

# Tackling the inverse problem in multidimensional granulation modelling



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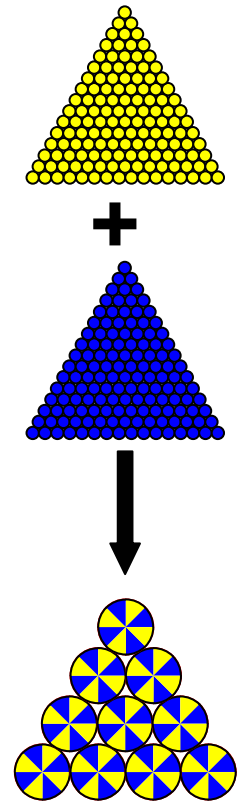


# Wet granulation – A start

- Products



- “Lump” small particles into bigger entities
  - Improve **handling** (storage, transport, safety,...)
  - Creation of **micro mixtures** (segregation, distribution of active components)
- **Here:** Model the process with multidimensional population balance model



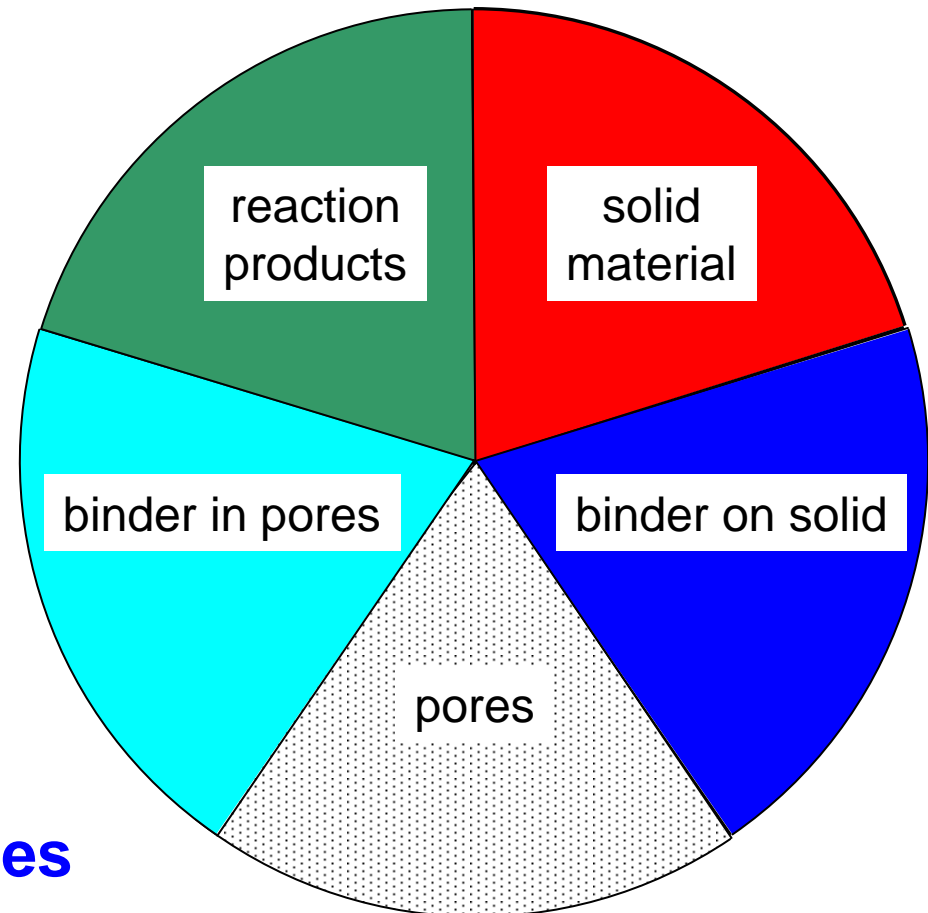
# Model – Particle description

- composition of a granule

- solid material
- binder on solid
- pores
- binder in pores
- reaction products

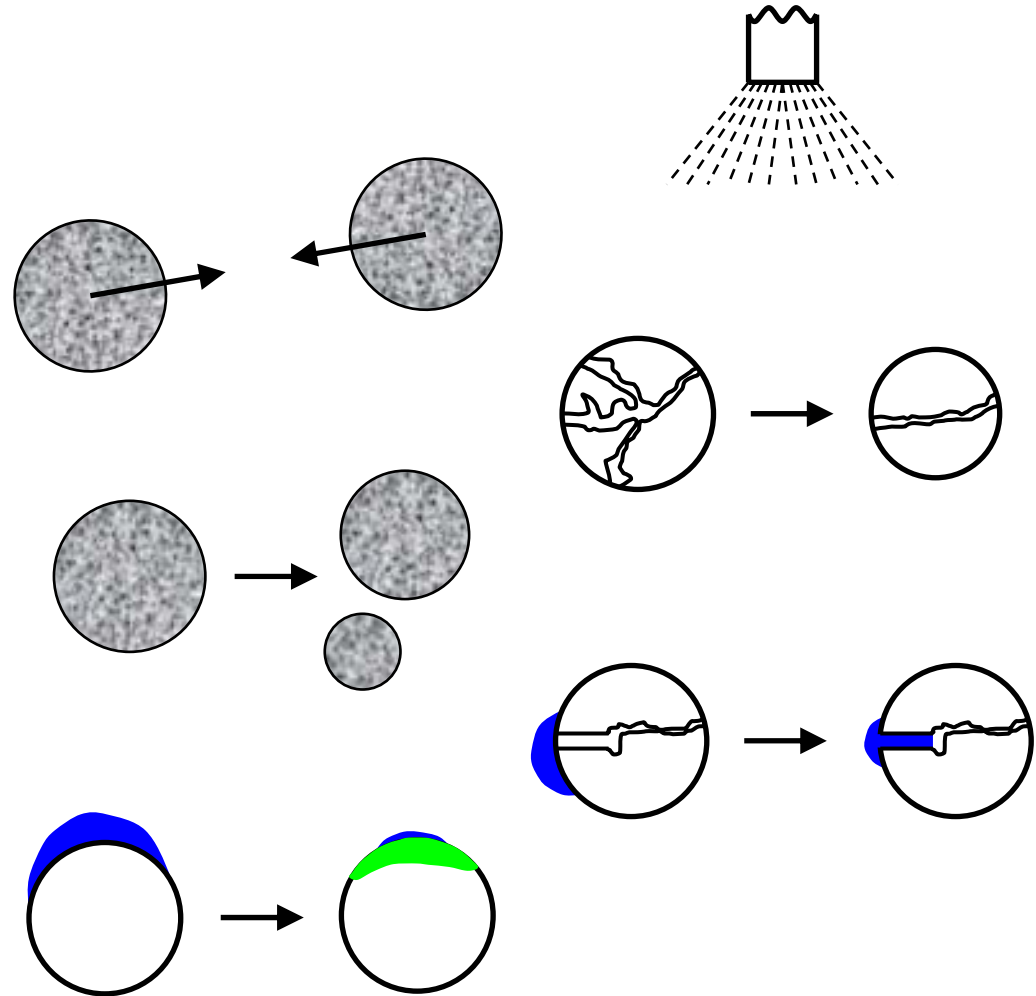
- particle shape: sphere

→ **5 independent variables**  
for particle description



# Model - Transformations

- Addition of binder
- Coalescence
- Compaction
- Breakage
- Penetration
- Reaction



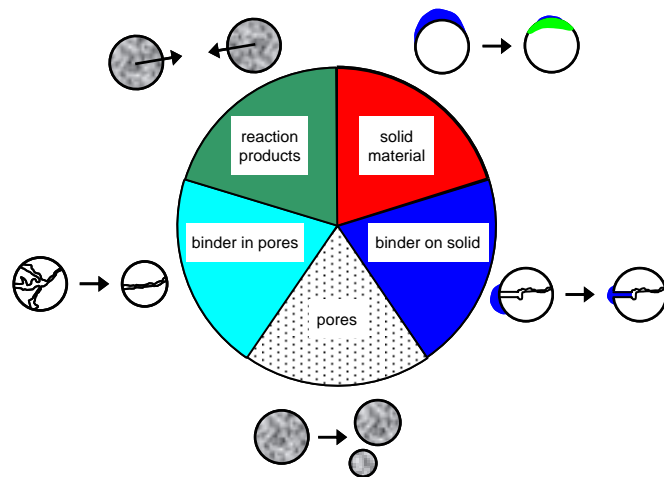
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# Using the model

- So what? – Run simulations!



Not quite!

$x_1 \dots x_K$

S

→ Establish  $x$ 's through exper

solid particles	$s_o$	$m^3$	ge.de/
	$\rho_{s_o}$	$kg/m^3$	
liquid droplets	$V_{droplet} = l_e$	$m^3$	
	$\eta_l$	$Pa \cdot s$	
	$\rho_{l_e}$	$kg/m^3$	
Mixer-granulator operating parameters			
	$n_{impeller}$	$s^{-1}$	
	$r_{impeller}$	$m$	
	$\tilde{u}_{col}$	-	
	$\tilde{u}_{imp}$	-	
Breakage			
	$\hat{k}_{att}$	$s \cdot m^{-5}$	
	$s_r^*$	-	
	$a$	-	
	$b$	-	
	$v_{frag,min}$	$m^3$	
	$v_{max,I}$	-	
	$v_{min,max}$	-	
	$\alpha$	-	
	$\beta$	-	
	$v_{max,II}$	-	
Chemical Reaction			
	$C$	-	
	$k_{react,e}$	$m/s$	
	$k_{react,i}$	$m/s$	
Coalescence			
	$e_{s_o}$	-	
	$e_{s_r}$	-	
	$h_a$	$m$	
	$\hat{K}_0$	$m^3$	
Compaction			
	$k_{porred}$	$s/m$	
	$\varepsilon_{min}$	-	
Penetration			
	$\hat{k}_{pen}$	$kg^{1/2} s^{-3/2} m^{-7/2}$	

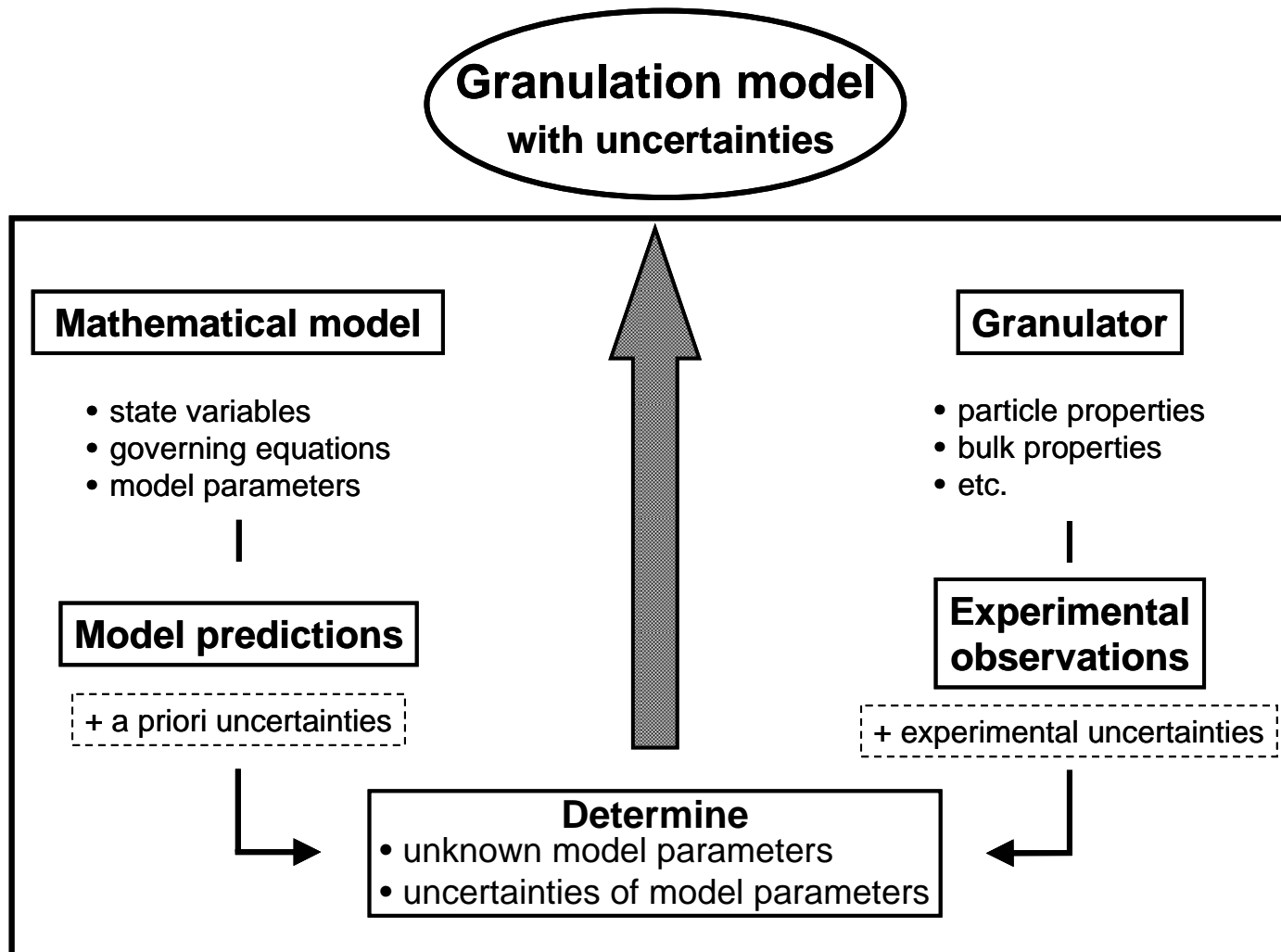


# Unknown parameters

- unknown parameters:
  - coalescence rate constant  $K_0 = ?$
  - compaction rate constant  $k_{\text{comp}} = ?$
  - breakage rate constant  $k_{\text{att}} = ?$
  - reaction rate constant  $k_{\text{reac}} = ?$



# The idea



# Theory

- Experimental data

$$\eta^{\text{exp}} = \eta_0^{\text{exp}} \pm \sigma^{\text{exp}}$$

- Model response with parameters  $x$

$$\eta = \eta(x) \quad \text{with} \quad x = (x_1, \dots, x_K)$$

- and  $x = x_0 + c\xi$  with uncertainty factor  $c$  and standard normally distributed  $\xi$

For a simple linear model,  $K=1$

$$\eta(x) = A + Bx$$

$$\eta(x, c, \xi) = A + B(x_0 + c\xi)$$

$$\mu(x_0) = E[\eta(x_0, c, \xi)] = A + Bx_0 \quad \sigma(c) = \sqrt{\text{Var}(\eta(x_0, c, \xi))} = \sqrt{B^2 c^2}$$





# Theory II

- Optimal values for parameters and associated uncertainties

$$(\mathbf{x}_0^*, \mathbf{c}^*) = \underset{\mathbf{x}_0, \mathbf{c}}{\operatorname{argmin}} \{ \Phi(\mathbf{x}_0, \mathbf{c}) \}$$

- with objective function

$$\Phi(\mathbf{x}_0, \mathbf{c}) = \sum_{i=1}^N \left( [\eta_{0,i}^{\text{exp}} - \mu_i(\mathbf{x}_0, \mathbf{c})]^2 + [\sigma_i^{\text{exp}} - \sigma_i(\mathbf{x}_0, \mathbf{c})]^2 \right)$$

- and constraints

$$\mathbf{x}_{0,k,\text{low}} \leq \mathbf{x}_{0,k} \leq \mathbf{x}_{0,k,\text{up}} \quad (k = 1, \dots, K)$$

$$0 \leq \mathbf{c} \leq \mathbf{c}^{(0)}$$



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# Get on with it

- **Problem:** Running the optimisation on the complex model is computationally expensive

→ find substitute: **Response surfaces**

= approximation of process behaviour

$$\eta(x) = y_{sim}(x) + \varepsilon$$

- How to use?



# Response surfaces

- approximate process (model) behaviour locally
- e.g., 1<sup>st</sup> order response surface

$$\eta(\mathbf{x}) = \beta_0 + \sum_{k=1}^K \beta_k x_k$$

with  $\eta(\mathbf{x}) =$  response surface

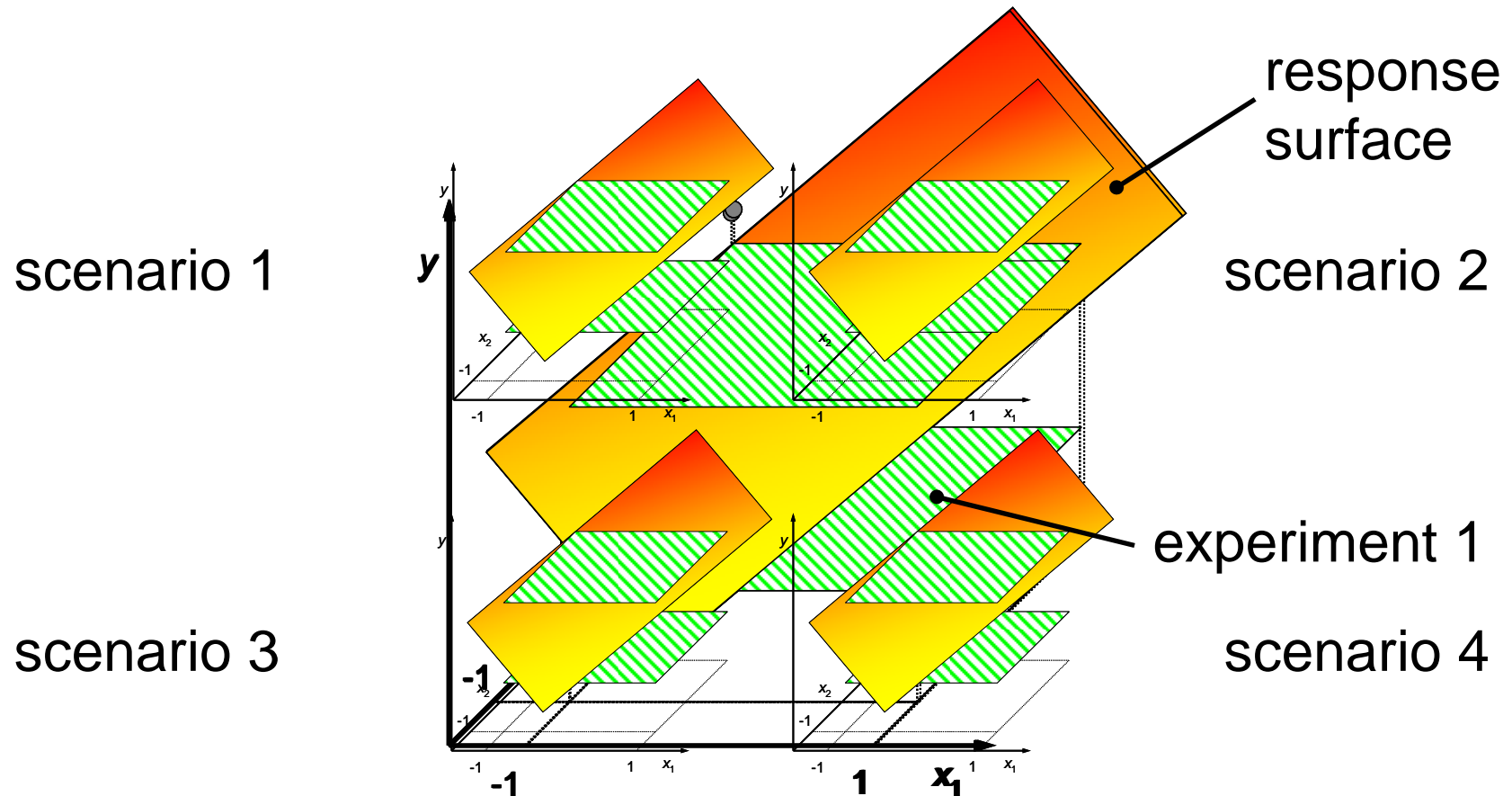
$K =$  number of variables

$\beta_0, \beta_k =$  parameters of surface

- $\beta_k$  : **process sensitivities** with respect to  $x_k$



# Methodology in short



parameters:  $\beta_0, \beta_k, k_{\text{reac}}$

$$\hat{y}(x) = \beta_0 + \sum_{k=1}^K \beta_k x_k$$



# Example

- experimental data from Simmons et al. (2006)
    - bench scale granulation of non-pareils (sugar) with PEG4000/water binder
    - run at 900 and 1200 rpm impeller speed
    - examine amount of agglomerates
    - 4 sampling times
- 8 experimental data sets

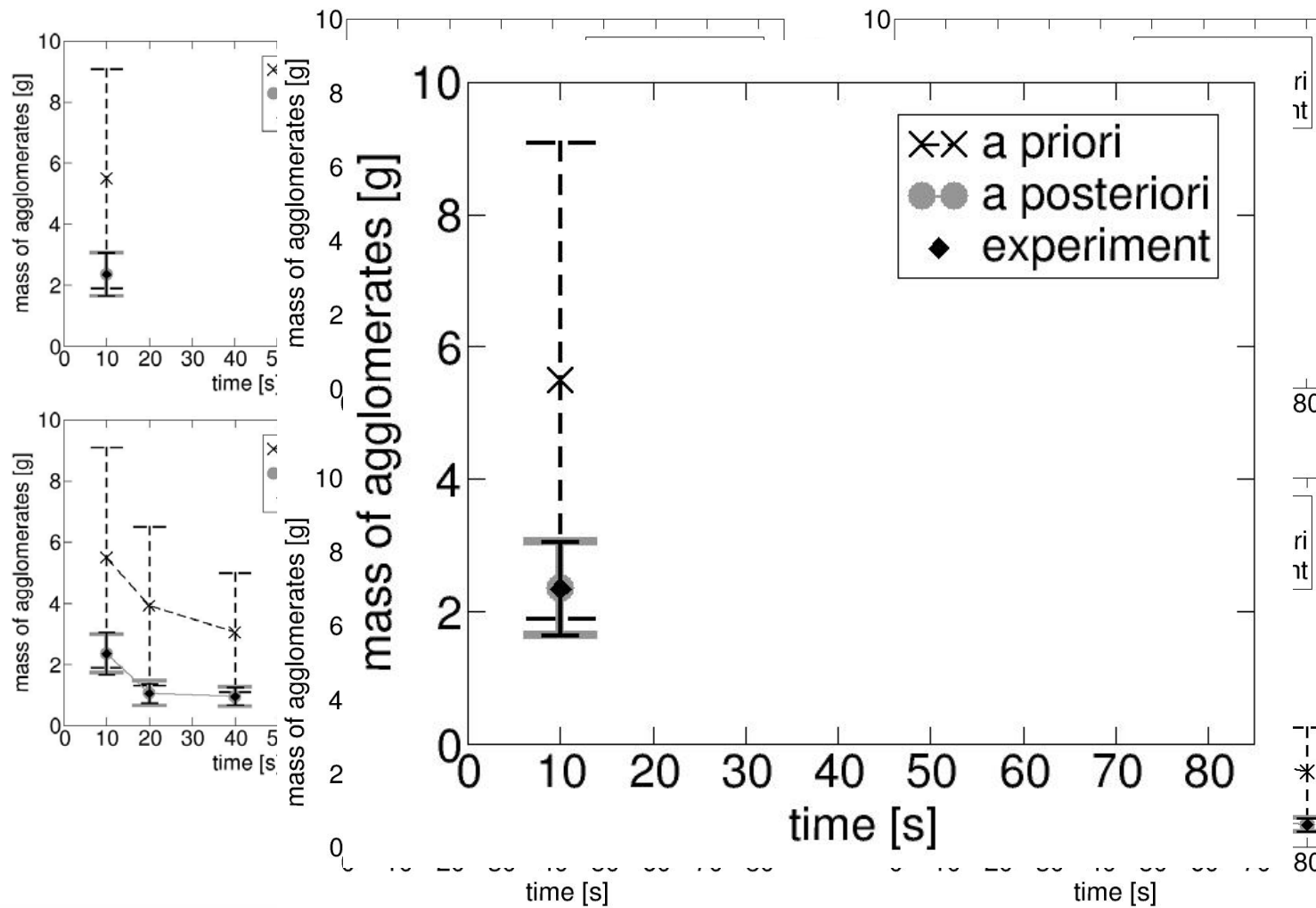


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# Example – 900 rpm data



uncertainties

1)  $\cdot 10^{-10} \text{ m}^3$

2) 4 s/m

$\cdot 10^7 \text{ s/m}^5$

$\cdot 10^{-9} \text{ m/s}$

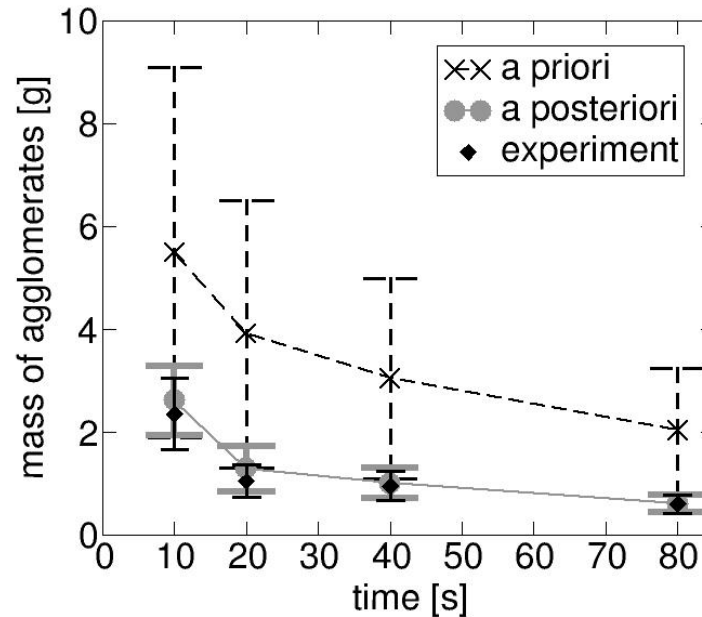


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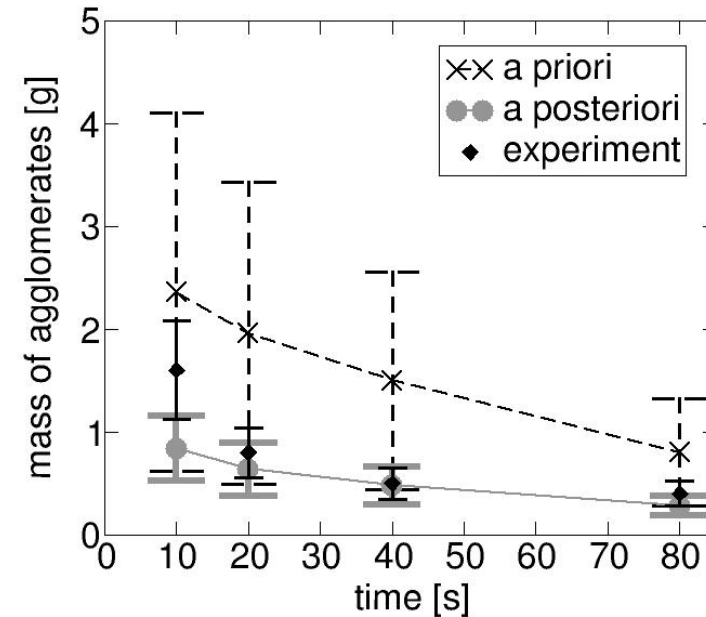
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# Example – all data



900 rpm



1200 rpm

Parameters and uncertainties using all 8 sets

$$K_0 = (1.23 \pm 0.00) \cdot 10^{-10} \text{ m}^3, \quad k_{\text{comp}} = 0.20 \pm 0.05 \text{ s/m}$$
$$k_{\text{att}} = (4.9 \pm 0.0) \cdot 10^7 \text{ s/m}^5, \quad k_{\text{reac}} = (4.000 \pm 0.003) \cdot 10^{-9} \text{ m/s}$$

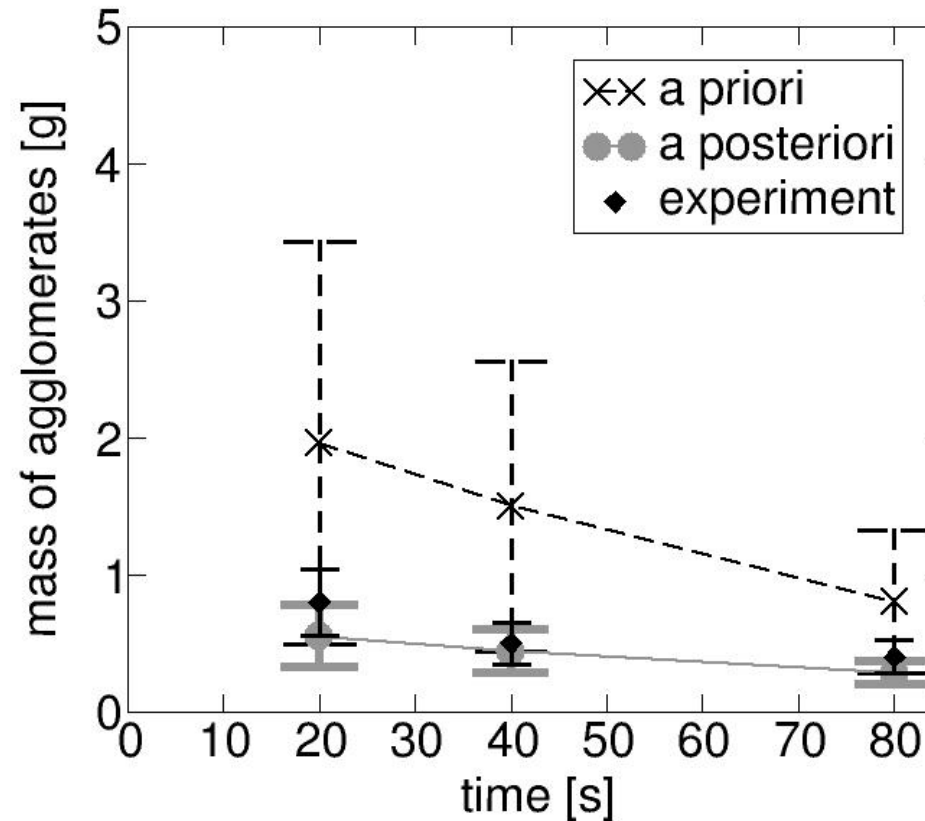


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# All data – a closer look



caution :

→ ~~existence of parameters~~



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# Summary

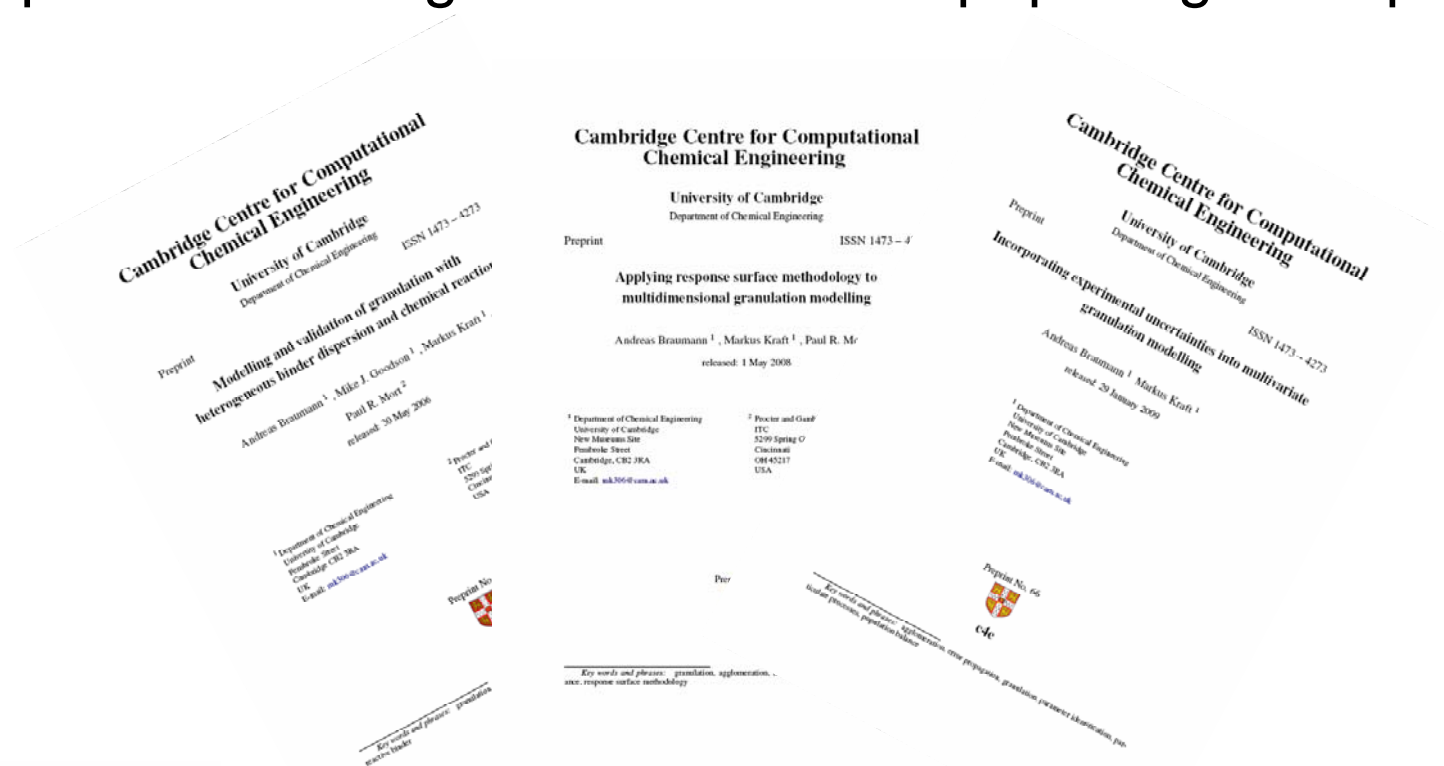
- Population balance model for granulation
  - 5 dimensional particle space
  - several subprocesses
- Approach for parameter and uncertainty estimation using experimental data
- Falsification of models



# Further information

## CoMo Group preprints

<http://como.cheng.cam.ac.uk/index.php?Page=Preprints>



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