Passivity Based Control and Time Optimal Trajectory Planning of a Single Mast Stacker Crane

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abstract: This paper is concerned with feedforward control and active rejection of vibrations of a single mast stacker crane used for the automatic storage or retrieval of load carries in high bay racks. Based on an infinite dimensional mathematical model the trajectory planning and tracking problems are discussed. It turns out that the theory of differential flatness in combination with a finite dimensional approximation of the system is a suitable tool to obtain an open loop control law and to derive time optimal trajectories. The application of backstepping to the infinite dimensional model leads to a passivity based controller for the stabilization of the tracking error. Finally, measurement results show the feasibility of this approach.

Keywords: Infinite Dimensional Systems; Flexible Structures; Flatness based Control; Passivity based Control; Trajectory Tracking; Time Optimal Trajectory Planning.

1. INTRODUCTION

In industry a lot of applications are represented by flexible structures, for example stacker cranes operating in a high bay warehouse. The flexibility in such devices arises because of light weight constructions which are preferably used to satisfy the economic objective to increase the efficiency of the storage system. However, this may cause some vibration problems which must be investigated, because they affect storage access time considerably.

This contribution deals with a flexible mechanical system as depicted in Fig. 1, representing a single mast stacker crane. Basically, it consists of a driving unit \( m_w \) which can move only in the \( X^3 \) direction. A vertical beam carrying a moving mass \( m_h \), in the following also referred to as lifting unit, is attached to this unit. Furthermore, a tip mass \( m_k \) is fixed at the free end of the beam. The task is to move the mass \( m_h \) from a starting point into a desired goal equilibrium point inside the \((X^1, X^3)\) plane as fast as possible. To achieve this specification a flatness based feedforward and a passivity based feedback control concept is proposed, see e.g. Macchelli and Melchiorri [2005], Thull et al. [2005].

This paper is organized as follows. In Section 2 an infinite dimensional model of the system is derived. Based on this model trajectory planning is discussed in Section 3. Here we will use the theory of differential flatness applied to a finite dimensional approximation of the nonlinear system to derive an open loop control. A passivity based control law to stabilize the tracking error is proposed next. This will be the main focus of Section 4. Section 5 is concerned with the problem of time optimal trajectory generation. The property of differential flatness will help us to formulate a nonlinear parametric optimization program which must be solved to obtain the optimal trajectories. Finally, some measurement results show the feasibility of this approach and complete the contribution.
2. THE MATHEMATICAL MODEL

Let us consider the mechanical system shown in Fig. 1. The driving unit is modeled as a rigid body \( m_w \) with position \( x^1 \). The displacement of the beam centerline is denoted by \( w \). We will assume that the beam satisfies the Euler-Bernoulli hypothesis, having length \( L \), uniform mass density \( \rho \), uniform cross section area \( A \) and uniform flexural rigidity \( EI \). The center of gravity of \( m_b \) moves along the beam centerline. As depicted in Fig. 1, the two masses \( m_b \) and \( m_k \) can move in the \((X^1,X^3)\) plane along the beam center line. The vectors \( G_k \) read as \( G_1 = (1,0,0) \) and \( G_2 = (0,0,1) \). The problem and the ordinary differential equations

To derive the equations of motion we apply Hamilton’s Lagrange multipliers is to extend the functional and we are able to define the action functional

Furthermore, the relations

must be fulfilled. It is worth mentioning that (2) is valid for the domains \( \Omega^1 = [0,x^2] \) and \( \Omega^2 = [x^2,L] \) respectively and that (3) correspond to the law of conservation of linear momentum concerning the masses \( m_w \), \( m_b \) and \( m_k \). The clamping and the strain conditions at \( X^1 = 0 \), \( X^1 = x^2 \) and \( X^1 = L \) are summarized in (4). The values \( x^2 \) and \( x^2 \) denote the right-hand and the left-hand limit of \( x^2 \).

3. FLATNESS BASED OPEN LOOP CONTROL

Next, the theory of differential flatness, well known and established for finite dimensional nonlinear systems and generalized for a special class of linear infinite dimensional systems, see e.g. Fliss et al. [1995], Rudolph [2003], is used in order to design the feedforward part of the control law. This will be done by means of a finite dimensional system approximation. Therefore, we apply the Ritz ansatz

for the displacement \( w \) with a function in space

and the new generalized coordinate \( q^i \). The function \( \Phi \) fulfills the equations (4) but it does not incorporate the shear force condition at \( X^1 = x^2 \) and \( X^1 = L \). This simplification has been justified by numerical investigations with a higher order ansatz that fulfills this requirement. The substitution of the ansatz into (1) together with Hamilton’s principle leads to the equations of motion

with \( i,j = 1,\ldots,3 \) and \( u = (u_1,F_2) \). We get a nonlinear mechanical system underactuated by one control. The elements of the symmetric inertia matrix \( M \) and those of the vector \( C \) are

with constants \( m_{11}, m_{22}, m_{12}, k_1 \) and \( k_2 \). The vectors \( G^k \) read as \( G^1 = (1,0,0) \) and \( G^2 = (0,0,1) \). The problem
is now to find a set of flat outputs \((y_1^f, y_2^f)\) for system (6). It is worth mentioning that the explicit state space representation of (6) is not state feedback equivalent to a linear time-invariant system, nevertheless it might be flat. The crucial point is now that the first two rows of (6) can be rewritten as

\[ M_{ij}(x^2)q^j + C^1_i(x^2, \dot{x}^2, \ddot{x}^2)q^j + C^2_i(x^2, \ddot{x}^2)\ddot{q}^j = G^1_iu^i \]

with \(i, j = 1, \ldots, 4\), state vector \(z = (x^1, \dot{x}^1, q^1, \ddot{q}^1)\) and the remaining input \(\bar{u} = F_1\). Assuming that \(y^f_1 = x^2\) we know the trajectory \(x^2(t) = y^f_1(t)\) and its time derivatives and thus, we can treat system (7) as a linear time-varying system

\[ \dot{z}^i = A^i_1(x^2, \dot{x}^2, \ddot{x}^2)z^j + b^i(x^2)\bar{u} \]

(8)

Hence, the construction of a flat output is now straightforward, for example by transformation into time-varying controllability normal form, see e.g. Rothfuß [1997]. A flat output for (8) is given by \(y^f_2 = h_i(x^2, \dot{x}^2, \ddot{x}^2, \dddot{x}^2)\). Because of their complexity the expressions \(h_i\) are not presented here. Summarizing these observations we propose the differential independent functions

\[ y^f_i = (h_i(x^2, \dot{x}^2, \ddot{x}^2, \dddot{x}^2), x^2) \]

(9)
as flat outputs for the original system (6). To show this we succeed in expressing all system variables \(q\) and \(u\) by \(y^f_i\) and its time derivatives which ensures by definition that (9) are indeed flat outputs of the original system (6).

Once the flat outputs are found, we are able to calculate an open-loop control by defining a sufficient smooth trajectory \(y^f_i(t) = y^f_i(0, t)\). To assure continuous state and input trajectories the function \(y^f_i\) must be at least four times, the function \(y^f_2\) at least six times continuously differentiable.

To verify this statement it is remarkable that system (6) is linearizable by static state feedback after adding four integrators at input \(u_2\).

4. CONTROLLER DESIGN

The aim of this section is to derive a control concept for our elastic structure which provides good tracking behavior and at the same time achieves good disturbance rejection. A frequently used approach is to design the controller for a finite dimensional system approximation. This often leads to undesirable spillover effects in particular when collocation between input and output is not given. To ensure collocation between in- and output we will apply a passivity based approach directly to the infinite dimensional system.

4.1 Stabilization of an equilibrium point

Because we are interested in the behavior of the error dynamics we consider the change of coordinates and inputs

\[ z^i_c = \Psi^i(z) = z^i - z^i_d, \quad u^c = u^i - u^i_d \]

(10)

with \(z = [x^1, w]\), \(u = [F_1, F_2]\) and restrict ourself to equilibrium points with constant trajectories for \(z_d\) and \(u_d\).

With this transformation the error system reads as

\[ \rho A\dddot{w} + EI\dddot{w} = 0 \]

(11)

together with

\[ m\dddot{x} + m\dddot{x} = F_1, e, \quad m\dddot{x} + m\dddot{x} = F_2, e, \quad m\dddot{x} + m\dddot{x} = \dddot{w}, e, \]

(12)

\[ EI(h_2(x^2, t) - \ddot{x}^2, (\dot{x}^2, t)) = m_k\dddot{x}^2, \quad EI(h_2(x^2, t) - \ddot{x}^2, (\dot{x}^2, t)) = 0, \]

(13)

and

\[ w_e(0, t) = x^1, \quad \partial_x w_e(0, t) = 0, \quad EI(\ddot{x}) = 0. \]

According to the total stored energy \(H = E_{\text{kin}} + E_{\text{pot}}\) a suitable energy functional for the error system is now given by

\[ H_e = \frac{1}{2}m_w(\dot{x}^1)^2 + \frac{1}{2}m_h(\dot{x}^3)^2 + \frac{1}{2}m_k(\dot{x}^4)^2 + \frac{1}{2}\int_0^L (\rho A(\ddot{w}^2) + EI(\dddot{w}^2)^2)dx^1 . \]

(14)

The calculation of the time derivative of (14) along a solution of (11) to (13) yields

\[ \frac{d}{dt}H_e = \dot{x}^1F_1, e + \dot{x}^2F_2, e . \]

Hence, the energy ports of our error system are given by the collocated pairs \((F_1, e, \dot{x}^1)\) and \((F_2, e, \dot{x}^2)\) and the control law

\[ F_1, e = -\alpha_1\dot{x}^1, \quad F_2, e = -\alpha_2\dot{x}^2 \]

(15)

with \(\alpha_1, \alpha_2 > 0\) might be applied to provide

\[ \frac{d}{dt}H_e = -\alpha_1(\dot{x}^1)^2 - \alpha_2(\dot{x}^2)^2 \leq 0 . \]

It is worth mentioning that here only damping injection is considered. For practical use, in case of unfavorable mass relationships or very strong friction effects at the driving unit, this control law provides a very weak disturbance rejection. To avoid this disadvantage we use the method of backstepping to construct a new energy functional which provides an improved control law. This approach was already successfully applied in e.g. d’Andréa Novel and Coron [2000], Thull et al. [2005] for some well known heavy chain systems.

First we start with the splitting of our system as shown in Fig. 2. We obtain the subsystem beam and lifting unit \(\Sigma_1\)

\[ \Sigma_2 \]

\[ \Sigma_1 \]

\[ F_1 \]

\[ Q \]

\[ F_2 \]

(11)

and the subsystem driving unit \(\Sigma_2\) with the coupling force \(Q = EI\dddot{w}(0, t)\) which represents the shear force at the
clamping of the beam. The energy functional of the error subsystem beam and lifting unit is now given by

\[ H_{\Sigma 1} = \frac{1}{2} m_b (\dot{x}_e^2 + \dot{\gamma}_e^2) + \frac{1}{2} m_e (\dot{x}_e^2 + \dot{\gamma}_e^2) + \int_0^L \rho A (\partial_t \dot{w}_e)^2 + \frac{1}{2} EI (\partial_t^2 \dot{w}_e)^2) dX. \]  

(16)

Again, taking the time derivative of (16) along a solution of (11) to (13) we obtain

\[ \frac{d}{dt} H_{\Sigma 1} = \dot{x}_e^1 Q_e + \dot{x}_e^2 F_{2,e}. \]

Furthermore, we can state that the energy exchange between the two subsystems is done via the port \((\dot{x}_e^1, Q_e)\). To establish stabilization of the positions \(x^1\) and \(x^2\), which was not taken into account yet, (16) is extended with the energy of two nonlinear springs, namely

\[ V_e = \alpha_1 H_{\Sigma 1} + \int_0^{x_1} c_1(\xi) d\xi + \int_0^{x_2} c_2(\xi) d\xi, \]

where the conditions \(c_1(\xi) > 0\) for \(\xi \neq 0\) and \(c_1(0) = 0\) must be met. The time derivative of (17) along a solution of the error system yields

\[ \frac{d}{dt} V_e = \dot{x}_e^1 (\alpha_1 Q_e + c_1(x_1^e)) + \dot{x}_e^2 (\alpha_1 F_{2,e} + c_2(x_2^e)) . \]

Let us consider \(\dot{x}_e^1\) as fictive control input of \(\Sigma_1\), then the control law

\[ \dot{x}_e^1 = -\alpha_1 Q_e + c_1(x_1^e), \]

\[ F_{2,e} = -\frac{1}{\alpha_1} c_2(x_2^e) - \frac{\alpha_2}{\alpha_1} \dot{x}_e^2 \]

with \(\alpha_1, \alpha_2 > 0\) would yield \(\frac{d}{dt} V_e \leq 0\). To include the real control input \(F_{1,e}\), analogously to the backstepping approach for nonlinear finite dimensional systems, the new functional

\[ V_e = \tilde{V}_e + \frac{1}{2} (\dot{x}_e^1 + \alpha_1 Q_e + c_1(x_1^e))^2 \]

is introduced. Once again differentiating (18) with respect to time along a solution of (11) to (13) yields

\[ \frac{d}{dt} V_e = -(\alpha_1 Q_e + c_1(x_1^e))^2 + (\dot{x}_e^1 + \alpha_1 Q_e + c_1(x_1^e))^2 + (\dot{x}_e^1 + \alpha_1 Q_e + c_1(x_1^e))^2 + \frac{\partial c_1}{\partial x_1} (\dot{x}_e^1 + \alpha_1 Q_e + c_1(x_1^e))^2 . \]

(19)

where the shortcut \(\dot{Q}_e = EI \partial_t^2 \dot{w}_e(0, t)\) is used. Now it is easy to show by substituting in (19), that the control law

\[ F_{1,e} = -m_w (\alpha_2 + 1) c_1(x_1^e) + (\dot{x}_e^1 + \alpha_1 Q_e + c_1(x_1^e))^2 + (\dot{x}_e^1 + \alpha_1 Q_e + c_1(x_1^e))^2 + (\dot{x}_e^1 + \alpha_1 Q_e + c_1(x_1^e))^2 + \frac{\partial c_1}{\partial x_1} (\dot{x}_e^1 + \alpha_1 Q_e + c_1(x_1^e))^2 . \]

(20)

\[ F_{2,e} = -\frac{1}{\alpha_1} c_2(x_2^e) - \frac{\alpha_3}{\alpha_1} \dot{x}_e^2 \]

with \(\alpha_1, \alpha_2, \alpha_3 > 0\) implies

\[ \frac{d}{dt} V_e = -(\alpha_1 Q_e + c_1(x_1^e))^2 - \alpha_2 (\dot{x}_e^1 + \alpha_1 Q_e + c_1(x_1^e))^2 - \alpha_3 (\dot{x}_e^1 + \alpha_1 Q_e + c_1(x_1^e))^2 \]

(21)

which provides the desired result

\[ \frac{d}{dt} V_e \leq 0. \]

Now, in contrast to (15), the positions \(x^1, x^2\), the shear force \(Q\) and their time derivatives appear in the control law. This means an additional measurement but it will improve the disturbance rejection considerably.

In contrast to the finite dimensional case to show stability of an infinite dimensional system in the sense of Lyapunov the conditions \(V_e > 0\) and \(\dot{V}_e \leq 0\) are only necessary and therefore, further investigations must be done.

4.2 Trajectory tracking

The trajectory tracking problem is more challenging because in this case the transformation (10) and consequently the error system become time variant and therefore time variant theory must be taken into account. The derivation of a tracking controller based on the time variant error system is part of future efforts.

In this paper we will assume that \(|\tilde{y}_d|\) is sufficiently small and use the control law (20) as tracking controller. The desired trajectories \(z^1_i\) in (10) are replaced by those of section 3. The feasibility of this assumption is shown by simulation and confirmed with measurement results in section 6.

5. TIME OPTIMAL MOTION PLANNING

The previous system analysis tells us that the finite dimensional system representation is differential flat with outputs \(y_i\). Hence, we are confronted with the problem of generating trajectories \(y_i\) which must comply with constraints and sometimes it is desirable to have trajectories that provide optimal performance according to some criteria.

In our particular case we ask for time optimal trajectories to move the system from an initial into a desired goal equilibrium point as fast as possible but within the system constraints. These constraints are once velocity \(v_{\text{max}}\), acceleration \(a_{\text{max}}\) and jerk constraints \(r_{\text{max}}\) of the driving and the lifting unit respectively. Moreover, we have to constrain the bending moment \(M_b\) of the beam at the clamping area because here the highest bending stress occurs. For example, this can be necessary to guarantee fatigue resistance. The limits of the input forces are not taken into account because they are not violated in this setup.

5.1 Problem Formulation

At first the outputs are parameterized in terms of B-spline functions. B-splines are frequently used basis functions because of some significant properties, e.g. their ease of enforcing continuity across knot points and ease of computing their derivatives, they are defined locally, etc. A
detailed discussion of B-splines and their implementation can be found in de Boor [1978]. We get
\[ y_{i,1} = B_{i,1}(t)p_1^1, \quad y_{i,2} = B_{i,2}(t)p_2^2, \]
i_1 = 1, \ldots, N_1, \quad i_2 = 1, \ldots, N_2, \] whereas \( B_{i,n}(t) \) denotes the B-spline basis function of the \( i \)-th knot with order \( n \) and \( p_1, p_2 \) the vectors of parameters to be calculated. As a consequence of this parameterization and the fact \( y_{i,1} \) and \( y_{i,2} \) are flat outputs, the system variables \( z = [x^1, F_1, M_b] \) are determined by \( p_1 \) and \( p_2 \), namely
\[ z = z(y_{i,1}, \partial^2 y_{i,1}, t) = z(p_1, p_2, t) \]
with \( k = 1, \ldots, 4 \). Next, \( N_c \) points are chosen uniformly over the time interval \([0, T_{\text{end}}]\), the constraints will be evaluated at these points. Moreover, we will not directly minimize the end time \( T_{\text{end}} \) but instead we will solve a sequence of subproblems each with fixed \( T_{\text{end}} \).

The subproblem can now be stated as the following nonlinear programming form:
\[
\min_p \| F_1(p, t_k) \|_\infty \quad \text{s.t.} \quad \begin{align*}
x^1(p, 0) &= x_0^1, \\
x^2(p, 0) &= x_0^2, \\
x^1(p, T_{\text{end}}) &= x_{\text{end}}^1, \\
x^2(p, T_{\text{end}}) &= x_{\text{end}}^2, \\
\| \partial_x x^1(p, t_k) \| &\leq v_{\text{max}}, \\
\| \partial^2 x^1(p, t_k) \| &\leq a_{\text{max}}, \\
\| \partial_x x^2(p, t_k) \| &\leq r_{\text{max}}, \\
M_b(p, t_k) &\leq M_{b,\text{max}},
\end{align*}
\]
with \( i = 1, 2 \), \( k = 1, \ldots, N_c \), parameters \( p = [p_1^1, p_2^2] \) and \( t_k \in [0, T_{\text{end}}] \). Each subproblem is solved for a fixed \( T_{\text{end}} \). If a solution is found \( T_{\text{end}} \) is appropriately reduced and the optimization problem is solved again. This procedure is repeated as long as in the next step no possible solution is found. This iteration determines \( \min(T_{\text{end}}) \) and finally the time optimization problem is solved.

5.2 Implementation and Results

The presented algorithm was implemented using C++ programming language. The nonlinear programming solver is chosen out of the NAG C library and is based on the SNOPT package described in Gill et al. [2002]. It uses a sequential quadratic programming (SQP) method and is designed for large-scale constrained optimization problems.

Fig. 3 shows the time optimal trajectories for the position of the lifting unit in the \( X^1-X^3 \) plane. Thereby, starting from position \( A \) the positions \( B, C \) and \( D \) were passed through successively. For example, the optimization for run \( A-B \) was done with the parameters \( N_1 = 60, N_2 = 40 \) and \( N_c = 599 \). In addition to Section 3 we want to constrain the jerk \( r \) as well and so the values \( n_1 = 5 \) and \( n_2 = 7 \) were chosen. Hence, the number of optimization parameters is 109 and the number of constraints is 2001. The CPU-time to solve one subproblem was about 2s and it took five iterations to find \( \min(T_{\text{end}}) \). More details to the several trajectories and their constraints will be presented, together with some measurement results, in the next section.

Roughly speaking time optimal trajectories are described by the characteristic that during the acceleration phases the lifting unit is at the lowest possible beam position. Consequently a higher acceleration of the driving unit can be achieved within the same maximal bending moment.

Therefore, if possible every time the lifting unit moves downwards and just at the last feasible moment it moves upwards again to reach the desired goal position.

6. MEASUREMENT RESULTS

The time optimal feed forward part together with the control law (20) were implemented at the laboratory experiment. Some setup data of interest are \( m_w = 13.1 \) kg, \( m_b = 0.86 \) kg, \( L = 0.54 \) m, \( EI = 15.1 \) Nm and \( \rho A = 2.1 \) kg/m. The constraints are fixed to \( v_1^{\text{max}} = 0.8 \) m/s, \( v_2^{\text{max}} = 0.4 \) m/s, \( a_1^{\text{max}} = 4 \) m/s\(^2\), \( a_2^{\text{max}} = 6 \) m/s\(^2\), \( r_1^{\text{max}} = 2 \) m/s\(^3\), \( r_2^{\text{max}} = 6 \) m/s\(^3\) and \( M_b^{\text{max}} = 0.75 \) Nm. The bending moment is measured via strain gage attached near the clamping of the beam. The shear force which is necessary in the control law could be measured directly by the help of a force sensor. Here it is approximated using the relationship between \( M_b \) and \( Q \) given by the first order ansatz (5), namely \( Q = M_b(\partial^2 \Phi(0, t)/\partial^2 t \Phi(0, t)) \).

Fig. 4 shows the desired and the measured trajectories of run \( A-B \) depicted in Figure 3. The optimal trajectories are calculated in such a way, that despite the motion of the lifting unit, the flexible structure moves from point \( A \) to \( B \) along the defined constraints. The lifting unit remains at the lowest beam position as long as possible and finally moves upwards with \( v_2^{\text{max}} \). Furthermore, it can be observed that the trajectories of the driving unit are adapted to the position of \( m_b \) as well. Here the acceleration is higher and therefore, the duration of this phase is shorter than the braking phase which is much softer.

Although the feedback part of the control law is designed for stabilizing an equilibrium point, it turns out that it acts also perfectly together with the feedforward part. The high frequent vibrations that arise in the measured bending moment signal, occur because of mechanical vibrations caused by some friction effects between lifting unit and beam. But it shows the robustness of the proposed control law that in spite of these uncertainties the closed loop is stable.
Fig. 4. Trajectory tracking performance of the proposed control law.

7. CONCLUSION

We have shown that the combination of flatness based feedforward and passivity based feedback control is an appropriate tool to achieve trajectory planning and tracking for a single mast stacker crane as shown in Fig. 1. In particular it is pointed out that the derivation of the open loop control by a finite dimensional system approximation assures good results. To improve disturbance rejection the method of backstepping is used to design a new Lyapunov function and as a consequence of this a new control law, where we successfully integrated the shear force \( Q \). In addition, the property of differential flatness helped us to formulate a nonlinear optimization problem to finally obtain the time optimal trajectories. The measurement results in Fig. 4 show the feasibility of this approach.

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