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Piezoelectric truss metamaterials: data-driven design and additive manufacturing

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The inherent directionality of piezoelectric materials is constrained by the symmetry of their crystal structure, which limits the property space in natural piezoelectric materials. To alleviate this limitation, one could leverage geometry or architecture at the mesoscale. Here, we present a framework for designing and 3D-printing piezoelectric truss metamaterials with customizable anisotropic responses. We employ generative machine learning to design truss metamaterials and achieve unconventional behaviors, including auxetic, unidirectional, and omnidirectional piezoelectricity. Then, we develop an in-gel-3D printing method to fabricate these structures using a composite slurry of photo-curable resin and lead-free piezoelectric particles. We achieve an improvement of over 48% in the specific hydrostatic piezoelectric coefficient in optimized metamaterials over bulk lead zirconate titanate (PZT), and the rare phenomenon of higher transverse piezoelectric coefficients than the longitudinal coefficient. Our approach enables customizable piezoelectric responses and paves the way towards the development of a new generation of electro-active animate materials.

Piezoelectric materials demonstrate a separation of positive and negative charge centers at the atomic scale in response to mechanical deformation, which facilitates the two-way coupling between mechanical and electrical energies. This makes them an indispensable component in a wide range of engineering applications across domains, such as biomedical^{1–3}, aerospace^{4,5}, electronics^{6,7}, building structures^{8,9}, and electro-chemistry¹⁰. The search for superior piezoelectric properties has been a topic of high interest among the scientific community. Yet, the progress has been primarily restricted to improving the performance of existing piezoelectric materials by either material synthesis-based methods^{11–15} or macro-scale design and optimization methods^{16–18}. With the advent of technologies such as micro- and nanoelectromechanical systems (MEMS and NEMS) and micro-robotics, the need for exotic and tunable electromechanically responsive systems is pushing the limits of functionalities beyond what current materials can offer. The primary restricting factor for the range of piezoelectric response in conventional piezoelectric materials is their crystal symmetry, which dictates the directionality of electromechanical coupling.

The piezoelectric coupling correlates six mechanical stresses or strains to the three electric field components. This correlation is defined by the piezoelectric tensor, which has 18 unique coefficients

correlating the electrical and mechanical state variables. Commonly used piezoelectric materials such as dielectric ceramics, e.g., lead zirconate titanate (PZT), BaTiO₃ (BT), PMN-PT, and polymers, e.g., polyvinylidene fluoride (PVDF), have only five non-zero piezoelectric coefficients, due to their crystal symmetry. These coefficients have fixed signs (positive or negative) which cannot readily be flipped. This makes certain desirable functionalities in piezoelectric materials rare. For example, the phenomenon of negative piezoelectricity, capable of generating compression in response to a positive electric field, has only been observed in PVDF and its co-polymers¹⁹. In addition to the individual coefficients, several of their desirable combinations are unavailable in conventional materials, such as observing piezoelectric response only in a particular direction and eliminating the noise signal from the other directions and zero longitudinal piezoelectricity to circumvent microphony^{20,21}. Thus, realizing the full anisotropic design space of piezoelectric materials could be beneficial for several applications and warrants the development of an effective design and fabrication route.

While changing or designing the arrangement of ions in the material crystals may not be physically feasible to achieve tailorable properties, it is possible to design the architecture of the material at a mesoscale. In this context, metamaterials, also known as architected

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materials, emerge as a promising way to design for tailored piezoelectric properties. Metamaterials are constructed by tessellations of periodic or aperiodic unit cells large enough to be manufacturable yet small enough to constitute a pseudo-homogeneous response at the macro scale. One notable class of these metamaterials is beam-based truss metamaterials (constructed by a spatial arrangement of connected struts), which have been shown to be extremely light-weight and mechanically strong^{22–24}.

The architecture of a truss metamaterial is defined by the combination of the geometry and topology of the unit cells, where the geometry means the position of the junctions or nodes of the struts, and topology is the connectivity of these nodes. Truss metamaterials have been vastly explored for exotic and tunable mechanical properties^{23–29}. The development of piezoelectric metamaterials requires special considerations in design and fabrication due to multiphysics coupling and the brittle nature of piezoelectric materials. Cui et al. have demonstrated the tailorable response of architected piezoelectric metamaterials and their applications in designing robotic metamaterials^{30,31}. While the presented concepts are promising, the design process has remained either ad hoc or based on a simple parametrization of a small design space. The main challenge is that of inverse design, namely, exploration of the infinite number of possible designs to discover the ones with desirable properties.

The inverse design problem has been tackled in mechanical metamaterials^{22,23}, however, for piezoelectric materials, it becomes more complicated due to multiphysics coupling and poling orientation dependence of piezoelectric properties. To solve this challenge, we propose a generative machine learning (ML) approach for designing the piezoelectric response in architected materials. We use two distinct descriptions of the truss metamaterials, where one provides diverse piezoelectric responses while the other facilitates a broad range of properties and better manufacturability. The design-property pairs of these truss metamaterials are used to train the ML framework, which can generate designs corresponding to desired properties. Having established the design framework, we further address the next major challenge in the development of piezoelectric metamaterials, which is their fabrication.

Additive manufacturing techniques, such as stereolithography (SLA), digital light processing (DLP), and fused filament deposition (FFF), offer great freedom in fabricating a broad range of topologies. Doing so with piezoelectric materials is challenging for several reasons, such as the requirement of high-temperature processing in ceramics and lower piezoelectricity in polymers, low mechanical integrity due to their brittleness, and the requirement of poling post-manufacturing. Printing polymer-ceramic piezoelectric composites with a photo-curable resin ink infused with piezoelectric particles has been shown recently for manufacturing spinodoid topologies with bi-continuous features³². This approach requires high precision to manufacture slender structural members of low relative density metamaterials, such as those based on truss architectures. Moreover, their printing requires recoating of the printing ink for each layer due to the necessary high particle loading, which leads to excess material usage and a cost-intensive setup^{30,32}. Another difficulty in achieving continuous SLA/DLP printing with piezoelectric inks arises from the contrast in optical and rheological properties of the piezoelectric particles and photo-curable resins. Direct ink writing (DIW) offers a better approach for this purpose, where composite inks can be extruded to print shapes in 3D space³³. This approach has been shown to be effective by Tao et al.³⁴ for printing piezoelectric materials with relatively simpler shapes. However, sustaining the accurate shape and ensuring connectivity becomes challenging when the whole structure cannot be printed continuously, such as the truss structures we propose.

To achieve this, we develop an in-gel 3D printing technique to enable the additive manufacturing of piezoelectric truss metamaterials. A photosensitive resin-based ink infused with piezoelectric microparticles is synthesized and used to extrude the desired shapes.

With an optimized print path and rheology, the printing technique enables connectivity of the struts while the shape accuracy is preserved due to structural support provided by the gel. Our inverse design framework, combined with the developed 3D printing methodology, could successfully design and fabricate unique piezoelectric metamaterials with tunable piezoelectric response. For example, we designed metamaterial unit cells with maximized hydrostatic piezoelectric coefficient, transverse piezoelectric coefficient higher than longitudinal, auxetic piezoelectric response, and selective piezoelectric coefficients.

Results

Piezoelectric truss metamaterials

The homogenized properties of truss metamaterials at the macro scale depend on the topology and the geometry of the unit cells. In the case of piezoelectric metamaterials, the orientation dependence of the piezoelectric property tensor of individual struts also becomes important. Figure 1a demonstrates the concept of piezoelectric truss metamaterials. The inclination of the individual struts activates multiple modes of piezoelectricity simultaneously, as demonstrated in Fig. 1a. We define a global coordinate system (X, Y, Z) and a local coordinate system for individual struts (x, y, z) where x -axis aligns with the length of the strut (Supporting Information Fig. S1). The two-level orientation dependence of the piezoelectric tensor arises from the fact that the piezoelectric tensor is defined with reference to the poling direction, which is parallel to the global Z axis. As the orientation of the individual struts with respect to the poling direction can be different, the piezoelectric coefficients can have different values for individual struts, which contributes to the complexity of the design-property relationship.

The two most common forms of constitutive equations of piezoelectricity are strain-charge and stress-charge forms, which can be written in Einstein summation notation as

$$\epsilon_{ij} = S_{ijkl}^E \sigma_{kl} + d_{ijk} E_k, \quad (1a)$$

$$D_i = d_{ijk} \sigma_{jk} + \kappa_{ik}^\sigma E_k, \quad (1b)$$

and,

$$\sigma_{ij} = C_{ijkl}^E \epsilon_{kl} - e_{ijk} E_k, \quad (2a)$$

$$D_i = e_{ijk} \epsilon_{jk} + \kappa_{ik}^\epsilon E_k. \quad (2b)$$

Here ϵ_{ij} and σ_{ij} denote the mechanical strain and stress tensors, and the electric field and electric displacement vectors are designated as E_k and D_k . C_{ijkl}^E , e_{ijk} , d_{ijk} , and κ_{ik}^σ denote mechanical stiffness, piezoelectric coupling coefficients in stress charge and strain charge forms, and dielectric permittivity of the material. The indices $i, j, k, l \in \{1, 2, 3\}$, while the superscripts $(\cdot)^E$, $(\cdot)^\sigma$, and $(\cdot)^\epsilon$ indicate the properties measured at a constant electric field, stress, and strain, which are omitted hereafter for the sake of brevity. To concur with the finite element framework used for homogenization, the stress-charge form of piezoelectricity in Eq. 2 is used. Theoretically, it has been shown that the piezoelectric tensor can be tuned by controlling the direction of poling of the material³⁵. We apply poling along the global Z axis, so the effective piezoelectric coefficient in the local (x, y, z) coordinate system of a beam, denoted by e' can be computed as

$$e'_{pqr} = R_{pi} R_{qj} R_{rk} e_{ijk}^{\text{base}}. \quad (3)$$

Here e_{ijk}^{base} are the components of the piezoelectric tensor of the base material, $\mathbf{R} \in \text{SO}(3)$ is the rotation matrix (see Supporting Information Eq. 2) containing the direction cosines of the local coordinate axes (x, y, z), with

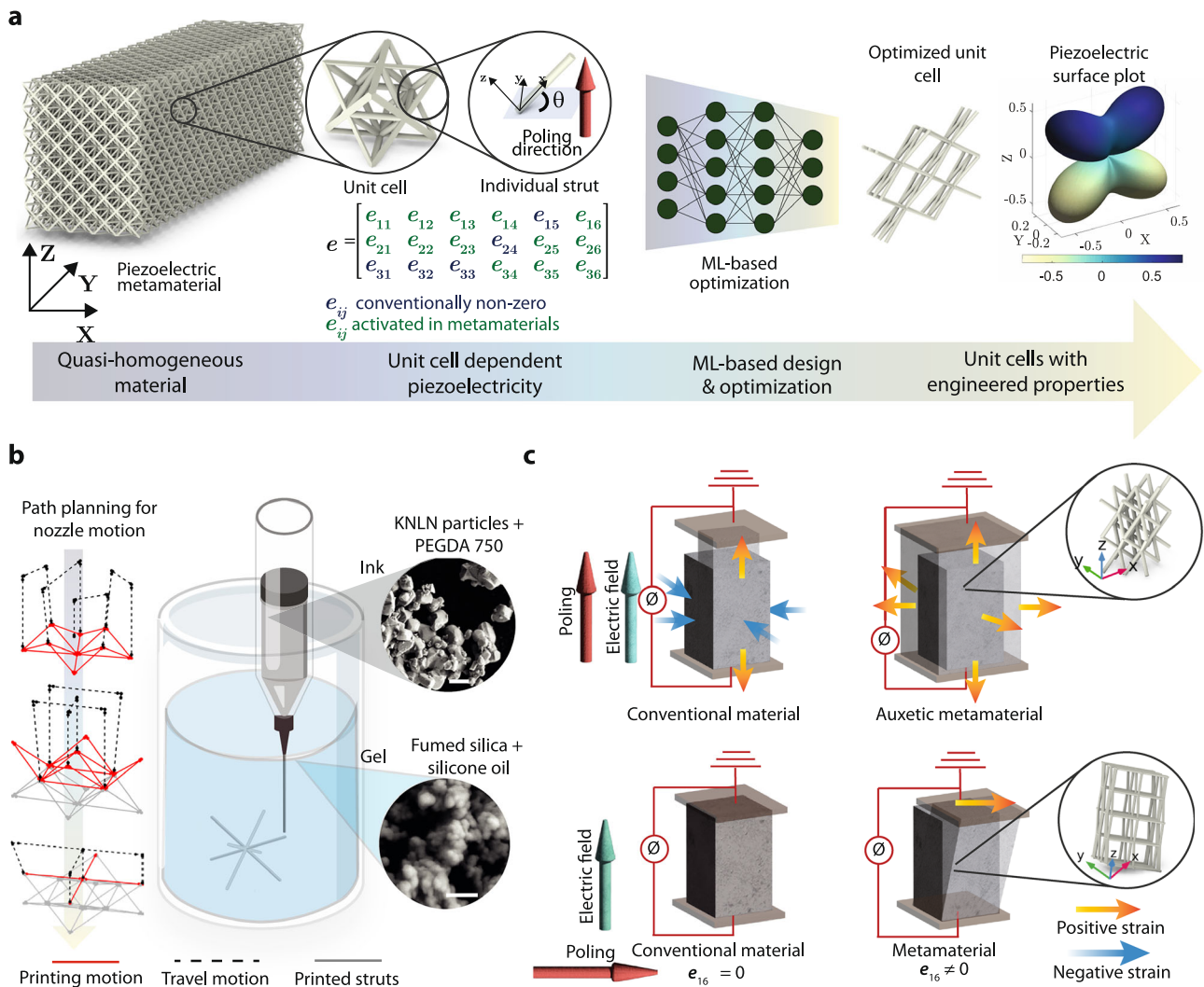


Fig. 1 | Concept of piezoelectric truss metamaterials for tailoring the electro-mechanical behavior. **a** Illustration of a piezoelectric metamaterial made by tessellating octet unit cells in three orthogonal directions. The different levels of structural features are shown in the insets. These mesoscale features can be tuned to selectively activate different components of the piezoelectric property matrix, in contrast to the only five non-zero coefficients in most of the common piezoelectric materials. ML is then used to design unit cells with desired properties, represented by the piezoelectric surface plot on the right. **b** In-gel direct-ink-writing-based 3D printing of the designed metamaterials is performed by extruding the piezoelectric ink in a support gel, by following the planned path for minimal travel motion. The piezoelectric composite ink is made of KNLN particles and PEGDA 700 matrix, as

shown in the inset (scale bar 2 μm). The ink is then extruded in a support gel made of fumed silica and silicon oil (scale bar 200 nm). **c** Examples of piezoelectric metamaterials made by the truss unit cells exhibiting behaviors not readily available in conventional materials. The first example (top) shows the auxetic behavior of a truss metamaterial. While conventional materials, under an applied electric field in the poling direction, expand in the direction of the electric field and contract along the two normal directions, the metamaterial made of the auxetic unit cell expands in all three directions. The second example (bottom) shows a metamaterial with a non-zero e_{16} coefficient, exhibiting shear deformation in response to an electric field applied in direction 1.

respect to global coordinate axes (X, Y, Z). Exploiting the symmetry of stress and strain components (e.g., $\sigma_{ij} = \sigma_{ji}$), Eq. 2 can be simplified and written in matrix form as

$$\begin{bmatrix} \sigma_{11} \\ \sigma_{22} \\ \sigma_{33} \\ \sigma_{23} \\ \sigma_{31} \\ \sigma_{12} \\ D_1 \\ D_2 \\ D_3 \end{bmatrix} = \begin{bmatrix} C_{11} & C_{12} & C_{13} & C_{14} & C_{15} & C_{16} & -e_{11} & -e_{21} & -e_{31} \\ C_{21} & C_{22} & C_{23} & C_{24} & C_{25} & C_{26} & -e_{12} & -e_{22} & -e_{32} \\ C_{31} & C_{32} & C_{33} & C_{34} & C_{35} & C_{36} & -e_{13} & -e_{23} & -e_{33} \\ C_{41} & C_{42} & C_{43} & C_{44} & C_{45} & C_{46} & -e_{14} & -e_{24} & -e_{34} \\ C_{51} & C_{52} & C_{53} & C_{54} & C_{55} & C_{56} & -e_{15} & -e_{25} & -e_{35} \\ C_{61} & C_{62} & C_{63} & C_{64} & C_{65} & C_{66} & -e_{16} & -e_{26} & -e_{36} \\ e_{11} & e_{12} & e_{13} & e_{14} & e_{15} & e_{16} & \kappa_{11} & \kappa_{12} & \kappa_{13} \\ e_{21} & e_{22} & e_{23} & e_{24} & e_{25} & e_{26} & \kappa_{21} & \kappa_{22} & \kappa_{23} \\ e_{31} & e_{32} & e_{33} & e_{34} & e_{35} & e_{36} & \kappa_{31} & \kappa_{32} & \kappa_{33} \end{bmatrix} \begin{bmatrix} \varepsilon_{11} \\ \varepsilon_{22} \\ \varepsilon_{33} \\ 2\varepsilon_{23} \\ 2\varepsilon_{31} \\ 2\varepsilon_{12} \\ E_1 \\ E_2 \\ E_3 \end{bmatrix} \quad (4)$$

In the above equation, the 18 unique coefficients of the piezoelectric tensor are written in the form of a 3×6 piezoelectric matrix, which will be used hereafter to denote the piezoelectric property space. The conversion of piezoelectric coefficients from stress-charge to strain-charge and vice versa are shown in Supporting Information Eq. 5. The effective electrical, mechanical, and electromechanical properties of a unit cell defined by its geometry and topology are computed using a finite element-based homogenization with periodic boundary conditions, as explained in the Supporting Information Section 1. However, a method to discover the geometry and topology of the unit cells corresponding to desired properties is needed to achieve a tailored piezoelectric response. Next, we develop an ML-based design methodology to achieve this.

ML-based optimization for tailored piezoelectricity

Truss metamaterials offer great tunability of physical properties due to virtually infinite possible topologies and geometries. This comes at the cost

of higher complexity in the design space and makes design and optimization difficult. The parameterization of these truss metamaterials is often discrete and high-dimensional. The complexity is further enhanced by the thin features, which make traditional optimization methods computationally expensive to implement for inverse designing to achieve desired properties. We explore the vast and highly complex design space of truss metamaterials and their respective piezoelectric properties by selecting two fundamentally different design spaces, each with its own advantages. These design spaces were originally developed for mechanical metamaterials and were used to train inverse design models for designing elastic properties^{22,23}. We adapt these design spaces and develop the optimization framework for designing the piezoelectric behavior of truss metamaterials. The first design space, based on geometric transformations, provides a broader range of piezoelectric responses and can activate all 18 piezoelectric coefficients. However, these geometric transformations lead to aspect ratios and rotations that are unfavorable for manufacturing. The second design space is restricted to a cubic symmetry and thus can only activate the five already non-zero coefficients of the base materials, with a broader range of values and improved manufacturability compared to the first design space. In the following, we present two different ML frameworks to suit the nature of the parameterizations and efficiently optimize in these design spaces.

Design space I: The first design space, introduced by Bastek et al.²², which was originally inspired by Zok et al.³⁶, is based on the superposition of a small set of elementary unit cells, which are subsequently stretched and rotated twice to get the final unit cell. Each unit cell is described by the following design parameters: (i) three elementary unit cells $L_1, L_2, L_3 \in \{1, \dots, 7\}$ chosen out of 7 options, (ii) their respective tessellations $t_1, t_2, t_3 \in \{1, 2\}$ (i.e., each cell tessellated as $1 \times 1 \times 1$ or $2 \times 2 \times 2$) following which they are superposed, (iii) the eigenvalues $U_1, U_2, U_3 > 0$ of the first affine stretch tensor $\mathbf{U} = \text{diag}(U_1, U_2, U_3)$, (iv) the first rotation tensor $\mathbf{R}_I \in \text{SO}(3)$, (v) the eigenvalues $V_1, V_2, V_3 > 0$ of the second affine stretch tensor $\mathbf{V} = \text{diag}(V_1, V_2, V_3)$, (vi) the second rotation tensor $\mathbf{R}_{II} \in \text{SO}(3)$, and lastly, (vii) the relative density ρ of the final unit cell achieved by choosing uniform diameter struts. The rotation tensors are defined by three unique parameters in axis-angle representation. Thus, each truss lattice is defined by the set of design parameters $\Theta = \{L_1, L_2, L_3, t_1, t_2, t_3, U_1, U_2, U_3, \mathbf{R}_I, V_1, V_2, V_3, \mathbf{R}_{II}, \rho\}$. Three example unit cells generated using this parametrization and their corresponding parameters are shown in Fig. S3a. Using this approach, one can obtain 262 unique topologies and virtually infinite designs. We used a set of 3,000,000 unit cell designs generated with this parametrization and performed numerical homogenization to obtain the pairs of unit cell designs and piezoelectric properties. Due to the rotation and stretching, we observe that all 18 components of the effective piezoelectric matrix \mathbf{e} are activated. Further, we note that certain components correlate highly with others (see Section 3 of the Supporting Information).

We train a forward model $\mathcal{F}_\omega : \Theta \rightarrow \mathbf{e}$ based on a neural network (NN) that maps both geometric and topological design parameters Θ of the truss lattices to the effective piezoelectric matrix \mathbf{e} . Here, ω denotes the trainable parameters (weights and biases) of the NN architecture. We use a multi-layer perceptron (MLP) architecture for the NN; see Supporting Information Section 2 for details. In Supporting Information Fig. S2a, we illustrate the prediction performance of the forward model with parity plots.

Once trained, the forward model can be used as a surrogate structure-property map for design optimizations, as shown in Fig. 2a. Here, we demonstrate this capability by designing a truss metamaterial with maximal hydrostatic coefficient e_h

$$e_h[\mathbf{e}] = e_{31} + e_{32} + e_{33}, \tag{5}$$

which quantifies the energy harvesting performance of a piezoelectric material under hydrostatic pressure, with higher being better. For converse piezoelectricity (actuation), e_h signifies the volumetric expansion under an electric field in direction 3. Since in most piezoelectric materials, e_{33} is positive while e_{31} and e_{32} are negative, very low magnitudes of e_h are

observed, leading to poor performance. To achieve an improved e_h , we formalize the design optimization problem as:

$$\begin{aligned} \Theta^* &\leftarrow \arg \max_{\Theta} e_h^*[\mathbf{e}], \quad \text{with } \mathbf{e} = \mathcal{F}_\omega(\Theta) \quad \text{and} \\ e_h^*[\mathbf{e}] &= e_h[\mathbf{e}] - \lambda_1 \underbrace{\left((e_{31} - e_{32})^2 + (e_{31} - e_{33})^2 + (e_{32} - e_{33})^2 \right)}_{\text{magnitude regularization}} \\ &\quad - \lambda_2 \underbrace{\left(\sum_{i=1}^6 e_{1i}^2 + \sum_{i=1}^6 e_{2i}^2 + \sum_{i=4}^6 e_{3i}^2 \right)}_{\text{shrinkage regularization}}, \end{aligned} \tag{6}$$

where Θ^* is the set of optimized design parameters. The second term with scaling hyperparameter $\lambda_1 > 0$ aids in regularizing the magnitudes of all three components so that we can achieve close-to-uniform volumetric expansion. The third term with scaling hyperparameter $\lambda_2 > 0$ serves as a shrinkage regularization to prevent distortions such as shear during actuation. To avoid extreme aspect ratios and improve the manufacturability of the optimized designs, we constrained the optimization process such that the stretch values \mathbf{U} and \mathbf{V} are scaled down by a factor of $\alpha \in (0, 1]$. The problem in (6) is solved numerically using gradient-based optimization, with Adam optimizer. All hyperparameters, training, and optimization protocols can be found in the Supporting Information Section 2.

Figure 2c shows the trace followed by the optimization framework, along with example lattices for different checkpoints. The categorical parameters related to topology lead to fluctuations in the unit cell designs and, consequently, the target property. Figure 2e shows the distribution of the dataset in the e_{31} vs. regularized hydrostatic coefficient e_h^* landscape. The optimized lattice lies outside the dataset, demonstrating the extrapolatory capability of the optimization and ML framework. Additional examples of tailored piezoelectricity are presented in the Supporting Information, Section 4, Fig. S6, which demonstrate the unconventional behaviors of full anisotropy and unidirectionality of piezoelectricity.

Design space II: The second design space, introduced by Zheng et al.²³, is based on a more general approach of generating highly complex unit cells with cubic symmetry. Each unit cell is created with a fixed set of virtual nodes, which can be offset freely along specific degrees of freedom, and the nodes are activated by connecting them to other nodes. Although the generated unit cells are of orthotropic nature, a vast range of stiffness values and orthotropic anisotropy can be achieved. Analogously, the design space is virtually infinite geometrically with at least millions of unique topologies if not more. Each unit cell is constructed by designing the lattice within the octant of a unit cube, which is then mirrored to obtain the unit cell with cubic symmetry and periodic tileability. There are 27 virtual nodes in the octant, i.e., 8 vertex nodes $\{\mathbb{V}_1, \mathbb{V}_2, \mathbb{V}_3, \mathbb{V}_4, \mathbb{V}_5, \mathbb{V}_6, \mathbb{V}_7, \mathbb{V}_8\}$, 12 edge nodes $\{\mathbb{E}_1, \mathbb{E}_2, \mathbb{E}_3, \mathbb{E}_4, \mathbb{E}_5, \mathbb{E}_6, \mathbb{E}_7, \mathbb{E}_8, \mathbb{E}_9, \mathbb{E}_{10}, \mathbb{E}_{11}, \mathbb{E}_{12}\}$, 6 face nodes $\{\mathbb{F}_1, \mathbb{F}_2, \mathbb{F}_3, \mathbb{F}_4, \mathbb{F}_5, \mathbb{F}_6\}$, and one body node $\{\mathbb{T}_1\}$. While the vertex nodes remain fixed, the edge nodes, face nodes, and body node can move freely on the corresponding edge, face, and inside of the octant, respectively. These node offsets are represented vectorially as $\mathbf{x} \in \mathbb{R}^{27 \times 3}$. The connectivity is represented by a 27×27 adjacency matrix $\mathbf{A} \in \{0, 1\}^{27 \times 27}$. For $i \neq j$, the entry $A_{ij} = 1$ if nodes i and j are connected by a strut, and $A_{ij} = 0$ otherwise. Diagonal entries follow the rule: $A_{ii} = 1$ if node i is part of the topology, and $A_{ii} = 0$ if it does not exist. A truss lattice is then uniquely defined by the graph $\Xi = (\mathbf{A}, \mathbf{x})$. Figure S3b shows three example unit cells generated with this parametrization. Similar to the first design space, we homogenized a set of 965,685 truss metamaterial designs and obtained their piezoelectric properties.

However, unlike the first design space, the graph-based representations are extremely discontinuous, high-dimensional, and combinatorial. To unify the discrete designs, the variational autoencoder (VAE) framework of Zheng et al.²³ is employed. The VAE consists of two neural networks. An encoder \mathcal{E}_ϕ , with trainable parameters ϕ , maps the graph of a truss lattice Ξ to a smooth low-dimensional latent space $\mathbf{z} \in \mathbb{R}^d$ of d -dimensions, i.e.,

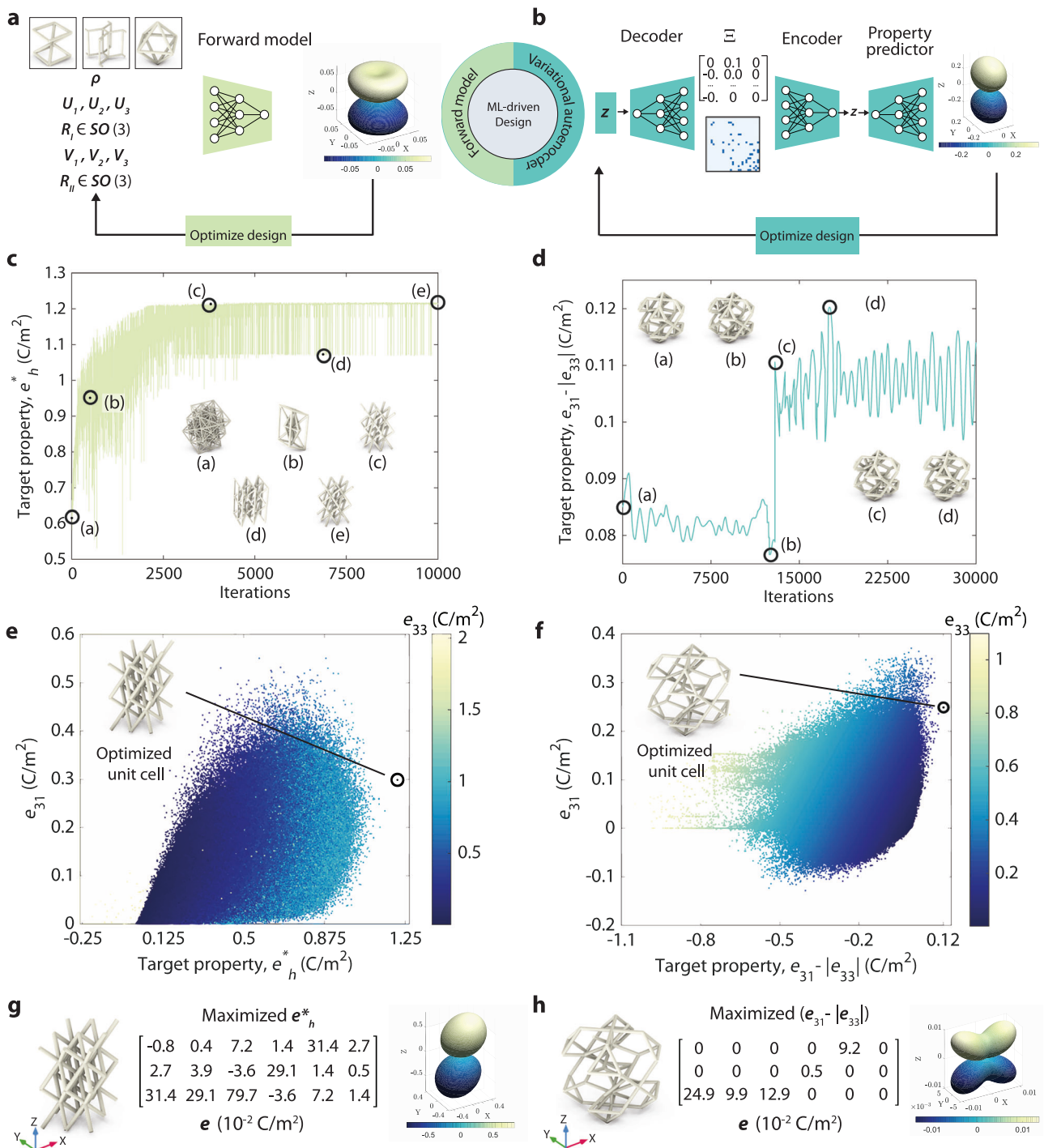


Fig. 2 | ML-based optimization of piezoelectric metamaterials. **a** The optimization framework employed for the design space I. A pre-trained forward neural network (NN) is used to perform gradient-based optimization. **b** The variational autoencoder (VAE) architecture employed to perform optimization in the lower-dimensional and continuous latent space for design space II. The additional encoding-decoding step is performed to ensure the validity of the predicted designs. The trace followed during the optimization, and different designs traversed during the path for (c).

Dataset I and (d). Dataset II. **e** The e_{31} vs. target property e_h^* for dataset I, and (f) the e_{31} vs. target property ($e_{31} - |e_{33}|$) landscape for dataset II. In both cases, the achieved optimized lattices, marked by the black circles, lie beyond the dataset and demonstrate the extrapolatory capability of the optimization framework. **g** An example optimized unit cell is shown, where the hydrostatic piezoelectric coefficient e_h^* is maximized. **h** An example optimized lattice with the optimization objective to maximize ($e_{31} - |e_{33}|$).

$z = \mathcal{E}_\phi(\Xi)$. A decoder \mathcal{D}_ζ with trainable parameters ζ reconstructs the graph of the truss lattice from the latent vector, i.e., $\Xi = \mathcal{D}_\zeta(z)$. The NNs are trained such that the truss lattices input into the encoder are accurately reconstructed by the decoder, while modeling the latent space to closely follow a standard normal distribution. Given sufficient reconstruction accuracy over a representative dataset, the latent space serves as an

informational bottleneck, providing a continuous and smooth, though abstract, vectorial representation of the design space. New designs can be generated by stochastically sampling the latent space via standard normal distribution and decoding into the graph representation, i.e., $\mathcal{D}_\zeta(z)$ with $z \sim \mathcal{N}(0, \mathbf{I})$. Simultaneously, a *property-predictor* neural network $\mathcal{P}_\psi : z \rightarrow e$ with trainable parameters ψ is trained to predict the effective

piezoelectric matrix e from the latent representation. Figure S2b of the Supporting Information illustrates the prediction performance of the forward model with parity plots. The encoder, decoder, and property predictor are trained jointly, as detailed in Supporting Information Section 2, enabling the latent space to be learned in a manner that facilitates efficient structure-property map exploration.

The pre-trained VAE framework is then used to optimize within the latent space to discover truss lattices that achieve a tailored electro-mechanical response, as shown in Fig. 2b. The optimization process starts with an initial guess \mathbf{z} from the latent space, which is decoded into the corresponding unit cell architecture $\Xi = \mathcal{D}_\zeta(\mathbf{z})$. To ensure the validity of the architecture²³, it is re-encoded back into the latent space $\mathbf{z}' = \mathcal{E}_\psi(\Xi)$. The property predictor is used to estimate the effective piezoelectric matrix, i.e., $e = \mathcal{P}_\psi(\mathbf{z}')$. The objective function is formulated in terms of e . A gradient-based optimization algorithm is then employed to iteratively update the latent vector \mathbf{z} ; the optimal \mathbf{z} is then decoded into optimal truss lattice design. As an example, we aim to maximize piezoelectric response in transverse mode while minimizing the influence of longitudinal mode, i.e.,

$$\mathbf{z}^* \leftarrow \arg \max_{\mathbf{z}} (e_{31} - |e_{33}|), \quad \text{with } e = \mathcal{P}_\psi(\mathbf{z}') \quad \text{and} \quad \mathbf{z}' = \mathcal{E}_\psi(\mathcal{D}_\zeta(\mathbf{z})). \quad (7)$$

Having such a piezoelectric response is advantageous in selectively activating transverse mode while minimizing the influence of longitudinal mode, and thus providing a unidirectional sensor or actuator even under multi-directional loading.

The trace of the optimization in Fig. 2d, along with example lattices from checkpoints along the trace, shows the different designs and their properties traversed by the optimization algorithm in the process. Figure 2f shows the training dataset properties in the e_{31} vs. target property ($e_{31} - |e_{33}|$) landscape. Similar to Fig. 2e, the optimized unit cell lies outside the bounds of the training dataset, showing an improvement of 18.42% compared to the best value of target property in the dataset. Details of the optimization and training protocols are presented in Supporting Information Section 2.

Figure 2g, h show the optimized unit cells for the representative design problem from both the design spaces, their corresponding piezoelectric matrices, and piezoelectric surface plots. The spatial variation of the surface plots is such that if a vector is drawn from the center of the plot to any point on the surface, the length of the vector represents the magnitude of the longitudinal piezoelectric coefficient, \tilde{e} in the direction of that vector. Here, $\tilde{e}(\mathbf{p})$ is defined as the piezoelectric coefficient coupling axial stress in direction \mathbf{p} to the electric field in direction \mathbf{p} . Along the direction $\mathbf{p} \in S^2$, where S^2 is a unit sphere in three dimensional space, $\tilde{e}(\mathbf{p})$ is evaluated as

$$\tilde{e}(\mathbf{p}) = \sum_{i,j,k=1}^3 e_{ijk} p_i p_j p_k. \quad (8)$$

While the first optimization example (Fig. 2g) shows a maximized e_{31} , the second example (Fig. 2h) shows a unidirectionality in electromechanical response, i.e., enhanced e_{31} and minimal e_{33} . In the former case, increased e_{31} and auxetic piezoelectric response provide higher energy harvesting and superior sensing capabilities due to the synergistic effect of transverse and longitudinal modes of operation, which are useful in underwater applications. In the latter case, isolating modes of operation in such a way is useful for transducers for a reduced effect of ambient acoustic and vibration noise, as well as in micro-robotics for specialized motion. Moreover, while e_{31} , and e_{32} are negative and e_{33} is positive for most of the conventional piezoelectric materials, in the truss metamaterials presented here, they are all positive, leading to auxetic piezoelectricity. This enables a synergistic interplay of these coefficients in multi-directional loading scenarios, unlike conventional piezoelectric materials, where the interplay is often destructive due to the opposing signs of these coefficients.

To complement the design methodologies and practical realization of truss metamaterials with tailored piezoelectricity, an efficient and accurate fabrication route is needed. Next, we present a custom technique for additively manufacturing these metamaterials.

In-gel direct ink writing of piezoelectric metamaterials

Additive manufacturing of piezoelectric truss metamaterials is challenging due to the brittle nature and polarization-dependent behavior of the base piezoelectric materials. Additionally, the 3D slender features with varying orientations complicate the process, especially when relying solely on ink rheology to produce free-standing structures.

To address this limitation, we developed a direct-ink-writing (DIW) methodology for lead-free piezoelectric transducers. The ink used for printing consists of 3% Li-substituted sodium potassium niobate $\text{K}_{0.485}\text{Na}_{0.485}\text{Li}_{0.03}\text{NbO}_3$ (KNLN), in combination with UV curable monomer poly-(ethylene glycol) diacrylate, PEGDA-700 (referred to as *piezo ink*). The KNLN particles, with an average size of 3 μm , were chosen for their high piezoelectric charge constants among lead-free piezo-ceramics and their biocompatibility. Stability and chemical compatibility between the ink and gel are crucial to prevent diffusion or reactions. A hydrophobic medium like silicone oil was chosen to prevent any diffusion of the hydrophilic PEGDA 700 monomer, which was observed in hydrophilic gels. The piezoelectric metamaterials were printed and embedded inside a supporting silicone matrix (referred to as *gel*). This method enables complex geometries without support structures, as struts are printed in a single direction, ensuring directional homogeneity similar to designed metamaterials and fast manufacturing. However, the extruder path must be optimized to avoid collisions and ensure end-connected struts. We used Fleury's algorithm-based approach proposed by Weeks et al.³⁷ to optimize the path traversed by the extruder, avoiding collisions and minimizing travel motions.

The design of the piezoelectric and the support medium must satisfy certain key chemical and rheological requirements to enable the successful embedded printing of lattices. Similar to the requirements of the materials used for DIW in Ammu et al.³³, both the piezo ink and gel need to be shear-thinning yield stress materials, allowing material flow under applied pressure from the extruder. Additionally, the support gel must be thixotropic, enabling it to recover its viscoelastic behavior after shear thinning. This ensures adequate support for the printed ink once the nozzle has moved past, allowing the gel to regain its elasticity and support the weight and shape of the printed ink. The detailed manufacturing steps for the ink and gel are described in the Materials and Methods section.

Ink design. Using the synthesized KNLN particles, we found that ceramic volume fractions in the range between 42 and 47% exhibit shear thinning behavior, along with being able to be printed using our DIW setup. Below this volume fraction, the yield stress of the piezo ink was too low, causing either uncontrolled flow from the nozzle, even when at rest, or sedimentation of the particles, leading to nozzle clogging and liquid phase migration. The low yield stress of the piezo ink led to fragmentation and beading of the extruded ink due to the yield stress of the ink being lower than the surface tension of the gel, similar to what was observed by Pairam et al.³⁸. On the other hand, if the piezo ink's elastic modulus and yield stress are too high, printed filaments get dragged through the matrix by the nozzle, leading to poor connectivity at the nodes as shown schematically in Fig. 3a. Hence, a volume fraction closer to the lower end of the range, at 43 vol%, was chosen to maintain a high enough yield stress to allow stable extrusion while not being too viscous. This allows for designing a shear-thinning, thixotropic gel medium whose viscosity can be matched to that of the ink.

Gel design. The gel was designed to have a high shear modulus to sufficiently support the printed filament's weight and prevent sagging,

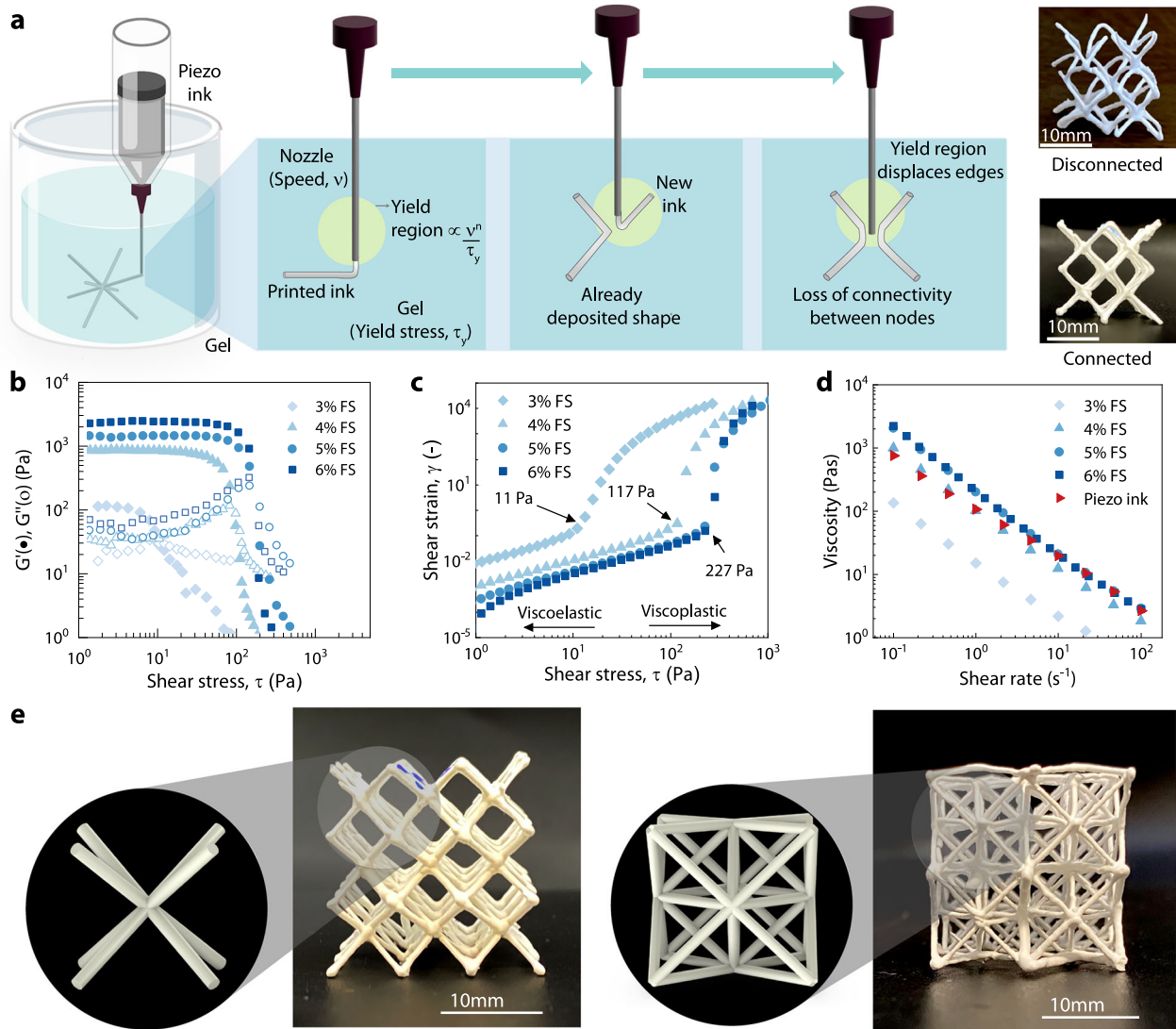


Fig. 3 | Direct ink writing of piezoelectric metamaterials. **a** Schematic of in-gel direct ink writing process, and the optimization of ink rheology and printing parameters to ensure the connectivity at the nodes, **(b, c)** shear strain variation with applied shear stress for different volume fractions of fumed silica in the gel. The yield stress points, i.e., 11 Pa, 117 Pa, and 227 Pa, are indicated in **(c)**, which are determined from the intersection of tangents at the inflection point of the log-log plots. **d** Varying viscosity of the gel with increasing shear stress for different volume fractions of fumed silica. **e** Printed samples of body centered cubic (BCC) (left) and

octet (right) lattices. While the tuned rheological properties of the gel allow for the printing of the BCC topology with overhanging struts, the fine-tuned printing speed and ink rheology allow for accurate connectivity of octet topology with a higher number of nodes. The high volume fraction of piezoelectric particles, which leads to the good piezoelectric performance, is detrimental to the printing resolution and can lead to some defects. Nevertheless, we demonstrate the ability to fabricate different topologies with good accuracy and structural integrity.

which can occur due to the density mismatch between the ink and the support matrix, especially for ceramics like KNLN with a high density of 4.2 g/cc. Fumed silica (FS) was chosen as the filler for the silicone matrix preparation due to its ability to provide an elastic network structure with sharp shear thinning behavior. Gels with less than 3% FS showed insufficient yield stress to be viscoelastic, leading to the investigation of gels with more than 3% FS as support media, as shown in Fig. 3b, where the shear storage (G') and loss (G'') moduli are plotted against applied shear stress.

The yield stress of the gel must be low enough to allow the nozzle to move through it freely and enable the gel to recover behind the nozzle. Printing inside a lower-yield stress gel, such as 3% FS gel with a yield stress of 11 Pa, resulted in highly distorted and unconnected lattices, as shown in Fig. 3a. This issue arises because of the large size of the stress field from the moving nozzle in a low-yield stress gel, which interferes with previously extruded ink. The size of the stress field is dependent on the Oldroyd

number, defined as³⁹

$$OD = \frac{\tau_y \Lambda^n}{\mathbb{K} U^m}, \tag{9}$$

where τ_y is the yield stress, \mathbb{K} is a constant determined by the matrix composition, n is a flow behavior index, Λ is the nozzle diameter, and U is the print speed.

The objective was to maximize the Oldroyd number and subsequently minimize the stress field around the nozzle. Increasing the volume fraction of FS in the gel resulted in stiffer gels, as evidenced by the increase in yield stress of the gels from 11 to 221 Pa (Fig. 3c), leading to improved shaping fidelity of the piezo ink. However, excessively high yield stress in the gel can cause static crevices that do not collapse or self-heal, resulting in print distortions. The self-healing and collapse of crevices depend on printing depth and the rheological properties of the support matrix. Gels with more

than 4% FS showed significant crevices and viscoplastic behavior. Therefore, a gel with 4% FS, yielding a stress of 117 Pa, was chosen for its sufficient self-healing properties upon nozzle translation.

Printing speed design. For the chosen piezoelectric ink of 43 vol% and gel of 4 wt% FS, the print speed was used as a final control parameter to minimize the yield region. Viscosity vs. shear rates for both the gel and ink were measured at typical printing speeds, showing a match in the shear rate region between 0.1 s^{-1} and 0.7 s^{-1} as seen in Fig. 3d.

The flow behavior index n for these materials was then calculated by using the Oswald-de Waele power law for shear-thinning fluids

$$\eta = \mathbb{K} \dot{\gamma}^{n-1}, \quad (10)$$

where, η is the viscosity, $\dot{\gamma} [\text{s}^{-1}]$ is the shear rate. Using n derived from the above equation, the range of printing speeds was determined based on the shear rate and nozzle diameter and was found to be between 2 mm/min and 50 mm/min. Low printing speeds (2 mm/min) minimized the yield region but resulted in higher print times and potential sedimentation and clogging of the piezo ink. Printing was conducted at two speeds: the bulk of the lattice (95% of each strut's length) was printed at a higher speed (50 mm/min) within the matched viscosity regime, while the remaining 5% near the nodes was printed at a lower speed (5 mm/min) to ensure minimal disturbance of the printed lines. This strategy enabled relatively quick printing while ensuring well-connected and non-distorted lattices at the nodes, as shown in Fig. 3a. Multiple lattices with different unit cells were then printed, as depicted in Fig. 3e. While our in-gel technique can print the truss structures with good accuracy, the complexity of the designs and sharp edges can influence the joints of struts. The oozing of ink from the nozzle during travel motions requires additional considerations. For the BCC and octet structures, it was sufficient to tune the printing speed and rheology to avoid unwanted *spikes* in the direction of the travel, which originate from the pressure build-up in the nozzle. This oozing of the ink can also be reduced by using a smaller particle size. However, this increases the viscosity of the ink and limits the overall volume fraction that can be achieved during printing. Other potential ways to avoid the overflow of the ink are implementing a sophisticated retraction mechanism for precise flow control or avoiding the use of gel altogether by simultaneous extrusion and curing of the ink during printing. We utilized a simple retraction mechanism by screw reversal for the optimized unit cell, where an excessive ink overflow was observed. Next, we explore the piezoelectric response of the metamaterials, the directionality enabled by their design, and compare their response with conventional bulk materials.

Tailored piezoelectricity of truss metamaterials

In this section, we demonstrate the directionality of the piezoelectric response in truss metamaterials by testing 3D-printed $2 \times 2 \times 2$ tessellations of octet unit cells. To show the architecture-driven electromechanical responses of these samples, we measured them in longitudinal and transverse modes. A standard Berlincourt d_{33} -meter was used to apply a harmonic force on the samples, and the generated voltage (ϕ) was recorded using a pico logger, as shown in Fig. 4a. Multiple electromechanical phenomena, such as triboelectricity and flexoelectricity, can often co-exist in such systems and might influence accurate piezoelectric characterization. To eliminate the possible influence of triboelectricity due to contact-separation and sliding of different surfaces, all the measurements were conducted while maintaining a static force. Due to the macro scale size of the samples, the generated strain gradients are low, and thus the influence of flexoelectricity can be considered negligible⁴⁰. With the robust measurement strategy, we turn to measure the directional response of the octet samples and show the directionality of response in piezoelectric truss metamaterials.

The amplitude of voltage in transverse and longitudinal modes is compared for the octet sample in Fig. 4b. While for the longitudinal mode, the force is applied to the electrodes, for measurements in the transverse mode, the samples are subjected to the harmonic excitation on the face

normal to direction 1. The well-known 50 Hz noise is observed to have a significant magnitude compared to the piezoelectric signal. This noise can be easily eliminated since the measurements were performed at a forcing frequency of 230 Hz, as later shown in Fig. 4d.

Due to lower resonant frequencies compared to bulk samples of similar dimensions, the piezoelectric response of piezoelectric truss metamaterials was observed to be highly sensitive to the forcing frequency. The voltage responses of an octet sample and a standard bulk PZT sample were measured at different frequencies in the range of 5–300 Hz, as shown in Fig. 4c. The PZT sample shows a linear increase in the voltage response with increasing forcing frequency, indicating that none of the resonance frequencies are close to the forcing frequencies. Thus, the equipment for measuring the piezoelectric coefficients of these samples is usually calibrated at these low frequencies for reliable estimates. For the octet sample, an exponential increase in voltage response with a peak at 250 Hz is observed, indicating mechanical resonance. This means that the harmonic force or impact measurements, commonly used for bulk piezoelectric samples, are not well-suited for characterizing the piezoelectric coefficients of the truss metamaterials. Therefore, to avoid the influence of mechanical resonance on characterization, we restrict ourselves to evaluating only the transient response of truss metamaterials in this low-frequency regime.

We further demonstrate the directionality of piezoelectric response in metamaterials by measuring the voltage response of octet samples in transverse and longitudinal modes. The frequency domain voltage response of the octet sample under a dynamic load of 230 Hz is compared in transverse and longitudinal modes in Fig. 4d. The voltage peak at 230 Hz in longitudinal mode is found to be approximately 65 times the peak in transverse mode due to the high d_{33}/d_{31} ratio of the octet unit cells, in contrast to the base composite material, where this ratio is approximately 2. This shows the effectiveness of the truss metamaterials in providing tailored piezoelectric response. An important factor to be considered for the physical realization of this concept is the compromise between the tunable piezoelectric properties and manufacturability constraints.

The two design spaces presented in this study provide different combinations of manufacturability and range of properties. Design space I can offer more diverse piezoelectric responses compared to design space II. Figure 4e shows the distribution of the fraction of non-zero coefficients of the piezoelectric matrix in design space I, which is defined as the ratio of number of non-zero piezoelectric coefficients in the piezoelectric matrix to the total number of coefficients. Most of the designs are found to have all 18 coefficients non-zero, indicated by a close to one probability distribution of the non-zero fraction of one. For design space II, most of the unit cells in the dataset are found to have a non-zero components fraction of 5/18, since only five of the 18 coefficients are non-zero, as shown in Fig. 4f. While limited to only five coefficients, all the unit cells in design space II have a cubic symmetry, leading to improved manufacturability. As shown in Figure Supporting Information Fig. S5a, the dataset from design space I shows a narrower range of coefficients e_{31} and e_{32} , restricted to only positive values. The distribution of e_{31} and e_{32} for dataset II in Figure Supporting Information Fig. S5b shows these coefficients ranging from negative to positive values. Thus, the trade-off between manufacturability and range of properties should be considered depending upon the manufacturing capabilities available and the intended application. These distinct features of the two design spaces stem from the different parameterizations used to construct them.

We explored the two design spaces for the influence of specific design parameters on the unit cell designs and their piezoelectric properties. Figure 4g shows the principal component analyses (PCA) conducted in the latent space of the VAE framework for design space II. Since the latent space was trained in conjunction with the property predictor NN, the latent space parameters are intertwined with the piezoelectric properties. A strong correlation between the principal component 2 and the piezoelectric coefficient e_{31} was observed. The principal component 1 has a weaker influence on e_{31} , while the geometry still evolves along principal axis 1, as shown in Supporting Information Fig. S7a, b, along with the evolution of the unit cell

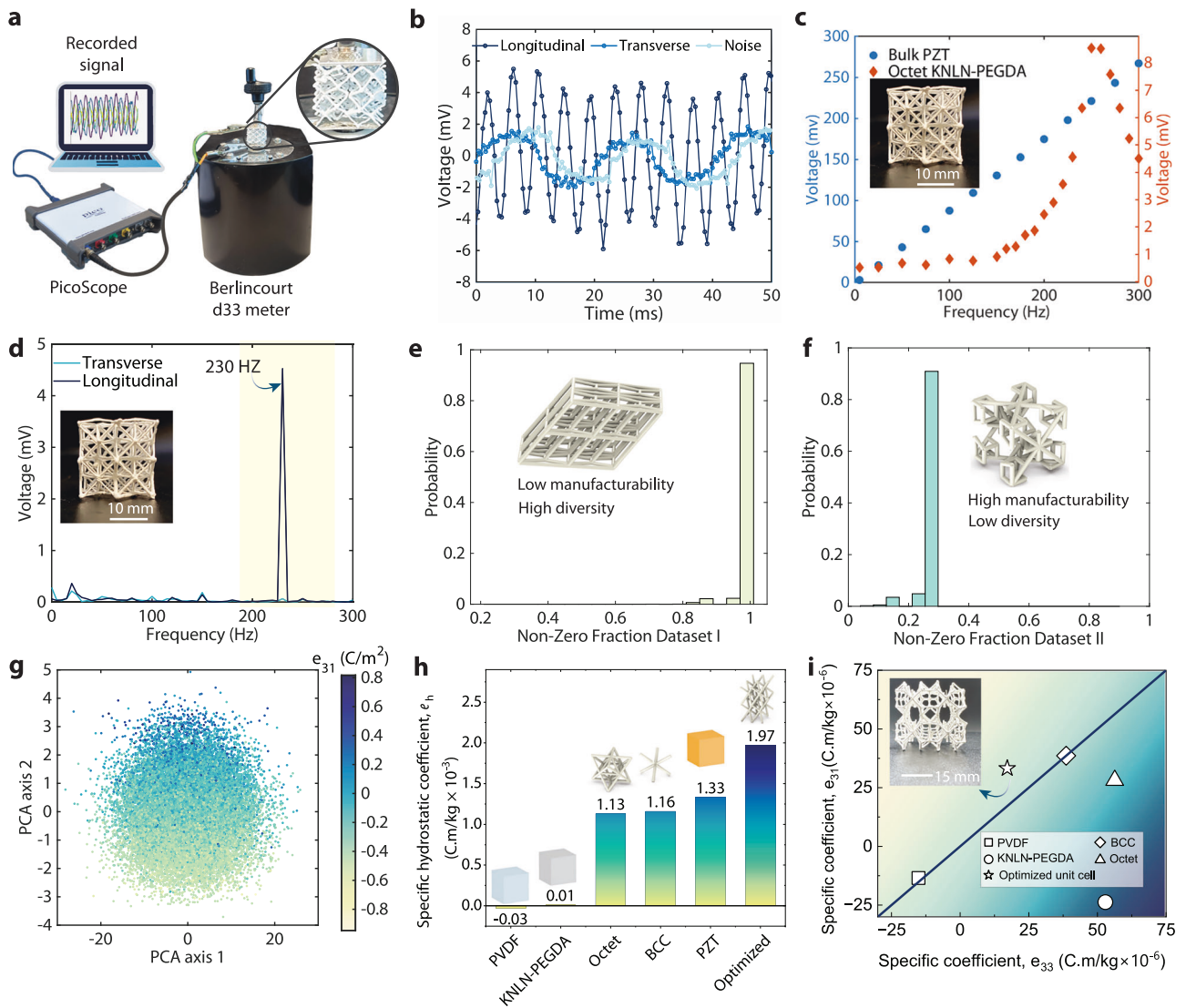


Fig. 4 | Evaluation of the tunable electromechanical response of piezoelectric truss metamaterials. **a** Schematic of the measurement setup used for recording the transient response of printed samples. The electrodes were fixed to the samples at the top and bottom surfaces. The force was applied either in the longitudinal direction or in the transverse direction, depending on the mode of operation. **b** Transient voltage response of octet metamaterial in longitudinal and transverse modes under a dynamic load of 0.5 N. **c** The measured frequency vs voltage response of an octet metamaterial and a bulk PZT sample. The bulk PZT sample shows a linear increase in voltage, while for octet samples, a peak is observed at 250 Hz. **d** The frequency domain response of the octet samples under a harmonic force of 0.5 N at 230 Hz. The voltage peak at 230 Hz in longitudinal mode is found to be significantly higher than the transverse mode, as predicted by the computational homogenization. **e, f** The comparison of the two design spaces utilized for optimization in terms of non-zero

fraction, which is the ratio of number of non-zero components in the piezoelectric matrix to the total number of components. Design space I offers all 18 piezoelectric coefficients but lacks manufacturability due to extreme aspect ratios and skewed geometries. Design space II activates only five piezoelectric coefficients with a broader range and better manufacturability due to cubic symmetry. **g** PCA performed in the latent space of dataset II, where, with the increasing values of principal component 2, the value of e_{31} increases. **h** The comparison of specific hydrostatic piezoelectric coefficient of the optimized unit cell with bulk piezoelectric materials, BCC, and octet metamaterials. **i** The e_{31} vs e_{32} plot for different piezoelectric materials shows that while most piezoelectric materials have e_{33} higher than e_{31} and lie below the 45° slope line, the ML-optimized unit cell is able to reverse this trend and has higher e_{31} . The printed sample with the ML-optimized unit cell is shown in the inset.

designs. Supporting Information Figure S7c shows PCA for the parameter space of design space I. We found that the principal components of design space I do not have a direct correlation with the piezoelectric coefficient e_{31} . This is due to the discrete nature of the design space I and small changes in design parameters resulting in sudden jumps in piezoelectric properties, as was also observed during the optimization in Fig. 2c. We take one example of optimization performed in each design space and compare their piezoelectricity with bulk materials.

We compared the piezoelectric performance of the metamaterials for application-relevant metrics. The optimized lattice obtained by maximizing the hydrostatic coefficient in design space I (Eq. (6)) outperforms the commonly used bulk materials, as well as the KNLN-PEGDA composite.

The optimized truss metamaterial offers a lightweight alternative for energy harvesting, actuation, and sensing applications under hydrostatic loads and uniform actuation, such as in underwater sensing applications. In applications where unidirectional sensing or actuation is required, the isolation of modes of piezoelectricity becomes crucial. We explored one such possibility, where higher transverse piezoelectricity is desired compared to longitudinal piezoelectricity, by solving the optimization problem in Eq. (7). Figure 4i shows the comparison of the optimized lattice with the bulk materials in the normalized e_{33} vs e_{31} landscape. The printed optimized lattice with $2 \times 2 \times 2$ tessellations is shown in the inset. While printing the optimized unit cell, the issue of oozing of the ink during travel motions, as discussed earlier, was found to be more pronounced. This is due to the smaller strut lengths and

more travel motions compared to the simple structures shown in Fig. 3e. We mitigated this by implementing a retraction step at the start of every travel motion, followed by an equal extrusion at the end of the travel move. For printing the optimized unit cell. Most of the common bulk materials, as well as the base material, lie below the unit slope line since the magnitude of e_{33} is higher than e_{31} for these materials, PVDF being an exception with a slightly higher value of e_{31} . The optimized lattice shows the highest value of normalized $e_{31} - |e_{33}|$ and lies beyond the unit slope line. This enables the pure transverse mode operation of piezoelectric materials even under mixed loading conditions, with potential applications in specialized motion in micro-robots, medical equipment, and unidirectional electromechanical transducers.

Discussion

Considering that most of the common bulk piezoelectric materials are restricted to the same symmetry class, our ML-based approach to designing new metamaterials offers clear advantages. Specifically, we achieved a hydrostatic piezoelectric coefficient for metamaterials higher than their bulk counterparts, and a significantly higher e_{31} value compared to e_{33} . The optimized properties of these metamaterials extend beyond the training design space, demonstrating the extrapolatory capabilities of our proposed ML framework.

Through optimization, we also discovered that designs constrained within cubic symmetry are limited to the standard five piezoelectric coefficients. Accessing additional coefficients requires skewed geometries or aspect ratios in the unit cells. Consequently, our strategy of exploring two design spaces maximizes both absolute performance and anisotropy or manufacturability. Our results show that truss metamaterials offer exceptional flexibility in tuning the piezoelectric response, activating all 18 coefficients, and enabling rare or unique combinations of coefficients not found in conventional materials. Principal Component Analysis of the latent space of design space II and design space I revealed differing behaviors. In design space II, the second principal component directly influenced the magnitude of e_{31} , while design space I exhibited no discernible trend.

In characterizing the response of 3D-printed octet metamaterials in longitudinal mode, we observed an exponential increase in the piezoelectric response at low frequencies, peaking in local resonance at 250 Hz. This finding suggests that standard characterization frequencies and methods for bulk piezoelectric materials may not be suitable for lattice structures. We propose that future studies further explore the frequency dependence of piezoelectric materials. Despite these considerations, lightweight lattice structures optimized using our ML framework outperform bulk materials in key metrics, such as specific hydrostatic performance or achieving dominant transverse piezoelectricity over longitudinal modes. To demonstrate the concept of piezoelectric truss metamaterials, we 3D-printed and tested the octet topology. While our 3D printing technique provides a way to print the truss metamaterials, further advancements, such as improved flow control of the ink, are warranted to translate this concept to real-world applications. Nonetheless, our developed optimization framework and 3D-printing technique provide a novel way to design piezoelectric materials, overcoming limitations in material and symmetry classes. These advancements enable the creation of multi-modal electromechanical devices with tailored behaviors, potentially useful in applications such as micro robotics and health monitoring devices. Our electromechanical testing also reveals the importance of considering structural dynamics in individual structures and designing specific mechanical and electrical boundary conditions for lattice-based piezoelectric materials.

This work advances the field of piezoelectricity by demonstrating the potential to optimize piezoelectric performance through structural design, transcending the constraints of conventional bulk materials' symmetries. Our proposed 3D-printing technique, coupled with a graph-based path planning framework, offers enhanced shaping freedom and precision, paving the way for advanced piezoelectric composite materials. These developments open new possibilities for electromechanical systems with engineered physical and responsive behaviors.

Methods

Numerical homogenization

For the base material used in simulations, isotropic elastic properties are assumed with Young's modulus $E^{\text{base}} = 60.606$ GPa and Poisson's ratio $\nu = 0.3$. For the piezoelectric properties, $e_{31}^{\text{base}} = e_{32}^{\text{base}} = -6.62281$ C/m², $e_{33}^{\text{base}} = 23.2403$ C/m², and $e_{15}^{\text{base}} = e_{24}^{\text{base}} = 17.0345$ C/m² are considered, while the relative dielectric permittivity is $\kappa_{11} = \kappa_{22} = \kappa_{33} = 1433.6$.

Ceramic synthesis

We synthesized the $\text{K}_{0.485}\text{Na}_{0.485}\text{Li}_{0.03}\text{NbO}_3$ (KNLN) ceramic powder through a conventional solid-state reaction with a two-step calcination process, reported previously⁴¹. Stoichiometric proportions of >99% Na_2CO_3 , K_2CO_3 , Li_2CO_3 , and Nb_2O_5 (Sigma Aldrich) were milled for at 200 rpm for 3 h with 5 mm yttria-stabilized ZrO_2 balls in hexane. Afterward, the milled powders were air-dried on a hot plate at 100 °C for 1 h, followed by calcination at 1050 °C for 3 h with a heating rate of 5 °C min⁻¹. After the initial calcination, the powder was milled in IPA for 3 h at 200 rpm, air-dried, and subjected to a second calcination at 925 °C for 20 h, with a slower heating rate of 1 °C min⁻¹. The resulting calcined powder was ball-milled again in IPA for 15 min, air-dried, sieved through a 90 μm mesh, and stored under vacuum. Particle morphology was examined using a Scanning Electron Microscope (SEM; JEOL JSM-7500F, Nieuw Venneep, Netherlands), and the particle size distribution was determined using a Malvern Mastersizer 3000. The analysis was performed on a 0.1% weight/volume aqueous suspension of the powders with sodium dodecyl sulfate as a surfactant. The mean value of the distribution represents the particle size of each type of powder.

Ink preparation

The synthesized ceramic powder was mixed with the UV-sensitive monomer poly(ethylene glycol) diacrylate with Mn 700 (Sigma Aldrich) in varying volume fractions. This mixture was combined using a planetary mixer (SpeedMixer DAC 150.1 FVZ) at 3500 rpm for 2 min, with 30-s intervals and a 15-s break to prevent excessive heating. Two photoinitiators, Irgacure 819 and 184, were added to maximize curing depth due to the high ceramic filler content, creating a UV-curable piezoelectric ink. The photoinitiators were added to the KNLN-PEGDA mixture in ratios of 1% and 2% by weight relative to the monomer and mixed at 1000 rpm for 5 min to ensure homogeneous distribution.

Support gel preparation

The gel is created by dispersing 4 wt% of fumed silica (Sigma Aldrich) in silicone oil. This mixture is processed in a planetary mixer (SpeedMixer DAC 150.1 FVZ) at 3500 rpm for 5 min to ensure thorough dispersion and uniformity.

Rheological characterization

The rheological properties of the inks were characterized using a rotational rheometer (Haake Mars III, Thermoscientific) with a 20 mm diameter serrated plate geometry using a gap height of 1 mm. Serrated plates prevent artifacts in measurements arising from wall slip, which has been studied to occur when highly loaded dispersions are subject to high localized deformation. Shear storage and loss moduli were determined as a function of shear strain via dynamic amplitude sweeps at a fixed frequency of 1 Hz with a stress sweep. The yield stress was measured via steady-state flow experiments with a sweep of shear stress and measuring. We used the point of change in slope of the log-log plot of shear stress vs strain to calculate the yield stress of the ink. The viscosity vs shear rates was measured between shear rates from 0.1 to 100 s⁻¹.

Printing and curing

A commercial desktop printer (Ultimaker 2+) was modified by replacing the original print head with a custom-made ink extrusion system consisting of a 25 ml syringe holder and a mechanically driven syringe pump. The inks loaded into syringes were attached with stainless steel nozzles of 1.5-inch

length with an inner diameter of 0.5 mm and were used for printing. The printed lines or parts were cured using an Omnicure S1500 (Lumen Dynamics) UV-lamp at 100% intensity at 20 cm for 30 s for the final part.

Poling

The printed samples are first cleaned in a Hexane solution to wash away the silicone oil and facilitate better electrode adhesion. A corona poling setup is used to pole the sample with 5 kV as the bottom plate potential and 50 kV applied to the pin electrodes. The base plate is heated to 120 °C to allow for the reorientation of the dipoles in the samples. Samples are poled for 6 h, after which they are cooled to room temperature with the electric field applied and used for measurements.

Piezoelectric characterization

Aluminum metal electrodes were attached to the faces, which were in the same direction as the poling direction (3), using a silver epoxy adhesive. The piezoelectric voltage was measured using a Berlincourt (PM 300, Piezo test, London, UK) piezometer with a static force of 2 N and a dynamic peak-to-peak sinusoidal excitation ranging between 0.05 N and 0.5 N at 110 Hz. The output voltage was measured using a Picoscope (5442D series).

Data availability

The dataset used for training the ML frameworks in this work is available at the repository⁴² (<https://doi.org/10.5281/zenodo.17041483>).

Code availability

The codes used for homogenization, ML training and optimization, and path planning are available at the repository⁴³, and <https://github.com/mmc-group/piezoelectric-truss-metamaterials-design>.

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Competing interests

Authors S.S., S.A., P.T., J.J., and K.M. declare no financial or non-financial competing interests. S.K. serves as an Editor of this journal and had no role in the peer-review or decision to publish this manuscript. S.K. declares no financial competing interests.

Additional information

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Author contributions

S.S. and S.K. developed the piezoelectric truss homogenization code. S.S. and P.T. trained the machine learning models and performed the design optimization. S.S. developed the path planning code for printing, S.K.A. performed the direct ink writing, and developed gel and ink formulation. S.S. and S.K.A. built the experimental setup for characterizing the fabricated structures and performed the measurements. S.S., J.J., K.M., and S.K. conceptualized the research. J.J., K.M., and S.K. supervised the research.