

A Nonlinear Model Predictive Controller for Reactivity Controlled Compression Ignition in Marine Engines

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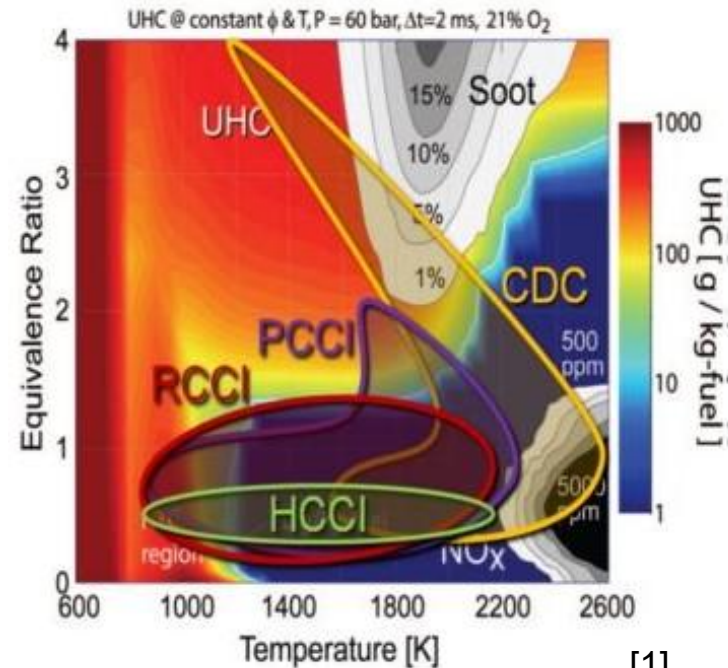
Background: Marine propulsion and RCCI

- Marine transport is essential to global trade, but it is also a significant source of global emissions.
- While new propulsion technologies (e.g., electric and hydrogen) are emerging, combustion engines are expected to remain the dominant prime movers for the foreseeable future.
- Among advanced combustion concepts, Reactivity Controlled Compression Ignition (**RCCI**) is emerging as a promising pathway for reducing emissions.



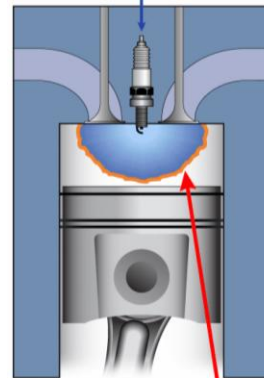
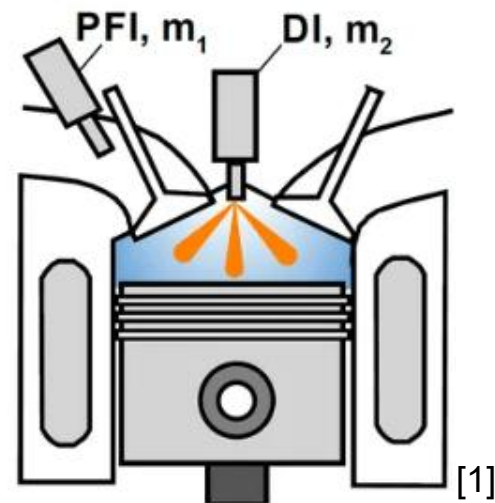
Background: Why RCCI?

- **Low-temperature Dual-Fuel** combustion concept
- Provides **low emissions** and **high thermal efficiency** when compared to conventional combustion strategies.
- Improved combustion control in comparison to other **LTC** strategies, such as **HCCI**, with similar benefits of **low NO_x** and **soot** emissions

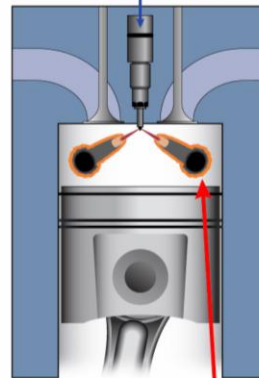


Background: How RCCI works?

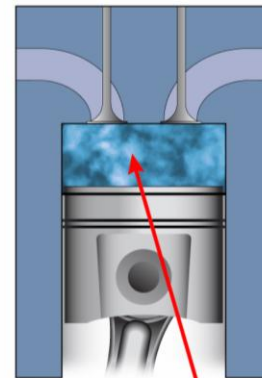
- Low-reactivity fuel (e.g. **methane**) via port-fuel injection (**PFI**)
- High-reactivity fuel (e.g. **diesel**) via early direct injection (**DI**)
- The fuel mixture is compressed and auto-ignited, resulting in spread-out low temperature combustion
- **Combustion timing** is controlled by adjusting **fuel reactivity**, i.e. the blend ratio of the two fuels



Hot-Flame Region:
NOx



Hot-Flame Region:
NOx & Soot



Low-Temperature Combustion:
Ultra-Low Emissions (<1900K)

Background: Previous control approaches

- **RCCI** has multi-input-output behavior, constraints, and nonlinear dynamics
 - -> Model predictive control (**MPC**) emerged as the leading strategy.
- Most RCCI control approaches rely on **data-based**, or linear, such as **LPV** models
- High nonlinearity and multiple dimensions require large and detailed maps.
 - -> High calibration and storage requirements for the model
- Fuel flexibility increases problem even further by introducing additional complexity

Proposed solution: Nonlinear prediction model

- Instead of a data-based or linearized model, a **nonlinear** combustion model could be used for RCCI combustion modeling.
- Nonlinear Model Predictive Control (**NMPC**) using a nonlinear combustion model enables a substantial reduction in model size.
 - -> A single model works in the whole engine operating range

Methodology: Data

Wärtsilä
SCE31

Experimental
data (steady)

Thermokinetic
multizone model
UVATZ [1]

Identification
data (steady
& transient)

Nonlinear RCCI
control oriented
model
COM

Table 1. Specification of the Wärtsilä SCE31 used for the UVATZ model validation.

Displacement	32.45 l
Nominal speed	720 rpm
Stroke/bore	1.39
Air system	External air compressor with air temperature & pressure control (up to 10 bar)
Valvetrain	4 valves with swirl + tumble ports variable intake valve closure (VIC) fixed exhaust valve opening (EVO)
Engine control	Rapid prototyping platform
Test fuels	ISO 8217 compliant LFO / LNG (MN=80)

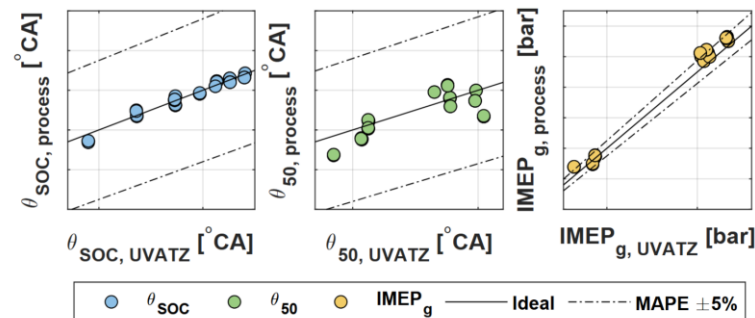
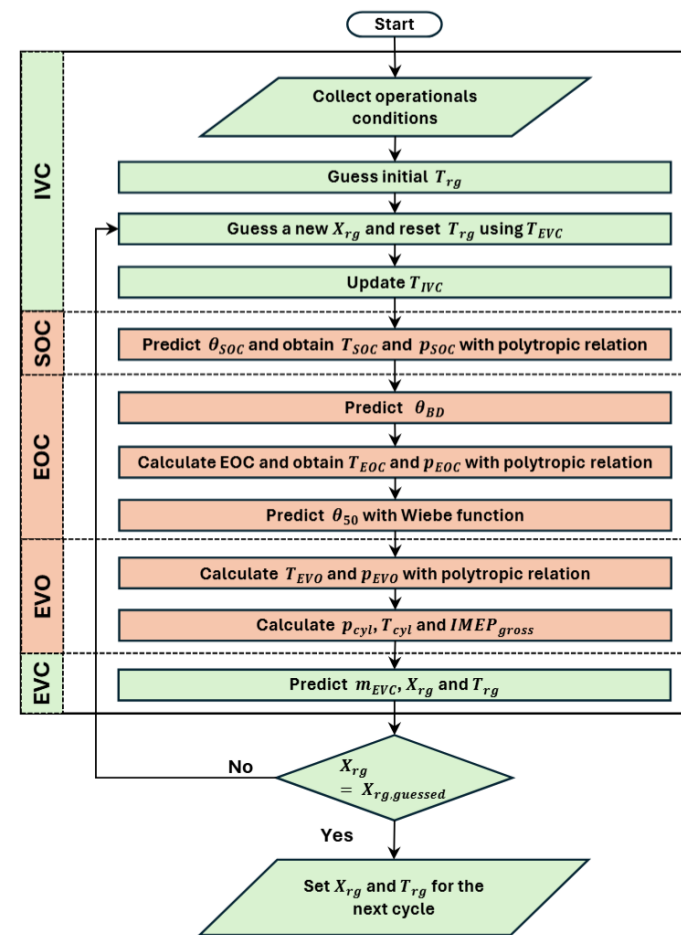


Fig. 2. Overall estimation accuracy of the nonlinear dynamic process model.

Methodology: Model

- **Process model** used in this work is a physics-based, nonlinear model which computes an **RCCI** operating cycle from intake to exhaust stroke. (**GREEN** + **RED**)
- To experiment using a static model to control the RCCI, only the closed cycle was used as **prediction model**. (**RED**)
- In control, the unmodeled cycle-to-cycle dynamics are accounted via feedback



The model is a combination of previous works in literature [1],[2].

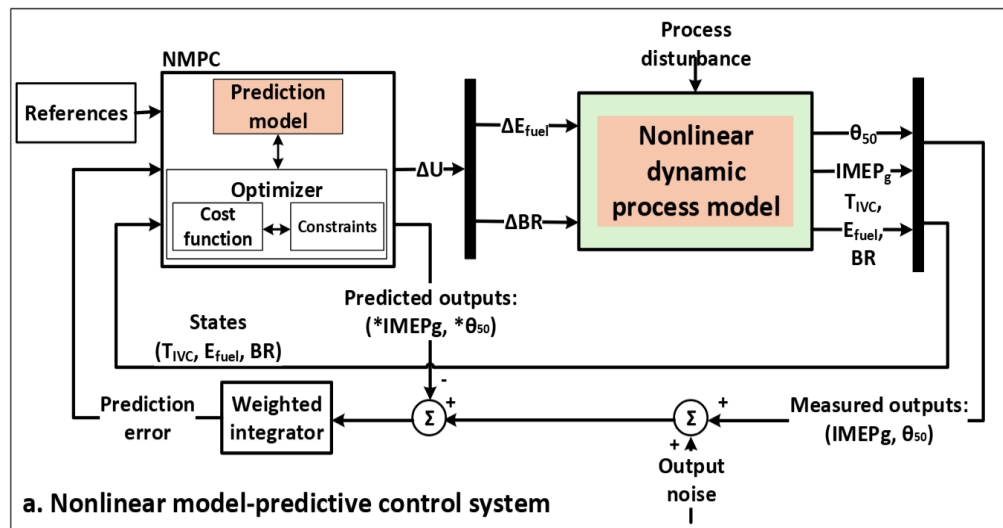
Controller setup: NMPC

- Prediction model is defined in equations 6 and 7.
 - $h()$ is the nonlinear combustion model
 - E is a correction term, integrated from prediction error
- Control- and prediction horizons set to two**
- SQP solver**, with two solver iterations

$$X_{k+1} = f(\cdot) = X_k + \Delta U_k \quad (6)$$

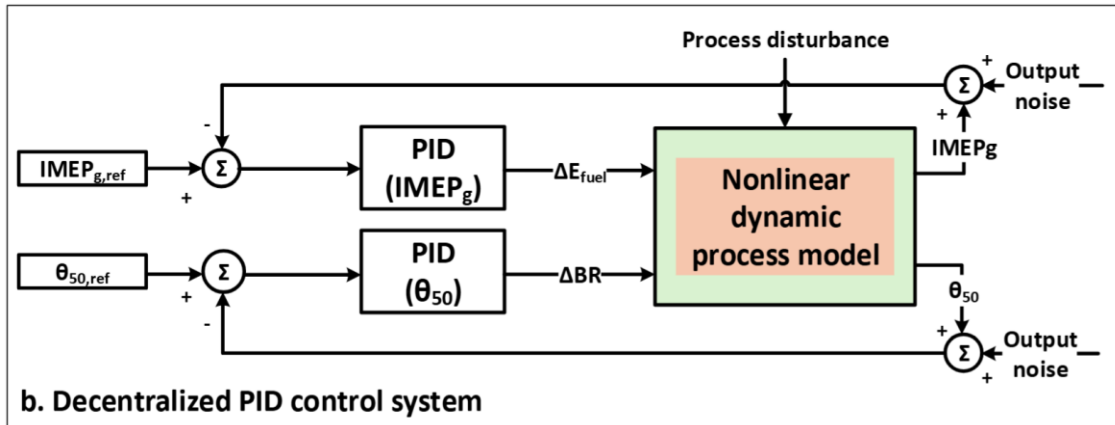
$$Y_k = g(\cdot) = h(X_k) + E_k \quad (7)$$

$$E_k = E_{k-1} + K_I(Y_k^* - Y_{k-1}) \quad (8)$$



Controller setup: PID

- **NMPC** controller is the one being developed, where as **PID** provides a simple comparison for evaluating the NMPC performance
- **IMEP** is controlled by **E_{fuel}**
- **CA50** is controlled by **BR**



Test setup

- Implemented and tested in **Matlab Simulink** simulation environment
- In all cases same engine transient from **mid-** to **high-** load
- Different disturbances and control constraints
 - Chosen to represent real engine behavior

Test nro. ^a	Setpoint change		Disturbances											
			θ_{50}		IMEP _g		T_{intake}		m_{EGR}		θ_{50}		IMEP _g	
			[°CA]	[bar]	[K]		[kg]		[°CA]		[bar]			
			μ	σ^2	μ	σ^2	μ	σ^2	μ	σ^2	μ	σ^2	μ	σ^2
I	-2	+9	0	1	0	10 ⁻⁴	-	-	-	-	-	-	-	-
II	-2	+9	-	-	-	-	0	0.5	0	0.2	-	-	-	-
III	-2	+9	-	-	-	-	-	-	-	-	-	-	-	-

^a I: Transient with process disturbance, II: Transient with output disturbance, III: transient with strict constraints.

Results: Test I & II

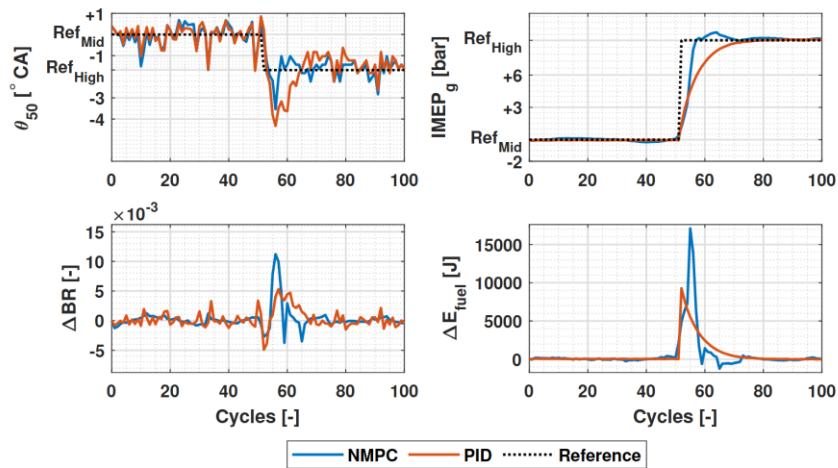


Fig. 4. Test I, load transient with process disturbance.

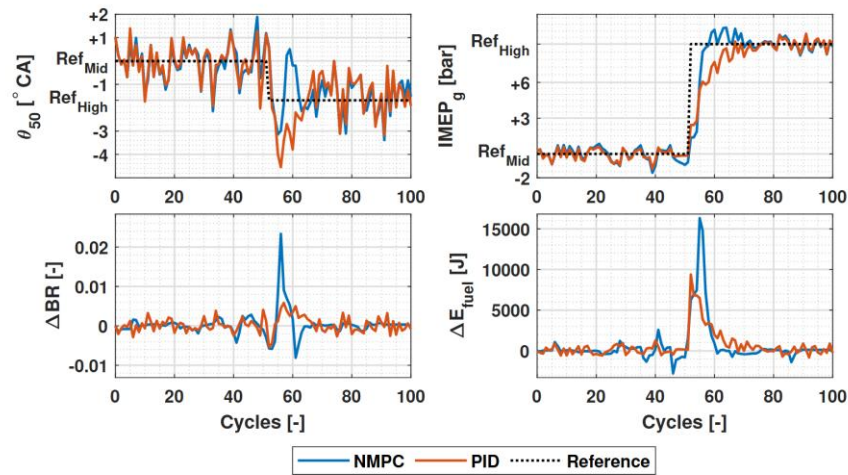


Fig. 5. Test II, load transient with output disturbance.

- NMPC has faster response and settling time, with less noise in CA50 output
- NMPC has smoother control signals

Results: Test III

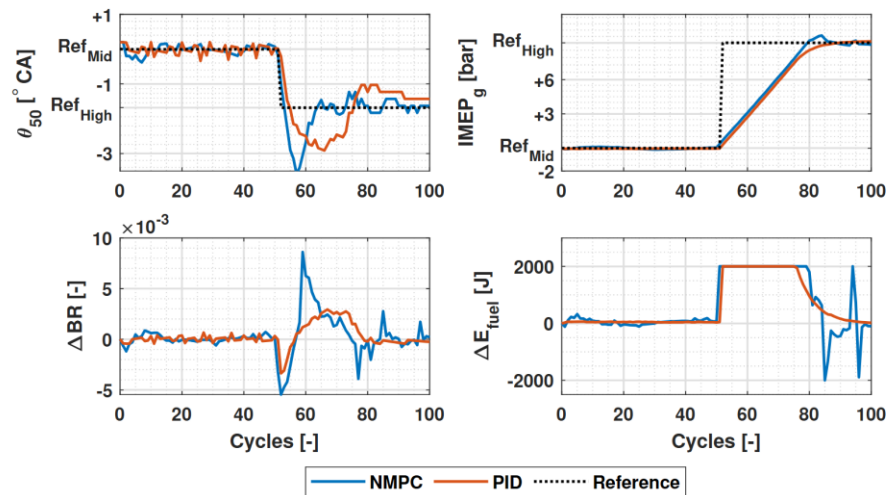


Fig. 6. Test III, load transient with strict constraints.

- NMPC again has faster settling time, with more aggressive control and overshoot

Results: Summary

- The developed **NMPC** achieved reasonable performance
- The controller achieved computation speed of below 1 second per cycle
- The engine operating point remained confined, so nonlinear behavior not so significant.
 - > **PID** performed OK in comparison

Table 4. Performance metrics averaged over all test cases.

	NMPC	PID
$MAE_{\theta_{50}}$ [$^{\circ}$ CA]	0.42	0.48
MAE_{IMEP_g} [bar]	0.80	0.89
$t_{f\theta_{50}}$ [cyc]	1	1
t_{rIMEP_g} [cyc]	10.67	16
$t_s(2\%)_{\theta_{50}}$ [cyc]	9.33	11.67
$t_s(2\%)_{IMEP_g}$ [cyc]	14.67	24.67
Undershoot $_{\theta_{50}}$ [$^{\circ}$ CA]	1.82	2.23
Overshoot $_{IMEP_g}$ [bar]	0.91	0
Final $\Delta E_{fuel}(\%)$	53.41	53.67
Sim. time [s]	66.20	4.06

Conclusions: What's next?

- Introducing more constraints to the controller, such as **combustion stability**, **emissions** and **maximum pressure**
- Improve the computation speed of the **NMPC**
- Preparing the system for **hardware-in-the-loop** and **engine-in-the-loop tests**

Thank you!

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