

WARP-BASED NEAR-REGULAR TEXTURE ANALYSIS FOR IMAGE-BASED TEXTURE OVERLAY

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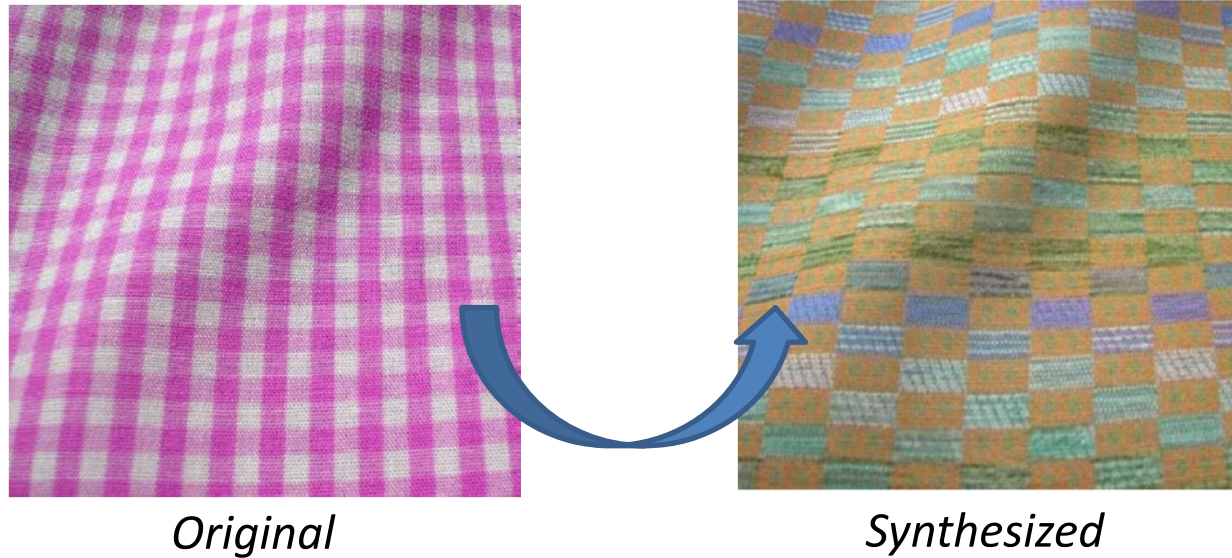
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Visual Computing
Berlin, Germany



16th International Workshop on Vision, Modeling and Visualization, VMV 2011



Image-based Texture Overlay



Preservation of

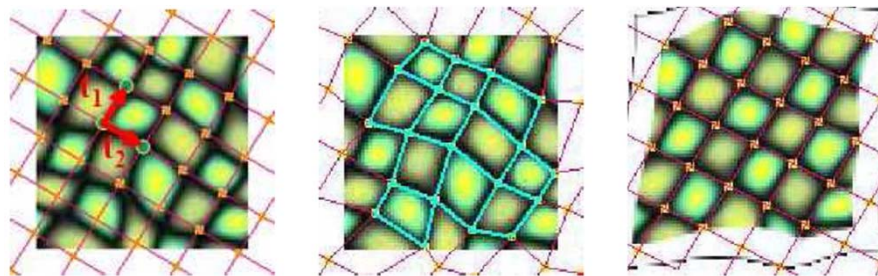
- texture distortion
- shading and reflection properties

Modification of

- texture albedo only

Related Work – Texture Overlay

Videos	Pilet2005 Scholz2006 Hilsmann2010	Estimation of texture distortion and shading properties in relation to a reference
Single Views using Shading Information	Fang2004 Guo2008 Yan2010	<ul style="list-style-type: none"> • No reference given • Texture-less surfaces • Estimation of texture distortion from shading
Single Views using Texture Information	Liu2004	<ul style="list-style-type: none"> • No reference given • Near regular textures • Interactive texture analysis



Related Work – Texture Overlay

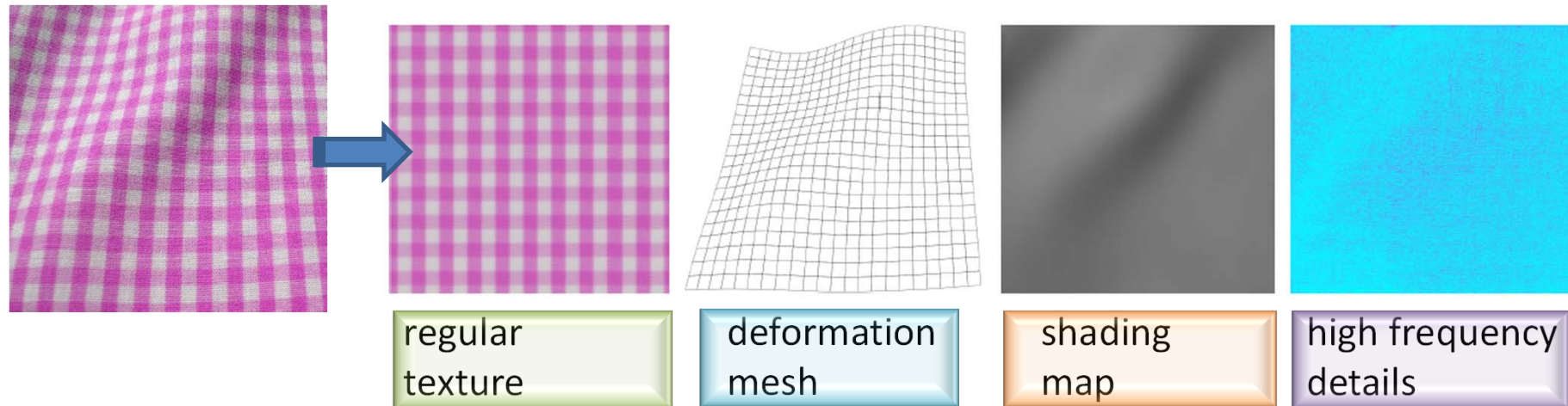
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Single Views using Texture Information	Liu2004 our method	<ul style="list-style-type: none"> • No reference given • Near regular textures • Interactive texture analysis • automatic warp-based geometric and photometric texture analysis

Near-Regular Textures



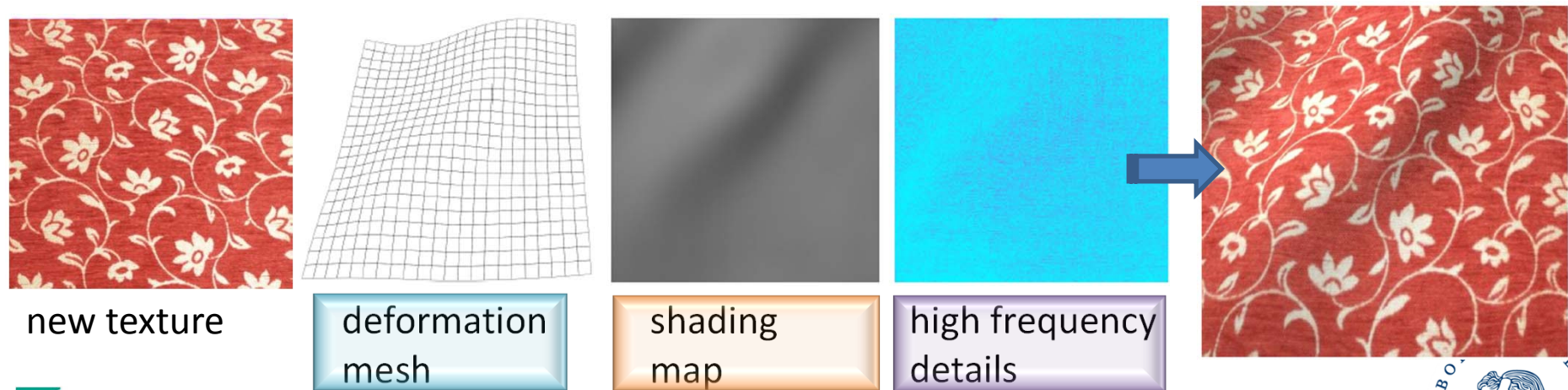
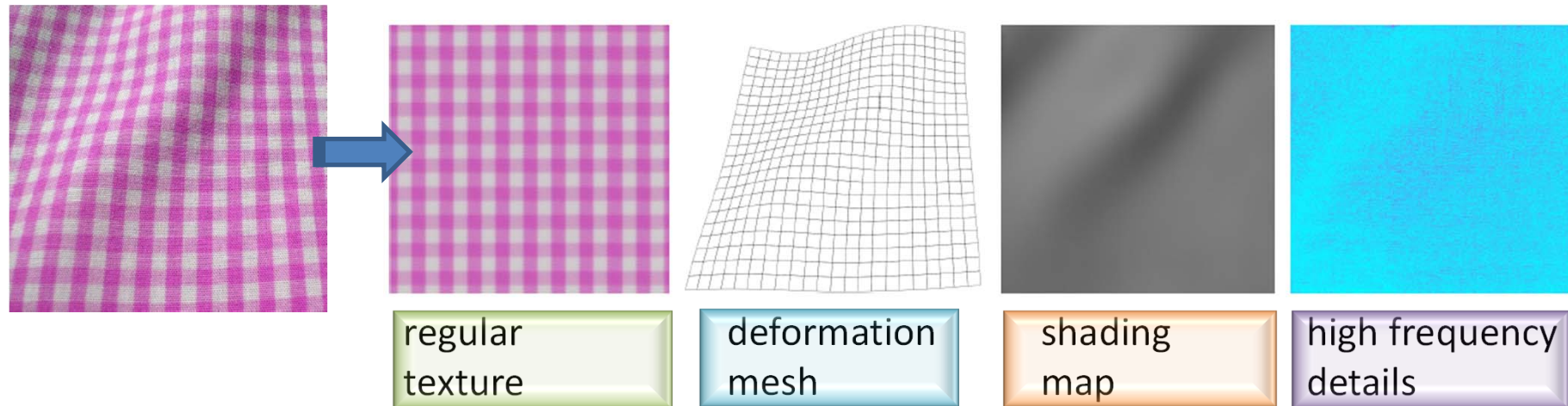
- Regular textures: constructed by regularly tiling the texture space with the same texture element
- Near-regular textures (Liu 2004):
 - Geometric and photometric deviation from regularity
 - Same topological regularities and relations

Texture Decomposition- Image Model



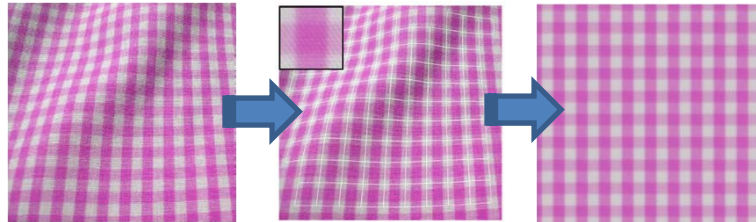
$$\mathcal{I}(\mathbf{x}) = \mathcal{W}_p(\mathbf{x}) \cdot \mathcal{T}(\mathcal{W}_g(\mathbf{x})) + \alpha \cdot \mathcal{HF}(\mathbf{x})$$

Texture Decomposition- Image Model

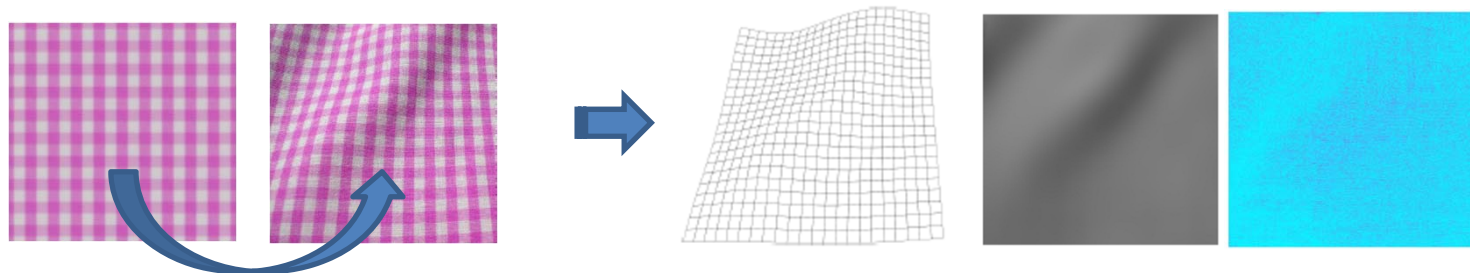


Overview

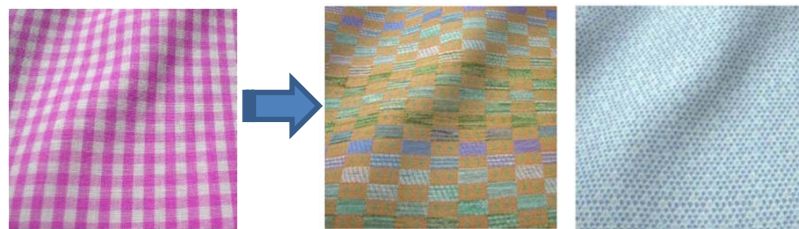
- Mean texel appearance and lattice estimation



- Texture decomposition through joint geometric and photometric registration



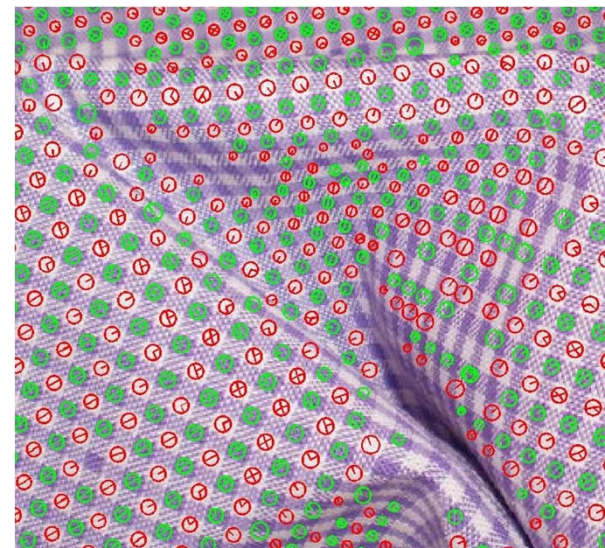
- Texture replacement



Mean Texel Appearance and Lattice Estimation

- Texels appear geometrically and photometrically distorted
- Nevertheless, they are still *similar*

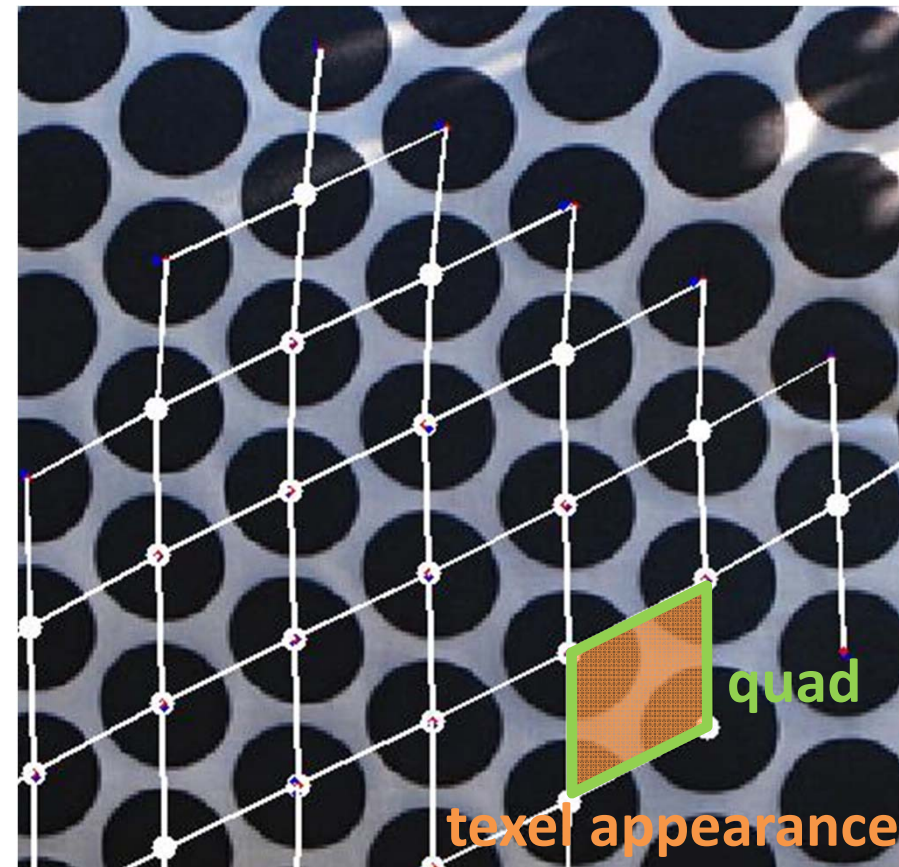
➔ Idea: unsupervised clustering of feature descriptors (e.g. SIFT)



Lattice Generation

- Each cluster consists of image points with similar SIFT descriptors
- Goal: generate degree-4-**lattice** that relates the points of each cluster by the same **topological relations** as in the undeformed texture (Liu et al 2004)

! there is more than one valid
● texel and tiling pattern

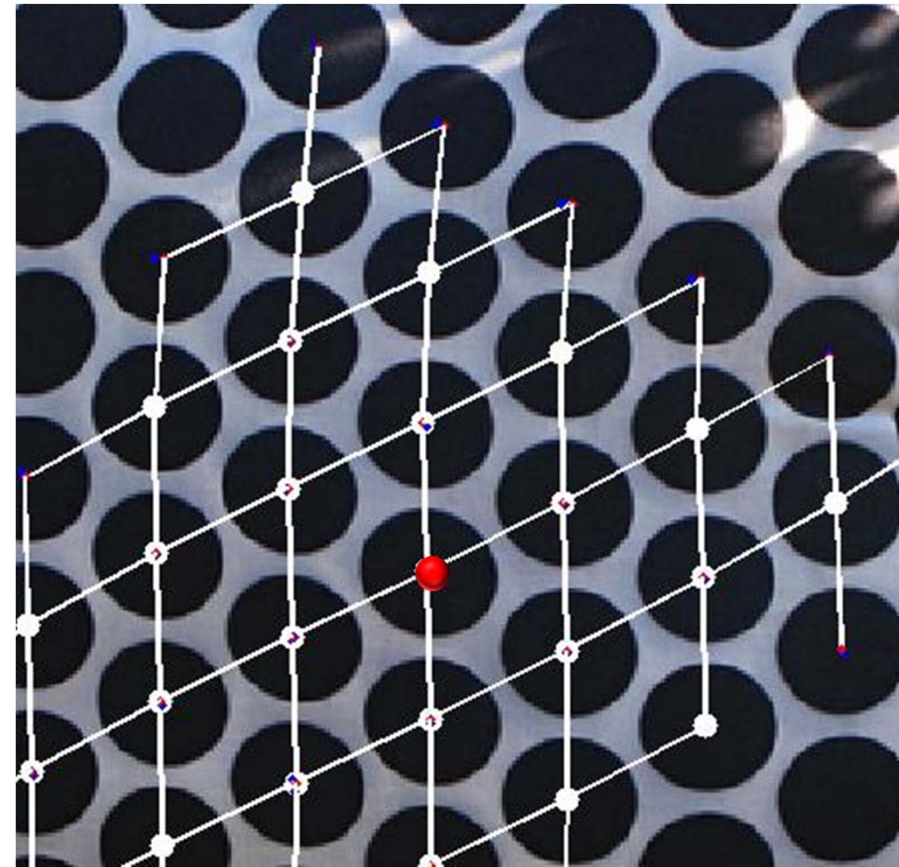


SIFT feature positions

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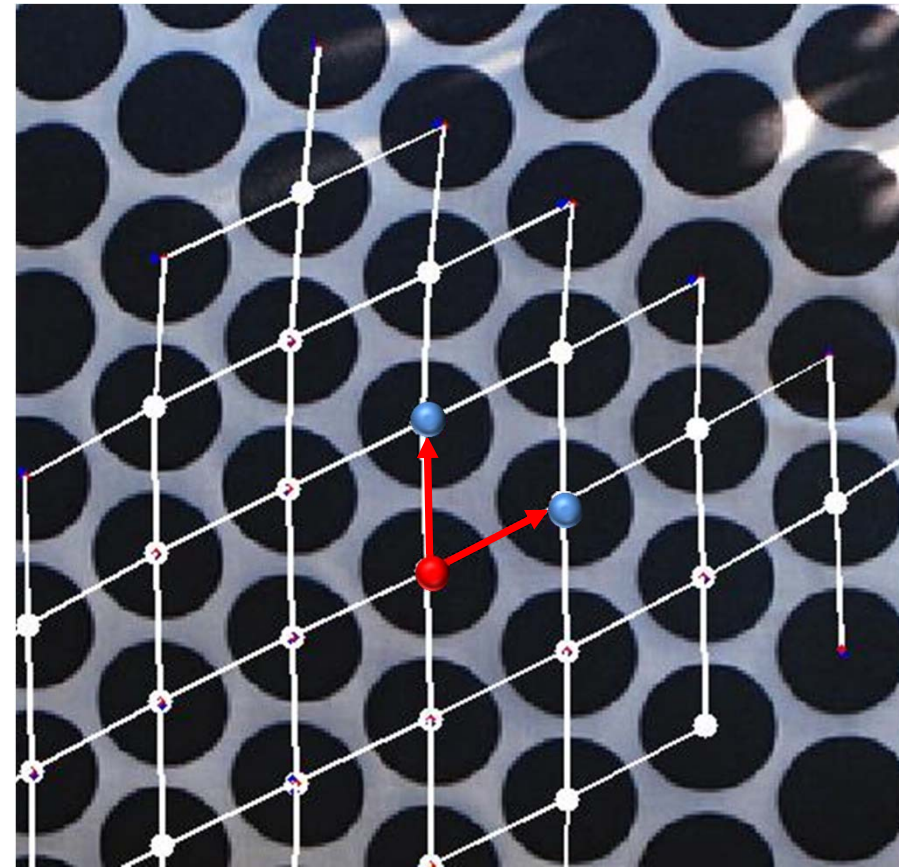


seed point \mathbf{p}_s

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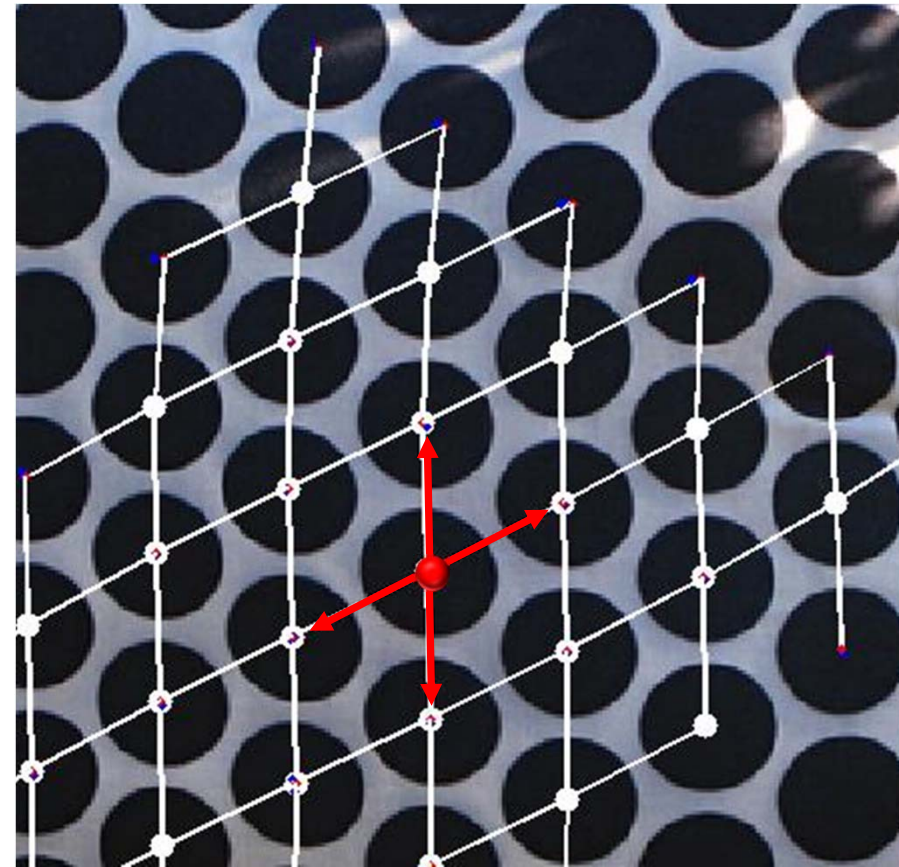


proposing vectors $\mathbf{V}_1, \mathbf{V}_2$

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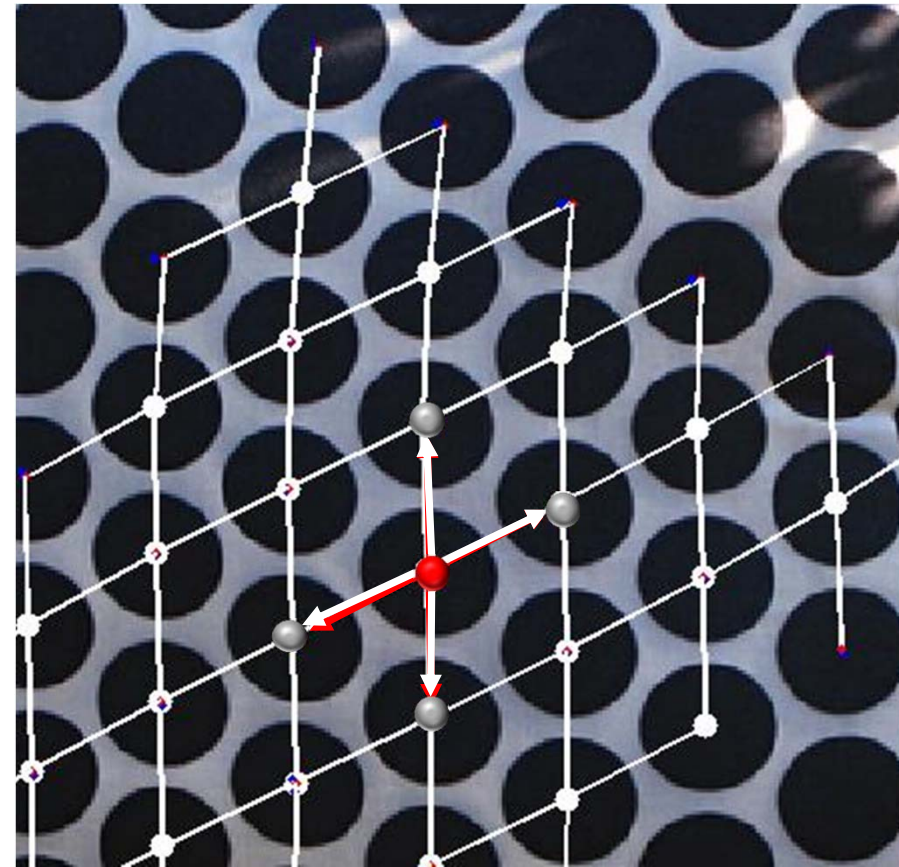


Find cluster points in a predefined distance from $\mathbf{p}_S \pm \alpha \mathbf{v}_1, \mathbf{p}_S \pm \alpha \mathbf{v}_2$
 $0.5 \leq \alpha \leq 2$

Lattice Generation

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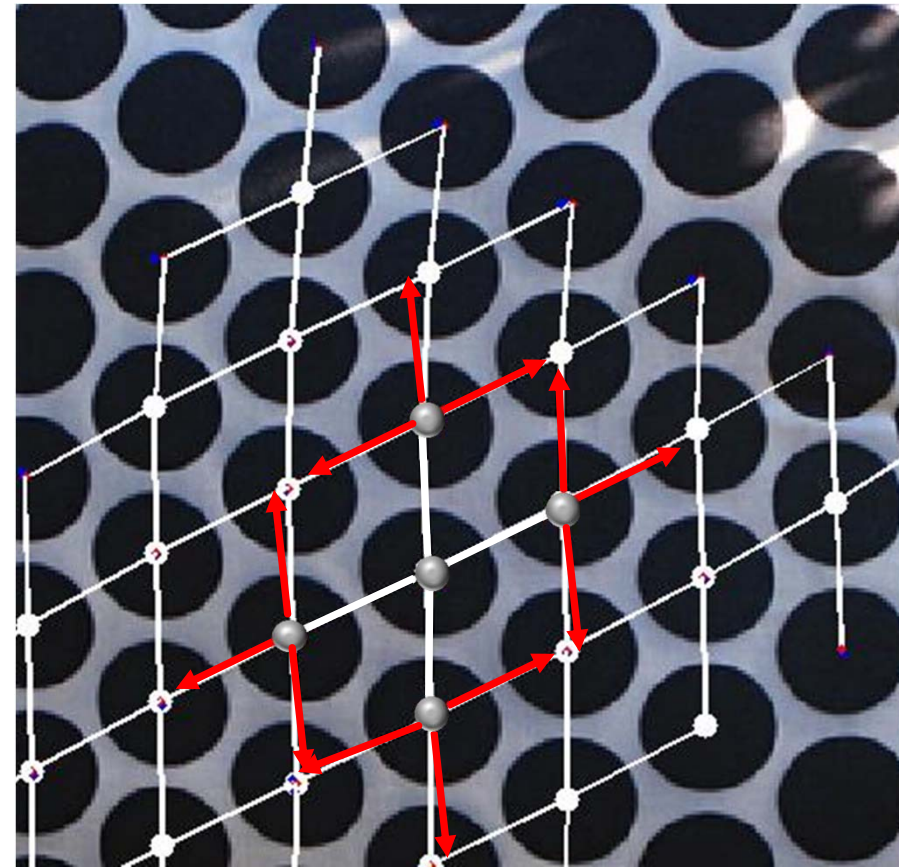


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Lattice Generation

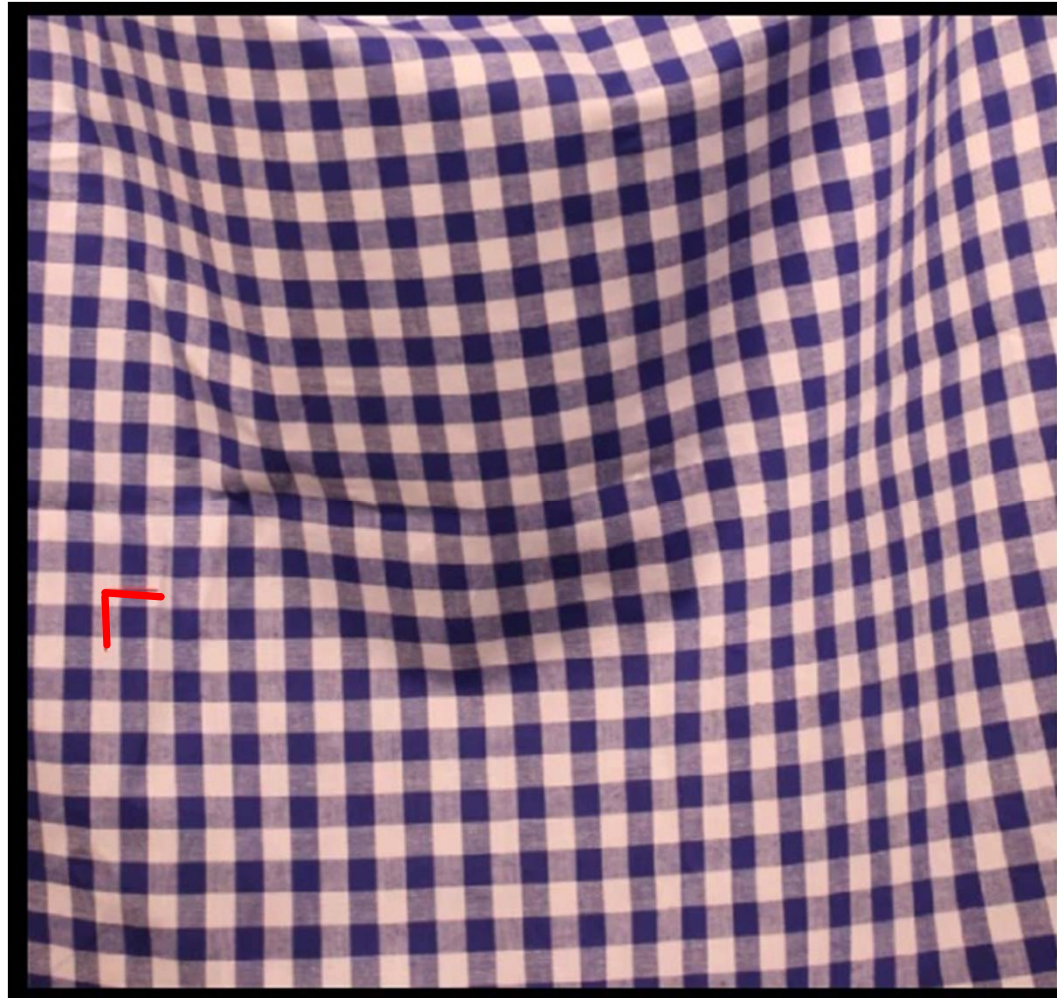
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for all children update proposing vectors V_1, V_2

Lattice Generation



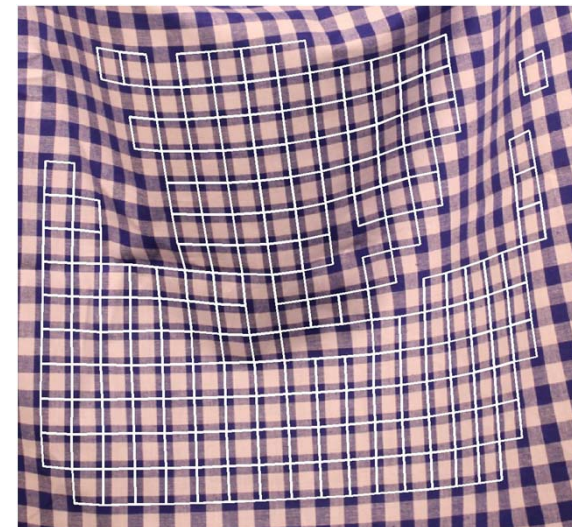
Mean Texel Appearance and Lattice Estimation

- Map all texels into a normalized texel coordinate system
- Normalize intensities
- Rectified and normalized mean texel
- MAD-based outlier rejection based on similarity to mean texel and additional continuity measure

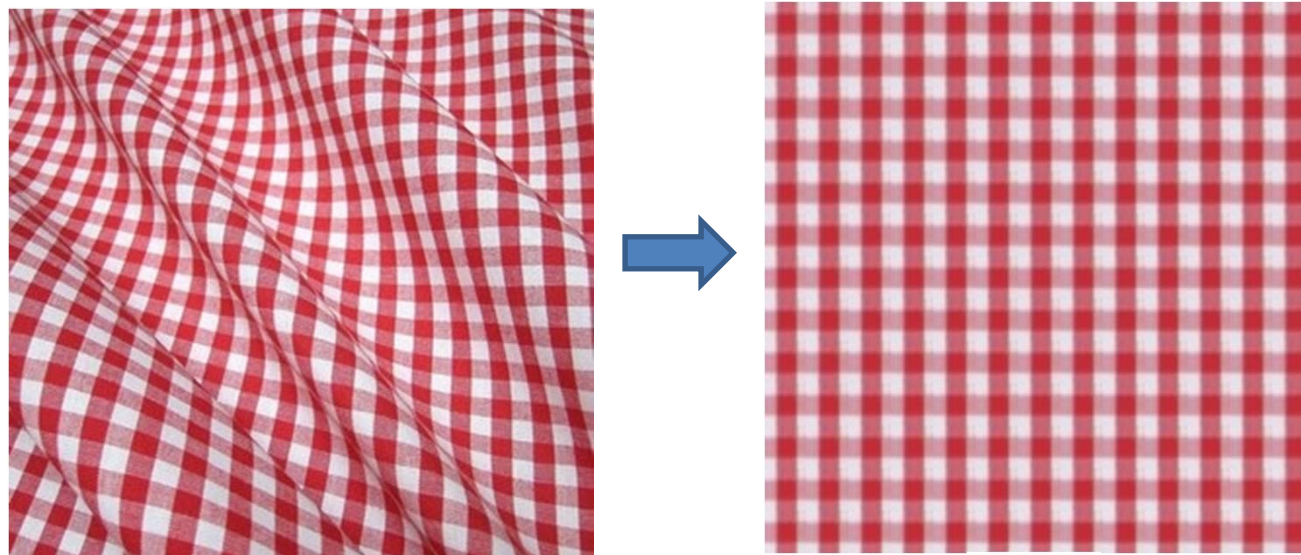
➔ From remaining normalized texels estimate mean texel appearance and coarse lattice structure



texels in normalized texel coordinate frame



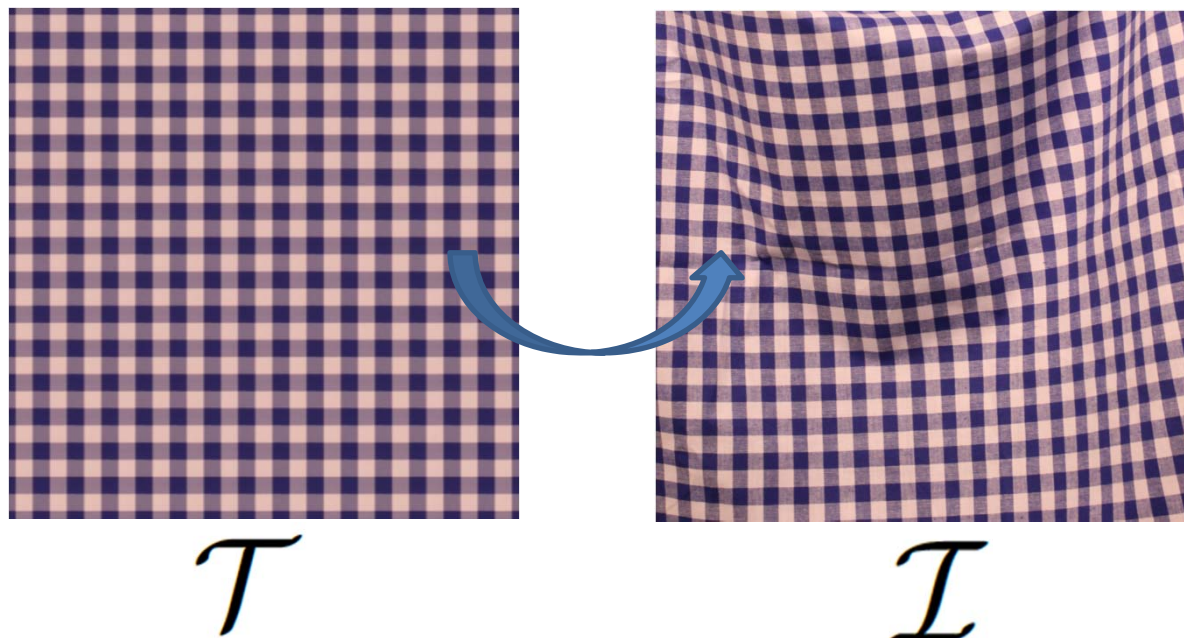
Results – Texel Appearance Estimation



Joint Deformation and Shading Optimization

- Joint optimization of a geometric and a photometric warp between the regular texture and the original image

➔ Refinement of texture deformation with a finer mesh and estimation of a shading map



Joint Deformation and Shading Optimization

- **Relaxed** Brightness Constancy Assumption

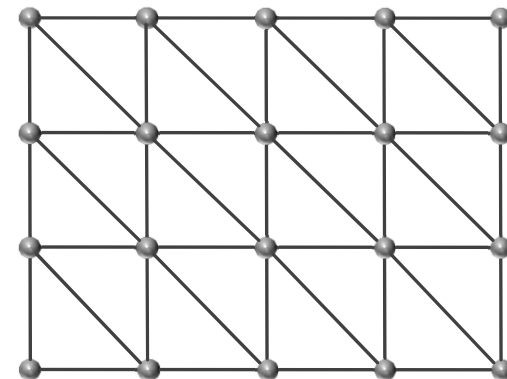
$$r_i(\theta) = \mathcal{W}_p(\mathbf{x}_i; \theta_p) \cdot \mathcal{T}(\mathcal{W}_g(\mathbf{x}_i; \theta_g)) - \mathcal{I}(\mathbf{x}_i)$$

- Mesh-based Warps

geometric warp $\mathcal{W}_g(\mathbf{x}_i; \theta_g) = \mathbf{x}_i + \mathbf{B}_g^i \cdot \theta_g$ ← Vertex displacements

photometric warp $\mathcal{W}_p(\mathbf{x}_i; \theta_p) = \mathbf{B}_p^i \cdot \theta_p$ ← Intensity scale

$\mathbf{B}_g^i \mathbf{B}_p^i$ Warp parameterization (e.g. barycentric or bilinear mesh coordinates)



A. Hilsmann and P. Eisert: "Joint Estimation of Deformable Motion and Photometric Parameters in Single View Video", *ICCV NORDIA Workshop*, Kyoto, Japan, Sept. 2009

Joint Deformation and Shading Optimization

- Joint optimization of a geometric and a photometric warp between the regular texture and the original image

$$\hat{\theta} = \arg \min_{\theta} \mathcal{E}_D(\theta) + \lambda^2 \mathcal{E}_S(\theta)$$

data $\mathcal{E}_D(\theta) = \sum_i \psi(r_i(\theta))$

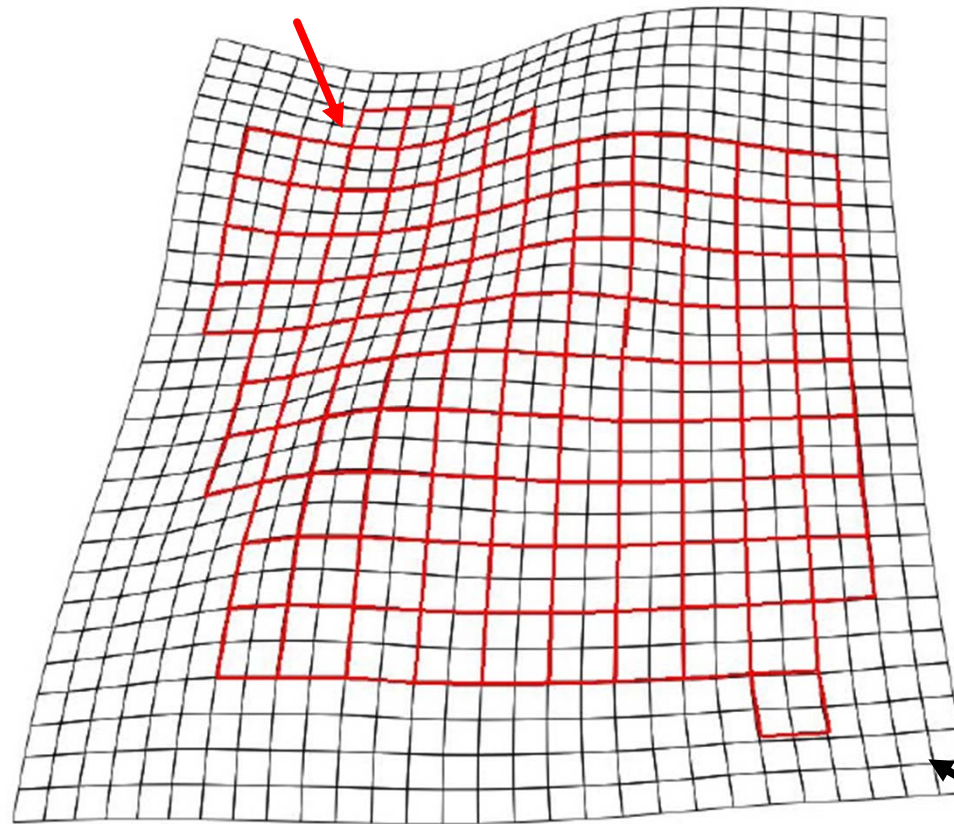
smoothness $\mathcal{E}_S(\theta) = \Gamma \cdot \theta$

$$\Gamma = \begin{bmatrix} \mathbf{L} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{L} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \lambda_p \mathbf{L} \end{bmatrix} \leftarrow \text{Mesh Laplacians}$$

A. Hilsmann and P. Eisert: "Joint Estimation of Deformable Motion and Photometric Parameters in Single View Video", *ICCV NORDIA Workshop*, Kyoto, Japan, Sept. 2009

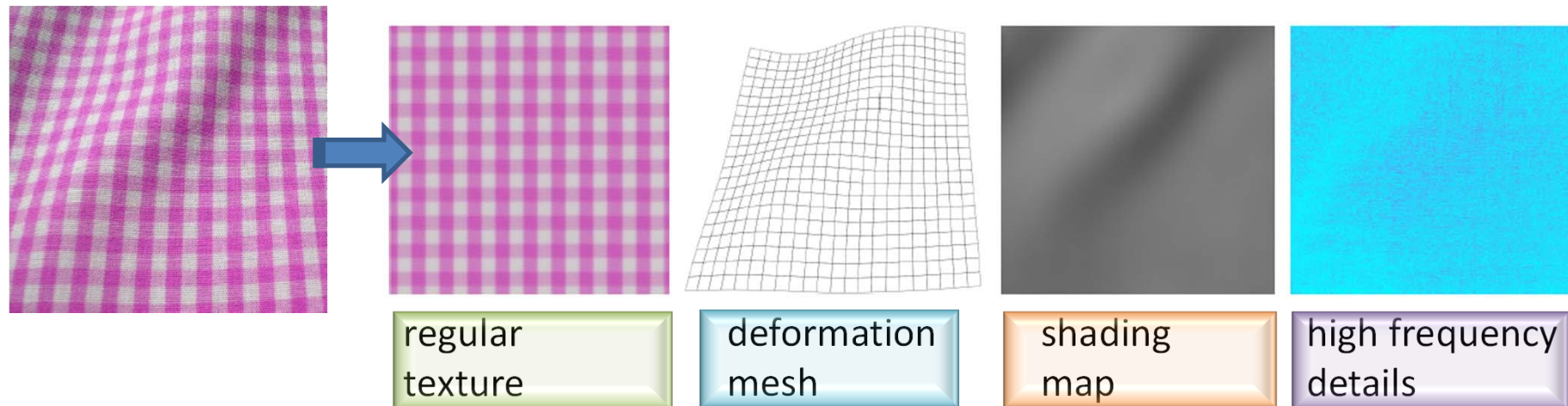
Mesh Initialization

Coarse Lattice



Fine mesh, refined with image based optimization

Texture Decomposition- Image Model



$$\mathcal{HF}(\mathbf{x}) = \mathcal{W}_p(\mathbf{x}; \hat{\theta}_p) \cdot \mathcal{T}(\mathcal{W}_g(\mathbf{x}; \hat{\theta}_g)) - \mathcal{I}(\mathbf{x})$$

Results – Retexturing



original

synthesized

Results – Retexturing



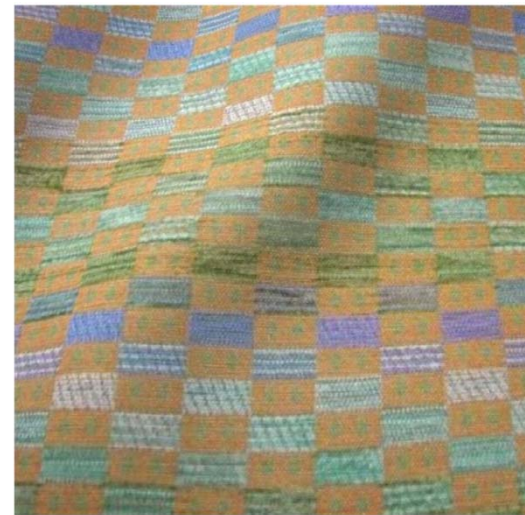
original



synthesized

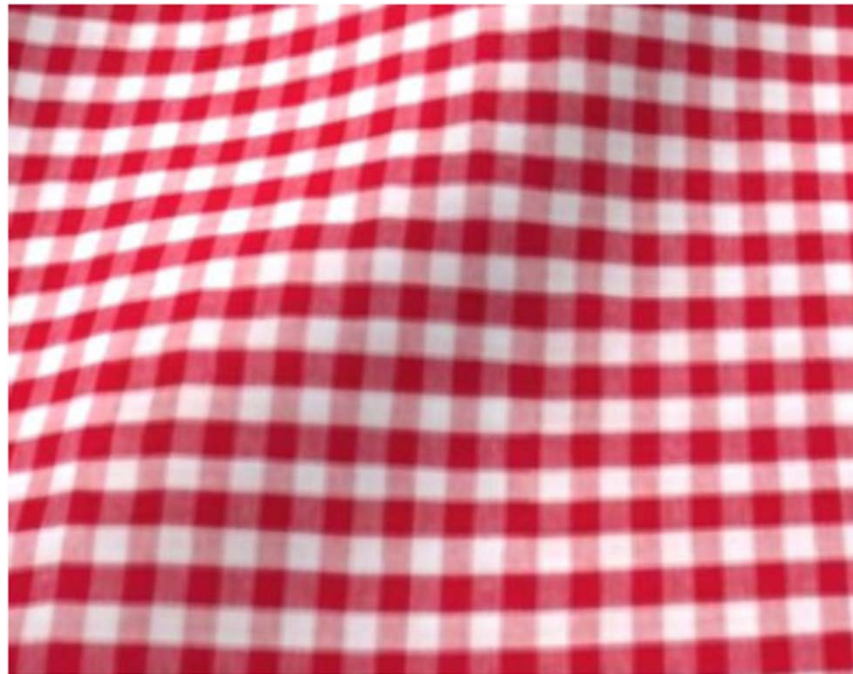


$\alpha = 0$



$\alpha = 1$

Results – Influence of Alpha

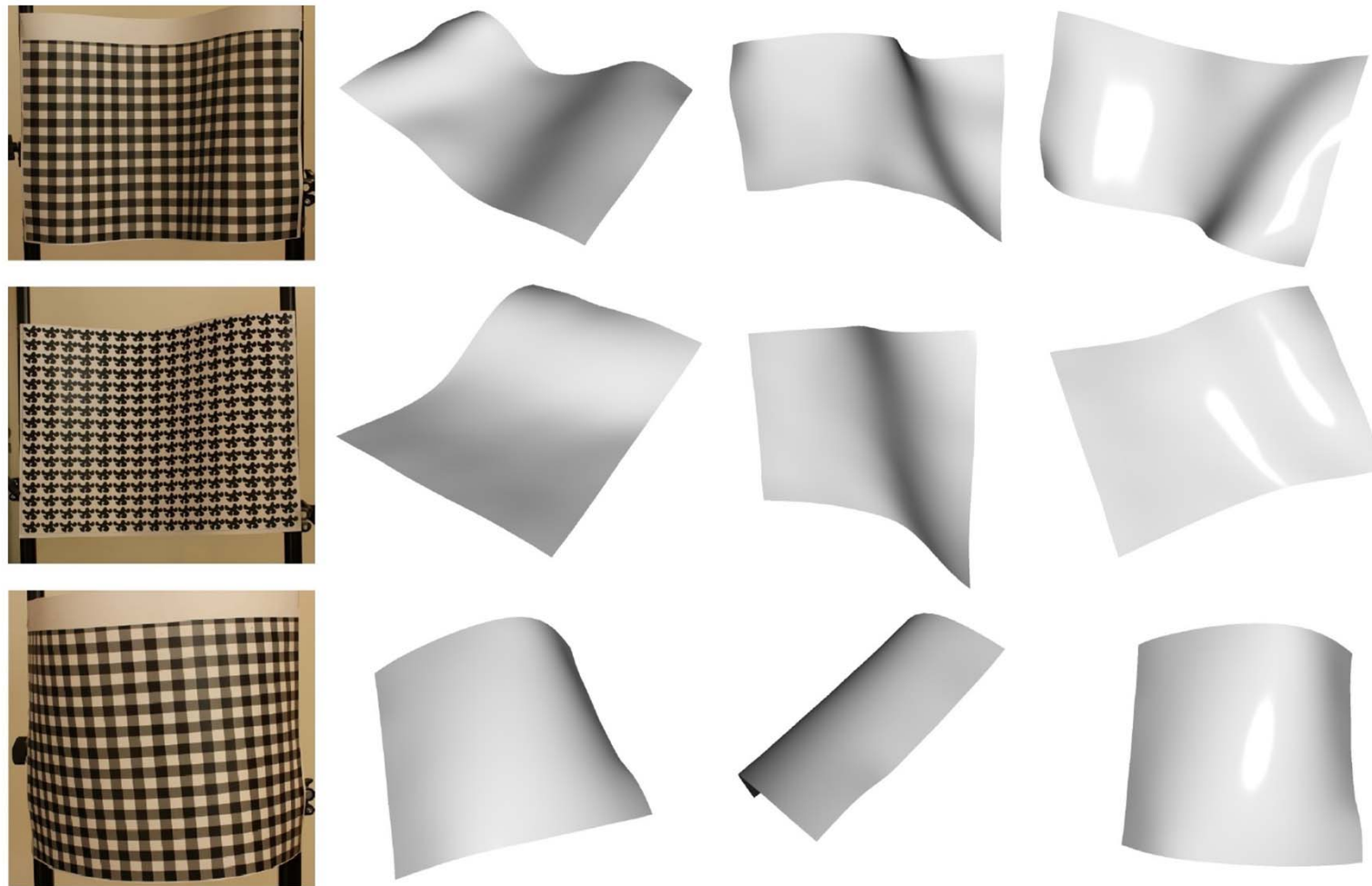


$\alpha = 0$



$\alpha = 1$

Application: Shape-from-Texture



A. Hilsmann et al: Template-free Shape-from-Texture with Perspective Cameras, Proc. BMVC 2011, Dundee, Scotland, Sep 2011

Conclusions

- Fully automatic decomposition of an image of an NRT for image-based retexturing
- Texel detection using unsupervised clustering of feature points (currently SIFT)
- Joint deformation and shading estimation as registration of a synthetic regular texture and the original image
- Self-occlusions, discontinuities?



Original image



reconstructed surface
from shape from texture



retextured results

THANK YOU