

# Stochastic weighted particle methods for fragmentation, coagulation and spatial inhomogeneity



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# Introduction

- In this work, we develop stochastic particle methods to solve population balance models with multiple compartments.
- Two popular stochastic particle methods:
  - Direct simulation algorithm (DSA)
  - Stochastic weighted algorithm (SWA)
- Application of stochastic particle methods in problems with spatial inhomogeneity is relatively new
- In particular, this work presents a family of fragmentation algorithms for SWA



# Introduction

- In this work, we develop stochastic particle methods to solve population balance models with multiple compartments.
- Two popular stochastic particle methods:
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- Application of stochastic particle methods in problems with spatial inhomogeneity is relatively new
- In particular, this work presents a family of fragmentation algorithms for SWA



# DSA (Direct Simulation Algorithm)

- Each computational particle represent the same number of real particles
- Coagulation events delete the original coalescing particles to create a new particle. This leads to a numerical issue where the ensemble will eventually deplete:
  - A **doubling algorithm** is used where the ensemble is duplicated when the number of particles falls below  $3/8$  of ensemble size
- Breakage increases number of particles
  - **Downsampling** may be necessary to avoid memory problems



# SWA (Stochastic Weighted Algorithm)

- A statistical weight is attached to each computational particle
- The statistical weight is representative of the number of particles
- Coagulation events do not change the number of computational particles, only the weights are adjusted
- No need for doubling algorithm – ideal for applications with high coalescence rates
- Inception (and breakage) introduce new particles and require **downsampling**



# Fragmentation: model definition

- A particle ( $x$ ) breaks into particles ( $y$ ) and ( $x-y$ ) with the frequency  $g(x)$ :

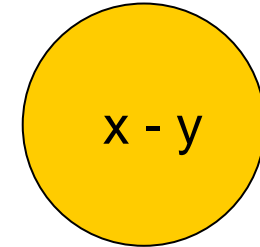
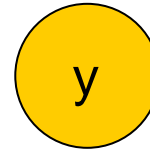
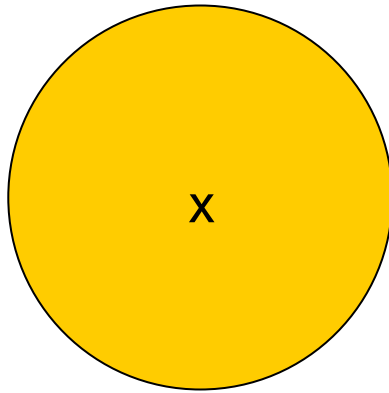
$$(x) \rightarrow (y), (x - y)$$

- The fragment particles ( $y$ ) and ( $x-y$ ) are defined by a probability density function

$$\beta(x, y)$$

- where  $(y) < (x)$ .

# Fragmentation: DSA



Particles break  
at the  
frequency:

$$g(x)$$

$$\text{Waiting time} = \sum g(x)$$

Size of  $y$  is  
selected  
according to:

$$\beta(x, y)$$

(symmetric)

The second  
fragment is  
determined as  
a function of  $y$ :

$$x - y$$

# Fragmentation: SWA

- Particle weights are no longer identical
- Main purpose: Perform breakage processes which change one particle at a time, i.e.

$$(x, w_x) \rightarrow (y, w_y)$$

- In order to simulate the process accurately, we need to define an appropriate weight transfer function,  $\gamma$  to calculate  $w_y$

$$\begin{aligned} w_y &= \gamma(x, w_x, y) \\ &= w_x \alpha(x, y) \end{aligned}$$



# Fragmentation: SWA

- Our algorithm will have the correct approximation if alpha is defined according to the restriction below:

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

- Two definitions are used in our studies:

SWA1

$$\alpha(x, y) = 2$$

Simplest solution but not efficient

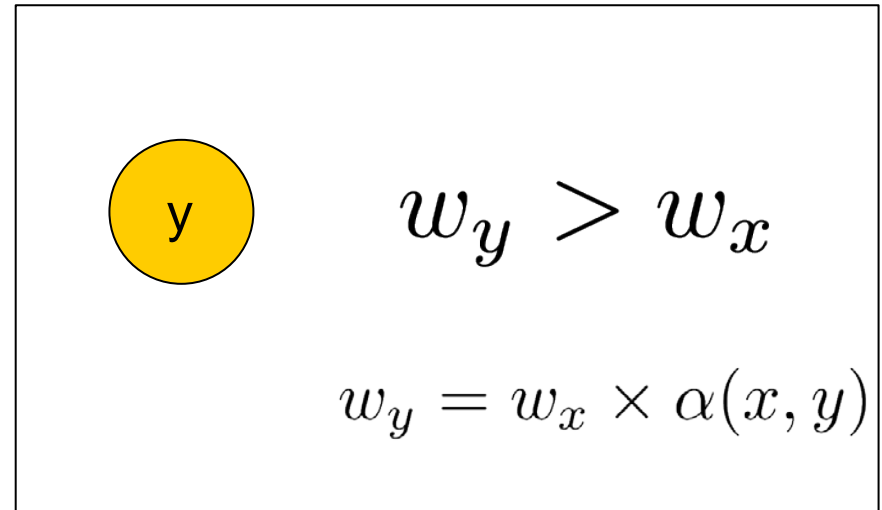
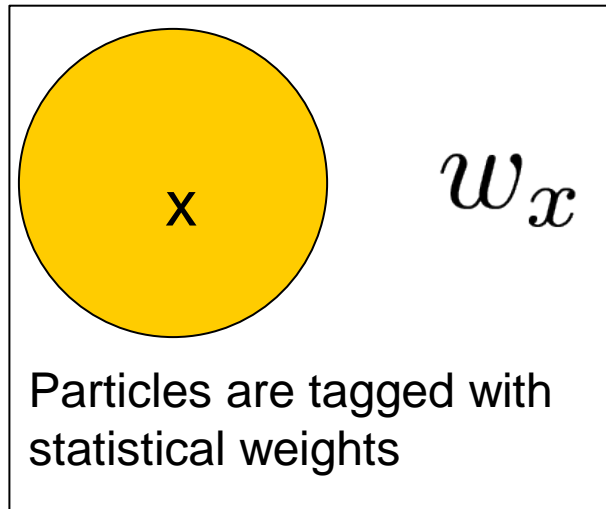
SWA2

$$\alpha(x, y) = \frac{m(x)}{m(y)}$$

Conserves total mass



# Fragmentation: SWA



Fragmentation frequency:  $g(x)$

Waiting time =  $\sum g(x)$

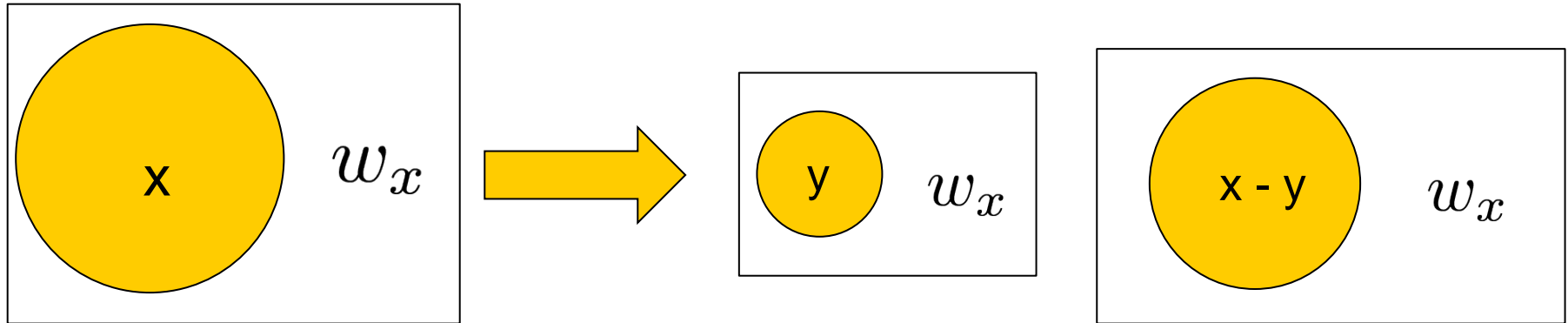
Size of  $y$  is selected according to:

$$\frac{w_x}{w_y} \beta(x, y)$$

(No longer symmetric)

$w_y$  is determined by the weight transfer function

# Fragmentation: SWA/DSA alternative



Fragmentation frequency:  $g(x)$

Size of  $y$  is selected according to:  $\beta(x, y)$

Waiting time =  $\sum g(x)$

Fragments are given the same weights, but this jump process increase the number of particles

SWA3

# Fragmentation: simulation algorithm

	DSA	SWA1	SWA2	SWA3
Waiting time	$\sum g(x)$			
Selection of particle to break	$g(x) / \sum g(x)$			
Selection of fragment particles	$\beta(x, y)$	$\frac{w_x}{w_y} \beta(x, y)$		$\beta(x, y)$
Jump process	$(x) \boxtimes$ $\boxtimes (y), (y-x)$	$(x, w_x) \boxtimes (y, w_y)$		$(x, w_x) \boxtimes$ $\boxtimes (y, w_x), (y-x, w_x)$
Weight transfer function	N/A	$w_y = 2w_x$	$w_y = \frac{m(x)}{m(y)} w_x$	N/A

# Coagulation: model definition

- A particle ( $x$ ) coagulates with particle ( $y$ ) to form a larger particle ( $x+y$ ):

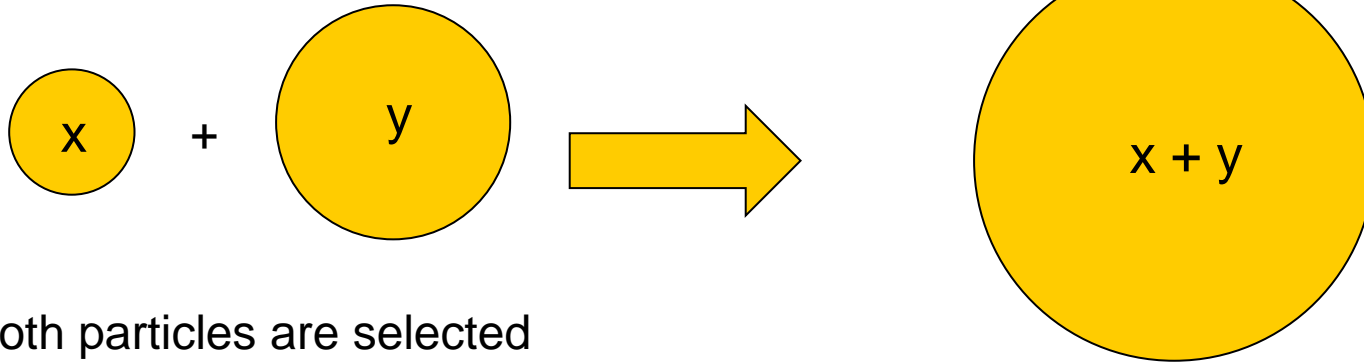
$$(x), (y) \rightarrow (x + y)$$

- The rate is specified by a kernel

$$K(x, y)$$



# Coagulation: DSA



Both particles are selected uniformly

Each pair coagulates at the rate =

$$\frac{K(x, y)}{n}$$

$$\text{Total waiting time} = \frac{1}{2} \sum_{x \neq y} \frac{K(x, y)}{n}$$

$n$  = normalisation parameter

DSA depletes the number of particles

Doubling algorithm is necessary to prevent ensemble from depletion

# Coagulation: SWA

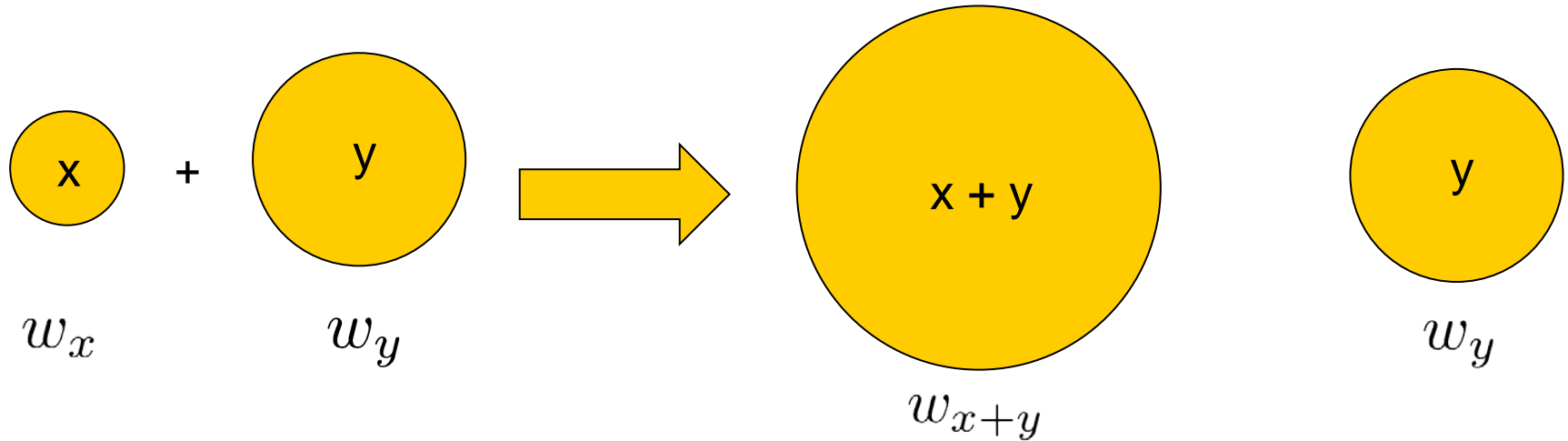
- Particle weights are no longer identical
- Main purpose: Perform coagulation jumps by changing one particle at a time, *i.e.*

$$(x, w_x), (y, w_y) \rightarrow (x + y, w_{x+y}), (y, w_y)$$

- At the rate  $\frac{\hat{K}(x, w_x, y, w_y)}{n} = \frac{K(x, y)w_y}{n}$
- The weight  $w_{x+y}$  is defined as

$$w_{x+y} = w_x \times \frac{m(x)}{m(x+y)}$$

# Coagulation: SWA



Each pair  
coagulates at  
the rate:

$$\frac{K(x, y)w_y}{n}$$

$w_{x+y}$  is determined  
by the weight  
transfer function

Total waiting  
time =  $\frac{\sum K(x, y)w_y}{n}$

Only 1 particle is changed at a time

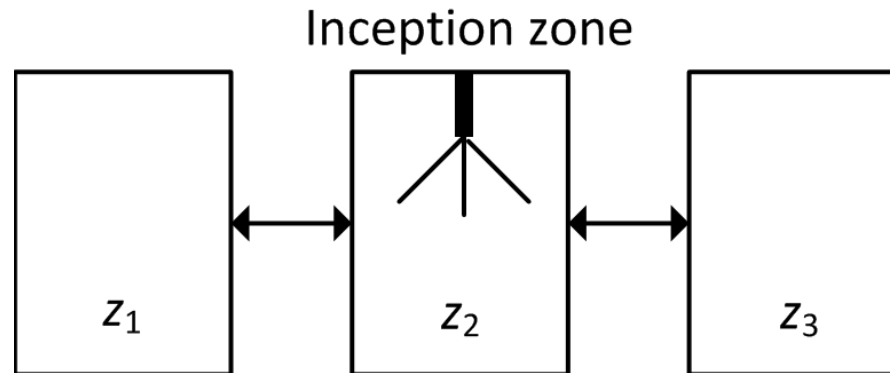
$n$  = normalisation parameter



# Coagulation: simulation algorithms

	DSA	SWA1/2/3
Waiting time	$\frac{1}{2} \sum_{x \neq y} \frac{K(x, y)}{n}$	$\sum_{i \neq j} \frac{K(x, y) w_y}{n}$
Selection of particle 1	Uniform	Uniform
Selection of particle 2	Uniform	$\frac{w_y}{\sum w}$
Jump process	$(x), (y) \square (x + y)$	$(x, w_x), (y, w_y) \square (x + y, w_{x+y}), (y, w_y)$
Weight transfer function	N/A	$w_{x+y} = \frac{m(x)}{m(x + y)} w_x$

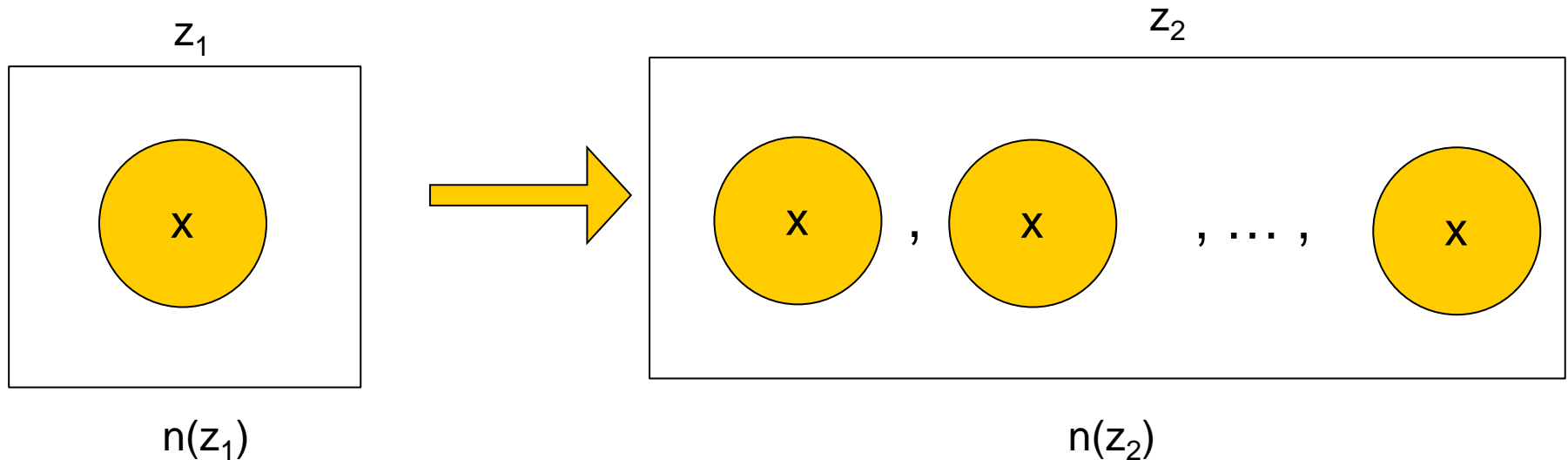
# Compartmental model - Transport



Normalisation parameter	$n(z_1)$	$n(z_2)$	$n(z_3)$
Residence time	$\varrho(z_1)$	$\varrho(z_2)$	$\varrho(z_3)$
Rate of particle leaving	$\frac{1}{\tau(z_1)}$	$\frac{1}{\tau(z_2)}$	$\frac{1}{\tau(z_3)}$
Particle destination	100% to $z_2$	50% to $z_1$ 50% to $z_3$	100% to $z_2$



# Transport: DSA



Number of copies is determined randomly by the ratio of normalisation parameters

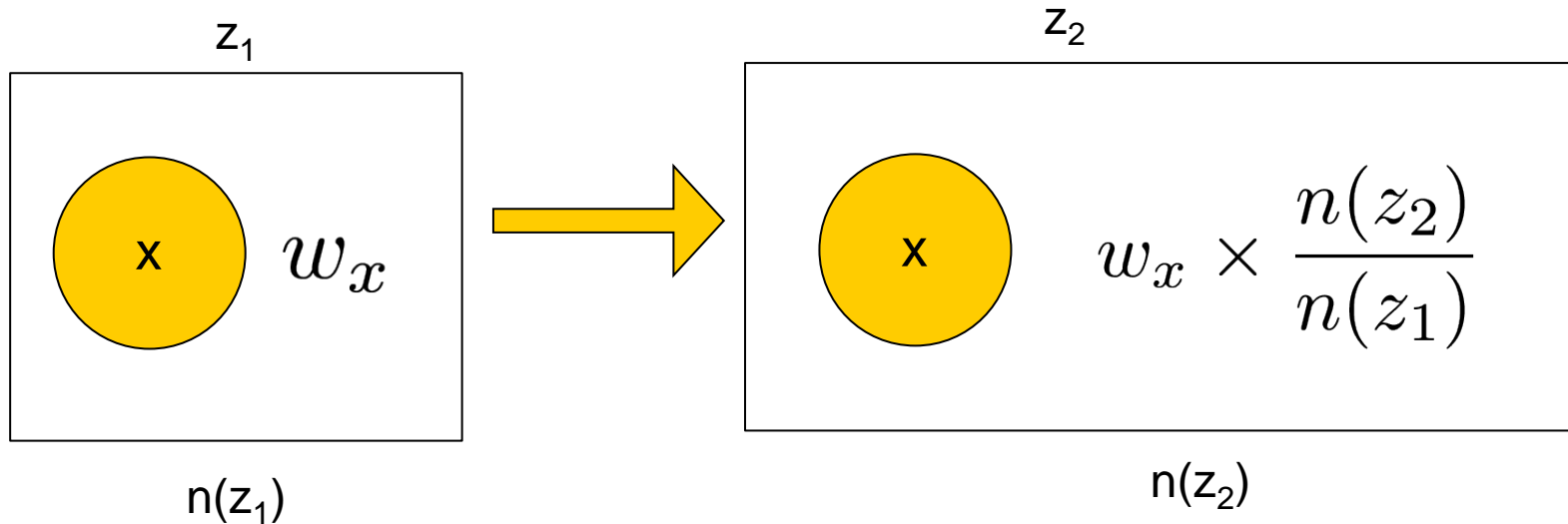
$$\text{Rate} = \frac{1}{\tau(z_1)}$$

$$n_c \approx \frac{n(z_2)}{n(z_1)} \longrightarrow \text{Usually not an integer}$$

Decide randomly between  $\lfloor \frac{n(z_2)}{n(z_1)} \rfloor + 1, \lfloor \frac{n(z_2)}{n(z_1)} \rfloor$

$$u < \frac{n(z_2)}{n(z_1)} - \lfloor \frac{n(z_2)}{n(z_1)} \rfloor$$

# Transport: SWA



$$\text{Rate} = \frac{1}{\tau(z_1)}$$

No need to determine the number of copies randomly with the presence of statistical weights

# Test system

- Type space  $\mathcal{X} = \{1, 2, 3, \dots\}$
- So the system can be written as a series of ODEs
- Analysed the performances of the algorithms at different rates
- Constant coagulation kernel (only differ in different compartments):

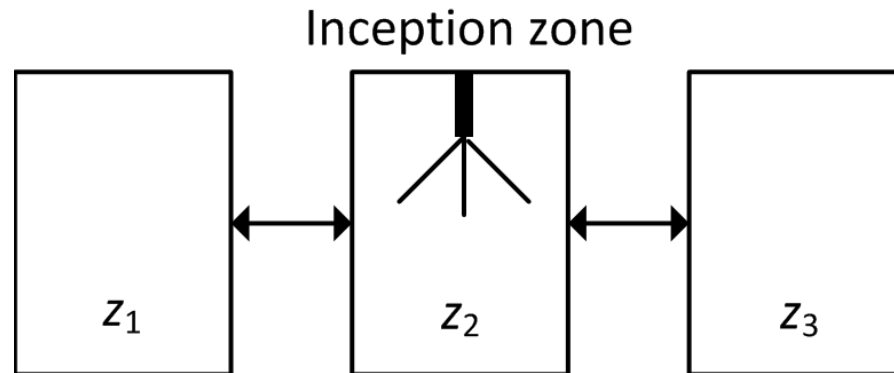
$$K(z, x, y) = k_{\text{coag}}(z)$$

- Fragmentation: Frequency proportional to size and a fragmentation probability density  $\beta$

$$g(z, x) = x k_{\text{frag}}(z) \quad \beta(x, y) = \frac{1}{x - 1}$$



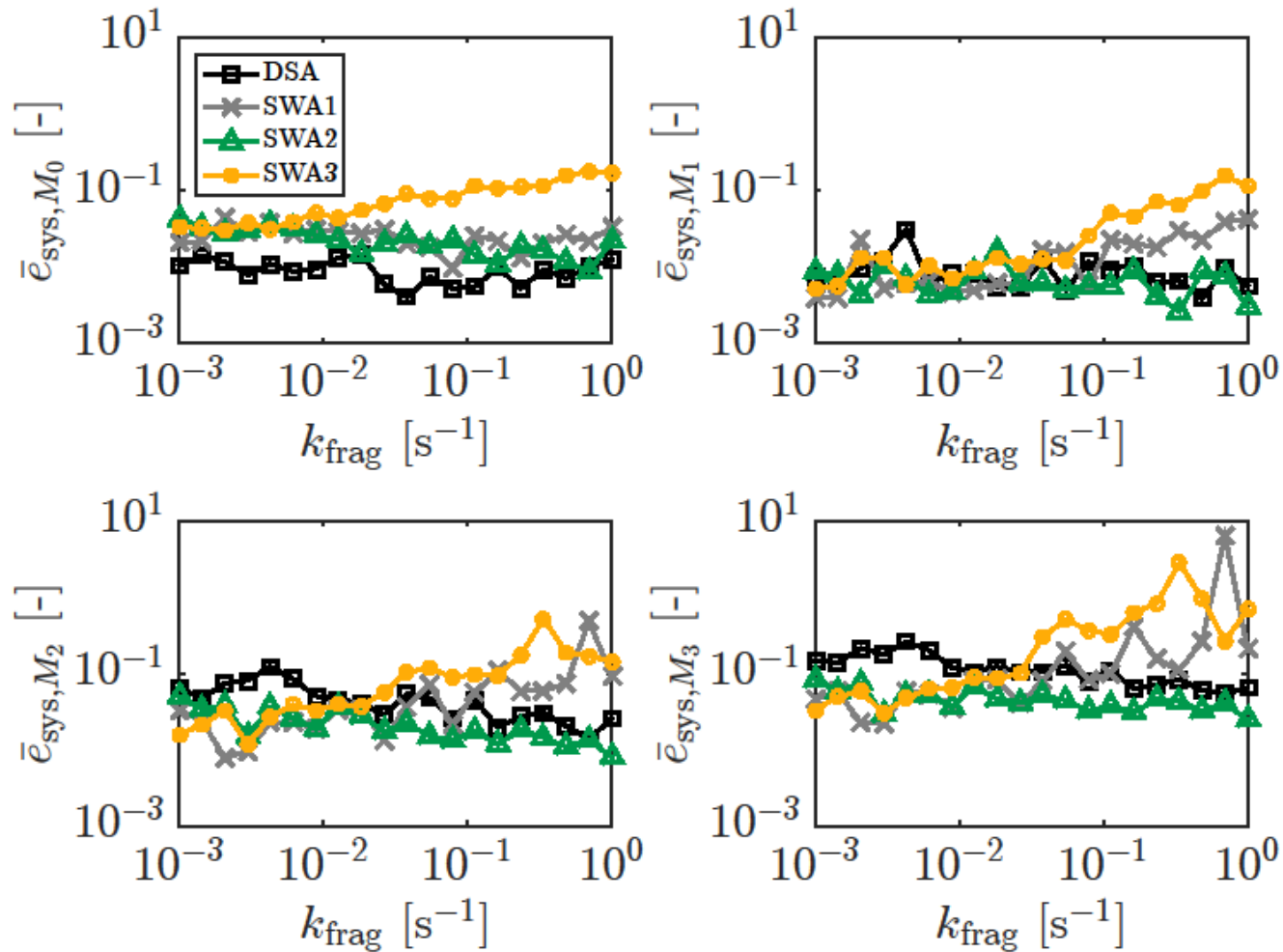
# Test system: compartmental model



Normalisation parameter	$n(z_1)$	$n(z_2)$	$n(z_3)$
Residence time	$\varrho(z_1)$	$\varrho(z_2)$	$\varrho(z_3)$
Rate of particle leaving	$\frac{1}{\tau(z_1)}$	$\frac{1}{\tau(z_2)}$	$\frac{1}{\tau(z_3)}$
Particle destination	100% to $z_2$	50% to $z_1$ 50% to $z_3$	100% to $z_2$

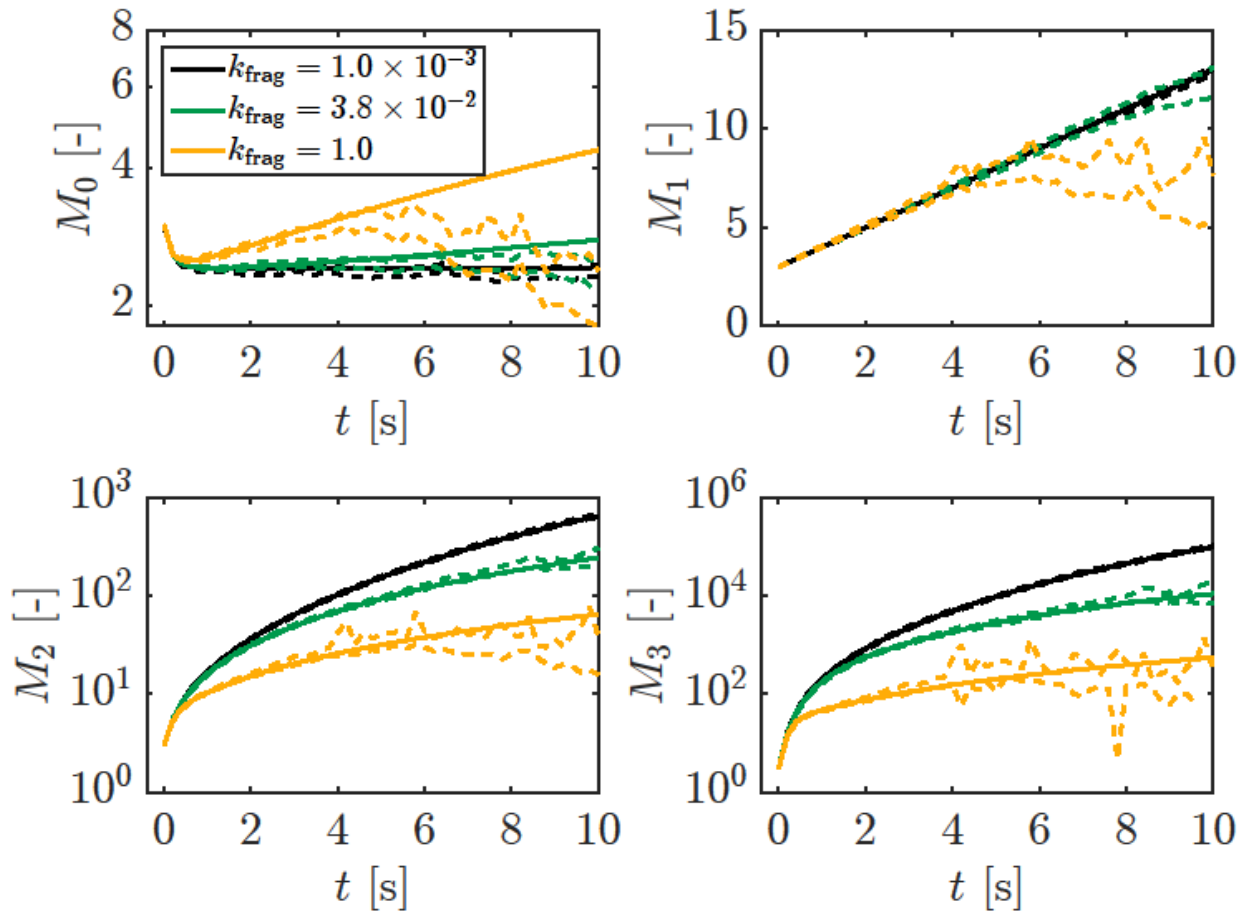


# Errors at different frag. rates for each algorithm



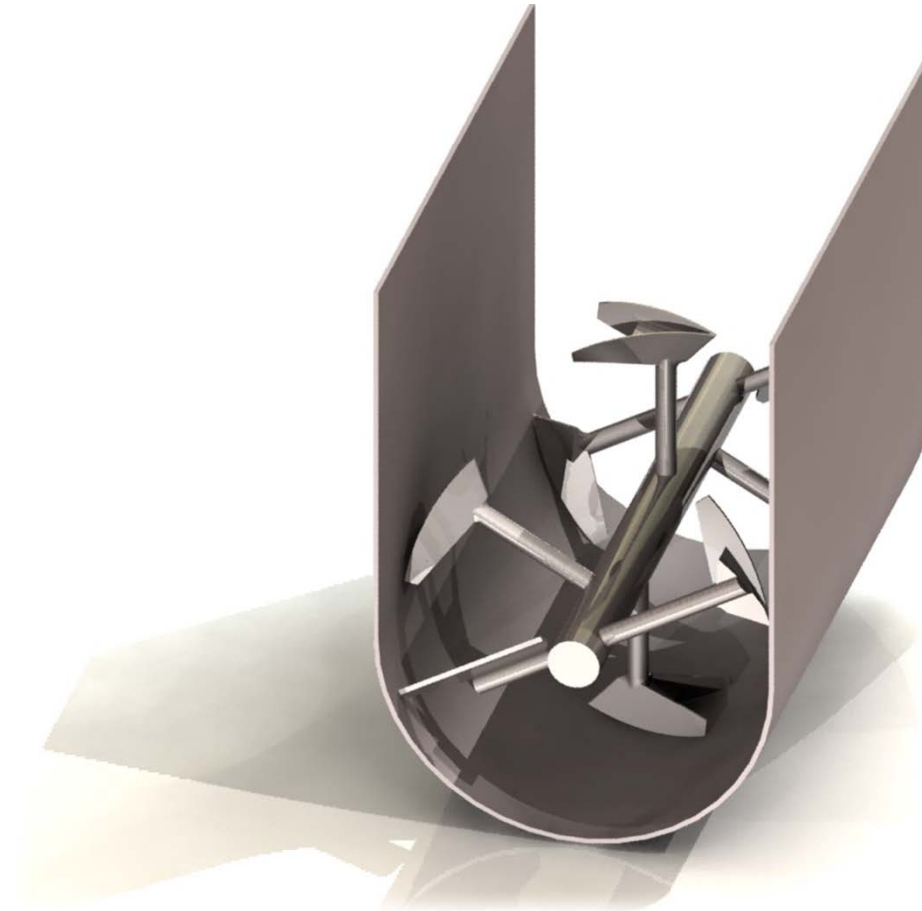
# SWA3 – worst algorithm

errors increase with frag. rate



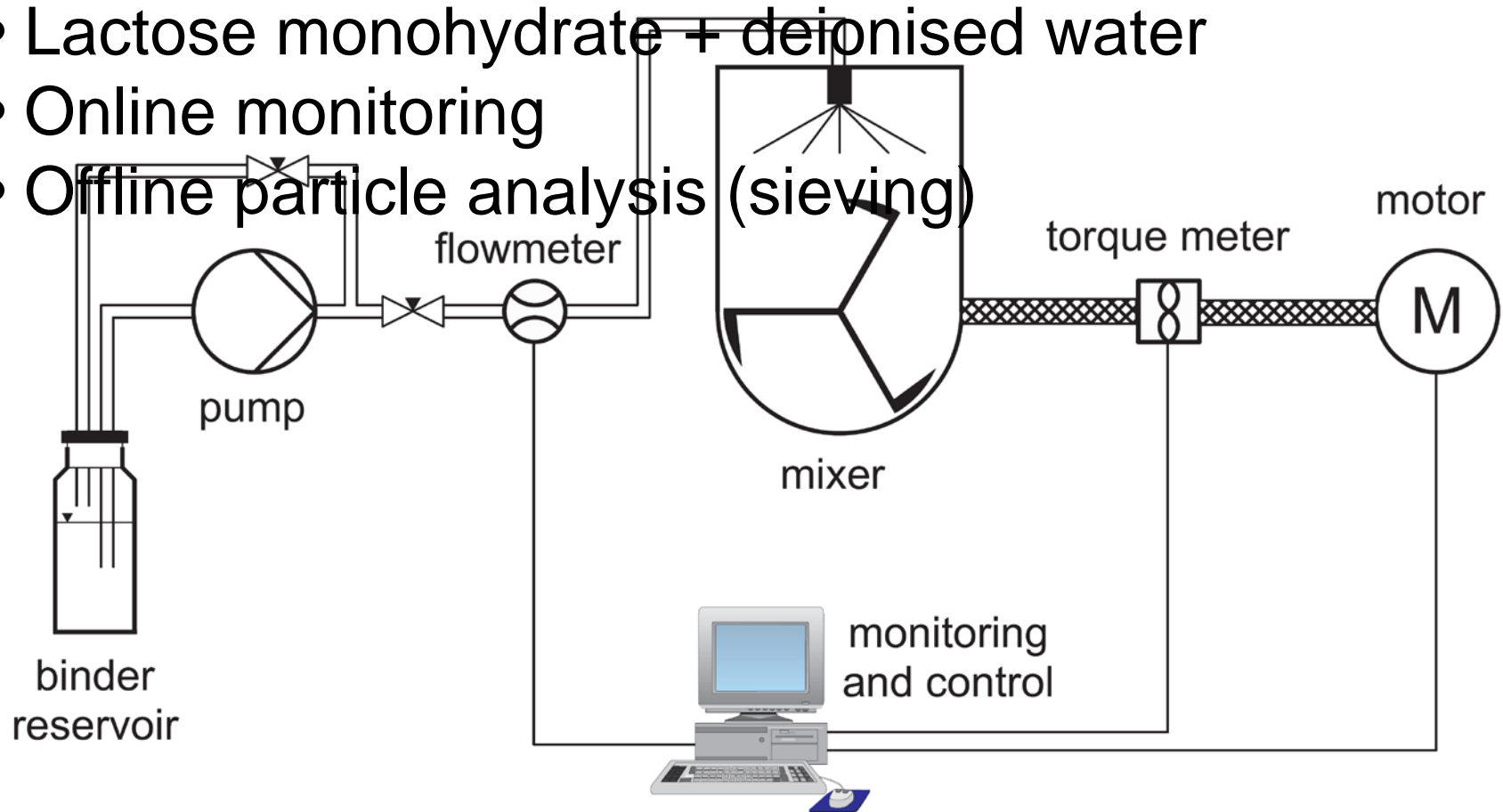


# Experimental system

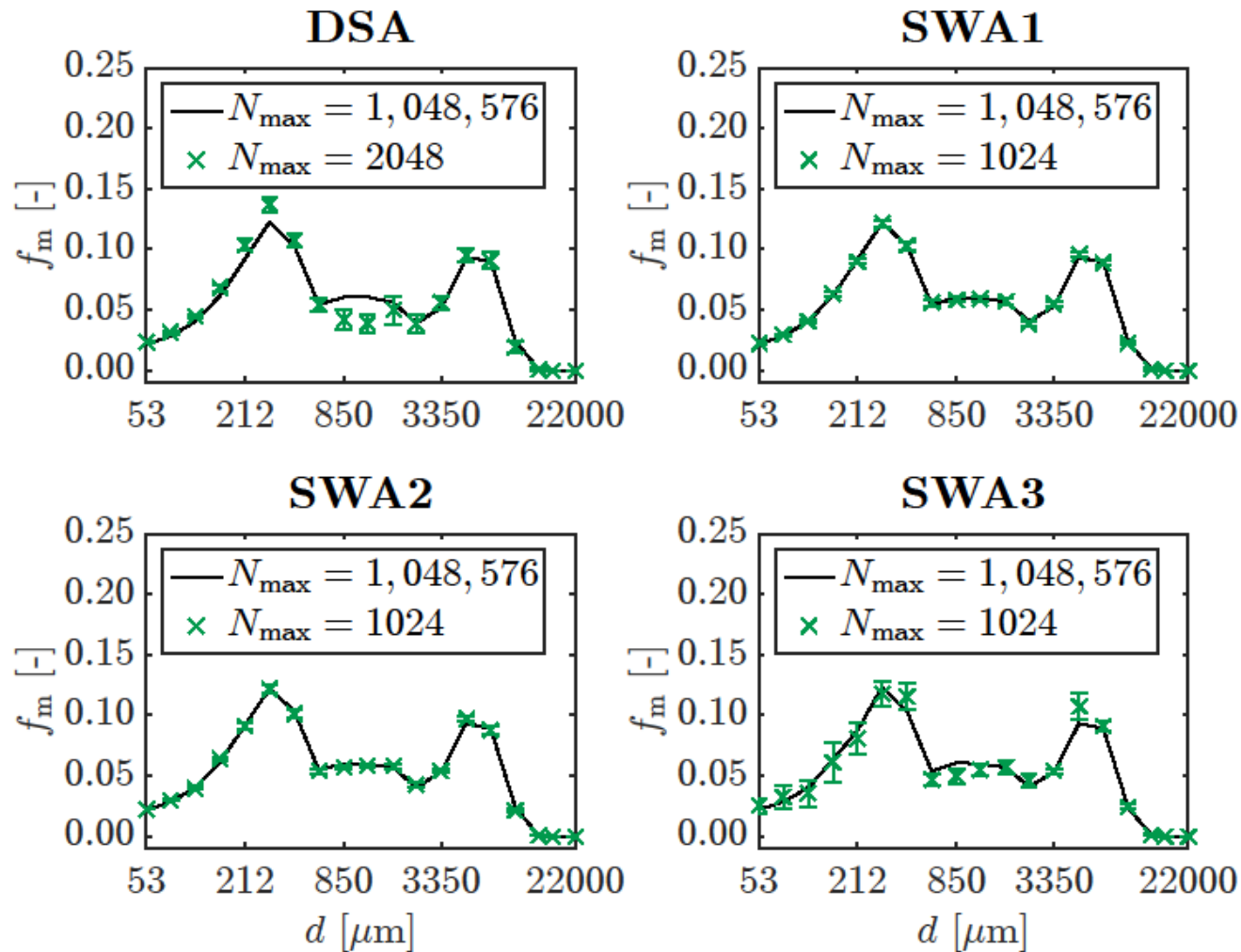


# Experimental system

- Bench scale system
- Lactose monohydrate + deionised water
- Online monitoring
- Offline particle analysis (sieving)

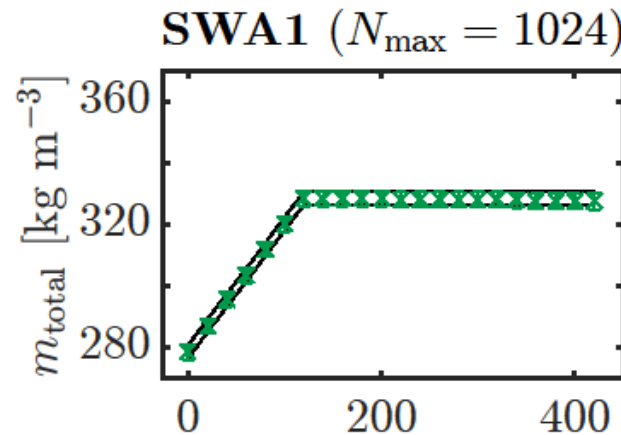
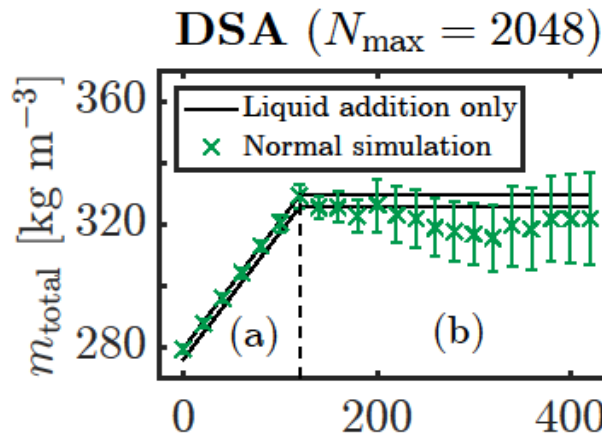


# Computational efficiency

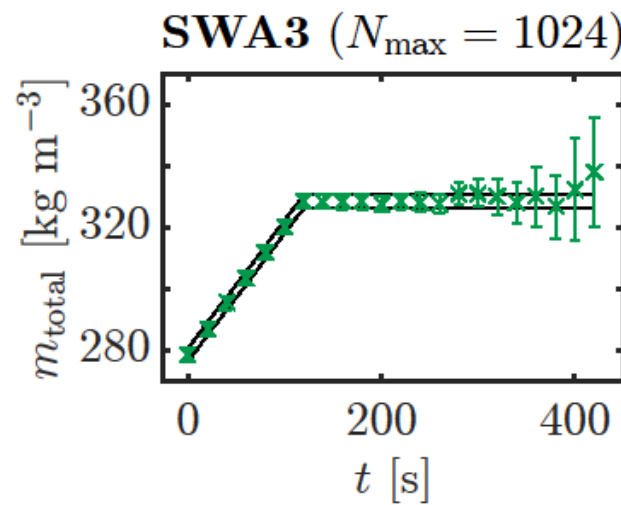
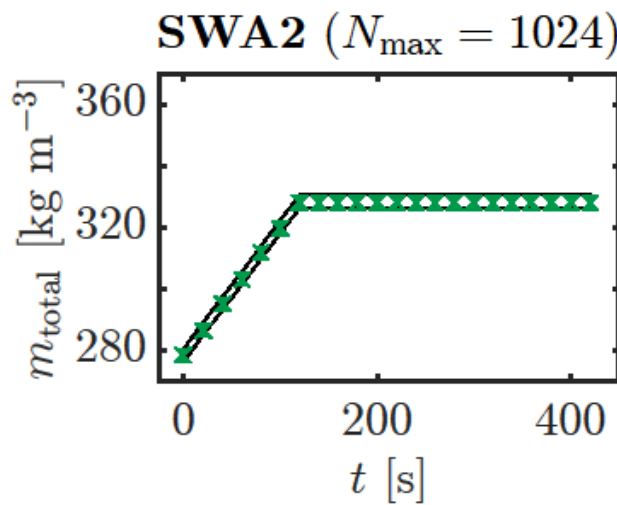


# Problems with DSA and SWA3

## - fluctuation of mass



DSA: Error mainly from transport



SWA3: Error mainly from particle deletions

# Conclusions

- A new family of fragmentation algorithms for weighted particles have been introduced in the context of granulation models
- All the algorithms converge to the same solution, but the new algorithms are more efficient
- The fragmentation algorithms are applied to a multi-dimensional population balance model
- It is found that the new algorithms provide significant numerical stability (e.g. negligible fluctuation in total mass)



# Acknowledgements



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