1 Introduction

At the German Aerospace Center (Deutsches Zentrum für Luft- und Raumfahrt e.V., DLR), a free piston internal combustion engine (ICE) is being developed. It is designed to convert chemical energy of a fuel into electrical energy. It can be utilized as a range extender in future electric vehicles. The engine consists of three main components:

1. A combustion section converting chemical energy into kinetic energy.
2. A linear generator converting kinetic energy into electric energy.
3. A gas spring storing energy and inverting the piston movement.

The combustion piston, generator rotor, and gas spring piston are rigidly coupled and perform a transversal oscillation between bottom dead center and top dead center. Compared to conventional combustion engines, this engine concept leads to fully flexible operating parameters such as variable stroke and variable compression ratio. Moreover, the shape of the piston trajectory is variable and in general different from the kinematically determined trajectory of a crank drive. For more information about the project, see Ref. [1].

The development of the free piston engine requires an approach, where every component is tested separately at first. Bringing all three components into service at the same time is an impossible task, because each of them is characterized by a large set of parameters which influence each other. In order to be able to test every single component, it is necessary to use an actuator, which replaces the missing components and, therefore, drives the piston. The actuator moves each component along a predefined desired trajectory so that it can be operated independently from the other two components. Pursuing this strategy, the components can be brought into service and adjusted separately before putting them together. A hydraulic linear actuator is chosen for this application as it is able to generate high forces and it allows to adapt the trajectories without any mechanical changes. Figure 1 shows the hydraulic engine test stand with all three components attached.

Challenging demands are made on the additional actuator. It is supposed to operate at high frequencies (30 Hz) with large strokes (90 mm) and high forces (even the inertia forces add up to 10 kN). Each of these values had been reached separately in the past, but no application is known justifying the need to combine them in a single operating point. The aforementioned values are higher than the ones being necessary for handling test rigs in automotive suspension development. For a summary of the state of the art in this field, see Deuschl [2]. The highest velocity recorded by Deuschl is 2.5 m/s, whereas the combined demands for the engine test stand are equivalent to an average velocity of 4.2 m/s and a maximum velocity of 8.5 m/s. In earthquake simulations [3], shaking tables are used to reaching strokes of 300 mm and maximum frequencies of 100 Hz but not in a single operating point. The limiting factor for shaking tables is the maximum acceleration of 7 g [3], while the accelerations demanded for the free piston linear generator (FPLG) test stand reach up to 160 g. Moreover, the requirements concerning the frequency bandwidth include the fundamental frequency of up to 30 Hz as well as three to seven multiples of it, because the desired plunger motion is a nonharmonic periodic signal.

To improve the performance of the system, the hardware setup is chosen as follows. The main component is a double-action hydraulic working piston, customized with an aluminum plunger in order to reduce the moving mass. It is driven by a three-stage proportional servo valve with a nominal flow rate of 1000 l/min. For an overview of the hydraulic system specifications, see Table 1. A high precision incremental displacement measurement system detects the plunger position. For the control task, a digital controller is implemented and run on a dSPACE real-time system.

The design and implementation of the plunger position controller will be discussed in this paper. Section 2 describes the control task as well as the plant and its mathematical model. Controller design is outlined in Secs. 3 and 4 (feedforward and feedback controller, respectively). In Sec. 5, measurement results are presented and a conclusion is made.

2 Mathematical Modeling and Control Task

In order to develop flatness-based feedforward controller, a model of the plant is built and implemented in MATLAB/SIMULINK. The hydraulic cylinder is described by its motion equation (Eq. (1)) and the differential equation of pressure generation for each chamber (found on the basis of Ref. [4] and exemplified for chamber A in Eq. (2)).

\[
\dot{p}_A - \dot{x}_p = \frac{(p_A - p_B) - d_p \dot{x}_p + F_{\text{Load}}}{V_0 + A_{\text{px}}^p}
\]

\[
\dot{D}_p = \frac{E_{\text{oil}}}{V_0 + A_{\text{px}}^p} (\dot{Q}_A - k_{\text{inj}} (p_A - p_B) - k_{\text{inj}} \dot{x}_p - A_p \ddot{x}_p)
\]

Therein, \(p_A\) and \(p_B\) are the oil pressures in the two chambers, \(x_p\) is the plunger position of the hydraulic actuator, \(A_p\) is the effective area of the plunger, and \(Q_A\) is the total volume flow into chamber A. \(F_{\text{Load}}\) is the force that is applied to the plunger by the three
components of the FPLG and can be regarded as a disturbance for the position controller discussed in this paper.

Internal leakage (leakage between the two chambers) consists of a pressure proportional and a velocity proportional component. This leakage model was originally developed within a swashplate pump model and afterward adapted for hydraulic working cylinders [5].

The parameters \( k_{ip} \) and \( k_{ipv} \) for the leakage model are listed in Table 2 among other model parameters. The parameters for leakage and friction cannot be measured directly nor are they available in any data sheets. In order to find appropriate values, the model is used as a part of the feedforward controller (see Sec. 3) and the model parameters are tuned such that the controller performance is satisfactory. Experimental results show that the entire plant model is comparatively insensitive to changes in those parameters, so that the effort for finding the values can be kept low. Nevertheless, the more accurate treatment of parameters and the robustness against dynamic changes of them are starting points for further performance optimizations, for example, by integrating quantitative feedback laws (QFT, [6,7]). These enhancements would be important especially for applying the model to plants with more significant friction or leakage effects.

Figure 2 shows the hydraulic running cylinder connected to the main valve stage. The main valve stage is the third stage of the entire servo valve. The first two stages (not shown in the figure) leads to an increase of \( \frac{A_0}{C_0} \) relation between the input voltage \( u \) and the output port pressure \( P \).

Second, a black box model is developed in order to represent the transfer behavior between the valve’s input signal \( u \) and the position of the main valve slide \( x \). The position \( x \) of the valve slide depends on the position of the valve slides of the first two stages. The three-stage valve is equipped with an internal controller. The aim of the valve-internal controller is to generate a proportional (“P”) relation between the input voltage \( u \) and the valve slide position \( x \). As neither the mechanical configuration of these valve stages nor the control algorithm and parameters are documented precisely, forming an accurate model based on physical laws is not feasible. Instead, this part of the plant is represented by a black box model. This nonlinear transfer behavior has been identified by

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**Table 1 Hydraulic system specifications**

<table>
<thead>
<tr>
<th></th>
<th>Value</th>
<th>Unit</th>
</tr>
</thead>
<tbody>
<tr>
<td>Actuator manufacturer</td>
<td>Hänchen</td>
<td></td>
</tr>
<tr>
<td>Actuator type</td>
<td>Servofloat 328</td>
<td></td>
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<tr>
<td>Plunger diameter</td>
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<td>Maximum stroke</td>
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<td>Hz</td>
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<td>Hydraulic supply pressure</td>
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<td>bar</td>
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<td>Valve manufacturer</td>
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<td>Valve type</td>
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<tr>
<td>Nominal flow rate</td>
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**Table 2 Model parameters**

<table>
<thead>
<tr>
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<th>Value</th>
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<tbody>
<tr>
<td>( A_p )</td>
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<td>cm²</td>
</tr>
<tr>
<td>( m_p )</td>
<td>3.2 + workload</td>
<td>kg</td>
</tr>
<tr>
<td>( d_p )</td>
<td>2·10³</td>
<td>Ns</td>
</tr>
<tr>
<td>( E_{oil} )</td>
<td>1.5·10⁷</td>
<td>m²N</td>
</tr>
<tr>
<td>( k_{ip} )</td>
<td>1.2·10⁻¹⁰</td>
<td>M²</td>
</tr>
<tr>
<td>( k_{ipv} )</td>
<td>2.4·10⁻⁴</td>
<td>mNs</td>
</tr>
<tr>
<td>( V_0 )</td>
<td>8.5</td>
<td>cm³</td>
</tr>
<tr>
<td>( r_v )</td>
<td>15</td>
<td>mm</td>
</tr>
<tr>
<td>( x )</td>
<td>0.49</td>
<td>—</td>
</tr>
</tbody>
</table>

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the valve manufacturer. In order to invert the model with manageable algebraic effort, a linear transfer function is required to represent this section of the plant. The linearization is possible in two different levels of accuracy.

System identification based on step response data leads to the conclusion that the amplitude response is most suitably described by a fourth order delay element in combination with a dead time element. This is the more accurate linearization, in the following referred to as $PTd_d$. The system description can be simplified by reducing the transfer behavior to a single dead time element (with a longer dead time constant than the dead time element in the $PT_d f_p$ system). In the following, this less accurately linearized system is referred to as $PT_d$. The following steps can be applied to both levels of model detail. Although the procedure of inverting the system is the same in both cases, the resulting equations become by far more unhandy in case of the $PT_d f_p$ system. As this paper only deals with the comparatively slow dynamics of the hydraulic valve lead to a low phase margin in the desired frequency range. Thus, it is impossible to control the system only by a feedback controller; a feedforward controller becomes necessary. Since the feedforward controller is based on a simplified model and disturbances occur, a feedback controller is needed in addition. Thus, to solve the control task of following a desired trajectory of the plunger position, a combination of both feedforward and feedback controller is used. A similar control structure for hydraulic plants has been used for different applications with larger time constants in the past [8,9].

The feedforward controller is designed using a model-based approach. The result of the system inversion will be an algorithm calculating the valve input voltage $u(t)$ that is necessary to track a desired trajectory $x_d(t)$ of the plunger. Therefore, the model derived in Sec. 2 of this paper has to be inverted. Discontinuity, dead time, and nonlinearity have to be treated within the procedure of system inversion. Especially, nonlinearity is an aspect that is part of a wide range of control problems [10,11].

In a first step, the system is inverted only for $x(t)$ (limited operating range in order to avoid the discontinuity). Inverting the dead time is avoided by introducing a time-shifted input voltage

$$\hat{u}_d(t) = u(t - T_d)$$

$T_d$ is the dead time given by the $PT_d$ model of the valve dynamics. Some minor simplifications are made concerning the bulk modulus of the oil in the chambers and the pressure drop at the orifices. The stiffness of the oil and the surrounding walls, pipes, and hoses is assumed to be independent of the plunger position. The pressure drop throughout the valve is assumed to be equally distributed to both orifices which are open at the same time (see Fig. 2). The result is a continuous and causal, yet nonlinear system of differential equations. The number of states is reduced to four (Eqs. (5)–(10))

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} x_p \\ x_v \\ p_A - p_B \\ x_v \end{bmatrix}$$

$$u = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} \hat{u}_d \\ F_{load} \end{bmatrix}$$

$$\dot{x}_1 = x_2$$

$$\dot{x}_2 = x_3 A_p - d_p x_2 + u_2$$

$$\dot{x}_3 = 2 \frac{E_{oil}}{V_0} (\hat{Q} - k_{ij} x_3 - k_{ii} x_2 - A_p x_2)$$

$$\dot{x}_4 = k_i u_1$$

$$\hat{Q} = \begin{cases} \begin{aligned} x \cdot 2 \pi \cdot r_1 \cdot x_1 & \quad \text{if } x_r > 0 \\ x \cdot 2 \pi \cdot r_1 \cdot x_4 & \quad \text{if } x_r < 0 \end{aligned} \end{cases}$$

Implemented in a simulation environment like MATLAB/SIMULINK, this set of equations can be used to simulate the system’s reaction $x$ to a given set of input values $u$. Finding $u$ for a desired trajectory $x_d$ is more complex, as the system (Eqs. (5)–(10)) cannot simply be solved for $u$. Instead, the system has to be inverted.

For inverting nonlinear systems, Fiess et al. introduced the concept of flatness in 1992 [12]. It can be shown that the system (Eqs. (5)–(10)) is differentially flat with the flat output $y = [x_1 \ u_2]'$.
The system is flat because all state variables and all system inputs can be expressed as a function of the flat output and a finite number of its derivatives. These conditions are checked by setting up the following two equations

\[ x = \psi_1(y, \dot{y}, \ldots, y^{(b)}) \]  
\[ u = \psi_2(y, \dot{y}, \ldots, y^{(b+1)}) \]  

In order to find the desired equation for the voltage \( u_1 \) as a function of the desired trajectory, these equations are combined and symbolically solved using a computer algebra system (CAS). The result is shown in Eq. (14). Following an analogous procedure for \( x_2 \leq 0 \) leads to a similar result (Eq. (15))

\[ u_1 = \begin{cases} 
\psi_1(y, \ldots, y^{(b)}) & \text{if } x_1 \geq 0 \\
\psi_2(y, \ldots, y^{(b)}) & \text{if } x_1 < 0 
\end{cases} \]

\[ \psi_1 = \frac{1}{k_v} \left[ \frac{1}{2\pi r_i} \right] \left[ \frac{\rho_v}{p_0} \cdot \frac{m_p y_1 + d_p y_1 - y_2}{A_p} \right] \right] \left[ \frac{V_0 m_p y_1^{(3)} + d_p y_1 - y_2}{2E_{oil}} \right] 
+ k_{ip} \frac{m_p \ddot{y}_1 + d_p \ddot{y}_1 - y_2}{A_p} 
+ k_{ip} \frac{y_1 + A_p \ddot{y}_1}{A_p} \]  

Equations (14) and (15) represent the inverted system equations for the two separate ranges of \( x_1 \). In contrast, Eq. (13) is formally no inverse system in the desired manner, because in addition to the flat output \( y \) and its derivatives, it requires the valve slide position \( x_1 \) as a variable on the right hand side. To solve this problem, the two cases have to be inverted separately and the cases are distinguished afterward. For the purpose of case distinction, the actual valve slide position is the most obvious criterion. An alternative that leads to more precise results is explained in the following. Both equations output the voltage \( u_1 \), which represents the set point for the valve slide position. A positive voltage leads to a positive valve slide position and vice versa. As \( u_1 \) is the variable being calculated by the equations, using itself to select the equation would lead to a circular reference. Analyzing the two equations (14) and (15) shows that the sign of \( u_1 \) only depends on the terms inside the brackets, which is the same in both cases. As a result, the sign of \( u_1 \) and with it the information, which of the two equations is used to calculate \( u_1 \), can be found by evaluating the term inside the brackets. The advantage of this more complex construction is that it takes into account the future (desired) valve position to select the correct model equation when calculating the future valve input voltage. This means that the calculation is always done for one consistent sampling point in time. As a result, discontinuities in the voltage signal are prevented.

The resulting algorithm outputs \( u_1(t) \), which is the desired voltage that a fictitious plant without dead time would need as an input in order to track the desired trajectory (see Fig. 4). In order to find \( u_1(t) \), which is usable for the entire physical plant, the dead time element would have to be inverted as well. It is well known that the inversion of dead time elements cannot be realized, because the resulting inverse system would be acausal. Therefore, only the valve flow, the working piston, and the proportional part (respectively, the \( PT_1 \) part) of the valve dynamics are treated as shown above, but the dead time element is not inverted. Instead, the data fed to the inverse model’s inputs are shifted in time. This means that the block calculating the voltage \( u_1 \) at time \( t \) does not expect the desired plunger position \( x_1 \) at time \( t \) as its input, but the desired plunger position at time \( t + T_{dt} \) which is in the future. The same demand applies for the load force. Thus, both the load force and the desired trajectory have to be known ahead of time. In case of the free piston engine test stand, this is possible, because the trajectory is periodic per definition and the forces can be predicted based on a model of the engine’s components.
Table 3 Methods for considering the error history in a feedback control signal

<table>
<thead>
<tr>
<th>Controller</th>
<th>Signal equation</th>
</tr>
</thead>
<tbody>
<tr>
<td>Integral</td>
<td>( e(t) + e(t - t_0) + e(t - 2t_0) + \cdots )</td>
</tr>
<tr>
<td>Repetitive</td>
<td>( e(t) + e(t - t_0) + e(t - 2t_0) + \cdots )</td>
</tr>
</tbody>
</table>

4 Feedback Control Design

In order to compensate unknown disturbances and model uncertainties, a feedback controller is superposed. The more precisely the feedforward part works, the lower the absolute value of the control signal generated by the feedback part will be. Two feedback control concepts are investigated in the following.

The first implemented feedback controller is the well-known proportional-integral-derivative (PID) controller. It is a widespread algorithm that is being used to solve many hydraulic control problems, e.g., Ref. [13]. When manually tuning the parameters, it turns out that the best parameter set depends on the operating point of the working piston. For good performance, the controller parameters have to be set depending on the desired operating frequency \( f_0 \), equivalent to rpm in a crank-driven engine) and the stroke. This is why the PID controller is expanded to a variable gain PID algorithm, varying its parameters according to a characteristic diagram with the input variables frequency and stroke and the output variables proportional controller gain, integral controller gain, and derivative controller gain. The setup is found experimentally by running the systems in different operating points and tuning the controllers gains for lowest tracking error in every operating point. The result is one characteristic diagram for each of the three parameters \( K_p, K_i \), and \( K_d \) as shown in Figs. 5–7.

On the other hand, a repetitive PD controller is considered. Repetitive control [13,14] is a scheme which can be used in control loops in which the disturbances and/or the setpoint signals are dominantly periodic in time with trial length \( 1/f_0 \). These prerequisites are fulfilled for the system regarded here. On the one hand, during steady state operation, the setpoint signal for the piston position is perfectly periodic. Even in case of transient operation, load changes are performed slowly compared to the cycle duration of the engine. On the other hand, a significant portion of the external forces acting on the hydraulic actuator is periodic due to the fact that they depend on the periodic piston position.

In comparison to the PID controller, the integrating branch is now replaced by a repetitive compensator. As the integrator in the PID is used to compensate the steady-state-error in time domain, the repetitive compensator compensates the steady-state-error in “trial domain” [15]. To illustrate the difference between the two methods, the discrete equations for calculating the integral/repetitive portion of the control signal are shown in Table 3. The equations show that the integral controller uses the entire history to compensate the steady-state-error, whereas the repetitive compensator regards only the current sample and the corresponding samples in previous trials.

5 Simulation and Measurement Results

Figure 9 shows the control signal generated by the feedforward controller for a desired sine wave trajectory with a frequency of 10 Hz and a stroke of 90 mm. First, the load forces are set to zero \((u_{F,F0})\). Afterward, the load forces are generated by a simple combustion model \((u_{F,FPLG})\). The combustion model (not subject to this paper) phenomenologically regards the release of energy and is similar to a Vibe function [16]. The peak around \( t = 0.08s \) is calculated by the feedforward controller due to a sudden increase of combustion pressure after ignition. Although the peak in the figure looks comparatively small, it leads to large hydraulic forces. The additional signal adds an additional displacement to the valve slide, so that an additional amount of oil is pumped into or out of the chambers. Due to the fact that the piston trajectory remains unchanged, the additional amount of oil with high stiffness generates large forces (40 kN in the shown example).

In order to validate the results of the inversion, the feedforward controller is used within a SIMULINK simulation. Therein, the model developed above is implemented as a plant. Also, the simplifications made prior to the inversion process have been implemented. As the plant is based on the same equations as the controller, a correct inversion will result in a perfect feedforward controller that tracks any trajectory without any deviation (as long as saturation does not become
A control concept for a hydraulic actuator with high demands concerning both stroke and frequency has been developed. For this purpose, a nonlinear model of the plant consisting of a hydraulic cylinder and a multistage servo valve was found and was inverted after some simplifications. Nonlinearity, discontinuity, and dead time contained in the plant were regarded when generating a feedforward control algorithm based on the inverse model. The feedforward controller calculates the optimal input signal for tracking a desired trajectory and takes into account the predicted trajectory of a prototype free piston linear generator.

6 Conclusion

A control concept for a hydraulic actuator with high demands concerning both stroke and frequency has been developed. For this purpose, a nonlinear model of the plant consisting of a hydraulic cylinder and a multistage servo valve was found and was inverted after some simplifications. Nonlinearity, discontinuity, and dead time contained in the plant were regarded when generating a feedforward control algorithm based on the inverse model. The feedforward controller calculates the optimal input signal for tracking a desired trajectory and takes into account the predicted load forces on the system. The validity of the implemented solution has been proven in simulations; the practical benefit has been verified experimentally. At the physical plant, a feedback controller is inevitably necessary. The performance of a repetitive controller was compared to the performance of a PID controller. Due to the non phase-minimal characteristics of the plant, the implemented repetitive controller does not lead to significant improvements compared to a conventional PID algorithm. In the most dynamic test runs, the maximum plunger velocity exceeded 8 m/s, which corresponds to an average velocity of 4 m/s. This is a significant expansion of the operating range of hydraulic actuators. The improved dynamic behavior of the hydraulic actuator allows for the development of free piston engines at higher frequencies and strokes. This is an important step for the development process of a prototype free piston linear generator.

Nomenclature

<table>
<thead>
<tr>
<th>Symbol</th>
<th>Description</th>
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<tbody>
<tr>
<td>( A_p )</td>
<td>effective area of hydraulic piston</td>
</tr>
<tr>
<td>( m_p )</td>
<td>overall moving mass (plunger, piston, etc.)</td>
</tr>
<tr>
<td>( d_p )</td>
<td>overall friction coefficient</td>
</tr>
<tr>
<td>( E_{oil} )</td>
<td>elastic modulus of hydraulic fluid</td>
</tr>
<tr>
<td>( V_0 )</td>
<td>hydraulic volume in neutral position of plunger</td>
</tr>
<tr>
<td>( k_{lep} )</td>
<td>coefficient of pressure proportional leakage</td>
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...
\[ k_{lv} = \text{coefficient of velocity proportional leakage} \]
\[ r_s = \text{radius of valve slide} \]
\[ \zeta = \text{orifice coefficient} \]
\[ T_d = \text{valve dead time} \]
\[ x_p\_{\text{des}} = \text{plunger position—desired} \]
\[ x_p = \text{plunger position—measured} \]
\[ x_i = \text{main valve slide position} \]
\[ e = \text{control deviation} \]
\[ F_{\text{Load}} = \text{load force} \]
\[ f_0 = \text{operating frequency FPLG} \]
\[ f_s = \text{sample frequency of the digital controller} \]
\[ p_0 = \text{hydraulic supply pressure} \]
\[ p_{A}, p_B = \text{hydraulic pressure in actuator chamber A/B} \]
\[ \dot{Q}_A, \dot{Q}_B = \text{total volume flow into chamber A/B} \]
\[ \dot{Q} = \text{total volume flow in reduced state system} \]
\[ \dot{Q}_{0\rightarrow A}, \dot{Q}_{0\rightarrow B} = \text{volume flow from pressure supply to chamber A/B} \]
\[ \dot{Q}_{A\rightarrow T}, \dot{Q}_{B\rightarrow T} = \text{volume flow from chamber A/B to tank} \]
\[ u_c = \text{control voltage of servo valve} \]
\[ u_f = \text{feedforward generated control voltage} \]
\[ u_b = \text{feedback generated control voltage} \]
\[ x, x_1, \ldots, x_4 = \text{state space: state variables} \]
\[ u, u_1, u_2 = \text{state space: input variables} \]
\[ y = \text{flat output} \]

References


