

Analysis of a twin-screw granulation process using a combined experimental and computational approach

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Introduction

► **Continuous wet granulation** using twin-screw granulators (TSG) is an important unit operation in future continuous manufacturing of pharmaceutical solid dosage forms.

► Several process and equipment settings govern the extent of different **rate processes** such as **aggregation** and **breakage** involved in granulation.

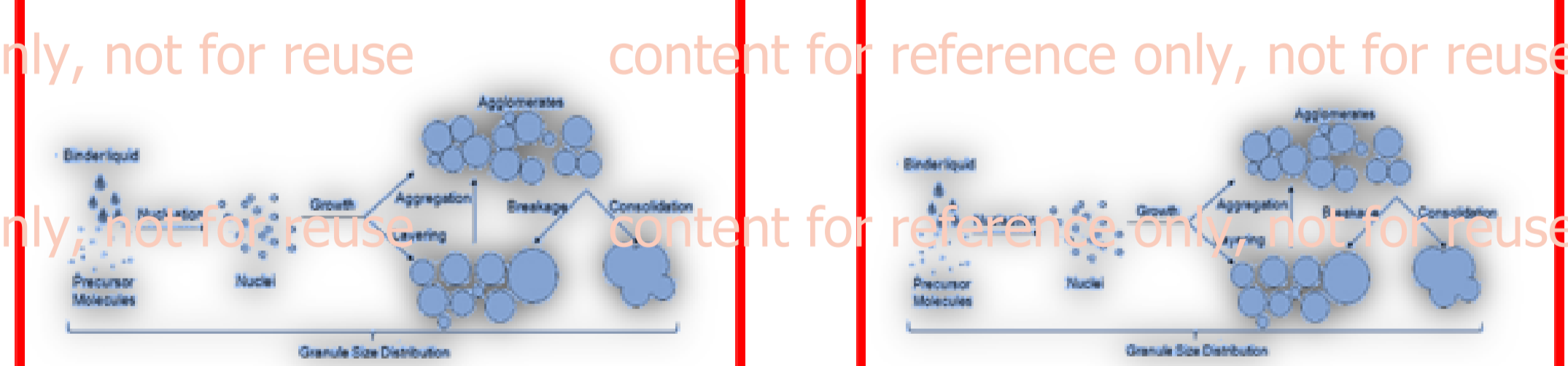
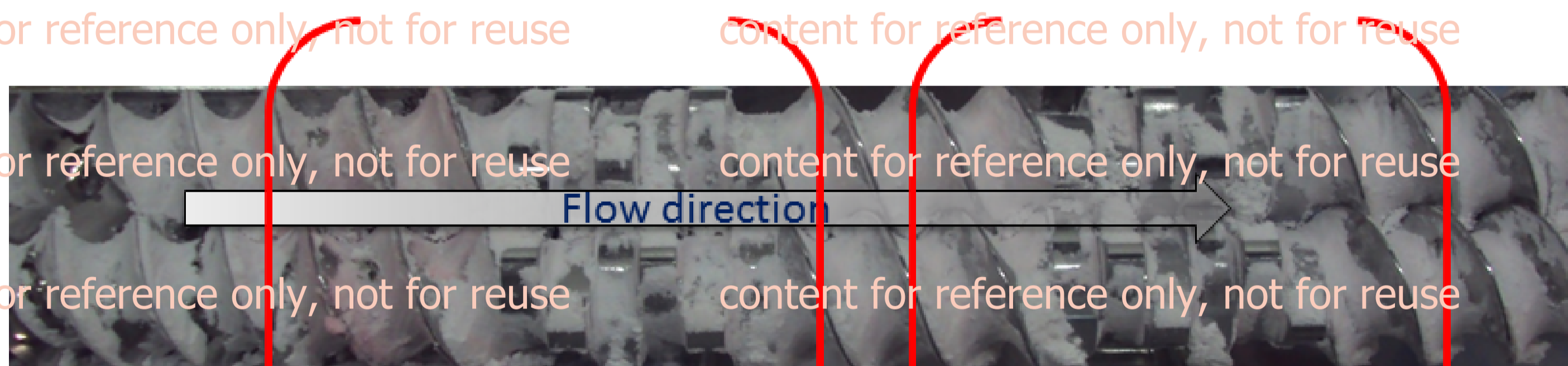
► However, little is in fact known about how these variables affect the evolution and kinetics of granule formation.

► **Combination of theoretical development and experimental validation** of a population balance modelling framework can bridge this gap to track the particle evolution during twin-screw granulation.

Twin-screw wet granulation

► Granulation time is short (in the order of seconds).

► At appropriate time scales and conditions, granulation is in steady state.



Mixing zone 1

Mixing zone 2

► Mixing zones in the TSG were assumed as well-mixed compartments.

PBM to track particle size evolution

$$\frac{\partial n(t, x)}{\partial t} = \frac{Q_{in}}{\tilde{V}} n_{in}(x) - \frac{Q_{out}}{\tilde{V}} n_{out}(x) \quad \text{GSD balance}$$

$$+ \frac{1}{2} \int_0^x \beta(t, x-\varepsilon, \varepsilon) n(t, x-\varepsilon) n(t, \varepsilon) d\varepsilon - n(t, x) \int_0^\infty \beta(t, x, \varepsilon) n(t, \varepsilon) d\varepsilon$$

aggregation rate

$$+ \int_0^\infty b(x, \varepsilon) S(\varepsilon) n(t, \varepsilon) d\varepsilon - S(x) n(t, x)$$

breakage fun. selection rate

selection rate

Kernels and solution method

Aggregation Kernel

$$\beta(x, y) = \beta_0$$

(Constant kernel)

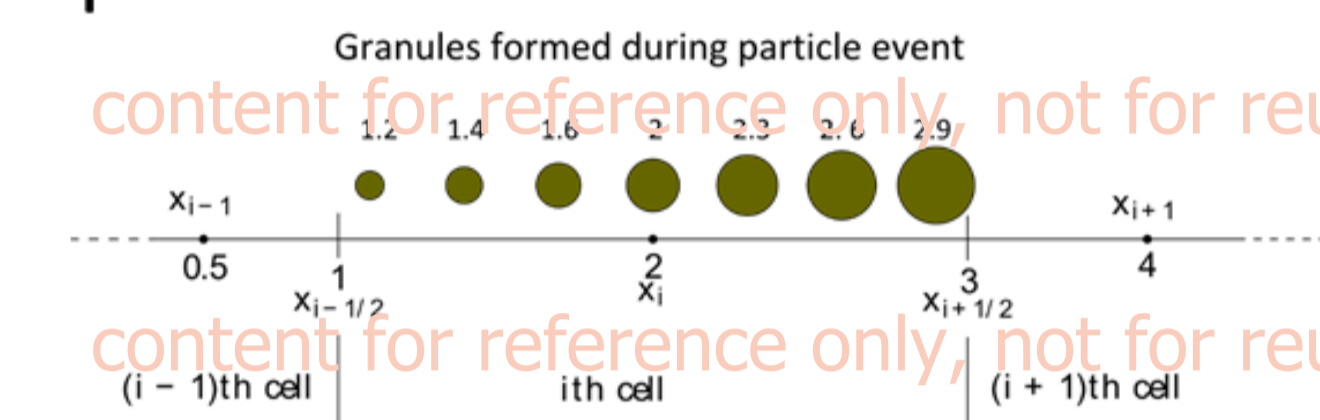
Breakage Kernel

$$S(y) = S_0(y)^\alpha$$

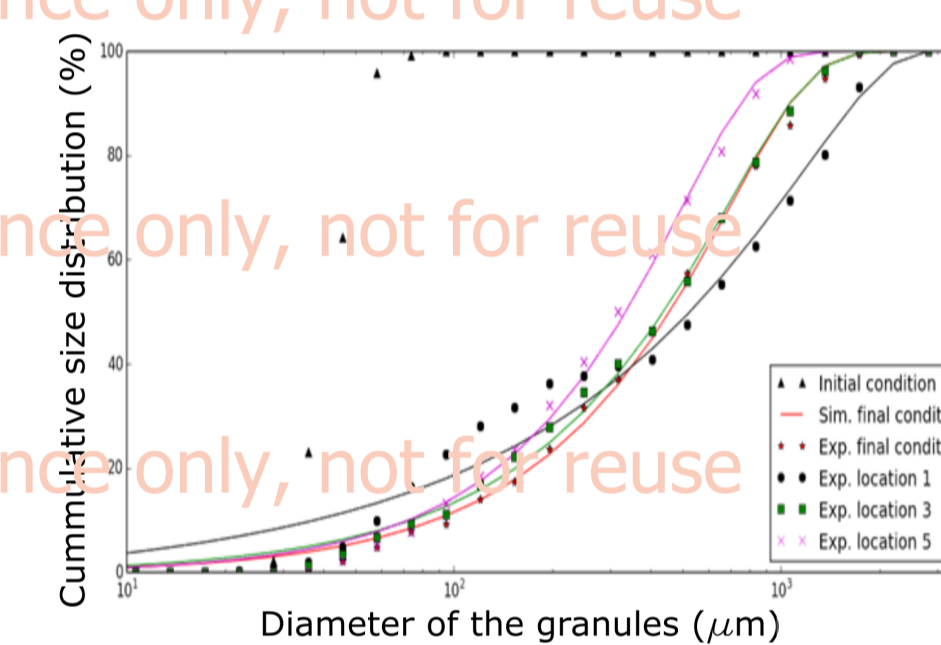
$$b(x, y) = \frac{\phi y^{1-\alpha} + (1-\phi) \alpha x^{\alpha-1}}{y^{\alpha+1} + \frac{(1-\phi) \alpha}{\alpha+1}}$$

(Austin, 2002)

Cell-average technique



Parameter estimation

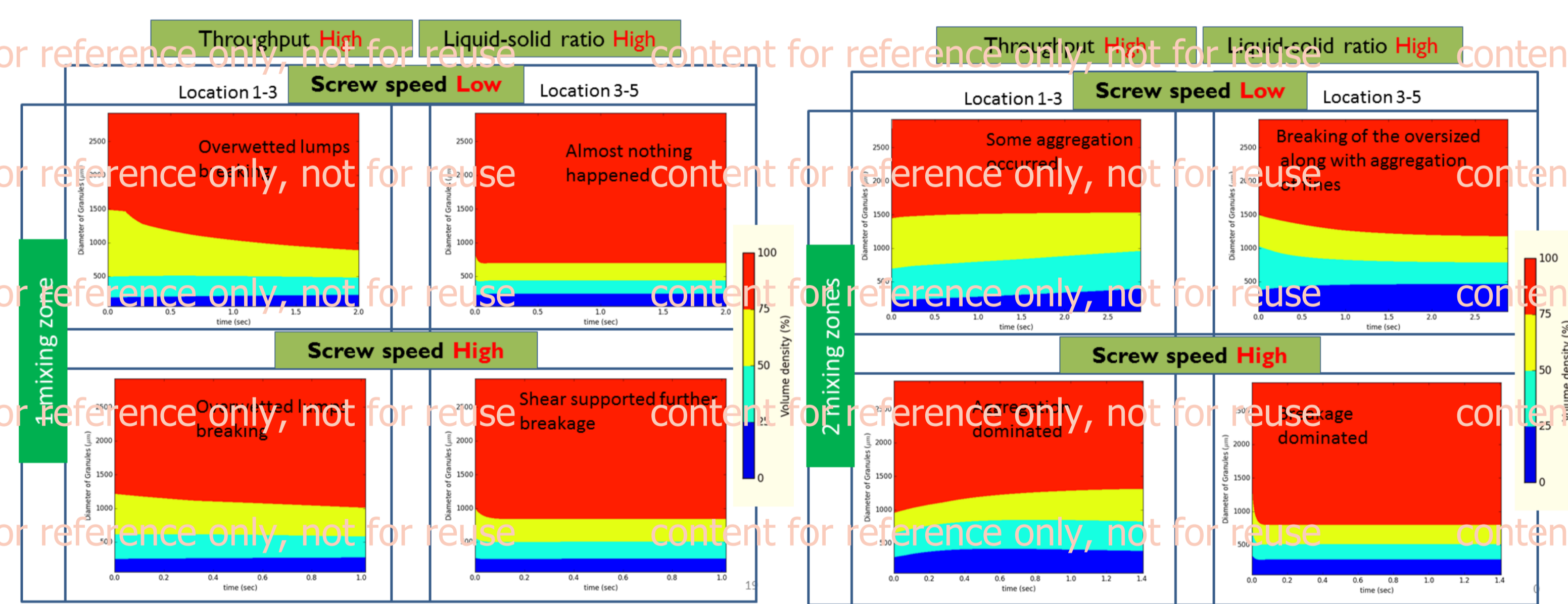


Parameter sets were derived by minimising the root mean square error (RMSE) via Monte Carlo Simulations

$$RMSE = \sqrt{\frac{\sum_{i=1}^n (y_{exp} - y_{sim})^2}{n}}$$

	1 st kneading block		2 nd kneading block	
	1	2	3	4
Zone	1-3	3-5	1-3	3-5
RMSE	2.424	2.317	2.716	3.929
R ²	0.989	0.987	0.984	0.983
β ₀	1.05E-03	3.12E-01	3.11E-03	1.20E-02
S ₀	0.030	3.304	0.023	0.062
α	5.01E-02	1.35E-02	1.21E-03	6.02E-04
γ	0.52	0.42	3.69	0.63
φ	0.97	0.72	0.93	0.98

Data postprocessing results



Take home message

By choosing appropriate process conditions regime separation can be achieved in TSG, allowing for improved design and operation of the continuous granulation process.

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