## Efficient Rendering of Heterogeneous Polydisperse Granular Media, Supplemental Material

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1 Implementation

In this section we provide additional implementation details.

Heterogeneous Grain Mixtures. We implement the assignment of grain types to bounding spheres as *Mixtures*. Such a *Mixture* implements a function which takes a bounding sphere's index and position as input arguments and maps these to natural numbers representing grain types. Useful implementations of *Mixtures* include *explicit* mixtures—e.g. from simulation data—which load an explicit mapping from disk, and various procedural mixtures, such as the homogeneous deterministic pseudo-random mixture used by Meng et al. [2015]. Additionally, we implemented a mixture which selects one of two grain types based on Perlin noise [Perlin 2002], or a checkerboard. We show examples of supported mixtures in Figure 1.

Fast Procedural Grain Instantiation. Similarly to Meng et al. [2015], we support procedural instantiation of grains filling a scene-provided watertight bounding mesh. We adopt their tile-based approach, where a pre-computed sphere packing inside a unit cuboid is repeated ad infinitum in a 3D grid. The key difference between our approach and their approach is which bounding spheres are considered to be *inside* the bounding mesh. They require bounding sphere centers to lie within the mesh, whereas we require the entire spheres to be contained in the mesh. Our slightly more strict definition allows us to only intersect those bounding spheres which lie along a ray segment passing through the mesh. Meng et al. [2015], in contrast, resort to explicitly labeling the tiles within the meshes bounding box in a per-scene pre-computation step, marking those tiles in which bounding spheres should be intersected. This can lead to wasted computation before and during rendering, since a single tile can contain thousands of bounding spheres, of which many may lie outside of the mesh. For each of these intersected spheres a check has to be performed to truly know whether it lies within the bounding mesh. For both their method and ours a simple way exists for checking whether a given bounding sphere lies within the mesh. Meng et al. [2015] shoot a ray from the center of the bounding sphere, and evaluate whether the ray intersects the mesh from inside or from outside. We traverse a kd-tree of the scene to check whether there are triangles closer to the center of the bounding sphere than the sphere's radius. Finally, after a bounding sphere which passes the aforementioned checks has been hit, a grain is instantiated and path-traced.

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**PPT and ST Rendering Pseudocode.** Algorithms 1 and 2 describe the procedure of path-tracing when hitting a grain's proxy, and shell tracing in continuous volumes, respectively.

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## 2 Shell Transport Functions

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In this section we provide additional material supporting our claims about STFs.

**2D Cosine Hemisphere.** In Figures 2 and 3 we show slices through the 2D directional STF component of the 4D STFs described by Moon et al. [2007] when computed on homogeneous continuous media derived from snow and dielectric spheres, respectively. As can be seen, the directional components of STF with large radii converge to cosine hemispherical distributions. Small shells generally have a small multiple-scattering component, and thus the worse approximation of the multiple-scattering lobe with a cosine-weighted hemisphere does not play as big a role.

**Investigation of Approximations.** In Figure 4 we show the effects of our various STF approximations on performance and bias. We provide TTUV and MRSE values for each approximation step we perform, and across a wide range of volume densities.

## **3** Grain-Scattering Distribution Functions

In this section we provide additional material supporting our claims about GSDFs.

**Investigation of Approximation.** In Figure 5 through Figure 32 we visualize the approximation error introduced by our GSDFs in the directional domain  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  for various grain types. Figures 9 through 32 follow a common color scale, whereas Figures 5 through 8 have their own color scale in order not to overexpose the visualization.

## References

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(a) Homogeneous mixture

(b) Checkerboard mixture

(c) Explicit mixture

**Figure 1:** Various supported grain mixtures. Grains can be mixed homogeneously (a), arranged in a procedural 3D checkerboard pattern (b) or distributed explicitly; e.g. governed by a simulation (c).

Algorithm 1 Path Tracing with Proxy Objects

 $\begin{array}{l} \mbox{function HANDLEPROXYVERTEX}(\mathbf{x}_{o}, \vec{\omega}_{o}) \\ (f_{g}, \mathbf{x}_{c}) \leftarrow \mbox{GETPROXYSPHERE}(\mathbf{x}_{o} \ ) \\ \vec{n}_{o} \leftarrow (\mathbf{x}_{c} - \mathbf{x}_{o}) / \|\mathbf{x}_{c} - \mathbf{x}_{o}\| \\ \beta_{o} \leftarrow \cos^{-1}(\vec{\omega}_{o} \cdot \vec{n}_{o}) \\ \alpha_{g} \leftarrow \alpha_{g}^{0}(\beta_{o}) + \alpha_{g}^{+}(\beta_{o}) \\ t \leftarrow \alpha_{g}t \\ \xi \leftarrow \mbox{sample uniformly from } [0, 1) \\ \mbox{if } \xi \leq \alpha_{g}^{0}(\beta_{o}) / \alpha_{g} \ \mbox{then} \\ \vec{\omega}_{i} \leftarrow \vec{\omega}_{o} \\ \mathbf{x}_{i} \leftarrow \mathbf{x}_{o} + 2(\vec{o}_{i} \cdot \vec{\omega}_{o}) \vec{\omega}_{o} \\ \mbox{else} \\ (\mathbf{x}_{i}, \vec{\omega}_{i}) \leftarrow \mbox{sample from } p_{g}(\cdot, \cdot | \beta_{o}) \\ L_{i} \leftarrow L_{i} + t \cdot \mbox{DIRECTLIGHT}(\mathbf{x}_{i}, p_{g}^{\vec{\omega}}(\cdot | \beta_{o})) \\ \mbox{end if} \\ \mbox{end function} \end{array}$ 

Algorithm 2 Path Tracing with Shell Transport Functions

1: **function** HANDLESHELLVERTEX( $\mathbf{x}_c, \vec{\omega}_c$ )  $(f_s, r) \leftarrow \text{GetLargestFittingShell}(\mathbf{x}_c)$ 2: 3: if  $r < r_1$  then 4: USEVPT $(\mathbf{x}_c, \vec{\omega}_c)$ 5: else  $\alpha_s \leftarrow \alpha_s^0(r) + \alpha_s^1(\alpha, g, r) + \alpha_s^m(\alpha, g, r)$ 6: 7:  $t \leftarrow \alpha_s t$  $\xi \leftarrow$  sample uniformly from [0, 1) 8: if  $\xi < \alpha_s^0(r)/\alpha_s$  then 9:  $\mathbf{x}_s \leftarrow \mathbf{x}_c + r\vec{\omega}_c$  $\vec{\omega}_s \leftarrow \vec{\omega}_c$ 10: 11: else if  $\alpha_s^0(r)/\alpha_s \leq \xi < \alpha_s^1(\alpha, g, r)/\alpha_s$  then 12: 13:  $d_1 \leftarrow \text{sample uniformly from } [0, 1)$  $\mathbf{x} \leftarrow \mathbf{x}_c + d_1 r \vec{\omega}_c$  $14 \cdot$ 15:  $f_p \leftarrow \text{GetMediumPhase}(\mathbf{x})$  $\vec{\omega}_s \leftarrow \text{sample from } f_p(\cdot)$ 16:  $\begin{aligned} & \theta_s \leftarrow \cosh^{-1}(\vec{\omega}_c \cdot \vec{\omega}_s) \\ & \theta_z \leftarrow r\sqrt{1 - d_1^2 \sin^2(\theta_s)/r^2} - d_1 \cos \theta_s \end{aligned}$ 17: 18:  $\mathbf{x}_s \leftarrow \mathbf{x} + d_2 \vec{\omega}_s$ 19:  $L_i \leftarrow L_i + t \cdot \text{DIRECTLIGHT}(\mathbf{x}, f_p(\cdot))$ 20: else 21:  $(\mathbf{x}_s, \vec{\omega}_s) \leftarrow \text{sample from } p_s^m(\cdot, \cdot | \alpha, g, r)$ 22:  $L_i \leftarrow L_i + t \cdot \text{DIRECTLIGHT}(\mathbf{x}_s, p_s^{m, \vec{\omega}}(\cdot | \alpha, g, r, \theta))$ 23: 24: end if 25: end if 26: end function

 $\triangleright \text{ Obtain GSDF } f_g \text{ and grain center } \mathbf{x}_c \\ \triangleright \text{ Compute proxy normal} \\ \triangleright \text{ Compute inclination angle} \\ \triangleright \text{ Compute grain albedo conditional on } \beta_o \\ \triangleright \text{ Adjust path throughput} \\ \triangleright \text{ Randomly choose scattering type} \\ \triangleright \text{ No scattering} \end{cases}$ 

Pass straight through bounding sphere
 Scattering
 Importance sample scattering
 Accumulate direct lighting

▷ There is no shell which fits
▷ Fall back to VPT

▷ Compute shell albedo conditional on α, g, r
 ▷ Update path throughput
 ▷ Randomly choose scattering type
 ▷ No scattering
 ▷ Pass through the shell

Single scattering
 Sample free-flight distance
 Compute single-scattering location

▷ Sample medium phase function
 ▷ Compute scattering angle
 ▷ Compute distance to shell surface
 ▷ Compute outgoing location on shell surface
 ▷ Compute direct light
 ▷ Multiple scattering
 ▷ Importance sample multiple scattering from tabulation
 ▷ Accumulate direct light



**Figure 2:** Directional multiple scattering component of the shell transport function corresponding to snow. This figure visualizes the dependence of the shape of the directional component on the shell's radius r (horizontal) and the teleportation angle  $\theta_t$  (vertical). As r increases, the shape of the directional component approaches a cosine hemisphere.



**Figure 3:** Directional multiple scattering component of the shell transport function corresponding to dielectric spheres. This figure visualizes the dependence of the shape of the directional component on the shell's radius r (horizontal) and the teleportation angle  $\theta_t$  (vertical). As r increases, the shape of the directional component approaches a cosine hemisphere.



**Figure 4:** We demonstrate the error introduced by each of the approximations we apply to STFs in the LUCY scene, which contains the Lucy statue filled with a homogeneous continuous participating medium derived from densely packed snow grains. The first column shows volumetrically path traced (VPT) reference images, which can be matched perfectly by the 4D STF in the case of homogeneous continuous media. The following three columns demonstrate the visual impact of our approximations: 2D STFs approximate  $p_s^{m,\vec{\omega}}$  as a cosine-weighted hemispherical distribution, using HG phase function significantly reduces the dimensionality of continuous-volume-appearance space to permit dense tabulation, and interpolation between tabulated STFs allows fitting STF tightly into the mesh. The last column shows the diffusion approximation employed by Meng et al. [2015], which yields higher error than all our approximations combined.



**Figure 5:** Front-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on sugar grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 6:** Back-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on sugar grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 7:** Front-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on brown sugar grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 8:** Back-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on brown sugar grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 9:** Front-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on ice grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 10:** Back-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on ice grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 11:** Front-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on snowflake grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 12:** Back-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on snowflake grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 13:** Front-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on white sand grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 14:** Back-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on white sand grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 15:** Front-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on yellow sand grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 16:** Back-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on yellow sand grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 17:** Front-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on brown sand grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 18:** Back-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on brown sand grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 19:** Front-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on dark brown sand grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 20:** Back-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on dark brown sand grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 21:** Front-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on black sand grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 22:** Back-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on black sand grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 23:** Front-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on flour grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 24:** Back-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on flour grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 25:** Front-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on hairball grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 26:** Back-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on hairball grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 27:** Front-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on cinnamon grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 28:** Back-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on cinnamon grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 29:** Front-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on dielectric sphere grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 30:** Back-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on dielectric sphere grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 31:** Front-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on diffuse sphere grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.



**Figure 32:** Back-facing directional GSDF approximation error  $e_{f_g}(\vec{\omega}_i, \vec{\omega}_o)$  on diffuse sphere grains. We visualize this 4-dimensional error function by plotting slices through  $\cos \beta_o$  (vertical) and  $\gamma_o$  (horizontal). The individual circular heatmaps encode  $\sin \beta_i$  as the distance from their center and  $\gamma_i$  as their rotational component.