Microstructure Synthesis via Neural Networks

Daria Fokina

Skoltech

July, 2019

Background

Microstructure - structure, that can be observed under the microscope

Microstructures are studied in:

- Medicine
- Space technologies
- Oil industry



Ceramics



Titanium alloy

The problem – upscaling



Soil slice

https://www.doitpoms.ac.uk https://www.researchgate.net https://www.researchgate.net Aim



Reconstructed image



Considered structure



Key points

- Microstructures have stochastic nature and can be viewed as a realisation of a random variable
- Multiscale modelling techniques are widely used for microstructures:
 - for modelling the response and life prediction of composite materials (C.Oskay, 2015)
 - for flow estimation in porous media (Ronaldo Giro, 2018)
 - for modelling of crack propagation in random heterogeneous media (Darith–Anthony Hun et al., 2019)

Style-GAN

- Number of layers increases during training
- Maximal resolution on the train set 256x256



StyleGAN architecture

Progressive Growing of GANs for Improved Quality, Stability, and Variation

Style-GAN

- Style feature y for AdalN operation: AdalN(x_i , y) = $y_{s,i} \frac{x_i - \mu(x_i)}{\sigma(x_i)} + y_{b,i}$
- Size of the output is equal to the size of train images
- Increase of size via image quilting



StyleGAN generator scheme A - affine transform, B - per-channel scaling of noise

A Style-Based Generator Architecture for Generative Adversarial Networks

Image quilting



Image stacking without quilting



Image stacking with quilting

Image quilting

- Quilting paths paths, minimising error on the overlap between two images
- Minimal error on (i,j)-th pixel of overlap:



Minimum boundary cut

$$E_{i,j} = \begin{cases} e_{i,j}, \ j = 0\\ e_{i,j} + \min(E_{(i-1),j}; E_{(i-1),(j-1)}; E_{(i-1),(j+1)}), \text{ otherwise} \end{cases}$$

 $e_{i,j} = (x_{i,j} - y_{i,j})^2,$ $x_{i,j}, y_{i,j}$ - (i,j)-th pixel of image 1 and 2 correspondingly

Wang tilings

• The plane is covered with tilings chosen randomly from a set of 16 tilings with 4 colors of edges



Process of tiles design

Aperiodic Compression and Reconstruction of Real World Material Systems Based on Wang Tiles

Experiments

Used structures



Alporas aluminium foam



Berea sandstone

Ketton limestone

Estimated values

Mechanical properties

- Poisson's ratio (ν)
- Young's modulus (E)

Minkowski functionals:

- Area density
- Perimeter density
- Euler2D density

Visual comparison of the results



Original image



Texture networks



Periodic unit cell (PUC)



Spatial GAN



Wang tilings



Style-GAN

Visual comparison of the results Alporas



Original image



Wang tilings result

Visual comparison of the results Alporas



Original image



Style-GAN result

Numerical results Mechanical properties

	Method		
	Original image	PUC	Wang tilings
E	0.0988 ± 0.0032	0.0966 ± 0.0112	0.0950 ± 0.0054
ν	0.3507 ± 0.0047	0.3460 ± 0.0190	0.3331 ± 0.0094

	Texture Networks	Spatial GAN	Style-GAN
E	0.0826 ± 0.0013	0.1120 ± 0.0094	0.0958 ± 0.0025
u	0.3191 ± 0.0049	0.3266 ± 0.0129	0.3634 ± 0.0084

Visual comparison of the results Berea



Original image

Wang tilings result

Style-GAN result

Visual comparison of the results Ketton



Original image

Wang tilings result

Style-GAN result

Numerical results

Minkowski functionals

Berea

	Original image	Wang tilings	Style-GAN
Area	0.7970 ± 0.0528	0.8292 ± 0.0481	0.8167 ± 0.0046
Perimeter	0.0633 ± 0.0073	0.0653 ± 0.0115	0.0667 ± 0.0030
Euler2D	-0.0009 ± 0.0004	-0.0017 ± 0.0005	-0.0013 ± 0.0002

Ketton

	Original image	Wang tilings	Style-GAN
Area	0.8753 ± 0.0206	0.9149 ± 0.0323	0.9112 ± 0.0167
Perimeter	0.0501 ± 0.0054	0.0409 ± 0.0127	0.0438 ± 0.0059
Euler2D	-0.0008 ± 0.0003	-0.0010 ± 0.0004	-0.0011 ± 0.0003

Thank you for attention!

Appendix: Mechanical properties computation

Linear elastic equation

$$-\nabla(C^{base}: [\epsilon(w^{kl}) + e^{kl}]) = 0 \text{ in } Y,$$

where $\epsilon(w^{kl})$ is unknown strain tensor,

$$C^{base}\tau = 2\mu(\tau - \lambda \operatorname{tr}(\tau)I) \cdot \rho, \quad (C:\epsilon)_{ij} = \sum_{k,l} C_{ijkl}\epsilon_{kl},$$

$$e^{kl} = \frac{1}{2}(e_k \otimes e_l + e_l \otimes e_k), \quad e_k \text{ - column of identity matrix,}$$

$$\mu = E/(2(1 + nu)), \quad \lambda = E\nu/(1 - \nu^2), \quad E = 1.0, \quad \nu = 0.3,$$

$$\rho \text{ - material density}$$

Appendix: Mechanical properties computation

Homogenised elasticity tensor:

$$C_{ijkl}^{H} = \frac{1}{|Y|} \int_{\omega} (e^{ij} + \epsilon(w^{ij}(y))) : C^{base} : (e^{ij} + \epsilon(w^{ij}(y)))dy$$

where:

$$C^{base}\tau = 2\mu(\tau - \lambda \mathrm{tr}(\tau)I) \cdot \rho,$$

$$\mu = E/(2(1 + \nu)), \ \lambda = E\nu/(a - \nu^2),$$

$$E = 1.0, \nu = 0.3$$

Appendix: Minkowski functionals

Three functionals for two-dimensional structure:

- Area
- Perimeter
- Euler characteristic $\chi = V E + F$, V number of vertices,
- *E* number of edges, *F* number of regions