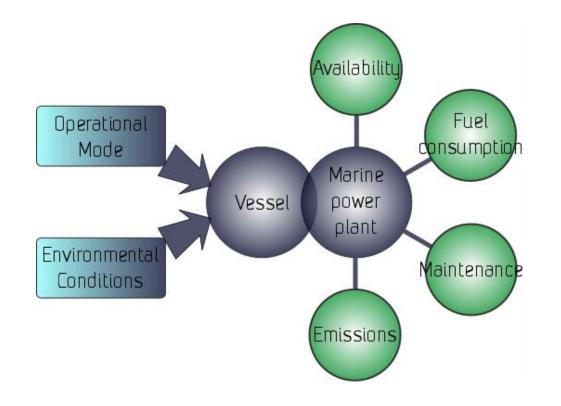


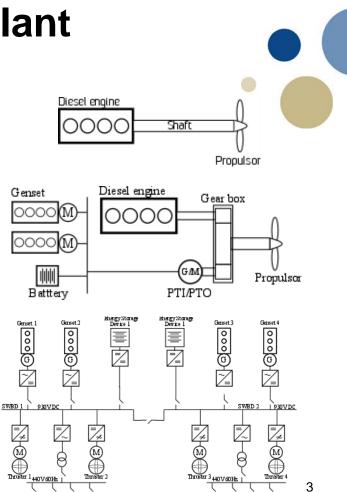
Transient Performance and Emissions of Turbocharged Diesel Engines for a Marine Power Plant

Numerical Simulation and Experimental Investigation

Introduction

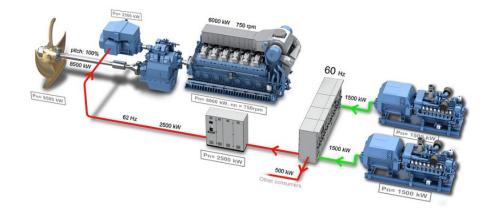
Marine Power / Propulsion Plant





Industry and Research Trend

- Hybrid Power System
- System integration
- Simulation based design and analysis

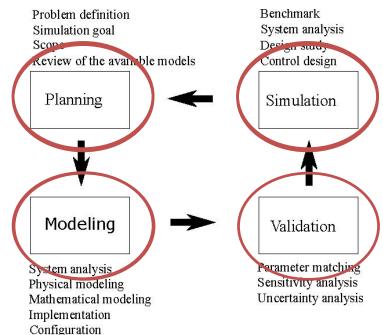




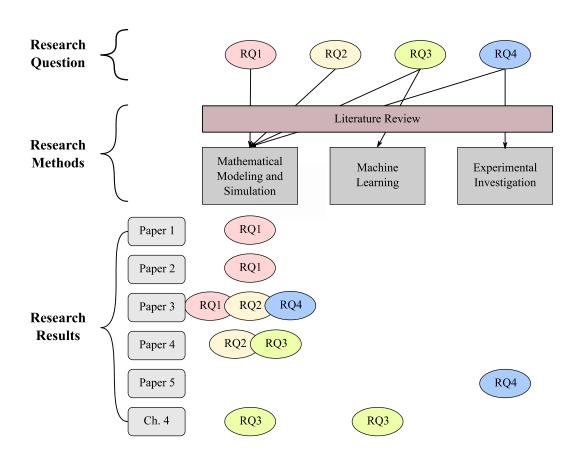
Research Questions

How can one build a mathematical model of a turbocharged diesel engine for various contexts of marine propulsion or power systems? Problem definition

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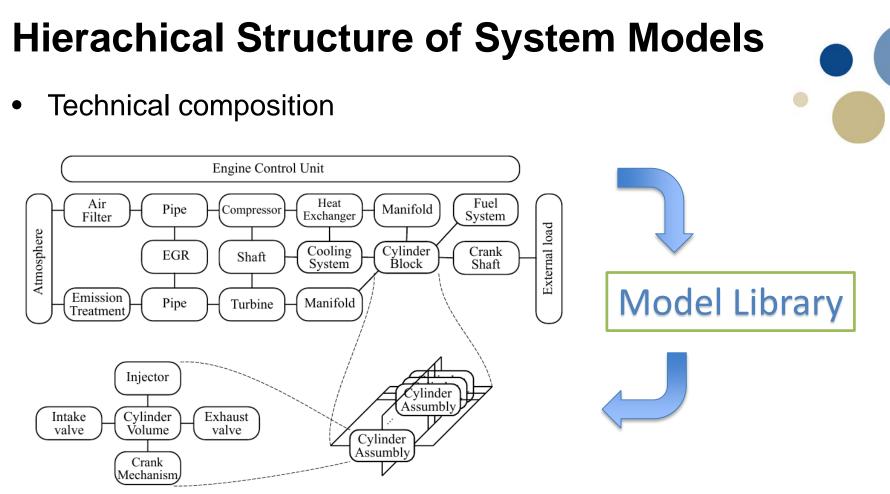
Research Methods and Results





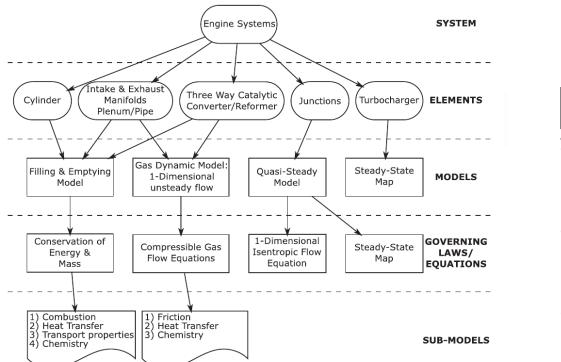


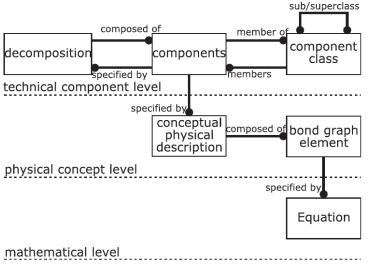
Modeling Framework for a Turbocharged Diesel Engine



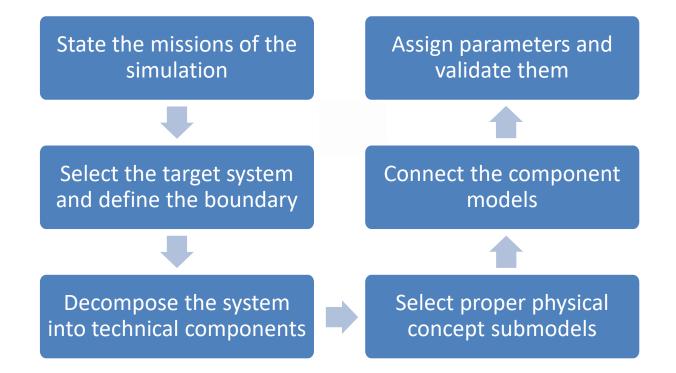
Hierachical Structure of System Models

• Level of Abstraction





Procedure for building a system model from the model library

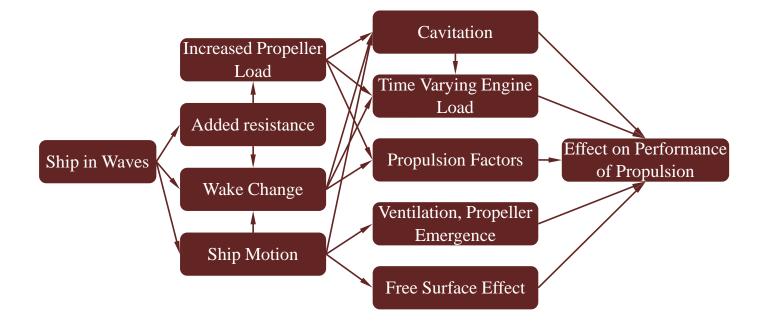




Diesel Engine Models for Propulsion System Simulation in Waves

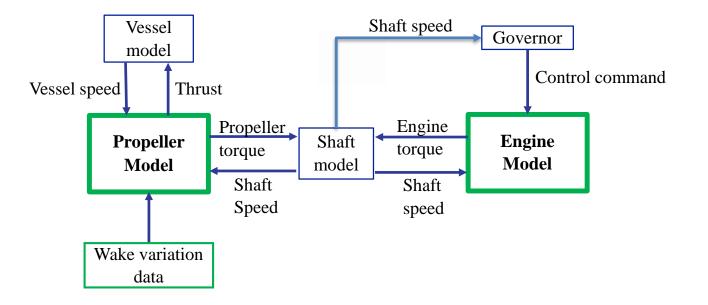
Motivation – Propulsion in Waves

• Propulsion in Waves

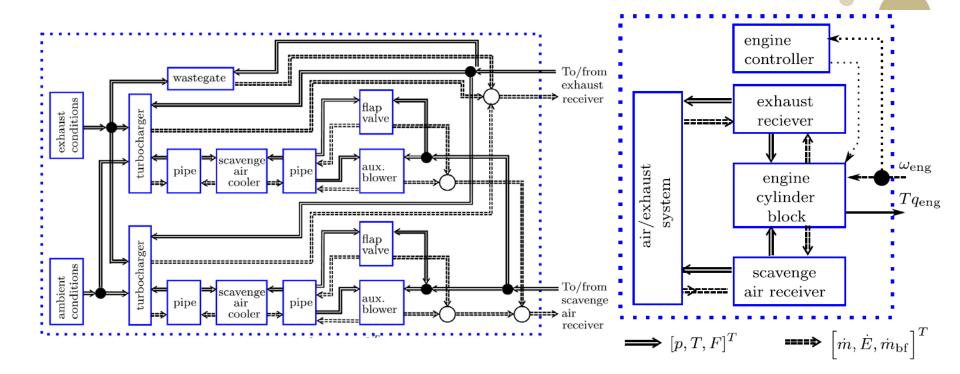


Coupled Engine Propeller Model

- Develop a hull-propeller-diesel engine simulator
- Understand the system dynamics in propulsion in waves
- Compare the transient fuel consumption with the steady-state mapping value



Engine Model Structure



Engine Model

• Modeling framework

Components		Process	Submodel	
Engine Cylinder	Control volume	Thermodynamic states	0D single zone	
		Heat transfer	Convection only	
	Scavenge port	Mass flow	Isentropic comp. flow	
	Exhaust valve	Valve lift	A look up table	
		Mass flow	Isentropic comp. flow	
	Crank mechanism	Transformation	Kinematics only	
	Gas exchange	Scavenging	Empirical model (S- shape model by Sher)	

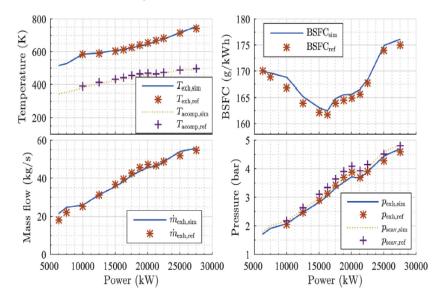
Case Vessel – KVLCC2

CFD and experimental wake data available for the KVLCC2 hull in three different wavelengths i.e. $\lambda/L = 0.6$, 1.1 and 1.6

Ship Partic	Ship Particulars Propeller Particulars		Engine Particulars		
Lbp (m)	320.0	D (m)	9.86	Model	Wartsila 8RT-flex68D
Lwl (m)	325.5	No of blades	4	Bore (mm)	680
Bwl (m)	58.0	Hub diameter (m)	1.53	Rated MCR (kW)	25,040
Depth (m)	30.0	Rotational speed (RPM)	95	Speed at rated power (RPM)	95
Draft (m)	20.8	A_e / A_0	0.431	Stroke (mm)	2720
∇ (m ³)	312622	(P/D)mean	0.47	Mean Effective Pressure (bar)	20
C _B	0.8098	Skew (°)	21.15	Number of cylinders	8
V (kts)	15.5	Rake (°)	0	Turbocharger	2 x ABB A175-L35

Validation of the model

• Steady-state



Main parameters for adjustment

- Valve timing, injection timing
- Turbine, compressor map

- Transient no data to fit
 - Sensitivity analysis

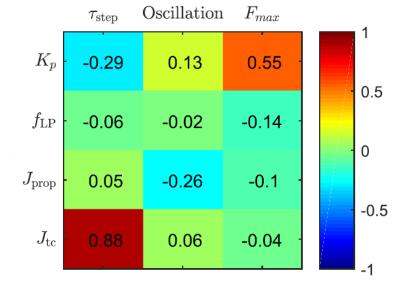


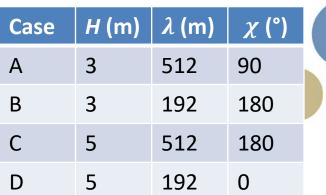
Fig. 20. Correlation matrix for the step response analysis.

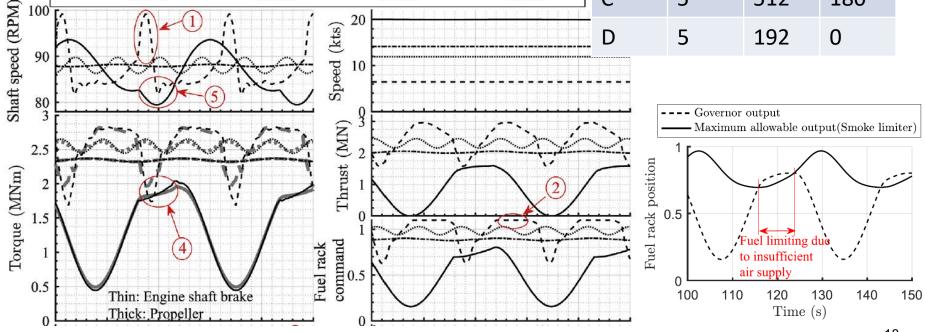
Simulation Results

..... case B

• System dynamics

----- case A





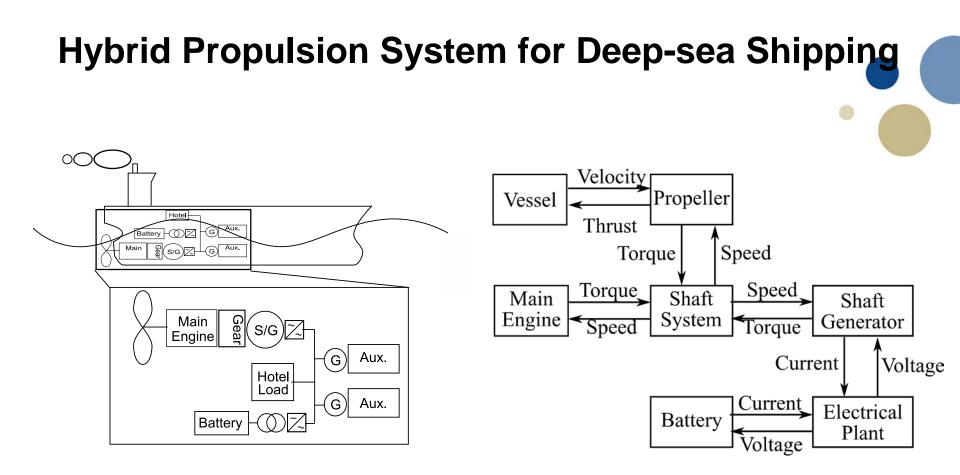
case D

---- case C

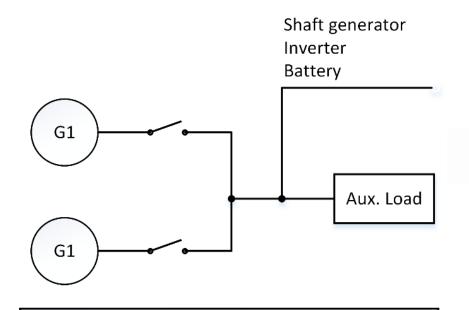
Simulation Results Transient simulation vs. steady state mapping • 3×10^{6} 176 R=-0.078354 **R=0.41542 R=-0.25555** 2.5 BSFC Difference (%) 2 1.5 0.54 0 192 512 45 90 135 180 3 352 0 Wave length (m) Wave encounter angle $(^{\circ})$ Wave amplitude (m)

Findings

- The simulator demonstrated the complex interactions between a hull, a propeller, a diesel engine and a control system.
- For transient response of the diesel engine with a governor and a smoke limiter, inertia of a turbocharger rotor plays important role as well as the control parameters.
- Steady-state mapping of the specific fuel consumption shows small deviations from the transient simulation with 0D model (fixed combustion profile).



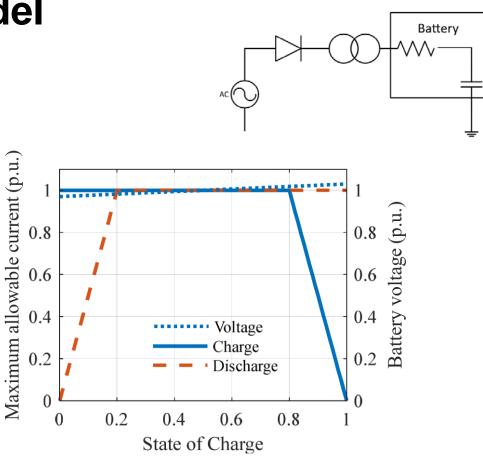
Electrical Plants



Power Management System

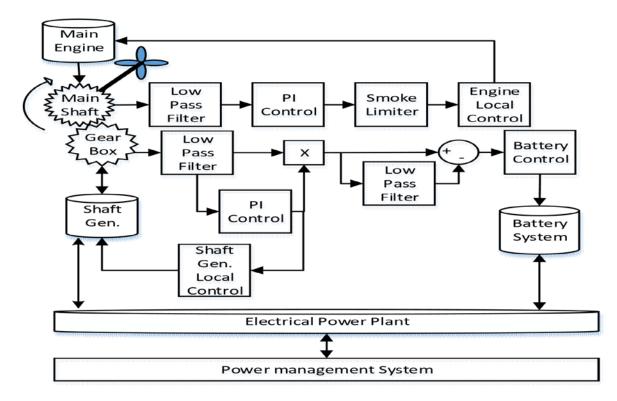
- Reference to Tom Arne Pedersen, "Bond graph modeling of marine power systems", 2009
- Transient three-phase voltage, current and frequency simulation using dq-frame modeling and transformation

Battery model





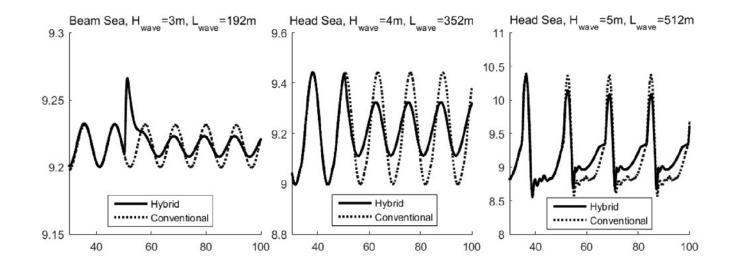
Control System



Reduction in amplitude of shaft speed and engine power
 fluctuation

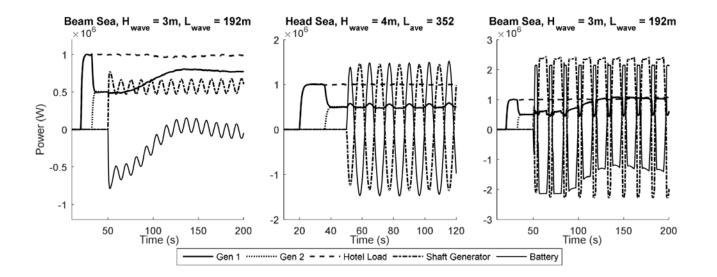
		Shaft speed	Engine	
			power	
Reduction	Mean	2.20		2.00
Ratio	Maximum	2.85		2.62
	Minimum	1.30		1.06
	Standard	0.36		0.45
	deviation			
Correlation	χwave	-0.35		-0.46
coefficient	$H_{\rm wave}$	-0.24		-0.46
to	$L_{\rm wave}$	0.24		0.13

• Time series simulation of the propulsion shaft speed

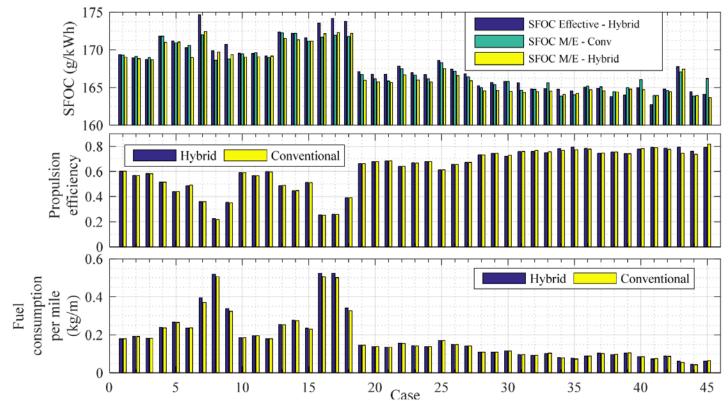




• Time series simulation of the electrical power balance



• Average Efficiency



Findings

- The simulation shows the possibility for reducing speed variation even for the mechanical propulsion using the existing installation by adding the battery system.
- The hybrid propulsion system utilized for peak-shaving in the extreme sea cases showed an equivalent level of efficiency.
- Optimization of the power plant including the main engine and the electrical power plant has potentials to improve the efficiency.
- Improvement for the battery control should be incorporated in the design of the power plant.



Experimental Investigation of the transient emissions

When is transient load dominant in the marine operation?

- Transit in waves
- DP operation in the extreme sea
- Active heave compensation in drilling or crane operation
- Crane operation => Step load

Cyclic load

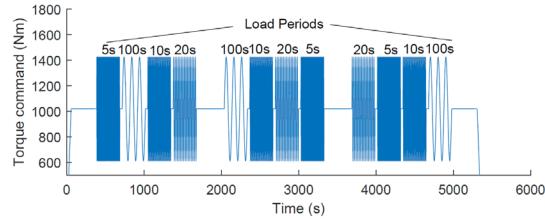


Research question

- Is there an effect of a cyclic transient load on the average fuel consumption and average NOx emission?
- If so, can we quantify the influence?

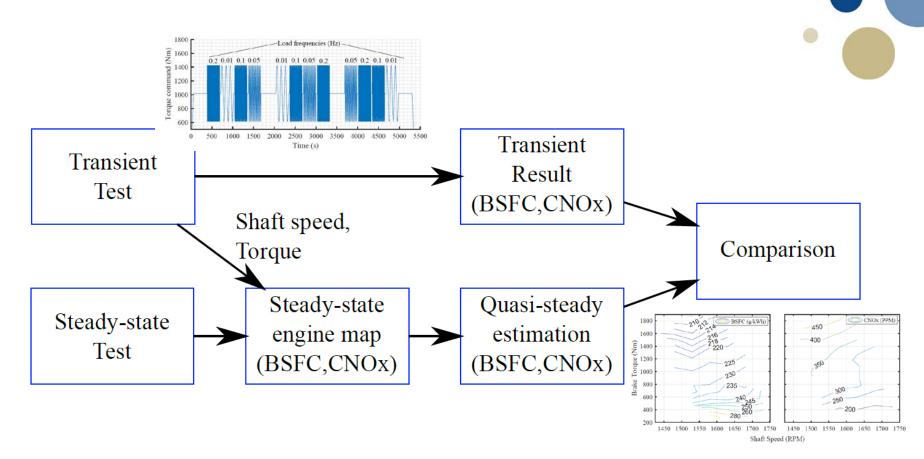
Methodology

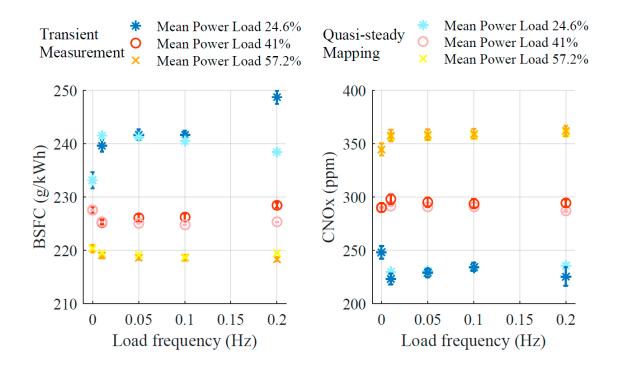
- Definition of the test case of transient load
 - Shaft speed(RPM): 1600
 - Load torque, combination of following
 - Period(s) : 5, 10, 20, 100
 - Amplitude(Nm): 407Nm
 - Mean Torque(Nm): 611, 1019, 1420

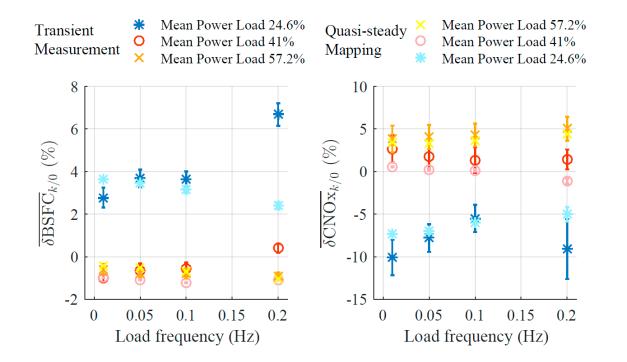




Methodology







Findings

- The effect of the cyclic transient load is found to be dependent on the average load level and becomes more visible as the load level is lowered.
- The quasi-steady mapping method provides an estimation of fuel consumption with a good accuracy, even without transient correction, for most cases.
- The effect of load smoothing for cyclic loads depends on the shape of the steady-state fuel consumption rate around the operating point.



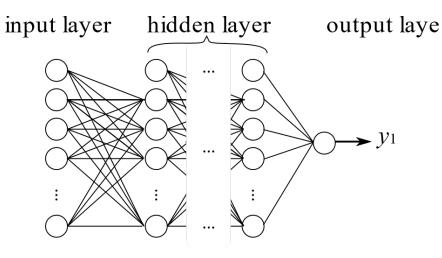
Use of Machine Learning Tools for Model Identification and model reduction

Motivation

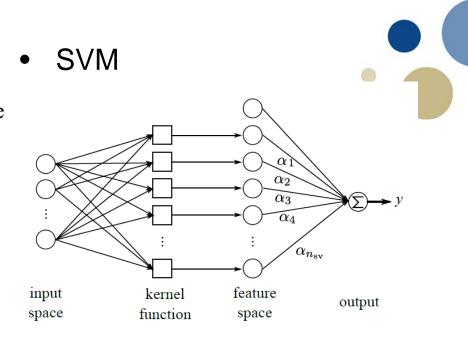
- Model identification
 - Highly non-linear relation between the time-varying parameters and the operating conditions
 - Compact representation of the dataset
- Model reduction
 - Replacement of the high fidelity model with the data-driven empirical model

ANN and SVM

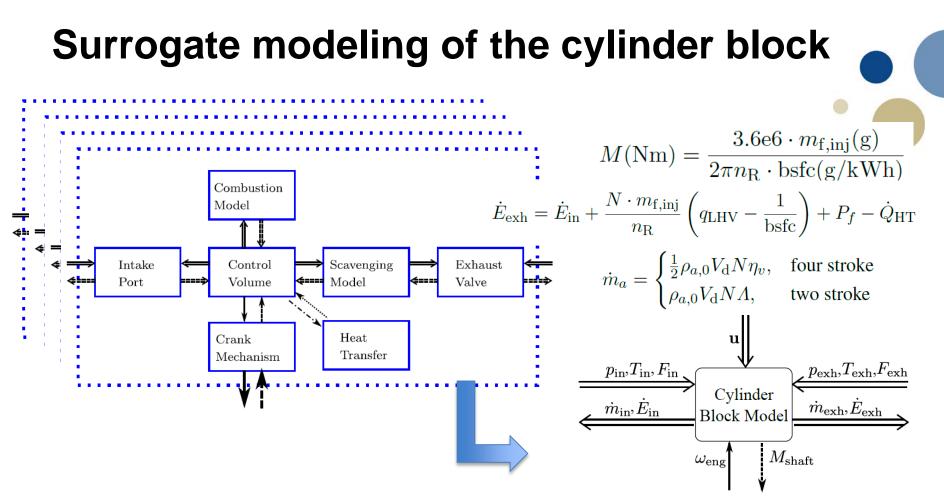
• ANN



$$x_{k,j} = g\left(\sum_{i=1}^{n} w_i \cdot x_{i,j-1} + b\right)$$



 $y = \sum_{i=1}^{n_{\rm SV}} \alpha_i \phi\left(\mathbf{x}, \mathbf{x'}\right) - \rho$





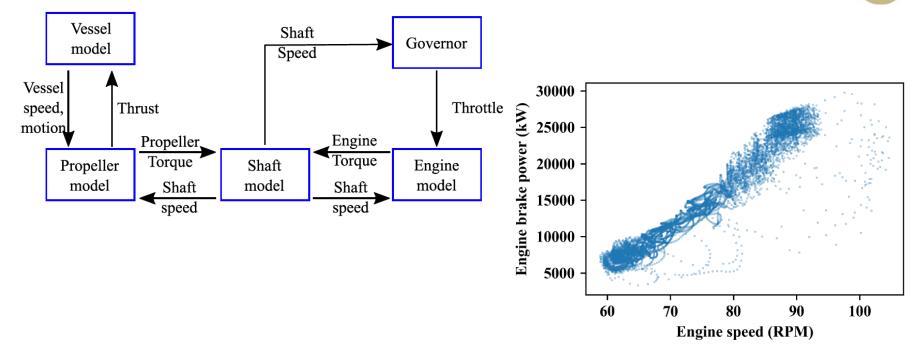
Generation of the training datasets from simulation

Training ANN and SVM models

Validation

Generation of the training datasets

• Propulsion in irregular waves



Input – Output Data Screening

Input

engine shaft speed $\omega_{
m eng}$ scavenge air temperature $T_{\rm scav}$ $T_{\rm exh}$ exhaust gas receiver temperature turbocharger speed $\omega_{
m tc}$ start of injection $\varphi_{\rm SOI}$ exhaust valve closing command $u_{\rm EVC}$ scavenge air pressure $p_{\rm scav}$ exhaust gas receiver pressure $p_{\rm exh}$ exhaust gas fuel-air equivalence ratio $F_{\rm exh}$ fuel command from the governor $u_{\rm gov}$ exhaust valve opening command $u_{\rm EVO}$

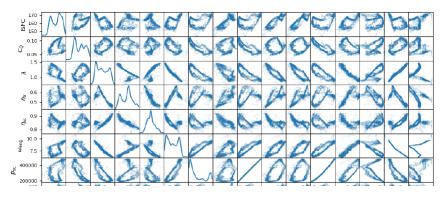
Output

ISFC Indicated specific fuel consumption

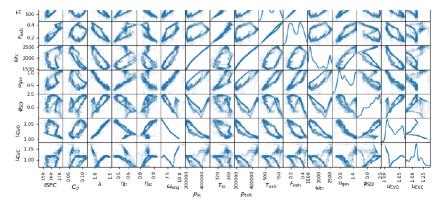
- C_Q Heat loss factor
- Λ Delivery ratio

$$SFC[g/kWh] = \frac{m_{f,cyc}}{\int_{t_{BDC,n}}^{t_{BDC,n+1}} P_{cyl}dt} \cdot 3.6e9$$
$$\Lambda = \frac{\int_{t_{IPO}}^{t_{IPC}} \dot{m}_{in}dt}{\rho_{scav} \cdot V_d}$$
$$C_Q = \frac{\int_{t_{BDC,n}}^{t_{BDC,n+1}} \dot{Q}dt}{m_{f,cyc} \cdot q_{LHV}}$$

Input – Output Data Screening



 $[\text{ISFC}, C_Q, \Lambda]^T = \mathbf{\Phi}\left(\omega_{\text{eng}}, p_{\text{scav}}, T_{\text{exh}}, u_{\text{gov}}, \varphi_{\text{SOI}}, u_{\text{EVC}}\right)$



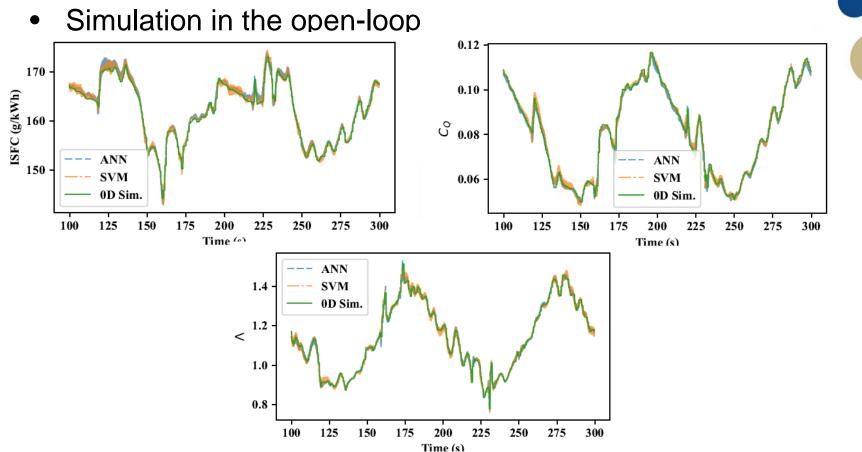


Training results

• Cross validation results (25%)

		ISFC (g/kWh)	C_Q	Λ	$\eta_{ m tr}$	$\eta_{ m sc}$
ANN	MAPE SAPE			0.61 0.74		
SVM	MAPE SAPE			0.58 1.38		

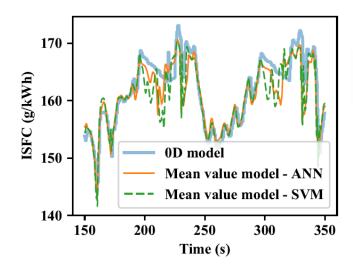
Verification

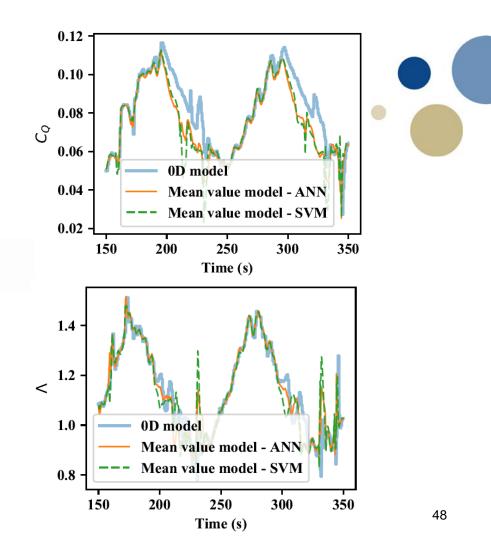


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Verification

• Simulation in the closed-loop



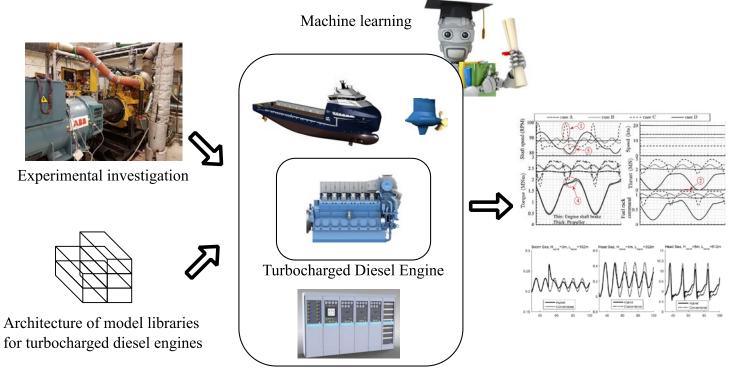


Findings

- ANN and SVM models showed good performance for prediction of the key parameters for mean value cylinder process model
- Generating training data by the total system simulator should have both high variance and correlation to the physical laws. Then, it becomes an effective method to avoid the curse of dimensionality
- Model reduction comes with unavoidable loss of fidelity. The effect of such loss should be evalueated on the system level.

Conclusion

• Main contribution of the thesis



Total System Simulator

Conclusion

- Further works
 - Development of transient NOx model for real-time simulation and validation
 - Development of gas engine model
 - Validation of the total system simulator by a full-scale test
 - Experimental investigation of transient effect under cyclic load for different type of engines

Thank you for your attention!