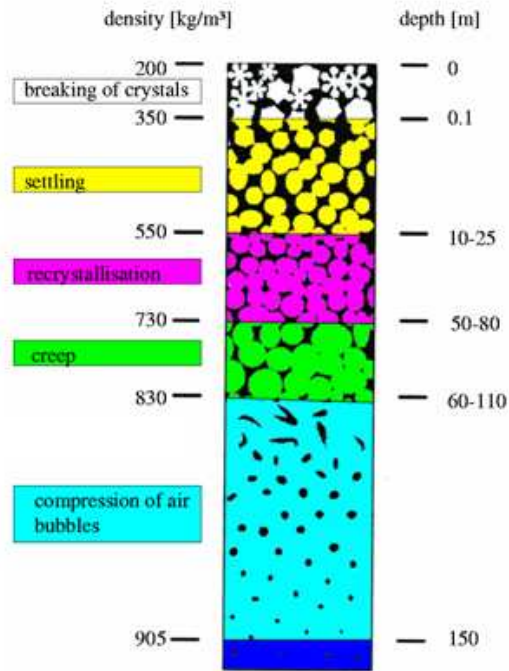

Anisotropy analysis of 3d point processes

Claudia Lautensack
with Aila Säykkä, Johannes Freitag, Katja Schladitz

Hochschule Darmstadt

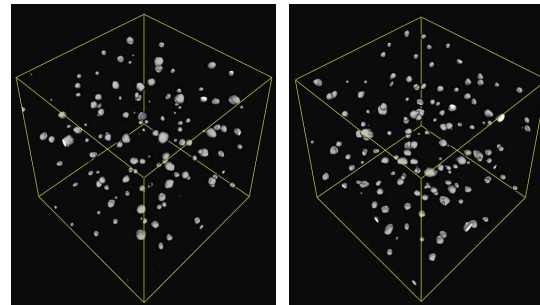
Smögen Workshop
Smögen, August 18-22, 2008

Motivation



Polar ice is compacted (sintered) snow. During the compression air pores are isolated in the ice.

Image: Freitag 1), Kipfstuhl 1), Stauffer 2)
1) Alfred-Wegener-Institute for polar and marine research, Bremerhaven
2) University Bern



Question

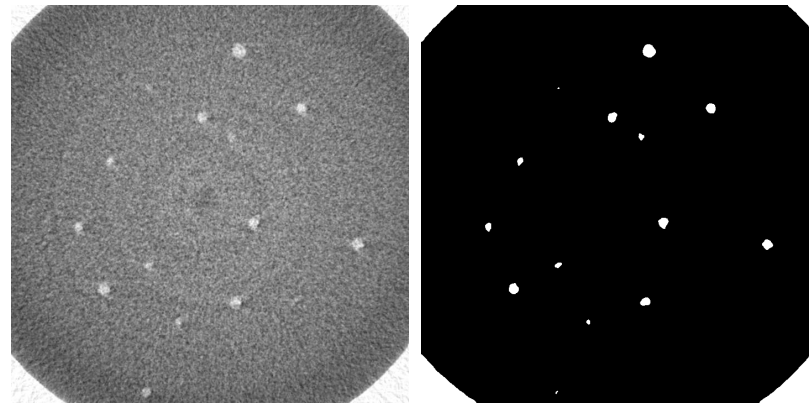
Does the location of the pores tell anything about the movements of the ice?

Data

tomographic images of ice samples
cylinder 15 mm height, 15 mm diameter
imaged inside a cold room at -15°C

images from three depths: 153 m, 353 m, 505 m

14 samples per depth



Model for pressing

shape of the pores not significant -> concentrate on pore centres

Transformation

volume preserving compression

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} \mapsto \begin{pmatrix} \frac{x}{\sqrt{c}} \\ \frac{y}{\sqrt{c}} \\ cz \end{pmatrix}, 0 < c < 1$$

Aim

- detection of the distortion
- estimation of the parameter c

Methods in 2d

- point-pair rose density (Stoyan, Beneš, 1991)
- directional pair correlation function (Stoyan, Stoyan, 1992)
- Fry-method (Fry, 1979)
- directional K-function, 0-contour of the density $\frac{d}{d\varphi}K(r, \varphi)$ (Stoyan, Kendall, Mecke, 1995)

Idea

partition of circle
compare estimates of summary statistics for different directions

Problems and ideas in 3d

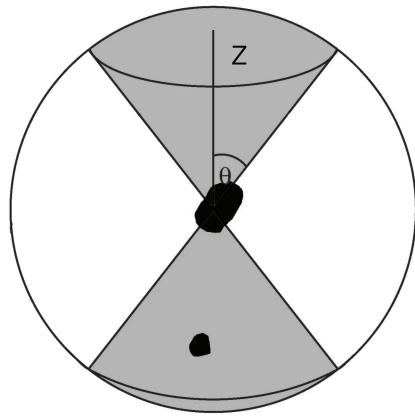
Requirements

- measurement of characteristics easy and robust
- methods for visualisation of results

Problems

- more parameters than in 2d
 - partition of the ball into suitable equal volume parts?
- > choose cumulative instead of density functions
- > adapt directions of investigation to the problem

Idea



- investigate point pattern in double cones aligned along coordinate axes
- compare observations
- large differences -> anisotropy

here: $\theta = \frac{\pi}{4}$

Advantage of cone

- + easy parametrisation w.r.t. spherical coordinates
- + can be rotated to arbitrary directions -> adaptable to data

Directional summary statistics: directional K

Directional version of the K -function

K_{dir}

Second reduced moment measure of the cone C
expected number of points within the double cone $x_0 + C$

Estimator

$$\lambda^2 \hat{K}_{\text{dir},u,\theta}(r) = \sum_{x \in \psi} \sum_{y \in \psi, y \neq x} \frac{\mathbb{1}_{C_u(r,\theta)}(x-y)}{|W_x \cap W_y|}, \quad r \geq 0,$$

W_x translation of the window W by the vector x
 $|B|$ volume of a set $B \subset \mathbb{R}^3$

Directional summary statistics: local G

Directional version of the nearest neighbour distance distribution function G

G_{loc} Distribution function of the distance from the typical point of the process to the closest point in the cone $x_0 + C$.

Estimator

$$\hat{G}_{\text{loc},u,\theta}(r) = \frac{\sum_{(x,d) \in \psi} \mathbf{I}_{[0,r]}(d) \mathbf{I}\{x + C_u(d, \theta) \subset W\}}{\sum_{(x,d) \in \psi} \mathbf{I}\{x + C_u(d, \theta) \subset W\}}, \quad r \geq 0.$$

Directional summary statistics: global G

Directional version of the nearest neighbour distance distribution function G

G_{glob}

Distribution function of the distance between x_0 and its nearest neighbour y conditioned on $y \in x_0 + C$.

Estimator

$$\hat{G}_{\text{glob},u,\theta}(r) = \frac{\sum_{(x,y) \in \Psi} \mathbf{1}_{C_u(r,\theta)}(x-y) \mathbf{1}_{W \ominus b(0, \|x-y\|)}(x)}{\sum_{(x,y) \in \Psi} \mathbf{1}_{C_u(\infty,\theta)}(x-y) \mathbf{1}_{W \ominus b(0, \|x-y\|)}(x)}, \quad r \geq 0.$$

less points investigated, stronger for higher intensities?

Isotropy tests

Given

n point patterns ψ_1, \dots, ψ_n .

$\hat{S}_x, \hat{S}_y, \hat{S}_z$

Estimators of one of the summary statistics with respect to the x -, y -, and z -direction.

Test statistics

$$T_{xy} = \int_{r_1}^{r_2} |\hat{S}_x(r) - \hat{S}_y(r)| dr, \text{ and}$$

$$T_z = \min \left(\int_{r_1}^{r_2} |\hat{S}_x(r) - \hat{S}_z(r)| dr, \int_{r_1}^{r_2} |\hat{S}_y(r) - \hat{S}_z(r)| dr \right),$$

where $[r_1, r_2]$ is a given interval.

Isotropy tests

Test statistics

$$T_{xy} = \int_{r_1}^{r_2} |\hat{S}_x(r) - \hat{S}_y(r)| dr, \text{ and}$$

$$T_z = \min \left(\int_{r_1}^{r_2} |\hat{S}_x(r) - \hat{S}_z(r)| dr, \int_{r_1}^{r_2} |\hat{S}_y(r) - \hat{S}_z(r)| dr \right),$$

Test

reject isotropy hypothesis at level α if $T_{z,i}$ is larger than 100(1 - α)% of the estimated T_{xy} values

Alternative

Monte Carlo test, if few replications available requires model for the data

Simulation study

Simulate regular point patterns in the unit cube.
Compress with pressing factor c .

Investigate different

- degrees of regularity: Matérn hard core point process and packing of balls (force biased algorithm)
- intensities: $\lambda = 500$ and $\lambda = 1000$
- hard core radii: $R = 0.025, 0.05, \text{ and } 0.075$
- pressing factors: $c = 0.7, 0.8, \text{ and } 0.9$

Simulation results: Matérn hard core

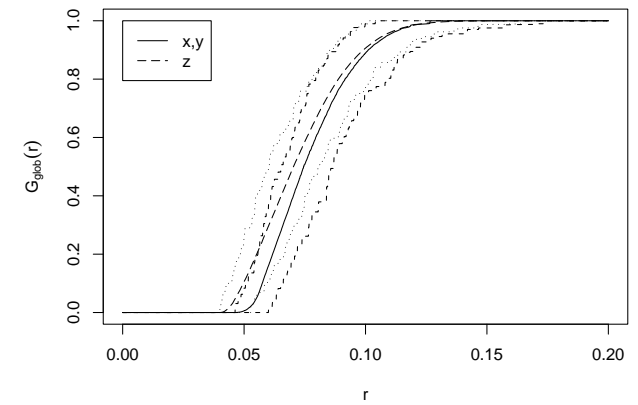
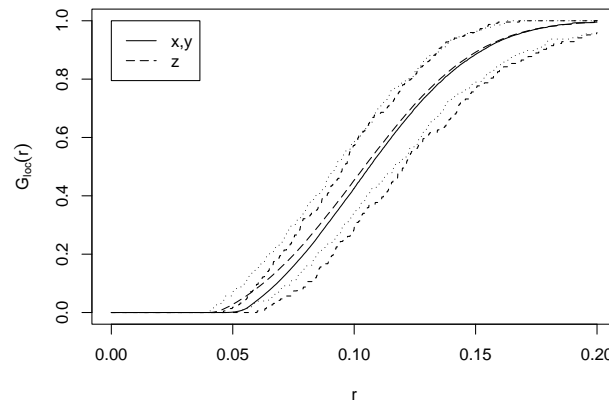
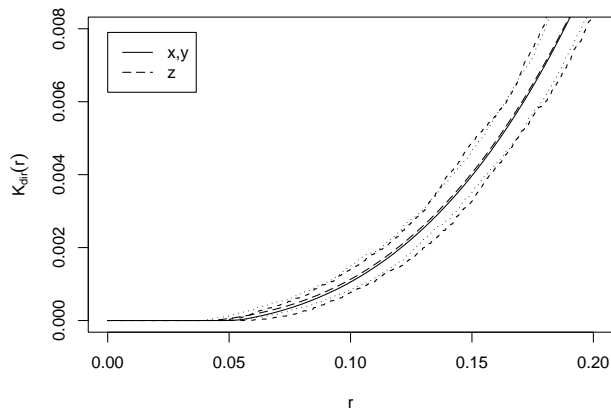
Parameters

$\lambda = 500$, $R = 0.05$, and $c = 0.8$

K_{dir}

G_{loc}

G_{glob}



Simulation results: Force biased

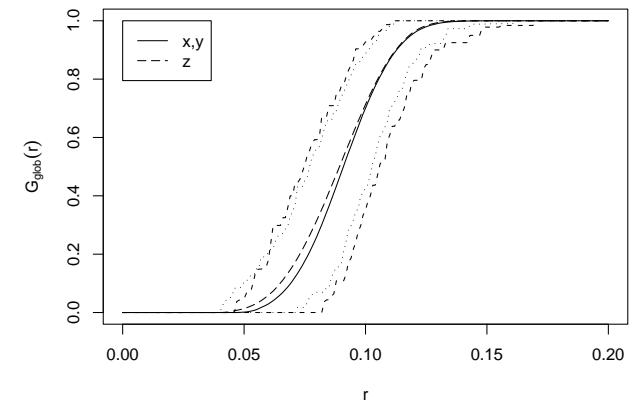
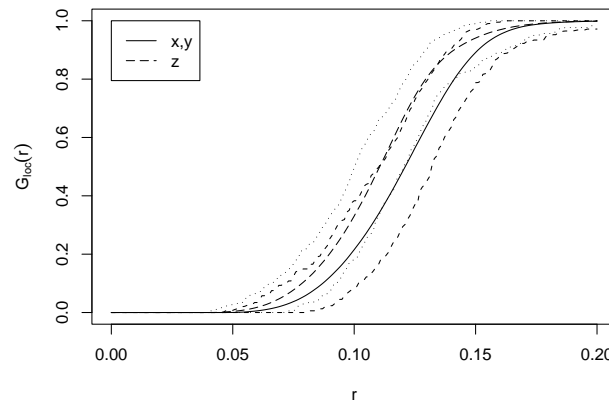
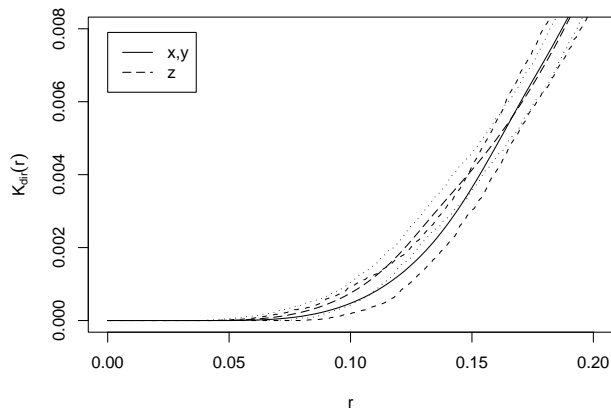
Parameters

$\lambda = 500$, $R = 0.05$, and $c = 0.8$

K_{dir}

G_{loc}

G_{glob}



Powers of tests
Investigate influence of

- degree of regularity, intensity, hard core radius, pressing factor
- interval of observation

test intervals

$[0, 1.1R]$, $[0, 4/3 R]$, $[0, 0.1]$ ($[0, 0.2]$, resp.)

R	0.05	0.05	0.05	0.075	0.075	0.075
r_2	0.1	0.067	0.055	0.2	0.1	0.0825
G_{loc}	17.2	73.8	97.2	91.7	99.8	100
G_{glob}	21.3	68.2	96.6	93.9	98.6	100
K_{dir}	19.9	79.7	98.8	36.9	100	100

powers of the tests on a 5% significance level, based on 1000 realisations

Existence of outliers

In real data the existence of outliers, i.e. points which violate the hard core condition, is likely. How does this influence the results?

Therefore:

- use point patterns from previous simulation
- choose 5 random points x_1, \dots, x_5 from each pattern
- include a further point y_i in balls of radius R centred in x_i

Result

- decreasing powers of the tests
- better results for K_{dir} and G_{loc} than for G_{glob} .
- larger integration intervals should be chosen.

Alternative: Investigate direction to nearest neighbour

Given

set of unit vectors $v_i = (x_i, y_i, z_i), i = 1, \dots, n$

Orientation matrix

$$A = \begin{pmatrix} \sum x_i^2 & \sum x_i y_i & \sum x_i z_i \\ \sum x_i y_i & \sum y_i^2 & \sum y_i z_i \\ \sum x_i z_i & \sum y_i z_i & \sum z_i^2 \end{pmatrix}$$

$$\lambda_1 \geq \lambda_2 \geq \lambda_3$$

eigenvalues of A

Test statistic

largest eigenvalue λ_1

Significance points

at 5% level: $\frac{1}{3} + \frac{0.873}{\sqrt{n}}$ for $n > 100$, Anderson, Stephens (1972)

Results

weaker performance than tests based on summary statistics

Summary of results

- tests using summary statistics better than analysis of direction to nearest neighbour

Higher power for

- higher regularity, larger intensities, stronger pressing
- integration interval should be chosen suitably
- in most cases: K_{dir} yields best results
- test with G functions more stable when changing interval
- tests also work for compressed clustered patterns

Estimation of the pressing factor

- rescale by $(\sqrt{d}, \sqrt{d}, \frac{1}{d})$ with $d \in [0.6, 1.1]$
- compute statistics

$$T_{\Sigma, d} = \int_{r_1}^{r_2} (|\hat{S}_{x,d}(r) - \hat{S}_{y,d}(r)| + |\hat{S}_{y,d}(r) - \hat{S}_{z,d}(r)| + |\hat{S}_{z,d}(r) - \hat{S}_{x,d}(r)|) dr,$$

Estimator for c

$$\hat{c} = \operatorname{argmin}_d T_{\Sigma, d}$$

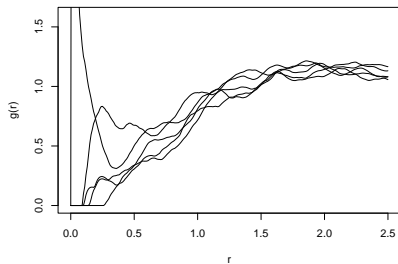
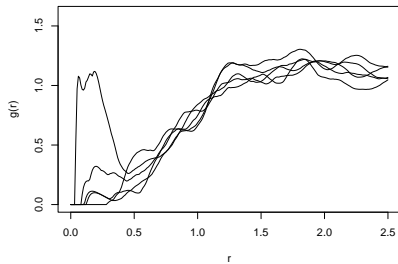
Results for simulated data

Parameters:

$$\lambda = 500, R = 0.05$$

r_2	c	\bar{c}_{loc}	MSE	\bar{c}_{glob}	MSE	\bar{c}_K	MSE
0.055	1.0	0.9828	3.944e-3	0.9795	6.763e-3	0.9875	4.850e-3
0.055	0.9	0.8810	5.813e-3	0.8855	5.138e-3	0.8933	3.631e-3
0.055	0.8	0.7780	5.363e-3	0.7818	4.544e-3	0.7923	2.844e-3
0.055	0.7	0.6880	1.850e-3	0.6913	1.981e-3	0.6860	1.988e-3
0.15	1.0	1.0005	2.625e-3	0.9385	2.081e-2	0.9955	1.813e-3
0.15	0.9	0.8953	2.806e-3	0.9060	1.819e-2	0.8993	1.669e-3
0.15	0.8	0.7968	2.306e-3	0.8655	2.511e-2	0.8030	1.013e-3
0.15	0.7	0.6958	1.631e-3	0.8128	3.222e-2	0.7015	7.125e-4

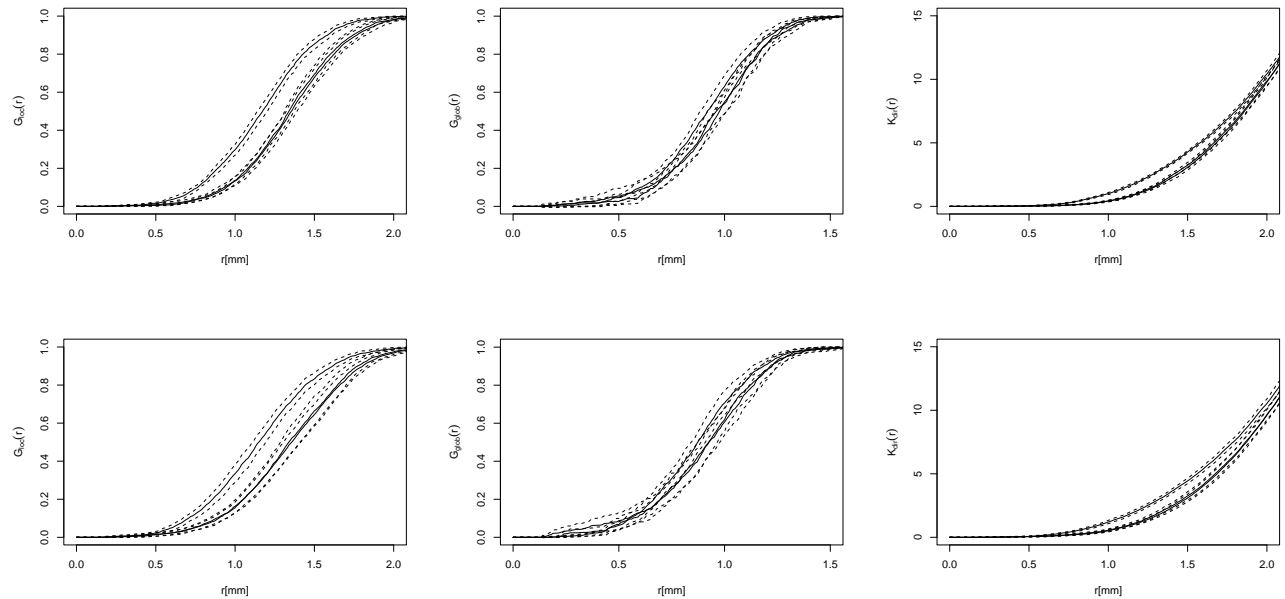
Ice: Choice of parameters



- between 329 and 733 pores per sample
 - study degree of regularity using isotropic pair-correlation function
 - investigation of distances to nearest neighbours shows existence of outliers
- > choose larger interval of observation

Ice: Isotropy test

Means and confidence bands of directional summary statistics



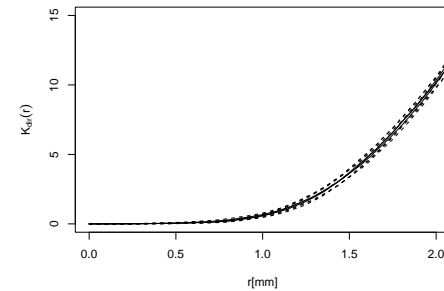
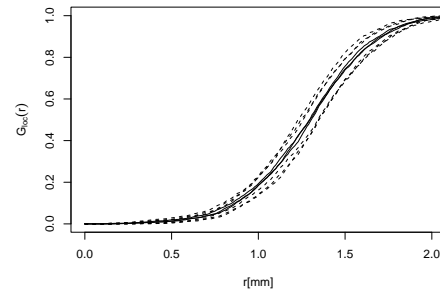
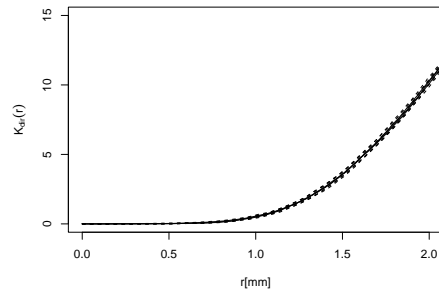
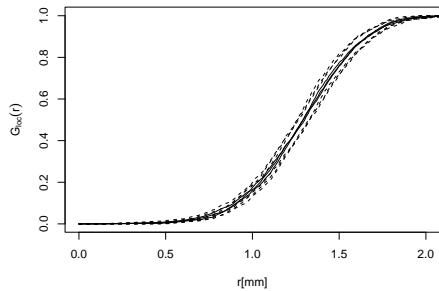
353 m

505 m

Pooling of the data as in Baddeley et. al. (1993)

Ice: Estimation of pressing factors

sample	153 m	353 m	505 m
N_V	380.29	454.79	516.71
λ	0.2528	0.3403	0.3241
\hat{c}_G	0.807	0.630	0.534
\hat{c}_K	0.821	0.641	0.545



Summary

We have

- detected the anisotropy within the ice
- estimated the pressing factor

Nye formula

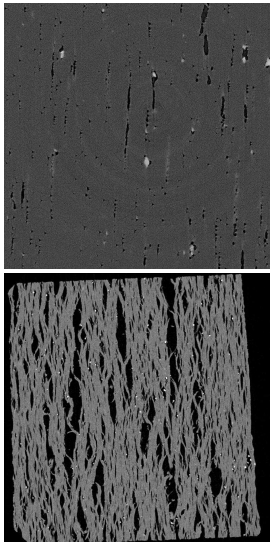
- simplified ice flow model
- > same trend but absolute values higher in our estimates (approx. 0.1)

However

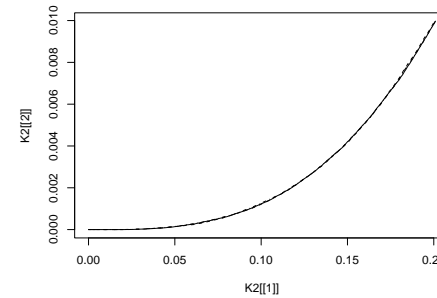
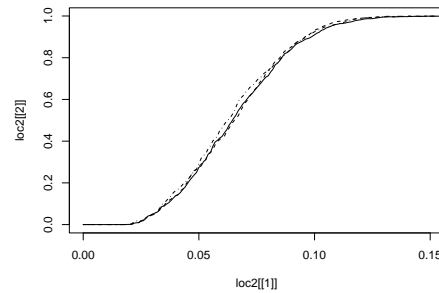
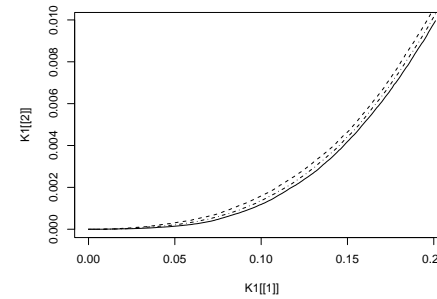
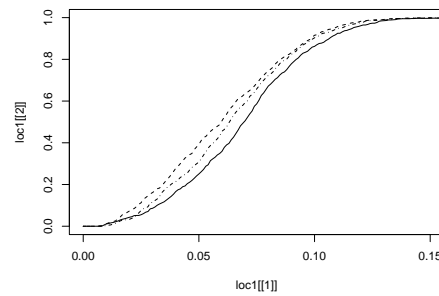
bedrock conditions not taken into account
unknown so far due to incomplete drilling
could shift expected pressing factors to higher values

Further example: Foam

aluminium foam: pressed before foaming process
degree of isotropy grows during foaming



grey: aluminium
white: foaming agent
black: pore



References

Anderson, Stephens (1972) Tests for randomness of directions against equatorial and bimodal alternatives. *Biometrika* 59/3), 613-621

Baddeley, Moyeed, Howard, Boyde (1993) Analysis of three-dimensional point patterns with replication. *Applied Statistics* 42, 641-668

Fry (1979), Random point distributions and strain measurement in rocks. *Tectonophysics* 60, p. 89-105, 1979

Lautensack, Särkkä, Freitag, Schladitz (2008) Anisotropy analysis of pressed point processes, ITWM report no. 141, 2008

Stoyan, Beneš(1991), Anisotropy analysis for particle systems. *Journal of Microscopy* 164, p. 159-168

Stoyan, Kendall, Mecke (1995), *Stochastic Geometry and its applications*. Wiley, Chichester

Stoyan, Stoyan (1992), *Fraktale, Formen, Punktfelder*. Akademie Verlag, Berlin