

# A BSSRDF Model for Efficient Rendering of Fur with Global Illumination Supplemental Materials

## 1 Micro-flakes approximation of scattered components

In this section, we elaborate on Section 4.4 of the main paper. We derive the parameters of a dipole model to approximate the scattered components of a hair/fur model based on the micro-flakes theory [Jakob et al. 2010]. Suppose  $\bar{r}$  is the average radius of the hair/fur volume. We approximate each hair/fur fiber as a set of spherical micro-flakes with radius  $\bar{r}$  compactly distributed along the hair/fur fiber as in Fig. 1. The phase function of each micro-flake is the normalized BCSDF of the hair/fur fiber that it approximates.

The micro-flakes theory estimates volume scattering parameters as follows:

$$\sigma_a = (1 - \mu)A(\omega_i)\rho, \quad \sigma_s = \mu A(\omega_i)\rho, \quad g = \int_{S^2} p(\omega_i, \omega_r)(\omega_i \cdot \omega_r) d\omega_r \quad (1)$$

where  $\mu$  is the albedo of each micro-flake,  $A(\omega_i)$  is the projected area of each micro-flake along direction  $\omega_i$ , the parenthesis is the dot product operator of  $\omega_i$  and  $\omega_r$ ,  $\rho$  is the micro-flakes density and  $p$  is the phase function.  $\omega_i$  and  $\omega_r$  are the incoming and outgoing directions respectively.

Using this approximation,  $\mu$  is the ratio of the energy scattered by the micro-flake to the total input energy, which can be estimated by a numerical integration of the BCSDF. Since the integral depends on the incoming longitudinal angle  $\theta_i$ , we estimate the average of the integrals over all potential angles in the range  $[0, \pi]$ . Similarly, for  $g$ , we numerically estimate the integral and average over all potential longitudinal angles. The projected area  $A(\omega_i)$  of each micro-flake is  $\pi\bar{r}^2$ , which is simply the projected area of the sphere. For the density  $\rho$ , we first estimate the number of micro flakes by  $n = L/(2\bar{r})$ , where  $L$  is the total length of all fibers. This is because each sphere takes up  $2\bar{r}$  of the total fiber length and all spheres are compactly distributed. Then, we estimate the volume  $V$  that the fibers take up in a coarse voxel grid and  $\rho = n/V$ . At this point, we have all the components to obtain  $\sigma_a$ ,  $\sigma_s$  and  $g$  for the dipole model.

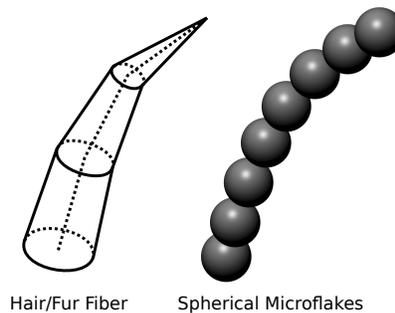


Figure 1: Approximating a hair fiber with spherical microflakes.

## 2 Importance sampling of dual scattering

### Problem overview

Dual scattering is essentially adding 2 lobes to the original hair/fur BCSDf. To sample dual scattering, we are essentially sampling the overall BCSDf, including  $R$ ,  $TT$ ,  $TRT$ ,  $TT^s$  and  $TRT^s$  lobes, together with the backward scattered lobe

$$S_b(\theta_i, \theta_r, \phi) = \frac{I_b(\phi)}{\pi \cos^2 \theta_i} \cdot d_b A_b \cdot G(\theta_r + \theta_i; \bar{\sigma}_b^2), \quad (2)$$

and the forward scattered lobe

$$S_f(\theta_i, \theta_r, \phi) = \frac{I_f(\phi)}{\cos^2 \theta_i} \cdot d_f A_f \cdot \sum_p (G(\theta_r + \theta_i; \bar{\beta}_p^2 + \bar{\sigma}_f^2) N_p^f). \quad (3)$$

### Method overview

We follow the high-level importance sampling scheme in Yan et al. [2017]. We first select one of all these lobes based on the energy each lobe contains, then we sample the selected lobe. The selection probability is denoted as  $P$ , which is proportional to the energy each lobe carries.

When sampling the selected lobe, we separately sample it longitudinally and azimuthally with probability density functions (PDFs)  $p_M$  and  $p_N$ , respectively. Then, the PDF of sampling the entire sphere is simply converted as  $p_M \cdot p_N / \cos(\theta_o)$ . The sampling weight is the BCSDf value of the selected lobe at the sampled direction  $\omega_o$ , divided by the probability  $P$  of selecting it.

### Sampling $S_b$ and $S_f$

First, we calculate the energy  $S_b$  and  $S_f$  carry, denoted as  $E_b$  and  $E_f$ . This is done simply by taking out the distribution related terms from Eqn. 2 and 3. We immediately have

$$E_b(\theta_i, \theta_r, \phi) = \frac{1}{\cos^2 \theta_i} \cdot d_b A_b, \quad (4)$$

and

$$E_f(\theta_i, \theta_r, \phi) = \frac{\pi}{\cos^2 \theta_i} \cdot d_f A_f \cdot \sum_p N_p^f. \quad (5)$$

Now, suppose we have already selected a lobe  $p$ . If  $p \in R, TT, TRT, TT^s, TRT^s$ , it is already handled by Yan et al. [2017] and we do not repeat the procedure here.

If the backward scattered lobe  $S_b$  is selected, we first importance sample it longitudinally according to the Gaussian  $G(\theta_r + \theta_i; \bar{\sigma}_b^2)$ , which is simply done by using the Box-Muller transform to sample a Gaussian centered at  $-\theta_i$  with variance  $\bar{\sigma}_b^2$ . And the PDF  $p_M$  is the Gaussian value at the sampled position.

Then, we uniformly sample  $\phi \in [-\pi/2, \pi/2]$  azimuthally with PDF  $p_N = 1/\pi$ , according to the backward hemisphere indicator  $I_b$ .

If the forward scattered lobe  $S_f$  is selected, since it consists of more sub-lobes (because the forward scattered lobe is essentially a modification of single hair/fur fibers' BCSDf with all lobes), we continue the lobe selection process within these sub-lobes to find one sub-lobe with probability  $Q$  (i.e. all probabilities of sampling the sub-lobes sum up to 1), then we're essentially sampling

$$\tilde{S}_f(\theta_i, \theta_r, \phi) = \frac{I_f(\phi)}{\cos^2 \theta_i} \cdot d_f A_f \cdot G(\theta_r + \theta_i; \bar{\beta}_p^2 + \bar{\sigma}_f^2) N_p^f / Q, \quad (6)$$

which is now similar to sampling the backward scattered lobe  $S_b$ .