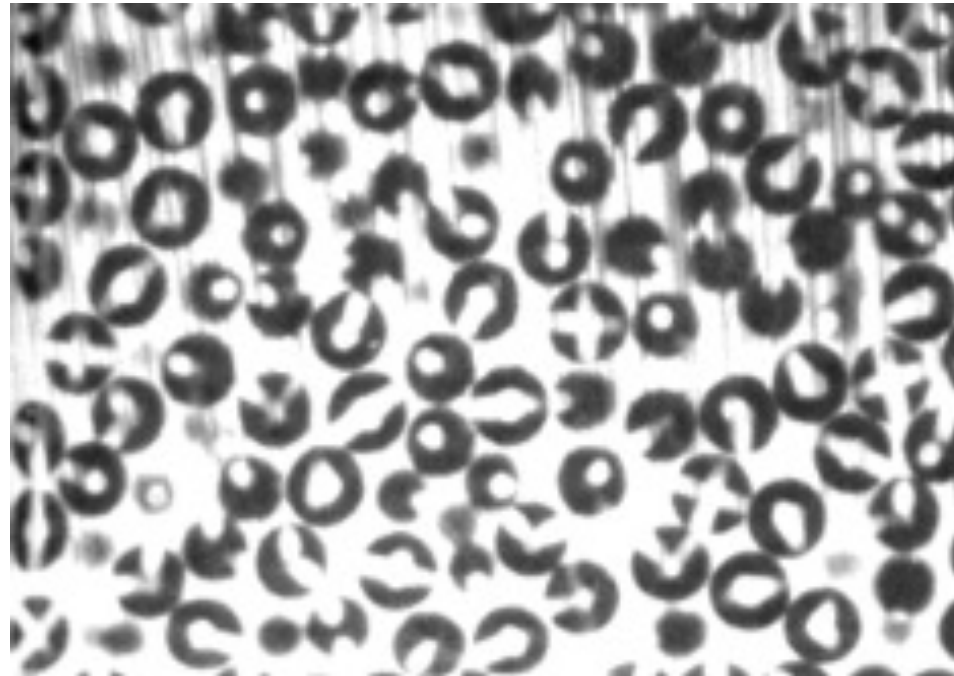
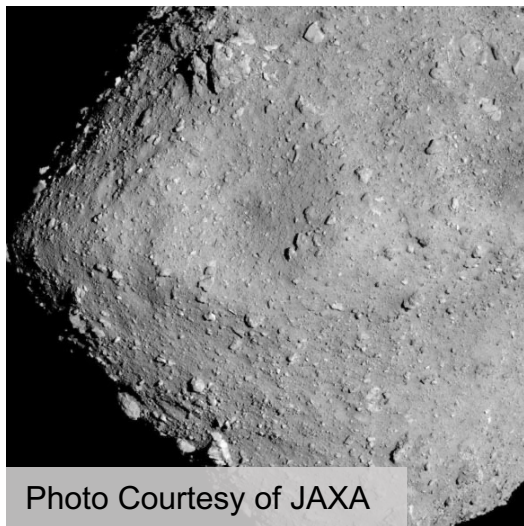

Experimentally measuring three-dimensional granular rotations.

Zackery A. Benson



Examples of granular materials

Asteroids and other ET objects can be modeled as granular material



- Collection of discrete particles that interact with a wide range of forces
- Inherently far from equilibrium
- Bulk properties depend heavily on material history



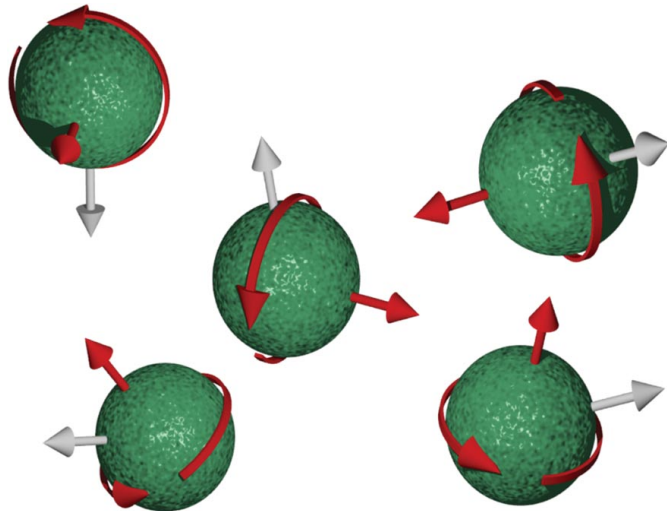
Ballast (bed of rocks) provides structural stability to railroad tracks

Rotations are needed for the complete study of particle dynamics

Rotational motion accounts for $\frac{1}{2}$ the total degrees of freedom

For dense systems, energy dissipation is dominated by frictional contacts instead of collisions

Collective rotations can emerge on multiple scales

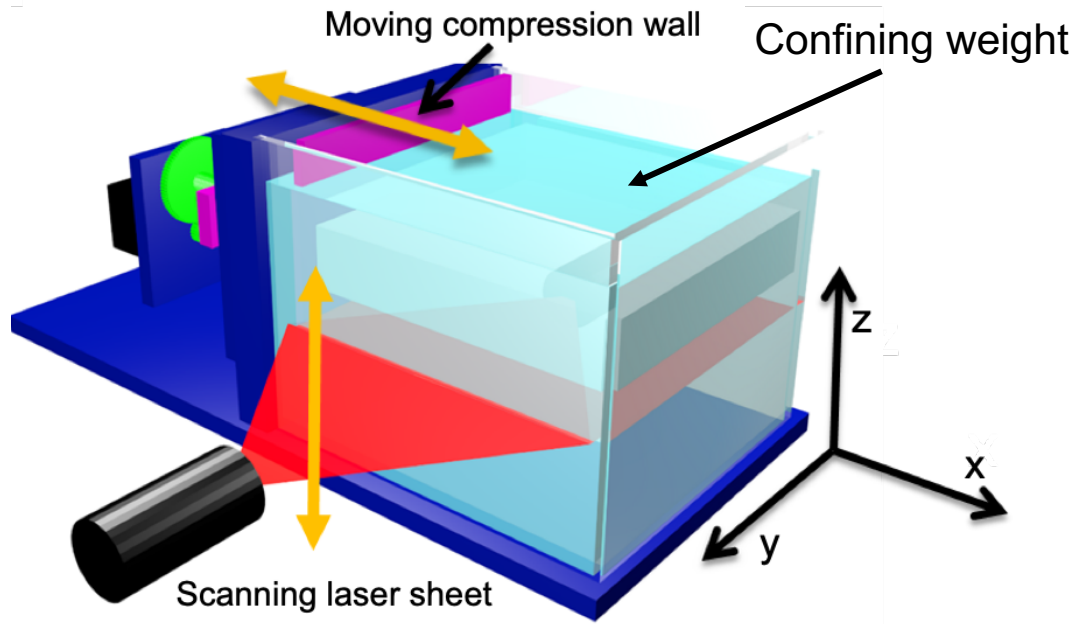


N. V. Brilliantov et. al. *PRL*, **98** (2007) 128001

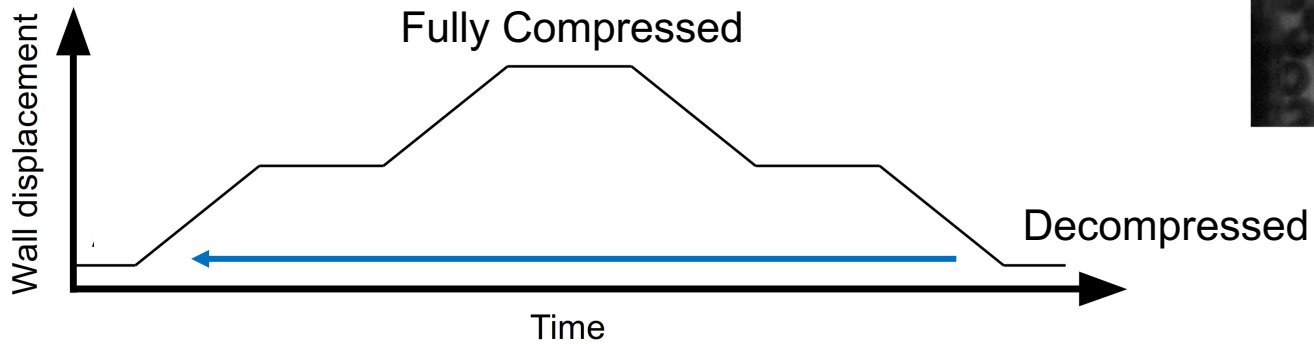
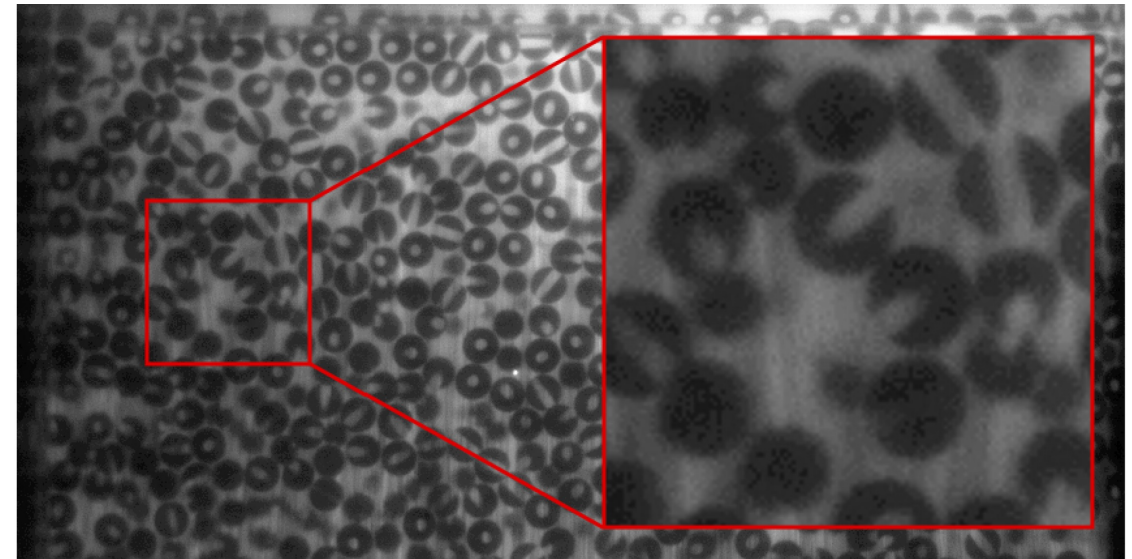


D. V. Stager et. al. *PRL*, **116** (2016) 254301

Our experimental system



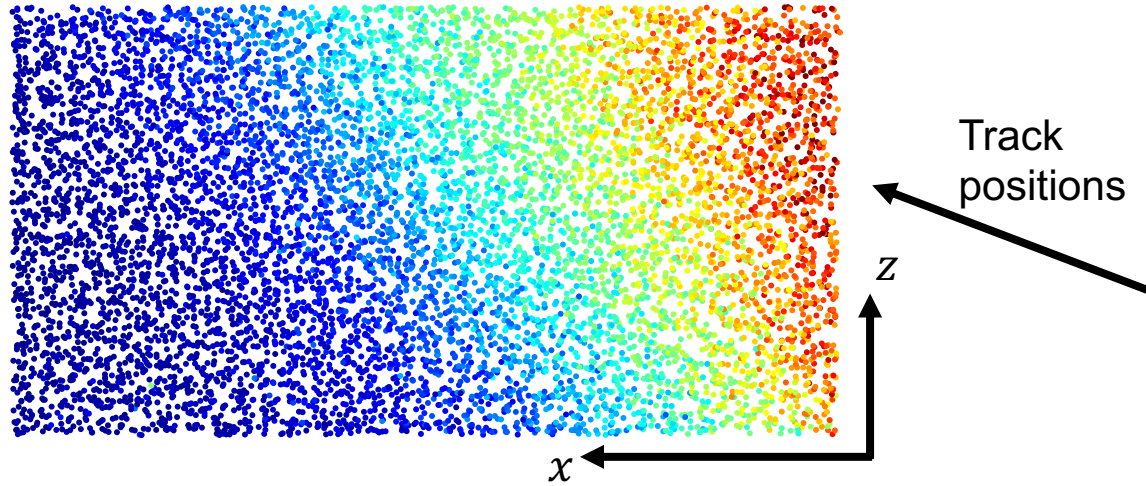
Material: 20,000 acrylic beads ($n=1.49$)
Radius = 0.25 cm
Fluid: Triton X100
Compression amplitude: 1% ($\sim 0.15\text{cm}$)
Packing fraction ≈ 0.6



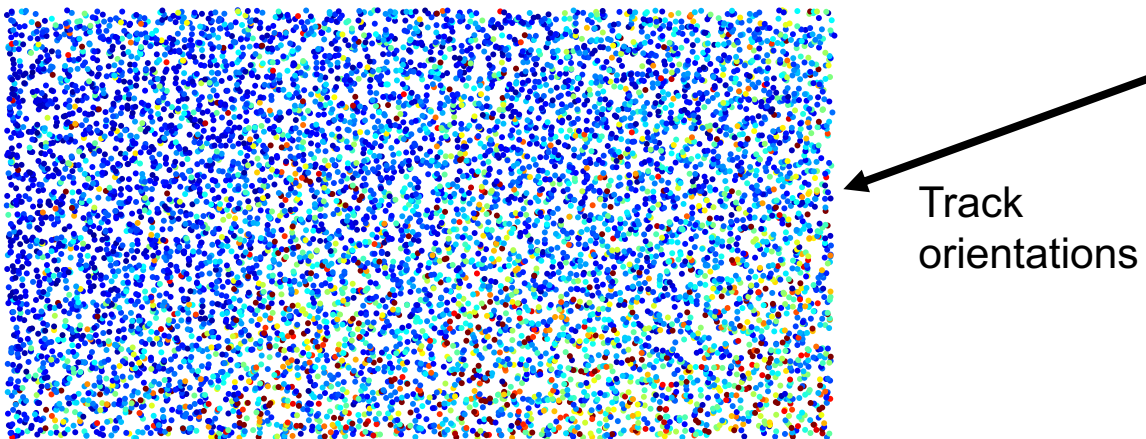
A. Peshkov et. al. *PRE* 100 (2019)

Capturing 3D rotational motion during cyclic compression

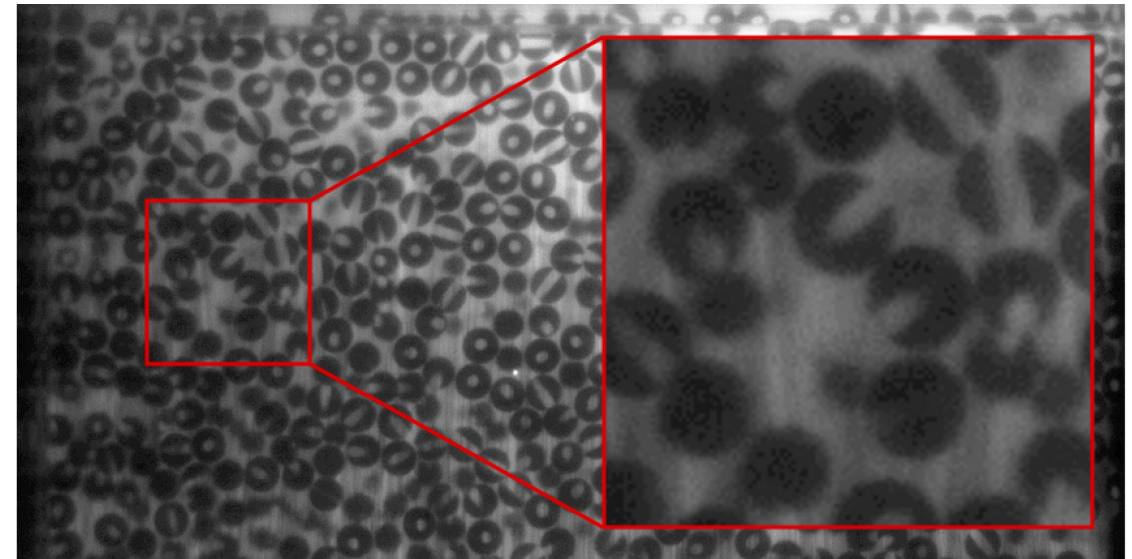
Mid Cycle Motion



Mid Cycle Rotations



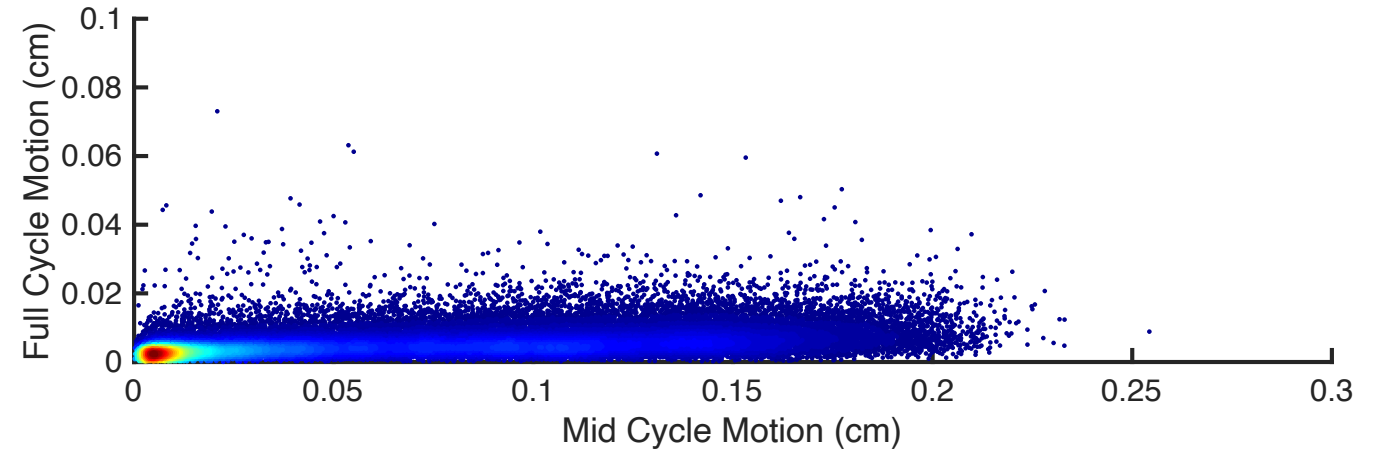
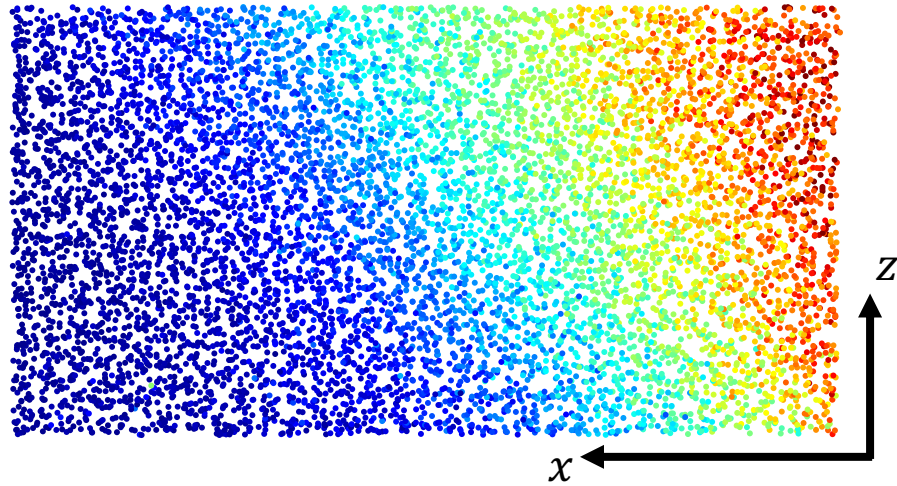
Material: 20,000 acrylic beads ($n=1.49$)
Radius = 0.25 cm
Fluid: Triton X100
Compression amplitude: 1% (~ 0.15 cm)
Packing fraction ≈ 0.6



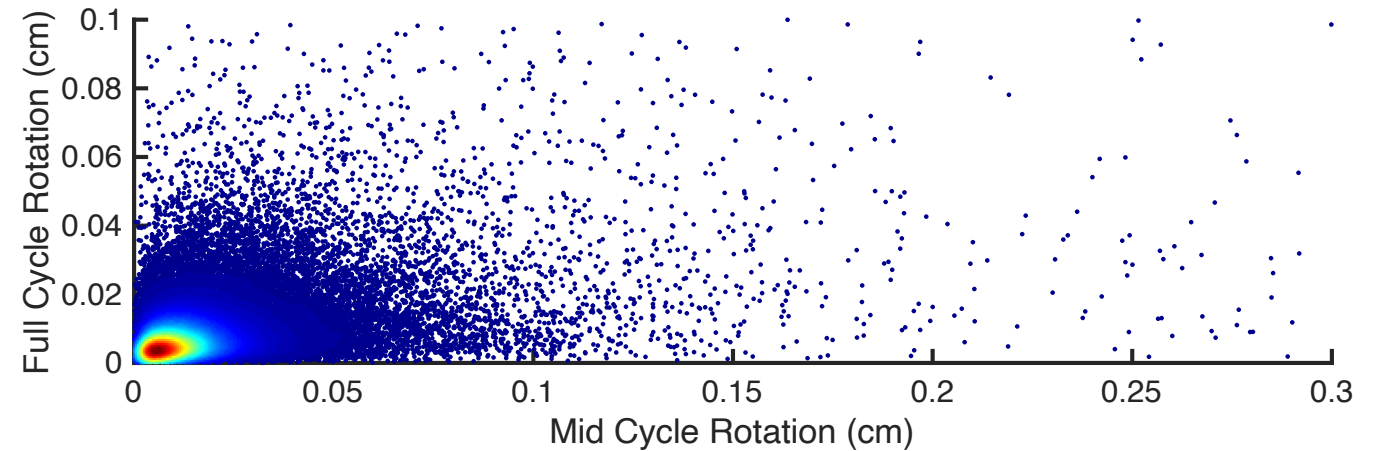
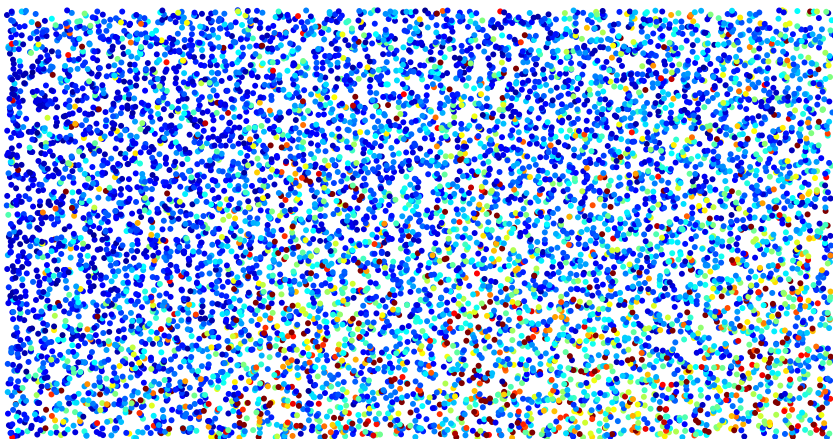
A. Peshkov et. al. *PRE* 100 (2019)

Tracking individual grains positions and orientations

Mid Cycle Motion

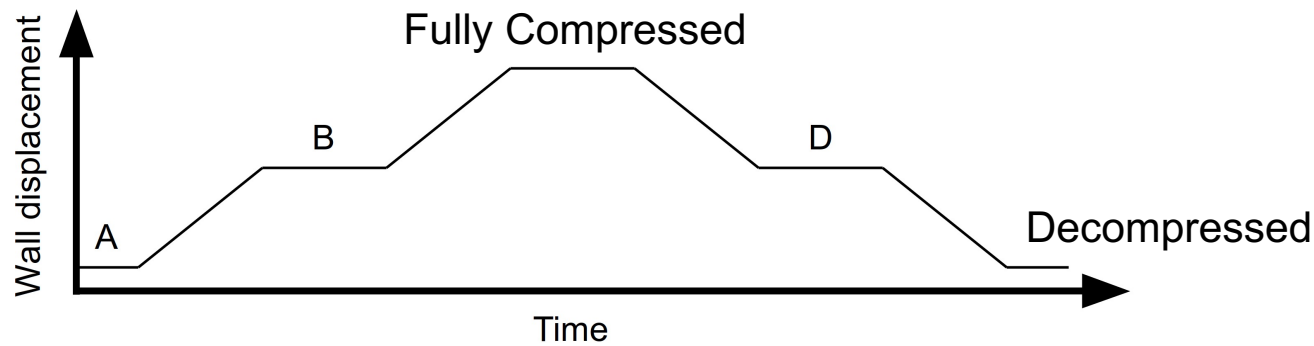
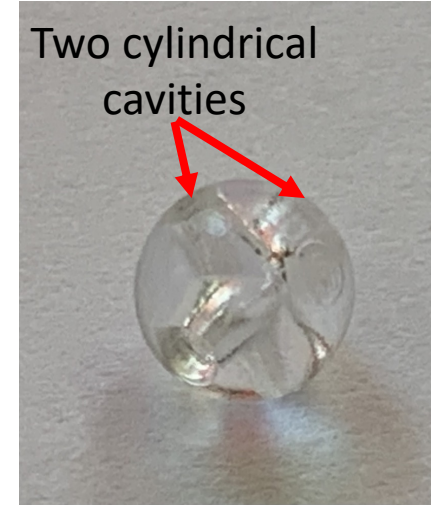
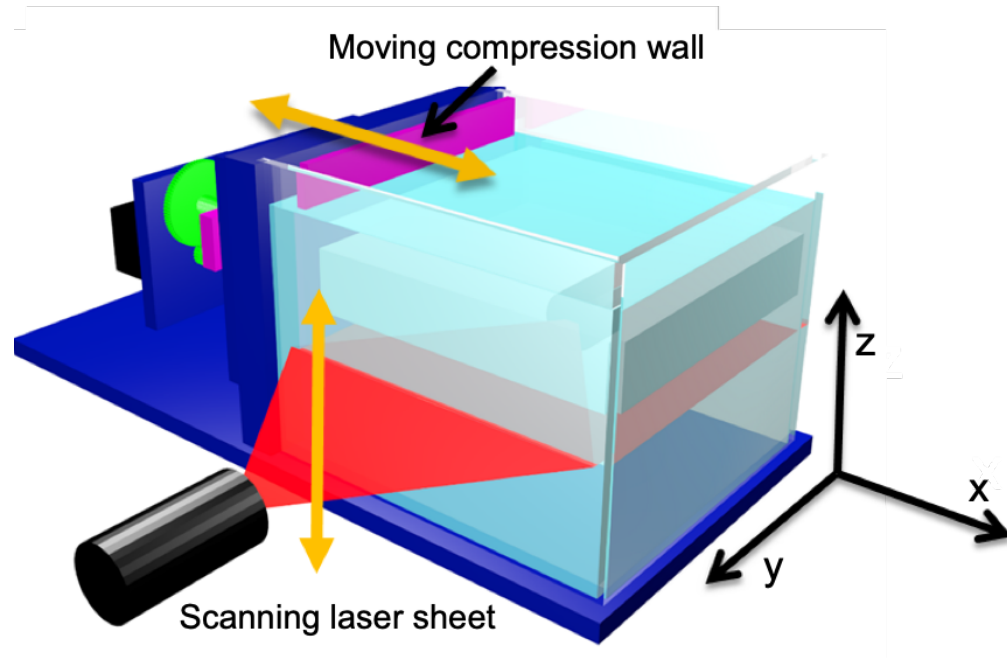


Mid Cycle Rotations

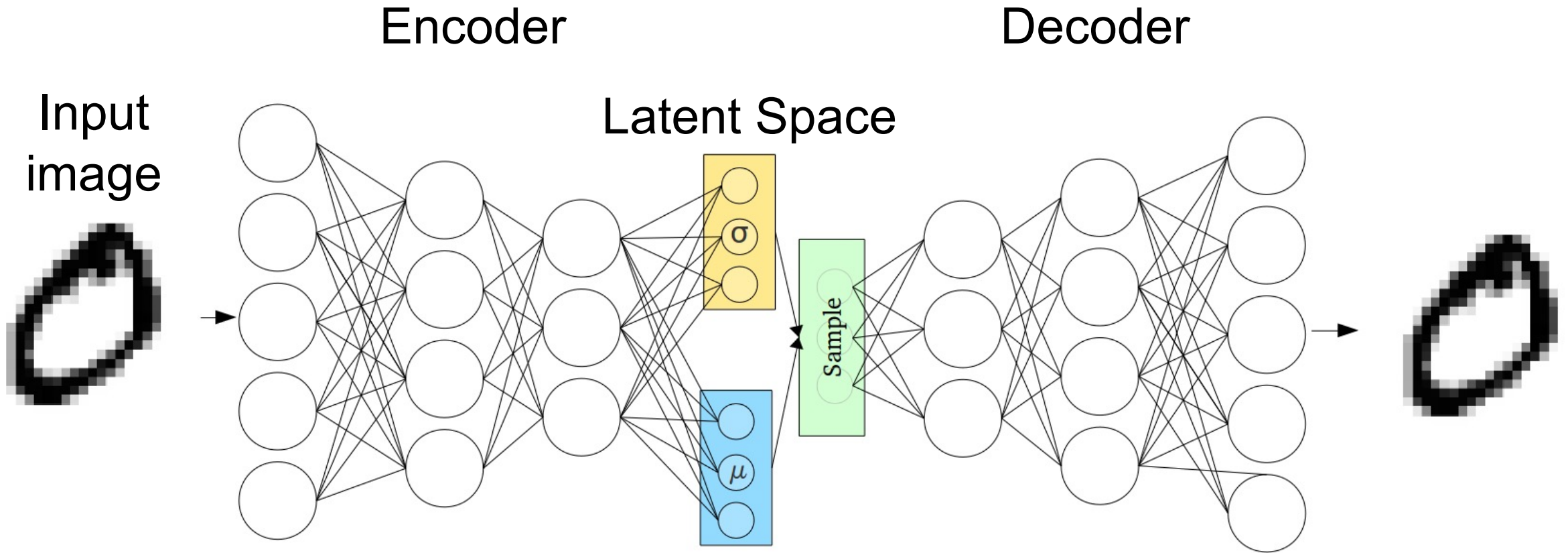


A. Peshkov et. al. *PRE* 100 (2019)

Two holes quantify all rotational degrees of freedom



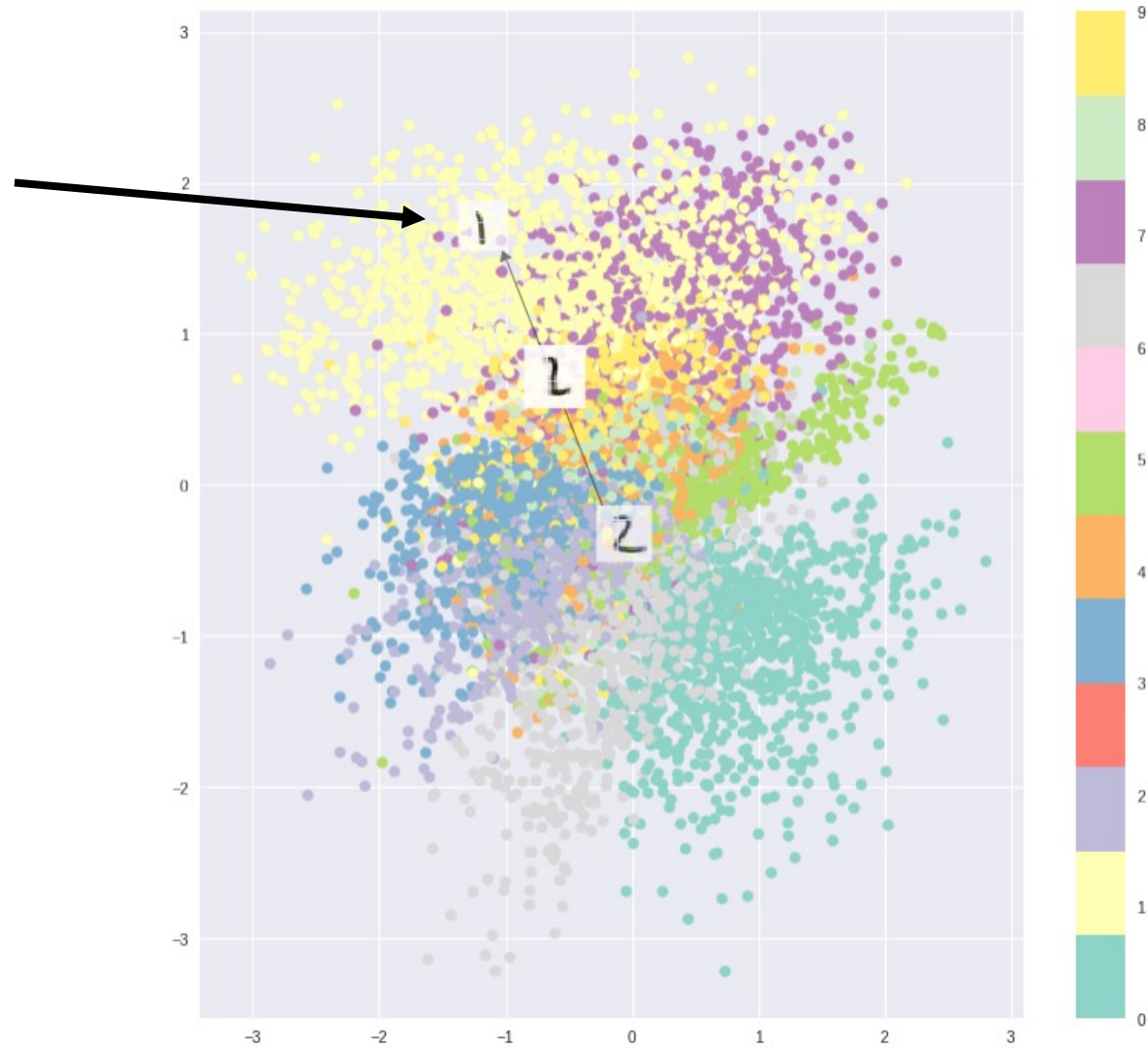
Application of variational auto encoders for image analysis



<https://towardsdatascience.com/intuitively-understanding-variational-autoencoders-1bfe67eb5daf>

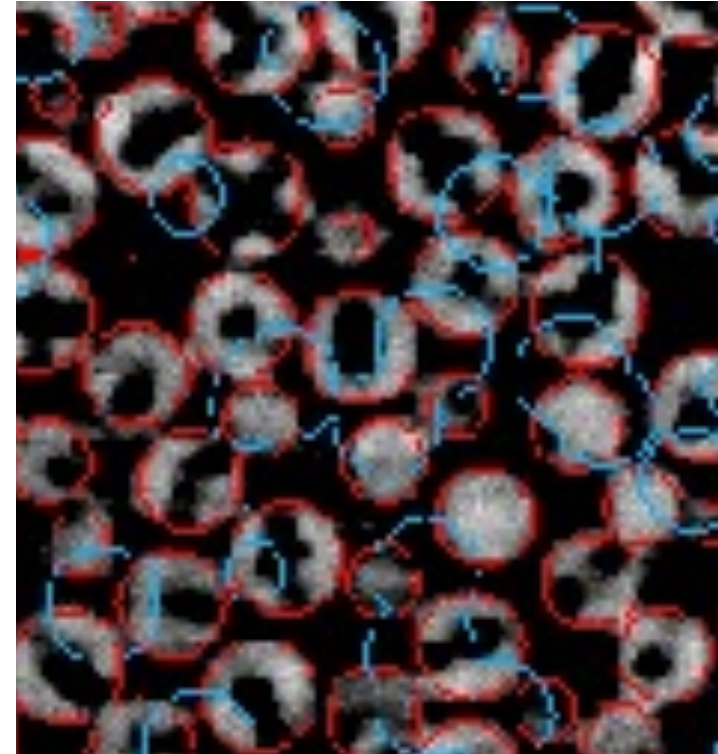
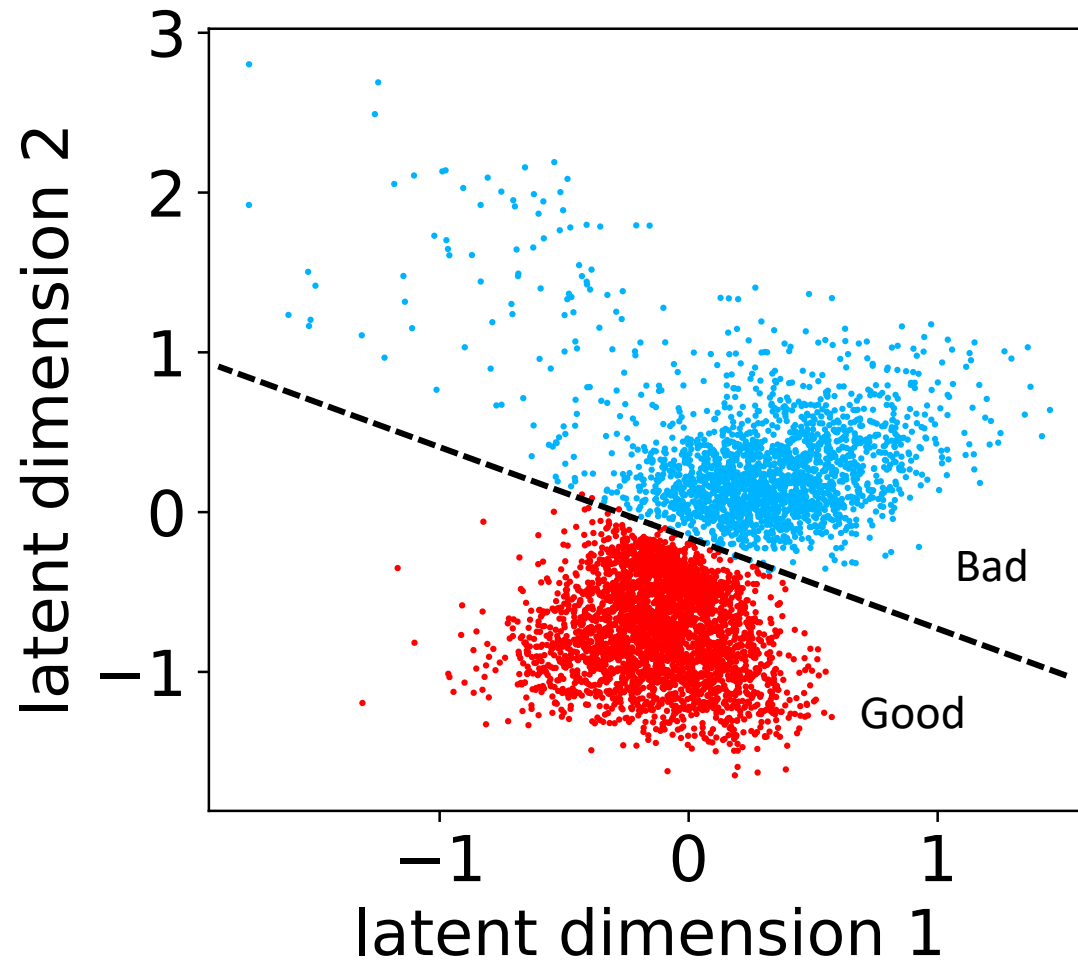
VAE encoding handwritten digits in a 2D latent space

This region corresponds to the digit 1

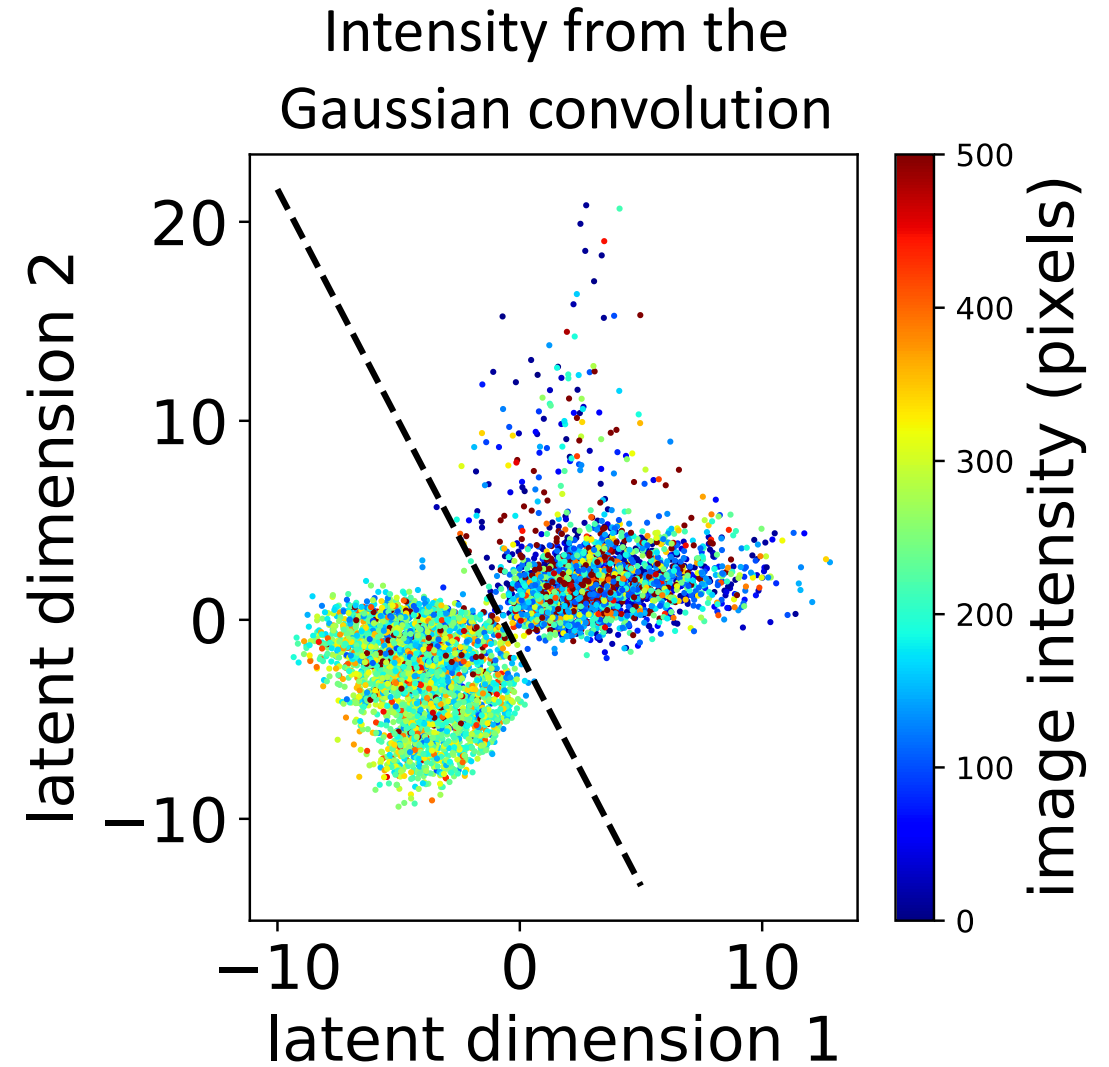
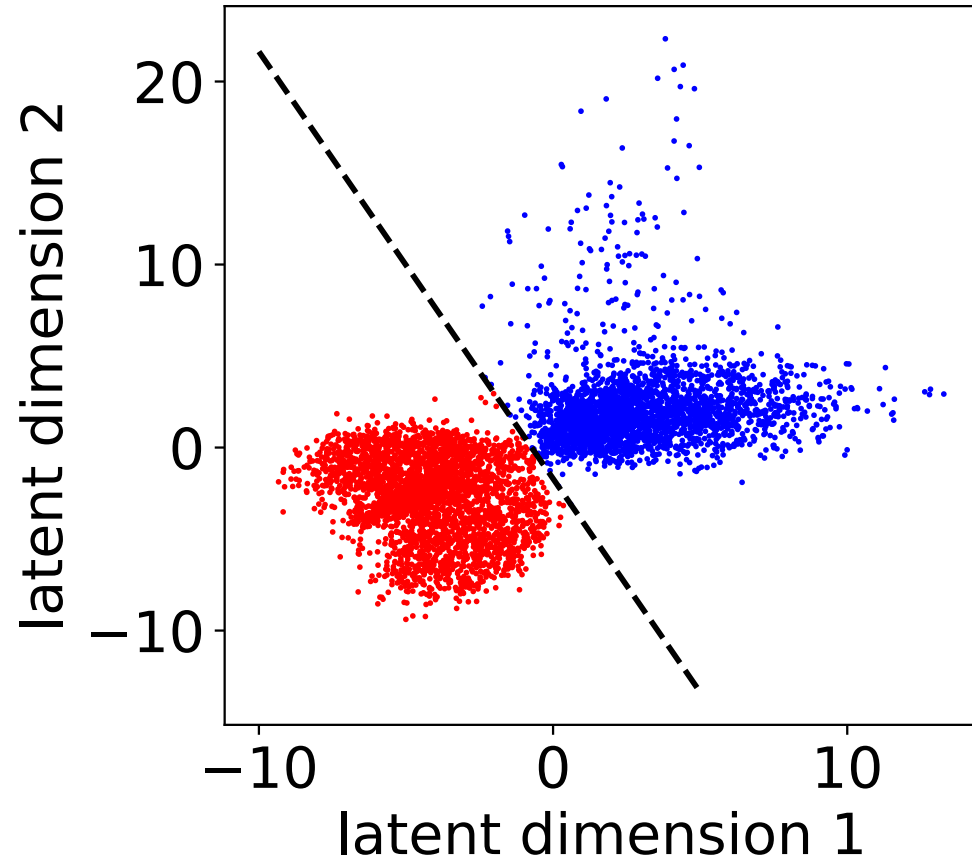


<https://towardsdatascience.com/intuitively-understanding-variational-autoencoders-1bfe67eb5daf>

Position identification with VAE

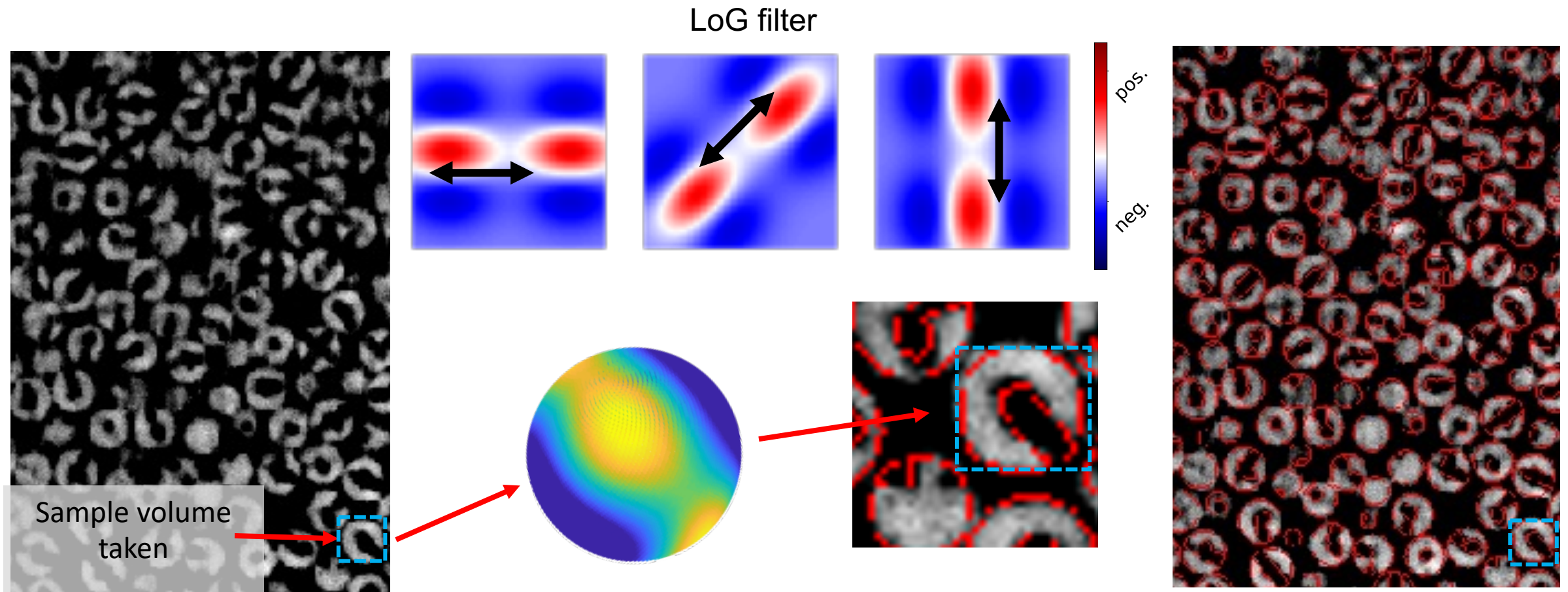


Application of VAE colored by pixel intensity



Latent space rotates during different training runs, but the feature separation remains the same

Orientation extraction



Computing rotations

The rotations for the experiment are calculated by the Kabsch algorithm.

$$C \rightarrow (\hat{p}_0 \quad \hat{q}_0) \cdot \begin{pmatrix} \hat{p}_1 \\ \hat{q}_1 \end{pmatrix}$$

Singular value decomposition

$$C = U\Sigma V^T$$

$$d = \text{sign}\{\det(VU^T)\}$$

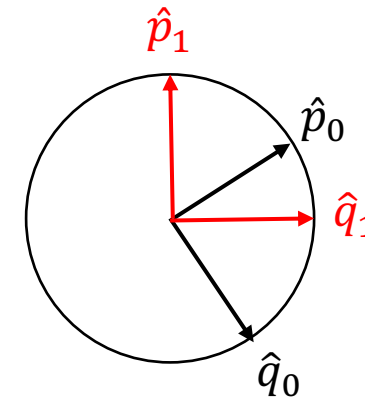
$$R = V \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & d \end{pmatrix} U^T$$

U and V^T → left and right singular vectors

Σ → contains the singular values

R → Rotation matrix

2D projection of rotating grain



Computing contact point rotations

Relative deformation of the contact point

$$\Delta \vec{U} = (\vec{v}_1 - \vec{v}_2) + (\vec{\omega}_1 \times \vec{d}_{12} - \vec{\omega}_2 \times \vec{d}_{21})$$

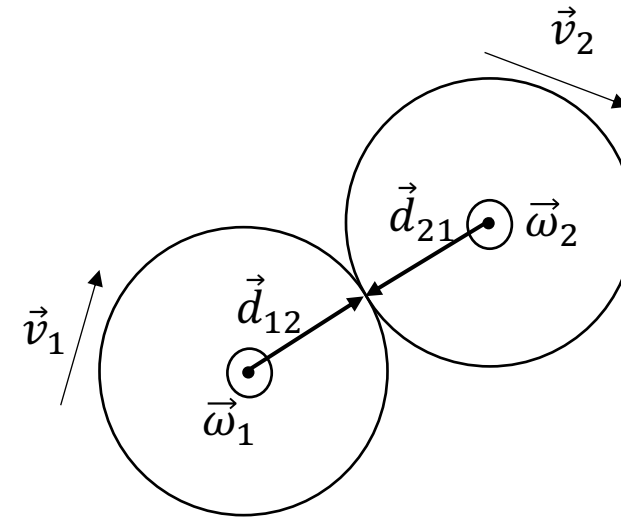
Take the tangential component

$$\Delta \vec{U}_{slid} = \Delta \vec{S} - (\Delta \vec{S} \cdot \hat{d}_{12}) \hat{d}_{12}$$

Rolling displacement

$$\Delta \vec{\omega} = \vec{\omega}_1 - \vec{\omega}_2$$

$$\Delta \vec{U}_{roll} = \Delta \vec{\omega} \times \vec{d}_{12}$$



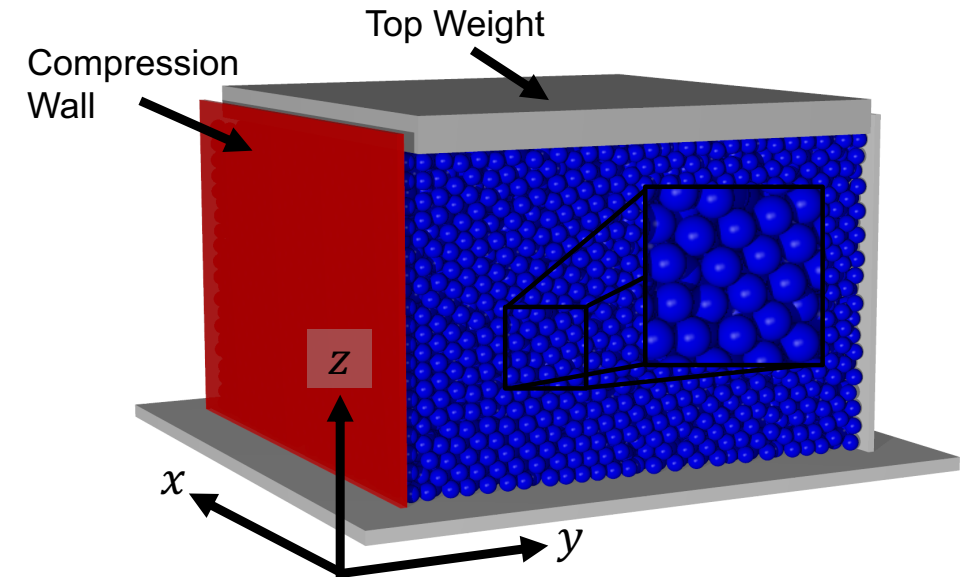
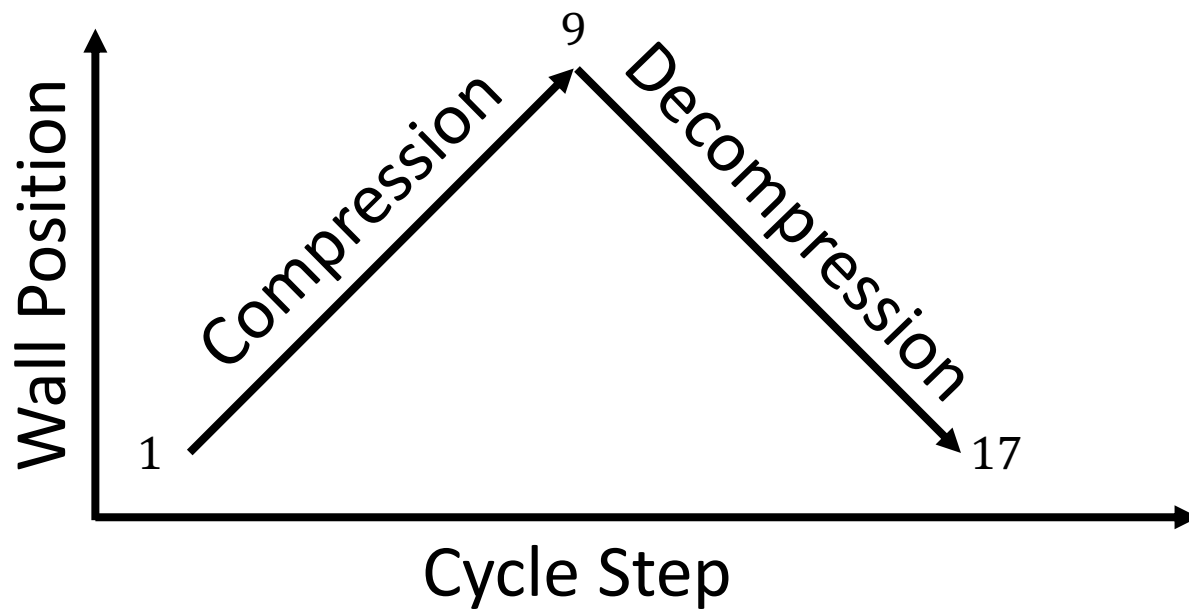
ω_i → angular velocity vector

\vec{v}_i → displacement vector

\vec{d}_{ij} → vector from center of i to the contact point

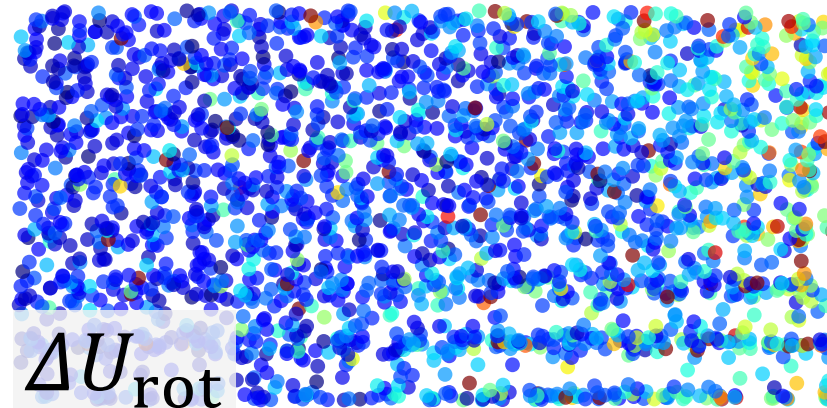
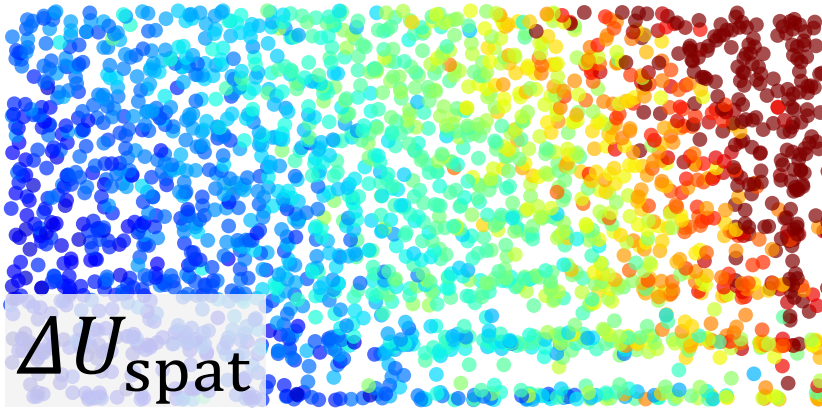
Compression protocol

17 images per full cycle
taken at equal intervals

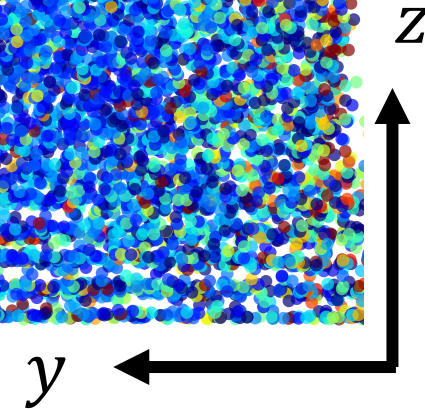
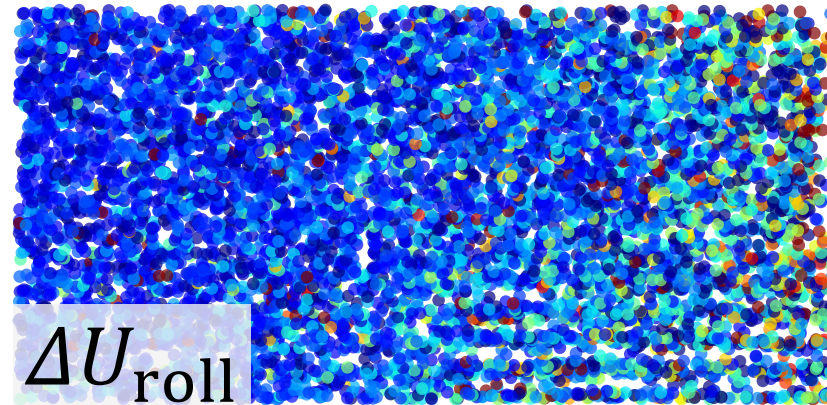
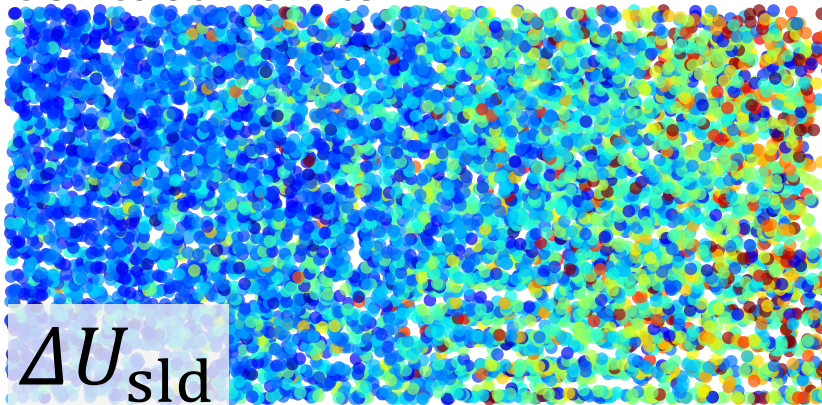


Spatial distribution of displacements at full compression

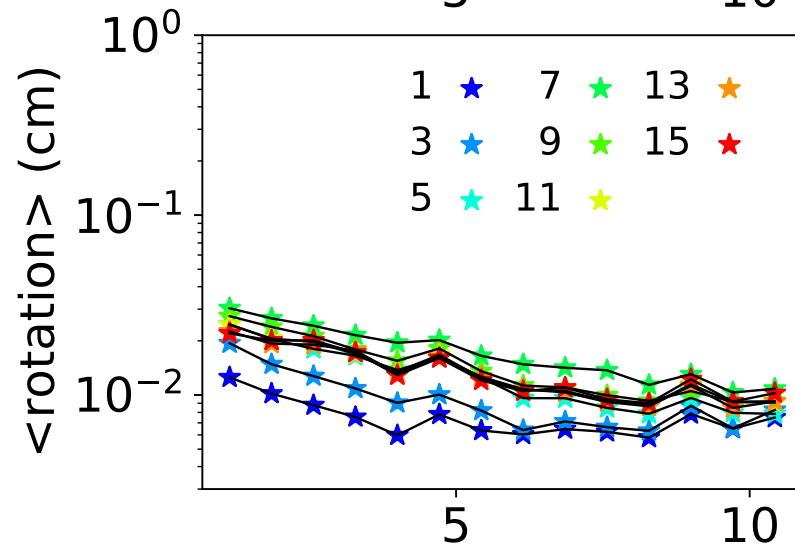
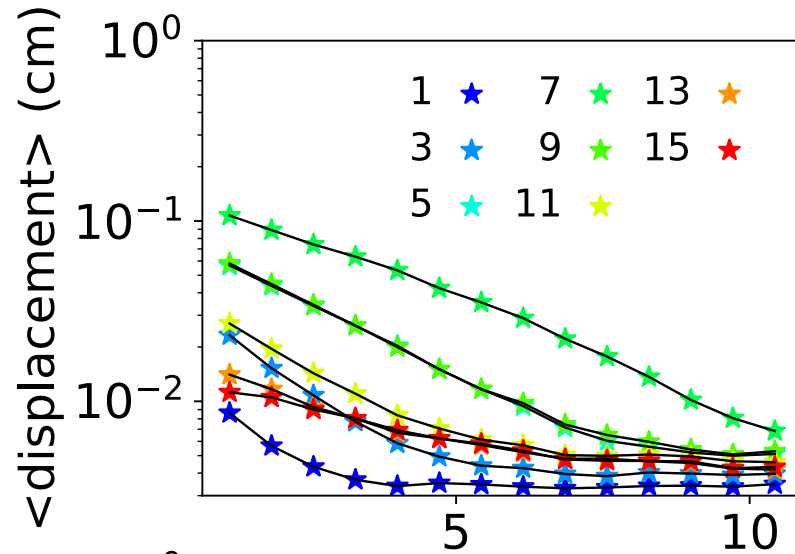
Grain Centers



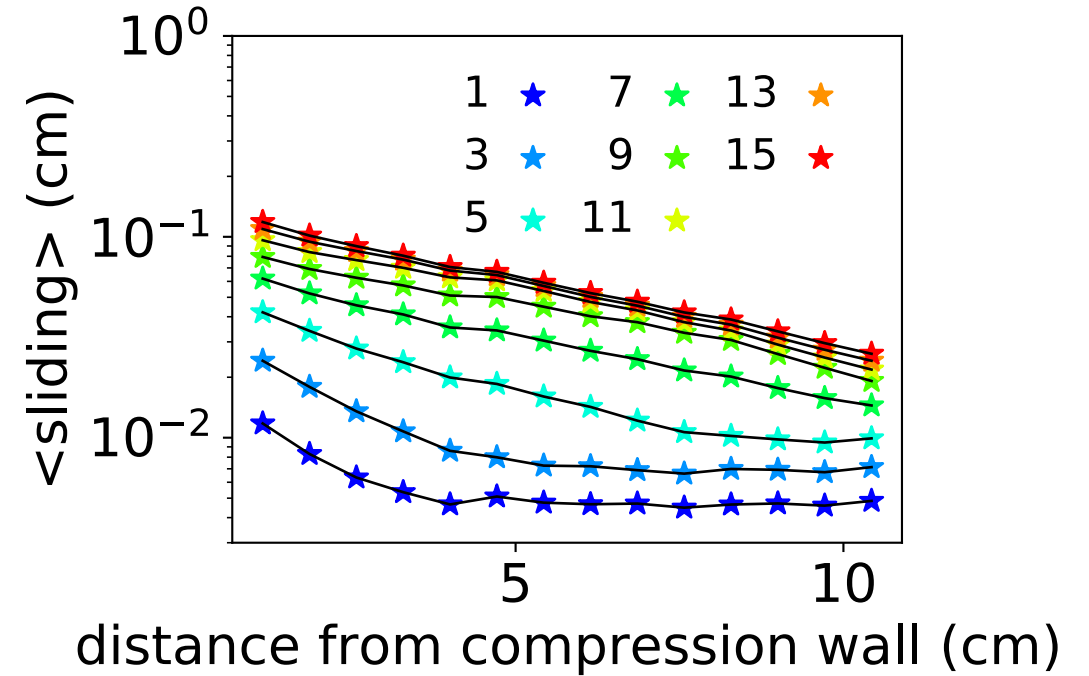
Contact Points



Types of motion penetrates at different lengths in the sample

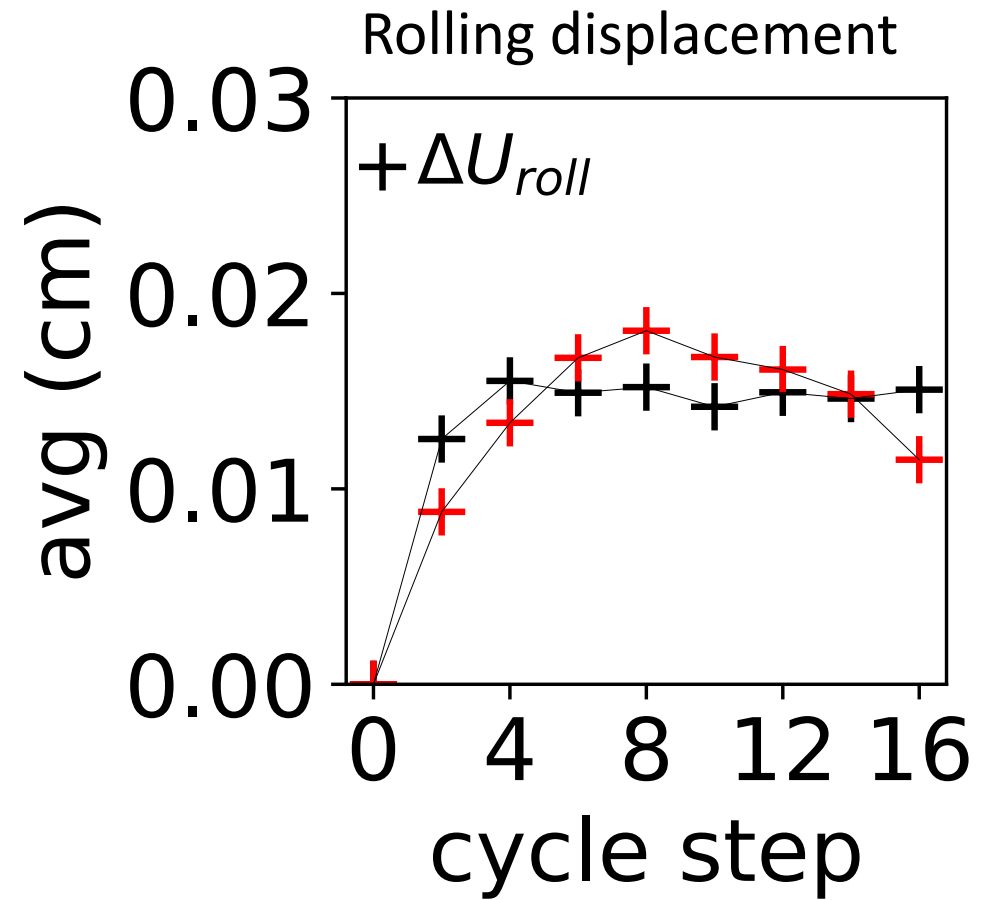
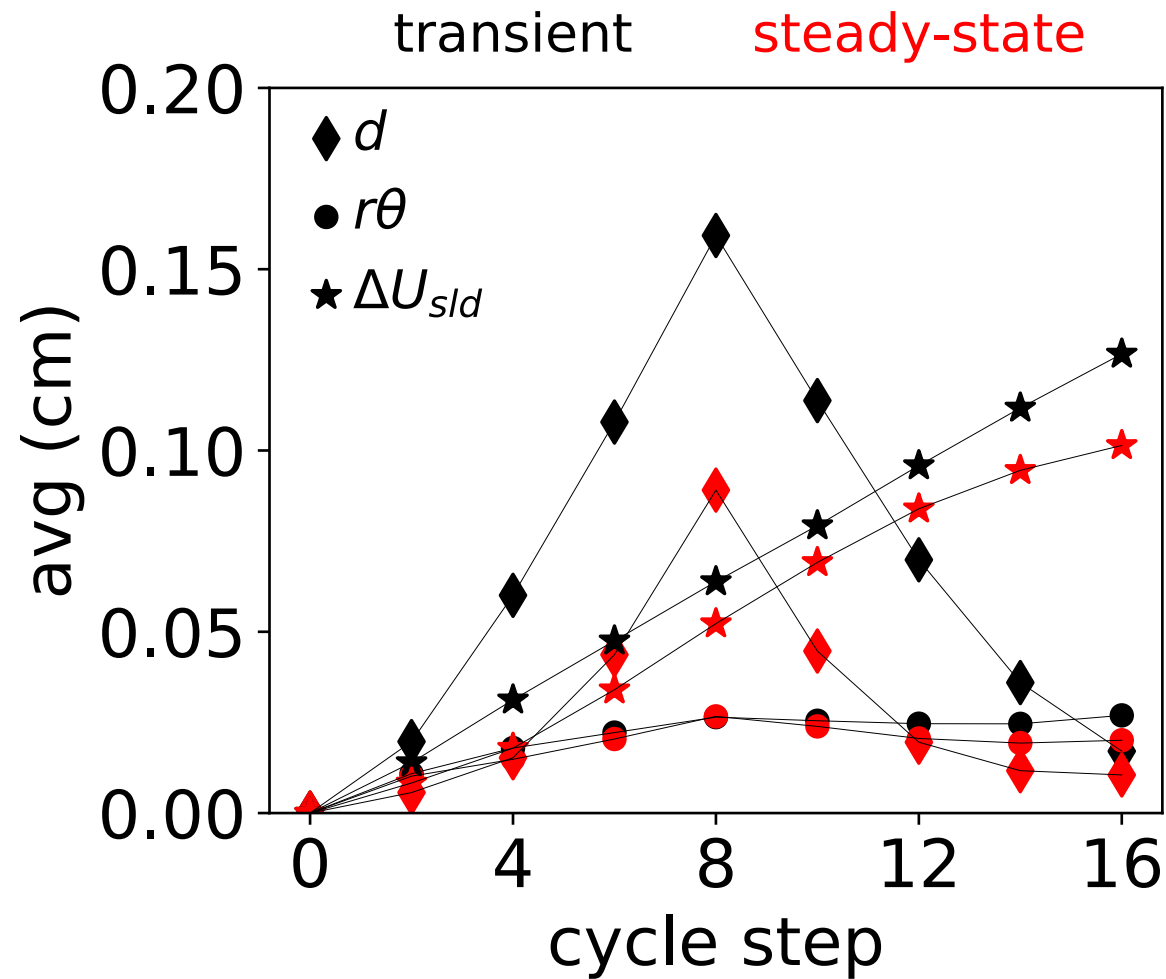


distance from compression wall (cm)

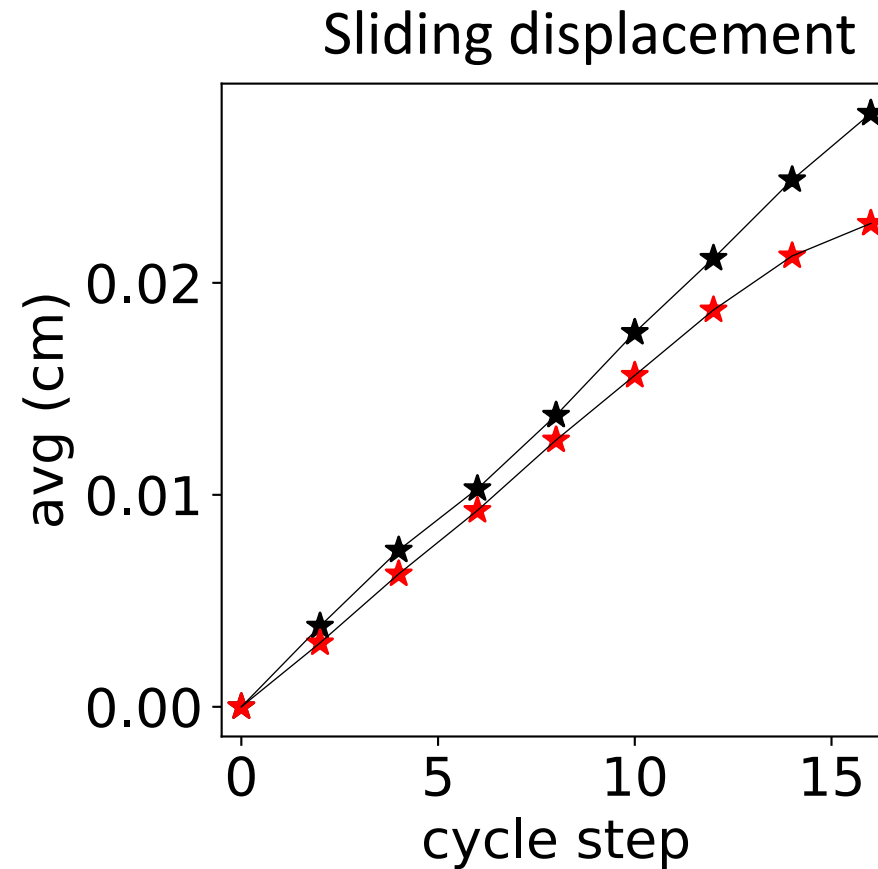
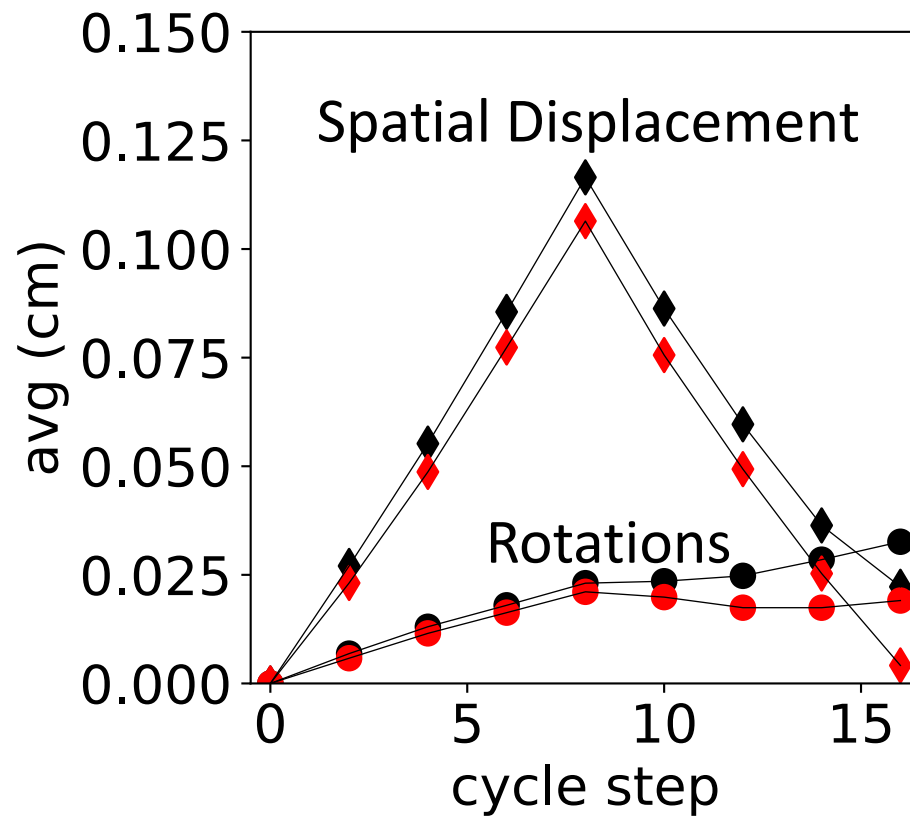


distance from compression wall (cm)

Mean displacements within a cycle



Comparison with DEM simulations



Conclusion and Acknowledgements

Measured 3D rotations of granular spheres.

Implemented VAE to aid in grain identification.

Quantified sliding displacements during cyclic compression.

Found agreement between simulations and experiments in rotational displacement.



Acknowledgements

Wolfgang Losert

Phillip Alvarez

Samira Aghayee

Abby Bull

Lenny Campanello

Sylvester Gates

Rachel Lee

Nick Mennona

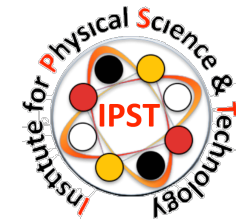
Kate O'Neill

Qixin Yang

Derek C. Richardson (UMD ASTR)

Anton Peshkov (U of Rochester, PHYS)

Nicole Yunger Halpern (Harvard University, PHYS)



Funding Sources

NSF GRFP, NSF DMR

Contact:

zbenson@umd.edu

