## The image as a virtual stage <br> 

Computational Photography
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## Today

- Inserting objects into legacy photos
- Uses single-view geometry and image-based lighting concepts
- Demo for using Blender


## Rendering Synthetic Objects into Legacy Photographs

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## The polygonal mesh

- Discrete representation of a surface
- Represented by vertices -> edges -> polygons (faces)



## Insert these...


...into this

...into this


## Inserting 3D objects into photographs

- Goal: Realistic insertion using a single LDR photo
- Arbitrary lighting environments
- Intuitive, quick and easy to create content
- Home
planning/redecoration
- Movies (visual effects)
- Video games



## Challenges

- Estimate a physical scene model including:

- Geometry

Walls/floor

- Surface properties
- Lighting info
- Camera parameters



## Earlier approaches with scene access

Manual authoring


## Earlier approaches with scene access

Manual authoring

[Fournier et al. '93]
Light probe, Inverse GI


## Earlier approaches without scene access



Outdoor illumination

[Lalonde et al. ‘09]


## System overview



Object insertion


Scene authoring


## Overview of getting geometry and lighting






## Manual input




## What the spatial layout provides



## Extruded geometry, billboards enable occlusion



Box, supporting surfaces enable object placement


Box, extruded geometry, lighting enables shadows, inter-reflections, caustics


Camera geometry ensures correct perspective



## Solving for camera viewpoint

Given 3 orthogonal VPs (at least two finite), can compute projection operator


## Solving for camera viewpoint

Given 3 orthogonal VPs (at least two finite), can compute projection operator: intrinsic matrix

$$
\begin{aligned}
& K=\left[\begin{array}{lll}
f & 0 & u_{0} \\
0 & f & v_{0} \\
0 & 0 & 1
\end{array}\right] K^{-1}=\left[\begin{array}{ccc}
1 / f & 0 & -u_{0} / f \\
0 & 1 / f & -v_{0} / f \\
0 & 0 & 1
\end{array}\right] \\
& e_{i}=(1,0,0)^{T}, e_{j}=(0,1,0)^{T}, e_{k}=(0,0,1)^{T} \\
& v_{i}=K R e_{i}, v_{j}=K R e_{j}, v_{k}=K R e_{k} \\
& (K R)^{-1} v_{i}=e^{i},(K R)^{-1} v_{j}=e^{j},(K R)^{-1} v_{k}=e^{k} \\
& e_{i}^{T} e_{j}=e_{j}^{T} e_{k}=e_{i}^{T} e_{k}=0 \\
& v_{i}^{T} K^{-T} R R^{-1} K^{-1} v_{j}=v_{j}^{T} K^{-T} R R^{-1} K^{-1} v_{k}=v_{i}^{T} K^{-T} R R^{-1} K^{-1} v_{k}=0 \\
& v_{i}^{T} K^{-T} K^{-1} v_{j}=v_{j}^{T} K^{-T} K^{-1} v_{k}=v_{i}^{T} K^{-T} K^{-1} v_{k}=0
\end{aligned}
$$

## Solving for camera viewpoint

Given 3 orthogonal VPs (at least two finite), can compute projection operator

$$
\begin{aligned}
& R=\left[\begin{array}{lll}
R_{1 c} & R_{2 c} & R_{3 c}
\end{array}\right] \\
& \lambda v_{i}=K R e_{i} \\
& R_{i c}=\lambda K^{-1} v_{i}
\end{aligned} \quad e_{i}=[1,0,0]^{T}
$$

## Projecting to image space

Given $\mathrm{K}, \mathrm{R}$, and a position in 3D, we can find its corresponding 2D image location:

$$
\lambda p_{2}=K R P_{3}
$$

## What about the reverse?

Given K, R, and a 2D position on the image, what do we know about its 3D location?

## What about the reverse?

Given K, R, and a 2D position on the image, what do we know about its 3D location?

$$
(K R)^{-1} p_{2}=\lambda P_{3}
$$

- Implies a line along which the 3D point lies
- Points on known surfaces can be localized


## Modeling occlusions



## User-defined boundary



- Tedious/inaccurate
- How can we make this better?


## Segmentation with graph cuts



$$
\text { Energy }(\mathbf{y} ; \theta, \text { data })=\sum_{i} \psi_{1}\left(y_{i} ; \theta, \text { data }\right) \sum_{i, j \in e d g e s} \psi_{2}\left(y_{i}, y_{j} ; \theta, \text { data }\right)
$$

## Segmentation with graph cuts



$$
\text { Energy }(\mathbf{y} ; \theta, \text { data })=\sum_{i} \psi_{1}\left(y_{i} ; \theta, \text { data }\right) \sum_{i, j \in e d g e s} \psi_{2}\left(y_{i}, y_{j} ; \theta, \text { data }\right)
$$

## Refined segmentation



## Spectral Matting



## Spectral Matting

- Create NxN matrix describing neighboring pixel similarity (Laplacian matrix, L)
- Extract "smallest" eigenvectors of L
- Soft segmentation defined by linear combination of eigenvectors
- Scribbles provide constraints to assign to foreground


## Spectral Matting



## Spectral matting



## Spectral matting



## Segmentations as "billboards"



## Segmentations as "billboards"



## Rendering via ray tracing



Insertion without relighting

...with relighting


## Estimating light

- Hypothesize physical light sources in the scene
- Physical $\rightarrow$ CG representations of light sources found in the real world (area lights, etc)
- Visible sources in image marked by user
- Refined to best match geometry and materials
- User annotates light shafts; direction vector
- Shafts automatically matted and refined


## Lighting estimation



## Lighting estimation



## Lighting estimation



## Light refinement

## Match original image to rendered image



## Initial light parameters



## Refined light parameters



## External light shafts



## External light shafts



## External light shafts

Shadow matting via Guo et al. [2011]


## Setting light shaft direction



## External light shafts



## Light shaft result



## Inserting objects

- Representation of geometry, materials and lights is now compatible with 3D modeling software
- Two methods of insertion/interaction
- Novice: image space editing
- Professional: 3D modeling tools (e.g. Maya)
- Scene rendered with physically based renderer (e.g. LuxRender, Blender's Cycles)


## Blender demo

## Final composite

Additive differential technique [Debevec 1998]


E (empty)


## Putting it all together

Video: https://vimeo.com/28962540

## Research directions

- Can we do better with
- Multiple images?
- Videos?
- Depth?
- Better scene understanding?
- How to insert image fragments?


## Fully Automated Scene Modeling

Karsch et al. 2014: http://vimeo.com/101866891

## Summary

- We can accurately predict how a 3D object would look in a depicted scene by recovering
- Viewpoint: camera matrix, single view geometry
- Scene geometry: single-view geometry
- Material: "intrinsic image approaches"
- Lighting: solve for lights such that rendering reproduces image
- Next classes: interest points, matching and alignment, and stitching

