

Texturing and Inpainting a Complete Tubular 3D Object Reconstructed from Partial Views

Julien Fayer^{a,b}, Bastien Durix^a,
Simone Gasparini^a, Geraldine Morin^a

^aIRIT, University of Toulouse

^bSAS InnerSense

June 7, 2018

Introduction: context

- Texturing a 3D model generated from a limited set of images



Figure: Reconstruction with the open-source state-of-the-art MVS pipeline AliceVision [AliceVision, 2017] with 4 images (right)

Introduction: context

- Texturing a 3D model generated from a limited set of images



Figure: Reconstruction with the open-source state-of-the-art MVS pipeline AliceVision [AliceVision, 2017] with 30 images (right)

Introduction: our proposal

- Goal: reconstruct a complete textured 3D object



- Hypothesis: object represented by a set of canal surfaces (*branches*) [Durix et al., 2016]

Introduction: our pipeline

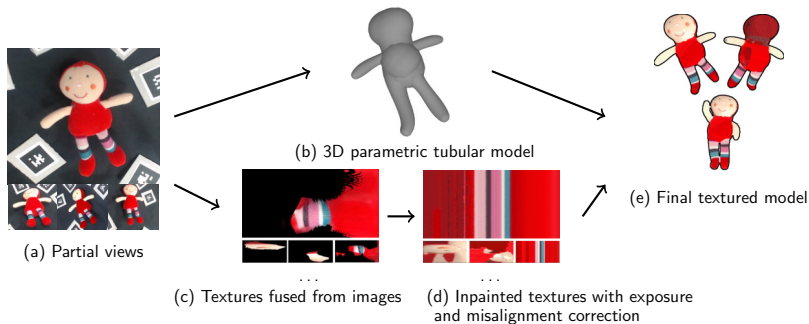


Figure: Pipeline

Outlines

- 1 Related works
 - Texturing 3D models generated by MVS
 - Texturing 3D models generated by other methods
- 2 Surface parametrization and texturing
 - Surface parametrization
 - Texturing a branch
- 3 Improving the textures
 - Exposure improvement
 - Alignment improvement
 - Texture completion
- 4 Experimental Results
 - Results
 - Limitations

1 Related works

- Texturing 3D models generated by MVS
- Texturing 3D models generated by other methods

2 Surface parametrization and texturing

- Surface parametrization
- Texturing a branch

3 Improving the textures

- Exposure improvement
- Alignment improvement
- Texture completion

4 Experimental Results

- Results
- Limitations

3D Texturing

Single-view

- Only locally optimal → visible artifacts and discontinuities seams
- Solutions: Blending the images, labeling [Sinha et al., 2008], energy minimization [Wachter et al., 2014]

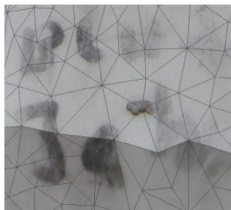


Figure: Solution of [Wachter et al., 2014]

Multi-views

- Weighted blending of a subset of source images [Callieri et al., 2008] → Suffering of loss of quality and of details + generating artifacts
- Solution: patch based synthesis [Bi et al., 2017]

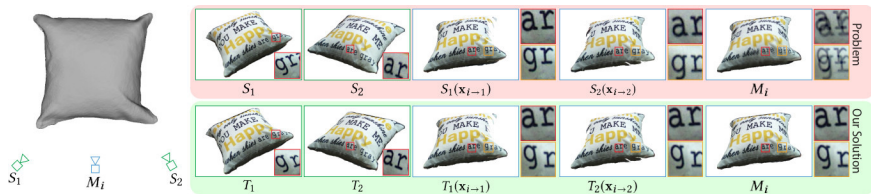


Figure: Patch based synthesis by [Bi et al., 2017]

Other reconstruction methods and texturing

- Shape from silhouette ([Carlos and Schmitt, 2004]): no parameterization \rightarrow difficult to texture
- Reconstruction of tubular shapes:
 - 3D model from a single image with user interaction: texture recovered by back-projection [Chen et al., 2013]
 - Reconstruction by skeletons [Durix et al., 2016]: camera and geometry alignments issues

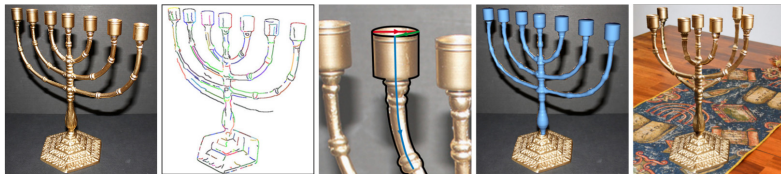


Figure: Results from [Chen et al., 2013]

Other reconstruction methods and texturing

- Advantage: complete reconstruction from partial views
- Contribution: texture reconstructed 3D model including occluded parts

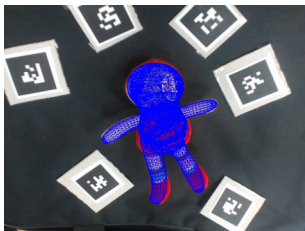


Figure: Estimated 3D skeleton [Durix et al., 2016]

1 Related works

- Texturing 3D models generated by MVS
- Texturing 3D models generated by other methods

2 Surface parametrization and texturing

- Surface parametrization
- Texturing a branch

3 Improving the textures

- Exposure improvement
- Alignment improvement
- Texture completion

4 Experimental Results

- Results
- Limitations

Surface parametrization

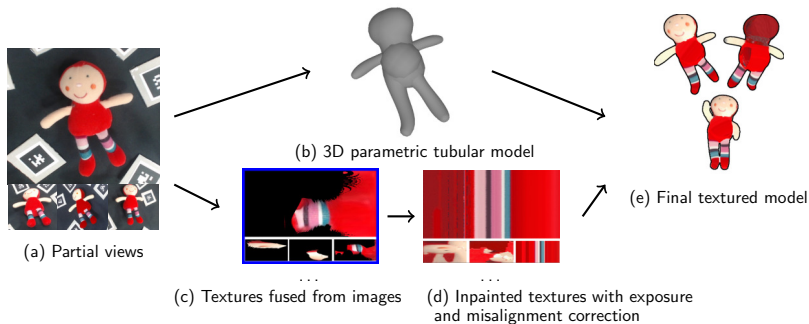


Figure: Pipeline

Texturing a branch from a single image

- Create a reference image texture I_b for each branch b
- Represented by a parametric canal surface $S(t, \theta)$ where $(t, \theta) \in A \times [0, 2\pi]$
- Occlusions handled by an adhoc z-buffer like method

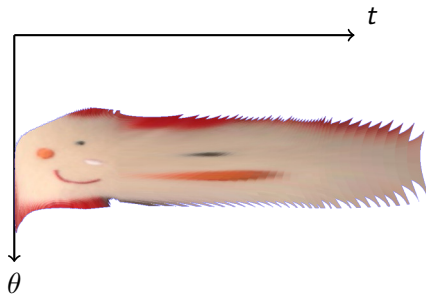


Figure: Surface parametrization

Texturing a branch from a single image

- Create a reference image texture I_b for each branch b
- Represented by a parametric canal surface $S(t, \theta)$ where $(t, \theta) \in A \times [0, 2\pi]$
- Occlusions handled by an adhoc z-buffer like method

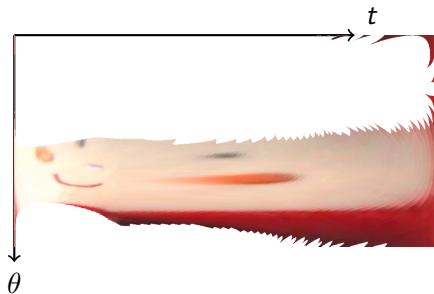


Figure: Surface parametrization

Texturing a branch from a single image

- Create a reference image texture I_b for each branch b
- Represented by a parametric canal surface $S(t, \theta)$ where $(t, \theta) \in A \times [0, 2\pi]$
- Occlusions handled by an adhoc z-buffer like method

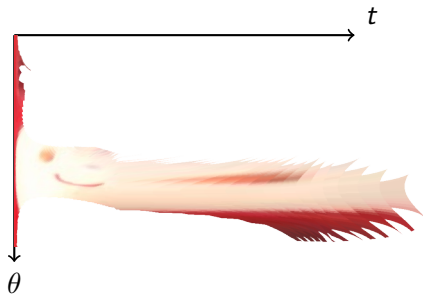
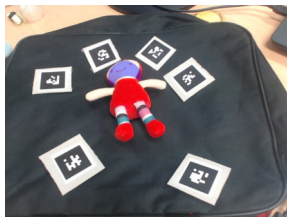


Figure: Surface parametrization

Texturing a branch from a single image

- Create a reference image texture I_b for each branch b
- Represented by a parametric canal surface $S(t, \theta)$ where $(t, \theta) \in A \times [0, 2\pi]$
- Occlusions handled by an adhoc z-buffer like method

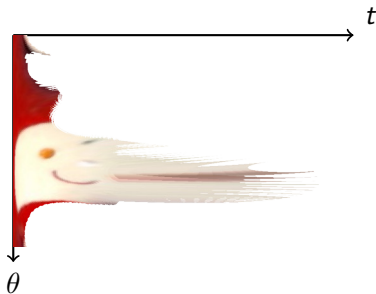


Figure: Surface parametrization

Texturing a branch from multiples images

- Apply the texture of the best image of a set V of images I_i
- Based on a radiometric confidence criterion of our previous work [Fayer et al., 2018]



Figure: Confidence of a reference image

Texturing a branch from multiples images

- Apply the texture of the best image of a set V of images I_i
- Based on a radiometric confidence criterion of our previous work [Fayer et al., 2018]

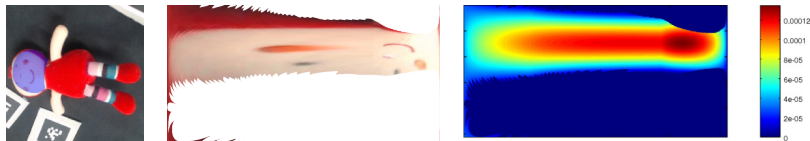


Figure: Confidence of a reference image

Surface parametrization and texturing

Texturing a branch from multiple images

- Each image I_i has a confidence map C_i
- For each pixel p of the final reference image I_b

$$I_b(p) = I_i(p) \quad i = \operatorname{argmax}_i (C_i(p))$$



Figure: Merged reference image

1 Related works

- Texturing 3D models generated by MVS
- Texturing 3D models generated by other methods

2 Surface parametrization and texturing

- Surface parametrization
- Texturing a branch

3 Improving the textures

- Exposure improvement
- Alignment improvement
- Texture completion

4 Experimental Results

- Results
- Limitations

Improving the textures

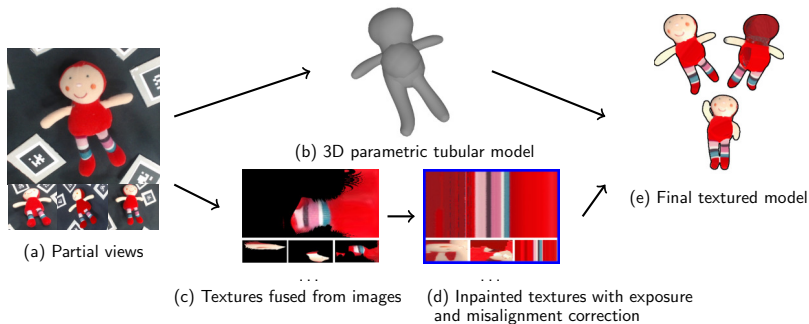


Figure: Pipeline

Improving the textures

- Remaining issues :
 - Exposure issues
 - Alignment issues
 - Completing the texture



Figure: Exposure differences between two reference images

Improving the exposure

- Inspired from [Zhang et al., 2016]
- Each pixel p of a reference image I_b has a radiance value $r(p)$
- Each image I_i has a exposure coefficient α_i
- Find $(\alpha_i, r(p))$ for each pixel p and each image I_i^b generated from I_i ie:

$$\min_{\alpha_i, r(p)} \sum_{i,p} \left(I_i^b(p) - \alpha_i r(p) \right)^2$$

Improving the exposure



Figure: Exposure correction example

Improving the alignment

- Remaining issues :
 - Exposure issues
 - Alignment issues
 - Completing the texture



Figure: Alignment issues of merged reference image

Improving the alignment

- Inspired from [Arikan et al., 2014]
- Minimize the following energy:

$$E = \sum_{p \in I_b} E_d(p, l) + \lambda \sum_{(p_1, p_2) \in \mathcal{N}} E_r(p_1, p_2, l_1, l_2)$$



Figure: Labels map before and after the minimization

Improving alignments

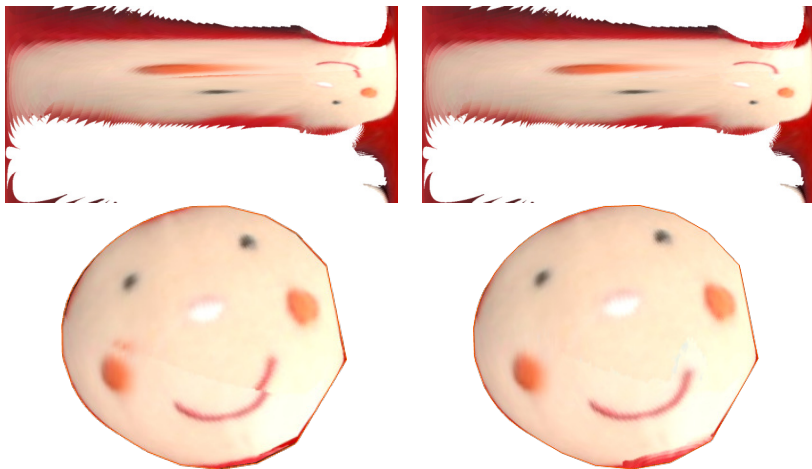


Figure: Alignment correction example

Completing the texture

- Remaining issues :
 - Exposure issues
 - Alignment issues
 - Completing the texture



Figure: Missing regions in reference image

Completing the texture

- Two kinds of textures:
 - Regular textures
 - Circular textures: color of the pixel $p(t, \theta)$ is independent of θ
- Apply two completion methods according to the texture type

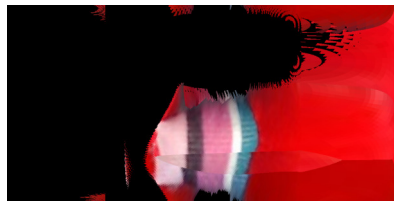


Figure: Examples of two textures: regular (left) and circular (right)

Inpainting: state-of-the-art

- Patch based methods: PatchMatch [Barnes et al., 2009], statistical analysis [He and Sun, 2012]



Figure: Patch based method example

Completing the texture

- Create a structure image
- Generate a new texture image with the structure image and the partial texture image

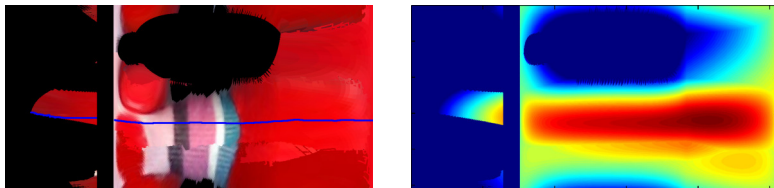


Figure: Circular texture completion

Completing the texture

- Create a structure image
- Generate a new texture image with the structure image and the partial texture image



Figure: Circular texture completion

Completing the texture

- Create a structure image
- Generate a new texture image with the structure image and the partial texture image

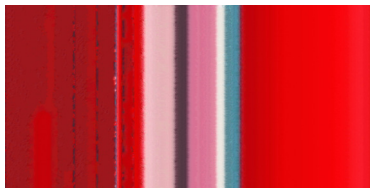


Figure: Circular texture completion

Completing the texture

- Create a structure image
- Generate a new texture image with the structure image and the partial texture image



Figure: Circular texture completion

1 Related works

- Texturing 3D models generated by MVS
- Texturing 3D models generated by other methods

2 Surface parametrization and texturing

- Surface parametrization
- Texturing a branch

3 Improving the textures

- Exposure improvement
- Alignment improvement
- Texture completion

4 Experimental Results

- Results
- Limitations

Context

- Dataset: 5 plushes, 3 to 5 photos for each plush
 - https://github.com/Ibujah/dataset_plushes
- Good candidates: isotropic material, circular surfaces around the direction of the main axis
- Results website:
<https://sketchfab.com/jfayer/collections>



Figure: Input plushes

Results



Figure: 3D models results

Results



Figure: Mesh completion

Results

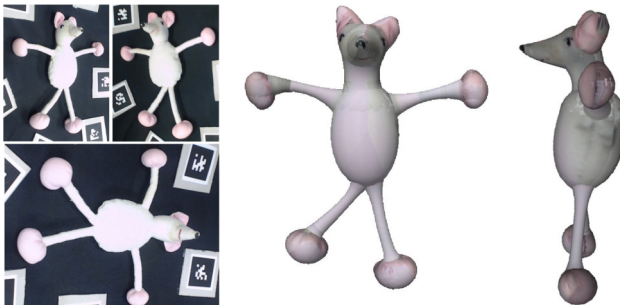


Figure: 3D models results

Results

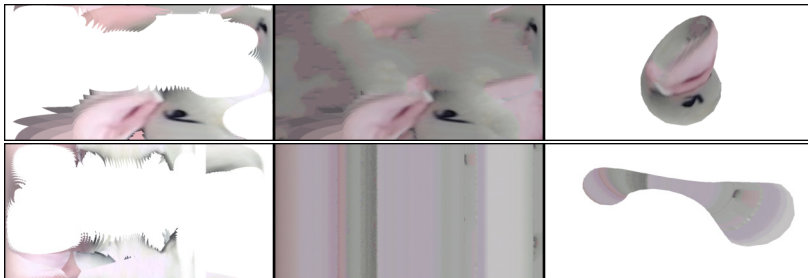


Figure: Mesh completion

Results

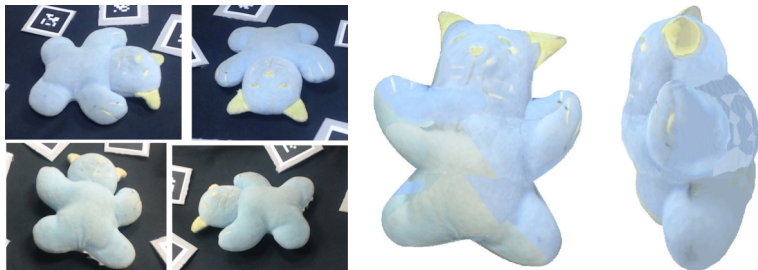


Figure: 3D models results

Results

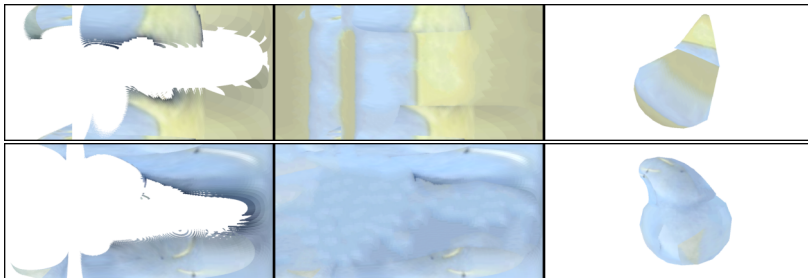


Figure: Mesh completion

Limitations



Figure: Limitations example

Conclusions

- Requiring only few images
- Texturing a complete 3D object
- Future directions:
 - Classify the texture into circular or regular
 - Generalize the work on exposure to smoothing of color
 - Handle shiny surfaces



AliceVision (2017).

Photogrammetric Computer Vision Framework.



Arikan, M., Preiner, R., Scheiblaue, C., Jeschke, S., and Wimmer, M. (2014).

Large-scale point-cloud visualization through localized textured surface reconstruction.
IEEE Transactions on Visualization and Computer Graphics, 20(9):1280–1292.



Barnes, C., Shechtman, E., Finkelstein, A., and Goldman, D. B. (2009).

PatchMatch: A randomized correspondence algorithm for structural image editing.
ACM Transactions on Graphics (Proc. SIGGRAPH), 28(3).



Bi, S., Kalantari, N. K., and Ramamoorthi, R. (2017).

Patch-based optimization for image-based texture mapping.
ACM Transactions on Graphics, 36(4):1–11.



Callieri, M., Cignoni, P., Corsini, M., and Scopigno, R. (2008).

Masked photo blending: Mapping dense photographic data set on high-resolution sampled 3D models.
Computers & Graphics, 32(4):464–473.



Carlos, H. and Schmitt, F. (2004).

Silhouette and stereo fusion for 3D object modeling.
Computer Vision and Image Understanding, 96(3):367–392.



Chan, T. F. and Shen, J. (2001).

Nontexture inpainting by curvature-driven diffusions.
Journal of Visual Communication and Image Representation, 12(4):436 – 449.



Chen, T., Zhu, Z., Shamir, A., Hu, S.-M., and Cohen-Or, D. (2013).

3sweep: Extracting editable objects from a single photo.
ACM Trans. Graph., 32(6):195:1–195:10.



Durix, B., Morin, G., Chambon, S., Roudet, C., and Garnier, L. (2016).

Skeleton-based Multiview Reconstruction.

In *Proceedings of the IEEE International Conference on Image Processing*, pages 4047–4051.



Fayer, J., Morin, G., Gasparini, S., Daisy, M., and Coudrin, B. (2018).

Radiometric confidence criterion for patch-based inpainting.

In *Proceedings of the International Conference on Pattern Recognition (ICPR 2018)*.
to appear.



He, K. and Sun, J. (2012).

Statistics of patch offsets for image completion.

In *Proceedings of the 12th European Conference on Computer Vision - Volume Part II, ECCV'12*, pages
16–29, Berlin, Heidelberg. Springer-Verlag.



Sinha, S. N., Steedly, D., Szeliski, R., Agrawala, M., and Pollefeys, M. (2008).

Interactive 3d architectural modeling from unordered photo collections.

ACM Trans. Graph., 27(5):159:1–159:10.



Waechter, M., Moehrl, N., and Goesele, M. (2014).

Let there be color! Large-scale texturing of 3D reconstructions.

In *Proceedings of the 2014 European Conference on Computer Vision (ECCV 2014)*, volume 8693 LNCS,
pages 836–850.



Zhang, E., Cohen, M. F., and Curless, B. (2016).

Emptying, refurbishing, and relighting indoor spaces.

ACM Trans. Graph., 35(6):174:1–174:14.

Annexe: AliceVision Results



Figure: AliceVision [AliceVision, 2017] results

Annexe: AliceVision Results



Figure: AliceVision [AliceVision, 2017] results

Annexe: AliceVision Results

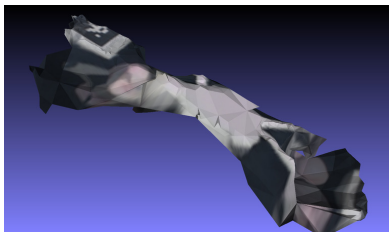


Figure: AliceVision [AliceVision, 2017] results

Annexe: AliceVision Results



Figure: AliceVision [AliceVision, 2017] results

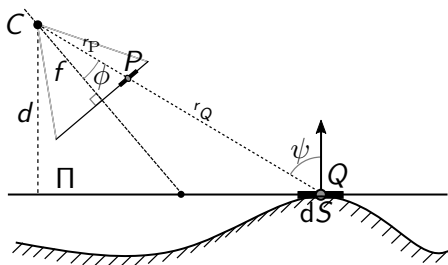


Figure: Confidence criterion applied to a viewpoint C and to a surface dS with tangent plane Π at point Q .

Annexe: Computation times

Plush	Blue	Mouse	Red	Bear	Rabbit
Perspective skeleton estimation	1.7 s	1.4 s	1.7 s	1.8 s	1.8 s
Triangulation	1.3 s	2.4 s	1.3 s	1.4 s	1.6 s
Z-buffering (CPU)	29 s	39 s	27 s	26 s	28 s
Texture extraction from images	9.9 s	15.1 s	9.6 s	8.4 s	9.8 s
Exposure correction	76 s	120 s	71 s	78 s	20 s
Registration correction	88 s	137 s	145 s	106 s	153 s
Texture completion	58 s	128 s	89 s	51 s	57 s
Total	264 s	443 s	345 s	273 s	271 s

Table: Computation times for some plushes