# Seeing mesoatomic distortions in softmatter crystals of a double-gyroid block copolymer 

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#### Abstract

Supramolecular soft crystals are periodic structures that are formed by the hierarchical assembly of complex constituents, and occur in a broad variety of 'softmatter' systems ${ }^{1}$. Such soft crystals exhibit many of the basic features (such as threedimensional lattices and space groups) and properties (such as band structure and wave propagation) of their 'hard-matter' atomic solid counterparts, owing to the generic symmetry-based principles that underlie both ${ }^{2,3}$.'Mesoatomic' building blocks of soft-matter crystals consist of groups of molecules, whose sub-unit-cell configurations couple strongly to supra-unit-scale symmetry. As yet, high-fidelity experimental techniques for characterizing the detailed local structure of soft matter and, in particular, for quantifying the effects of multiscale reconfigurability are quite limited. Here, by applying slice-and-view microscopy to reconstruct the micrometrescale domain morphology of a solution-cast block copolymer double gyroid over large specimen volumes, we unambiguously characterize its supra-unit and sub-unit cell morphology. Our multiscale analysis reveals a qualitative and underappreciated distinction between this double-gyroid soft crystal and hard crystals in terms of their structural relaxations in response to forces-namely a non-affine mode of sub-unitcell symmetry breaking that is coherently maintained over large multicell dimensions. Subject to inevitable stresses during crystal growth, the relatively soft strut lengths and diameters of the double-gyroid network can easily accommodate deformation, while the angular geometry is stiff, maintaining local correlations even under strong symmetry-breaking distortions. These features contrast sharply with the rigid lengths and bendable angles of hard crystals.


Three-dimensional (3D) tomographic imaging is the definitive experimental technique for determining the morphology of complex nanostructures. Tomography can be performed with a variety of microscopic methods, provided that there is a suitable match between the imaging resolution and the feature size. For block copolymer (BCP) structures in which periodicities are typically in the $10-100-\mathrm{nm}$ regime, with domain features on the scale of $2-20 \mathrm{~nm}$, electron microscopy techniques are generally required. To date, nearly all 3D tomograms of bulk-phase BCPs have been made using transmission electron microscopy (TEM) tomography ${ }^{4-6}$. Although powerful, this technique is quite limited with regard to the range of sample thicknesses, and inevitably incurs both sample deformation from microtomy and information loss associated with the restriction of tilt angles ${ }^{7}$.

Here we use the considerable advantages afforded by a wholly distinct approach to the tomography of nanostructured 3D morphologies:
slice-and-view scanning electron microscopy (SVSEM; also named focused ion beam scanning electron microscopy, or FIB-SEM) ${ }^{8-10}$. Crucially, by comparison with TEM tomography, SVSEM tomography can provide a much larger reconstruction in all three spatial dimensions, facilitating 3D analysis of volumes many orders of magnitude larger than those of the typical unit cell and allowing3D fast Fourier transform (FFT) from selected volumes within the overall reconstruction. We study a polystyrene-polydimethylsiloxane (PS-PDMS) double-gyroid BCP. The double gyroid is composed of two independent, interpenetrating enantiomorphic tubular networks of one type of block (PDMS), separated by a slab-like domain ${ }^{11,12}$ (whose shape is loosely approximated by the $G$ minimal surface ${ }^{13}$ ) that is constituted by the second, majority block (PS). Although a double gyroid would nominally be classified as cubic (cDG; space group of $I a \overline{3} d$ ) in accordance with equilibrium theories of $\mathrm{BCPs}^{14,15}$, a more critical analysis of the morphology

[^0]
c

d

(100) section

Fig. 1|Supra-unit cell structure of the PS-PDMS double-gyroid tubular network phase. a, Real-space SVSEM reconstruction of PDMS domains (rendered red and blue), corresponding to about 2,000 unit cells of the (non-cubic) double gyroid. The white box at the bottom right highlights the size of a $2 \times 2 \times 2$ unit-cell subvolume. b, A magnified view showing a different triclinic $2 \times 2 \times 2$ volume cropped from within the large volume shown in panel a. The binarized (PS versus PDMS) raw SVSEM voxels are further divided into red and blue PDMS voxel networks, with the PS domains rendered transparent. The triclinic unit cell parameters ( $a, b, c, \alpha, \beta$ and $\gamma$ ) measured from real space are also shown. $\mathbf{c}$, Rendering of the 3D FFT analysis of a region containing approximately 160 unit cells, with Bragg-like spots highlighted as quasispheriodal volumes, showing how the data intersect with $a(100)$ plane. The central 000 peak is indicated as a red spot. See Supplementary Video 2 for an animated view of the 3D diffractogram. d, 2D logarithmic intensity plot of a (100) section from the 3D Fourier data. In addition to the allowed $\{220\}_{\mathrm{tDG}},\{400\}_{\mathrm{DDG}}$, $\{420\}_{\text {tDG }}$ and $\{440\}_{\text {tDG }}$ reflections (indexed in white), there are $\{110\}_{\mathrm{tD}},\{200\}_{\mathrm{tDG}},\{310\}_{\mathrm{TD}}$ and $\{530\}_{\mathrm{tDG}}$ reflections that would be forbidden in the cubic double gyroid (indexed in red).
reveals that the slow solution-cast material organizes into distinct triclinic variants of the cubic phase.

A3D rendering of a $2.70 \times 2.70 \times 0.64 \mu \mathrm{~m}^{3}$ volume containing about 2,000 unit cells from within a large double-gyroid monodomain is shown in Fig. 1a. The two constituent tubular PDMS networks (shown in red and blue)-which each enclose about $20 \%$ of the total volumeare defect-free, consistent with their disjointed segmentation. Prior standard small-angle $X$-ray scattering (SAXS) and TEM analysis indicated that this sample has a cDG structure ${ }^{16}$; however, careful analysis of the morphology throughout the 'ultra-large' volumes accessible to 3DSVSEM reconstruction reveals a decidedly non-cubic unit-cell symmetry. Figure 1 b shows an experimental reconstructed $2 \times 2 \times 2$ unit-cell volume, including triclinic unit-cell parameters. The symmetry of the particular cells in this grain deviates greatly from cubic, with the largest and smallest lattice parameters differing by $12 \%$ (for example, 130 nm versus 116 nm ) and with pairs of translation vectors deviating by up to $14^{\circ}$ from orthogonality. The synchronized slices normal to the [001] direction from the experimental reconstruction and from the corresponding deformed self-consistent field (SCF) double-gyroid model are compared in Supplementary Video 1. As shown in Extended Data Fig. 1, unit-cell parameters exhibitonly small deviations throughout agiven many-cubic micrometre-scale grain. Other grains exhibit distinct triclinic variants. These triclinic variants deviate from cubic symmetry by up to about $20 \%$ in both length and angle. We denote this morphology as 'variable triclinic double gyroid' (vtDG) in order to indicate unit cells that are essentially coherent withingrains, but vary substantially from grain to grain. As shown in Extended Data Fig. 2, directions and magnitudes of deviations from cubic symmetry in distinct regions of the sample are uncorrelated with slicing directions, ruling out the possibility that the measured anisotropy is an artefact of SVSEMimaging or reconstruction.

Structural symmetries can also be assessed using 3DFFT of the SVSEM data. The detailed distribution of intensity from a particular ( $h k l$ ) Bragg plane depends on the orientation and spacing distributions of the (hkl) planes within the volume of the sample transformed. We use selected


Fig. 2 |Sub-unit cell IMDS curvature and distance metrics. a, Bragg-filtered reconstruction of the IMDS for one unit cell (for a description of the IMDS isosurface, see Methods). b, Graph plotting the normalized Gaussian $(K)$ and mean $(H)$ curvatures of the IMDS and their respective probability functions $(P)$, based on a region of 35 unit cells. The curvatures are normalized by $\langle D\rangle=130 \mathrm{~nm}$ (the average lattice parameter cell dimension measured by SAXS). The diagonal and vertical dashed blue lines indicate the curvature distributions for a constant minimal $G$ surface thickness (CMT) IMDS and a constant mean curvature (CMC) IMDS. c, The same unit cell as in panel a but with the IMDS made semitransparent, revealing the two skeletal graphs that are found by a thinning algorithm (see Methods).d, Distance distributions for the minority and majority domains. The minority-block thickness is measured from the IMDS to the closest distance to the skeletal graph. The majority-block thickness is measured as half the distance from a point on the red IMDS to the closest point on the blue IMDS.


Fig. 3|Sub-unit cell length and angular metrics. a, Region of roughly 100 unit cells, where a topological thinning of the segmented SVSEM tomogram has been used to create the two skeletal graphs of the double-gyroid structure. The viewing direction is approximately [100]. b, A small region from the theoretical cubic unit cell, depicting the skeletal graph and the surrounding IMDS, highlighting an internode strut with one cubic unit cell (top) and its dihedral rotation when viewed along the strut (bottom). The solid dark red central IMDS piece contains two nodes of the graph, and by viewing along the strut connecting the nodes, one can measure the dihedral angle, $\theta_{\mathrm{d}}$. c, Polar plot of the dihedral angle for the red $\left(<\theta_{\mathrm{d}}>=+70.9^{\circ}\right)$ and blue $\left(<\theta_{\mathrm{d}}>=-70.8^{\circ}\right)$ experimental networks, showing the narrow distribution of angles in each
network.d, A portion of the skeletal graph from panela, with the struts coloured according to their length, showing a factor of roughly three in variability. e, Spherical plot of the strut lengths versus their orientation in the laboratory frame.f, The same data projected onto a Mercator plot, where $\Theta$ and $\Phi$ are the polar and azimuthal angles, respectively, of the strut orientation as shown in e. The $<110>$ strut directions of a cDG lattice are shown as red circles. $g$, Inverse correlation between strut length and mean 'tube' radius of the IMDS measured at the midpoint along the strut (with red and blue colours indicating two distinct networks), showing the transverse contraction (dilation) upon length stretching (compression) of tubular struts.
filter (see Methods)-an approach that is well established in 2D highresolution TEM ${ }^{21}$. An isosurface constructed from the Bragg-filtered SVSEM data shows the two IMDS regions, each containing PDMS and enclosing roughly $20 \%$ of the unit volume (Fig. 2a), allowing measurement of the mean $(H)$ and Gaussian $(K)$ curvature distributions (Fig. 2b).

Heuristically, we can compare this experimental $H$ versus $K$ distribution with two limiting theoretical geometries: a constant matrix thickness (CMT) surface, which is surface displaced (normally) by a constant from the $G$ minimal surface ${ }^{13}$; or a constant mean curvature (CMC) surface ${ }^{22}$. Although a CMC shape has been suggested ${ }^{13,22}$ on the grounds that it minimizes the IMDS area for a fixed volume fraction ${ }^{11,23}$, the CMT surface minimizes the entropic penalty of variable stretching of the majority component at the expense of a slight increase in the interfacial area ${ }^{24}$. The curvature distributions for mathematical cDG surfaces of both types are shown in Extended Data Fig. 3a, b. The curvature distribution of a CMC surface is localized to a vertical band $H D=2.23$, while that of a CMT surface follows Steiner's linear relationship, $H D=-(t / 2 D) K D^{2}$ or $H D=-0.103 K D^{2}$ (where $D$ is the cell repeat length and $t$ is the constant thickness of the slab-like matrix domain) $)^{1}$. Relative to these reference surfaces, the experimental curvature distribution is closer to the CMC distribution. Also shown in Extended Data Fig. 3c-f are the $H$ and $K$ distributions for SCF theoretical calculations of cDG as a function of segregation strength. Although the IMDS shape of these model equilibrium states is always intermediate to the CMC and CMT shapes in terms of the curvature distribution, we note that the shapes are relatively CMC-like in weakly segregated gyroids and trend towards CMT-like when interblock repulsions are increased. This observation, in combination with the CMC-like distribution measured in Fig. 2b, might suggest that the experimental IMDS shape is inherited from a state in which the PS domain vitrifies and fixes the shape of the ordering structure during solvent evaporation. It remains far from clear, however, if and how closely the shape of the partially solvated and non-equilibrium




Fig. $4 \mid$ Models of sub-unit cell morphology. a-c, Orthographic views of portions of skeletal graphs for cubic (a), affine triclinic (b) and non-affine triclinic (c) models derived from equilibrium SCF calculations with cubic or triclinic symmetry. Here the colour indicates the strut length between nodes, normalized by the mean cell repeat length $D$. d, Probability distributions for the strut length/mean cell repeat length computed for affine and non-affine triclinic
models.e, Probability distributions for internode angles for the same computed networks, with colours as in panel d.f, Intensity of reflections (circles) made by 3DFFT of: the SVSEM reconstruction (left; roughly 160 unit cells); the affine triclinic SCF model (centre; 64 unit cells); and the non-affine triclinic SCF model (right; 64 unit cells). For reference, the allowed peaks for cDG are plotted with open squares.
double-gyroid morphology can be modelled by an equilibrium theory for neat diblocks. We note that the IMDS shape of a BCP double gyroid has been analysed previously ${ }^{4-6}$, yet resolution limitations for the IMDS curvature of the TEM tomography measurement made identification of the surface shape signature inaccessible.

The inhomogeneous geometry of the double gyroid implies a heterogeneous distribution of domain thicknesses ${ }^{12,25}$, corresponding to variable degrees of chain extension throughout the structure. To characterize the variable thickness of the minority domains, we derive askeletal graph from the SVSEM reconstruction; this graph consists of 1D struts threading through geometric 'centres' of tubular domains and meeting at threefold junctions (Fig. 2c) ${ }^{26}$. We define a 'minor block thickness' as the shortest distance from a point on the IMDS to the interior skeleton, while a 'major block thickness' is half of the shortest distance from a point on one IMDS to the IMDS of the opposing network (Fig. 2d, inset). Distributions of the minor (PDMS) and major (PS) block thicknesses are shown in Fig. 2d. Previous explanations for double-gyroid formation in BCP have emphasized that the constraints involved in packing polymer blocks at constant density require variation in the stretch length, most
prominently derived from the greater distance from the IMDS to the threefoldjunction node than from the corresponding IMDS-to-sketelon distance at the mid-portion of an internode strut ${ }^{24,27}$. This notion of 'packing frustration' is consistent with the broad spread of minor-block length. However, the measured distribution of thicknesses for the major PS block also exhibits substantial spread, indicating that-contrary to the present heuristic picture ${ }^{24}$-packing of the majority block is also frustrated. Non-uniform matrix thickness thus arises as a consequence of the 'tug of war' between majority-block stretching and the countervailing forces that favour more uniform minor-block lengths, as well as from the drive towards area-minimizing IMDS shapes ${ }^{23,28}$, consistent with the more CMC-like distribution observed in Fig. 2b.

Although the IMDS shape and domain thickness are necessarily inhomogeneous even in the ideal cDG, the internode struts in the ideal cDG are uniform in length and all orient along $<110>$ directions. The experimental network morphology can be further analysed using the skeletal graphs (Fig. 3a). We first consider the dihedral angle (Fig.3b). In an ideal cDG network this angle is $\pm 70.5^{\circ}$ (modulo $180^{\circ}$ ), where the sign characterizes the chirality of the two enantiomorphic single-gyroid
networks ${ }^{26}$. Remarkably, the experimental dihedrals for the positive and negative networks deviate little (with a root-mean-squared variance of less than $11^{\circ}$ ) from the ideal cubic geometry values (Fig. 3c). We also show (Extended Data Fig. 4) for the interstrut angles in the experimental vtDG a deviation of only about $20 \%$ from the perfect threefold coordination $\left(120^{\circ}\right)$ of cDG. This degree of local angular order in the PDMS networks contrasts starkly with the pronounced variability in the length of tubular struts as measured by skeletal edges. Strut length can vary by up to 300\% (Fig. 3d). Figure 3e, f analyses the lengths of PDMS struts according to their orientation, and indicates a strong correlation between the orientation and the $12<110>$ directions of a cDG graph. Struts in a given orientation are relatively homogeneous in length, but show prominent length variations between distinct orientations-exceptionally large given the more modest (roughly 20\%) discrepancy between triclinic and cubic cell geometry. Notably, in Fig. 3 g we show concomitant contraction/ dilation of the transverse tubular radius with stretching/compression of the strut length.

To understand the origin of the anomalously large variability among PDMS strut lengths versus the relative constancy of strut angles, we consider the microdomain structures and their associated skeletal graphs derived from SCF models of the tDG. The first model, affine triclinic, is generated by affinely deforming the SCF cDG structure (Fig. 4a) into a particulartDG shape equal to that of theexperimentally determined unit cell (Fig. 4b). A second model, non-affine triclinic, instead uses the same experimental triclinic cell boundary conditions to compute an equilibrium SCF double-gyroid morphology (Fig. 4c). As shown in Fig. 4d, e, the affine triclinic deformation of the double gyroid leads to spread of the strut lengths and angles by roughly 10-20\%, comparable in scale to the imposed strains deforming the cubic cell to triclinic symmetry. Remarkably, if instead we consider the predicted double-gyroid morphology that equilibrates within the same triclinic cell, the network structure adopts an increased degree of length dispersity (Fig. 4d)-well beyond the nominal lengths derived from cubic to triclinic distortion-and yet a reduced degree of angle dispersity (Fig. 4e). Moreover, comparison of the spectrum of Fourier peaks from the experimental versus the SCF affine triclinic and non-affine triclinic models reveals extraordinary correspondence between the experimental structure and the non-affine triclinic model (Fig. 4f).

Taken together, these observations suggest a strong thermodynamic coupling of sub-unit-cell morphology to symmetry breaking at the supra-unit-cell scale of the double-gyroid phase. Strut lengths (and diameters) are relatively soft and thus easily accommodate deformation, presumably through relatively rapid intradomain transport of polymer chains. By contrast, the angular geometry of the gyroid network is stiff, favouring local correlations that are maintained even under strong symmetry-breaking deformations. This suggests a heuristic model for the non-affine structure of symmetry-broken soft-matter networks, consisting of periodic networks of tensed struts (so-called Steiner networks, which are1D analogues of Plateau borders) ${ }^{29,30}$ that adjust lengths locally yet maintain force-balancing angular coordination at the nodes, in order to minimize the stretching that occurs in response to imposed changes in unit-cell symmetry.

Our observations have been made possible by the accessibility of ultralarge volumes to SVSEM tomography, in combination with the Bragg averaging of selected volumes, in order to achieve enhanced resolution of sub-unit-cell features. New distance and angle metrics applied to the complex double-gyroid phase allow deeper insight into the complex energetic competition, as reflected in the distinctive structural distortions in actual samples that invariably result from solvent evaporation and grain boundary incompatibility. The symmetrybreaking distortions are predicted to have impacts on, for example, the photonic/phononic band properties of double-gyroid assemblies ${ }^{2,31}$. Our research opens up a new way of unambiguously characterizing a variety of soft-matter systems that assemble under different processing
conditions into a variety of soft crystals (beyond networks), illuminating their formation mechanisms, supra-cell and sub-cell structures and structure-property relationships.

## Online content

Any methods, additional references, Nature Research reporting summaries, source data, extended data, supplementary information, acknowledgements, peer review information; details of author contributions and competing interests; and statements of data and code availability are available at https://doi.org/10.1038/s41586-019-1706-1.

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## Methods

## Material and sample preparation

We synthesized the polystyrene-poly(dimethylsiloxane) (PS-PDMS) diblock copolymer ${ }^{16}$ by sequential anionic polymerization of styrene and hexamethylcyclotrisiloxane. The polymer has number average molecular weights of $43.5 \mathrm{~kg} \mathrm{~mol}^{-1}$ (for PS) and $29.0 \mathrm{~kg} \mathrm{~mol}^{-1}$ (for PDMS), with an overall composition of $40 \%$ PDMS (by volume) and a polydispersity index of 1.04. The sample studied was cast slowly (over the course of one week) from a $10 \mathrm{wt} \%$ solution ( 2 ml ) in toluene. After drying, the sample was heated to $60^{\circ} \mathrm{C}$ for 3 days in a vacuum in order to remove any residual solvent.

On the basis of characteristic 2D TEM and radially averaged SAXS, the PS-PDMS diblock copolymer has been reported to have a double-gyroid morphology ${ }^{16}$. We further characterized the small piece of sample that we used for SVSEM with the synchrotron X-ray at Sector 12-ID-B of the Advanced Photo Source in the Argonne National Laboratory. Given the SAXS pattern (Extended Data Fig.5), the structure can indeed be nominally associated with a double-gyroid morphology, with an average cubic repeat of $D=130 \mathrm{~nm}$. However, we observe a prominent low $q$ peak associated with the $\{110\}$ planes that is forbidden for the cubic $I a \overline{3} d$ space group. Given the size of the incident X-ray beam and the sample thickness, this SAXS pattern must come from about $10^{9}$ unit cells. Before SVSEM imaging, the sample was attached to a $45^{\circ}$ SEM stub with doublesided conductive carbon tape, and the outer surface was then coated with a $50-\mathrm{nm}$ layer of platinum.

## Slice-and-view SEM data acquisition

Extended Data Fig. 6 shows the workflow involved in SVSEM. This is a tomographic method enabled by advances in ion milling, monochromated field emission electron beams and electron imaging detectors, as well as by precise stage motion and sophisticated software routines for correlation and registration of the image stack ${ }^{32-34}$. During data collection (Extended Data Fig. 6a), ion milling is combined with electron imaging (Extended Data Fig. 7a): an incident high-energy ion beam is used to make an impact on and mill away a thin slice of the near-surface region of the sample (Extended Data SFig. 7b, i); then an electron beam is directed at the surface and a secondary-electron image is recorded (Extended Data Fig. 7b, ii). This ion-slice, electron-image sequence is repeated until a sufficient thickness of the sample has been serially imaged. The 3D tomogram is then reconstructed by vertical stacking of the aligned 2D SEM images. Three large-volume tomographic data sets from distinct double-gyroid grains are available at an online data repository (https://doi.org/10.7275/wv24-3j62).

We used a Thermo Fisher Helios NanoLab 660 SEM/FIB DualBeam system for data acquisition. A focused gallium-ion ( $\mathrm{Ga}^{+}$) beam with an energy of 30 KeV and a relatively low beam current of 80 pA was used to mill the sample surface in order to minimize damage from the FIB beam ${ }^{35}$. A1-KeV electron beam with beam current of 50 pA was used to image the sample surface with a through lens (TLD) secondary-electron detector (secondary-electron images taken with different incident energies are shown in Extended Data Fig. 8). Notably, the stronger scattering from the higher atomic number of silicon atoms in the PDMS and the resulting additional secondary-electronemission is sufficient to provide excellent intrinsic contrast between the PS and PDMS domains without staining. We used X-shaped fiducials to register the FIB and secondary-electron images during the automatic slice-and-view process, and drilled a deep hole (or holes) into the sample along the direction normal to the observing surface with FIB (at an acceleration voltage of 30 kV and beam current of 0.23 nA ) for fine registration of secondary-electronimages (for details, see Extended Data Fig. 9a-c). For FIB slicing, we set the slice thickness at 3 nm . Further monitoring during FIB image acquisition found the actual slice thicknesses to be $2.96 \pm 0.01$ nm per slice (for details, see Extended Data Fig. 9d). A potential relative rotation of SEM images during SVSEM
acquisition was also monitored and excluded (Extended Data Fig. 10 and Supplementary Video 3).

## SCFT and IMDS models

We performed self-consistent (mean) field theory calculations of diblock copolymer melts using a polymer self-consistent field (PSCF) code (http://pscf.cems.umn.edu/) as in ref. ${ }^{36}$ for cubic double gyroid (cDG) and non-affine triclinic double gyroid (natDG) morphologies, with a single-chemical-parameters family: that is, a volume fraction $f$, product of interblock repulsion $\chi$, and degree of polymerization $N$. The initial density-field profiles for a double gyroid were based on the example files distributed with the software, and later targeted for the $40 \%$ minority (PDMS) and $60 \%$ majority (PS) volume fractions.

For the natDG we generated an initial density field that is identical to the experimentally reported lattice dimensions. We performed SCFT calculations by iteratively changing the density fields while calculating the free energy at each step and progressing towards a free-energy minimum. Then, for CDG and natDG, we further adjusted the lattice dimensions while maintaining the unit-cell length ratios in order to find a metastable density field subject to imposed symmetry constraints. For natDG, we fixed $a / b=0.97 / 1, a / c=1.10 / 1$ and angles $\alpha, \beta, \gamma$ as $75.9^{\circ}, 84.7^{\circ}$, $89.1^{\circ}$, which derive from the unit-cell parameters computed from the optimal(inverse) reciprocallattice dimension from experimentalSVSEM measurements, and are therefore close to those of the local volume region shown in Fig. 1b. Here we did not attempt to achieve an equivalent segregation strength $(\gamma N)$ to match PS-PDMS systems, because it is unclear which equilibrium values correspond best to the conditions in which the sample (as it is undergoing solvent evaporation) becomes ordered but remains fluid in both domains. Instead, we aimed to capture the basic mechanism of the sub-unit cell non-affine structure without imposing a close match in interblock repulsion. Hence, for fixed symmetries (cDG or natDG) we consider the cell dimension that minimizes the free energy for fixed ratios of unit-cell dimensions and angles. The affinely sheared triclinic DG (atDG) was generated from cDG, by applying a transformation matrix that consisted of lattice vectors matching the natDG at the same composition and interaction parameters. Finally, for each of the cDG, atDG and natDG, we generated image stacks of the density fields for further geometric analysis.

For the IMDS models in Extended Data Fig. 3a, b, we generated the CMC surface of a tubular double-gyroid network by using Surface Evolver ${ }^{37}$ and following the procedure in ref. ${ }^{22}$. Thus we generated discrete CMC surfaces corresponding to the IMDS by minimizing the area and energy associated with a target mean curvature $\left(H_{0}\right)$ for the IMDS (that is, $\int\left(H(x)-H_{0}\right)^{2} \mathrm{~d} A$, where $H(x)$ is the mean curvature of the surface at point $x$, and $d A$ is the infinitesimal area at point $x$ ) by imposing volume constraints such that the minority component volume enclosed by the tubular networks is $40 \%$ of the cubic cell. Similarly, for the CMT IMDS, we took a discretized surface of the gyroid minimal surface and then pushed off all points by the same distance along the normal ( $\pm \hat{\mathbf{n}})$, such that the resulting volume enclosed by surfaces on either side of the minimal surface matched the $60 \%$ volume fraction of PS in the unit cell. See the supporting software at the online data repository (https://doi. org/10.7275/wv24-3j62) for details on generating image stacks from density fields.

## Morphological analysis

The following subsections describe the steps (Extended Data Fig. 6bg) that ultimately led to a detailed geometric analysis (Extended Data Fig. 6h). The custom computer codes used for these tasks are noted in each subsection, and the file 'README.txt' distributed with the supporting software outlines further instructions regarding the use of these codes with source data. All of the supporting software codes and 'README.txt' are at the online data repository (https://doi.org/10.7275/ wv24-3j62).
window was centred on the selected reciprocal-space lattice point. The Bragg mask is given by:

$$
B(i, j, k)=\sum_{\mathbf{k} \in \mathbf{G}_{h k l}^{\prime}} \exp \left[\frac{\left(i \delta k_{x}-k_{x}\right)^{2}}{2 \sigma_{x}^{2}}+\frac{\left(j \delta k_{y}-k_{y}\right)^{2}}{2 \sigma_{y}^{2}}+\frac{\left(k \delta k_{z}-k_{z}\right)^{2}}{2 \sigma_{z}^{2}}\right]
$$

Where $\sigma_{x}, \sigma_{y}$ and $\sigma_{z}$ are widths of Gaussian windows in the mask and $\mathbf{G}_{h k l}^{\prime}$ are the set of vectors in the (non-cubic) reciprocal lattice that locate each of the intensity peaks in the 3D FFT. We applied the filter pointwise to the FFT data and then applied an inverse FFT in order to obtain the filtered real-space volume data for further analysis. To carry out SVD analysis, we selected reciprocal-space lattice points on the basis of the corresponding overall intensity of each diffraction family of the FFT data and their associated cubic $q$ value (Extended Data Fig.11). Here we chose reciprocal-space lattice points such that their corresponding diffraction families have an overall intensity greater than $10^{-7}$ (normalized by the strongest $\{211\}_{\mathrm{tDG}}$ family) and their associated cubic $q$ values are smaller than $0.2 \mathrm{~nm}^{-1}$ to make the Bragg filtering mask (that is, the $\{110\}_{\mathrm{tDG}},\{200\}_{\mathrm{tDG}},\{210\}_{\mathrm{tDG}},\{211\}_{\mathrm{tDG}},\{220\}_{\mathrm{tDG}},\{310\}_{\mathrm{tDG}},\{321\}_{\mathrm{tDG}},\{400\}_{\mathrm{tDG}}$ families), while the standard deviation of the Gaussian window is two pixels (that is, $\sigma_{x}=2 \delta k_{x}, \sigma_{y}=2 \delta k_{y}, \sigma_{z}=2 \delta k_{z}$ ). A comparison of the 3DFFT pattern of the raw volume data and the 3D FFT pattern of the volume data after SVD treatment is shown in Extended Data Fig. 6d, e. See the supporting software 2.

Network skeletal graphs and analysis. See Extended Data Fig. 6g, h (with regard to dihedral angles, strut lengths and strut orientations). Using ImageJ (https://imagej.nih.gov/ij/), we binarized the greyscale image-stack data in order to identify the tubular networks formed by the minority domains, and separated them from the majority-block-filled matrix by using a threshold such that the volume fractions of the two binary components matched with the experimentally reported volume fractions. Note that although the same analysis could be applied to post-Bragg filtered data, the data in Fig. 3 consider a larger volume than is accessible to memory limitations of FFT filtering in Mathematica (see https://doi.org/10.7275/wv24-3j62). Hence, in order to analyse largevolume networks, we extracted the skeleton directly from the rawSVSEM data (comparative analyses of skeletons from pre- and post-filtered data in a smaller volume confirm that Bragg filtering has a negligible impact on network statistics).

We then reduced these networks into 1D skeletal graphs-that is, straight-line bonds that connect nodes which are threefold coordinated or higher; no fourfold or higher-fold nodes were identified in this way (indicating the absence of topological defects ${ }^{8}$ ). The initial task of reducing filtered 3D volume data into 1D lines was done using the inbuilt skeletonization feature in ImageJ. This procedure followed ref. ${ }^{40}$, and it reduces binarized volume data into a 1D curve (also referred as a medial axis) that is a collection of voxels. To identify the skeletal graphs, we subjected the 1D curve to further refinements. We did this using a custom Mathematica code, whereby we first converted the voxel collections into a graph by taking the voxel coordinates as vertices, and then connected each voxel to its adjacent neighbours in a $3 \times 3 \times 3$ voxel neighbourhood. For the next refinement, we fixed the vertices that lie on the boundary and iteratively removed vertices that have only one nearest neighbour which effectively removed branches of the 1D curve that did not connect to a node. Finally, we converted the remaining 1D curve into a straight line of bonds by iteratively removing vertices with two neighbours and then connecting them to one another. The end of this process usually results in having small clusters of vertices at the site of a node, which we rectified by replacing them with a single vertex, ultimately resulting in the skeletal graph with the same topology of the network that we started out with.

We then applied an optimization procedure to ensure that the skeleton lines lie along the regions of maximal density in the 3D volume data. This was achieved using an algorithm described in ref. ${ }^{26}$, which
defines an optimization functional in $\phi=\sum_{\langle i j\rangle} \frac{1}{L_{i j}} \int_{i}^{j} \mathrm{~d} s \phi(\mathbf{x})$ that averages the local intensity $\phi(\mathbf{x})$ over the skeleton bonds, where $\langle i j\rangle$ denotes the skeleton bond connecting $\mathbf{x}_{i}$ and $\mathbf{x}_{j}$, and $s$ represents the arc length along each skeletal bond. We used a cubic spline interpolation to create the density (or intensity $f$ ) function $\phi(\mathbf{x})$, based on reconstructed 3D volume data from images. We maximized this function with respect to node positions $\mathbf{x}_{i}$ in order to optimize the skeleton position over density $\phi$. We analysed the structure of the optimized skeletal graphs by calculating the skeleton dihedral angles and the length and orientation of the bonds that make up the graphs. For a given triplet of consecutive bonds, we defined the two planes and their normal as $\hat{\mathbf{n}}_{\alpha \beta}=\left(\hat{\mathbf{r}}_{\alpha} \times \hat{\mathbf{r}}_{\beta}\right) /\left|\left(\hat{\mathbf{r}}_{\alpha} \times \hat{\mathbf{r}}_{\beta}\right)\right|$ and $\hat{\mathbf{n}}_{\beta \gamma}=\left(\hat{\mathbf{r}}_{\beta} \times \hat{\mathbf{r}}_{\gamma}\right) /\left|\left(\hat{\mathbf{r}}_{\beta} \times \hat{\mathbf{r}}_{\gamma}\right)\right|$, where $\hat{\mathbf{r}}_{\alpha}, \hat{\mathbf{r}}_{\beta}$ and $\hat{\mathbf{r}}_{\gamma}$ are the unit vectors along the bonds. The dihedral angle is defined as the angle between these plane normals, with $\sin \theta=\left(\hat{\mathbf{n}}_{\alpha \beta} \times \hat{\mathbf{n}}_{\beta \gamma}\right) \cdot \hat{\mathbf{r}}_{\beta}$, $\cos \theta=\hat{\mathbf{n}}_{\alpha \beta} \cdot \hat{\mathbf{n}}_{\beta \gamma}$. We calculated this measure for all consecutive triplets of bonds in the skeletal graph. We also calculated the bond length as $l_{i j}=\left|\mathbf{r}_{i j}\right|$ where $\mathbf{r}_{i j}=\mathbf{x}_{i}-\mathbf{x}_{j}$, with $i$ and $j$ denoting the nodes (end points) of the struts, and the spherical angle coordinates $\Theta$ and $\Phi$ describing the length and orientation anisotropy. The node angle $\psi=\cos ^{-1}\left(\hat{\mathbf{r}}_{\alpha} \cdot \hat{\mathbf{r}}_{\beta}\right)$ is computed for each of the three pairs $(\alpha, \beta)$ of struts that meet at a single node; this is done for all nodes in the unit cell. Data from this analysis are presented as a polar histogram of dihedral angles in Fig. 3c, a histogram of node angles in Extended Data Fig. 4, Mercator plots of orientation in Fig. 3e, f, and anisotropy of strut length in Fig. 4d. See the supporting software 3,4 .

Calculation of curvature. See Extended Data Fig. 6g, h (with regard to IMDS curvature). We computed the mean curvature ( $H$ ) and Gaussian curvature ( $K$ ) of the IMDS. The IMDS is represented as a triangulated mesh, which we identified by finding a surface of the linear interpolation of density data at $\phi_{40}$ that separated the 3D volume into three types of domain with $20 \%, 20 \%$ and $60 \%$ volume.

We further used two-step conditioning by first applying an edgelength regularization to the mesh, and then constraining the mesh vertices to lie on the isosurface of a third-order interpolation of the density to ensure that mesh vertices represent a surface that is at least second-order differentiable. To regularize the triangle edge lengths, we minimized a regularization functional defined as $F_{\text {reg }}=\sum_{\langle i j\rangle}\left(L_{i j}-\bar{L}\right)^{2}$. We optimized this functional via a gradient-descent approach by taking the gradient with respect to the triangle vertex positions, and applied a constraint such that all resulting vertices lie on the surface by subtracting the component of gradient parallel to vertex normal. To constrain the mesh vertices, we created a third-order Hermite interpolation of density $\phi$ and we constrained each triangle vertex to lie along the $\phi=0.4$ isosurfaces within this interpolation. We accomplished this by minimizing $(\phi-0.4)^{2}$ for each vertex, again using the gradient-descent approach. We finally used the patch curvature function in the MatlabFile Exchange developed by D.-J.Kroon (https://www.mathworks.com/matlabcentral/ fileexchange/32573-patch-curvature) in order to compute curvatures on the optimized triangulated mesh that represents the IMDS. This algorithm calculated the principal curvatures $\kappa_{1}$ and $\kappa_{2}$ associated with each triangulated vertex by fitting a paraboloid to that vertex and its nearest neighbours, with the paraboloid axis constrained along the vertex normal. From the principal curvatures, the mean curvature $H=\frac{\left(\kappa_{1}+K_{2}\right)}{2}$ and Gaussian curvature $K=K_{1} \kappa_{2}$ can be calculated for each vertex. Curvature-distribution data are shown in Fig. 2b and Extended Data Fig. 3. See also the supporting software 5, 6,7.

Calculation of skeleton-IMDS and IMDS-IMDS distances and strut diameters. See Extended Data Fig. 6g, h (with regard to the
skeleton-IMDS and IMDS-IMDS distances). We used the optimized IMDS triangulated mesh and discretized skeletal graph to compute skeleton-IMDS and IMDS-IMDS distances and the effective diameters of the tubular networks. We carried out skeletal-graph discretization by choosing a discretization length $\mathrm{d}=\left\langle l_{b}\right\rangle / 100$, where $\left\langle l_{b}\right\rangle$ is the average bond length and for each bond we chose $l_{b} / d$ evenly spaced points along the bond. We calculated distances between the skeleton and IMDS by finding the nearest skeleton point for each vertex on the IMDS triangulated mesh, and IMDS-IMDS distances by considering separately the two IMDS surfaces resulting from individual networks and finding the nearest vertex in one IMDS from each vertex on the other. To calculate the effective strut diameter, we found the (quasi-ellipsoidal) 1 D intersection of the computed IMDS and a 2D plane that bisects a strut. An 'average tube radius' is computed by dividing the length of the 1D path (that is, the circumference) by $2 \pi$. The skeleton-IMDS and IMDS-IMDS distances are plotted in Fig. 2d, and the effective strut diameter versus strut length in Fig. 3g. See also the supporting software 8, 9.

## Data availability

SVSEM and SCF modelling data are available at https://doi.org/10.7275/ wv24-3j62.

## Code availability

Supporting software codes are available at https://doi.org/10.7275/ wv24-3j62.
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Author contributions The research was designed and supervised by E.L.T. and G.M.G. The BCP was synthesized by A.A., and R.-M.H. was responsible for processing samples. The SVSEM technique was developed by H.G., M. Z. and X.F., and carried out by X.F. and H.G. K.Y. provided modifications to the SVSEM software. Numerical algorithms for morphology analysis were developed by C.J.B. with I.P., and applied by C.J.B., X.F. and A.R. SCF computations were carried out and analysed by A.R. The manuscript was written by E.L.T, G.M.G., X.F., C.J.B. and A.R.

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| Unit cell <br> position | $a$ <br> $(\mathrm{~nm})$ | b <br> $(\mathrm{nm})$ | $c$ <br> $(\mathrm{~nm})$ | $\alpha$ | $\beta$ | $\gamma$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 129 | 132 | 116 | 76.4 | 84.3 | 90.0 |
| 2 | 128 | 131 | 115 | 76.7 | 85.5 | 90.6 |
| 3 | 130 | 131 | 115 | 75.7 | 86.3 | 91.1 |
| 4 | 129 | 130 | 116 | 73.6 | 85.2 | 93.0 |
| 5 | 129 | 132 | 116 | 74.4 | 86.3 | 92.3 |

Extended Data Fig. 1 | Variation in the unit-cell parameters of triclinic unit cells within one grain. Unit-cell parameters at five different places (1-5) within one grain are measured in real space. The result indicates that the structure is
coherent but non-cubic, with unit-cell parameters exhibiting only small deviations throughout the many-cubic micrometre grain.

## Article




Strain
$-0.206$
0.206

Eigenvectors:
$\begin{array}{rrr}(0.801 & 0.529 & -0.264) \\ (0.327 & -0.028 & 0.945) \\ (0.493 & -0.848 & -0.195)\end{array}$

Eigenvectors:

| $(-0.381$ | 0.782 | $-0.493)$ |
| :---: | :---: | :---: |
| $(-0.056$ | 0.512 | $0.857)$ |
| $(-0.923$ | -0.354 | $0.152)$ |



Eigenvectors:
$\left(\begin{array}{lll}0.488 & 0.356 & 0.797\end{array}\right)$
$\left(\begin{array}{lll}-0.667 & -0.437 & 0.604)\end{array}\right.$
(0.563 $-0.826 \quad 0.024$ )

Extended Data Fig. 2 |Strain eigenvector mapping from a cDG lattice to a vtDG lattice within the slicing coordinate frame of reference. Directions and magnitudes of deviations from cubic symmetry in different grains of the sample are not correlated with the ion-milling (slicing) direction (Z).



Extended Data Fig. $4 \mid$ Histogram showing the internode angles of experimental vtDG (from SVSEM) and non-affine triclinic SCF models.


Extended Data Fig. 5 |SAXS pattern from a region of the bulk polygranular PS-PDMS sample. The structure can be nominally associated with a doublegyroid morphology, with an average cubic lattice parameter of $D=130 \mathrm{~nm}$. Diffraction from the $\{110\}_{\mathrm{tDG}}$ and $\{200\}_{\mathrm{tDG}}$ families, which are forbidden for the cubic $I a \overline{3} d$ space group, are observed, indicating the non-affine deformation of the cubic double-gyroid lattice.

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Extended Data Fig. 6 |The workflow for collection and analysis of SVSEM tomography data. See Methods for more details.


Extended Data Fig. 7 |Acquisition and processing of SVSEM images. a, Illustration of the SVSEM reconstruction method. In step 1, low-energy incident electrons ( 1 KeV ) are used to image the near-surface region of a bulk sample. In step $2, \mathrm{a} \mathrm{Ga}^{+}$beam is used to slice a roughly 3 -nm-thick section from the sample surface. In step 3, electrons are again used to image the ion-beammilled sample surface. The process is repeated. With a large enough number of

(ii) Electron imaging
images (more than 200), the 3D morphology can be constructed via alignment of the stack of slices.b, Different sample stage positions are used for undistorted imaging during slice-and-view. $\mathbf{i}$, For ion milling during slicing, the sample observing surface is parallel to the ion beam.ii, For electronimaging, the sample observing surface is perpendicular to the electron beam.

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Extended Data Fig. $8 \mid$ Secondary-electron images acquired with different electron-accelerating voltages. Corresponding raw greyscale pixel-intensity distributions are presented blow the electron images. With a lower accelerating

voltage, there is a clear binary separation of the pixels into a dark peak (left) and a bright peak (right). Each image is from a freshly sliced region.
(a) FIB Image for registration

(b) SEM Image for registration


Extended Data Fig. 9 |Alignment fiducials and monitoring of slice thickness during experiments.a, An $X$-shaped fiducial (within the red square) in ion-beam view is used for the registration of FIB slicing. $\mathbf{b}$, An X-shaped fiducial (within the yellow square) in electron-beam view is used for registration of electron imaging. The round cross-section of the perpendicularly drilled hole (within the orange square) is used for the fine registration of secondary-electron images.

## (c) SEM Image for data acquisition


(d) Monitoring the slice thickness

c, A secondary-electron image used for data acquisition, showing an ion-milled hole (within the orange square) used for fine registration. d, Monitoring of slice thickness: we measured the distance between the milling surface of the $n$th slice and the milling surface of the first slice (total slice thickness $d$ ) from FIB images, and then plotted $d$ versus $(n-1)$. This reveals a linear relationship with a slope of $2.96 \pm 0.01$, which is the averaged slice thickness (in nm).

## Article

(a)

(b)


Hole 1
(c)


Hole 2

## Extended Data Fig. $10 \mid$ Monitoring of potential SEM image rotation during

 FIB-SEM image collection. a, SEM raw data image of the region of interest, with two holes drilled normal to the slice surface by the FIB.b, c, Side-view snapshots of 3D reconstructed holes 1 and 2 (the corresponding rotational videos are in Supporting Video 3). The image stack ( 80 slices) was aligned using hole 1. The reconstruction of hole 2 is still symmetric, indicating no image rotation.

Extended Data Fig. 11|Important Fourier components of the experimental double-gyroid structure. The overall intensity of each diffraction family was normalized by the strongest $\{211\}_{\mathrm{tDG}}$ family from the 3 DFFT of the region containing approximately 160 unit cells (the same region as in Fig. 4f, which shows the intensity of each individual peak). The overall normalized intensity data are plotted against their associated cubic $q$ value for those planes. For our reconstructions, we use a Bragg filter that selects peaks above an intensity threshold of $10^{-7}\left(I_{\text {min }}\right)$ for associated $q$ values smaller than $0.2 \mathrm{~nm}^{-1}$.


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