

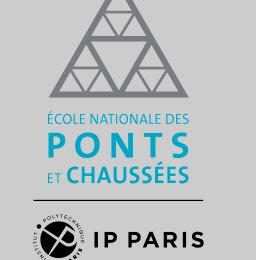


# A Reduced Ordel Model for Aerosol Coagulation

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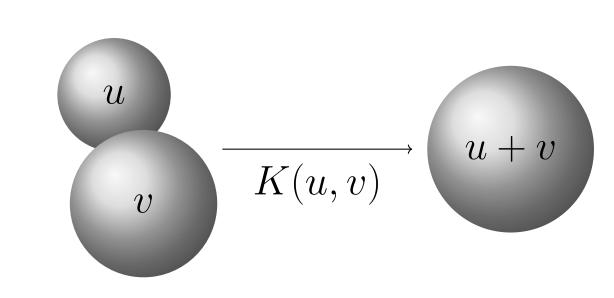
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### Coagulation

Coagulation is a process affecting the size distribution of particles, describing the rate of collisions of particles and the formation of larger particles over time.

It behaves non-linearly and non-locally [Smo16], leading to prohibitive computational demands for high-resolution simulations.



$$\frac{\partial n}{\partial t}(v,t) = \frac{1}{2} \int_0^v K(u,v-u) \ n(u,t) \ n(v-u,t) \, \mathrm{d}u$$

$$- n(v,t) \int_0^\infty K(v,u) \ n(u,t) \, \mathrm{d}u$$
(1)

## Relevance to many domains of physics

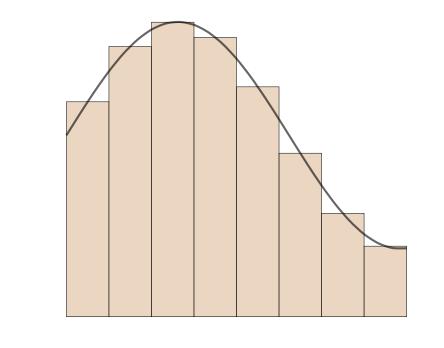
Coagulation is involved in many types of interacting particle systems, such as

- Atmospheric aerosol and air quality [SP16]
- Cloud droplets [CS98]
- Polymerization in colloidal systems [Ada95]
- Protoplanet accretion [Cha43]

## Full order model

Let  $E = Span\{\phi_1, \dots, \phi_N\}$  a reference approximation subspace, for instance a finite element approximation supported on a fine mesh.

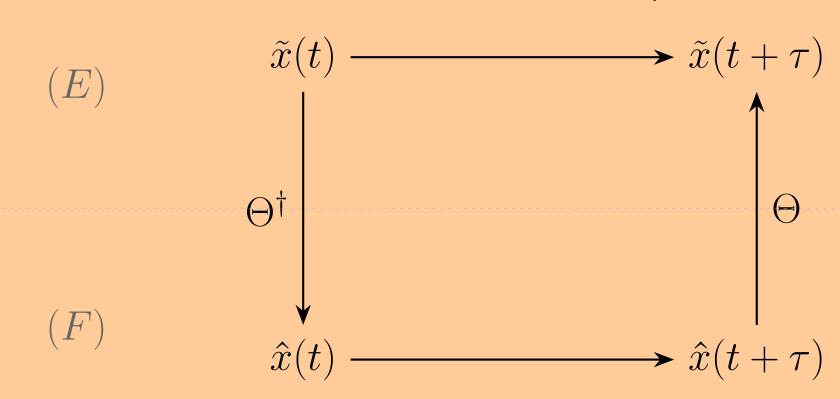
Let  $\tilde{n}(v,t) = \sum_{i=1}^{\infty} \phi_i(v)\tilde{x}_i(t)$  the approximation of (1) in E. We are interested in the solution set  $\mathcal{S}$  corresponding to a given a set of initial conditions  $\mathcal{X}_0$ 



$$\mathcal{S} = \{ \tilde{n}(v,t) \mid \tilde{n}(v,0) \in \mathcal{X}_0 \}$$
 (2)

#### Model order reduction

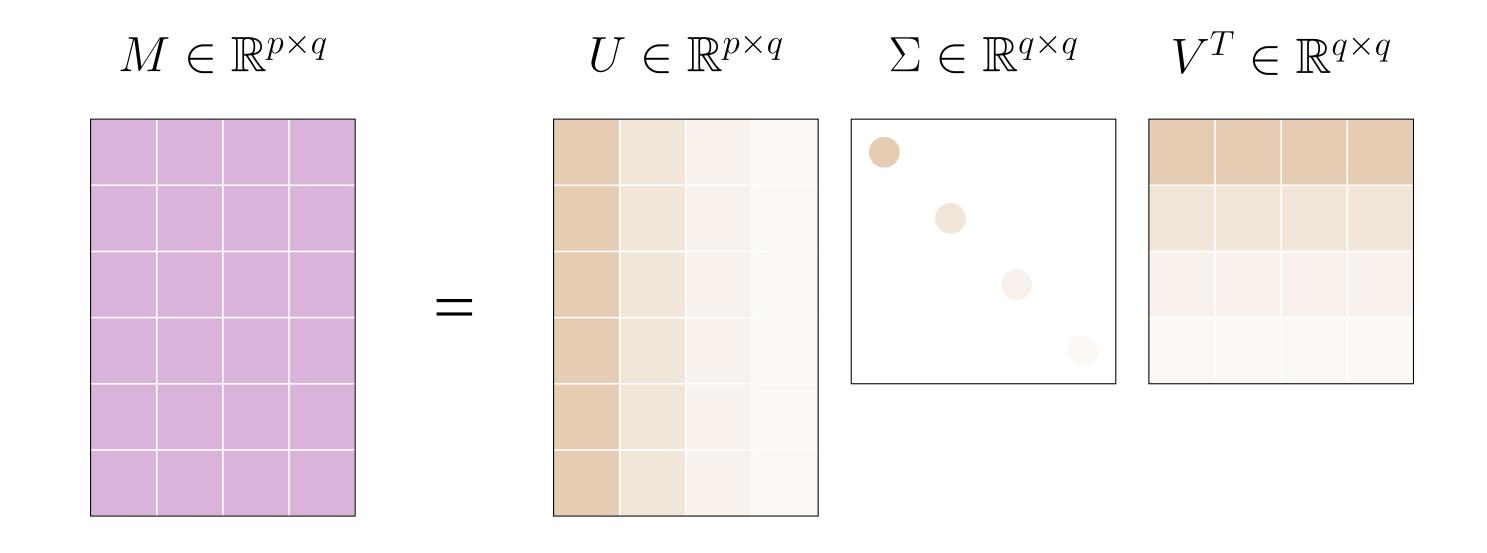
Our goal is to find a subspace F with reduced dimension such that S is well approximated, but time-evolution in F can be performed at a lower cost.



Let  $F = Span\{\varphi_1, \dots, \varphi_R\} \subset E$  with  $\varphi = \phi\Theta$ , with  $\Theta \in \mathbb{R}^N$ . We are looking for the optimal  $\Theta$  such that  $\|\tilde{n} - \hat{n}\|_2$  is minimized.

### Identification of a suitable subspace

Singular Value Decomposition (SVD):



The truncated SVD up to rank R provides the optimal R dimensional approximation in a least square sense.

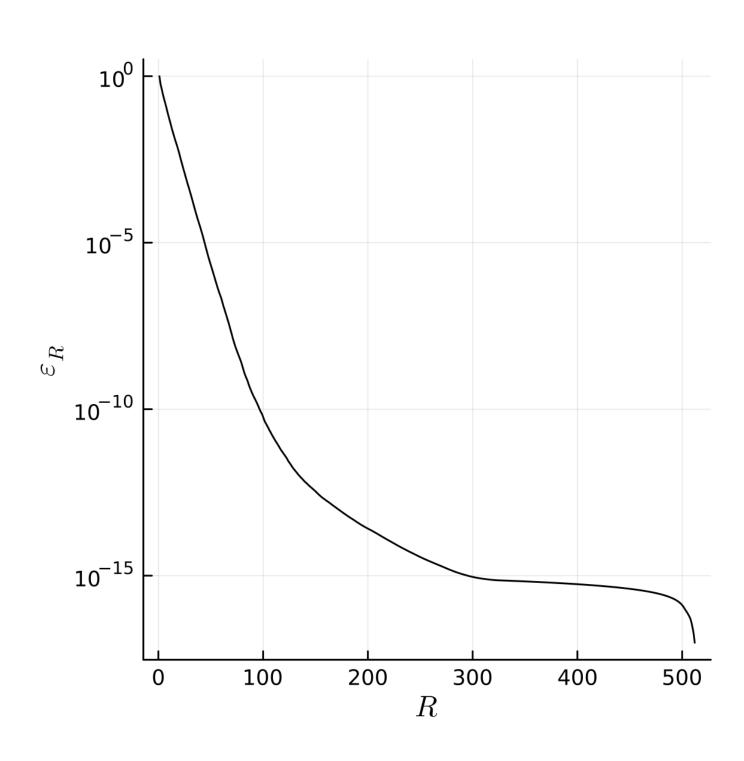


Figure 1.  $L^2$  error associated with rank-R approximation of S.

# Reduction of the time-evolution operator

The time-evolution operator for the ROM can be precomputed offline such that the *online* cost scales only with R, independently of N.

Full order model 
$$\left[\frac{\partial \tilde{x}}{\partial t}\right]_{i} = \sum_{j,k=1}^{N} \tilde{\chi}_{ijk} \, \tilde{x}_{j} \tilde{x}_{k} \tag{3}$$

Reduced order model  $\left[\frac{\partial \hat{x}}{\partial t}\right]_{\cdot} = \sum_{i,j=1}^{R} \hat{\chi}_{ijk} \hat{x}_{j} \hat{x}_{k}$ 

with precomputed tensors

$$\tilde{\chi}_{ijk} = \sum_{i'=1}^{N} \left[ g_{\phi}^{-1} \right]_{i,i'} \langle \phi_{i'}, \mathcal{B}[\phi_{j}, \phi_{k}] \rangle \qquad \tilde{\chi} \in \mathbb{R}^{N \times N \times N}$$

$$\hat{\chi}_{iik} = \sum_{i'=1}^{R} \left[ g_{\phi}^{-1} \right] \quad \langle \varphi_{i'}, \mathcal{B}[\varphi_{i}, \varphi_{k}] \rangle \qquad \hat{\chi} \in \mathbb{R}^{R \times R \times R}$$

#### **Mass-conservation**

Mass conservation can be casted as an orthogonality condition:

$$M(t) = \int_{0}^{\infty} v \, n(v, t) \, \mathrm{d}v \tag{7}$$

$$\frac{dM}{dt} = \int_{0}^{\infty} v \frac{\partial n}{\partial t}(v, t) \, dv = \left\langle v, \frac{\partial n}{\partial t} \right\rangle = 0 \tag{8}$$

A sufficient condition for mass conservation is that  $v \mapsto v \in F$ . Classical reduction does not lead to a mass conservative ROM. However, it can always be enforced from a non conservative one, at the cost of one extra degree of freedom.

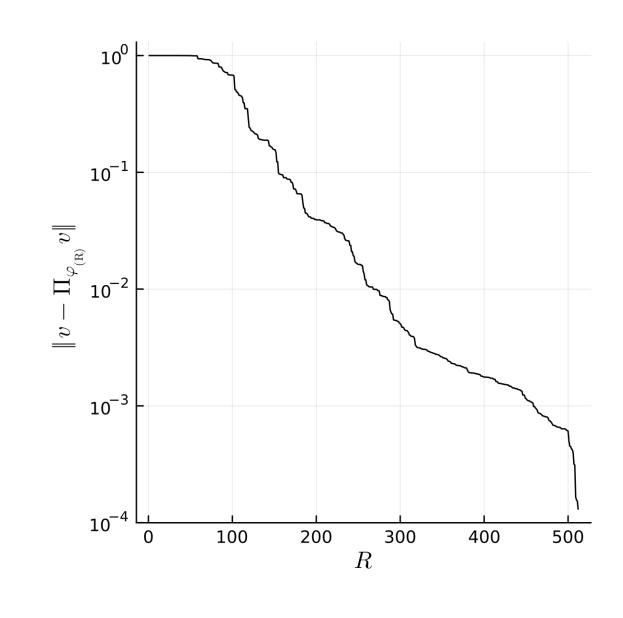


Figure 2. Projection error associated to mass imbalance

# Certification through a posteriori error estimates

Quality of the ROM outputs can be certified by using *a posteriori* error estimates [HRS16]. The norm of the residual between the dynamics projected on E and F can be computed from the knowledge of  $\hat{x}$  only, at a cost independent of N.

$$\hat{r}[\hat{n}](v) = \Pi_E \mathcal{B}[\hat{n}, \hat{n}](v) - \Pi_F \mathcal{B}[\hat{n}, \hat{n}](v)$$

$$R = \|\hat{r}\| = \left(\sum_{i,j,k,l=1}^{R} \Delta_{ijkl} \,\hat{x}_i \hat{x}_j \hat{x}_k \hat{x}_l\right)^{1/2} \tag{10}$$

#### References

[Ada95] Y Adachi. Dynamic aspects of coagulation and flocculation. Advances in Colloid and Interface Science, 56:1-31, March 1995. [Cha43] S. Chandrasekhar. Stochastic Problems in Physics and Astronomy Reviews of Modern Physics, 15(1):1–89, January 1943. S. Cueille and C. Sire. Droplet nucleation and smoluchowski's equation with growth and injection of particles. Physical Review E, 881, 1998. [HRS16] Jan S Hesthaven, Gianluigi Rozza, and Benjamin Stamm Certified Reduced Basis Methods for Parametrized Partial Differential Equations. SpringerBriefs in Mathematics. Springer International Publishing, Cham, 2016. [Smo16] M. V. Smoluchowski. Drei vortage uber diffusion, brownsche bewegung und koagulation von kolloidteilchen. Physik, 17:557-585, 1916. John H. Seinfeld and Spyros N. Pandis.

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