Reconstructing the World's Museums

Jianxiong Xiao
MIT

Yasutaka Furukawa
Google

Ciao, Firenze!

Hello, Florence!

Maps

Photorealistic Maps

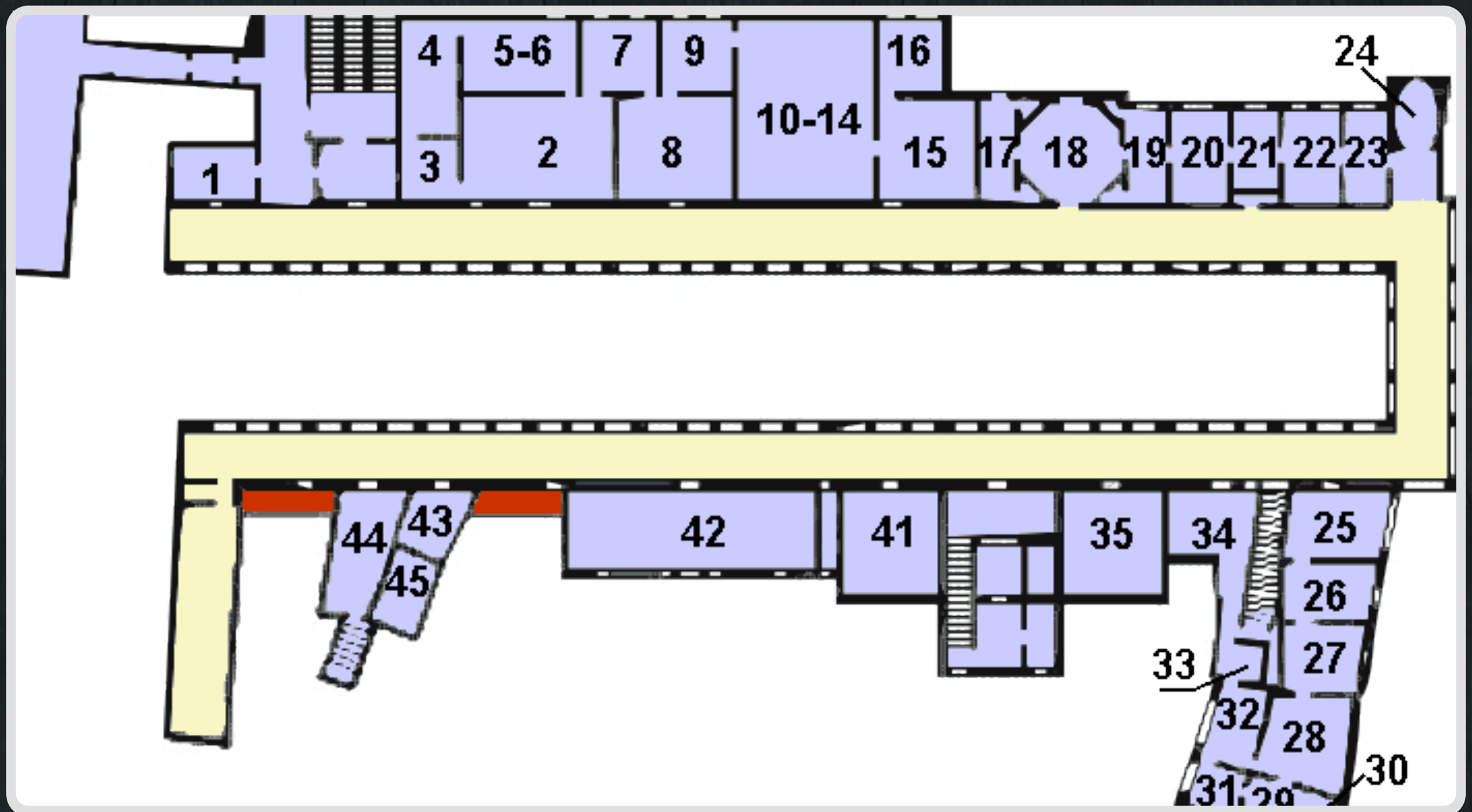
That is great!

What about indoors?

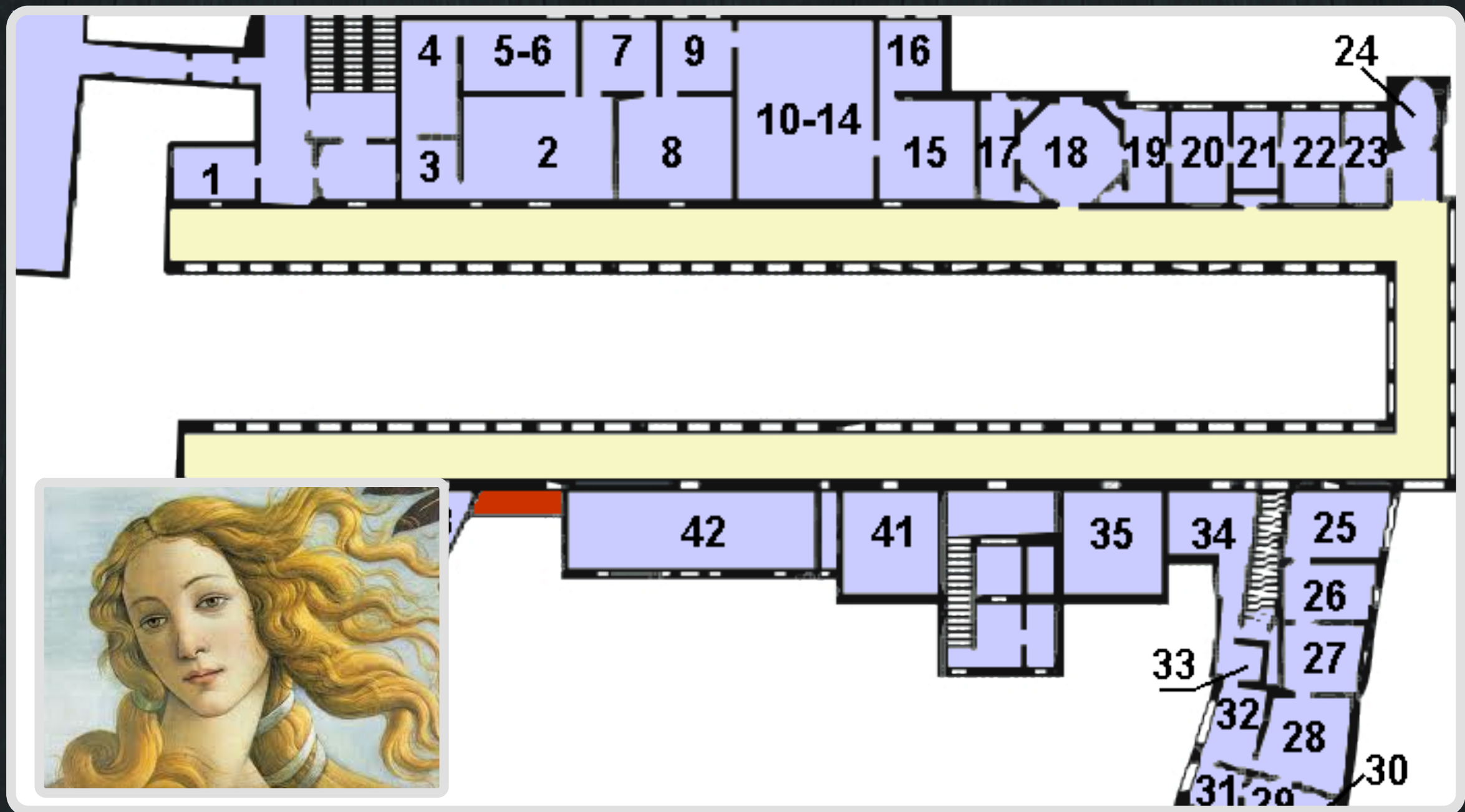
Florence = Renaissance Center of Art

Uffizi museum = one of the most famous art galleries

Uffizi Museum

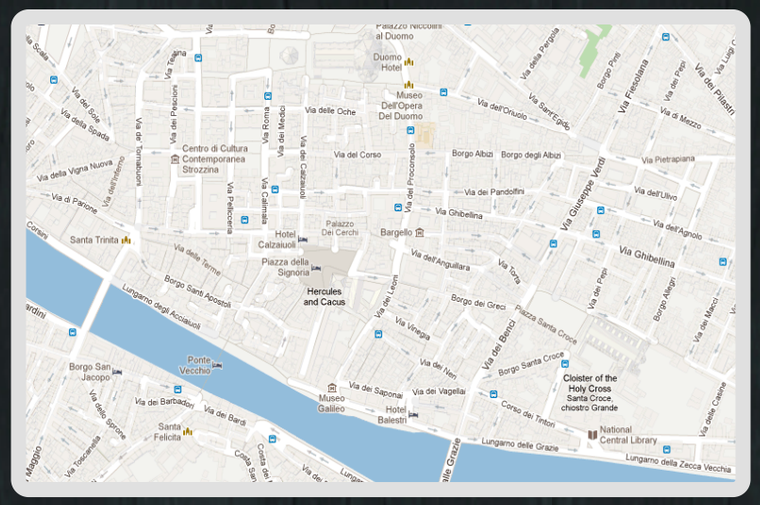


Uffizi Museum



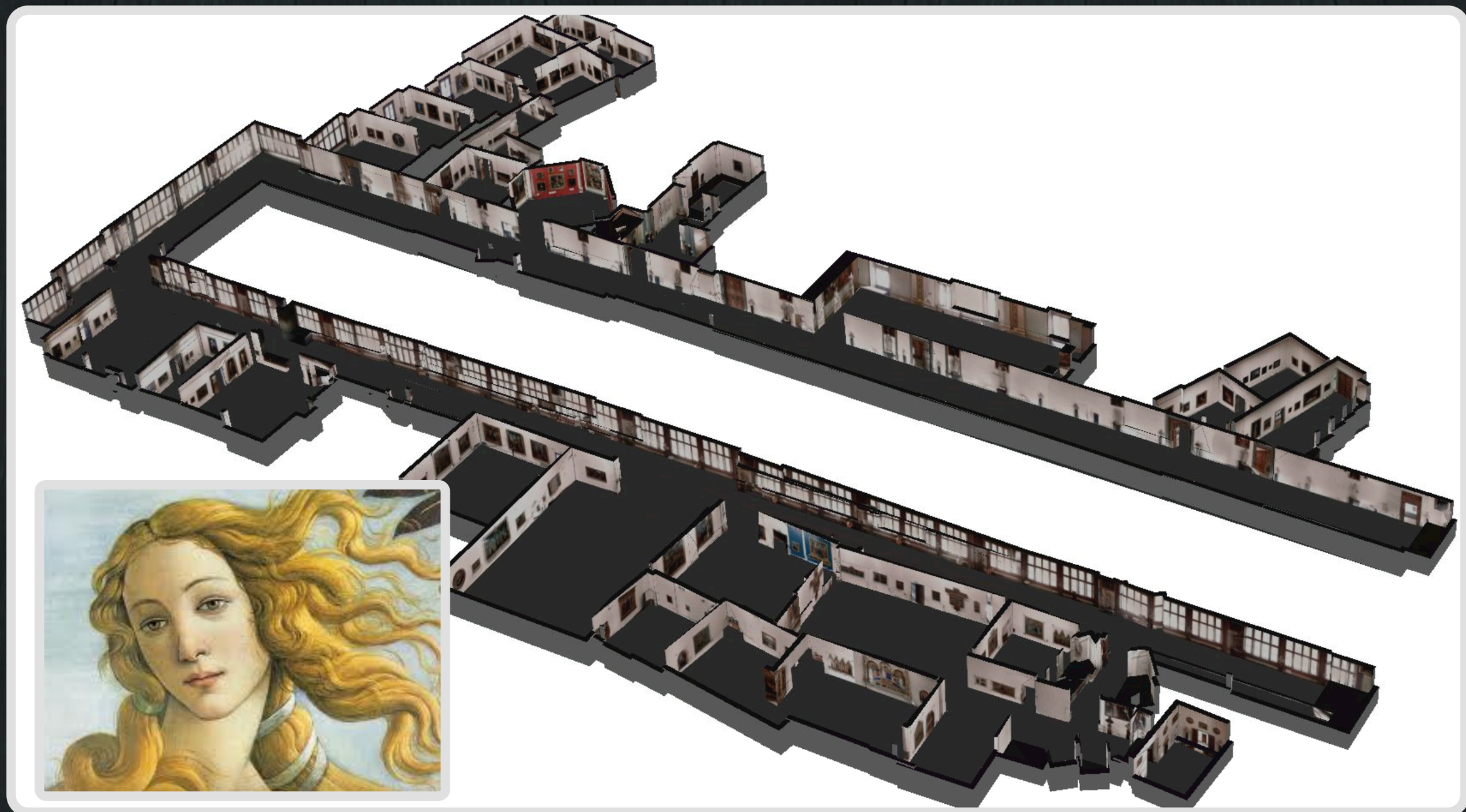
Where is “The Birth of Venus”?

Cartography

	2D Line Drawing	3D Realistic Maps
Outdoor		
Indoor		?

Photorealistic Indoor Maps

Photorealistic Indoor Maps



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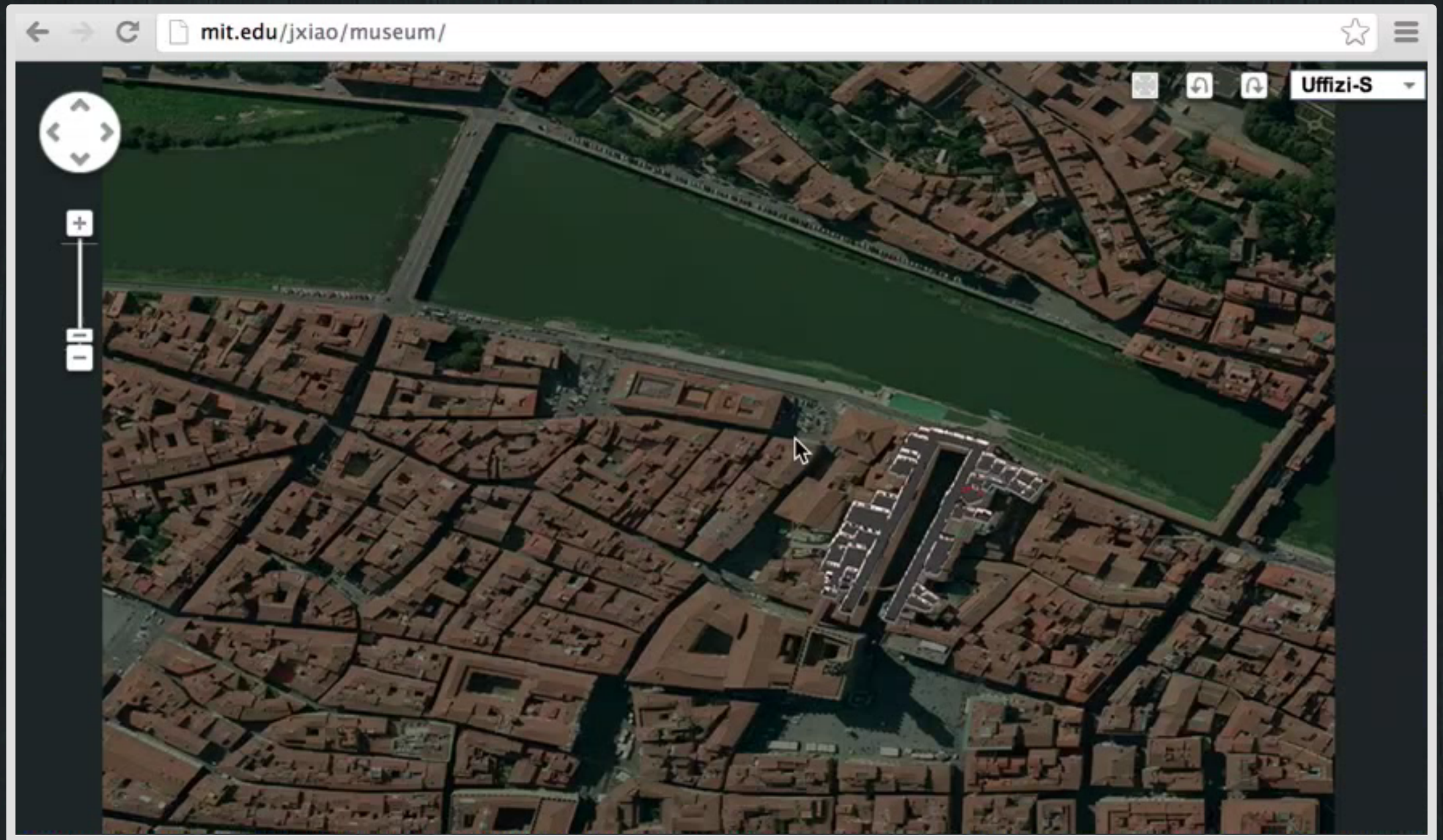
Photorealistic Indoor Maps



Data-driven Brute-force Approach

Data-driven Brute-force Approach

SUN Database. Xiao et al. 2010.

Data-driven Brute-force Approach

2-step algorithm

Data-driven Brute-force Approach

1. Remove ceiling.
2. Take pictures from aerial viewpoints.

What is wrong?

What is wrong?

I don’t own the museums.

Old Approach

1. Remove ceiling.
2. Take pictures from aerial viewpoints.

New Approach

0. Own the museums.
1. Remove ceiling.
2. Take pictures from aerial viewpoints.

New Approach

0. Reconstruct the museums.
1. Remove ceiling.
2. Take pictures from aerial viewpoints.

Our Goal

- Global texture-mapped 3D model
- Optimize for aerial viewing
- Enable effective indoor navigation

Size matters!

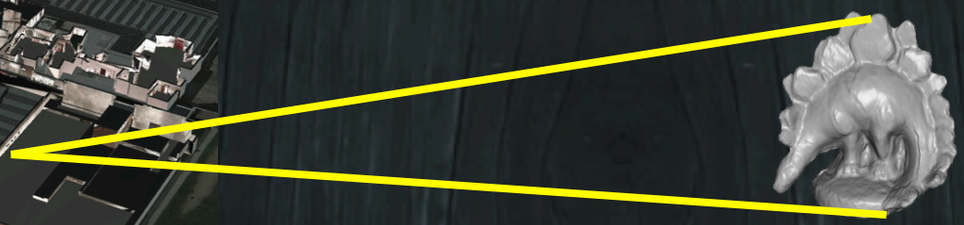
Existing methods

[Xue 11], [Hongxing 10], [Kolev 10],
[Li 10], [Song 10], [Xi 10], [Xiao 09], [Furukawa 09],
[Jancosek 09], [Vu 09], [Xiao 08], [Bradley 08],
[Furukawa 08], [Zach 08], [Xiao 07] , [Gargallo 07],
[Goeselle 07], [Habbecke 07], [Merrell 07],
[Sinha 07], [Sormann 07], [Stark 07], [Vogiatzis 07],
[Zach 07], [Zaharescu 07], [Furukawa 06],
[Goeselle 06], [Hornung 06], [Tran 06],
[Strecha 06], [Pons 05], [Vogiatzis 05],
[Hernandez 04], [Kolmogorov 02], ...

- Require accurate calibration
- Produce a “dense” model
  - Susceptible to errors
  - Not scalable
- Mostly small scale

22,610 triangles

> 1,000,000 triangles



System Pipeline

1. Take pictures inside the rooms
2. Reconstruct the 3D shape
3. Render from aerial viewpoints

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Art Project

www.GoogleArtProject.com

System Pipeline

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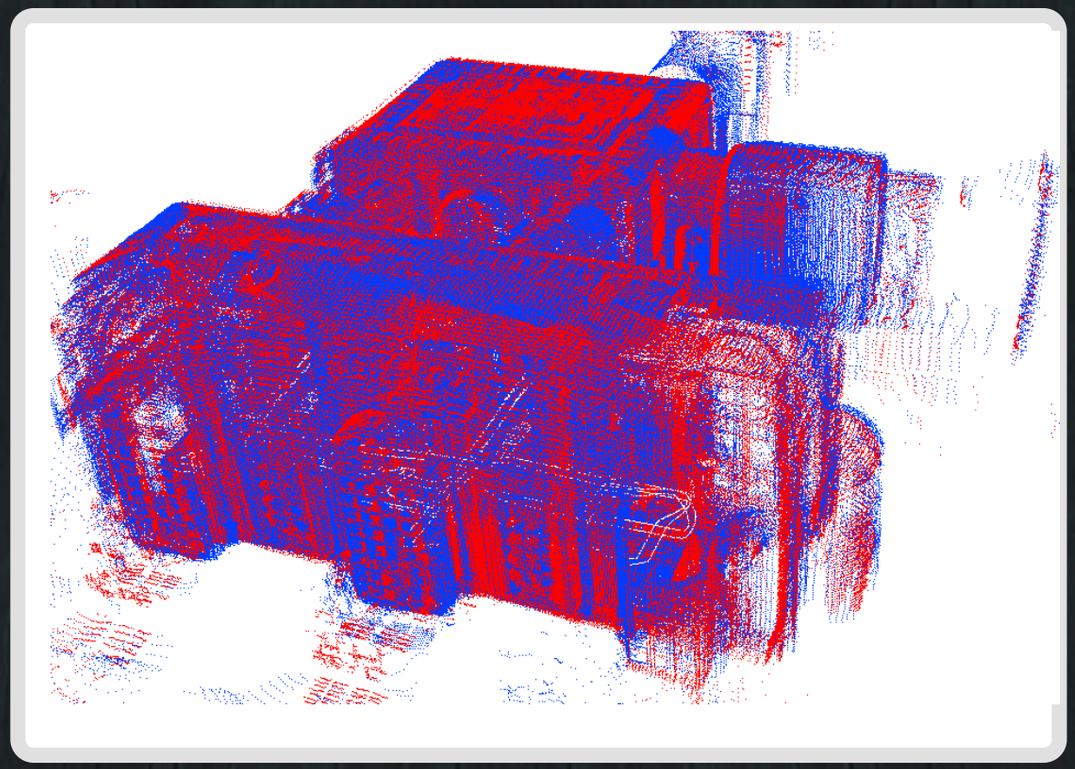

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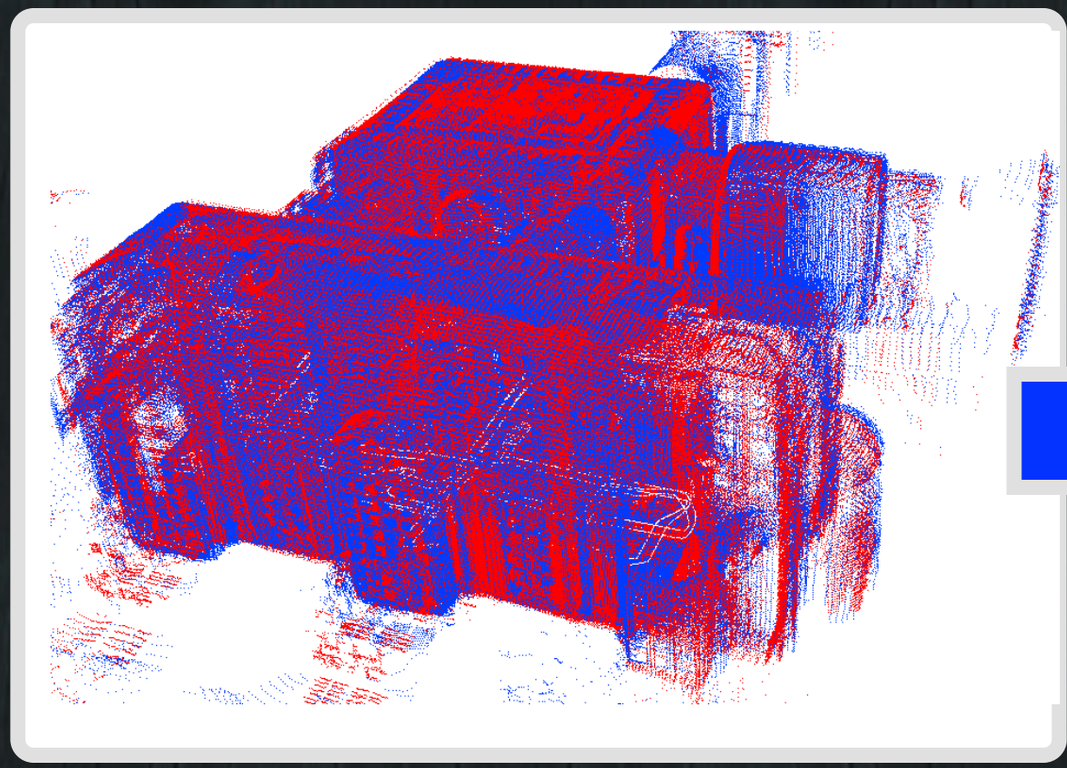

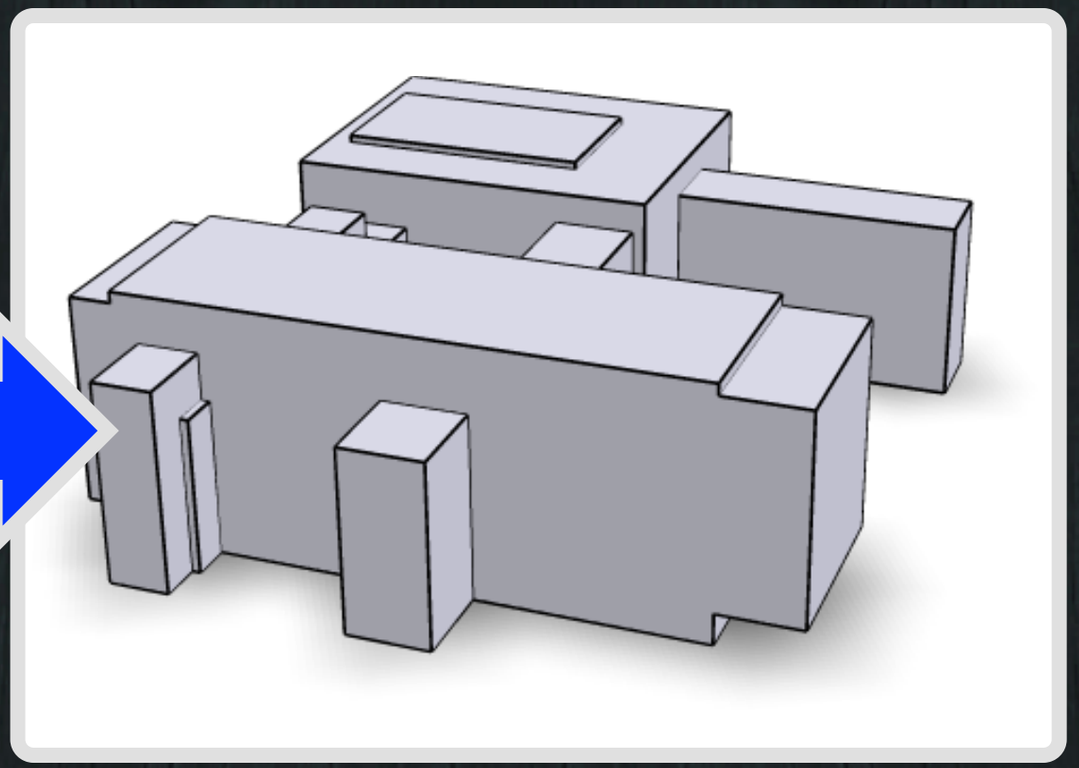

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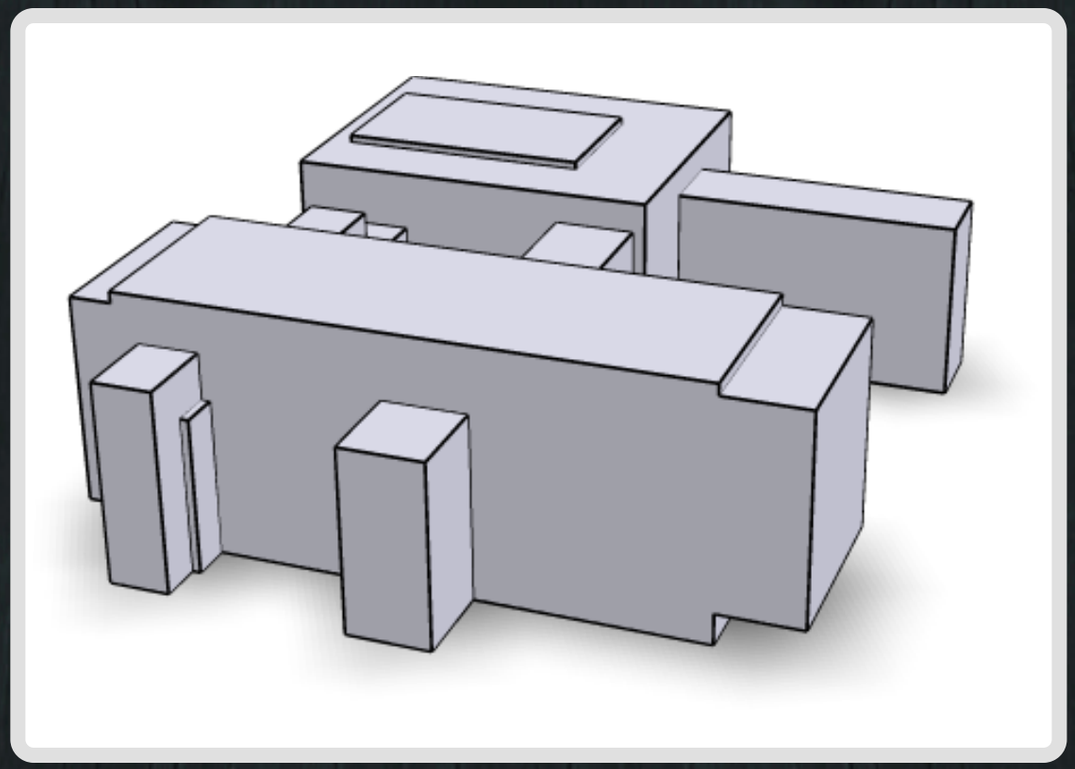

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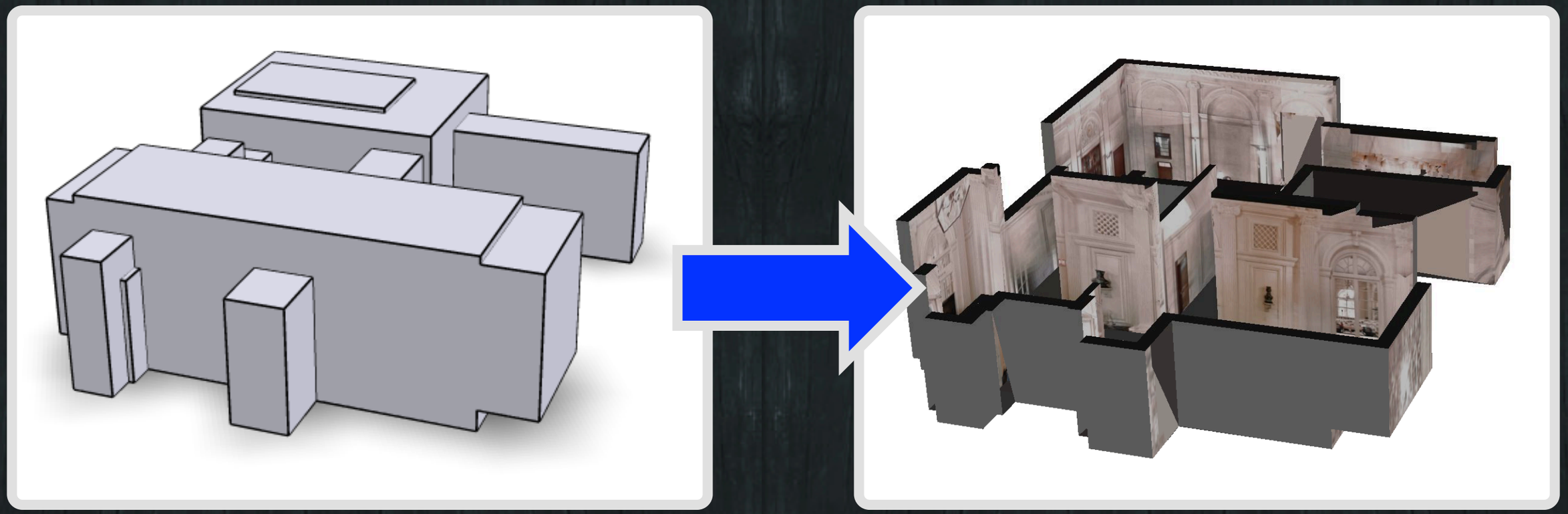


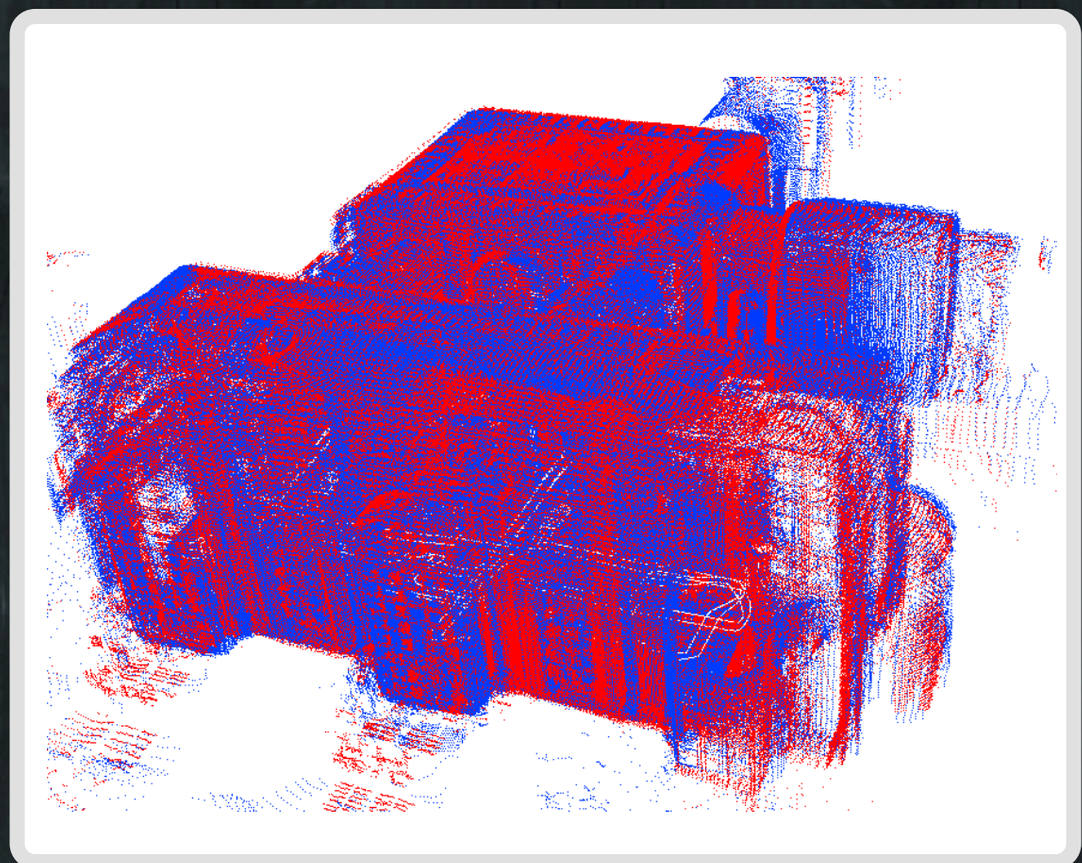
Results

Effective Navigation

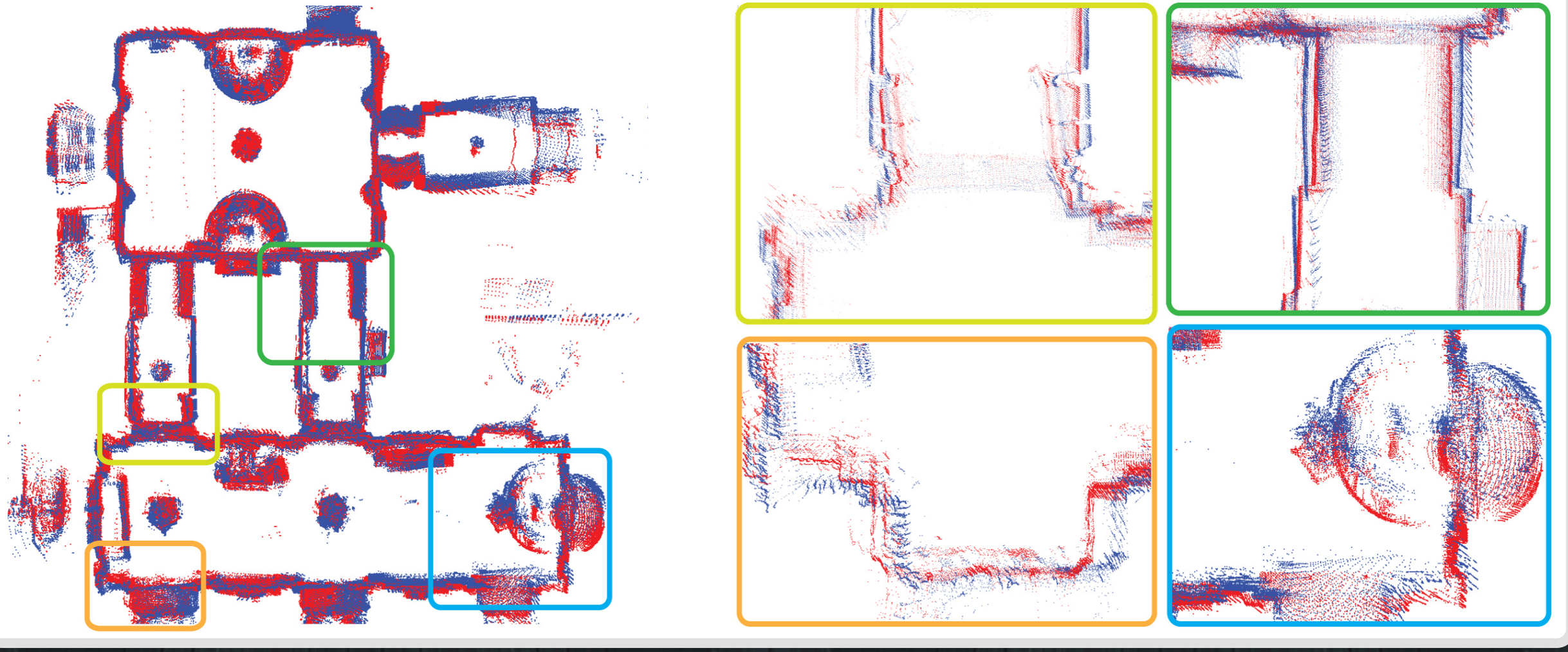
Reconstruction

Input

- Images (> 40,000)
- Laser points (> 200,000,000)
- Input collected over multiple sessions

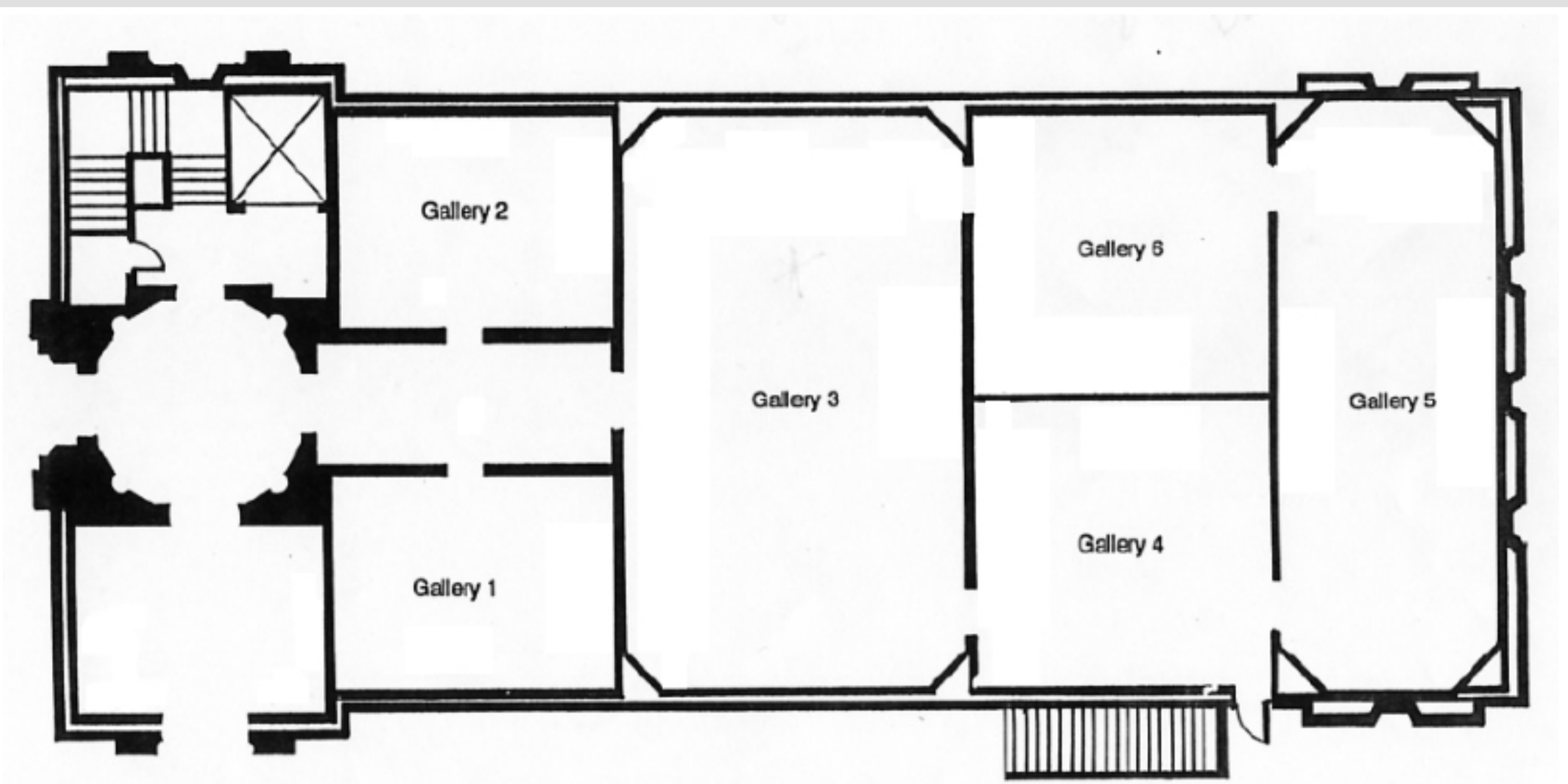


Noisy Laser Points



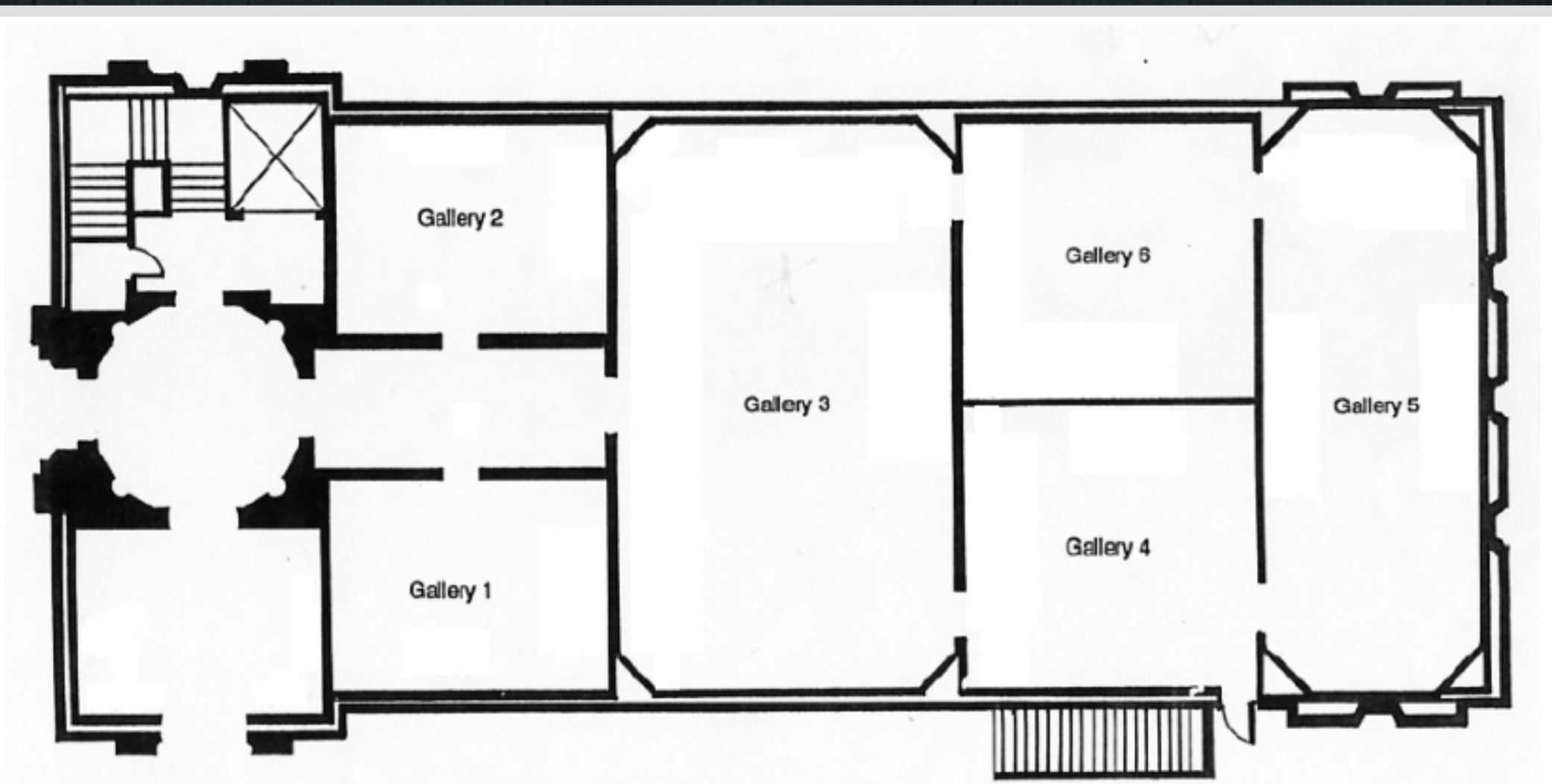
top-down view of input laser points
(two different colors represent two vertical laser range sensors)

Challenges for Indoor Scenes



Prevalence of thin structures

Opportunities for Indoor Scenes



Structural regularities
(planarity/orthogonality)

Exploit Structural Regularity

Image-based Street-side City Modeling, Xiao et al, 2009.

Xiao, Fang, Tan, Zhao, Ofek and Quan, 2008

Furukawa, Curless, Seitz and Szeliski 2009

Liu, Carlberg, Chen, Chen Kua and Zakhor, 2010

Adan and Huber, 2011

Gallup, Frahm and Pollefeys, 2010

Sinha, Steedly and Szeliski, 2009

Birchfield and Tomasi, 1999

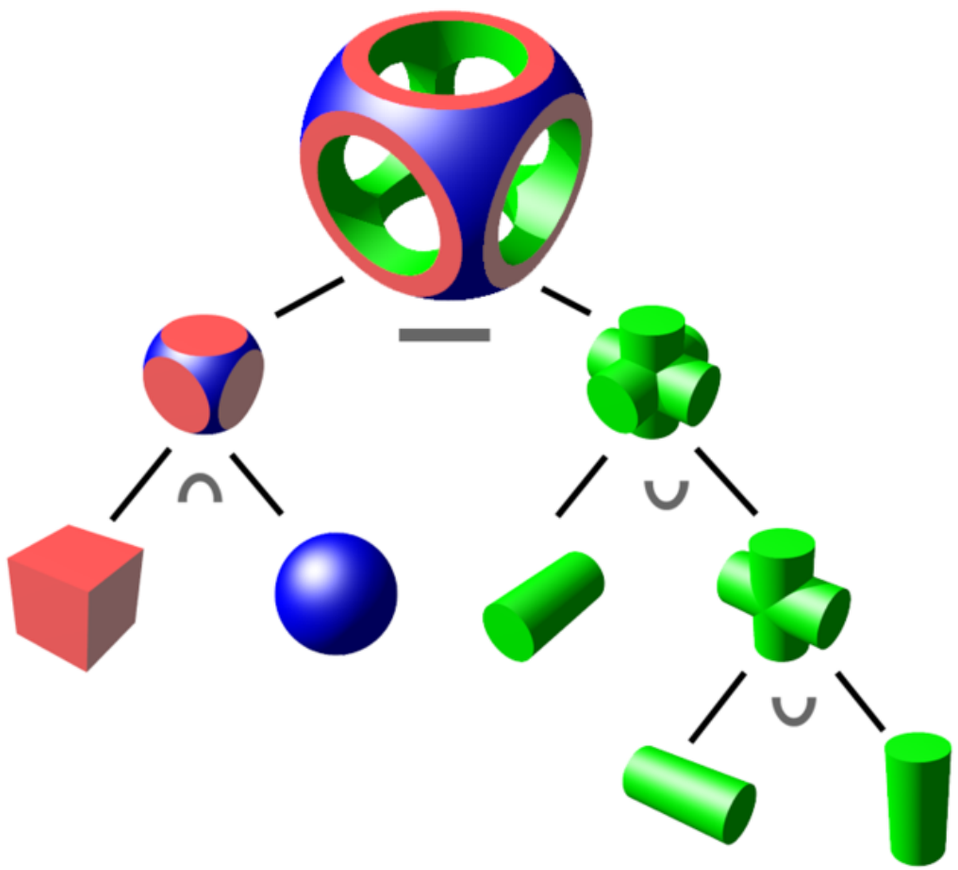
Wang and Adelson, 1994 …

Summary on Previous Work

- Structure prior enforced in 2D
  (depth or façade)
- Visualization limited to ground level
- Assume near-perfect calibration

InverseCSG Algorithm

Constructive Solid Geometry (CSG)



InverseCSG Algorithm

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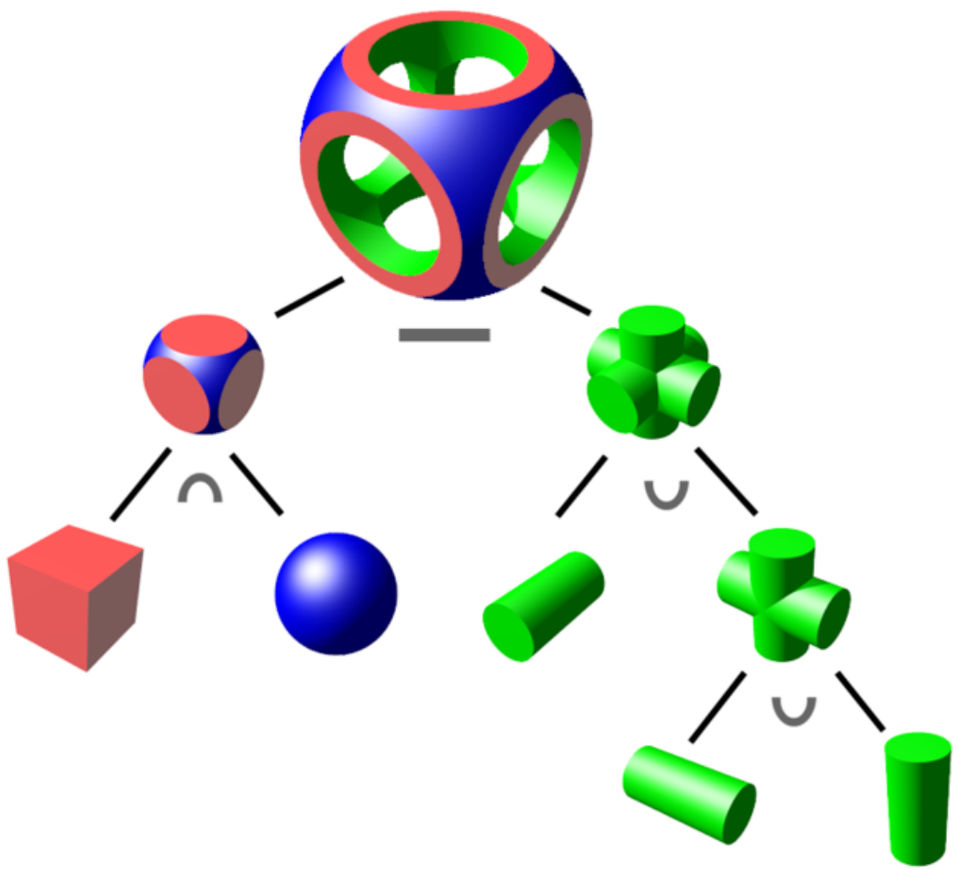
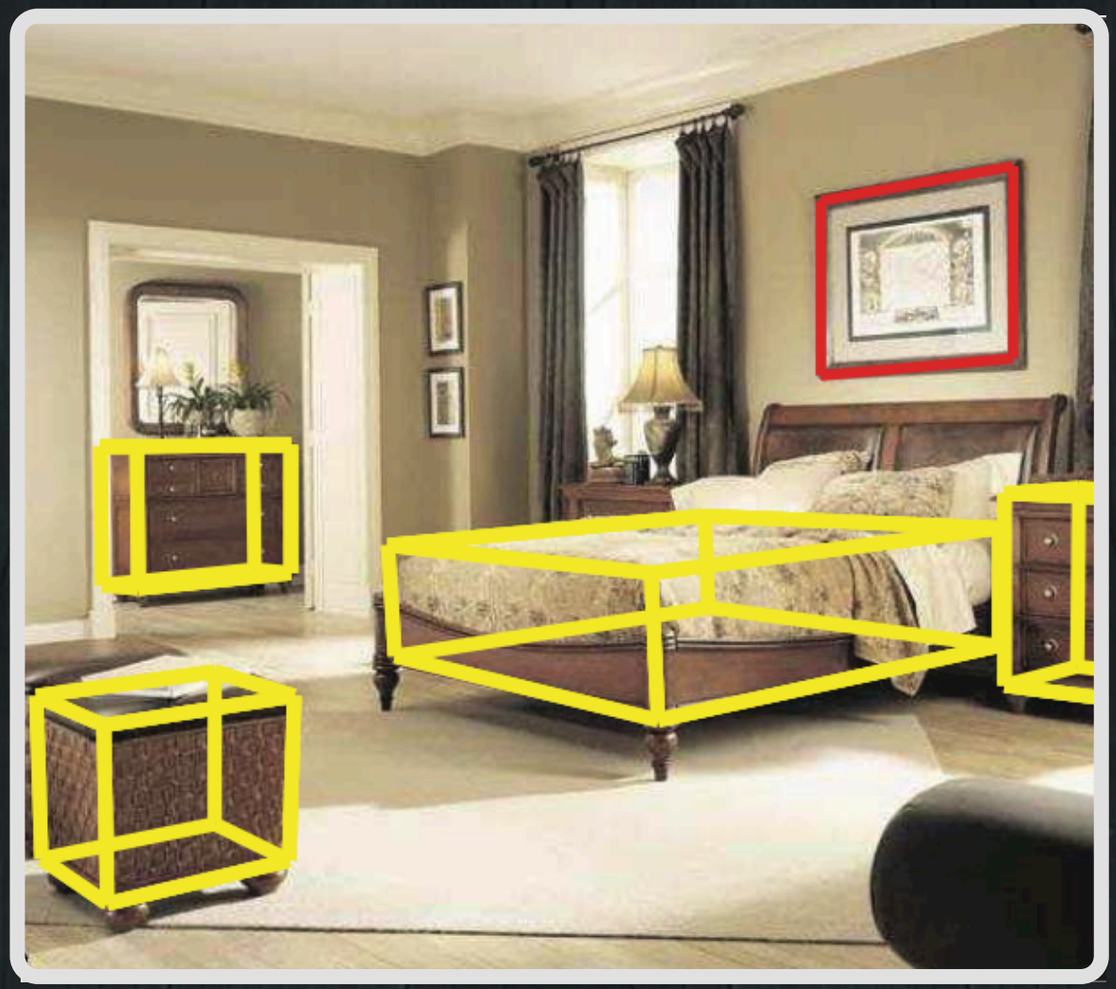


Cuboids as primitives



Xiao et al. NIPS 2012
Xiao et al. Siggraph Asia 2012a

Bottom-up & Top-down

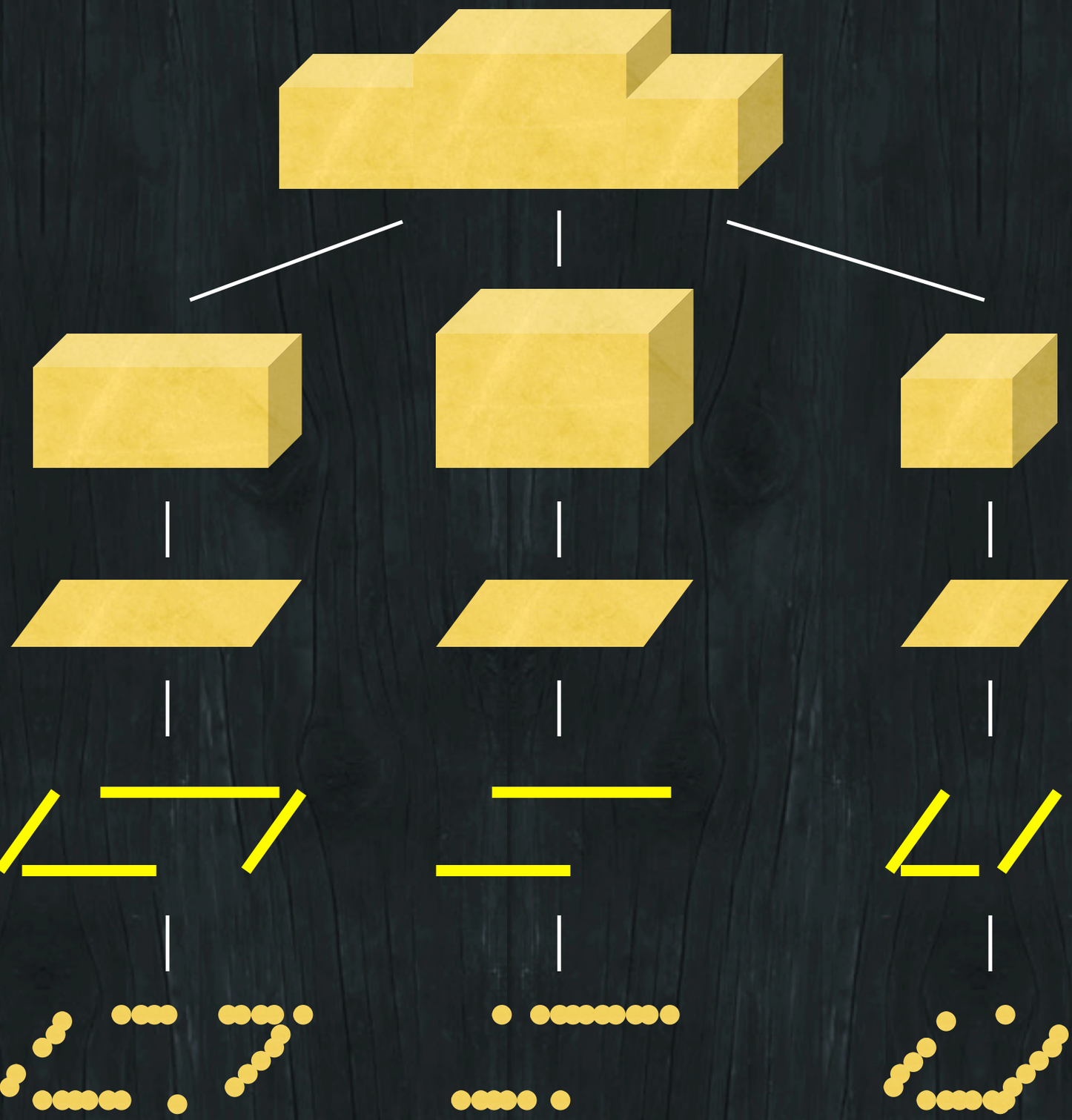
model

3D cuboid

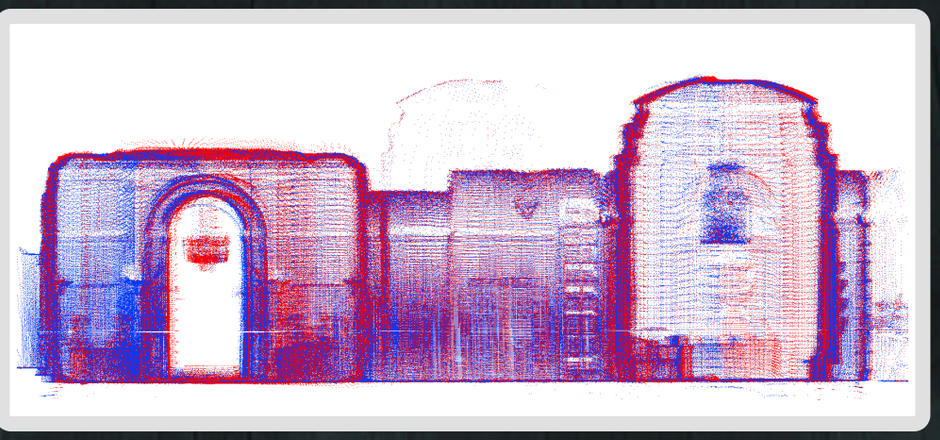
2D rectangle

2D line

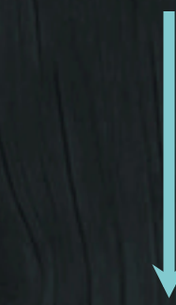
point



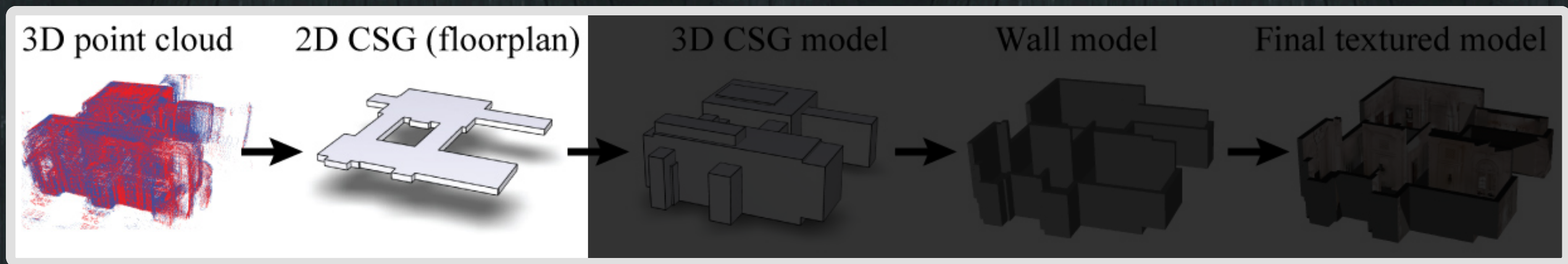
Cut into Slices



side view



gravity



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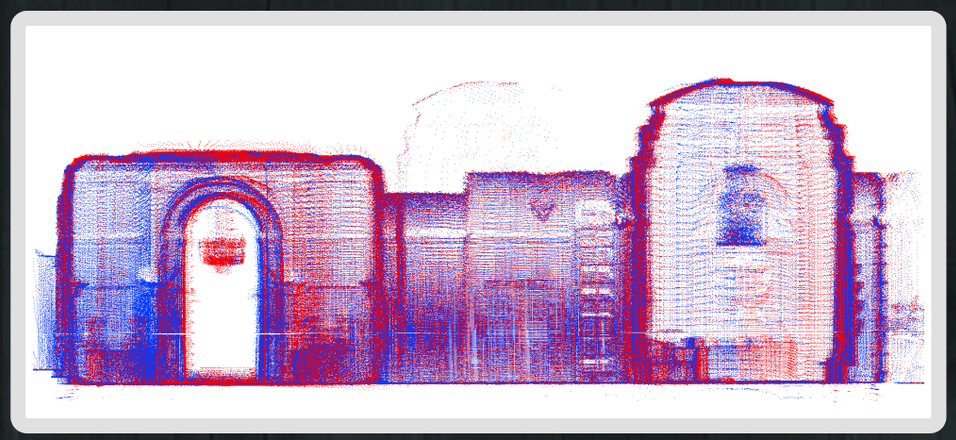

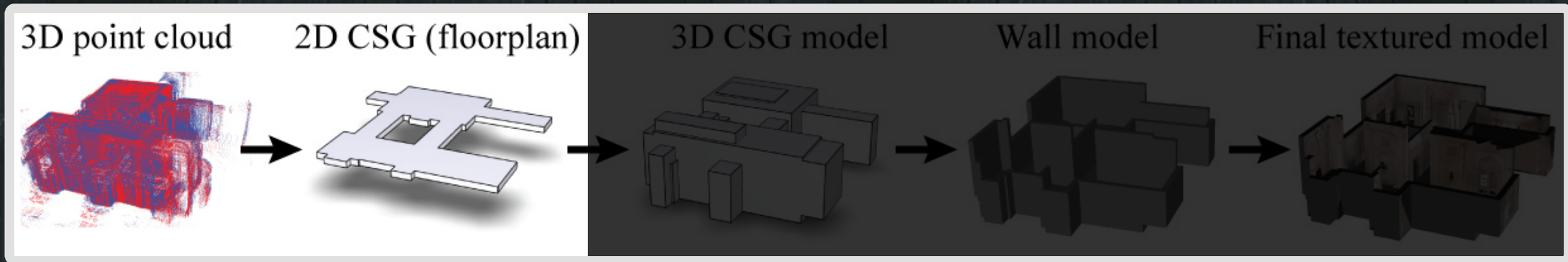

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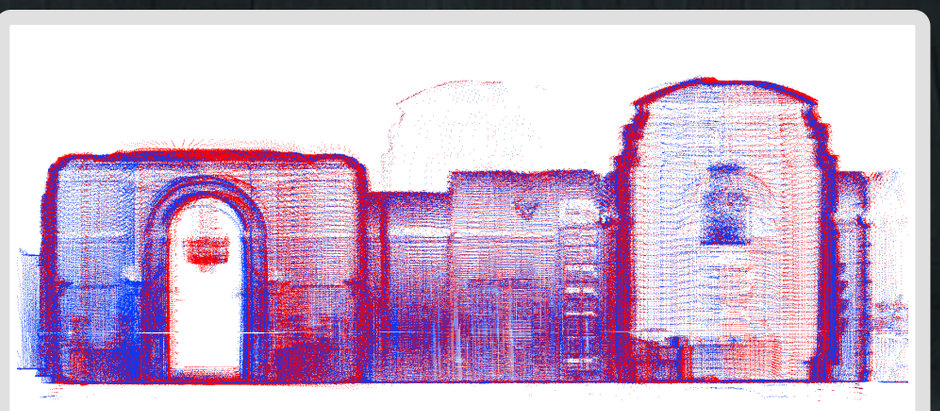
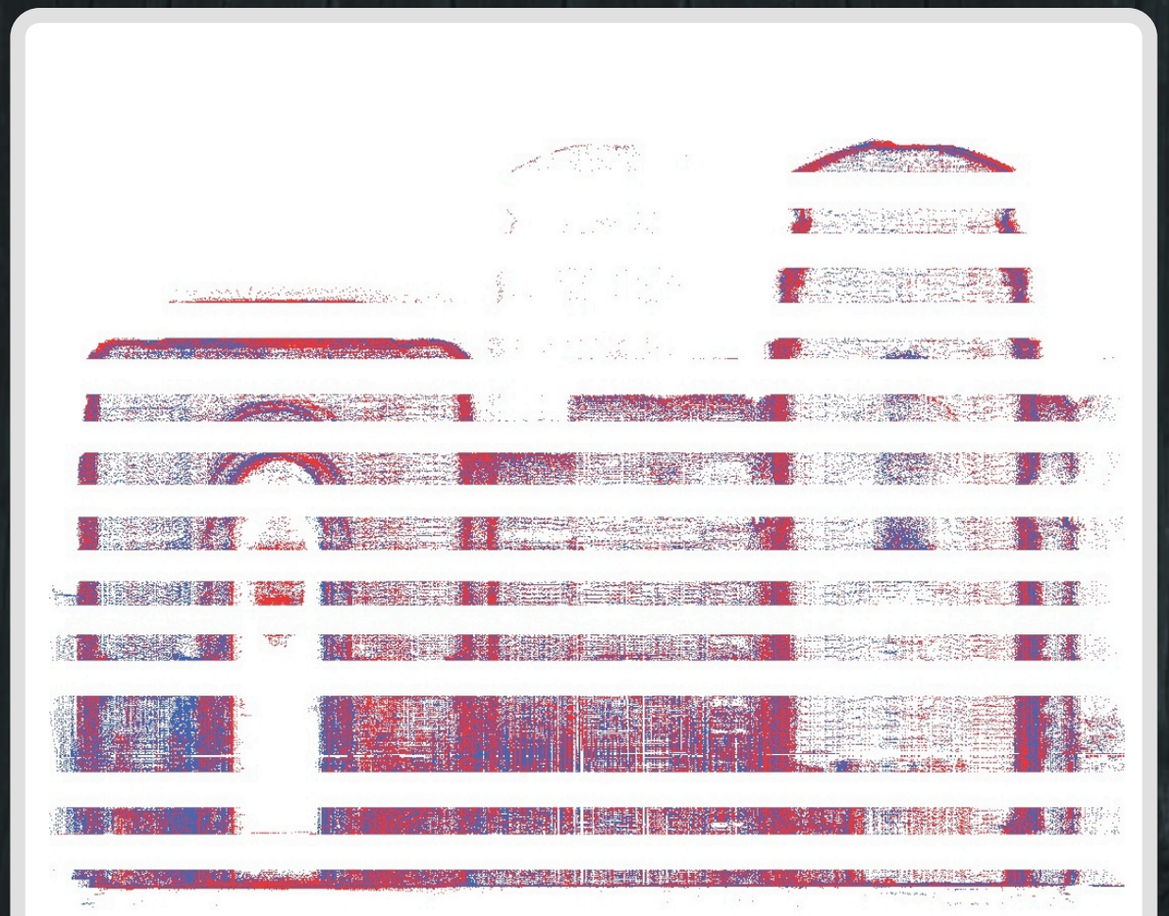
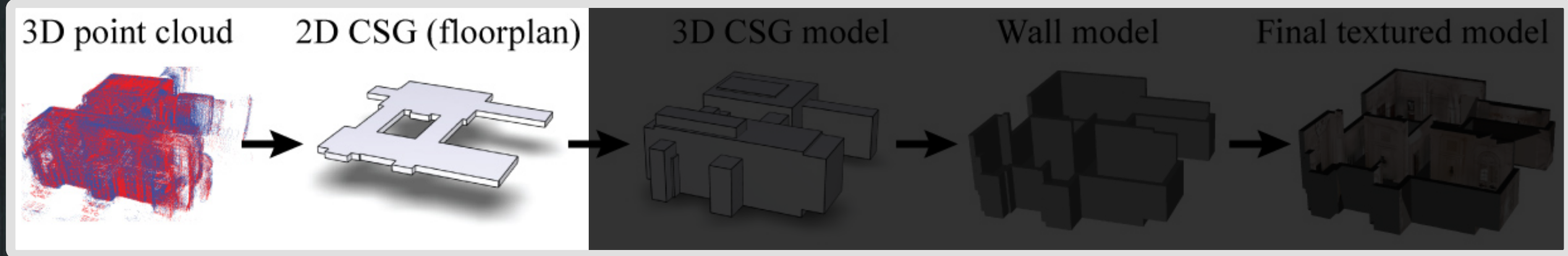


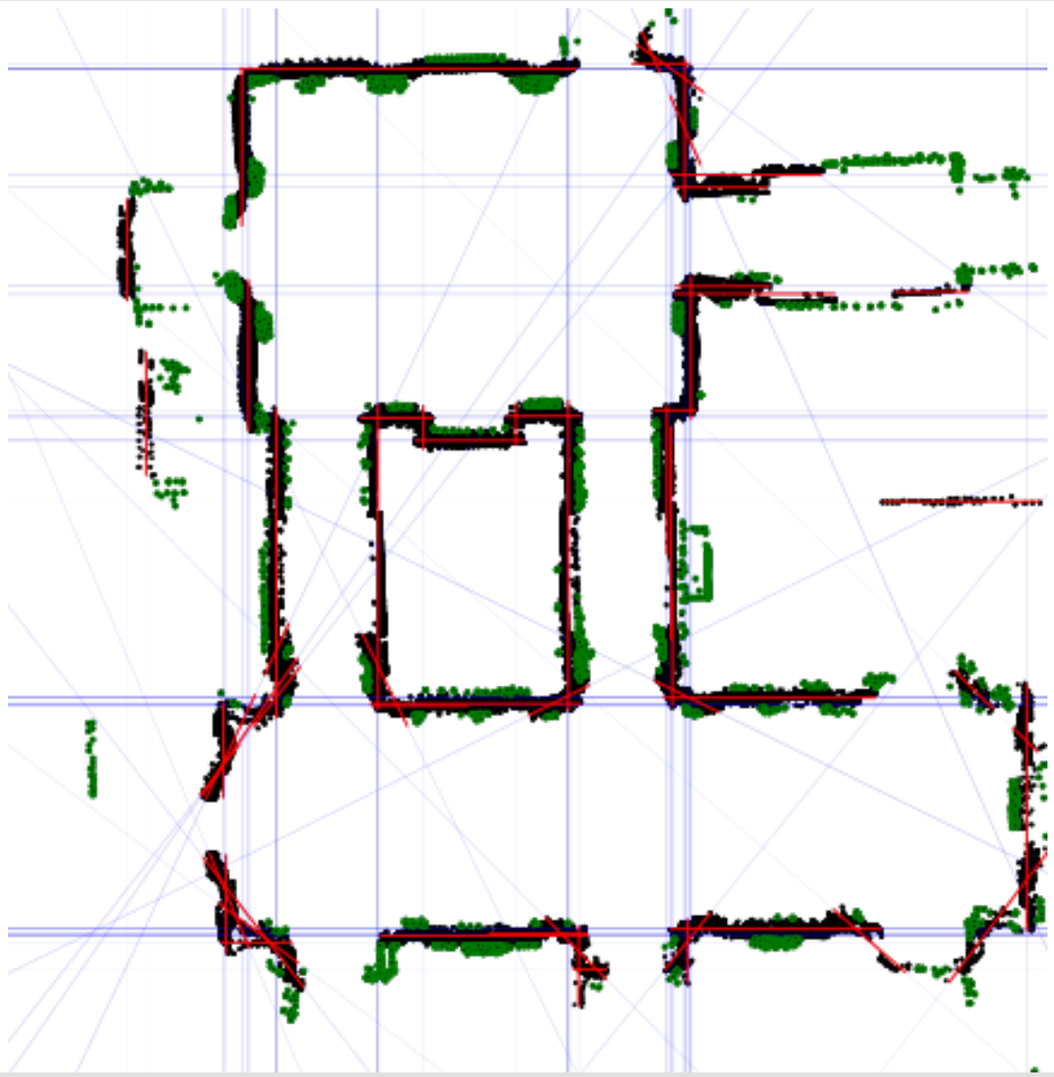


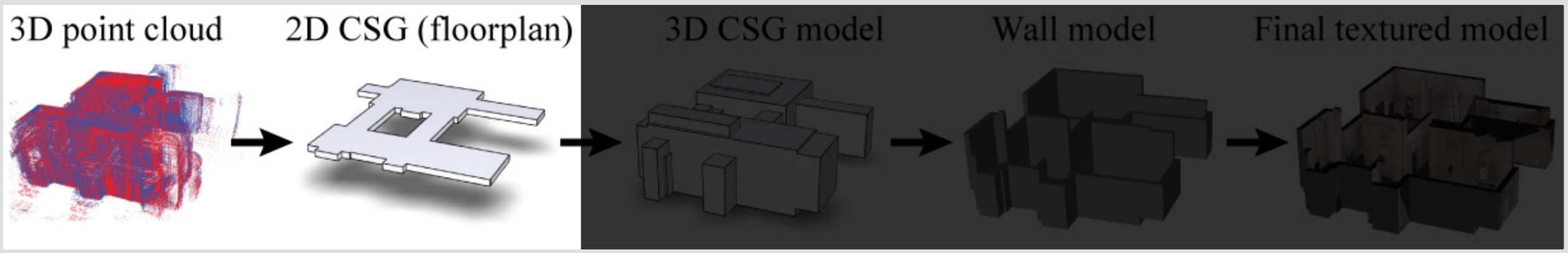


2D CSG Reconstruction

1. Generate primitives

2. Choose a subset





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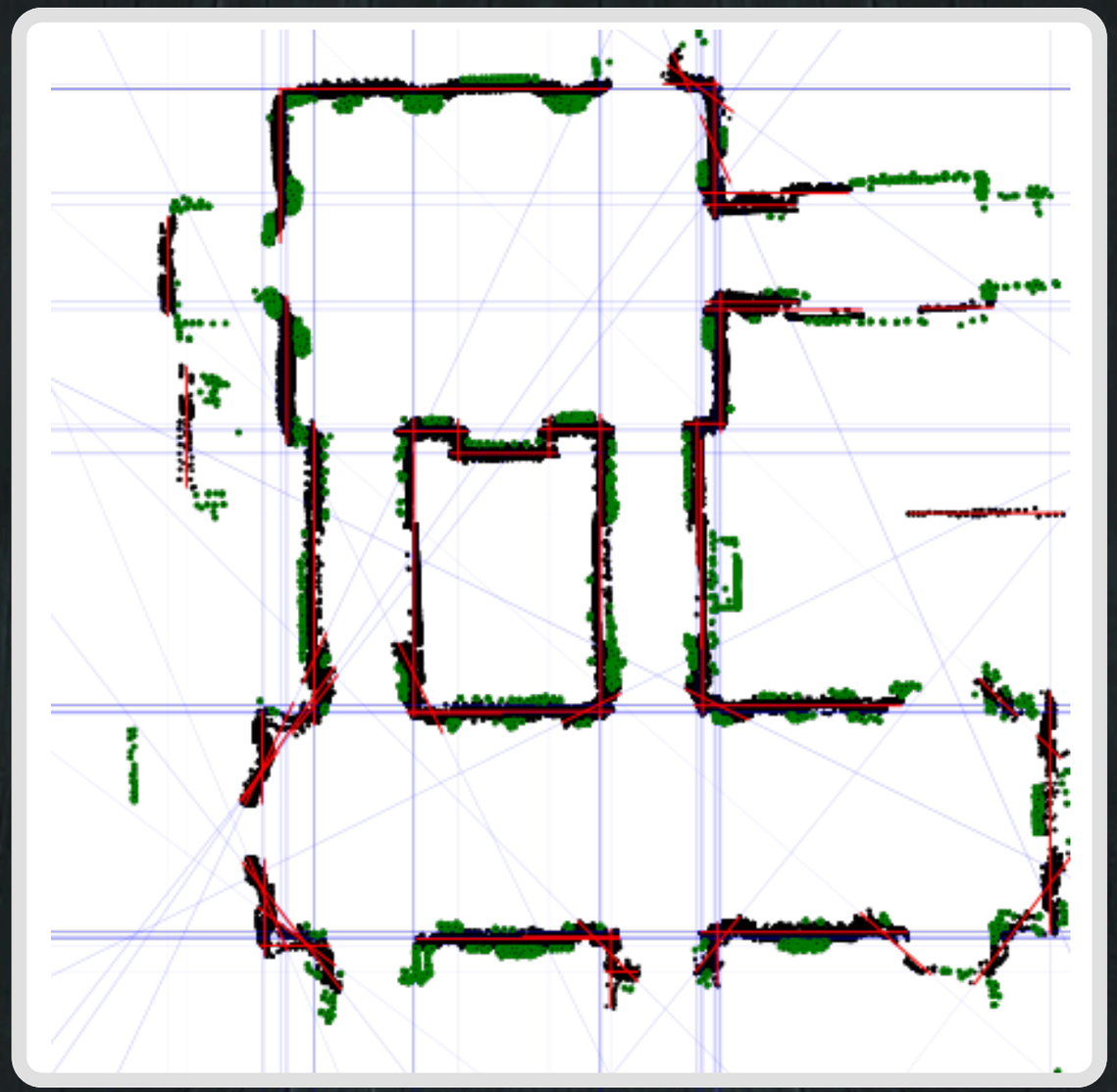


point → line

2D CSG Reconstruction

1. Generate primitives

point → line

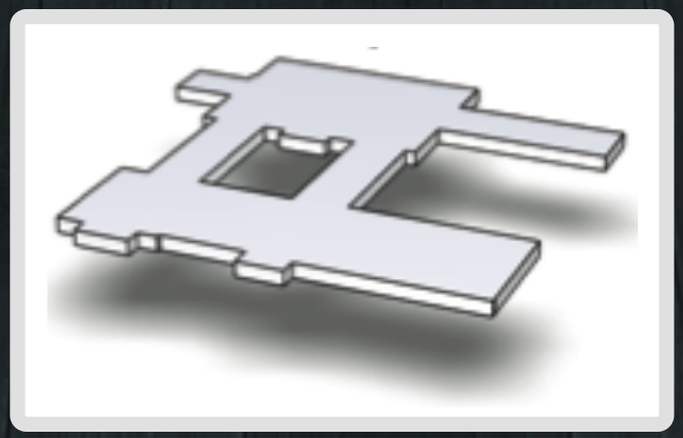
line → rectangle

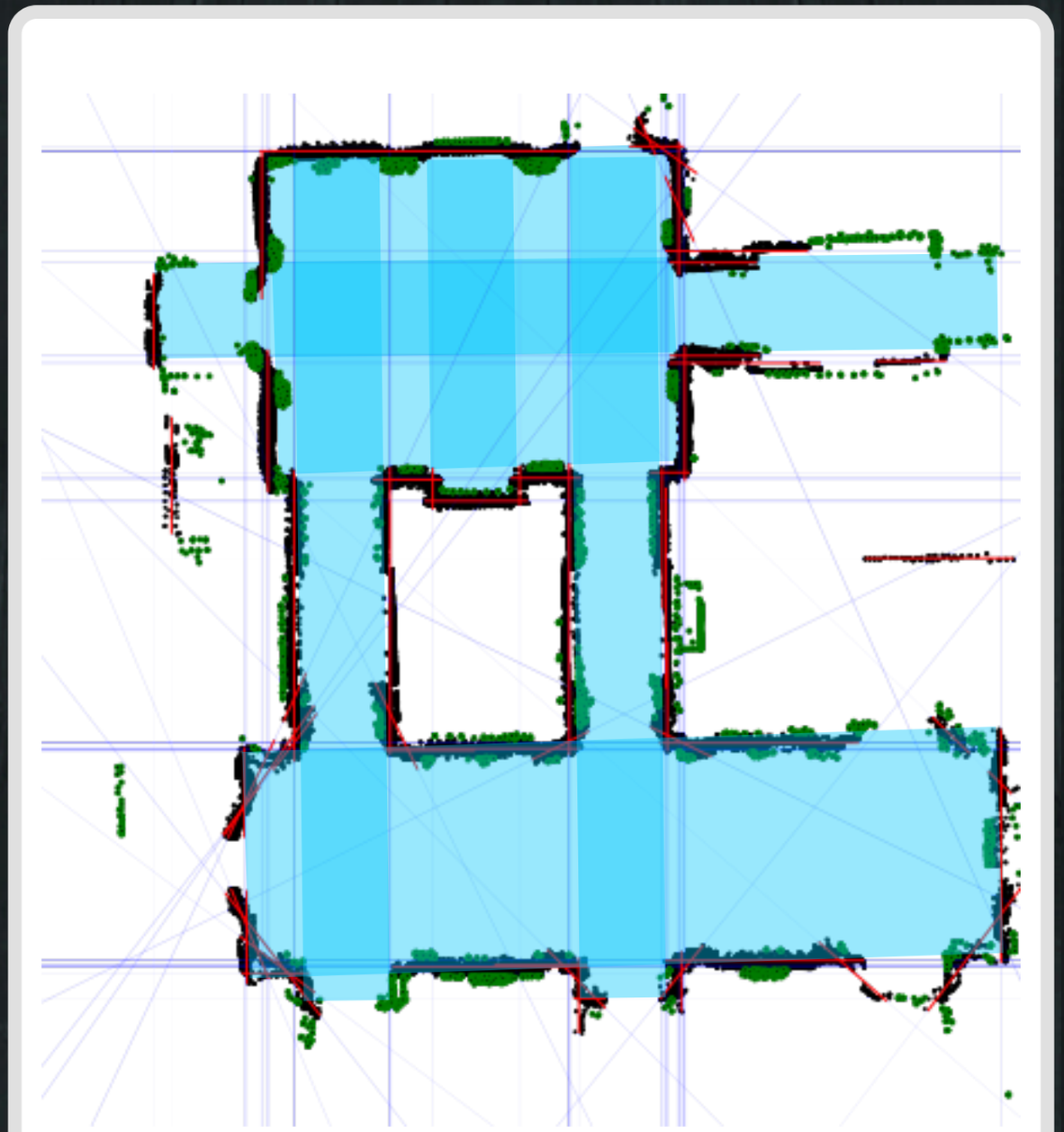
From 4 line segments

2D CSG Reconstruction

1. Generate primitives

2. Choose a subset



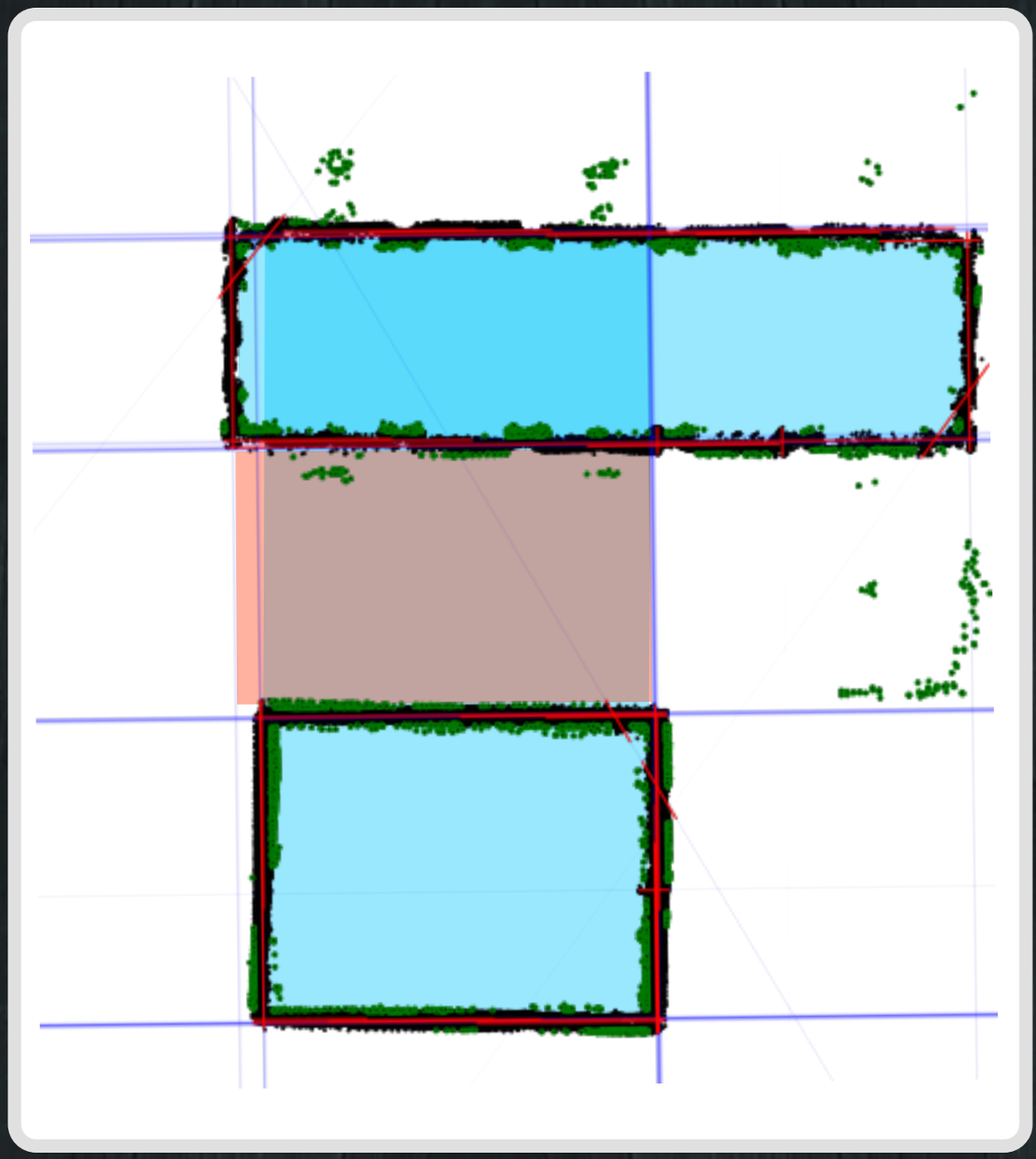


2D CSG Reconstruction

1. Generate primitives

2. Choose a subset

Repeat in each slice



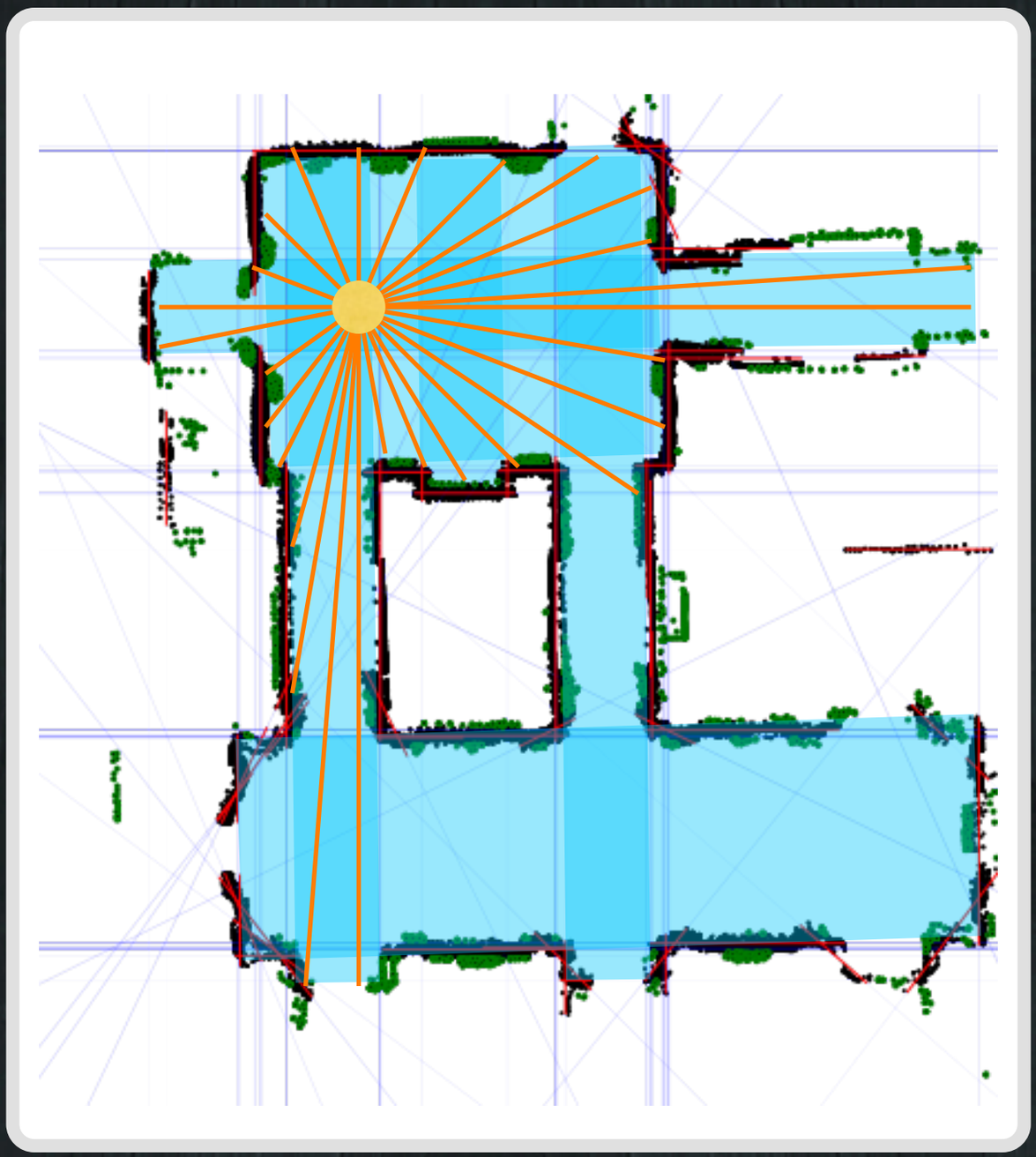
2D CSG Reconstruction

Explain the data

- Free space
- Laser points

Simple

- Regularization



Objective Function

$$E_1(T) = \frac{\{\text{Sum of free-space scores inside } T\}}{\{\text{Total sum in the domain without negative scores}\}}$$

$$E_2(T) = \frac{\{\#\text{ of points on the surface of } T\}}{\{\text{total \# of points}\}}$$

$$E_3(T) = \frac{\{\text{perimeter of } T \text{ near laser points (within 0.2 meters)}\}}{\{\text{total perimeter of } T\}}$$

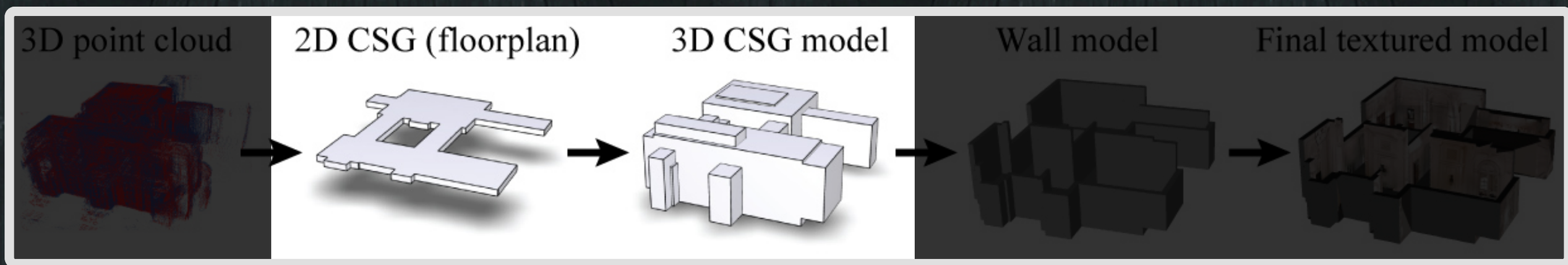
$$E(T) = w_1 E_1(T) + w_2 E_2(T) + w_3 E_3(T)$$

Free space Laser points Regularization

3D CSG Reconstruction

1. Generate primitives (cuboids)

2. Choose a subset (out of primitive candidates)



3D CSG Reconstruction

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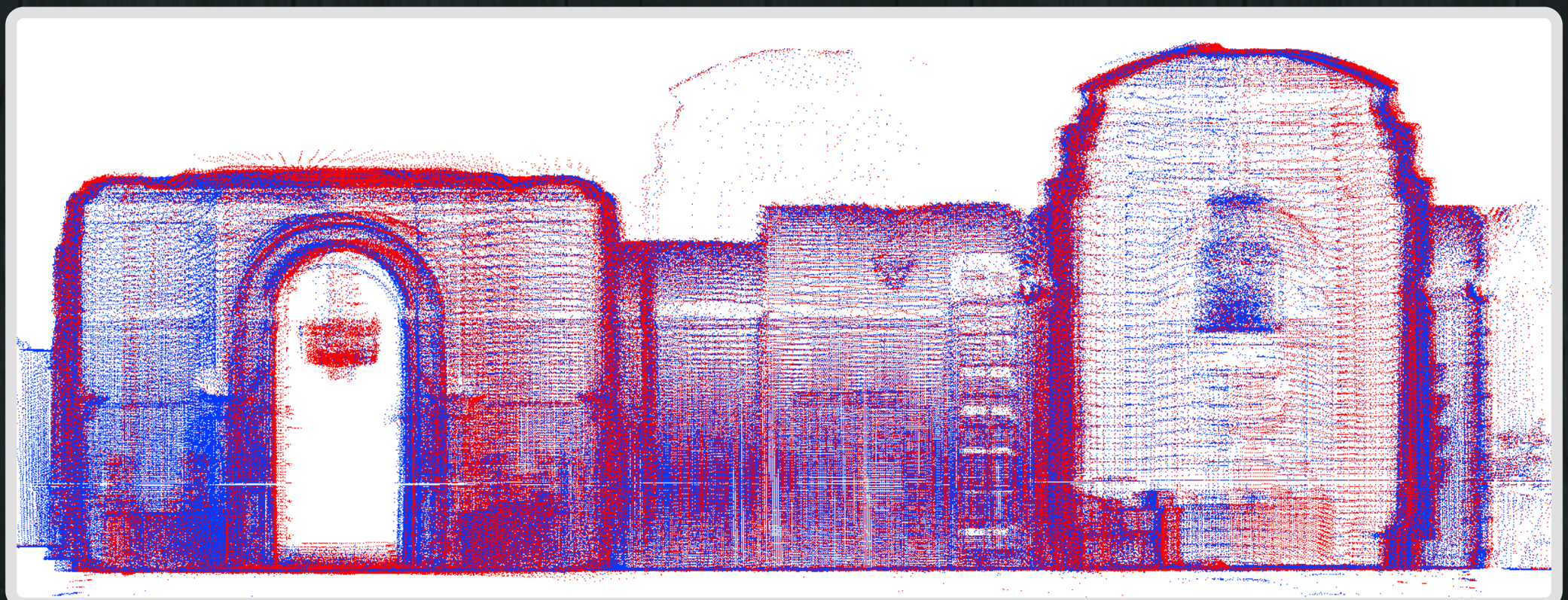

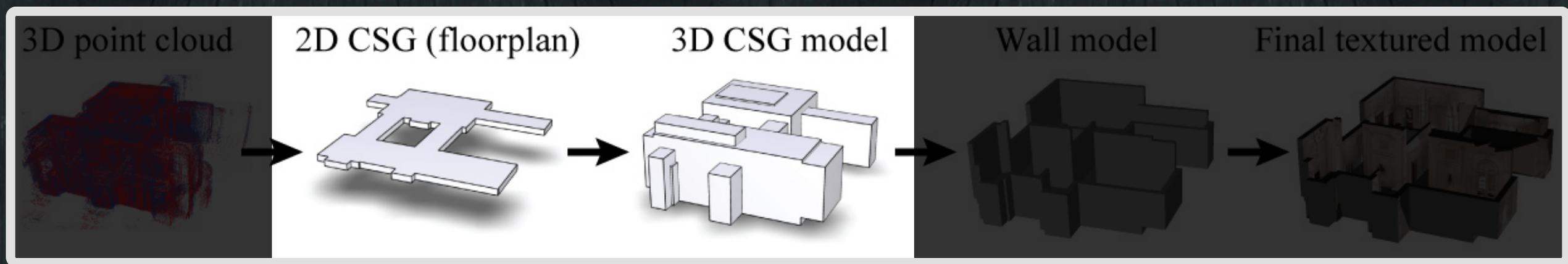

3D CSG Reconstruction

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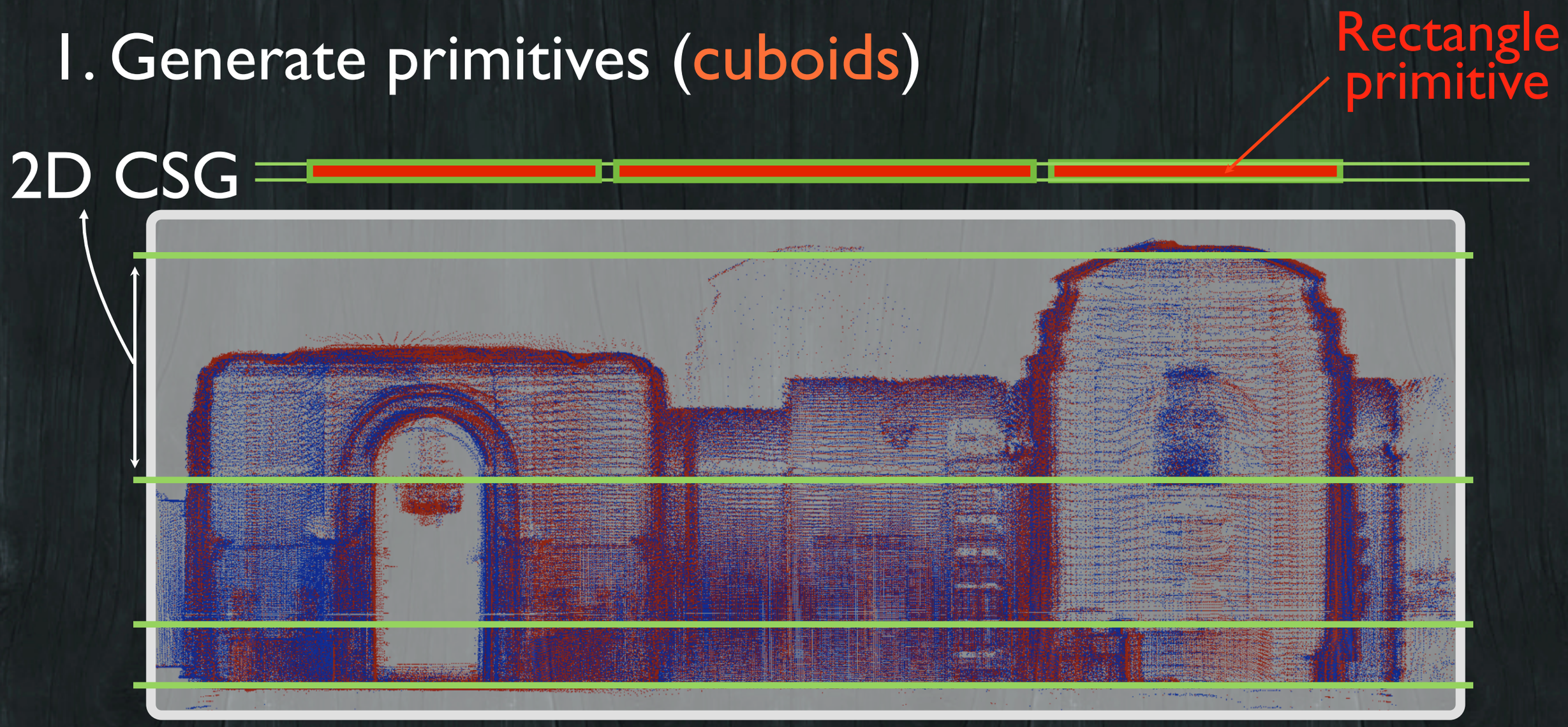

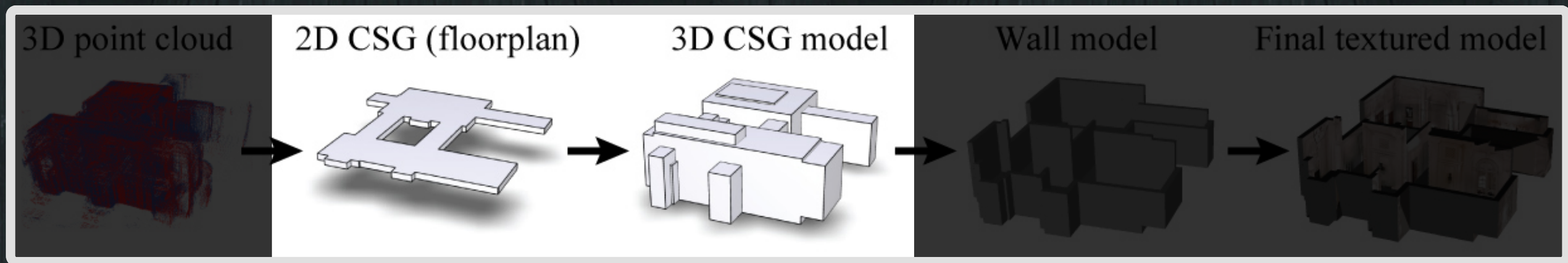

3D CSG Reconstruction

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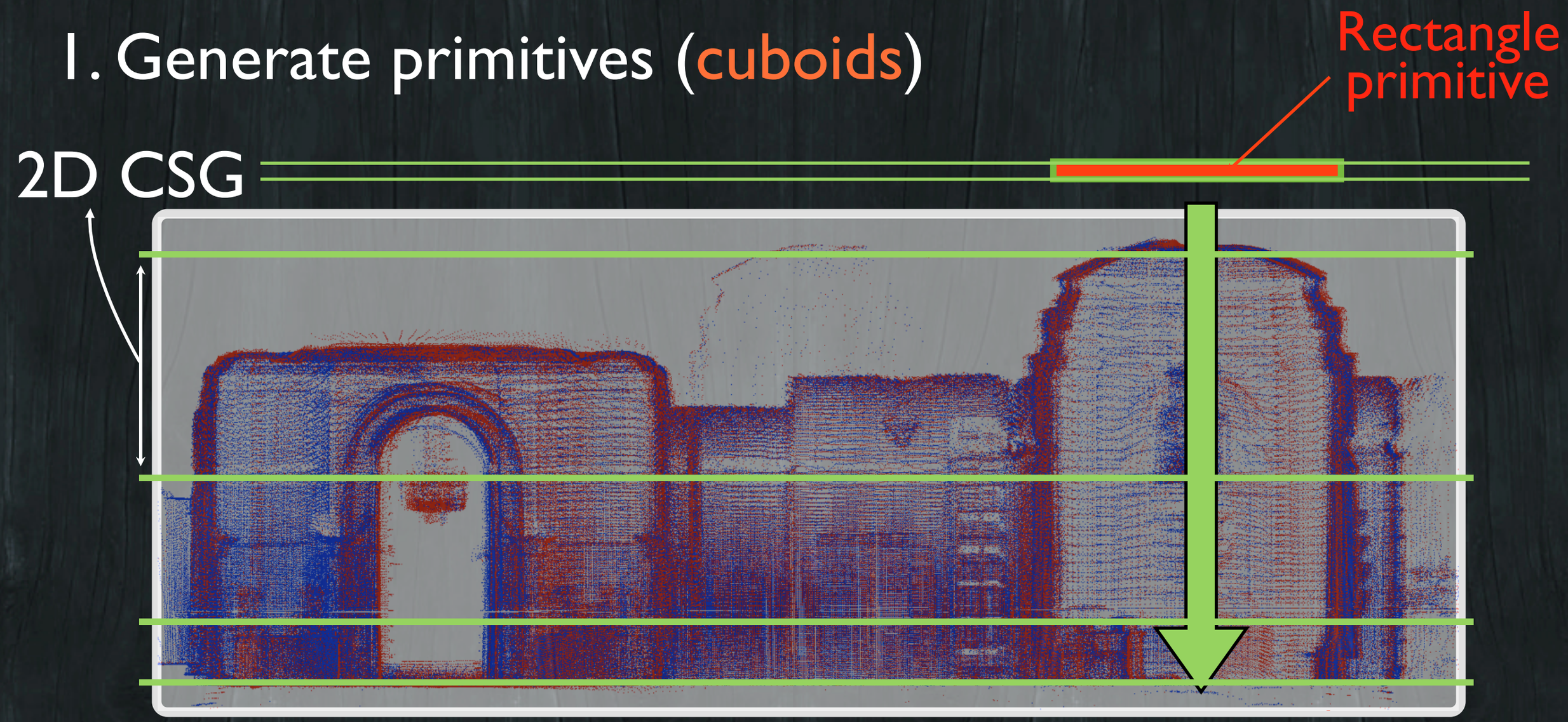

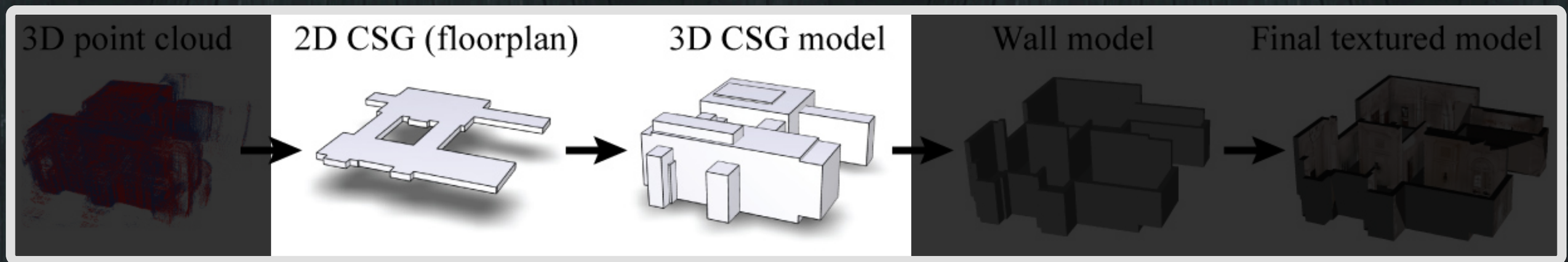

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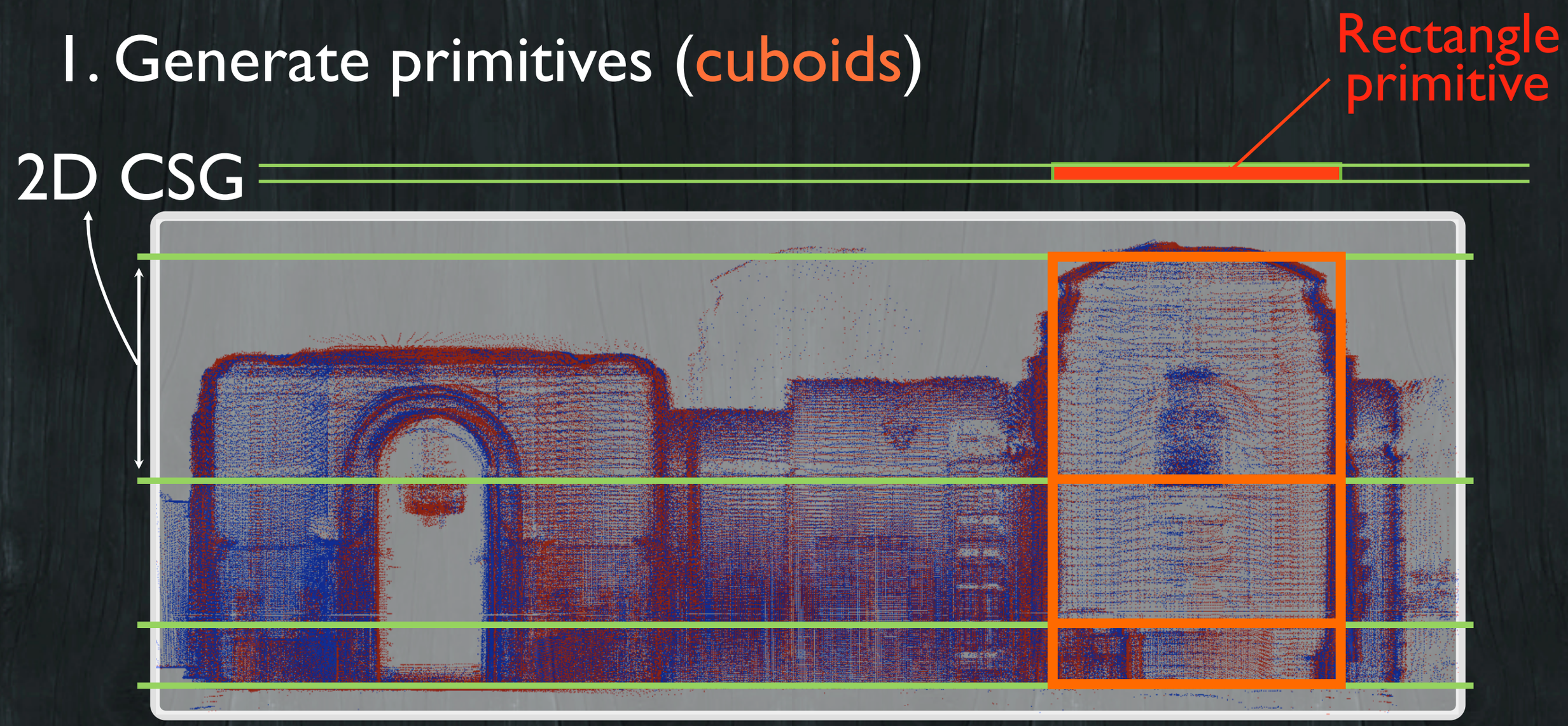
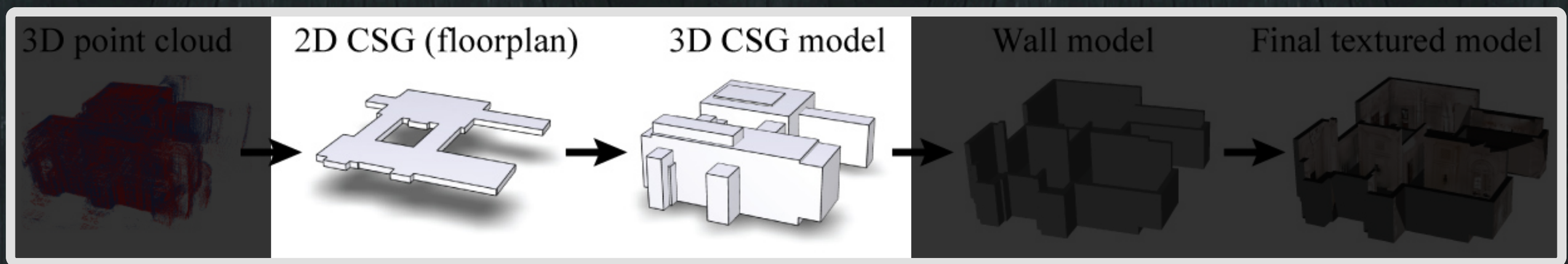
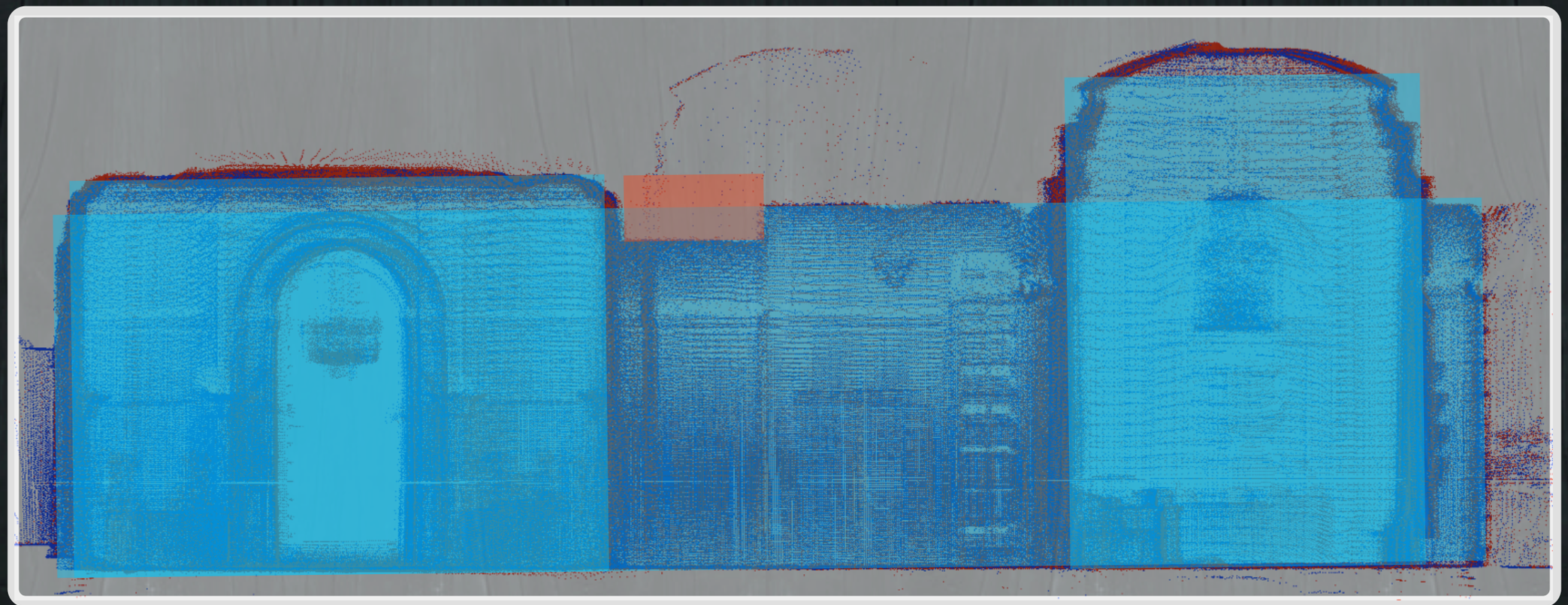
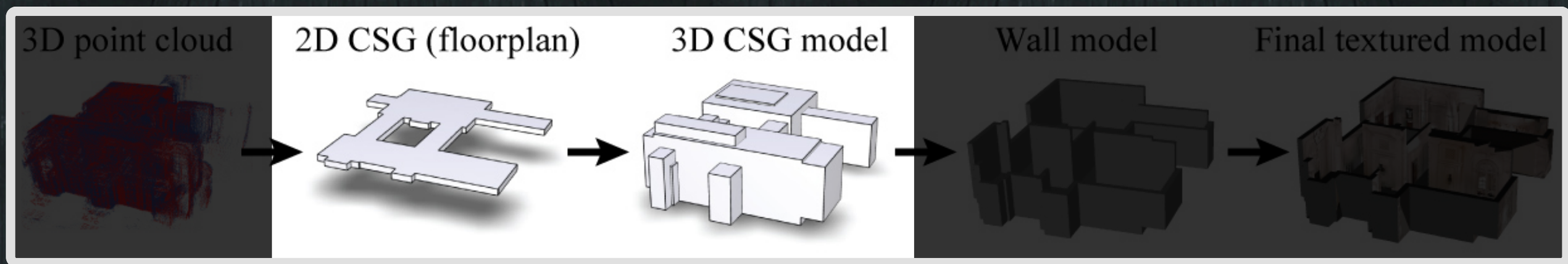




3D CSG Reconstruction

1. Generate primitives (cuboids)
2. Choose a subset

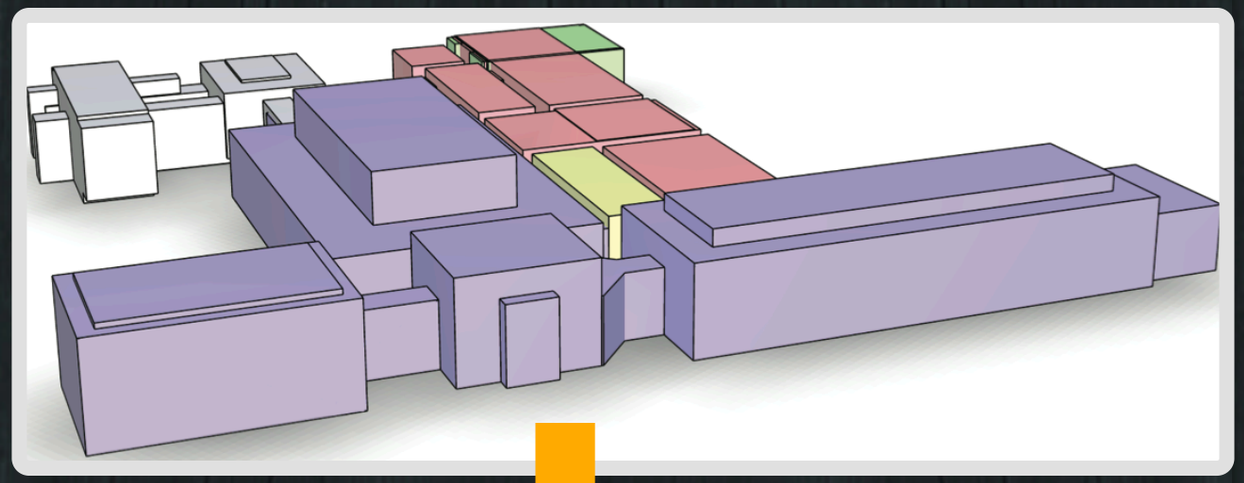




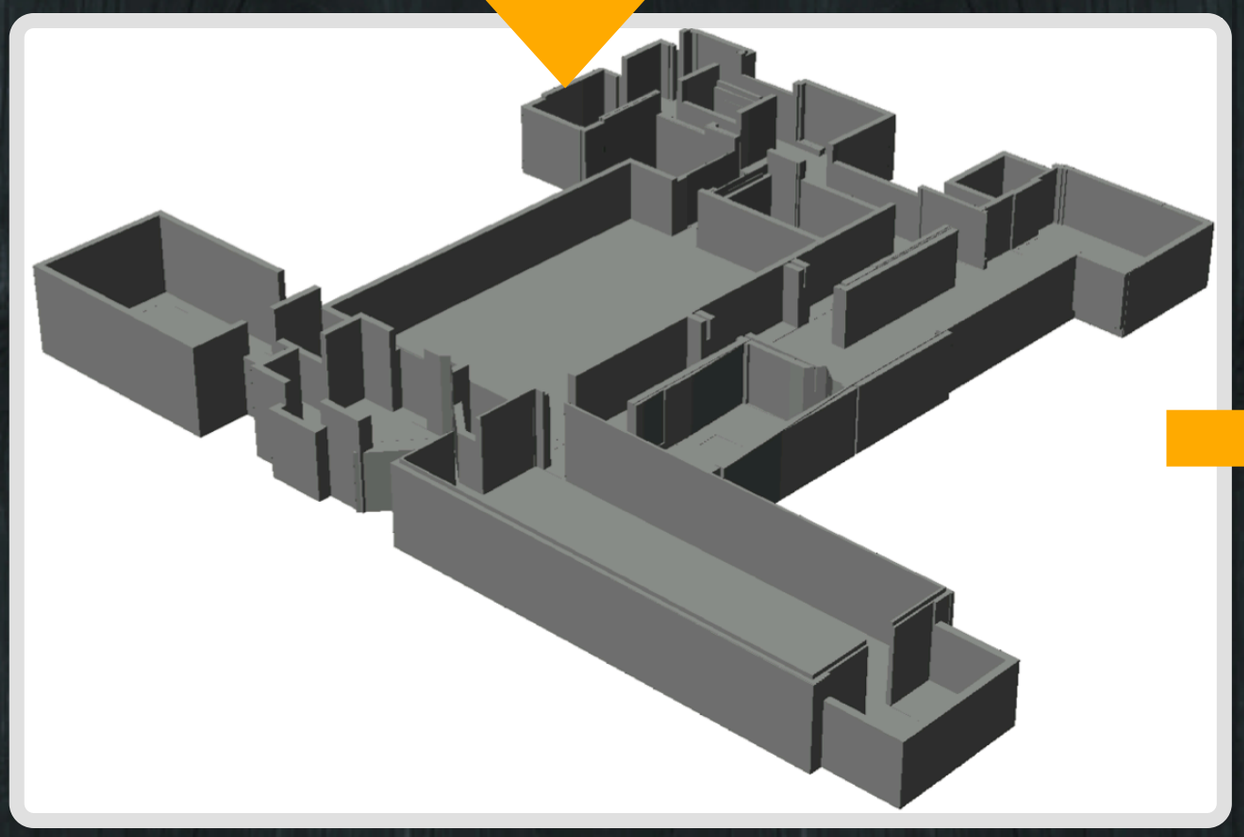
Algorithm on Run

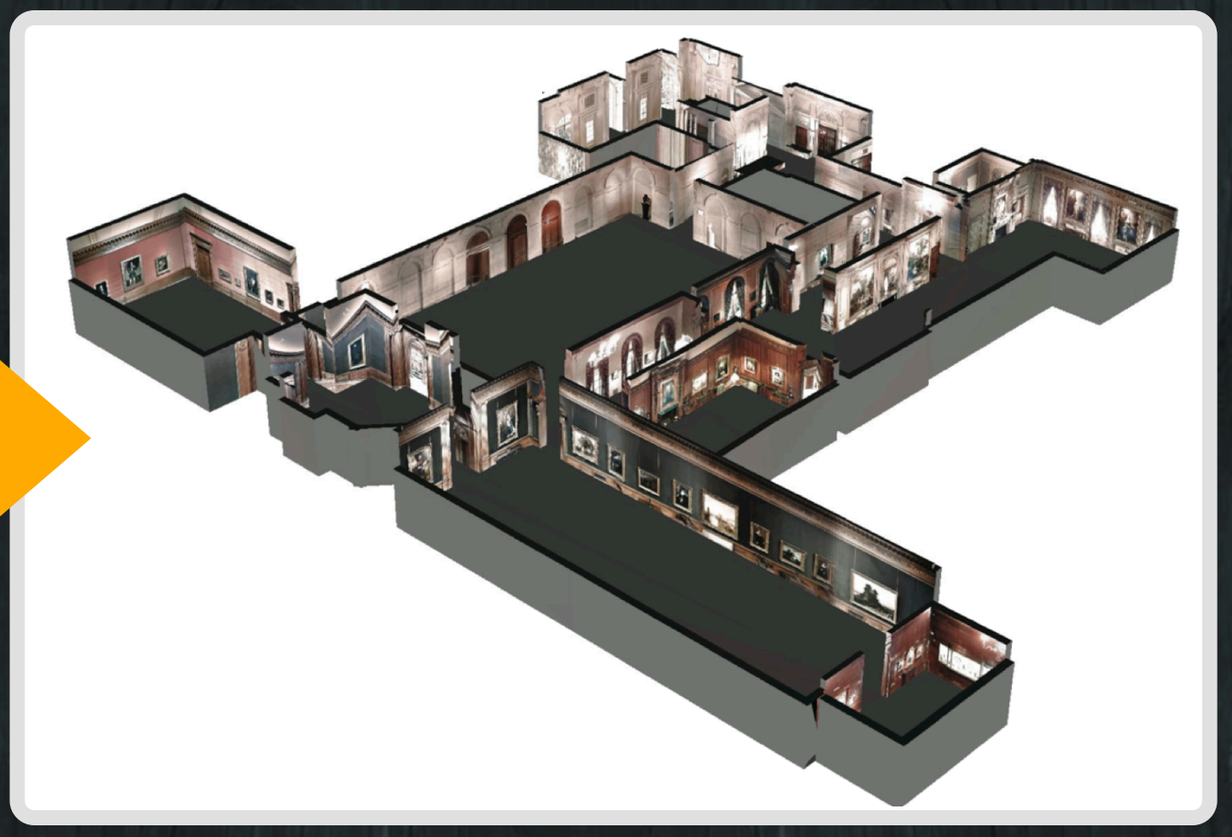
Step-by-step visualization
of
3D CSG model reconstruction

Last Step



1. Remove Ceiling
2. Texture Mapping

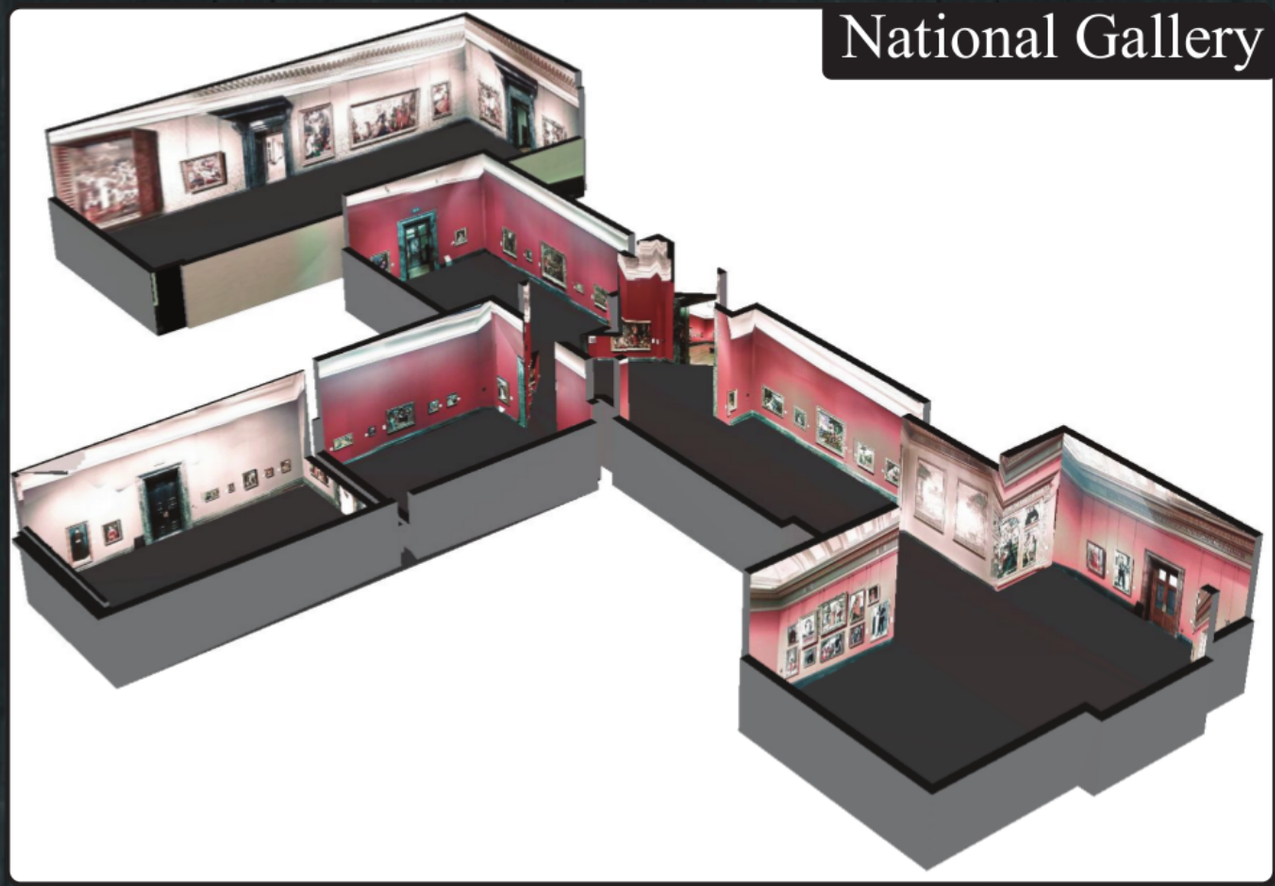




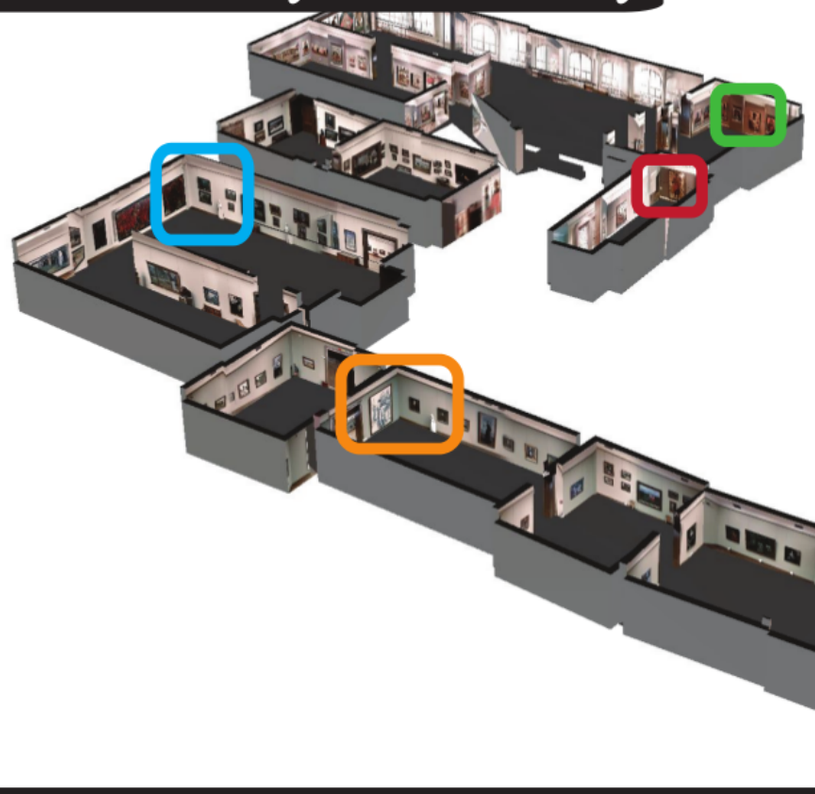
Per plane texture stitching [Xiao et al. 2008, Xiao et al. 2009].

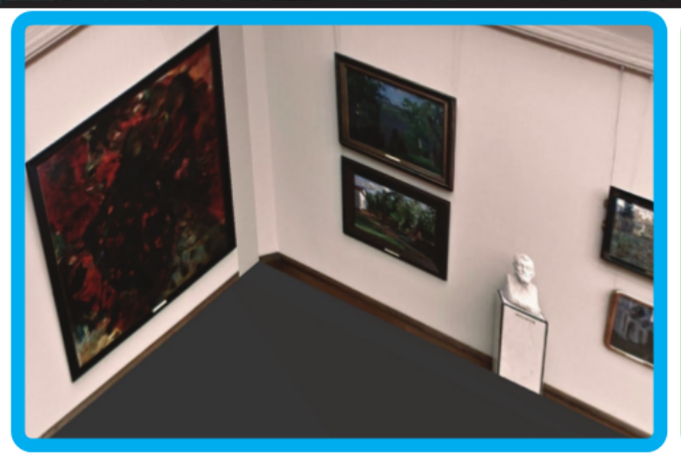
The Frick Collection

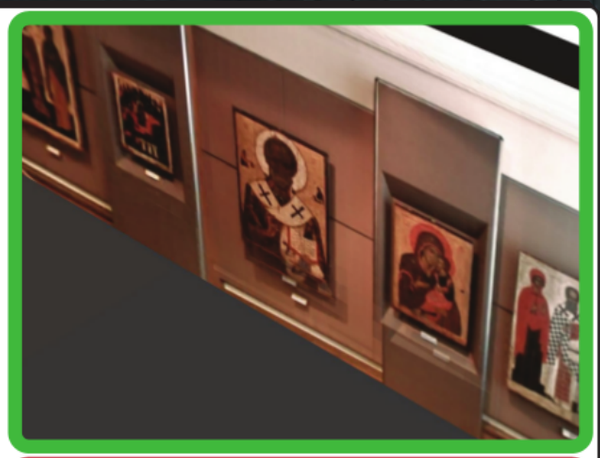
National Gallery

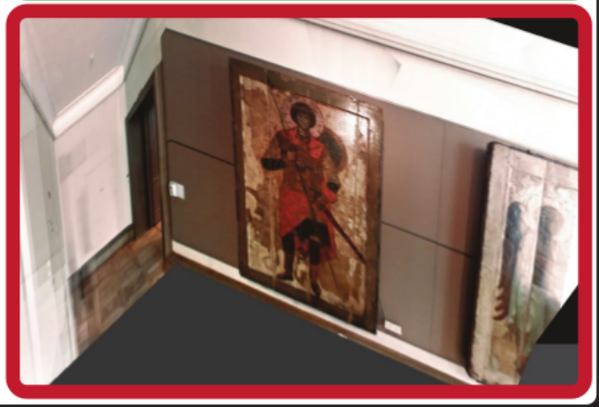


The State Tretyakov Gallery







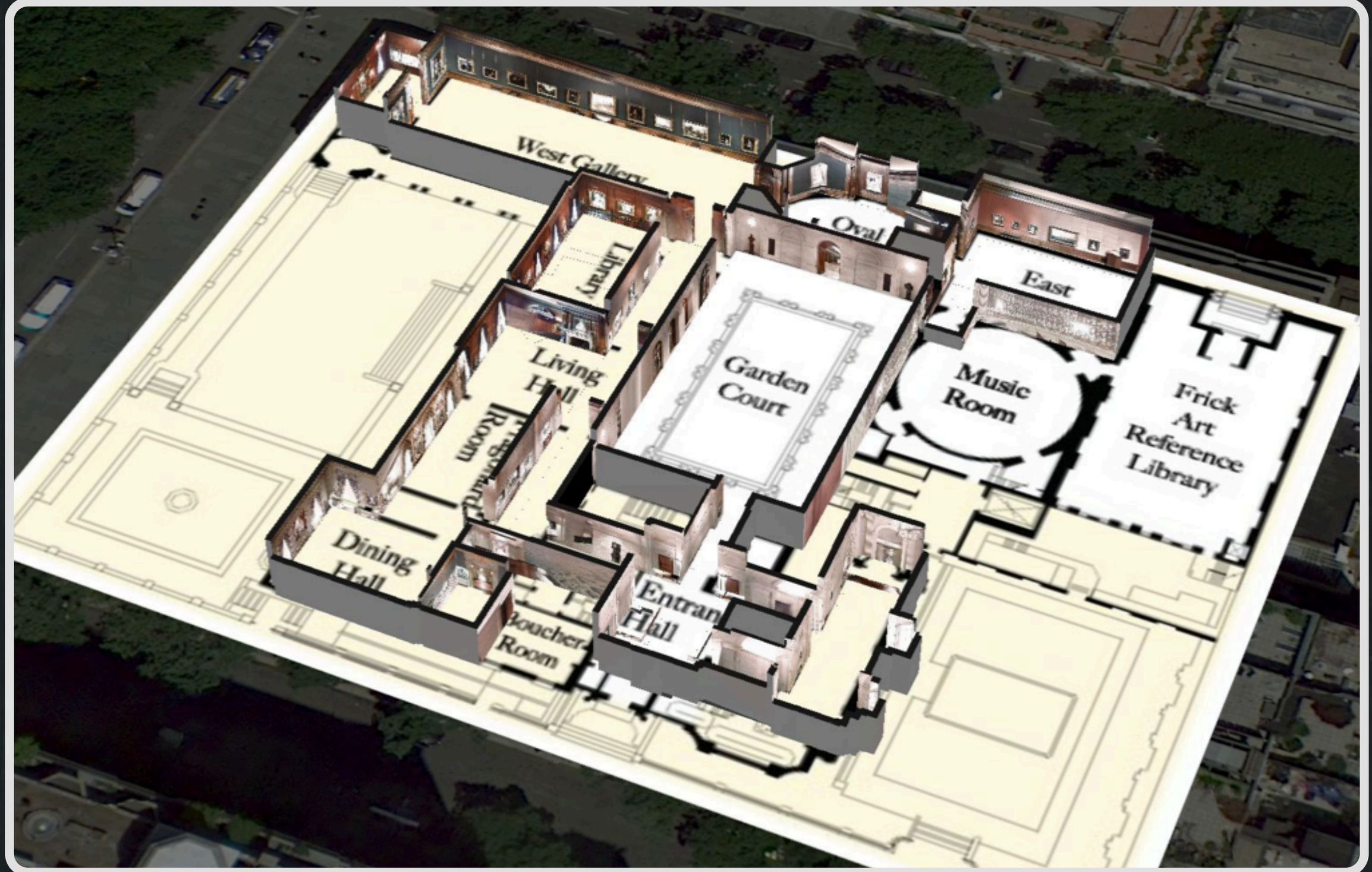


View-dependent Construction

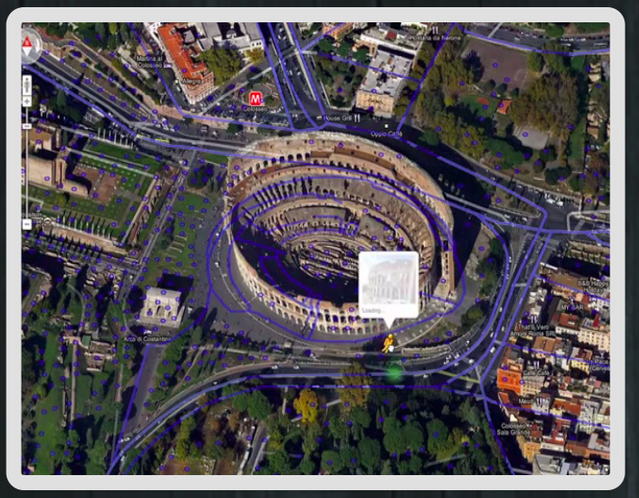
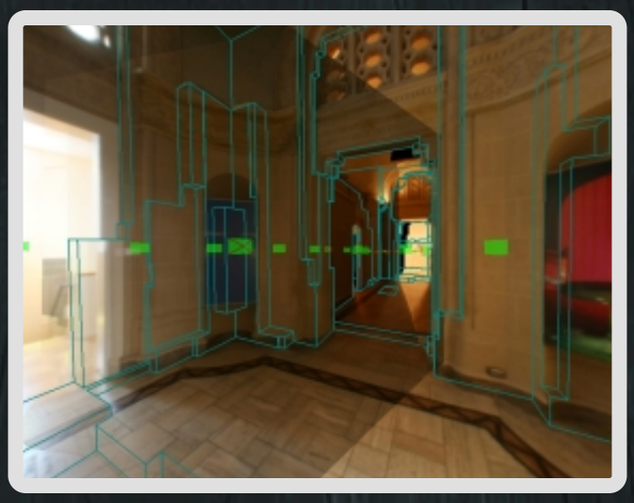
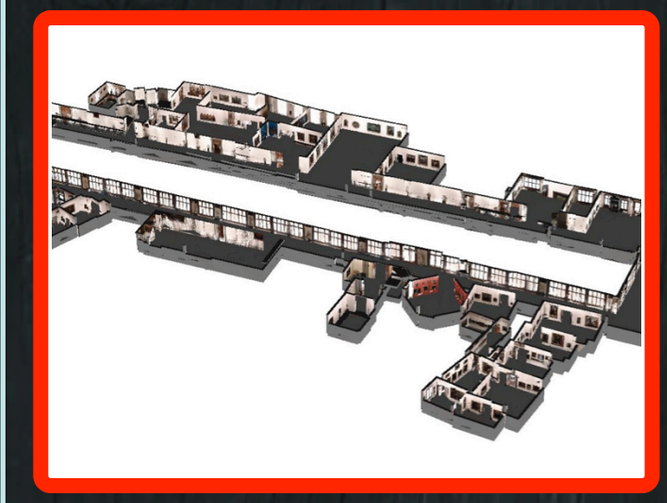
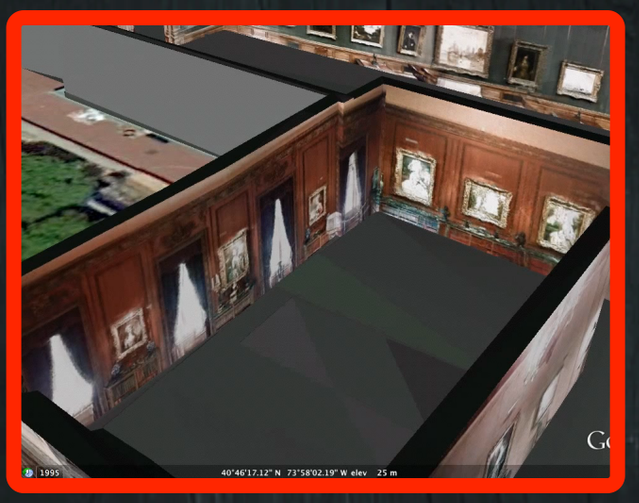
View-independent model

Aerial Maps

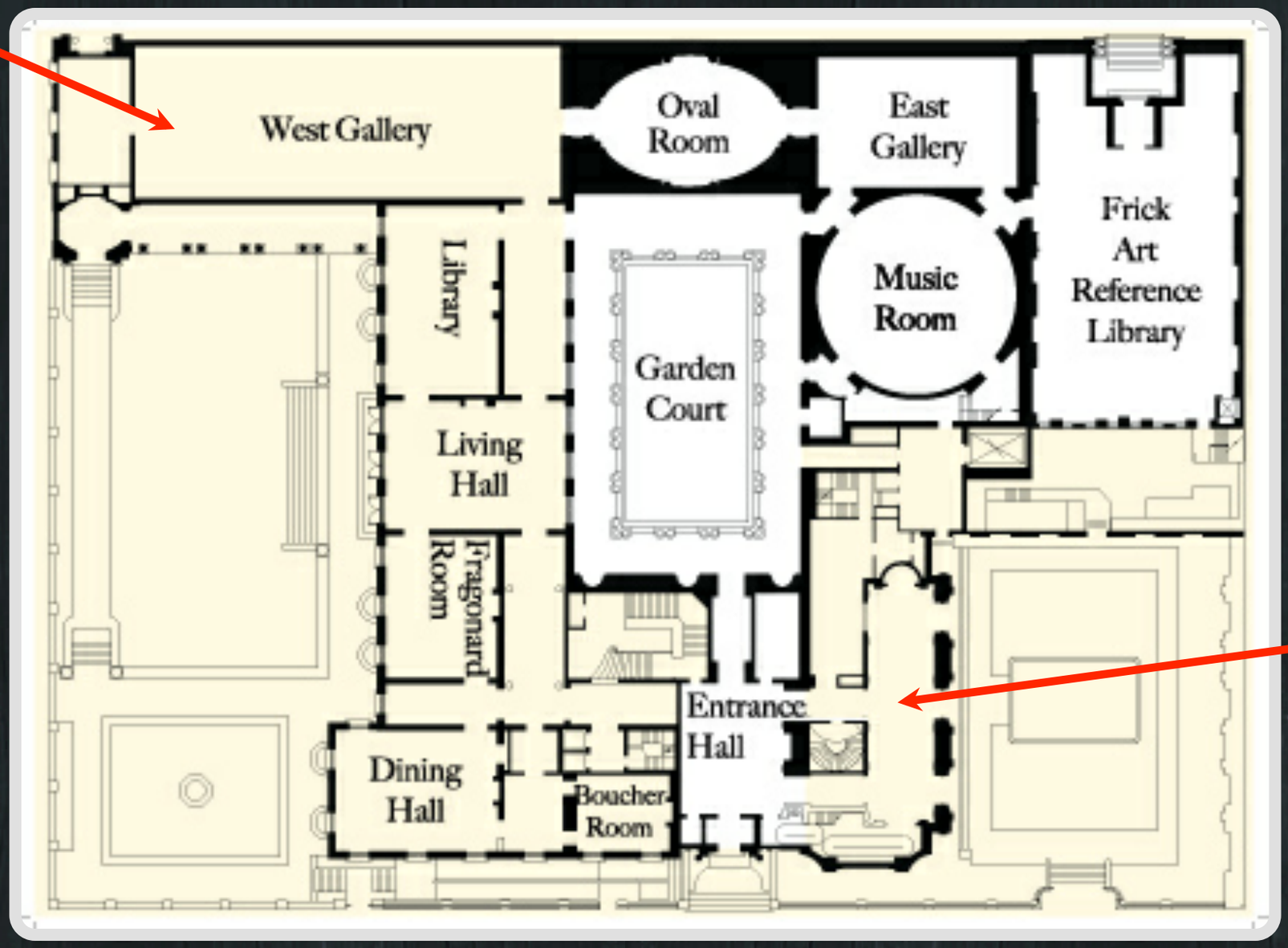
Hybrid Maps


West Gallery
Library
Living Hall
Reception Room
Oval
Garden Court
East
Music Room
Frick Art Reference Library
Dining Hall
Boucher Room
Entrance Hall

Ground vs. Aerial → Ground+Aerial

	Ground	Aerial	Ground+Aerial
Outdoor	 Google Streetview	 Google/Bing/NASA …	 Google MapsGL
Indoor	 Furukawa et al.	 This paper	 This paper

Enhance Navigation



Frick Collection Gallery (New York City)


5x
Google
© 2012 Google
Report a problem


The Frick Collection
Exit Street View
© 2012 Google
Google earth

Take Home Message

Take Home Message

Technical Contribution

Conceptual Contribution

Take Home Message

Technical Contribution

Inverse CSG for
Large-scale Reconstruction

Conceptual Contribution

Take Home Message

Technical Contribution

Inverse CSG for
Large-scale Reconstruction

Conceptual Contribution

Indoor Photorealistic Maps
+ Aerial → Ground Transition
for Effective Navigation

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Let machine see

Visualization

Let human see better

Goals of Computer Vision

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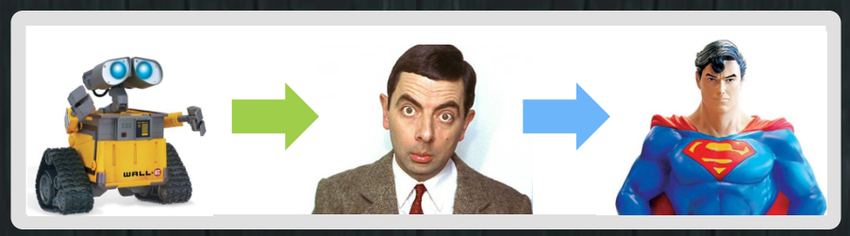
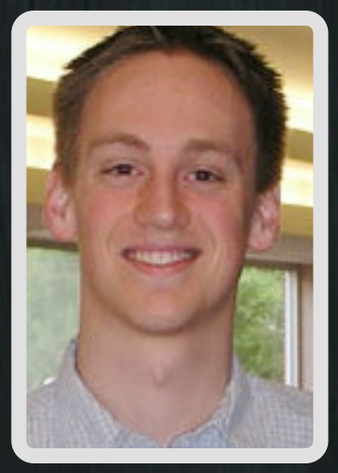
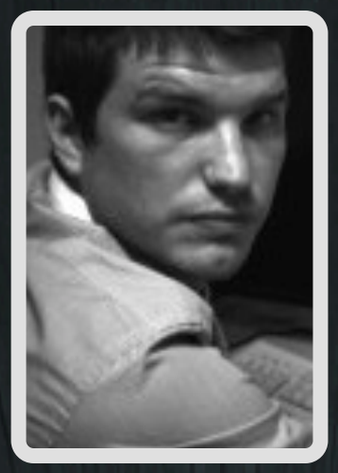


Acknowledgements

Steve Seitz



David Gallup

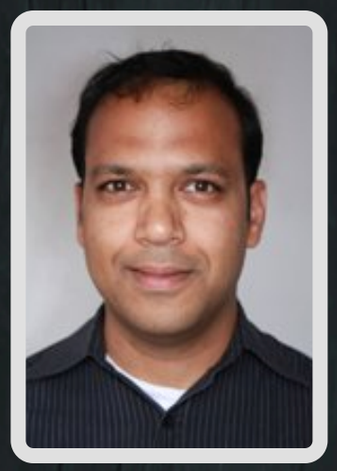


Carlos Hernandez

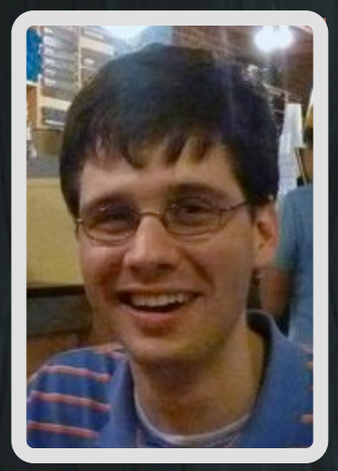
Art Project

Research at Google

PhD Fellowship Program



Maneesh Agrawala



Andrew Owens

Antonio Torralba

Conference Travel Grant

Visart
Where Computer Vision Meets Art

Where is “the birth of venus”? Google Art from bird’s-eye view

Jianxiong Xiao

Time: Friday Oct 12, 2:30PM

Location: Room C 2F Affari

Demo: http://mit.edu/jxiao/museum/