

# Structural invariance of stochastic fibrous networks

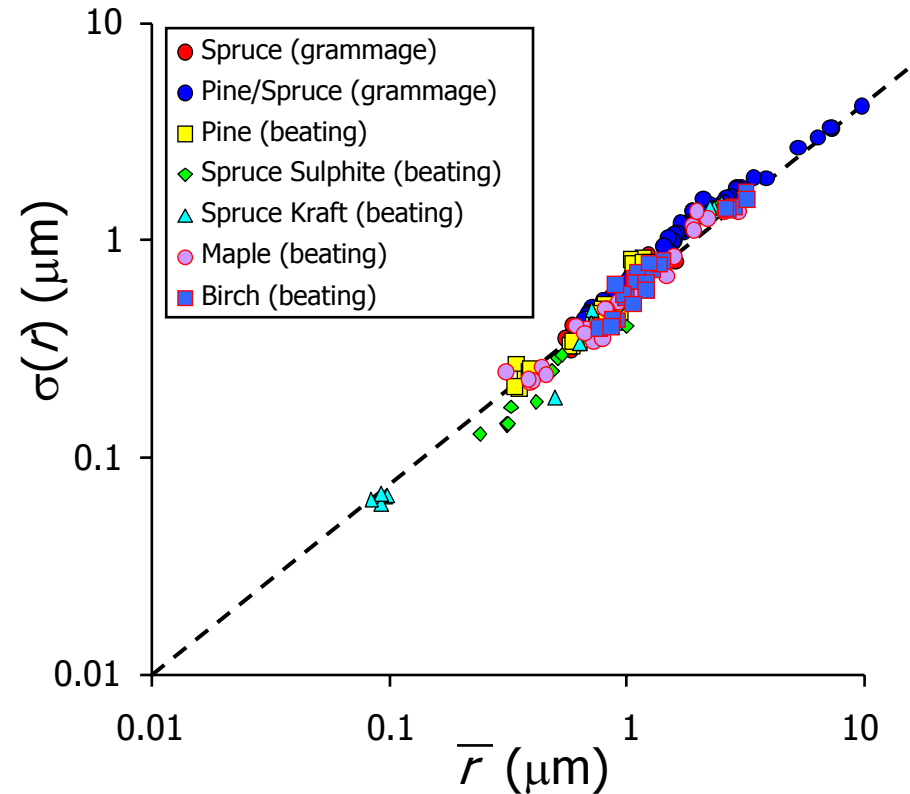
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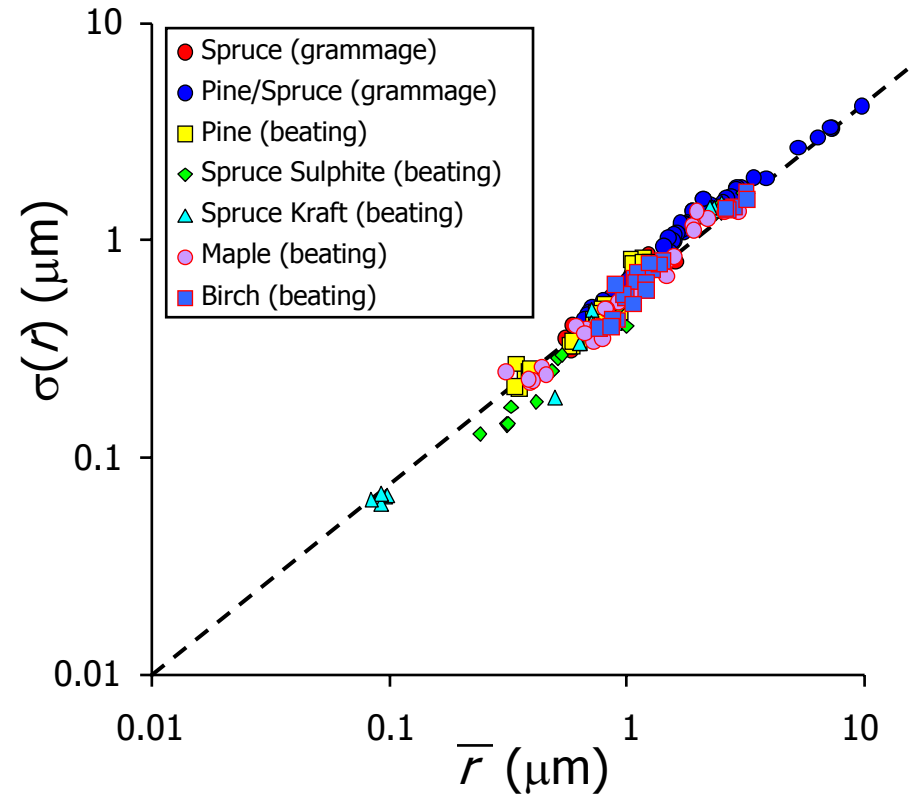
# Structural invariance of voids: Pore size distribution

- Standard deviation of pore radii proportional to the mean:
  - Property of the gamma distribution
  - Measurements in real paper dominated by pore height distribution (exponential)
  - Insensitive to formation



# Structural invariance of voids: Pore size distribution

- Standard deviation of pore radii proportional to the mean:
  - Property of the gamma distribution
  - Measurements in real paper dominated by pore height distribution (exponential)
  - Insensitive to formation
- Polygons tend to be 'roundish'
  - Conjecture, FRS13: correlated free-fibre-lengths



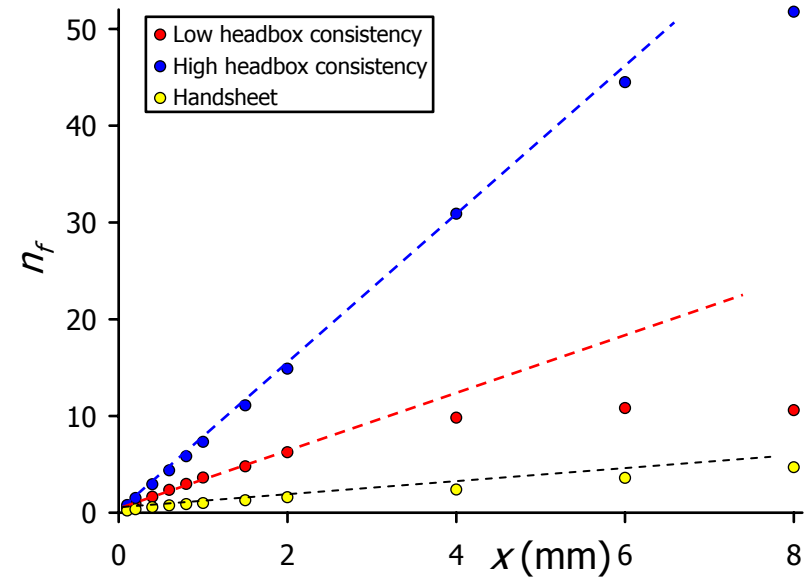
# Structural invariance of mass: Formation number

- Variance of local grammage,  $\sigma_x^2(\tilde{\beta})$ , of random fibre networks known analytically

- Formation number (variance ratio):

$$n_f = \frac{\sigma_x^2(\tilde{\beta})^{\text{measured}}}{\sigma_x^2(\tilde{\beta})^{\text{random}}}$$

- $n_f$  vs. inspection zone size ( $x$ ) 'linear' for  $x < 4$  mm



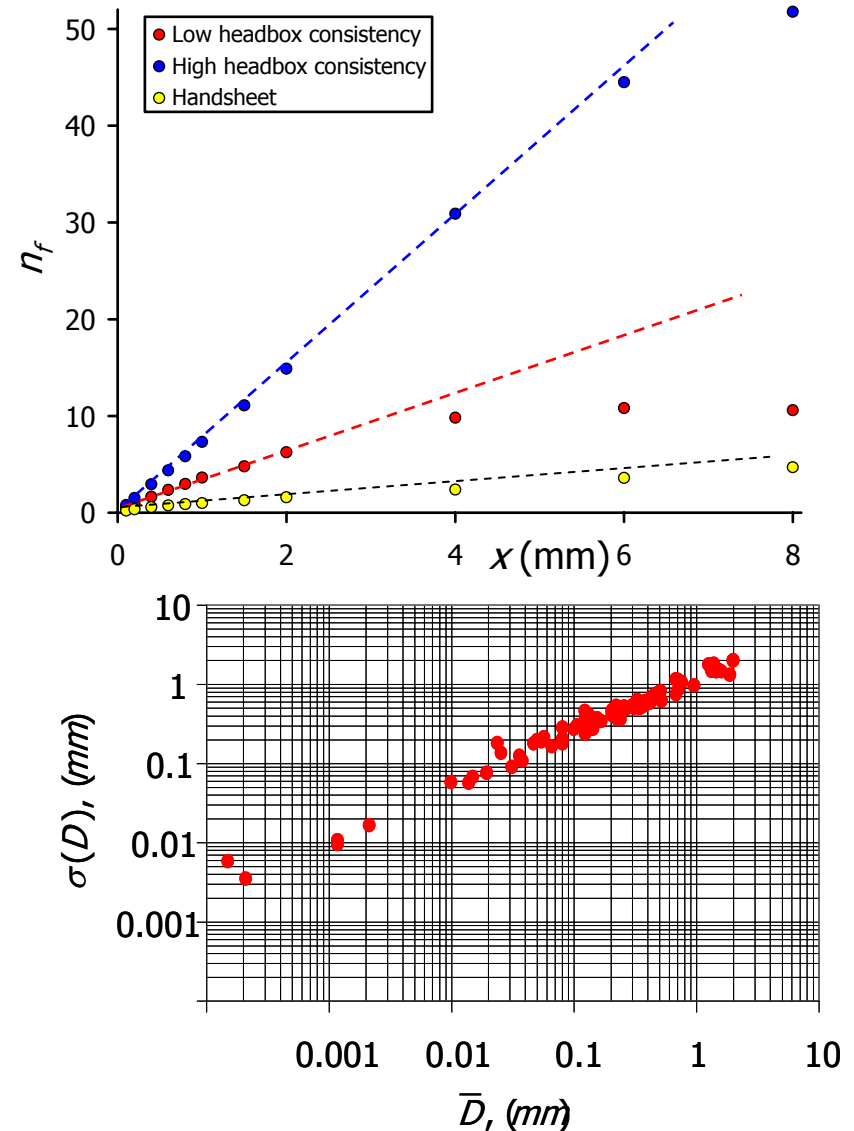
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- $n_f$  vs. inspection zone size ( $x$ ) 'linear' for  $x < 4$  mm
- Random disk model:

$$\sigma(D) \propto \bar{D}^{2/3}$$



# Objectives and Approach

## Objective:

- **To explain the seemingly narrow class of structures realised in papermaking processes:**  
pore shape, pore size distribution and mass distribution

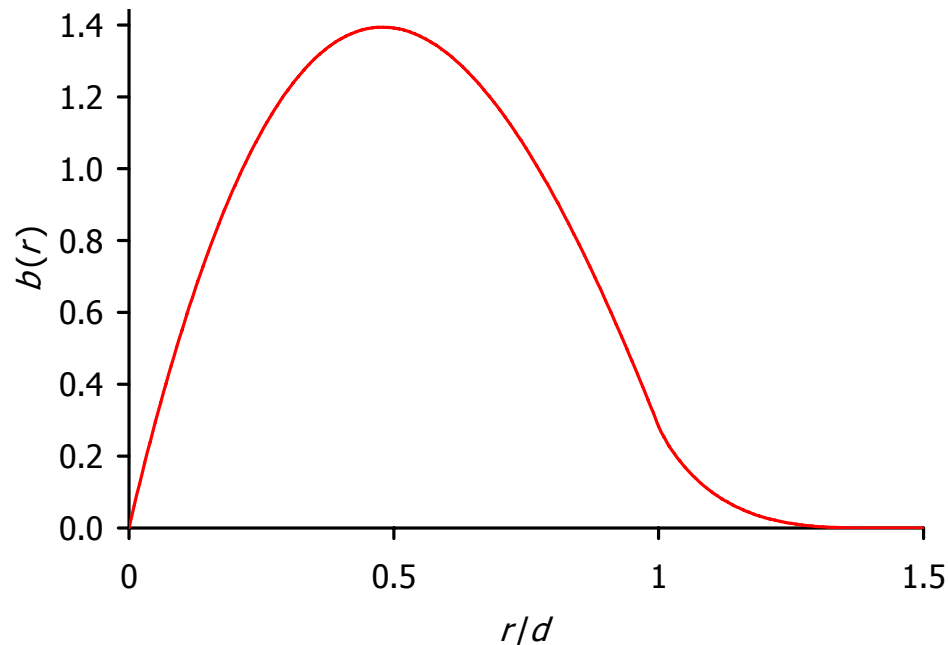
## Approach:

- **Simulation** of random and clustered processes of:
  - **Points**, to represent fibre centres
  - **Infinite lines** to generate free-fibre-lengths
  - **Finite lines**, to represent fibres and generate mass maps.

# Fibre centre statistics: Poisson points

Probability density of distance,  $r$ , between fibre centres occurring in a square of side,  $d$ , is known for the **random** (Poisson) case (Ghosh, 1951):

$$b(r, d) = \begin{cases} \frac{2r}{d^4} (\pi d^2 - 4dr + r^2) & \text{for } 0 \leq r \leq d \\ \frac{2r}{d^4} \left( 4d\sqrt{r^2 - d^2} - r^2 - d^2(2 + \pi - 4\sin^{-1}(d/r)) \right) & \text{for } d < r \leq \sqrt{2}d \end{cases}$$



# Fibre centre statistics: Non-Poisson points

For flocculated and dispersed cases, we use simulation to deposit 50,000 fibre centres in a unit square.

**Clustered:** Compound Poisson process

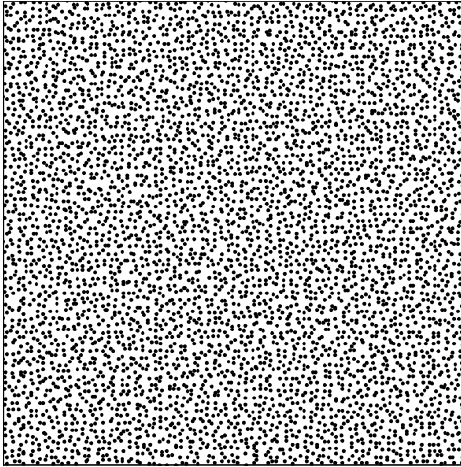
- Cluster centres distributed as a 2D Poisson process
- Number of centres per cluster has a Poisson distribution
- Inputs:
  - mean number of centres per cluster
  - cluster diameter

**Dispersed:** Small random perturbation from square lattice

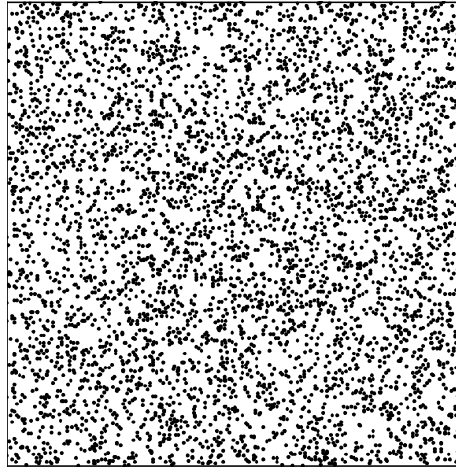


# Fibre centres: Distributions of points and fibres

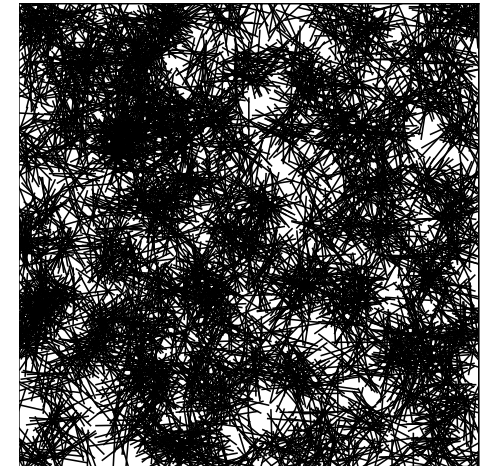
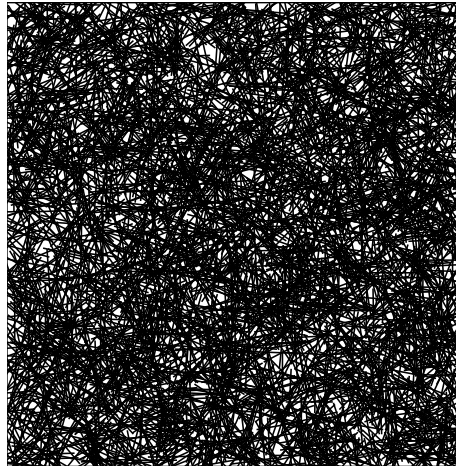
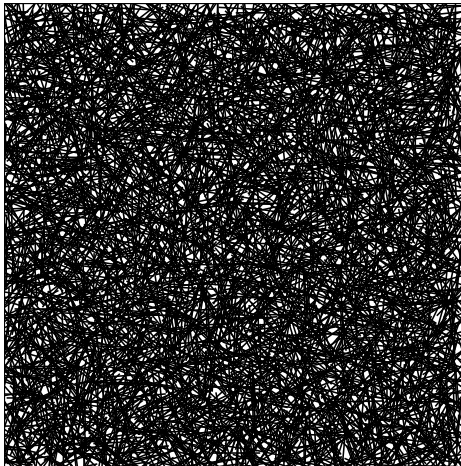
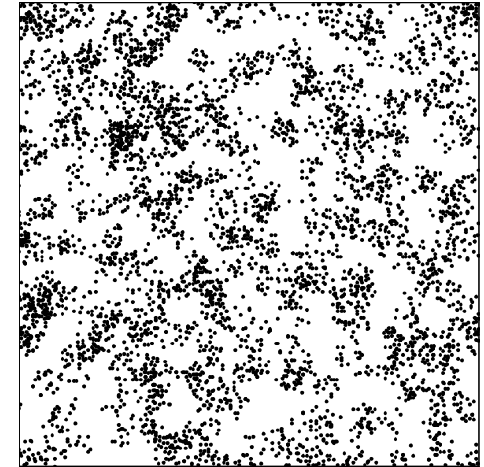
Dispersed



Random

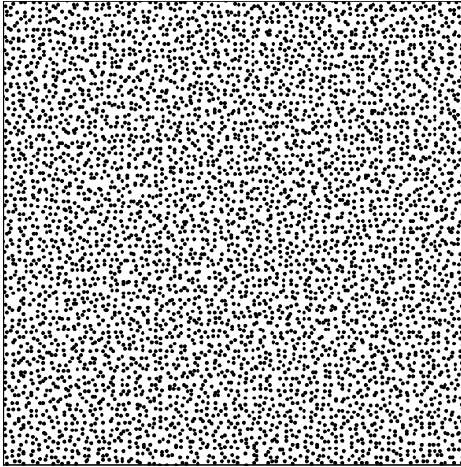


Clustered

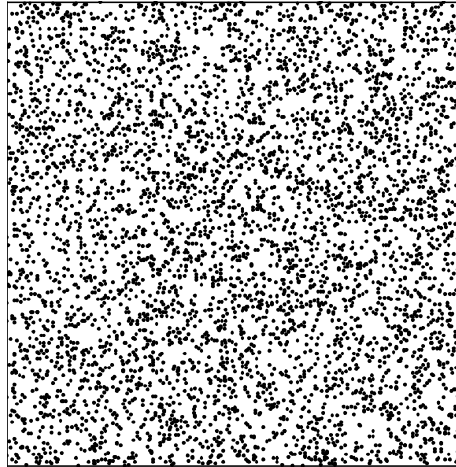


# Fibre centres: Distributions of points and their separations

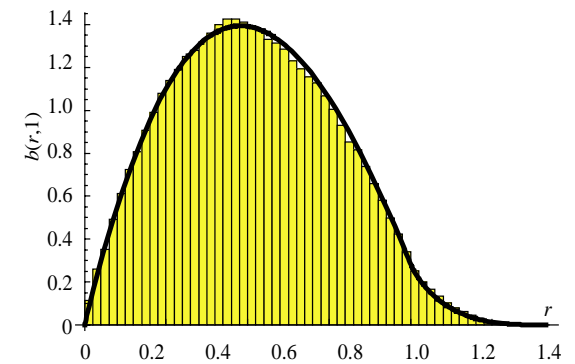
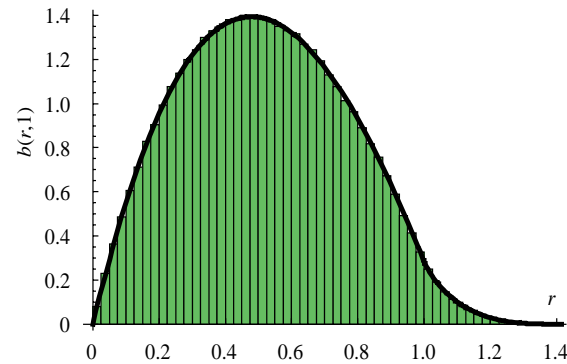
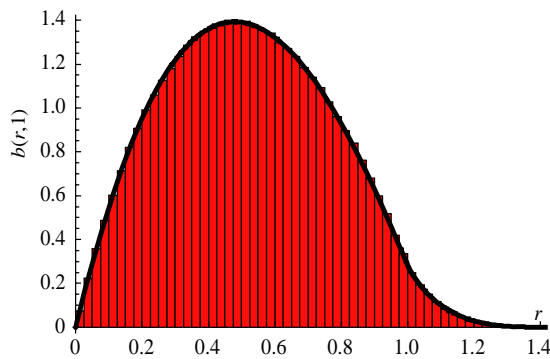
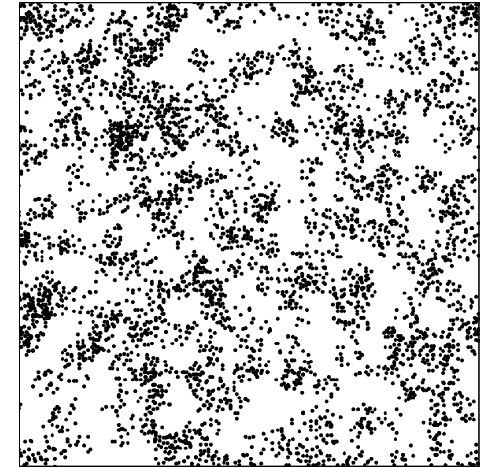
Dispersed



Random



Clustered

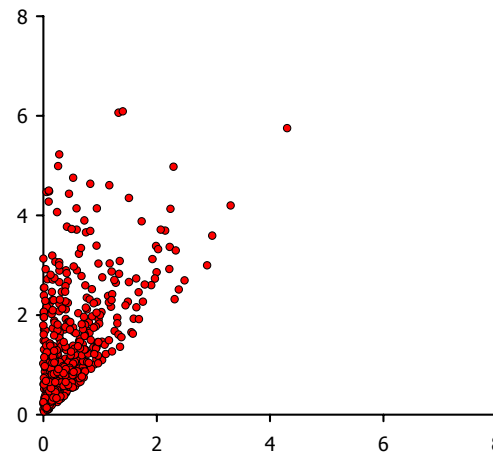
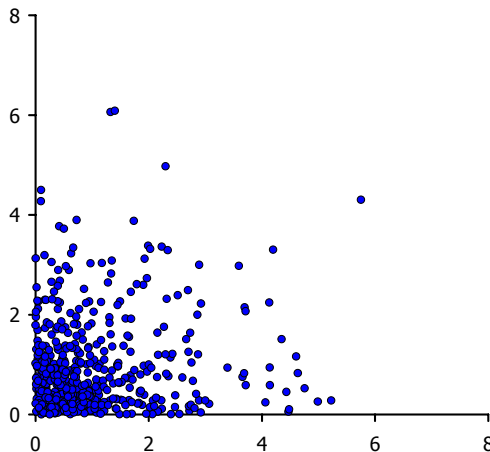
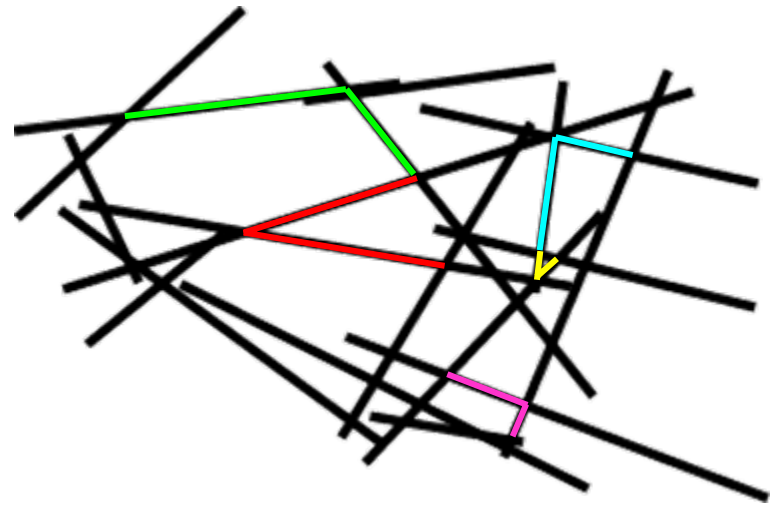


# Random networks: Correlation of adjacent free-fibre-lengths?

- Conjecture, FRS13: adjacent free-fibre-lengths are correlated, such that polygons tend to be 'roundish'.
- Free-fibre-length distribution is exponential. For unit mean

$$f(x) = e^{-x} \text{ and } \sigma^2(x) = \bar{x} = 1$$

- To reveal correlation between pairs of free-fibre-lengths,  $\{x_i, y_i\}$ , we must order the pairs such that  $x_i \leq y_i$



# Correlation of adjacent free-fibre-lengths: Independent pairs

- First we must obtain the correlation for **independent** sorted pairs
- Start with a pair of numbers,  $\{x_i, y_i\}$ , drawn from the exponential distribution
- Convert each pair into an ordered pair  $(x_i, y_i)$  such that  $x_i \leq y_i$
- Two distributions arise, one for  $x$  and one for  $y$ :

$$g_{\leq}(x) = \frac{1}{2} e^{-2x} \quad \text{with } \bar{x} = \frac{1}{2} \text{ and } \sigma^2(x) = \frac{1}{4}$$

$$g_{\geq}(y) = 2 e^{-2y} (1 - e^{-y}) \quad \text{with } \bar{y} = \frac{3}{2} \text{ and } \sigma^2(y) = \frac{5}{4}$$

# Correlation of adjacent free-fibre-lengths: Independent pairs

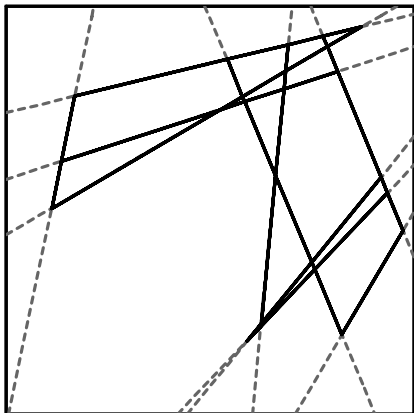
- The correlation  $\rho$  is given by

$$\begin{aligned}\rho &= \frac{\overline{xy} - \bar{x}\bar{y}}{\sigma(x)\sigma(y)} \\ &= \frac{1}{\sqrt{5}}\end{aligned}$$

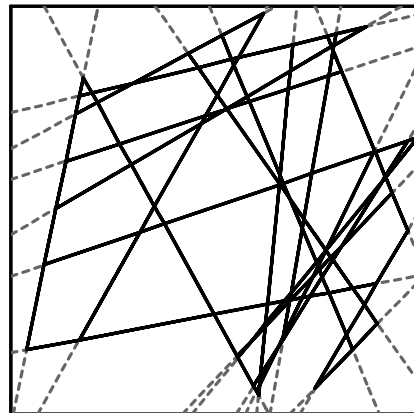
- So for **independent** pairs  $\rho \approx 0.447$ . We use simulation to obtain  $\rho$  for stochastic line networks. If the simulation yields  $\rho > 0.447$ , then we have intrinsic correlation in our networks.

# Correlation of adjacent free-fibre-lengths: Random networks

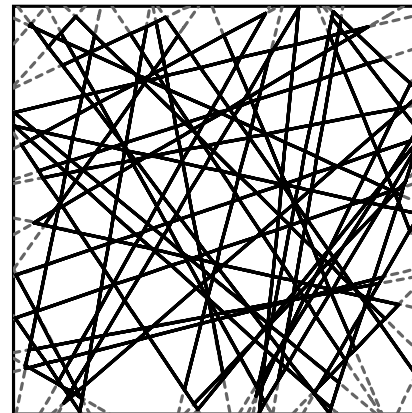
- Simulation solves equations of random lines drawn in a unit square to compute coordinates of crossings and polygon sides (free-fibre-lengths)
- Adjacent polygon sides are paired
- Polygon sides crossing boundaries of unit square are discarded
- Correlation of sorted and paired polygon side lengths computed
- Correlation tracked as number of lines increased



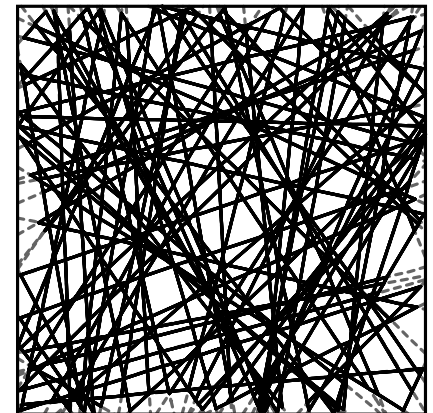
10 lines



20 lines

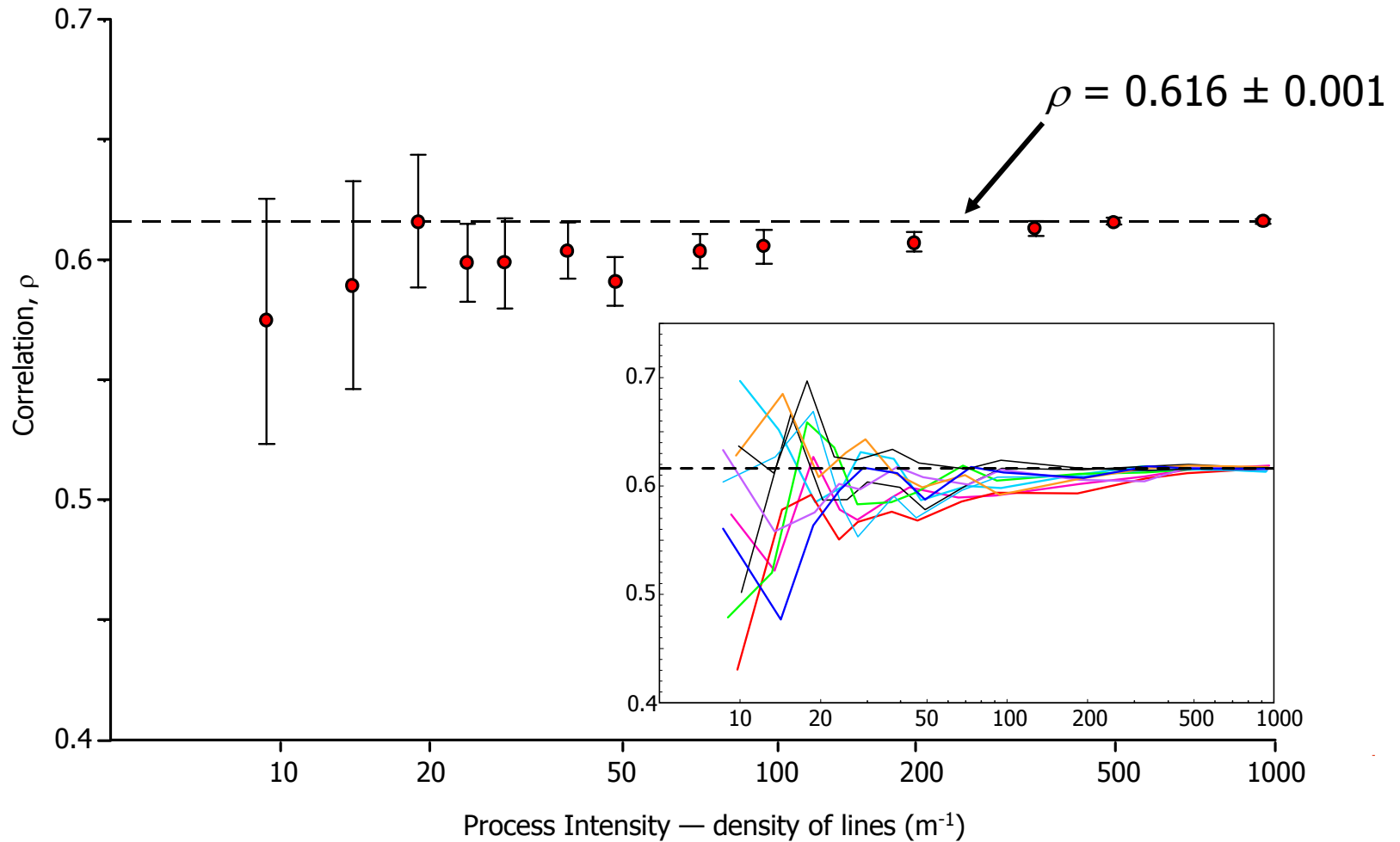


50 lines



100 lines

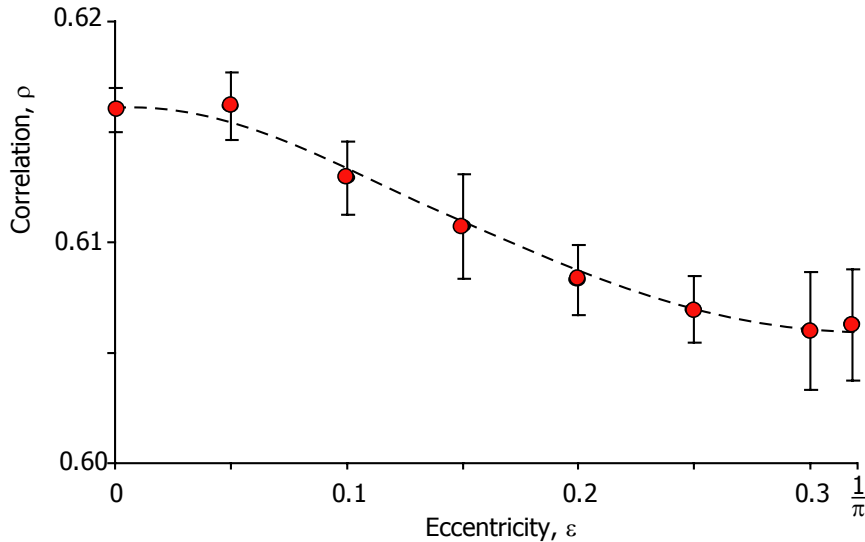
# Correlation of adjacent free-fibre-lengths: Random networks



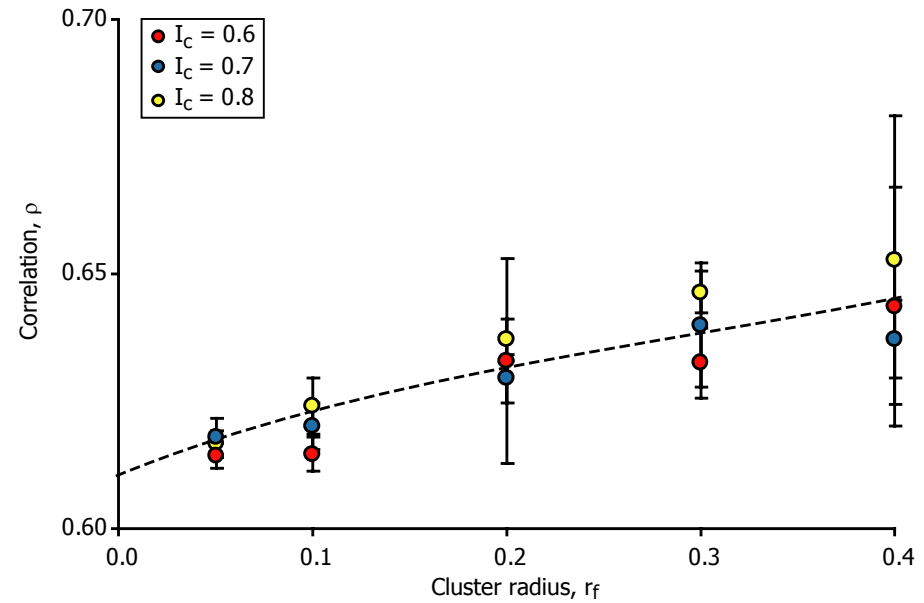
# Correlation of adjacent free-fibre-lengths: Oriented & clustered

- Very small influence of fibre orientation using 1-parameter cosine distribution

$$f(\theta) = \frac{1}{\pi} - \varepsilon \cos(2\theta) \quad 0 \leq \theta \leq \frac{1}{\pi}$$



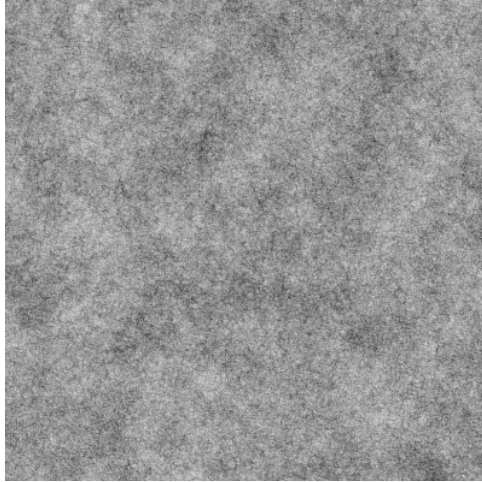
- Very small influence of clustering. Lines pass through clusters of points with varying cluster radius,  $r_f$  and cluster intensity,  $I_c$



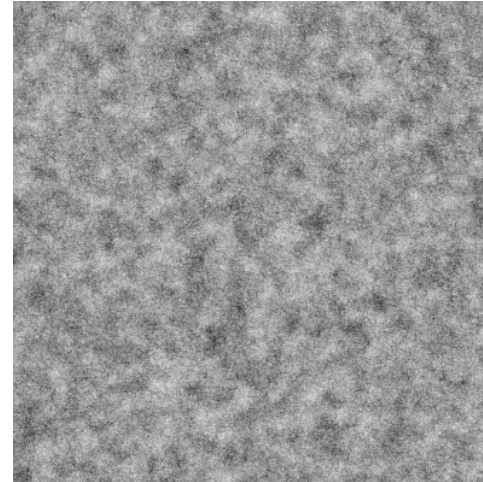


# Formation simulator

Cloudy



Grainy



- Formation easily visually classified as 'cloudy' (large flocs, low floc grammage) or 'grainy' (small flocs, high floc grammage)
- Experimentally difficult to vary scale and intensity of flocculation independently
- Simulator generates 4 cm × 4 cm grammage maps where scale and intensity of clusters (flocs) can be varied independently

# Formation simulator

## Inputs:

- Grammage,  $\bar{\beta}$
- Fibre properties:
  - Length,  $\lambda$
  - Coarseness,  $\delta$
  - Width,  $\omega$
- Mean floc radius,  $r_f$
- Floc intensity,  $0 \leq I \leq 1$
- Expected number of fibres per cluster,  $\overline{n_c}$

# Formation simulator for flocculated networks

## Inputs:

- Grammage,  $\bar{\beta}$
- Fibre properties:
  - Length,  $\lambda$
  - Coarseness,  $\delta$
  - Width,  $\omega$
- Mean floc radius,  $r_f$
- Floc intensity,  $0 \leq I \leq 1$
- Expected number of fibres per cluster,  $\overline{n_c}$

## Simulation:

- Number of fibres per cluster,  $n_c$  is a Poisson variable with mean,  $\overline{n_c}$
- Mean grammage,  $G$ , of each cluster is assumed constant (*cf.* Farnood *et al.* 1995)

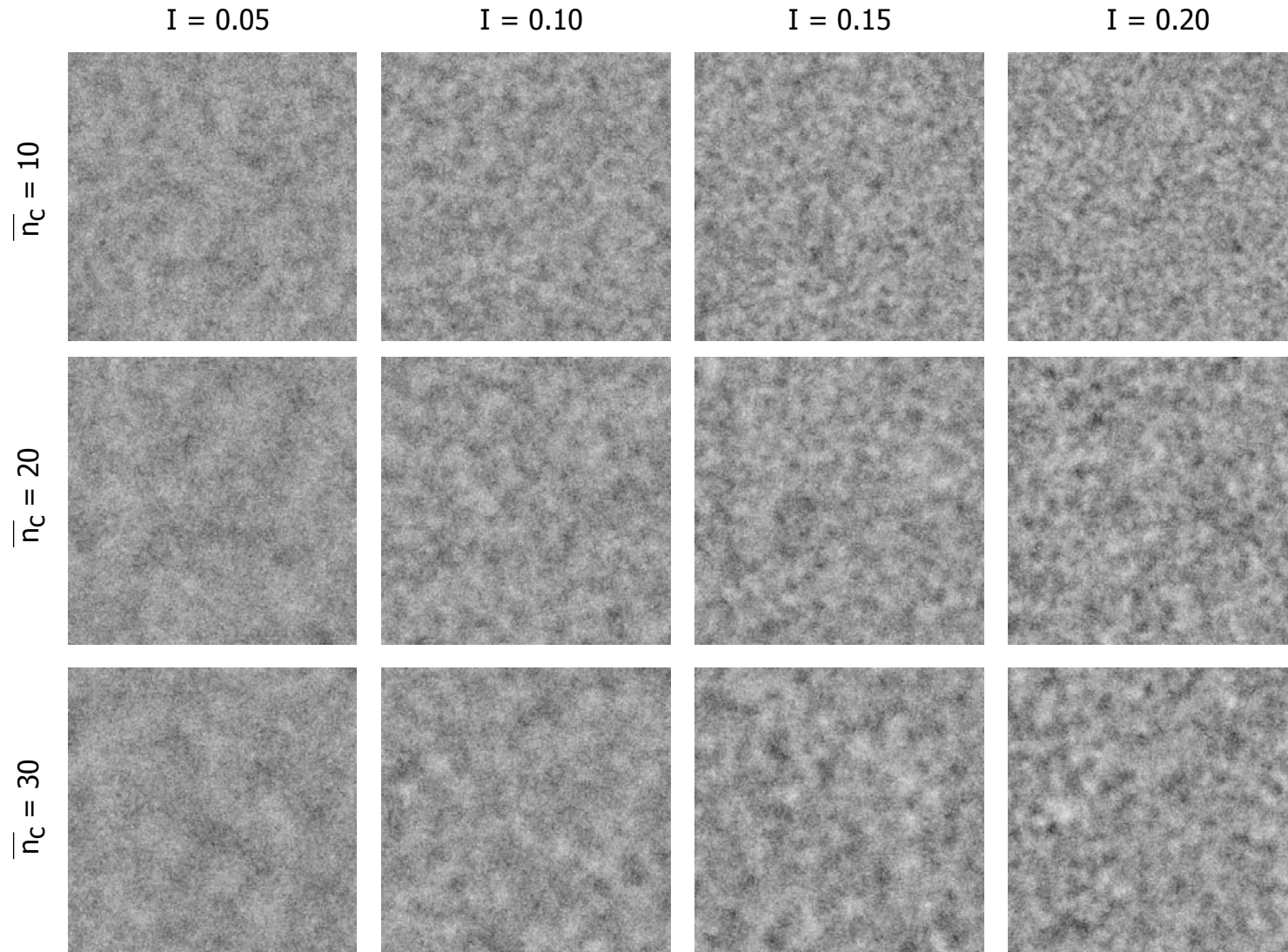
$$G = I \beta_{\text{fib}} = \frac{I \delta}{\omega}$$

- Radius of each cluster is

$$r = \sqrt{\frac{n_c \lambda \omega}{\pi I}}$$

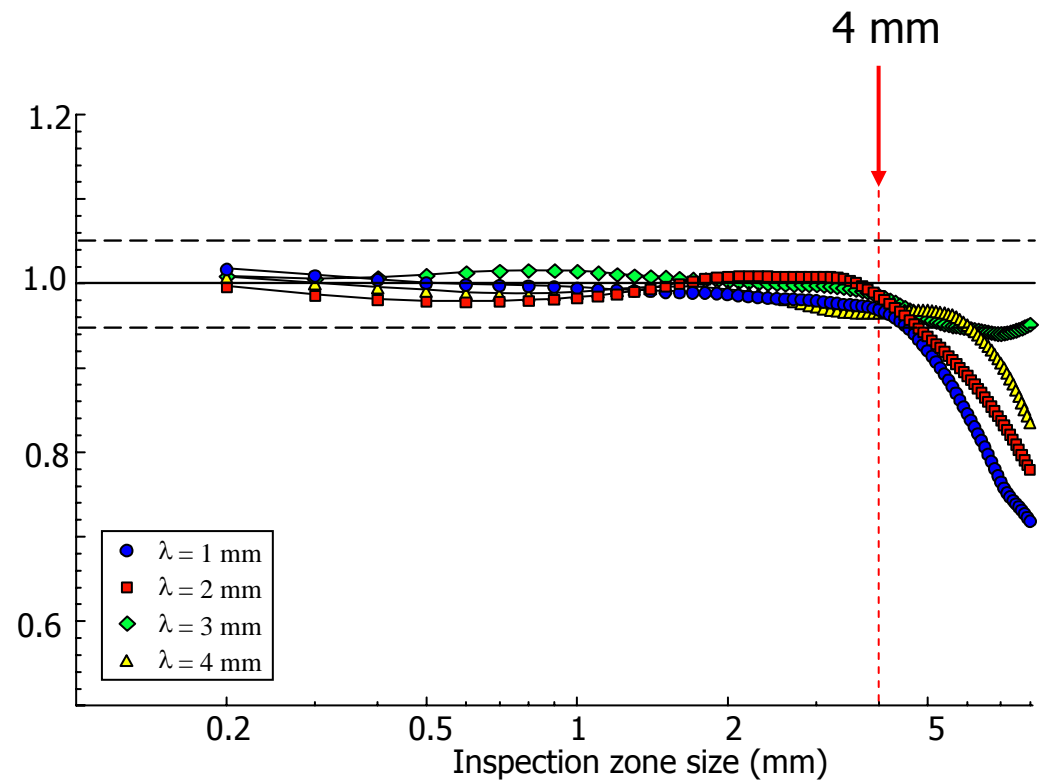
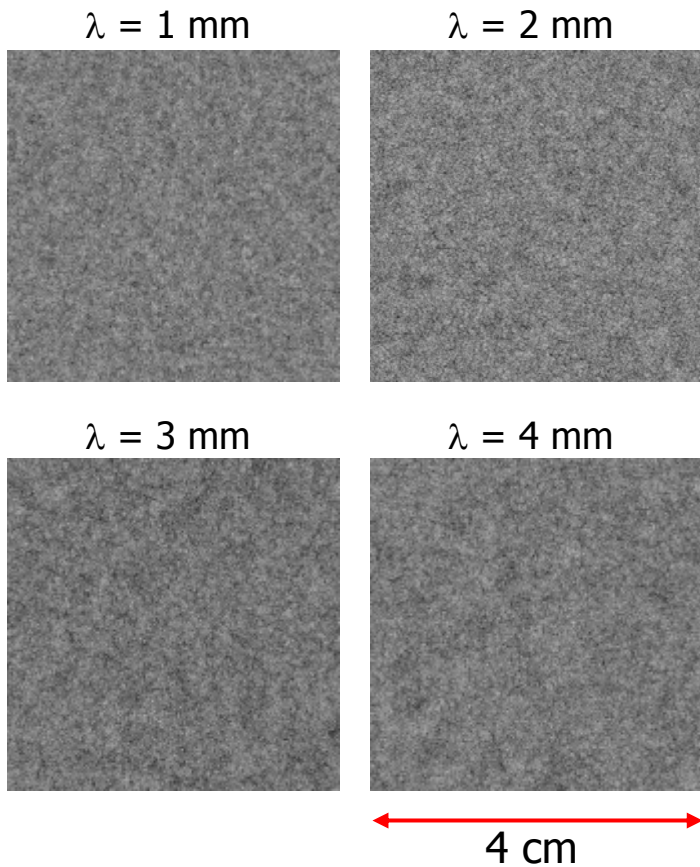
- $n_c$  fibre centres deposited within circles of radius  $r$ .
- For each fibre, contribution to mass of each pixel calculated.

# Formation simulator: Example outputs



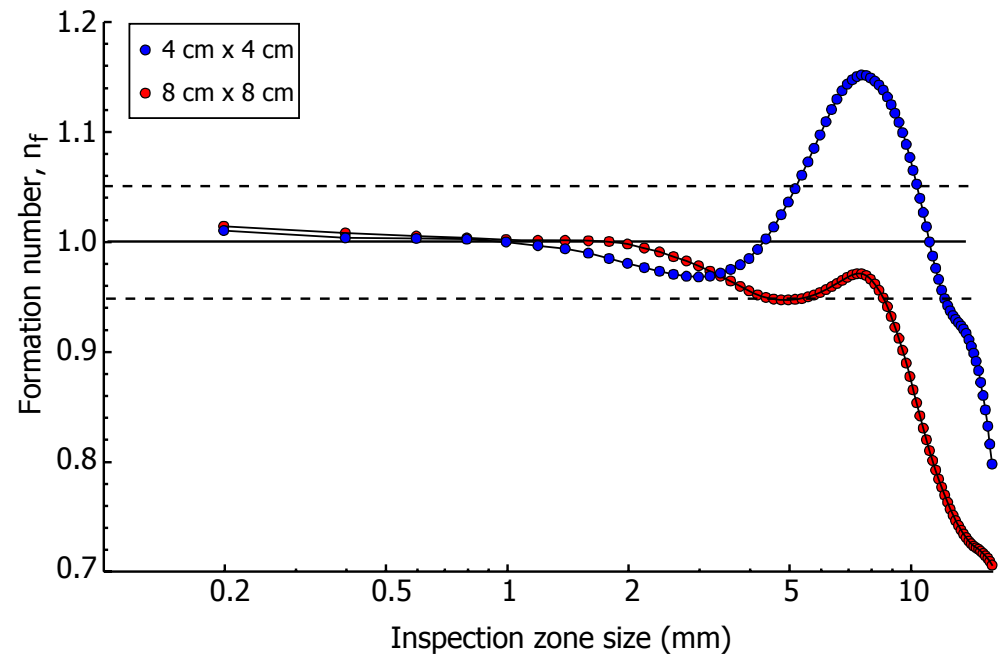
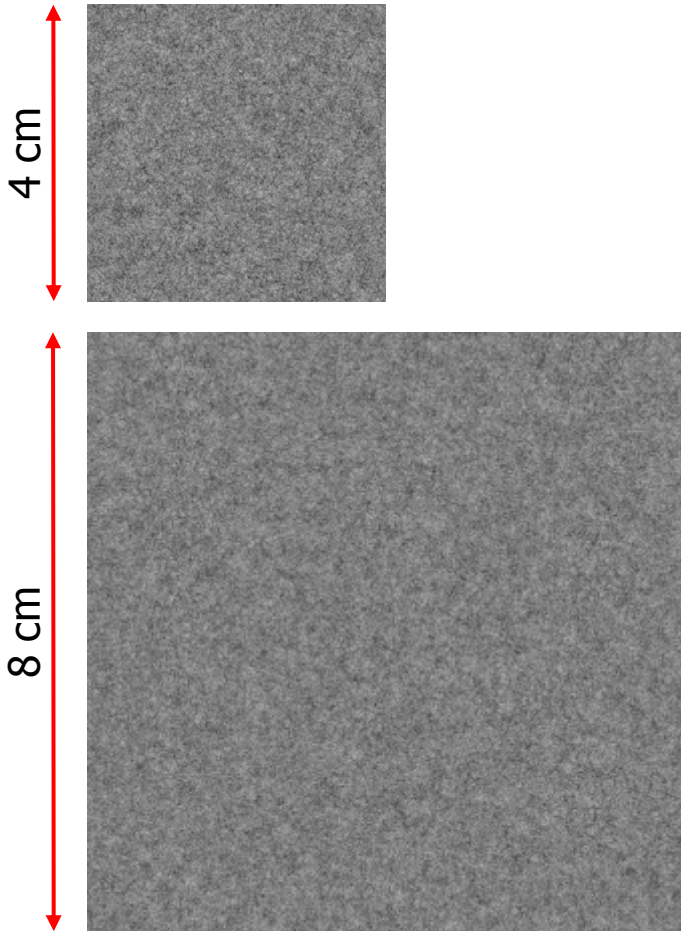
# Formation simulator: Random case, fibre length effect

- Simulator generated Poisson random networks ( $n_c = 1$ ) for different fibre lengths and computed  $n_f$



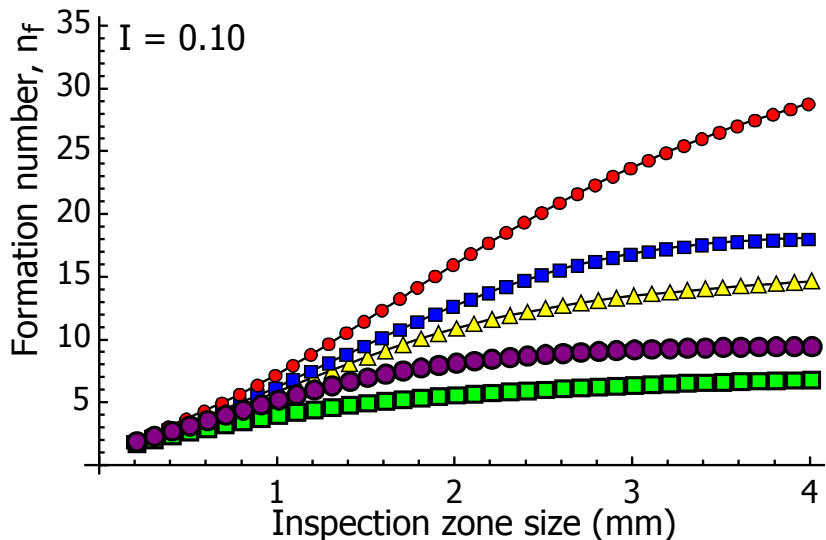
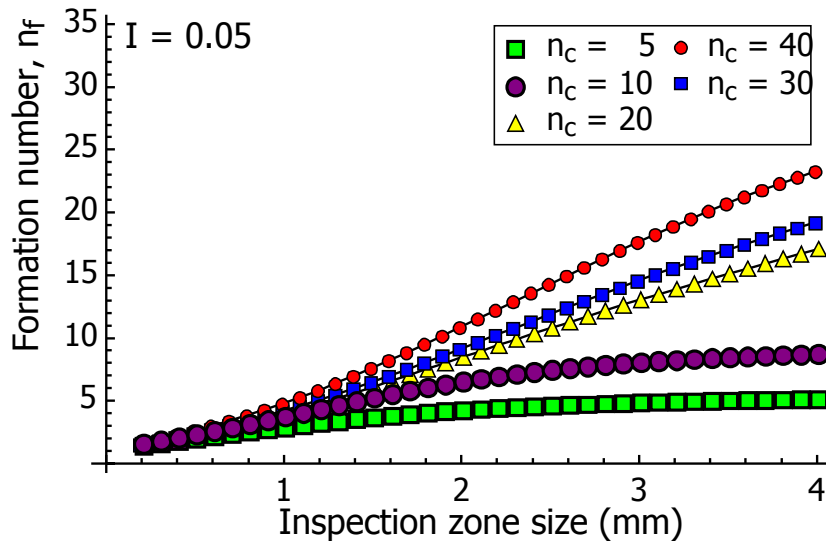
# Formation simulator: Random case, scale sensitivity

- Different areas



- Sufficient data reliably to compute variance only for scales up to about 10% of imaging area

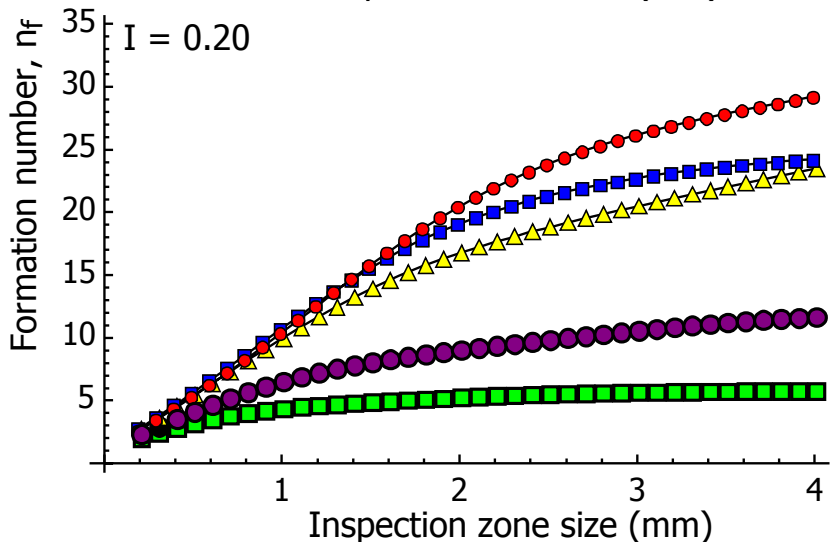
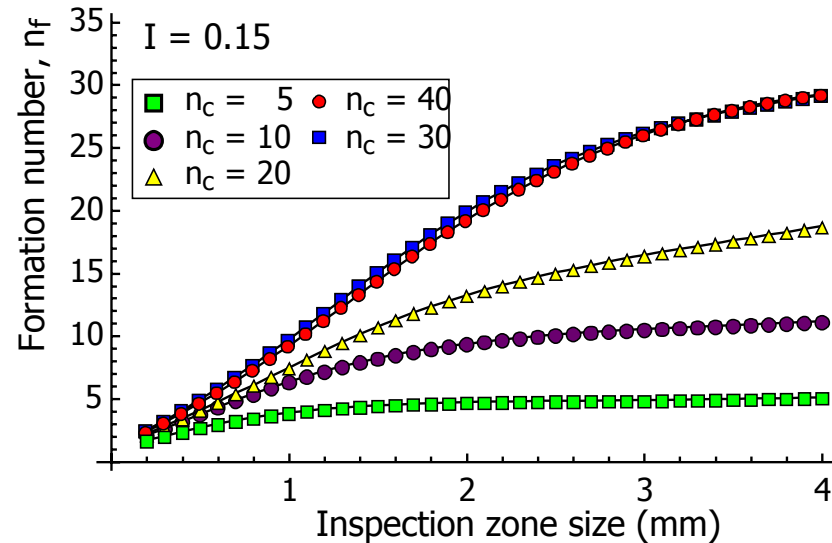
# Formation simulator: Scale dependence of $n_f$



- $n_f$  increases with
  - inspection zone size
  - flocculation intensity,  $I$
  - number of fibres per cluster,  $n_c$
- Dependence on zone size nonlinear
  - literature data sparse
  - linear regression gives  $r^2 > 0.9$  on these data

$$\lambda = 1 \text{ mm}; \omega = 20 \text{ } \mu\text{m}; \delta = 2 \times 10^{-7} \text{ kg m}^{-1}; \beta = 60 \text{ g m}^{-2}$$

# Formation simulator: Scale dependence of $n_f$



- $n_f$  increases with
  - inspection zone size
  - flocculation intensity,  $I$
  - number of fibres per cluster,  $n_c$
- Dependence on zone size nonlinear
  - literature data sparse
  - linear regression gives  $r^2 > 0.9$  on these data
- No simple scaling law found, but
  - when  $n_c$  and  $I$  are both large,  $n_f$  exhibits similar dependence on zone size at small scales

$$\lambda = 1 \text{ mm}; \omega = 20 \text{ } \mu\text{m}; \delta = 2 \times 10^{-7} \text{ kg m}^{-1}; \beta = 60 \text{ g m}^{-2}$$



# Analytic approximation of formation number $n_f$

For a Poisson structure of sparse disks with grammage,  $G$ , and uniform diameter,  $D$ , the variance of local grammage for square zones of side  $x$  is

$$\sigma^2_{\text{disks},x}(\tilde{\beta}) = \bar{\beta} G \int_0^{\sqrt{2x}} \alpha_{\text{disks}}(D,r) b(r,x) dr$$

where

$$\alpha_{\text{disks}}(D,r) = \frac{2}{\pi D} \left( D \cos^{-1}(r/D) - r \sqrt{1 - (r/D)^2} \right)$$

Also

$$n_f(x) = \frac{\sigma^2_{\text{disks},x}(\tilde{\beta})}{\sigma^2_{\text{fibres},x}(\tilde{\beta})}$$

Where  $\sigma^2_{\text{fibres},x}(\tilde{\beta})$  is the variance of local grammage of a random fibre network and is known analytically:

$$\sigma^2_{\text{fibres},x}(\tilde{\beta}) = \bar{\beta} \beta_{\text{fib}} \int_0^{\sqrt{2x}} \alpha(r,\omega,\lambda) b(r,x) dr$$

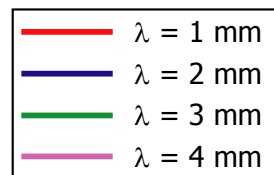
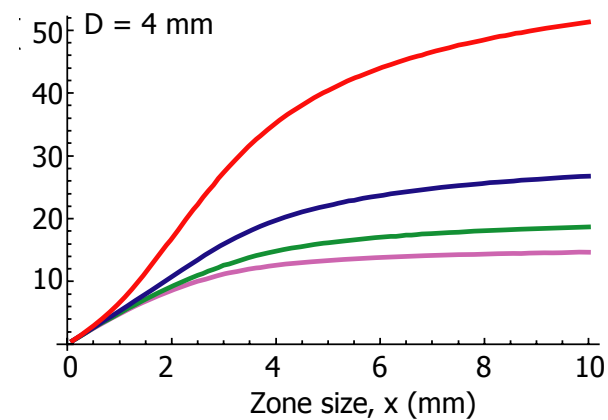
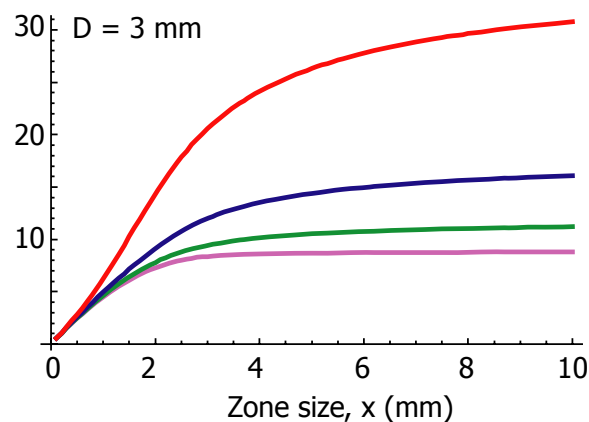
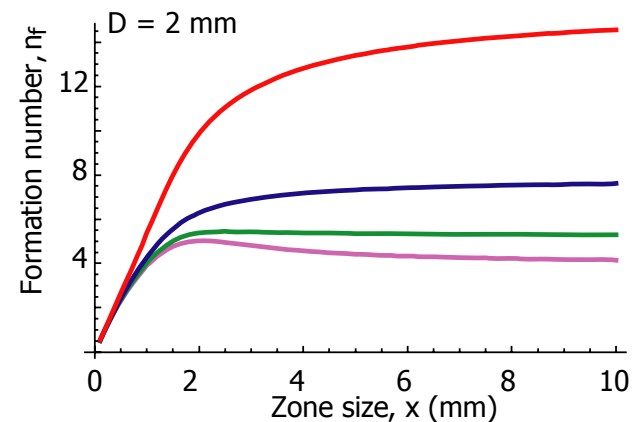
# Analytic approximation of formation number $n_f$

$$\begin{aligned} n_f(x) &= \frac{\sigma_{\text{disks},x}^2(\tilde{\beta})}{\sigma_{\text{fibres},x}^2(\tilde{\beta})} \\ &= I \frac{\int_0^{\sqrt{2x}} \alpha_{\text{disks}}(D,r) b(r,x) dr}{\int_0^{\sqrt{2x}} \alpha(r,\omega,\lambda) b(r,x) dr} \end{aligned}$$

# Analytic approximation of formation number $n_f$

$$n_f(x) = \frac{\sigma_{\text{disks},x}^2(\tilde{\beta})}{\sigma_{\text{fibres},x}^2(\tilde{\beta})}$$

$$= I \frac{\int_0^{\sqrt{2x}} \alpha_{\text{disks}}(D,r) b(r,x) dr}{\int_0^{\sqrt{2x}} \alpha(r,\omega,\lambda) b(r,x) dr}$$



# Analytic approximation of formation number asymptotics

For large inspection zones

$$\lim_{x \rightarrow \infty} n_f(x) = n_c$$

At small inspection zones, the initial slope is

$$\begin{aligned} \lim_{x \rightarrow 0} \frac{dn_f}{dx} &= I \frac{\alpha_{\text{disks}}(D, r) b(r, x)}{\alpha(r, \omega, \lambda) b(r, x)} \Big|_{r \rightarrow x \rightarrow 0} \\ &= I \end{aligned}$$

So the initial slope of a plot of formation number against inspection zone size depends on the intensity of flocculation and the asymptotic value depends on the expected number of fibres per floc.

# Conclusions

- **Pores ARE roundish:** Natural stochastic clusters, which generate local free-fibre-length correlations, overwhelm any effects on pore shape of fibre orientation or flocculation.
- **Trapped polygon void sizes:** Local free-fibre-length correlations force coefficient of variation for in-plane pore sizes to be insensitive to flocculation and orientation.
- **Trapped formation:** Variance ratio to random ( $n_f$ ) asymptotic to number of fibres per cluster ( $n_c$ ) and convergence rate is intensity of flocculation ( $I$ ).