



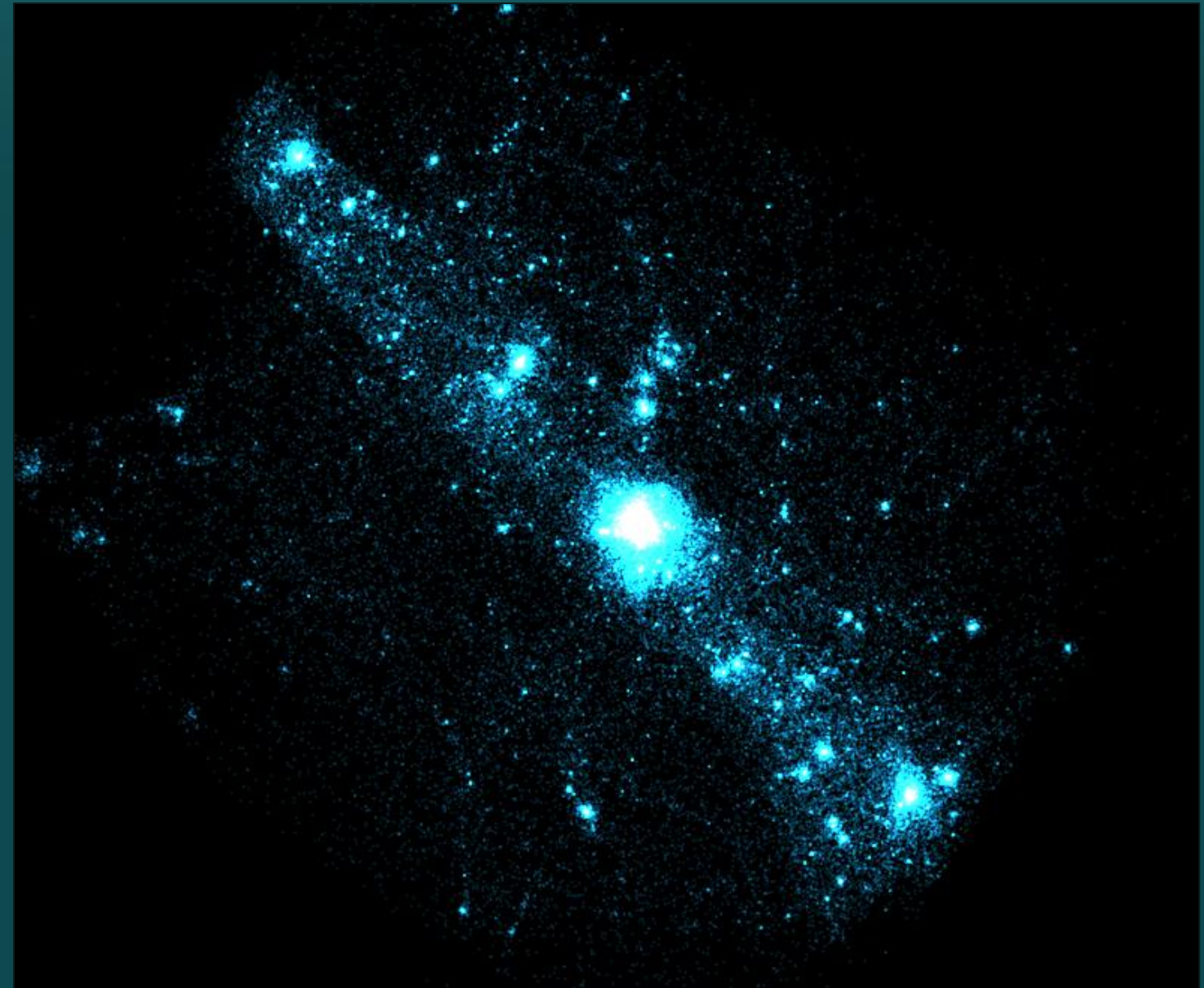
Efficient Structure-Aware Selection Techniques for 3D Point Cloud Visualizations with 2DOF Input



Lingyun Yu Konstantinos Efsthathiou
Petra Isenberg Tobias Isenberg

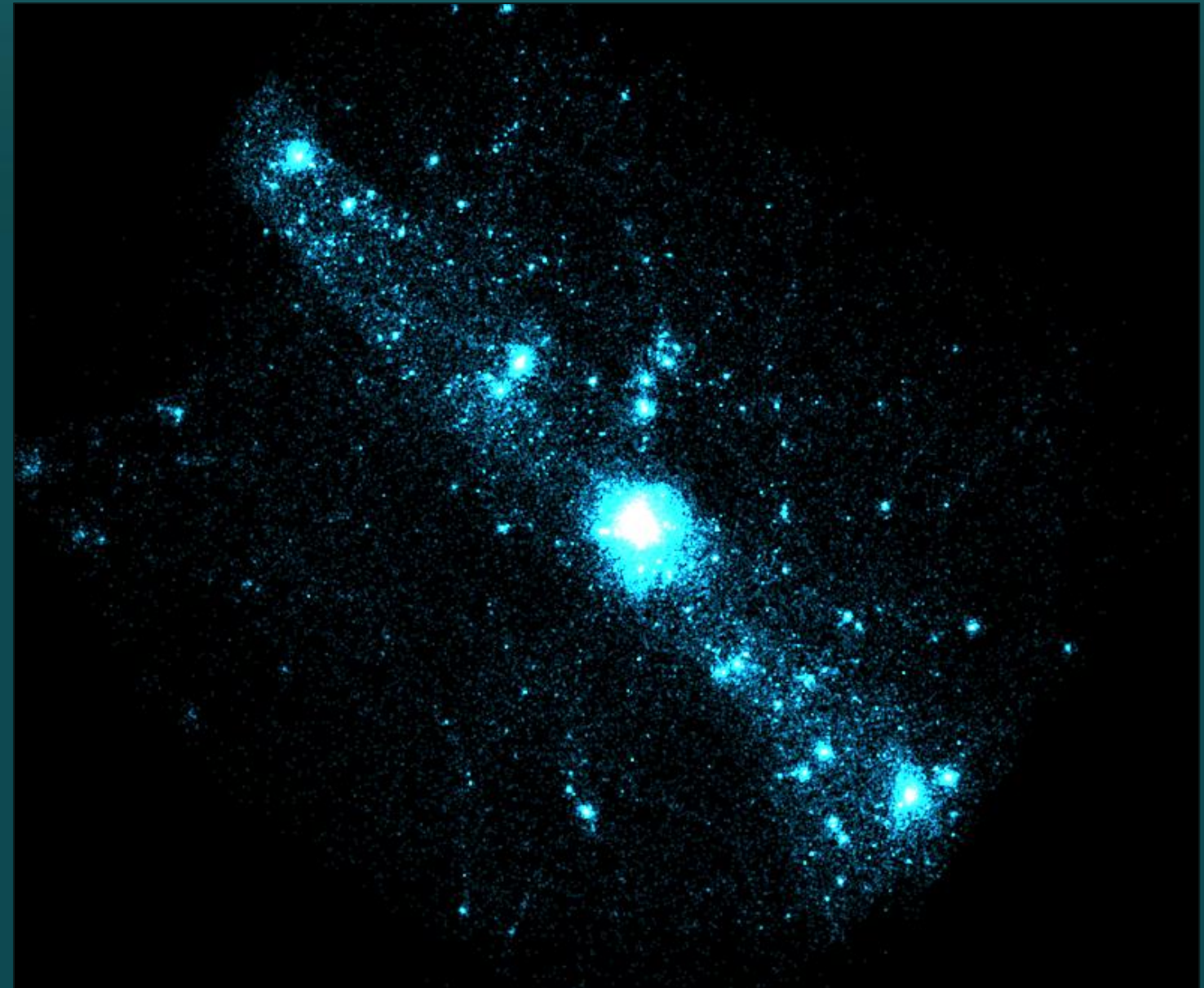
The Problem: Selection of 3D Subspaces

- 3D spatial data—basis of many visualization research questions
- **problem:** why/how to efficiently select subspaces in 3D?



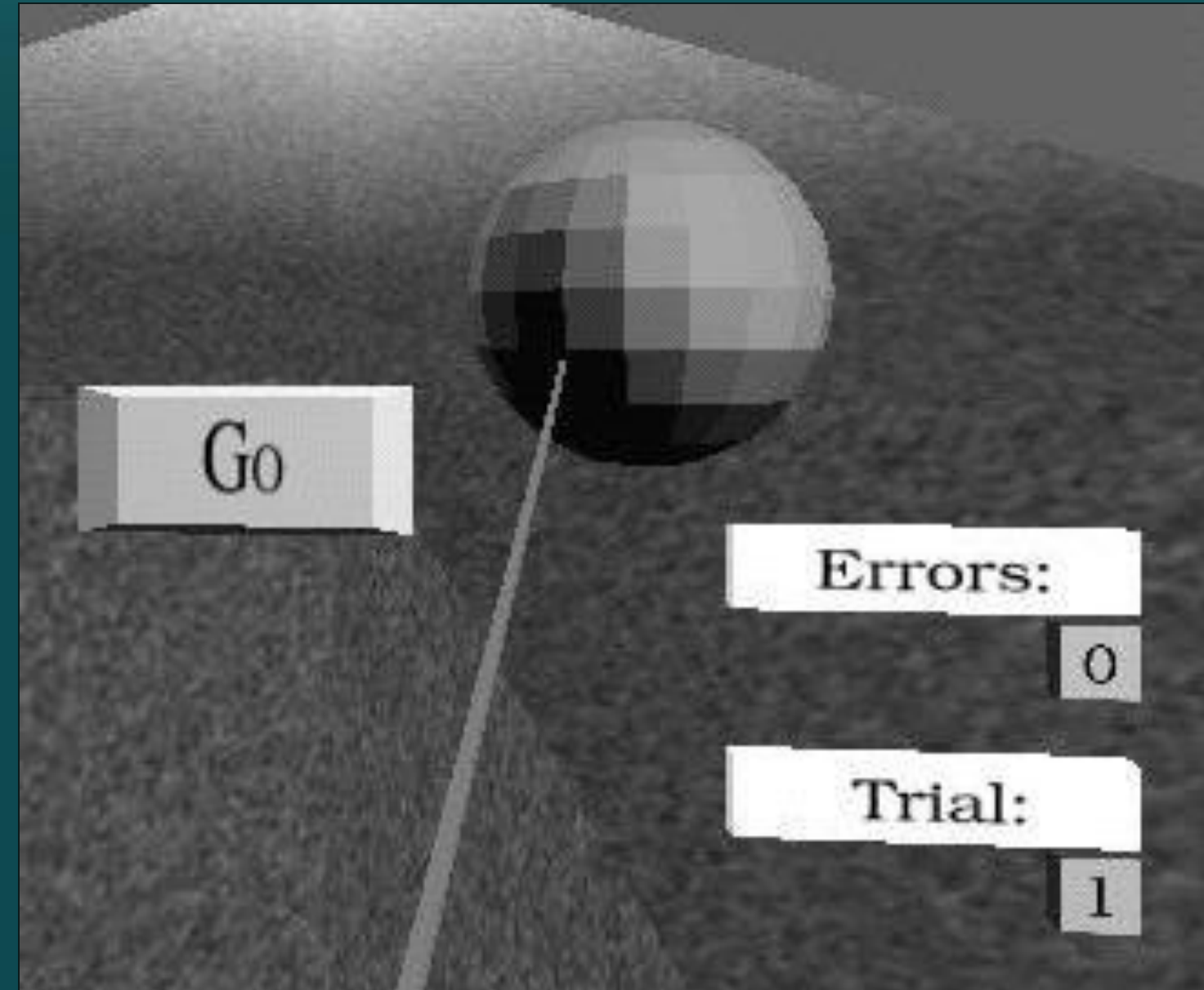
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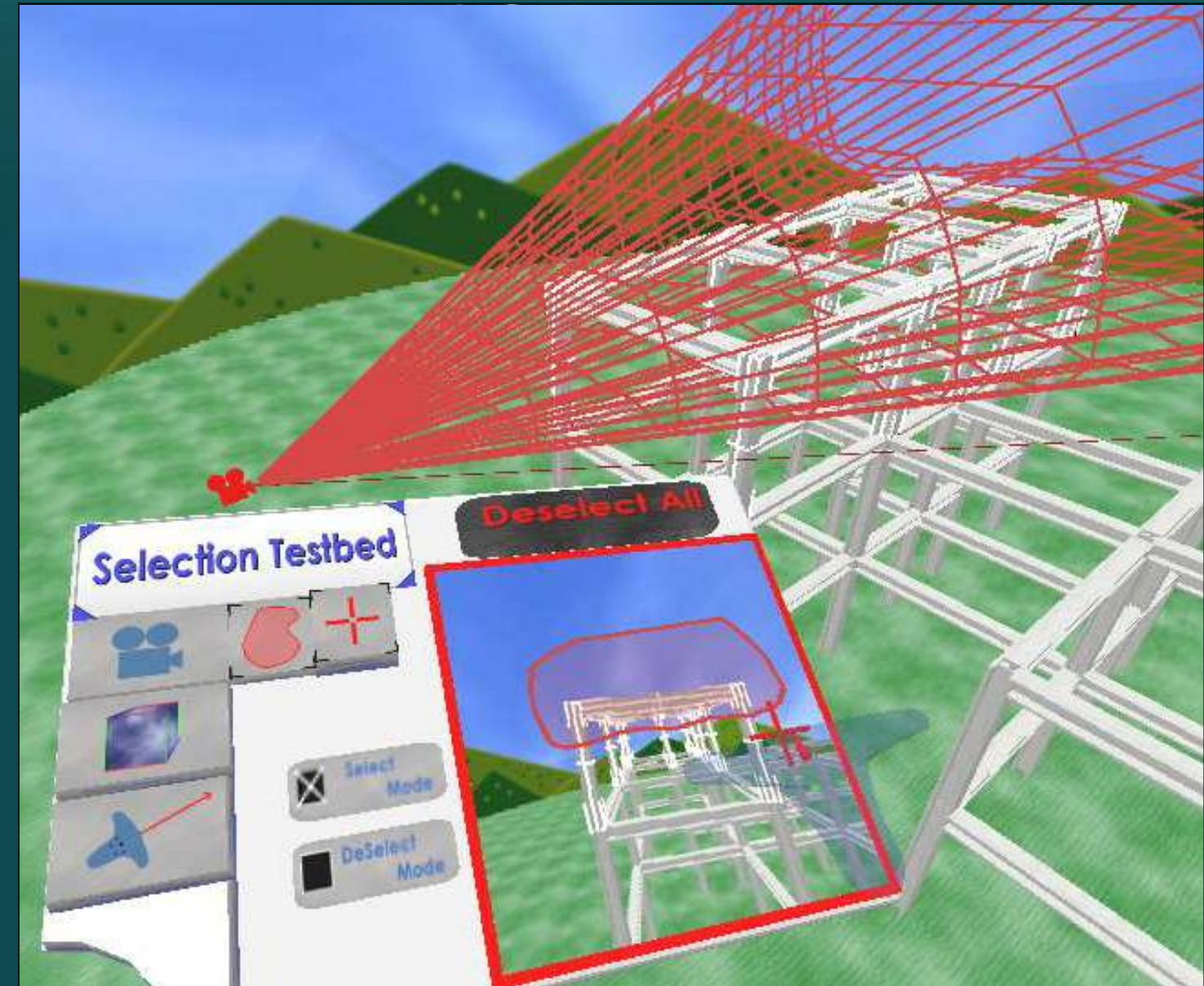
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[Wingrave & Bowman, 2005]

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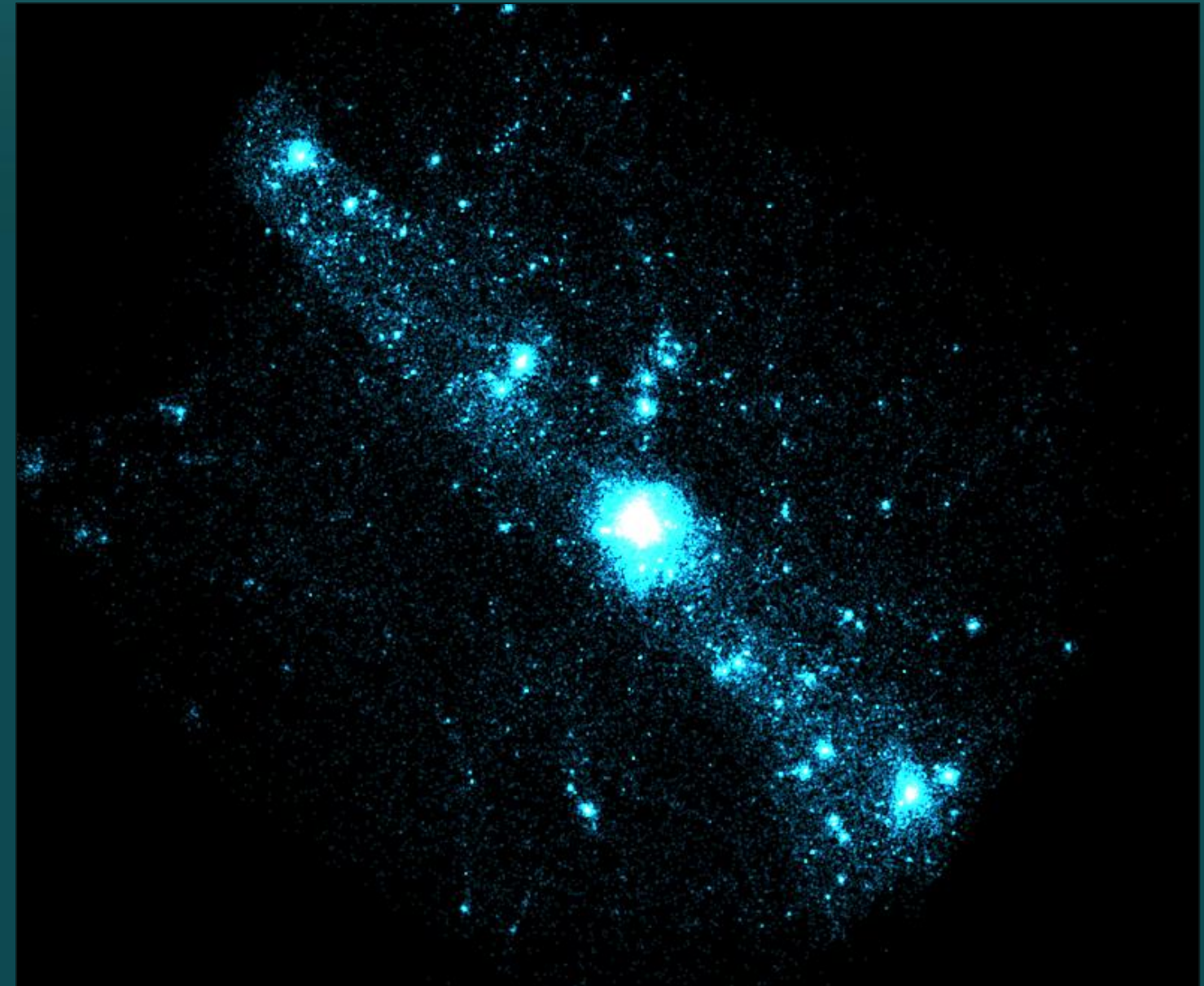
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[Lucas & Bowman, 2005]

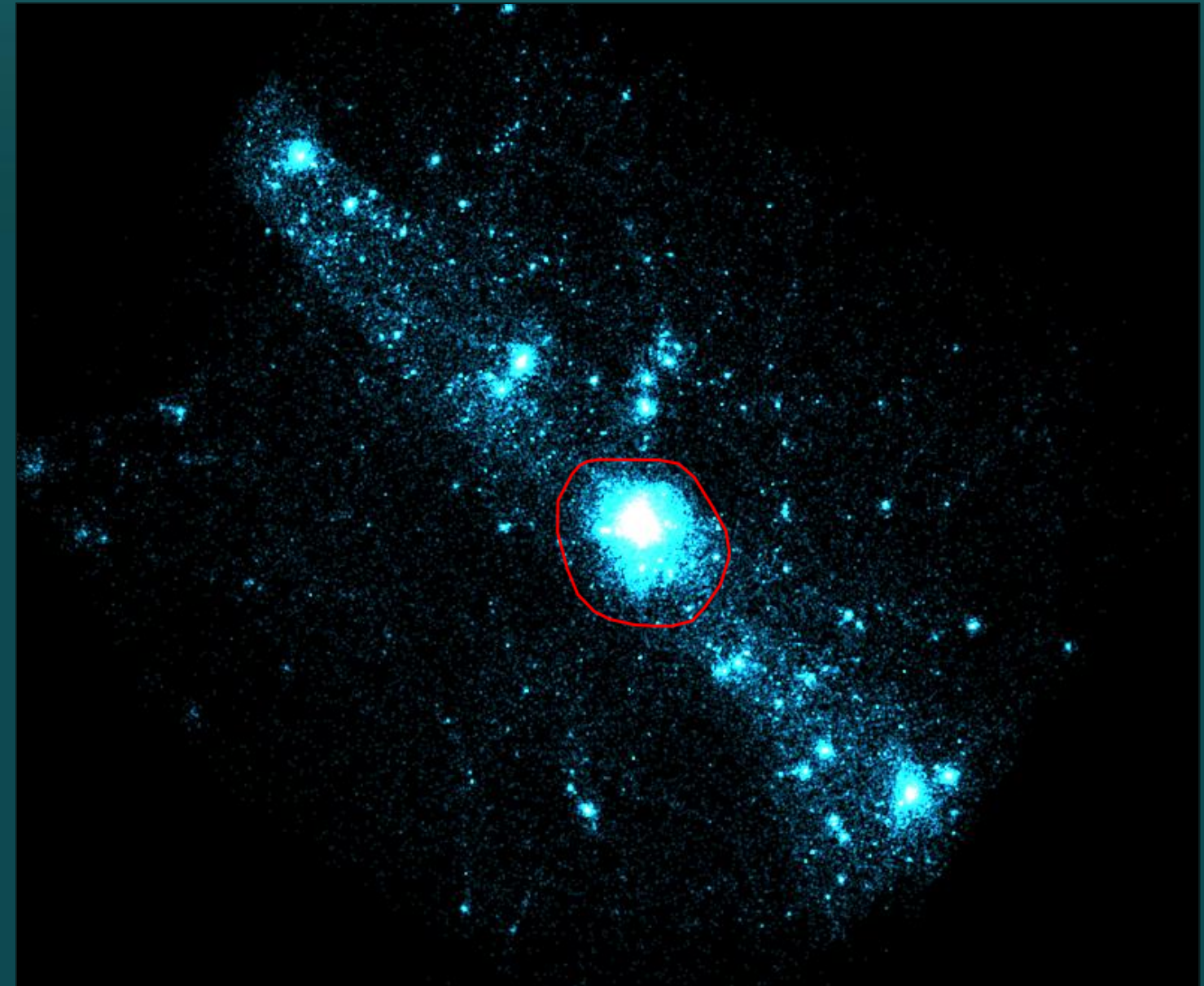
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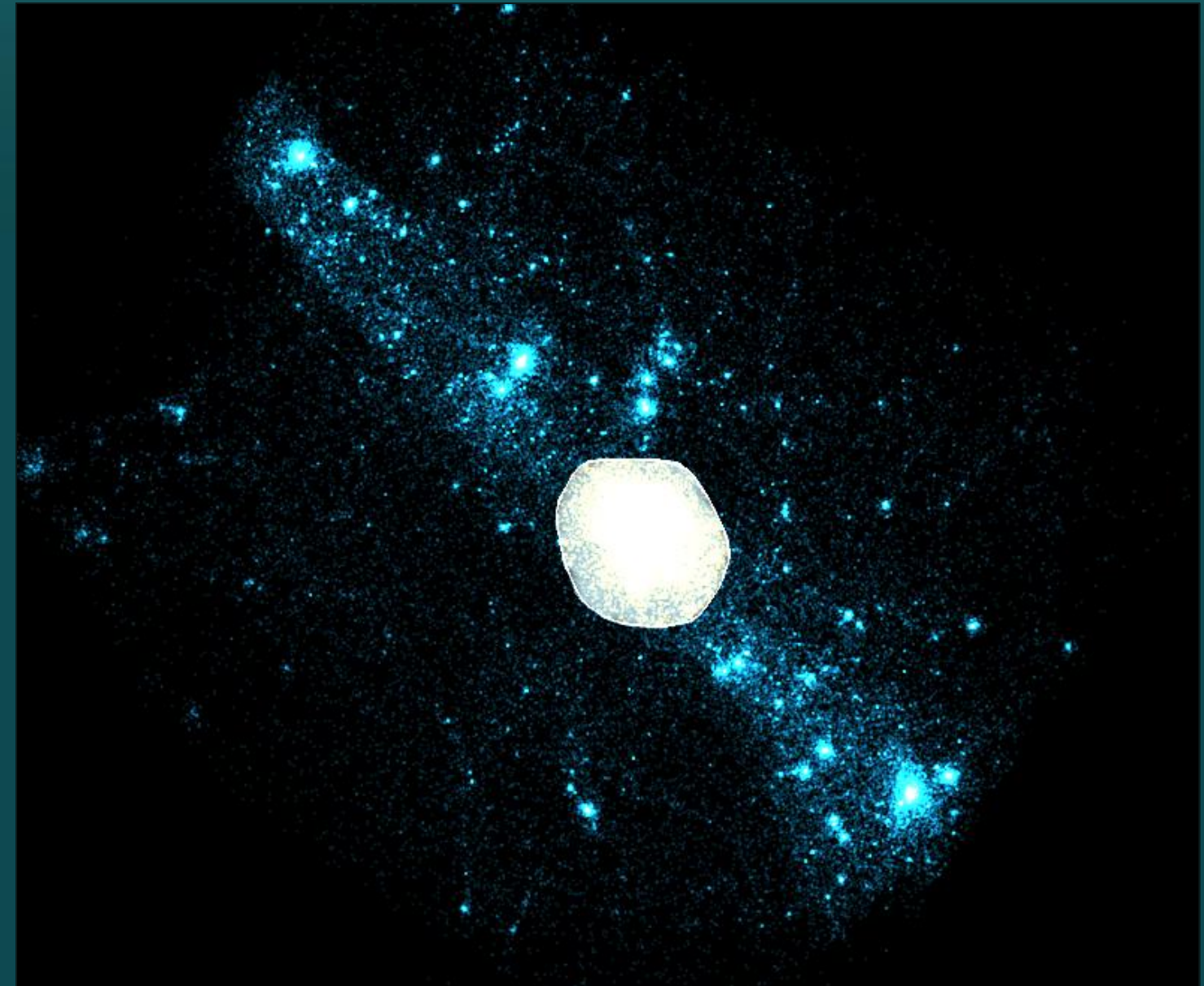
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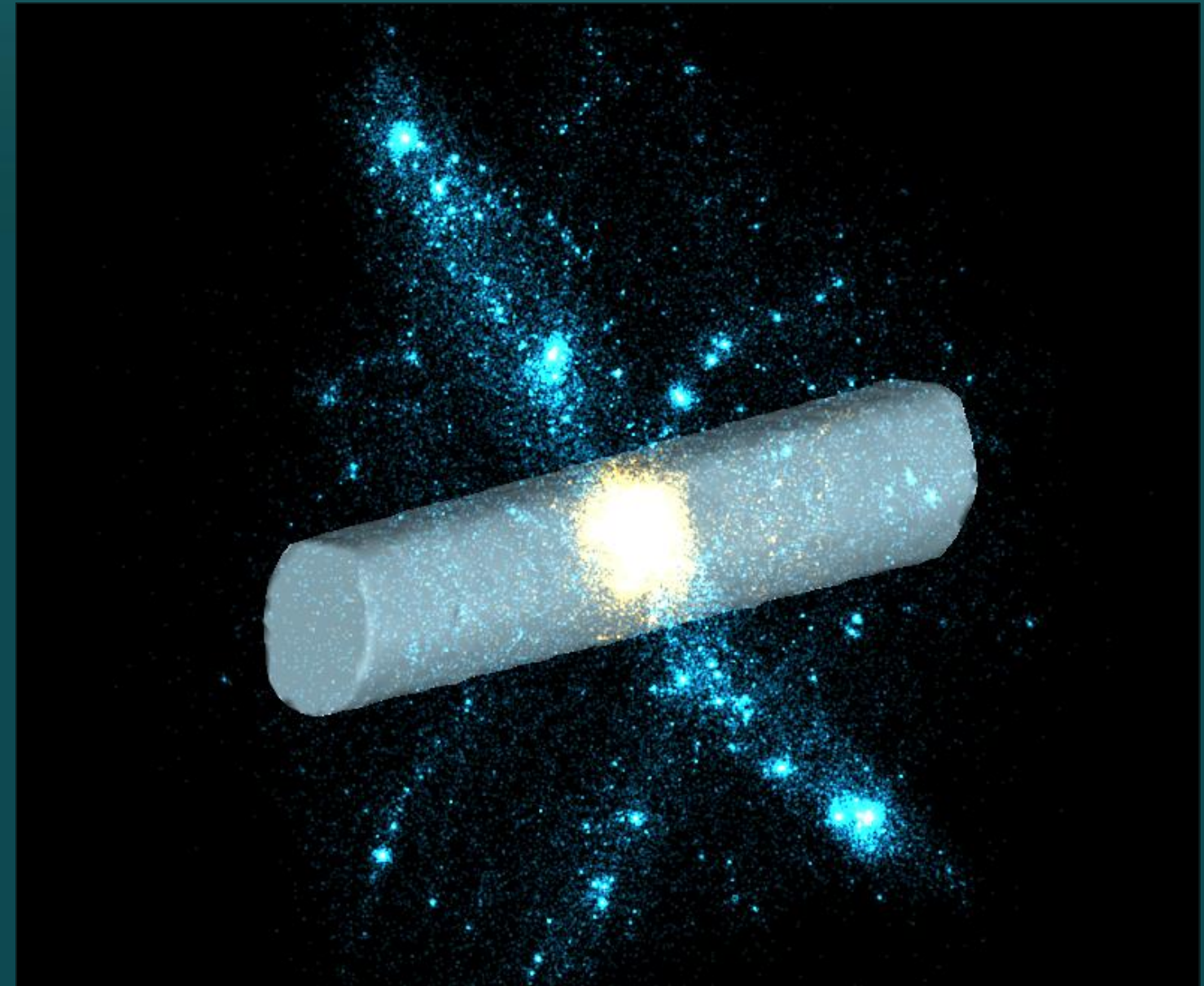
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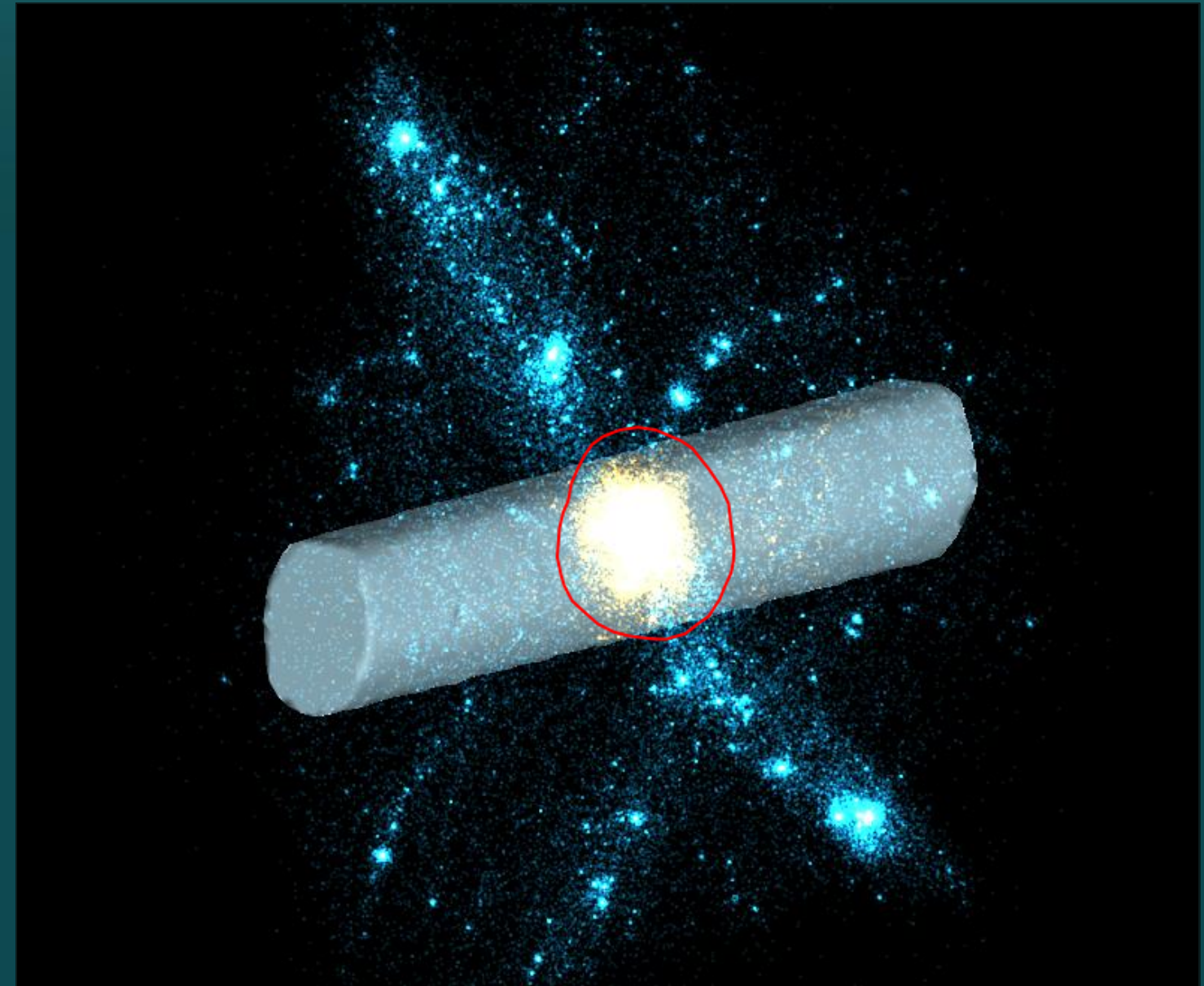
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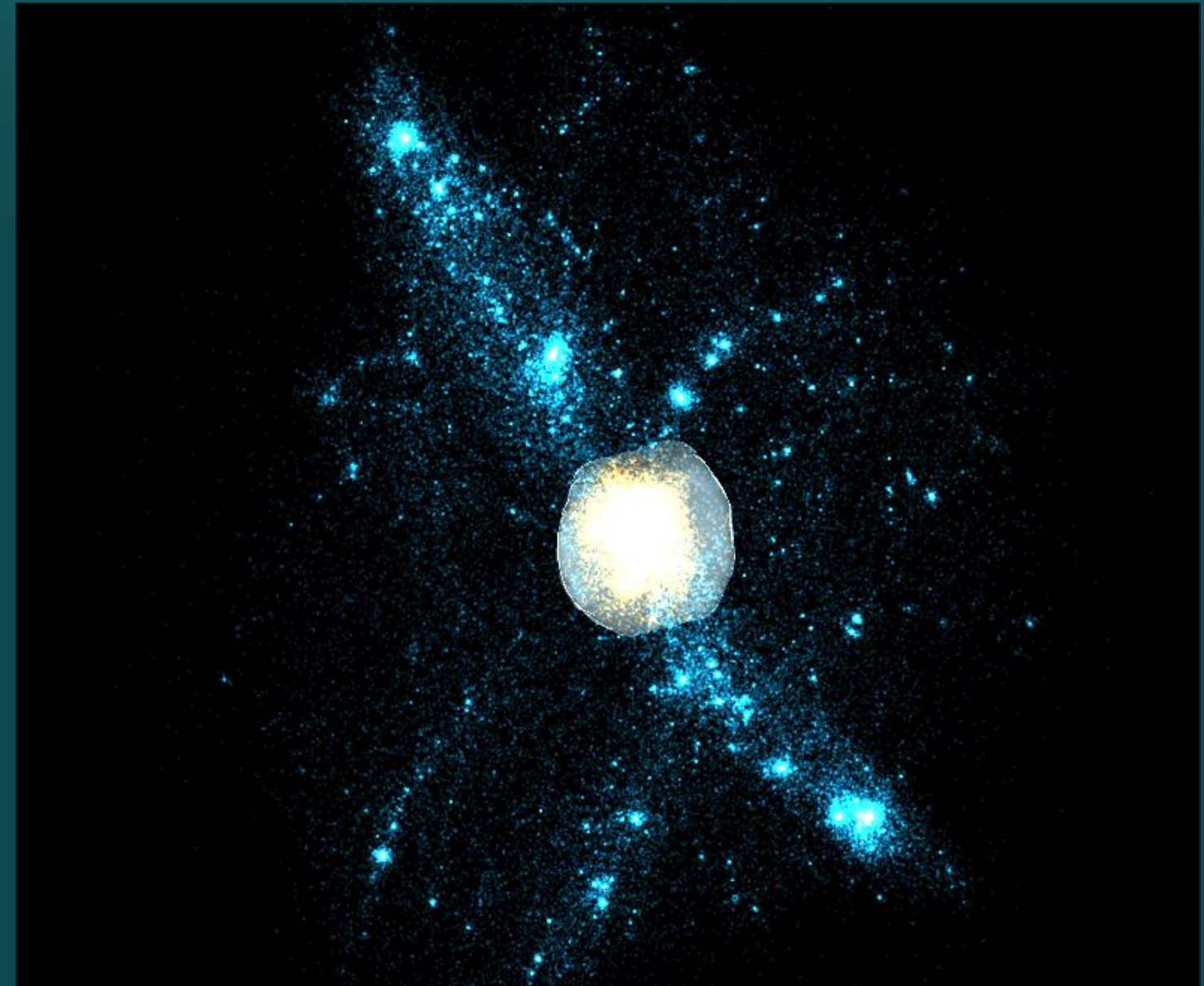
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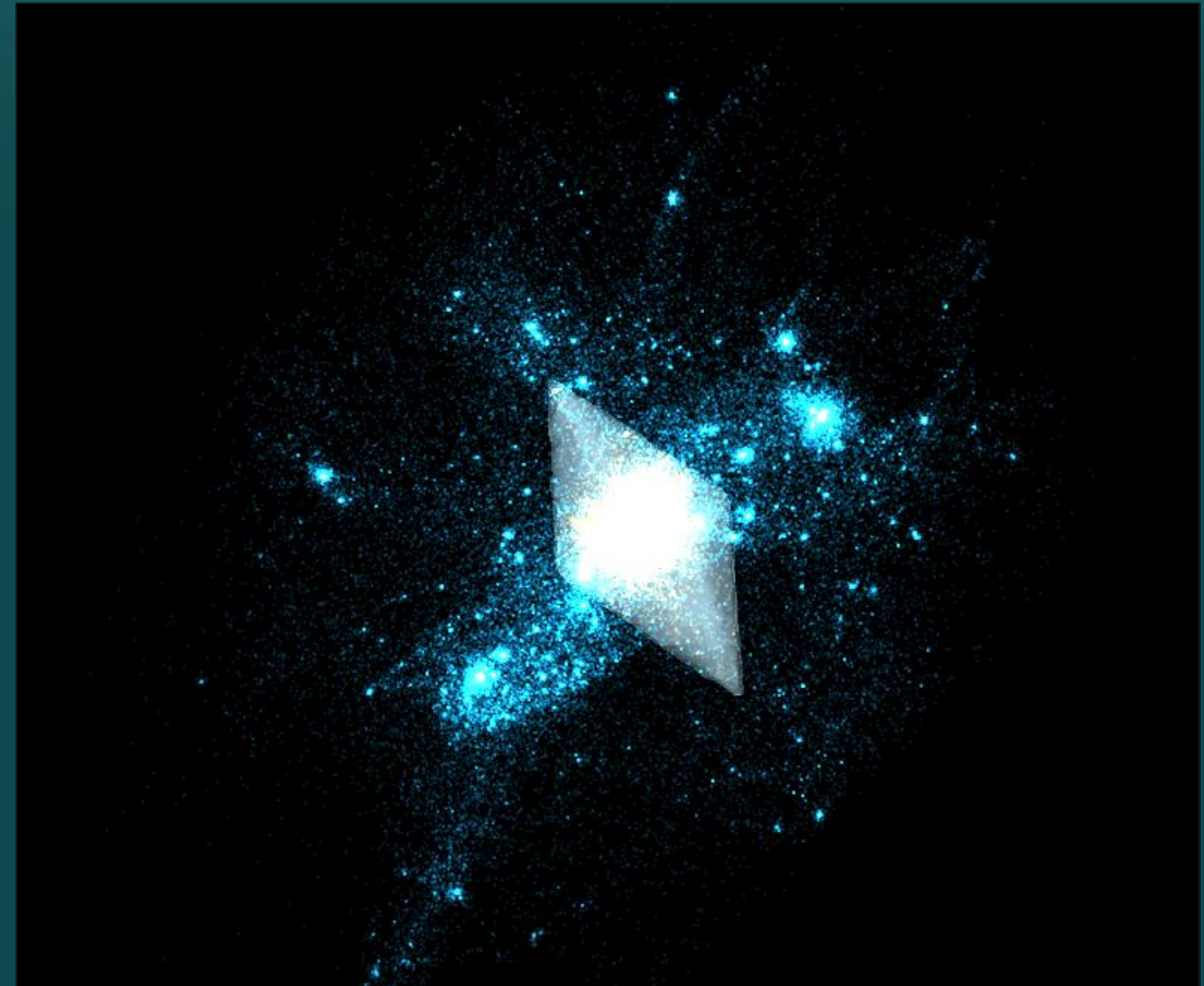
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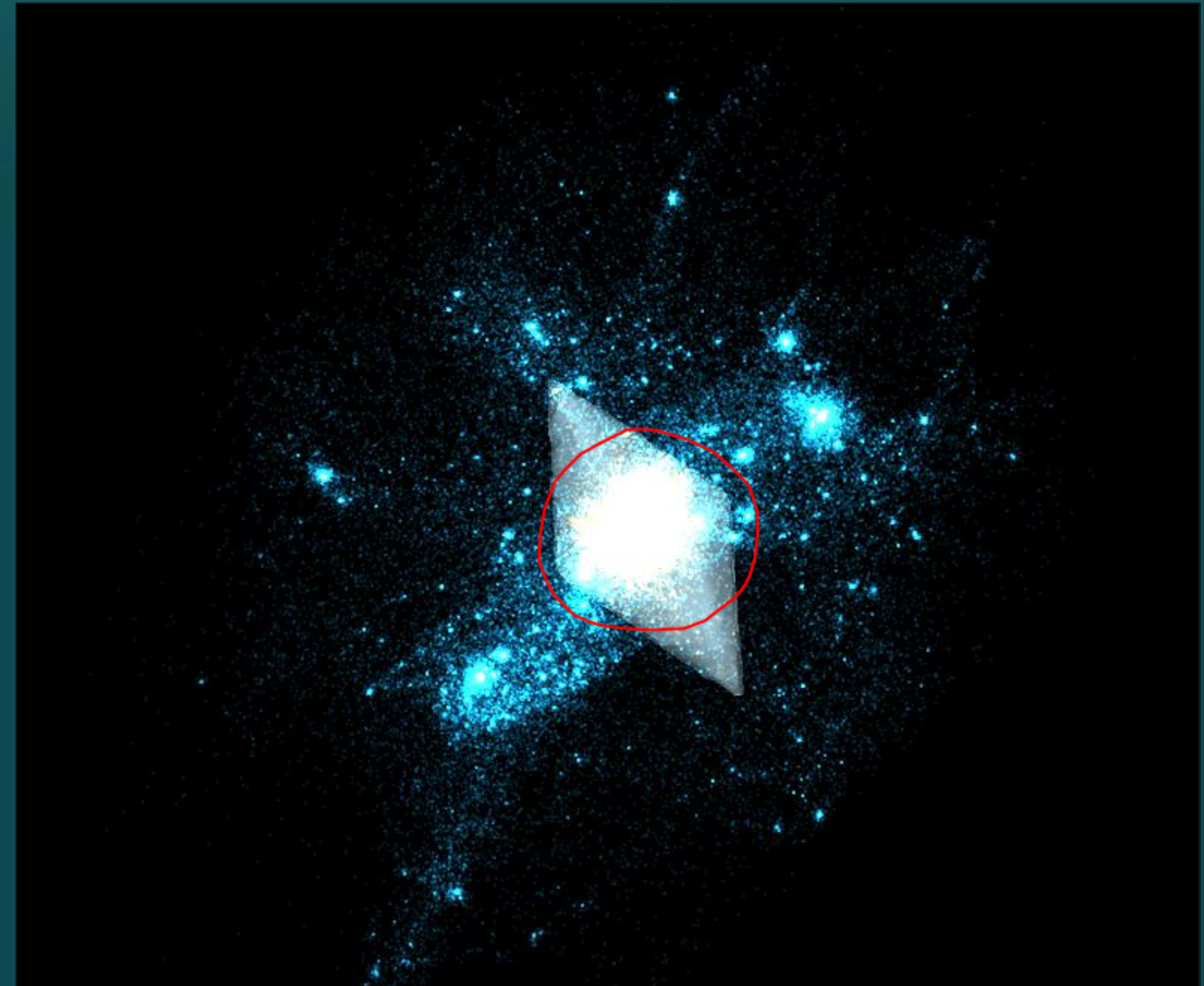
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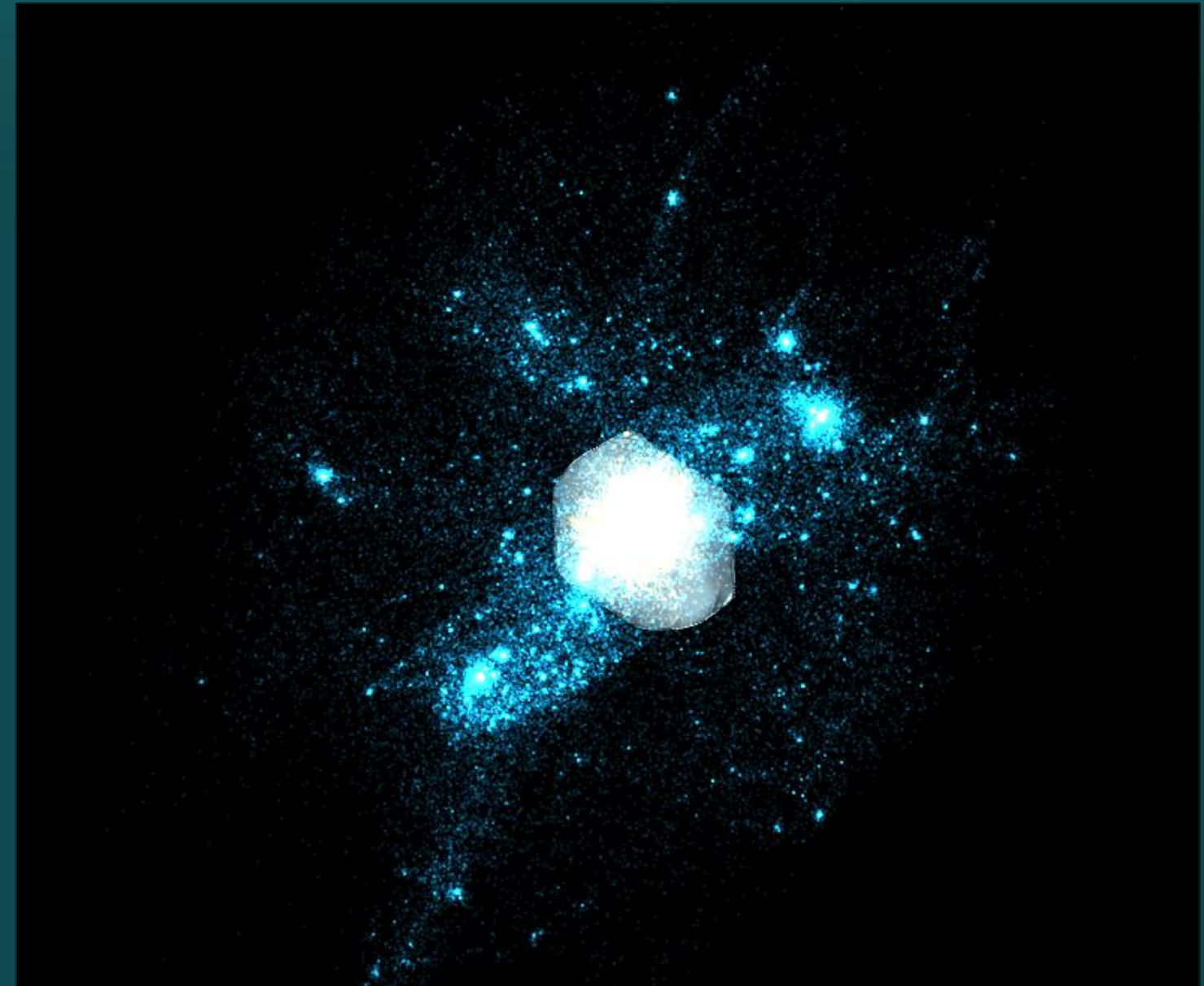
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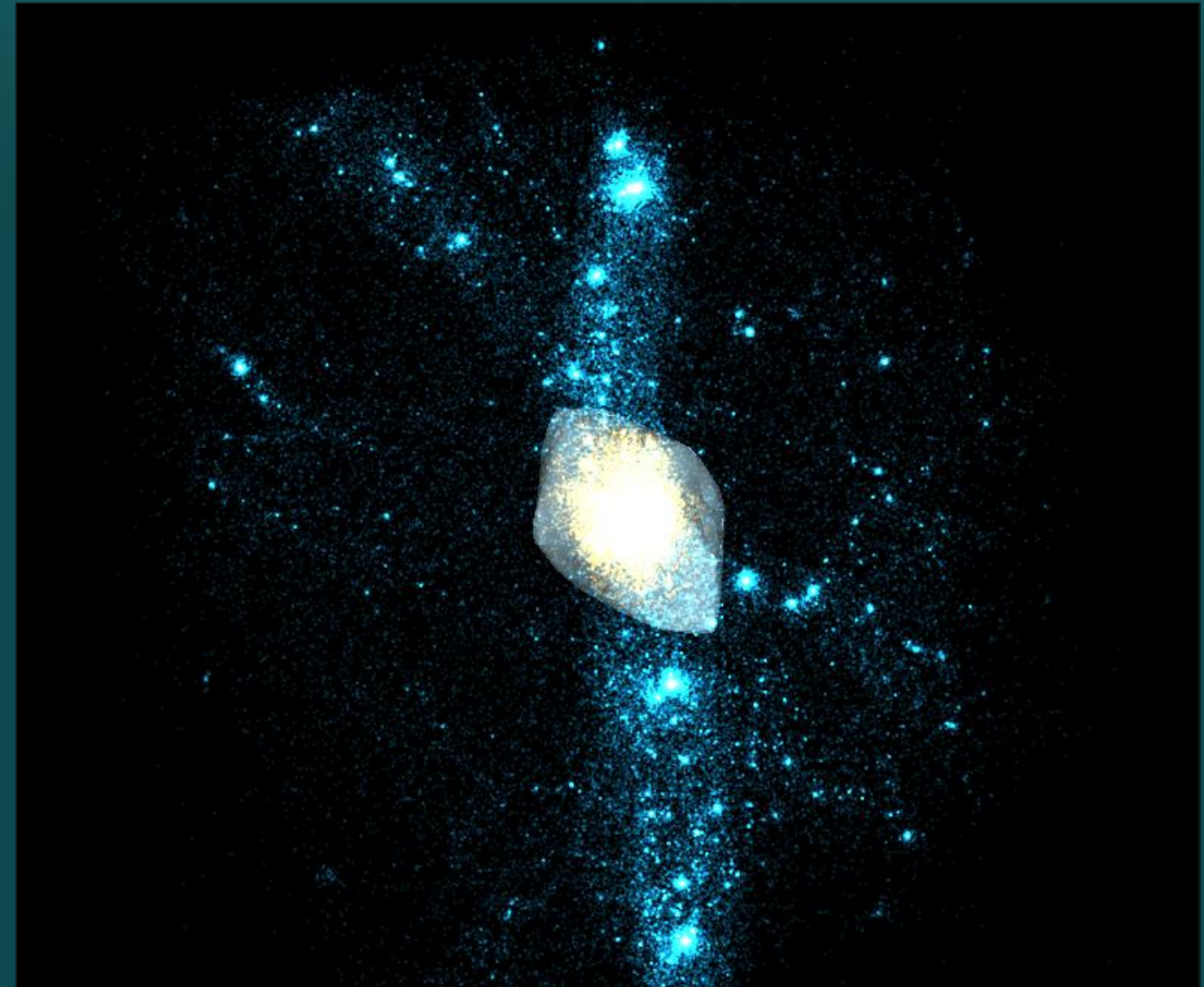
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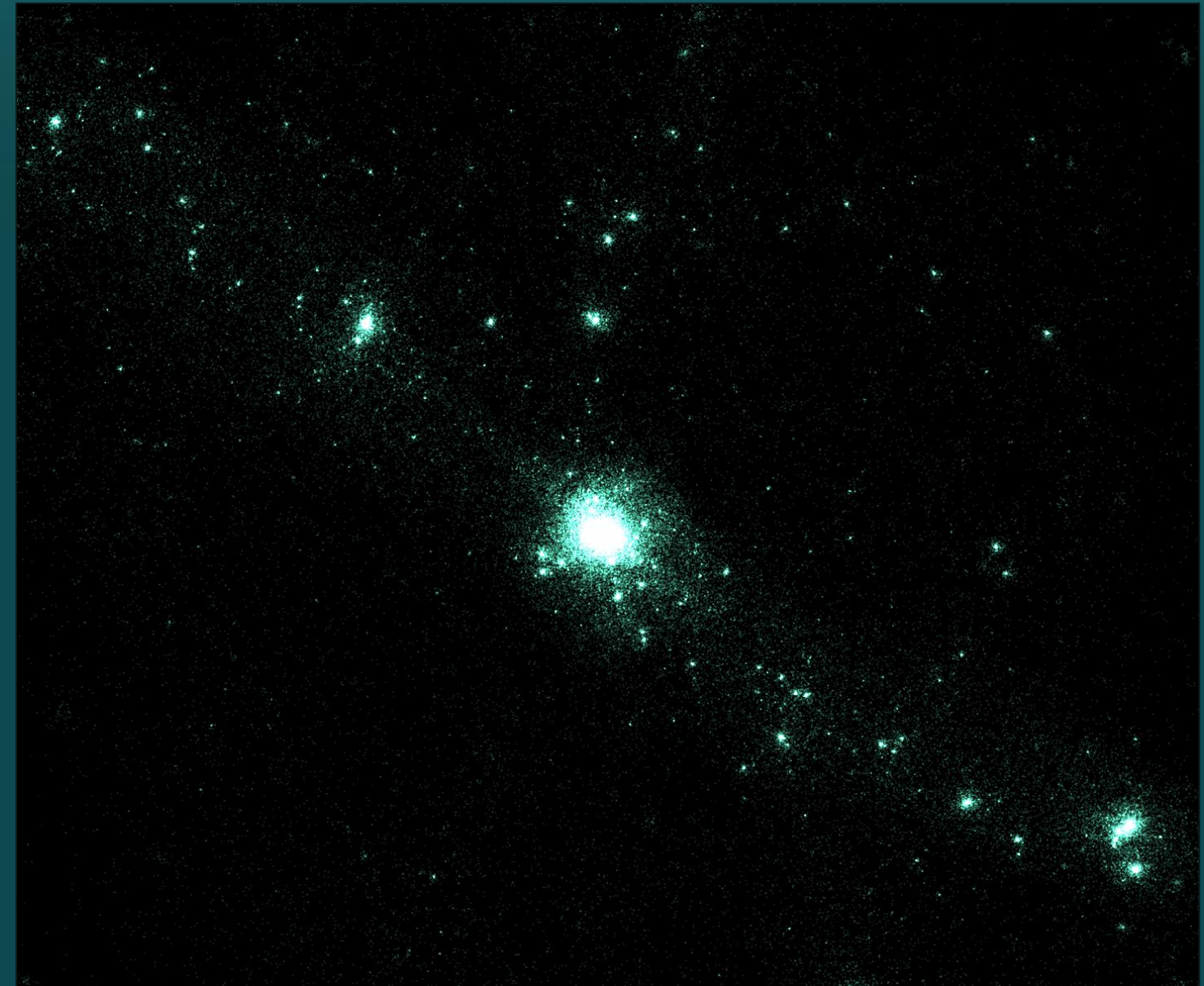
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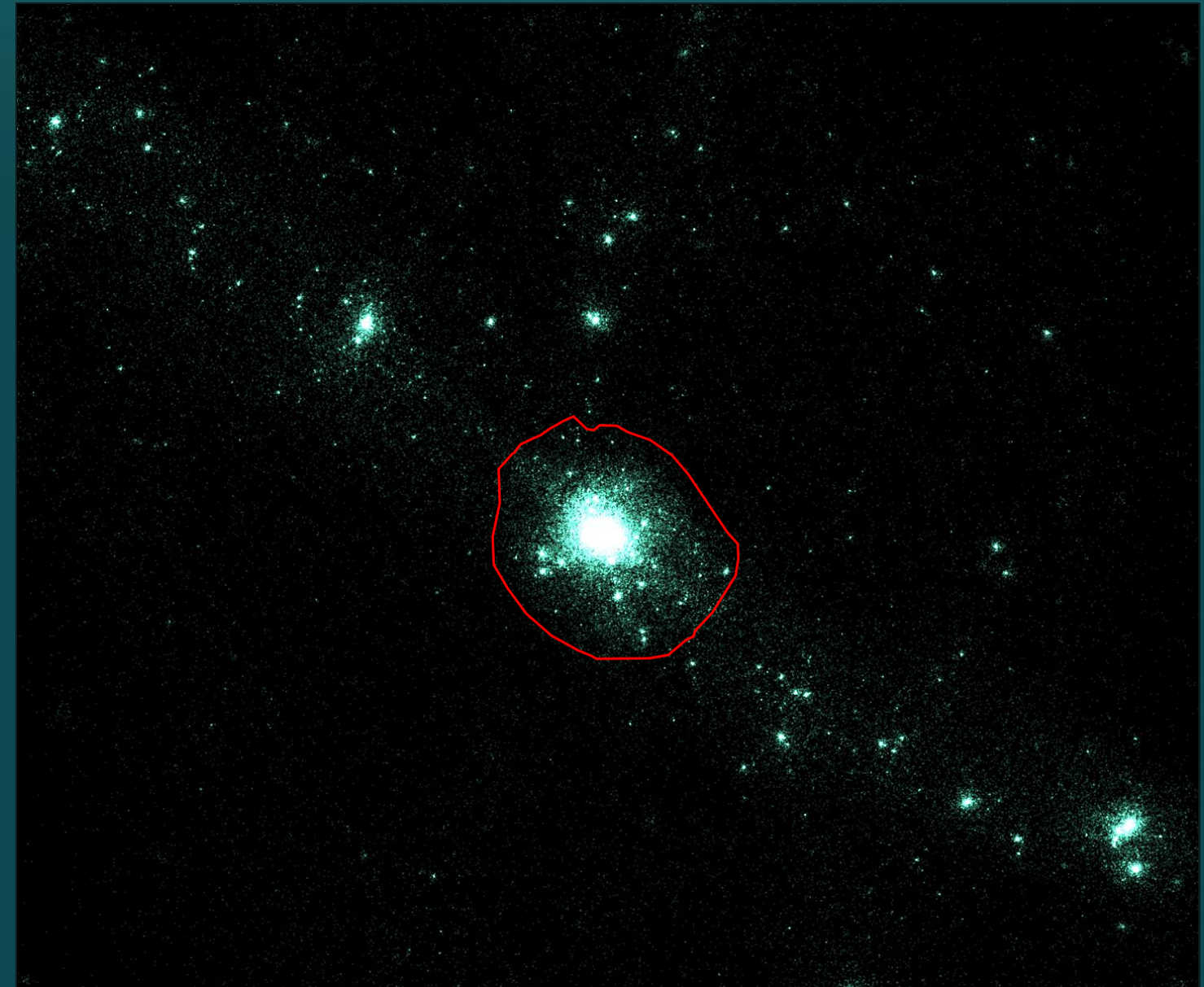
A New Interactive Selection Technique

- ***spatial selection*** rather than object-based selection
- ***two-dimensional input*** (PC, touch displays)
- ***2D lasso*** interaction: ***intended*** selection
- ***structure-aware selection*** in 3D depth



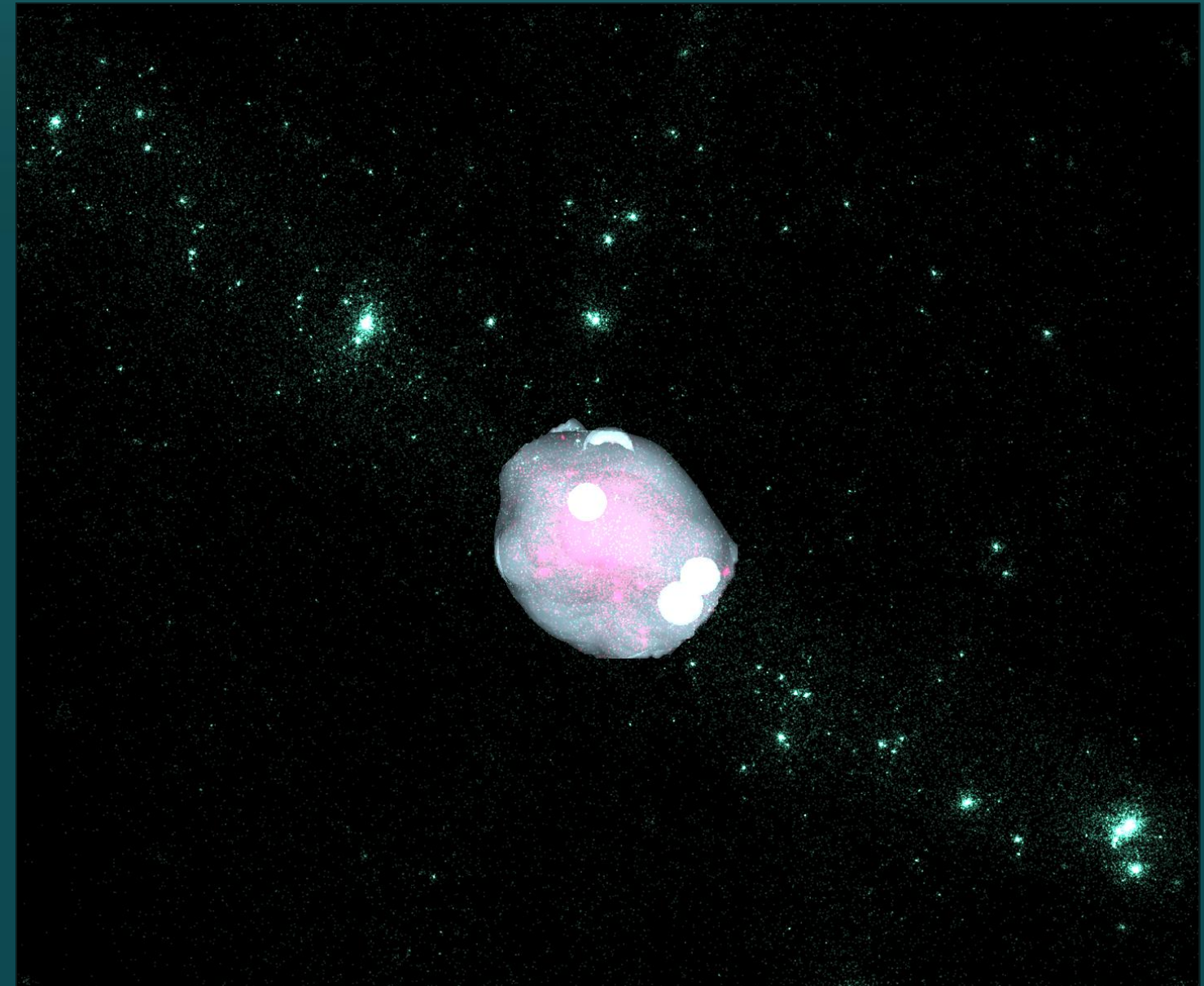
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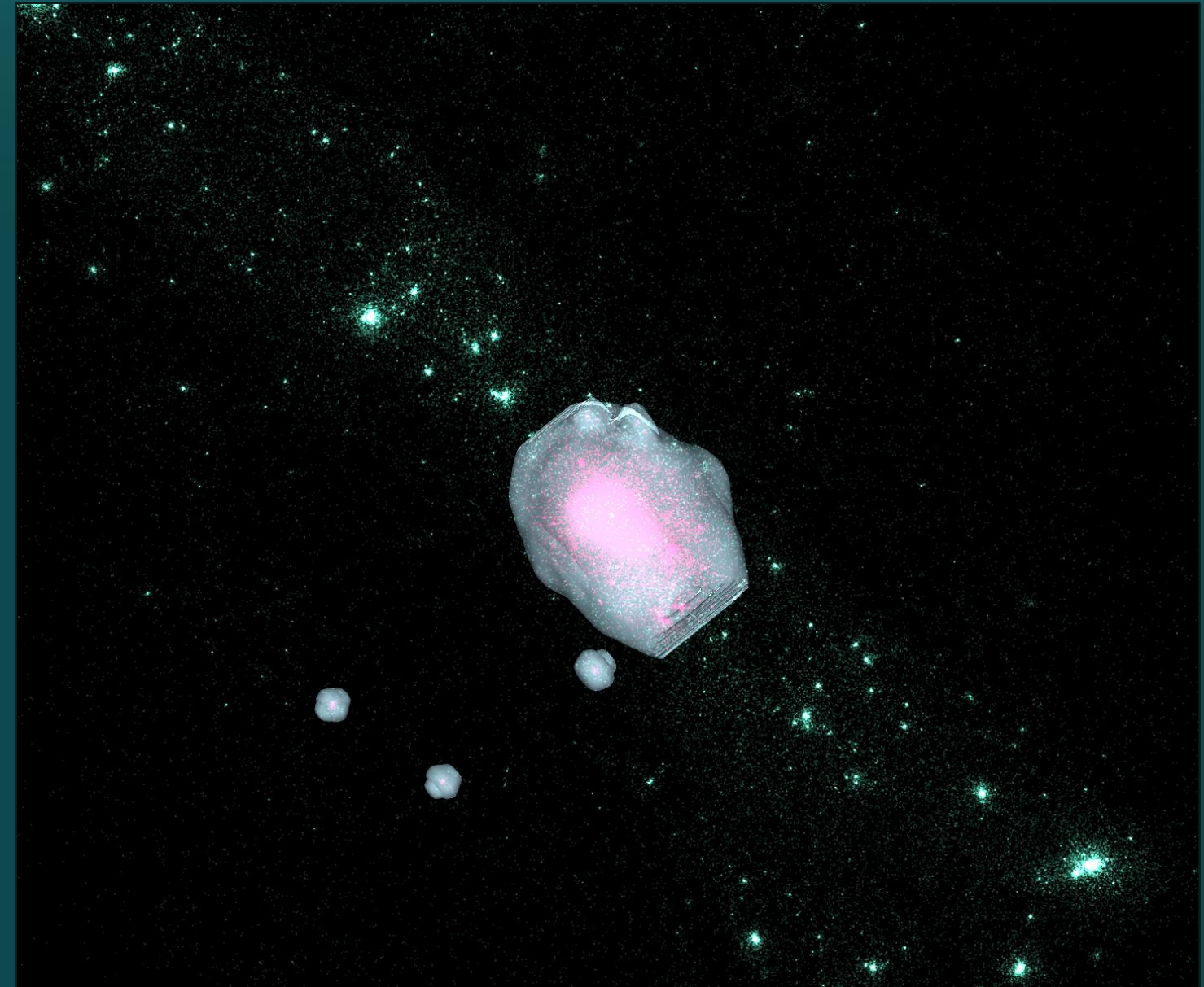
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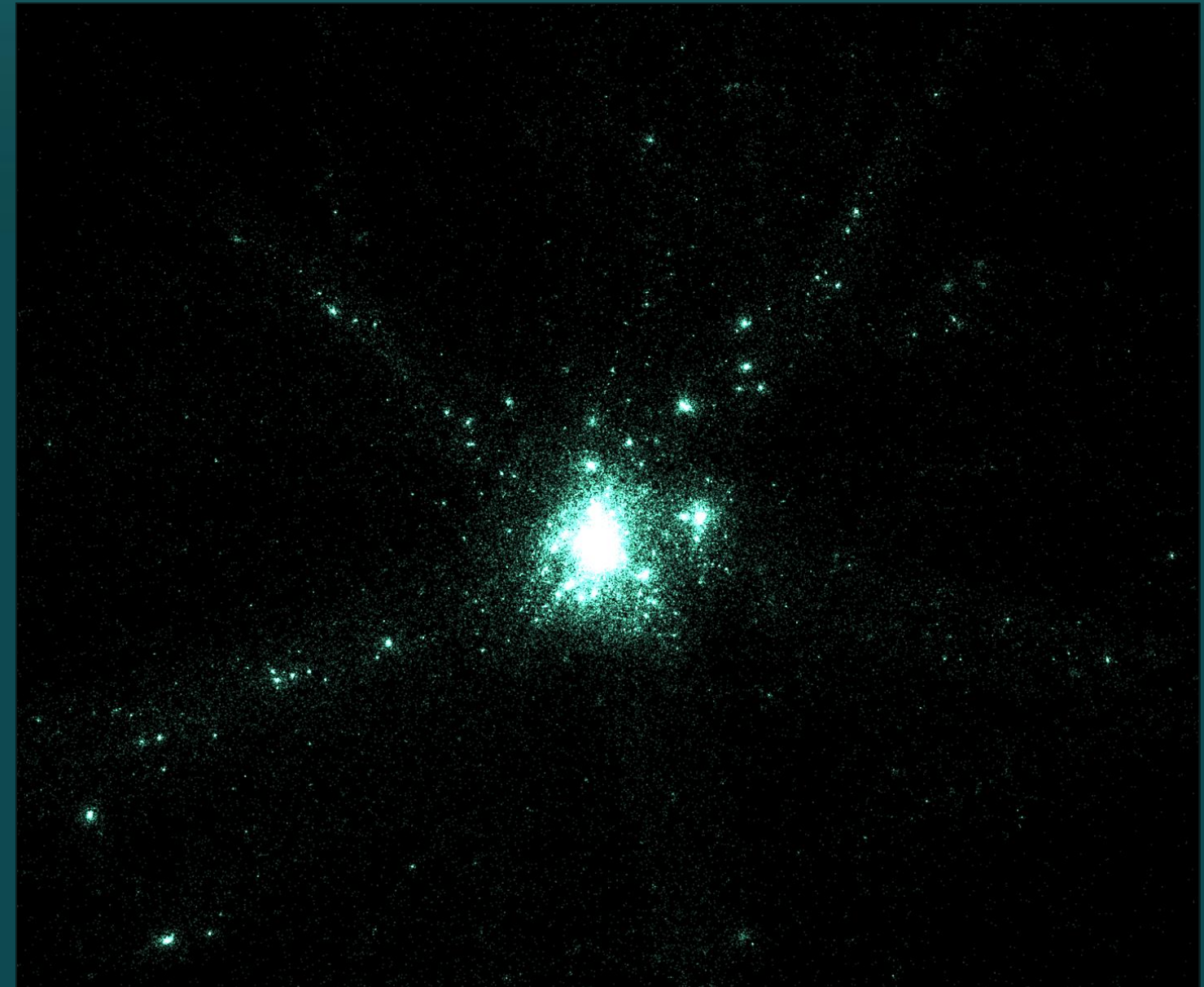
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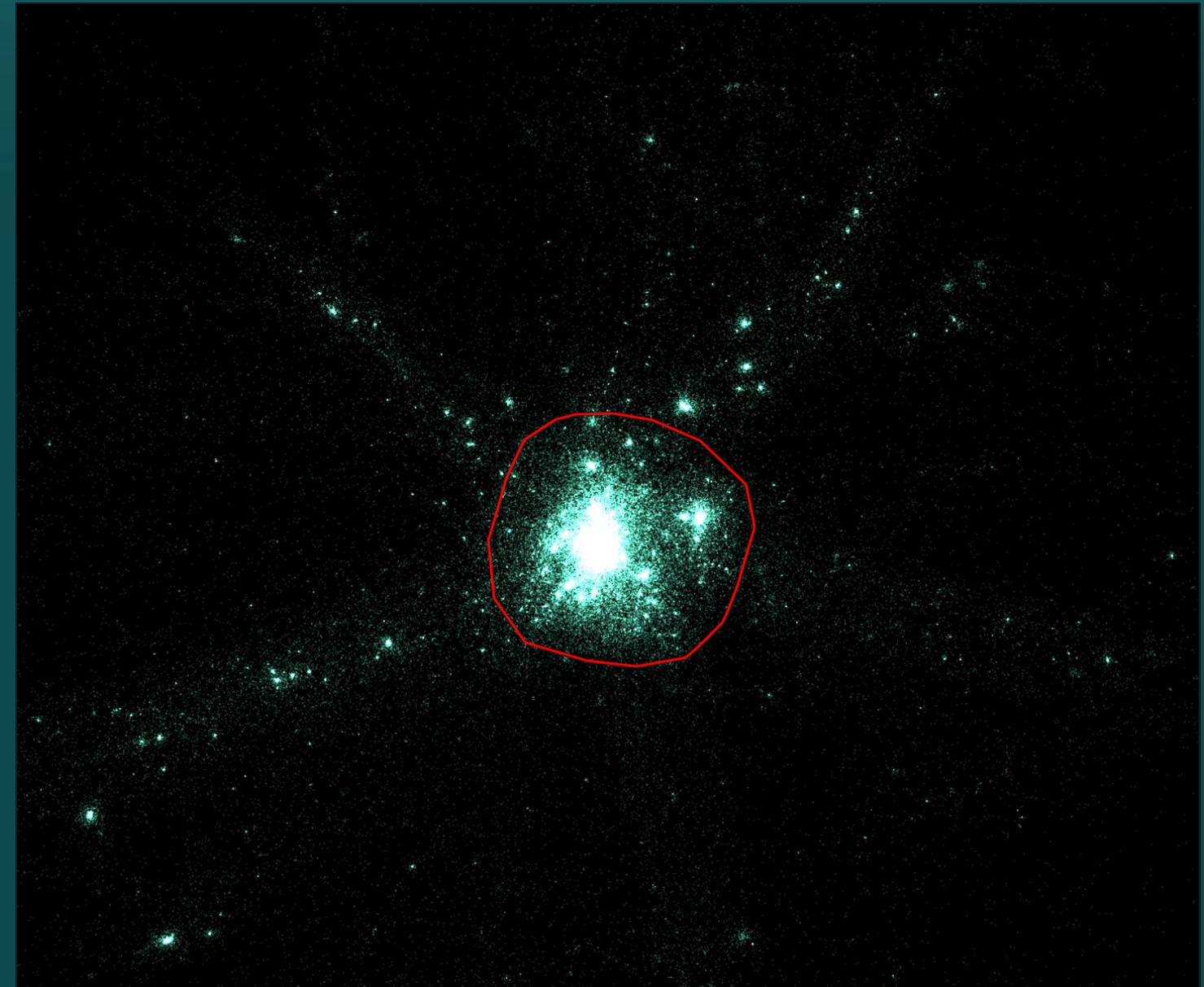
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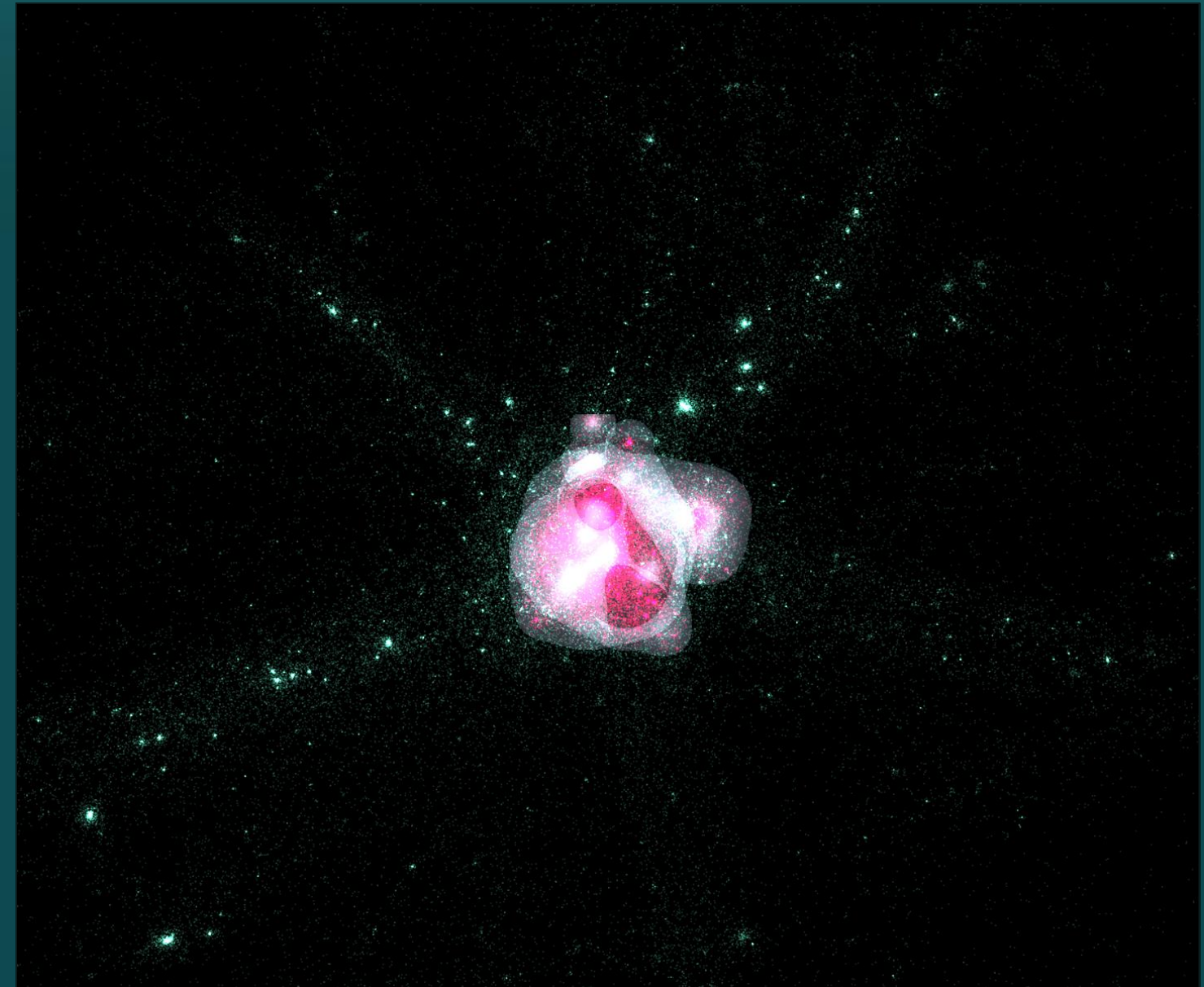
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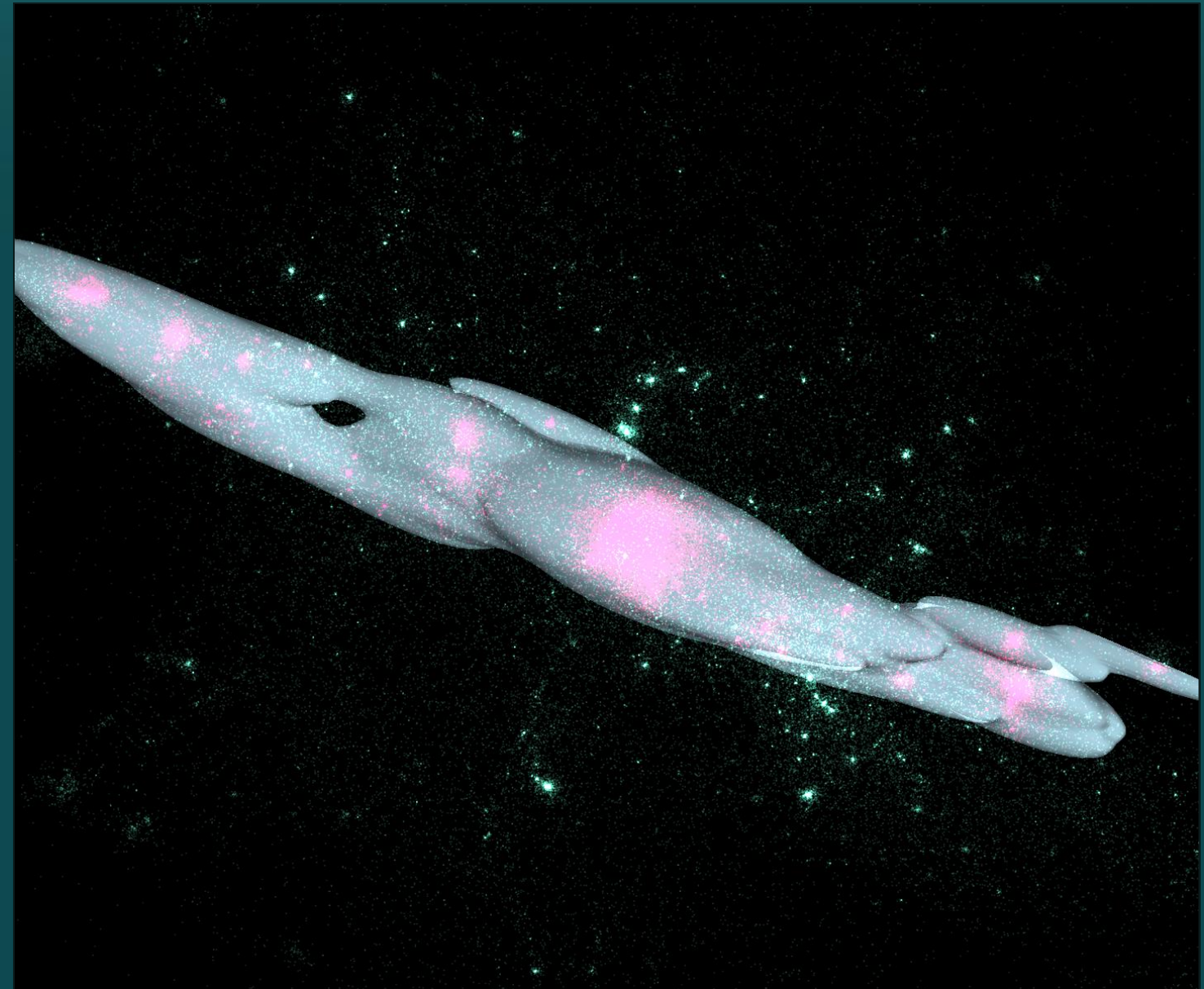
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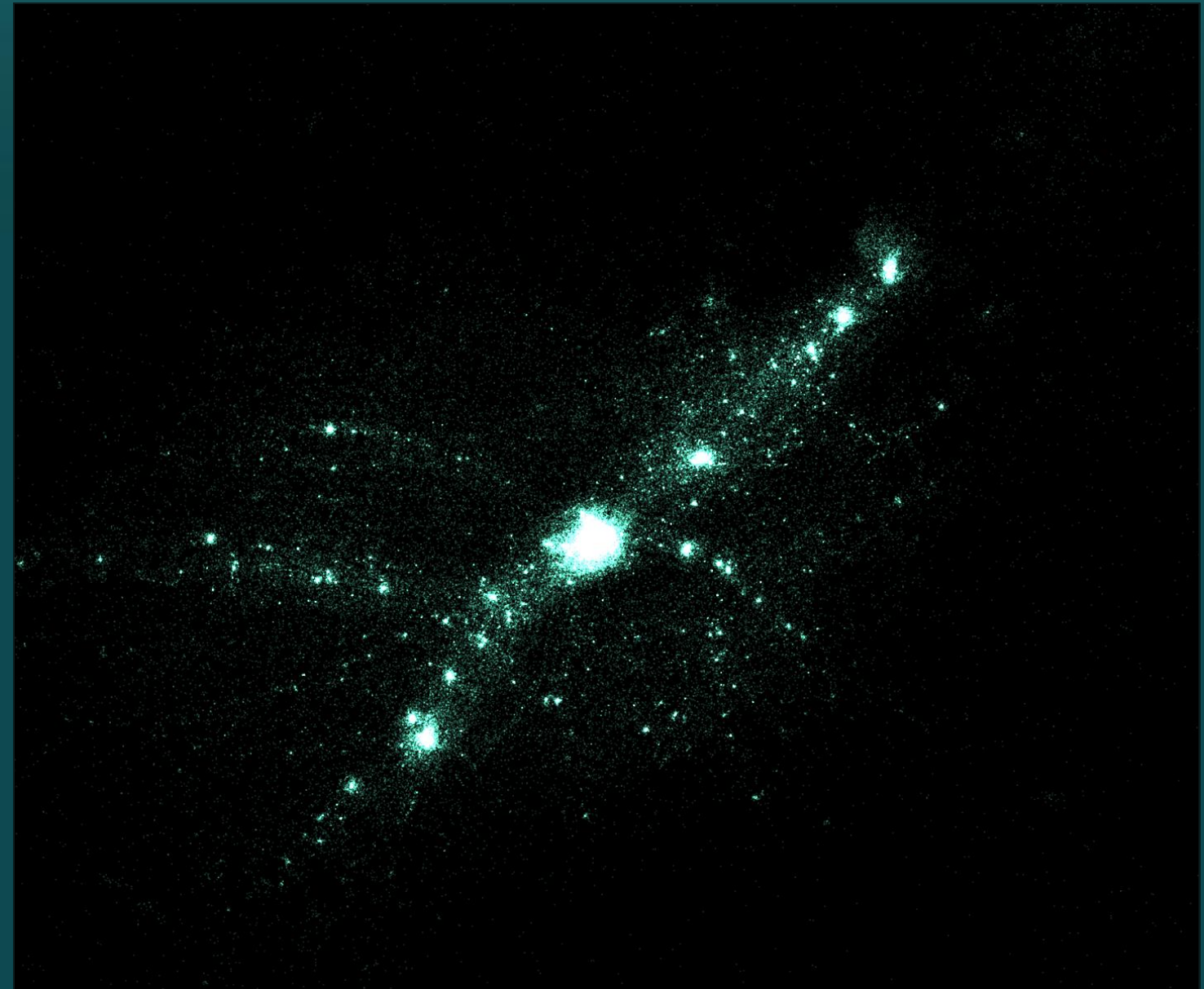
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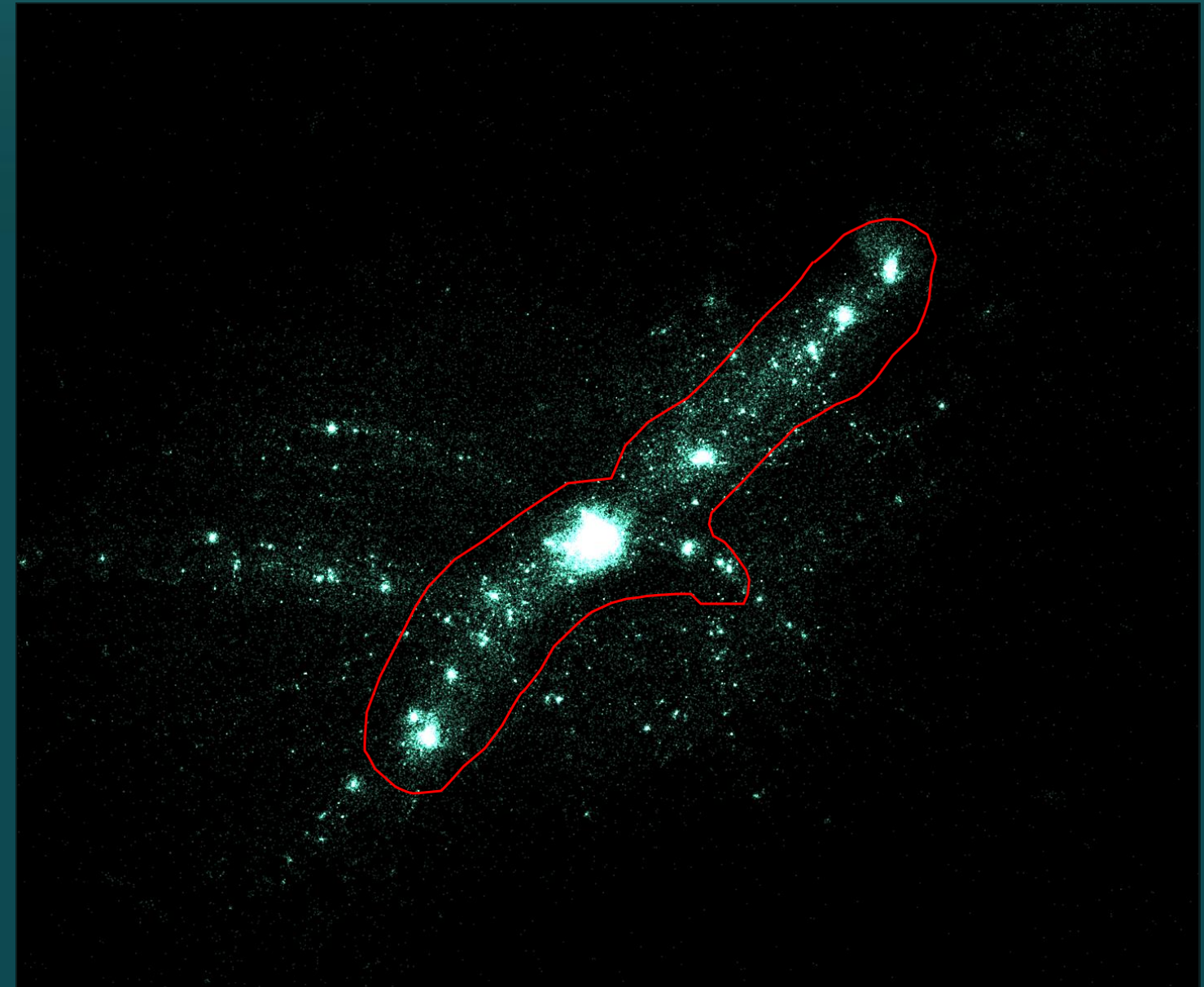
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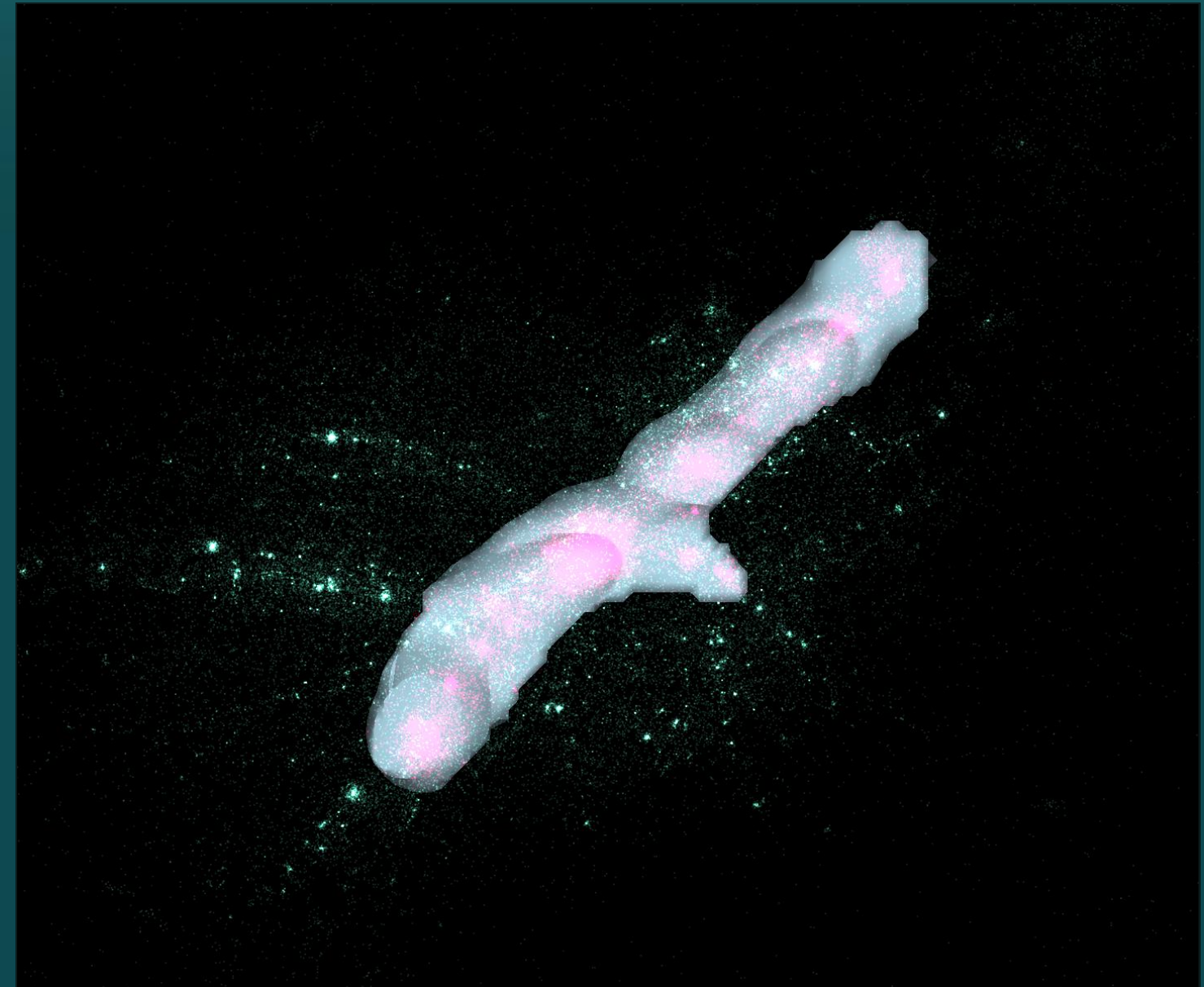
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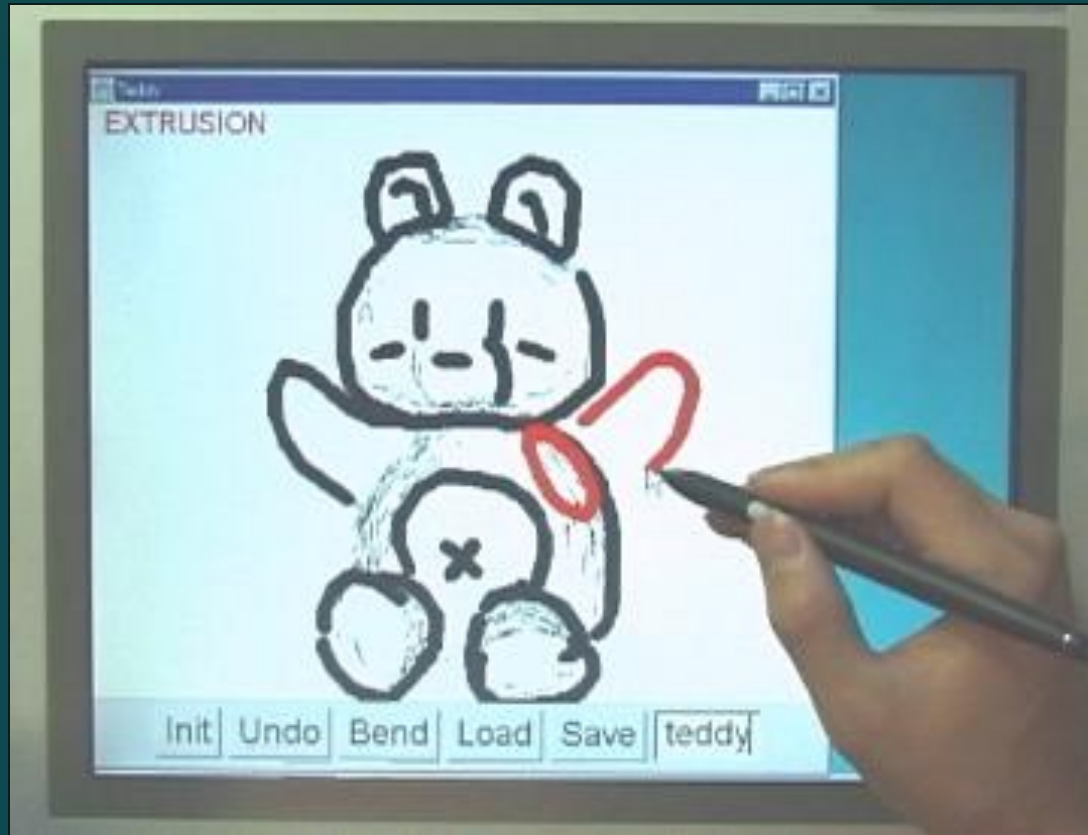
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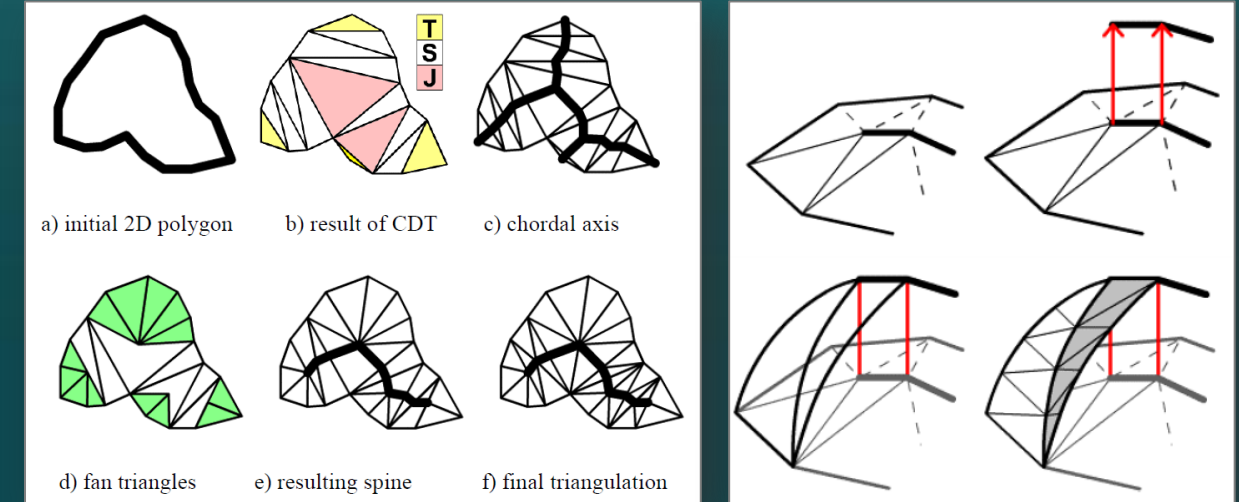
- observation: similar constraints as in sketch-based modeling:
→ definition of 3D space based on 2D input



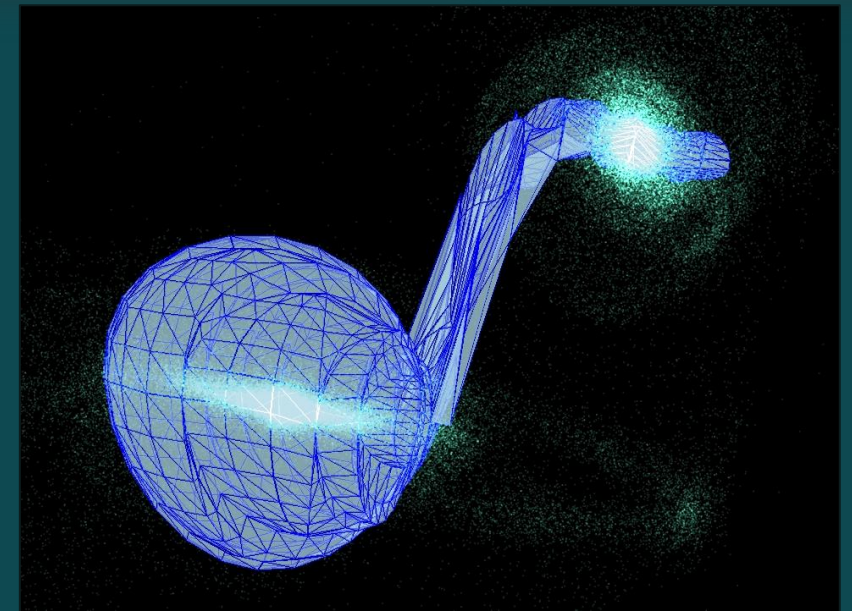
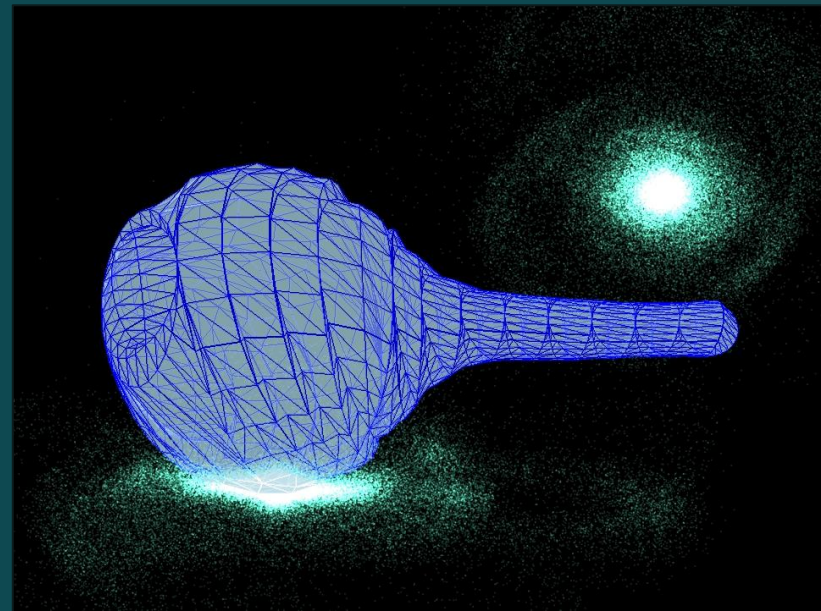
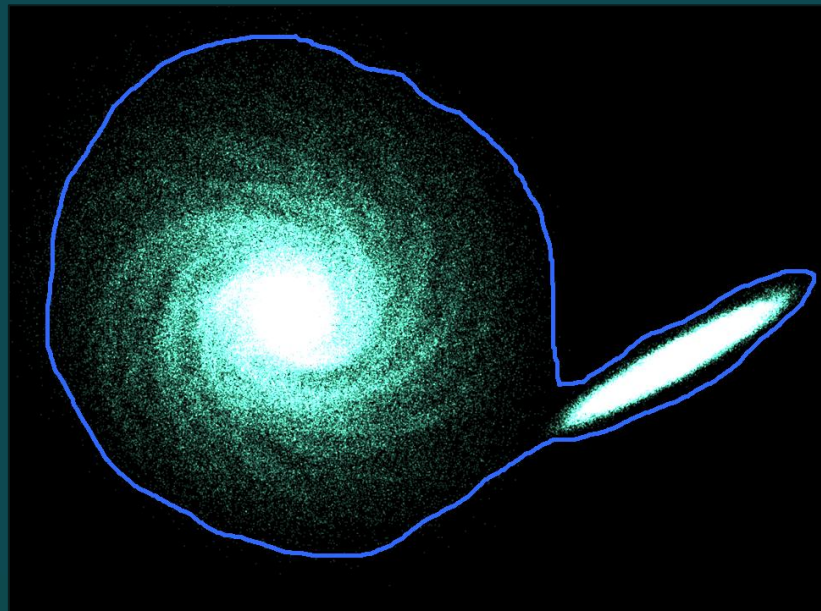
[Igarashi et al., 1999]

TeddySelection

- sketch-based modeling:
Teddy by Igarashi et al. [1999]
 - model in 3D based on sketched outline
 - not directly usable for selection

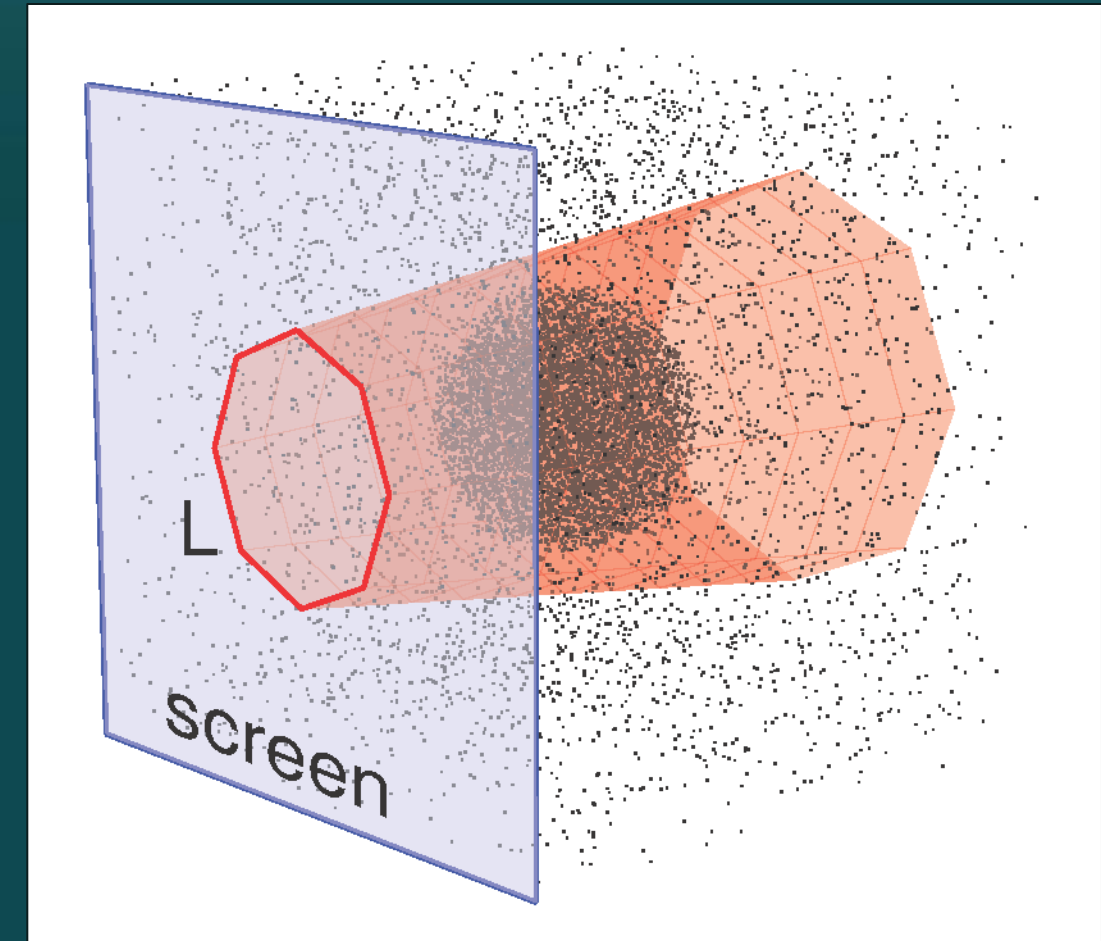


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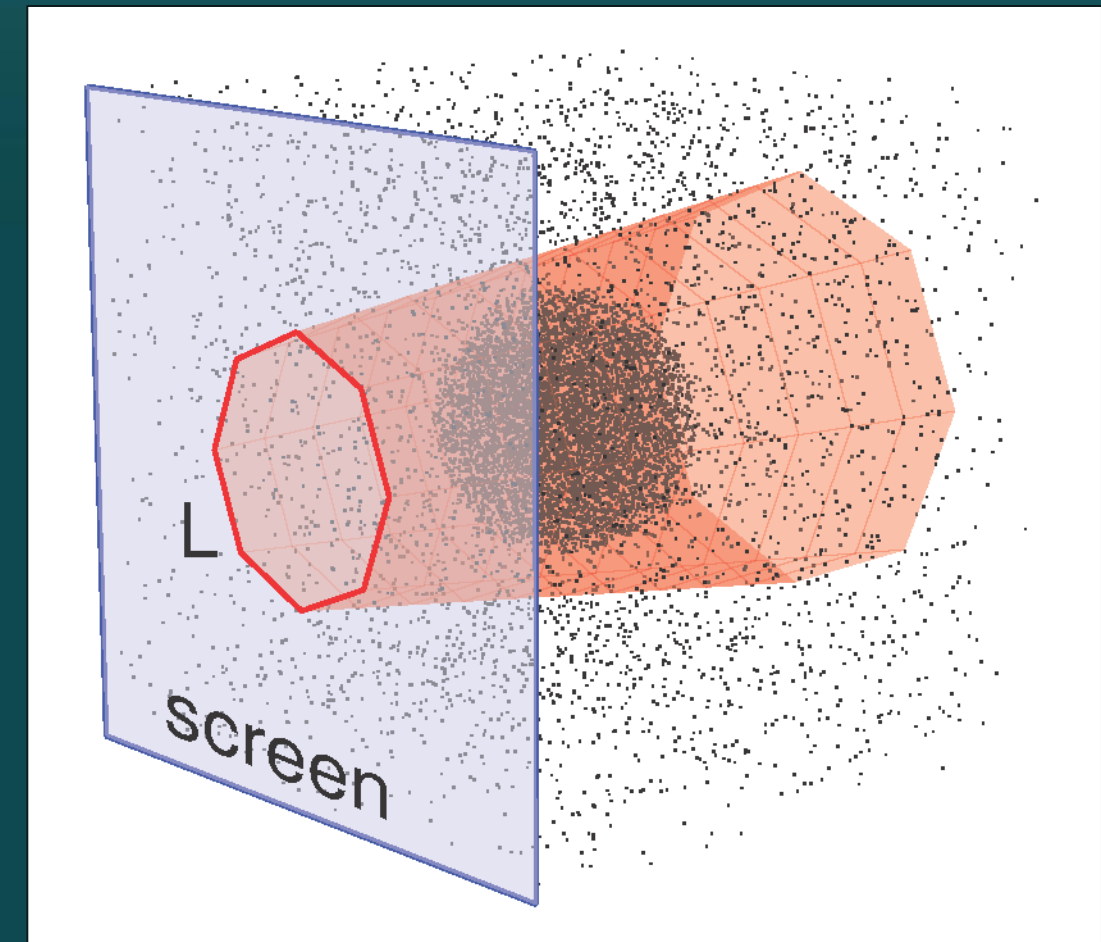
TeddySelection: Principle

1. draw lasso



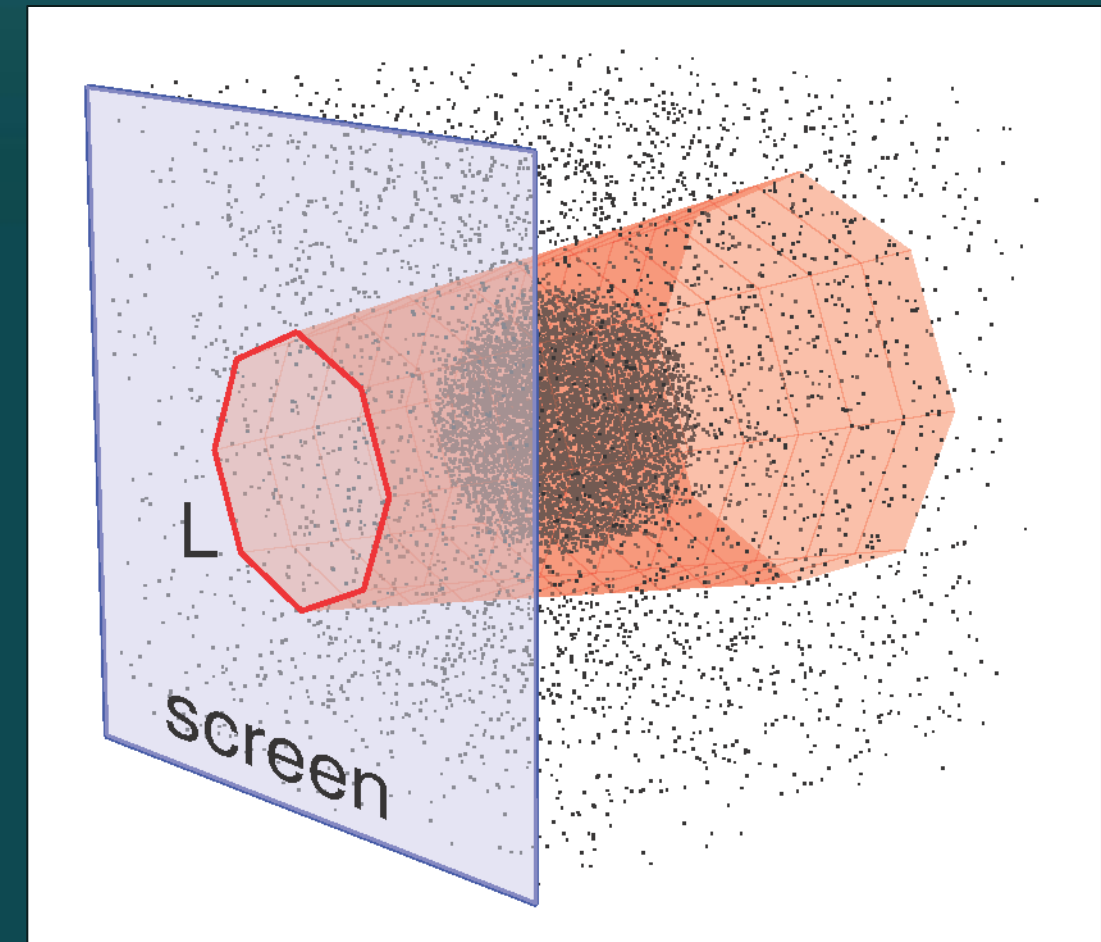
TeddySelection: Principle

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2. input polygon triangulation



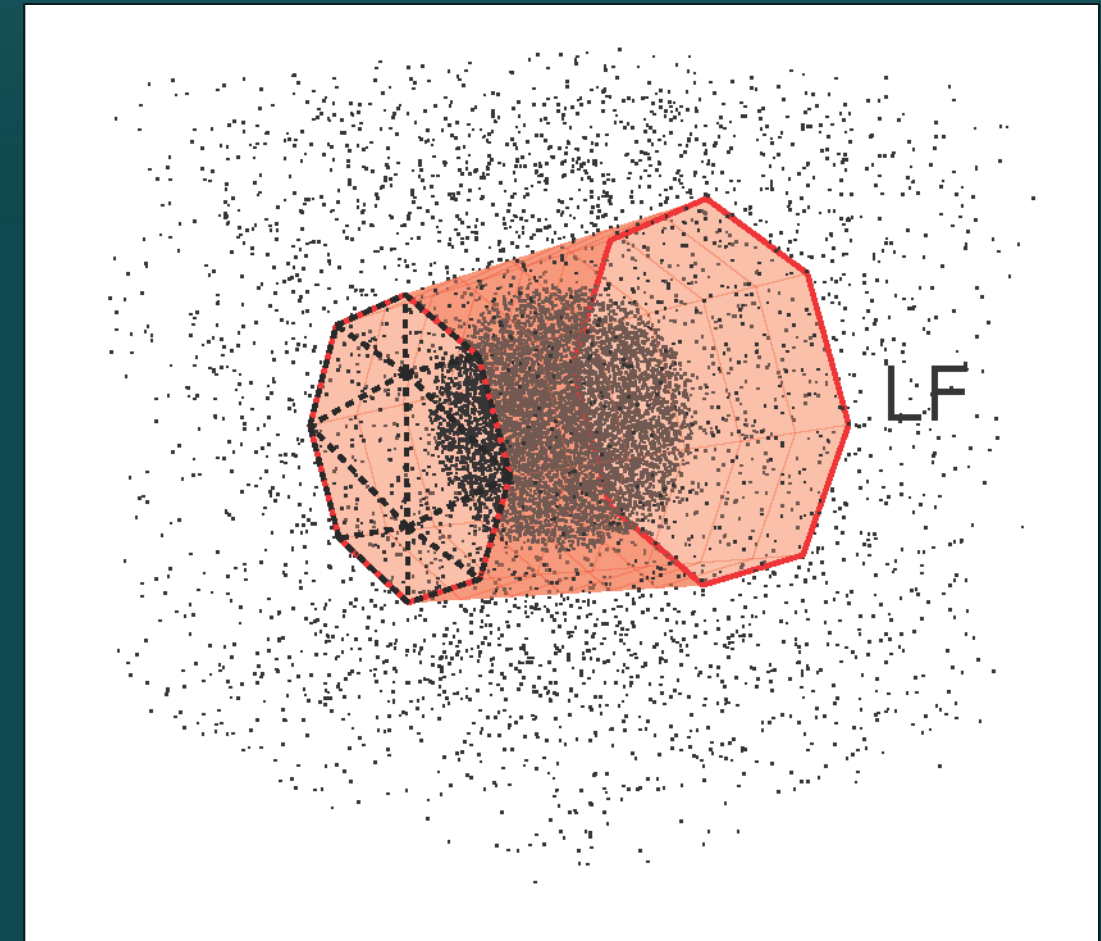
TeddySelection: Principle

1. draw lasso
2. input polygon triangulation
3. mapping particles to triangles



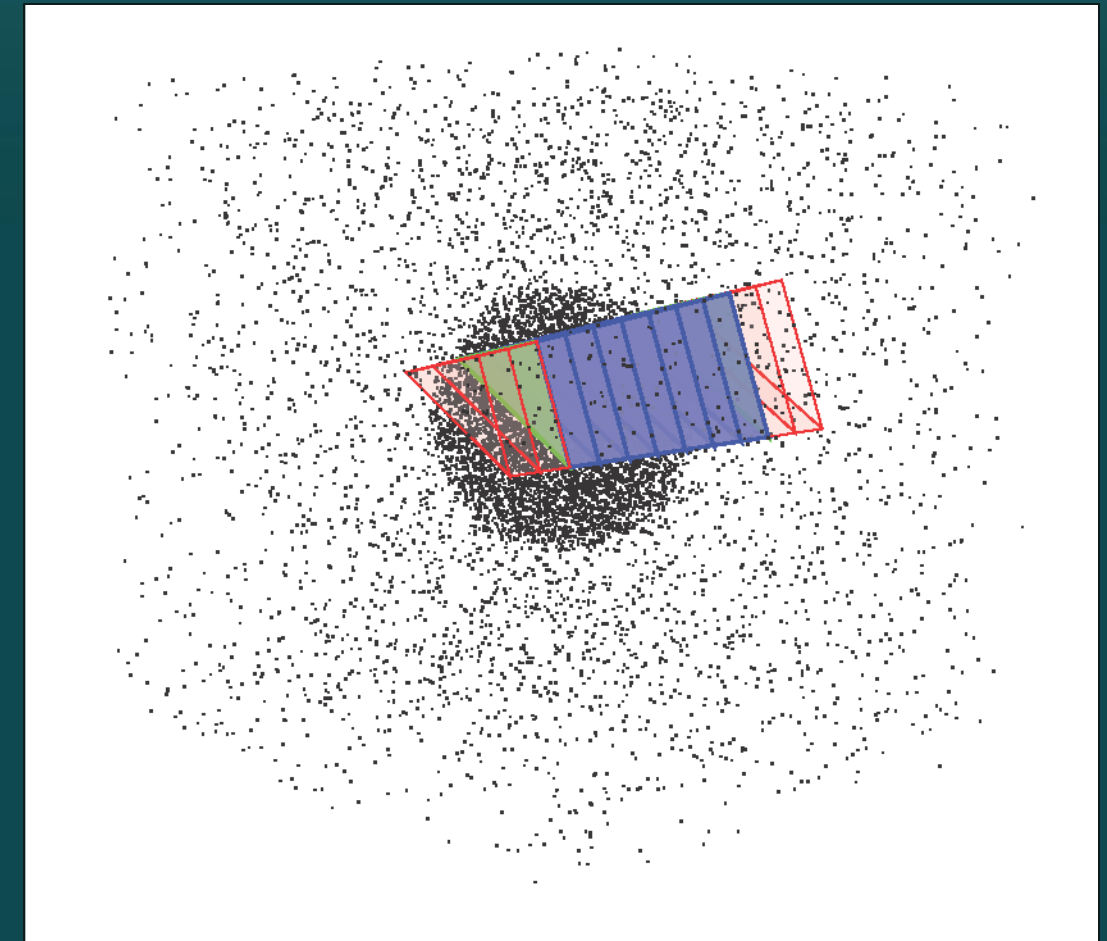
TeddySelection: Principle

1. draw lasso
2. input polygon triangulation
3. mapping particles to triangles
4. selection mesh construction
 - 1st binning to fit generalized cylinder to data



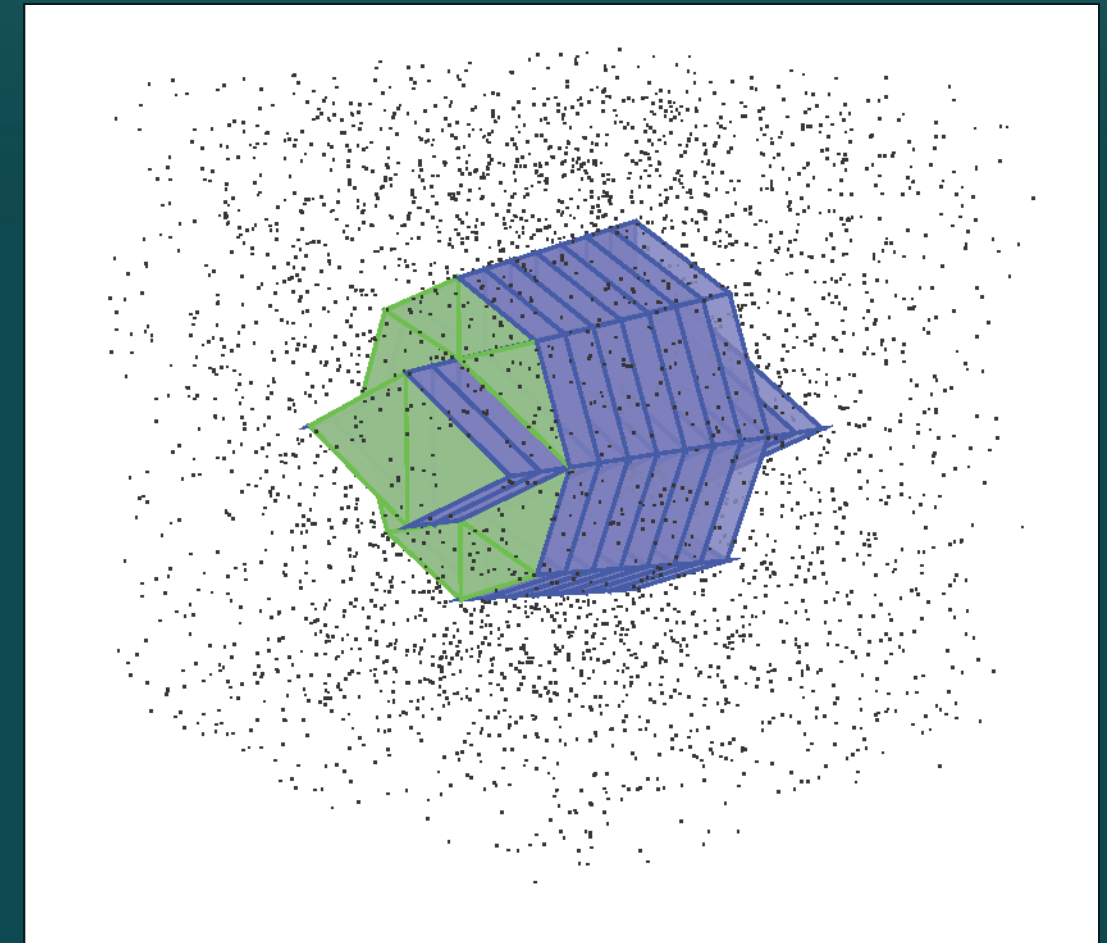
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 - 1st binning to fit generalized cylinder to data
 - 2nd binning per triangle, find closest/furthest dense bin (density $> 4 \cdot$ expected density)



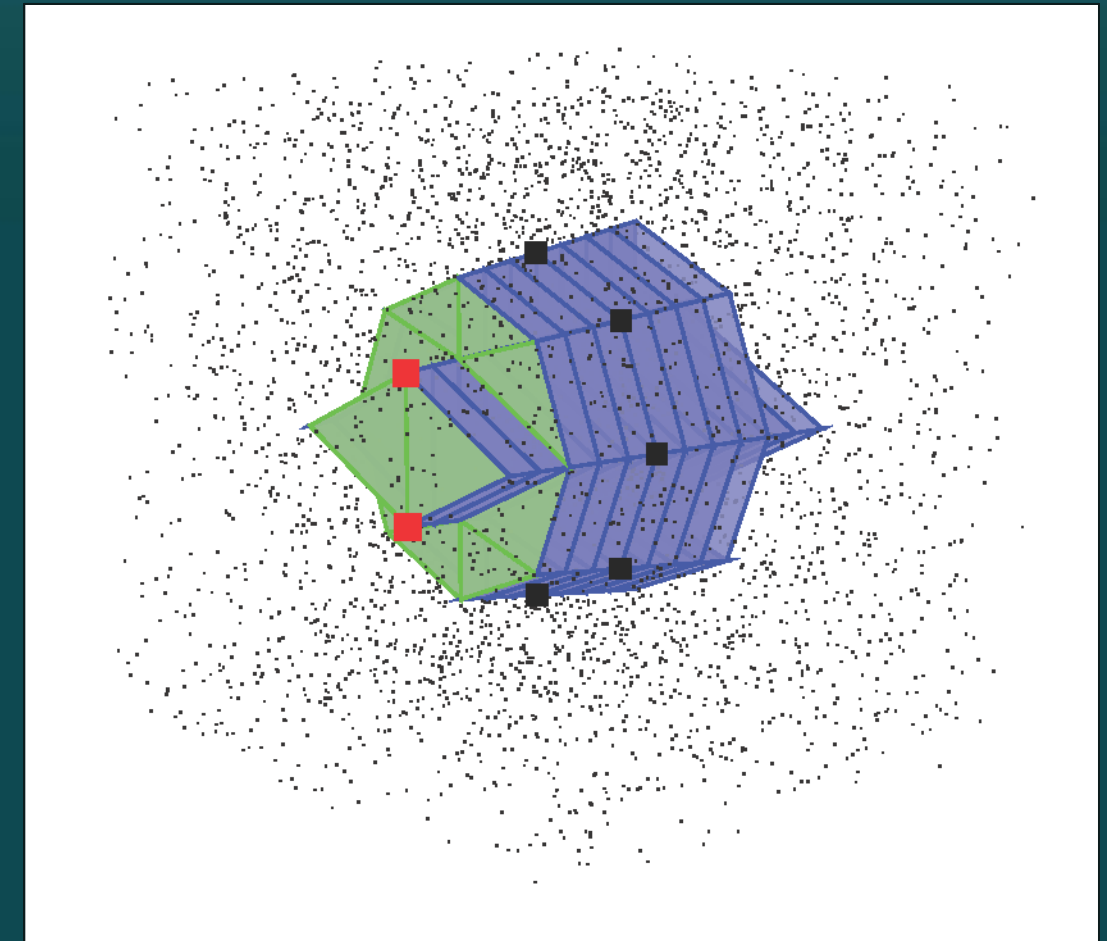
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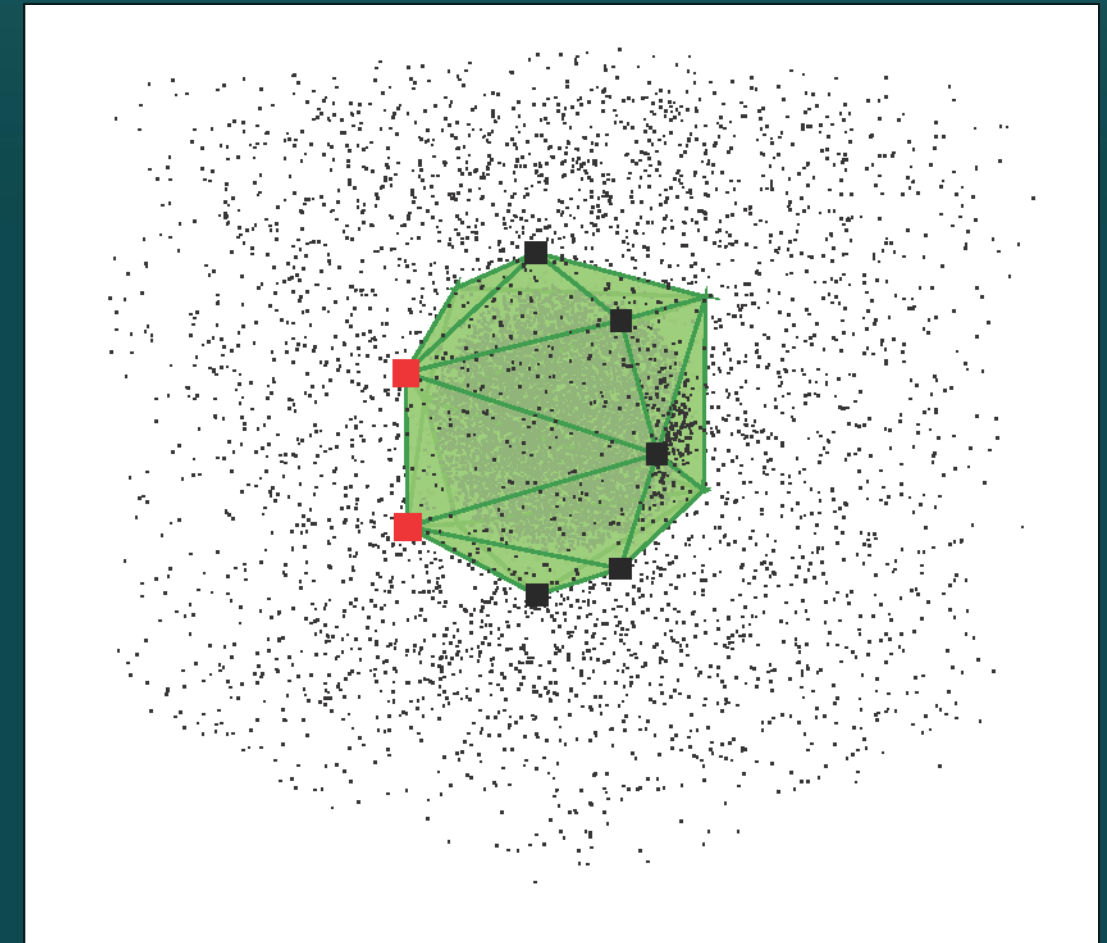
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 - place outline vertices at average distance

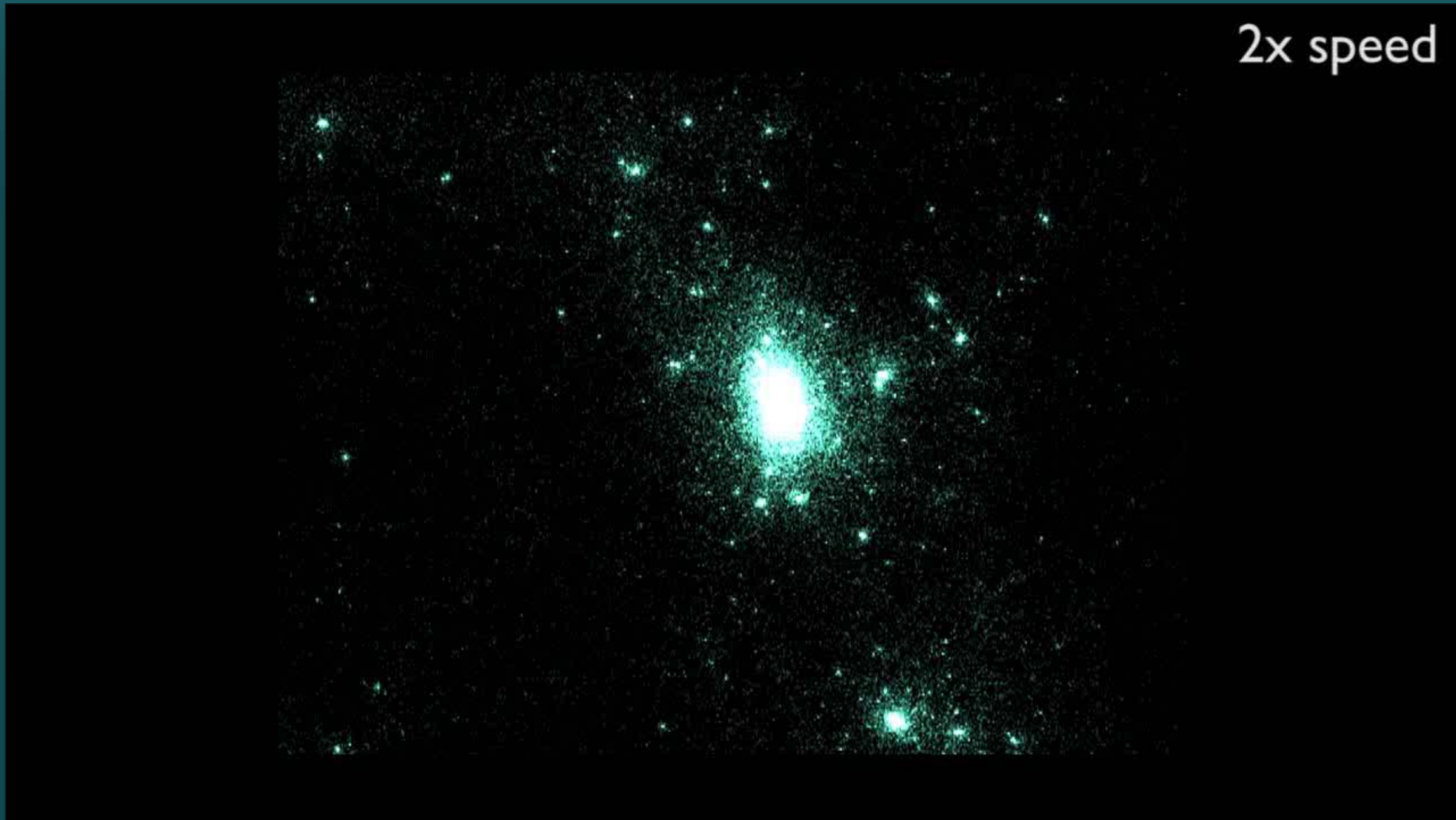


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 - place outline vertices at average distance
 - inflate 2D mesh based on binning data

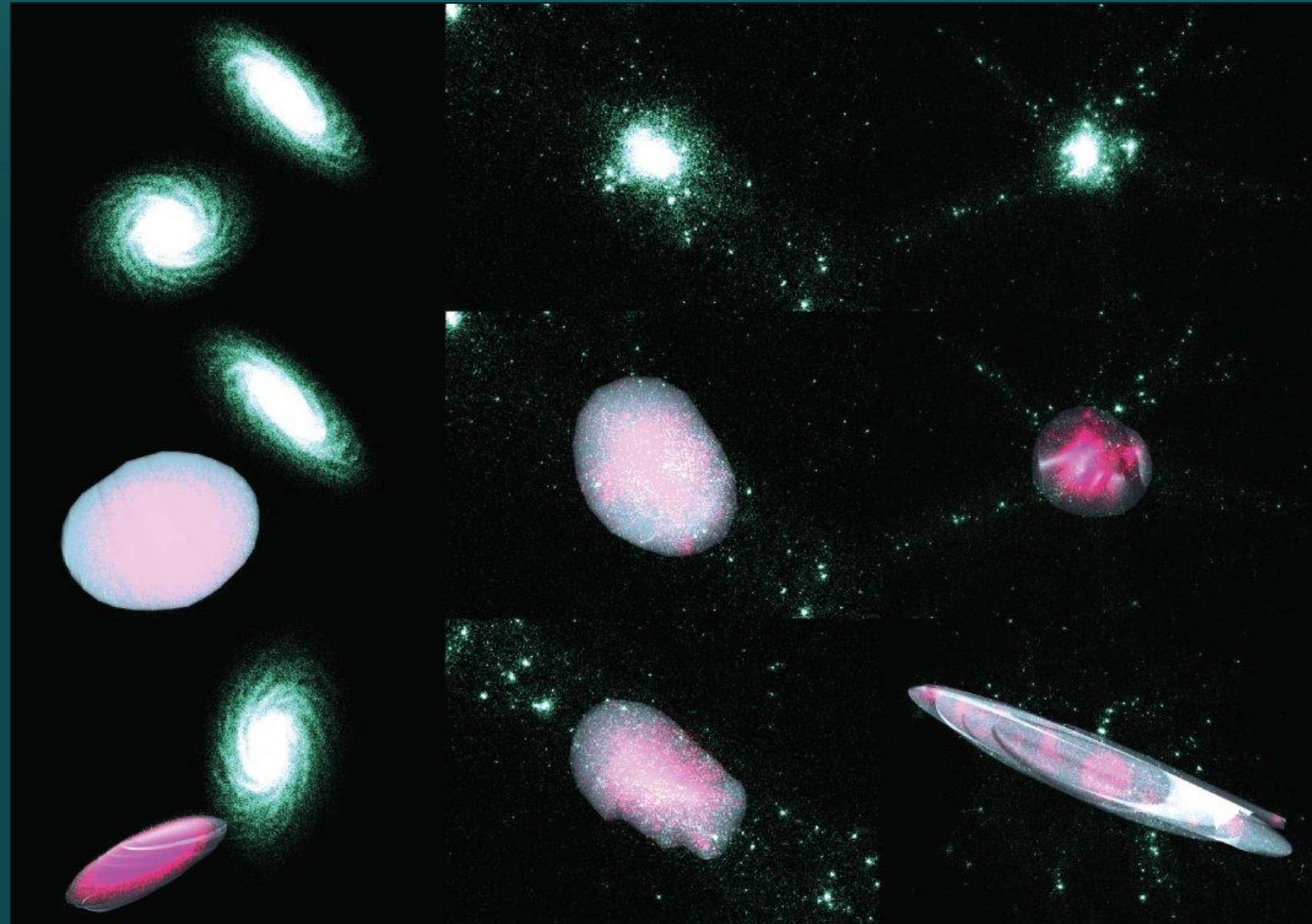


Video: TeddySelection



TeddySelection: Pros & Cons

- benefit
 - structure-aware selection
 - compact selection volume
 - fast selection (≈ 0.2 sec.)
- criticism
 - problems in sparse regions
 - volume always connected, does not work well for many small clusters



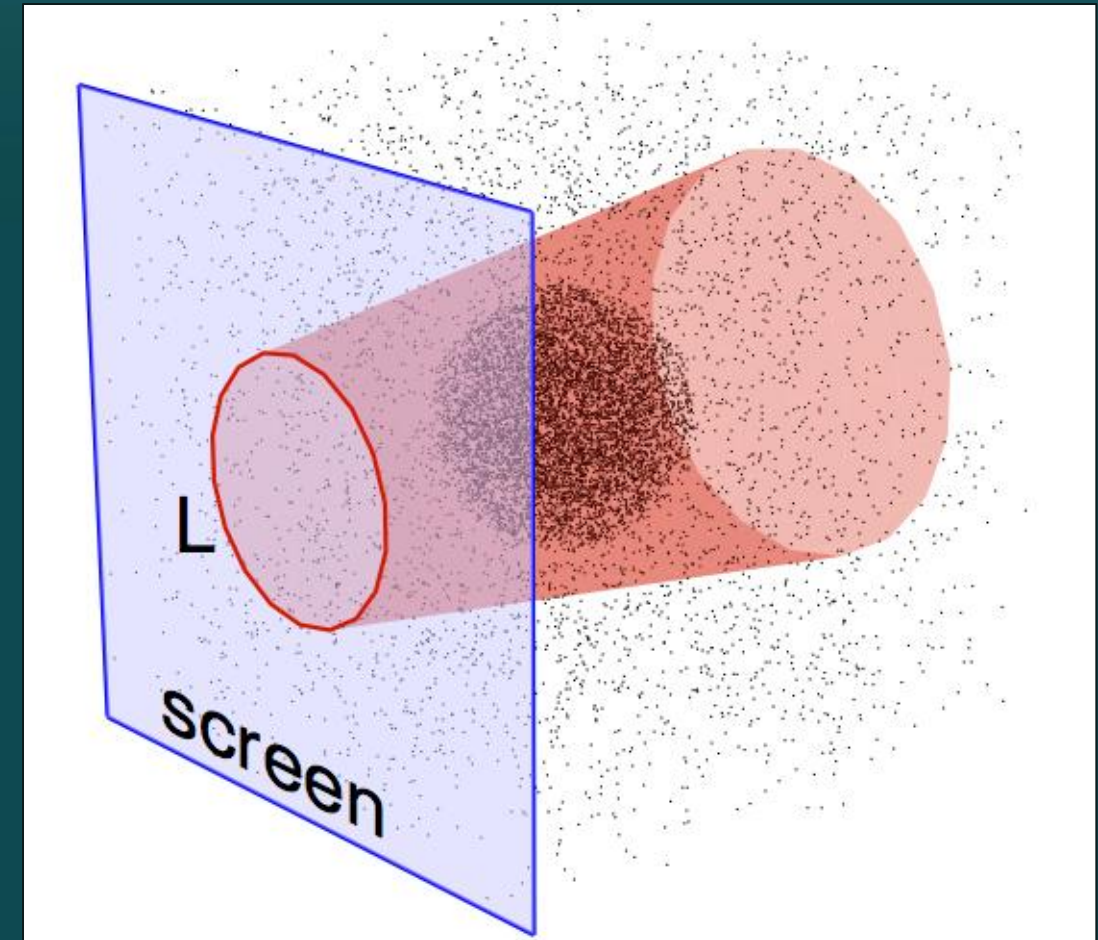
CloudLasso

- goals
 - same selection procedure as before
 - overcome limitations of TeddySelection
→ be able to treat clusters
- concept
 - base the selection volume on global particle density estimation
 - i.e., selection mesh based on density field
→ marching cubes algorithm



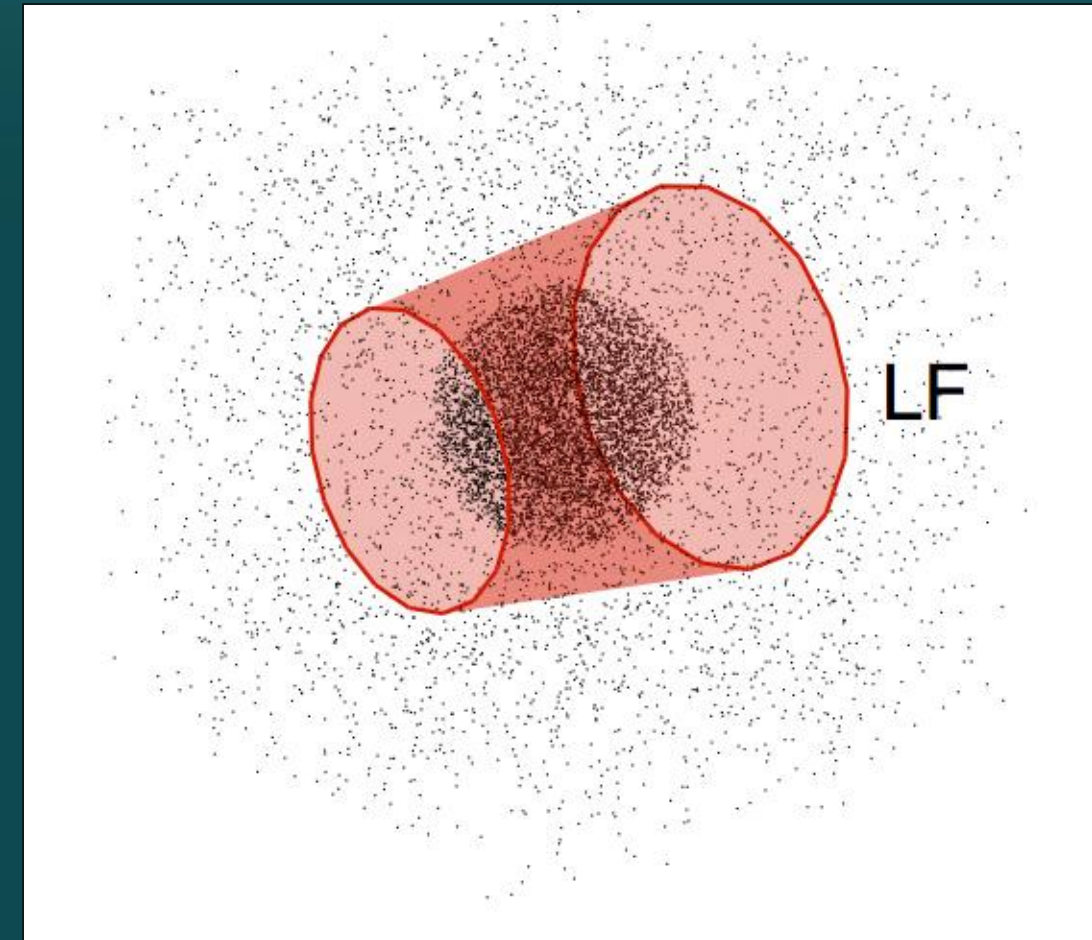
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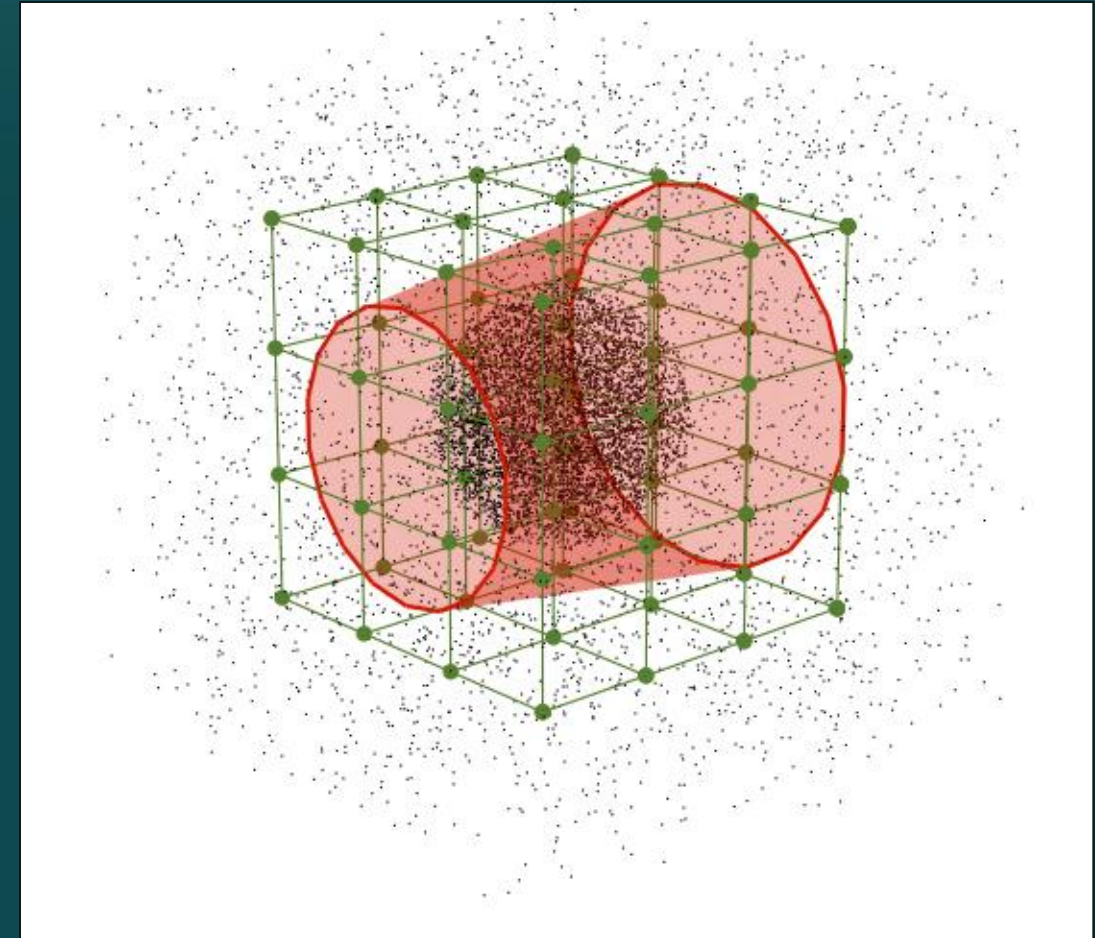
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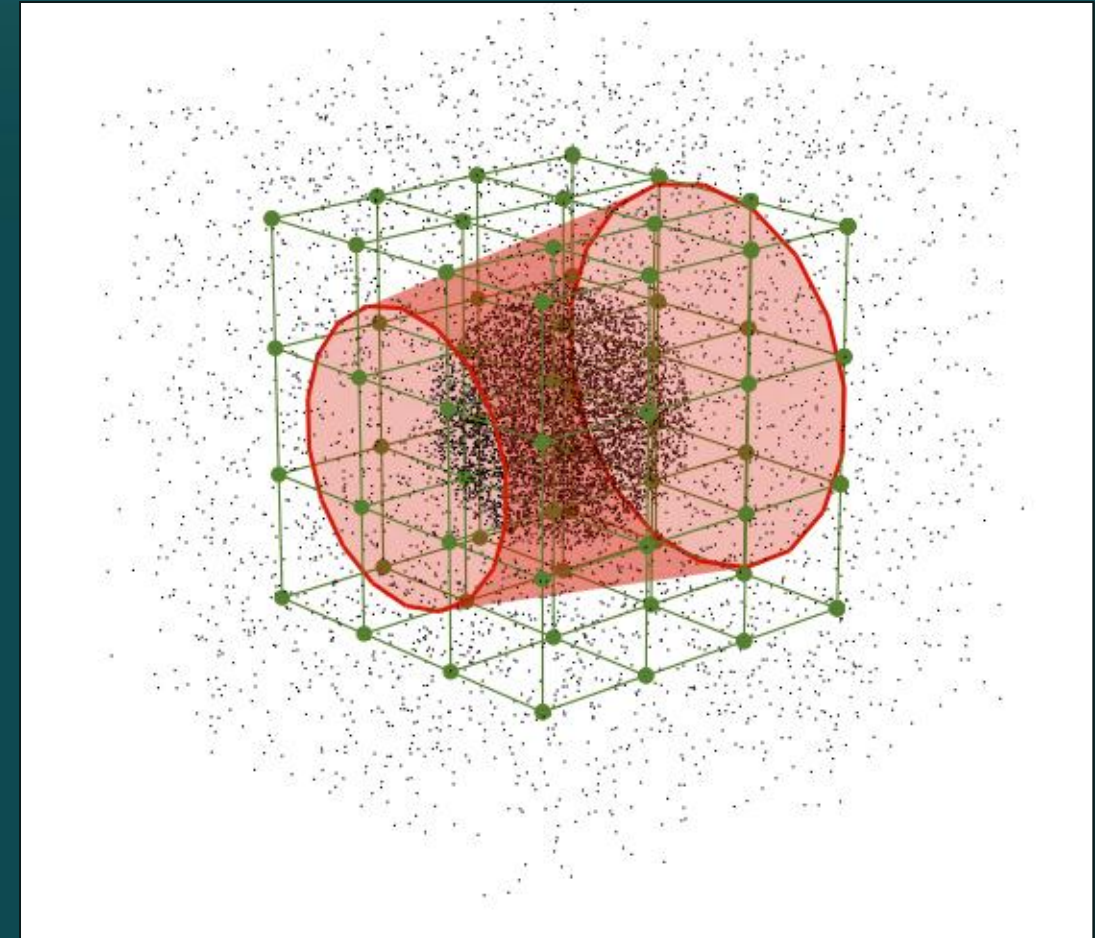
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 - fit regular grid ($64 \times 64 \times 64$) to enclose the lasso frustum



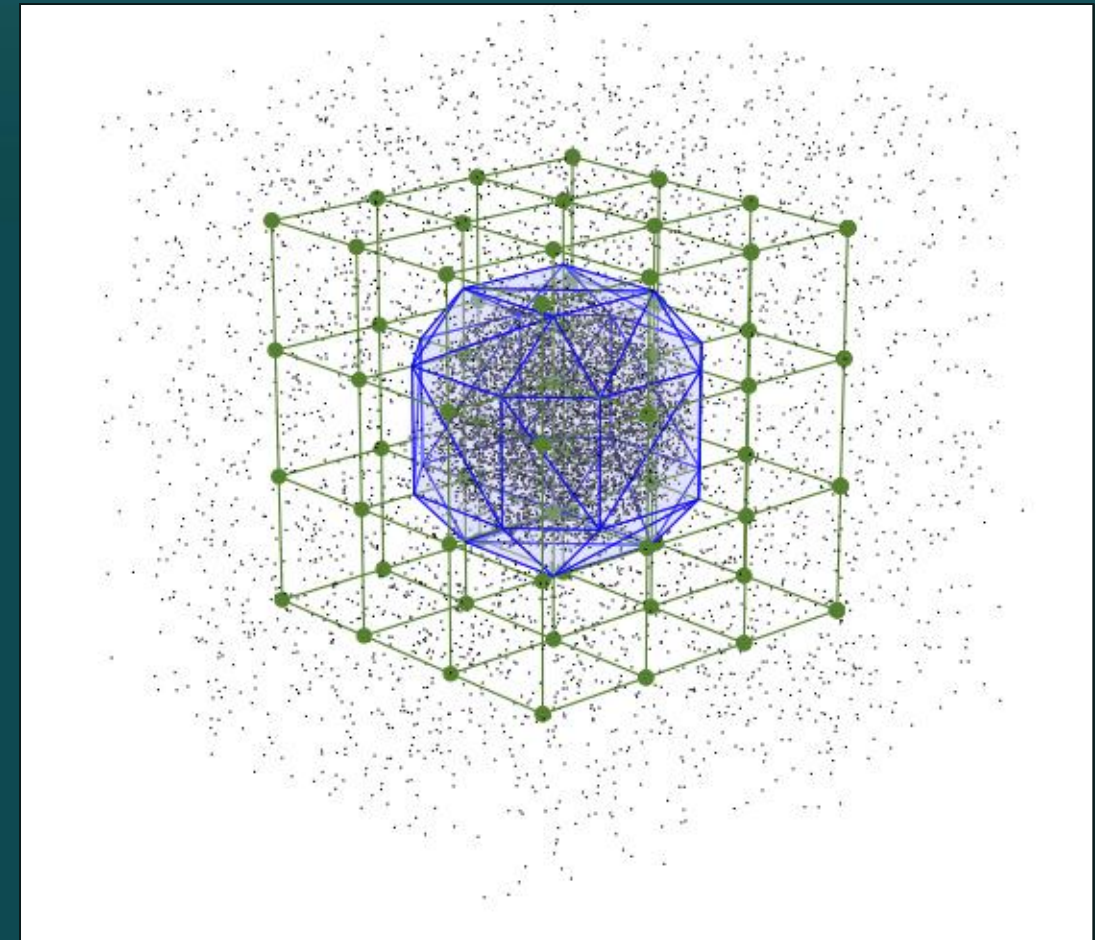
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 - use kernel density estimation on grid



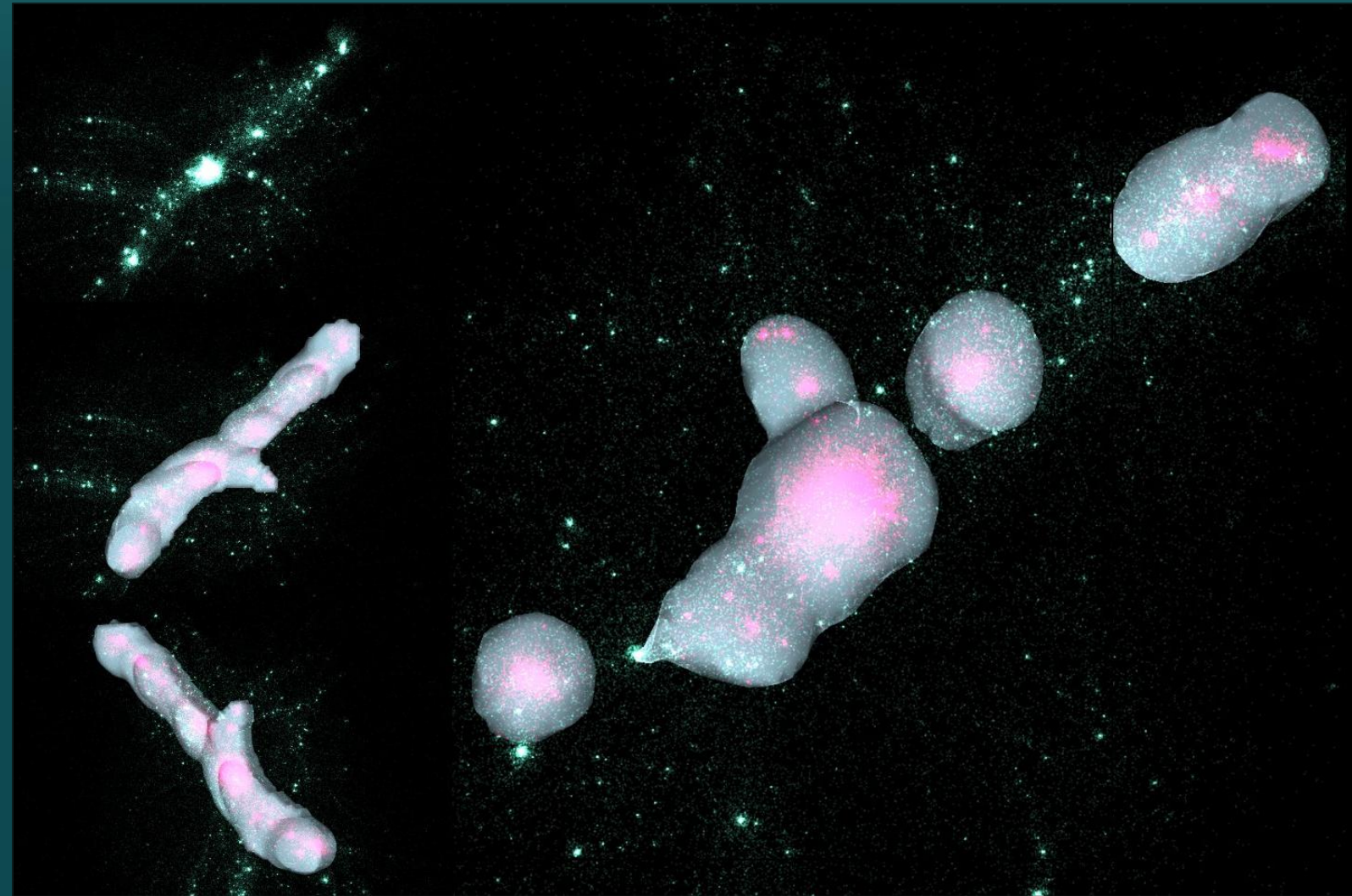
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2. selection mesh construction
 - 1st binning to fit generalized cylinder to data
 - fit regular grid ($64 \times 64 \times 64$) to enclose the lasso frustum
 - use kernel density estimation on grid
 - run marching cubes algorithm, but ensure to ignore parts outside lasso
- threshold adjustment possible interactively



CloudLasso: Results

- structure-aware selection
- separate clusters
- interactive adjustment of selection threshold
- performance:
Marching Cubes: ≈ 0.4 sec.
density estimation:
4–6 sec. for $\approx 2 \cdot 10^5$ particles

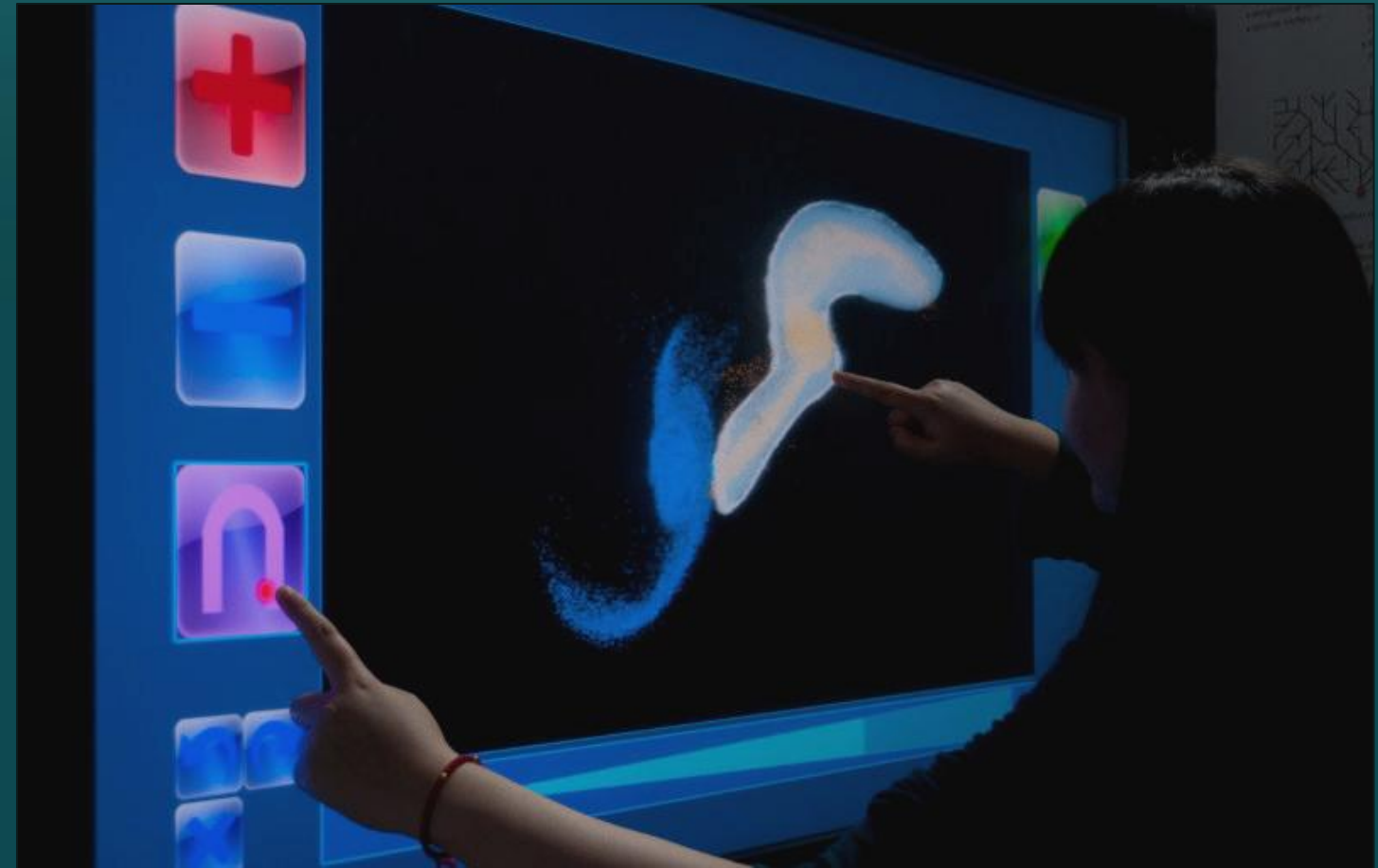


Video: CloudLasso Selection & Interaction



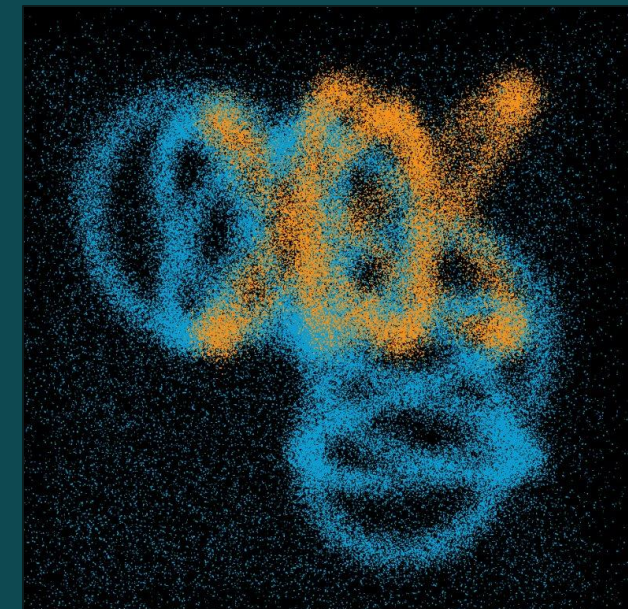
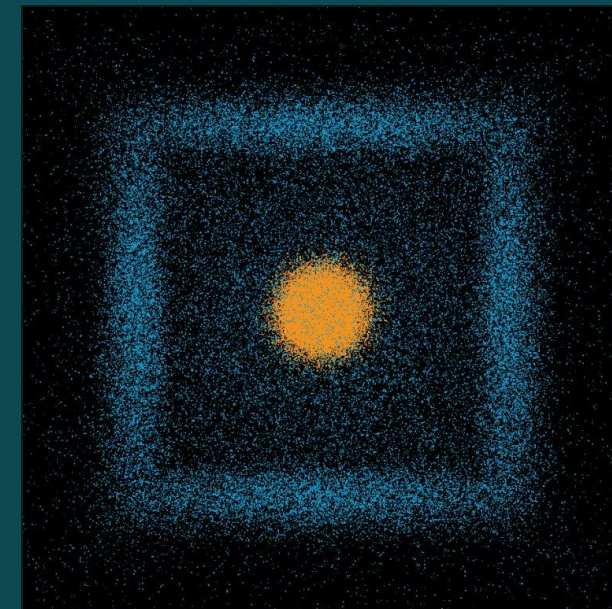
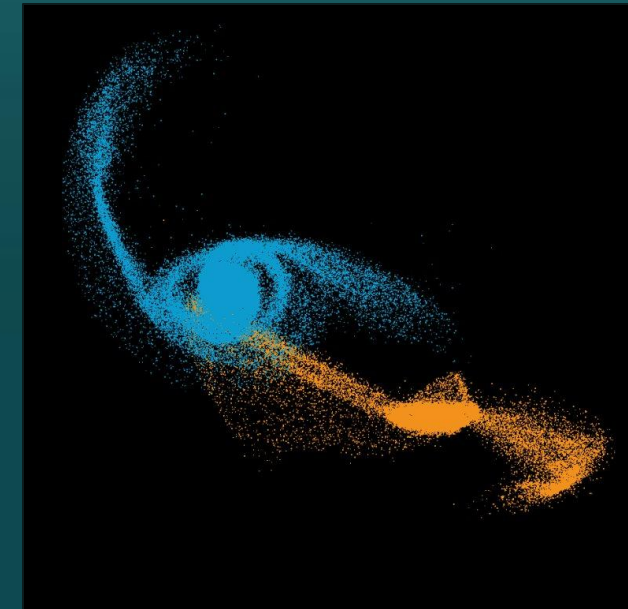
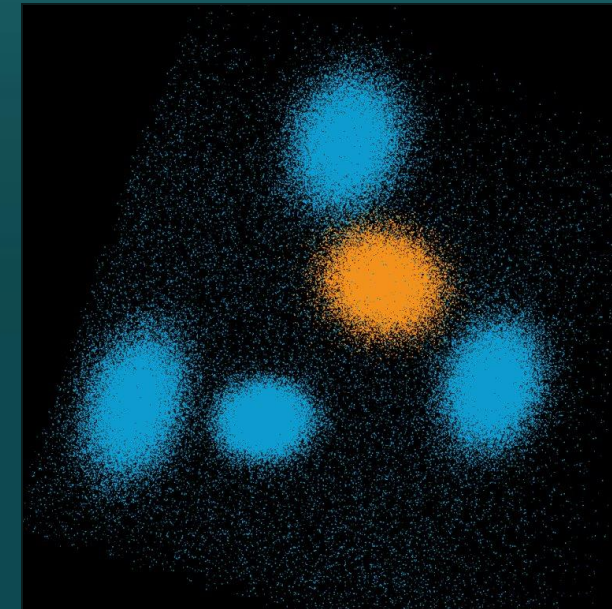
Evaluation & Validation: User Study

- informal feedback positive
- quantitative study to confirm
- restriction to 2 selection methods:
 - CylinderSelection
 - CloudLasso
 - base line (Tablet Freehand Lasso)
- both could be fine-tuned
- Boolean operations possible

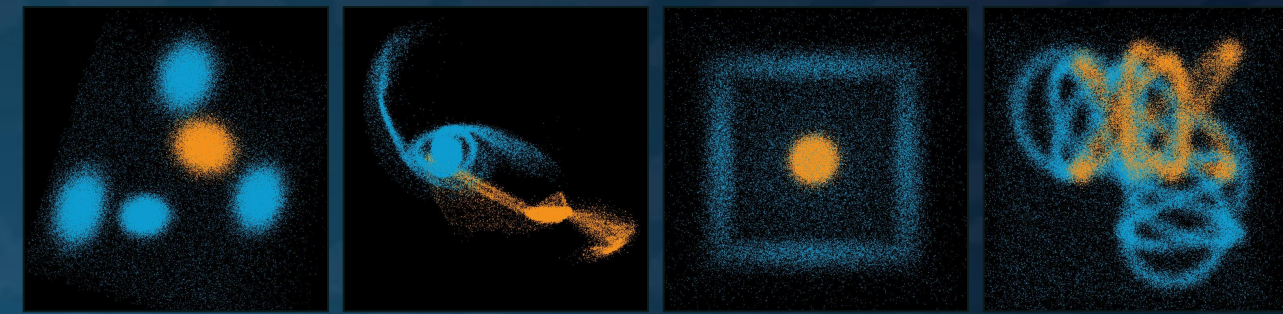


Study Design

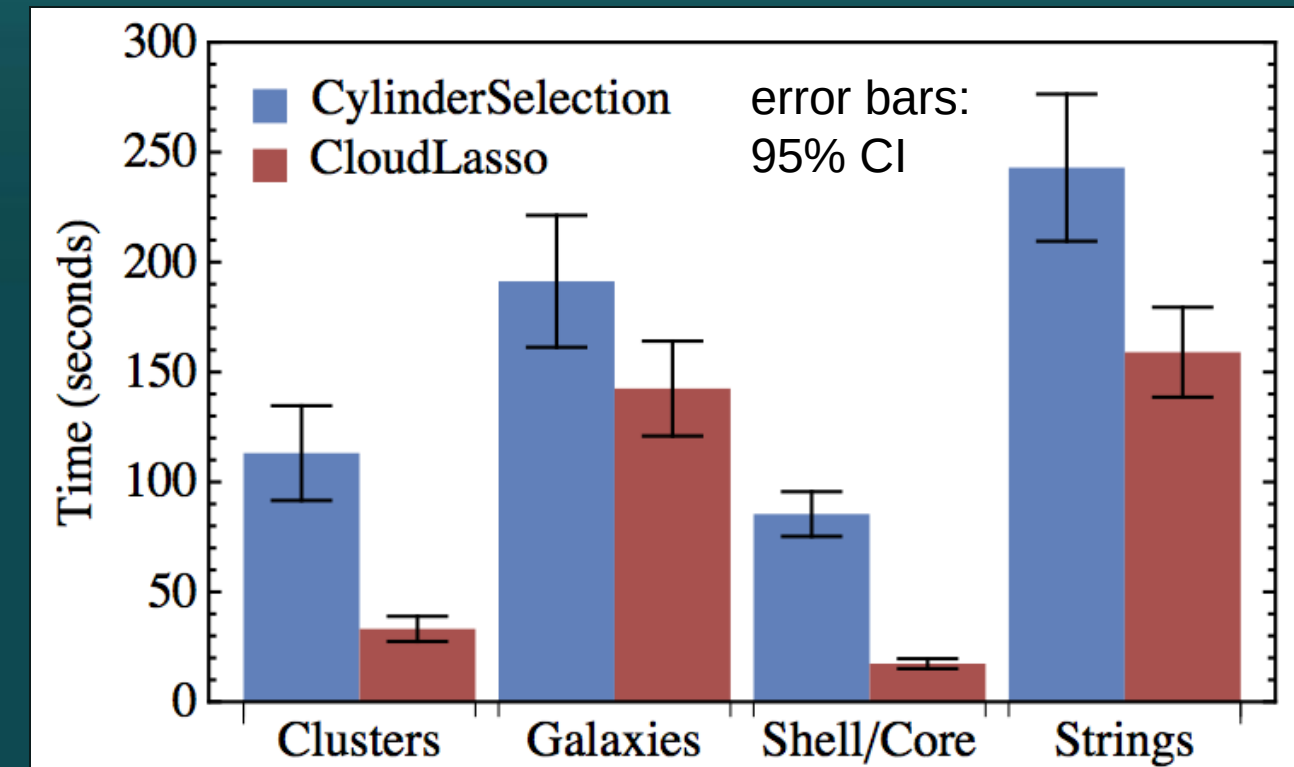
- 12 participants (4 female)
- 4 selection tasks (datasets)
- measurement of **time**, **error**, and **selection volume**
- questionnaire for subjective opinion



Study Results

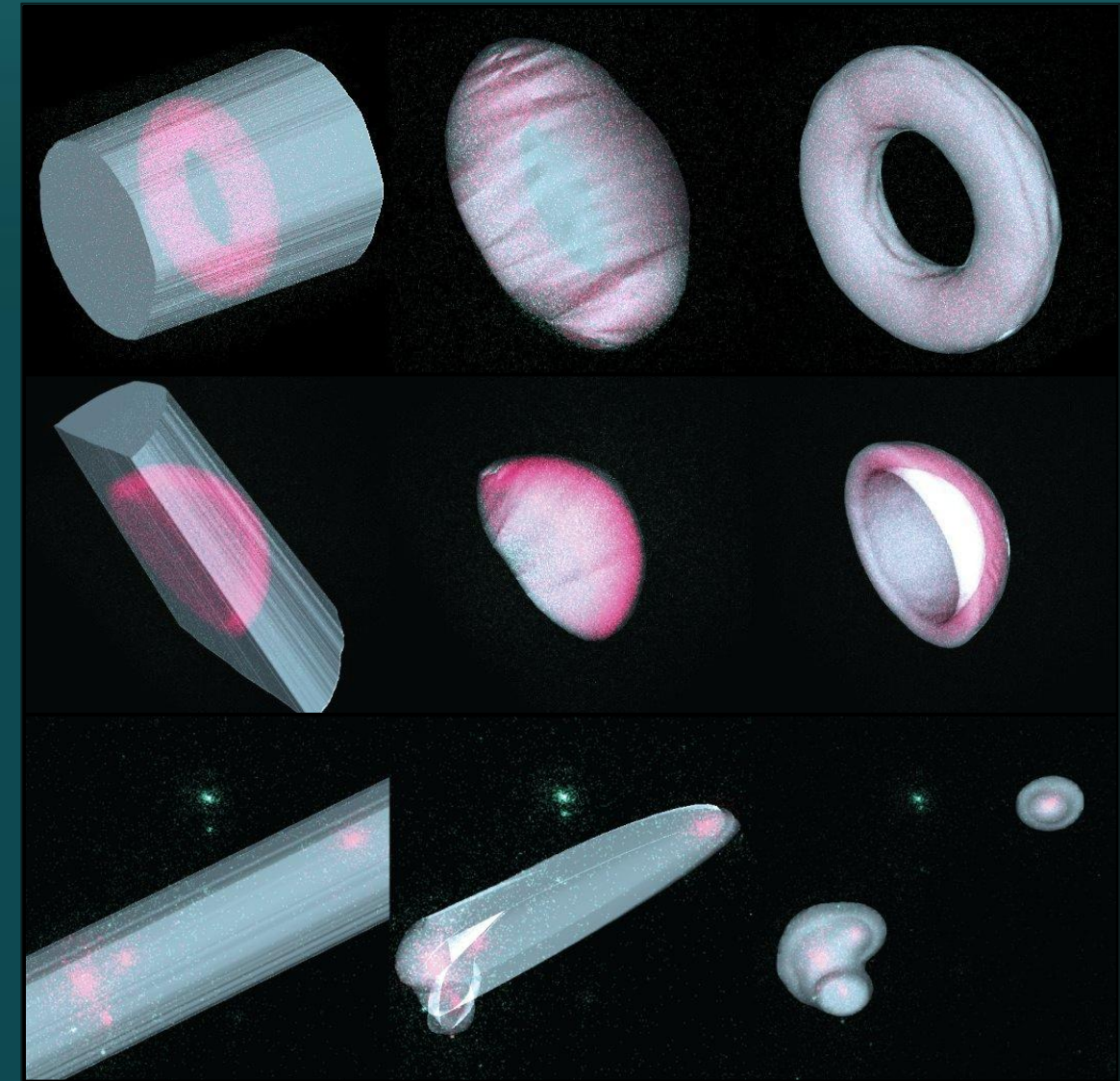


- CloudLasso (CL) always faster than CylinderSelection (CS)
 - significant except galaxies
- two error metrics F_1 & MCC
 - CloudLasso always less error than CS
 - F_1 significant except galaxies
 - MCC significant for clusters & shell/core
- CloudLasso volume always smaller, significant for strings dataset
- CloudLasso the preferred technique for all participants



Discussion: CloudLasso vs. TeddySelection

- both **spatial & structure-aware** selection
- both based on lasso principle
- TeddySelection: connected selection
- CloudLasso: individual clusters
- CloudLasso can handle difficult cases
- both can be coupled with Cylinder-Selection using Boolean operations

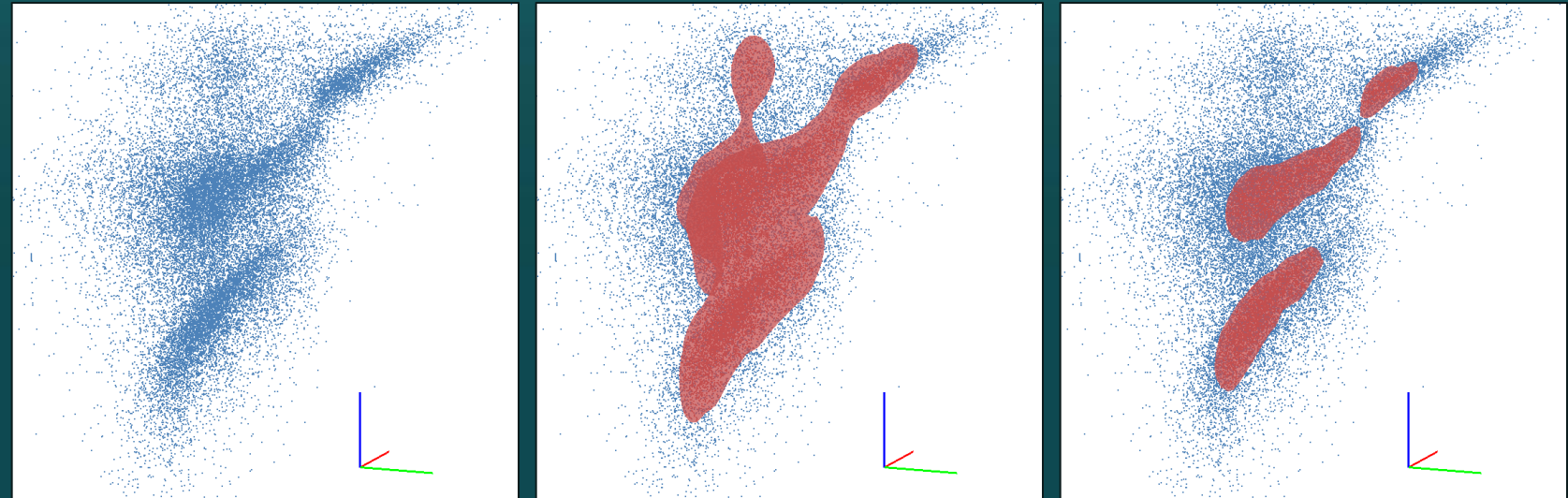


Limitations

- performance
 - CloudLasso requires grid-based density estimation
 - slower than interactive speeds ($\approx 4\text{--}6$ seconds for $\approx 2 \cdot 10^5$ particles)
 - but parallelizable / GPU; only needs to be computed once per scale level
- several parameter choices (e.g., # of bins)
 - parameters seem stable, not changed in our experiments
 - initial density threshold of CloudLasso suggested by algorithm
- set difference (subtraction) just with structure-aware selection not good
 - include operations with CylinderSelection results (e.g., for subtraction)

Application Domains / Future Work

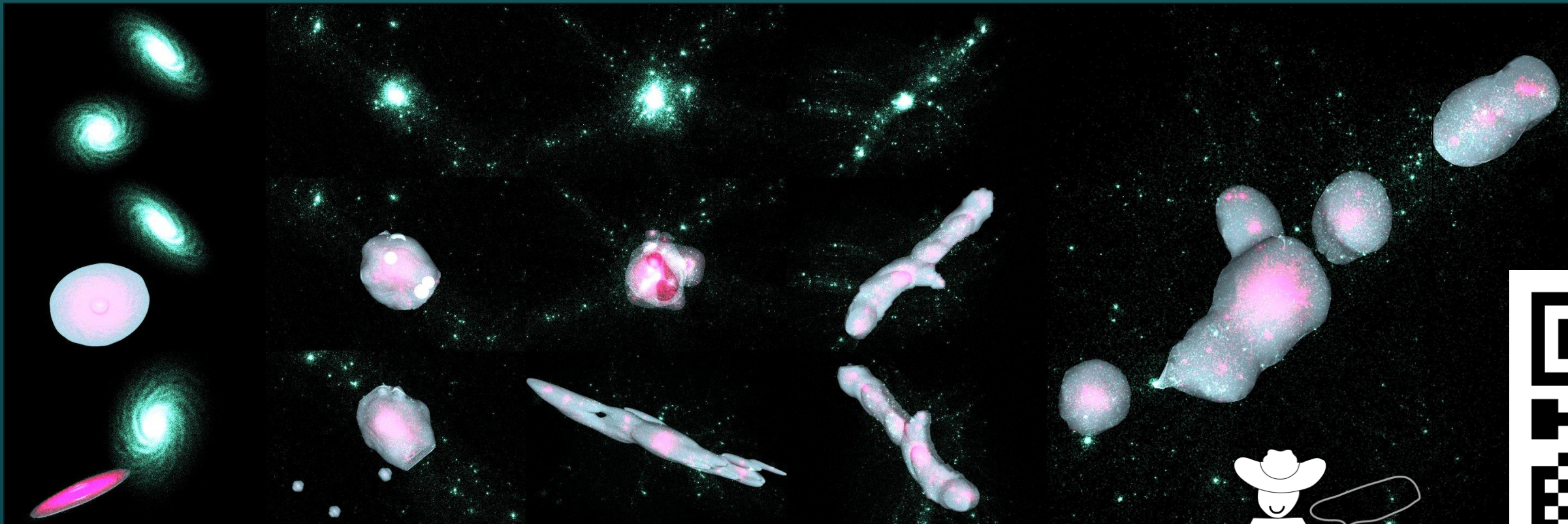
- any particle-based dataset
- also abstract data such as 3D scatter plots
→ linked views
- huge datasets possible
- selection metrics other than density possible
- applicable to volume data with minor changes



Conclusion

- TeddySelection & CloudLasso:
new **spatial, structure-aware** 3D selection techniques
- input: lasso drawn in 2D; output: 3D subspaces
- support complex spatial selections
- applications in many fields of visualization
- study showed that CloudLasso is superior to the traditional cylinder-based selection both in performance and overall preference
- smart selection techniques essential for **interactive** visualization

Thanks for your attention!



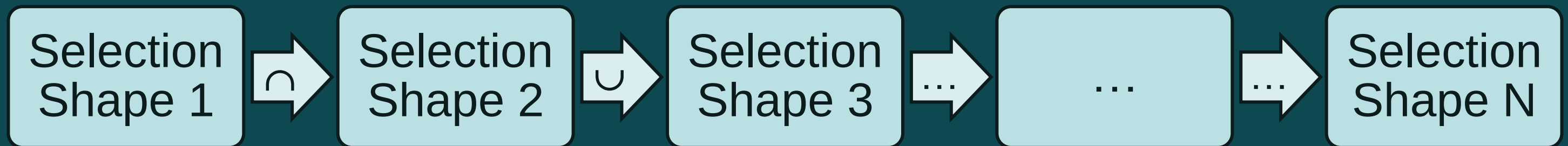
*Trivia: LingYun Yu's nick name is Yun which in Mandarin (云) means "cloud".
So it's really Yun's Lasso and she is the LassoGirl ... ;-)*



Video & demo:
<http://tobias.isenberg.cc/VideosAndDemos/Yu2012ESA>

What about HUGE datasets?

- interactive selection based on a well-chosen sample
- use of both LassoSelection and CylinderSelection
- generation of selection shapes, sequence of Boolean operations
- off-line application to the whole huge dataset (batch process)



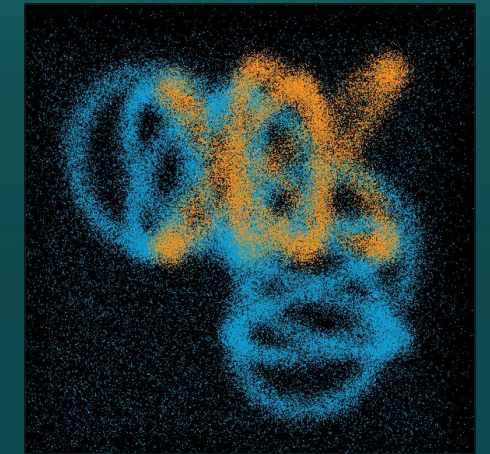
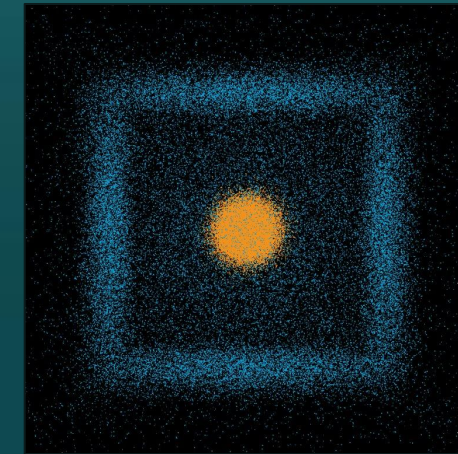
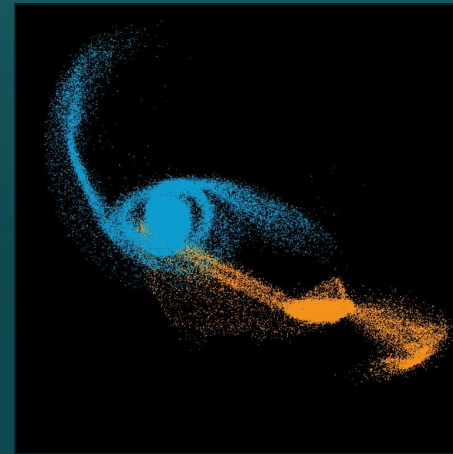
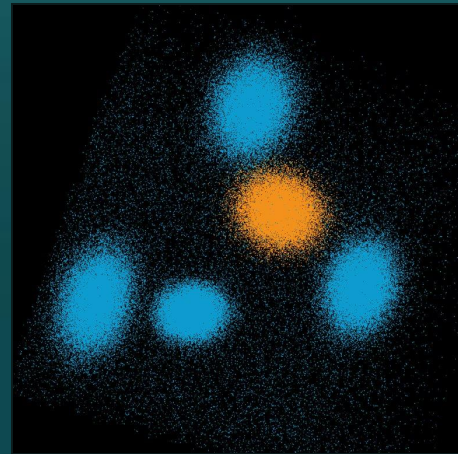
What about properties other than density?

- density makes most sense for particle data
- other properties may make sense, e.g., for volume data
- property needs to be defined continuously
- need means to compute property on a grid for CloudLasso
- output always a mesh that encloses a volume in 3D space
- particles/voxels inside that volume are selected

What about precision issues?

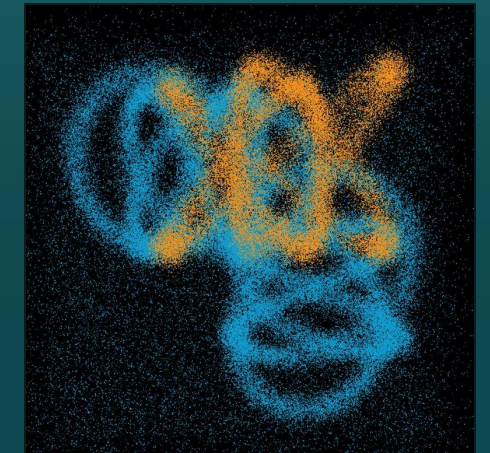
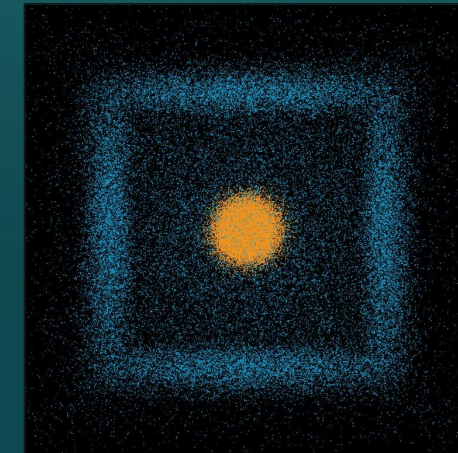
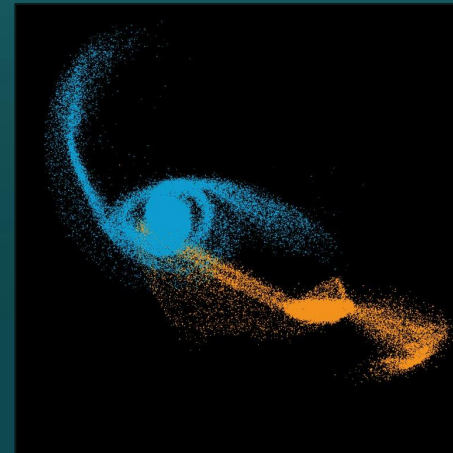
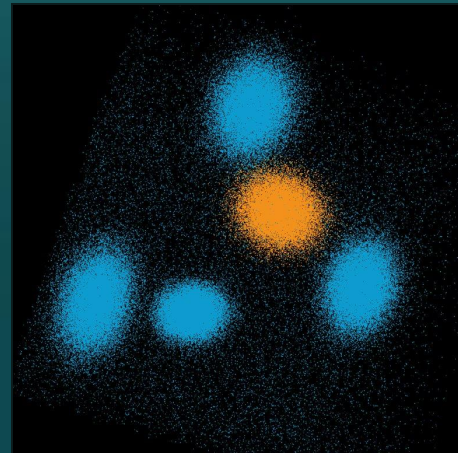
- precise input possible (mouse, pen, algorithmic)
- adjustment possible after selection operation
- iterative selections possible
- selection is structure-aware, thus needs less precision

Study Results – Errors: F_1



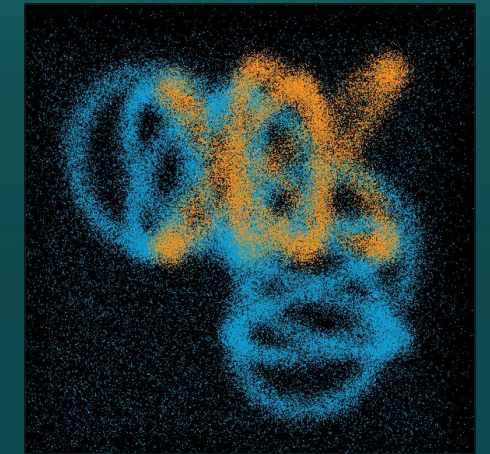
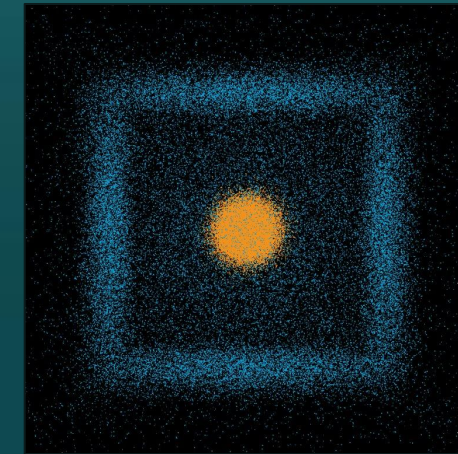
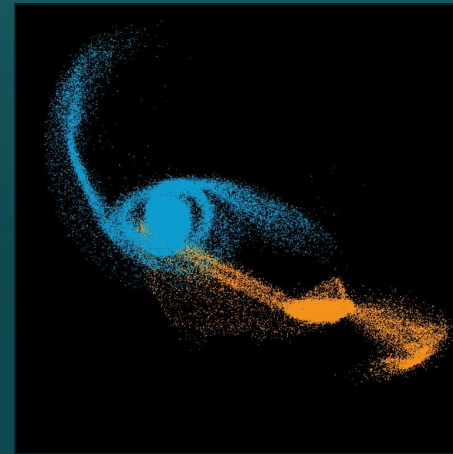
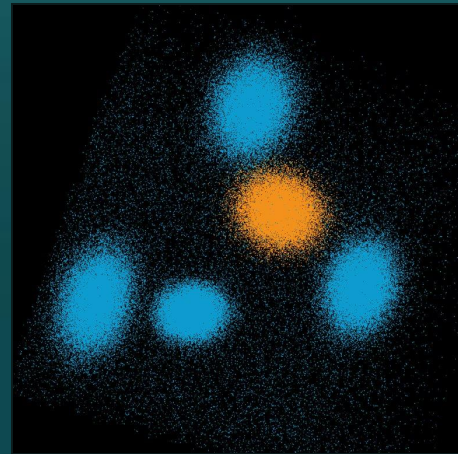
	Clusters	Galaxies	Shell/Core	Strings
CloudLasso	.9789	.9866	.9980	.7494
CylinderSelection	.9759	.9855	.9960	.7303
Z	2.67	0.71	3.06	2.04
p	<.01	.48	<.01	.041

Study Results – Errors: MCC



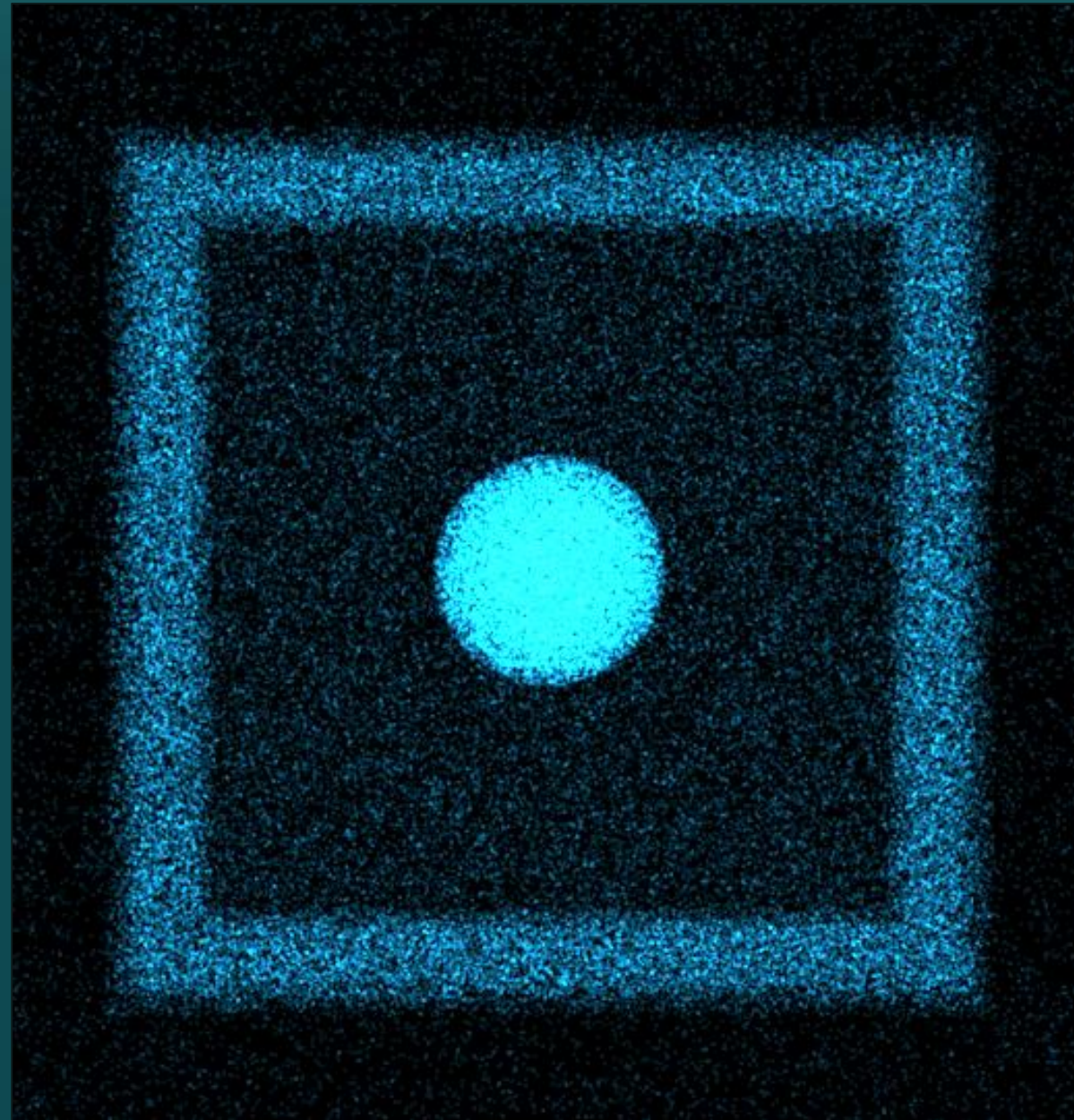
	Clusters	Galaxies	Shell/Core	Strings
CloudLasso	.9765	.9733	.9974	.6519
CylinderSelection	.9731	.9712	.9948	.6305
Z	2.75	0.78	3.06	1.89
p	<.01	.43	<.01	.06

Study Results – Errors: $V_{\text{selected}} / V_{\text{real}}$

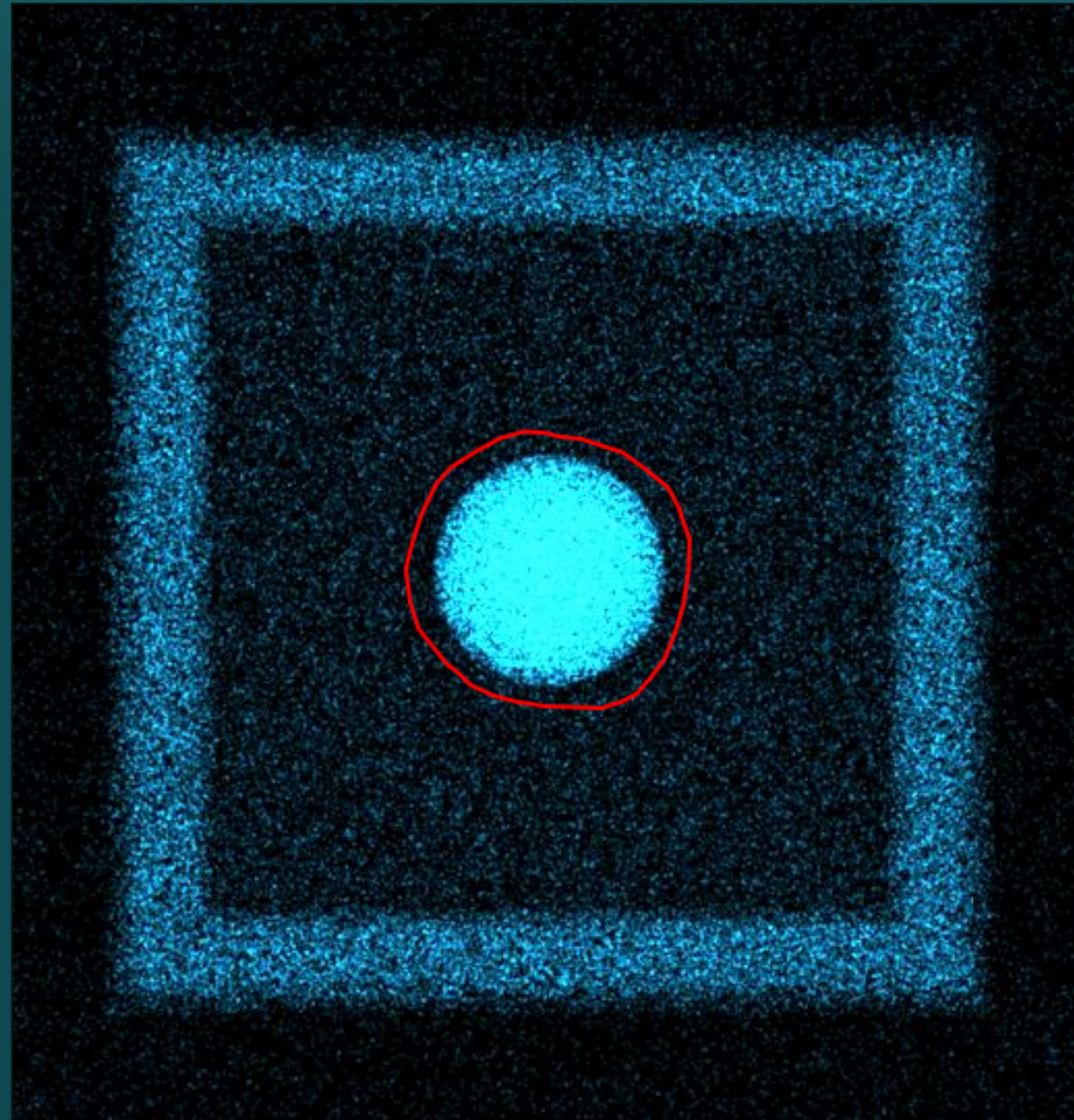


	Clusters	Galaxies	Shell/Core	Strings
CloudLasso	1.244	4.055	1.327	1.852
CylinderSelection	1.303	5.855	1.360	2.691
Z	0.94	1.49	1.57	2.98
p	.347	.14	.875	<.01

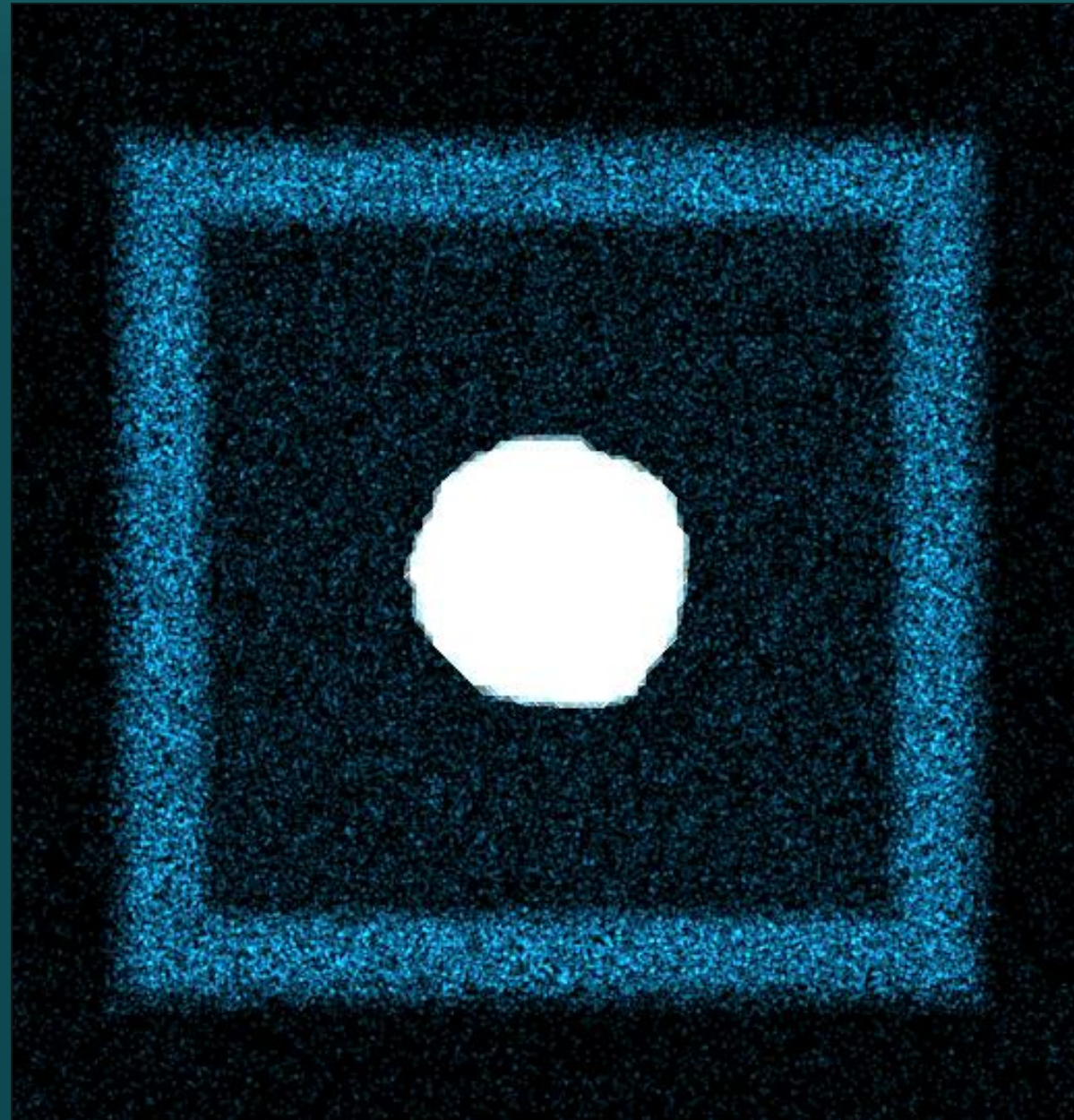
Problem with Structure-Aware Subtraction



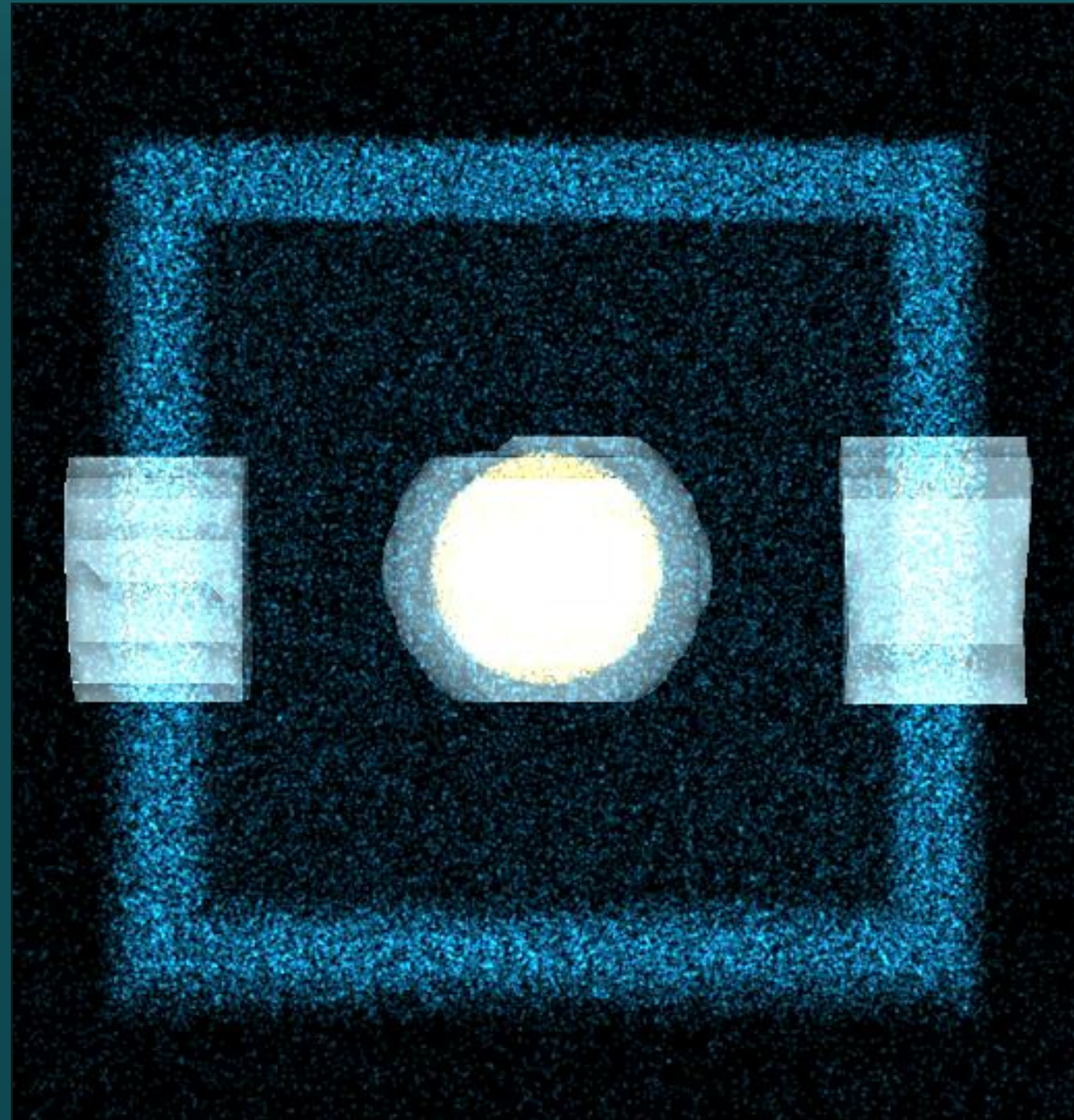
Problem with Structure-Aware Subtraction



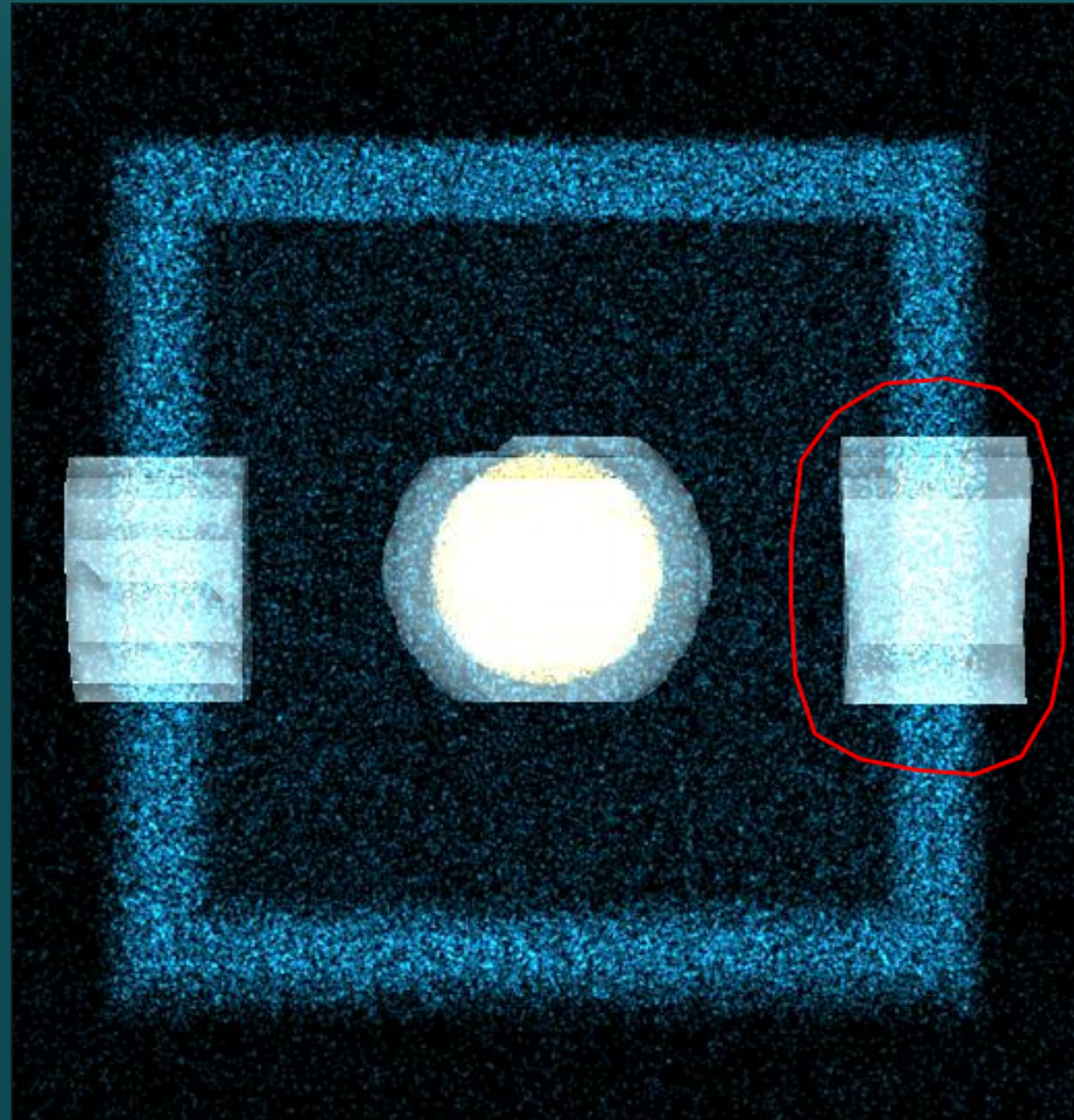
Problem with Structure-Aware Subtraction



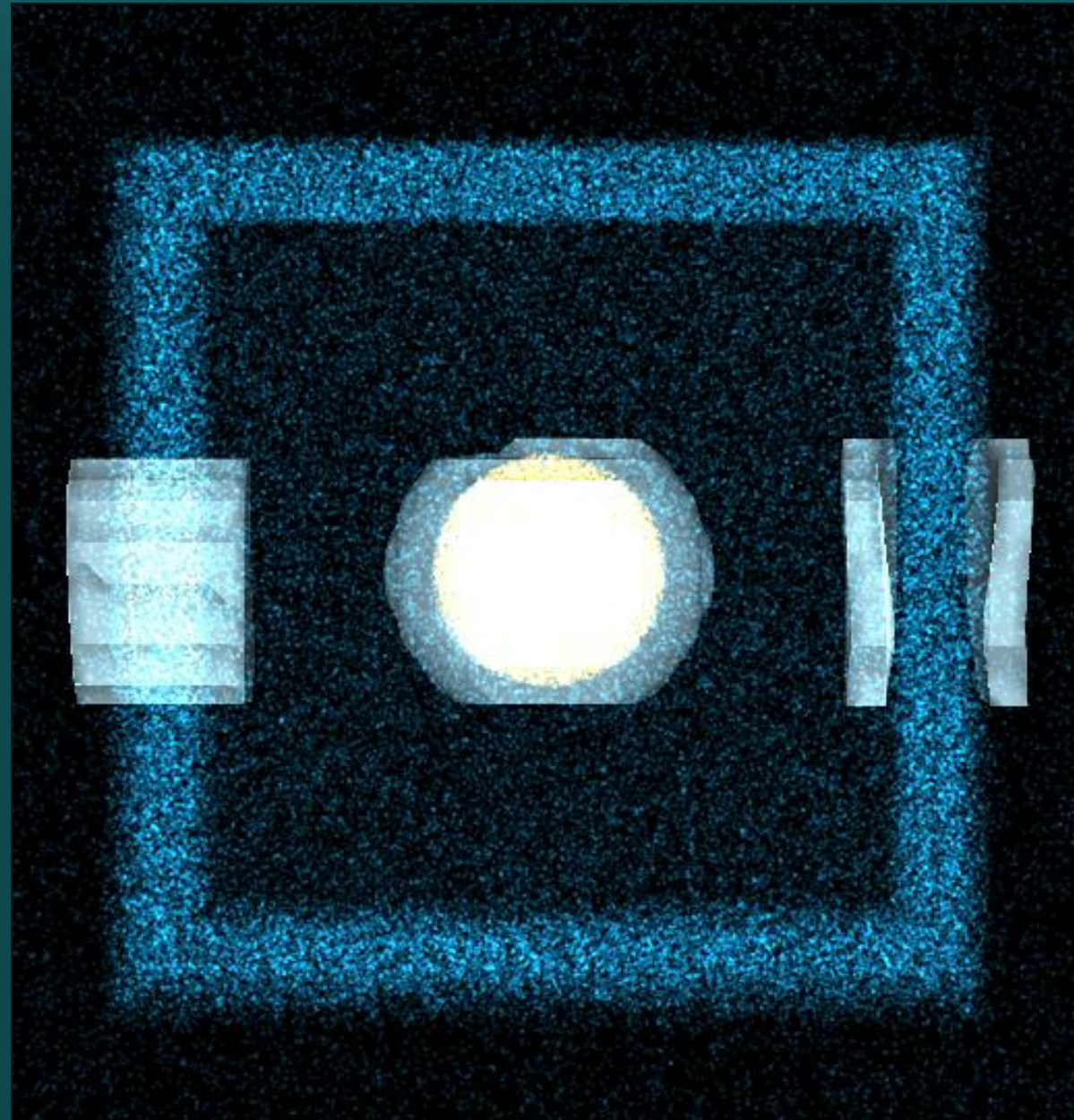
Problem with Structure-Aware Subtraction



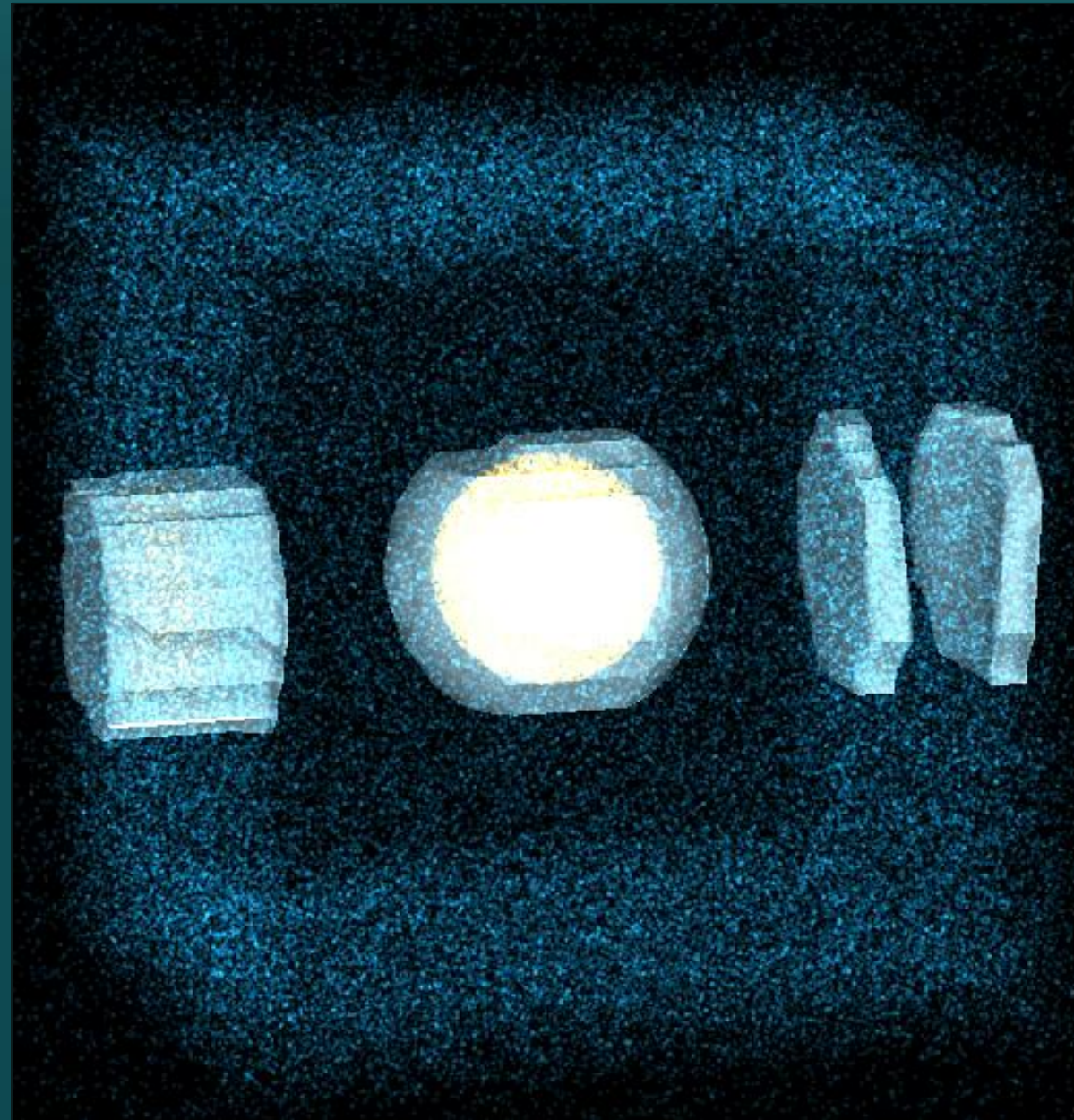
Problem with Structure-Aware Subtraction



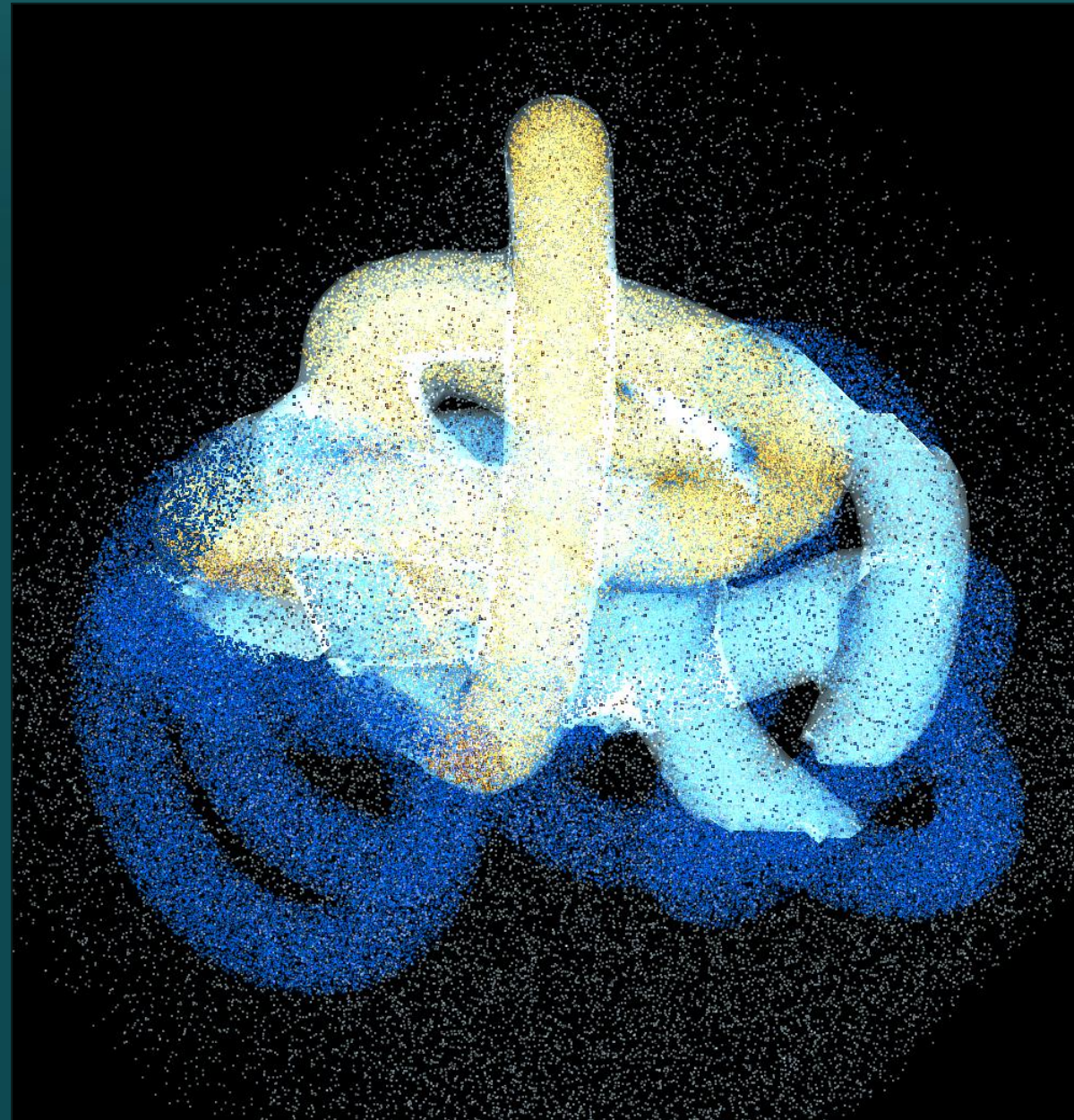
Problem with Structure-Aware Subtraction



Problem with Structure-Aware Subtraction



Problem with Structure-Aware Subtraction



Problem with Structure-Aware Subtraction

