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KU Leuven/ESAT/PSI

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Real-time Photometric Stereo

Context

Context



Real-time Photometric Stereo

Structured light

Photogrammetry

Photometric stereo



Photometric stereo using light domes



Real-time Photometric Stereo

Photometric stereo using light domes

Photometric strengths :

- Determine high-frequency 3D details
- Determine surface reflectance

Minidome : single camera version with 260 LED lights.



Photometric stereo (PS) – the challenge

The goal of basic PS is to extract the 3D shape (normals) knowing the illumination and local surface albedo and the type of reflectance

e.g. assuming a *Lambertian* surface:

I = a(L.n)

Real-time Photometric Stereo

Photometric stereo (PS) - the challenge

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n can be solved if the surface is illuminated from 3 directions in turn

Photometric stereo (PS) – the challenge

The goal of basic PS is to extract the 3D shape (normals) knowing the illumination and local surface albedo and the type of reflectance

e.g. assuming a *Lambertian* surface:

I = a(L.n)

More unknowns can be solved (*non-linearly*) if there are more measurements (more light sources)

Real-time Photometric Stereo

Photometric stereo (PS) - the challenge

The previous conditions can be relaxed A solution for the diffuse (Lambertian) component still possible by removing specular outliers among multiple illuminations; this also allows for a more sophisticated reflectance model: diff + spec comp

e.g. Verbiest and Van Gool, *Photometric Stereo with Coherent Outlier Handling and Confidence Estimation,* IEEE Conf. on Computer Vision and Pattern Recognition '08

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Real-time Photometric Stereo

Robust Photometric stereo

Outcome of the software pipeline, for all surface points:

Surface normals (unlike PTM)

Albedos (color) : 'diffuse material color'

Specular lobes

Robust Photometric stereo

From these

> 3D reconstruction by normal integration, allowing users to make metric measurements



real-time shader filtering for study, research and inspection of objects and artefacts.

Real-time Photometric Stereo

Example:



Cuneiform tablet with stamp

Example:

Cuneiform tablet with stamp



Real-time Photometric Stereo

Example:



Literature



Minidome applications



- Coin collection from museum of Pisa
- specularity / reflections





Minidome applications

Paintings

- Van Dyck painting : surface reveals history and possibly identity.
- Brushstrokes of older painting are visible underneath hat and cape.







Minidome applications

Entymology

- Insects: traditionally 'difficult to scan'
- Need for non-destructive measurement techniques.



Minidome applications

Fossils (NHMBerlin)





Minidome applications



Anthropology

Ivory piece





Antropology

Kent graffiti plate : copper plate found in Kent (UK), origin was never known, until minidome shaders revealed 15th century Ship graffiti (3 master).



Real-time Photometric Stereo

Operational status:

The new minidome setup is more rugged: 3D printing / silicon mall reproducible and therefore now available @ 11 kEuro



Operational status:



E.g. 30 Mpix image sequence (x260), runs > 10 minutes on 8 cores.

Processing is off-line, as batch job.

Success of recording (influenced by camera aperture and shutter speed settings) can only be evaluated afterwards.

Real-time Photometric Stereo

Operational status:

Processing speed 2m – 15m, depending on resolution and CPU power.

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GPU implementation offers the possibility for on-line evaluation and real-time processing.

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GPU implementation offers the possibility for on-line evaluation and real-time processing. *Diffuse only now*

Real-time Photometric Stereo

GPU implementation :



GPU implementation:

- Processing of PS is parallel per pixel, based on (at most) 260 observations (LEDs)
- Each pixel is executed by one available thread on the GPU, but each thread deals with multiple pixels
- Careful optimization between thread processing power / memory type: shared, register, cache, global (red is faster but limited) / memory coalescing

GPU implementation :



Real-time Photometric Stereo

GPU implementation :

GPU implementation:

- Further optimizations: 260 images call for quite a lot of memory and may not fit on a typical GPU.
- 1) images are sliced into sections and processed one by one.
- 2) processing takes advantage of the possibility of transfer concurrency while a previous stored slice is processed



GPU implementation :

Results:

Speedup close to factor 1000. Depends on graphics card specs.

Tabel shows performance increase on GTX 780ti. On GX980 speedup increases with another factor > 1.5. On TitanX another > 1.5.. Unit: milliseconds

	CPU	GPU
1.2MP	158*10^3	265
1.7MP	252*10^3	267
2.7MP	401*10^3	431
3MP	445*10^3	480
6MP	891*10^3	966
28MP	4185*10^3	5009

Real-time Photometric Stereo

Additional examples (30Mpix)



Future and ongoing work



Real-time Photometric Stereo

Conclusions

- Robust Photometric stereo algorithm on GPU
- Allows for live processing and visual feedback during recording
- Future work on more complex material representation such as BRDFs (also see S. Georgoulis, M. Proesmans, and L. Van Gool, A Gaussian Process Latent Variable Model for BRDF Inference, ICCV, 2015)

