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## Modeling, Simulation and Identification of a Mobile Concrete Pump

authored by **J. Henikl, W. Kemmetmüller, M. Bader, and A. Kugi**  
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## Modelling, simulation and identification of a mobile concrete pump

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Due to the light-weight construction of modern large-scale manipulators used, e.g., in mobile concrete pumps, the elasticity of the construction elements plays a significant role in the dynamic behaviour of the system. Therefore, current research is concerned with control strategies for active damping of elastic vibrations and trajectory planning. For this purpose, tailored mathematical models are required. Apart from the mathematical modelling, the identification of the model parameters constitutes a challenging task. This is mainly due to the large number of parameters to be identified and, considering the large scale, due to the fact that the boom movement cannot be measured by means of standard sensors. This paper presents a systematic approach for the mathematical modelling and identification of hydraulically actuated large-scale manipulators. The feasibility of the overall approach is demonstrated by means of measurement results of a mobile concrete pump.

**Keywords:** flexible multibody system; hydraulic cylinder; parameter identification; machine vision

### 1. Introduction

New materials and improved construction and production methods enable the design of large-scale manipulators which possess a significantly reduced weight compared to classical designs. Although these new manipulators can carry the same load as traditional constructions, the reduced stiffness makes the systems vulnerable to vibrations. Therefore, the elasticity of the manipulators has to be systematically incorporated into the mathematical modelling, the analysis and the controller design.

For the modelling of flexible multibody systems, well-developed methods exist in the literature, see e.g., [1,2]. Most investigations in this field are related to flexible robot systems with electromechanical actuators. In large-scale manipulators like mobile concrete pumps, hydraulic actuators comprising hydraulic cylinders and valves are commonly used. Their non-linear dynamic behaviour and the non-linear kinematics of the assembly lead to a complex relationship between piston force and the resulting torque in the joint. Moreover, the dynamics of the pressure in the cylinder chambers is coupled with the piston position and velocity and thus with the motion of the cylinder assembling points on the flexible structure.

The combination of flexible multibody systems and hydraulic actuators has been extensively studied in the literature in recent years, see e.g., [3–6]. Therein, different strategies for modelling the elasticity of the boom are proposed. In [4] and [5], the elastic beam elements are approximated by a number of rigid bodies which are connected via

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joints and spring elements. This leads to a well-known procedure for the derivation of the equations of motion as it is common in robotic applications. The position of the additional joints and the spring rates have, however, no direct relation to real physical parameters.

Another approach is to model the elastic segments in form of Euler–Bernoulli beams. This leads to a distributed-parameter representation which has to be approximated by a finite-dimensional model. In this context, the Ritz method is often used in order to obtain a system of ordinary differential equations for the elastic system, see e.g., [7], [3] and [6].

In this work, the latter approach will serve as a basis for the mathematical modelling. In order to ensure fast simulation times of the transient behaviour of the model and to simplify the controller design, the dimension of the model should be kept as small as possible. The right choice of the basis functions for the Ritz method turns out to be the crucial point to achieve a high model accuracy when using only a small number of basis functions. In [7], the eigenfunctions of a homogeneous Euler–Bernoulli beam are chosen as basis functions and [6] recommends the use of Legendre polynomials. These approaches are all feasible, but they do not account for the inhomogeneous structure of the beam elements of the considered application. Therefore, a high number of basis functions are required to accurately describe the beam deflection dynamics. In particular, the static bending of the beam elements cannot be described with sufficient accuracy by a low number of eigenfunctions or polynomial functions. In [3], a survey of the choice of the basis functions for hydraulically actuated large-scale manipulators is given. Therein, the use of the static bending line of the homogeneous massless beam with an external unity force is proposed as first basis function. In addition, it is also suggested to utilize the numerically calculated eigenfunctions for inhomogeneous beams.

In this paper a set of two basis functions for each beam element is proposed, where the numerically calculated static bending function is used as the first basis function and an orthogonal polynomial serves as a second basis function. As will be shown in the paper, this leads to an adequate approximation of both the static and the dynamic behaviour of the elastic boom.

The choice of the elastic degrees of freedom of the boom also has an important influence on the complexity of the resulting equations, which describe the interconnection between the mechanical and the hydraulic part. For example, the model of [4] results in a simple relationship among piston force, joint torque and the rigid body angle while in contrast, the approach of [6] leads to a complex relationship of the generalized forces depending on the elastic deformations. In the present contribution, an adequate formulation of the static bending function combined with an orthogonal polynomial ensures a simple kinematic relationship between the motion of the cylinder and the elastic boom. In particular, it is shown that the suggested selection of the degrees of freedom results in a cylinder piston position which solely depends on the rigid body angle. This is of special interest, since this allows a simple incorporation of the static friction of the hydraulic cylinders into the mathematical model.

In [8], it is shown that classically designed hydraulic systems used for the actuation of mobile concrete pumps exhibit major weaknesses for the realization of high performance controllers. For this reason, an alternative hydraulic set-up was proposed that fulfils both the dynamic as well as the safety requirements. In this paper, the application of this proposed hydraulic architecture is presumed. This yields a relatively simple mathematical description of the hydraulic components. The identification of the model parameters turns out to be a non-trivial task since the dynamic behaviour of the individual boom segments is highly coupled. A systematic identification procedure is proposed, which is illustrated for the boom of a mobile concrete pump shown in Figure 1. For this purpose, specific static and dynamic experiments were performed, where inclination and pressure sensors,



Figure 1. Boom of a mobile concrete pump.

strain gauges and incremental encoders were utilized. However, the absolute movement of the boom cannot be measured with this setting. Therefore, the use of non-contact measurement principles like laser, infrared or ultrasonic sensors come into play, but they are practically not feasible since a large number of sensors would be required.

This is why we decided to track the movement of particular spots at the boom by means of a camera system by exploiting the methods of machine vision. Commercially available solutions for visual 3D motion like VICON<sup>1</sup> are able to track objects with a precision of 50  $\mu\text{m}$  at 357 Hz with multiple cameras in real time within a closed room as it was used in [9] to control flying quadcopters. Considering, however, the price and the set-up time, such a system would be far oversized, in particular in view of the fact that the motion of the boom for the identification just takes place in a plane of the 3D space. Two high-resolution cameras were used to capture the large working space of the boom. In order to increase the precision, circular markers were attached at the spots to track the boom.

This paper is an extended version of [10], where in addition to the derivation of the mathematical model the identification by means of a tailored vision system and the experimental validation is presented in detail.

The paper is organized as follows. In Section 2, the derivation of the equations of motion for the elastic boom is shown. The mathematical description of the hydraulic actuation and the friction model are discussed in Section 3. In Section 4, the proposed modelling approach is applied to an industrial mobile concrete pump. For this, the simplifications and assumptions used in the model are summarized, the identification procedure of the model parameters is elaborated and the vision system used to measure the boom movement is described. The simulation results of the mathematical model are compared with measurements in Section 5. Finally, Section 6 gives a short conclusion.

## 2. Modelling of the mechanical subsystem

In this section, the general approach for the derivation of the equations of motion for a large-scale elastic manipulator with multiple beam segments will be presented. Figure 2 illustrates the considered planar flexible manipulator with  $N = 4$  beams and the corresponding hydraulic actuators. The degrees of freedom are the rigid-body angles  $\varphi_i$  and the transversal deflections  $w_i(x_i)$  of the beams,<sup>2</sup>  $i = 1, \dots, N$ .

The overall motion of the system is described with respect to the inertial frame  $0_0x_0y_0$ . Furthermore, each beam is equipped with a local coordinate frame  $0_ix_iy_i$ , where the angle  $\varphi_i$  describes the rigid body motion of the  $i$ th beam with respect to the beam  $(i-1)$ . By means of a suitable choice of the orientation of the local coordinate frames, the cylinder

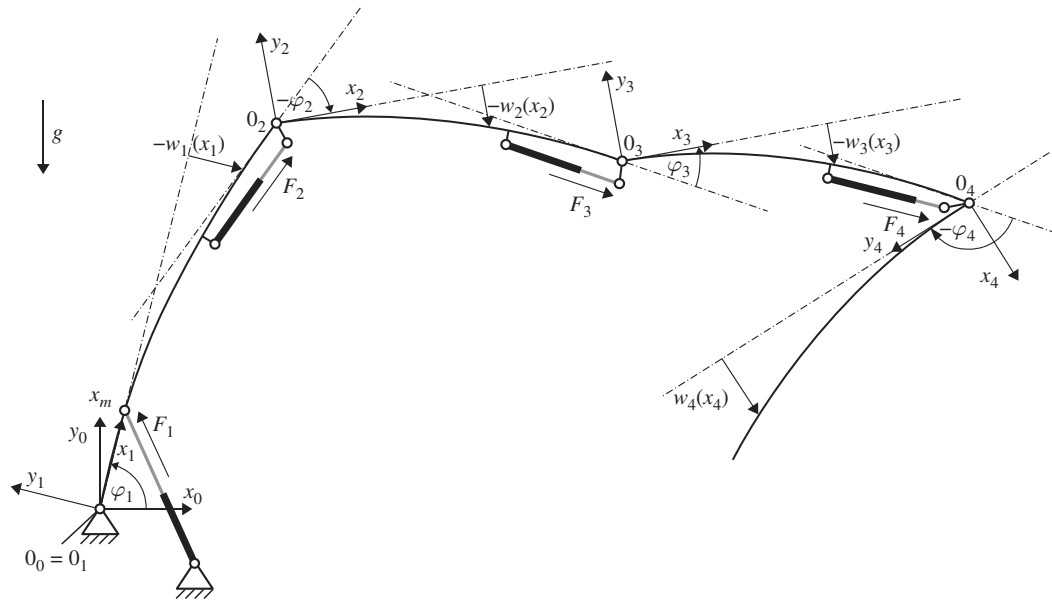


Figure 2. Flexible manipulator with hydraulic actuation.

piston position is solely a function of the respective rigid body angle  $\varphi_i$ . The inputs of the mechanical part are the cylinder forces  $F_i$ .

For the derivation of the mathematical model, the kinetic and potential energy of the system are required. For this, the positions and the velocities of each mass particle have to be described in the inertial frame  $0_0x_0y_0$ . The kinematic relations between the local and the inertial coordinate frames are given by rotation matrices and translation vectors. The inertial coordinates of a mass particle of the  $i$ th beam located at  $\mathbf{r}_i^i(x_i) = [x_i, w_i(x_i)]^T$ , described in the  $i$ th local coordinate frame (subscript), can be calculated by

$$\mathbf{r}_0^i(x_i) = \mathbf{R}_0^i \mathbf{r}_i^i(x_i) + \mathbf{d}_0^i. \quad (1)$$

The matrix  $\mathbf{R}_0^i = \mathbf{R}_0^1 \mathbf{R}_1^2 \dots \mathbf{R}_{i-1}^i$  with

$$\mathbf{R}_0^1 = \begin{bmatrix} \cos(\varphi_1) & -\sin(\varphi_1) \\ \sin(\varphi_1) & \cos(\varphi_1) \end{bmatrix} \quad (2)$$

and

$$\mathbf{R}_j^{j+1} = \begin{bmatrix} 1 & -w'_j(L_j) \\ w'_j(L_j) & 1 \end{bmatrix} \begin{bmatrix} \cos(\varphi_{j+1}) & -\sin(\varphi_{j+1}) \\ \sin(\varphi_{j+1}) & \cos(\varphi_{j+1}) \end{bmatrix} \quad (3)$$

for  $j = 1, \dots, N - 1$  describes the rotation of the local coordinate frame  $0_i x_i y_i$  with respect to the inertial frame  $0_0 x_0 y_0$ . The lengths of the corresponding beams are denoted by  $L_j$ . Similar to [7] small deflections and thus

$$\arctan\left(\frac{\partial w_i}{\partial x_i}\bigg|_{x_i=L_i}\right) \approx \frac{\partial w_i}{\partial x_i}\bigg|_{x_i=L_i} = w'_i(L_i) \quad (4)$$

are presumed. The displacement  $\mathbf{d}_0^i$  between the local coordinate frame  $0_i x_i y_i$  and the inertial frame  $0_0 x_0 y_0$  reads as

$$\mathbf{d}_0^i = \mathbf{R}_0^{i-1} \mathbf{d}_{i-1}^i + \mathbf{d}_0^{i-1}, \quad (5)$$

with  $\mathbf{d}_0^1 = [0, 0]^T$  and  $\mathbf{d}_j^{j+1} = [L_j, w_j(L_j)]^T$  for  $j = 1, \dots, N$ .

Based on the kinematics of the system, the kinetic energy is given by  $T = \sum_{i=1}^N T_i$ , where  $T_i$  is the kinetic energy of the  $i$ th beam

$$T_i = \frac{1}{2} \int_0^{L_i} \sigma_i(x_i) (\dot{\mathbf{r}}_0^i(x_i))^T \dot{\mathbf{r}}_0^i(x_i) dx_i \quad (6)$$

and  $\sigma_i(x_i)$  denotes the position-dependent mass distribution.

Using the Euler–Bernoulli assumptions, the potential energy due to the beam deflection and gravitation can be described by  $V = \sum_{i=1}^N V_i$  with

$$V_i = \int_0^{L_i} \sigma_i(x_i) g r_{0,y}^i(x_i) dx_i + \frac{1}{2} \int_0^{L_i} EI_i(x_i) \left(\frac{\partial^2 w_i(x_i)}{\partial x_i^2}\right)^2 dx_i, \quad (7)$$

where  $g$  is the acceleration of gravity,  $E$  denotes Young’s modulus,  $I_i(x_i)$  is the position-dependent second moment of inertia and  $r_{0,y}^i(x_i)$  is the component of  $\mathbf{r}_0^i(x_i)$  in the direction of  $y_0$ . Viscous damping of the beams is included by means of the Rayleigh dissipation function

$$R = \frac{1}{2} \sum_{i=1}^N \left( d_{\varphi_i} \dot{\varphi}_i^2 + d_{w_i} \int_0^{L_i} \dot{w}_i^2(x_i) dx_i \right), \quad (8)$$

with the coefficients  $d_{\varphi_i} > 0$  representing the damping in the joints and  $d_{w_i} > 0$  accounting for the damping of the beam deflection movement.

Due to the distributed-parameter character of the beams, a mathematical description of the system leads to a set of partial differential equations. To obtain a finite-dimensional approximation, the Ritz method will be employed in the sequel. As already mentioned before, the static bending profiles of the beams are used as first basis functions. The benefit of this choice is the accurate approximation of the elastic deformations at static loads already with the first basis functions alone. Thus, the number of basis functions required for the simulation model can be kept small compared to other choices of the basis functions, e.g., the eigenfunctions of a homogeneous Euler–Bernoulli beam. The shape of the static bending line is determined by the mass distribution of the beam and its external load, which can be represented by a transverse force and torque at the end of the beam. For all beams with the exception of the last one, it can be assumed that the external load is dominating in comparison to the self-weight. Due to the construction of the joints the external load is mainly determined by the transverse forces. Hence, the static bending profiles of the massless beams with position-dependent second moment of inertia  $I_i(x_i)$

and an external force at the free end  $\bar{f}_L$  serve as basis functions. For the last beam, the bending profile resulting only from its own mass distribution is used.

Thus, the static bending profiles  $\bar{w}_i(x_i)$ ,  $i = 1, \dots, N - 1$  are determined by solving the differential equations

$$\frac{\partial^2}{\partial x_i^2} \left( EI_i(x_i) \frac{\partial^2 \bar{w}_i(x_i)}{\partial x_i^2} \right) = 0 \tag{9}$$

for an Euler–Bernoulli beam in a left-side clamped configuration, see Figure 3. Due to the assembling of the first hydraulic cylinder at  $x_1 = x_m$  (cf. Figure 2), a supported configuration is considered for the first beam, see Figure 4. The boundary conditions for the ordinary differential equations are given by

$$\bar{w}_1(0) = 0, \tag{10a}$$

$$\lim_{x_1 \rightarrow x_m^-} \bar{w}_1(x_1) = 0, \tag{10b}$$

$$\lim_{x_1 \rightarrow x_m^+} \bar{w}_1(x_1) = 0, \tag{10c}$$

$$\lim_{x_1 \rightarrow x_m^-} \frac{\partial \bar{w}_1(x_1)}{\partial x_1} = \lim_{x_1 \rightarrow x_m^+} \frac{\partial \bar{w}_1(x_1)}{\partial x_1}, \tag{10d}$$

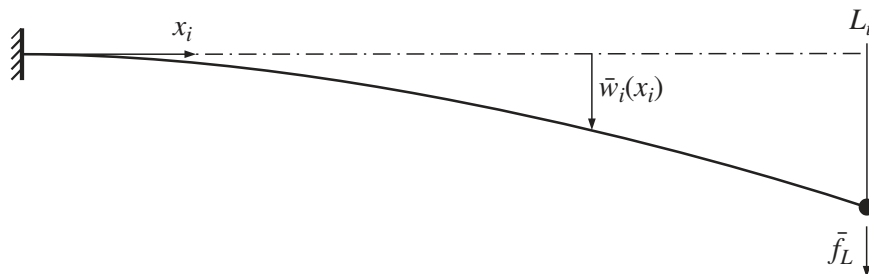


Figure 3. Static bending profile: beam  $i = 2, \dots, N - 1$ .

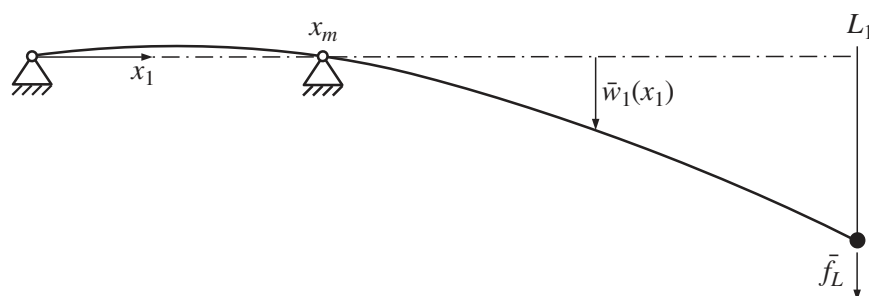


Figure 4. Static bending profile: beam 1.

$$\left( EI_1(x_1) \frac{\partial^2 \bar{w}_1(x_1)}{\partial x_1^2} \right) \Big|_{x_1=0} = 0, \quad (10e)$$

$$\lim_{x_1 \rightarrow x_m^-} EI_1(x_1) \frac{\partial^2 \bar{w}_1(x_1)}{\partial x_1^2} = \lim_{x_1 \rightarrow x_m^+} EI_1(x_1) \frac{\partial^2 \bar{w}_1(x_1)}{\partial x_1^2}, \quad (10f)$$

$$\left( EI_1(x_1) \frac{\partial^2 \bar{w}_1(x_1)}{\partial x_1^2} \right) \Big|_{x_1=L_1} = 0, \quad (10g)$$

$$\left( \frac{\partial}{\partial x_1} \left( EI_1(x_1) \frac{\partial^2 \bar{w}_1(x_1)}{\partial x_1^2} \right) \right) \Big|_{x_1=L_1} = \bar{f}_L \quad (10h)$$

for the first boom segment and

$$\bar{w}_i(0) = 0, \quad (11a)$$

$$\left( \frac{\partial \bar{w}_i(x_i)}{\partial x_i} \right) \Big|_{x_i=0} = 0, \quad (11b)$$

$$\left( EI_i(x_i) \frac{\partial^2 \bar{w}_i(x_i)}{\partial x_i^2} \right) \Big|_{x_i=L_i} = 0, \quad (11c)$$

$$\left( \frac{\partial}{\partial x_i} \left( EI_i(x_i) \frac{\partial^2 \bar{w}_i(x_i)}{\partial x_i^2} \right) \right) \Big|_{x_i=L_i} = \bar{f}_L \quad (11d)$$

for the boom segments  $i = 2, \dots, N - 1$ . The geometric boundary conditions of the joints and the clamped end are represented by Equations (10a–d) and (11a), (11b), respectively. Equations (10e) and (10f) result from the torque-free joints and Equations (10g), (10h) and (11c), (11d) are the boundary conditions for the free ends. For the external force, an arbitrary value  $\bar{f}_L \neq 0$  can be presumed. The value of  $\bar{f}_L$  only scales the result and has no influence on the shape of the bending line. The differential equation for the last beam is given by

$$\frac{\partial^2}{\partial x_N^2} \left( EI_N(x_N) \frac{\partial^2 \bar{w}_N(x_N)}{\partial x_N^2} \right) = -g\sigma_N(x_N) \quad (12)$$

and the boundary conditions, see Figure 5, read as

$$\bar{w}_N(0) = 0, \quad (13a)$$

$$\left( \frac{\partial \bar{w}_N(x_N)}{\partial x_N} \right) \Big|_{x_N=0} = 0, \quad (13b)$$

$$\left( EI_N(x_N) \frac{\partial^2 \bar{w}_N(x_N)}{\partial x_N^2} \right) \Big|_{x_N=L_N} = 0, \quad (13c)$$



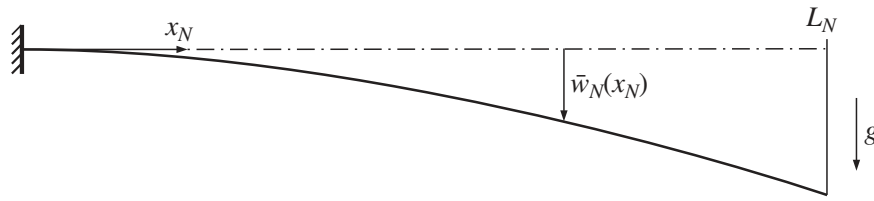


Figure 5. Static bending profile: beam  $N$ .

$$\left( \frac{\partial}{\partial x_N} \left( EI_N(x_N) \frac{\partial^2 \bar{w}_N(x_N)}{\partial x_N^2} \right) \right) \Big|_{x_N=L_N} = 0. \quad (13d)$$

Since the differential Equations (9)–(13a–d) can only be solved numerically, a polynomial approximation of the static bending profile of each beam is performed, i.e.,

$$\eta_1^b(x_1) = \sum_{j=2}^{N_1^b} \eta_{1,j}^b (x_1 - x_m) x_1^{j-1}, \quad (14a)$$

$$\eta_i^b(x_i) = \sum_{j=2}^{N_i^b} \eta_{i,j}^b x_i^j, \quad i = 2, \dots, N, \quad (14b)$$

with the order of the polynomials  $N_1^b$  and  $N_i^b$ , respectively. The coefficients  $\eta_{1,j}^b$  and  $\eta_{i,j}^b$  are determined by minimizing the quadratic error between the numeric solutions of Equations (9)–(13a–d) and the polynomial approximation (14a–b). The formulation of  $\eta_1^b(x_1)$  and  $\eta_i^b(x_i)$  has been chosen in such a way that the relative motion of the mounting points of the hydraulic actuators on the beams can be solely described by the rigid body angles  $\varphi_i$ . As a matter of fact, the polynomials (14) fulfil the geometric boundary conditions (10a–d), (11a), (11b) and (13a), (13b).

In order to increase the model accuracy especially for the dynamical system behaviour and to account for higher bending modes of the beams, polynomials of the form

$$\eta_1^o(x_1) = (x_1 - x_m) x_1^2 + \eta_{1,2}^o (x_1 - x_m) x_1 \quad (15a)$$

$$\eta_i^o(x_i) = x_i^3 + \eta_{i,2}^o x_i^2, \quad i = 2, \dots, N \quad (15b)$$

are used as second basis functions. It can be easily seen that the polynomials (15) also satisfy the geometric boundary conditions. To ensure the orthogonality to the first basis functions (14), the coefficients  $\eta_{1,2}^o$  and  $\eta_{i,2}^o$  result from the orthogonality condition

$$\int_0^{L_i} \eta_i^b(x_i) \eta_i^o(x_i) dx_i = 0, \quad i = 1, \dots, N. \quad (16)$$

The approximated elastic beam deflections are defined as a superposition of the introduced basis functions

$$w_i = h_i^b(t) \eta_i^b(x_i) + h_i^o(t) \eta_i^o(x_i), \quad i = 1, \dots, N, \quad (17)$$

with the time-varying coefficients  $h_i^b(t)$  and  $h_i^o(t)$ . The overall degrees of freedom of the mathematical model are composed of the rigid body angles  $\varphi_i$  and the elastic degrees of freedom  $h_i^b, h_i^o$ , which are summarized in the vector

$$\mathbf{q} = [\varphi_1, \dots, \varphi_N, h_1^b, \dots, h_N^b, h_1^o, \dots, h_N^o]^T. \quad (18)$$

The connection to the hydraulic part is given by the cylinder forces  $F_i$ , which serve as inputs to the mechanical system. Due to the special choice of the local coordinate frames  $0_i x_i y_i$ , the cylinder piston positions  $s_{p,i}$  are solely depending on the rigid body angles  $\varphi_i$ ,

$$s_{p,i} = f_i(\varphi_i), \quad i = 1, \dots, N. \quad (19)$$

The vector of the generalized forces  $\mathbf{Q}$  can be calculated by means of d'Alembert's principle in the form

$$\mathbf{Q} = \left[ F_1 \frac{\partial f_1(\varphi_1)}{\partial \varphi_1}, \dots, F_N \frac{\partial f_N(\varphi_N)}{\partial \varphi_N}, 0, \dots, 0, 0, \dots, 0 \right]^T. \quad (20)$$

The functions  $f_i(\varphi_i)$  are determined by the geometry of the joint construction and are strictly increasing in  $\varphi_i$  in the operating range of the joints.

The application of the Euler–Lagrange equations

$$\frac{d}{dt} \frac{\partial T}{\partial \dot{\mathbf{q}}} - \frac{\partial T}{\partial \mathbf{q}} + \frac{\partial V}{\partial \mathbf{q}} + \frac{\partial R}{\partial \dot{\mathbf{q}}} = \mathbf{Q}^T \quad (21)$$

finally yields the mathematical model of the mechanical part of the system. The resulting equations of motion can be written in the compact form

$$\mathbf{M}(\mathbf{q})\ddot{\mathbf{q}} + \mathbf{c}(\mathbf{q}, \dot{\mathbf{q}}) + \mathbf{D}\dot{\mathbf{q}} + \mathbf{g}(\mathbf{q}) = \mathbf{Q}, \quad (22)$$

with the positive definite inertia matrix  $\mathbf{M}(\mathbf{q})$ , the vector of Coriolis and centrifugal forces  $\mathbf{c}(\mathbf{q}, \dot{\mathbf{q}})$ , the damping matrix  $\mathbf{D}$  and the vector of forces  $\mathbf{g}(\mathbf{q})$  related to the potential energy.

### 3. Modelling of the hydraulic subsystem

As described in Section 1, large-scale manipulators are typically controlled by means of hydraulic actuators. With the hydraulic architecture proposed in [8], the dynamic behaviour of the hydraulic actuators only depends on the hydraulic cylinders and the proportional directional control valves, see Figure 6. The related differential equations for the chamber pressures  $p_1$  and  $p_2$  take the form

$$\dot{p}_1 = \frac{\beta}{V_{01} + A_1 s_p} (-A_1 v_p + q_1), \quad (23a)$$

$$\dot{p}_2 = \frac{\beta}{V_{02} - A_2 s_p} (A_2 v_p - q_2), \quad (23b)$$

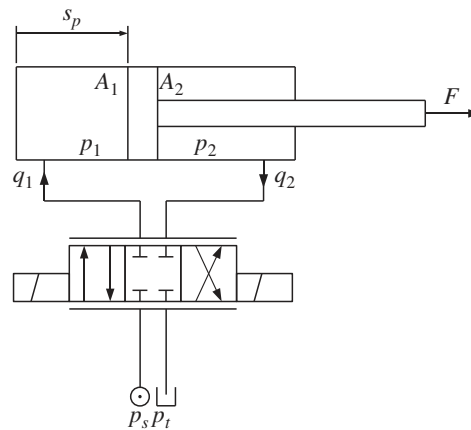


Figure 6. Schematic diagram of the hydraulic actuator.

with the bulk modulus of oil  $\beta$ , the piston areas  $A_1$  and  $A_2$ , the offset volumes  $V_{01}$  and  $V_{02}$ , and the velocity  $v_p = \dot{s}_p$ . The volume flows  $q_1$  and  $q_2$  are determined by the valve position  $s_v$  and the pressure difference,

$$q_1 = \begin{cases} \alpha \sqrt{\frac{2}{\rho}} A_{v1}(s_v) \sqrt{p_s - p_1} & s_v \geq 0 \\ \alpha \sqrt{\frac{2}{\rho}} A_{v1}(s_v) \sqrt{p_1 - p_t} & s_v < 0, \end{cases} \quad (24a)$$

$$q_2 = \begin{cases} \alpha \sqrt{\frac{2}{\rho}} A_{v2}(s_v) \sqrt{p_2 - p_t} & s_v \geq 0 \\ \alpha \sqrt{\frac{2}{\rho}} A_{v2}(s_v) \sqrt{p_s - p_2} & s_v < 0, \end{cases} \quad (24b)$$

where  $\alpha$  denotes the contraction coefficient,  $\rho$  the oil density,  $p_s$  the supply pressure,  $p_t$  the tank pressure and  $A_{v1}(s_v)$  and  $A_{v2}(s_v)$  are the opening areas of the valve to chamber 1 and 2, respectively. The resulting cylinder force  $F$  acting on the mechanical part of the system reads as

$$F = p_1 A_1 - p_2 A_2 - F_R(v_p), \quad (25)$$

where the first part accounts for the pressure force generated by the cylinder and  $F_R(v_p)$  summarizes the friction forces of the cylinder. It is well known that friction plays an important role in hydraulic actuators. In particular a certain amount of static friction is inevitable for the type of hydraulic cylinders typically used in the considered application. In order to analyse the influence of friction on the dynamic behaviour and the control performance, a static friction model is incorporated into the mathematical model. As can be seen from Figure 7, the friction force is composed of a viscous part  $r_V v_p$ , a Coulomb part  $r_C \text{sign}(v_p)$  and the static friction  $r_S$ .

The static friction phenomenon leads to a switched system behaviour, i.e., if the piston velocity is zero ( $v_p = 0$ ) and the absolute value of the sum of the forces on the piston is smaller than  $r_S$ , then the piston sticks and the whole system loses a degree of freedom.

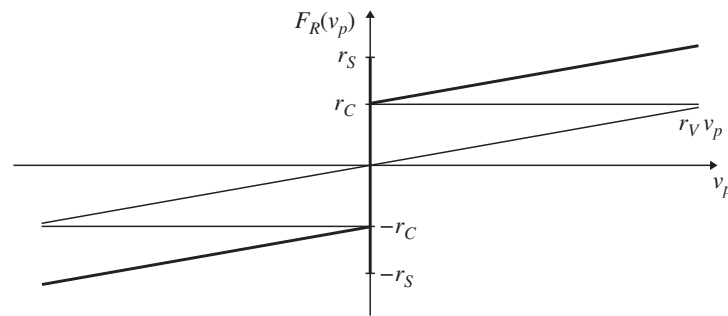


Figure 7. Static friction model.

Due to the special choice of the coordinate frame and the basis functions, the detection of zero velocity  $v_{p,i} = 0$  of the  $i$ th piston is equivalent to check for  $\dot{\varphi}_i = 0$ . Furthermore, if the  $i$ th cylinder is sticking, the equations of motion result from Equation (22) by using the constraints  $\dot{\varphi}_i = 0$  and  $\ddot{\varphi}_i = 0$ . To verify the sticking condition and calculate the absolute value of the sum of the forces acting on the piston, the force  $\bar{F}_i$  of the mechanical part is calculated from the  $i$ th row of Equations (20) and (22)

$$\bar{F}_i = \left( \sum_{j=1}^N \mathbf{M}(\mathbf{q})[i,j] \ddot{\mathbf{q}}[j] + \mathbf{c}(\mathbf{q}, \dot{\mathbf{q}})[i] + \sum_{j=1}^N \mathbf{D}[i,j] \dot{\mathbf{q}}[j] + \mathbf{g}(\mathbf{q})[i] \right) \left( \frac{\partial f_i(\varphi_i)}{\partial \varphi_i} \right)^{-1}. \quad (26)$$

Since the functions  $f_i(\varphi)$  are strictly increasing in the operating range, the invertibility of  $\partial f_i(\varphi_i)/\partial \varphi_i$  is ensured. Summarizing, the  $i$ th piston is sticking if  $v_{p,i} = 0$  and the condition

$$|p_{1,i} A_{1,i} - p_{2,i} A_{2,i} - \bar{F}_i| \leq r_{S,i}, \quad (27)$$

holds.

#### 4. Parameter identification for a mobile concrete pump

In this section, the industrial mobile concrete pump with four joints according to Figure 1 with an operating range of about 40 m is considered. The mathematical model can be derived along the lines of Sections 2 and 3 provided that the following assumptions hold:

- The concrete pump is considered as a planar manipulator. The rotational movement and the lateral beam deflections are neglected.
- The truck's movement due to the dynamic load of the boom is neglected.
- The boom segments are modelled as Euler–Bernoulli beams.
- The pumping of wet concrete is not considered.

However, in view of the last assumption it is worth noting that for the control design the vibrations induced by the pumping are considered as disturbances acting on the system. The model parameters are determined by means of data sheets, design drawings and experimental measurements. In contrast to the parameters of the hydraulic components, which are well known from data sheets, the parametrization of the

boom model turns out to be more involved. Besides the geometrical data, the position-dependent mass distribution and the second moment of inertia can be calculated from design drawings. In this context, it is rather difficult to systematically account for externally assembled components on the boom segments like hydraulic cylinders, hydraulic lines and concrete delivery pipes. In particular, their influence on the resulting stiffness of the overall system can hardly be determined analytically. Furthermore, the damping coefficients are unknown and have to be identified by dynamical measurements.

#### 4.1. Experimental set-up

In the experimental set-up, in addition to the machine vision system, which will be presented in Section 4.2, the following sensors were used:

- Inclination sensors at both ends of each boom segment.
- Two pressure sensors at each hydraulic cylinder.
- Incremental encoders to measure the cylinder piston movements.
- Strain gauges attached to the end of each boom segment where the hydraulic actuator is located, in order to determine the beam deflection.

The strain gauges measure the elongation of the surface of the boom segment. Using the Euler–Bernoulli assumption, the relation between the elongation and the beam deflection is given by

$$\varepsilon_i(x_{SG,i}) = k_{SG,i} \frac{\partial^2 w_i(x_i)}{\partial x_i^2} \Big|_{x_i=x_{SG,i}} + b_{SG,i}, \quad (28)$$

with the gain  $k_{SG,i}$  the bias  $b_{SG,i}$  and the strain gauge position  $x_{SG,i}$ . In the case of a homogeneous beam, the gain coefficient  $k_{SG,i}$  is given by the distance of the surface to the symmetry line of the beam. However, in the case of an inhomogeneous beam it is difficult to determine  $k_{SG,i}$  analytically. Since in practice it is nearly impossible to attach the strain gauge to a completely unbent boom segment, the strain gauge signal contains a bias component  $b_{SG,i}$ , which also depends on the temperature of the beam. Therefore, the gain and bias coefficients have to be identified by means of measurements.

#### 4.2. Vision system

The parameters of the mathematical model of the boom, in particular the mass distribution, the stiffness and damping coefficients and the strain gauge coefficients, could be identified based on static measurements at several configurations of the boom and dynamic experiments with a prescribed excitation of the system. However, due to the large scale of the boom there was the desire to also validate the absolute movement of the boom segments, which cannot be captured by the standard sensors installed. Therefore, a tailored vision system was developed which allows to measure the absolute movement of particular spots on the boom. For this purpose, markers placed at the end of each boom segment and the slewing gear, see Figure 8, were recorded by cameras, and their location in the work space was calculated by means of the respective pixel coordinates on the images.

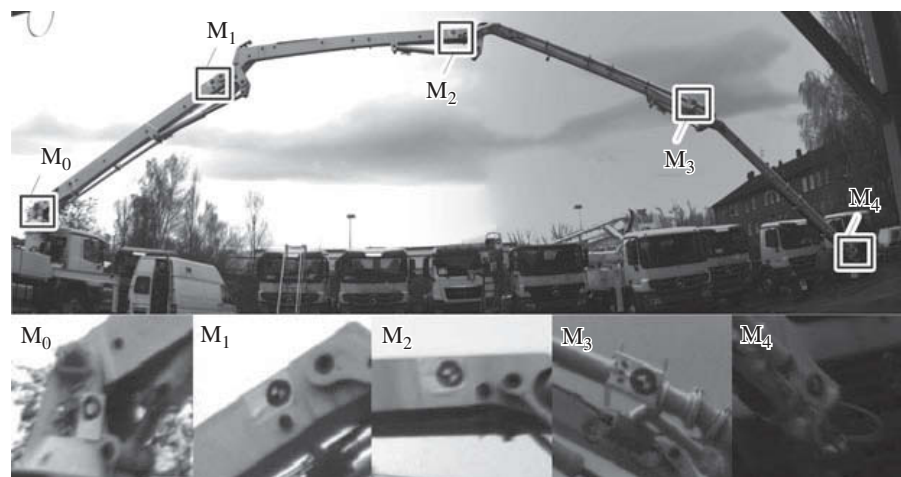


Figure 8. Markers placed on the boom.

In order to capture the large work space of the boom, two 5-mega-pixel greyscale image FireWire cameras<sup>3</sup> were used. The  $2452 \times 2054$  pixel resolution and a frame rate of 15 Hz allow for measurements within a working space of 20 m with an average pixel precision of about 0.01 m/pixel. The open source framework ROS<sup>4</sup> [11] for robotics applications was utilized to record synchronized camera images and to visualize the measurements. After comparing various marker types, including [12–14], circular markers without encoded identifiers were selected. The detection of the circles and their projected elliptical shapes was done by using the standard computer vision library OpenCV.<sup>5</sup> With the algorithm presented in [15] and [12], the precision of the marker position estimation can be increased up to sub-pixel level. This enables the detection of boom movements of even 2 mm. In addition to the markers on the boom, further markers located at fixed positions were used for offline estimation of static camera poses. In order to start the tracking algorithm, all measurement points must be initialized in the camera image.

#### 4.2.1. Feature detection

In order to speed up the marker detection process, the algorithm has an initial feature location and predicts the next location throughout the process. For this reason, the marker detection has to be applied only to a small window around the predicted feature location. During the initialization phase, pixel locations and the estimated size of the circular feature have to be defined. The detection algorithm then detects multiple ellipse hypotheses in each predicted search window. For this purpose, a function which uses the estimated location and the expected radius sorts all detected ellipses within the search window and places the most likely one first. This ellipse is then used as the basis for the next search window. Figure 9 shows multiple search windows with the most likely ellipse in red.

#### 4.2.2. 3D data generation

The geometric relation between the marker position in the workspace and their projected image coordinates are given in this section. The transformation of a point  $\mathbf{p}_W = [p_{W,x}, p_{W,y}, p_{W,z}]^T$ , defined in the workspace coordinate frame  $0_W x_W y_W z_W$ , to the

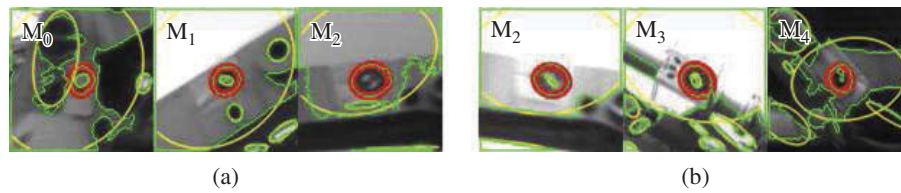


Figure 9. Search windows with detected edges in green, ellipses in yellow and the most likely marker ellipse in red. (a) Left camera. (b) Right camera.

respective coordinates  $\mathbf{p}_C = [p_{C,x}, p_{C,y}, p_{C,z}]^T$  in the camera coordinate frame  $0_C x_C y_C z_C$ , which is defined by the camera centre, is given by the translation vector  $\mathbf{d}_C^W \in \mathbb{R}^{3 \times 1}$  and the rotation matrix  $\mathbf{R}_C^W \in \mathbb{R}^{3 \times 3}$ ,

$$\mathbf{p}_C = \mathbf{R}_C^W \mathbf{p}_W + \mathbf{d}_C^W. \quad (29)$$

In the literature, this transformation is typically described by the so-called external camera matrix, see e.g., [16].

Assuming a pinhole camera model, the projection of the point  $\mathbf{p}_C$  onto the 2D image frame  $0_I u_I v_I$  is given by

$$\mathbf{p}_I = \begin{bmatrix} p_{I,u} \\ p_{I,v} \end{bmatrix} = \begin{bmatrix} -f_u \frac{p_{C,x}}{p_{C,z}} + o_u \\ -f_v \frac{p_{C,y}}{p_{C,z}} + o_v \end{bmatrix}, \quad (30)$$

with the internal camera parameters  $f_u = f/s_u$ ,  $f_v = f/s_v$ ,  $o_u$  and  $o_v$ . This projection is typically referred to as intrinsic camera matrix, see e.g., [16]. The focal length  $f$  describes the distance of the camera centre to the sensor chip or the image frame, respectively,  $s_u$  and  $s_v$  represent the effective physical dimensions of one pixel on the camera chip and the distance between the image centre and the origin of the image frame is given by  $o_u$  and  $o_v$ .

In order to compensate for lens distortions and to be able to use a simple pinhole camera model, every frame captured has to be preprocessed first to get an undistorted image before it is further used in any detection process, see e.g., [16].

The parameters for the compensation of the lens distortions as well as the internal camera parameters were estimated by means of an offline calibration procedure. The calibration was done using a checker-board pattern of known size and the ROS camera calibration node.<sup>6</sup> For the estimation of the three linearly independent quantities parametrizing the rotation matrix  $\mathbf{R}_C^W$  and the three coefficients of the translation vector  $\mathbf{d}_C^W$ , an iterative method based on the Levenberg–Marquardt [17] optimization was utilized, which is implemented in the before mentioned OpenCV library. For this purpose, the workspace coordinates of at least three known reference points  $\mathbf{p}_W$  and their visual appearance in the image frame coordinates  $\mathbf{p}_I$  are required. Thus, with the transformation (29) and the projection (30), six non-linear equations are given to determine the six unknown coefficients for  $\mathbf{R}_C^W$  and  $\mathbf{d}_C^W$ . With the use of additional reference points, the error due to the measurement uncertainty of the reference coordinates could be minimized. In Figure 10, five reference points  $\mathbf{p}_{W,j}^e$  for  $j = 1, \dots, 5$ , used for the estimation of the camera pose, are shown. The coordinates were measured manually by a laser distance sensor. The reference

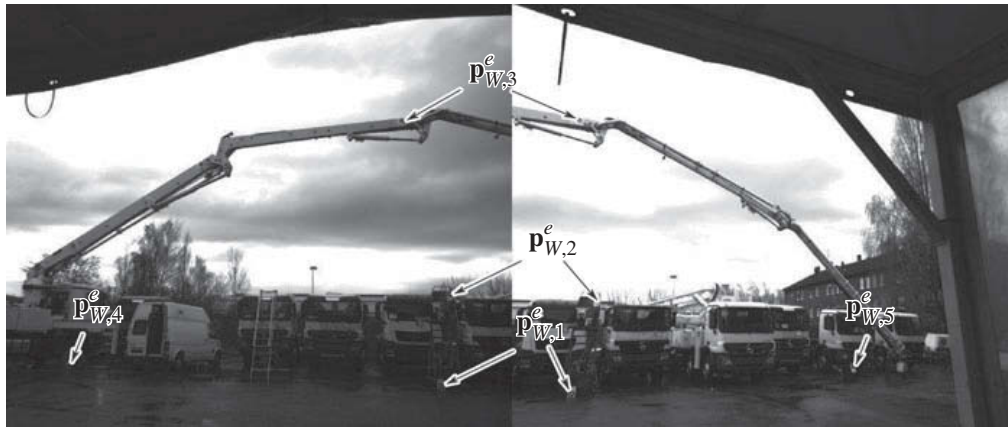


Figure 10. Reference points.

points  $\mathbf{p}_{W,l}^e$  for  $l = \{1, 2, 3, 4\}$  and  $\mathbf{p}_{W,r}^e$  for  $r = \{1, 2, 3, 5\}$  were used for the left and the right camera, respectively.

With these results, the relation of the coordinates of a marker centre  $\mathbf{p}_W^m$  described in the workspace coordinate frame and the respective coordinates  $\mathbf{p}_I^m$  is given. The goal here is, however, the inverse problem. A 3D point linked to a 2D image point has to be found. This can be done by creating a straight line, see e.g., [18] that intersects the camera centre and the image frame at the respective 2D image point. Considering Equation (29) and  $\mathbf{p}_C^e = \mathbf{0}$ , the camera centre described in the workspace frame is given by

$$\mathbf{p}_W^c = -(\mathbf{R}_C^W)^{-1} \mathbf{d}_C^W. \quad (31)$$

With Equation (30) and  $p_{C,z} = f$ , the 2D image point at the image frame in front of the camera, defined by the pixel coordinates of the marker centre  $\mathbf{p}_I^m = [p_{I,u}^m, p_{I,v}^m]^T$ , is located at

$$\hat{\mathbf{p}}_C^m = \begin{bmatrix} -\frac{1}{f_u} (p_{I,u}^m - o_u) f \\ -\frac{1}{f_v} (p_{I,v}^m - o_v) f \\ f \end{bmatrix} \quad (32)$$

and

$$\hat{\mathbf{p}}_W^m = (\mathbf{R}_C^W)^{-1} \hat{\mathbf{p}}_C^m - (\mathbf{R}_C^W)^{-1} \mathbf{d}_C^W, \quad (33)$$

respectively. With this and Equation (31), the direction of the line is given by

$$\mathbf{v}_W = \hat{\mathbf{p}}_W^m - \mathbf{p}_W^c = f (\mathbf{R}_C^W)^{-1} \begin{bmatrix} -\frac{1}{f_u} (p_{I,u}^m - o_u) \\ -\frac{1}{f_v} (p_{I,v}^m - o_v) \\ 1 \end{bmatrix} = f \bar{\mathbf{v}}_W \quad (34)$$

and thus, the line could be described by the equation



$$\mathbf{p}_W^l(\lambda) = \mathbf{p}_W^c + \lambda \mathbf{v}_W \quad (35)$$

for  $\lambda \in \mathbb{R}$ . The final 3D marker location  $\mathbf{p}_W^m$  can be computed under the assumption that the boom and the marker moves only in the plane defined by

$$\mathbf{n}_\Sigma^T \mathbf{p}_W = \lambda_\Sigma, \quad (36)$$

with the normal vector  $\mathbf{n}_\Sigma^T$  to the plane and a scalar  $\lambda_\Sigma$  defining the position of the plane. The normal vector  $\mathbf{n}_\Sigma^T$  and the scalar  $\lambda_\Sigma$  were determined by means of the reference points lying in the plane, see Figure 10. With this, the scalar  $\lambda_\tau$  has to be found that defines the intersection of the line (35) with the plane (36),  $\mathbf{p}_W^c + \lambda_\tau \mathbf{v}_W = \mathbf{p}_W^m$ . Considering  $\mathbf{n}_\Sigma^T(\mathbf{p}_W^c + \lambda_\tau \mathbf{v}_W) = \lambda_\Sigma$ ,  $\lambda_\tau$  is given by

$$\lambda_\tau = \frac{\lambda_\Sigma - \mathbf{n}_\Sigma^T \mathbf{p}_W^c}{\mathbf{n}_\Sigma^T \mathbf{v}_W}, \quad (37)$$

and therefore  $\mathbf{p}_W^m$  reads as

$$\mathbf{p}_W^m = \mathbf{p}_W^c + \frac{\lambda_\Sigma - \mathbf{n}_\Sigma^T \mathbf{p}_W^c}{\mathbf{n}_\Sigma^T \mathbf{v}_W} \mathbf{v}_W = \mathbf{p}_W^c + \frac{\lambda_\Sigma - \mathbf{n}_\Sigma^T \mathbf{p}_W^c}{\mathbf{n}_\Sigma^T \bar{\mathbf{v}}_W} \bar{\mathbf{v}}_W. \quad (38)$$

### 4.3. Identification procedure

Due to the model complexity and the high number of system parameters, the identification has to be performed in several steps. The following procedure was applied to the simulation model of the mobile concrete pump.

First, a fine tuning of the mass distribution was obtained by means of static experiments. For this purpose, the cylinder forces respectively the chamber pressures and the inclination at the beginning of each boom segment were measured at different angles  $\varphi_1 = \bar{\varphi}_{1,m}$  of the first joint and a sprawled boom, i.e.,  $\varphi_2 = \varphi_3 = \varphi_4 = 0$ . For the prescribed constant angles  $\bar{\varphi}_{1,m}$ , the corresponding measured cylinder forces and inclinations are referred to as  $\bar{F}_{i,m}(\bar{\varphi}_{1,m})$  and  $\bar{\psi}_{b,i,m}(\bar{\varphi}_{1,m})$ ,  $i = 1, \dots, N$ . In order to compare the measurement results with the simulation model, the stationary solution of Equations (20) and (22) is determined, i.e.,  $\mathbf{g}(\bar{\mathbf{q}}) = \bar{\mathbf{Q}}$  and, given  $\varphi_2 = \varphi_3 = \varphi_4 = 0$ , solved for  $\bar{h}_1^b, \dots, \bar{h}_N^b, \bar{h}_1^o, \dots, \bar{h}_N^o$  and  $\bar{F}_1, \dots, \bar{F}_N$  as a function of  $\bar{\varphi}_1$ . In general, the inclination at the beginning of the boom segment reads as

$$\psi_{b,i} = \sum_{n=1}^i \varphi_n + \sum_{n=1}^{i-1} w'_n(L_n). \quad (39)$$

Thus, for  $\varphi_2 = \varphi_3 = \varphi_4 = 0$  and  $w_n(x_n)$  according to Equation (17), the stationary relation of Equation (39) takes the form

$$\bar{\psi}_{b,i} = \bar{\varphi}_1 + \sum_{n=1}^{i-1} \left( \bar{h}_n^b (\eta_n^b)'(L_n) + \bar{h}_n^o (\eta_n^o)'(L_n) \right). \quad (40)$$

With Equation (40), the stationary cylinder forces  $\bar{F}_i, i = 1, \dots, N$ , can be parametrized in terms of the respective inclination, i.e.,  $\bar{F}_i(\bar{\psi}_{b,i})$ . Now the nominal mass distribution of the boom segments  $\sigma_i(x_i), i = 1, \dots, N$  is scaled with a multiplicative constant  $\xi_i, i = 1, \dots, N$ , which serves as a parameter to be identified. For this, the least squares problem

$$\min_{\xi_i} \left( \bar{F}_i(\bar{\psi}_{b,i,m}) - \bar{F}_{i,m}(\bar{\psi}_{b,i,m}) \right)^2 \quad (41)$$

has to be recursively solved beginning at the last boom segment  $i = N$ .

In order to identify the stiffness and the damping behaviour of the boom, dynamic experiments with a defined excitation of the system were performed in the next step. With the measurements of the cylinder forces, the strain gauges and the marker movements of the machine vision system, the mean equivalent Young's modulus  $E$  as well as the damping coefficients  $d_{\varphi_i}$  and  $d_{w_i}$  were determined such that the simulation model matches the occurring oscillation frequencies and the decay of the envelopes. The strain gauge coefficients  $k_{SG,i}$  and  $b_{SG,i}$  were also identified by means of static measurements, where a similar procedure as for the adjustment of the mass distribution was employed.

## 5. Experimental results

The following subsection presents the validation of the simulation model by means of experimental results. The numerically calculated basis functions (14) and (15), employed in the simulation model, are illustrated in Figure 11. In Figure 12, the results of static measurements in comparison to the simulation model are given. The cylinder forces and beam flections have been measured at different angles  $\varphi_1$  of the first joint and a sprawled boom with joint angles  $\varphi_2, \varphi_3$  and  $\varphi_4$  close to 0. The cylinder forces are normalized to the

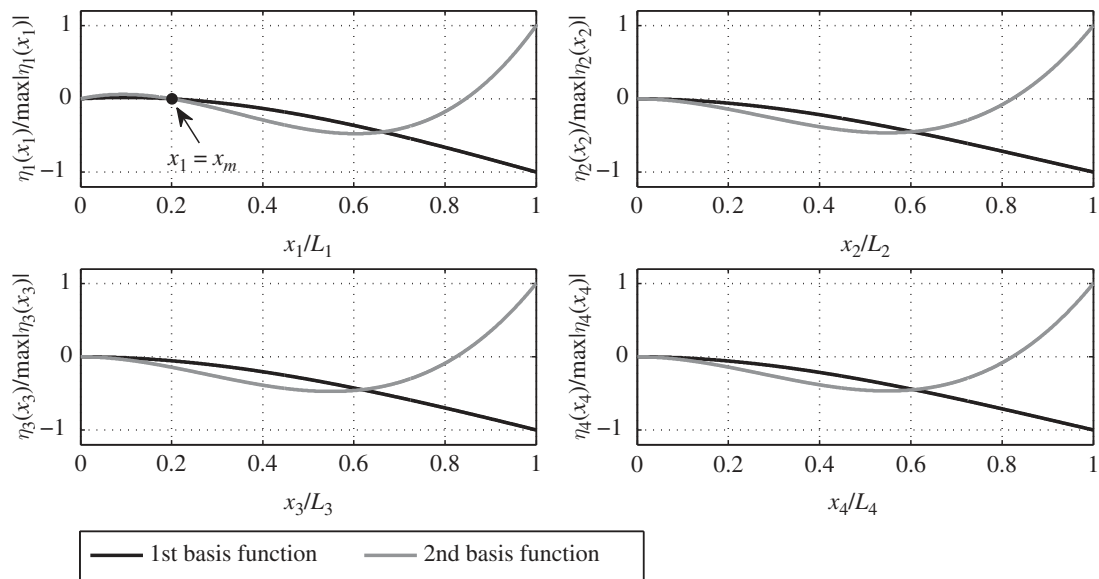


Figure 11. Basis functions of the boom segments normalized to the beam length and to the maximum function value.

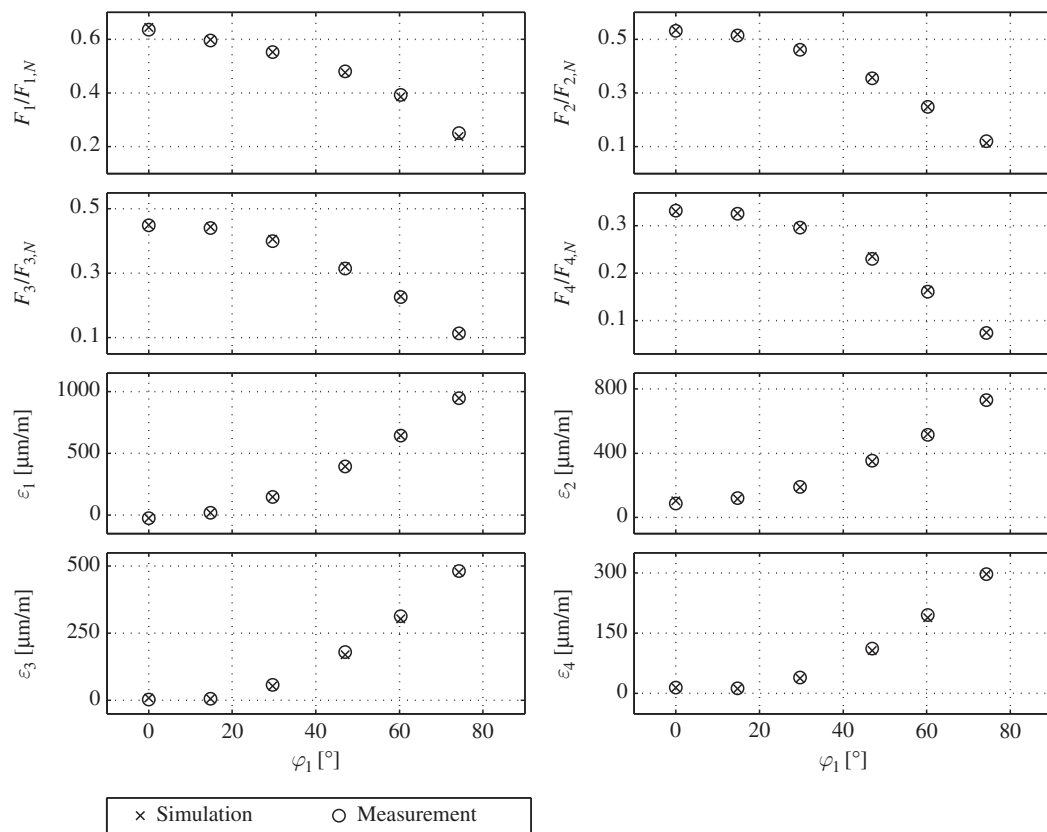


Figure 12. Static measurements at a sprawled boom: cylinder forces normalized to the maximum forces and strain gauge measurements as a function of the first joint angle.

maximum forces, which are determined by the supply pressure and the piston area,  $F_{i,N} = p_s A_{1,i}$ .

In order to validate the dynamic system behaviour, the system was excited by a sudden release of a 90 kg load at the end of the boom. In Figure 13, the comparison of the measured and simulated response of the normalized cylinder forces, the strain gauge signals and relative movements of the cylinder pistons at a typical boom configuration on the construction site is shown, see Figure 1. The stationary deviations of the strain gauge signals can be explained by thermal expansion due to different experiment days.<sup>7</sup>

Finally, in Figure 14 the measurement results of the machine vision system are presented.<sup>8</sup> The simulated and tracked movement of the markers, described in the inertial frame, is illustrated. It can be seen that the amplitudes of the oscillations agree very well. With this, an essential requirement for the mathematical model, which serves as a basis for the design of active damping control strategies, is fulfilled. Looking at the absolute deviations to the inertial frame, one has to bear in mind that already small deviations of the angle  $w'_1(L_1)$  at the end of the first boom segment have a big influence on the absolute marker positions. Furthermore, due to lens distortions at the outer image section as well as the restricted validity of the assumption that all markers move in the same plane, the precision of the absolute machine vision measurements is limited. However, considering the size of the boom, the biggest vertical and horizontal deviations of about 60 cm for  $y_{m,4}$  and 15 cm for  $x_{m,3}$  can be considered small in relation to the horizontal range of 27 m.

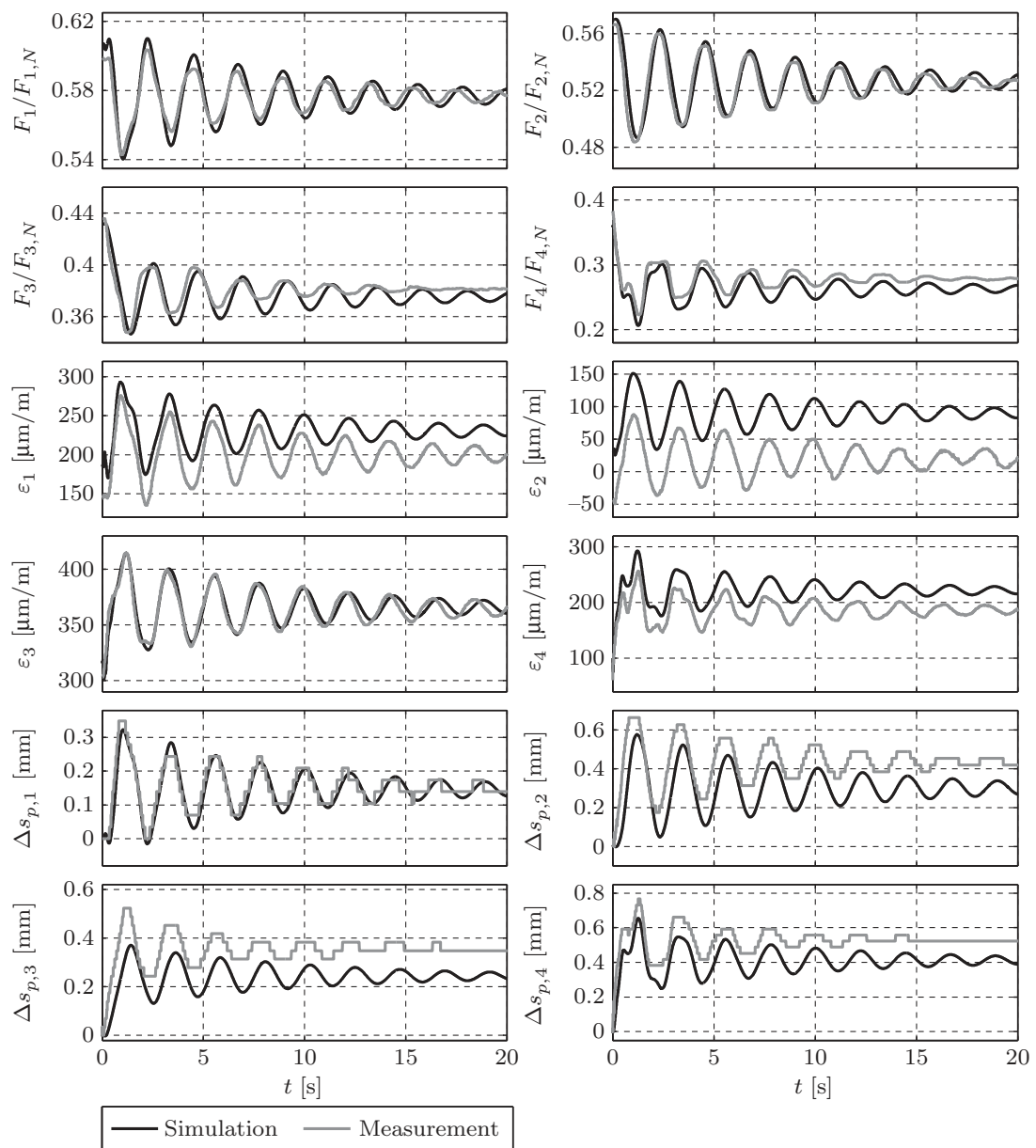


Figure 13. Load release at the end-effector: cylinder forces, strain gauge signals and cylinder piston movements.

Since in view of the envisaged controller design the main goal of the model is to capture the essential dynamic behaviour of the mobile concrete pump, the impreciseness in terms of the absolute deviations is of minor importance.

Figure 14 shows the relatively high resolution and low noise level of the signals, taking into account the large work space. Figure 15 depicts the relative vertical movement of a marker attached to the slewing gear. It can be seen that the vision system is even able to capture the 2 mm amplitude of the reaction of the truck caused by the sudden release of the load. With the use of the effective ellipse search algorithm, the image processing time for each frame is far lower than the sampling time defined by the frame rate. Hence, it would even be possible to use this system for real-time applications.

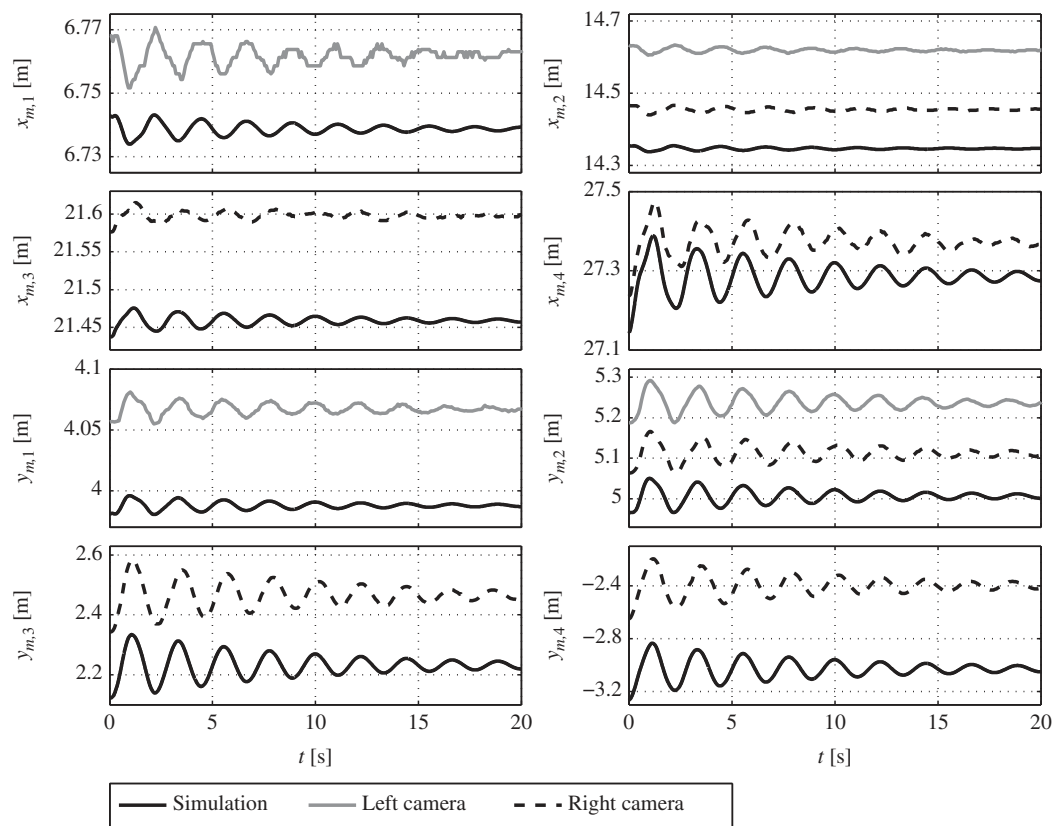


Figure 14. Load release at the end-effector: horizontal and vertical distance of the markers to the inertial frame.

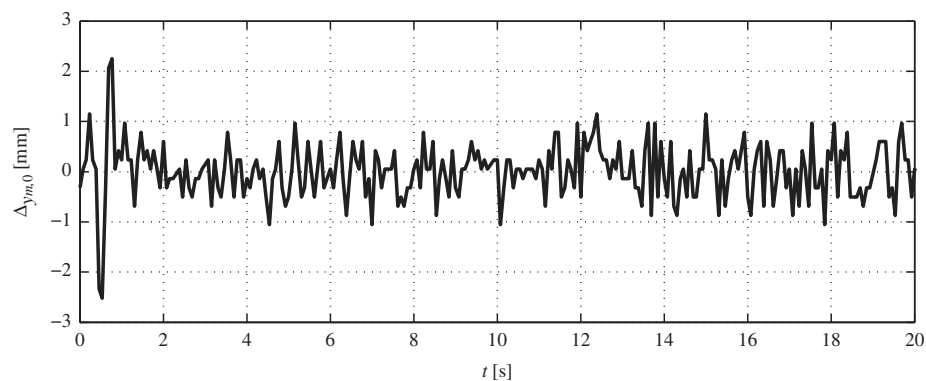


Figure 15. Load release at the end-effector: relative vertical movement of a marker attached to the truck.

## 6. Conclusions

In this paper, a systematic approach for the mathematical modelling of large-scale lightweight manipulators with hydraulic actuation was proposed. It was shown that a suitable choice of the basis functions for the approximation of the elastic bending of the beams has significant advantages for the incorporation of static friction into the model. Furthermore, simulation and measurement results of a mobile concrete pump

show the accuracy and the feasibility of the proposed modelling approach already for a small number of basis functions. A procedure for the identification of the uncertain model parameters is given. For this purpose, a tailored machine vision system was presented for the measurement of the absolute boom movement. It was demonstrated that this system enables the tracking of motions with amplitudes of approximately 2 mm even for a large observation area.

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### Supplemental data

Supplemental data for this article can be accessed [here](#).

### Notes

1. VICON: <http://www.vicon.com>
2. For the sake of readability, the time dependence of the degrees of freedom is not explicitly stated in this paper.
3. Allide Vision: Pike F-505, <http://www.alliedvisiontec.com>
4. ROS (Robot Operating System): <http://www.ros.org>
5. OpenCV (Open Source Computer Vision): <http://opencv.willowgarage.com/wiki/>
6. ROS Camera Calibration: [www.ros.org/wiki/camera\\_calibration](http://www.ros.org/wiki/camera_calibration)
7. With the linear thermal expansion coefficient for steel of  $\alpha_{T,Steel} \approx 11 \dots 13 \times 10^{-6} \text{K}^{-1}$  a temperature difference over night of 10 K yields an expansion of about 110 ... 130  $\mu\text{m/m}$ .
8. A video illustrating the capturing of the vertical motion of the markers is available on [http://www.acin.tuwien.ac.at/fileadmin/cds/videos/Motion\\_Capturing\\_Mobile\\_Concrete\\_Pump.mp4](http://www.acin.tuwien.ac.at/fileadmin/cds/videos/Motion_Capturing_Mobile_Concrete_Pump.mp4)

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