

# Advanced Methods for Condition Monitoring and Control of Large Engines in Maritime Applications



# Let's Start With Some History...

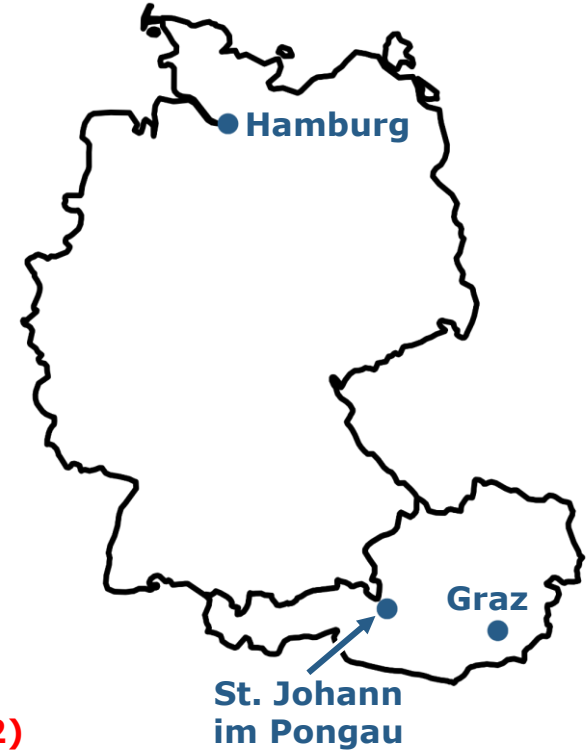
...My History with Hamburg



When I think of Hamburg, this is what I picture\*:



**Queen Elizabeth 2 (QE2)**  
Maiden call in Hamburg (1972)



# Let's Start With Some History...

## ...On the Queen Elizabeth 2 (QE2)



- **QE2 facts** (just a few)
  - British transatlantic ocean liner (Cunard Line)
  - In service 1969–2008
  - 70,327 GT | 293.5 m long | 32.1 m wide
  - Service speed ~28–32 knots
  - Engine room
    - Originally steam turbine powerplant
    - Refit 1986/87 to diesel electric with 9 medium-speed engines (95.6 MW in total)
  - Sophisticated alarm and monitoring system
- **What has changed since the QE2 era in condition monitoring?**
  - Rise of artificial intelligence (AI) and machine learning (ML)
  - Emergence of edge computing for decentralized, real-time processing



# Condition Monitoring of Large Engines



Where Does Artificial Intelligence Fit In?

## Condition monitoring process

Condition detection

Condition comparison

Diagnosis



Virtual sensors



Reference models



Anomaly detection



Digital twins

Large language models

Early failure indicators

Fault

### Key message:

Many opportunities for  
**Artificial Intelligence**  
and **Machine Learning**



Wear/damage compensation

**Control system**



Model predictive control

Parameter tuning

Model-free ML control

Preventive & condition-based

Corrective

**Maintenance**

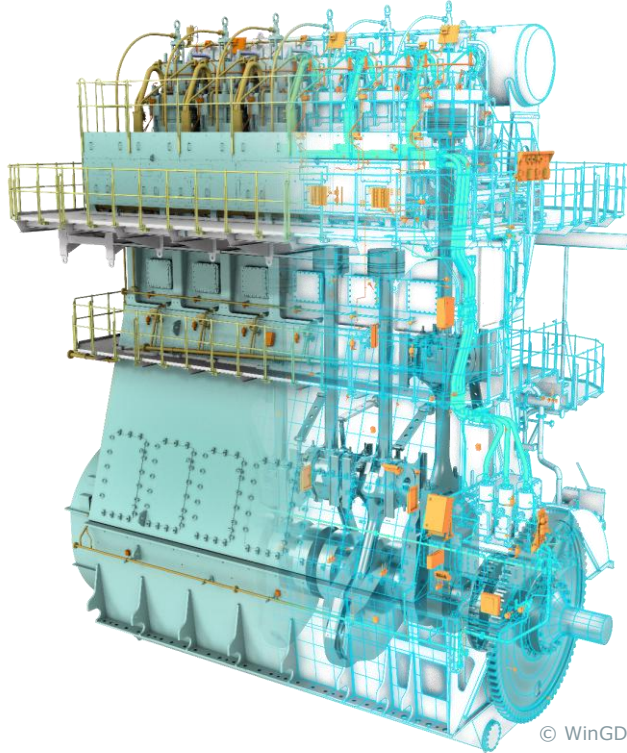


Remaining useful life prediction

Optimal maintenance scheduling

# Condition Monitoring of Large Engines

## Condition Monitoring Systems



© WinGD

- **Engine level systems**

- Typically provided by engine manufacturer/designer
- Instrumentation of entire engine measuring, e.g.,
  - Engine speed
  - Engine load
  - Cylinder pressures
  - Bearing temperatures
  - Fuel mode
  - Injection settings
  - Valve timings
- Contextual data on navigation, weather, etc.
- Standalone platforms offered for data analytics



# Condition Monitoring of Large Engines

## Condition Monitoring Systems



- **Component level systems**

- Typically provided by component manufacturer
- Component-specific instrumentation
  - Standalone system with local focus
  - No engine level data access/input
  - Operational context information partly detectable
- Components can become “intelligent” through

Advanced instrumentation + Data analytics (AI/ML) → Value-added data (VAD)

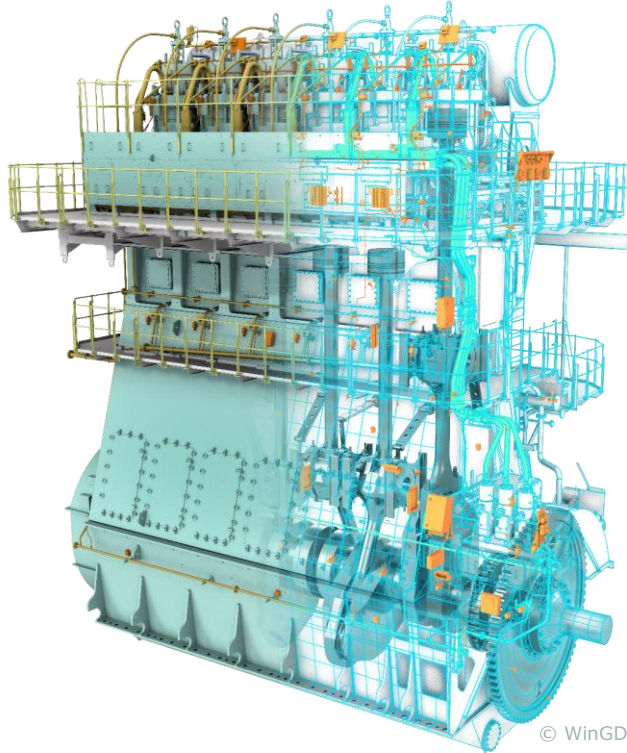
- Standalone platforms offered for data analytics



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# Condition Monitoring of Large Engines

How to Benefit From Both Engine and Component Level?



© WinGD

## Data sharing

Intelligent  
component VAD



Engine operation  
and  
contextual data

## Scenarios

Permanent data  
sharing

Learn self-  
contextualization



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# Advanced CM of Large Engines

## How to Start Such an Endeavor?



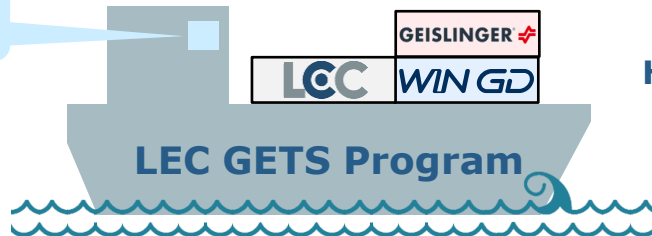
### Engine level

- Large two-stroke low-speed engines by **WINGD**
- Fuel types: LNG, Diesel, NH<sub>3</sub>, MeOH, EtOH, LPG
- WinGD Integrated Digital Expert (WiDE) providing
  - Engine data monitoring, diagnostics and troubleshooting capabilities
  - Onboard and offshore support

### Component level

- Torsional vibration (TV) dampers by **GEISLINGER** 
- Mounted at crankshaft free end
- Geislinger monitoring system (GMS) providing
  - Continuous acquisition and
  - Evaluation of torsional vibrations } Intelligent TV damper

*Where to steer in the sea of data?*



### Simulations

Historical data



Test beds

Field operation



# Advanced CM of Large Engines

## Where to Get Field Operation Data From?

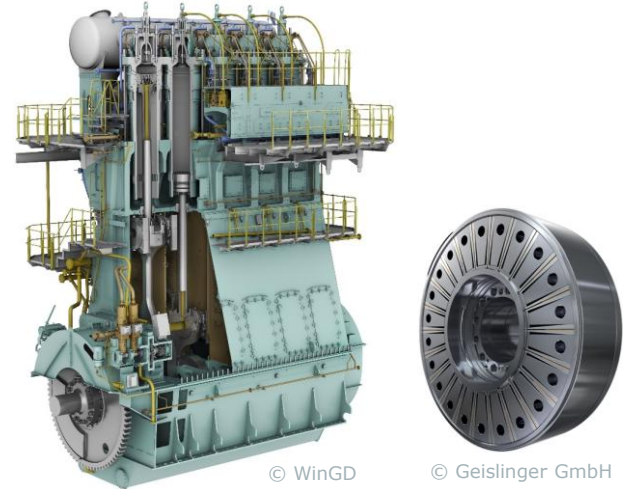


- **Pilot installation**

- Ocean-going tanker vessel in regular operation
- Equipped with
  - 5X72DF WinGD dual fuel engine featuring WiDE
  - Geislinger TV damper featuring GMS

- **Database generation**

- Navigation, voyage, engine, and vibration data
  - Collected onboard with WiDE and GMS
  - Transmitted via secure channels
- Post-hoc data synchronization (1 sample/min)
- Additional data preprocessing for further analysis



→ **Database of 253 informative features covering more than one year of vessel operation**

# Advanced CM of Large Engines

## Use Case: Virtual Load Sensor

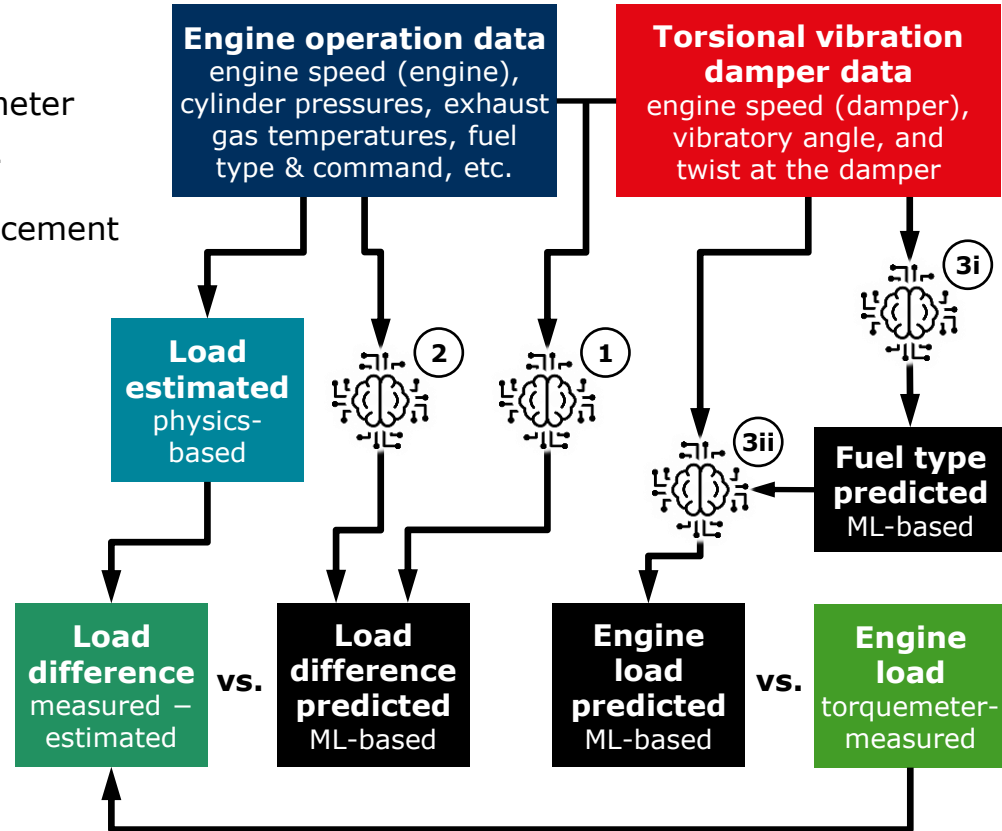


- **Motivation**

- Engine load is key performance parameter
- Currently measured with torquemeter
- Virtual load sensor as backup or replacement

- **Modeling scenarios**

1. Combined data sources
2. Engine operation data only
3. TV damper data only (incl. fuel type prediction)



# Virtual Load Sensor

## Machine Learning Framework and Results

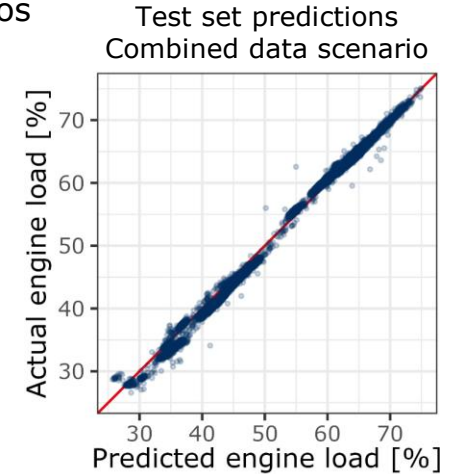


- **Model training and comparison**

- Nested k-fold cross-validation (CV) framework employed
- Two voyages as final test set (data not used for model training)
- Various AI/ML methods compared under multiple feature scenarios

- **Results**

- Best CV results with combined data scenario (Mean absolute percentage error, MAPE, of 0.59 %)
- Error on unseen test data: MAPE of 1.14 %
- TV damper only scenario
  - CV error: MAPE of 0.83 %
  - Test error: MAPE of 1.25 %
  - Fuel type classification: 0.99 Matthews correlation coef.



→ **Results demonstrate benefit of data sharing, as well as potential for self-contextualization**

# AI-Enhanced CM = Success?

Great Potential But Some Challenges Remain



- **Challenges of AI-enhanced CM for maritime applications**

- Generalization to changing operating conditions and fleet-wide application
- Field data acquisition cumbersome (often third-party ownership)
- Scarcity of faults/failures during real-world operation
- Trustworthiness, explainability and legal aspects
- Cybersecurity and system resilience
- Technical vs. commercial feasibility
- Environmental impact of AI
- ...

**Where lies the future  
of AI-enhanced CM?**

**Of course, tackle all  
these issues!**

**Edge-based solutions  
with tinyML**

# Edge-Based CM With TinyML

## LEC's Motivation



- **TinyML in general**

- ML subfield focused on deploying lightweight models on microcontrollers and other edge devices
- Data analytics performed close to data source (reducing data volume and latency)

- **Advantages for large engine CM**

- Bring existing cloud-based solutions back to the engine or component
- Enable data-intensive condition detection methods
- Allows for immediate sensor/measurement self-validation

- **Some use cases in engine development**

- Sensor validation in telemetry-based condition detection systems
- Condensation of high-frequency acoustic emission data
- Real-time control reaction in case of combustion anomalies





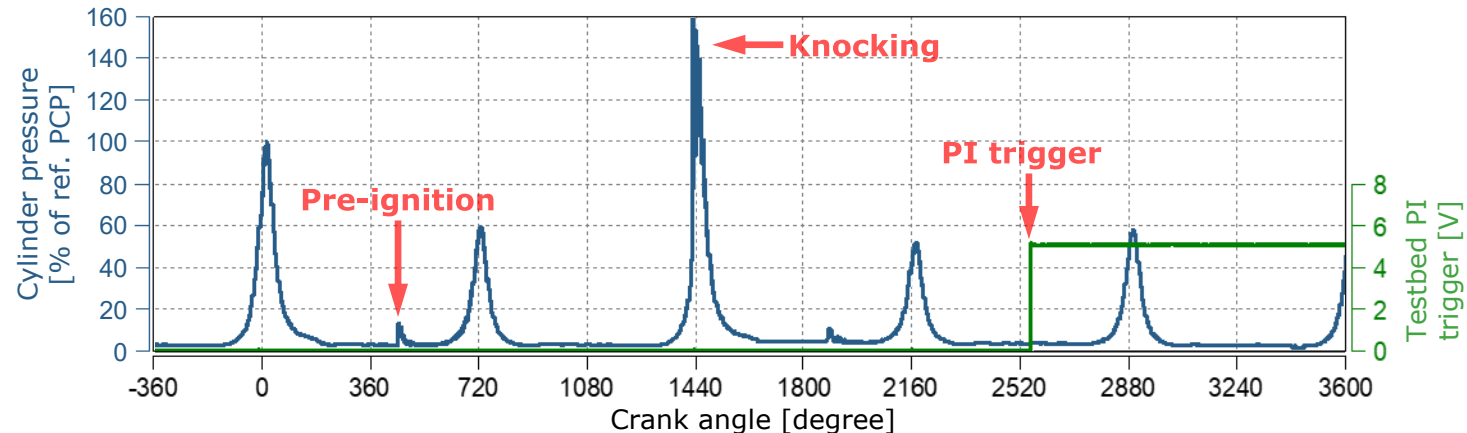
# Advanced CM of Large Engines

Use Case: Edge-based *Knock Block*



- **Motivation**

- Pre-ignition (PI) events during high-load hydrogen operation often followed by knocking
- However, PI detection from test bed software is too slow for triggering a countermeasure



# Advanced CM of Large Engines

## Use Case: Edge-based *Knock Block*

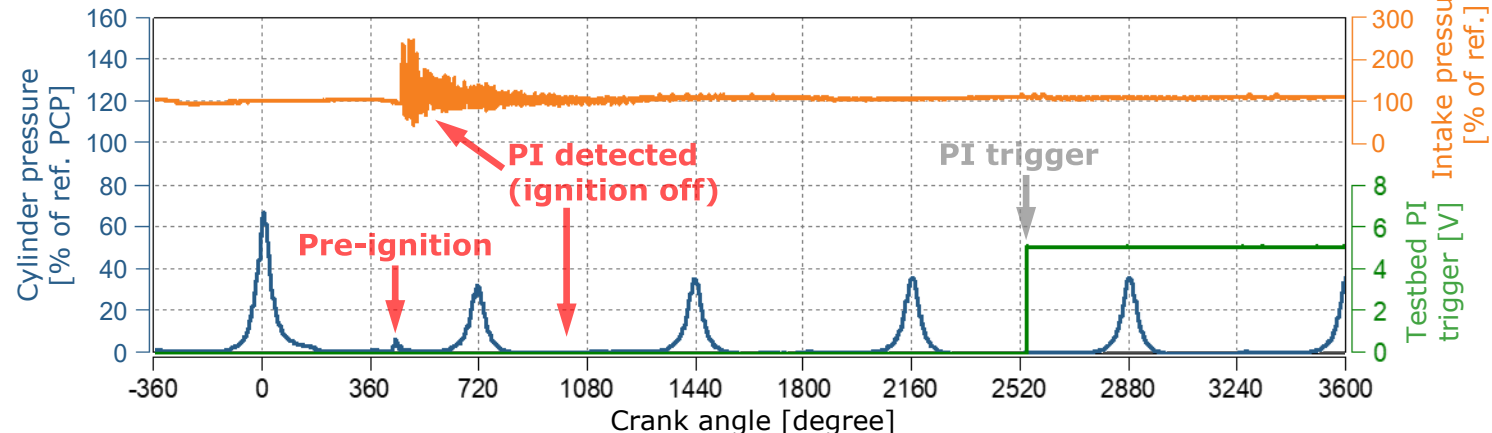


- **Motivation**

- Pre-ignition (PI) events during high-load hydrogen operation often followed by knocking
- However, PI detection from test bed software is too slow for triggering a countermeasure

- **Knock Block**

- Edge computing is used to detect PI via intake manifold pressure (microcontroller-based, few € cost)
- For immediately following cycle, injection is switched off → knocking is avoided



# Advanced CM of Large Engines

## A Journey Worth Taking?



- **Condition monitoring process** offers many opportunities for advancement

Condition detection

Condition comparison

Diagnosis

- Great potential of **AI-enhanced CM**, yet some challenges remain 
- **Edge-based analytics** promising to
  - support in engine development,
  - reduce AI's environmental impact in field applications,
  - and make CM more resilient and reliable overall

**“It's worth remembering that it is often the small steps, not the giant leaps, that bring about the most lasting change.”**

Queen Elizabeth II  
(the person, not the ship)

# Thank you for your attention!

The logo consists of the letters 'LEC' in a bold, white, sans-serif font, centered within a large, light green circle. The background of the slide is a teal color with a pattern of small, white, stylized icons representing various scientific and technological fields like energy, transport, and sustainability.

## Evolutionary Green Energy

## and Transport Systems

## for a Sustainable Tomorrow

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