

Paper reading(18) - CL-686

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Title: Model Predictive Idle Speed Control: Design, Analysis, and Experimental Evaluation

Idle Speed: It is the rotational speed of the engine when vehicle is not in motion.

2 controllers: Single input(throttle) and multi-input(throttle and spark timing).

- ① MPC prediction model is developed, model type and order of model are design parameters selected.
- ② States are augmented, The cost function weights and the prediction horizon are the design parameters selected. And obtained controller is simulated to evaluate the performance and the relation between closed-loop performance and tuning parameters.
- ③ The MPC controller is synthesized in a suitable way for implementation in automotive ECUs. Computational feasibility of the controller is verified.
- ④ Experimental tests are performed to validate the nominal and non-nominal performance of the closed-loop system

$$\dot{N}(t) = \frac{1}{J} \frac{30}{\pi} (M_e(t) - M_L(t)) \quad (12a)$$

$$M_e(t) = \kappa_{\text{spk}}(t - t_{ds}) M_{e,\delta}(t - t_d) - M_{\text{fr}}(t) - M_{\text{pmp}}(t) \quad (12b)$$

$$\kappa_{\text{spk}}(t) = (\cos(\alpha(t) - \alpha_{\text{MBT}}(t)))^\xi \quad (12c)$$

$$t_d(t) = \frac{60}{N(t)} \quad (12d)$$

$$\begin{aligned} \dot{M}_{e,\delta}(t) = & -\gamma_2 \frac{RT_{\text{im}}}{V_1} \frac{N(t)}{\gamma_1} M_{e,\delta}(t) + \gamma_2 \frac{RT_{\text{im}}}{V_{\text{im}}} \gamma_3 \vartheta(t) \\ & - \frac{\gamma_0 \cdot \gamma_1}{N^2} \dot{N}(t) \end{aligned} \quad (12e)$$

Figure: Engine model from 1st principle near idle speed

Engine Model for ISC(Contd...)

Where,

$N[RPM]$: engine speed,

$M_e[Nm], M_L[Nm]$: engine brake torque, load torque on crankshaft,

$k_{spk}, \alpha[rad]$: torque ratio, spark ignition angle,

$M_{e,\delta}(t)[Nm]$: indicated engine torque,

$v[deg]$: throttle position

Control oriented model

From the above model,

To remove nonlinearity - control variable is torque ratio k_{spk}

Control inputs are throttle position and torque ratio achieved via spark retard.

Linearized model: $Y(s) = G_{thr}(s)e^{st_d} U_1(s) + G_{spk}(s)e^{st_{ds}} U_2(s)$ (Direct result from another reference paper)

where,

$$y(t) = N(t) - \bar{N}, u_1(t) = v(t) - \bar{v}, u_2(t) = k_{spk}(t) - k_{spk}^-$$

u_1, u_2 are throttle and spark input.

Discretized state space model

$$\begin{aligned}x_{thr}(k+1) &= A_{thr}x_{thr}(k) + B_{thr}u_{thr}^{\delta}(k) \\ y_{thr}(k) &= C_{thr}x_{thr}(k)\end{aligned}\tag{1}$$

$$\begin{aligned}x_{spk}(k+1) &= A_{spk}x_{spk}(k) + B_{spk}u_{spk}^{\delta}(k) \\ y_{spk}(k) &= C_{spk}x_{spk}(k)\end{aligned}\tag{2}$$

where, $k \in \mathbb{Z}$, $x_{thr}, x_{spk} \in \mathbb{R}^2$

$u_{spk}^{\delta}(k), u_{thr}^{\delta}(k)$ are delay free torque ratio and throttle command.

Control oriented model(Contd...)

By cascading a fourth-order airflow delay model with (1), and a first-order spark delay model with (2), the complete linear model of the engine is obtained.

$$\begin{aligned}x^p(k+1) &= A^p x^p(k) + B^p u^p(k) \\ y_p(k) &= C^p x^p(k)\end{aligned}\tag{3}$$

where, for multi input controller

$$x^p = \begin{bmatrix} x_{thr}^f \\ x_{spk}^f \end{bmatrix}, u^p = \begin{bmatrix} u_{thr} \\ u_{spk} \end{bmatrix}, A^p = \begin{bmatrix} A_{thr}^f & 0 \\ 0 & A_{spk}^f \end{bmatrix},$$

$$B^p = \begin{bmatrix} B_{thr}^f \\ B_{spk}^f \end{bmatrix}, C^p = [C_{thr}^f \quad C_{spk}^f],$$

$$x_{thr}^f \in \mathbb{R}^6, x_{spk}^f \in \mathbb{R}^3 \text{ and } u^p \in \mathbb{R}^2$$

for single input controller

$$x^p = x_{thr}^f, u^p = u_{thr}, A^p = A_{thr}^f, B^p = B_{thr}^f, C^p = C_{thr}^f$$

Started with single input (throttle) controller

adding the dynamics $q_N(k+1) = q_N(k) + T_s(y_p(k) - r_{y_p}(k))$

where $q_N \in \mathbb{R}$ is the discrete-time integral of the output. $r_{y_p} \in \mathbb{R}$ is the desired speed set-point offset.

After adding this to eqn (3) of single input controller, state becomes

$$\begin{aligned} x &= \begin{bmatrix} x_p \\ q_N \end{bmatrix}, x \in \mathbb{R}^7 \\ A &= \begin{bmatrix} A_p & 0 \\ T_s C_p & 1 \end{bmatrix}, B = \begin{bmatrix} B_p \\ 0 \end{bmatrix}, C = \begin{bmatrix} C_p & 0 \\ 0 & 1 \end{bmatrix} \end{aligned} \quad (4)$$

MPC design(Contd...)

Constraints and Cost function

Output constraints: $-200 \leq y_p(k) - r_{y_p}(k) \leq 200$

Input Constraints: $0 \leq u + u_{FF} \leq 10$, where $u_{FF} = \bar{v}$ and $v(t) = u_{FF} + u(t)$

Cost function:

$$\begin{aligned} J(y(k), u(k), u(k-1)) \\ = \sum_{i=0}^{h-1} (y(i|k) - r_y)^T Q (y(i|k) - r_y) + \Delta u(i|k) S \Delta u(i|k) \end{aligned} \quad (5)$$

where, $(i|k)$ indicates i steps ahead from k ,

$h \in \mathbb{Z}$ is prediction horizon,

$y(k)$ and $u(k)$ are output and input sequences from 0 to $h-1$ predicted at step k , $r_y = [r_{y_p} \ 0]^T$ is the output setpoint,

$\Delta u(i|k) = u(i|k) - u(i-1|k)$, Q and S are positive definite matrices.

Optimal control problem

$$\begin{aligned} \min_{\sigma, u(k)} \quad & \rho\sigma^2 + \sum_{i=0}^{h-1} (y(i|k) - r_y)^T Q (y(i|k) - r_y) + \Delta u(i|k)^T S \Delta u(i|k) \\ \text{s.t.} \quad & x(i+1|k) = Ax(i|k) + Bu(i|k) \\ & y(i|k) = Cx(i|k), i = 0, \dots, h-1 \\ & u_{\min} \leq u(i|k) \leq u_{\max}, i = 0, \dots, h_u - 1 \\ & y_{\min} - \sigma \mathbf{1} \leq y(i|k) \leq y_{\max} + \sigma \mathbf{1}, i = 0, \dots, h_c \\ & u(i|k) = u((h_u) - 1|k), i = h_u, \dots, h-1 \\ & \sigma \geq 0 \\ & u(-1|k) = u(k-1), x(0|k) = x(k) \end{aligned} \tag{6}$$

The output constraints are “softened” by the additional optimization variable $\sigma \in \mathbb{R}$ and $\rho > 0$

Redefining the constraints

The constraints values are computed by the software running in the vehicle ECU.

For single input,

$$\begin{aligned}y_{min} &= \begin{bmatrix} -200 + r_{yp}(k) \\ -\infty \end{bmatrix} \\u_{min} &= -u_{FF} \\y_{max} &= \begin{bmatrix} 200 + r_{yp}(k) \\ +\infty \end{bmatrix} \\u_{max} &= 10 - u_{FF}\end{aligned}\tag{7}$$

For multi-input,

$$\begin{aligned} y_{min} &= \begin{bmatrix} -200 + r_{yp}(k) \\ -\infty \\ -\infty \end{bmatrix}, u_{min} = \begin{bmatrix} -u_{FF} \\ -0.1 \end{bmatrix} \\ y_{max} &= \begin{bmatrix} 200 + r_{yp}(k) \\ +\infty \\ +\infty \end{bmatrix}, u_{max} = \begin{bmatrix} 10 - u_{FF} \\ \Delta k_{spk}^{max}(k) \end{bmatrix} \end{aligned} \quad (8)$$

Kalman filter

For the idle speed controllers, a stationary Kalman filter has been used for estimating x^p

$$\begin{aligned}x^p(k|k-1) &= A^p x^p(k-1) + B^p u^p(k-1) \\x^p(k) &= x^p(k|k-1) + K_{KF}(y_p(k) - C^p x^p(k|k-1))\end{aligned}\tag{9}$$

where K_{KF} is the stationary Kalman filter gain

$K_{KF} = P C^p T (C^p P C^p T + R_{KF})^{-1}$ and P is the solution of the algebraic Riccati equation

$P = A^p P A^p T + Q_{KF} - A^p P C^p T (C^p P C^p T + R_{KF})^{-1} C^p P A^p T$. And Kalman filter is tuned by Q_{KF} and R_{KF} , that represents here a process and measurement noise covariances.

Closed loop simulations

Defined parameters are,

$$\bar{N} = 600 \text{ r/min}, h = 30, h_c = 3, h_u = 3$$

$\epsilon_N(t) = \hat{N}(t) - N(t)$, Torque disturbance is of 20 Nm

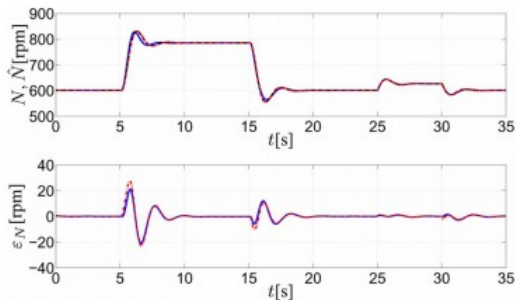


Figure: Upper plot: engine speed vs time, Lower plot: error in control speeds vs time, nonlinear model (dashed line), continuous-time linearized model (solid line), discrete-time linearized model (dashed-dotted line)

Closed loop simulations(Contd...)

For single-input MPC, spark retard k_{spk} was kept constant and throttle v was the only available degree of freedom

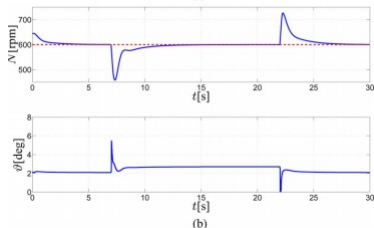
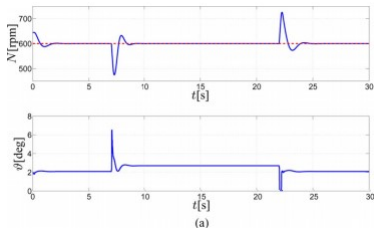


Figure: Single-input MPC with nonlinear model

Part (a) - aggressive tuning - Underdamped behaviour

Part (b) - slower overdamped behavior

Closed loop simulations(Contd...)

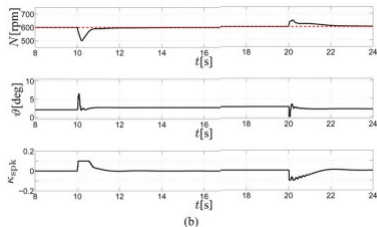
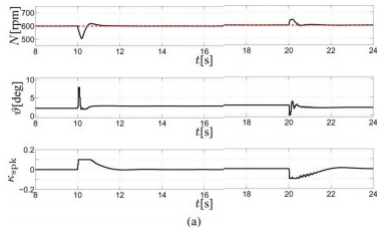


Figure: Multi-input MPC with nonlinear model

In both cases the multi-input controller achieve better disturbance rejection compared to the corresponding single-input MPC in terms of maximum deviation from the setpoint and settling time.