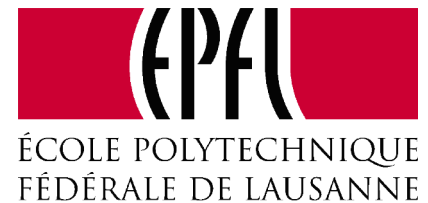


# Using Shape Priors to Regularize Intermediate Views in Wide-Baseline Image-Based Rendering



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# Organization of the presentation

- Context
- Current limitations in model/image-based rendering
- Contributions in wide-baseline rendering
  - Match epipolar segments based on virtual prior
  - Determine prior about the virtual shapes
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# Conventional video production fixes the viewpoint to one of the cameras that capture the scene





# Weaknesses of existing solution

- Main issues:
- Restricted camera coverage (cameras on the same side, outside the scene)
  - Non-smooth transitions between fixed cameras.



Camera-1



Camera-7



Camera-2



Camera-6

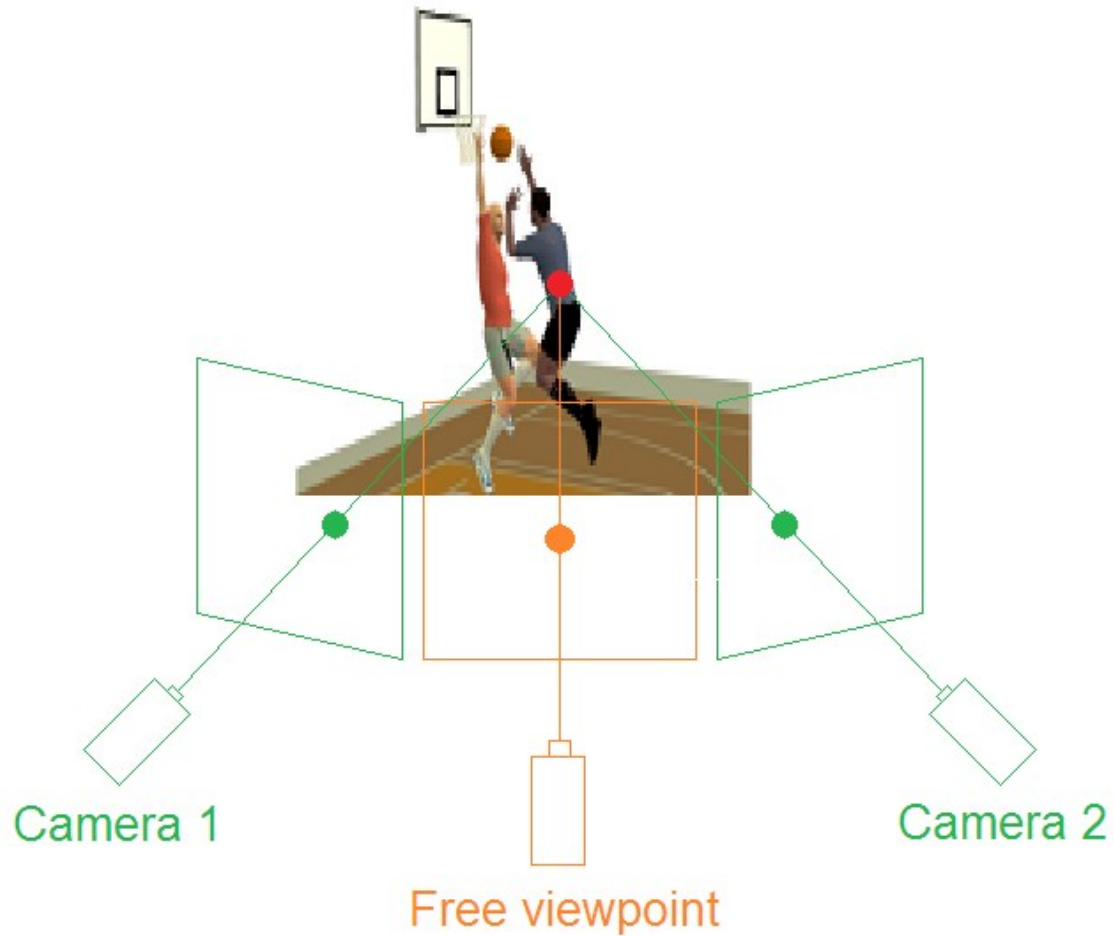


Camera-3



Camera-4

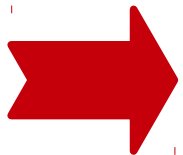
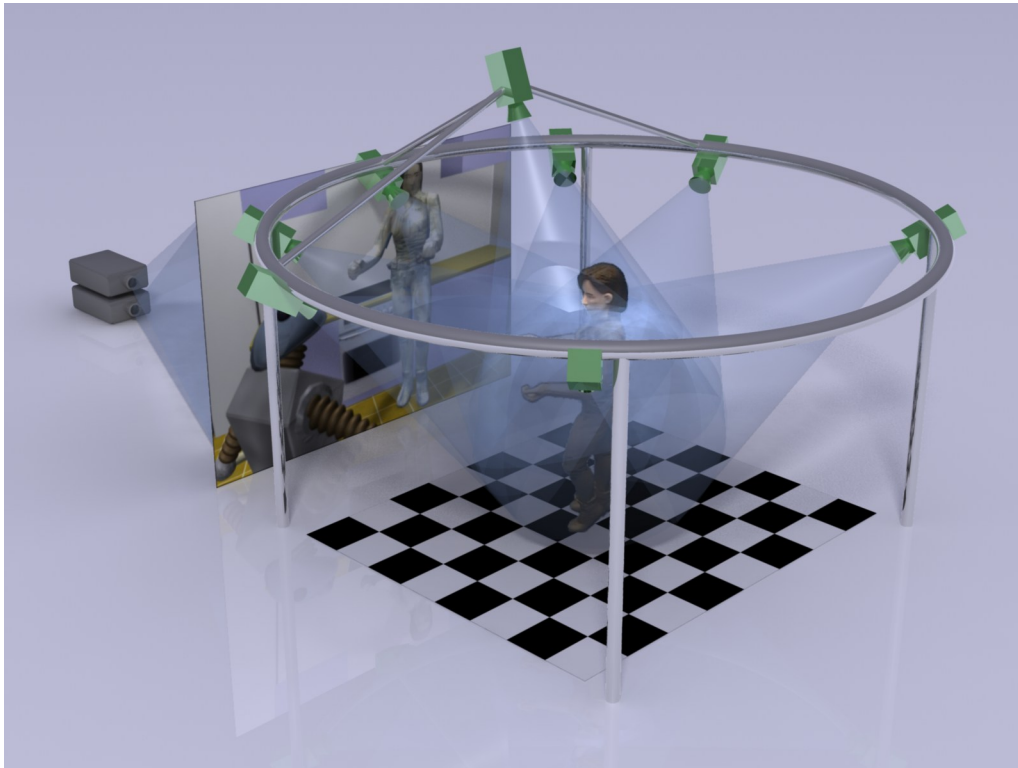
# What is free viewpoint rendering ?



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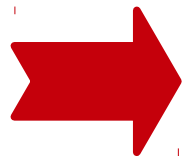
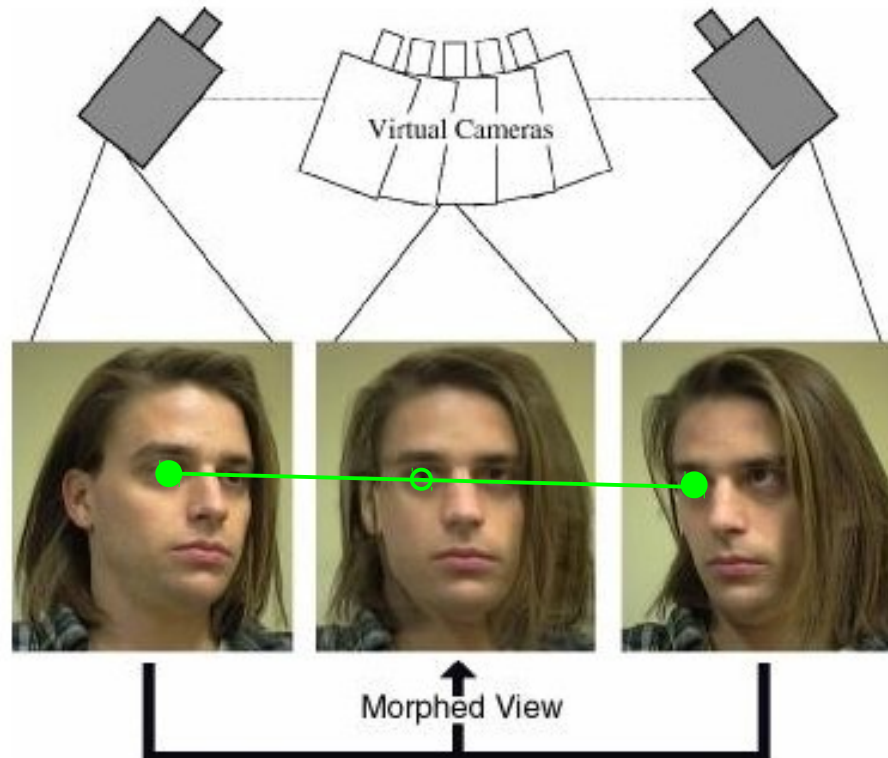
# Model-based rendering [1-4] projects an estimated 3D model on the virtual camera



**Realistic virtual view (accurate 3D model) only possible with dense camera coverage [5] or far views**

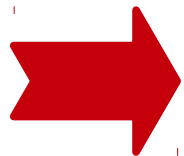
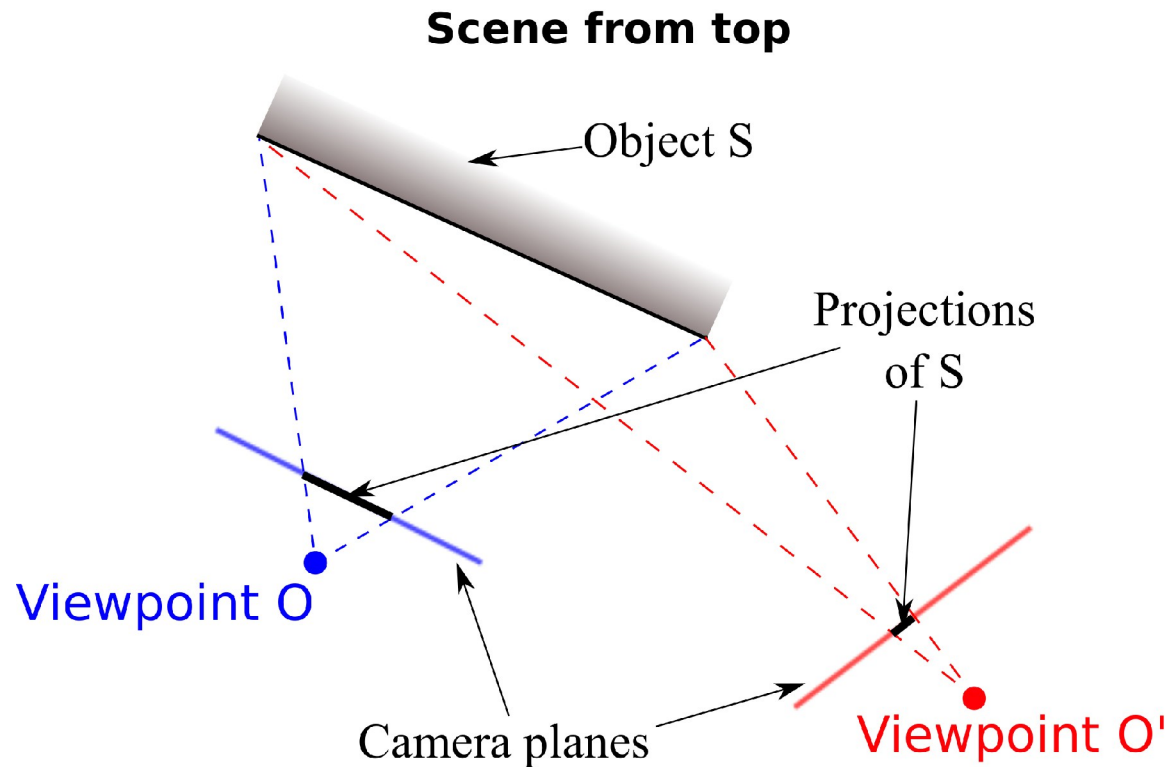


# Image-based rendering [6] interpolates intermediate virtual images based on dense matches



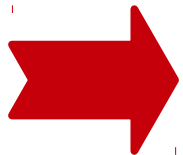
**Assumption : one-to-one pixel correspondences [7-8]**

**In wide-baseline, several pixels in one basis image can correspond to one pixel in the other basis view (foreshortening effect)**



**State-of-the-art stereo matching [9-11] makes the assumption of one-to-one correspondences**

**Also, because there is no correspondence in the occluded areas, the virtual view has important holes**



**Filling the holes is an ill-posed problem and can lead to unrealistic virtual views**

# Organization of the presentation

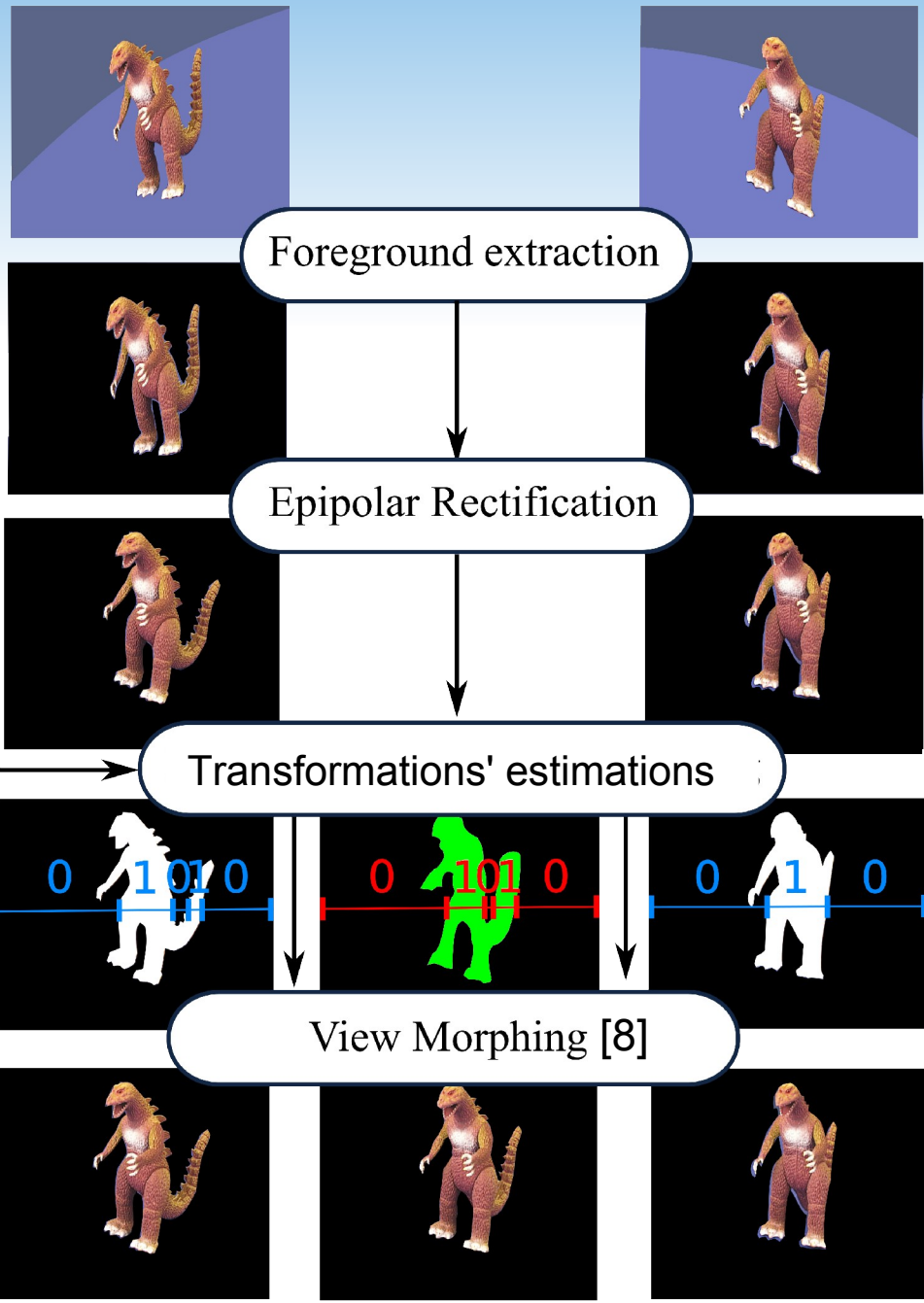
- Context
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# Organization of the presentation

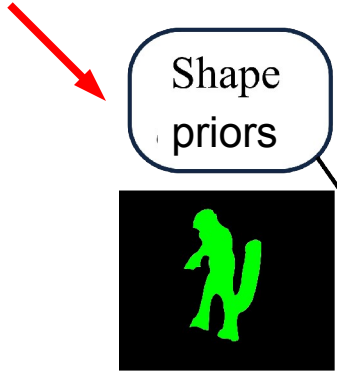
- Context
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*Left view*

*Right view*



**2)** Add prior information on the virtual views to disambiguate the matching



**1)** Consider epipolar foreground segments to tolerate foreshortening effect

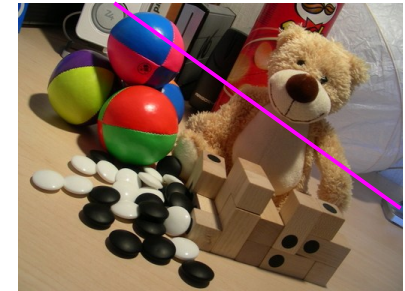


# What is an epipolar line segment ?

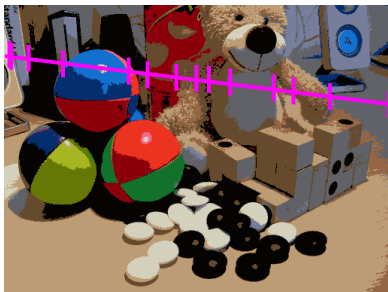
- New (and simple) image representation that describes the
  - Geometric constraints between the calibrated cameras



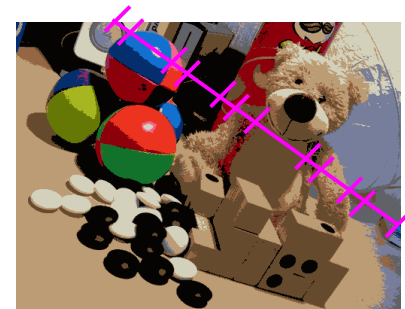
Epipolar geometry [12]



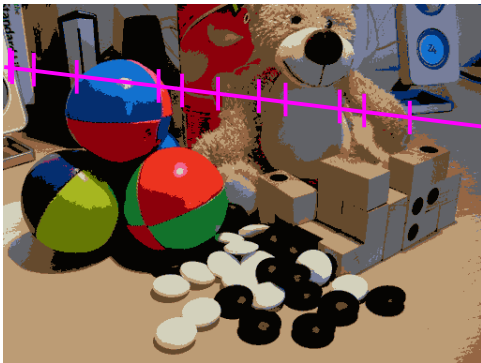
- Semantic content of the observed scene



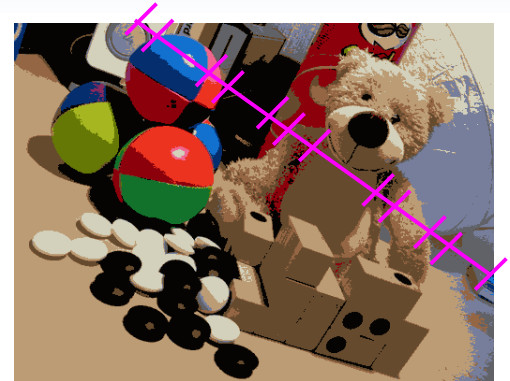
(Color) segmentation [13]



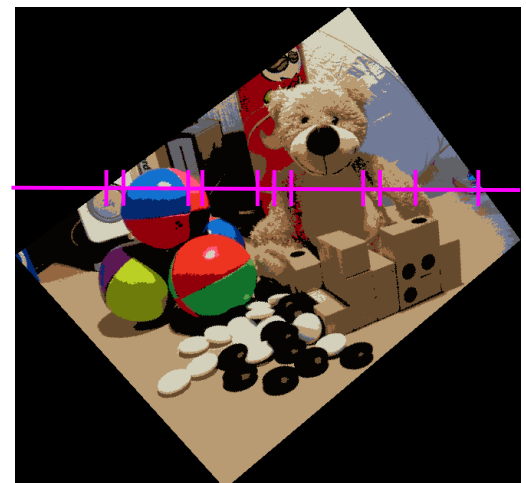
# Advantages of epipolar line segments ?



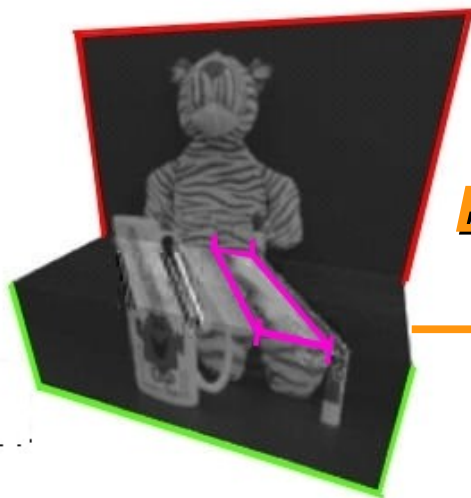
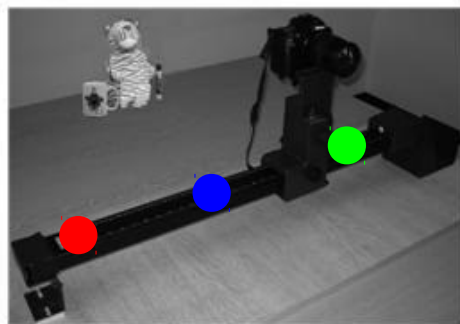
Model the foreshortening effect



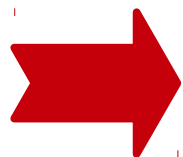
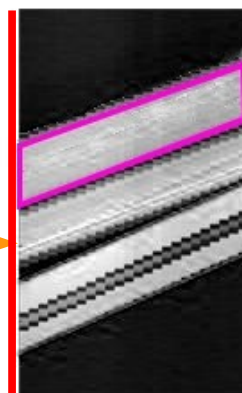
1D matching problem



# What happens to the epipolar line segments when the viewpoint changes?



*Epipolar  
plane  
image*



**Epipolar line segments undergo a 1D translation and 1D scaling**

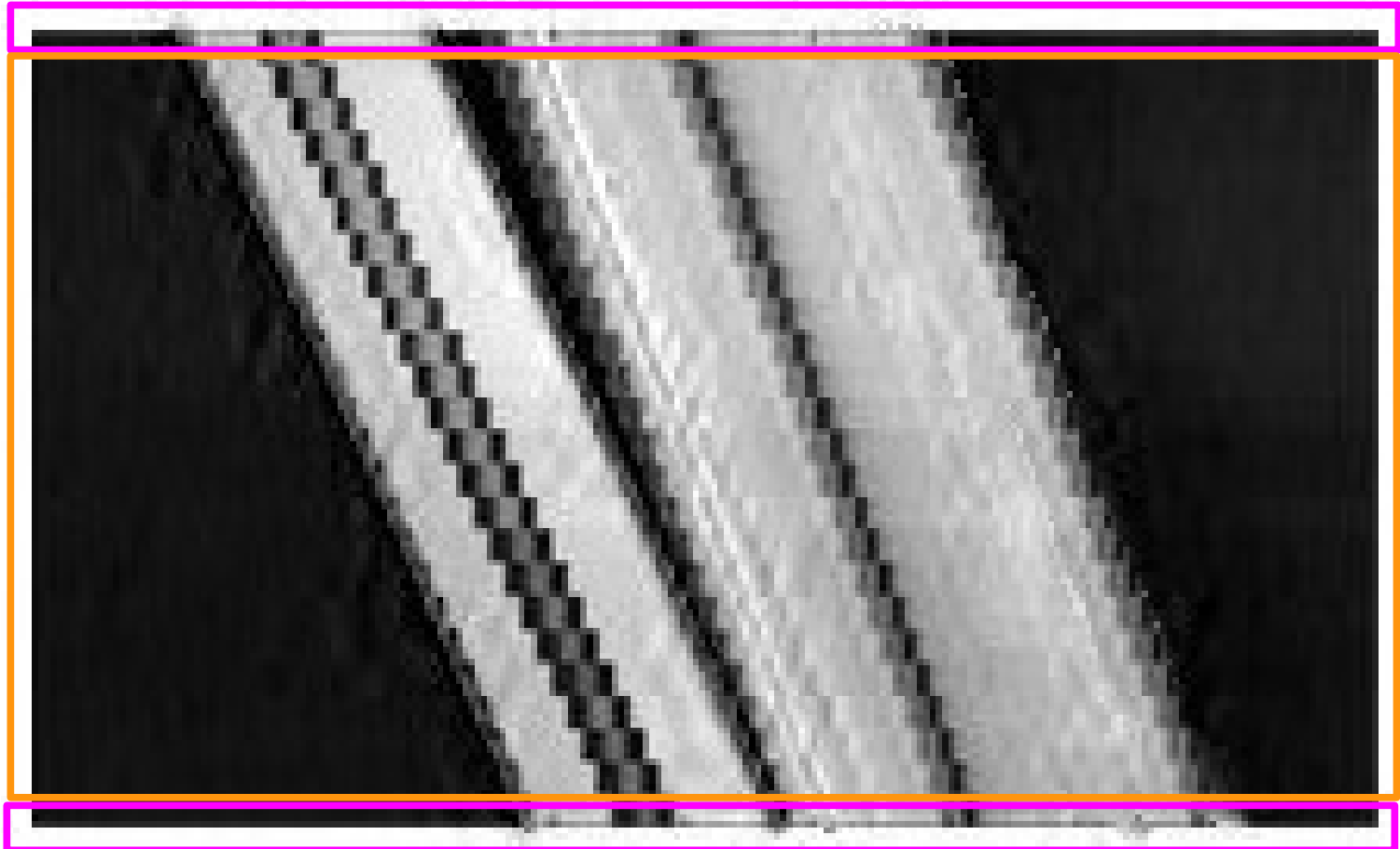
**View interpolation problem :**

**Given two epipolar sequences**

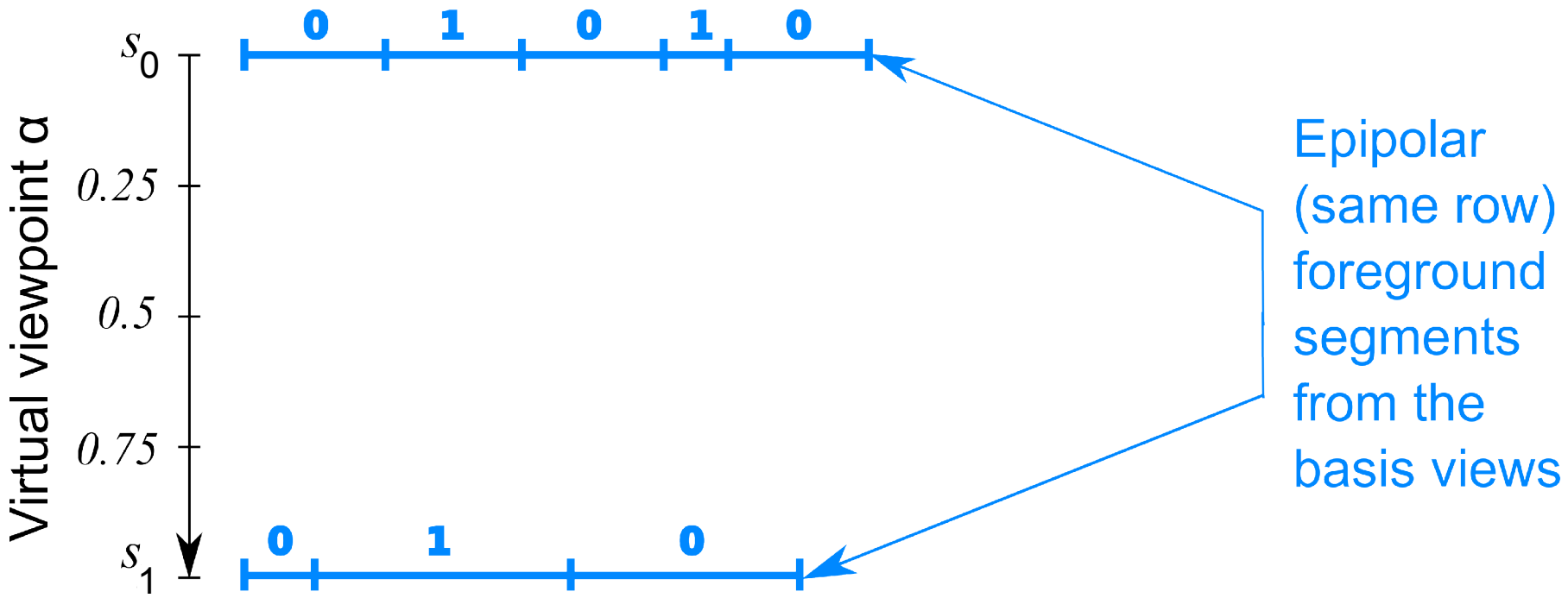


## View interpolation problem :

Given two epipolar sequences, reconstruct the EPI

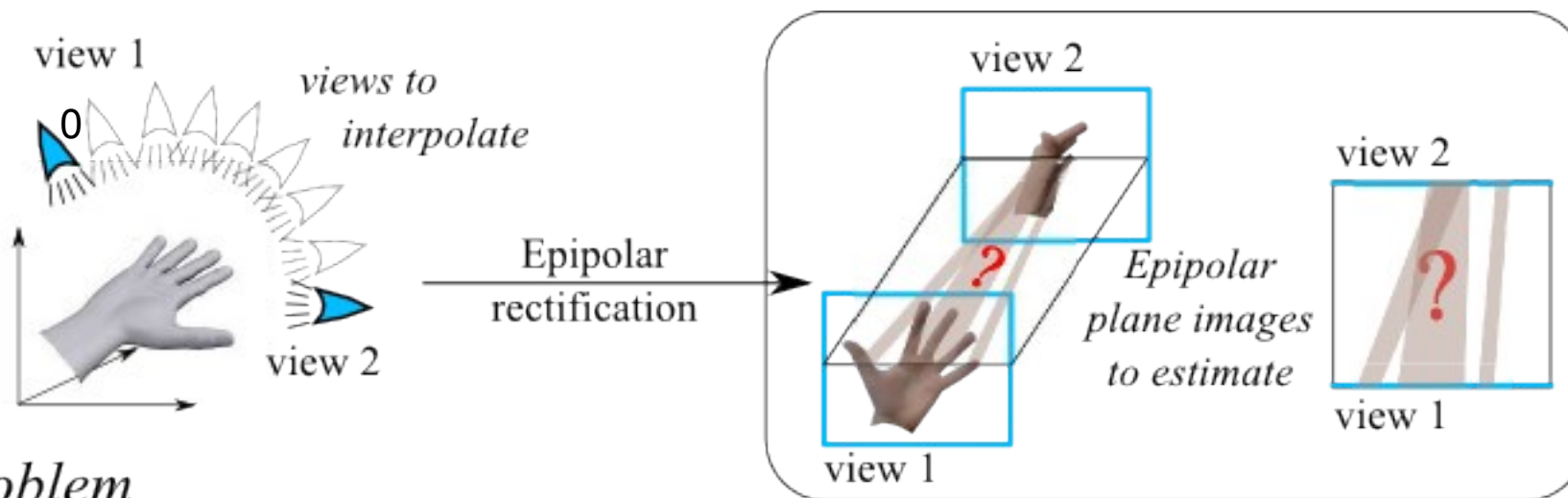


# Due to occlusions, EPI reconstruction is an ill-posed problem



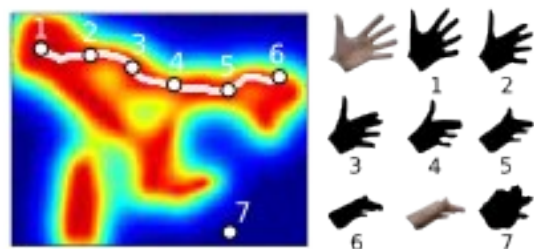


We propose to regularize the reconstruction so that the silhouettes of the reconstructed object tends to belong to the object's manifold.

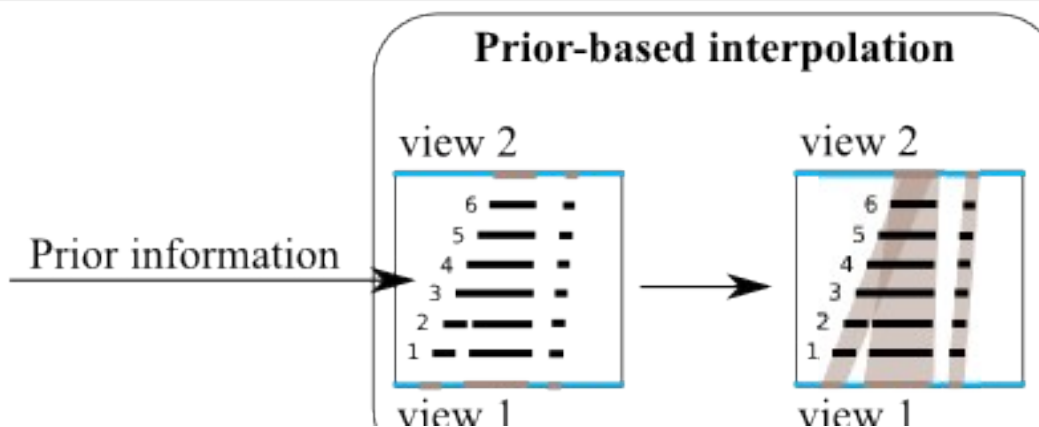


*Problem*

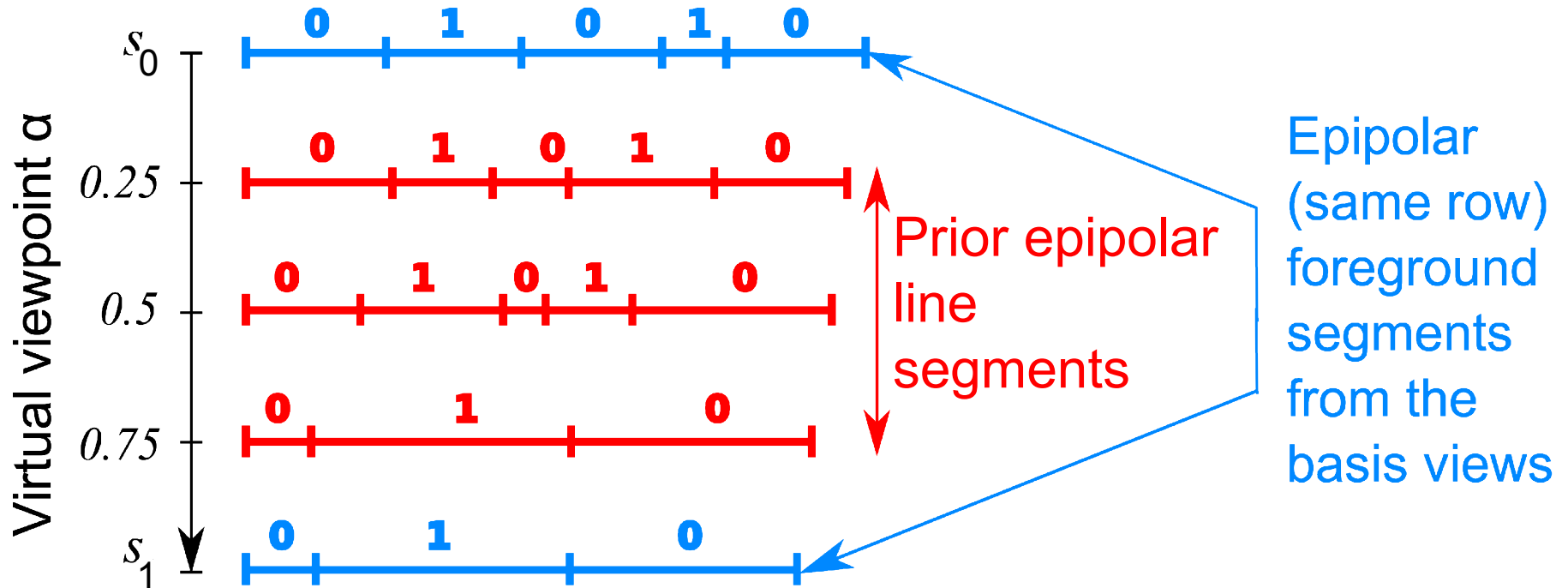
*Our solution*



*Learnt object manifold*

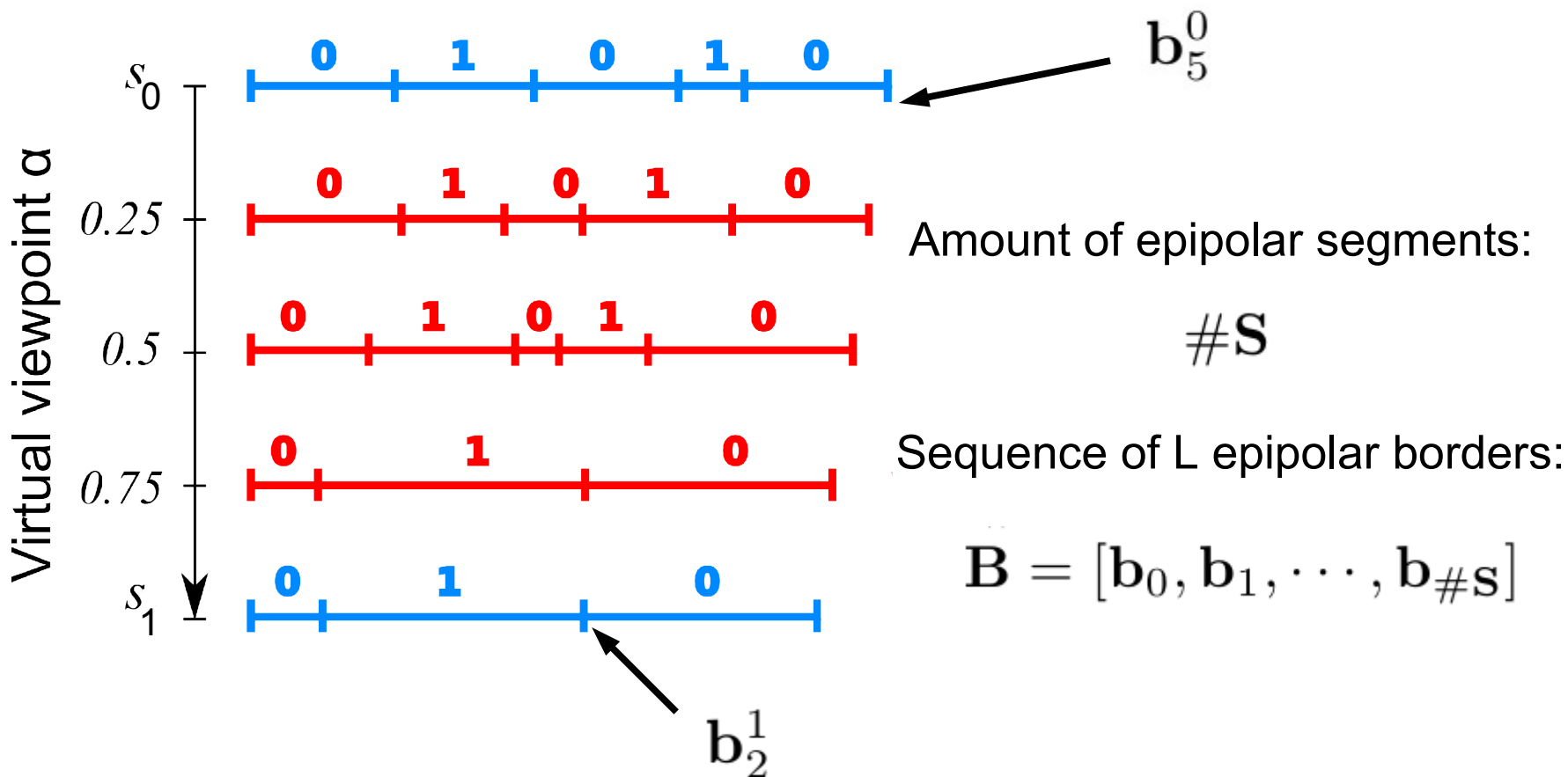


# Imagine we have a prior knowledge about the epipolar line segments in the intermediate views...



**Constrain (regularize) the possible transformations of the epipolar line segments**

# Instead of determining the segments' transformations, we estimate the displacement of their borders



# Step 1 : Identify, in the basis views, the matched epipolar borders and the occluded ones

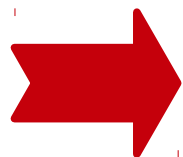
$$\min_{\mathbf{U} \in \{0,1\}^{L_0 \times L_1}} \sum_{i=0}^{L_0} \sum_{j=0}^{L_1} \mathbf{U}_{ij} \cdot \mathbf{D}_{ij} + \lambda \cdot \mathbf{C}_{ij}(1 - \mathbf{U}_{ij})$$

Cost of matching  
 $\mathbf{b}_i^0$  with  $\mathbf{b}_j^1$

Cost of considering  
 $\mathbf{b}_i^0$  (or  $\mathbf{b}_j^1$ ) as occluded

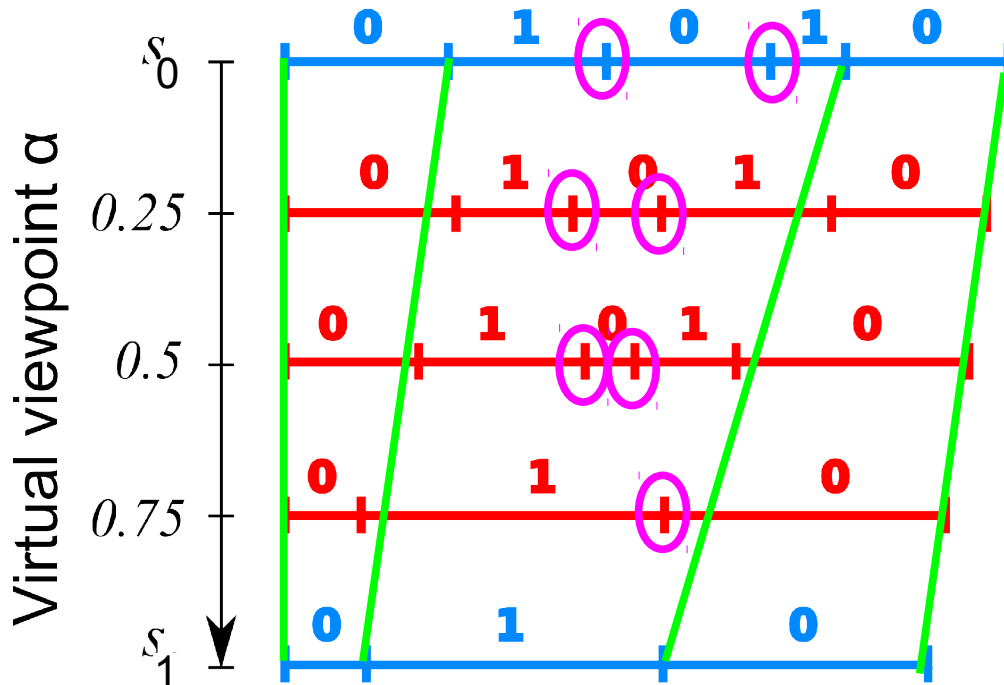
with the indicator matrix  $\mathbf{U} \in \{0, 1\}^{L_0 \times L_1}$

$$\text{s.t. } \begin{cases} \sum_{i=0}^{L_0} \mathbf{U}_{i,j} \leq 1 & \forall j \in \{0, \dots, L_1\} \\ \sum_{j=0}^{L_1} \mathbf{U}_{i,j} \leq 1 & \forall i \in \{0, \dots, L_0\} \end{cases}$$



Solved by dynamic programming (Needleman and Wunsch algorithm [14])

## Step 2 : Based on the **matched epipolar borders**, determine the **prior borders that are occluded**

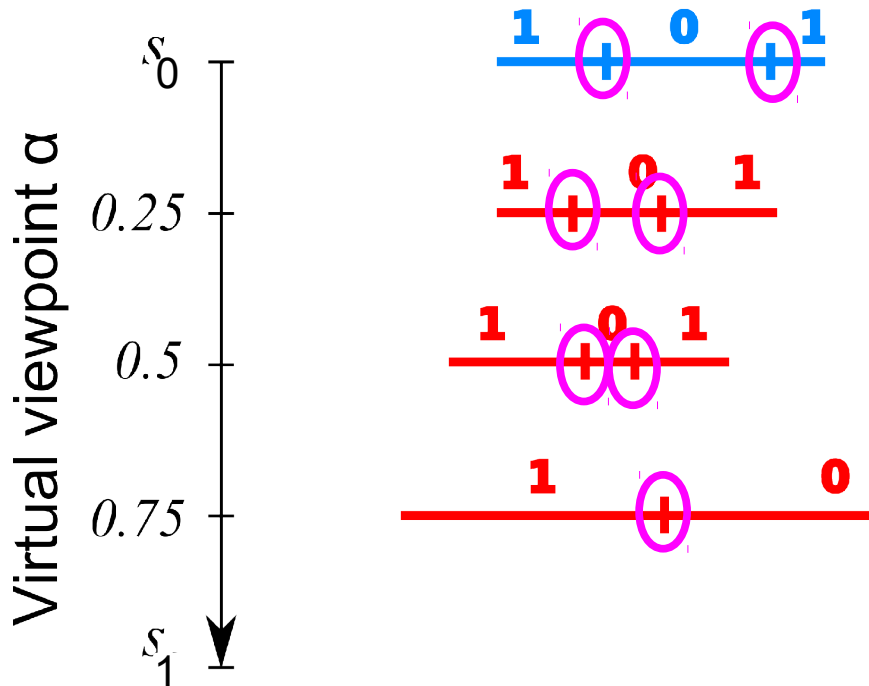


Prior borders are considered as occluded if their L1 distance to any of the interpolated path is higher than a threshold  $T$ .

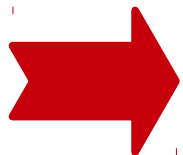
$T$  is set to an arbitrary (very) small value (e.g. 5% of the width of the interpolated image).

**→ The set of occluded prior borders is corrupted by matched prior borders**

# Step 3 : Estimate the trajectories of the occluded borders



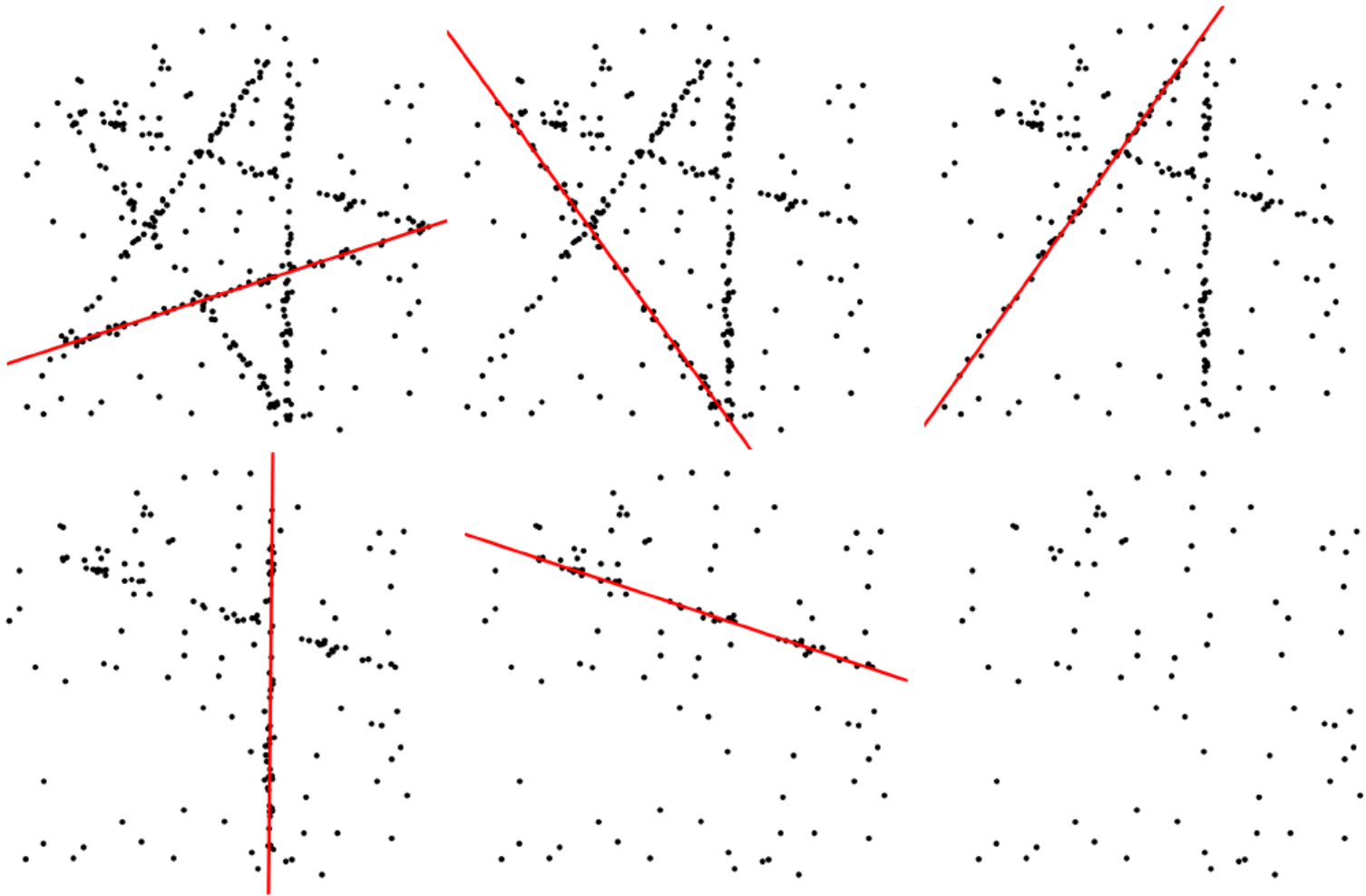
From the corrupted set of occluded prior borders



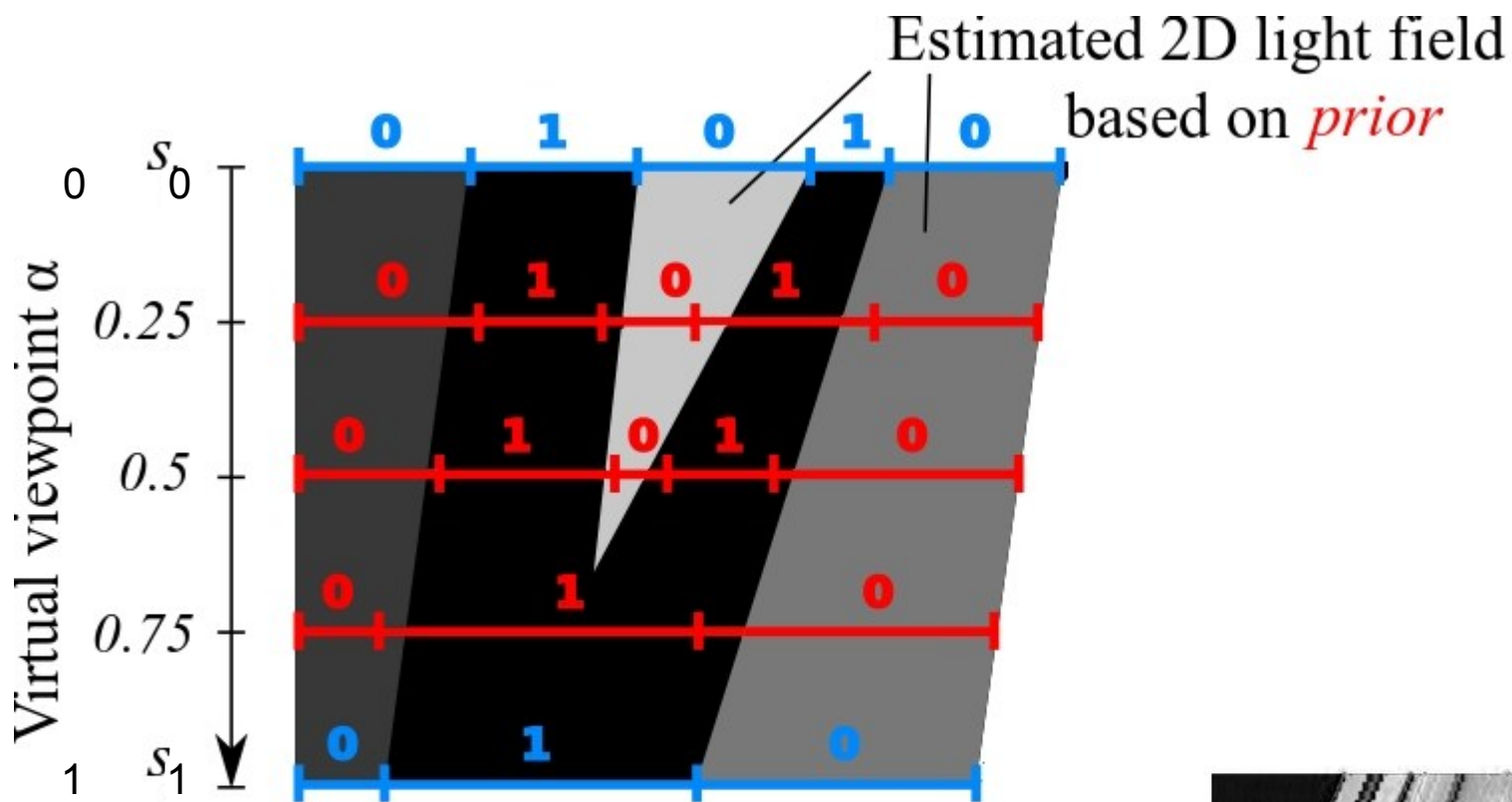
Apply RANSAC sequentially on borders with the same foreground/background transition



# Sequential RANSAC [15]



These steps enables to estimate the how epipolar line segments transform in-between the views



# Organization of the presentation

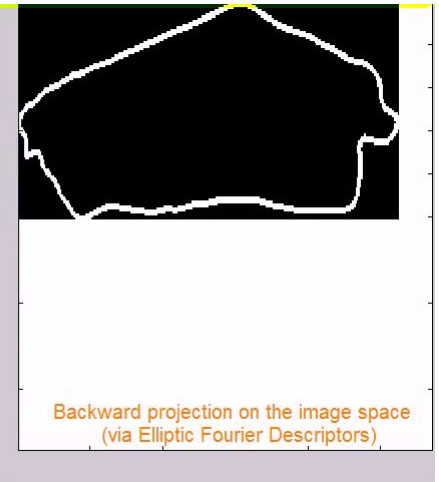
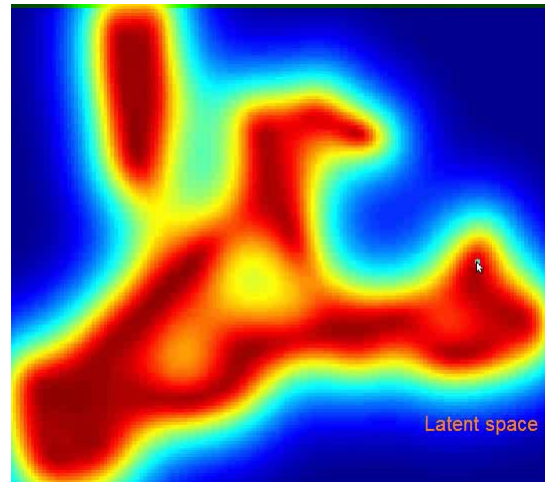
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# Defining shape priors in the intermediate views

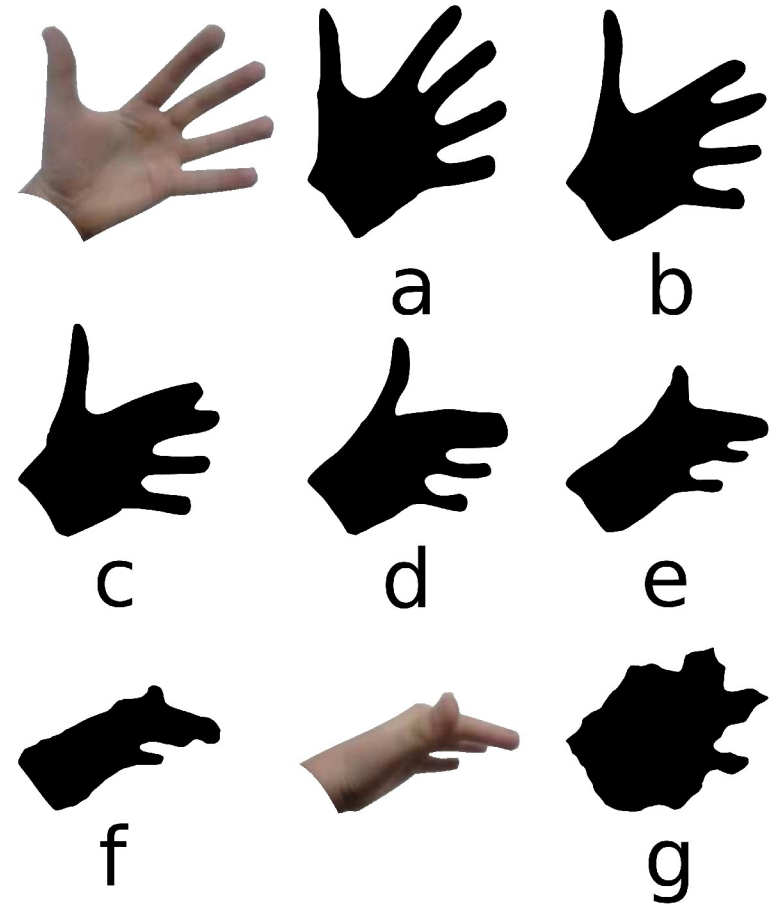
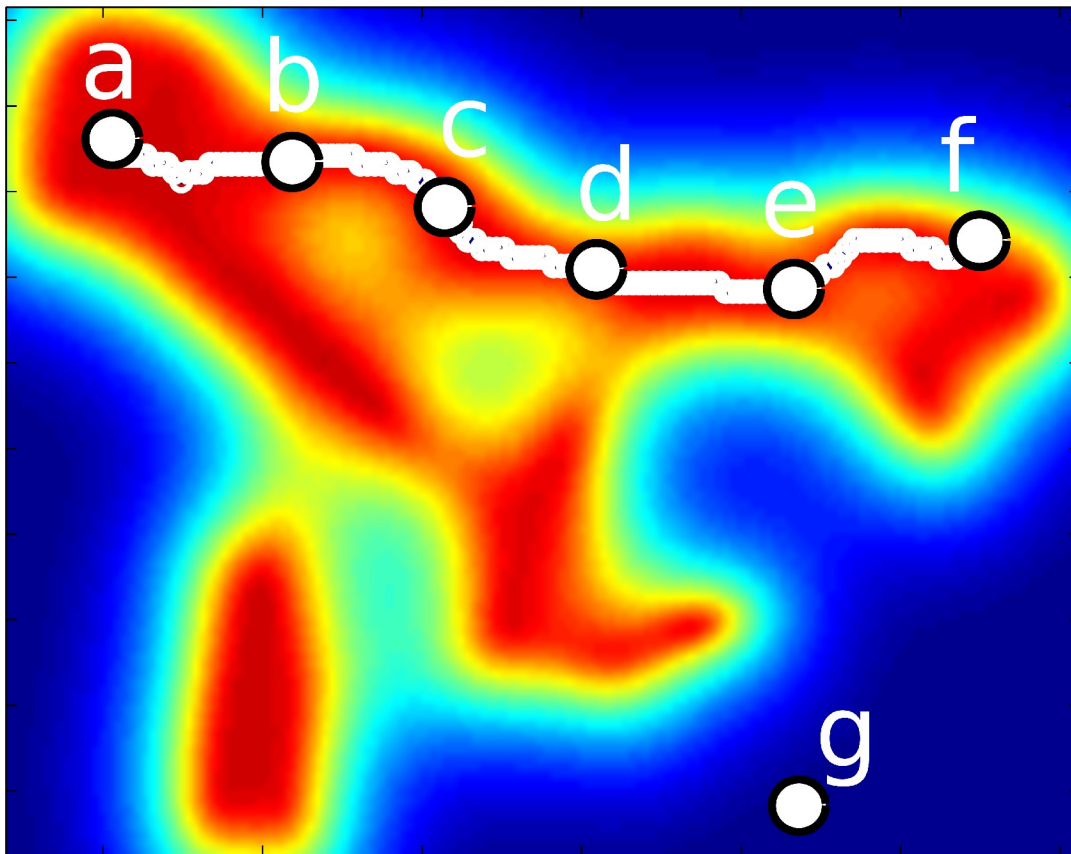
- 1) Learn a low-dimensional space representative of the shape of the analyzed object
- 2) Project the shapes of the basis views onto the low-dimensional space
- 3) Interpolate the prior intermediate shapes in this low-dimensional latent space and project them back onto the (high-dimensionality) shape space

# Learn a low-dimensional space representative of the shape of the analyzed object

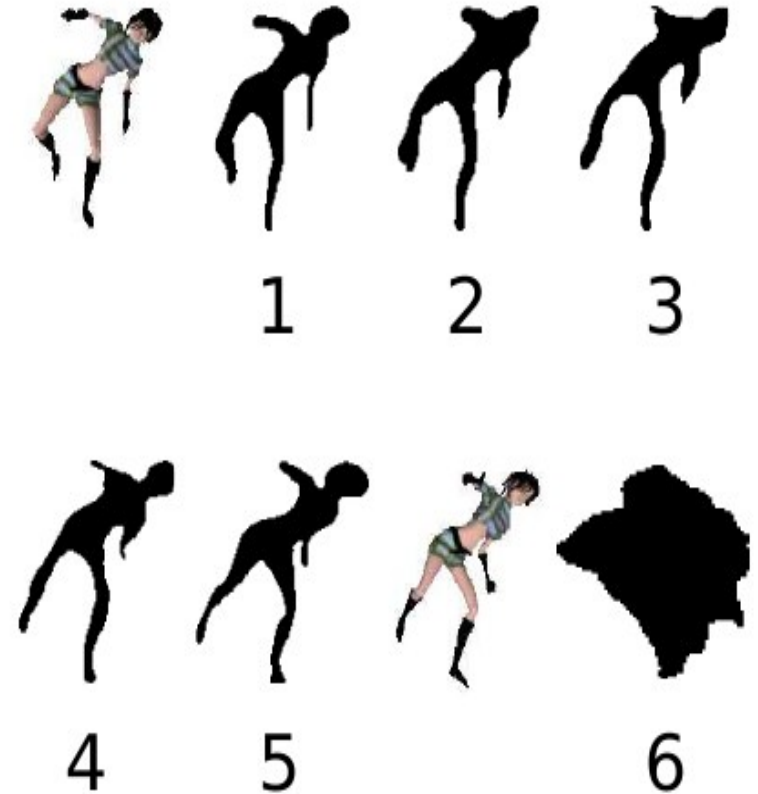
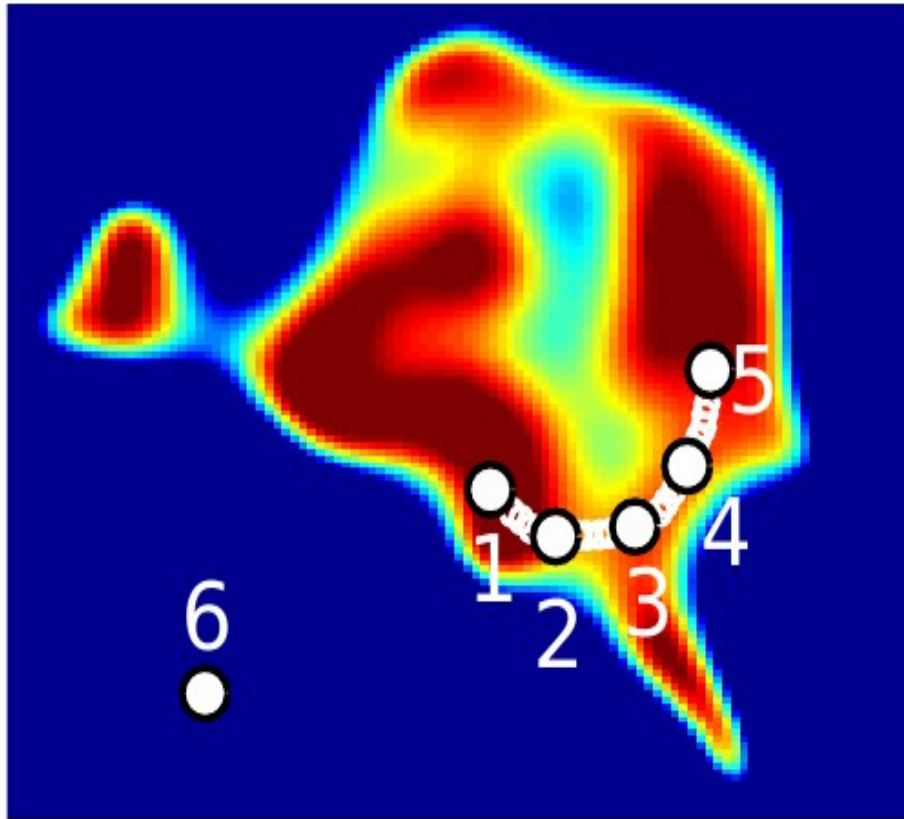
- Description of a given shape of an object based on Elliptic Fourier shape Descriptors (EFD) [16]
- Learn the latent space of the object by non-linear dimensionality reduction (by GPLVM [17]) of the EFD of multiple object's occurrences



Interpolate [18] the prior intermediate shapes in this low-dimensional latent space and project them back



Interpolate [18] the prior intermediate (virtual) shapes in this low-dimensional latent space and project them back



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- **Experimental validation**



# Result based on Zitnick et al., SIGGRAPH'04 [11]



# Dynamic programming [14] on epipolar foreground segments



# Proposed method (50 harmonics for EFD and 6 intermediate priors to regularize the matching)



# Proposed method (50 harmonics for EFD and 6 intermediate priors to regularize the matching)



$\alpha = 0$

$\alpha = 0.25$

$\alpha = 0.5$

$\alpha = 0.75$

$\alpha = 1$

# Conclusion

- Free viewpoint rendering enables to virtually navigate across a scene
- The foreshortening effect, as well as the occlusions, make the state-of-the-art methods fail in wide-baseline configurations
- We propose to match foreground segments between two very different viewpoints to tolerate the foreshortening effect
- We propose to add prior information on the intermediate views to disambiguate the ill-posed matching
- This allows to determine realistic vanishing and appearing trajectories of occluded parts
- Perspective : regularize based on color segments instead of foreground segments

# Some references

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- (2) W. Matusik, C. Buehler, R. Raskar, S. J. Gortler, and L. McMillan, “Image-based visual hulls”, in Proc. of the ACM GRAPHITE, pages 369–374, 2000.
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- (14) I. J. Cox, S. L. Hingorani, S. B. Rao, and B. M. Maggs, “A maximum likelihood stereo algorithm”, in CVIU, 63(3):542-567, 1996.
- (15) R. Toldo and A. Fusiello, “Robust multiple structures estimation with j-linkage”, in ECCV, pages 537-547, Springer, 2008.
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- (17) N. D. Lawrence and J. Quinonero-Candela, “Local distance preservation in the GPLVM through back constraints”, in ACM NIPS, pages 513–520, 2006.
- (18) E. W. Dijkstra, “A note on two problems in connexion with graphs”, in Numerische mathematik, 1(1):269–271, 1959.

Thank you very much for your attention...

Questions?



