

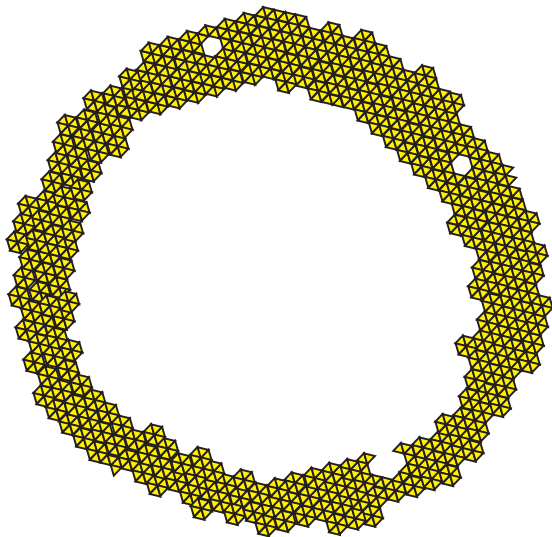
# Topological data analysis

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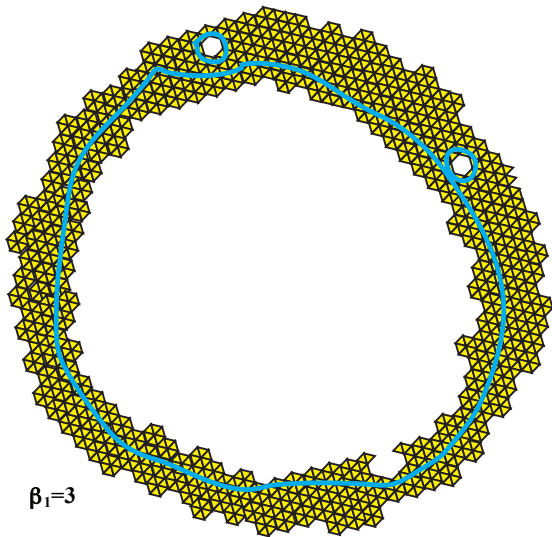
## Example: pointcloud of a circle

First idea: pick some nice  $\epsilon$  to work with  $VR_\epsilon(X)$ .



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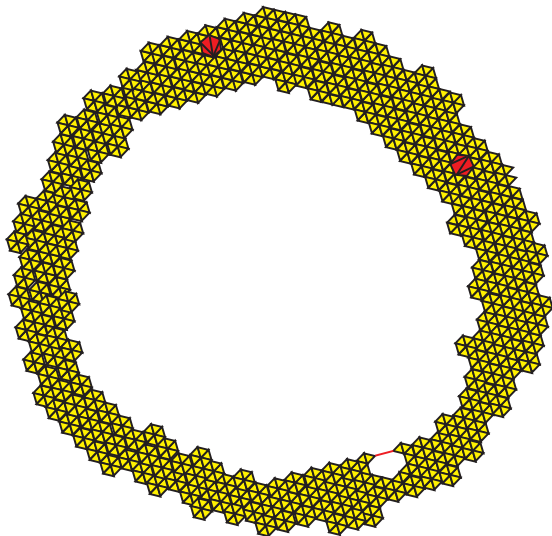
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$\beta_1=3$

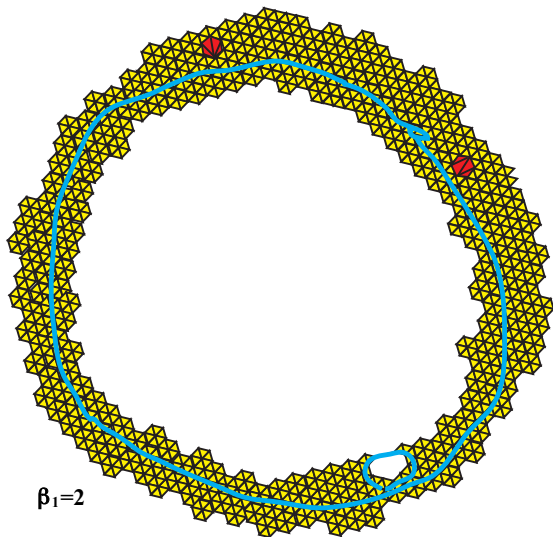
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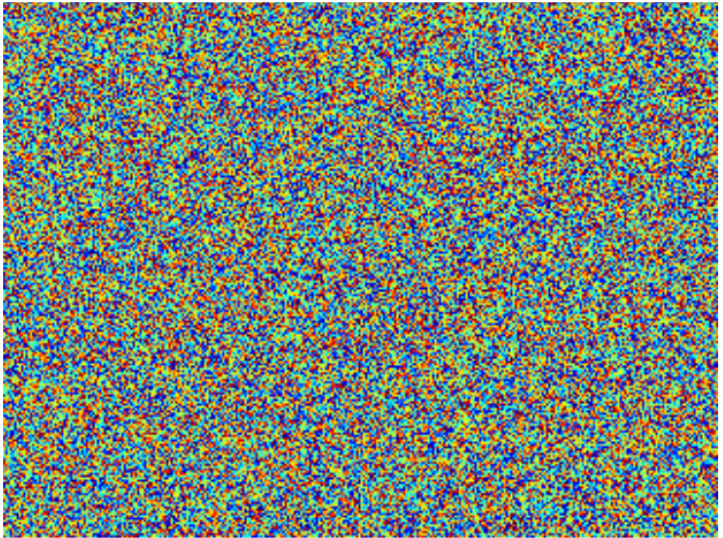


## Example: pointcloud of a circle

First idea: pick some nice  $\epsilon$  to work with  $VR_\epsilon(X)$ .



$\beta_1=2$





## Example: natural images

Lee-Mumford-Pedersen investigated whether a statistically significant difference exists between natural and random images.

Natural images form a “subspace” of all images. Dimension of ambient space e.g.  $640 \times 480 = 307\,200$ .

This space of natural images should have:

- ▶ high dimension: there are many different images.
- ▶ high codimension: random images look nothing like natural ones.

## Natural 3x3 patches

Instead of studying entire images, we consider the distribution of  $3 \times 3$  pixel patches.

Most of these will be approximately constant in natural images. Allowing these drowns out structure.

Lee-Mumford-Pedersen chose 8 500 000 patches with high contrast from a collection of black-and-white images used in cognition research. Each  $3 \times 3$ -patch is considered a vector in  $\mathbb{R}^9$ .

Normalised brightness:  $\mathbb{R}^9 \rightarrow \mathbb{R}^8$ . Normalised contrast:  $\mathbb{R}^8 \rightarrow \mathcal{S}^7$ .

Subsequent topological analysis by Carlsson–de Silva–Ishkanov–Zomorodian.

## Pixel patches in $S^7$

The resulting patches are dense in  $S^7$  – so we consider high-density regions.

Pick out 25% densest points. We can pick a parametrised method to measure density:

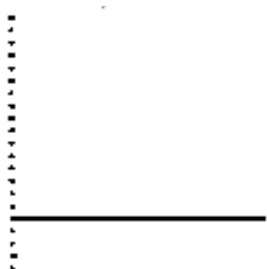
### Definition

$k$ -codensity  $\delta_k(x)$  of a point  $x$  is the distance to its  $k$ th nearest neighbour.

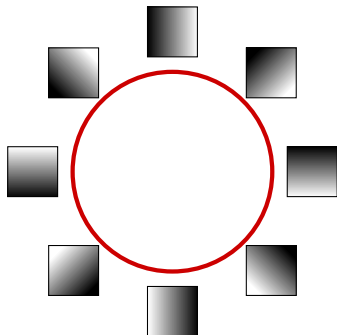
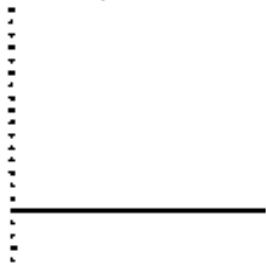
$k$ -density  $d_k(x)$  is  $1/\delta_k(x)$ .

High  $k$  yields a smoothly changing density measure capturing global properties. Low  $k$  yields a wilder density measure capturing local properties.  $k$  acts as a kind of focus control.

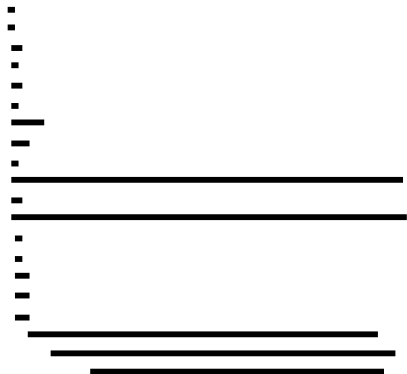
# 300-density



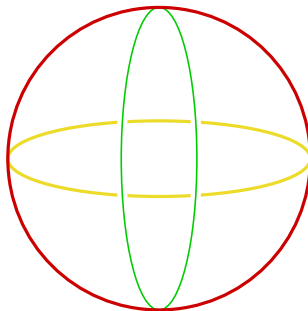
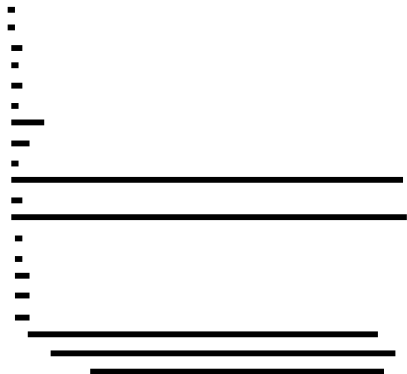
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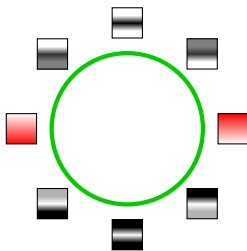
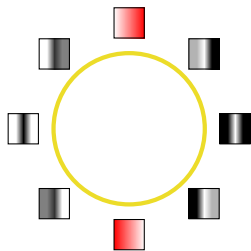
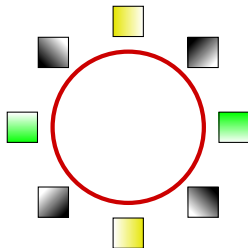
# 15-density



# 15-density



# Three circles

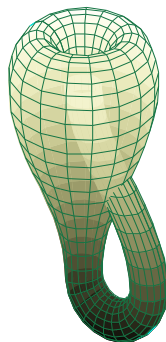
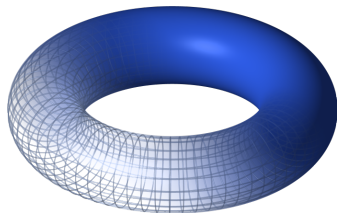


# Identifying the subspace of natural pixel patches

Raising the cut-off bar yields, with coefficients in  $\mathbb{F}_2$

$$\beta_0 = 1 \quad \beta_1 = 2 \quad \beta_2 = 1$$

Assuming the shape is a surface, this corresponds to one of

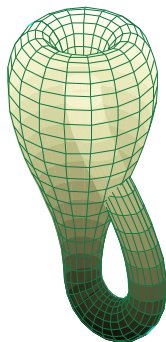


# Identifying the subspace of natural pixel patches

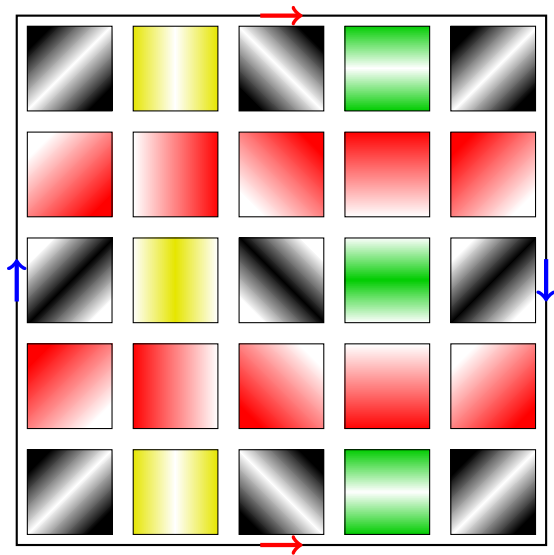
Raising the cut-off bar yields, with coefficients in  $\mathbb{F}_3$

$$\beta_0 = 1 \quad \beta_1 = 1$$

Thus, the relevant shape is:



# Klein bottle of pixel patches



# Applications of this analysis

## Image compression

A  $3 \times 3$ -cluster may be described using 4 values:

- ▶ Position of its projection onto the Klein bottle
- ▶ Original brightness
- ▶ Original contrast

## Texture analysis

Textures yield distributions of occurring patches on the Klein bottle. Rotating the texture corresponds to translating the distribution. [J Perea]

# Coordinatization methods

My own work is on automating the above process by finding topological methods to recover intrinsic coordinate maps.

## Idea

- ▶ Starting from a dataset  $X$ : compute its persistent homology  $H(VR_*(X); k)$
- ▶ Guess a simplicial complex  $Y$  with corresponding homology
- ▶ Find maps  $X \rightarrow Y$  or  $VR_*(X) \rightarrow Y$  that lift to the expected correspondance.

# First results

Joint with Vin de Silva and Dmitriy Morozov.

We can use that the circle is an Eilenberg-Mac Lane space, and thus

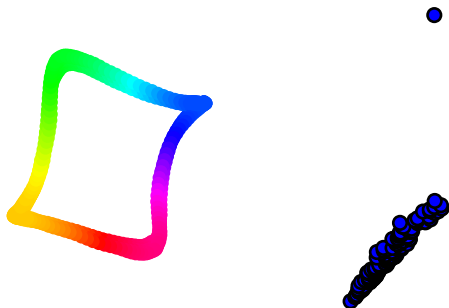
$$H^1(X; \mathbb{Z}) = [X, S^1]$$

We have established a definition of persistent cohomology, and produced techniques, algorithms and software for computing circle-valued coordinate functions using cohomology and a smoothing step.

## Future directions

- ▶ Approach more generic coordinatizations by studying optimal chains in  $H_0(\text{hom}(CX, CY)) = \bigoplus_p \text{hom}(H_p X, H_p Y)$ .
- ▶ Apply the circular coordinates work to periodic and recurrent systems and signals. Currently looking at data sets from: meteorology, climate research, gait research, music.
- ▶ Use circular coordinates for quality control on existing analysis methods for periodic signals.

# Questions?



Delay embedding of a window from a clarinet tone, using circular coordinates and a persistence diagram to quality control the delay embedding.