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Fast Compressive 3D Single-pixel Imaging

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Abstract: In this work, we demonstrate a modified photometric stereo system with perfect pixel registration, capable of reconstructing continuous real-time 3D video at ~ 8 Hz for 64×64 image resolution by employing evolutionary compressed sensing.

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1. Introduction

Photometric stereo is a well-known 3D imaging technique, which defines the surface normals according to measured intensity differences between images taken under different incident lighting directions [1, 2]. This approach demands the scene to remain completely static whilst the lighting condition changes in order to prevent surface reconstruction errors, which limits its scope in real-time applications. Single-pixel imaging, which determines the apparent lighting conditions of a scene by employing single-pixel detectors as imaging devices and a spatial light modulator to provide either structured detection or structured illumination, is an alternative computational 3D imaging technique to provide perfect pixel registration [3]. However, the size of projection patterns required in the single-pixel system may lead to lengthy acquisition process.

Orthogonal and pseudo-random bases have been employed which significantly reduces the acquisition time, however, the finite modulation rates of micro-electromechanical-systems (MEMS) technology, typically ~ 20 kHz, places restrictions on the achievable frame rates even for relatively low resolution images. A number of compressed sensing schemes have been developed in recent years to enable real-time compressed sensing for large image resolutions. In this work, we employ one of these compressive strategies, known as evolutionary compressed sensing [4, 5], to demonstrate a real-time 3D single-pixel imaging system at ~ 8 Hz for 64×64 image resolution, 4 times faster than a conventional raster-scanning sampling strategy.

2. Imaging System Design

In the system, as illustrated in Figure 1, we choose a 3W white LED, a digital micro-mirror device (DMD) with 1024×768 pixels, and a camera lens with a 24mm focal length to provide structured illumination at a rate of 22 kHz. A data acquisition board (DAB) is used to convert analog intensity measured by photodetectors (PD) into digital signals at a rate of 250 kHz which subsequently were processed to reconstruct 2D images via computational imaging, following a 3D image based on photometric stereo.

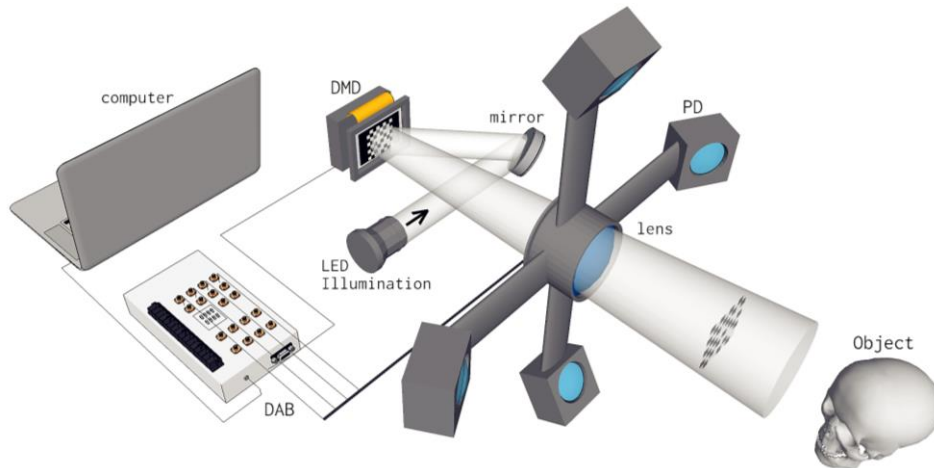


Fig. 1. 3D single-pixel imaging system. The system contains a DMD, a camera lens and four spatially separated photodetectors fixed surrounding it. The DMD chip is illuminated by a white LED and the structured light patterns are projected through the lens onto the object. A plastic polarizer sheet is used in front of the photodetectors horizontally and the camera lens vertically to remove specular reflection on the object.

To maximize sampling efficiency, a Hadamard matrix in the order of 2^{2k} , which provides orthogonal structured patterns to avoid overlap information, is preloaded for the DMD. In our system, a 4096×4096 Hadamard matrix is used to generate 4096 of 64×64 2D Hadamard derived patterns to modulate the light source. And by this analogy, we generate 16384 of 128×128 2D Hadamard derived patterns.

We order the Hadamard patterns based on their corresponding mean signal intensities from the photodetectors, and utilize the top-ranking patterns to form 2D images. Based on evolutionary compressive sensing algorithm, we replace a small percentage (we choose 10% patterns in this experiment) of low-ranking patterns among the top-ranking patterns with the ones that are randomly selected from the rest Hadamard patterns each time for fuzzy self-adjustment due to the variations between adjacent frames. By combining those 2D images, we then performed photometric stereo to gain 3D images.

3. Results

In Figure 2, we perform 3D reconstruction of a skull at 128×128 pixel resolution by using five different numbers of pattern pairs: 16384 pattern pairs, 12288 pattern pairs, 8192 pattern pairs, 4096 pattern pairs, and 2048 pattern pairs, equivalent to 100% (zero-compression), 75%, 50%, 25% and 12.5% compression ratio.

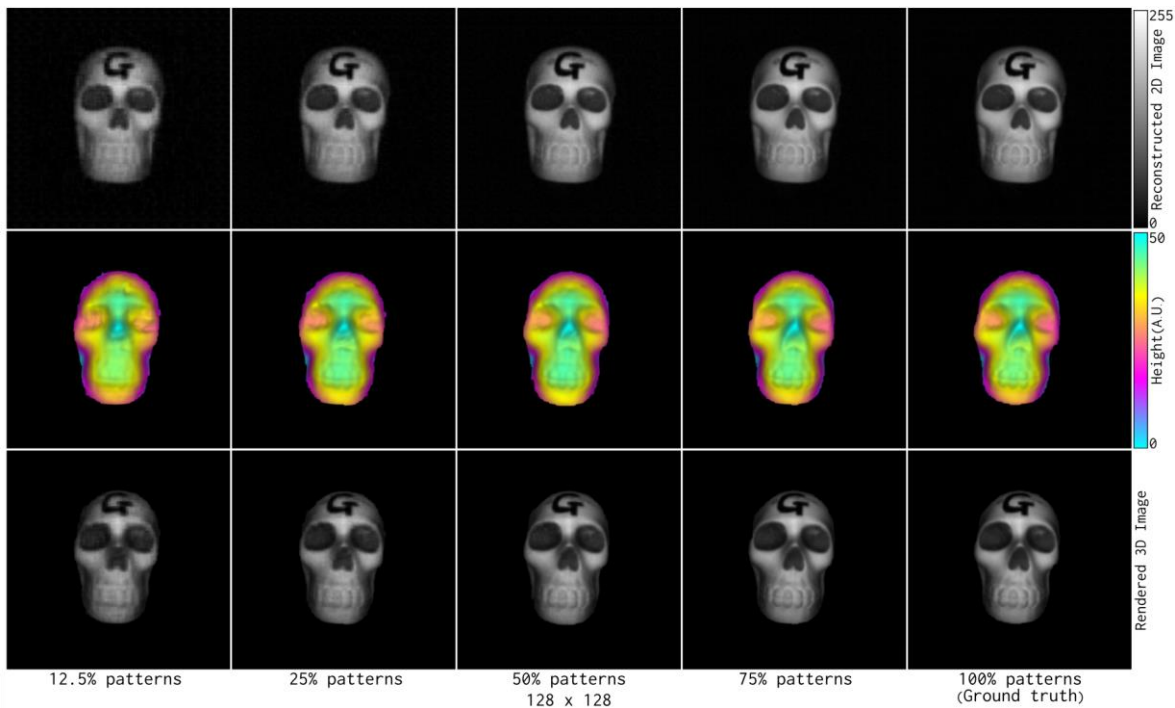


Fig. 2. 3D reconstruction comparison with evolutionary compressed sensing. The object was reconstructed at 128×128 pixel resolution with five different compression ratios: 12.5%, 25%, 50%, 75% and 100%.

We choose the zero-compression data as the ground truth to compare the relative root-mean square (RMS) errors of the height value in those reconstructed 3D images, and report the performance in Table 1. The result shows that the RMS error of the object's height value increases when using less pattern pairs (higher compression).

Table 1. Relative RMS error comparison at 128×128 pixel resolution.

128 x 128 pixel resolution					
Patterns Used	12.50%	25%	50%	75%	100% (Ground truth)
Relative RMS error	3.586	3.443	2.229	1.571	0

In Figure 3, we present a sample of real-time compressed 64×64 pixel resolution 3D video frames in 1 second for a skull physically rotated about the z-axis and the 3D reconstructed model rotated simultaneously about the y-axis on the front controller at an angle range of ± 30 degrees. Each frame is produced based on 1024 patterns (25% compression ratio), which equals to the same amount of patterns for a zero-compression 32×32 pixel resolution 3D

image reconstruction. The frame rate of this 3D video is $\sim 8\text{Hz}$, four times faster than the zero-compression 64×64 pixel resolution one.

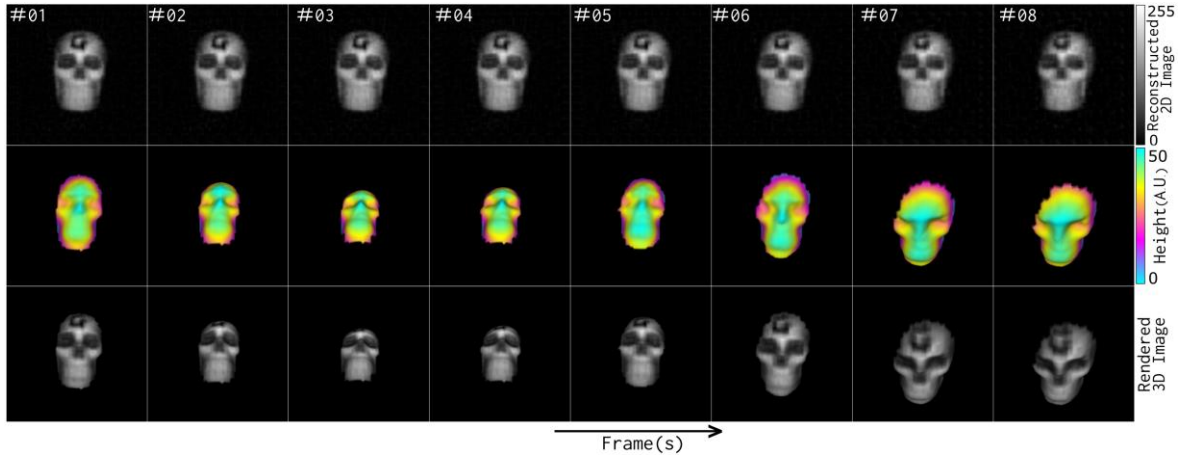


Fig. 3. Sample of a compressed video-rate 64×64 pixel resolution 3D frames in 1 second. Each 3D frame is produced based on 1024 patterns (25% compression ratio) acquired at a frame rate of $\sim 8\text{Hz}$ for an object physically rotated about the z-axis, whilst the 3D reconstructed model is rotated about the y-axis at an angle range of $[-30, 30]$ controlled by the system.

In Figure 4, we demonstrate a 10-second real-time 3D video frames of a Santa object reconstructed at 128×128 pixel resolution at a frame rate of 0.9Hz , using 4096 Hadamard patterns (25% compression ratio). The object is physically rotated about the z-axis.

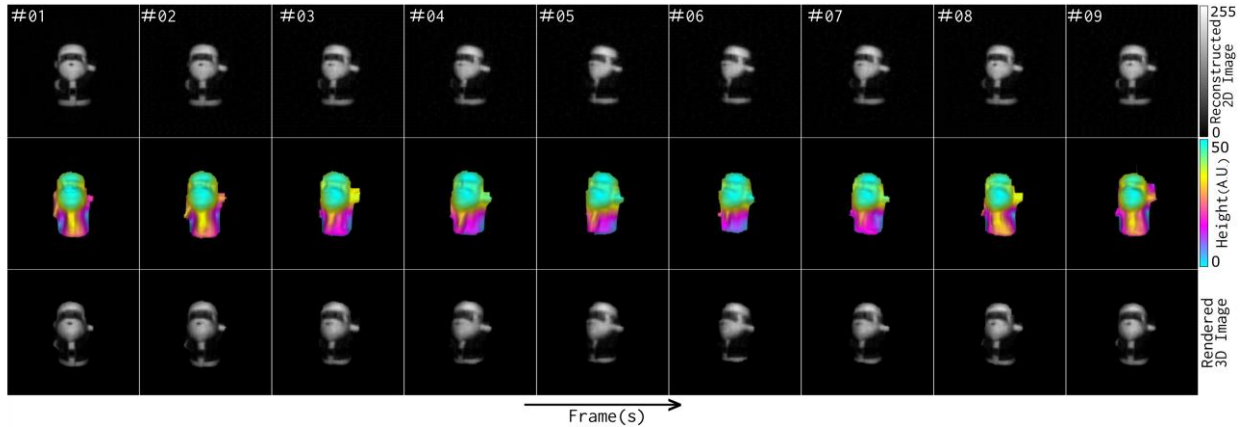


Fig. 4. Sample of a 10-second 128×128 pixel resolution 3D video frames of a Santa object. Each 3D frame is produced based on 4096 patterns (25% compression ratio) acquired at approximately a frame rate of $\sim 1\text{Hz}$ while the Santa object is physically rotated about the z-axis.

References

1. R. J. Woodham, "Photometric method for determining surface orientation from multiple images," *Opt. Eng.* **19**, 139–144 (1980).
2. Y. Zhang, M. G. Graham, R. Hay, R. W. Bowman, M. J. Padgett, M. P. Edgar, "A fast 3D reconstruction system with a low-cost camera accessory," *Sci. Rep.* **5**, e10909 (2015).
3. B. Sun, M. P. Edgar, R. Bowman, L. E. Vittert, S. Welsh, A. Bowman, M. J. Padgett, "3D computational imaging with single-pixel detectors," *Science* **340**, 844–847 (2013).
4. Y. Zhang, M. P. Edgar, B. Sun, N. Radwell, G. M. Gibson, M. J. Padgett, "3D single-pixel video," *J. Opt.* **18**(3), 035203(2016).
5. M. P. Edgar, G. M. Gibson, R. W. Bowman, B. Sun, N. Radwell, K. J. Mitchell, S. S. Welsh, M. J. Padgett, "Simultaneous real-time visible and infrared video with single-pixel detectors," *Sci. Rep.* **5**, e10669 (2015).