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4 Ease-to-use approach of BRDF recovery from 360°

- 5 Light Probe
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1 Abstracts

- $\mathbf{2}$ We present an easy-to-use measurement of BRDF for material appearance of three-dimensional object 3 based on the estimation of surface reflectance properties with 360-degree light probe. Conventional 4 methods of BRDF measurement such as gonio-reflectometry have some problems relating to the size, $\mathbf{5}$ cost, and difficulty of operation. Appearance from Motion Method proposed by Dong et al. is 6 appropriate to conquer above problems. Their method attempted to estimate the surface reflectance $\overline{7}$ under unknown incident light component. However, a use of multiple unknown parameters causes the 8 loss of convergence, and sometimes sacrifices the accuracy and computational cost. Therefore, we 9 improved their algorithm of Appearance from Motion method by using an omni-directional camera as 10 a light probe. We demonstrate valuable results: Higher reproducibility and lower cost of computation 11 than those of the conventional method. 1213**Keywords**
- 14 BRDF, Measurement, 360-degree light probe, Reproduction
- 15

1 **1. Introduction**

The important information of surface reflectance is useful for the field of computer graphics. This information is defined as bi-directional reflectance distribution function (BRDF) to calculate an appearance of object. Especially, since Spatially-Varying BRDF (SVBRDF) has information of spatially variation of the objects, therefore, SVBRDF is used for the rendering of virtual object with multiple textures. Various profile of this BRDF changes the material appearance based on dichromatic reflection model. Specular reflection represents the surface smoothness by its degree of reflection.

8 Conventionally, the use of BRDF information was limited in the field of computer science. 9 Recent revolution of three-dimensional (3D) reproduction gives us the useful chance of BRDF 10 information in commercial production. Actually, there are many applications of internet shopping, 11 reproducing in digital mock-up, arts, and regenerative medicine. The most expected contribution of 12 BRDF information is application for rapid proto-typing that is compatible to final product with same 13 property of surface roughness. Therefore, the method of BRDF measurement attracts attention as the 14 essential tool for advancement of 3D reproduction.

Various measurement methods have been proposed for the accurate BRDF of real object. However, the currently used method or instruments such as gonio-reflectometer has large scale, needs high calculation cost and difficult to acquire BRDF information. Moreover, it is difficult for non-expert user to measure the appearance of the object. For this reasons, the measurement system, which is commercially available, rapid and compact, is required for handling the appearance.

In this paper, therefore, we propose a rapid and compact measurement system that is expected to be equipped with 3D reproduction system. Our method refers to Appearance from Motion method proposed by Dong et al.¹), whereas we make a goal to ascend the industrial value by adding an extra procedure which is the use of environment map captured by omni-directional camera instead of the estimation of an incident lighting component. This improvement provides the accuracy and simplification compared with the conventional method.

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27 **2. Related Work**

28Appearance reproduction is well-studied with BRDF, which is physical model for 29representing photon propagation. There are many kinds of appearance measurement method or 30 instruments as previous works ²⁾⁻⁵⁾. In general, the recovery of object appearance needs three 31information, such as geometry (normal vector), surface reflectance (BRDF), and incident lighting 32component $^{(6)}$. When the shape of object is assumed as a flat plate to fix the geometrical parameter, the 33 gonio-reflectometer can be useful to acquiring the BRDF parameters for reproduction of various 34 appearances. Unfortunately, accurate BRDF measurement needs 6 DOF movement and precious 35position arrangement. Therefore, a tangled repeat setting and long measurement time are required. In 36 order to decrease the time and effort, multiple lighting or wide view camera system was proposed to

1 detect much reflectance information at one shot ⁷⁾⁻¹⁴⁾. Although these ideas can reduce the 2 measurement cost dramatically, the flexibility of the BRDF measurement was lost by reason of fixed 3 and limited measurement equipment.

4 A breakthrough idea for this complicated BRDF measurement was proposed by Dong et al. $\mathbf{5}$ in ACM SIGGRAPH Asia 2014¹⁾. Their method can reconstruct the BRDF information at spatially 6 varying isotropic surface (SVBRDF) by solving ill-posed problem between the reflectance and 7 illumination without the knowledge of incident light. Only capturing the video while rotating the 8 objects, non-expert users are also accessible and easy to control, because acquisition scheme of this 9 method is greatly simplified. In this method, they estimated three unknown variables; distribution of 10 specular highlights, diffuse albedo, and lighting component, under the condition that the shape of 11 object is known. Moreover, their group expanded their method to estimate the shape information with 12the estimation of reflectance and illumination by using the temporal trace processing from the 13 measured data 15).

14Although it is arguable for the estimation of all information since the measurement 15technology for 3D shape is remarkably developed ¹⁶, their state-of-the-art method is undeniably useful 16 for the reconstruction of object appearance. This method proposed by Dong et al. was based on the 17attempt to set various restrictions to solve this complex problem and demonstrated feasible result of 18 acquiring the surface reflectance. However, the estimation of several variables such as albedo, specular, 19 and illumination in their method causes less convergence and more computational cost. Moreover, 20lighting component calculated by their method indicated uncertain artifacts in the result of their 21estimation. For example, the edge of estimated lighting components are affected by target object or 22immoderately saturated in Fig. 17 of SIGGRAPH paper, since their method converges their calculation 23when the edge of illumination was sharpened. Here, it is noticed that the estimation of lighting 24component will have an insignificant importance because incident lighting is easy to measure by using 25omni-directional cameras that become compact and low price. Therefore, we improve and simplify 26the Dong's method for the guarantee of convergence and rapid estimation under known lighting 27condition, which is derived environmental image captured by omni-directional cameras.

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29 **3. Method**

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3.1 Acquisition and Calibration

Our proposed method requires the measured images of object that is rotating on the table, environmental image to provide an incident lights on the target object, and geometry data of target object. We captured images of object on the rotating table by digital camera (D5100, Nikon) equipped with varifocal lens (AF-S NIKKOR 24-120mm, Nikon) as shown in Fig.1. Intrinsic camera parameters were precomputed by using the method of Zhang et al. ¹⁷⁾ and the captured image was saved every one degree of rotating angle as a radiometrically linear RAW image with single exposure. Environmental

1 image as the 360-degree light probe was also captured by omni-directional camera (Theta, Ricoh). In $\mathbf{2}$ order to use this image captured with omni-directional camera as the incident light component, it is 3 necessary to consider the radiometrically linearity and color matching with digital camera. To solve 4 this problem, we reproduced high dynamic range (HDR) image by capturing 11 images with different $\mathbf{5}$ exposure time. HDR composition was carried out with the method proposed by Debevec et al.¹⁸). After that, we used Munsell color chart for linear correction and performed the color matching between 6 7 digital camera and omni-directional camera. Through these processing, we derived incident light 8 component from omni-directional camera as shown in Fig.2. The feature of object was measured in 9 sequence by using a rotational stage (SGSP-60YAW, Sigma-Koki). In this experiment, we prepared 10 real object which was simple plate or cylinder as the measurement target. Since the shape of these 11 objects is fully axisymmentrical, we also prepared the polygon model based on the measured data with 12real size of each object. The polygon model was registered to the first frame of image sequences by 13hand with GUI application. Here, it is noted that the subsequent frames are automatically registered 14between real object and its polygon data, because the rotational position of each frames are known and 15the center of rotation is consistent.

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17 **3.2 Assumption and input data**

18 Our proposed method refers to the pioneering idea proposed by Dong et al.¹⁾ We assume 19 that surface reflectance is isotropic and it is expressed by microfacet reflectance model. As the 20 dichromatic BRDF model is used in our system, the surface reflectance at a point x is defined by Eq.1.

$$f(\boldsymbol{\omega}_{i},\boldsymbol{\omega}_{o};x) = \frac{\rho_{d}(x)}{\pi} + \rho_{s}(x)f_{s}(\boldsymbol{\omega}_{i},\boldsymbol{\omega}_{o};x)$$
(1),

where $\boldsymbol{\omega}_i$ and $\boldsymbol{\omega}_o$ are the incident and reflected directions, ρ_d and ρ_s are the diffuse and specular reflectance, and $f_s(\boldsymbol{\omega}_i, \boldsymbol{\omega}_o; x)$ is the specular reflectance function, which is proposed by Asikhmin et al. as follows ¹⁹.

$$f_s(\boldsymbol{\omega}_{\mathbf{i}}, \boldsymbol{\omega}_{o}) = \frac{D(\boldsymbol{\omega}_{\mathbf{h}})G(\boldsymbol{\omega}_{\mathbf{i}}, \boldsymbol{\omega}_{o})F(\boldsymbol{\omega}_{\mathbf{i}}, \boldsymbol{\omega}_{o})}{4(\mathbf{i}\cdot\mathbf{n})(\mathbf{o}\cdot\mathbf{n})}$$
(2),

24where $D(\boldsymbol{\omega}_{\rm h})$ indicates the normal distribution function (NDF) of the halfway direction $\boldsymbol{\omega}_{\rm h}$ with 25microfacet, $F(\mathbf{\omega}_i, \mathbf{\omega}_0)$ indicates Fresnel reflectance function (we assumed a fixed index of refraction of 261.3 for all materials as same as Dong et al.), and $G(\mathbf{\omega}_i, \mathbf{\omega}_0)$ indicates the shadowing and masking 27function. Because specular highlights are quite effective factor for appearance reproduction, it is very 28important to select the spatial distribution of specular highlights, namely NDF. We observed two major 29trends of spatial distribution of highlights in real world. Ceramic plates have weak tail in highlights as 30 shown in Fig.3 (a). On the other hands, wooden flooring plate has relatively strong tail in highlights 31as shown in Fig.3 (b). Therefore, we employed two NDFs owing to represent these characteristics. 32First, we adopt Beckmann distribution model to express the weak tail in highlight as shown in Fig.3

(a). In other case, GGX model is adopted to express the strong tail in highlight as shown in Fig.3 (b) 2 $^{20)}$. These models determine the spatial distribution by the angles between halfway direction ω_h and 3 surface normal. For the simplification of acquiring the surface reflectance properties, we seek the 4 roughness parameters of each model as the objective function instead of the 1-dimensional tabulated 5 function that monotonically decreased used in previous study.

6 By solving the minimization problem, it is possible to obtain an appropriate value (surface 7 reflectance properties) with residual simultaneously. It can be regarded as the smaller residual is 8 suitable for representing the surface reflectance properties. For the reduction of computational cost, 9 we employed Smith's shadowing and masking term, which has relatively less computation cost. In our 10 method, three variables for each surface point are determined separately. To derive these parameters, 11 we solve minimization problem denoted by Eq.3,

$$\arg\min_{(\rho_a,\rho_a,D)_s} \sum_{t} \sum_{x} ||I(\boldsymbol{\omega}'_{o}, x, t) - L(\boldsymbol{\omega}'_{o}, x, t)||^2$$
(3)

where $I(\boldsymbol{\omega}_{o}', x, t)$ is the observation value of rotating target object, $L(\boldsymbol{\omega}_{o}', x, t)$ is the outgoing radiance at a surface point x at time t. $\boldsymbol{\omega}_{o}'$ indicates the outgoing (viewpoint) vector, and prime symbol is the direction in global coordinate. $L(\boldsymbol{\omega}_{o}', x, t)$ is determined by registered object geometry $\mathbf{n}(x,t)$ and incident light component $E(\boldsymbol{\omega}_{i}')$ captured with omni-directional camera as denoted in Eq.4.

$$L(\boldsymbol{\omega}_{o}', \boldsymbol{x}, t) = \int_{\Omega} f_{r}(\boldsymbol{\omega}_{i}(\boldsymbol{x}, t), \boldsymbol{\omega}_{o}(\boldsymbol{x}, t); \boldsymbol{x}) E(\boldsymbol{\omega}_{i}')(\boldsymbol{n}(\boldsymbol{x}, t) \cdot \boldsymbol{\omega}_{i}') d\boldsymbol{\omega}_{i}'$$
(4),

16 where $\mathbf{\omega}_i$ and \mathbf{n} indicates the lighting vector and surface normal at the surface point *x*. $\mathbf{\omega}_i$ is the sampling 17 area from environmental map captured by omni-directional camera. Based on the prior work, we set 18 the range of sampling area less than 0.5π from the mirror reflection with respect to the viewpoint 19 vector. This idea is very simple but it is effective to reduce the computational cost. Figure 4 shows the 20 illustration of acquiring incident light component from the environmental map which is captured with 21 omni-directional camera.

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23 **3.3 NDF Recovery**

Our proposed method solves Eq.3 with each surface point separately. Thus, we rewrite Eq.3
to denote the equation of a single surface point *x*,

$$\arg\min_{(\rho_a,\rho_a,D)_s} \sum_{t} ||T(\boldsymbol{\omega}'_o,t) - L(\boldsymbol{\omega}'_o,t)||^2$$
(5)

26 where $T(\boldsymbol{\omega}_0', t)$ is equal to $I(\boldsymbol{\omega}_0', x, t)$, which represents the temporal variation at the surface point x as

shown in Fig.5. It is difficult to solve Eq.5 because there are still three variables. Therefore, we take

temporal gradient of observation and outgoing radiance on the surface point x. Temporal gradient of

29 the outgoing radiance can be expressed as Eq.6.

Fig.3

Fig.4

$$\nabla_{t} L(\boldsymbol{\omega}_{o}, t) = \nabla \int_{\Omega} f_{r}(\boldsymbol{\omega}_{ix}(t), \boldsymbol{\omega}_{ox}(t)) E(\boldsymbol{\omega}_{i}')(\boldsymbol{n}_{x}(t) \cdot \boldsymbol{\omega}_{i}') d\boldsymbol{\omega}_{i}'$$

$$= \nabla_{t} \int_{\Omega} \frac{\rho_{dx}}{\pi} E(\boldsymbol{\omega}_{i}')(\boldsymbol{n}_{x}(t) \cdot \boldsymbol{\omega}_{i}') d\boldsymbol{\omega}_{i}'$$

$$+ \nabla_{t} \int_{\Omega} \rho_{xx} f_{xx}(\boldsymbol{\omega}_{ix}(t), \boldsymbol{\omega}_{ox}(t)) E(\boldsymbol{\omega}_{i}')(\boldsymbol{n}_{x}(t) \cdot \boldsymbol{\omega}_{i}') d\boldsymbol{\omega}_{i}'$$

$$= \frac{\rho_{dx}}{\pi} \nabla_{t} \int_{\Omega} E(\boldsymbol{\omega}_{i}')(\boldsymbol{n}_{x}(t) \cdot \boldsymbol{\omega}_{i}') d\boldsymbol{\omega}_{i}'$$

$$+ \rho_{xx} \nabla_{t} \int_{\Omega} f_{xx}(\boldsymbol{\omega}_{ix}(t), \boldsymbol{\omega}_{ox}(t)) E(R_{xx(t)}(\boldsymbol{\omega}_{i})) \boldsymbol{\omega}_{k} d\boldsymbol{\omega}_{i}'$$
(6).

1 There are two terms in Eq.6, diffuse reflection component and specular reflection component. Because 2 temporal variation of diffuse component is comparatively small, temporal gradient of the diffuse 3 component can be assumed approximately zero. Hence, specular component, which is principally 4 effected by lighting, is the dominant factor of outgoing radiance denoted as Eq.7.

$$L(\boldsymbol{\omega}_{o}', \boldsymbol{x}, t) \approx \rho_{s}(\boldsymbol{x}) \int_{\Omega} f_{sx}(\boldsymbol{\omega}_{i}, \boldsymbol{\omega}_{o}(t)) \boldsymbol{\omega}_{iz} E(R_{nx(t)}(\boldsymbol{\omega}_{i})) d\boldsymbol{\omega}_{i}'$$
(7).

5 Taking that approximation, we can simply solve the minimization problem denoted in Eq.8 with 6 temporal gradient of the observation and outgoing radiance denoted in Eq.7.

$$\arg\min_{D_{a'}}\sum_{t}||\nabla_{t}T(\boldsymbol{\omega}_{o}',t)-\int_{\Omega}f_{sx}(\boldsymbol{\omega}_{i},\boldsymbol{\omega}_{o}(t))\boldsymbol{\omega}_{iz}E(R_{nx(t)}(\boldsymbol{\omega}_{i}))d\boldsymbol{\omega}'||^{2}$$
(8),

7 where D'_x is the unnormalized NDF biased by specular albedo ρ_{sx} . Finally, the NDF D_x is recovered 8 from D'_x via unit integration. In our implementation, we employ two NDFs, Beckmann distribution 9 and GGX distribution. It is expected that residual of minimization problem may be reduced if the 10 characteristics of NDF is suitable for representing the surface reflectance of target object. We adopt 11 the result which has smaller residual value for verifying the accuracy of our proposed method.

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13 **3.4 Albedo Recovery**

14 Given the recovered NDF D_x from previous formulation, we can now determine 15 the diffuse component and specular component of outgoing radiance denoted as Eq.9.

$$T_{dx}(\boldsymbol{\omega}_{o},t) = \int_{\Omega} f_{r}(\boldsymbol{\omega}_{ix}(t),\boldsymbol{\omega}_{ox}(t))E(\boldsymbol{\omega}_{i}')(\boldsymbol{n}_{x}(t)\cdot\boldsymbol{\omega}_{i}')d\boldsymbol{\omega}_{i}'$$

$$= \frac{\rho_{dx}}{\pi} \int_{\Omega} E(\boldsymbol{\omega}_{i}')(\boldsymbol{n}_{x}(t)\cdot\boldsymbol{\omega}_{i}')d\boldsymbol{\omega}_{i}'$$

$$+ \rho_{xx} \int_{\Omega} f_{xx}(\boldsymbol{\omega}_{ix},\boldsymbol{\omega}_{ox})E(\boldsymbol{\omega}_{i}')(\boldsymbol{n}_{x}(t)\cdot\boldsymbol{\omega}_{i}')d\boldsymbol{\omega}_{i}'$$

$$= \rho_{dx} T_{dx}(\boldsymbol{\omega}_{o}',t) + \rho_{xx} T_{xx}(\boldsymbol{\omega}_{o}',t)$$
(9).

16 $T_{dx}(\boldsymbol{\omega}_{0}, t)$ and $T_{sx}(\boldsymbol{\omega}_{0}, t)$ are diffuse trace and specular trace of the outgoing radiance. It can be regarded 17 that the observation trace $T(\boldsymbol{\omega}_{0}, t)$ is the weighed sum of normalized diffuse trace and specular trace. 18 As a result, recovery of the albedo components are formulated as non-negative least squares 19 minimization problem as shown in Eq.10.

$$\underset{(\rho_{d},\rho_{o})_{s}}{\operatorname{argmin}}\sum_{t}||T_{x}(\boldsymbol{\omega}_{o}',t)-(\rho_{dx}T_{dx}(\boldsymbol{\omega}_{o}',t)+\rho_{sx}T_{sx}(\boldsymbol{\omega}_{o}',t))||^{2}$$
(10).

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2 4. Verification

3 4.1 Evaluation for estimated BRDF

In order to validate our proposed method, we prepare synthetic image generated by Mitsuba 4 $\mathbf{5}$ renderer ²¹). We rendered three kinds of objects by changing surface reflectance as shown in upper row 6 of Fig.6. These images rendered with the office window environment map (Fig.2) and used Beckmann 7 distribution, and lower row images were rendered by using the acquired surface reflectance from our 8 proposed method. Comparing upper to lower images, it can be confirmed that the calculated surface 9 reflectance properties can make a reasonable image with our proposed method. Though the calculated 10 parameters of the most right one in upper and lower row are different, visual difference is slightly 11 small.

12 We also evaluated the validity of our proposed method by comparing with the NDF 13 acquired by gonio-reflectometer (GCMS-4, Murakami Color Research Laboratory). To measure the 14surface reflectance of the objects, we take three type of flat board objects, blue ceramic, wooden 15flooring plate and red gum (Also shown in Fig.3). Figure 7 shows the results of measurement by using 16gonio-reflectometer and our proposed method, where blue line in the Fig.6 is measured by gonio-17reflectometer, and red line and green line in Fig.7 are Beckmann distribution model and GGX 18 distribution model, respectively. As we expected, surface reflectance of each target object is recovered 19by appropriate NDF to represent each material. Figure 8 shows the reproduced image by using 20acquired NDF value. It is apparent that NDFs of each of the subjects are estimated accurately as same 21as the gonio-reflectometer and rendered images of appearance is observed as a quite similar to original 22board objects.

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4.2 Evaluation for SVBRDF measurement

25Next, we challenged to reproduce the SVBRDF profile which is set up in the surface of 26aluminum can with multiple texture. Figure 9 shows the picture of real object (left) and measured 27result of diffuse albedo map (right) by using our system. From the result, our system can obtain the 28meaningful albedo map which can recognize as a combination of different textures. It can distinguish 29each texture and positional relationship, unfortunately, it is hard to make reasonable result for 30 measurement of SVBRDF. This result indicates that our measurement system has some 31incompleteness and improvement for SVBRDF measurement. For its causes and solutions, we will 32describe in the latter discussion part in this paper.

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4.3 Subjective appearance evaluation

In this section, we verified the BRDF recovery by using real 3D objects, which has distinctive surface appearance. Four types of 3D object with different surface profile were selected,



such as green gum ball, urethane coated bowl, white ceramic vase and blue plastic cone. Figure 10
 shows the results of this verification under the same condition and measurement instrument as Fig.7.

3 The left side image in Fig.8 shows real objects, and right side image shows rendered objects by using

4 Mitsuba renderer with Reinhard's tone mapping operator ²²). The calculation for matching the

5 distribution model was performed by using the Beckman and GGX models, and we selected the best

6 matching model between estimated and modeled reflectance.

Here, it is noted that we have no choice but to evaluate the appearance by appreciation since it is difficult to acquire the ground truth of these 3D object. Therefore, we applied subjective evaluation to verify each rendering result. The color corrected display (CX241-CNX, EIZO) was used, and fifteen participants with normal color vision assessed by using five grade evaluation, (5:quite similar, 4:rather similar, 3:similar, 2:rather not similar, 1: not similar). Figure 11 shows the result of histogram for five grade in this subjective evaluation.

13 As a comprehensive evaluation, our proposed method is possible to measure and reproduce 14the surface appearance of material even if the target is 3D object. The highest score was granted for 15the result of white ceramic vase. From the participant's comment, a clear surface reflectance of 16 specular represents the appearance of ceramic material gradely. On the other hand, the reproduction 17of blue plastic cone had the lowest score, since the contrast between specular and diffuse color was 18 ill-balanced in comparison to real 3D object. Commonly, it is difficult to use the best tone mapping in 19 changing from 32bit to 8bit signals because the measured BRDF is expressed by high dynamic range 20data. In this experiment, we made the visual appearance close to real object by controlling the exposure 21of omni-directional images. However, a slight difference of tone balance provides an unpleasant 22impression in rendering reproduction such as the case of blue plastic cone and green gum ball. The 23other results of urethane coated bowl was satisfied by almost participants. They appreciated the 24matching of specular distribution on the surface of 3D object, although ill-balanced contrast between 25specular and diffuse color is slightly concerned. Therefore, for visual verification, the consideration 26for model matching is necessary to attach importance as either specular or diffuse reflectance. In our 27case, the colored object should be attached importance to the diffuse reflection and monotone object 28should be attached importance to the specular reflection.

29

30 5. Discussion

5.1 Measurement setting and SVBRDF recovery

In this paper, it is blindingly clear that our proposed system can measure accurate BRDF when the target object consists of unique material. For this result, we infer that position-independent and integrated best solution is obtained by the calculation in an optimized process. On the other hand, more rigid measurement about position matching is required for the SVBRDF measurement, since the BRDF profiles are changed according to the measurement position. Therefore, all of the geometries

in our work are formulated by combination of simple geometries, such as a sphere, cuboid, cylinder
 and so on. Moreover, many number of trial for position alignment were executed for the SVBRDF
 measurement, although this alignment was performed manually. However, this result indicates that
 our measurement system has some incompleteness.

 $\mathbf{5}$ From the result as shown in Fig.9, we can observe several unnatural features which are 6 related to the position misalignment. At first, we notice a blurred edge at the part of trade name into 7 the rotated direction. Moreover, we found that the detail part of "d" or "u" character have a serious 8 blur compared with the first character "R". This unnatural result of blur leads us to some assumptions 9 that the rotated table is hard to keep accurate one degree of angle, the center of object is out of 10 alignment from the rotated center, and the shape of real object is partly distorted compared with 11 polygon geometry. It is obvious that the blur image is generated as the mixture of diffuse albedo and 12specular reflection since the BRDF at misaligned point is estimated by using the different normal 13 vector. In order to decrease the mixture of diffuse and specular reflection, the use of polarizer may be 14appropriate⁴⁾. Unfortunately, this workaround idea invites an additional measurement in order to 15change the direction of polarization, and more than all, the user always should prepare the polarizer 16 every time.

17Other unnatural feature in Fig.9 is found that the irregularity of up and down is existed in 18 the horizontal linear edge. This misalignment infers that the slight movement of up and down side may 19 be generated every movement by the rotating table. As just described, it is necessary to improve the 20precision of position alignment and stability of movement with careful consideration of requisite 21resolution in SVBRDF measurement. As an essential solution for this problem, it may be useful to 22measure the shape of object together by using digital camera. However, it might bring the trade-off 23between an accuracy of SVBRDF and computational cost, since a complicate and corrective 24calculation is necessary to measure the shape of object which has specular reflection on its surface. 25Recent novel instrument of 3D measurement may be possible to obtain the accurate shape even if 26target has the specular reflection $^{16)}$. Unfortunately, it is necessary to match the position and shape 27between measured shape and measurement point of BRDF finally.

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5.2 Calculation cost reduction

Another priority of our proposed method is reduction of computational cost. Since our proposed method uses environmental map as the incident light component, the computational cost is simply reduced. Not to require the estimation of incident light component, we can estimate each surface reflectance independently. This independence of calculation will provide several merits that the parallel computing technique is easy to apply for our proposed method, and it can accelerate estimation process dramatically. In order to confirm our superiority, we tried to compare the performance of calculation between conventional and our method. Unfortunately, it is only provided 1 the fragmentary specification of computation and the concept of algorithm in Dong's paper. Therefore, $\mathbf{2}$ we make two kinds of hypothetical comparison; one is clock speed competition and the other is 3 simulation for each algorithm. Table 1 shows that the comparison of clock speed of ours and 4 conventional methods. Clock speed of conventional method is calculated by their paper. Our $\mathbf{5}$ implementation environment uses Intel Corei7 processor with 16 GB memory. On the other hand, 6 conventional method uses dual Intel Xeon E5-2690 processor with 64GB memory. Our computational 7 specification is obviously less powerful than their condition, however we can achieve faster 8 calculation for BRDF recovery.

9 For the simulation of each algorithm, we estimate the computational cost of both methods 10 with the object as shown in Fig.7(a). With same polygon size of object, same measured pixel, and 11 same optimization algorithm, both BRDF and environmental illumination map was estimated as the 12conventional method. On the other hand, only the BRDF was estimated by using our proposed method 13 under the condition that the incident lighting component was known. As the result of simulation, our 14method achieves this process about 10 seconds, and conventional Dong's method is necessary to 15calculate about 101 seconds. As a conclusion, we can acquire precious BRDF value about ten times 16 faster than conventional method.

17

18 **6.** Conclusion and future work

19 In this paper, we present a simple BRDF measurement method with well-convergence and 20rapid estimation by adding the lighting component derived from omni-directional camera. Our 21proposed method accomplished compatibility between accuracy and speed-up of calculation by using 22environmental image as light probe. The accuracy of our measurement was verified by comparing the 23surface reflectance between our system and gonio-reflectometer, and it is confirmed that the surface 24reflectance of flat board objects is almost completely reproduced by appropriate NDF with fitting to 25the Beckmann and GGX distribution model. The accuracy of visual reproduction is verified by 26comparing the appearance between real and reproduced 3D object, and it is confirmed that most of the participants selected their answers as "rather similar", or "similar" in the subjective evaluation. 2728Moreover, our proposed method achieves the speed ten times faster than that of conventional method 29owing to the simplification of incident light components.

30 Geometric data selection can also affect the result of both NDF and albedo recovery of 31 SVBRDF. If the misalignment and variation between measuring point and shape position are existed, 32 the image of reproduced surface texture becomes blurred. Therefore, it is necessary to take careful 33 position alignment and stability of rotational movement for the acquisition of shape and reflection data. 34 The misalignment and variation between measuring point and shape position make the image of 35 reproduced surface texture blurred. Here, it is noted that they have a trade-off relationship both 36 accuracy and calculation cost. 1 Comments by participants in the evaluation of our proposed method prompted us to 2 evaluate the target object with more complex shape. Therefore, high density measurement of geometry 3 data and precise position matching between 3D model and environmental image are necessary as a 4 future work. Moreover, an appropriate control of HDR measured data should be developed for 5 practical use of our method. The HDR measurement for light probe and object feature was embedded 6 in proposed method. Therefore, a precise tone mapping for the visual reproduction of material 7 appearance is also necessary as a future work.

8

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1 Caption of table

- $\mathbf{2}$
- 3 Table 1
- 4 Result of comparison for computational speed between conventional and proposed method
- $\mathbf{5}$

1	Caption of figures		
2			
3	Figure 1		
4	Illustration of our experimental setup for BRDF measurement and incident light capturing by 36		
5	degree omni-directional camera.		
6			
7	Figure 2		
8 9	Example of light probe which is HDR image captured by 360-degree omni-directional camera.		
10	Figure 3		
11	Various spatial distribution of highlights in real world material, (a) Ceramic plate, (b) Wooden flooring		
12	plate, (c) Red gum plate.		
13			
14	Figure 4		
15	Illustration of acquiring incident light component from the HDR image captured with omni-directional		
16	camera.		
17			
18	Figure 5		
19	Temporal variation of pixel value during the measurement with a rotational table.		
20			
21	Figure 6		
22	Results of synthetic image generated by Mitsuba renderer based on the calculated three variables,		
23	which are roughness, diffuse albedo, and specular albedo. Each numerical value indicates the		
24	comparison between measured ground truth and calculated result by using our method.		
25			
26	Figure 7		
27	Comparison results of reflectance distribution between our proposed method and gonio-reflectometer.		
28			
29	Figure 8		
30	Reproduced images by using acquired NDF value. (a) Blue ceramic plate, (b) Wooden flooring plate,		
31	(c) Red gum plate.		
32			
33	Figure 9		
34	Result of SVBRDF measurement by using aluminum can with multiple texture, (a) Real object, (b)		
35	diffuse albedo map.		
36			

- 1 Figure 10
- 2 Images for subjective evaluation. The Left images show the captured image of 3D real objects, and
- 3 the right images show the rendered results by our proposed method.
- 4
- 5 Figure 11
- 6 Result of subjective evaluation for our rendered images.
- 7

- 1 Table
- $\mathbf{2}$
- 3
- 4 Table 1
- $\mathbf{5}$

Method	Maximum clock speed	Minimum clock speed
Dual Intel Xeon (Conventional)	396.8GFlops	6.2GFlops
Intel Core i7 (Ours)	53.28GFlops	13.2GFlops

6



1 Figure 3





12 Figure 4







16 Figure 8



- 18
- Figure 9 19



(b)

