

# Camouflaging an object from many viewpoints

Andrew Owens

Connelly Barnes

Alex Flint

Hanumant Singh

Bill Freeman







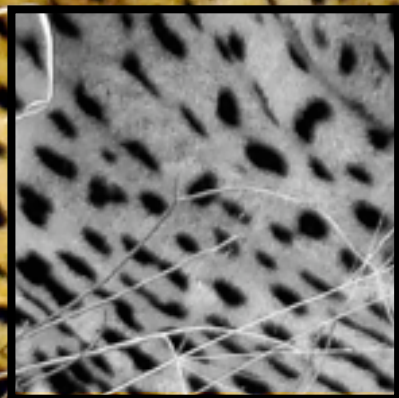


# Object detection





Object  
“non-detection”







<http://www.youtube.com/watch?v=JSq8nghQZqA>



# Camouflage problem





# Camouflage problem

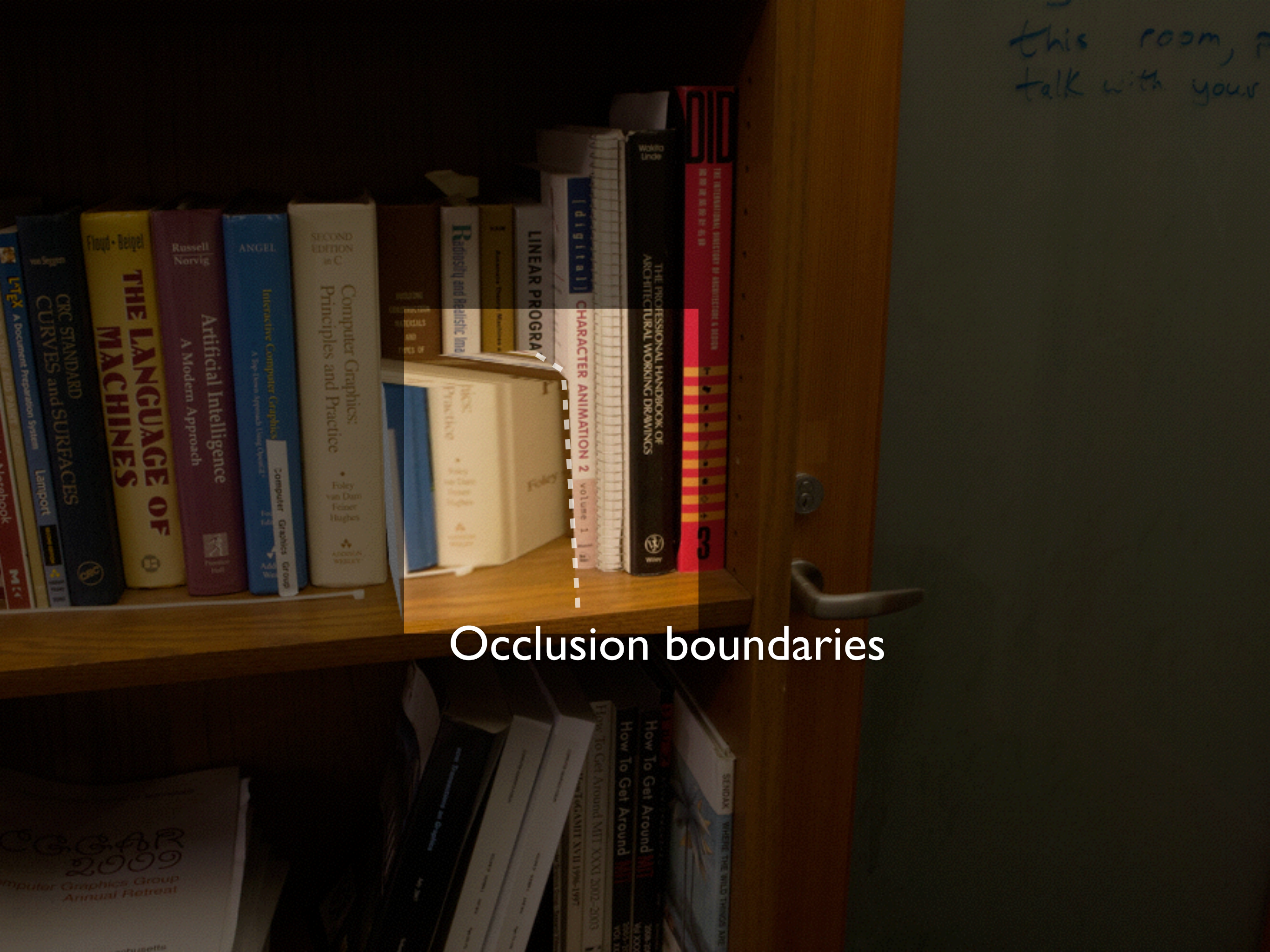


What pattern will make it hard to see?





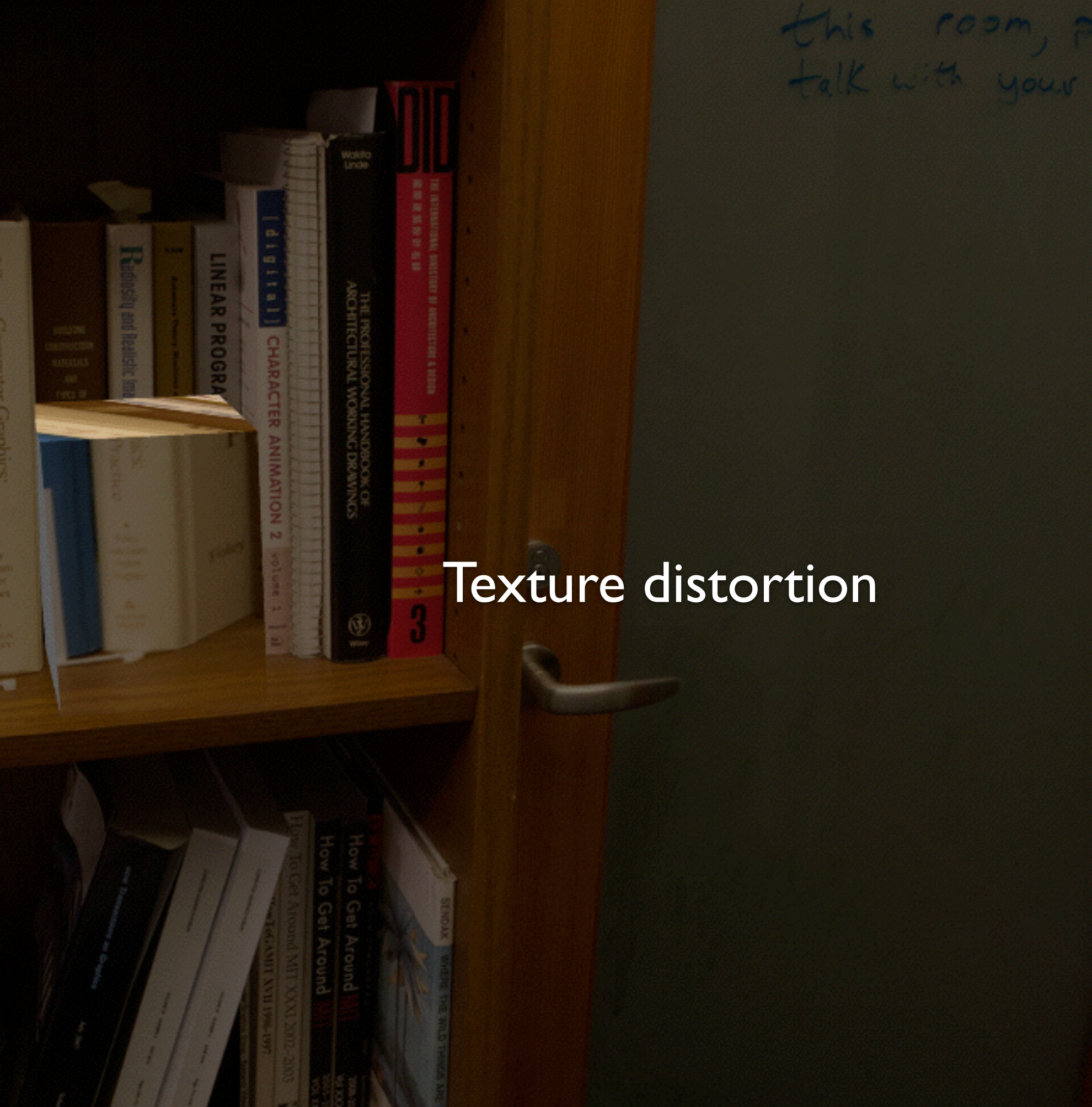




this room, p  
talk with your

Occlusion boundaries





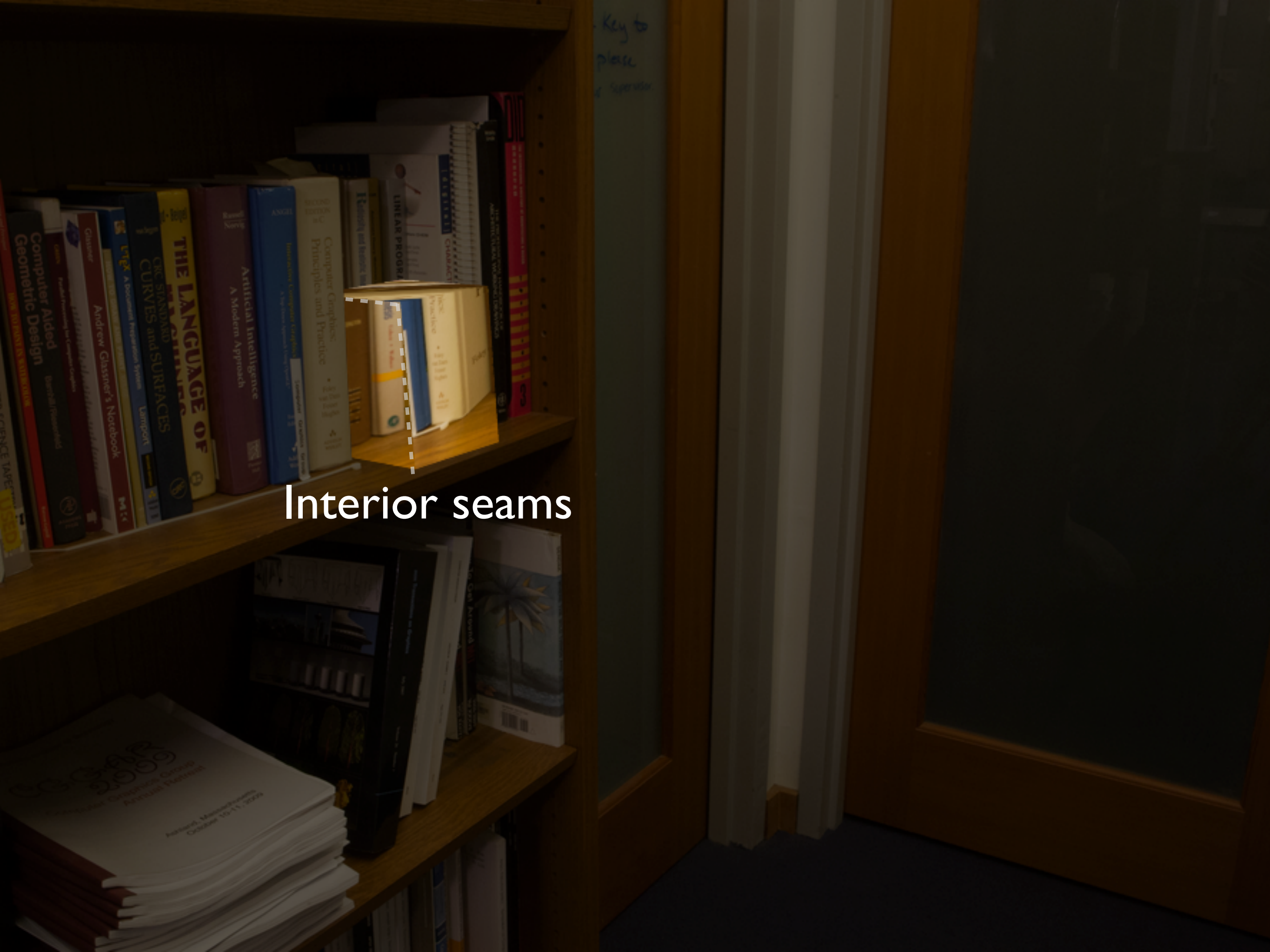
Texture distortion



this room, p  
talk with your







Interior seams



# Camouflage model



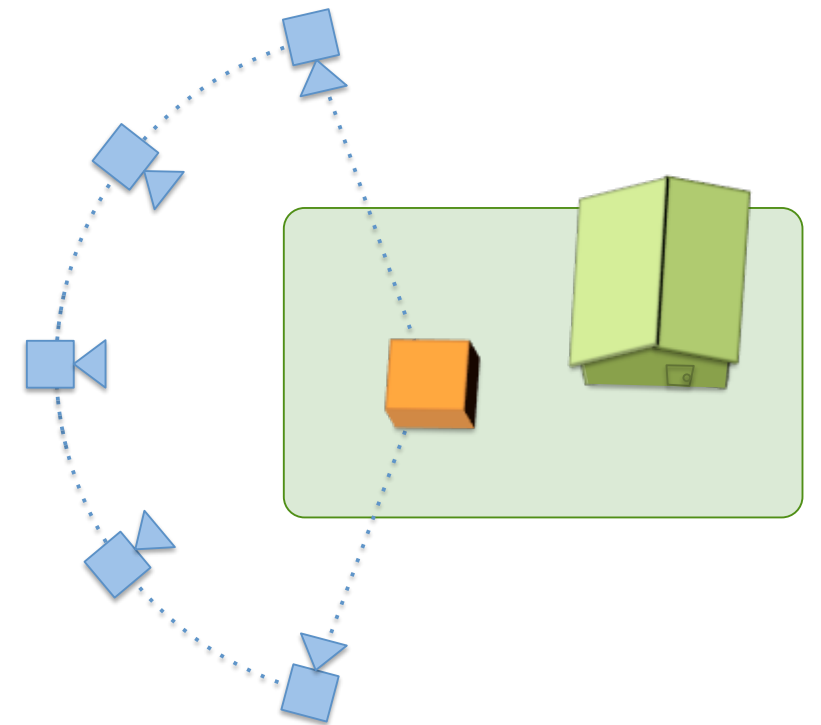
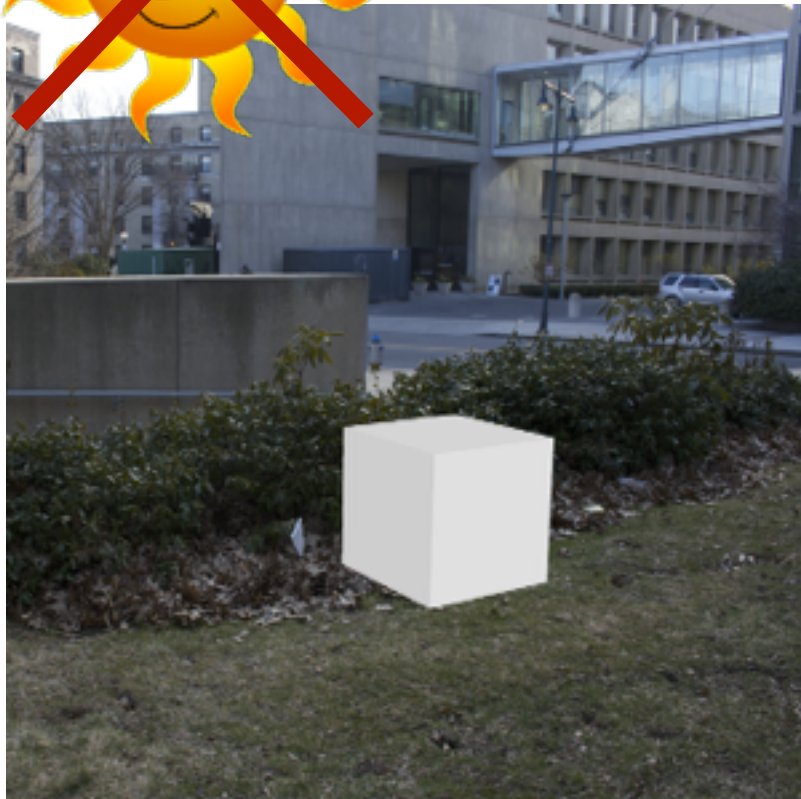
# Capturing a scene for camouflage



≈ 15 photos + camera pose



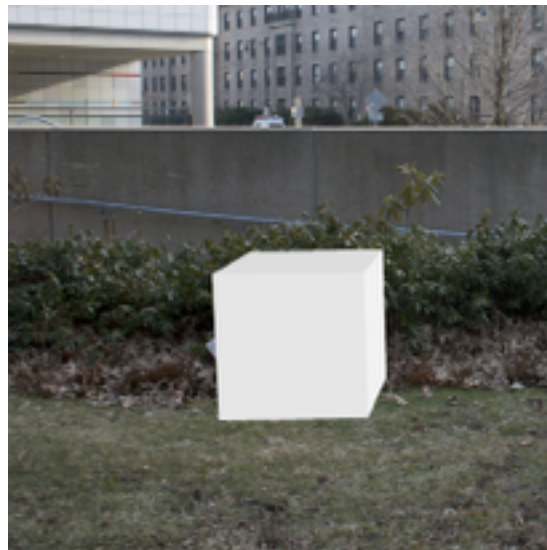
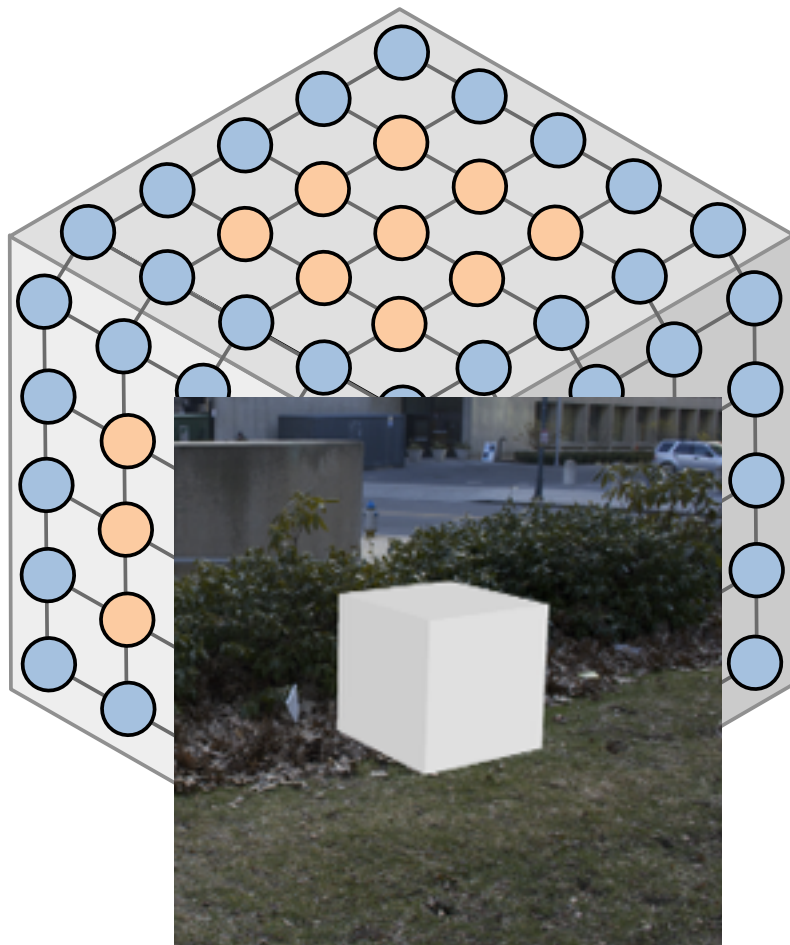
# Capturing a scene for camouflage



≈ 15 photos + camera pose

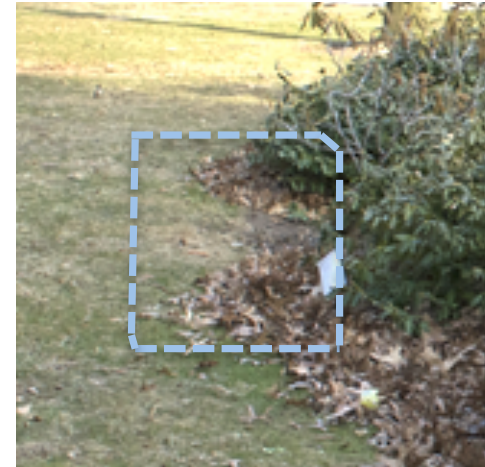
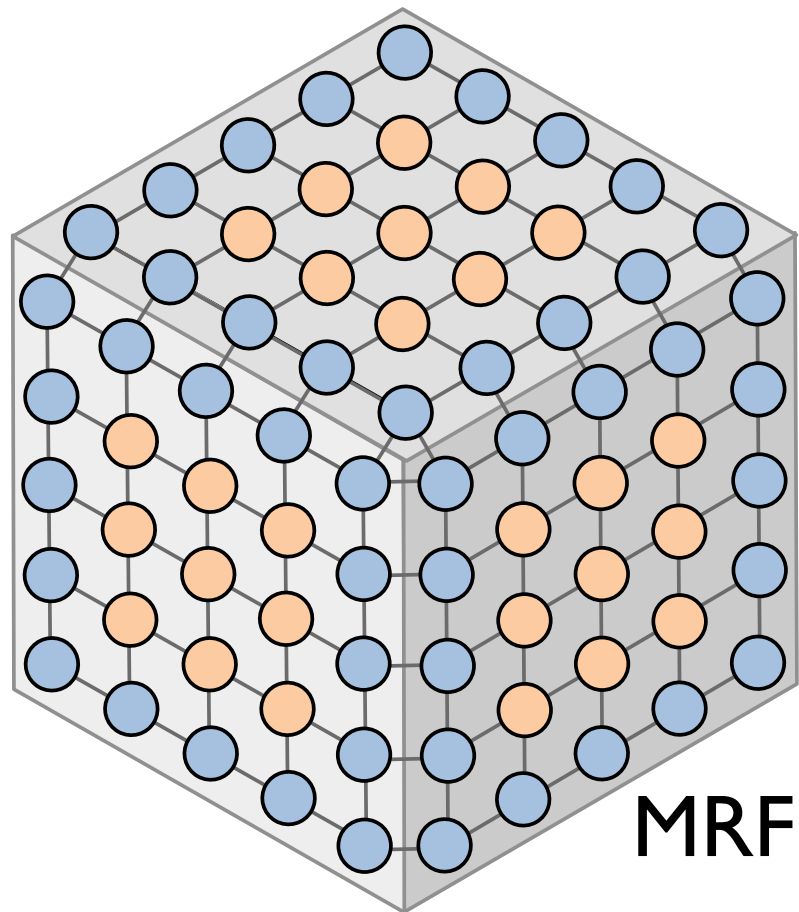


# Texture synthesis model





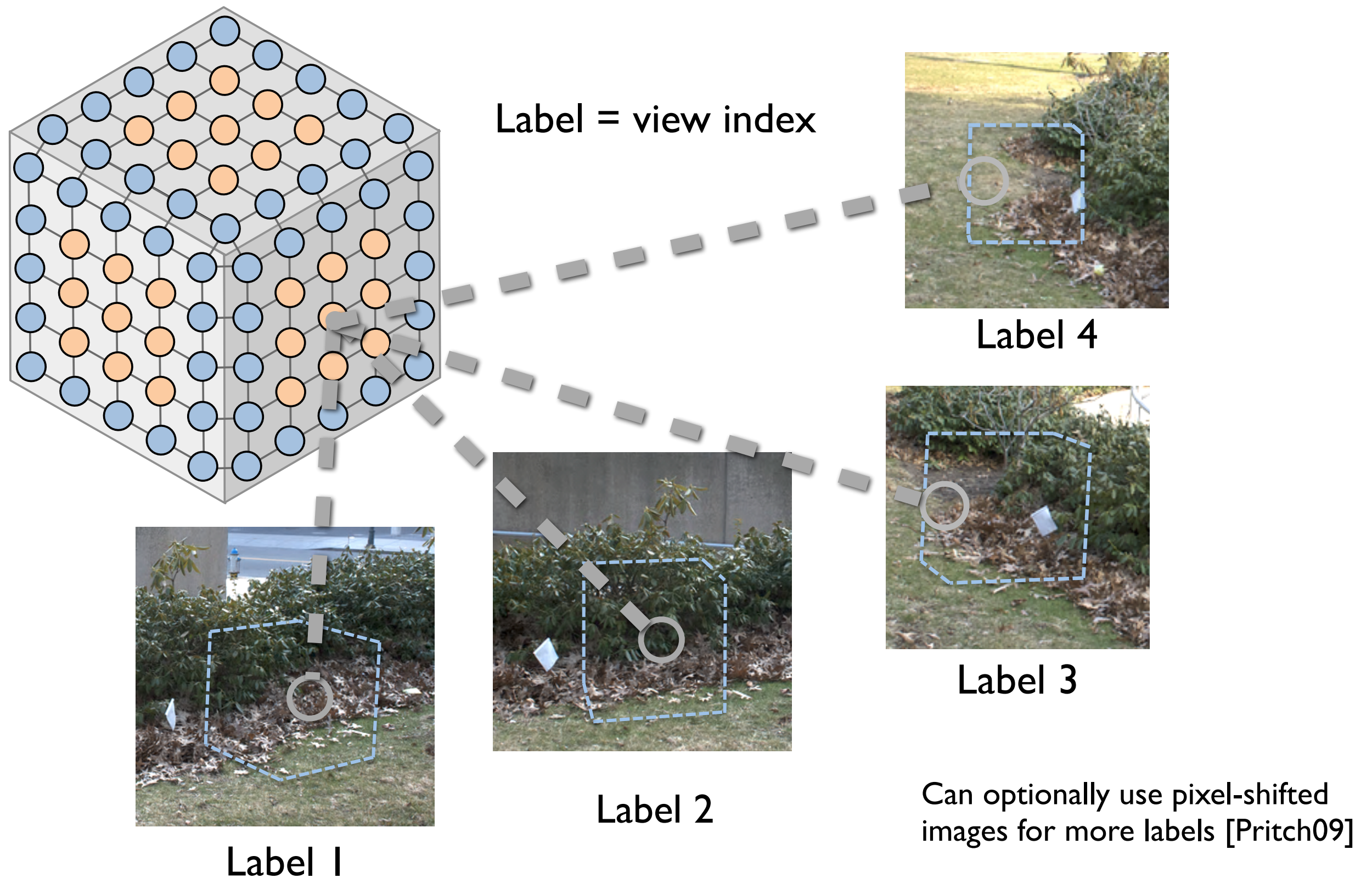
# Texture synthesis model



Shift-map image editing [Pritch09]  
Seamless montage [Gal10]  
Photomontage [Agrawala04]

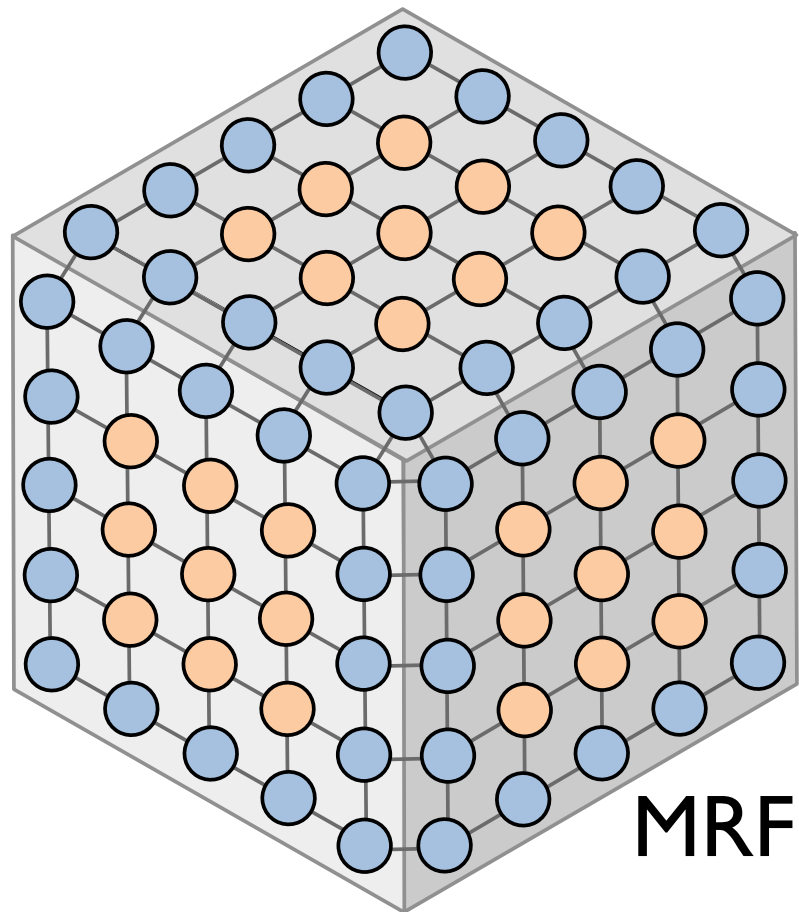


# Texture synthesis model





# Texture synthesis model



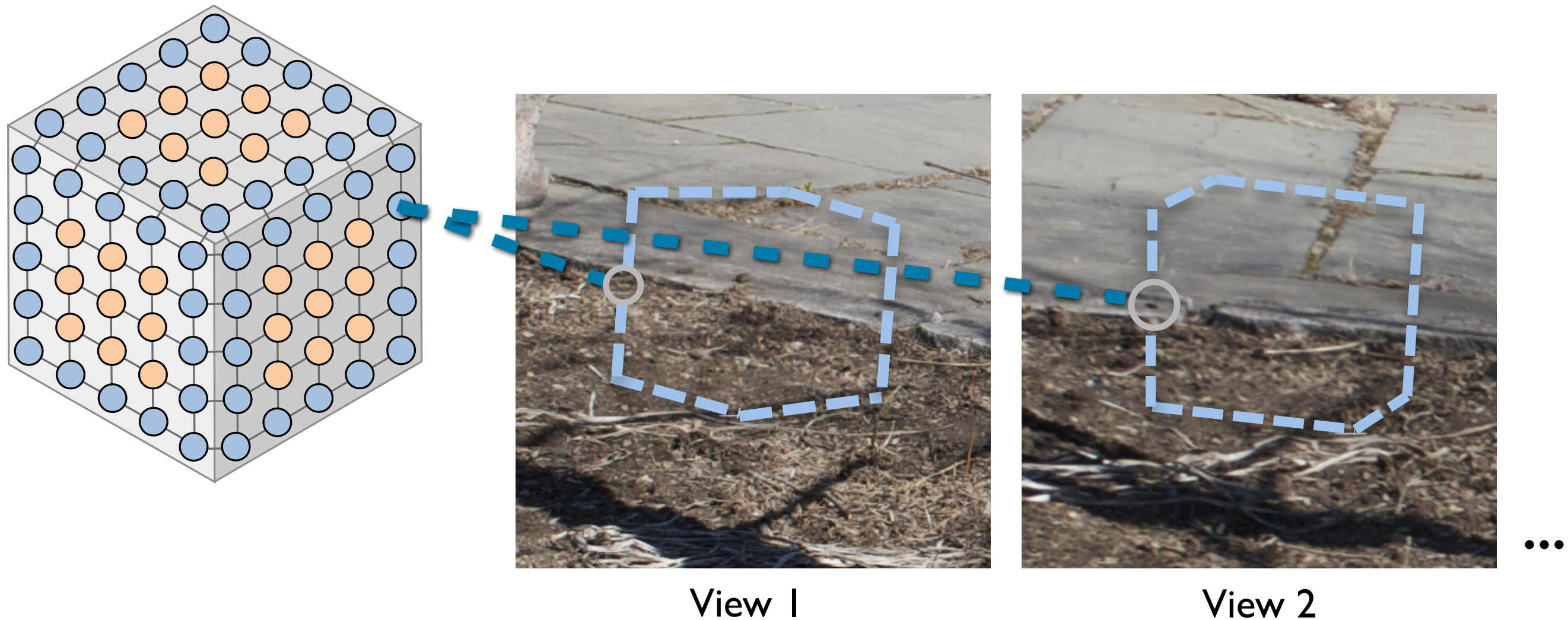
$$\sum_i E_i(x_i) + \sum_{i,j} \underbrace{E_{ij}(x_i, x_j)}_{\text{discourages interior seams}}$$

A blue arrow points from the boxed  $E_i(x_i)$  term in the equation down to the definition below.

$$E_i(x_i) = E_i^{\text{occlusion}}(x_i) + E_i^{\text{distortion}}(x_i)$$



# Occlusion cues



Measure how conspicuous the boundaries will be.

$$E^{\text{occlusion}}(x_i) = \frac{1}{n} \sum_{j=1}^n \left\| \underbrace{c_i(x_i)}_{\text{Label color}} - \underbrace{\tilde{c}_j}_{\text{Color seen in view } j} \right\|$$



# Texture distortion



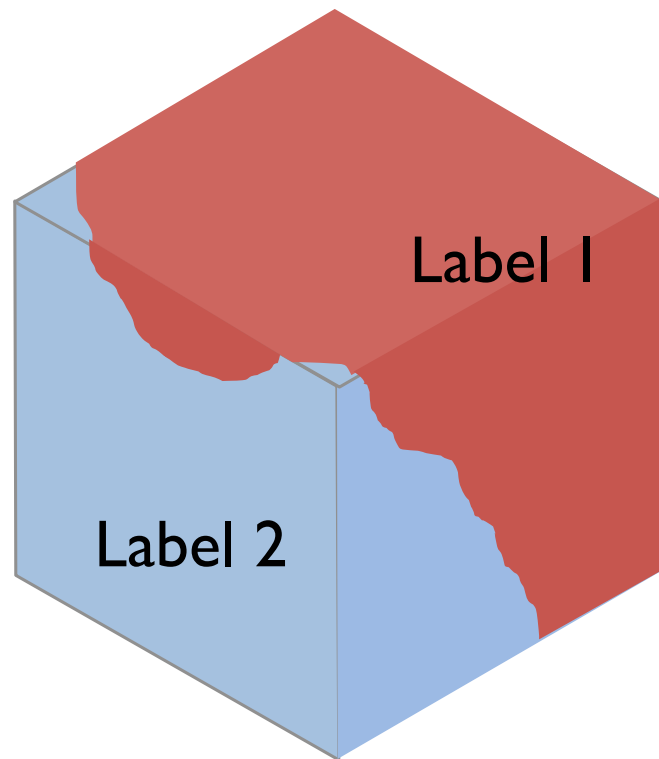
Discourage textures that will look distorted from other views.

$$E_i^{\text{distortion}}(x_i) = \frac{1}{n} \sum_{j=1}^n \underbrace{\rho(x_i, j)}_{\text{Distortion in view } j}$$

$$\rho(x_i, j) \text{ penalizes large eigenvalues of } J = \begin{pmatrix} \frac{\partial x'}{\partial x} & \frac{\partial x'}{\partial y} \\ \frac{\partial y'}{\partial x} & \frac{\partial y'}{\partial y} \end{pmatrix}$$



# Interior seams



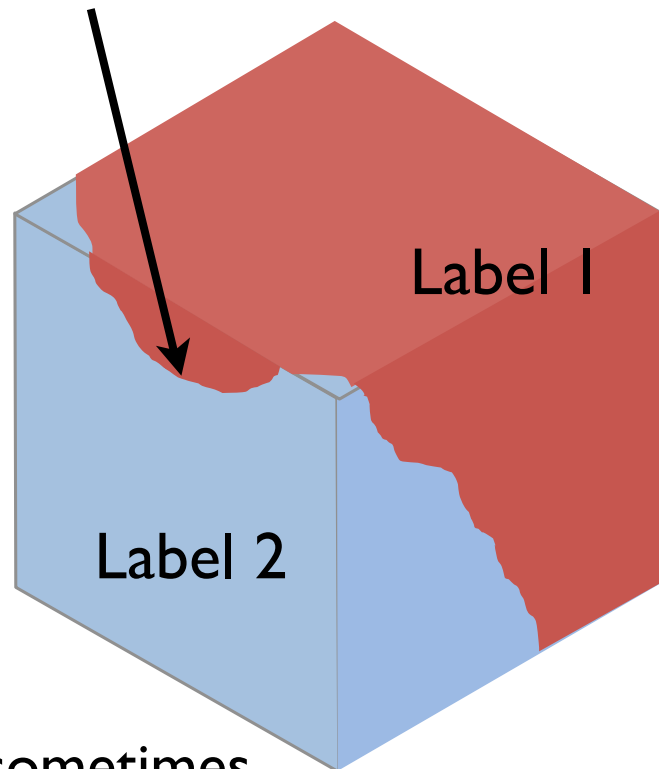
## Interior MRF

- Allow seams to occur anywhere on the object
- Avoid texture gradients
- Shift-map smoothness function [Pritch09]



# Interior seams

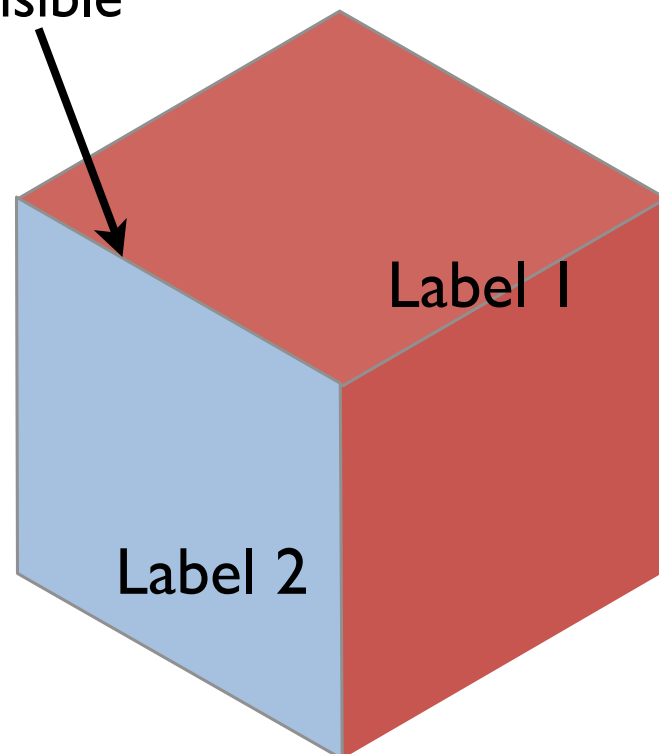
Visible in every view



## Interior MRF

- Allow seams to occur anywhere on the object
- Avoid texture gradients
- Shift-map smoothness function [Pritch09]

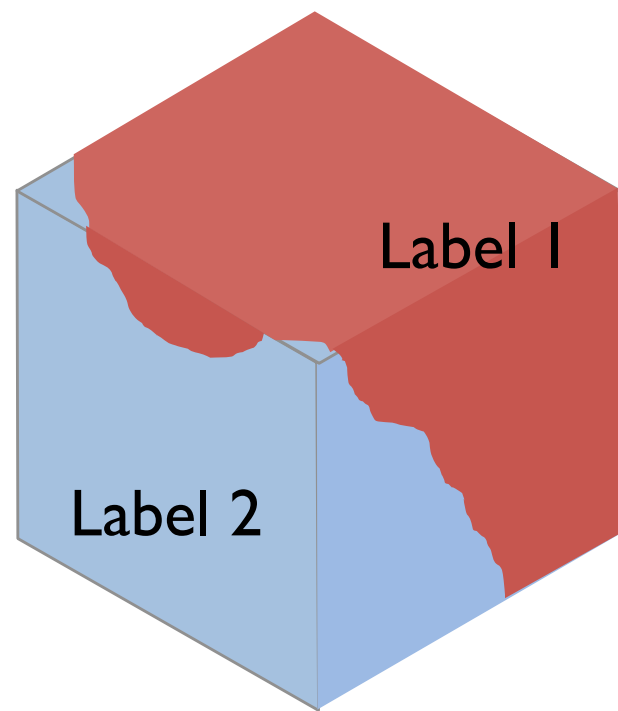
Only sometimes visible



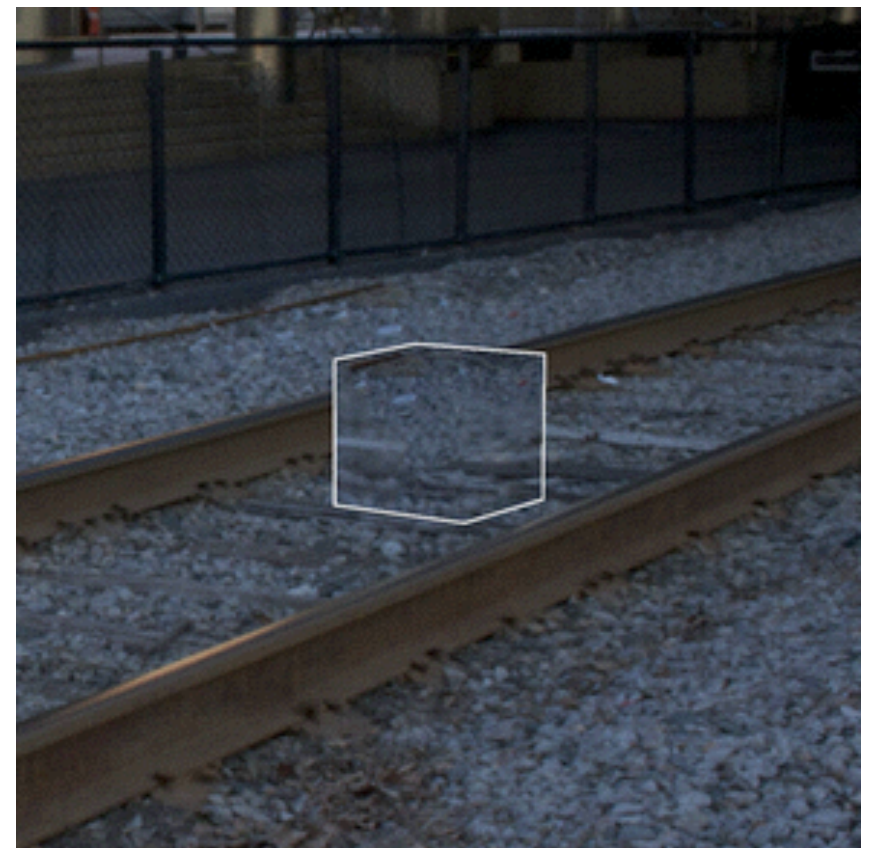
## Boundary MRF

- Only allow seams at face boundaries
- Assign same label to all texels on a face
- Typically uses two labels for whole cube

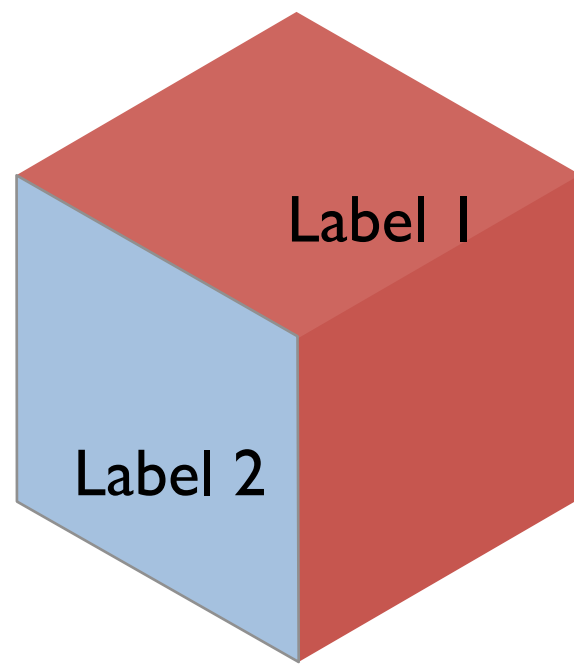




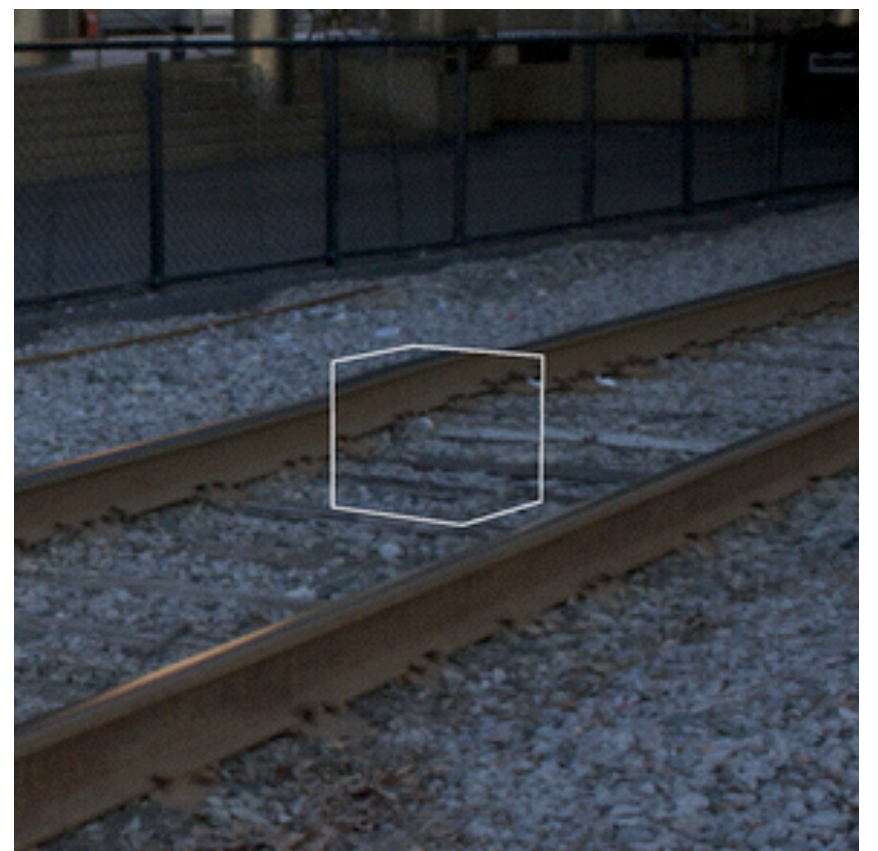
**Interior MRF**



Stitch together a texture from many images.



**Boundary MRF**



Hide perfectly from a representative viewpoint.



# Camouflage game!

More at <http://camouflage.csail.mit.edu>











A photograph of a modern building with a large red sculpture in the foreground. The building is a multi-story concrete structure with many windows. In the foreground, there is a large, abstract red sculpture made of thick, intersecting beams. A person is standing near the sculpture. The ground is covered in dry, yellowish grass. The sky is clear and blue.

No box!











# Experiments



Mean  
(baseline)



Random



- Mechanical Turk
- Measure *confusion rate* and *time-to-find*
- Show random viewpoint (varied per subject)
- Dataset containing 37 scenes



# Experiments

Model	Confusion	Time to find	$n$
Mean	9% $\pm$ 0	1.70s $\pm$ 0.04	328
Random	25% $\pm$ 0	2.59s $\pm$ 0.08	288
Interior MRF	<b>30% <math>\pm</math> 1</b>	2.90s $\pm$ 0.10	299
Boundary MRF	<b>30% <math>\pm</math> 1</b>	<b>3.25s <math>\pm</math> 0.13</b>	284

} 1.5s

- Models significantly outperform *Mean* baseline
- Similar confusion rates
- Boundary MRF takes longest to find.



# Applications



By Joshua Callaghan



Trompe-l'œil









au bon pain au bon pain

au bon pain au bon pain





Code and data will be released on our webpage: <http://camouflage.csail.mit.edu>

**Thank you!**

**Acknowledgements:** Funding by NSF CISE/IIS award I212928 and an NDSEG Fellowship. Leopard image credit: Wikipedia.

Thanks: Yichang Shih, Abe Davis, Phillip Isola, Yair Weiss, Ruth Rosenholtz's group