

# Rates of convergence for Smoluchowski's coagulation equation

Ravi Srinivasan  
Duke Univ./Univ. of Texas at Austin  
Dec. 07, 2009

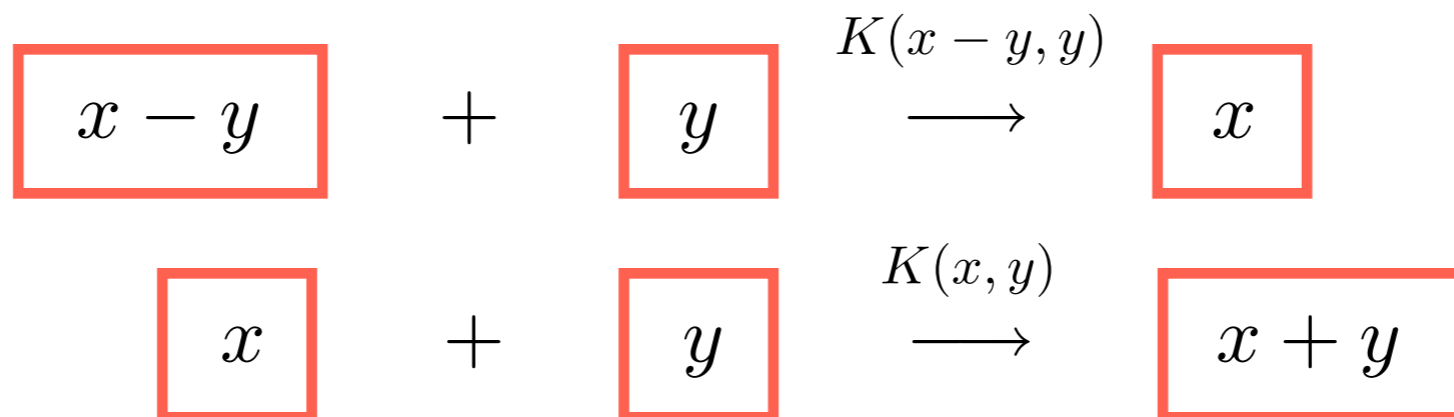
- R. Srinivasan, <http://arxiv.org/abs/0905.3450> (2009)

## Smoluchowski's coagulation equation

$n(t, x)$  = number density of clusters of size  $x > 0$  at time  $t \geq 0$

$K(x, y) = K(y, x)$  symmetric rate kernel

$$\partial_t n(t, x) = \frac{1}{2} \int_0^x K(x-y, y) n(t, x-y) n(t, y) dy - \int_0^\infty K(x, y) n(t, x) n(t, y) dy$$



Smoluchowski's equation has been used as a mean-field model for a variety of agglomeration phenomena:

- coagulation of colloids
- formation of clouds and smog
- kinetics of polymerization
- mass aggregation in astrophysics
- schooling of fishes
- merging of banks
- random graph theory
- ballistic aggregation of shocks in Burgers turbulence ( $K=x+y$ )

More generally, consider measure-valued solutions under weak formulation (moment identity), with suitable test functions  $\phi$ :

$n(t, dx)$  = number measure of clusters of size in  $[x, x + dx)$

$$\begin{aligned} & \partial_t \int_{(0, \infty)} \phi(x) n(t, dx) \\ &= \frac{1}{2} \int_{(0, \infty)} \int_{(0, \infty)} (\phi(x+y) - \phi(y) - \phi(x)) K(x, y) n(t, dy) n(t, dx) \end{aligned}$$

Many recent studies have focused on the homogeneous, “solvable” rate kernels

$$K(\alpha x, \alpha y) = \alpha^\gamma K(x, y), \forall \alpha > 0$$

$$K = 2, x + y, xy$$

$$\gamma = 0, 1, 2$$

- For these kernels one can obtain exact solutions via the Laplace transform

dynamic scaling



Smoluchowski's  
coagulation equation

$$n(t, dx)$$

classical limit theorems  
in probability

$$S_n = X_1 + \dots + X_n$$

(i) Well-posedness of dynamical system on space of prob. measures

- Menon & Pego (CPAM, 2004)

- Well-posedness for general homogeneous K:

Fournier & Laurençot, Escobedo; Mischler & Rodriguez Ricard

(ii) Existence of  $\Gamma$ -parameter family of self-similar solutions, domains of attraction:

- Menon & Pego (CPAM, 2004)

SSS w/exp. decay



Gaussian

SSS w/power law decay

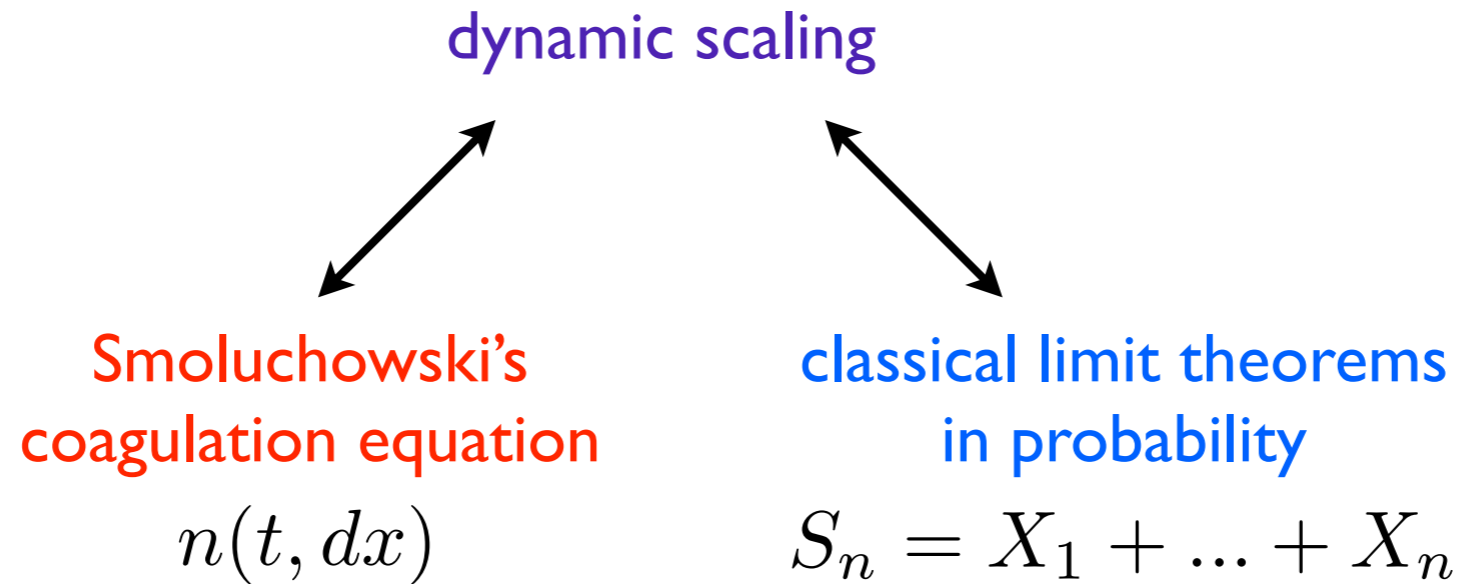


Lévy stable laws

convergence to SSS,  
regular variation



central limit  
theorem



(iv) *Interplay between moment hypothesis on the initial data and stronger modes of convergence*

- Menon & Pego (SIAM Review, 2006): For SSS with exp. decay, uniform convergence of densities under dynamic scaling

There are more correspondences, which we do not discuss here:

(v) *Attractor of the dynamical system modulo scaling*

- Menon & Pego (2008)

eternal solutions



infinitely divisible  
distributions

## Continuing the analogy...

rates of convergence to  
SSS with exponential  
tail



Berry-Esséen theorem  
for convergence to  
Gaussian

$$\mathbb{E}X_i = 0, \mathbb{E}X_i^2 = 1, \mathbb{E}X_i^3 = \rho < \infty$$

$$F_n(x) = \mathbb{P} \left( \frac{X_1 + \dots + X_n}{\sqrt{n}} \leq x \right)$$



$$\sup_{x \in \mathbb{R}} |F_n(x) - \mathcal{N}(x)| \leq \frac{3\rho}{\sqrt{n}}$$



## Continuing the analogy...

rates of convergence to  
SSS with exponential  
tail

?



Berry-Esséen theorem  
for convergence to  
Gaussian

$$\mathbb{E}X_i = 0, \mathbb{E}X_i^2 = 1, \mathbb{E}X_i^3 = \rho < \infty$$

$$F_n(x) = \mathbb{P} \left( \frac{X_1 + \dots + X_n}{\sqrt{n}} \leq x \right)$$



$$\sup_{x \in \mathbb{R}} |F_n(x) - \mathcal{N}(x)| \leq \frac{3\rho}{\sqrt{n}}$$

- Initial time  $t_0 = 1$  ( $\gamma = 0$ )  
 $t_0 = 0$  ( $\gamma = 1, 2$ )

$$\int_{(0, \infty)} x^\gamma n(t_0, dx) = 1 < \infty \quad \longrightarrow \quad \text{well-posedness}$$

- Dynamic scaling

$$m_\gamma(t) = \int_{(0, \infty)} x^\gamma n(t, dx)$$

$$m_0(t) = t^{-1}, \quad m_1(t) = 1, \quad m_2(t) = (1 - t)^{-1}$$

$$\lambda_0(t) = t, \quad \lambda_1(t) = e^{2t}, \quad \lambda_2(t) = (1 - t)^{-2}$$

- Self-similar solution with exponential tail

$$n(t, dx) = \frac{m_\gamma(t)}{\lambda_\gamma(t)^\gamma} \hat{n}_{*,\gamma}(d\hat{x}) \quad \hat{x} = \frac{x}{\lambda_\gamma(t)}$$

$$\hat{n}_{*,0}(d\hat{x}) = e^{-\hat{x}} d\hat{x} \quad \hat{x} \hat{n}_{*,1}(\hat{x}) = \hat{x}^2 \hat{n}_{*,2}(\hat{x}) = \frac{1}{\sqrt{2\pi}} \hat{x}^{-1/2} e^{-\hat{x}/2} d\hat{x}$$

$$\int_{(0, \infty)} x^{\gamma+1} n(t_0, dx) = 1 < \infty \quad \longrightarrow \quad \text{in domain of attraction of SSS with exponential tail}$$

- Time change

$$\tau_\gamma(t) = \int_{t_0}^t m_\gamma(s) ds$$

$$\tau_0(t) = \log(t), \quad \tau_1(t) = t, \quad \tau_2(t) = \log(1 - t)^{-1}$$

- Rescaled solution converges to SSS (weak convergence of measures)

$$\hat{n}(\tau_\gamma, d\hat{x}) := \frac{\lambda_\gamma(t)^\gamma}{m_\gamma(t)} n(t, \lambda_\gamma(t) d\hat{x})$$

$$\hat{n}(\tau_\gamma, d\hat{x}) \xrightarrow{\tau_\gamma \rightarrow \infty} \hat{n}_{*, \gamma}(d\hat{x})$$

$$\int_{(0,\infty)} x^{\gamma+2} n(t_0, dx) := \mu_{\gamma+2} < \infty \longrightarrow \text{convergence rate}$$

- Exponential convergence in terms of distribution functions

$$F_{\gamma}(\tau_{\gamma}, \hat{x}) = \int_{(0,\hat{x})} \hat{y}^{\gamma} \hat{n}(\tau_{\gamma}, d\hat{y}), \quad F_{*,\gamma}(\hat{x}) = \int_{(0,\hat{x})} \hat{y}^{\gamma} \hat{n}_{*,\gamma}(d\hat{y})$$

**Theorem (Srinivasan, 2009):** For any  $\tau_{\gamma}(t) \in [0, \infty)$

$$\sup_{\hat{x} > 0} |F_{\gamma}(\tau_{\gamma}, \hat{x}) - F_{*,\gamma}(\hat{x})| \leq C(\mu_{\gamma+2})(1 + \tau_{\gamma})e^{-\tau_{\gamma}}$$

- Holds for a broad class of initial data with minimal assumptions (existence of an additional higher moment)

- Near optimal: For monodisperse initial data  $n_0(dx) = \delta_1(dx)$ ,

$$\sup_{\hat{x} > 0} |F_\gamma(\tau_\gamma, \hat{x}) - F_{*,\gamma}(\hat{x})| = O(e^{-\tau_\gamma})$$

- For  $K=2$ : Cañizo, Mischer, Mouhot (2008) showed exponential convergence to SSS in a weighted Sobolev norm for initial densities satisfying decay assumptions on its derivatives, also using Fourier methods

## Main ingredients of proof for $K=2$ :

- Consider Fourier-Laplace representation of number measure

$$s \in \bar{\mathbb{C}}_+ := \{z \in \mathbb{C} : \operatorname{Re}(z) \geq 0\}$$

$$u(\tau, s) = \int_{(0, \infty)} e^{-s\hat{x}} \hat{n}(\tau, d\hat{x}), \quad u_*(s) = \int_{(0, \infty)} e^{-s\hat{x}} \hat{n}_*(d\hat{x})$$

- Smoothing argument (Feller, Vol. II)

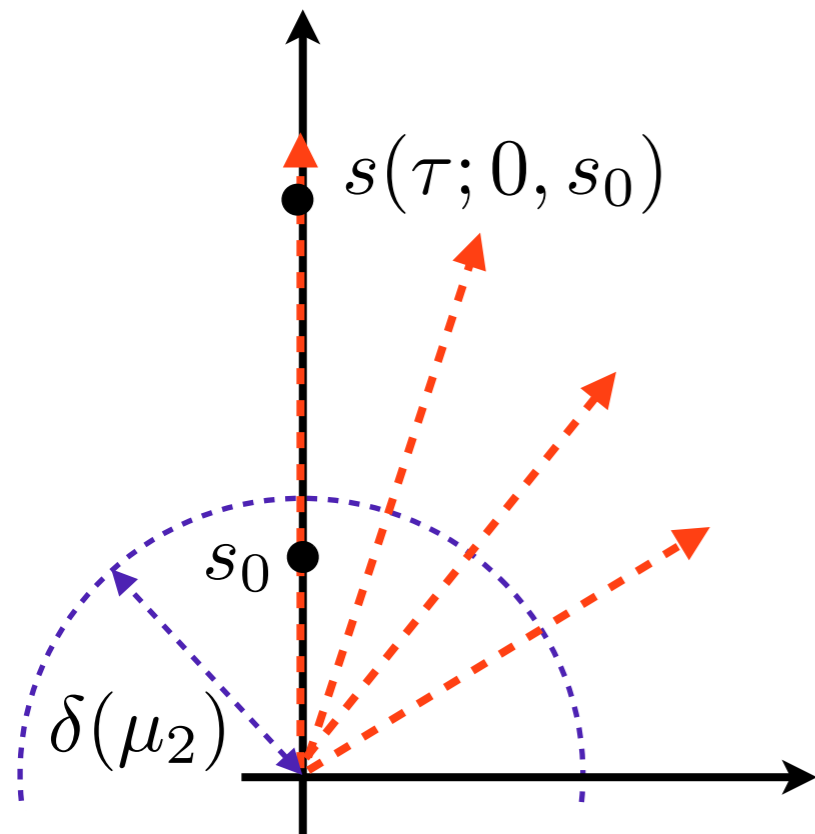
$$\sup_{\hat{x} > 0} |F(\tau, \hat{x}) - F_*(\hat{x})| \leq \frac{1}{\pi} \sup_{\hat{x} > 0} \left| \int_{-iT}^{iT} \frac{e^{s\hat{x}}}{s} (u(\tau, s) - u_*(s)) ds \right| + \frac{24}{\pi T}$$

↑  
main term to  
estimate

(i) Smoluchowski's equation  $\longleftrightarrow$  PDE for Laplace transform

$$\partial_\tau u + s \partial_s u = -u(1 - u)$$

For  $K=2$ , characteristics  $s(\tau; 0, s_0) = e^\tau s_0$  do not depend on initial data.



$$u(\tau, s) = e^{-\tau} \frac{u_0(s_0)}{1 - u_0(s_0)(1 - e^{-\tau})}$$

$$u_*(s) = e^{-\tau} \frac{u_*(s_0)}{1 - u_*(s_0)(1 - e^{-\tau})}$$

Main term with  $T = \delta e^\tau$ :

$$\left| \int_{-i\delta}^{i\delta} \frac{e^{s(\tau; 0, s_0) \hat{x}}}{s(\tau; 0, s_0)} \frac{(u_0(s_0) - u_*(s_0))}{(1 - u_*(s_0)(1 - e^{-\tau}))(1 - u_0(s_0)(1 - e^{-\tau}))} ds_0 \right|$$

(ii) Moment hypothesis (decay of tails of initial data) gives approximation for difference of Laplace transforms near origin

$$\begin{aligned}u_0(0) &= -u'_0(0) = 1, & u''_0(0) &= \mu_2 \\u_*(0) &= -u'_*(0) = 1, & u''_*(0) &= 2\end{aligned}$$

$$|u_0(s_0) - u_*(s_0)| \leq \left(1 + \frac{\mu_2}{2}\right) |s_0|^2$$

This approximation is good in the region  $|s_0| \leq \delta(\mu_2)$  with

$$2\delta(\mu_2) = \sqrt{1 + 2\left(1 + \frac{\mu_2}{2}\right)^{-1}} - 1$$

(iii) Plug in estimates—main contribution:

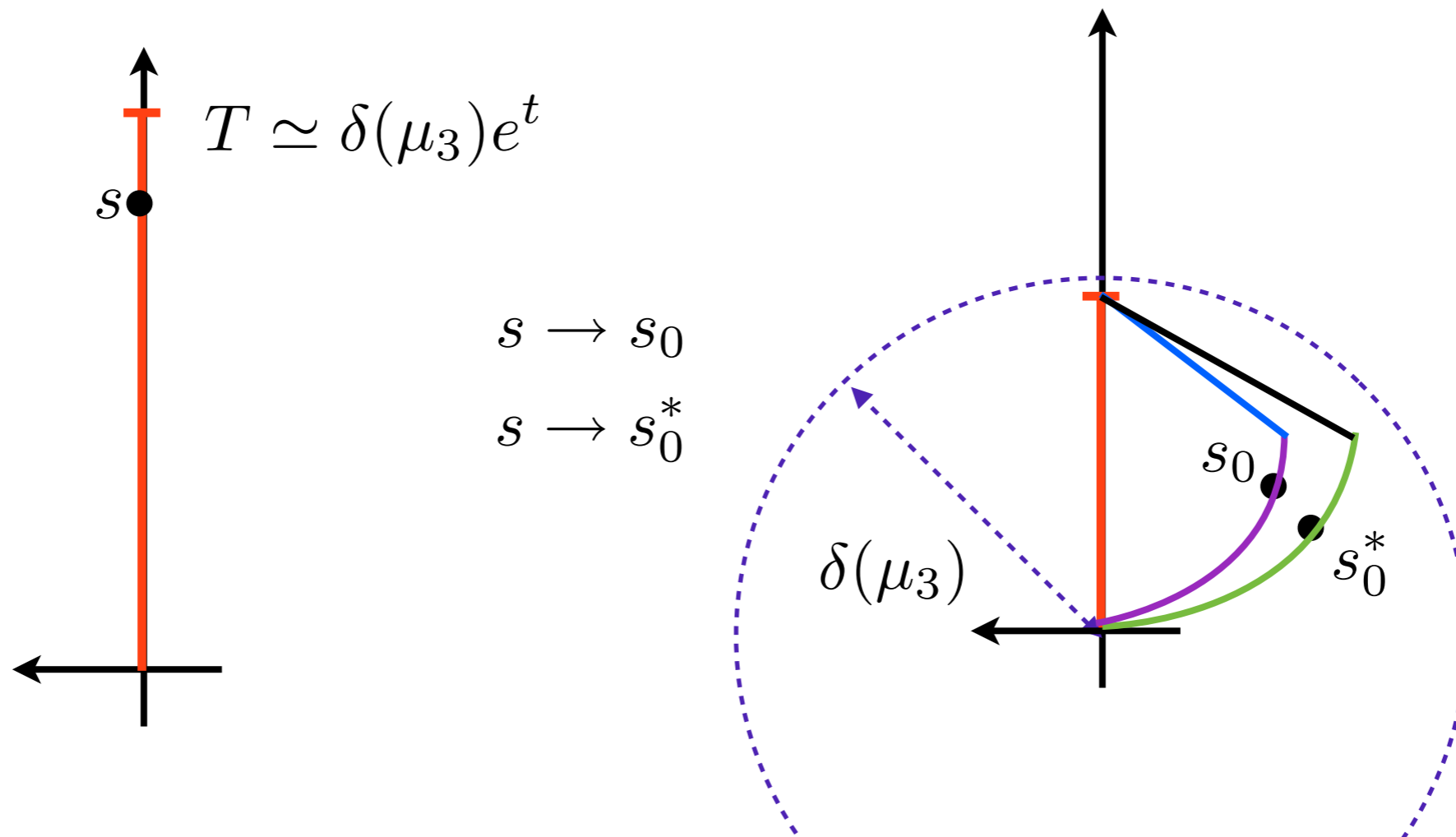
$$\begin{aligned}\left| \int_{\delta e^{-\tau}}^{\delta} \right| &\leq 2\left(1 + \frac{\mu_2}{2}\right) (1 + \delta)^2 e^{-\tau} \int_{\delta e^{-\tau}}^{\delta} \frac{1}{|s_0|} d|s_0| \\ &= 2\left(1 + \frac{\mu_2}{2}\right) (1 + \delta)^2 \tau e^{-\tau} \quad \text{done!}\end{aligned}$$



## What about $K=x+y$ ?

- Idea of proof is same as for  $K=2$ . But characteristics depend on initial data and are no longer rays, but curves in the complex plane. We use a contour deformation argument as in Menon-Pego (2006):

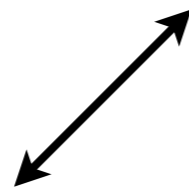
$$\begin{aligned} \varphi(t, s) &= \int_{(0, \infty)} (1 - e^{-s\hat{x}}) \hat{n}(t, d\hat{x}) \\ u(t, s) &= \int_{(0, \infty)} e^{-s\hat{x}} \hat{x} \hat{n}(t, d\hat{x}) \end{aligned} \longrightarrow \begin{aligned} \partial_t \varphi + (2s - \varphi) \partial_s \varphi &= \varphi \\ \partial_t u + (2s - \varphi) \partial_s u &= -u(1 - u) \end{aligned}$$



## What about $K=xy$ ?

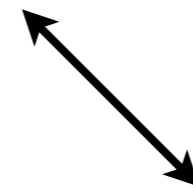
- Solutions can be obtained from the case  $K=x+y$  by a well-known change of variables given in Drake (1972). We therefore get the convergence rate for  $K=xy$  for free.

dynamic scaling



Smoluchowski's  
coagulation equation

$$n(t, dx)$$



classical limit theorems  
in probability

$$S_n = X_1 + \dots + X_n$$

Many thanks to Govind Menon (Brown) & Bob Pego (CMU)