

Simple 3D Reconstruction of Single Indoor Image with Perspective Cues

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- 1 Introduction
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Problem Statement

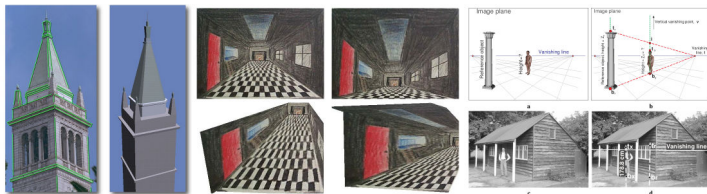
Some Image-Based Modeling Methods:

- Reconstruct from different views (e.g. space carving)
- Reconstruct from single view with different lighting (e.g. photometric methods)
- Reconstruct using visual cues (e.g. stereo vision and motion)

3D Reconstruction from a single image is a difficult problem since we have very limited information.

Previous Work: Methods With Assistance from Users

- Facade by Debevec
- Tour Into the Picture by Youichi et al.
- Single View Metrology by Criminisi



Previous Work: Automatic methods

- Automatic TIP by Boulanger
- Photo Pop-up by Hoeim
- Vanishing point detection using Bayesian network
- Markov random field by Delage et al.

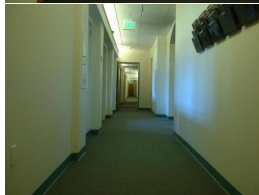
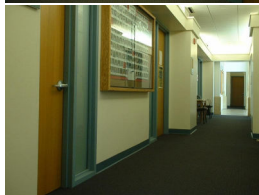


Our Goal

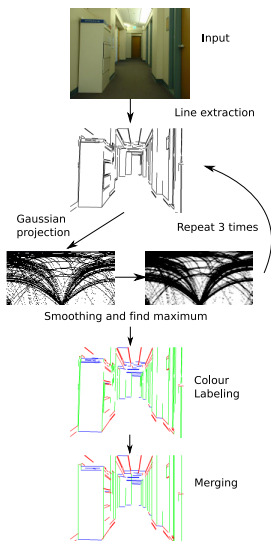
- Focus on specific class of images
- Reconstruct basic geometry
- Emphasis on speed and simplicity

Assumptions

- One-point perspective indoor images consist of orthogonal planes.
- Presence of floor, ceiling, and walls.
- Well-composed images.



Vanishing Point Detection

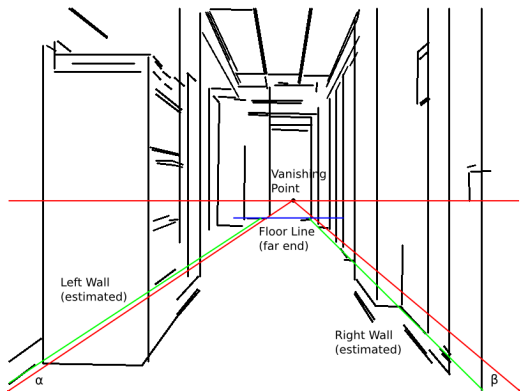


Workflow

- 1 Canny edge detector
- 2 Gaussian hemisphere projection
- 3 Labeling based on vanishing points
- 4 Merging excessive lines

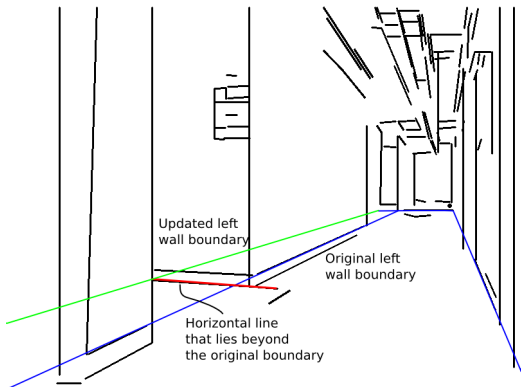
Floor Detection

- Vanishing point marks horizon.
- Horizon and lower corners of the image define a search range for floor boundaries.



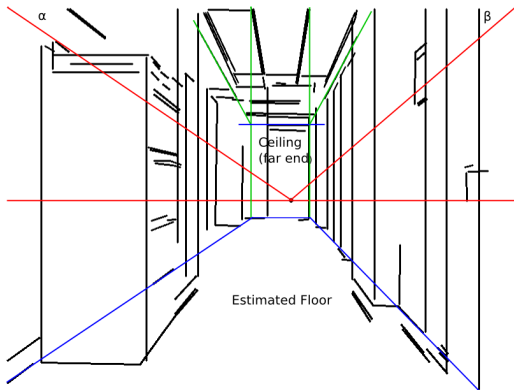
Floor Detection

- Update boundaries after the initial estimation.



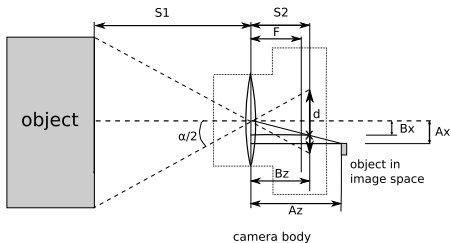
Ceiling Detection

- Horizon and upper corners define a search range for ceiling boundaries.
- User floor boundaries to locate end points of ceiling boundaries.



Focal Length and Room Size

- Approximate the focal length using the optics of photography
- $F \approx$ diagonal size of the photograph



$$\frac{1}{F} = \frac{1}{S_1} + \frac{1}{S_2}$$

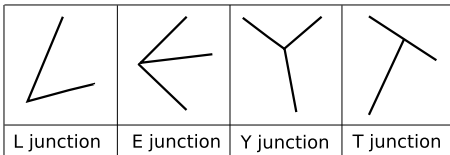
$$S_1 \gg S_2$$

$$\frac{B_x}{A_x} = \frac{B_z}{A_z}$$

Multi-pass Junction Point Detection

Junction Points

- A junction point is a good indicator of the orientation of flat surfaces.
- Too many junction points and missing ones in the image.

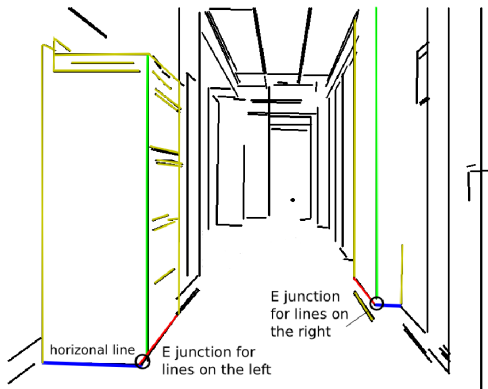


Divide and Conquer

- Process easy cases first
- Remove processed lines
- Infer missing lines

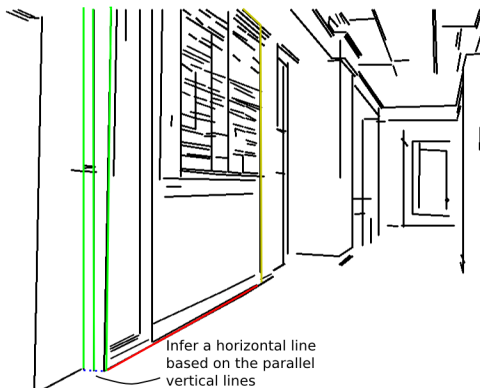
First Pass

- Scan horizontal lines on the floor.
- Focus on E junctions.
- Trace L junctions that can become E junctions.



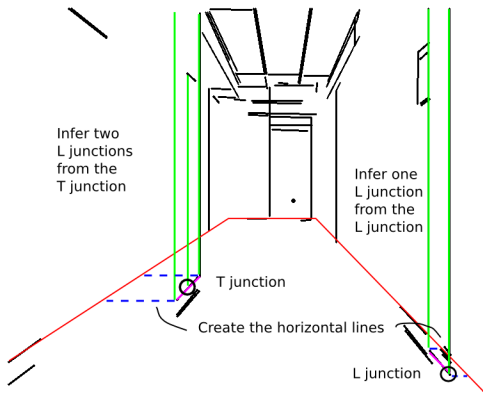
Second Pass

- Sometimes the edge detector would miss horizontal lines.
- We can only infer certain ones.
- Repeat the first pass after adding the inferred lines.



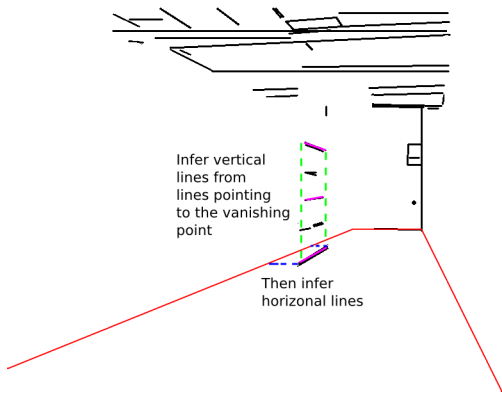
Third Pass

- No horizontal line on the floor in this pass.
- Scan vertical lines on the floor to find L junctions.
- T junctions can also be formed because we removed lines in previous passes.



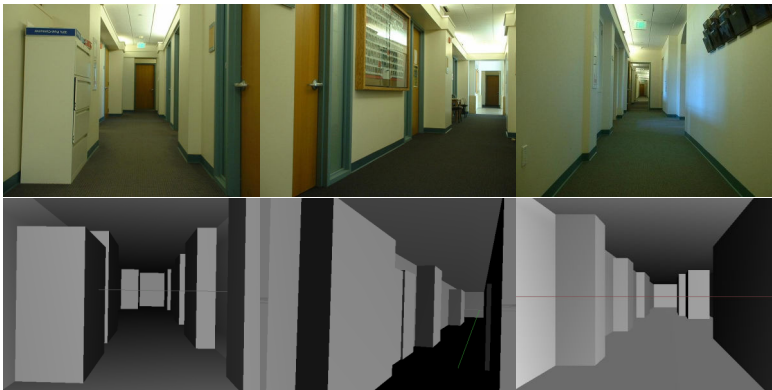
Fourth Pass

- On the floor only lines pointing to the vanishing point are left.
- We may miss some vertical lines because of edge detector and line removal.
- We can infer some vertical lines.



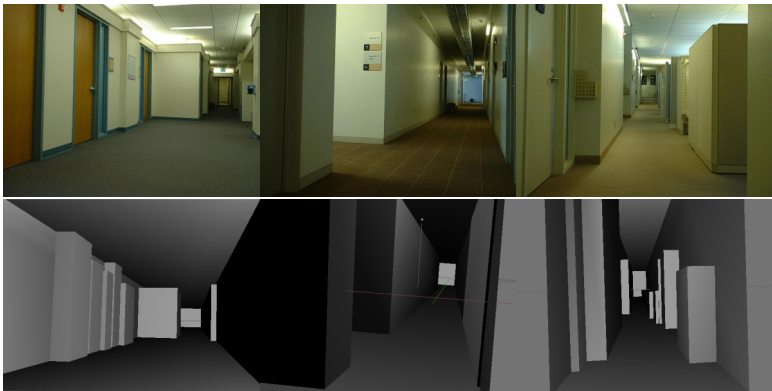
Results and Conclusions

- Smooth and complete structure given the presence of floor, ceiling, and walls.
- Simple to implement and fast to run (2-3 milliseconds per line segment).



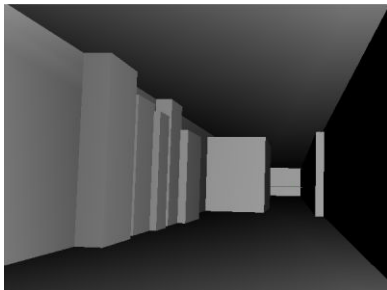
Results and Conclusions

- The result however is not accurate due to the amount of estimation involved.

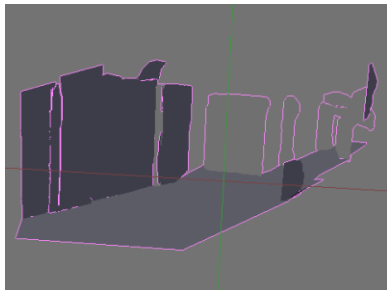


Comparisons

- The reconstructed models are continuous because we make strong assumptions about the input.



(a)

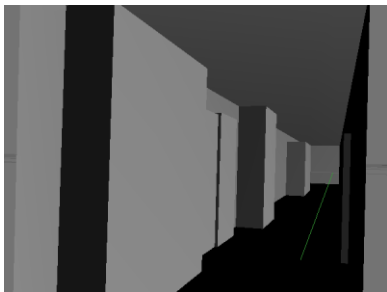


(b)

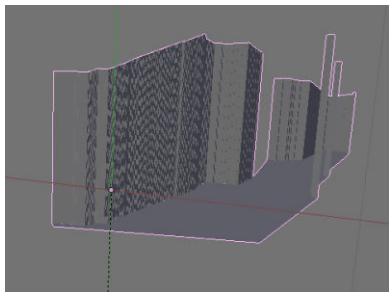
Figure: Compared with Markov Random Field Method

Comparisons

- Our models don't require complex mesh representation, and have no missing ceiling or wall.



(a)



(b)

Figure: Compared with Dynamic Bayesian Network Method

Thank you!
Questions?

