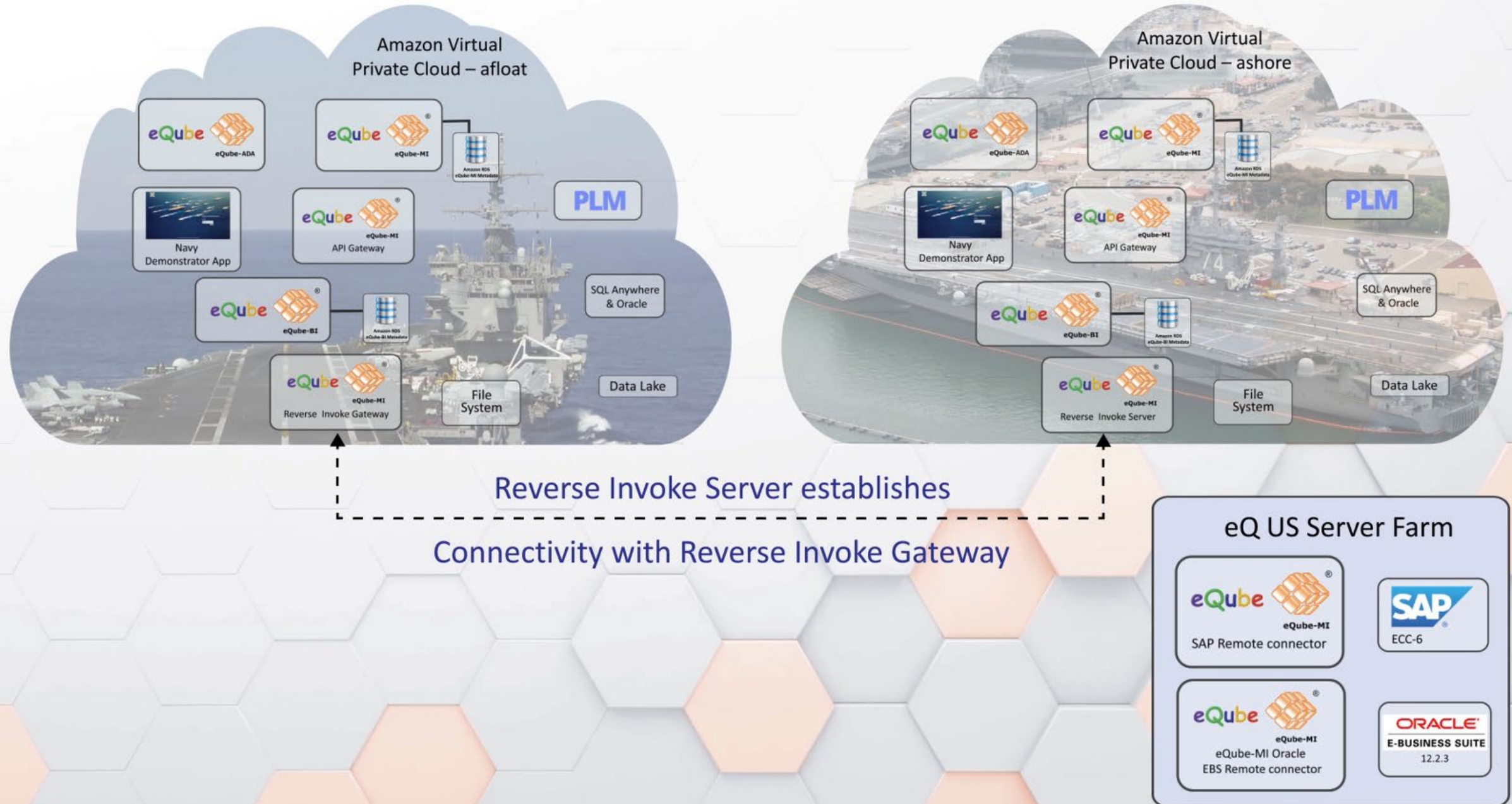
# About this demonstration

Flight recovery log, Workflow logs, PR, Engineering Change and Maintenance Action are synchronized to Ashore





## Synchronization between afloat and ashore



## Scenario 1 – Receive aircraft with no Incident

- Afloat - Create Flight Recovery Log & download flight data
- Review Analysis done by eQube-ADA
- Synchronize with ashore
  - Flight Recovery data
  - Ingest sensor data to data lake



## Scenario 2 – Receive aircraft with auto-detected Incident and repair on board

- Afloat - create Flight Recovery Log & download flight data
- User does not log problem
- eQube-ADA review analysis however, detects a problem and raises a PR
- Disposition to repair on board, perform repair and record aircraft configuration
- Synchronize with ashore
  - Flight Recovery data
  - Ingest sensor data to data Lake
  - Problem Report
  - Aircraft configuration changes
  - Inventory changes



## Scenario 3 – Receive aircraft with Incident and request Eng'g assistance

- Create Flight Recovery Log & download flight data
- Create and review Problem Report (PR) of Incident
- Disposition Engineering Support - required from ashore
- Synchronize with ashore
  - Flight Recovery data
  - Ingest sensor data to data lake
  - Problem Report

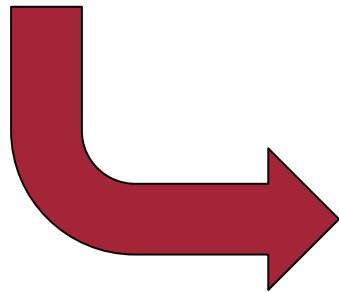
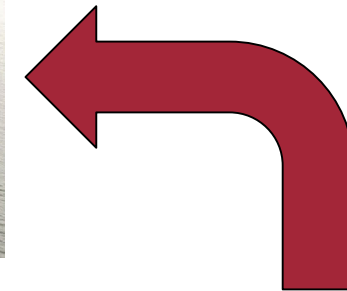
# MBPS – Day on the Life

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Ashore



<https://vpc2-east-aws.1eq.com/eQubeMI/navy>



Afloat




# eQube® Navy NOBLE IDE Demonstrator Goals

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- Quickly build “for purpose” apps, unlocking the data and make various operations efficient
  - eQube Data Virtualization layer enables integrated view of data by federating data across multiple systems with Predictive Analytics
- Expose and utilize APIs – leverage NOBLE, CANES, M&SWP, ALE and ACS investments
- Create single view of data that can be governed and managed.
- Will perform and scale to support current and future applications that Navy will use.
- Will support all D-DIL modes
- Work across diverse application deployment landscapes – On premise, cloud, Hybrid, etc.
- Service orchestration capability as well as ability to orchestrate workflows in COTS applications such as PLM, MRO and ERP.
- Security using PKI, CAC cards etc.
- Simplified administration and deployment Containerization, Elastically Scaled
- Perform machine learning and leverage automated analytics

# Digital Backbone – Actionable Insight



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eQ Technologic Inc,

# Predictive Maintenance Analysis

Navy Demonstrator Report Overview

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# Analysis techniques used

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## ■ Engine Reliability/Survival score

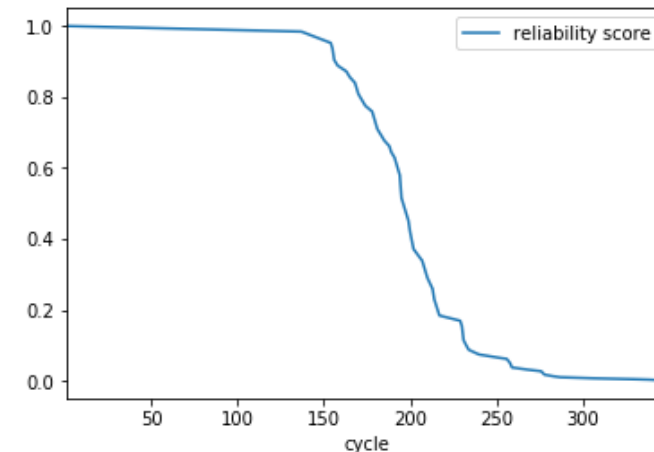
- Model predicts survival probability for entire engine lifespan
- Takes into account the sensor readings at latest cycle while predicting scores
- Calculated using cox-proportional hazard model on historical engine failure data

## ■ Remaining useful life

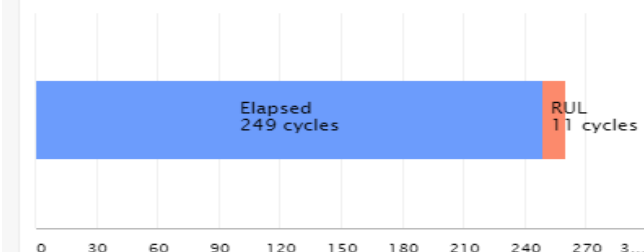
- Prediction model (XGBoost - regression) built to learn the remaining useful life based on sensor readings on train data
- Deployed model used to predict remaining useful life for uploaded log data

96%

Historical Engine Failure Probability  
(based on historical engine failure data)



Engine Elapsed vs Remaining Useful Life( RUL )

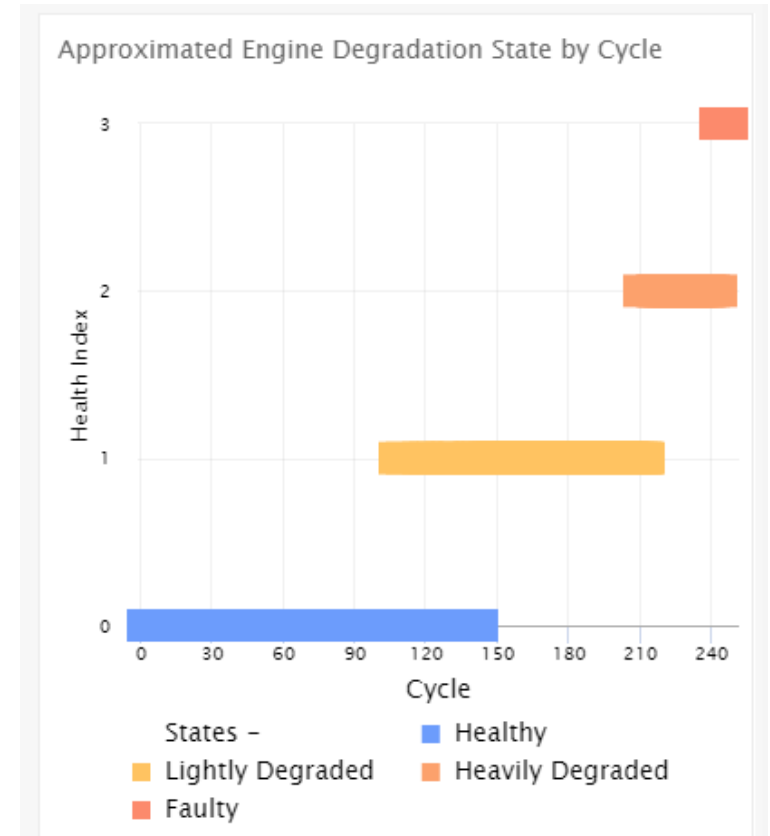
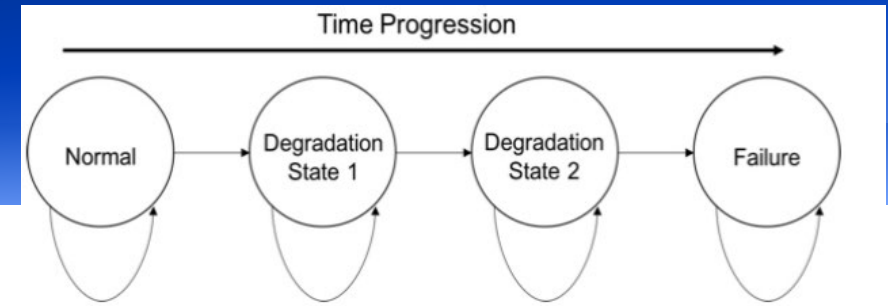
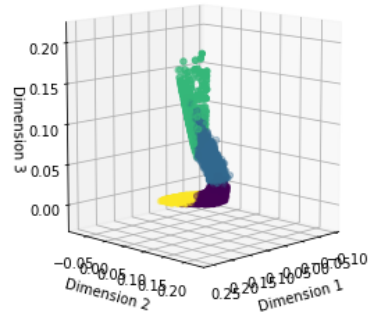
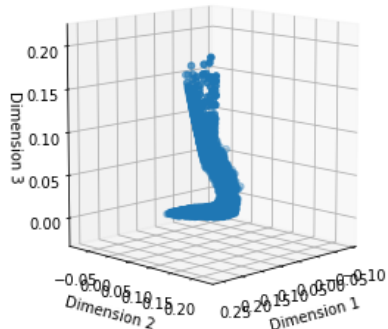


# Analysis techniques used cont..

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- Overall Engine degradation

- Estimated engine health across its elapsed life
- 4 unique health states were identified
- Degradation States will overlap on some cycles because during transition the engine will usually show properties of both cycles
- All sensors considered for analysis
  - Spectral embedding was used to reduce dimensionality of the data
  - Spectral clustering was used to cluster data into unique states

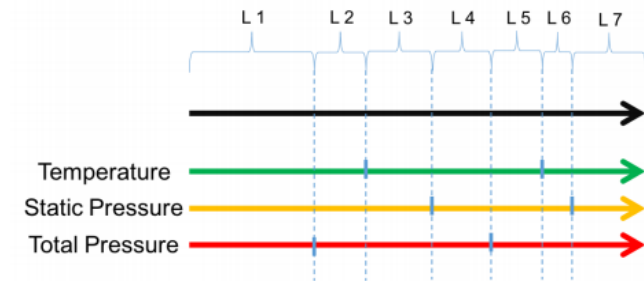
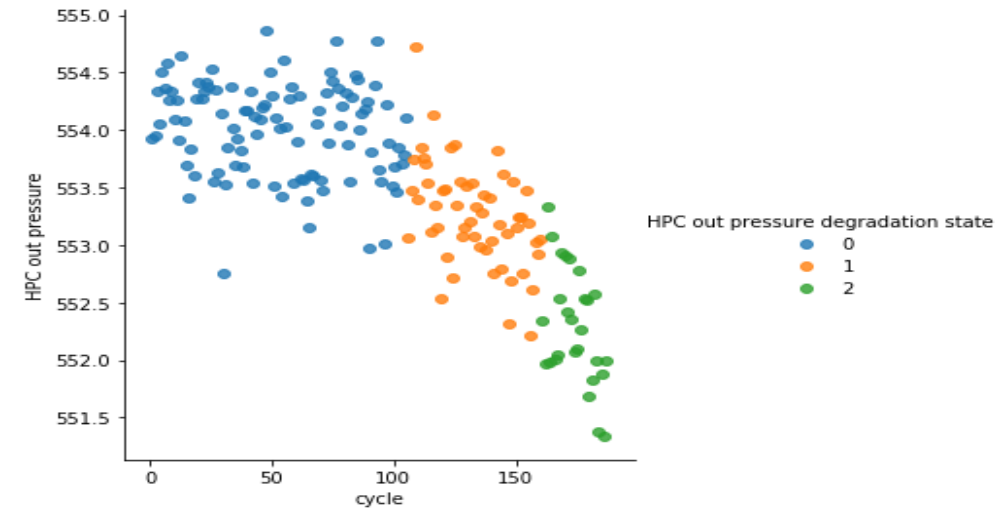




# Analysis techniques used cont..

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- Part Degradation from Sensor states
  - Hidden markov models used to segment individual sensor data
  - Individual Part degradation were estimated from individual sensor states and defined heuristics
  - Thus a score was assigned to each part and the highest degraded part is found out



**FIGURE** The overall machinery deterioration level division and identification based on each fin-grained KPI deteriorating states according to predefined rules.

Major Factors contributing the Engine Health Degradation

High Pressure Compressor Faulty

62%

Static Pressure to Fuel Ratio  
Heavily Deteriorated

521.24 psia  
Normal Average

520.70 psia  
Current Average

Out Pressure  
Heavily Deteriorated

553.15 psia  
Normal Average

552.21 psia  
Current Average

Fan  
Lightly Degraded

11%

Physical Speed  
Lightly Deteriorated

2388.13 rpm  
Normal Average

2388.13 rpm  
Current Average

In Pressure  
Healthy

14.62 psia  
Normal Average

14.62 psia  
Current Average

# Analysis techniques used cont..

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- **Sensor Anomalies**

- Anomalies identified using deep learning based auto encoder models
  - Normal state of each sensor identified using hmm models
  - Auto encoder model trained to learn the normal state of the data
- The deployed model used on uploaded logs to identify the anomalies
  - The model also output the expected value of the sensor which is plotted alongside the actual value

