

Temporal Upsampling for Image Sequences Using a Non-Local Means Algorithm

Masterstudium:
Visual Computing

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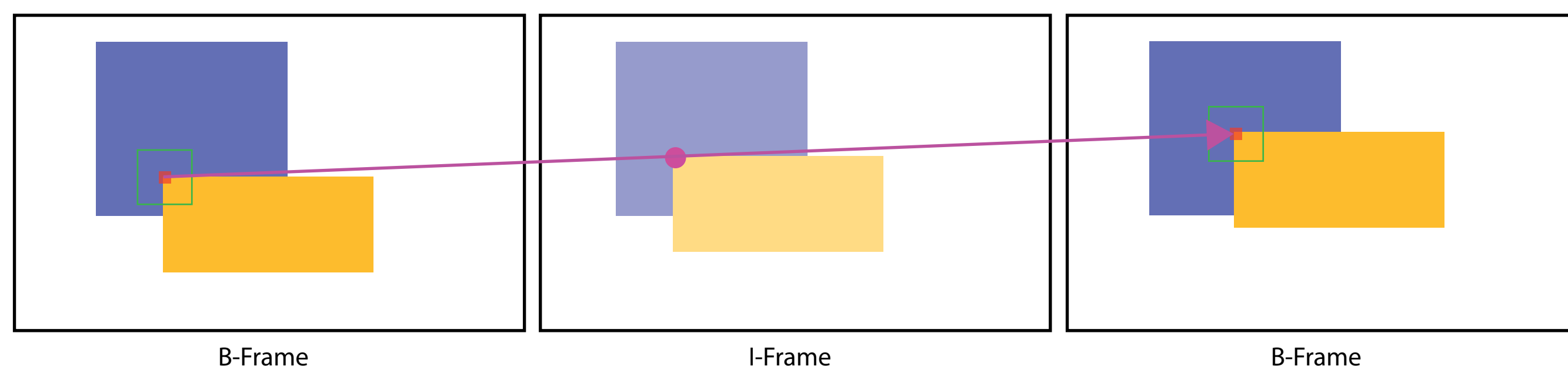
Problem and Motivation

- CGI Industry lately favors global-illumination (GI) and higher frame-rates, both of them increase visual quality but take longer for the image sequences to render.
- Temporal coherence (upsampling) methods for GI algorithms increase performance compared to regular rendering.
- Image-based upsampling techniques generate frames based on ground-truth ones and are in general significantly faster than the previous mentioned ones.
- However, upsampling techniques can not handle reflections and refractions correctly, because they result in motion of image features which is incoherent to the objects motion in the scene.

Contribution

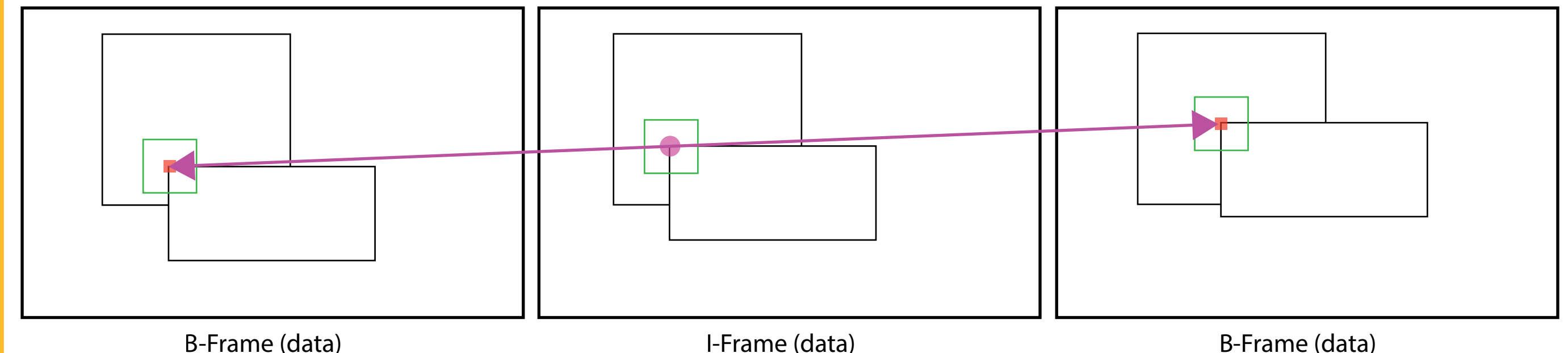
- Two image-based upsampling techniques, that use temporal coherence to create additional frames.
- Independency from rendering algorithm allows for portability.
- Preserves image features such as specular or transparent surfaces better than previous work.
- First to utilize the non-local means denoising algorithm for up-sampling frames, because it does not rely on object movement, in contrast to previous work, and it can be optimized for parallel execution.
- More accurate than some image-based previous work.

Motion-Vector Technique



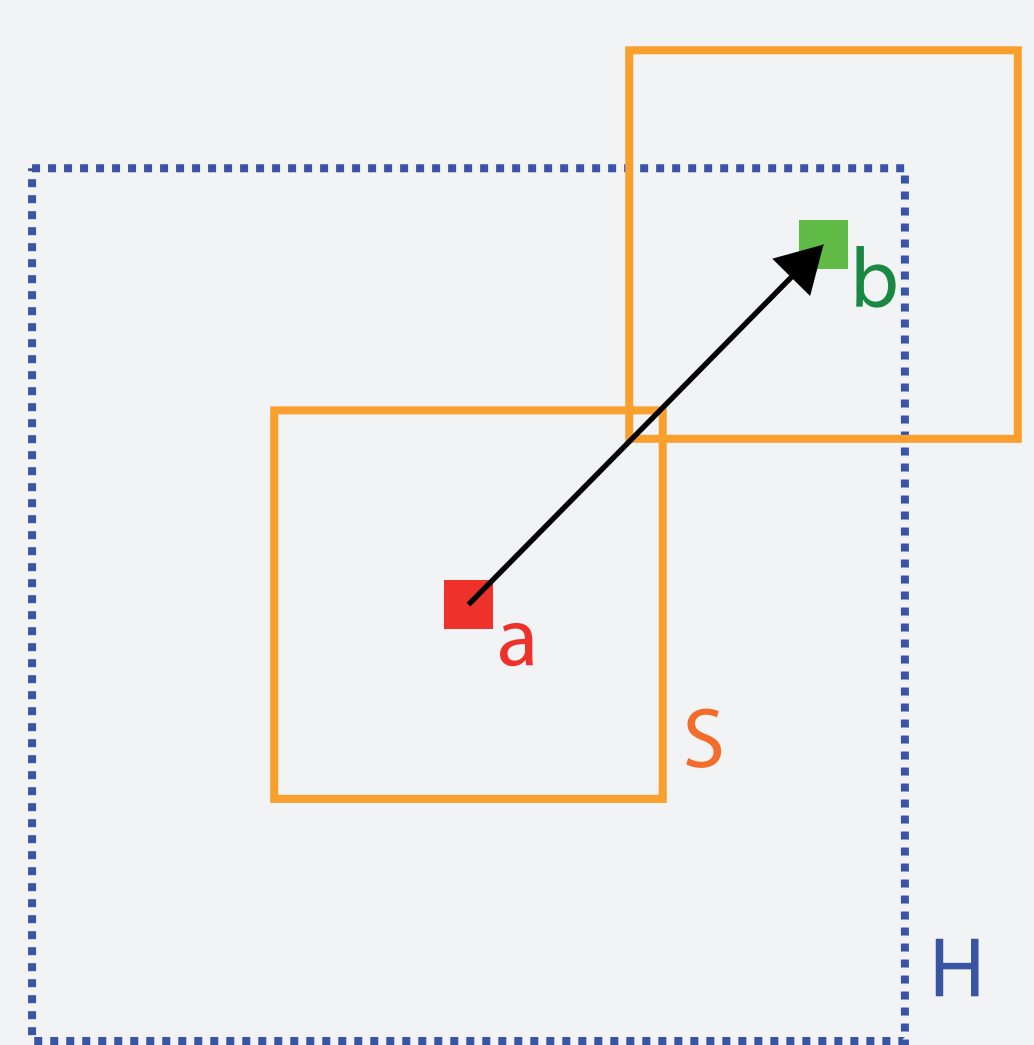
- Take 2 B-frames (fully rendered ones).
- Find similar pixels from one frame to the other based on the color, using the non-local means algorithm (with the difference in color as ξ). Those similarities can be interpreted as motion vectors.
- For each pixel in the I-frame (the interpolated one) find the source pixel in the B-frames based on the motion vectors.
- Mix the source pixel together or use the scene-assisted method of Yang et al. [1] as fallback.

Pixel-Similarity Technique



- Render a stubby frame (only surface data) of the I-frame.
- Search in adjacent B-frames for points with similar surface data, using the non-local means algorithm.
- The similarity measure ξ is made up of:
 - the diffuse surface similarity: point-location.
 - the specular surface similarity: location after reflection.
 - the transparent surface similarity: location after refraction.
- Color the I-frame based on the similarities.

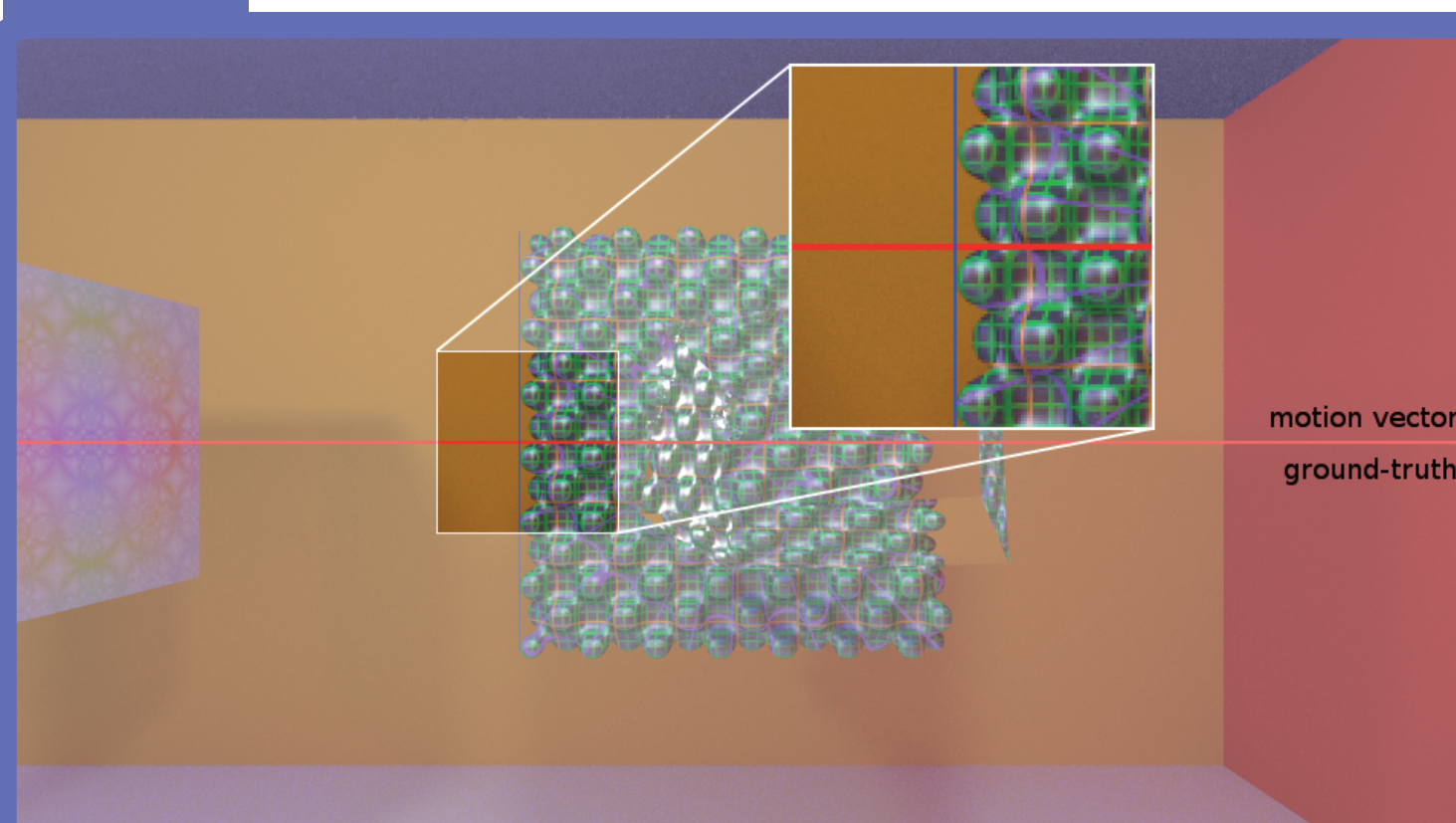
Adapted Non-Local Means Algorithm



- Searches for similar pixels b in a pre-defined neighbourhood H around a .
- Two pixels a and b are similar if the pixels in their surrounding S match.
- A match in the surrounding is calculated by the similarity measure ξ .
- How similar two pixels are, is defined by the sum of the similarity measures results.
- Can be optimized for parallel execution.
- Adaption stores the n most similar pixels.

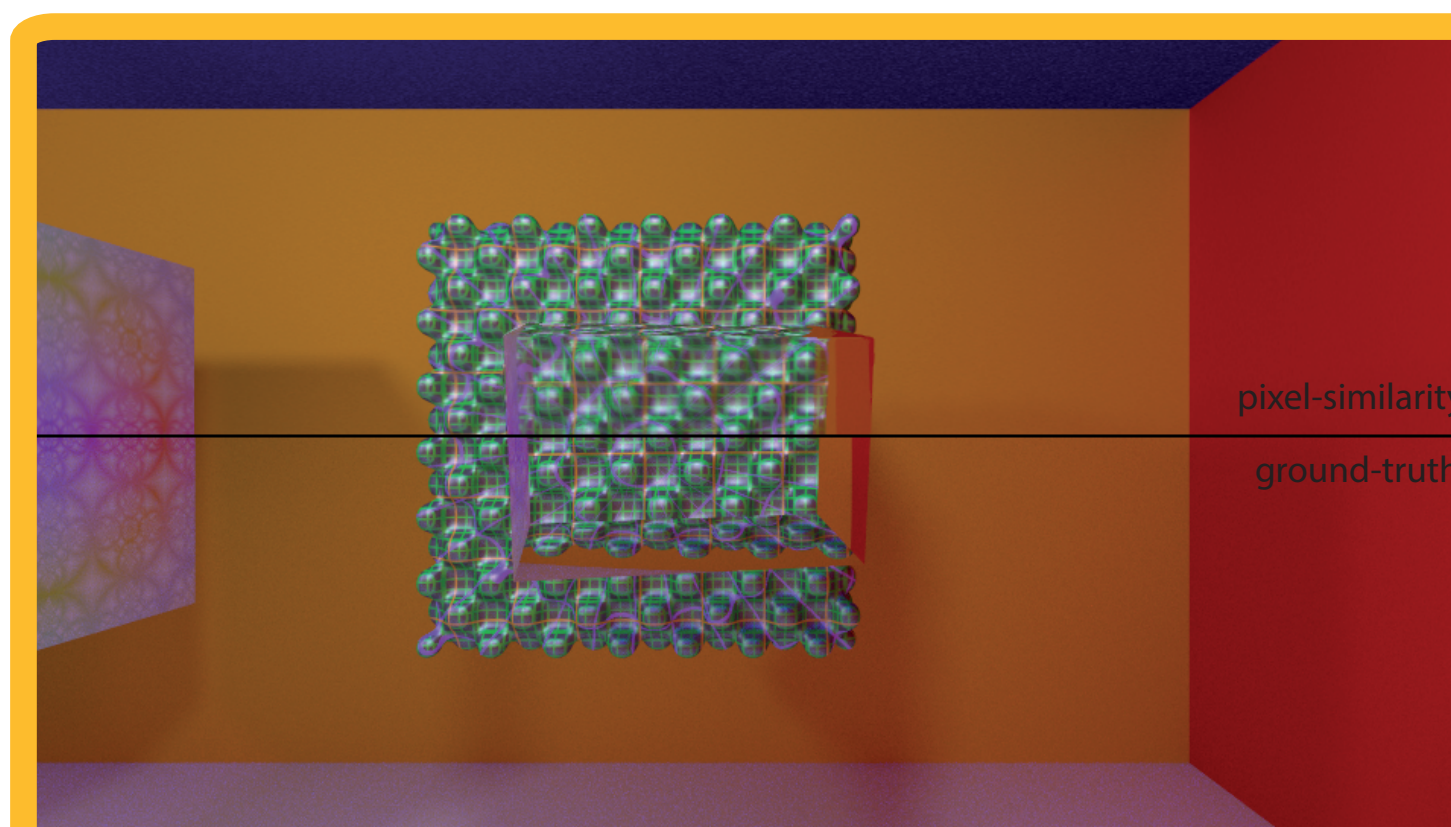
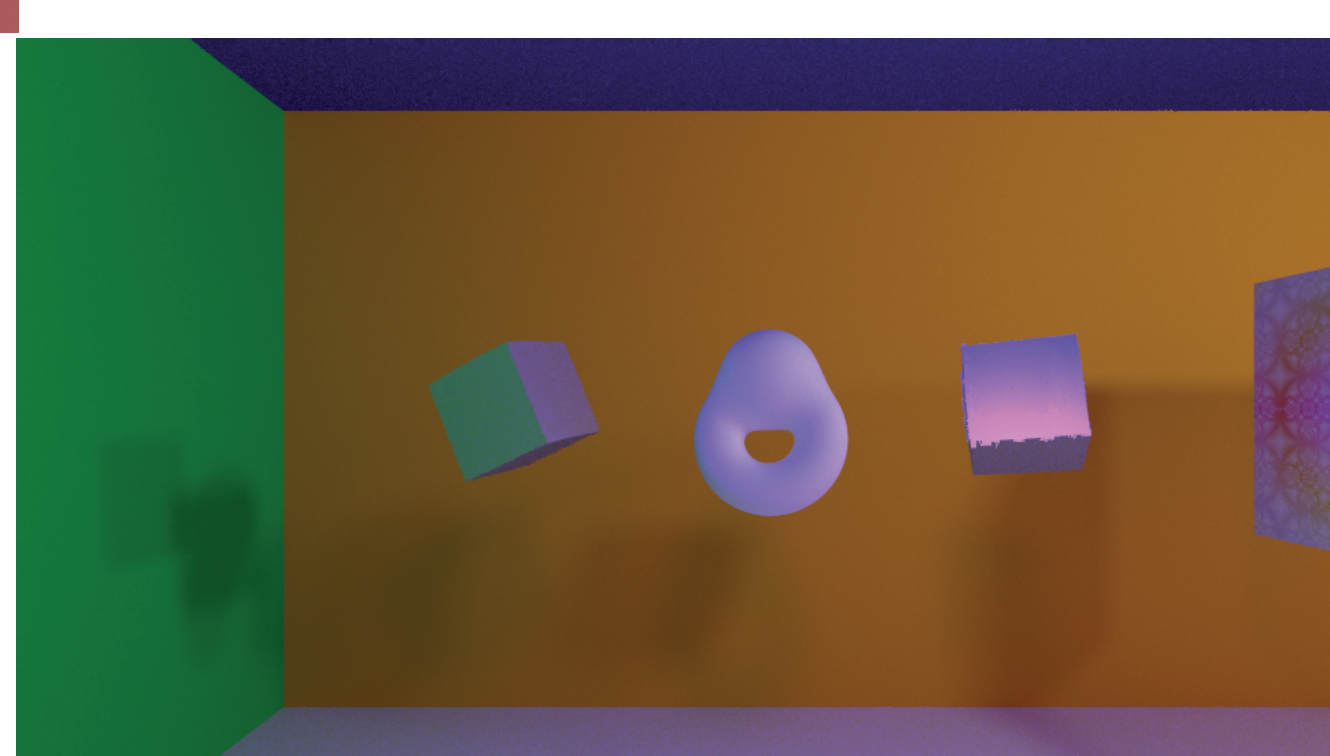
Unoptimized adapted non-local means:

```
for each pixel  $a$  in image do
  for each  $h$  in neighbourhood  $H$  do
     $\Sigma = 0$ ;
    for each  $s$  surrounding  $S$ 
      calculate  $\Sigma += \xi(a+s, a+h+s)$ 
    end
    keep if  $\Sigma$  maximal
  end
end
```



- Handles transparent objects better than other methods.
- Not data-dependent, unlike pixel-similarity technique.

- Color bleeding when new geometry shows, for example due to rotation.
- Edge location is off due to non-linear movement.



- Handles reflections better than other methods.
- Edge-location always similar to ground-truth.

- Depends on availability of surface data. When the appropriate data is not available, it produces artifacts, e.g. at transparent surfaces.

