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Real-Time Optimal Energy Management of Electrified Engines

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Abstract: The electrification of engine components offers significant opportunities for fuel economy improvements, including the use of an electrified turbocharger for engine downsizing and exhaust gas energy recovery. By installing an electrical device on the turbocharger, the excess energy in the air system can be captured, stored, and re-used. This new configuration requires a new control structure to manage the air path dynamics. The selection of optimal setpoints for each operating point is crucial for achieving the full fuel economy benefits. In this paper, a control-oriented model for an electrified turbocharged diesel engine is analysed. Based on this model, a structured approach for selecting control variables is proposed. A model-based multi-input multi-output decoupling controller is designed as the low level controller to track the desired values and to manage internal coupling. An equivalent consumption minimization strategy is employed as the supervisory level controller for real-time energy management. The supervisory level controller and low level controller work together in a cascade which addresses both fuel economy optimization and battery state-of-charge maintenance. The proposed control strategy has been successfully validated on a detailed physical simulation model.

Keywords: electrified engines, fuel economy optimization, real-time energy management, decoupling control, dynamics analysis

1. INTRODUCTION

Motivated by the increasing fuel costs and environmental pressures, countries around the world are passing increasingly stricter legislations on the fuel efficiency of internal combustion engines. Heavy duty vehicles (HDVs) contribute almost a quarter of the fuel consumption in transportation, and most are equipped with diesel engines due to their higher thermal efficiency and good durability. With state-of-the-art technologies, an estimated improvement in fuel economy of 35% - 50% of HDVs is considered realistic, with most of the potential concentrated on engine efficiency improvement and engine hybridization Boyd and Vandenberghe (2010). The electrified turbocharger is a critical technology in engine hybridization. It combines a variable geometry turbocharger (VGT) and an electric machine (EM) within a single housing. The EM is capable in bi-directional power transfer, so excess power can be recuperated by the EM to supply electrical accessories or to be stored in a battery for later usage. On the other hand, the EM accelerates the turbocharger to improve engine response, especially for transient torque demands. Due to its location in the air system, reasonably small electric systems can have a large effect on transient behavior and engine efficiency. In this sense the technology is superior to conventional hybrid systems that act on the engine crankshaft.

The electrified turbocharger has attracted considerable development interest. The mainstream diesel engine and turbocharger manufacturers have developed their own prototype electrified turbochargers, see Bailey (2000), Arnold et al. (2005), and Ibaraki et al. (2006). Investigations into the efficiency characterization of an electrified turbocharger through experiments with a heavy duty diesel engine have been made by Terdich and Martinez-Botas (2013) and Terdich et al. (2014). There is a broad agreement that the development of a systematic strategy in both real-time energy management and multi-input multi-output (MIMO) control is essential for exploring the maximum benefits of the electrified turbocharger.

A suitable control and energy management structure for the electrified turbocharged diesel engine (ETDE) is still to be established. In Ibaraki et al. (2006), the EM is controlled in an open loop manner. In Glenn et al. (2010), the EM, VGT, and exhaust gas recirculation (EGR) valve are controlled independently without considering the internal couplings in the engine or the generation setpoints. In 2014, a consortium led by Caterpillar Inc. developed a cutting-edge electrified turbocharger called electric turbo assist (ETA). Based on the ETA, several control methods and testing systems were developed in Loughborough University in simulations Zhao et al. (2015); Zhao and Stobart (2015); Zhao et al. (2016) and experiments Zhao et al. (2014, 2013); Winward et al. (2016); Yang et al. (2016).
This paper presents an attempt to address this gap. The main contributions are:

(1) An explicit principle for the selection of control variables of an ETDE is proposed based on the dynamics analysis. From this, clear guideline for the generation of setpoints follow.

(2) A two-level structure of energy flow management and air path regulation is presented. On the supervisory level, the optimal setpoints of control variables are computed to distribute energy flows in the optimal way. On the low level, a MIMO controller is designed to implement the optimal energy flow distribution.

(3) An equivalent consumption minimization strategy (ECMS) algorithm is designed as the supervisory level controller to guarantee the real-time optimization of fuel economy while keeping the sustainable battery usage.

(4) A model-based non-smooth robust controller is designed as the low level controller to regulate the air path dynamics and to address the internal couplings among actuators.

(5) The proposed control strategy is successfully applied on a physical engine model for validation.

The paper is organized as follows. Following the introduction in section 1, the electrified turbocharger model is described in section 2. The control problem is formulated in section 3, and the control-oriented dynamic system is analyzed in section 4. The supervisory level controller design is presented in section 5, and the low level controller in section 6. Validation results of the strategy are demonstrated in section 7, followed by conclusions in section 8.

2. ELECTRIFIED TURBOCHARGER MODEL

The structure of an ETDE is illustrated in Fig. 1, while its variables and related parameters are defined in Table 1. A switched reluctance motor (SRM) is selected as the EM. It is an excellent option in extra-high speed applications thanks to its simple structure, see Bilgin et al. (2015). The EM can work in both assisting (motoring) mode and harvesting (generating) mode. In assisting mode, the EM extracts energy from the battery to improve the engine transient response, or provide steady state boost pressure to enhance low speed torque. In harvesting mode, the additional turbine torque resulting from excess exhaust energy causes a power flow to the generator.

The dynamics of the turbocharger can be modeled as a first-order lag power transfer function with time constant \( \tau \):

\[
\dot{P}_c = \frac{1}{\tau} \left( P_t + P_{em} - P_{bl} - P_{wl} - P_c \right),
\]

where \( P_0 \) and \( P_{wl} \) are the power to overcome bearing losses and windage losses, respectively. The mechanical efficiency \( \eta_m \) is introduced to quantify the energy losses. Therefore, (1) can be represented as

\[
\dot{P}_c = \frac{1}{\tau} \left( \eta_m (P_t + P_{em}) - P_c \right).
\]

\( W_c \) is related to \( P_c \) by

\[
W_c = \frac{\eta_c}{c_p T_0} \left( \frac{P_c}{p_{am}} \right)^\gamma - 1,
\]

and \( P_t \) can be expressed by \( W_t \):

\[
P_t = \eta_c c_p T_{exh} \left( 1 - \left( \frac{p_{am}}{p_{exh}} \right)^\gamma \right) W_t.
\]

3. PROBLEM FORMULATION

The development of the real-time energy management strategy is separated into an off-line stage and an online stage. In the off-line stage, the control variables are to be selected, which use the additional degree of freedom introduced by the EM actuator for best effect. In addition, the boundaries of selected control variables are to be chosen so as to guarantee that all the setpoints
3.1 Energy Flow Management

The primary task of energy flow management is to formulate the cost function based on fuel efficiency. The total fuel power is

$$P_t = H_{HV} W_f,$$

where $H_{HV}$ is the lower heating value of the fuel. For energy is generated from both engine cylinders and battery, the instantaneous equivalent consumed power of an ETDE is given by

$$P_{eq} = P_e + s(SOC) P_b,$$

where $P_e$ is the battery power, and $s(SOC)$ is a non-dimensional electricity-fuel energy equivalent factor, which applies a necessary penalty on battery state-of-charge (SOC) to discard pure electric operation.

Since $P_{eq}$ is an instantaneous variable, it can be selected as the cost function

$$J(y^*) = P_{eq},$$

where $y^*$ denotes the setpoints for controlled variables. The on-line optimization problem is formulated as

$$\min : J(y^*),$$

s.t.: $y_{min} \leq y^* \leq y_{max}$,

where the limits on $y^*$ denote the permissible range of the setpoints set.

3.2 Air Path Regulation

The objective of air path regulation is the tracking of the computed setpoints $y^*$. Because the air path is highly nonlinear, it is challenging to build an effective global control strategy that covers all possible operating regions. Developing piecewise-linear controllers at different operating points is a feasible approach, where the operating points are determined by the independent variables of $N$ and $T_i$. At a specified operating point, the linear model of the ETDE air path can be expressed in the form of discrete state space equations:

$$\begin{align*}
    \left\{ \begin{array}{l}
    x(k+1) = Ax(k) + Bu(k) \\
    y(k) = Cx(k)
    \end{array} \right.,
\end{align*}$$

where $x \in \mathbb{R}^n$ is the state vector, $u \in \mathbb{R}^m$ is the input vector, $y \in \mathbb{R}^p$ is the output vector, $(A, B)$ is a controllable pair, and the coefficient matrices $A, B,$ and $C$ are obtained via system identification. The entire engine operating region is segregated into several subzones and the identified model at the operating point is effective in the located subzone. The number of subzones depends on the required precision of the identified model.

4. CONTROL-ORIENTED DYNAMICS ANALYSIS

The selection of control variables and the boundaries of their setpoints will be underpinned by the dynamics analysis of an ETDE.

4.1 Selection of Control Variables

The key to reducing exhaust emissions in terms of nitrogen oxides (NOx) and particulate matter (PM) is keeping rational values of $F_1$ and $\lambda$, which can be obtained from the independent variables $W_e$ and $W_{egr}$ according to their expressions in Table 1. The independent variables can also be replaced by $P_{in}$ and $W_{egr}$. However, to achieve optimal fuel efficiency and exhaust emissions in an ETDE, controlling only $P_{in}$ and $W_{egr}$ is not enough. Since the EM impacts the exhaust manifold dynamics directly, a further performance variable in the exhaust manifold has to be considered. In the available literature, no explicit principle for selecting control variables of an ETDE and for the analysis of its influence on fuel economy could be found.

A proposed criterion for selecting control variables is that they must not be strongly coupled. This requirement is necessary for independent control of the selected variables. To analyze the relationship between different control variables, a calibration is implemented on a physical diesel engine model at 1800 rpm, 600 Nm by tuning $\chi_{vgt}$ and $P_{em}$. The results with no EGR are illustrated in Fig. 3. The setting on $\chi_{vgt}$ and $P_{em}$ is shown as Fig. 3(a), where $P_{em}$ changes from the maximum generating power to the maximum motoring power, with the value of -5 kW and 5 kW, respectively. The $\chi_{vgt}$ changes from 0.1 to 0.9, where 0 implies the VGT vane is fully closed and offers the least flow area to the exhaust flow. Fig. 3(b) shows the strong coupling between boost pressure and turbine speed. The turbine speed behaves as a 2nd-order polynomial function with the boost pressure, which shows the strong coupling between them. On the other hand, the exhaust pressure varies much according to different settings on $\chi_{vgt}$, as shown in Fig. 3(c). With the VGT vane closing from 0.9 to 0.1, both boost pressure and exhaust pressure increase, as well as the condition when $P_{em}$ increases with the fixed $\chi_{vgt}$.

The coupling between boost pressure and turbine speed can also be found using a theoretical causality analysis. Supposing a 2 inputs - 2 outputs control structure, where $P_{in}$ and $\omega$ are controlled by $\chi_{vgt}$ and $P_{em}$, respectively. If the desired turbine speed $\omega^*$ increases, then $P_{em}$ would increase accordingly to drive up $\omega$. At the same time, $P_{in}$ would be driven up because of the enhanced boost assist. To track a steady value of $P_{in}$, then $\chi_{vgt}$ would open further to reduce $P_{in}$. As a result, $\omega$ is decreased to give a lower pressure ratio in the exhaust manifold. Therefore, $\omega$ has a strong coupling with $P_{in}$ and its setpoint cannot
be selected independently. Based on both test results and dynamics analysis, the control variables are selected as
\[ y = [p_{in}, p_{exh}, W_{egr}]^T, \tag{10} \]
while the control inputs of
\[ u = [\chi_{vgt}, P_{em}, \chi_{egr}]^T. \tag{11} \]

4.2 Setting of Boundaries for Setpoints

In the online implementation of energy flow management, \( y^* \) can vary in a range to find the optimal value for best fuel economy. Prior to the online implementation, an offline calibration on the boundaries of \( y^* \) is required to guarantee \( y^* \) are reachable.

The setpoints of the control variables are denoted as \( p_{in}^*, p_{exh}^*, \) and \( W_{egr}^* \), respectively. The values of \( p_{in} \) and \( W_{egr} \) can be obtained from the engine calibration settings within the engine control unit (ECU). Therefore, only \( p_{exh}^* \) is to be determined. A calibration-based approach is implemented to get the boundary values of \( p_{exh}^* \), which consists of two steps.

In the first step, a calibration is fulfilled at a specified operating point, followed by the development of a MIMO controller based on the calibration data. In the second step, \( p_{exh}^* \) is set as an incrementally increasing series. The developed MIMO controller is applied on the ETDE model to track \( p_{exh}^* \). The value of \( p_{exh}^* \) is feasible if it can be tracked well. If the controller fails to track the given \( p_{exh}^* \) value in some cases, it implies the \( p_{exh}^* \) value is beyond its boundary.

A calibration of the boundary of \( p_{exh}^* \) has been made at 1800 rpm, 700 Nm, while \( p_{exh}^* \) increases from 265 kPa to 289 kPa step-wise, with steps of 2 kPa. Each \( p_{exh}^* \) value continues for 15 s. Up to 283 kPa, \( p_{exh}^* \) is tracked well. However, strong oscillations happen on all three variables when \( p_{exh}^* \) exceeds the threshold of 283 kPa after 165 s, which means the value of \( p_{exh}^* \) in this period is hard to reach. The spikes on \( p_{in}^* \) and \( W_{egr}^* \) are due to the ECU generated setpoints being unfiltered. At the very beginning, though 265 kPa can be tracked well, the control inputs especially the EM power are not fully stabilized, which also indicates the \( p_{exh}^* \) value is very close to its lower limit. Therefore, the available range of \( p_{exh}^* \) is selected as [265 kPa, 283 kPa] at 1800 rpm, 700 Nm. Similarly, the boundaries \( p_{exh}^* \) at other operating points are also determined in the same way. At the other untested operating points, the boundaries are obtained by means of gain scheduling.

5. SUPERVISORY LEVEL CONTROLLER DESIGN

The supervisory level controller is designed in the form of ECMS, to minimize fuel consumption while maintaining the battery SOC value. The ECMS realizes the optimization problem by minimizing an instantaneous cost function, so it behaves as a closed-loop controller. The energy equivalent factor in (6) is designed as
\[ s = K_P \Delta SOC, \tag{12} \]
where \( K_P \) is a positive constant, and
\[ \Delta SOC = SOC - SOC^*, \tag{13} \]
while \( SOC^* \) is the desired value of battery SOC.

The optimal \( p_{exh}^* \) is obtained from the lookup table according to the minimal \( p_{egr} \), where the lookup tables are generated from the calibration. The fuel power and EM power with different exhaust pressure settings at 1800 rpm, 700 Nm are listed in Table 2. Without considering electrical energy losses, it is assumed that \( P_{em} \) equals to
Table 2. Lookup table of $P_a$ and $P_{em}$ with respect to $P_{exh}^*$

<table>
<thead>
<tr>
<th>$P_{exh}$ (kPa)</th>
<th>$P_a$ (kW)</th>
<th>$P_{em}$ (kW)</th>
</tr>
</thead>
<tbody>
<tr>
<td>265</td>
<td>241.09</td>
<td>1.67</td>
</tr>
<tr>
<td>267</td>
<td>241.62</td>
<td>1.61</td>
</tr>
<tr>
<td>269</td>
<td>242.05</td>
<td>1.46</td>
</tr>
<tr>
<td>271</td>
<td>242.48</td>
<td>1.22</td>
</tr>
<tr>
<td>273</td>
<td>242.84</td>
<td>0.96</td>
</tr>
<tr>
<td>275</td>
<td>243.16</td>
<td>0.69</td>
</tr>
<tr>
<td>277</td>
<td>243.49</td>
<td>0.38</td>
</tr>
<tr>
<td>279</td>
<td>243.76</td>
<td>0.06</td>
</tr>
<tr>
<td>281</td>
<td>244.09</td>
<td>0.13</td>
</tr>
<tr>
<td>283</td>
<td>244.37</td>
<td>-0.255</td>
</tr>
</tbody>
</table>

Fig. 4. The cost function with respect to $P_{exh}^*$

Fig. 5. The diagram of supervisory level controller

$P_b$. Based on the data in Table 2, $P_{eq}$ varies with $P_{exh}^*$ and $s$, as shown in Fig. 4. When $s$ has different values, the minimal $P_{eq}$ is represented by different $P_{exh}^*$ value.

The diagram of the supervisory level controller is illustrated as Fig. 5, where $Q_b$ and $Q_{max}$ denote the battery stored energy and battery capacity, respectively. In Fig. 5, $P_{exh}^*$ is generated according to the minimized $P_{eq}$, whose value is computed by the feedback variables $P_a$ and $P_b$, and the updated cost factor $s$. The generated $P_{exh}^*$ is sent to the cascaded low level controller, together with $P_{in}$ and $W_{egr}$ from the ECU, as the setpoints to be tracked.

Fig. 6. Control structure of the ETDE

6. LOW LEVEL CONTROLLER DESIGN

The single-input single-output (SISO) PI controllers are widely used in the control of conventional turbocharged diesel engines, while $P_{in}$ and $W_{egr}$ are controlled by $\chi_{egr}$ and $\chi_{vgt}$, respectively. The gain values of PI controllers are generally tuned in off-line calibration. However, in the ETDE, the dynamics has been significantly changed, therefore the controllers require to be re-designed. Since it is time-consuming to re-tune the gain values individually, designing a decoupling MIMO controller is critical. The 3-inputs 3-outputs control structure of the ETDE is illustrated as Fig. 6.

The ETDE model is formed by the state space equations:

$$\begin{bmatrix} \dot{x} \\ y \\ z \end{bmatrix} = \begin{bmatrix} A & B_1 & B_2 \\ C_1 & D_{11} & D_{12} \\ C_2 & D_{21} & D_{22} \end{bmatrix} \begin{bmatrix} x \\ w \\ u \end{bmatrix}, \quad (14)$$

where $x$, $u$, and $y$ have been defined in (9). Moreover, $w \in \mathbb{R}^{m_1}$ is the vector of exogenous inputs, $z \in \mathbb{R}^{p_1}$ is the vector of regulated outputs.

The dynamic output feedback control law $u = K(s)y$ is to be designed. A nonsmooth $H_\infty$ synthesis method is employed to build $u$ since it is very convenient in practice, see Apkarian and Noll (2006). This method can address a mixed set of time- and frequency-domain criteria including settling time, steady state error, stability margin, and noise rejection. The constraints can be grouped as hard (must have) constraints and soft (nice to have) objectives. The controller synthesis is transformed to an optimization problem as

$$\text{minimize : } \max_i \{ ||T_{w_i \rightarrow z_i}(P, K)|| \}, \quad (15)$$

subject to : $\max_j \{ ||T_{w_j \rightarrow z_j}(P, K)|| \} \leq 1, \quad (16)$

where $g$ is the vector of tunable gain values, $||T_{w \rightarrow z}||$ denotes the closed loop map from input $w$ to output $z$, and $||\cdot||$ can be either $H_\infty$ norm or $H_2$ norm, $i$ and $j$ denote the index of soft objectives and hard constraints, respectively. The optimization problem (15) and (16) means the minimization of the worst-case value of the soft objectives while satisfying all the hard constraints. In the formulation, all terms have been normalized, see Apkarian (2013).

In this research, the soft objectives are settling time and steady state error. The hard constraints are stability margins including the gain margin and phase margin. The coefficient matrices in (14) are obtained via calibration tests. The systune solver in Matlab is adopted to compute the decoupling control law.

7. SIMULATION RESULTS

Simulations have been carried out on a physical plant model built in Dynasty, a proprietary multi physics
simulation software package used within Caterpillar. The controller and ETDE model are built in Simulink and Dynasty, respectively. The engine manifolds are modeled as one-dimensional components, permitting to capture the pulsations caused by engine actuators operation. All the cylinders are modeled separately to indicate the energy transfer from engine to turbocharger, such that the transient performance is simulated accurately.

The compressor and turbine are represented as map based models. The ETA is also represented as a map-based model, whose maximum motoring/generating power are found from a two-dimensional map, with the inputs of \( \omega \) and power electronics controller setting. The controllers are built in Matlab/Simulink for the available tools in Matlab can facilitate the controller design. To guarantee steady communications, the sampling frequency of the controller in Matlab and the controlled plant in Dynasty are set as identical. The investigated engine is a CAT six-cylinder, 7.01-Litre heavy duty engine with rated power of 225 kW at 2200 rpm and a rated torque of 1280 Nm at 1400 rpm. The engine has been fitted with an experimental turbo-charger for the purposes of this work. The maps used in the paper are all generated from off-line experimental calibration on the instrumented engine. The battery capacity is set as 0.1 kWh considering the balance between usage flexibility and cost.

The operating points are selected as 1800 rpm, 260 Nm, and 1800 rpm, 700 Nm, covering the low load region and high load region, which are typical working conditions in the trenching test cycle for heavy duty engines. A block load condition between the switching of the two operating points is also given, while the gain values are scheduled according to the load.

### 7.1 Simulation Results of the Low Level Controller

In generating the decoupling control law, the criteria are to be specified. In the soft objectives, the criteria on settling time and steady state error are set as 1s and 0.5%, respectively. In the hard constraints, the criteria on gain margin and phase margin are set as 3dB and 5 rad/s, respectively. The controller is tested under block loads between 1800 rpm, 700 Nm, and 1800 rpm, 260 Nm. The transient period is 1s. The operating conditions are demonstrated in Fig. 7, where the block load condition repeats 4 cycles. In each cycle, the setpoint of \( p_{exh} \) at 1800 rpm, 700 Nm is set as 265 kPa, 270 kPa, 275 kPa, and 280 kPa, respectively. Meanwhile, the setpoint of \( p_{exh} \) at 1800 rpm, 260 Nm is kept as 215 kPa in each cycle. The setpoints of \( p_{in} \) and \( W_{egr} \) are generated from ECU. The tracking of the control variables in transients is very fast with small spikes, as can be observed in Fig. 8. The decoupling controller only needs to schedule gain values rather than the system model when operating point changes, which reduces the risk of system instability.

### 7.2 Simulation Results of the Two-Level Controllers

The performance of the two-level controllers in energy management is validated under block loads, while the operating condition is shown as Fig. 9. When the initial values of the battery SOC are set as 20% and 80%,...
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