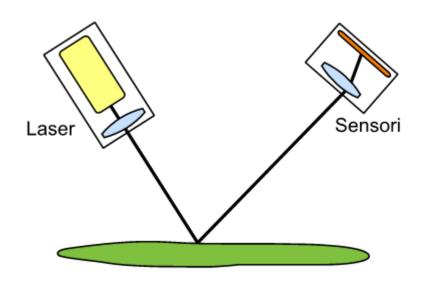
#### Simulation

Variance reduction Example

#### M C Example

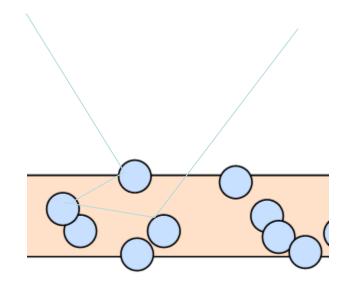
- Consider scattering of laser beam from a material layer
- MSc thesis of Jukka Räbinä 2005
- Goal is to simulate different statistics of the scattered image using Monte-Carlo

# Experimental set up



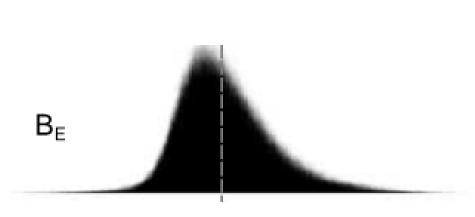
## Scattering

- Simulate propagation of ray in cloud of particles
- Basically ray tracing
- Positions and scattering directions of particles are random



#### Goal of simulation

- Compute the intensity, center of mass etc of the scatter image captured by camera
- I.e. an integral of a function involving the intensity distribution.



## Performing the simulation

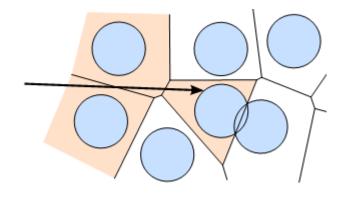
- We have to simulate several "images" and compute desired quantities with confidence intervals
- How to implement a single "image" and how does this reflect in confidence intervals
  - Point of view from variance reduction

## Straightforward approach

- Create a random particle cloud
  - Sizes prescribed, centers randomly distributed in given layer
- Simulate a fixed number of rays
  - Each ray scatters from particle surface to a random direction (case dependent distributions)
  - Count rays that reach the camera

## Straight forward approach

- Hard to find the hits to the particle cloud (additional data structures needed)
- Only few particles hit the camera

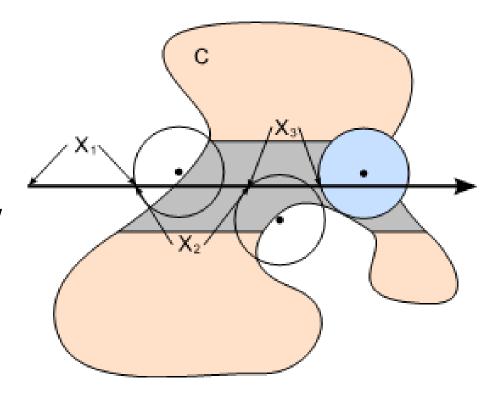


## Straight forward approach

- Limited possibilities for variance reduction
  - In practice only antithetic variables when generating the laser beam
  - Due to many collisions only small correlation between antithetic rays
- Smallish number of images (distinct particle clouds) shows up as variance in the results

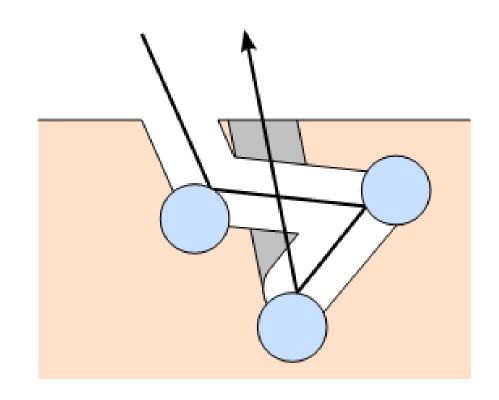
- Can the a priori fixed random particle cloud be replaced somehow
  - We can derive the expected mean free path of rays in the cloud
    - Requires statistical analysis/understanding of the situation
  - Particles/collision can be generated dynamically
    - Draw the free path and angle of attack to the next particle -> next collision can be modelled

- Draw next free path (Expdistributed) and distance of center point from ray-line (Unif-distrib.) -> we can define the centerpoint for new particle
- If center in the area of interest, generate particle and compute collision, otherwise draw a new free path to the same direction
- The collision is modelled as for a fixed particle



- Computation is lighter (60-85%)
  - Only needed particles are generated
  - No search is needed to find collisions
- For each ray a new collection of particles
  - Smaller variance (by 50%) as images are based on average particle cloud instead of a fixed cloud

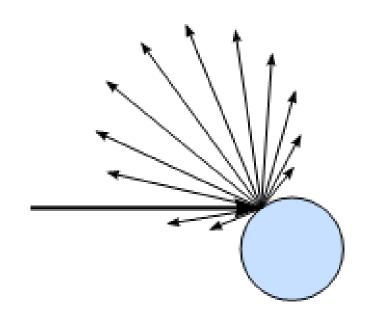
- Results are biased if ray history is not accounted in generation of particles
- Direct backscattering should not be blocked on the way back
- The previous path (and observed particle free zone has to be remembered
- Computing time grows (max 50%)
- Still twice faster and more accurate than basic approach



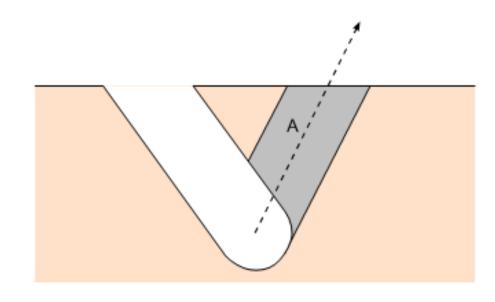
- Use of dynamic particles did not change the frequency of rays hitting the camera
  - Ray probagates unattennuated through the simulation
  - Only a small fraction reaches the camera
  - Can we increase the number of rays contributing to the image

- On each collision divide the intensity to two parts
  - Compute the expectation of the intensity scattering to the direction of the camera
  - Ray with this intensity is sent towards the camera (given direction, random mean path and accounting for known particles)
  - Rest of the energy is scattered as one ray to a random direction

 Requires that particles scatter the rays (no specular reflections)



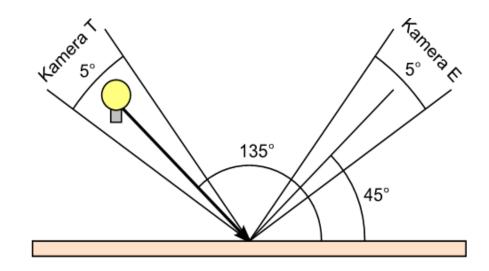
- Does the ray heading towards camera get through
- If we know of a particle that blocks the route to the camera, the ray is lost for sure
- Otherwise we draw a free path and see if it leads out from the particle layer



- Big fraction of collisions can send some energy towards the camera
  - About twice more computation
- More hits to a single pixel
- Hits have smaller intensity (so each hit has smaller effect to the image)
- -> Smaller variance (less than 1/5 compared to dynamic particles)

## Simulation experiment

- Send parallel rays with normally distributed intensity
- Collect the (few) rays scattered to the camera



#### Simulated results

- Same amount of rays and images using three methods (S static, D dynamic particles, E expectations.)
- E method about 250 times more efficient than the static approach
- Most of efficiency comes from reduced variance

A <sub>E</sub>	Intensiteetti
S	$1.81881 \pm 0.00547$
$D_{\infty}$	$1.81603 \pm 0.00274$
$\mathbb{E}_{\infty}$	$1.81688 \pm 0.00044$

	Aika T <sub>A</sub>
S	1505s
$D_{\infty}$	391 <i>s</i>
$\mathbb{E}_{\infty}$	840s