

Development and Application of a Description-based Interface for 3D Object Reconstruction

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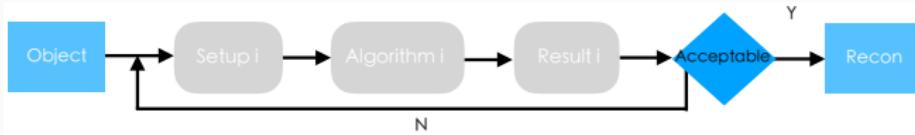
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Motivation

Motivation: traditional 3D Reconstruction

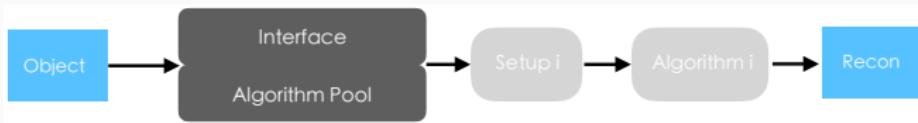


Challenges:

- Hardware: controlled environment, calibration
- Algorithms: vision background
- Results: keep trying until an acceptable result

Motivation: interface to 3D Reconstruction

What if we can create an interface above the 3D algorithms, which can select an appropriate algorithm based on users' description, and achieve a successful reconstruction result.



Contribution

Contribution

Development of an interface for 3D reconstruction problem, which hides algorithmic details and allows users to describe conditions surrounding the problem. This description can be interpreted so that an appropriate algorithm is chosen to obtain a successful reconstruction result.

Contribution (cont'd)

This contribution is significant because:

- No single algorithm can work for a diverse categories of objects. The interface, to some extent, can cover a wider range of object categories by incorporating multiple algorithms.
- An description is provides that hides the algorithmic details, thus understanding of the algorithm, or conditions to apply a specific algorithm is not a prerequisite.

Related Work

Related Work: softwares

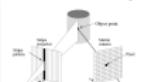
Some notable examples of open source or commercial softwares:

| Case | Software | Example | Problem |
|------------------|-----------|---|---------------------------|
| General VL | OpenCV | Mat findHomography(); | Generic APIs |
| Software bundles | SfM + MVS |   | Cater to specific objects |

Challenge

Not that we don't have enough tools, but the barrier to take advantage of these tools is high.

Related Work: algorithms

| Category | Method | Problem |
|-----------------------|---|----------------------------|
| Shape from Stereo |  | Texture, Specular |
| |  | Albedo, Specular |
| Shape from Intensity |  | Albedo, Specular, Geometry |
| |  | Albedo, Geoemtry |
| Shape from Silhouette |  | Geoemtry |
| |  | Geoemtry |

Related Work: algorithms

Challenges

1. No algorithm works for objects with diverse properties;
2. The knowledge of whether an algorithm works is not known a priori.

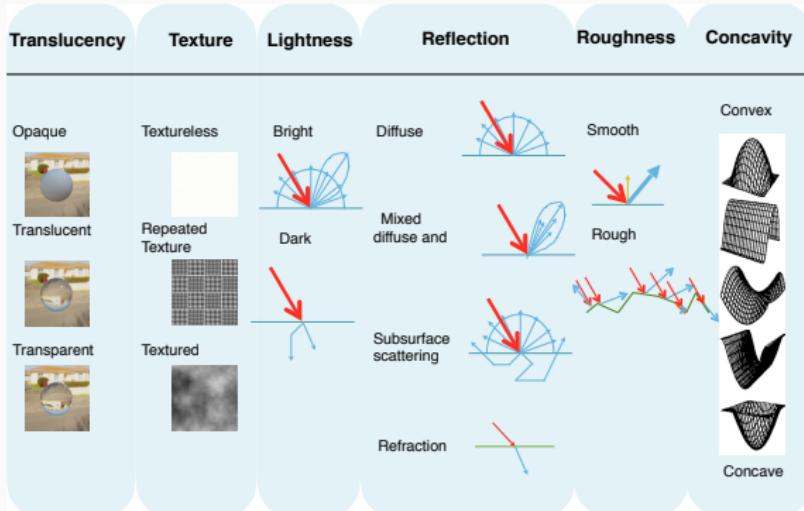
Development of Interface

Roadmap

- Taxonomy: understand the problem space;
- Description: describe the problem space;
- Mapping: discover the connection between a sub-volume in problem space to algorithm.

Taxonomy: properties of problem space

- *algorithm-centered* taxonomy categorizes algorithms based on algorithmic details, as discussed in **Related Work**;
- *object-centered* taxonomy categorizes algorithms based on the problem conditions that the algorithm can reliably work under.



Taxonomy: problem conditions

Assumptions:

- Active methods require high surface albedo (bright), in order to demonstrate the effectiveness of these methods, we focus on bright surfaces only.
- Diffuse is caused solely by surface roughness since sub-surface scattering is ignored.

| Condition | Texture | Lightness | Reflection | Roughness | Label | | | | |
|-----------|------------------|--------------|------------|------------|-------------|-----------|------------|-----------|----------|
| 1 | Textureless (Tl) | Textured (T) | Dark (D) | Bright (B) | Diffuse (D) | Mixed (M) | Smooth (S) | Rough (R) | TI-B-D-R |
| 2 | Yes | | | Yes | Yes | | Yes | | TI-B-M-S |
| 3 | | Yes | | Yes | Yes | | | Yes | T-B-D-R |
| 4 | | Yes | | Yes | | Yes | Yes | | T-B-M-S |

Description: model and representations

| Model | Representation |
|-------------|-------------------------------|
| Texture | <i>Texture randomness</i> |
| Lightness | <i>Albedo</i> |
| Specularity | <i>Specular/diffuse ratio</i> |
| Roughness | <i>SD of facet slopes</i> |

Table 1: Representations of the 3D reconstruction problem.

Description: expression

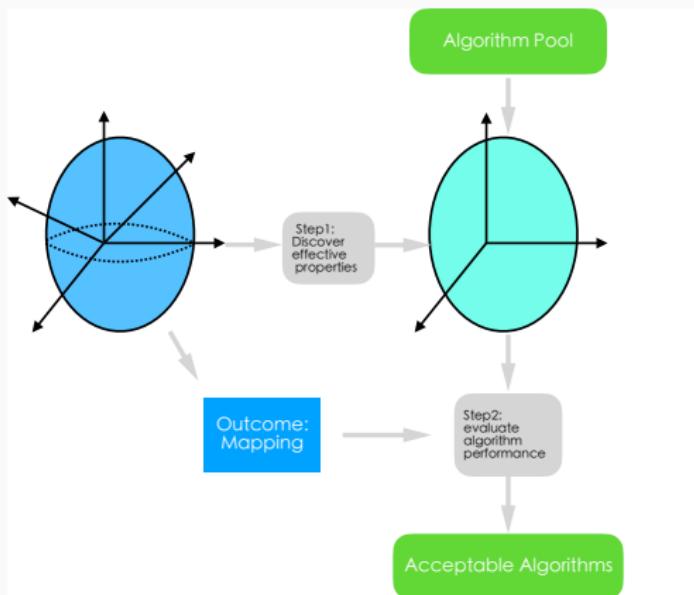
We use three discrete scales to parameterize these properties: *low* (0.2), *medium* (0.5), and *high* (0.8).

| Object | Texture | Albedo | Specular | Rough | Label |
|---------|---------|---------|----------|---------|----------|
| Class 1 | low/med | high | low/med | high | TI-B-D-R |
| Class 2 | low/med | high | high | low/med | TI-B-M-S |
| Class 3 | high | high | low/med | high | T-B-D-R |
| Class 4 | high | high | high | low/med | T-B-M-S |
| Class 5 | high | low/med | low/med | high | T-D-D-R |
| Class 6 | high | low/med | high | low/med | T-D-M-S |

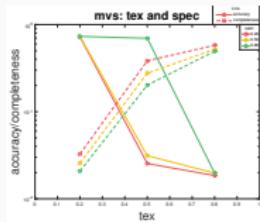
Table 2: Expression of the reconstruction problem for the object classss.

Mapping

Investigate the problem conditions under which the algorithms can reliably work. This structure of this chapter is as follows



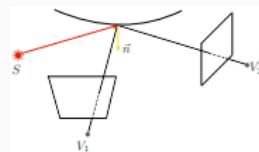
Mapping: notable findings 1



(a). Algo. performance



(c) V_1



(b) Image formation



(d) V_2

Figure 1: (a) shows the algorithm performance w.r.t. texture and specularity. (b) shows the reflection of light off a specular surface. V_1 received the diffuse component while V_2 receives the specular component. (c), (d) shows the images observed from these two views. The specular area (red circle) observed in V_2 is visible in V_1 .

Mapping: notable findings 2

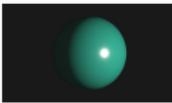
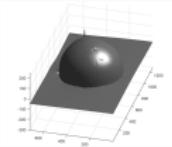
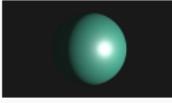
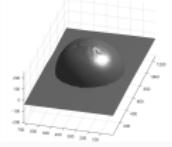
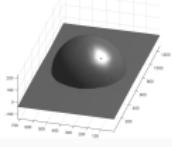
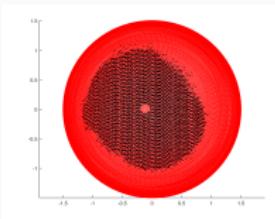
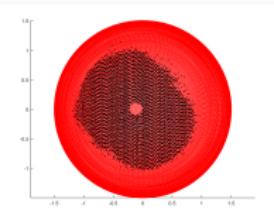
| Image | Normal map | Height map | Angular error |
|---|---|---|---|
|  |  |  |  |
| (a). rough: 0.2 | | | |
|  |  |  |  |
| (b). rough: 0.5 | | | |
|  |  |  |  |
| (c). rough: 0.8 | | | |

Figure 2: The effect of roughness on PS. Albedo is set as 0.8, and specular is set as 0.8. (b) demonstrates that a medium level roughness would lead to worse normal estimation since it blurs the specular lobe.

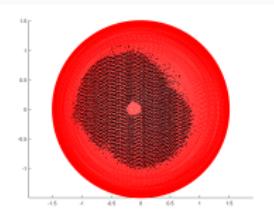
Mapping: notable findings 3



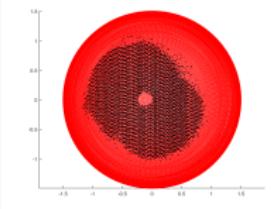
(a) specular: 0.2



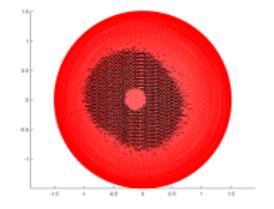
(b) specular: 0.5



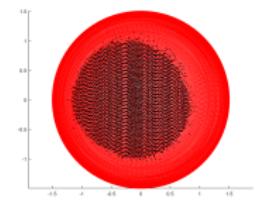
(c) specular: 0.8



(d) roughness: 0.2



(e) roughness: 0.5



(f) roughness: 0.8

Figure 3: (a)-(c): the roughness is set as 0.2, and specular has a negative effect on completeness; (d)-(e): the specular is set as 0.8, roughness has a positive effect on completeness.

Conclusions

- PMVS can work on specular surfaces;
- EPS and GSL fails on highly specular areas, and a blurred specular area causes worse results.

Evaluation of Interface

Interpretation: Key Evaluation Questions

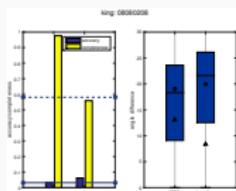
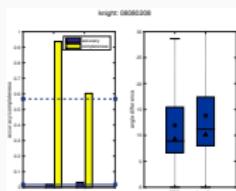
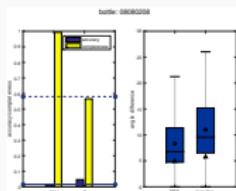
- Evaluation of mapping: is the mapping robust to changes of shape?
- Evaluation of interpreter: can the proof-of-concept interpreter return a successful reconstruction given the correct description?

Interpretation: evaluation of mapping

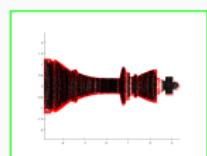
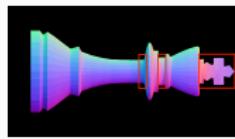
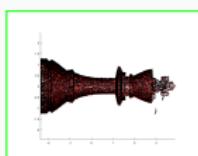
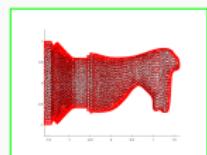
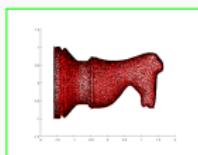
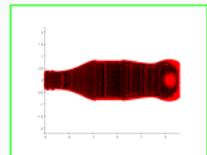
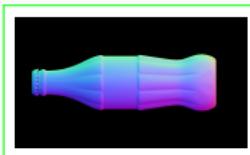
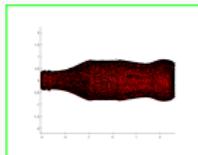
- Shape variation: too vast and complicated to model;
- Instead focus on one geometric property: surface concavity, and see how robust is the mapping with respect to concavity changes.

Interpretation: evaluation of mapping (cont'd)

Quantitative results



Qualitative results



PMVS

EPS

GSL

Figure 4: Problem condition: 08080208, mapped algorithms: PMVS, EPS, GSL

Interpretation: evaluation of mapping (cont'd)

Conclusion:

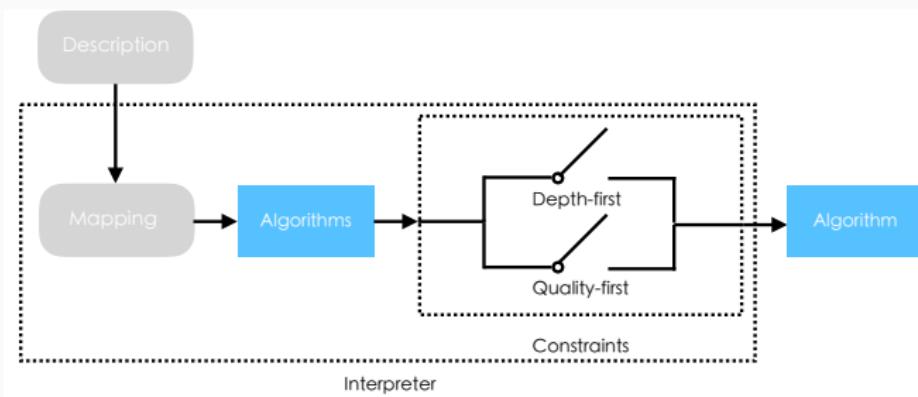
- Mappings to PMVS and GSL are robust to concavity changes whereas those to EPS are not.

Suggestions:

- Develop more advanced description, incorporating concavity into the description
- Use more underlying algorithms that are robust to concavity changes.

Proof-of-concept interpreter

An interpreter selects an appropriate algorithm based on description of problem condition and constraints.



Interpretation: evaluation of interpreter

| Desc # | Bust | Vase1 | Barrel | Vase0 | Selected Algo. |
|--------|---|---|---|---|----------------|
| 1 |  |  |  |  | GSL |
| 2 |  |  |  |  | EPS |
| 3 |  |  |  |  | GSL |
| 4 |  |  |  |  | PMVS |

Interpretation: real-world objects

| class # | 1 | 2 | 3 | 4 |
|-------------|---|---|--|---|
| description | textureless diffuse bright | textureless mixed d/s bright | textured diffuse dark/bright | textured mixed d/s dark/bright |
| object |  |  |  |  |

Figure 5: The representatives of the four classes of objects used for evaluation.

Interpretation: evaluation of interpreter (cont'd)

| Desc # | Statue | Cup | Pot | Vase | Selected Algo. |
|--------|---|---|---|---|----------------|
| 1 |  |  |  |  | GSL |
| 2 |  |  |  |  | EPS |
| 3 |  |  |  |  | GSL |
| 4 |  |  |  |  | PMVS |

Conclusions

Conclusions

- the proposed description is able to give correct reconstruction for non-concave objects
- To deal with more complicated objects, we need more complicated properties, or ways to describe the objects, but the challenge is the easy mathematical representation might not be available.
- Using the simple descriptive language and proof-of-concept interpreter, we demonstrate the possibility of using descriptive properties to hide algorithmic details.

Take-away

message

Computer vision should focus on more than just algorithms, but easier accessibility.