Dynamic Control of a SI Engine with Variable Intake Valve Timing

Jun-Mo Kang and J. W. Grizzle

Abstract

Engines equipped with a means to actuate air flow at the intake valve can achieve superior fuel economy performance in steady state. This paper shows how modern nonlinear design techniques can be used to control such an engine over a wide range of dynamic conditions. The problem is challenging due to the nonlinearities and delays inherent in the engine model, and the constraint on the air flow actuator. The controller is designed on the basis of a mean-value model, which is derived from a detailed intake stroke model. The control solution has two novel features. Firstly, a recovery scheme for integrator wind-up due to input constraints is directly integrated into the nonlinear control design. The second novel feature is that the control Lyapunov function methodology is applied to a discrete-time model. The performance of the controller is evaluated and compared with a conventionally controlled engine through simulation on the detailed engine model.

Keywords

Discrete-time nonlinear Control, Variable intake valve, Control Lyapunov function.

I. Introduction

In the design of an engine controller, one must optimize and make tradeoffs between fuel economy, drivability (torque management) and emissions. Since an automobile must meet stringent federal emissions regulations in order to be sold, emissions control often is the most important factor. The customer, however, will consider fuel economy and torque response in making a selection.

The three way catalytic converter is the current technology for meeting emissions regulations. When operated near the stoichiometric point, emission conversion efficiencies of 98 % for hydrocarbons, carbon monoxide and oxides of nitrogen can be achieved. However, as seen in Figure 1, deviations of ± 0.2 air-fuel ratio (A/F) will cause the conversion efficiency of at least one of the emission components to drastically decrease. Thus an important control objective is to maintain the air-fuel ratio near stoichiometry.

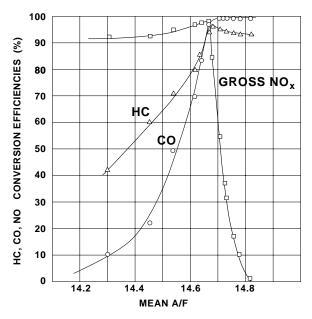


Fig. 1. Steady state conversion efficiency of TWC.

In a standard spark ignition engine, the primary actuator is the fuel injector, which is typically located at the intake port. The mass flow rate of air entering the intake manifold is measured with a hot wire anemometer, and the fuel injected into the engine is adjusted to achieve a stoichiometric mixture; this is clearly a feedforward control action. In

Jun-Mo Kang is a Ph.D. candidate in Electrical Engineering and Computer Science Department, University of Michigan, Email: junmo@eecs.umich.edu.

J. W. Grizzle is with the Control Systems Laboratory, Electrical Engineering and Computer Science Department, University of Michigan, Ann Arbor, MI 48109-2122, Tel: (734)-763-3598, FAX: (734)-763-8041, Email: grizzle@umich.edu.

order to compensate for inevitable errors in air-fuel ratio, the air-fuel ratio is measured in the exhaust stream with an exhaust gas oxygen (EGO) sensor, and a PI feedback control loop is then used to achieve zero steady state error for constant throttle position and engine speed.

Extensive research has been done to improve A/F control performance of the system. Part of this research has focused on accurate estimation of transient air flow, thereby improving the accuracy of the feedforward controller. Another possibility is to control the air flow into the intake manifold with an electronic throttle [1], [2], or the air flow into the cylinders. This latter actuation can be achieved by adjusting the cam timing of the intake valves [3], by implementing independent electro-hydraulically controlled intake valves [4], by secondary (or port) throttles [5], or by using secondary valves in series with conventional intake valves [6]. The common element of these actuators is that they allow control of the air flow into the cylinders by adjusting the effective area of the intake valves. Three of these methods, namely variable intake cam timing, variable intake valve control, and series secondary valves can also be used to improve fuel economy. This is because, by controlling the breathing process of the engine, it is possible to raise the average manifold pressure, and thereby reduce pumping losses in the engine [4], [6].

The local aspects of joint air and fuel control have been studied in [5] by designing a linear controller based around a specific operating point. The torque (drivability) and A/F responses were superior or equal to that of a conventional engine (with fuel PI control) for small step changes in the primary throttle position. The major problem encountered with the linear analysis was that the resulting closed-loop system went unstable for large changes in the primary throttle position. This can be traced to two causes: the nonlinearities in the engine model and saturation in the air flow actuator.

This paper will attempt to address these issues by developing a more global control strategy. This design has two novel features. Firstly, an integrator anti-wind-up scheme is directly integrated into the nonlinear control design. The basic idea is to inject an extra reference signal to stabilize the integrator whenever the control inputs saturate, thereby avoiding integrator wind-up. The second novel feature is the application of the control Lyapunov (clf) methodology to a discrete-time model [7].

An overview of the engine model used in this study is presented in the next section. Control objectives are summarized in Section III. The nonlinear control design is carried out in Section IV. A performance analysis via simulation is presented in Section V. The controller design will be performed on a mean-value model, whereas its performance will be evaluated via simulation on a more detailed model which captures the air flow dynamics during an intake event.

II. Engine Model

A. Breathing process model for un-actuated air flow

The intake manifold representation considered here follows [8]. It is a continuous, nonlinear, 1.6 L, 4-cylinder model. The intake breathing process of the conventional engine, based upon the ideal gas law and the conservation of mass¹, can be described by

$$\frac{dp_m}{dt} = \frac{RT_m}{V_m} \left[\dot{m}_\phi - \sum_{i=1}^4 \dot{m}_{c_i} \right] \tag{1}$$

$$\frac{dp_{c_i}}{dt} = \frac{1}{V_{c_i}} \left[RT_m \dot{m}_{c_i} - \dot{V}_{c_i} p_{c_i} \right] , \quad i = 1, \cdots, 4.$$
 (2)

where, p_m is the intake manifold pressure (Pascal) and p_{c_i} is the pressure (Pascal) in the i^{th} cylinder, \dot{m}_{ϕ} and \dot{m}_{c_i} are the mass air flow rate (Kg/s) into the manifold and that pumped out of the manifold into the i^{th} cylinder, respectively. V_m and V_{c_i} are the volume (m^3) of the intake manifold and that of the i^{th} cylinder, R is the specific gas constant (J/KgK), and T_m is the manifold temperature (K). The i^{th} cylinder volume varies as a function of crank-angle:

$$V_{c_i}(\theta) = \frac{V_d}{2} \left(1 - \cos(\theta - \frac{720}{4}(i-1)) \right) + V_{cl}$$
(3)

$$\theta = \left(\int_0^t \frac{N}{60} 360 \, d\tau \right) \bmod 720^\circ, \tag{4}$$

where V_d is the maximum cylinder displaced volume (m^3) , V_{cl} is the cylinder clearance volume (m^3) , θ is crank-angle in degrees, and N is the engine speed in RPM. The quantity p_{c_i} in (2) is periodically initialized to the exhaust pressure (110 KPa) at the i^{th} intake valve open timing (IVO). The mass air flow into the manifold, \dot{m}_{ϕ} , is approximated as a function of upstream pressure (p_o) and the downstream pressure, which is manifold pressure. Upstream pressure is

¹See Appendix for specific parameter values used in the study.

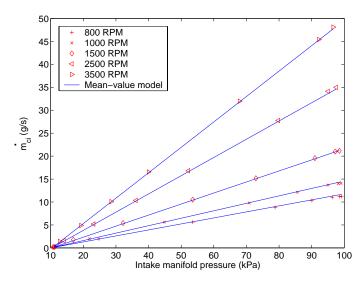


Fig. 2. Comparison between averaged mass air flow rate m_{ci} and the m_c from regression model.

assumed to be atmospheric (i.e., $p_o = 100 \text{ KPa}$):

$$\dot{m}_{\phi} = A_{\phi}(\phi)d(p_m, p_o) \tag{5}$$

$$A_{\phi}(\phi) = 1.268 \times 10^{-4} (-0.215 - 2.275\phi + 0.23\phi^{2}) \tag{6}$$

$$d(p_m, p_o) = \begin{cases} 1 & \text{if } p_m \le p_o/2\\ \frac{2}{p_o} \sqrt{p_m p_o - p_m^2} & \text{if } p_m > p_o/2 \end{cases},$$
 (7)

where $A_{\phi}(\phi)$ is the effective area (m^2) of throttle body, as a function of primary throttle angle (ϕ) in degrees. The mass air flow into the i^{th} cylinder, \dot{m}_{c_i} , is expressed as

$$\dot{m}_{c_i} = A_v(L_{v_i})d(p_{c_i}, p_m),\tag{8}$$

where A_v is the effective area (m^2) of an intake valve, which is modeled as a linear function of valve lift (mm), L_{v_i} ,

$$A_v(L_{v_i}) = \alpha L_{v_i}. (9)$$

The scale factor α is identified as 0.0175 in [8] for the experimental engine under consideration.

The valve lift motion is characterized by open timing (IVO), maximum lift (IVL), and open duration (IVD). For a conventional engine, the valve lift is a sinusoidal function of these parameters and crank-angle during an intake event [8]:

$$L_{v_i}(\theta) = IVL \cdot \sin^2\left(\frac{180}{IVD}(\theta - 90(i-1) - IVO)\right). \tag{10}$$

In this study, the valve specifications in [8] are used: $IVO = -8^{\circ}$, IVL = 8.1mm, and $IVD = 234^{\circ}$.

B. Mean-valued breathing process model

The above model describes the evolution of the various pressures and mass flow rates of the breathing process within an engine event. In general, for control design purposes, it is preferable to adopt a phenomenological, mean-valued model by averaging pressures and mass flow rates over an engine event [3], [9]. This typically results in a simpler model and one where the time scales are better adapted to those of the actuation processes. For these reasons, the mass air flow rate into the cylinder, (8), is first averaged based on simulation results, and then, via regression, represented as a function of manifold pressure (Pascal) and engine speed N (RPM) as

$$\dot{m}_c(p_m, N) = -1.7474 \times 10^{-3} + 5.6837 \times 10^{-6} p_m - 1.6529 \times 10^{-3} N + 1.5666 \times 10^{-7} N p_m. \tag{11}$$

For validation, Figure 2 compares the averaged mass air flow rate m_{ci} with constant manifold pressure and engine speed against the regression model (11).

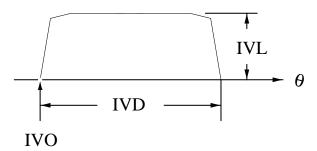


Fig. 3. Profile of hydraulically actuated valve.

C. Cylinder air charge control

To account for air flow actuation, the mass flow rate into the cylinder is modified as follows:

$$\dot{m}_a = \beta \dot{m}_c, \tag{12}$$

where β represents the normalized linear scale factor of mass air flow rate, and is limited between² 0.1 and 1. This results in the mean-value breathing process model:

$$\frac{dp_m}{dt} = \frac{RT_m}{V_m} (\dot{m}_\phi - \beta \dot{m}_c). \tag{13}$$

Depending on the air charge actuation scheme used, more or less manipulation may be necessary to put the model in this form. This will be discussed shortly. However, each scheme has the qualitative feature of being able to increase or decrease the air admitted into the cylinder, over a certain range.

On a practical basis, the choice of the particular air charge actuation scheme will be based on many factors. For example, in view of fuel economy, control of the cylinder air charge via intake valve open timing or intake valve open duration reduces pumping losses by allowing increased intake port pressure, which is essentially equivalent to intake manifold pressure [4]. On the other hand, secondary throttles choke the air flow at the intake ports, thereby decreasing the intake port pressure, which results in increased pumping losses [10]. Other issues such as reliability and cost must also be considered. From the control point of view taken in this paper, the primary difference between the various schemes lies in the speed of response of the associated actuator dynamics. The analysis carried out in this paper is valid for any actuation scheme that can be modeled by (12) with an actuator response that is essentially instantaneous with respect to the time duration of an engine event. For definiteness, this paper assumes a hydraulically actuated cam, whose valve motion profile is depicted in Figure 3.

The key concept is to control the air flow by independently adjusting three parameters IVO, IVL, and IVD of the intake valve motion. The cylinder air charge, m_{a_i} (Kg), is then determined as a function of these parameters via

$$m_{a_i} = \frac{60}{360} \int_{IVO}^{IVO+IVD} \dot{m}_{c_i} \frac{1}{N} d\theta = \frac{60}{360} \int_{IVO}^{IVO+IVD} A_v \left(L_{v_i}(IVL, \theta) \right) d(p_{c_i}, p_m) \frac{1}{N} d\theta \tag{14}$$

For simplicity in this study, IVO and IVL are fixed at top dead center and 8 mm respectively, leaving IVD as the unique control parameter. As an example, Figure 4 shows a typical mass air flow rate curve as a function of IVD, with respect to crank-angle. Then IVD_{β} , which is defined to be the intake valve open timing which approximately achieves

$$m_{a_i} = \frac{60}{360} \int_{IVO}^{IVO+IVD_{\beta}} A_v (L_{v_i}(\theta)) d(p_{c_i}, p_m) \frac{1}{N} d\theta \approx \beta \dot{m}_c T,$$
 (15)

where T is the time taken for the piston to travel from top dead center (TDC) to bottom dead center (BDC), can be determined through simulation, and mapped as a function of engine speed and intake manifold pressure.

D. Mean-value feedgas and torque model

The discrete-event nature of the combustion process introduces transport delays, which are dependent on engine speed. This motivates discretizing the overall model synchronously with engine events [11], [12]. That is, the independent variable is transformed from time to crank-angle, and the model is then discretized at a constant rate in the crank-angle domain. Here, the model is discretized with period π radians in crank-angle, which corresponds to one engine event

²The lower bound of β should be determined to avoid misfire owing to lack of oxygen in the mixture.

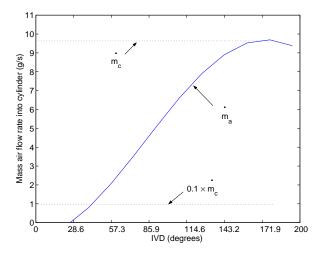


Fig. 4. Mass air flow rate into cylinder as a function of IVD at constant engine speed 1200 RPM and intake manifold pressure 60 KPa.

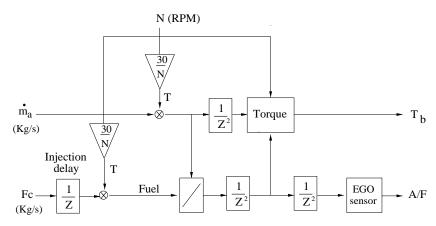


Fig. 5. Feedgas and torque model.

(elapsed time of revolution for the intake stroke, for example). This procedure introduces speed dependent terms in the dynamics, but it permits standard stability analysis to be applied.

The calculation delay in the injection of fuel and the transport delays are included in the model. The dynamics of the EGO sensor is modeled by a first order difference equation; in the time domain, its time constant is 0.20 sec.

The steady state engine brake torque is affected by many parameters such as ignition delay, EGR and so on. The general relations between these parameters and brake torque are derived from experimental data and curve fitting methods. Unfortunately, a torque model for the engine in consideration [8] is not available at this moment, and for this reason, that of [9] is adopted in the model:

$$T_b = -181.3 + 379.36 \times 10^3 m_a + 21.91 (A/F) - 0.85 (A/F)^2 + 0.26 \sigma_s - 0.0028 \sigma_s^2 + 0.0027N -0.00000107N^2 + 0.000048N \sigma_s + 2.55 \times 10^3 \sigma_s m_a - 0.05 \times 10^3 \sigma_s^2 m_a + 2.36 \sigma_s m_e$$
 (16)

where

 m_a : mass air charge (Kg/intake event)

A/F: air-fuel ratio

N: engine speed (RPM) m_e : EGR (g/intake event)

 σ_s : degrees of spark advance before top dead center.

For simplicity, it is assumed that there is no EGR (i.e., $m_e=0$) and ignition delay (σ_s) is set to 30°. The above model was identified [9] at air-fuel ratios between 13.6 and 15.6, engine speeds between 800 RPM and 6000 RPM, intake manifold pressures between 35 and 100 KPa, and torque from 14 to 135 Nm.

The complete block diagram of the feedgas and torque model is shown in Figure 5, and that of the overall engine model is shown in Figure 6.

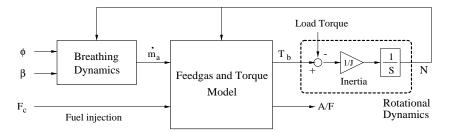


Fig. 6. The block diagram of overall engine model.

III. CONTROL PROBLEM DESCRIPTION

The major objectives of the control design are:

- 1. exploit the air flow actuation capability to achieve higher manifold pressure, thereby reducing pumping losses and improving fuel economy;
- 2. achieve a torque response that is as similar as possible to a conventional engine so that there is no perceptible loss in drivability;
- 3. minimize air-fuel ratio excursions from stoichiometry to maximize the simultaneous conversion efficiency of the catalyst, thereby minimizing overall emissions.

The control inputs are effective valve area factor, β , and (amount of) fuel injection, F_c . It is assumed that the air-fuel ratio is measured by a linear EGO sensor placed in the exhaust stream, just ahead of the catalyst. In addition, it is assumed that some means of measuring torque is available.

As stated, the problem has two-inputs, two-measured outputs and three performance objectives. This imbalance is treated by "squaring down" the performance objectives. At stoichiometry, torque depends primarily on mass air flow. At low primary throttle angle, a static mass air flow model is constructed so as to closely match the steady state torque of the joint-air-and-fuel-controlled engine to that of the conventional engine, while maintaining the intake manifold pressure greater than 50 KPa, in steady state. This also guarantees control authority over cylinder mass air flow rate [5]; see Figure 7. In this regime, the parameter β is near 0.5 to 0.6. At high primary throttle angles, the manifold pressure is already high in a conventional engine, and hence, the static mass air flow model is simply designed to closely match the steady state torque of the conventional engine. The static mass air flow model, and hence the static torque model as well, is a function of the primary throttle angle and engine speed.

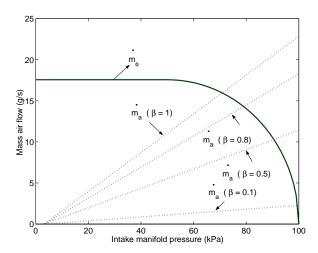


Fig. 7. Steady state mass air flow rate corresponding to the scale factor β at a primary throttle angle of 30°, and an engine speed of 1500 RPM.

The control problem is now defined as in Figure 8: the objective is to design a controller that achieves zero steady state error in commanded torque and stoichiometric air-fuel ratio for constant primary throttle position, while avoiding integrator wind-up. The commanded torque is taken to be a low pass filtered version of the static torque model. The time constant of a first-order low-pass filter, τ_r , is to be determined to trade off drivability (speed of torque response) with emissions (deviations in air-fuel ratio from stoichiometry).

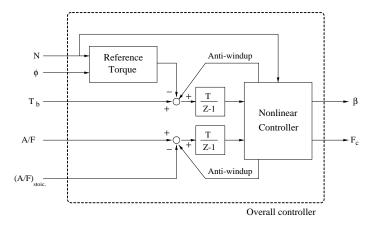


Fig. 8. Controller structure.

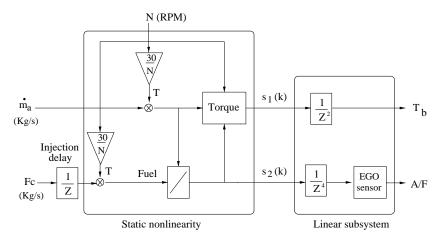


Fig. 9. Equivalent feedgas and torque model.

IV. NONLINEAR FEEDBACK CONTROL DESIGN

This section follows a recent approach to the design of nonlinear controllers, namely control Lyapunov functions (clf). In particular, a Hammerstein-like equivalent model of the feedgas and torque model is introduced and a nonlinear feedback controller is developed based on a positive semi-definite clf. This allows a systematic design procedure for a state feedback controller, and an observer for implementing the state feedback controller. First, a simple state feedback controller is designed based on the intake manifold dynamics plus the nonlinear portion of the Hammerstein-like model, and then the controller is extended to include the linear subsystem. Finally, an asymptotic observer is designed, and stability of the resulting closed-loop system is discussed.

One of the novelties in this work is the use of control Lyapunov functions on a discrete-time system model. Most of the work in this area has focused on continuous-time models.

A. State feedback control

The feedgas and torque model of Figure 5 includes delays and nonlinearities (air-fuel division and torque generation). In the sense of input-output equivalence, it can be rearranged to an equivalent Hammerstein model with a delayed input, as shown in Figure 9. The feedgas and torque model is then a delay plus a static nonlinearity followed by a decoupled linear subsystem.

To fix the main ideas of the clf design, a controller is first designed for an engine model consisting of the intake manifold dynamics followed by the injection delay and static nonlinearity of the feedgas and torque model (the linear subsystem will be initially ignored). The control signals are effective valve area factor, β , and ζ , which is the inverse (amount of) fuel flow rate (s/Kg). To aid in the feedback design, the torque generation equation (16) is linearized around stoichiometry. Since β is limited by 0.1 and 1, and the fuel injection rate is practically constrained (from 0.01 g/s to 20 g/s), discrete state equations are given by

$$p_m(k+1) = p_m(k) + \frac{RT_m}{V_m} T(\dot{m}_{\phi}(p_m(k), N) - \operatorname{sat}_{0.1}^1(\beta) \dot{m}_c(p_m(k), N))$$
(17.a)

$$x_1(k+1) = \frac{1}{T} \operatorname{sat}_{0.05}^{100}(\zeta)$$
 (17.b)

$$T_b(k) = 410.86 \times 10^3 T \dot{m}_c(p_m(k), N) \operatorname{sat}_{0.1}^1(\beta) - 2.98 (T x_1(k) \dot{m}_c(p_m(k), N) \operatorname{sat}_{0.1}^1(\beta) - A/F_s)$$

$$+ \psi(N)$$
(17.c)

$$A/F(k) = Tx_1(k)\dot{m}_c(p_m(k), N)\operatorname{sat}_{0,1}^1(\beta)$$
 (17.d)

where

 x_1 : delayed fuel injection

: intake event duration, $\frac{30}{N}$ (sec), N in RPM

 A/F_s : stoichiometric air-fuel ratio

 $\psi(N) = -37.44 + 0.00414N - 0.00000107N^2$

$$sat_b^a(u) = \begin{cases}
 a & \text{if } u \ge a \\
 u & \text{if } b < u < a \\
 b & \text{if } b \ge u
\end{cases}$$

Figure 10 shows that the approximation error between (16) and the linearized torque is less than 1 Nm for air fuel ratios between 13.6 and 15.6, where the torque model was identified in [9].

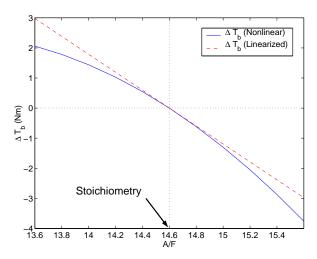


Fig. 10. Torque deviation from nominal torque given mass air charge and engine speed.

Due to the constraints on the inputs, an integrator anti-wind-up scheme needs to be integrated into the controller design. The key idea used here is to actively adjust the reference signals in order to stabilize the integrators. The difference equations of the integrators are modified to include reference adjustment as follows:

$$q_{1}(k+1) = q_{1}(k) + T(T_{b}(k) - r - r_{1}(k))$$

$$= q_{1}(k) + T(410.86 \times 10^{3} T \dot{m}_{c}(p_{m}(k), N) \operatorname{sat}_{0.1}^{1}(\beta)$$

$$-2.98(Tx_{1}(k) \dot{m}_{c}(p_{m}(k), N) \operatorname{sat}_{0.1}^{1}(\beta) - A/F_{s}) + \psi(N) - r - r_{1}(k))$$
(18)

$$q_{2}(k+1) = q_{2}(k) + T(A/F(k) - A/F_{s} - r_{2}(k))$$

$$= q_{2}(k) + T(Tx_{1}(k)\dot{m}_{c}(p_{m}, N)\operatorname{sat}_{0.1}^{1}(\beta) - A/F_{s} - r_{2}(k))$$
(19)

where

r: reference torque (Nm), as a function of primary throttle angle (ϕ) and engine speed (N) r_1 , r_2 : reference signal adjustments that are to be determined.

Since $q_1(k+1)$ and $q_2(k+1)$ have common terms, it is natural to choose a candidate Lyapunov function as

$$V_{L1} = V_1^2 = (q_1 + 2.98q_2)^2 (20)$$

so that these two states are bounded relative to one another; that is, if one of them is bounded, then so is the other. In the next step, another candidate Lyapunov function with parameter κ is chosen to force one of the integrator states, q_2 , to be bounded relative to the state x_1 :

$$V_{L2} = V_2^2 = (\kappa q_2 + x_1)^2 \tag{21}$$

Thus, because p_m is always bounded, if it can later be proven that any one of x_1 , q_1 or q_2 is bounded, then all of the states are bounded.

A composite, quadratic, positive semi-definite Lyapunov function is then given by

$$V_L(x) = V_{L1}(x) + V_{L2}(x) = V_1^2(x) + V_2^2(x) \ge 0$$
, where $x = (p_m, x_1, q_1, q_2)$. (22)

The difference equation can be computed to be

$$\Delta V_L(x(k)) = (V_1^2(x(k+1)) - V_1^2(x(k))) + (V_2^2(x(k+1)) - V_2^2(x(k)))
= (V_1(x(k+1)) - V_1(x(k)))(V_1(x(k+1)) + V_1(x(k)))
+ (V_2(x(k+1)) - V_2(x(k)))(V_2(x(k+1)) + V_2(x(k)))$$
(23)

where

$$V_1(x(k+1)) - V_1(x(k)) = T(410.86 \times 10^3 T \dot{m}_c(p_m(k), N) \operatorname{sat}_{0.1}^1(\beta) - 2.98 r_2(k) + \psi(N) - r - r_1(k))$$
(24)

$$V_{2}(x(k+1)) - V_{2}(x(k)) = \kappa T(Tx_{1}(k)\dot{m}_{c}(p_{m}(k), N)\operatorname{sat}_{0.1}^{1}(\beta) - A/F_{s} - r_{2}(k)) + \frac{1}{T}\operatorname{sat}_{0.05}^{100}(\zeta) - x_{1}(k).$$
(25)

The control signals are designed as

$$\beta(x) = \frac{1}{410.86 \times 10^3 T \dot{m}_c(p_m, N)} (-\psi(N) + r - \frac{c_1}{T} V_1(x))$$
(26)

$$\zeta(x) = -\kappa T^2(Tx_1\dot{m}_c(p_m, N)\operatorname{sat}_{0.1}^1(\beta(x)) - A/F_s) + T(x_1 - c_2V_2(x)), \tag{27}$$

so that

$$V_1(x(k+1)) - V_1(x(k)) = -c_1 V_1(x(k))$$
(28)

$$V_2(x(k+1)) - V_2(x(k)) = -c_2 V_2(x(k)). (29)$$

With $0 < c_1 < 2$ and $0 < c_2 < 2$, this achieves $\Delta V_L(x)$ negative semi-definite if the control signals are within the constraints:

$$\Delta V_L(x) = -c_1(2 - c_1)V_1^2(x) - c_2(2 - c_2)V_2^2(x) \le 0.$$
(30)

If the control signals exceed their constraints, r_1 and r_2 become active and are designed as

$$r_1 = 410.86 \times 10^3 T \dot{m}_c(p_m, N) \operatorname{sat}_{0.1}^1(\beta) - 2.98r_2 + \psi(N) - r + \frac{c_1}{T} V_1(x)$$
(31)

$$r_2 = Tx_1 \dot{m}_c(p_m, N) \operatorname{sat}_{0.1}^1(\beta) - A/F_s + \frac{1}{\kappa T} \left(\frac{1}{T} \operatorname{sat}_{0.05}^{100}(\zeta) - x_1 + c_2 V_2(x)\right), \tag{32}$$

so that (30) is consistently preserved regardless of the input constraints. Note that if the control signals satisfy their constraints, r_1 and r_2 are zero and do not affect the integrators. The goal now is to understand what (30) implies about the stability of the closed-loop system. When $\Delta V_L \equiv 0$, $V_1 \equiv V_2 \equiv 0$ from (30) and the control signals become

$$\beta_{\Delta V_L=0} = \frac{1}{410.86 \times 10^3 T \dot{m}_c(p_m, N)} (-\psi(N) + r) \tag{33}$$

$$\zeta_{\Delta V_L=0} = (1 - \kappa T^2 \dot{m}_c(p_m, N) \operatorname{sat}_{0,1}^1(\beta_{\Delta V_L=0})) T x_1 + \kappa T^2 A / F_s.$$
(34)

Then, given constant primary throttle angle and constant engine speed, the discretized, mean-value breathing dynamics (13) is asymptotically stable in the sense of Lyapunov with control signal (33) if the sampling rate is sufficiently fast. To show this, mass air flow rate \dot{m}_{ϕ} and $\mathrm{sat}_{0.1}^{1}(\beta_{\Delta V_{L}=0}) \times \dot{m}_{c}$ are shown in Figure 11, along with the equilibrium intake manifold pressure, denoted as p_{s} . The discrete breathing dynamics can be expressed as

$$p_e(k+1) = p_e(k) + \frac{RT_m}{V_m} T(\dot{m}_{\phi}(p_e(k) + p_s) - \operatorname{sat}_{0.1}^1(\beta_{\Delta V_L=0}) \dot{m}_c(p_e(k) + p_s))$$

$$= p_e(k) + f(p_e(k)), \tag{35}$$

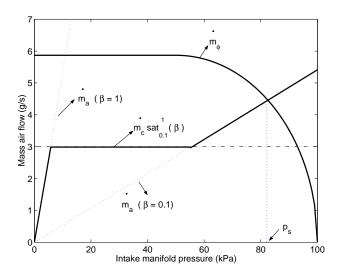


Fig. 11. Mass air flow rate at a primary throttle angle of 20°, and an engine speed of 3500 RPM.

where $p_e(k) = p_m(k) - p_s$. A candidate positive definite Lyapunov function for p_e is

$$V_p(p_e) = p_e^2 > 0, (36)$$

whose difference equation is

$$\Delta V_p(p_e(k)) = (p_e(k+1) - p_e(k))(p_e(k+1) + p_e(k)) = f(p_e(k))(2p_e + f(p_e(k))). \tag{37}$$

Since $f(p_e)$ is a static nonlinearity lying in the second and fourth quadrants, and $|2p_e| > |f(p_e)|$ if the sampling rate is sufficiently fast³, it can be shown that $2p_e + f(p_e)$ lies in the first and third quadrants. Thus $\Delta V_p(p_e)$ is negative definite, which proves asymptotic stability of p_m in the sense of Lyapunov.

With control signals (33) and (34), the state x_1 evolves as:

$$x_1(k+1) = \frac{1}{T} \operatorname{sat}_{0.05}^{100} ((1 - \kappa T^2 \dot{m}_c(p_m(k), N) \operatorname{sat}_{0.1}^1(\beta_{\Delta V_L = 0})) T x_1(k) + \kappa T^2 A / F_s).$$
(38)

The parameter κ is now chosen so that

$$\left|1 - \kappa T^2 \dot{m}_c(p_m, N) \operatorname{sat}_{0.1}^1(\beta_{\Delta V_L = 0})\right| < 1,$$
 (39)

in order that x_1 be asymptotically stable in the sense of Lyapunov. Since $V_1 \equiv V_2 \equiv 0$ in the manifold $W = \{x | \Delta V_L(x) = 0\}$, (20) and (21) imply asymptotic stability in the sense of Lyapunov of the integrator states. Then by [7], the closed-loop system is stable in the sense of Lyapunov, since it is asymptotically stable in the manifold $Z = \{x | V_L(x) = 0\}$, which is equal to W. This also proves asymptotic stability since all states are bounded and approach W by LaSalle's Theorem [14].

Remark: the air-fuel ratio is kept to stoichiometry in steady state as long as the fuel required for stoichiometry is within the allowed constraints.

The above idea can be easily extended to the full order model. The complete state equations of the feedgas and torque model shown in Figure 9 are given by

$$p_m(k+1) = p_m(k) + \frac{RT_m}{V_m} T(\dot{m}_{\phi}(p_m(k), N) - \operatorname{sat}_{0.1}^1(\beta) \dot{m}_c(p_m(k), N))$$
(40.a)

$$x_1(k+1) = \frac{1}{T} \operatorname{sat}_{0.05}^{100}(\zeta)$$
 (40.b)

$$x_2(k+1) = Tx_1(k)\dot{m}_c(p_m(k), N)\operatorname{sat}_{0.1}^1(\beta)$$
 (40.c)

$$x_3(k+1) = x_2(k) (40.d)$$

$$x_4(k+1) = x_3(k) (40.e)$$

$$x_5(k+1) = x_4(k) (40.f)$$

³The sampling period used here of one revolution of the crankshaft can be easily decreased, if necessary [13].

$$x_6(k+1) = (1 - \frac{T}{\tau_s})x_6(k) + \frac{T}{\tau_s}x_5(k)$$
 (40.g)

$$x_7(k+1) = 410.86 \times 10^3 T \dot{m}_c(p_m(k), N) \operatorname{sat}_{0.1}^1(\beta) - 2.98(T x_1(k) \dot{m}_c(p_m(k), N) \operatorname{sat}_{0.1}^1(\beta) - A/F_s) + \psi(N)$$

$$(40.h)$$

$$x_8(k+1) = x_7(k), (40.i)$$

$$T_b(k) = x_8(k) \tag{40.j}$$

$$A/F(k) = x_6(k), \tag{40.k}$$

$$q_1(k+1) = q_1(k) + T(T_b(k) - r - r_1(k))$$
(40.1)

$$q_2(k+1) = q_2(k) + T(A/F(k) - A/F_s - r_2(k)),$$
 (40.m)

where τ_s is the time constant of the EGO sensor. As a next step, V_{L1} and V_{L2} in (20) and (21) are simply extended and replaced with

$$V_{L1} = V_1^2 = (q_1 + 2.98(q_2 + T(x_2 + x_3 + x_4 + x_5) + \tau_s x_6) + T(x_7 + x_8))^2$$
(41)

$$V_{L2} = V_2^2 = (\kappa (q_2 + T(x_2 + x_3 + x_4 + x_5) + \tau_s x_6) + x_1)^2.$$
(42)

Then with the positive semi-definite Lyapunov function

$$V_L(x) = V_{L1}(x) + V_{L2}(x) = V_1^2(x) + V_2^2(x) \ge 0$$
, where $x = (p_m, x_1, \dots, x_8, q_1, q_2)$, (43)

difference equations (23), (24) and (25) are preserved, and accordingly, the same argument can be repeated to show asymptotic stability of p_m and x_1 in the manifold $W = \{x | \Delta V_L(x) = 0\}$, which is equal to $Z = \{x | V_L(x) = 0\}$. It follows therefore that $s_1(k)$ and $s_2(k)$, defined in Figure 9, both converge to constants. Since the linear subsystem of the model in Figure 9 is asymptotically stable, this guarantees that $x_2(k), \ldots, x_8(k)$ converge to constants. Hence, by (41) and (42), the integrator states converge as well. Thus the system is asymptotically stable in the manifold Z, and this proves stability in the sense of Lypaunov of the closed-loop system by [7]. This then proves asymptotic stability in the sense of Lyapunov of the closed-loop system since all the states are bounded and approach W by LaSalle's Theorem [14].

B. Observer based feedback implementation

Since not all of the states are directly measurable, an observer is required in order to implement the feedback of Section IV-A. It is assumed that intake manifold pressure is measured by a MAP sensor. Since x_1 is simply a computation delay and is known, a Kalman filter is designed for the linear subsystem of Figure 9. The filter gain, L, is chosen to achieve an asymptotically stable error dynamics

$$x_e(k+1) = (A - LC)x_e(k),$$
 (44)

where $x_e = x - \hat{x}$. Then for any positive definite matrix M, there exists a unique positive definite matrix Q [15] such that

$$(A - LC)^T Q(A - LC) - Q = -M.$$

$$(45)$$

For the observer-based controller, (41) and (42) are now replaced with

$$V_{L1} = V_1^2 = (q_1 + 2.98(q_2 + T(\hat{x}_2 + \hat{x}_3 + \hat{x}_4 + \hat{x}_5) + \tau_s \hat{x}_6) + T(\hat{x}_7 + \hat{x}_8))^2$$
(46)

$$V_{L2} = V_2^2 = (\kappa (q_2 + T(\hat{x}_2 + \hat{x}_3 + \hat{x}_4 + \hat{x}_5) + \tau_s \hat{x}_6) + x_1)^2.$$
(47)

In order to analyze the closed-loop stability properties, a candidate positive semi-definite Lyapunov function is chosen as

$$V_e(x) = x_e^T Q x_e \ge 0$$
, where $x = (p_m, x_1, \dots, x_8, q_1, q_2, \hat{x}_2, \dots, \hat{x}_8)$, (48)

so that the difference equation, $\Delta V_e(x(k))$, is negative semi-definite:

$$\Delta V_e(x(k)) = V_e(x(k+1)) - V_e(x(k)) = x_e^T(k+1)Qx_e(k+1) - x_e^T(k)Qx_e(k)$$

$$= x_e^T(k)((A-LC)^TQ(A-LC) - Q)x_e(k) = -x_e^T(k)Mx_e(k) \le 0.$$
(49)

In the manifold $Z_e = \{x | V_e(x) = 0\}$, the closed-loop system is asymptotically stable in the sense of Lyapunov, as proven in the previous subsection. This proves stability around equilibria by [7]. Then by LaSalle's Theorem [14], the states approach the largest positively invariant set contained in $W_e = \{x | \Delta V_e(x) = 0\}$. Since $W_e = Z_e$ from (48) and (49), it follows that the closed-loop system is asymptotically stable in the sense of Lyapunov.

V. Performance Analysis

A. Parameter design

This subsection presents a way to select the parameters κ , c_1 and c_2 . In the nominal case, that is, if the effective valve area β is well within its constraint, the condition (39) becomes

$$\left|1 - \kappa T^2 \dot{m}_c(p_m, N) \cdot \beta_{\Delta V_L = 0}\right| = \left|1 - \kappa T \frac{-\psi(N) + r}{410.86 \times 10^3}\right| < 1,\tag{50}$$

and, since the static reference torque r is a function of primary throttle angle and engine speed, κ is scheduled as

$$1 - \kappa(\phi, N)T \frac{-\psi(N) + r(\phi, N)}{410.86 \times 10^3} = 0.1,$$
(51)

in order to maintain condition (39). This will allow a consistent fuel convergence rate over the various operating points when ΔV is close to zero. As a next step, c_1 and c_2 are designed to shape the sensitivity function of the system's linearization about equilibria. For this purpose, the engine model and the observer-based controller are linearized at a nominal operating point, and c_1 and c_2 are tuned to minimize the cost

$$J = \sum_{k=1}^{n} e^{w(\sigma(S_k) - 1)}, \tag{52}$$

where, $\sigma(S_k)$ is the maximum singular value of the input sensitivity function, sampled at discrete points between 0 and one-half of the Nyquist frequency $(\frac{\pi}{T})$, n is the number of sample points, and w is a weight which was chosen to be 10. In this way, the sensitivity function is shaped to be below 1 over the specified frequency range, while the closed-loop bandwidth increases as much as possible.

As an example, at a constant engine speed of 1500 RPM, primary throttle angle equal to 25° and effective valve area factor equal to 0.7, c_1 and c_2 were tuned to 0.08 and 0.06, respectively. Figure 12 shows the maximum singular values of the sensitivity function at different primary throttle angles that are obtained with this method. It is seen that the sensitivity functions are nearly below 1, but degrade as manifold pressure decreases. This is because the air charge actuator begins to lose authority over the air flow as manifold pressure decreases to 50 KPa [5], [3].

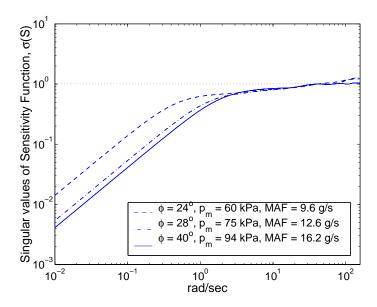


Fig. 12. Singular values of input sensitivity functions at engine speed 1500 RPM; MAF represents mass air flow rate into the cylinder.

B. Simulations

The performance of the controller designed above was first evaluated through the mean-value model. The time constant, τ_r , of the torque reference model was set to 0.05. The stoichiometric air-fuel ratio, A/F_s , is set to 14.6, and the engine speed was held constant at 1500 RPM. The torque and A/F responses were compared to the conventional engine, with fuel managed by a standard feedforward plus PI controller. The feedforward signal was generated in the usual way, and the PI gain was chosen so that A/F excursions are minimized. The results, displayed in Figure 13,

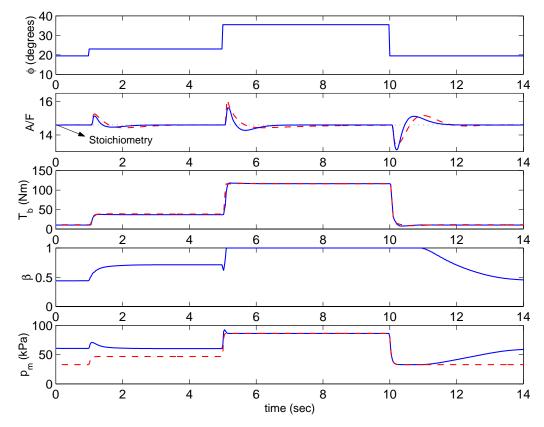


Fig. 13. Simulation results with mean-value model at constant engine speed 1500 RPM. Solid line represents engine with joint air and fuel control and the dashed line the conventional engine with a feedforward plus PI controller.

show that the engine with joint air and fuel control achieves similar torque response to the conventional engine. At 2 seconds, a $+3.5^{\circ}$ step change was given at a primary throttle angle 19.5° , and it is seen that the effective valve area slowly increases to smooth the air flow change caused by the driver, resulting in superior A/F performance over the conventional engine. At 5 and 10 seconds, $+12.5^{\circ}$ and -16° step changes were given. With the throttle step increase, the effective valve area factor reaches its maximum constrained value. In each case, for the engine with joint air and fuel control, the A/F response has a faster convergence rate to stoichiometry than that of the conventional engine. However, there is a slightly larger undershoot at tip-out. Figure 13 also displays the intake manifold pressure. The possibility of controlling the cylinder air charge process has resulted in the ability to maintain a higher manifold pressure than that of the conventional engine, yielding a potential reduction of pumping losses at low primary throttle angles. As discussed previously, this potential would be realized if the actuation were implemented with variable valve timing or height, secondary valves, or via variable cam timing, but would not be realized in the case of secondary throttles. The steady state torque value is approximately equal to that of the conventional engine, even at higher intake manifold pressure, because mass air flow into the cylinders saturates at intake manifold pressures lower than around 50 KPa.

In the second simulation, shown in Figure 14, the performance of the controller was evaluated on the more detailed intake stroke model in Section II. In addition, the fuel puddle dynamics developed in [16] is included in the fuel path after the injection delay, and the engine speed was allowed to vary through a rotational dynamics model. Instead of a MAP sensor, a MAF sensor is used to measure the mass air flow rate into throttle body, and the algorithm in [13] is employed to estimate the intake manifold pressure; see Appendix A for details. In this simulation, the physical variable IVD is plotted instead of the virtual control variable, β ; the effective valve area factor, β , exceeds the constraint from 1.2 to 4.1 seconds. It is seen that the controller achieves a torque response similar to that of the conventional engine, but superior A/F performance and higher intake manifold pressure.

VI. Conclusions

In this paper, a discrete-time, nonlinear controller was developed for joint air and fuel management in a SI engine with variable valve timing. A mean-value model was derived from a detailed intake stroke model, and used for the control design. The control law was based on a conceptually simple control Lyapunov function, and includes a recovery scheme for integrator anti-wind-up. The performance of the closed-loop system was evaluated via simulation on the detailed intake stroke model. It was seen that joint air and fuel management in a SI engine has the potential to achieve

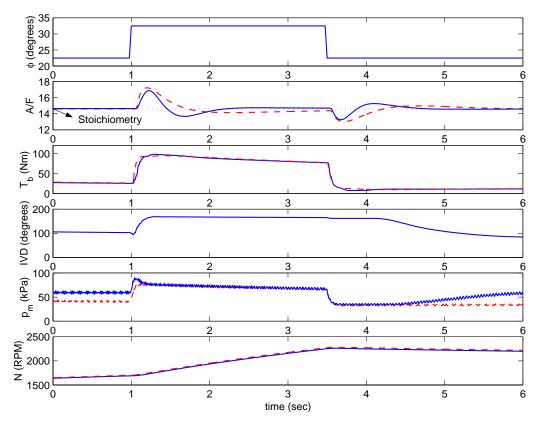


Fig. 14. Simulation results with the detailed intake stroke model, fuel puddle dynamics and varying engine speed. Solid line represents engine with joint air and fuel control and the dashed line the conventional engine with a feedforward plus PI controller. Note that the variation in manifold pressure during an intake event is captured in this simulation.

a faster A/F convergence rate to stoichiometry and lower pumping losses than a conventionally controlled engine, with a torque response that is similar to that of a conventionally controlled engine. This implies that improved fuel economy and emissions performance can be obtained through joint air and fuel control without losing drivability.

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APPENDIX A: INTAKE MANIFOLD PRESSURE ESTIMATE VIA MAF SENSOR

The algorithm in [13], which estimates the intake manifold pressure from direct measurement of \dot{m}_{ϕ} from the MAF sensor (hot-wire anemometer), is briefly summarized here. The sensor dynamics is approximated as a first order lag with time constant τ_m (=0.13 sec):

$$m(k+1) = (1 - \frac{T}{\tau_m})m(k) + \frac{T}{\tau_m}\dot{m}_{\phi}(p_m(k), N),$$
 (I)

where m represents the sensor's output (in Kg/s). The estimate of intake manifold pressure, \hat{p}_m , is

$$\hat{p}_{m}(k+1) = \hat{p}_{m}(k) + \frac{RT_{m}}{V_{m}} T(\dot{m}_{\phi}(p_{m}(k), N) - \operatorname{sat}_{0.1}^{1}(\beta) \dot{m}_{c}(\hat{p}_{m}(k), N))
= \hat{p}_{m}(k) + \frac{RT_{m}}{V_{m}} T(\frac{\tau_{m}}{T} m(k+1) - (\frac{\tau_{m}}{T} - 1) m(k) - \operatorname{sat}_{0.1}^{1}(\beta) \dot{m}_{c}(\hat{p}_{m}(k), N)).$$
(II)

To remove the m(k+1) term, a new variable χ is defined as

$$\chi(k) = \hat{p}_m(k) - \frac{RT_m}{V_m} \tau_m m(k). \tag{III}$$

This yields

$$\chi(k+1) = \chi(k) + \frac{RT_m}{V_m} T(m(k) - \operatorname{sat}_{0.1}^1(\beta) \dot{m}_c(\hat{p}_m(k), N))$$
 (IV)

$$\hat{p}_m(k) = \chi(k) + \frac{RT_m}{V_m} \tau_m m(k). \tag{V}$$

APPENDIX B: NOMENCLATURE

t: time (sec)

 θ : crank-angle (degrees)

R : specific gas constant (=287 J/KgK) T_m : intake manifold temperature (=316 K) V_m : volume of intake manifold (=0.001 m^3)

 V_d : cylinder displacement volume (=4 × 10⁻⁴ m^3) V_{cl} : cylinder clearance volume (=4 × 10⁻⁵ m^3)

 $\begin{array}{lll} p_o & : \text{ atmospheric pressure (100 KPa)} \\ p_m & : \text{ intake manifold pressure (Pascal)} \\ p_{c_i} & : \text{ pressure in } i^{th} \text{ cylinder (Pascal)} \\ V_{c_i} & : \text{ volume of } i^{th} \text{ cylinder } (m^3) \end{array}$

 \dot{m}_{c_i} : mass air flow rate into i^{th} cylinder (Kg/s)

N: engine speed (RPM)

 ϕ : primary throttle angle (degrees)

 \dot{m}_{ϕ} : mass air flow rate into intake manifold (Kg/s)

 A_{ϕ} : effective area of throttle body (m^2) A_v : effective area of intake valve (m^2) L_{v_i} : intake valve lift of i^{th} cylinder (mm)IVO : intake valve open timing (degrees)

IVL : intake valve lift (mm)

IVD : intake valve open duration (degrees)

 \dot{m}_c : mass air flow rate into cylinder of conventional engine (Kg/s) \dot{m}_a : mass air flow rate into cylinder of air flow actuated engine (Kg/s)

 F_c : fuel injection rate (Kg/s)

 β : effective area factor of intake valve ζ : inverted fuel injection rate (s/Kg), $\frac{1}{F_c}$ T : time taken for intake event (sec)

 τ_r : time constant of reference torque (=0.05 sec) τ_s : time constant of EGO sensor (=0.20 sec) τ_m : time constant of MAF sensor (=0.13 sec)

A/F : air-fuel ratio

 A/F_s : stoichiometric air-fuel ratio

 T_b : break torque (Nm) r: reference torque (Nm) κ, c_1, c_2 : control parameters

 S_k : sampled sensitivity function

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