Topological synthesis of fluidic pressure-actuated robust compliant mechanisms

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Abstract: This paper presents a robust density-based topology optimization approach for synthesizing pressure-actuated compliant mechanisms. To ensure functionality under manufacturing inaccuracies, the robust or three-field formulation is employed, involving dilated, intermediate and eroded realizations of the design. Darcy’s law in conjunction with a conceptualized drainage term is used to model the pressure load as a function of the design vector. The consistent nodal loads are evaluated from the obtained pressure field using the standard finite element method. The objective and load sensitivities are obtained using the adjoint-variable approach. A multi-criteria objective involving both the stiffness and flexibility of the mechanism is employed in the robust formulation, and min-max optimization problems are solved to obtain pressure-actuated inverter, gripper, and contractor compliant mechanisms with different minimum feature sizes. Limitations of the linear elasticity assumptions while designing mechanisms are identified with high pressure loads. Challenges involved in designing finite deformable pressure-actuated compliant mechanisms are presented.

Keywords: Pressure-driven compliant mechanisms; Soft robots; Robust formulation; Geometric nonlinearity; Follower forces; 3D-Printing

1 Introduction

Compliant mechanisms (CMs) are established concepts in industry and academia offering various advantages over traditional linkage-based counterparts, e.g., less wear and tear, low manufacturing and assembly cost, repeatability and high precision, lack of frictional losses, to name a few. Due to such promising advantages, their usage is continuously rising in a wide variety of applications (Ananthasuresh, 2021; Frecker et al., 1997; Howell, 2001; Kumar et al., 2019a, 2021b). These mechanisms characterized via monolithic designs are termed CMs, since their functionality arises from the elastic deformations of their flexible (compliant) members in response to the input forces. Finding the optimum balance between output deformation and stiffness when designing CMs is a nontrivial task (Cao et al., 2018). Topology optimization

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2This pdf is the personal version of an article whose final publication is available at Mechanism and Machine Theory
(TO) has been shown to be an effective approach for designing such mechanisms (Zhu et al., 2020). TO is a computational design technique able to achieve the optimized material distribution within a given design domain by extremizing the conceptualized (desired) objectives under a known set of physical and geometrical constraints (Sigmund and Maute, 2013). In a general structural setting, the design domain is parameterized using finite elements (FEs). Each FE is assigned a material density design variable $\rho_i \in [0, 1]$. Here, $\rho_i = 1$ and $\rho_i = 0$ represent solid and void states of the $i$th FE, respectively. Ideally, FEs with $\rho = 1$ should constitute the optimized CMs.

Actuating forces of CMs can be recognized as either design-dependent, e.g., pneumatic, hydraulic pressure loads, or design-independent, e.g., constant forces. Design-dependent pressure loads alter their magnitude, location and/or direction as the design boundary on which they act evolves during the TO process. Consequently, such loads pose many challenges for the TO formulation, e.g., locating a valid surface to apply the loads, relating pressure field to the design vector, and evaluating consistent nodal forces and their sensitivities with respect to the design vector (Kumar et al., 2020).

**Pressure-actuated CMs (Pa-CMs)** constitute a relatively novel category of mechanisms that find application in e.g. pneumatically or hydraulically actuated soft robots (mechanisms) (Chen et al., 2018; Luo et al., 2020). Note however that, in order to attain maximum flexibility, CMs designed via classical TO are prone to exhibit single-node-connected hinges and gray density FEs ($0 < \rho < 1$) in the optimized designs (Sigmund, 1997; Yin and Ananthasuresh, 2003). Such features cannot be realized, and in order to render the design manufacturable, post-processing of the optimized geometry is necessary which can severely affect the optimized performance. This challenge is found in Pa-CMs designed by TO as well, where it can be even more detrimental given the close relation between boundary shape and loading. A schematic figure of a Pa-CM is depicted in Fig. 1 wherein the pressure loading boundary moves from its initial surface $\Gamma_p$ to the final (optimized) surface $\Gamma_p^b$. Furthermore, Fig. 1 depicts a single-node hinge and region with gray FEs. Given the negative effect such features have on the manufacturability and performance of Pa-CMs, it is important to control and avoid them. The need to generate Pa-CM designs whose actual performance closely matches the simulated optimized performance forms the motivation for the present study.

2 Background and Approach

Hammer and Olhoff (2000) were first to present an approach to design pressure-loaded structures. A fictitious thermal loading setting was exploited to solve pressure-loaded design problems by Chen and Kikuchi (2001). Sigmund and Clausen (2007) used the mixed-finite element method Zienkiewicz and Taylor (2005) with a three-phase material (solid, void, fluid) formulation in their approach. The mixed-finite element approaches require satisfaction of the Babuska-Brezzi condition for the stability in the FE analysis Zienkiewicz and Taylor (2005). Chen et al. (2001) employed the approach presented in Chen and Kikuchi (2001) to design Pa-CMs. Panganiban et al. (2010) used a nonconforming FE method which is not a standard FE method. Vasista and Tong (2012) employed the solid isotropic material with penalization (SIMP) and the moving isosurface threshold methods in their approach. de Souza and Silva (2020) also employed the method proposed by Sigmund and Clausen (2007). In our previous study Kumar et al. (2020), the authors presented a design method using Darcy’s law in conjunction with a drainage term. In that work we also demonstrated the importance of load sensitivities in designing of the Pa-CMs. The method proposed in Kumar et al. (2020) uses the standard FE

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3We henceforth for brevity write pressure loads instead of design-dependent pressure loads.
formulation, provides consistent sensitivities and was found to work well in generating Pa-CMs for two- and three-dimensional as well (Kumar and Langelaar, 2021). Therefore, it will also be used in the present study.

For the design optimization process to be useful and reliable, it is important that as-fabricated Pa-CMs perform similar to the prediction made by the numerical simulation used in the optimization process. However, with existing methods a significant decline in actual performance can arise compared to the numerical predictions. This is primarily due to three factors: (A) inaccurate and/or approximate conversion of one-node-connected hinges to thin-flexible regions (Fig. 2a) and inaccuracies introduced by unrealistic representation of thin, flexible regions in FE models used in TO, (B) CMs being overly sensitive to manufacturing inaccuracies or arbitrariness in design extraction (Fig. 2b), and (C) the use of small displacement analysis and linear elasticity assumptions for the Pa-CM designs.

Firstly, one-node-connected hinges (Fig. 2a), artificially stiff locations, appear due to deficiencies in the FE analysis with quadrilateral FEs that permit load transfer with zero rotational stiffness (Sigmund, 1997; Yin and Ananthasuresh, 2003). Such hinges pose challenges in accurate design interpretation of the optimized mechanisms, since real compliant hinges will always have a finite rotational stiffness (Sigmund, 1997; Yin and Ananthasuresh, 2003). One of the methods for approximating a one-node-connected location for fabrication may be as depicted in Fig. 2a, which results in a thin-flexible region and thus, the performance of the numerical design (Fig. 2a-a1) will differ from that of the fabricated one (Fig. 2a-a2). Various ways have been proposed to prevent formation of single-node-connected hinges in CMs (Cao et al., 2015a; Poulsen, 2003; Saxena and Saxena, 2007; Wang et al., 2011; Yin and Ananthasuresh, 2003), but these have not yet been applied to and evaluated for Pa-CM TO.

Secondly, in a standard density-based TO setting with a gradient-based optimizer, it is difficult to obtain pure 0-1 solutions (Fig. 1). Therefore, extraction of the optimized designs based on the considered density threshold (Fig. 2b) is required, which invariably alters the final designs and thus, the performances with respect to the numerical predictions. Fig. 2b illustrates a
scenario to indicate how the different threshold material densities lead to different material (contour) layouts (Fig. 2b-b1 and Fig. 2b-b2) and thus, the corresponding fabricated Pa-CM and potentially also its loading (significantly) differs from that obtained via TO.

Thirdly, a typical Pa-CM or CM may experience large deflection and also, contact between branches, i.e., self-contact (Kumar et al., 2017) and external/mutual contact (Kumar et al., 2016, 2019b) thus, TO design approaches must include nonlinear mechanics (with contact formulation) to predict mechanism performance (Kumar et al., 2016, 2019b). However, nonlinear structural analysis poses various challenges in TO (van Dijk et al., 2014; Kumar et al., 2021b), which can even get more pronounced in combination with pressure loads whose magnitude, direction and/or location vary and follow the surfaces/facets where they are applied. To model the characteristics of pressure loads, one needs to include the follower force concepts in the design approach (Zienkiewicz and Taylor, 2005), which demands a dedicated and in-depth investigation within a TO setting, which is out of the scope of this paper. Therefore, instead of addressing this point at the TO stage, we choose to investigate and assess its influence based on the Pa-CM designs generated using linear modeling.

To address the outlined challenges, in this study we take the following approaches:

1. The robust or three-field formulation (Wang et al., 2011) is adopted to address problems A (approximate conversion of one-node-connected hinges to thin-flexible regions) and B (CMs being overly sensitive to manufacturing inaccuracies or arbitrariness in design extraction), by its ability to impose a minimum length scale and to reduce the sensitivity of the final design to post-processing or manufacturing errors.

2. Nonlinear FEA is used to analyze optimized Pa-CM designs with a neo-Hookean hyper-elastic material model and increased pressure loads to investigate the large deformation behavior and to determine the limitations of the linear elastic assumptions.

While the robust formulation (Wang et al., 2011) is expected to solve factors A and B given results reported in literature (Wang et al., 2011), this has as of yet not been investigated or confirmed for Pa-CM, which is the specific interest of this study. Next to this, the nonlinear FEA i.e. geometric nonlinearity (due to finite deformation in the mechanisms) and material nonlinearity (as rubber-like materials are typically used to fabricate such mechanisms, cf. (Schmitt et al., 2018)) study is not intended as a solution to the lack of large-displacement analysis during Pa-CM TO, but our aim is to provide a quantitative assessment of the severity of the
error introduced by this simplification, in representative design cases and also, the challenges involved in designing finite-deformable Pa-CMs. In addition, nonlinear finite element analyses are used to assess the sensitivity of standard and robust Pa-CMs to design extraction choices/manufacturing errors. Due to computational limitations and for clarity of presentation this study has been carried out in a 2D setting. Its findings nonetheless are expected to apply to general 3D cases as well.

In particular, the current paper offers following new aspects:

- A robust topology optimization approach to optimize design-dependent fluidic Pa-CMs. The Darcy law in conjunction with the conceptualized drainage term (Kumar et al., 2020) is employed for the pressure field modeling, whereas material layout modeling is performed using the three-field (dilated, intermediate and eroded fields) formulation (Wang et al., 2011).

- Demonstration of the robustness and efficacy of the proposed method by designing various pressure-driven compliant mechanisms, e.g., inverter, gripper, and contractor mechanisms. The optimized compliant mechanisms contain a large space for pressure to inflate them like soft robots, which makes the optimized mechanisms uniquely different from the other pressure-actuated compliant mechanisms previously designed by topology optimization.

- Assessment of the optimized Pa-CMs obtained via linear elastic assumptions under high pressure loads. Several challenges are demonstrated and discussed for designing finite deformation fluidic Pa-CMs.

- A method to extract the optimized designs to facilitate CAD modeling for further analyses and 3D printing/fabrication.

This paper is organized as follows. Sec. 3 presents modeling of the pressure loads as a function of the design vector using the Darcy law with a drainage term, in line with Kumar et al. (2020). Using a transformation matrix, the consistent nodal loads are evaluated. The TO formulation with the robust approach together with the corresponding sensitivity analysis is described in Sec. 4. Next, in Sec. 5 numerical examples of designing robust pressure-actuated inverter and gripper mechanisms are presented. The optimized Pa-CM designs are extracted, and nonlinear FE analyses are performed in ABAQUS with high pressure loads to investigate large deformation behavior of the CMs. Lastly, conclusions are drawn in Sec. 7.

3 Design-dependent pressure load modeling

In this section, modeling of the pressure field as a function of the design variables, finite element formulation and consistent nodal loads evaluation are summarized for the sake of self-consistency. For a detailed description, we refer to our previous paper (Kumar et al., 2020).

The material boundaries of a given design problem evolve as TO progresses. Thus, it becomes challenging especially at the beginning of the optimization to locate an appropriate boundary to apply the fluidic pressure load. In addition, a design-dependent and continuous pressure field are expected to help the TO process. Further, at the initial stage of the optimization, one can consider each element as a porous medium, and boundaries with the prescribed input pressure and zero pressure loads are already provided. Therefore, the Darcy law is adopted herein to model the pressure field wherein the flow coefficient of each element is interpolated using a smooth Heaviside function (Kumar and Langelaar, 2021; Kumar et al., 2020). As per
where $H$ (Kumar et al., 2020) solid and void phases interpolated by a smooth Heaviside projection function control the values of the medium and the fluid viscosity $\mu$ respectively. In a typical density-based TO setting, an FE displays two states, therefore the are determined using a transformation matrix $T$ where the drainage coefficient $D$ is the ratio of input pressure at depth $\Delta s$, i.e., $p|_{\Delta s} = rp_{in}$ and the penetration depth $\Delta s$ can be set equal to the width or height of a few FEs. Using $Q_{\text{drain}}$, Eq. 1 transpires per Kumar et al. (2020) as

$$\nabla \cdot q - Q_{\text{drain}} = 0.$$  

In a discrete FE setting, one writes the weak form of Eq. 5 for an FE with domain $\Omega_e$ as (Kumar et al., 2020)

$$\int_{\Omega_e} \left( K \nabla p \cdot \nabla p + D N_p^T N_p \right) d\Omega_e \quad p_e$$

$$\sum_{\Gamma_e} D N_p^T p_{\text{ext}} \cdot n_e \, d\Gamma_e - \int_{\Gamma_e} N_p^T q \cdot n_e \, d\Gamma_e$$

where $p_e$ is the pressure field to be evaluated, $B_p = \nabla N_p$ with $N_p = [N_1, N_2, N_3, N_4]$ are the bi-linear shape functions for a quadrilateral FE. $A_p = f$ is the global form of Eq. 6 with $p$ is the global pressure load vector. In this work, $p_{\text{ext}}$ and $q$ are set to zero, therefore $f = 0$, i.e., $A_p = 0$. Using the obtained global pressure field $p$, the consistent global nodal forces $F = -T_p$ are determined using a transformation matrix $T$ whose elemental form $T_e$ is related to that of nodal force $F_e$ as (Kumar et al., 2020)

$$F_e = T_e p_e = -\int_{\Omega_e} N_u^T B_p d\Omega_e p_e.$$
where \( \mathbf{N}_u = [N_1 \mathbf{I}, N_2 \mathbf{I}, N_3 \mathbf{I}, N_4 \mathbf{I}] \) with \( \mathbf{I} \) the identity matrix in \( \mathbb{R}^2 \). In summary, one employs Eq. 6 for determining the pressure field, whereas the corresponding consistent nodal force vector for an FE is evaluated using Eq. 7. Note that through the use of smooth Heaviside functions the loads are a differentiable function of the density variables. This allows performing load sensitivity analysis readily, as detailed in Kumar et al. (2020) and further elaborated in Sec. 4.2.

4 Topology optimization formulation

The three-field \((\rho, \tilde{\rho}, \bar{\rho})\) representation of the design domain is considered (Lazarov et al., 2016). The filtered design variable \( \tilde{\rho}_j \) of element \( j \) is determined using weighted average of the design variables \( \rho \) pertaining to neighboring FEs lying within a circular region of radius \( r_{\text{fill}} \) (Bruns and Tortorelli, 2001). Mathematically,

\[
\tilde{\rho}_j = \frac{\sum_{k=1}^{N_e} v_k \rho_k w(x_k)}{\sum_{k=1}^{N_e} v_k w(x_k)}
\]

where \( N_e \) is the total number of neighboring elements of the \( j^{th} \) FE, and \( v_k \) is the volume of neighboring element \( k \). The weight function \( w(x_k) = \max(0, 1 - \frac{d}{r_{\text{fill}}}) \), wherein \( d = ||x_j - x_k|| \) is a Euclidean distance between centroids \( x_j \) and \( x_k \) of elements \( j \) and \( k \), respectively. \( r_{\text{fill}} \) is called filter radius for the considered design problems. The derivative of \( \tilde{\rho}_j \) (Eq. 8) with respect to \( \rho_k \) is

\[
\frac{\partial \tilde{\rho}_j}{\partial \rho_k} = \frac{v_k w(x_k)}{\sum_{i=1}^{N_e} v_i w(x_i)}.
\]

The physical design variable \( \bar{\rho}_j \) is defined as (Wang et al., 2011)

\[
\bar{\rho}_j(\tilde{\rho}_j, \beta, \eta) = \frac{\tanh (\beta \eta) + \tanh (\beta (\tilde{\rho}_j - \eta))}{\tanh (\beta \eta) + \tanh (\beta (1 - \eta))},
\]

where \( \beta \in [0, \infty) \) and \( \eta \in [0, 1] \) control the steepness and the threshold of the projection function, respectively. To achieve the optimized solutions close to black and white designs, typically \( \beta \) is increased from an initial value \( \beta_{\text{int}} = 1 \) to a maximum value \( \beta_{\text{max}} \) using a continuation strategy. \( \eta = 0 \) ensures the minimum length scale on the solid phase (Wang et al., 2011), whereas that of solid phase is obtained using \( \eta = 1 \). Note that when using \( \eta = 0 \) and \( \eta = 1 \), Eq. 10 yields the Heaviside step approximation function given in Guest et al. (2004) and the modified Heaviside step approximation function mentioned in Sigmund (2007), respectively. The derivative of \( \bar{\rho}_j \) with respect to \( \tilde{\rho}_j \) is

\[
\frac{\partial \bar{\rho}_j}{\partial \tilde{\rho}_j} = \frac{1 - \tanh(\beta (\tilde{\rho}_j - \eta))^2}{\tanh (\beta \eta) + \tanh (\beta (1 - \eta))}.
\]

Having noted the derivatives in Eqs. 9 and 11, the chain rule is used to determine the derivatives of a function \( f \) with respect to \( \rho_k \) as

\[
\frac{\partial f}{\partial \rho_k} = \sum_{j=1}^{N_e} \frac{\partial f}{\partial \tilde{\rho}_j} \frac{\partial \tilde{\rho}_j}{\partial \rho_k},
\]

where \( \frac{\partial f}{\partial \tilde{\rho}_j} \) is evaluated using the adjoint-variable method (see Sec. 4.2). We use the modified SIMP (Simplified Isotropic Material with Penalization) method to interpolate the Young’s modulus of each FE using its physical design variable \( \tilde{\rho}_j \) as

\[
E_1(\tilde{\rho}_j) = E_0 + (\tilde{\rho}_j)^\zeta (E_1 - E_0), \quad \tilde{\rho}_j \in [0, 1]
\]
where \( E_1 \) and \( E_0 \) are Young’s moduli of the solid and void phases of an FE, respectively. The material contrast, i.e., \( \frac{E_1}{E_0} = 10^{-6} \) is set, and the penalty factor \( \zeta = 3 \) is used in order to steer the topology optimization towards a ‘0-1’ solution.

### 4.1 Robust formulation

The robust formulation is employed wherein three physical density fields, i.e., dilated \( \bar{\rho}^d \), intermediate (blueprint) \( \rho^i \) and eroded \( \rho^e \), are considered for the design domain (Wang et al., 2011). Erosion and dilation are morphological image operators, which can be used in TO for e.g. robustness and feature size control (Sigmund, 2007). Assuming uniform manufacturing errors, maximum and minimum manufacturing limits are indicated by the dilated and eroded designs respectively, whereas the intermediate (blueprint) ones denote the desired manufacturing limit. Here, \( 0.5 + \Delta \eta \), 0.5 and \( 0.5 - \Delta \eta \) in Eq. 10 are used in place of \( \eta \) to evaluate \( \rho^e \), \( \rho^i \) and \( \rho^d \), respectively. The deviation \( \Delta \eta \in [0, 0.5] \) is a user defined parameter, which in combination with the filter radius \( r_{fil} \) determines the minimum length scale on the solid and void phases (Trillet et al., 2021).

The optimization problem is formulated as a min-max problem (Wang et al., 2011)

\[
\begin{align*}
\min_{\rho} : & \max \left( f_0(\bar{\rho}^d(\rho)), f_0(\rho^i(\rho)), f_0(\rho^e(\rho)) \right) \\
\text{s.t.} : & A(\rho^l)\mathbf{p}(\rho^l) = 0, \quad l = d, i, e \\
& K(\rho^l)u(\rho^l) = F = -Dp(\rho^l) \\
& V(\rho^i(\rho)) - V^*_d \leq 0 \\
& 0 \leq \rho \leq 1 
\end{align*}
\]

(14)

where \( f_0 \) is a multi-criteria objective aimed at obtaining effective compliant mechanisms (Saxena and Ananthasuresh, 2000) defined by \( -\mu \frac{MSE}{SE} \), with \( MSE = v^\top K u \) and \( SE = \frac{1}{2} u^\top K u \). \( MSE \) and \( SE \) represent the mutual strain energy and strain energy of the mechanism, respectively. \( \mu \) is the scaling factor used to scale the objective for optimizer compatibility. Note that the multi-criteria objective proposed by Frecker et al. (1997) finds an optimal trade-off between the flexibility and stiffness of the mechanisms. \( V(\rho^i(\rho)) = \sum_{m=1}^{n_e} V_m \rho^i_m \), where \( V_m \) is the volume of \( m \)th element whose dilated density is \( \rho^i_m \). The volume constraint is imposed using the dilated design wherein the actual volume of the dilated design is updated after a specific number of optimization iterations such that the volume of the intermediate design becomes equal to the permitted one at the end of the optimization when the volume constraint becomes active (Wang et al., 2011). Further, \( V^*_d = \frac{V^*_e}{V(\rho^e(\rho))} V(\rho^i(\rho)) \), where \( V^*_e \) denotes upper limit of the volume fraction of the dilated design, \( V^*_e \) and \( V(\rho^i(\rho)) = \sum_{m=1}^{n_e} V_m \rho^i_m(\rho) \) are the prescribed and actual volumes of the intermediate design, respectively.

The robust formulation (Eq. 14) requires solutions to three state equations pertaining to \( u, p, v \) fields and also furnishes three optimized designs with only one design vector \( \rho \). Readers may refer to the paper by Trillet et al. (2021) for a complete discussion on the minimum feature size with the three-field design representation technique. The discreteness of the optimized solutions is measured using a gray scale indicator \( M_{nd} \) defined as (Sigmund, 2007)

\[
M_{nd} = \frac{\sum_{e=1}^{n_e} 4(\bar{\rho}_e)(1 - \rho_e)}{n_e},
\]

(15)

where \( n_e \) is the total number of elements employed to discretize the design domain.
Figure 3: Symmetric half design domains of the mechanisms. (a) Inverter design domain, (b) Gripper design domain and (c) Contractor design domain. The dimensions of the design domains are $L_x = 0.2$ m and $L_y = 0.1$ m, where $L_x$ and $L_y$ indicate dimensions in $x-$ and $y-$directions, respectively. For the inverter and gripper mechanisms, the input pressure load is applied on the left edge, and zero pressure is applied at the other edges except the symmetric boundary. The contractor mechanism is actuated from both the left and right edges. Symmetric boundaries and fixed boundaries are also depicted. Workpiece stiffnesses are represented via the output springs with stiffnesses $k_{ss}$. In gripper mechanism domain, a non-design void region having area $L_x \times L_y$ and a solid region having area $L_x \times L_y$ are used at the right lower part. A solid non-design domain of size $L_x \times L_y$ is present in the middle of the symmetric half domain of the contractor mechanism.

Table 1: Various parameters used in this paper.

| Parameter                                         | Notation | Value                                      |
|---------------------------------------------------|----------|--------------------------------------------|
| Young’s modulus of actual material                 | $E_1$    | $3 \times 10^9$ N m$^{-2}$                |
| Poisson’s ratio                                    | $\nu$    | 0.40                                       |
| Out-of-plane thickness                             | $t$      | 0.001 m                                    |
| Penalization                                       | $\zeta$  | 3                                           |
| Young’s modulus of a void FE ($\rho = 0$)          | $E_0$    | $E_1 \times 10^{-6}$ N m$^{-2}$            |
| External move limit$^2$                            | $\Delta \rho$ | 0.1 per iteration                          |
| Input pressure load                                | $p_{in}$ | $1 \times 10^5$ N m$^{-2}$                |
| $K(\rho)$ step location                           | $\eta_k$ | 0.3                                        |
| $K(\rho)$ slope at step                            | $\beta_k$ | 10                                         |
| $H(\rho)$ step location                           | $\eta_h$ | 0.2                                        |
| $H(\rho)$ slope at step                            | $\beta_h$ | 10                                         |
| Flow coefficient of a void FE                      | $k_v$    | $1$ m$^4$ N$^{-1}$ s$^{-1}$                |
| Flow coefficient of a solid FE                     | $k_s$    | $k_v \times 10^{-7}$ m$^4$ N$^{-1}$ s$^{-1}$|
| Drainage from solid                                | $h_s$    | $(\ln r)^2$ k$\_1$                         |
| Remainder of input pressure at $\Delta s$         | $r$      | 0.1                                        |

$^2$ The external move limit is used to update $\mathbf{x}_{\text{minvec}}$ and $\mathbf{x}_{\text{maxvec}}$ of the MMA outside of the $\text{mmasub}$ function call.

Figure 4: Color schemes employed in this paper to plot material, pressure and displacement fields are shown in (a), (b) and (c) respectively. $p_{max} = 1$ bar and $p_{min} = 0$ bar represent the maximum and minimum values of the pressure load. Maximum and minimum values of the magnitude of displacement field are indicated via $u_{min} = 0$ and $u_{max}$ respectively.
0.25 \leq \rho \to 11 \leftarrow \rho \geq 0.75 \text{ (a)(a1) (a2)}

Figure 5: An optimized pressure-actuated inverter mechanism obtained using the method presented in Kumar et al. (2020) is displayed in (a). The optimized design contains thin flexure regions surrounded by gray elements that are depicted in insets. Using the two different thresholds, the approximated designs are displayed in (a1) and (a2).

4.2 Sensitivity analysis

We use the Method of Moving Asymptotes (MMA) (Svanberg, 1987), a gradient-based optimizer, for solving the optimization problem (Eq. 14). A standard setting available in the MMA optimizer is used to solve the min-max optimization problem. The Lagrangian $L$ using the objective function and constraints can be written as

$$L = f_0(\bar{\rho}) + \lambda_1^T (Ku + Hp) + \lambda_2^T (Ap) + \lambda_3^T (Kv - F_d) + \Lambda (V - V^*_d), \quad (16)$$

where $\lambda_i_{i=1,2,3}$ and $\Lambda$ are the Lagrange multipliers. Using the adjoint equations corresponding to Eq. 16, i.e., $\frac{\partial C}{\partial u} = 0$, $\frac{\partial C}{\partial p} = 0$, and $\frac{\partial C}{\partial v} = 0$, one finds the Lagrange multipliers $\lambda_1$, $\lambda_2$ and $\lambda_3$ as (Kumar et al., 2020)

$$
\begin{align*}
\lambda_1^T &= -\frac{\partial f_0(u,v,\rho)}{\partial u}K^{-1} \\
\lambda_2^T &= -\lambda_1^T HA^{-1} \\
\lambda_3^T &= -\frac{\partial f_0(u,v,\rho)}{\partial v}K^{-1}
\end{align*}
\tag{17}
$$

The design equation corresponding to Eq. 16 is

$$\frac{\partial L}{\partial \rho} = \frac{\partial f_0}{\partial \rho} + \lambda_1^T \frac{\partial K}{\partial \rho} u + \lambda_2^T \frac{\partial A}{\partial \rho} p + \lambda_3^T \frac{\partial K}{\partial \rho} v + \Lambda \frac{\partial (V(\bar{\rho}^d(\rho)) - V^*_d)}{\partial \rho}, \quad (18)$$
We refer the inverter mechanisms in the first, second, third and fourth rows as IV1 Pa-CMs, IV2 Pa-CMs, IV3 Pa-CMs and IV4 Pa-CMs respectively.

Figure 6: The pressure-actuated inverter mechanisms. Filter radius $5.4h$ is used for the optimized results shown in rows 1 and 2, whereas for the results displayed in rows 3 and 4, it is set to $8.4h$. Note $h = \min \left( \frac{L_x}{N_{ex}}, \frac{L_y}{N_{ey}} \right)$. We refer the inverter mechanisms in the first, second, third and fourth rows as IV1 Pa-CMs, IV2 Pa-CMs, IV3 Pa-CMs and IV4 Pa-CMs respectively.
with complementarity condition $\Lambda (V(\rho^d(\rho)) - V^*_d) = 0, \Lambda \geq 0$. Using $f_0 = -\mu \frac{MSE}{SE}$ and in view of Eq. 17, $\Theta = \Theta_1 + \Theta_2$ transpires as

$$\Theta_1 = \mu \left[ u^\top \frac{\partial K}{\partial \rho} \left( -u \frac{MSE}{2(SE)^2} + \frac{v}{SE} \right) \right], \quad (19)$$

$$\Theta_2 = \mu \left[ \left( \frac{MSE}{(SE)^2} u^\top + \frac{-v^\top}{SE} \right) HA^{-1} \frac{\partial A}{\partial \rho} p \right]. \quad (20)$$

where $\Theta_1$ and $\Theta_2$ represent objective and load sensitivities respectively. Now, using Eq. 18 in association with the chain rule given in Eq. 12, one can find the sensitivity of $L$ with respect to the design vector, i.e., $\frac{dL}{d\rho}$. This formulation facilitates straightforward evaluation of the load sensitivities (Eq. 19) that affect the optimized designs of Pa-CMs (Kumar et al., 2020) and therefore, are important to consider while designing such mechanisms.

5 Numerical examples and discussions

This section evaluates the presented robust approach by designing pressure-actuated inverter and gripper CMs. The symmetric half design domains for designing these mechanisms are displayed in Fig. 3. $L_x = 0.2$ m and $L_y = 0.1$ m are set, where $L_x$ and $L_y$ represent the dimension in $x-$ and $y-$directions, respectively. 1 bar pressure load is applied on the left edge of the domains, whereas remaining edges excluding the symmetric ones experience zero pressure loading. The fixed parts of the domains and their symmetry boundaries are also depicted. Springs with spring stiffnesses $k_{ss} = 1 \times 10^4$ N m$^{-1}$ represent the workpiece at mechanism output locations (Fig. 3). Table 1 summarizes the design parameters used in the optimization. We use $N_{ex} \times N_{ey} = 200 \times 100$ bi-linear quadrilateral FEs to parameterize the symmetric half design domains (Fig. 3), where $N_{ex}$ and $N_{ey}$ indicate FEs in $x-$ and $y-$directions, respectively.

One can also employ honeycomb tessellation (hexagonal FEs) (Kumar, 2022) for the design representation. A density-based TO approach with one design variable for each FE is employed with plane stress conditions. The design variable is considered constant within each FE. The external move limit of the MMA optimizer is set to 0.1. The color schemes displayed in Fig. 4 are used to plot material, pressure and displacement fields in this paper.
\( \Delta \eta = 0.15 \) : 
\[ M_{\text{nd}} = 0.21\%, V_f = 0.17 \]

\( \Delta \eta = 0.05 \) : 
\[ M_{\text{nd}} = 0.26\%, V_f = 0.19 \]

\( \Delta \eta = 0.15 \) : 
\[ M_{\text{nd}} = 0.54\%, V_f = 0.16 \]

\( \Delta \eta = 0.05 \) : 
\[ M_{\text{nd}} = 0.37\%, V_f = 0.18 \]

Figure 9: The robust pressure-actuated gripper mechanisms. Filter radius 5.4h is used for the optimized results displayed in rows 1 and 2, whereas for the results shown in rows 3 and 4 are obtained with filter radius 8.4h. Note \( h = \min \left( \frac{L_x}{N_{\text{res}}}, \frac{L_y}{N_{\text{res}}} \right) \). We refer the gripper mechanisms in the first, second, third and fourth rows as GP1 Pa-CMs, GP2 Pa-CMs, GP3 Pa-CMs and GP4 Pa-CMs respectively.
(a) Eroded design, $\Delta = -0.13\,\text{mm}$
   \[ \Delta\eta = 0.15; \] 
   \[ M_{nd} = 0.26\%, \, V_f = 0.173 \]

(b) Intermediate design, $\Delta = -0.09\,\text{mm}$
   \[ M_{nd} = 0.65\%, \, V_f = 0.20 \]

(c) Dilated design, $\Delta = -0.07\,\text{mm}$
   \[ M_{nd} = 0.27\%, \, V_f = 0.23 \]

(d) Eroded design, $\Delta = -0.076\,\text{mm}$
   \[ \Delta\eta = 0.05; \] 
   \[ M_{nd} = 0.21\%, \, V_f = 0.19 \]

(e) Intermediate design, $\Delta = -0.065\,\text{mm}$
   \[ M_{nd} = 0.30\%, \, V_f = 0.20 \]

(f) Dilated design, $\Delta = -0.060\,\text{mm}$
   \[ M_{nd} = 0.22\%, \, V_f = 0.21 \]

(g) Eroded design, $\Delta = -0.14\,\text{mm}$
   \[ \Delta\eta = 0.15; \] 
   \[ M_{nd} = 0.46\%, \, V_f = 0.15 \]

(h) Intermediate design, $\Delta = -0.078\,\text{mm}$
   \[ M_{nd} = 0.30\%, \, V_f = 0.020 \]

(i) Dilated design, $\Delta = -0.058\,\text{mm}$
   \[ M_{nd} = 0.41\%, \, V_f = 0.25 \]

(j) Eroded design, $\Delta = -0.083\,\text{mm}$
   \[ \Delta\eta = 0.05; \] 
   \[ M_{nd} = 0.33\%, \, V_f = 0.18 \]

(k) Intermediate design, $\Delta = -0.063\,\text{mm}$
   \[ M_{nd} = 0.68\%, \, V_f = 0.20 \]

(l) Dilated design, $\Delta = -0.052\,\text{mm}$
   \[ M_{nd} = 0.4\%, \, V_f = 0.22 \]

Figure 10: The pressure-actuated contractor mechanisms. Filter radius $5.4h$ is used for the optimized results shown in rows 1 and 2, whereas for the results displayed in rows 3 and 4, it is set to $8.4h$. Note $h = \min \left( \frac{L_x}{N_{ex}}, \frac{L_y}{N_{ey}} \right)$. We refer the inverter mechanisms in the first, second, third and fourth rows as CT1 Pa-CMs, CT2 Pa-CMs, CT3 Pa-CMs and CT4 Pa-CMs respectively.
Figure 11: Objective convergence plots for IV2 and IV3 Pa-CMs are displayed in (a) and (b) respectively. The black, blue and red curves indicate the dilated, intermediate and eroded designs convergence history. We use the same color scheme for showing convergence curves henceforth.

5.1 Traditional pressure-actuated inverter mechanism

The symmetric half design domain displayed in Fig. 3a is considered, and the optimization formulation presented in Kumar et al. (2020) is employed for designing the inverter mechanism. The filter radius $r_{\text{fill}}$ is set to $2 \times \max\left(\frac{L_x}{N_{\text{ex}}}, \frac{L_y}{N_{\text{ey}}}\right)$. 20% material volume is permitted. Fig. 5a depicts the optimized inverter mechanism with its final pressure field. $M_{\text{rel}} = 8.9\%$ and $\Delta = 0.0235\;\text{mm}$ are found. Insets in Fig. 5a display the thin flexure regions constituted of gray elements. These geometrical anomalies pose challenges in manufacturing and thus, they are undesirable. When the optimized design is approximated using: (i) $\rho \in [0.75, 1] \rightarrow 1$ and $\rho \in [0, 0.75) \rightarrow 0$ as shown in Fig. 5a1, a design with thin and potentially challenging to manufacture regions is obtained with $\Delta = 0.0243\;\text{mm}$ and (ii) $\rho \in [0.25, 1] \rightarrow 1$ and $\rho \in [0, 0.25) \rightarrow 0$ as displayed in Fig. 5a2, a design with $\Delta = 0.0205\;\text{mm}$ is obtained, which is considerably lower than the displacement obtained for the actual design (Fig. 5a). These approximations also alter the topologies. Therefore, to circumvent these issues and also, to obtain optimized solutions close to 0-1 such that contours of the designs can be extracted without performing any approximation for fabrication purposes, as mentioned before, the robust formulation is employed in all following examples (Wang et al., 2011).

5.2 Pressure-actuated robust mechanisms

The various optimized designs of the pressure-actuated inverter, gripper and contractor CMs are presented using the robust formulation (Eq. 14). In each case, we get three optimized designs, i.e., the dilated, intermediate and eroded continua, and in those, the intermediate designs are intended for fabrication.

The permitted volume fraction for the intermediate design is set to 0.20 for all the cases. The maximum number of MMA iterations is fixed to 400. In the projection filter (Eq. 10), $\beta$ is altered from 1 to 128 using a continuation scheme wherein $\beta$ is doubled after each 50 MMA iteration and once it reaches to 128, it remains so for the remaining optimization iterations. The volume update for the dilated design is performed at each 25th MMA iteration.

In case of the inverter CMs, it is desired that the mechanisms provide deformation in the opposite
Figure 12: Volume fraction convergence plots for IV2 and IV3 Pa-CMs are displayed in (a) and (b) respectively.

Figure 13: Objective convergence plots for GP2 and GP3 Pa-CMs are shown in (a) and (b) respectively.

Figure 14: Volume fraction convergence plots for GP2 and GP3 Pa-CMs are shown in (a) and (b) respectively.
Figure 15: Objective convergence plots for CT2 and CT3 Pa-CMs are displayed in (a) and (b) respectively.

Figure 16: Volume fraction convergence plots for CT2 and CT3 Pa-CMs are displayed in (a) and (b) respectively.
Figure 17: Intermediate optimized designs of inverter (Row 1), gripper (Row 2) and contractor (Row 3) Pa-CMs. Results in column 1, column 2 and column 3 are obtained using $V_f^* = 0.1$, $V_f^* = 0.20$ and $V_f^* = 0.3$ respectively.

Figure 18: Deformed profiles of the optimized inverter, gripper and contractor CMs are displayed with 50 times magnified displacement obtained from linear analysis.
direction of the pressure loading direction, whereas gripping motions are desired in response to the pressure load for the gripper CMs. At the output location of the gripper mechanism, a void area of size $\frac{L}{5} \times \frac{L}{5}$ and a solid region of area $\frac{L}{5} \times \frac{L}{40}$ are considered to facilitate gripping of a workpiece. For the contracting mechanism, a contraction motion is sought while applying pressure loads on the left and right edges. A solid non-design domain of size $\frac{L}{40} \times \frac{L}{4}$ is present in the middle of the symmetric half domain of the contractor mechanism. Different filter radii and $\Delta \eta$ are used to find the mechanism optimized designs.

The full final inverter, gripper and contractor mechanisms are obtained by suitably transforming the symmetric half optimized results, and they are depicted in Fig. 6, Fig. 9, and Fig. 10 respectively. Note the absence of one-node hinges—these would lead to disconnected structures in the eroded design with very poor performance, and hence the optimizer avoids such problematic features entirely. The topology for the dilated, intermediate and eroded is the same for all the presented cases. The eroded designs feature thin members, whereas the dilated designs consist of thicker branches, which is expected. Thicknesses of the members of the intermediate designs are between those of the respective eroded and dilated designs. $M_{nd}$ and final volume fraction for each optimized design are also mentioned. The obtained solutions in Fig. 6, Fig. 9 and Fig. 10 are very close to 0-1 designs as their $M_{nd}$ values are low. This is a very desirable result, since now the design interpretation no longer significantly changes the geometry of the design and its performance.

One can note that IV2 Pa-CMs have different topologies than those of IV1, IV3 and IV4 Pa-CMs (Fig. 6). Fig. 7 and Fig. 8 display the topologies of IV1 Pa-CMs and IV2 Pa-CMs at different MMA iterations, respectively. Only the degree of manufacturing error, expressed by parameter $\Delta \eta$, differs between these two cases. From Figs. 7a-7b and Figs. 8a-8b, until 6 MMA iterations, design evolution of IV1 and IV2 mechanisms is similar. At the 7th MMA iteration, a minor difference can be noted in the material layouts of these mechanisms (Figs. 7c and 8c) and that eventually has lead to entirely different outcomes. Parameter dependence is expected given the non-convex nature of topology optimization problems. Moreover, IV2 represents the case with the smallest manufacturing error, thus has the highest design freedom and allows for the thinnest features. This apparently allows the optimization process to pursue a different solution with more localized (but not one-node connected) hinges. From a performance viewpoint, design IV1 is superior, hence in case IV2 an inferior local optimum is obtained. This may be the downside of the larger design freedom in this case, as it also can lead to a larger number of local optima. Note that the found output stroke $\Delta$ of inverter mechanisms obtained using the proposed formulation (Fig. 6) is better than that of the inverter mechanism (Sec. 5.1) obtained using the previous method presented in Kumar et al. (2020). This is because the current formulation tends to give relatively better distributed compliance with crisp boundaries, and as per Yin and Ananthasuresh (2001) such mechanisms perform relatively better. Because of the improved boundary definition, after post-processing the difference in performance is likely to increase even further. Design extraction is discussed in Sec. 5.4.

The convergence of the objective function is shown in Fig. 11, Fig. 13 and Fig. 15, and the corresponding evolution of the volume fraction in Fig. 12, Fig. 14 and Fig. 16. Stepwise changes are associated to updates of the parameter $\beta$. In all cases, the final volume fractions of the optimized intermediate designs are observed to be equal to the permitted volume fraction, i.e., 0.20. Near the 200th MMA iteration the objective values corresponding to the eroded designs IV3 Pa-CMs (red curve in 11b) are close to zero, which indicates the instantaneous disconnect-ednesses in those eroded designs. It is noticed that the optimized mechanisms obtained with $\Delta \eta = 0.15$ have larger minimum length scale than those obtained with $\Delta \eta = 0.05$ at the same filter radius. In addition, the optimized mechanisms with same $\Delta \eta$ but higher filter radius have larger minimum length scale. Therefore, the minimum feature size increases with increase in $\Delta \eta$.
and is also a function of the filter radius, which are known properties of the robust formulation (Trillet et al., 2021; Wang et al., 2011).

The layout of the optimized mechanisms is different than the traditional counterparts except that of IV2 Pa-CMs. The obtained designs contain a large space for fluid to inflate. This is reminiscent of designs of pneumatically-actuated soft robots. The obtained output deformations of the optimized eroded mechanisms are higher than the intermediate designs in each case (Fig. 6, Fig. 9 and Fig. 10), however, as mentioned before, such designs have lower manufacturing limits and as-fabricated designs may be fragile. Figures 11 and 12 indicate the convergence history of the objective and volume constraints for the optimized IV2 and IV3 mechanisms (Fig. 6). Those for the GP2 and GP3 mechanisms (Fig. 9) are displayed in Fig. 13 and Fig. 14 respectively. The objective and volume convergence plots for CT2 and CT3 mechanisms are displayed in Fig. 15 and Fig. 16, respectively. One notices a smooth convergence, and the volume constraint is active for the intermediate design at the end of the optimization for the mechanisms. The objective values of the intermediate designs are lower than those of the corresponding eroded and dilated designs (Figs. 11, 13 and 15), indicating that intermediate designs are the best performing ones in the view of the considered multi-criteria objective that determines a balance between the output displacement and strength of the mechanism. The objectives of respective eroded and dilated designs are close to each other at the end of the optimization (Figs. 11, 13 and 15). The convergence curves of all cases show similar characteristics and relatively orderly behavior (aside from the expected continuation-induced steps). Showing these characteristics is the main point. The deformed profiles for the full intermediate inverter (IV2 and IV3), gripper (GP2 and GP3) and contractor (CT3) mechanisms at 50 times magnified linear deformation are displayed in Fig. 18. We select \{IV2, IV3\} and \{GP2, GP3\} mechanisms to study noting the dissimilarities in their topologies (Figs. 6 and 9), whereas CT3 mechanism is selected (randomly) for contractor mechanism analyses. The same set of mechanisms is considered in Sec. 6 for large deformation analyses with high pressure loads. These Pa-CMs perform as expected, however the deformation profiles are far from those obtained when using nonlinear mechanics with high pressure loads, as studied in Sec. 6.

5.3 Pa-CMs for different volume fractions

In this section, we demonstrate effects of different volume fractions on the optimized Pa-CMs. \(\Delta \eta = 0.15\), filter radius \(r_{\text{fill}} = 8.4h\) with \(h = \min \left( \frac{L_x}{N_{ex}}, \frac{L_y}{N_{ey}} \right)\) and \(V_f^* = 0.1\) and 0.3 are taken herein.

Figure 17 depicts the intermediate optimized designs of inverter (Row 1), gripper (Row 2) and contractor (Row 3) Pa-CMs. One can note that topologies of the optimized designs for inverter are different with \(V_f^* = 0.1\), \(V_f^* = 0.2\), and \(V_f^* = 0.3\), however those of the contractor mechanisms are same. For the gripper mechanisms, the topologies are different with \(V_f^* = 0.1\), and \(V_f^* = 0.2\). By and large, the topologies of optimized CMs depend upon the given volume as different amounts of available material enable different optimal design solutions. In addition, with the lower volume fraction e.g. \(V_f^* = 0.1\), the optimized designs contain relatively more area for fluidic pressure load to inflate like soft robots. Therefore, the final topologies as well as the regions within the optimized CMs to inflate under fluidic pressure loads depend upon the permitted resource volume. One notices that the objective improves as the volume fraction of the mechanisms is increased. This implies that the optimized mechanisms converge towards better combinations of output displacement and stiffness, according to the stated objective (Eq. 14). With the increase in volume fraction, strain energy of the mechanism decreases, i.e. stiffness of the mechanisms increase. The output deformation \(\Delta\) decreases as the volume fraction
and stiffness increases of the Pa-CM. $\Delta$ can also potentially depend upon the final topology of the mechanism, as noted in gripper Pa-CMs (row 2 of Fig. 17). It can be concluded that in the current formulation, the volume fraction is an important parameter to explore in Pa-CM design studies.

5.4 Extracting the optimized designs

We present a method to extract the optimized Pa-CMs for generating CAD models and for performing further analysis with high pressure loads in ABAQUS herein. We adopt the following steps:

1. Prepare a VKT file using the optimized physical density field, nodal coordinates and element connectivity matrix.

2. Import the VKT file into a data visualization software, ParaView (Ahrens et al., 2005). Use the command CellDatatoPointData which converts cell (element) information to point (node) information. Extract the final design at the threshold density 0.85 value using the IsoVolume function, and save it as a portable network graphics (PNG) file.

3. Import the PNG file to a vector graphics software, InkScape. Use Trace Bitmap to trace the boundary and save the traced design as a DXF file.

4. Import the DXF file to AutoCAD software and save as an IGES file for importing into ABAQUS for the further analyses.
Figure 20: Deformed profiles of the gripper CM at different pressure loads. GP2 Pa-CMs (a) 10 bar, (b) 25 bar, (c) 50 bar and GP3 Pa-CMs (d) 10 bar, (e) 25 bar, (f) 50 bar.

Figure 21: Deformed profiles of the contractor CM at different pressure loads. CT3 Pa-CMs (a) 10 bar, (b) 25 bar, (c) 50 bar.
Large deformation analyses and challenges

In this section, the optimized compliant mechanisms are tested with high pressure loadings to investigate their behaviors under large deformation cases. In the FE analysis, geometric nonlinearity will now be considered. In addition, instead of the linear material model, a neo-Hookean material model with the following strain energy function $W$ (Zienkiewicz and Taylor, 2005) is employed

$$W = \frac{G}{2} [\text{tr}(FF^T) - 3 - 2\ln J] + \frac{\lambda}{2} (\ln J)^2,$$

(21)

where $F = \nabla_0 u + I$ is the deformation gradient and $G = \frac{E_1}{2(1+\nu)}$ and $\lambda = \frac{2G\nu}{1-2\nu}$ are Lame constants. $\nabla_0 u$ denotes gradient of the displacement field $u$ with respect to reference coordinates $X$, and $\nu$ is Poisson’s ratio. $J = \det(F)$, and $I$ is the unit tensor. Typically, rubber-like materials are used for pneumatically-actuated mechanisms Schmitt et al. (2018) and to numerically model such materials, a neo-Hookean material description can be employed (Zienkiewicz and Taylor, 2005).

Using the fundamentals of the continuum mechanics, the Cauchy stress tensor $\sigma$ can be determined from the strain energy function noted in Eq. 21 as

$$\sigma = \frac{G}{J}(FF^T - I) + \frac{\Lambda}{J}(\ln J)I.$$

(22)

Following the nonlinear FE formulation, the displacement vector $u$ is determined by solving

$$R(u) = F_{\text{int}}(u) - F_{\text{ext}}(u) = 0,$$

(23)

where $R(u)$ is the residual force and $F_{\text{ext}}(u)$ is the external force arises due to the pressure loading. The internal force vector $F_{\text{int}}^e(u)$ at the element level is determined as

$$F_{\text{int}}^e = \int_{\Omega^e} B_{UL}^\top(d\epsilon(u)) \sigma(u) d\Omega^e,$$

(24)

where $B_{UL}(u)$ and $\sigma_e$ are the updated Lagrangian strain-displacement matrix and the Cauchy stress tensor of an FE $\Omega^e$, respectively. Eq. 23 can be solved using a Newton-Raphson (N-R) iterative process. Note that, $F_{\text{ext}}(u)$ varies as it arises from pressure loading which follows the surface where it is applied upon, i.e., it is a follower force and thus, contributes in the tangent stiffness of the nonlinear equations and cannot be omitted from the topology optimization. In addition, the flow coefficient matrix $A^e$ (Eq. 6) varies with the deformation. Further analysis robustness and efficiency under distortion and/or unrealistic deformation of the low-density elements with high pressure load (van Dijk et al., 2014) pose challenges in a non-linear finite element TO setting.

The elemental $K_{\text{ext}}^e$ can be determined as (see Appendix B)

$$K_{\text{ext}}^e = \int_{\Gamma_p} pN^\top(n \otimes a^\alpha - a^\alpha \otimes n)N_{\alpha \alpha} da$$

(25)

where $p$ represents the magnitude of the pressure load, $n$ is the normal vector and $a^1$, and $a^2$ are the contra-variant tangent vectors to the pressure surface $\Gamma_p$ (see Appendix A). $n = \frac{a_1 \times a_2}{|a_1 \times a_2|}$, where $a_1$, and $a_2$ are the co-variant tangent vectors (see Appendix A). $\otimes$ represents the tensor product, $da$ is the elemental area and $\alpha = 1, 2$. Note that with a design dependent load case, $n = n(u, \rho)$ which gives additional load sensitivities and a nonlinear external force stiffness matrix. Herein, we use ABAQUS for the nonlinear finite element analyses to show the limitations of Pa-CMs optimized via linear elastic assumptions and also, to note additional challenges.
Table 2: Output displacements of the optimized Pa-CMs under high pressure loadings

| Mechanisms/Pressure load | 10 bar | 25 bar | 50 bar |
|--------------------------|--------|--------|--------|
| IV2 Pa-CM                | -31.30 | -39.60 | -43.0  |
| IV3 Pa-CM                | -8.93  | -10.99 | -12.10 |
| GP2 Pa-CM                | -13.40 | -15.38 | -15.67 |
| GP3 Pa-CM                | -7.63  | -12.0  | -14.80 |
| CT3 Pa-CM                | -3.42  | -10.48 | -15.30 |

The intermediate optimized designs of the inverter (IV2 and IV3), gripper (GP2 and GP3) and contractor (CT3) mechanisms are selected for the nonlinear analyses in ABAQUS as mentioned in Sec. 5.2. First, the boundaries of the optimized designs are extracted, and corresponding 2D CAD models are generated (see Sec. 5.4). Thereafter, using these CAD models, nonlinear FE analyses while considering follower force characteristics of the pressure loads are performed with input pressure 10 bar, 25 bar and 50 bar in ABAQUS (note, the design pressure load is 1 bar). Fig. 19, Fig. 20 and Fig. 21 display the deformed profiles of the CMs with high pressure loadings. As pressure loads increase the deformation of the Pa-CMs also increase, which is expected and natural. Table 2 indicates the output displacements of the mechanisms at different pressure loads. At 50 bar, IV3 Pa-CM experience self-contact (Fig. 19f) that indicates that one may have to include self-contact conditions (Kumar et al., 2019b, 2021a; Kumar, P., 2017) between the branches of the mechanisms when dealing with high pressure loadings for the large deformation cases. For the inverter mechanisms, the deformation profiles in Fig. 18b and Fig. 19f are different, and the output displacement has reduced by 90.25 %. Likewise, the deformed continua in Fig. 18d and Fig. 20f are not the same. Although the gripper still exhibits the intended functionality, the magnitude of the jaw displacement has reduced by 76.11 %. These indicate limitations and shortcomings of the Pa-CMs obtained assuming linear elasticity concepts. Therefore, ideally, one has to include full nonlinear mechanics (with contact) within the design approach for high pressure loadings wherein Pa-CMs can experience large deformations and even self and/or mutual contact. In addition, integration of the required actuators with the mechanism/soft robot design optimization process provides an additional set of challenges (Cao et al., 2015b), which is another future research direction.

7 Closure

With the aim to bridge the gap between optimized and as-fabricated designs, this paper presents a robust density-based topology optimization approach to generate pressure-actuated compliant mechanisms. The robust formulation, i.e., the eroded, intermediate (blueprint) and dilated projections for the design description is employed for the first time to this problem class, in combination with a representation of the pressure loads using the Darcy law in combination with a drainage term. Well-functioning pressure-actuated robust inverter, gripper, and contractor mechanisms are obtained with different robustness levels and filter radii to illustrate different minimum length scales of the mechanisms. Single-node hinges, that cannot be fabricated but frequently appear when using the traditional topology optimization approach, are no longer found in the obtained designs. Moreover, it is observed that the robust approach leads to improved mechanism performance with better boundary crispness. The approach solves three
sets of equilibrium equations for each field (pressure, displacement and virtual displacement) and uses a continuation approach for the projection parameter $\beta$ that requires a large number of the optimization iterations and thus, computational cost increases. The approach provides three physical material density vectors with one design variable vector. Intermediate designs can be used for manufacturing purposes.

The obtained optimized designs are close to binary. This eliminates the loss of performance observed when grayscale topology optimization results are post-processed into CAD geometries for fabrication. The intermediate designs are used to study behavior of the mechanisms with high pressure loads while accounting for geometric and material nonlinearity. It is found that the scaled linear deformation profiles and those obtained with full nonlinear analyses do not match well. Moreover, at high pressures self-contact of mechanism branches occurs. These observations indicate that for further development of topology optimization for pressure-actuated compliant mechanisms, next to the robust formulation, considering nonlinear mechanics and self-contact may be indispensable.

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Declaration of Competing Interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

Appendix A Surface description

To model follower forces, a ubiquitous option is to employ curvilinear coordinates setting Wriggers (2006) which is summarized herein.

Imagine a material point $x \in \Gamma_p$, where $\Gamma_p$ is a pressure load surface of the evolving design $^4\Omega$ (Fig. A.1). In a parametric setting, the point can be described as

\[ x = x(\xi^\alpha), \quad \xi^\alpha \in \Gamma_p \]  

\[ ^4\text{deformed configuration of } \Omega_0 \]
where $\xi^\alpha|_{\alpha=1,2}$ are curvilinear coordinates lying in the 2D parametric space. Please note, Greek indices in Eq. (A.1) and henceforth adopt a set of values $\{1, 2\}$. Repeated indices follows Einstein’s summation rule. The tangent vectors at the point $x \in \Gamma_p$ can be determined as

$$a_\alpha = \frac{\partial x}{\partial \xi^\alpha}, \quad \alpha = 1, 2$$  \hspace{1cm} (A.2)

Though these tangents may not necessarily be orthonormal, they form the basis of tangent plane to $\Gamma_p$. $a_\alpha$ refer co-variant tangent vectors that are related via an associated metric tensor $a_{\alpha\beta}$ as

$$a_{\alpha\beta} = a_\alpha \cdot a_\beta.$$  \hspace{1cm} (A.3)

The unit normal at $x \in \Gamma_p$ is evaluated as

$$n = \frac{a_1 \times a_2}{\|a_1 \times a_2\|} = \frac{a_1 \times a_2}{j_a},$$  \hspace{1cm} (A.4)

where $j_a = \sqrt{\det a_{\alpha\beta}}$ can be established (Sauer et al., 2014; Wriggers, 2006). The set $\{a_1, a_2, n\}$ constitutes a basis of $\mathbb{R}^3$. In addition, one can define another set of tangents $a^\beta$ termed contra-variant tangents as

$$a_\alpha \cdot a^\beta = \delta_{\alpha\beta},$$  \hspace{1cm} (A.5)

where $\delta_{\alpha\beta}$ is the Kronecker delta. The metric tensors of co-variant and contra-variant tangents are related as

$$a^{\alpha\beta} = [a_{\alpha\beta}]^{-1},$$  \hspace{1cm} (A.6)

where contra-variant metric tensor $a^{\alpha\beta} = a^\alpha \cdot a^\beta$. Using contra-variant tangents and the unit normal (Eq. A.4) one can have another set of basis vectors $\{a^1, a^2, n\}$ of $\mathbb{R}^3$. The co-variant and contra-variant tangents are also related as $a_\alpha = a_{\alpha\beta} a^\beta$ and $a^\alpha = a^{\alpha\beta} a_\beta$.

![Figure A.2: Components of vector $x$ along co-variant and contra-variant tangents](image)

Using triad $\{a_1, a_2, n\}$ and/or $\{a^1, a^2, n\}$ a vector $x$ can be written as (Fig. A.2)

$$x = x^\alpha a_\alpha + x_n n = x_\alpha a^\alpha + x_n n$$  \hspace{1cm} (A.7)

where $x^\alpha = x \cdot a^\alpha$, $x_\alpha = x \cdot a_\alpha$ and $x_n = x \cdot n$ represent contra-variant, co-variant and normal components of the vector, respectively. Note that $x^\alpha = a^{\alpha\beta} r_\beta$ and $r_\alpha = a_{\alpha\beta} r^\beta$. Parametric derivative of the co-variant tangents can be evaluated as

$$a_{\alpha,\beta} = \frac{\partial a_\alpha}{\partial \xi^\beta}.$$  \hspace{1cm} (A.8)
Appendix B  Evaluation of tangent stiffness matrix $K^e_{\text{ext}}$

Linearization of the elemental pressure load $F^e_{\text{ext}}$ at $u$ in the direction of $\Delta u$ is performed to evaluate $K^e_{\text{ext}}$. One writes

$$f^e_{\text{ext}}(u + \Delta u) \approx f^e_{\text{ext}}(u) + \Delta f^e_{\text{ext}}(u),$$  \hspace{.5cm} \text{(B.1)}

Where (Zienkiewicz and Taylor, 2005)

$$f^e_{\text{ext}} = \int_{\Gamma_p} N^\top p \text{d}a,$$ \hspace{.5cm} \text{(B.2)}

$p$ is the magnitude of the follower force. Evidently,

$$\Delta f^e_{\text{ext}}(u) = \int_{\Gamma_p} N^\top p \text{d}(\text{da}) + \int_{\Gamma_p} N^\top \Delta p \text{d}a.$$ \hspace{.5cm} \text{(B.3)}

Herein $\Delta p = 0$ as pressure is considered height independent. Therefore,

$$\Delta f^e_{\text{ext}}(u) = \int_{\Gamma_p} N^\top p \text{d}(\text{da}).$$ \hspace{.5cm} \text{(B.4)}

In view of Eq. A.4, we have

$$\Delta(\text{nda}) = \Delta ((a_1 \times a_2)\text{d}\square) = \Delta (a_1 \times a_2)\text{d}\square$$ \hspace{.5cm} \text{(B.5)}

where $\text{ja}_a\text{d}\square = \text{da}$ and $\text{d}\square = d\xi^1d\xi^2$.

**B.1 Evaluation of $\Delta(a_1 \times a_2)\text{d}\square$**

Herein, steps for evaluating $\Delta(a_1 \times a_2)\text{d}\square$ are described.

$$\Delta(a_1 \times a_2)\text{d}\square = (\Delta a_1 \times a_2 + a_1 \times \Delta a_2)\text{d}\square$$

$$= \sum_I \left( N_{1I} \frac{\Delta u_I \times a_2 + a_1 \times \Delta u_I N_{2I}}{T_1} \right) \quad \text{[\because \Delta u_I = \Delta x_I]}$$ \hspace{.5cm} \text{(B.6)}

Decomposing $\Delta u_I$ using its components along tangential and normal direction as $\Delta u_I = \Delta u_I^1 a_1 + \Delta u_I^2 a_2 + \Delta u_I^n n$ allows evaluation $T_1$ and $T_2$ as

$$T_1 = \Delta u_I \times a_2 = (\Delta u_I^1 a_1 + \Delta u_I^2 a_2 + \Delta u_I^n n) \times a_2)$$

$$= j_a (\Delta u_I^1 n - \Delta u_I^n a^1)$$

$$= j_a (n \otimes a^1 - a^1 \otimes n) \Delta u_I,$$ \hspace{.5cm} \text{(B.7)}

where $\otimes$ indicates the tensor product. Likewise,

$$T_2 = j_a (n \otimes a^2 - a^2 \otimes n) \Delta u_I.$$ \hspace{.5cm} \text{(B.8)}

With $T_1$ and $T_2$, Eq. (B.6) yields

$$\Delta(a_1 \times a_2)\text{d}\square = (n \otimes a^\alpha - a^\alpha \otimes n) N_{\alpha} \Delta u_c \text{d}a$$ \hspace{.5cm} \text{(B.9)}

Therefore, in view of Eq. (B.5) and Eq. (B.9), one writes Eq. (B.4) as

$$\Delta f^e_{\text{ext}}(u) = K^e_{\text{ext}} \Delta u_c$$ \hspace{.5cm} \text{(B.10)}
where

\[ \mathbf{K}_{\text{ext}}^e = \int_{\Gamma_p} p \mathbf{N}^T (\mathbf{n} \otimes \mathbf{a}_p - \mathbf{a}_p \otimes \mathbf{n}) \mathbf{N}, a \, da. \] (B.11)

**B.2 Expression \( \mathbf{K}_{\text{ext}}^e \) in 2D**

\( \mathbf{K}_{\text{ext}}^e \) is evaluated by interpolating the edge of a quadrilateral element (Fig. B.1) using linear Lagrangian shape functions (Eq. B.12) wherein

\[ N_1 = \frac{1}{2} (1 - \xi_p), \quad N_2 = \frac{1}{2} (1 + \xi_p), \quad \xi_p \in [-1, 1] \] (B.12)

Point \( \mathbf{x}_p = N_1 \mathbf{x}_1 + N_2 \mathbf{x}_2 \), length of the element \( l_e = ||\mathbf{x}_2 - \mathbf{x}_1|| \), tangent vector \( \mathbf{a}_p = \frac{\mathbf{x}_2 - \mathbf{x}_1}{l_e} \) and unit normal \( \mathbf{n} = \mathbf{e}_3 \times \mathbf{a}_p \) can be evaluated. With respect of these values, we can have

\[ \mathbf{K}_{\text{ext}}^e = \int_{\Gamma_p} p \mathbf{N}^T (\mathbf{n} \otimes \mathbf{a}_p - \mathbf{a}_p \otimes \mathbf{n}) \mathbf{N}, a \, da \] (B.13)

with \( \mathbf{N} = [N_1 \mathbf{l}_2, \ N_2 \mathbf{l}_2] \) and \( \mathbf{N}, a = 0.5 \times [-\mathbf{l}_2, \ \mathbf{l}_2] \) and \( \mathbf{l}_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \).

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