# X-ray Scattering for CB and CB Nanocomposites

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## X-ray Scattering for CB and CB Nanocomposites

## -Overview of X-ray Scattering

# -Nanocomposite Dispersion and Distribution Carbon Black in Elastomers and other Polymers

X-ray Scattering

2-D Narrow size range Pinhole XRD/SAXS/(USAXS) 1-D 4+-orders of size Bonse-Hart Camera USAXS











Kammler HK, Beaucage G, Kohls DJ, Agashe N. Ilavsky J *J Appl. Phys.* 97(2005) (Article 054309).

















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### **Descriptors**

Aggregate size, R Primary particle size,  $d_p$ Degree of aggregation, z Short-circuit path length, p Connective path length, s Fractal dimension,  $d_f$ Tortuosity,  $d_{min}$ Connectivity, c Branch fraction,  $\phi_{br}$ Branch number,  $n_{br}$ 



Beaucage, G. Determination of branch fraction and minimum dimension of mass-fractal aggregates. *Physical Review E – Statistical, Nonlinear, and Soft Matter Physics* 2004, 70, 031401-1–031401-10.

Ramachandran, R., Beaucage, G., Kulkarni, A. S., McFaddin, D., Merrick-Mack, J., & Galiatsatos, V. Branch content of metallocene polyethylene. *Macromolecules* 2009, 42, 4746–4750.

Rai, D., Beaucage, G., Jonah, E. O., Britton, D. T., Sukumaran, S., & Härting, M. Quantitative investigations of aggregate systems. *J. Chem. Phys.* 2012, 137, 044311–1–044311-6.



Crumple paper: c fixed at 2;  $d_{\min}$  increases from 1 to  $\leq 1.5$ Rip arms off an aggregate:  $d_{\min}$  fixed; c decreases towards 17 Algorithm:

Input z and a sticking probability Randomly grow aggregates Compute the scattering parameters, p, R, n<sub>Br</sub> etc. Iterate by varying sticking probability until computed matches experimental

A. Mulderig, G. Beaucage, K. Vogtt, H. Jiang and V. Kuppa, Quantification of branching in fumed silica, J. Aerosol Sci. 2017, 109, 28–37.

		1	1				
Sample	А	В	С	D	E	F	
R	19	19	26	24	37	91	
d <sub>p</sub> (nm)	36	21	17	14	12	10	
Cp	1.54	1.53	1.51	1.50	1.52	1.51	
Z	157	177	483	567	819	6430	
р	36	35	46	38	37	91	
S	66	75	195	265	819	6430	
d <sub>min</sub>	1.21	1.20	1.17	1.14	1.00	1.00	
с	1.41	1.46	1.62	1.75	1.86	1.95	
d <sub>f</sub>	1.70	1.75	1.90	1.98	1.86	1.95	
$n_{br} (f = 3)$	19	23	68	106	409	3215	
φ <sub>br</sub> , %	77	80	91	93	95	99	

#### Experimental

#### Simulation

Z	157	177	483	567	819	6430
Sticking probability	0.40	0.37	0.11	0.23	0.82	1.00
R	18	19	26	29	40	93
р	31	33	45	42	40	93
S	68	75	191	298	819	6430
d <sub>min</sub>	1.2	1.2	1.17	1.11	1	1
с	1.47	1.48	1.62	1.7	1.82	1.93
d <sub>f</sub>	1.76	1.75	1.9	1.89	1.82	1.93
n <sub>br</sub>	21	26	68	121	409	3215
φ <sub>br</sub> , %	80%	81%	91%	93%	95%	99%

A. Mulderig, G. Beaucage, K. Vogtt, H. Jiang and V. Kuppa, Quantification of branching in fumed silica, J. Aerosol Sci. 2017, 109, 28–37.

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A. Mulderig, G. Beaucage, K. Vogtt, H. Jiang and V. Kuppa, Quantification of branching in fumed silica, J. Aerosol Sci. 2017, 109, 28–37.



Mulderig, A.; Narayanan, V.; McGlasson, A.; Rackaitis, M.; Ilavsky, J. Impact of an Emergent Hierarchical Filler Network on Nanocomposite Dynamics. Macromolecules 2018, 51, 7893–7904.Copyright 2018 American Chemical Society



A. Mulderig, G. Beaucage, K. Vogtt, H. Jiang and V. Kuppa, Quantification of branching in fumed silica, J. Aerosol Sci. 2017, 109, 28–37.

### At concentrations above c\*

When concentration is greater than Mass of CB aggregate By Volume of aggregate

 $c^* = M/V$  of CB aggregate

Then the aggregate can't be seen

It is screened above a screening length,  $\xi$ 

Below  $\xi$  we see the structure

Above  $\xi$  we see uniform contrast



### The RPA Equation

$$\frac{\phi}{I(q,\phi)} = \frac{\phi_0}{I(q,\phi_0)} + \phi \upsilon$$

Reduced high-q intensity remains unchanged, Reduced low-q intensity is diminished with concentration increase

The q => 0 intercept is a measure of the strength of binary interactions (the Second Order Virial Coefficient)

This assumes no specific interactions. A mean field is considered.





#### Mean Field System



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**Specific Interaction System** 



McGlasson, A.; Rishi, K.; Beaucage, G.; Chauby, M.; Kuppa, V.; Ilavsky, J.; Rackaitis, M. Quantification of Dispersion for Weakly and Strongly Correlated Nanofillers in Polymer Nanocomposites. *Macromolecules* 2020

For specific interactions we consider the structure factor S(q) $I(q) = F^2(q) S(q)$ 

 $S(q) = I(q)/\phi \; (\phi_0/I_0(q))$ 







McGlasson, A.; Rishi, K.; Beaucage, G.; Chauby, M.; Kuppa, V.; Ilavsky, J.; Rackaitis, M. Quantification of Dispersion for Weakly and Strongly Correlated Nanofillers in Polymer Nanocomposites. Macromolecules 2020

$$S_{\text{RPA}}(q) = \frac{1}{1 + \phi v(I_0(q, \phi_0)/\phi_0)} \qquad S_{\text{PBG}}(q, \xi) = \int_0^\infty P(\xi) \times \left[\frac{1}{1 + p\theta(q, \xi)}\right] d\xi \qquad S_{\text{PBG}}(q, \xi) = \int_0^\infty P(\xi) \times \left[\frac{1}{1 + p\theta(q, \xi)}\right] d\xi$$
$$q_{\text{mod}} = q \times \exp\left\{\delta \frac{(q - q_{\text{peak}})}{q}\right\}$$

8 Data from Anderson, B. J. and Zukoski, C.F. Rheology and Microstructure of Polymer Nanocomposite Melts: Variation of Polymer Segment- Surface Interaction. Langmuir 2010, 26, 8709-8720.

9 Data from Jin, Y.; Beaucage, G.; Vogtt, K.; Jiang, H.; Kuppa, V.; Kim, J.; Ilavsky, J.; Rackaitis, M.; Mulderig, A.; Rishi, K.; et al. A Pseudo- Thermodynamic Description of Dispersion for Nanocomposites. Polymer 2017, 129, 32-43.

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## TiCl<sub>4</sub> Diffusion Flame Nucleation and Growth

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#### Titania Diffusion Flame from TiCl<sub>4</sub>

Beaucage G, Agashe N, Kohls D, Londono D, Diemer B

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Silica Diffusion Flame Axial Particle Growth follows Classic Diffusion Limited Surface Growth, d ~ t<sup>1/2</sup>



## Nanoparticle Dispersion and Distribution in Nanocomposites

# Kinetically mixed systems

## Mean field



# Specific interactions



# Clustering due to immiscibility

## Current state-of-the-art academic nanocomposite



100

Kumar, S.K., Jouault, N., Benicewicz, B. and Neely, T., 2013. Nanocomposites with polymer grafted nanoparticles. *Macromolecules*, *46*(9), pp.3199-3214. Kumar, S.K., Benicewicz, B.C., Vaia, R.A. and Winey, K.I., 2017. 50th anniversary perspective: are polymer nanocomposites practical for applications?. *Macromolecules*, *50*(3), pp.714-731.

Asai, M., Zhao, D. and Kumar, S.K., 2017. Role of grafting mechanism on the polymer coverage and self-assembly of hairy nanoparticles. *ACS Nano*, *11*(7), pp.7028-7035.

## The original nanocomposite

Polydisperse aggregates

Melt state processed

Kinetically mixed **immiscible** 





Objective: Complex hierarchical structure Tear resistance Static charge dissipation

Why/how do added nanoparticles impact structures 100-1,000 times larger?



Song, L; Wang, Z; Tang, X.; Chen, L.; Chen, P.; Yuan, Q.; Li, L. Visualizing the Toughening Mechanism of Nanofiller with 3D X-ray Nano-CT: Stress-Induced Phase Separation of Silica Nanofiller and Silicone Polymer Double Networks Macromolecules 50 7249-7257 (2017). Greg Beaucage, University of Cincinnati gbeaucage@gmail.com


Rishi, K., Beaucage, G., Kuppa, V., Mulderig, A., Narayanan, V., McGlasson, A., Rackaitis, M. and Ilavsky, J., 2018. Impact of an emergent hierarchical filler network on nanocomposite dynamics. *Macromolecules*, *51*(20), pp.7893-7904. Mulderig, A., Beaucage, G., Vogtt, K., Jiang, H. and Kuppa, V., 2017. Quantification of branching in fumed silica. *Journal of Aerosol Science*, *109*, pp.28-37.

# **Dispersion/Distribution can have a size-scale dependence**

# Macroscopically disperse systems might be clustered on the nanoscale





# Clustering due to immiscibility

## How can dispersion/distribution be quantified on the nano-scale?

# Kinetically mixed systems

# Mean field



# Specific interactions



# Clustering due to immiscibility



Pedersen, J. S.; Sommer, C. Temperature Dependence of the Virial Coefficients and the Chi Parameter in Semi-Dilute Solutions of PEG. In *Scattering Methods and the Properties of Polymer Materials*; Springer Berlin Heidelberg: Berlin, Heidelberg, 2005; pp 70–78.

Vogtt, K.; Beaucage, G.; Weaver, M.; Jiang, H. Thermodynamic Stability of Worm-like Micelle Solutions. Soft Matter 2017, 13 (36), 6068-6078.

Jin, Y.; Beaucage, G.; Vogtt, K.; Jiang, H.; Kuppa, V.; Kim, J.; Ilavsky, J.; Rackaitis, M.; Mulderig, A.; Rishi, K.; Narayanan, V. A Pseudo-Thermodynamic Description of Dispersion for Nanocomposites. *Polymer (Guildf)*. 2017, *129*, 32–43.

#### Immiscible: Miscible: **Carbon Black and Silica in Elastomer Organic Pigment with Triton X100** ~50nm y = mx + bm = -0.0474 ± 0.00767 3 $b = 3.25 \pm 0.315$ $A_2 (cm^3 mol/g^2) (x10^{-8})$ = 0.950 A<sub>2</sub>, 10<sup>-9</sup> mole cm<sup>3</sup>/g<sup>2</sup> A<sub>2</sub> Values largest particle size for stable **CB** Data dispersion ~ 68nm 140ND (High cis) PI B45 (Low cis) repulsive regime SiO<sub>2</sub> Data 0 attractive regime □ 140ND (High cis) Δ PI B45 (Low cis) 0 20 60 80 40 100 10 20 30 50 40 $d_p$ (nm) d<sub>p</sub>, nm Thermally driven nano-dispersion

**Thermal Distribution versus Kinetic Distribution** 

### (Stokes drag coefficient)

### Mechanically driven nano-dispersion (Lever arm)

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Mulderig, A.; Beaucage, G.; Vogtt, K.; Jiang, H.; Jin, Y.; Clapp, L.; Henderson, D. C. Structural Emergence in Particle Dispersions. *Langmuir* **2017**, *33* (49), 14029–14037. Jin, Y.; Beaucage, G.; Vogtt, K.; Jiang, H.; Kuppa, V.; Kim, J.; Ilavsky, J.; Rackaitis, M.; Mulderig, A.; Rishi, K.; Narayanan, V. A Pseudo-Thermodynamic Description of Dispersion for Nanocomposites. *Polymer (Guildf)*. **2017**, *129*, 32–43.

# van der Waals Analogy

*a* reflects the attractive energy of interaction between aggregates.

$$\Pi = \frac{kT}{(V+b)} - \frac{a}{V^2}$$

*b* is the excluded volume *a* is the attractive enthalpic interaction leading to phase separation

$$\Pi = kT\left(\frac{1}{V} + \frac{B_2}{V^2} + \cdots\right)$$

# Thermally driven colloidal dispersion



Energy *«* Temperature

$$B_2(T) = b - \frac{a}{kT}$$

$$B_2(t) = \frac{A_2(t)M^2}{N_A} = b^* - \frac{a^*}{t}$$

#### Mechanically dispersed nano-fillers



#### Energy *∝* Mixing Time

$$B_2(t) = b^* - \frac{a^*}{\gamma}$$

 $A_2$  arrives from using mass concentration rather than molar or number concentration



Rishi, K., Beaucage, G., Kuppa, V., Mulderig, A., Narayanan, V., McGlasson, A., Rackaitis, M. and Ilavsky, J., 2018. Impact of an emergent hierarchical filler network on nanocomposite dynamics. *Macromolecules*, *51*(20), pp.7893-7904.

Trappe, V. and Weitz, D.A., 2000. Scaling of the viscoelasticity of weakly attractive particles. *Physical review letters*, 85(2), p.449.

Mulderig, A., Beaucage, G., Vogtt, K., Jiang, H. and Kuppa, V., 2017. Quantification of branching in fumed silica. Journal of Aerosol Science, 109, pp.28-37.



Hashimoto, T., Amino, N., Nishitsuji, S. and Takenaka, M., 2019. Hierarchically self-organized filler particles in polymers: cascade evolution of dissipative structures to ordered structures. Polymer Journal, 51(2), pp.109-130.

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Baeza, G.P., Genix, A.C., Degrandcourt, C., Petitjean, L., Gummel, J., Couty, M. and Oberdisse, J., **2013**. Multiscale filler structure in simplified industrial nanocomposite silica/SBR systems studied by SAXS and TEM. *Macromolecules*, *46*(1), pp.317-329.



Richards, J.J., Hipp, J.B., Riley, J.K., Wagner, N.J. and Butler, P.D., **2017**. Clustering and percolation in suspensions of carbon black. *Langmuir*, *33*(43), pp.12260-12266.



Filippone, G., Romeo, G. and Acierno, D., **2010**. Viscoelasticity and structure of polystyrene/fumed silica nanocomposites: filler network and hydrodynamic contributions. *Langmuir*, *26*(4), pp.2714-2720. Filippone, G. and Salzano de Luna, M., **2012**. A unifying approach for the linear viscoelasticity of polymer nanocomposites.

Macromolecules, 45(21), pp.8853-8860.

### van der Waals model for incompatible polymer nanocomposites

*a* reflects the attractive energy of interaction between aggregates.

$$\Pi = \frac{kT}{(V+b)} - \frac{a}{V^2}$$

*b* is the excluded volume *a* is the attractive enthalpic interaction leading to phase separation

$$\Pi = kT\left(\frac{1}{V} + \frac{B_2}{V^2} + \cdots\right)$$



Rishi, K.; Narayanan, V.; Beaucage, G.; McGlasson, A.; Kuppa, V.; Ilavsky, J.; Rackaitis, M. A Thermal Model to Describe Kinetic Dispersion in Rubber Nanocomposites: The Effect of Mixing Time on Dispersion. Polymer (Guildf). **2019**, 175, 272–282.





Rishi, K.; Narayanan, V.; Beaucage, G.; McGlasson, A.; Kuppa, V.; Ilavsky, J.; Rackaitis, M. A Thermal Model to Describe Kinetic Dispersion in Rubber Nanocomposites: The Effect of Mixing Time on Dispersion. Polymer (Guildf). 2019, 175, 272–282.

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Wetting time depends on <u>viscosity</u> and <u>primary particle size</u>

x-intercept reflects "wetting time"  $B_2 = 0, t^* = a^*/b^*$ 



Mooney Viscosity (M.U.)

Rishi, K.; Narayanan, V.; Beaucage, G.; McGlasson, A.; Kuppa, V.; Ilavsky, J.; Rackaitis, M. A Thermal Model to Describe Kinetic Dispersion in Rubber Nanocomposites: The Effect of Mixing Time on Dispersion. Polymer (Guildf). 2019, 175, 272–282. 4



#### Wetting time depends on <u>viscosity</u> and <u>primary particle size</u>

x-intercept reflects "wetting time"  $B_2 = 0, t^* = a^*/b^*$ 14 N110 based nanocomposites 12 Series "B" Series "SB with 26% styrene" 10 Series "SB with 38% styrene" 8 ť (min) 6 4 2  $r^2 = 0.93$ 0 40 50 60 80 20 30 70 90

Mooney Viscosity (M.U.)

Rishi, K.; Narayanan, V.; Beaucage, G.; McGlasson, A.; Kuppa, V.; Ilavsky, J.; Rackaitis, M. A Thermal Model to Describe Kinetic Dispersion in Rubber Nanocomposites: The Effect of Mixing Time on Dispersion. Polymer (Guildf). 2019, 175, 272–282. 4

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# Morphology from rheology

• How does this multi-hierarchical model relate to oscillatory rheometry?



Filippone, G., Romeo, G. and Acierno, D., 2010. Viscoelasticity and structure of polystyrene/fumed silica nanocomposites: filler network and hydrodynamic contributions. *Langmuir*, *26*(4), pp.2714-2720.
Filippone, G. and Salzano de Luna, M., 2012. A unifying approach for the linear viscoelasticity of polymer nanocomposites. *Macromolecules*, *45*(21), pp.8853-8860.



 $\phi_{\rm eff}/\phi=\beta=(1+e)/(1+\varepsilon)=(1+\rho DBPA)/(1+\varepsilon)$ 

Einstein, 1909 (add Einstein reference here)

H. M. Smallwood, J. Appl. Phys. 1944, 15, 758.

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S. Wolff, M.-J. Wang, Rubber Chem. Technol. 1992, 65, 329.

G. Huber, T. A. Vilgis, Kgk Kautschuk Gummi Kunststoffe 1999, 52, 102.

G. Huber, T. A. Vilgis, Macromolecules 2002, 35, 9204.



Rishi, K., Beaucage, G., Kuppa, V., Mulderig, A., Narayanan, V., McGlasson, A., Rackaitis, M. and Ilavsky, J., 2018. Impact of an emergent hierarchical filler network on nanocomposite dynamics. *Macromolecules*, *51*(20), pp.7893-7904.

Vilgis, T.A. and Winter, H.H., 1988. Mechanical self similarity of polymers during chemical gelation. *Colloid and Polymer Science*, 266(6), pp.494-500.



Rishi, K., Beaucage, G., Kuppa, V., Mulderig, A., Narayanan, V., McGlasson, A., Rackaitis, M. and Ilavsky, J., 2018. Impact of an emergent hierarchical filler network on nanocomposite dynamics. *Macromolecules*, *51*(20), pp.7893-7904.

Macroscopic network dynamics is related to nanoscale clusters through the network  $d_f$ 



Einstein/Guth-Gold: G' = G'<sub>0</sub> (1 + 2.5  $\phi_f$ )

- At intermediate frequencies, deviation of semi-dilute rheology from dilute under same shear conditions ascertained by scaling dilute sample by Einstein-Smallwood factor
- At low oscillation frequencies, G'~G" indicates gel-like behavior\*

Network/Winter  $G' \sim G'' \sim \omega^{\alpha}$ 

Rishi, K., Beaucage, G., Kuppa, V., Mulderig, A., Narayanan, V., McGlasson, A., Rackaitis, M. and Ilavsky, J., 2018. Impact of an emergent hierarchical filler network on nanocomposite dynamics. *Macromolecules*, *51*(20), pp.7893-7904.





Okoli, U.; Rishi, K.; Beaucage, G.; Kammler, H. K.; McGlasson, A.; Michael, C.; Narayanan, V.; Grammens, J. *Dispersion and Dynamic Response for Flame-Synthesized and Chemically Modified Pyrogenic Silica in Rubber Nanocomposites*; in press 2022. *Polymer*.

#### Nanoscale control over hierarchical response

Rishi, K., Beaucage, G., Kuppa, V., Mulderig, A., Narayanan, V., McGlasson, A., Rackaitis, M. and Ilavsky, J., 2018. Impact of an emergent hierarchical filler network on nanocomposite dynamics. *Macromolecules*, *51*(20), pp.7893-7904.

### The impact of the local filler network on conductivity



Rishi, K., Gogia, A., Cui, X., Beaucage, G., Tang, A., Kumar, J., Kuppa, V. Structure-conductivity relationships in carbon black rubber nanocomposites. In preparation



#### Mixing Geometry: Dispersive (Breakup) versus Distributive (Arrangement) Mixing

Sample Name	Processing equipment	Screw Speed,	Accumulated Strain,
		Ń (rpm)	Yacc
GM-300	Twin Screw Extruder utilizing Gear	300	2,000
	Mixers (TSE - GM)		
GM-400	Twin Screw Extruder utilizing Gear	400	2,500
	Mixers (TSE - GM)		
KB-300	Twin Screw Extruder utilizing	300	2,700
	Kneading Blocks (TSE - KB)		
KB-400	Twin Screw Extruder utilizing	400	3,510
	Kneading Blocks (TSE - KB)		
SSE	Single Screw Extruder	30	2,020
Mixer	Brabender Mixer with Banbury blades	60	1,575

#### N110 Carbon Black in Polystyrene



Veigel D, Rishi K, BeaucageG, Galloway J, Campanelli H, Ilavsky J, Kuzmenko I, Fickenscher M, Okoli U *Nanocomposite dispersion and distribution in melt mixers* Submitted *Polymer* (2022).

#### **Distributive Mixing** $\equiv$ **Structural breakup and changes**





Distributive Mixing doesn't occur in laminar flow (*Re* << 2000 even at high strain rate) (Dispersive breakup can still occur if the particles are large enough)

 $Re = \frac{\rho v G}{M}$ 

η



Veigel D, Rishi K, BeaucageG, Galloway J, Campanelli I H, Ilavsky J, Kuzmenko I, Fickenscher M, Okoli U *Nanocomposite dispersion and distribution in melt mixers* Submitted *Polymer* (2022).



#### **Mixing Geometry and Shear Rate => Hierarchical Emergence**



#### **Mixing Geometry and Shear Rate => Hierarchical Emergence**

 $B_2$  measures the effective volume per aggregate which increases with breakup (dispersion) and better distribution

 $L_{\rm f,avg}$  seems to be most sensitive to distributive mixing and not sensitive to breakup/dispersion of nanoparticles



Inverse accumulated strain,  $1/\gamma_{acc}$ 

#### Nanoparticle Dispersion in Nanocomposites

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- There are multiple levels and not 1 that assemble from small to large sizes
- Dispersion has a size-scale dependence
- Large-scale network morphology relies on immiscible nanoscale clustering governed by a balance of thermodynamics and kinetics.
- This constitutes a multiscale physics approach
- The point of compatibilization is to control immiscibility that leads to hierarchical structures not to produce miscibility on the nano-scale
- Hierarchical dynamics can be directly linked to hierarchical structure
- We hope to couple structural measures of thermodynamics and MD modeling to predict structural emergence in complex hierarchical materials

### Acknowledgements

### BRIDGESTORE

Krauss Maffei



Color & Comfort

University of Dayton Research Institute



lyondellbasell





Rishi, K., Beaucage, G., Kuppa, V., Mulderig, A., Narayanan, V., McGlasson, A., Rackaitis, M. and Ilavsky, J., 2018. Impact of an emergent hierarchical filler network on nanocomposite dynamics. *Macromolecules*, *51*(20), pp.7893-7904.

# Modeling of Multi-Hierarchical Emergence

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**Figure 1.** a) X-ray tomograph of carbon network in reinforced elastomer by Song et al. [3] b) Illustration of four structural levels in a carbon black/elastomer nanocomposite: micron-scale network; nano- to colloidal-scale cluster; nanoscale aggregate; primary particles. Percolation of the aggregates occurs at about 5 weight percent and leads to enhanced dynamic mechanical properties, while percolation of the clusters into a micron-scale network occurs at about 20 weight percent and leads to bulk conductivity and tear resistance.

3. Song, L; Wang, Z; Tang, X.; Chen, L.; Chen, P.; Yuan, Q.; Li, L. *Visualizing the Toughening Mechanism of Nanofiller with 3D Xray Nano-CT: Stress-Induced Phase Separation of Silica Nanofiller and Silicone Polymer Double Networks* Macromolecules **50** 7249-7257 (2017).





**Figure 2.** Pigment nano-crystals (blue parallelogram) can exist as elemental particles or as clusters. Clusters of nano-crystals form primary particles depending on the solubility of the non-ionic surfactant (red hydrophilic/yellow oleophilic surfactant segments). The primary particles can exist as independent clusters at low temperatures near  $T_{CA}$ , or as ramified aggregates, above  $T_{CA}$  but below  $T_{LCST}$  for the surfactant. Aggregation is controlled by the nan-crystal cluster size, temperature, and bond enthalpy. [1]

Rishi K; Mulderig A; Beaucage G; Vogtt K; Jiang H *Thermodynamics of Hierarchical Aggregation in Pigment Dispersions* Langmuir **35** 13100-13109 (2019).
 Vogtt K; Beaucage G; Rishi K; Jiang H; Mulderig *A Hierarchical approach to aggregate equilibria* Phys. Rev. Res. **1** 033081 (2019).

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$$\Delta G_{2(1)} = RT \ln \left( \frac{N_1^T}{N_2^T} \right) = -RT \ln \left( \frac{1}{Z_{2(1)}} \right) = -RT \ln \left( K_{2(1)} \right)$$
 Vogtt Equation (1,2)

The number distribution of primary clusters,  $N_1(n_0)$ , consisting of  $n_0$  elemental crystals and the number distribution of aggregates of these primary clusters,  $N_2(n_1)$ , comprising  $n_1$  primary particles

 $N_1^{\rm T}$  is the total number of primary particles and  $N_2^{\rm T} = \int_0^\infty N_2(n_1) dn_1$  is the total number of aggregates at a given temperature  $N_1^{\rm T}/N_2^{\rm T}$  = number average degree of aggregation  $K_{2(1)}$  is the dissociation constant

Rishi K; Mulderig A; Beaucage G; Vogtt K; Jiang H *Thermodynamics of Hierarchical Aggregation in Pigment Dispersions* Langmuir **35** 13100-13109 (2019).
 Vogtt K; Beaucage G; Rishi K; Jiang H; Mulderig *A Hierarchical approach to aggregate equilibria* Phys. Rev. Res. **1** 033081 (2019).

Greg Beaucage, Kabir Rishi, Andrew J Mulderig, Karsten Vogtt University of Cincinnati



**Figure 5**. Rearrangement of simulated<sup>26</sup> nanoparticle pigment PY14 aggregates with increasing temperature. (Simulated structures are based on USAXS data. [17]) At higher temperatures, the overall change in free energy to remove an elemental crystallite from a primary particle is decreased. Simultaneously, the overall change in free energy to remove a primary particle from an aggregate is increased across the temperature series. The total aggregate mass (proportional to the number of elemental particles in an aggregate) is observed to reduce with temperature as the aggregates rearrange to form new equilibrium fractal structures. [2]

Vogtt K; Beaucage G; Rishi K; Jiang H; Mulderig A Hierarchical approach to aggregate equilibria Phys. Rev. Res. 1 033081 (2019).
 Mulderig, A; Beaucage, G; Vogtt, K; Jiang H; Kuppa V Quantification of branching in fumed silica J. Aerosol Sci. 109 28-37 (2017).
### A model for equilibrium clustering and its application to multi-hierarchical growth.

Greg Beaucage, Kabir Rishi, Andrew J Mulderig, Karsten Vogtt University of Cincinnati





Figure 4. Pigment blue 15:3 (beta copper phthalocyanine).

Triton X-100 surfactant (PEO, hydroxyl and aromatic/alkane end groups.)

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**Figure 3.** Thermal behavior for aqueous pigment yellow  $14/Triton^{TM}$  X-100 showing Boltzmann dependence predicted by the Vogtt model [1,2] for reversible equilibria in multi-hierarchical structures. a) Primary nano-crystal cluster degree of aggregation,  $z_{1(0)}$ , as a function of inverse temperature. The critical temperature is close to the LCST temperature for Triton<sup>TM</sup> X-100. b) Aggregate degree of aggregation,  $z_{2(1)}$ , as a function of inverse temperature. The critical temperature is close to volume ratio for large primary particles. ( $T_{CA}$  is below the freezing point of water.)

Rishi K; Mulderig A; Beaucage G; Vogtt K; Jiang H *Thermodynamics of Hierarchical Aggregation in Pigment Dispersions* Langmuir **35** 13100-13109 (2019).
 Vogtt K; Beaucage G; Rishi K; Jiang H; Mulderig *A Hierarchical approach to aggregate equilibria* Phys. Rev. Res. **1** 033081 (2019).

#### A model for equilibrium clustering and its application to multi-hierarchical growth.

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#### Work in Progress

We can measure  $z_{3(2)}$ ,  $z_{2(1)}$ , and  $z_{1(0)}$  in USAXS which yields dissociation free energies  $\Delta G_{3(2)}$ ,  $\Delta G_{2(1)}$ ,  $\Delta G_{1(0)}$ These can be used in a simulation of aggregate growth. But this doesn't yield the topology, only the mass of the different structural levels.

Topology comes from the transport kinetics (reaction limited growth/diffusion limited growth).

- 1) Specify the particle concentration and size.
- 2) Distribute the particles allowing for excluded volume.
- 3) Use LAMMPS MD environment with experimental free energies of aggregation and structural sizes and Langevin dynamics

Thermodynamics:  $z_{agglomerate}$ ;  $z_{aggregate}$  (degree of aggregation),  $d_p$ , Sauter mean diameter of primary particles Kinetics: topology,  $d_f$ ,  $d_{min}$ , c, number of branches, branch length

### **Clustering can lead to locally higher concentrations**



# **Clustering due to immiscibility**

# Kinetically dispersed systems

# Mean field



# Specific interactions



### **Descriptors**

Aggregate size, R Primary particle size,  $d_p$ Degree of aggregation, z Short-circuit path length, p Connective path length, s Fractal dimension,  $d_f$ Tortuosity,  $d_{min}$ Connectivity, c Branch fraction,  $\phi_{br}$ Branch number,  $n_{br}$ 



Beaucage, G. Determination of branch fraction and minimum dimension of mass-fractal aggregates. *Physical Review E – Statistical, Nonlinear, and Soft Matter Physics* 2004, 70, 031401-1–031401-10.

Ramachandran, R., Beaucage, G., Kulkarni, A. S., McFaddin, D., Merrick-Mack, J., & Galiatsatos, V. Branch content of metallocene polyethylene. *Macromolecules* 2009, 42, 4746–4750.

Rai, D., Beaucage, G., Jonah, E. O., Britton, D. T., Sukumaran, S., & Härting, M. Quantitative investigations of aggregate systems. *J. Chem. Phys.* 2012, 137, 044311–1–044311-6.

#### Scattering

Unified Fit: 
$$I(q) = \sum_{i=1}^{n} \left\{ G_i e^{-\frac{q^2 R_{g,i}^2}{3}} + e^{-\frac{q^2 R_{g,i+1}^2}{3}} B_i q^{*-P_i} \right\}; \quad q^* = \frac{q}{\operatorname{erf}\left(\frac{1.06q R_{g,i}}{\sqrt{3}}\right)^3} \quad q = \frac{4\pi}{\lambda} \sin\left(\frac{\theta}{2}\right)$$

Parameters from Unified Fit used to determine topological parameters:

$$\begin{split} d_{f} &= P_{2}; \quad z = \frac{G_{1}}{G_{2}} + 1; \quad d_{\min} = \frac{B_{2}R_{g,2}^{d_{f}}}{C_{p}\Gamma(d_{f}/2)G_{2}}; \quad C_{p} = \text{polydispersity factor} \\ R &= \frac{R_{g}}{d_{p}} = z^{1/d_{f}} = p^{1/d_{\min}} = s^{1/c} \qquad \varphi_{br} = \frac{z - p}{p} \qquad n_{br} = \frac{z^{\left[\left(\frac{9}{4d_{f}} - \frac{5}{4c}\right) + \left(1 - \frac{1}{c}\right)\right]} - 1}{2} \end{split}$$

Beaucage, G. Approximations Leading to a Unified Exponential/Power-Law Approach to Small-Angle Scattering J. Appl. Cryst. 1995, 28 (6), 717–728.

Herrmann, H. J., & Stanley, H. E. The fractal dimension of the minimum path in two- and three-dimensional percolation. *Journal of Physics A: Mathematical and General* **1988**, 21, L829–L833.

Meakin, P., Majid, I., Havlin, S., & Stanley, H. E. Topological properties of diffusion limited aggregation and cluster-cluster aggregation. *Journal of Physics A: Mathematical and General* **1984**, 17, L975–L981.

Witten, T. A., & Sander, L. M. Diffusion-limited aggregation. *Physical Review B* 1983, 27, 5686–5697.

Sorensen, C. M. Light scattering by fractal aggregates: A review. Aerosol Sci. Tech. 2001, 35, 648–687.

Algorithm:

Input z and a sticking probability Randomly grow aggregates Compute the scattering parameters, p, R, n<sub>Br</sub> etc. Iterate by varying sticking probability until computed matches experimental

A. Mulderig, G. Beaucage, K. Vogtt, H. Jiang and V. Kuppa, Quantification of branching in fumed silica, J. Aerosol Sci. 2017, 109, 28–37.

	-				
А	В	С	D	E	F
19	19	26	24	37	91
36	21	17	14	12	10
1.54	1.53	1.51	1.50	1.52	1.51
157	177	483	567	819	6430
36	35	46	38	37	91
66	75	195	265	819	6430
1.21	1.20	1.17	1.14	1.00	1.00
1.41	1.46	1.62	1.75	1.86	1.95
1.70	1.75	1.90	1.98	1.86	1.95
19	23	68	106	409	3215
77	80	91	93	95	99
	A 19 36 1.54 157 36 66 1.21 1.41 1.70 19 77	A         B           19         19           36         21           1.54         1.53           157         177           36         35           66         75           1.21         1.20           1.41         1.46           1.70         1.75           19         23           77         80	A         B         C           19         19         26           36         21         17           1.54         1.53         1.51           157         177         483           36         35         46           66         75         195           1.21         1.20         1.17           1.41         1.46         1.62           1.70         1.75         1.90           19         23         68           77         80         91	A         B         C         D           19         19         26         24           36         21         17         14           1.54         1.53         1.51         1.50           157         177         483         567           36         35         46         38           66         75         195         265           1.21         1.20         1.17         1.14           1.41         1.46         1.62         1.75           1.70         1.75         1.90         1.98           19         23         68         106           77         80         91         93	A         B         C         D         E           19         19         26         24         37           36         21         17         14         12           1.54         1.53         1.51         1.50         1.52           157         177         483         567         819           36         35         46         38         37           66         75         195         265         819           1.21         1.20         1.17         1.14         1.00           1.41         1.46         1.62         1.75         1.86           1.70         1.75         1.90         1.98         1.86           19         23         68         106         409           77         80         91         93         95

#### Experimental

#### Simulation

Z	157	177	483	567	819	6430
Sticking probability	0.40	0.37	0.11	0.23	0.82	1.00
R	18	19	26	29	40	93
р	31	33	45	42	40	93
S	68	75	191	298	819	6430
d <sub>min</sub>	1.2	1.2	1.17	1.11	1	1
с	1.47	1.48	1.62	1.7	1.82	1.93
d <sub>f</sub>	1.76	1.75	1.9	1.89	1.82	1.93
n <sub>br</sub>	21	26	68	121	409	3215
φ <sub>br</sub> , %	80%	81%	91%	93%	95%	99%

A. Mulderig, G. Beaucage, K. Vogtt, H. Jiang and V. Kuppa, Quantification of branching in fumed silica, J. Aerosol Sci. 2017, 109, 28–37.



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A. Mulderig, G. Beaucage, K. Vogtt, H. Jiang and V. Kuppa, Quantification of branching in fumed silica, J. Aerosol Sci. 2017, 109, 28–37.



Mulderig, A.; Narayanan, V.; McGlasson, A.; Rackaitis, M.; Ilavsky, J. Impact of an Emergent Hierarchical Filler Network on Nanocomposite Dynamics. Macromolecules 2018, 51, 7893–7904.Copyright 2018 American Chemical Society



A. Mulderig, G. Beaucage, K. Vogtt, H. Jiang and V. Kuppa, Quantification of branching in fumed silica, J. Aerosol Sci. 2017, 109, 28–37.

## Mean field (CB) and specific interactions (Silica)



McGlasson, A., Rishi, K., Beaucage, G., Chauby, M., Kuppa, V., Ilavsky, J. and Rackaitis, M., 2020. Quantification of dispersion for weakly and strongly correlated nanofillers in polymer nanocomposites. *Macromolecules*, 53(6), pp.2235-2248.

Rishi, K., Beaucage, G., Kuppa, V., Mulderig, A., Narayanan, V., McGlasson, A., Rackaitis, M. and Ilavsky, J., 2018. Impact of an emergent hierarchical filler network on nanocomposite dynamics. *Macromolecules*, 51(20), pp.7893-7904.

Rishi, K.; Pallerla, L.; Beaucage, G.; Tang, A. Dispersion of Surface-Modified, Aggregated, Fumed Silica in Polymer Nanocomposites. J. Appl. Phys. 2020, 127 (17), 174702. 86

## **Specific interactions (Silica)**

Positive  $a^*$  can lead to correlated silica aggregates, New scattering function to fit these curves (solves an impossible task)



McGlasson, A., Rishi, K., Beaucage, G., Chauby, M., Kuppa, V., Ilavsky, J. and Rackaitis, M., 2020. Quantification of dispersion for weakly and strongly correlated nanofillers in polymer nanocomposites. *Macromolecules*, *53*(6), pp.2235-2248.

Rishi, K.; Pallerla, L.; Beaucage, G.; Tang, A. Dispersion of Surface-Modified, Aggregated, Fumed Silica in Polymer Nanocomposites. J. Appl. Phys. 2020, 127 (17), 174702.



Structural ordering is related to the Debye screening length and the dielectric constant of the polymer



Rishi, K.; Pallerla, L.; Beaucage, G.; Tang, A. Dispersion of Surface-Modified, Aggregated, Fumed Silica in Polymer Nanocomposites. J. Appl. Phys. 2020, 127 (17), 174702.

## **Aggregates to Clusters**





Okoli, U.; Rishi, K.; Beaucage, G.; Kammler, H. K.; McGlasson, A.; Michael, C.; Narayanan, V.; Grammens, J. Dispersion and Dynamic Response for Flame-Synthesized and Chemically Modified Pyrogenic Silica in Rubber Nanocomposites; 2022. Submitted to Composites Sci. & Tech.



McGlasson, A., Rishi, K., Beaucage, G., Chauby, M., Kuppa, V., Ilavsky, J. and Rackaitis, M., 2020. Quantification of dispersion for weakly and strongly correlated nanofillers in polymer nanocomposites. *Macromolecules*, *53*(6), pp.2235-2248. Okoli, U.; Rishi, K.; Beaucage, G.; Kammler, H. K.; McGlasson, A.; Michael, C.; Narayanan, V.; Grammens, J. *Dispersion and Dynamic Response for* 

Flame-Synthesized and Chemically Modified Pyrogenic Silica in Rubber Nanocomposites; submitted 2022. Composites Sci. & Tech.



## **Clustered aggregates to bulk network**

McGlasson, A., Rishi, K., Beaucage, G., Chauby, M., Kuppa, V., Ilavsky, J. and Rackaitis, M., 2020. Quantification of dispersion for weakly and strongly correlated nanofillers in polymer nanocomposites. *Macromolecules*, *53*(6), pp.2235-2248.

Rishi, K., Beaucage, G., Kuppa, V., Mulderig, A., Narayanan, V., McGlasson, A., Rackaitis, M. and Ilavsky, J., 2018. Impact of an emergent hierarchical filler network on nanocomposite dynamics. *Macromolecules*, 51(20), pp.7893-7904.

Rishi, K.; Pallerla, L.; Beaucage, G.; Tang, A. Dispersion of Surface-Modified, Aggregated, Fumed Silica in Polymer Nanocomposites. J. Appl. Phys. 2020, 127 (17), 174702. 91

#### Method –

Si190 Precipitated silica, 3 grades of Aerosil® fumed silica, 6 grades of surface modified fumed silica, 4 grades of carbon coated silica

### SBR, PDMS and Polystyrene

Mixing in a Brabender Plasticorder, Single screw extruder and vortex mixing

F	Matrix / Processing Technique	Processing specs.	Equipment specifications	Geometry Constant, Ѱ	Total Accumulate d strain, $\gamma = Nt\Psi$
E	SB80-38 / Brabender Mixing	<i>N</i> = 60 rpm <i>t</i> = 12 min	$d_{wall} = 40 \text{ mm}$ $d_{rotor} = 35 \text{ mm}$ $\mathfrak{H} = \frac{d_{wall}}{d_{rotor}} = 1.14$ $n = 1^{a}$	$\frac{4\pi(\mathfrak{H})^{2/n}}{n\{(\mathfrak{H})^{2/n}-1\}} = 53.6 \text{ b}$	38600
	PS / Single screw Extrusion	<i>N</i> = 5 rpm <i>t</i> = 12 min	d <sub>screw</sub> = 11.7 mm L = 343 mm H(343) = 1 mm	$\frac{\pi \{d_{screw} - 2H(L)\}}{H(L)} = 33.6 \text{ c}$	2020
	PDMS / Vortex Mixing	<i>N</i> = 3200 rpm <i>t</i> = 0.5 min	$R \gg r$	$\frac{2\pi(2R^2)}{(R^2 - r^2)} = 12.6 \text{ d}$	20100

<sup>a</sup> power law index for styrene butadiene rubber
<sup>b</sup> Couette flow equation from Bousmina et al.
<sup>c</sup> equation from Hassinger et al.
<sup>d</sup> equation from Mezger

Rishi K, Pallerla L, Beaucage G, Tang A *Dispersion of surface-modified, aggregated, fumed silica in polymer nanocomposites* J. Appl. Phys. **127** 174702 (2020).

![](_page_92_Figure_0.jpeg)

**FIG. 11.** A plot showing the bound polymer content  $(\Delta b^*)$  determined from the excluded volumes of the filler aggregates before and after dispersion as a function of surface methyl content ( $N_{CH_a}$ ). The dashed line indicates that  $\Delta b^*$  is proportional to N<sub>CH</sub>, determined from FTIR.

 $b^*$  can be calculated as the excluded volume for an aggregate,  $zV_0$  without bound rubber

#### $b^*$ increases with bound rubber.

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![](_page_92_Figure_5.jpeg)

Figure 9. Plot of the particle interaction parameter,  $a^*$ , expressed in cm<sup>3</sup>/aggregate as a function of the linear sum of the surface carbon content  $(N_c)$  and surface hydroxyl content  $(N_{OH})$  weighted differently. a\* is an attractive potential so negative values indicate relative repulsion between aggregates that increases with surface carbon content. That is, surface carbon enhances aggregate/polymer attraction relative to aggregate/aggregate attraction. The fit parameters, A',  $K'_{C}$ , and  $K'_{OH}$  were obtained through least squares minimization.

 $a^*$  reflects the attractive energy of interaction between aggregates.  $\Pi = \frac{RT}{(V+b)} - \frac{a}{V^2}$ , attractive potential

Rishi, K.; Pallerla, L.; Beaucage, G.; Tang, A. Dispersion of Surface-Modified, Aggregated, Fumed Silica in Polymer Nanocomposites. J. Appl. Phys. 2020, 127 (17), 174702.