



Cannonballs in (dusty) turbulence

Drag, Accretion & Concentrations

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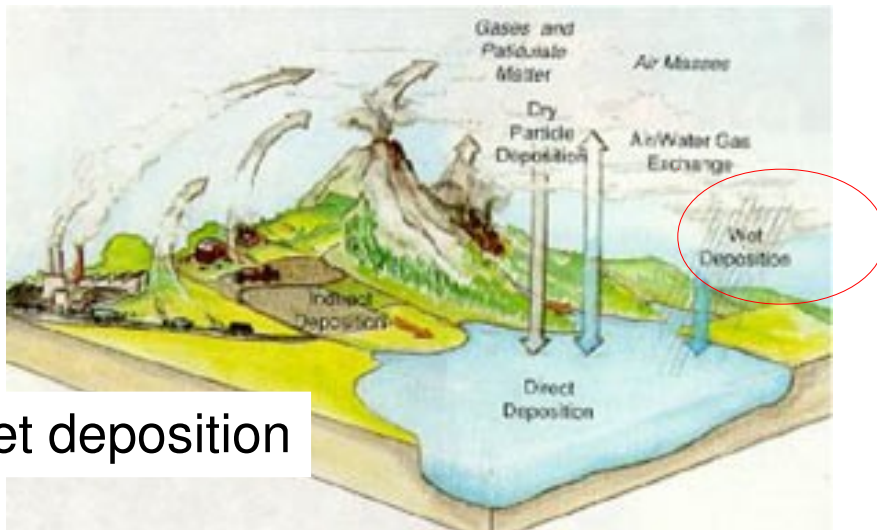
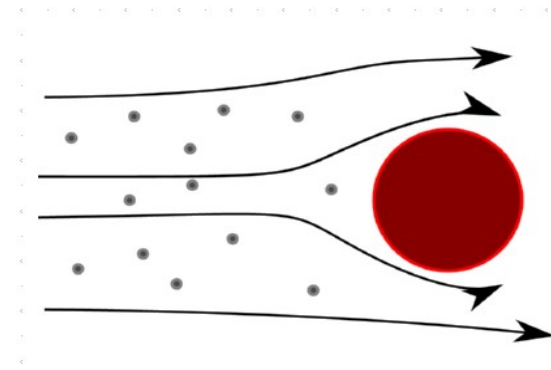
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Motivation - Applications

CONTEXT: Large particle moving through a (turbulent) fluid seeded with small inertial particles:



Wet deposition



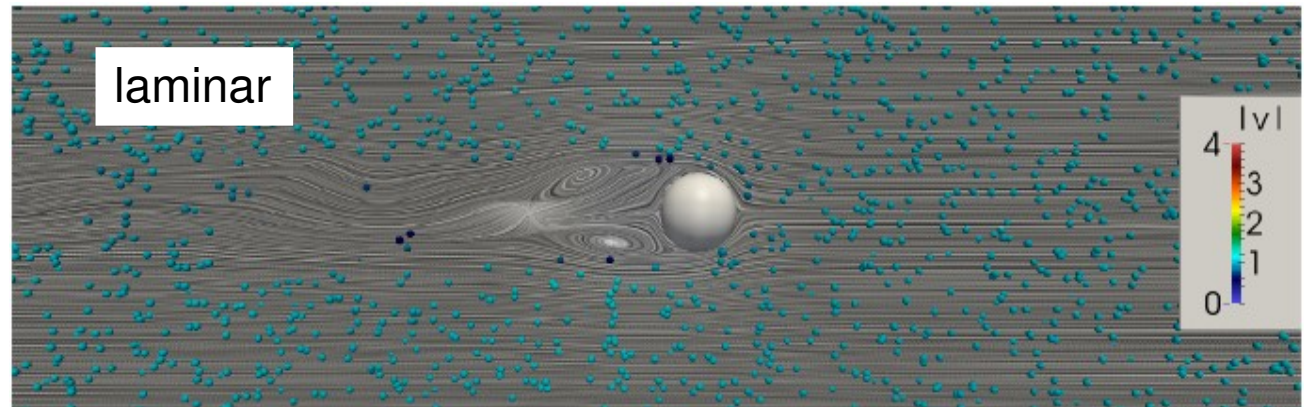
Sekiya and Takeda (2003)

Collision rates of small particles with big body

Motivation – Physical problem

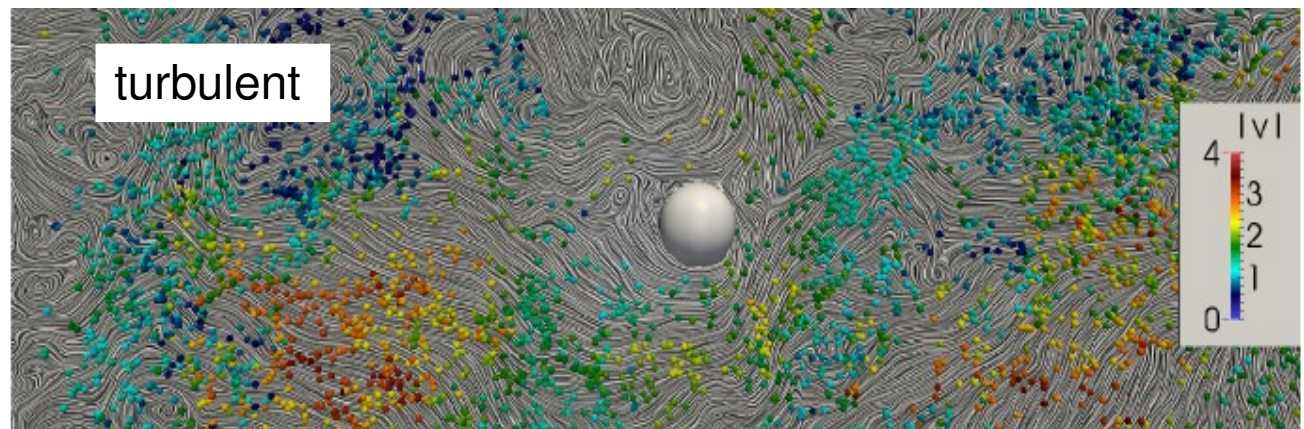
Interaction of big body with stream of small inertial particles

- Small particle collision rates
- Wake interactions

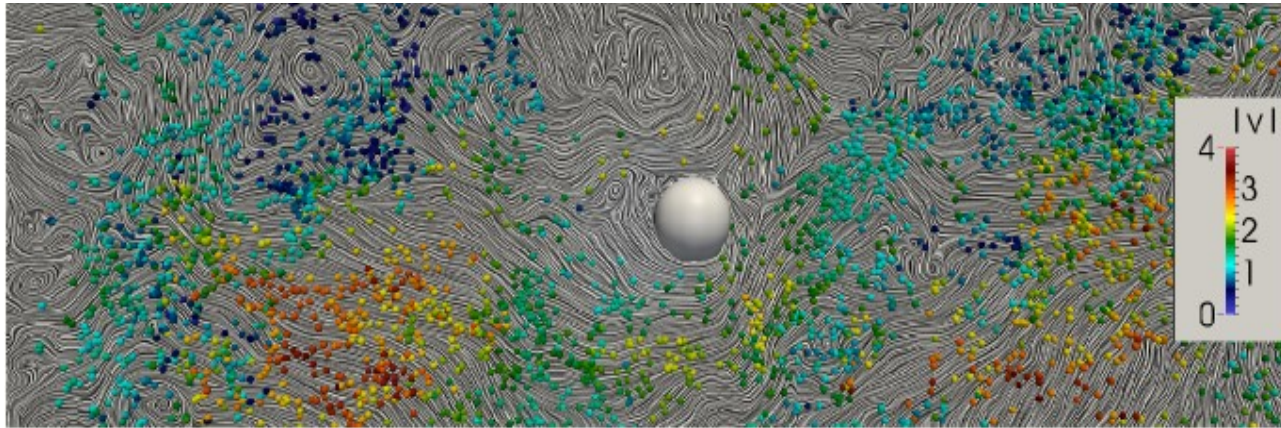


Influence of turbulent velocity fluctuations

- Drag force
- Wake modification
- Collision rates



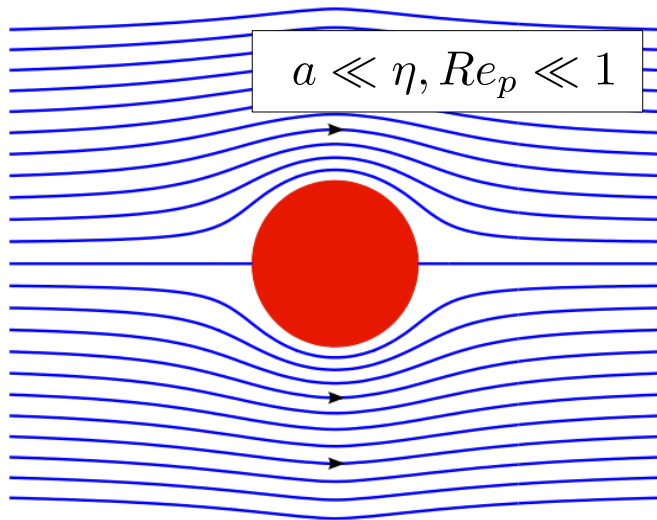
Numerical challenges



What do we need?

- Equations and solver for fluid motion
No-slip boundary condition on big particle
High Reynolds numbers
- Equations for small particle motion
Integration of many trajectories for statistics

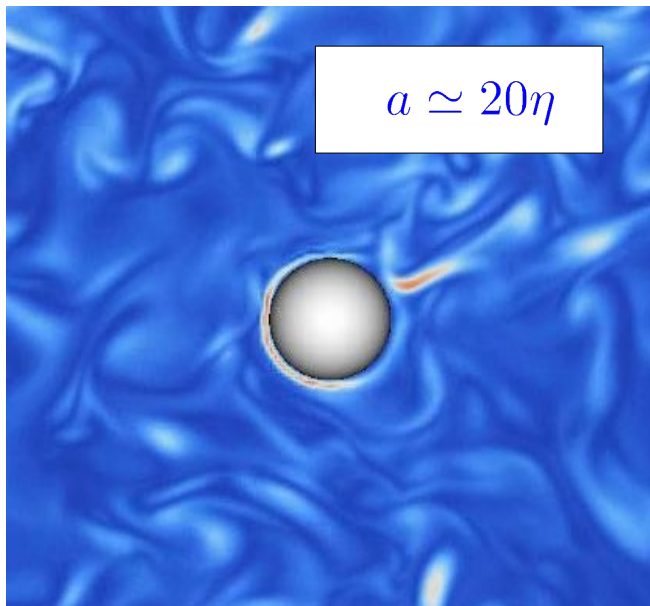
Small vs large particles



Analytically known drag force and flow structure

Stokes drag
$$\frac{d\mathbf{v}}{dt} = \frac{3\nu}{a^2}(\mathbf{u} - \mathbf{v})$$

$\mathbf{u}$ known $\rightarrow$ ODE for small particle trajectories



What if particle the particle is bigger?

$\rightarrow$ handle the boundary layer!

The numerical problem

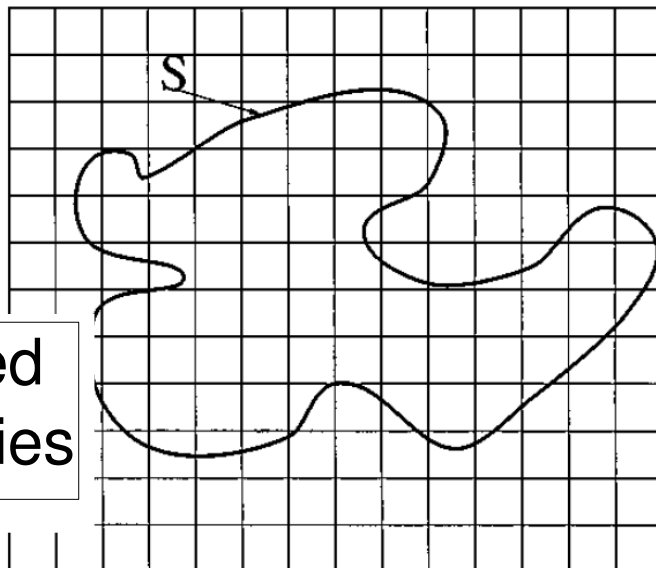
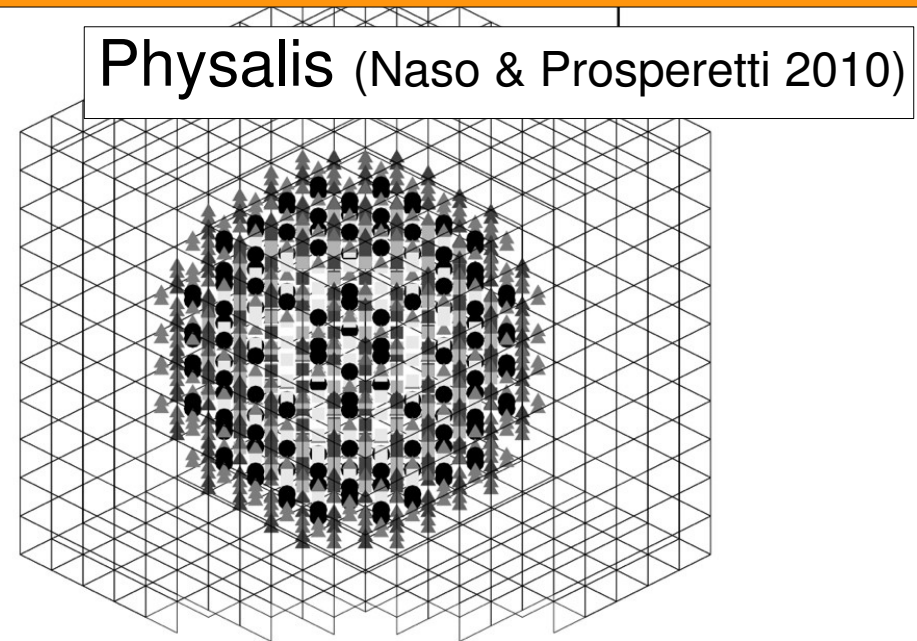
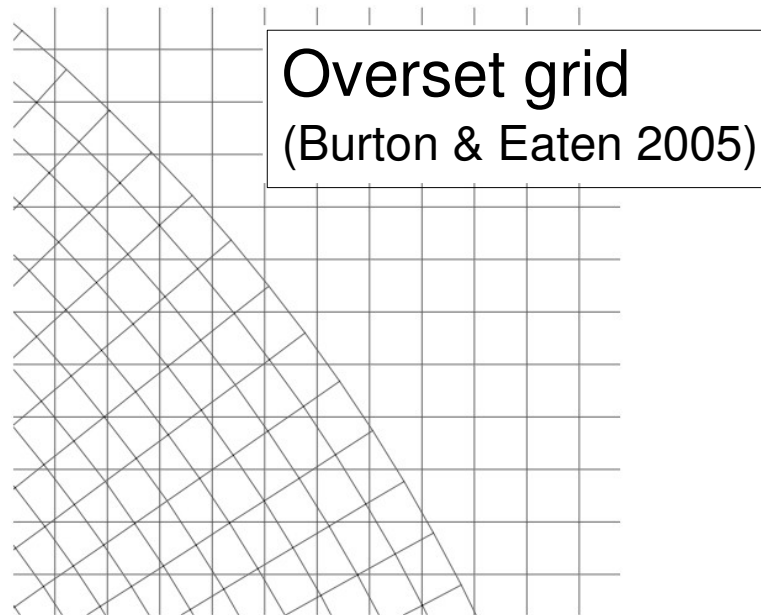


FIG. 1. Sketch of the immersed boundary.

- Solving dynamics on complete grid
- Imposing boundary conditions on S

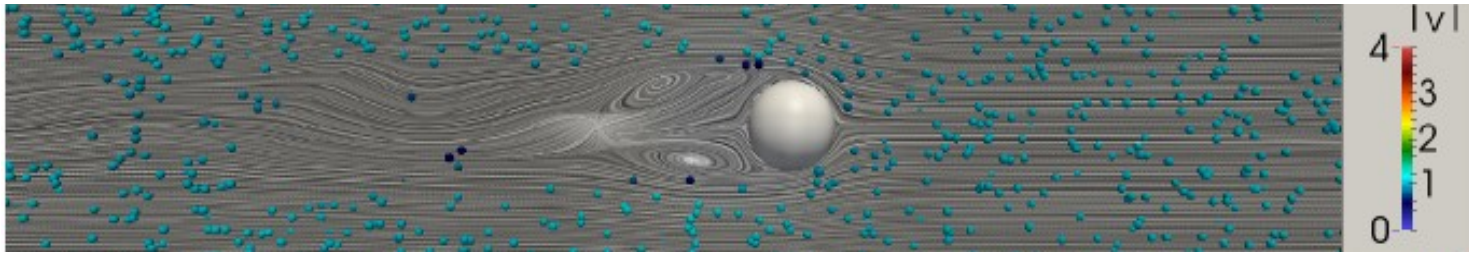
Advantages :

- Arbitrary shapes of S
- Easy to implement
- Compatible with Fourier-spectral codes

Shortcomings :

- Accuracy (order, boundary layer)

Numerical method



- Fluid: Navier-Stokes equations by pseudo-spectral method

$$\partial_t \mathbf{u} + \mathbf{u} \cdot \nabla \mathbf{u} = -\nabla p + \mathbf{f} + \nu \nabla^2 \mathbf{u}, \quad \nabla \cdot \mathbf{u} = 0 \quad \langle \mathbf{u} \rangle = U \mathbf{e}_x$$

$\mathbf{u}$: velocity field, ν : kinematic viscosity, $\mathbf{f} = \mathbf{f}_e + \mathbf{f}_b$: forcing (external + boundary)

- Small particles : heavy particles with low Re $\rightarrow$ Stokes drag

$$\ddot{\mathbf{X}} = \frac{1}{\tau} [\mathbf{u}(\mathbf{X}, t) - \dot{\mathbf{X}}] \quad \tau = 2\rho_p a^2 / (9\rho_f \nu) \quad \text{particle response time}$$

$$\text{St} = \tau / (d/U)$$

Effect of inertia based on particle crossing time

- No-slip boundary conditions: penalty method (direct forcing, Fadlun et al. 2000) :

$$\mathbf{V} = \mathbf{u}|_{\text{sphere}} = 0$$

discretized Navier-Stokes equation : $\mathbf{u}^{n+1} - \mathbf{u}^n = \Delta t (\text{rhs}^n + \mathbf{f}_b^n)$
 replacing $\mathbf{u}^{n+1} \rightarrow \mathbf{V}^{n+1}$ yields force $\mathbf{f}_b^n = -\text{rhs}^n + (\mathbf{V}^{n+1} - \mathbf{u}^n) / \Delta t$

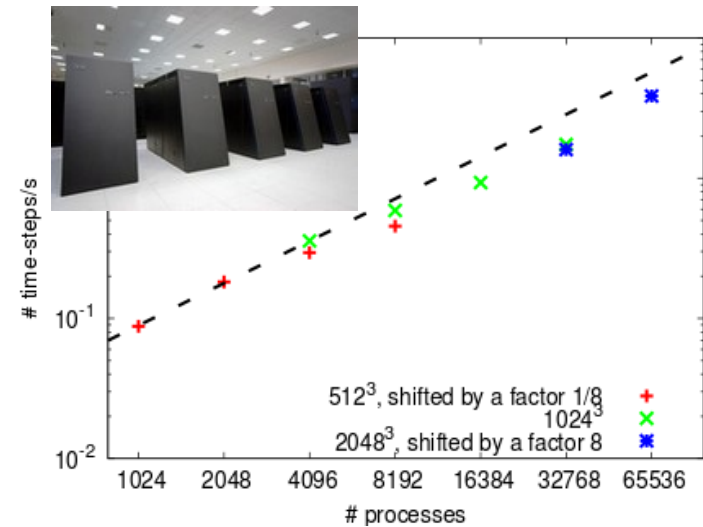
Numerical method – boundary conditions

Advantages:

- Standard Fourier-solver for incompressible turbulence
- Massive parallel codes
- Easy to implement

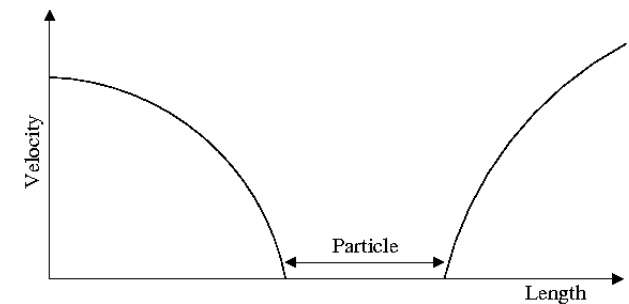
Difficulties:

- Truncated Fourier representation and discontinuities
- No-slip condition and divergence-freeness at the boundaries



Discontinuities in the derivative of the velocity field

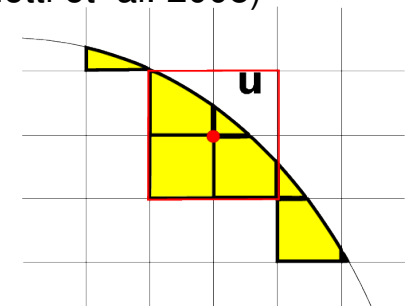
- lost spectral accuracy (non-smooth)
- Gibbs oscillations at the discontinuities



Smoothed particle: volume-fraction scheme (Fadlun et al. 2000, Pasquetti et al. 2008)

- reduces Gibbs-oscillation
- allows for a moving particle
- in the spirit of a volumic forcing

$$\chi = \frac{1}{2h} \int_{(i-1)h}^{(i+1)h} \chi dx$$



Numerical method – no-slip & solenoidal

Main troubles :

- Operations boundary conditions (B) and divergence-freeness (P) do not commute !
- Fourier : poisson-equation for p with periodic boundary conditions

Solution : pressure predictor (Brown et al. 2001)

$$\tilde{u}^{n+1} = u^n + \Delta t [-(u^n \cdot \nabla)u^n - \nu \Delta u^n - \nabla p^n] \quad \Delta p^n = \nabla \cdot [(u^n \cdot \nabla)u^n]$$

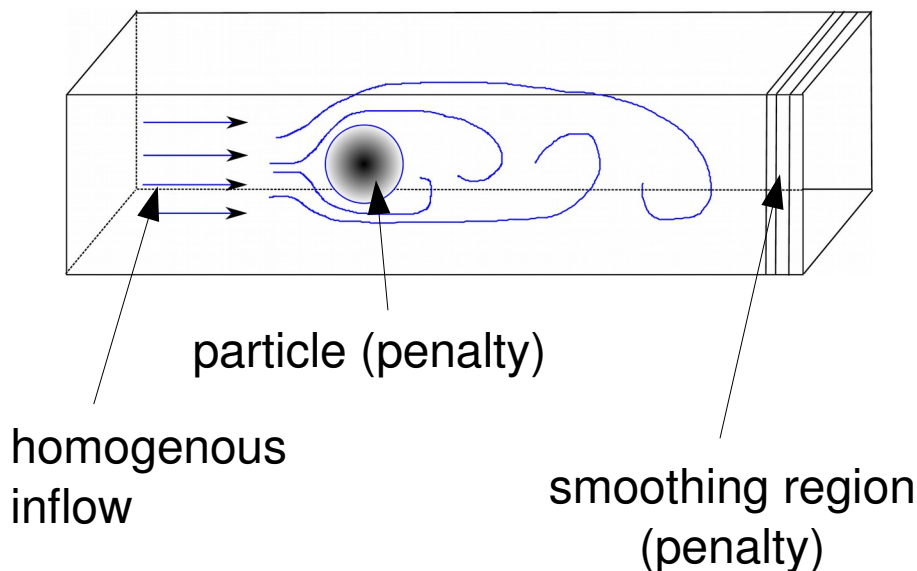
$$\tilde{\tilde{u}}^{n+1} = B(\tilde{u}^{n+1}) \equiv \tilde{u}^{n+1} + f_b^n$$

$$u^{n+1} = P(\tilde{\tilde{u}}^{n+1})$$

Usually no predictor ∇p – only projection

Benchmark-setup :

Numerical windtunnel



- regular grid with e.g. 256x256x1024 grid-points
- periodic boundary conditions at the surfaces
- penalty for particle and smoothing region

Numerical method – no-slip & solenoidal

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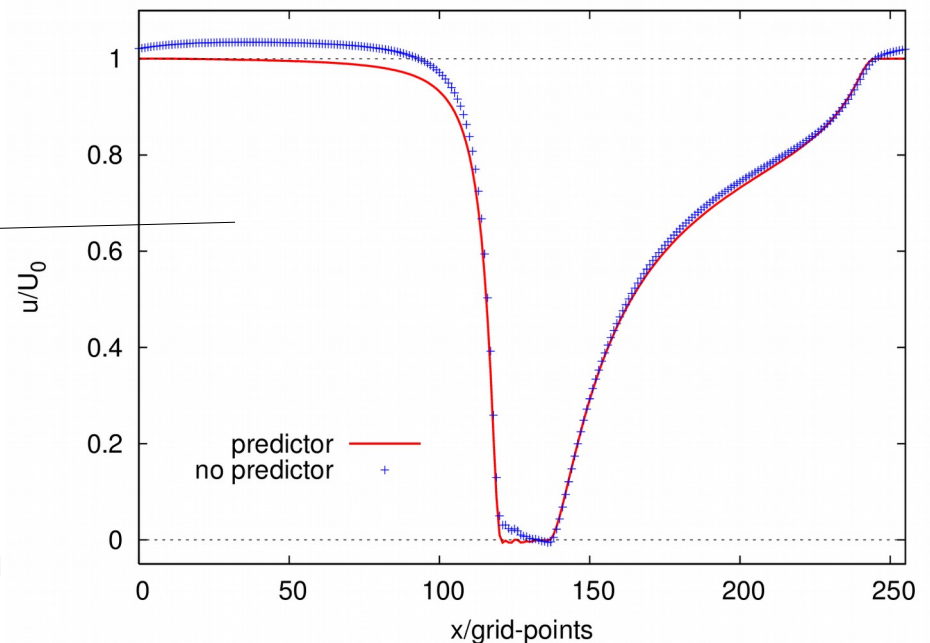
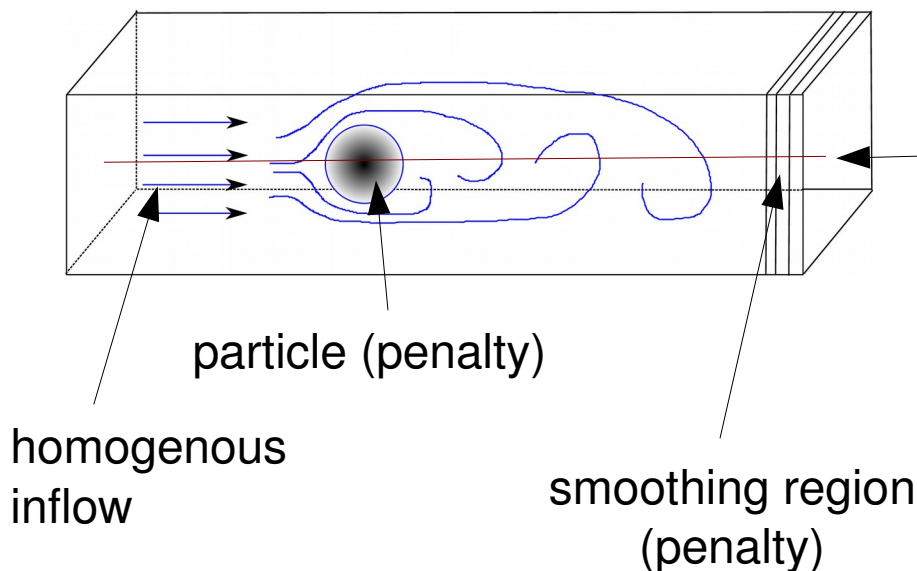
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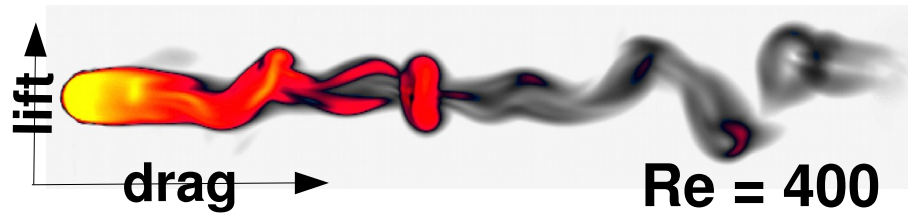
Benchmark-setup :

Numerical windtunnel

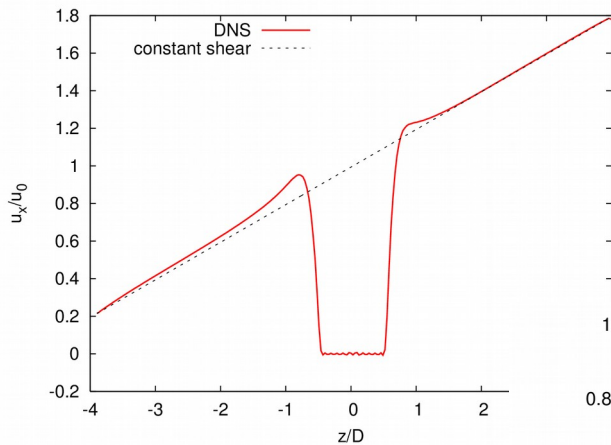
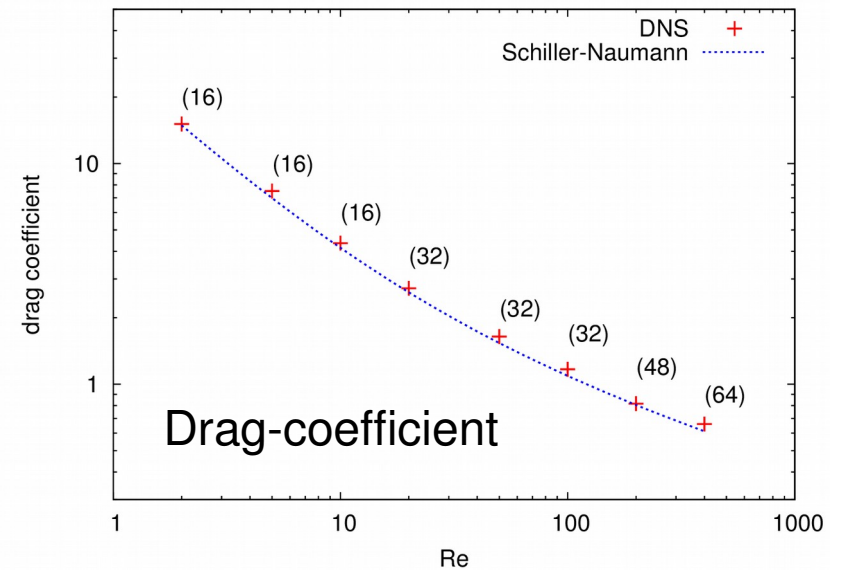


Numerical method - benchmarks

Forces on the particles

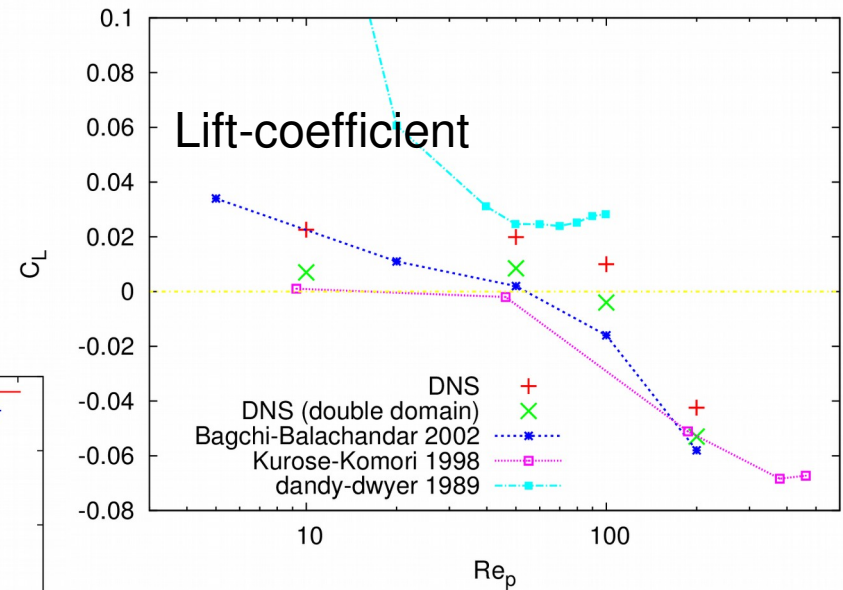
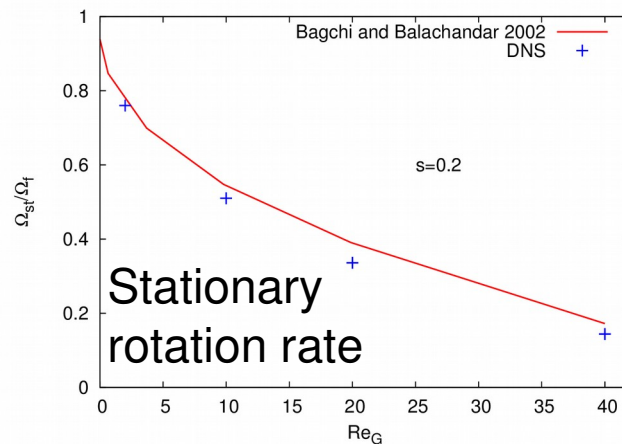


Drag-coefficient $C_D = F_D / (0.5 U_\infty^2 \pi r^2)$



Constant shear:

$$s = \nabla u D / u_\infty$$



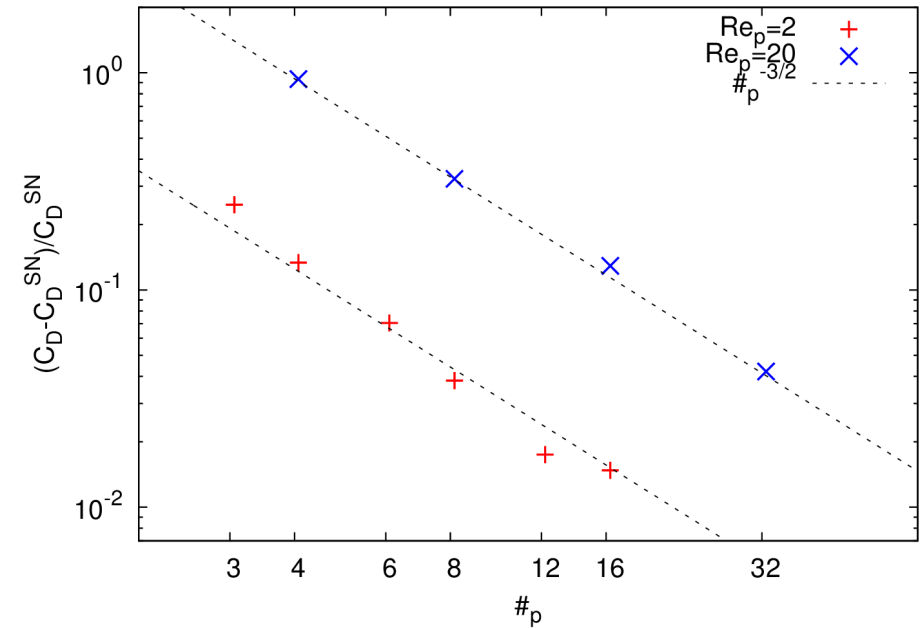
Numerical method - convergence

Convergence rate:

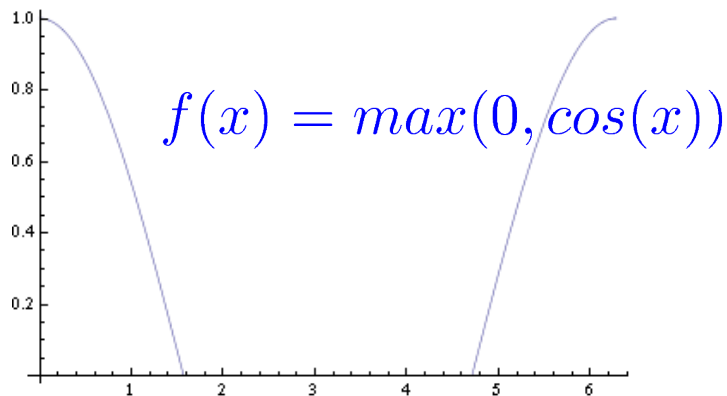
Boundary layer thickness $\delta_b \sim Re^{-1/2}$

Resolution : $D/\Delta_x \sim \delta_b/\Delta_x$

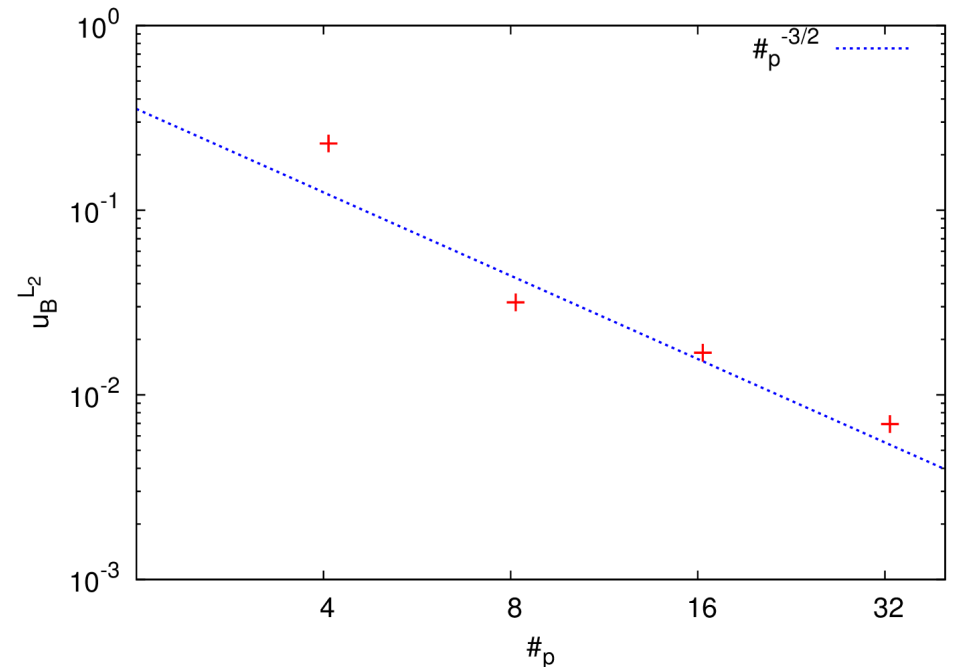
for $Re = const$

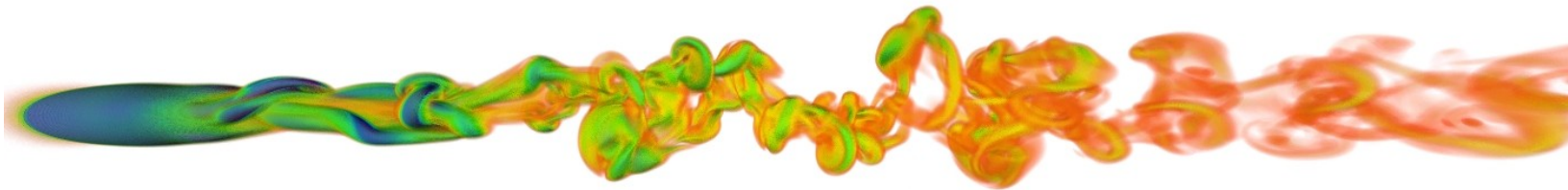


Why -3/2?



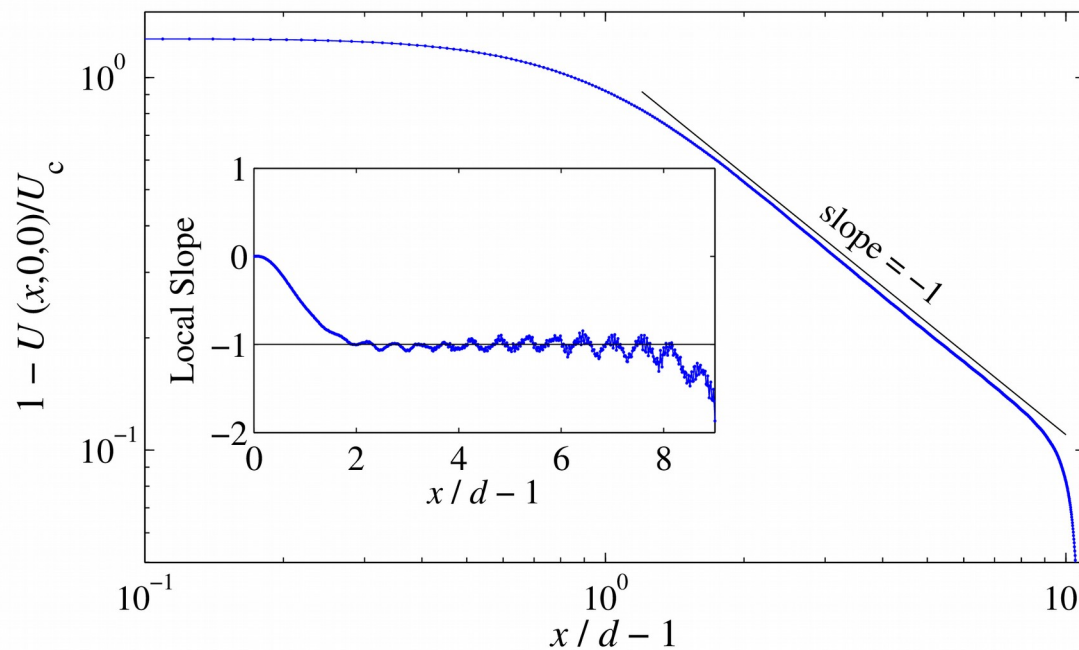
L_2 -Norm $\left(\int_B f_K(x)^2 dx \right)^{1/2} \sim K^{-3/2}$





Mean velocity deficit : $\Delta U = 1 - U(x, 0, 0)/U_c \sim x^{-1}$

(Wu & Faeth 1994, Bagchi & Balachandar 2004)



Results

- 1) Particle collision rates
- 2) Particle-wake interactions
- 3) Drag force

Small particle collision rates

Problem Parameters : Particle Reynolds number

$$100 \leq Re_p = \frac{d U_c}{\nu} \leq 1000$$

Fluid Reynolds number

$$0 \leq Re = \frac{u_{rms} L}{\nu} \leq 3000$$

Turbulent intensity

$$0 \leq I = \frac{u_{rms}}{U_c} \leq 1.2$$

Dust Stokes number

$$0 \leq \tau \leq 80$$

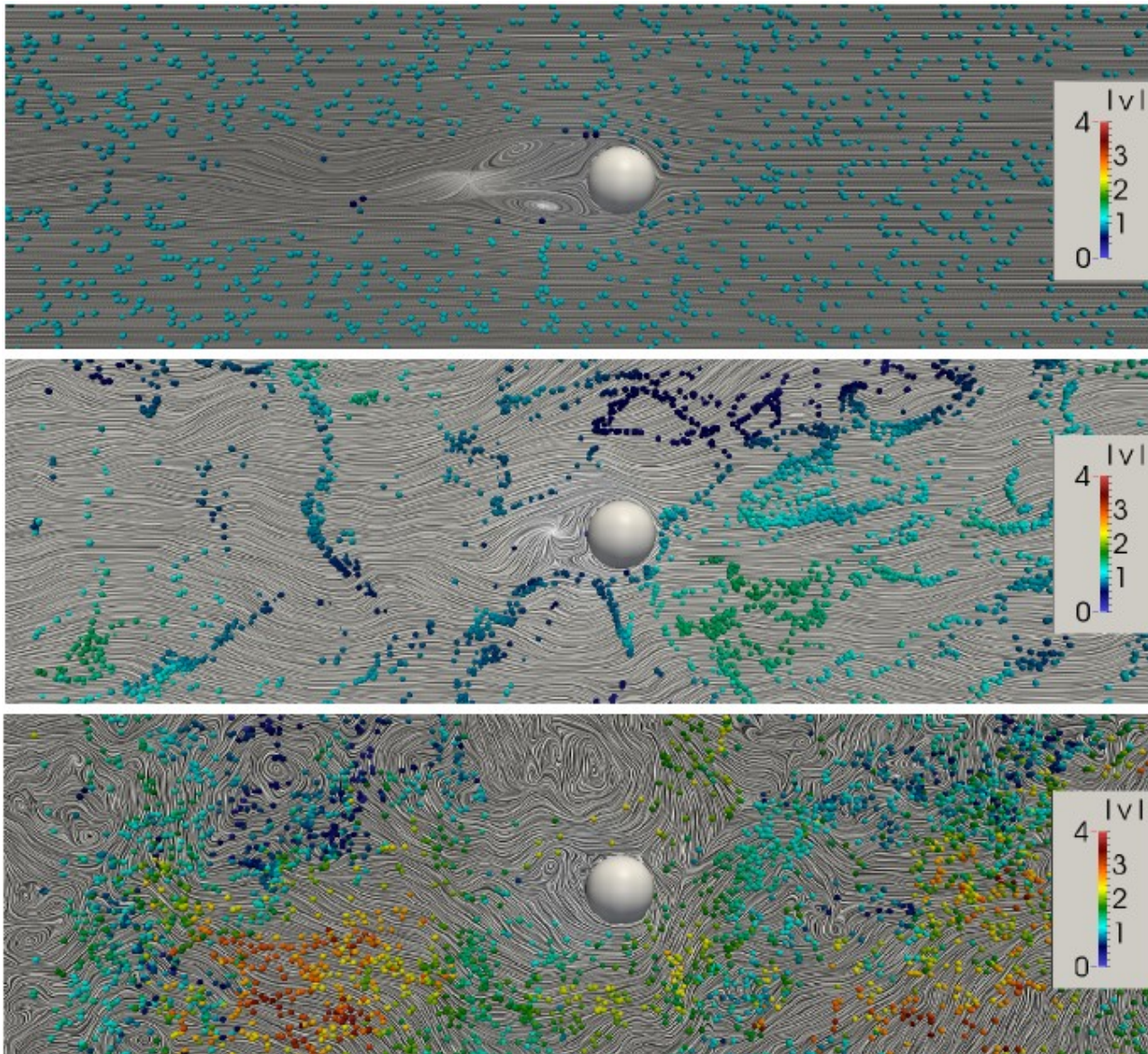
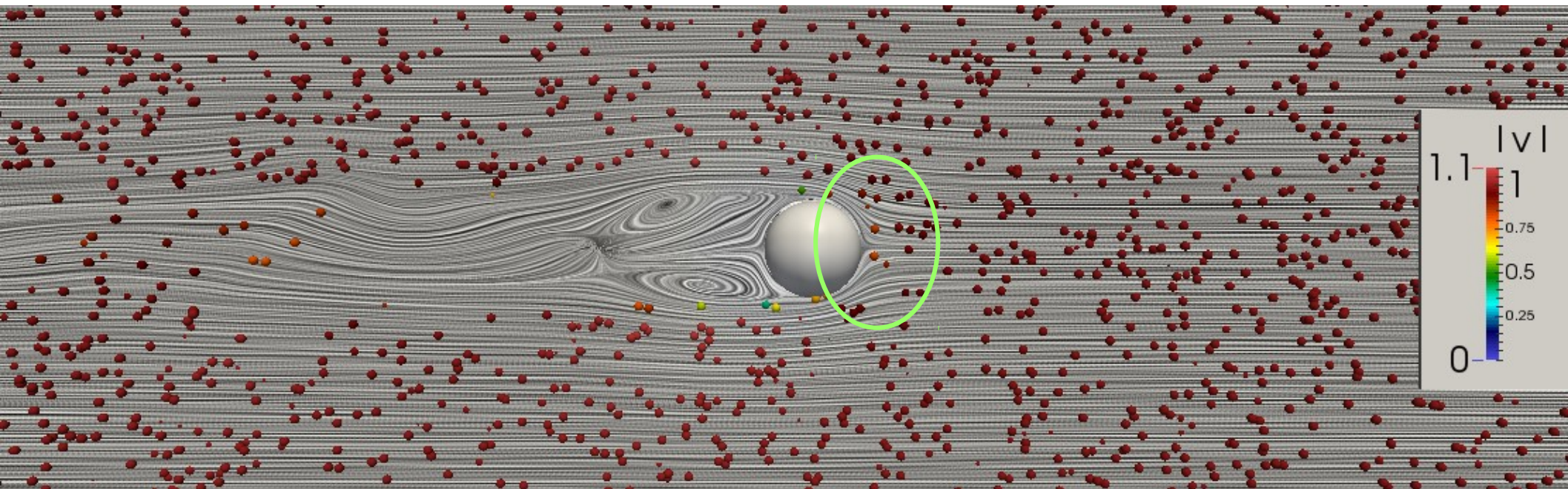


Fig. 1. Instantaneous streamlines of the gas flow and dust particles ($St = 0.8$) for $I=0$ (laminar), $I = 0.29$ (weakly turbulent), and $I = 1.18$ (strongly turbulent) from top to bottom. Note that while $St = 0.8$ for all shown cases, $St_\eta = 1.25$ for $I = 0.29$ and $St_\eta = 9.8$ for $I = 1.18$ so that the clustering properties of the shown particles change from one I to another.

Small particle collision rates

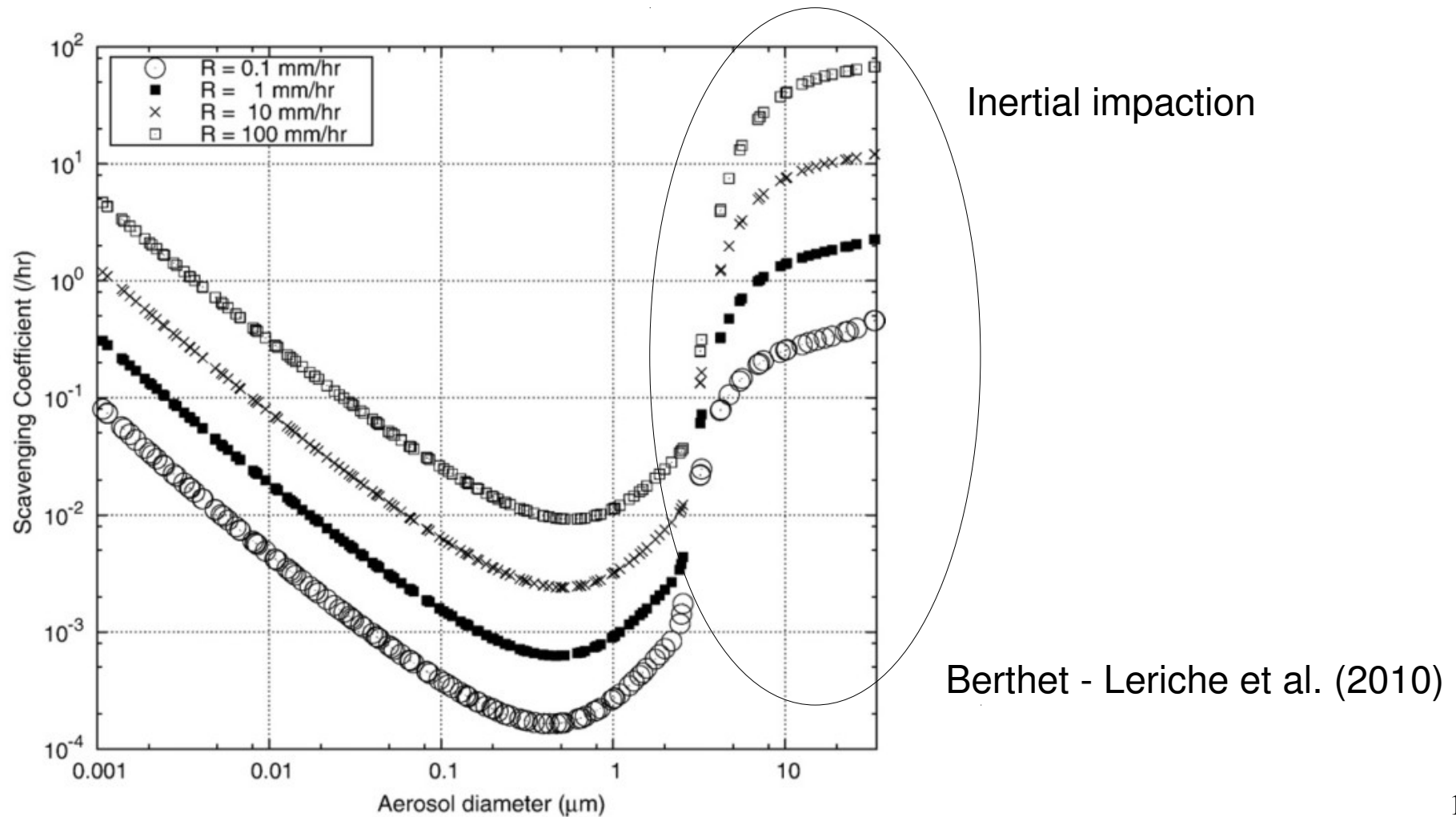
Laminar (uniform) inflow



Small particle collision rates

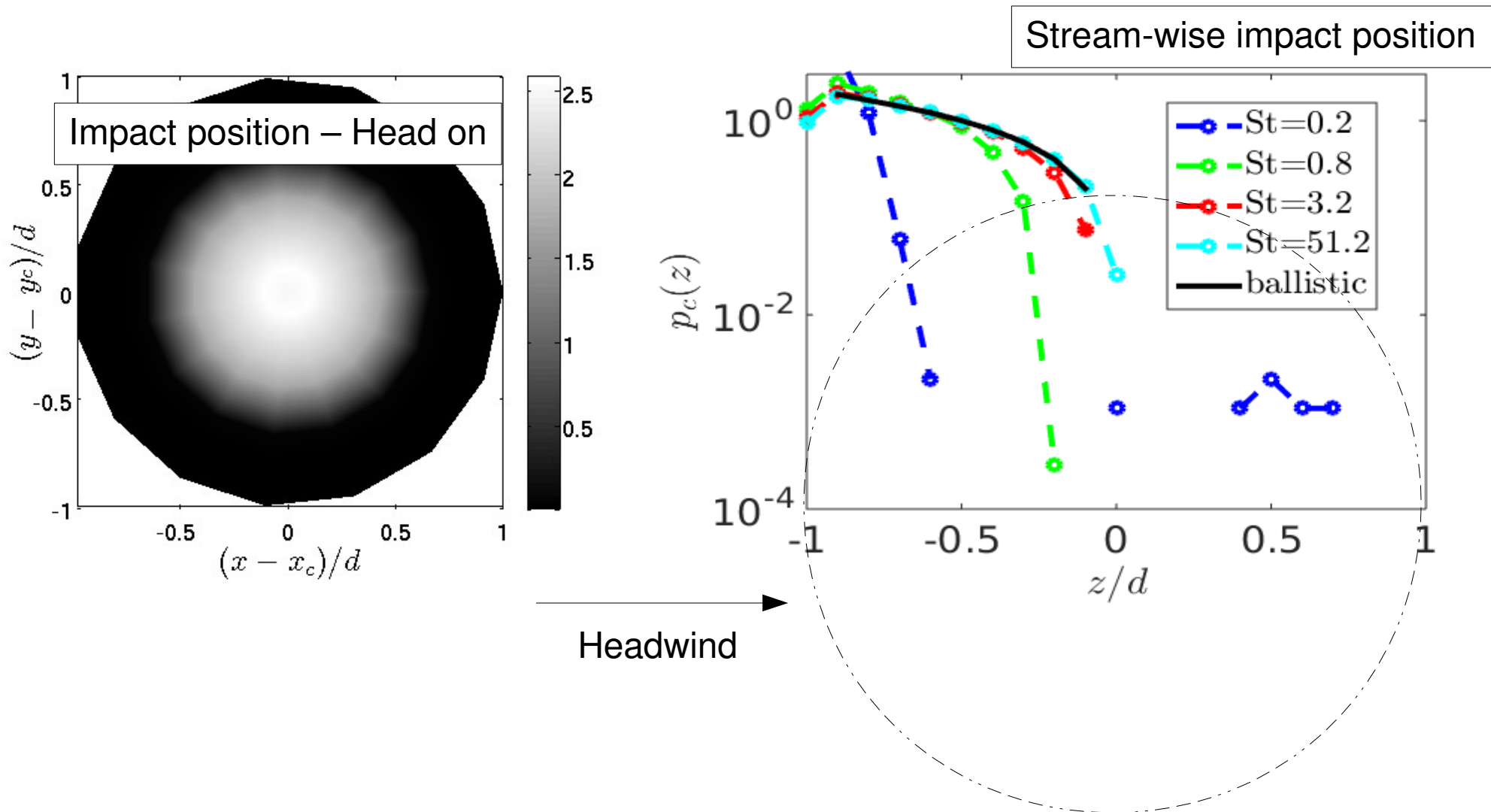
Application: Wet deposition of aerosols by falling raindrops

$$\gamma(d) = \int_0^\infty \pi/4 U_t(D) E(d, D) n(D) dD$$



LAMINAR collision statistics

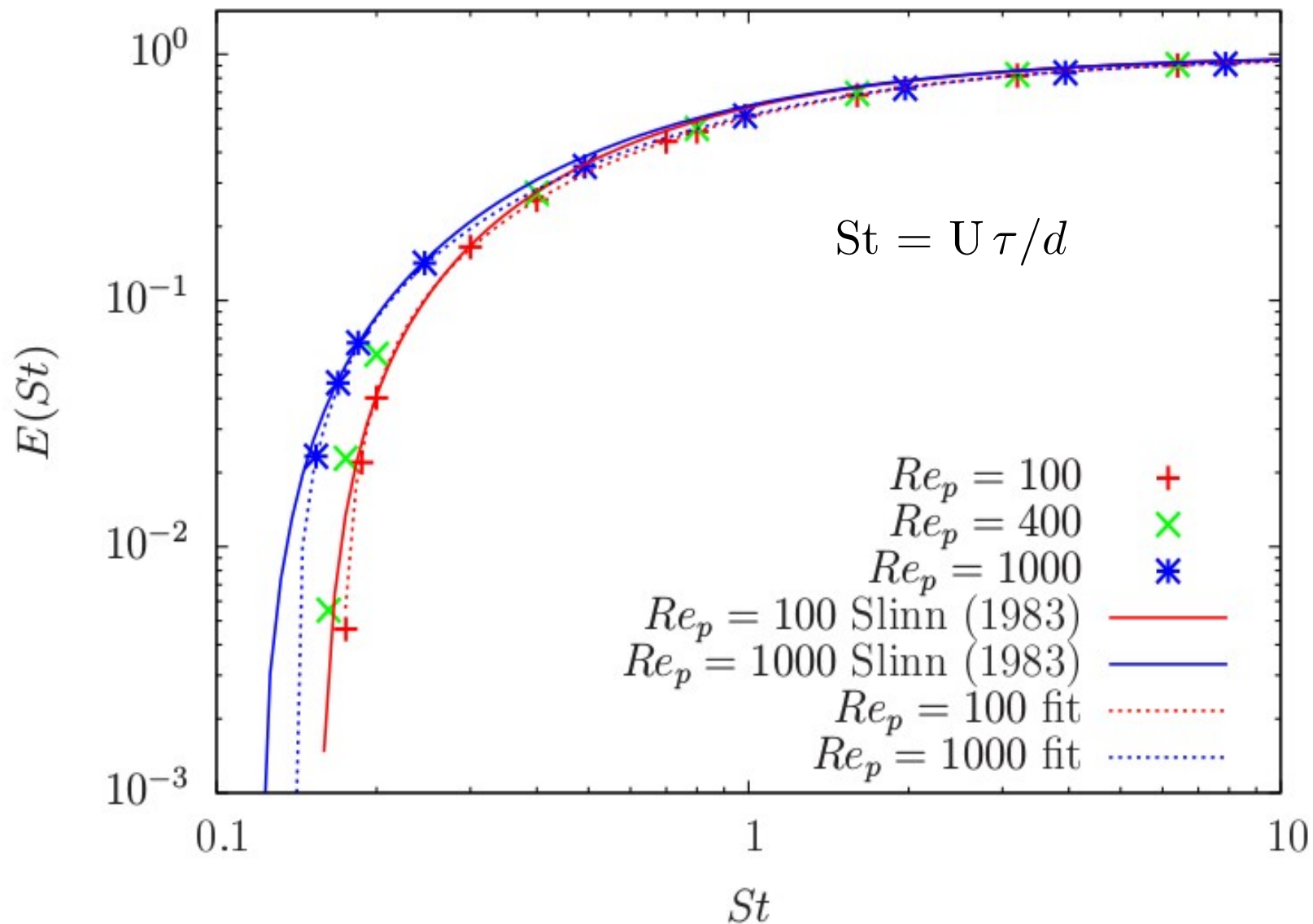
Where do collisions happen ?



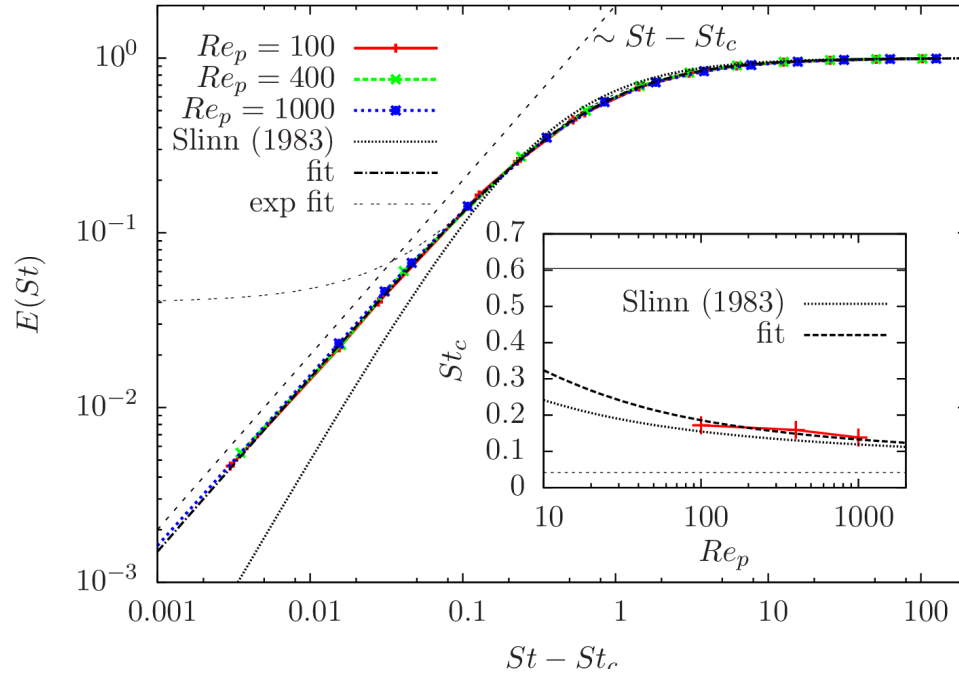
- Collisions are concentrated around the axis of symmetry
- Virtually no backward collision happen

LAMINAR Collision statistics

Collision rate



LAMINAR Collision statistics



Small St limit :

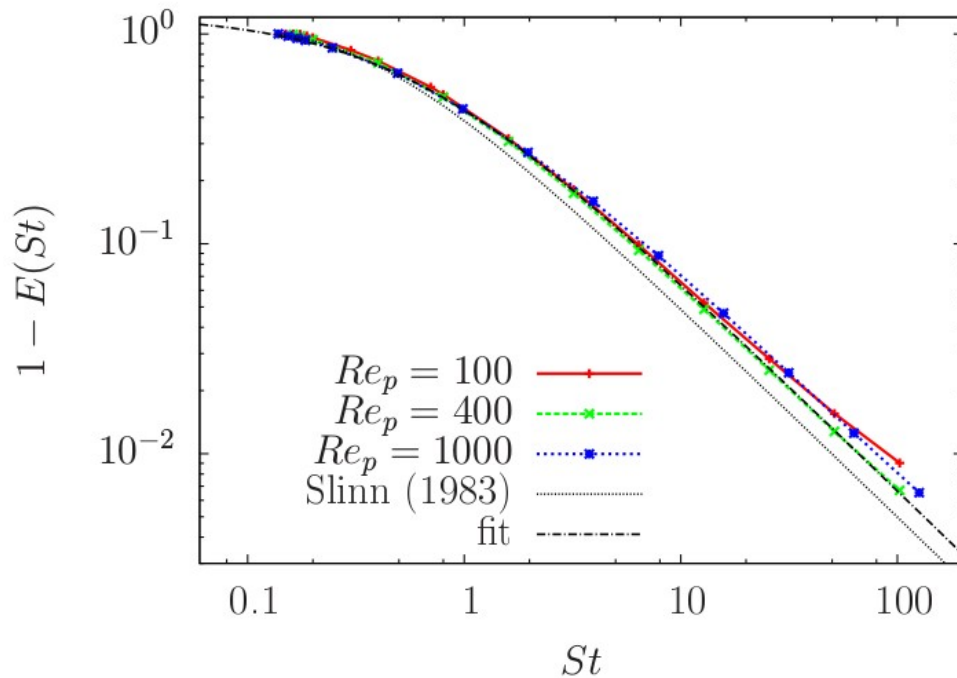
- Critical Stokes number St_c
- Euler flow : $St_c = 1/24$
- Stokes flow : $St_c = 0.6$
- Re-collapse with linear slope

fit :
$$\frac{St - St_c}{St - St_c + 2/3}$$

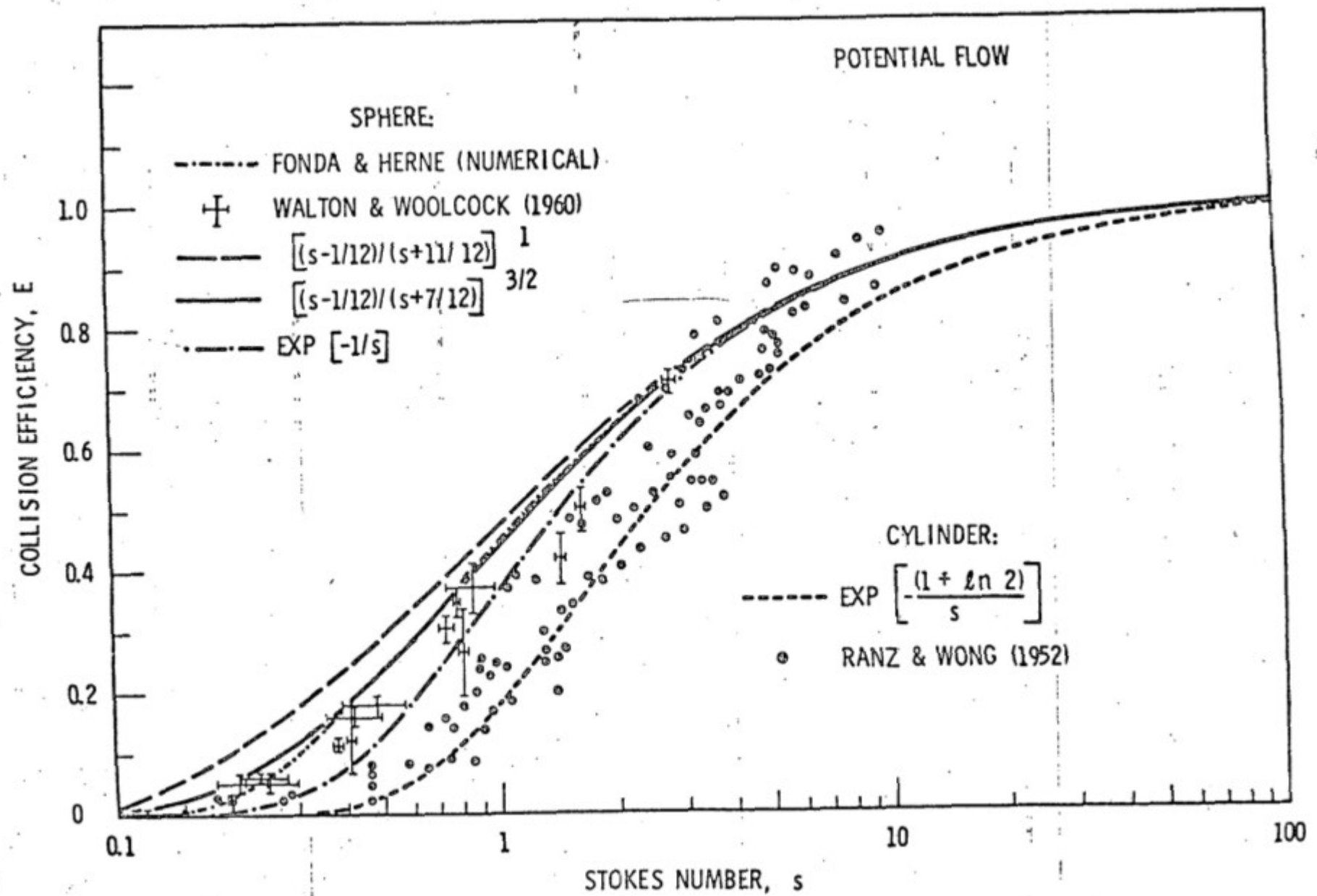
Slinn :
$$\left(\frac{St - St_c}{St - St_c + 2/3} \right)^{3/2}$$

Large St limit :

- Re-collapse with $E(St) \sim 1 - 1/St$

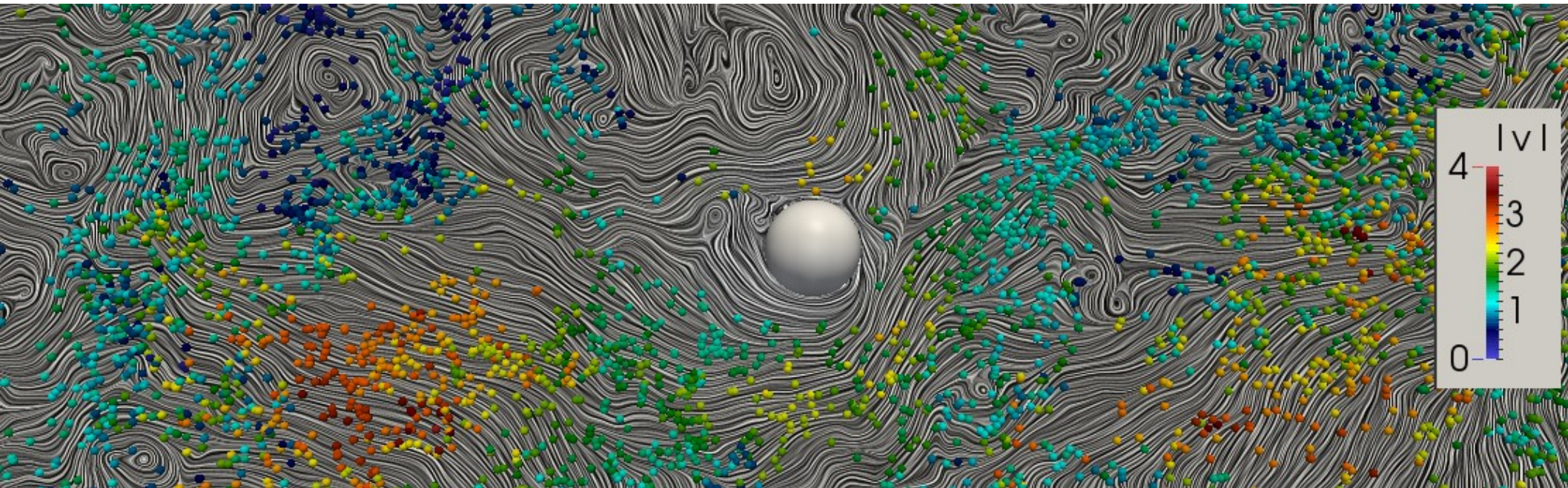


Slinn's fit (1974)

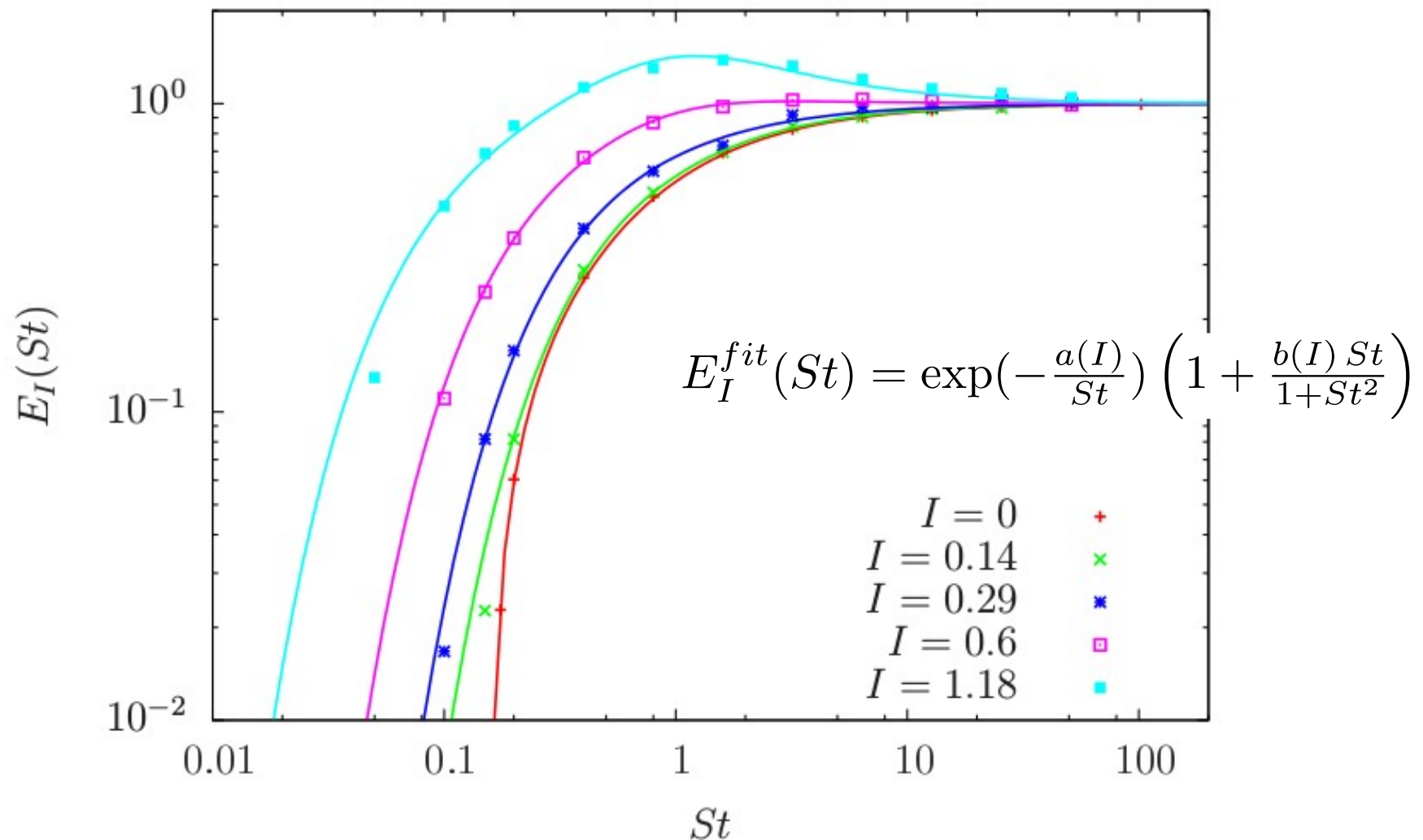


Small particle collision rates

Turbulent inflow



TURBULENT Collision Statistics



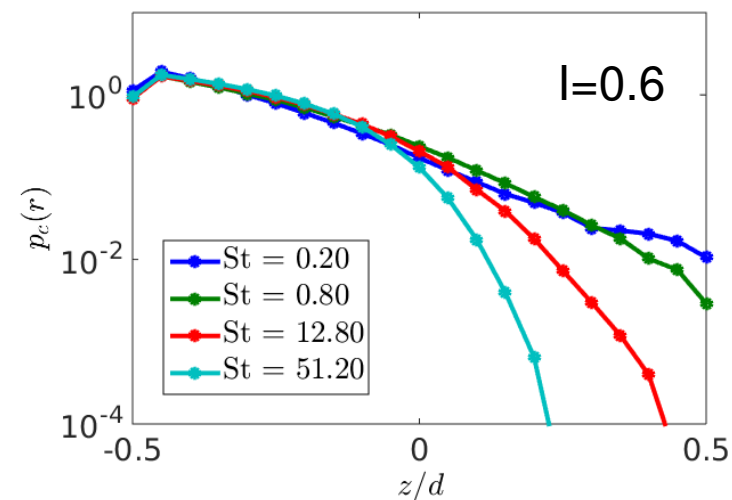
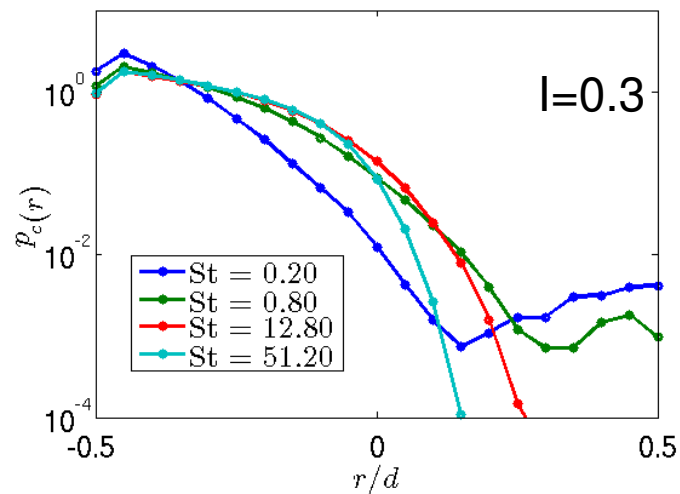
- Turbulence increases significantly the collision efficiency
- Collision efficiency can be larger than unity
- Probably no critical Stokes number

Turbulence effects on collision rate I

1) Fluctuating head wind

- Increase of collision rate (higher Re $\rightarrow$ higher collision rate)
- Velocity distribution $\rightarrow$ no critical Stokes number

2) Randomization of impact direction and position ($\rightarrow$ backward collisions)

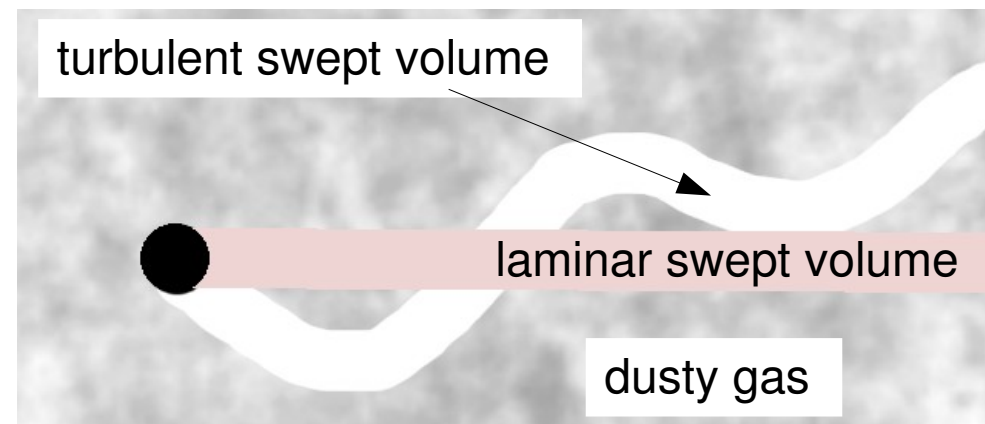


3) Increase of swept volume

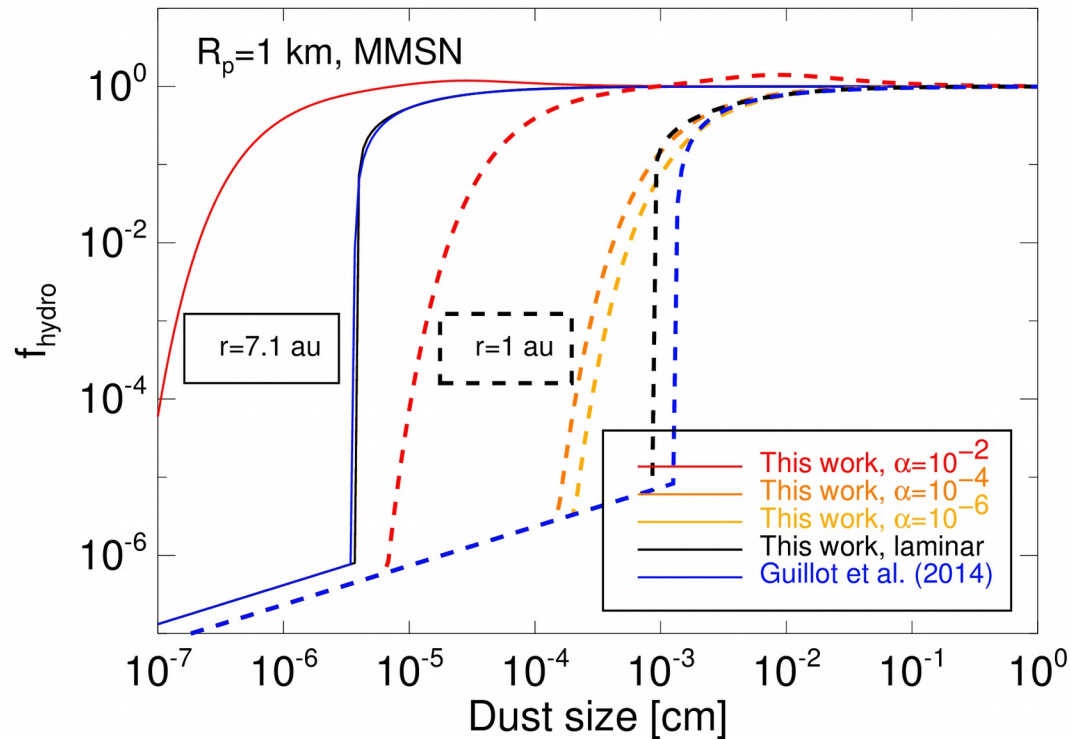
turbulent volume > laminar volume
simple computation for sinusoidal motion:

$$E_I(St) = E_0(St) \left(1 + \frac{I^2}{1+St^2} \right)$$

Allows for collision rate > 1 !



Consequences for planet formation

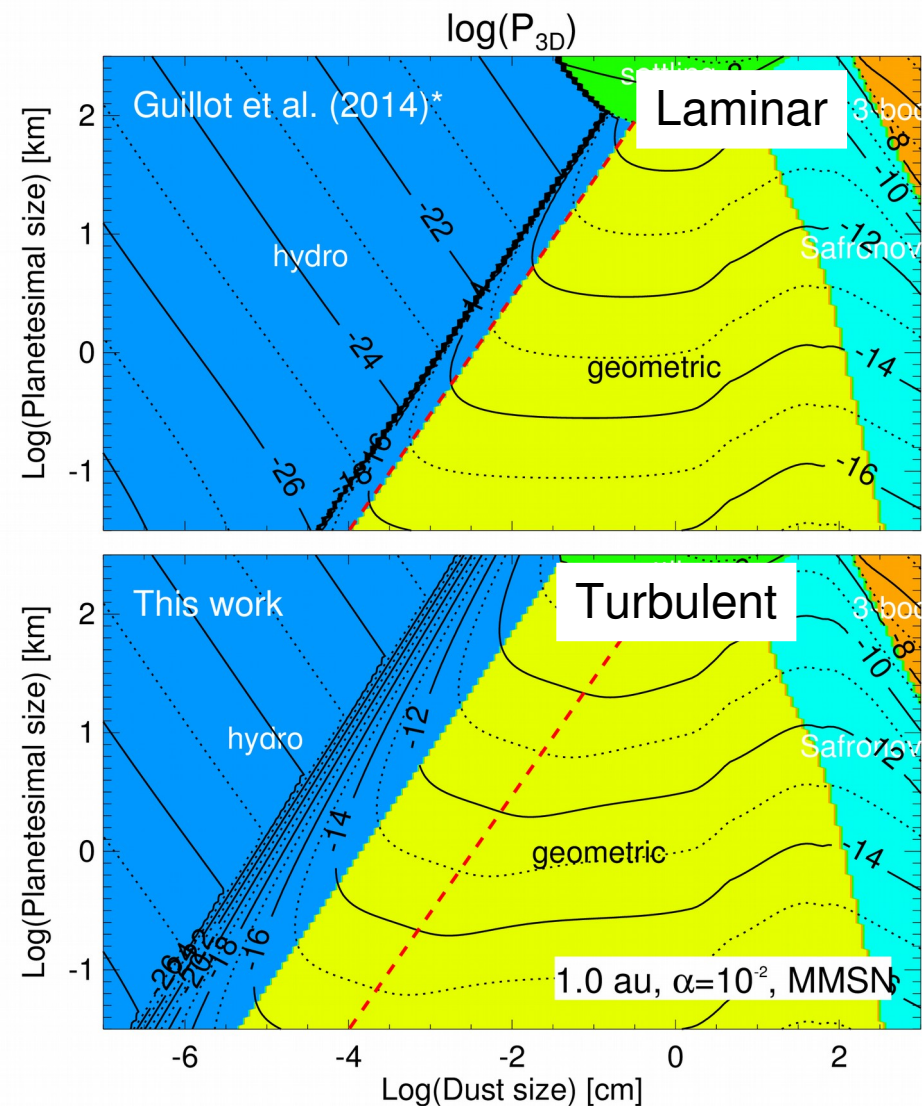


Collision efficiency for different orbital distances and turbulent intensities

alpha parameter: Turbulent eddy viscosity

$$\nu_t = \alpha c_s H$$

Extension of the geometric regime to smaller dust sizes by turbulence



Importance of collision speeds

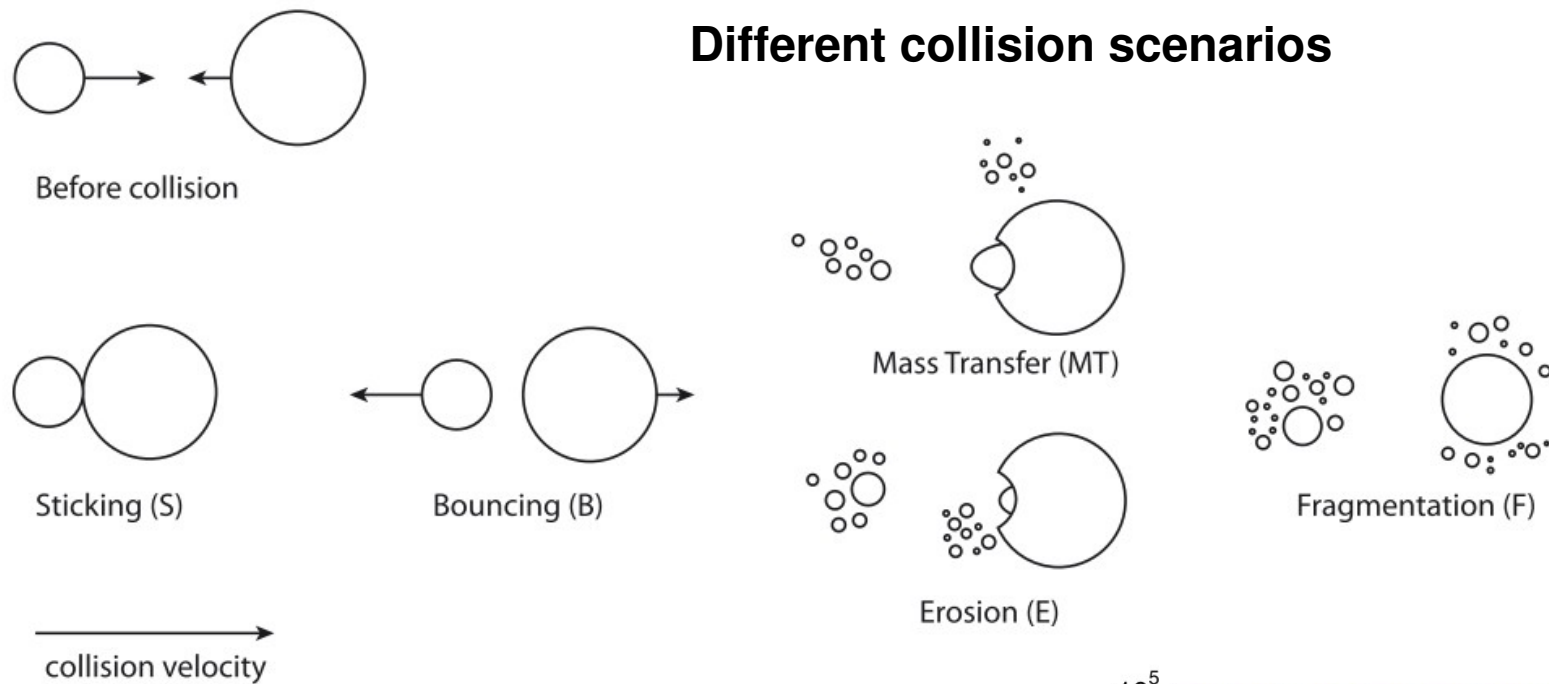
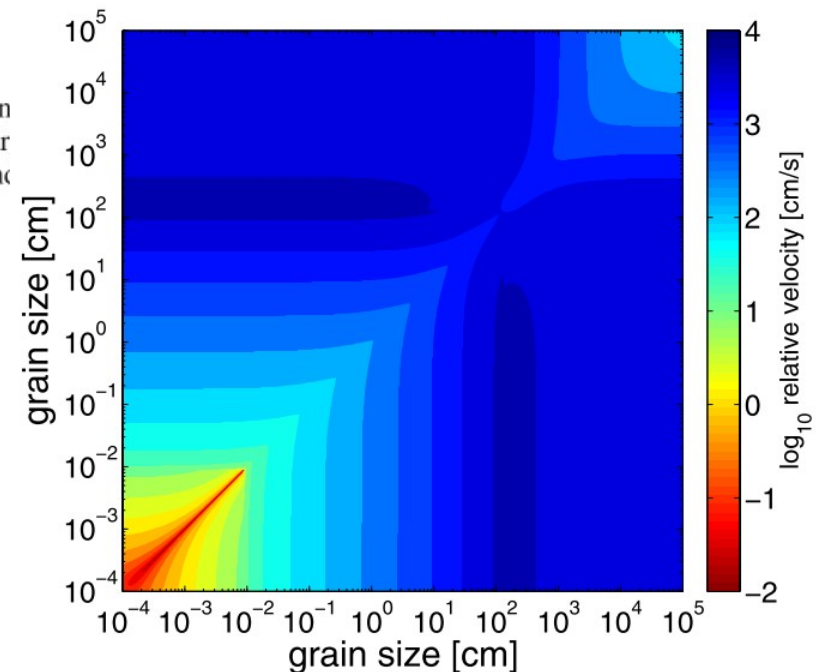


Fig. 2. Sketch of the five possible outcomes described in Sect. 3 sorted in rough order of collision velocity, and we define a mass transfer collision as leading to net growth for the target particle. This outcome is extremely dependent on the mass-ratio of the particles, thus adding a second

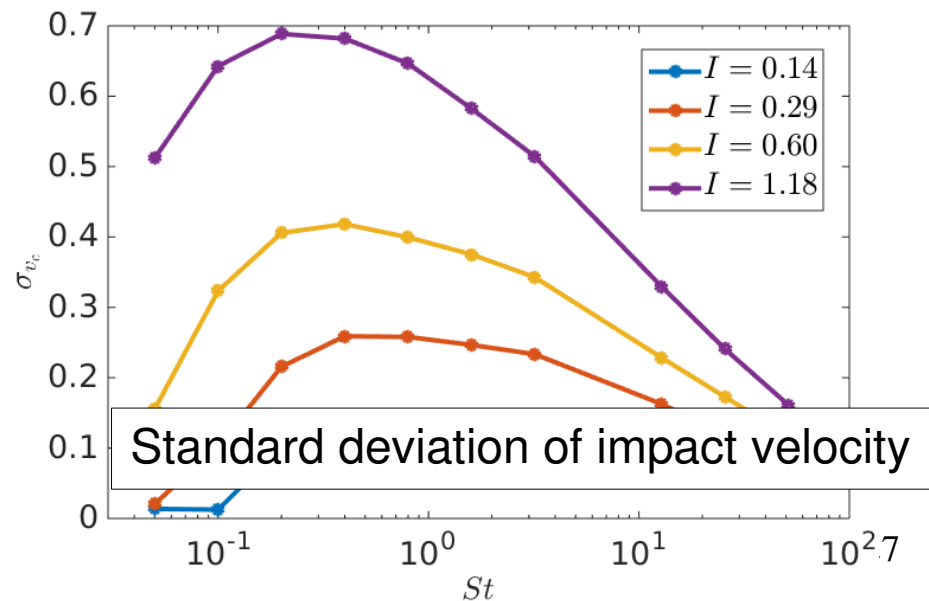
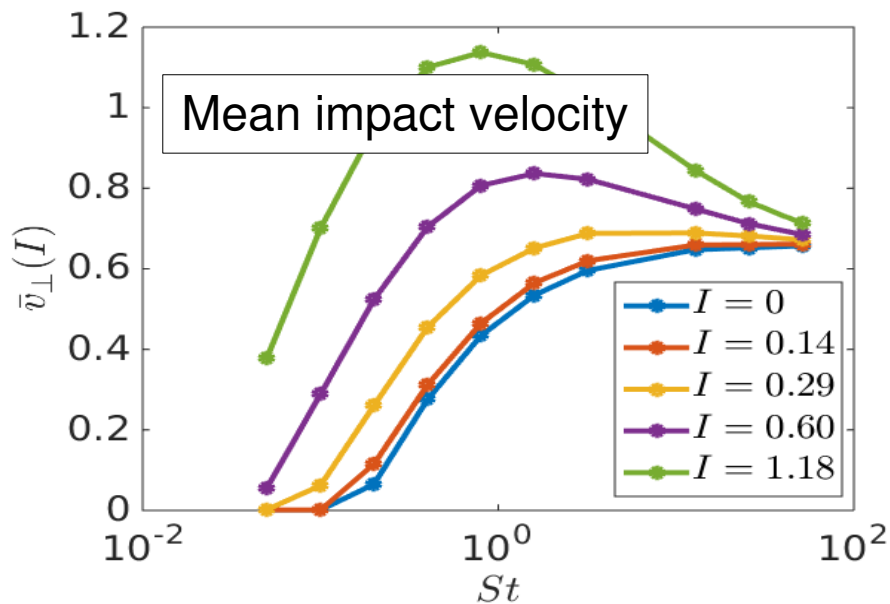
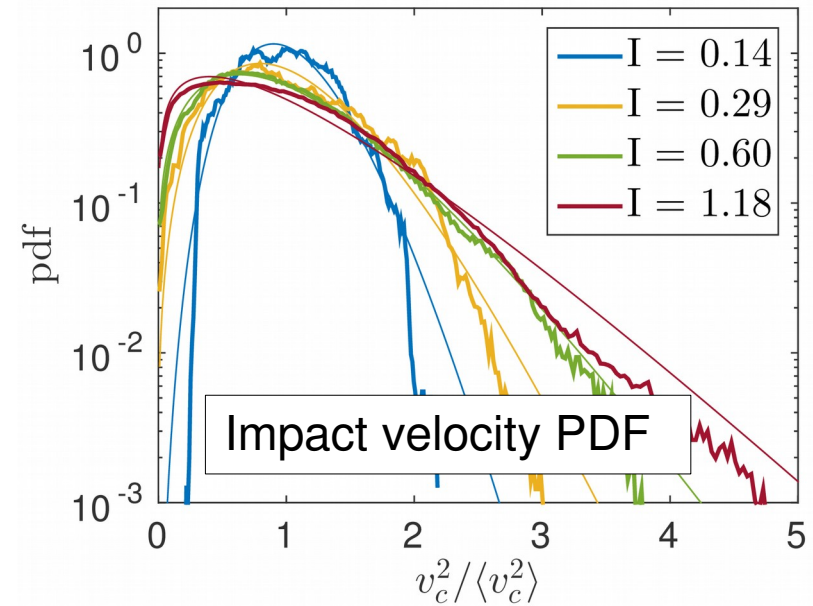
Population evolution models use velocity maps:

Windmark et al. (2012)



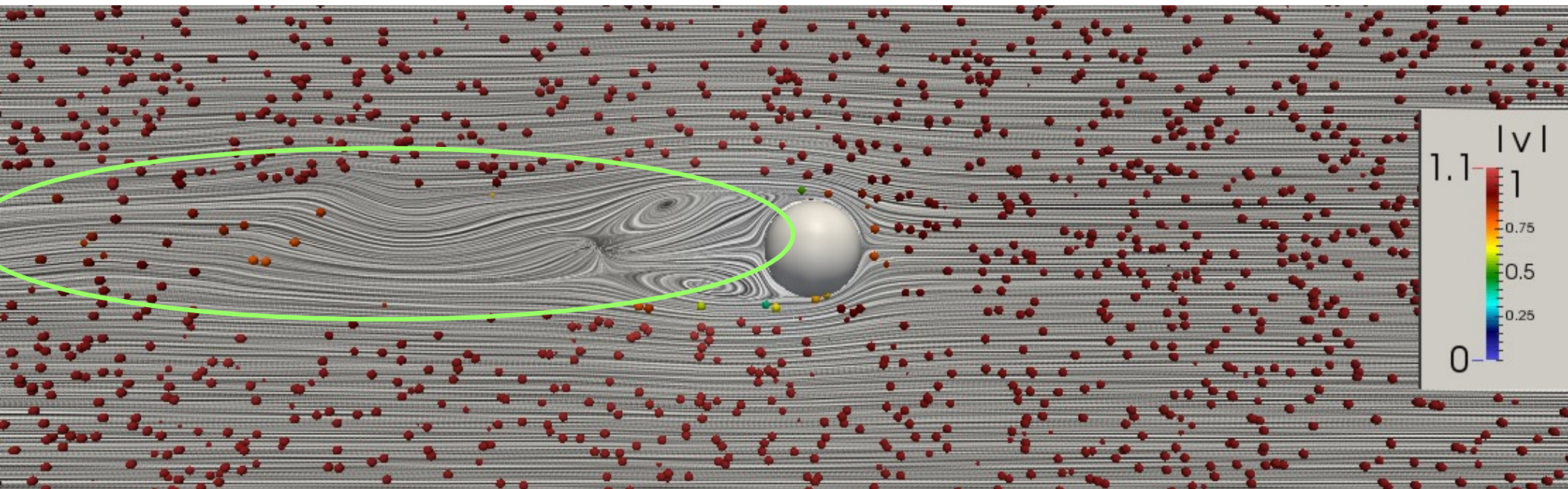
Measured collision speeds

- Turbulent fluctuations lead to velocity distribution
- Distribution fairly modeled by non-central chi-squared distribution
- Turbulence increases mean impact speed with maximum close to $St=1$
- Standard deviation approx. proportional to turbulence intensity



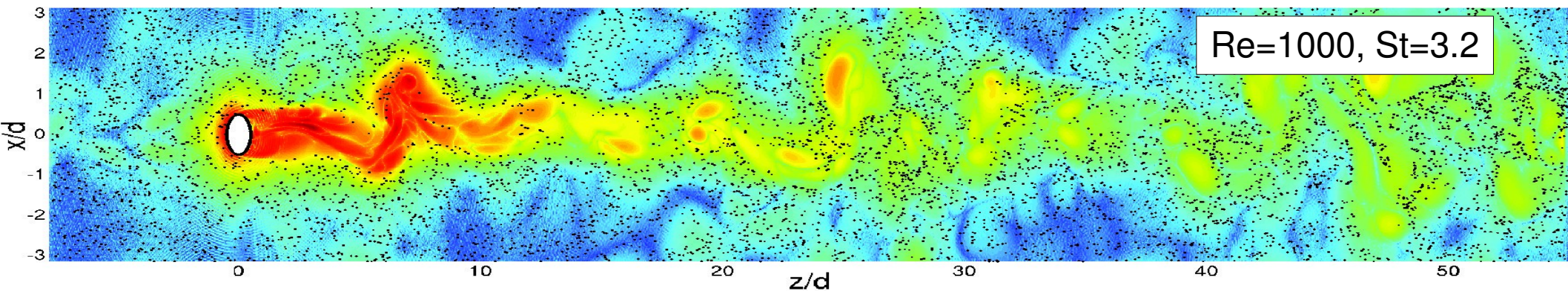
Small particle wake interactions

Laminar (uniform) inflow



Dust in the wake

What happens to particles that passed the sphere ?



Okubo-Weiss parameter

$$\Delta = \left(\frac{\det[\hat{\sigma}]}{2} \right)^2 - \left(\frac{\text{tr}[\hat{\sigma}^2]}{6} \right)^3$$

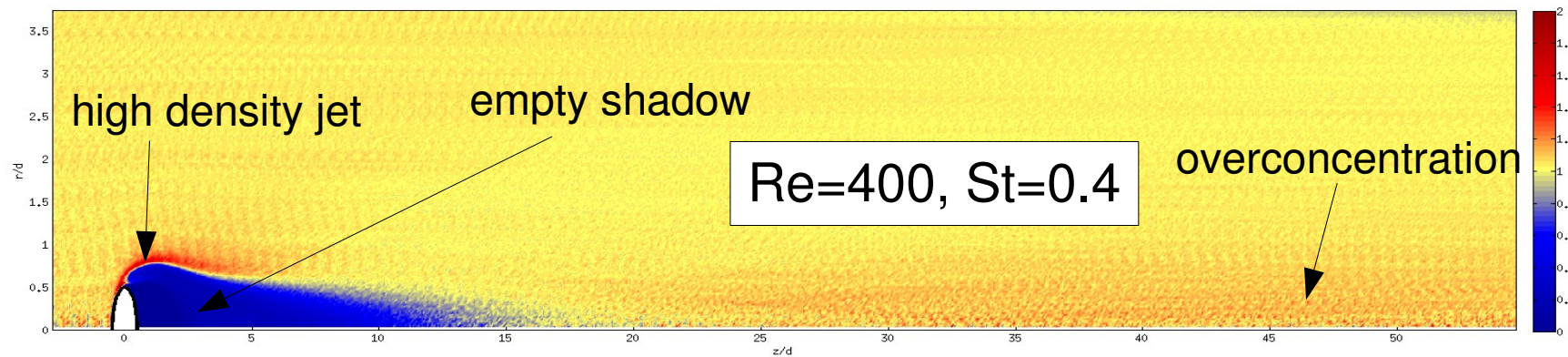
$\Delta < 0$ straining regions

$\Delta > 0$ rotational regions

$\hat{\sigma} = \partial_i u_j$ strain matrix

Particles interact with the shed structures – they are ejected

Mean particle densities



Experimental measurement (Jacober and Matteson 1990)

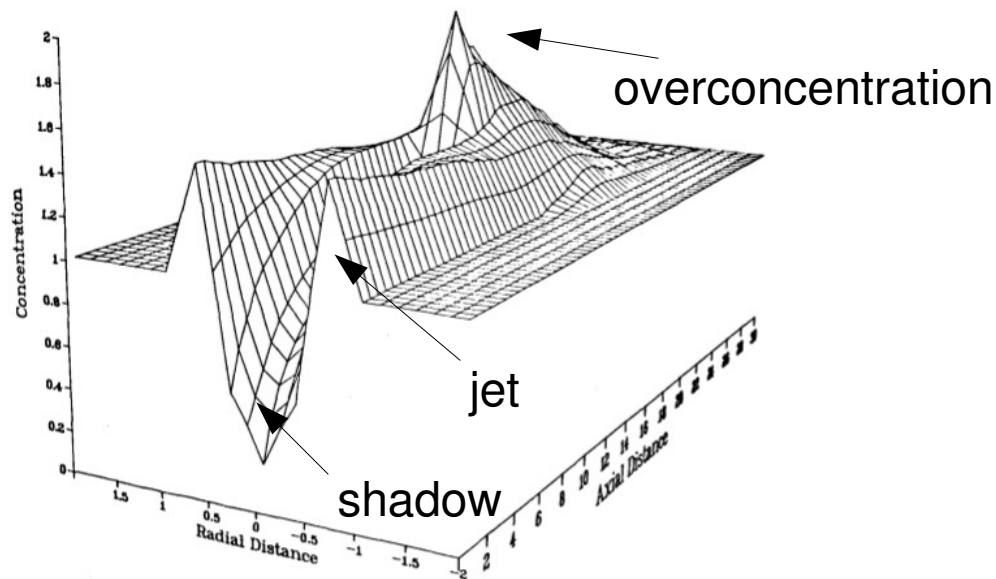
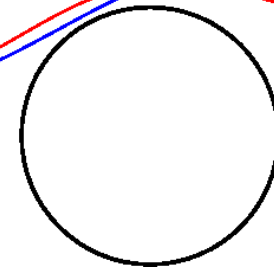


FIGURE 7. Axial and radial particle concentration distribution in the sphere wake for $10\text{-}\mu\text{m}$ particles and a free stream velocity of 6.1 m/s (radial and axial distances in radii).

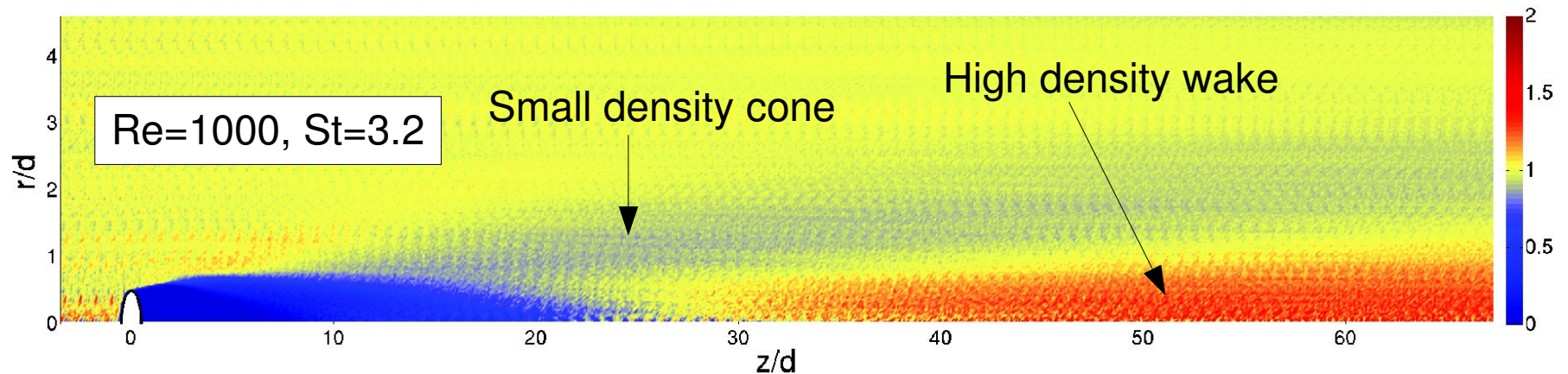
Jet creation mechanism

Heavy particles-
trajectories

Fluid-
trajectories

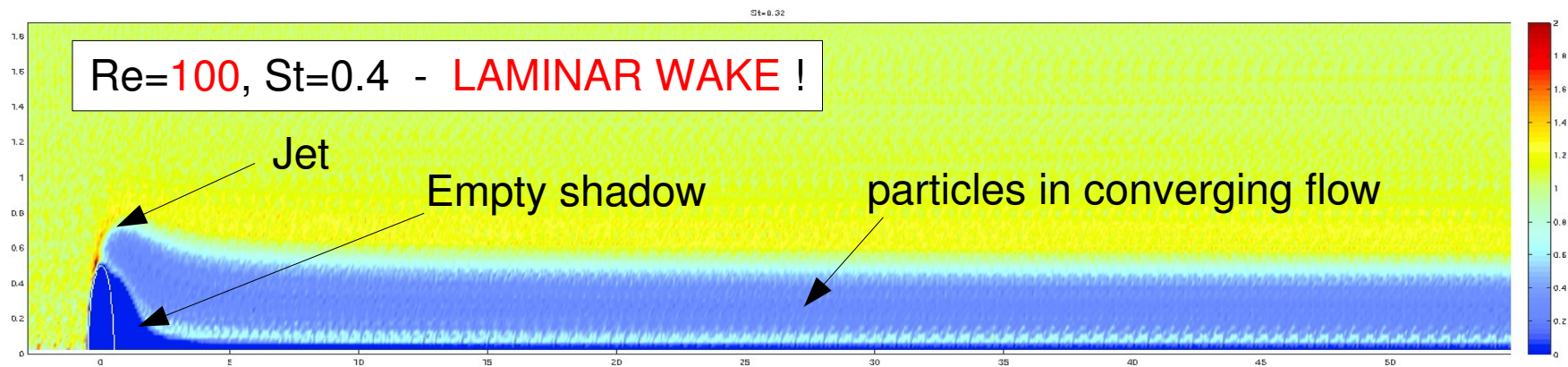


The wake : Mean particle densities



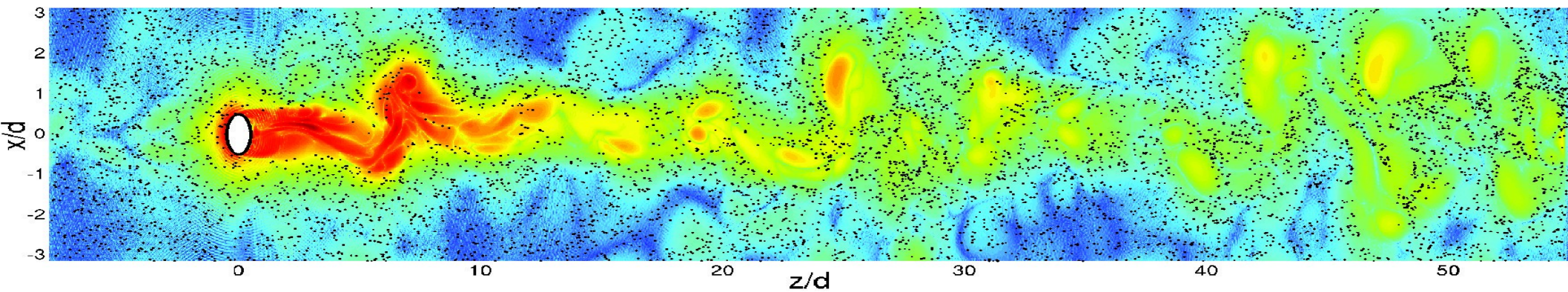
Jacobson and Matteson : wake concentrations from collapsing jet

But: Concentrations only for turbulent wakes !

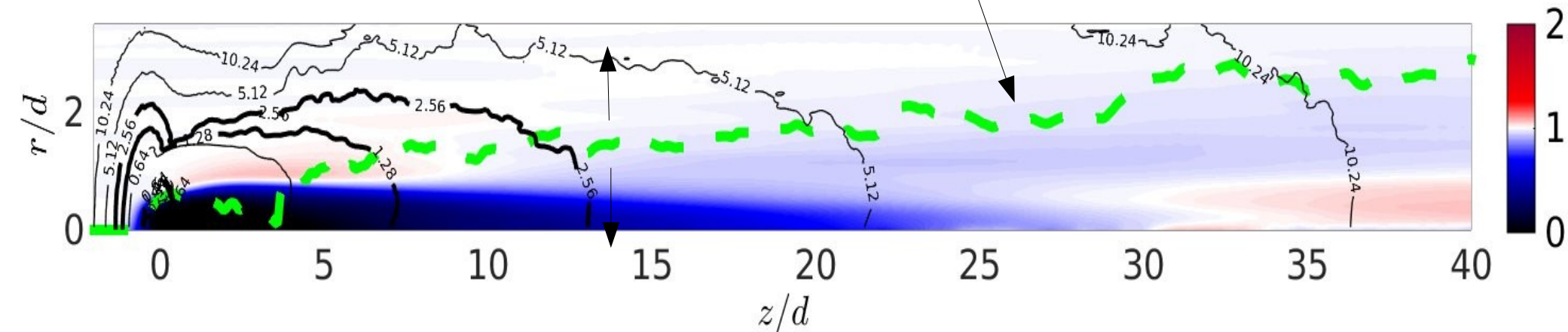


Concentration in the wake due to turbulent fluctuations 31

Wake concentration mechanism



Maximum of Δ depends on stream-wise position



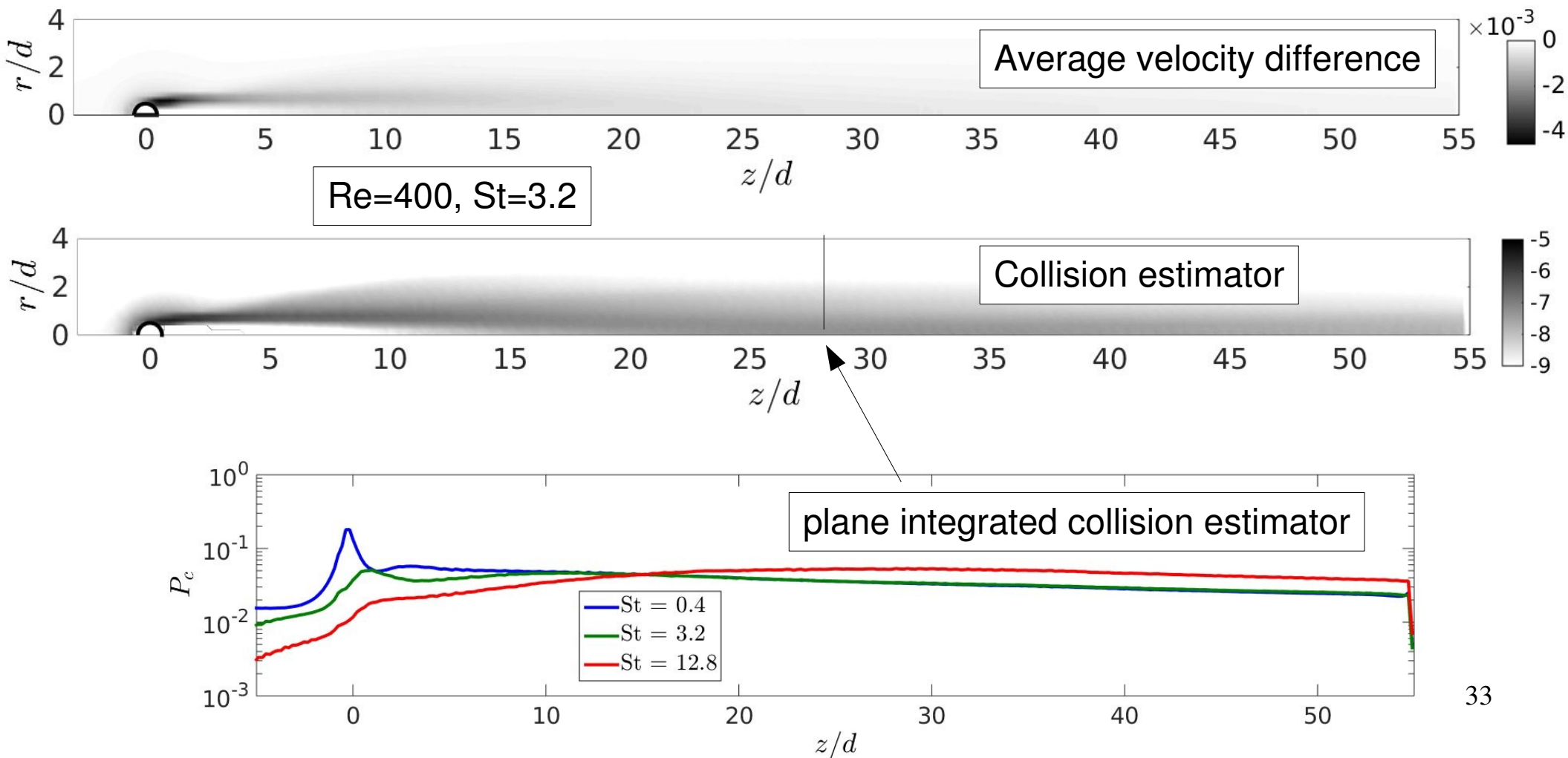
Resonant ($St \simeq 1$) ejection from vortices determines starting point of overconcentration

Inter-particle collisions

Rough collision estimator : $p_c(x) = -\langle n(x)^2 \rangle \langle w^-(x) \rangle$

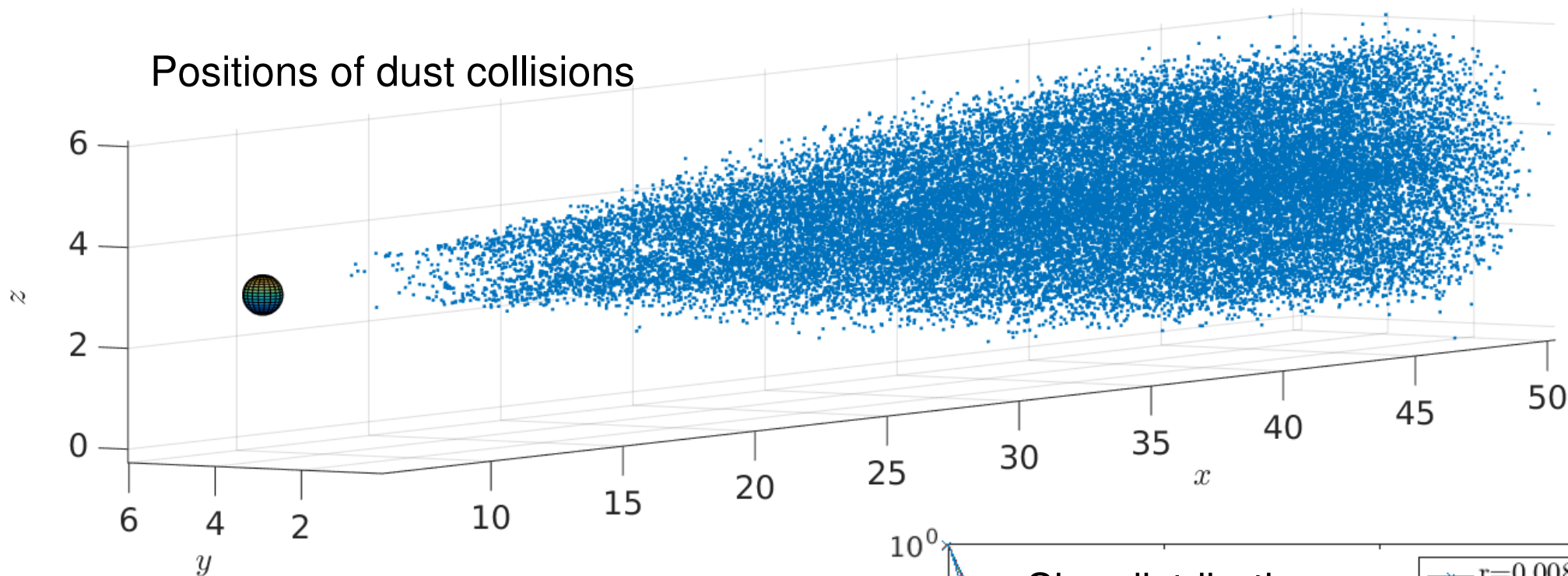
$n(x)^2$: number of particle pairs

approaching particles $\rightarrow w^-(x)$ longitudinal velocity difference

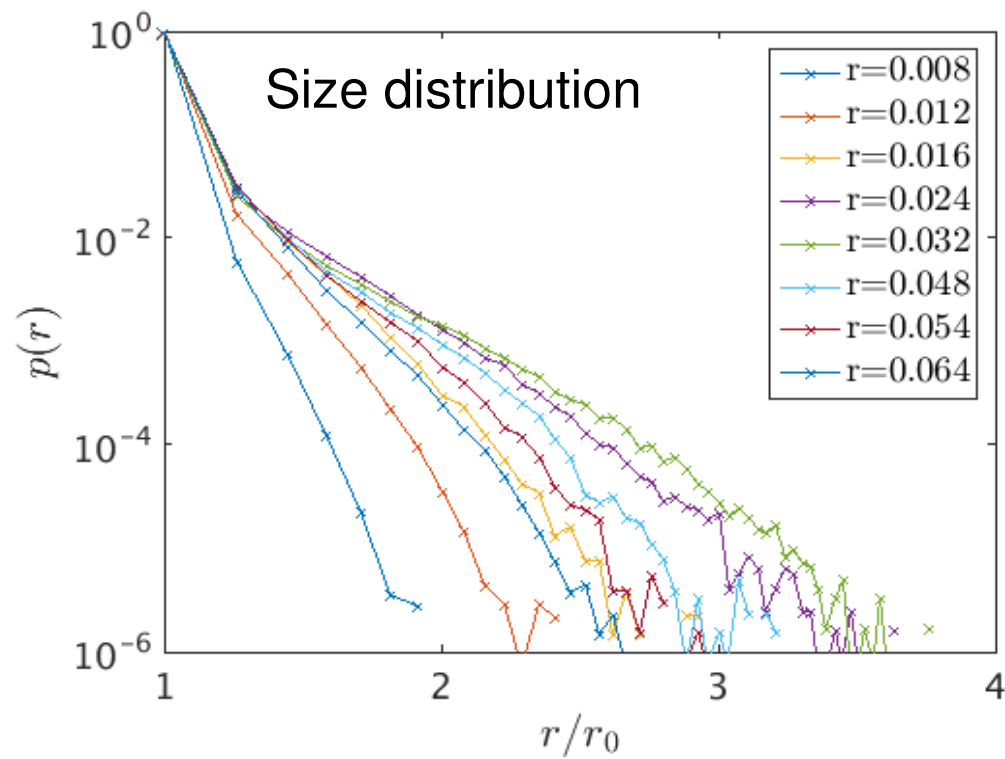


Soon with “real” inter-dust collisions

Positions of dust collisions

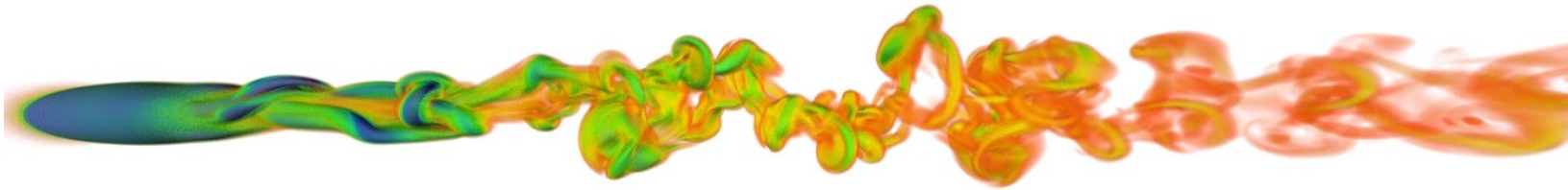


- Functional form of size-distribution
- Estimate processing efficiency of swarm of big particles (planetesimals)



The drag force

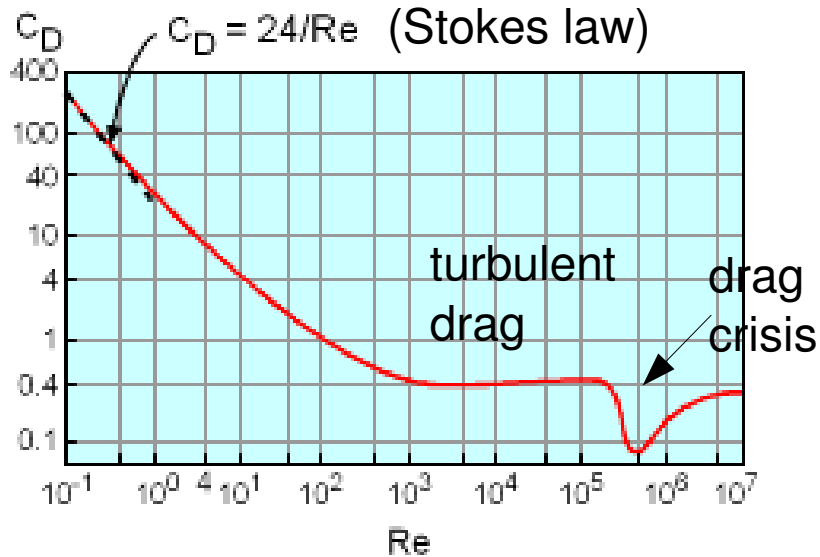
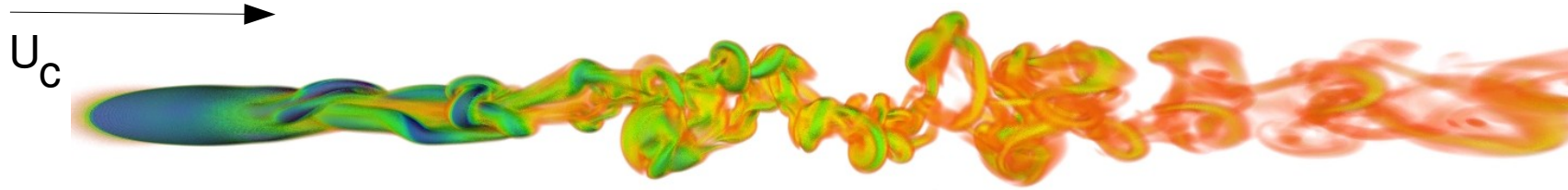
Particle in a laminar flow



Particle in a turbulent flow



The drag coefficient and flow profiles



Complicated drag force and boundary layer already for a constant inflow

Drag-coefficient

$$C_D = \frac{8F}{\rho U_c^2 \pi d^2}$$

$$Re_p = \frac{d U_c}{\nu}$$

F

U_c

d

ρ

ν

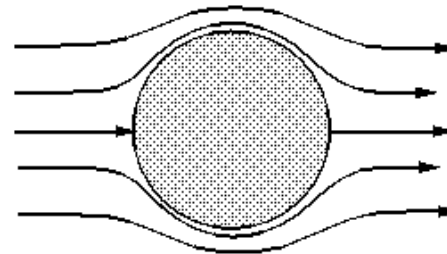
drag force

inflow speed

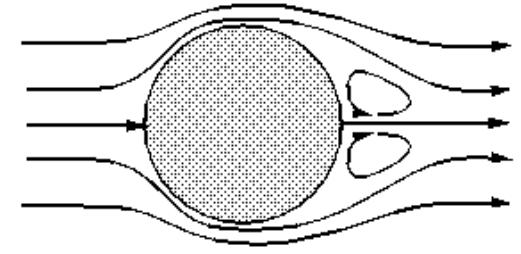
sphere diameter

fluid density

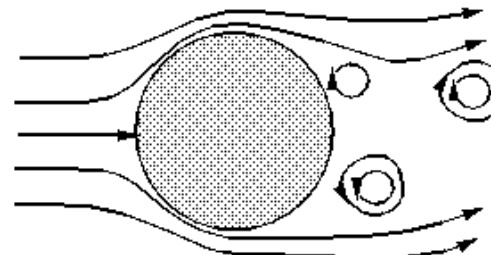
fluid viscosity



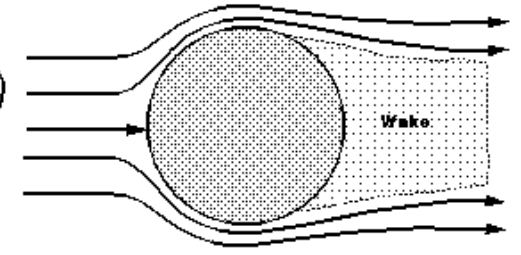
$$Re_p \lesssim 23$$



$$23 \lesssim Re_p \lesssim 350$$



$$350 \lesssim Re_p \lesssim 10^5$$



$$Re_p \gtrsim 10^5$$

The drag coefficient and flow profiles



NOW: Large sphere moving through turbulent fluid

- How does the drag force depend on turbulent fluctuations?
- How is the fluid affected?

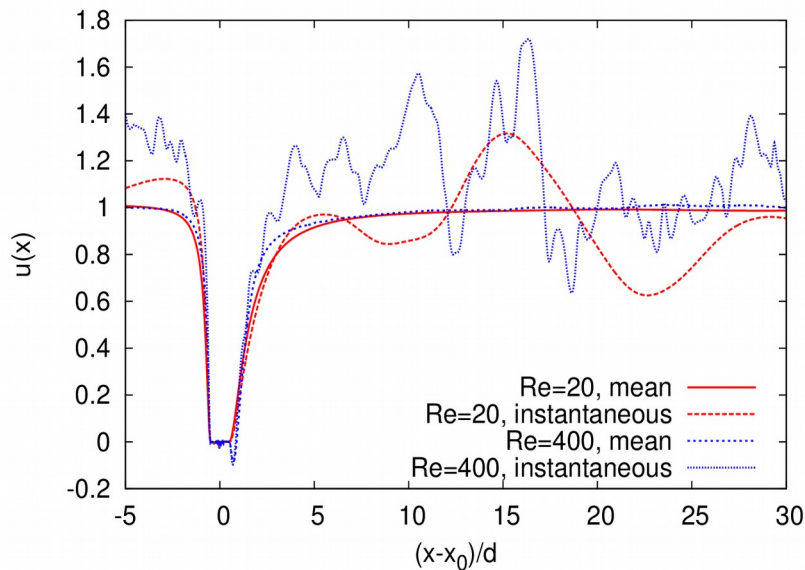
Dimensionless parameters :

- Particle Reynolds number $20 \leq Re_p = \frac{d U_c}{\nu} \leq 400$
- Fluid Reynolds number $0 \leq Re = \frac{u_{rms} L}{\nu} \leq 930$
- Turbulent intensity $0 \leq I = \frac{u_{rms}}{U_c} \leq 0.6$

Drag force from immersed boundary method: $\int_S \nabla \cdot \mathbf{T} dV = \int_S \mathbf{f}_b dV$

$$\mathbf{T} = -p\mathbf{I}_3 + \nu(\nabla \mathbf{u} + \nabla \mathbf{u}^T) \quad \text{stress-tensor}$$

Drag force – Large scale effects



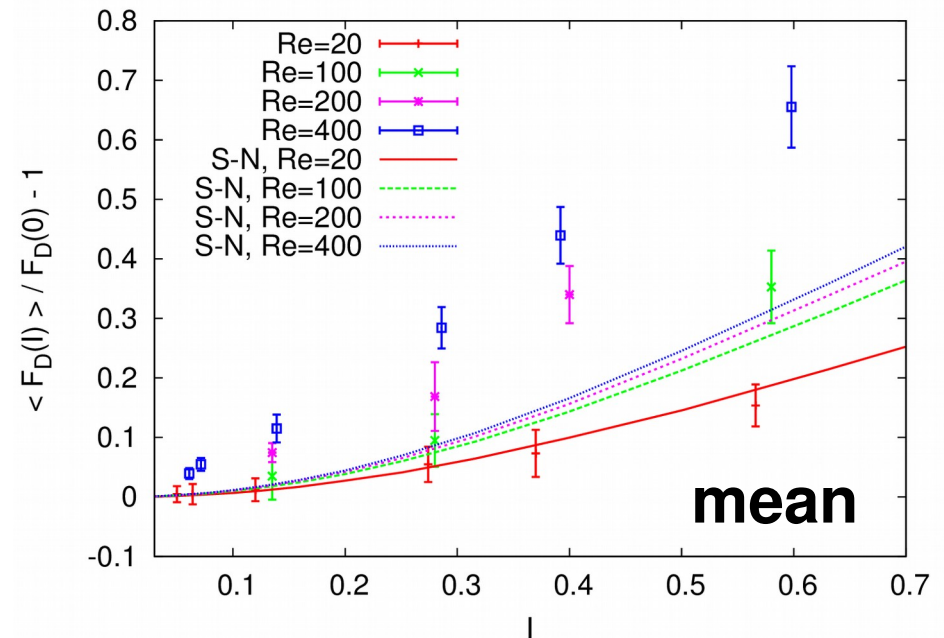
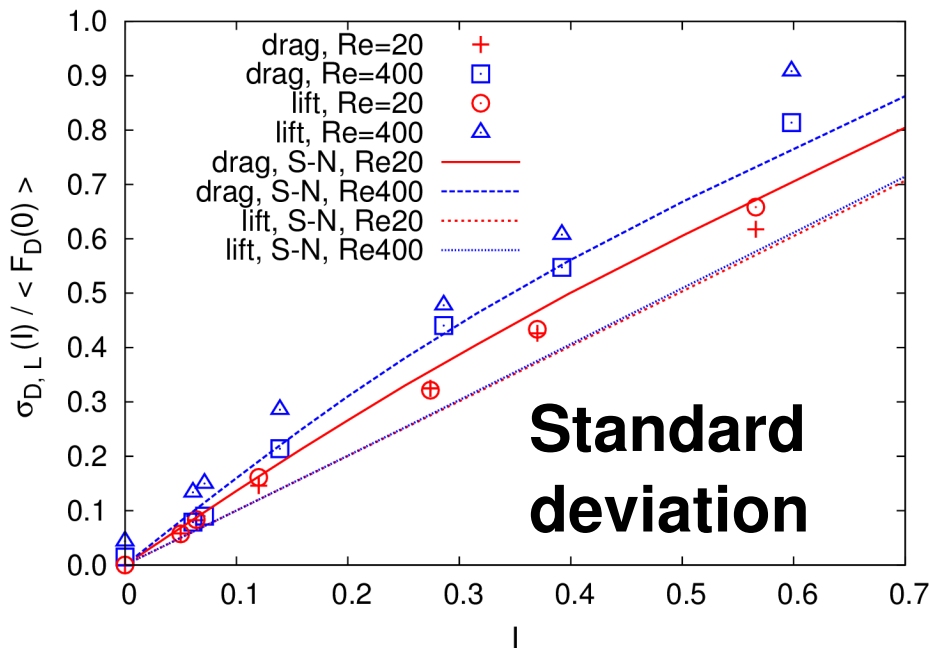
Drag is not linear

→ Effective drag force (Bagchi and Balachandar, 2003 & Kim and Balachandar, 2010)

Fitting formular of Schiller and Naumann (1933)

$$\mathbf{F} \approx \mathbf{F}^{\text{SN}}(\mathbf{U}) = 3\pi d \nu \left[1 + 0.15 \left(\frac{|\mathbf{U}| d}{\nu} \right)^{0.687} \right] \mathbf{U}.$$

Test:
$$\overline{\mathbf{F}}^{\text{SN}}(I) = \frac{1}{(\sqrt{2\pi} u_{\text{rms}})^3} \int \mathbf{F}^{\text{SN}}(\mathbf{U}) e^{-|\mathbf{U}|^2 / (2u_{\text{rms}}^2)} d^3\mathbf{U}$$



Drag force – Small scale effects

Another approach:

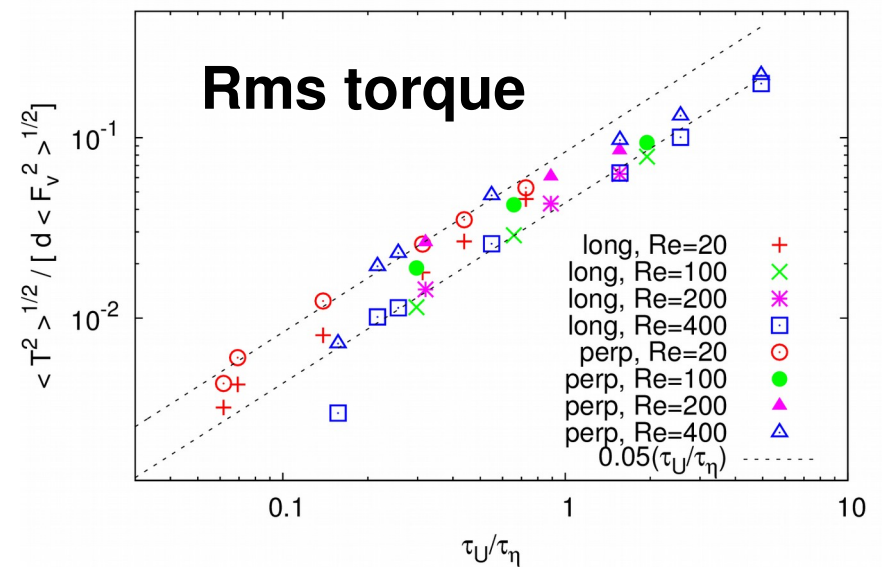
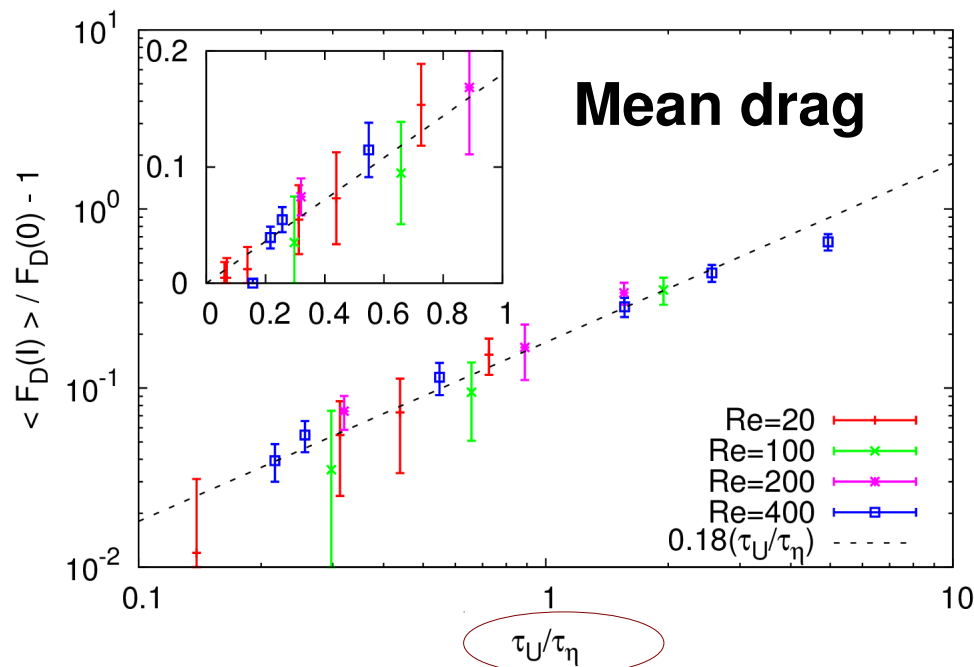
Drag increase by **modified velocity gradients** at the particle surface

Laminar gradient $\sim U_c/d$

$\tau_U = d/U_c$ Sweeping time

Turbulent gradient $u_\eta/\eta = \tau_\eta^{-1} = \nu^{-1/2}\varepsilon^{1/2}$

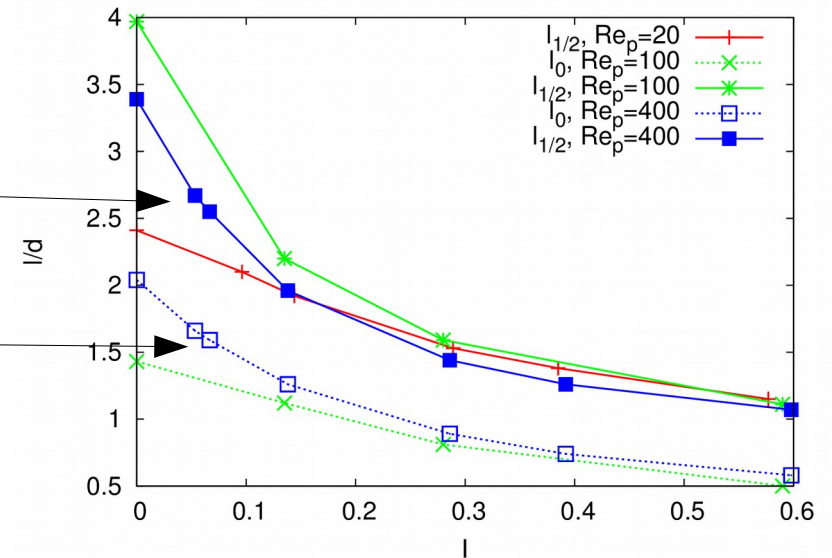
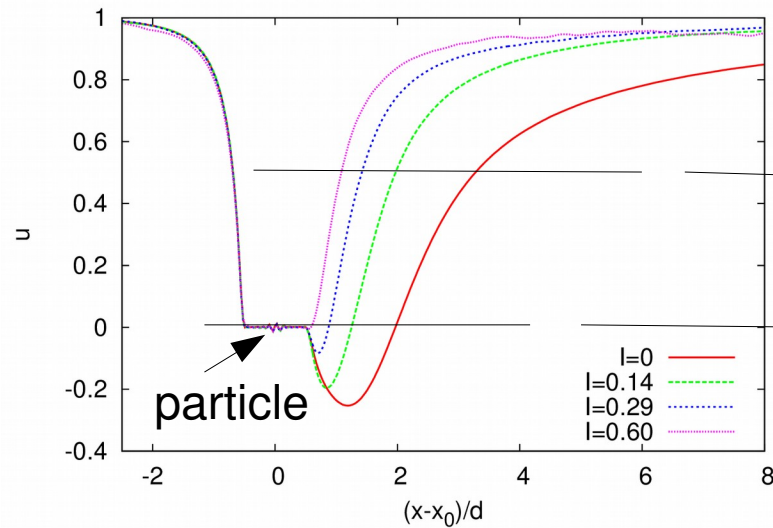
$\tau_\eta = (\nu/\varepsilon)^{1/2}$ Kolmogorov time scale



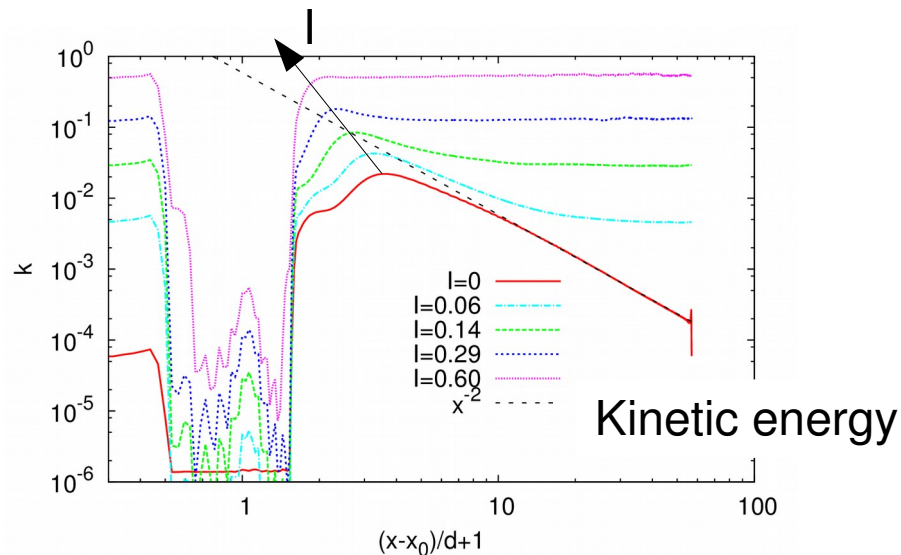
Small and large scales increase the drag on the particle

Shortening of the wake

Average velocity profile ($Re=400$)



Turbulent fluctuations significantly reduce wake length



- falls off quadratically
- maximum shifts to the particle surface

Conclusions

Outer turbulence leads to

- more dust – particle collisions
- higher dust impact speeds
- increased drag forces
- shortening of the wake

Turbulent wake leads to

- dust concentrations behind the sphere
- inter-dust collisions and heavy modification of the dust size distribution

Homann, Bec, Grauer (2013) JFM 721:155

Homann & Bec (2015) Phys. Fluids 27:053301

Homann, Guillot, Bec, Ormel, Ida, Tanga (2016) to appear in A&A

Perspectives

- Effect of rotation of sphere?
- Gravitational focusing?
- Effect of back-reaction of the particles?

Numerical implementation aspects

How to handle millions (or billions!) of particles in a numerical code?

Particle storage in HPC simulations

Array of Structure (AoS)

vs

Structure of Array (SoA)

```
struct Particle
{
    int id;

    double pos[3];
    double vel[3];

    float mass;
};

Particle particle;
particle.id = 1;

list<Particle> particleList;
particleList.push_back(Particle());
```

```
struct ParticleArray
{
    vector<int> id;

    vector<double> posX;
    vector<double> velX;

    vector<float> mass;

    void resize(int i) {
        id.resize(i);
        posX.resize(i);
        ...
    }

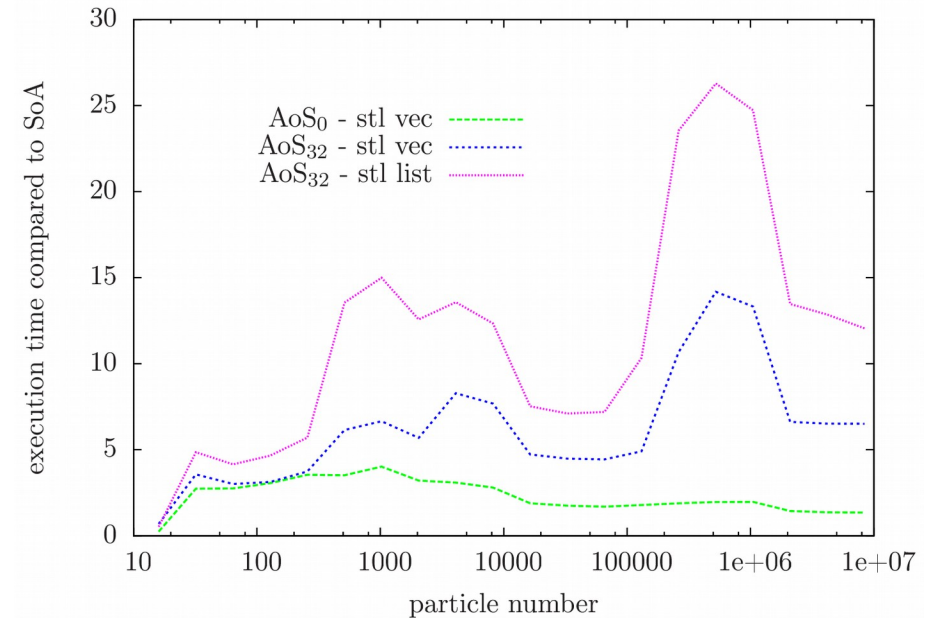
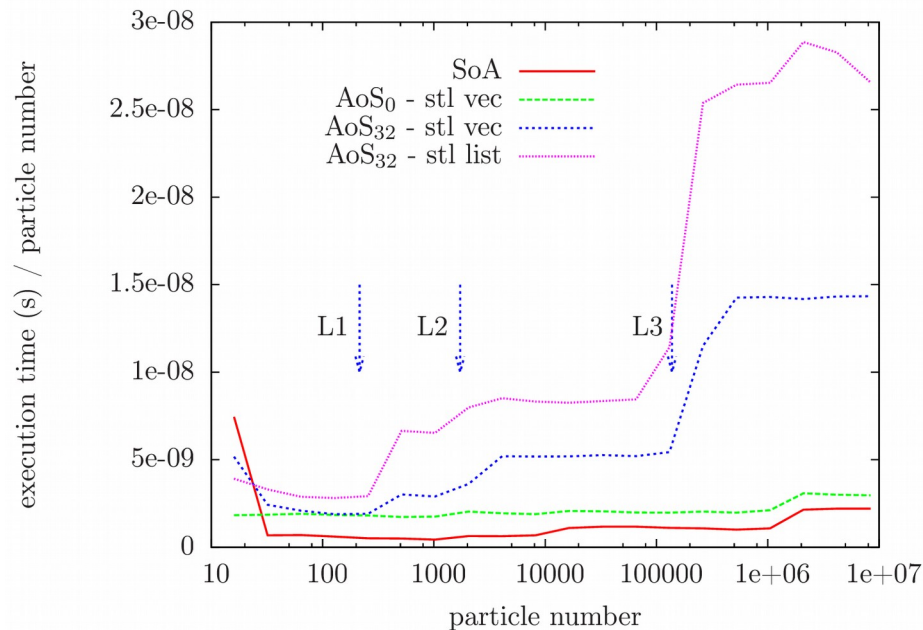
    void init() { ... }
    void output() { ... }
    ...
};
```

AoS (Particle = Object) is convenient:

- Modification of particle properties (charge, history)
- Creation/Deletion of particles
- I/O and MPI

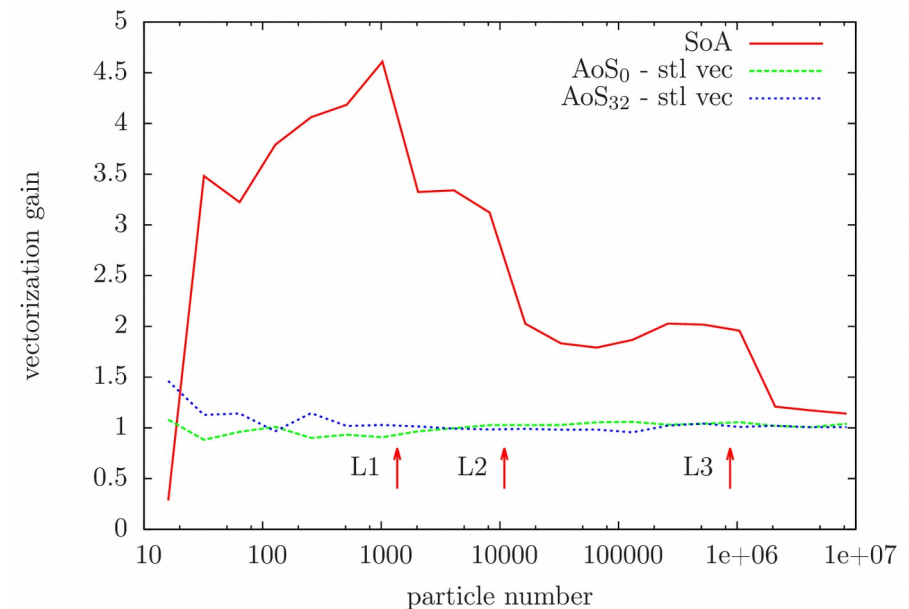
Particle storage: Benchmark for CPUs

Euler-type for-loop over all particles: $\vec{x}+ = dt \vec{v}$



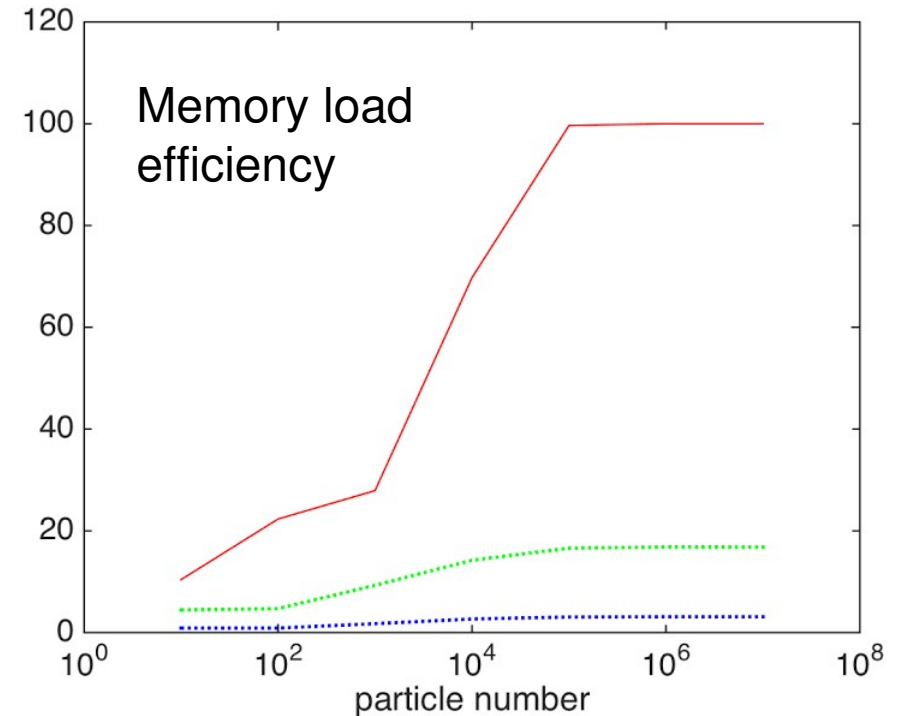
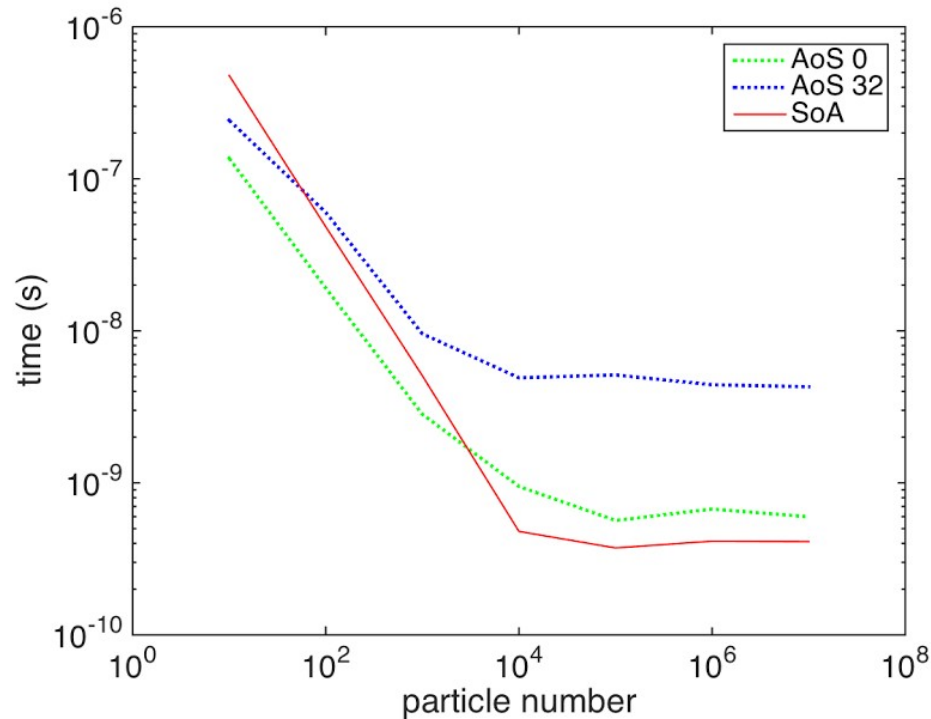
AoS are (much) slower than SoA!

- Vectorization
accelerating operations on cached data
- Cache
Pollution from unused particle properties



Particle storage: Benchmark for **GPUs**

Euler-type for-loop over all particles: $\vec{x}+ = dt \vec{v}$



AoS are slower than SoA for large particle numbers

- Memory bandwidth: Pollution from unused particle properties
- Cache benefits at small particle numbers

A fast and flexible library (SoAX)

```
// Define attributes by SOAX macro
SOAX_ATTRIBUTE(id, 'N');      // identity (number)
SOAX_ATTRIBUTE(pos, 'P');     // position
SOAX_ATTRIBUTE(vel, 'V');     // velocity
SOAX_ATTRIBUTE(mass, 'M');    // mass

// std::tuple holding attributes
// of specified type and dimension
typedef std::tuple<id<int,1>,
                  pos<double,3>,
                  vel<double,3>,
                  mass<float,1>> Attributes;

// create structure of array for 42 particles
Soax<Attributes> soa(42);

// access attributes:
soa.id(0) = 0;                // set identity
soa.pos(i,0) = 100;           // set x-coordinate

// array-wise operations using
// expression templates (x = vy - vz)
soa.posArr(0) = soa.velArr(1) - soa.velArr(2);

// extract and treat a particle as an object
auto particle = soa.getElement(7);
particle.id() = 42;
particle.pos(0) = 3.14;
soa.push_back(particle[]);
```

- Optimal Performance (same as c-arrays)
- Handyness of Arrays of Structure (particle = object)

Implementation of SoAx

- Inheritance
- C++11
- **Template meta-programming**
Simple example: Compute the factorial at compile time

```
template <int N>
struct Factorial
{
    enum { value = N * Factorial<N - 1>::value };
};

template <
struct Factorial<0>
{
    enum { value = 1 };
};
```

Factorial<11>::value is a compile time constant

Implementation of SoAx

Template meta-programming

Second example: User provided function on arrays

Set all particle data to a given value (here 42):

Code to write by user:

```
p.apply<SetToValue>(42);
```

+

Function object (SetToValue):

```
struct SetToValue  
{
```

=

Code generated at compile time

```
    template<class T, class Type>  
    static void doIt(T& t, Type value) {  
        auto tRef = *t;  
        for(int i=0;i<t->size();i++)  
            t[i] = value;  
    }  
};
```

Implementation of SoAx

Application of Functor (SetToValue) to all arrays by recursive instantiation:

```
template<class Tuple, std::size_t N, class DoItClass>
struct TupleDo {
```

```
    template<class ... Args>
    static void doIt(Tuple& t, Args... args)
    {
        TupleDo<Tuple, N-1, DoItClass>::doIt(t, args...);
        DoItClass::doIt(std::get<N-1>(t), args...);
    }
};
```

```
template<class Tuple, class DoItClass>
struct TupleDo<Tuple, 1, DoItClass> {
```

```
    template<class ... Args>
    static void doIt(Tuple& t, Args... args)
    {
        DoItClass::doIt(std::get<0>(t), args...);
    }
};
```

Conclusions

Use Structures of Arrays!

Homann & Laenen (2016) in preparation for Comp. Phys. Comm.

Thank you for your attention!

