# Cosmic Web Reconstruction through Density Ridges

#### Yen-Chi Chen

Shirley Ho Peter E. Freeman Christopher R. Genovese Larry Wasserman

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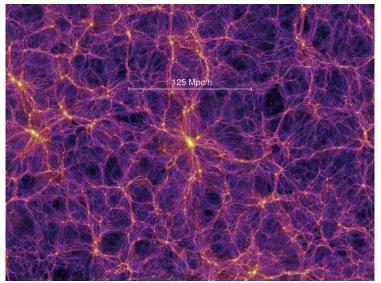
> > August 31, 2015

- Introduction
- Model and Algorithm
- Analysis

# Introduction

- Model and Algorithm
- Analysis

### Cosmic Web: What Does Our Universe Look Like

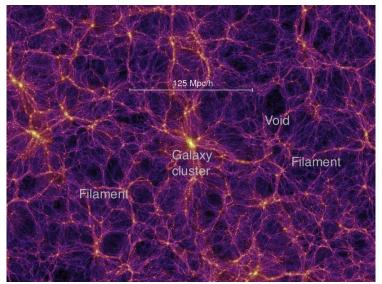


Credit: Millennium Simulation

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**Cosmic Web Reconstruction** 

### Cosmic Web: What Does Our Universe Look Like



Credit: Millennium Simulation

Why filament?

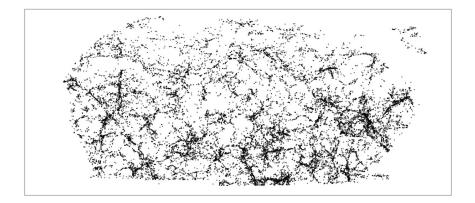
• Galaxies tend to concentrate around filaments.

Why filament?

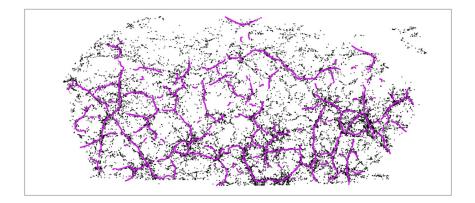
- Galaxies tend to concentrate around filaments.
- Several properties of a galaxy are influenced by filaments.

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### An Example



### An Example



### Statistical Model for Filaments: Density Ridges

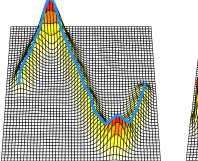
#### Formally, we define a filament to be a ridge of the density.

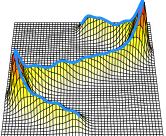
# Example: Ridges in Mountains



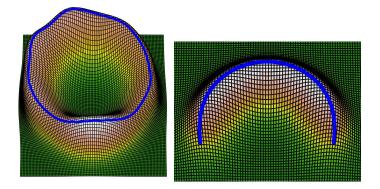
Credit: Google

### Example: Ridges in Smooth Functions

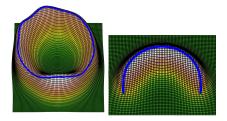




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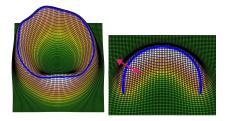


# Ridges: Local Modes in Subspace



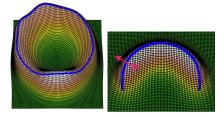
• A generalized local mode in a specific 'subspace'.

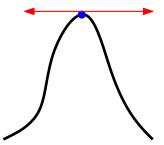
# Ridges: Local Modes in Subspace



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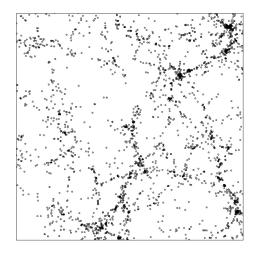
# Ridges: Local Modes in Subspace



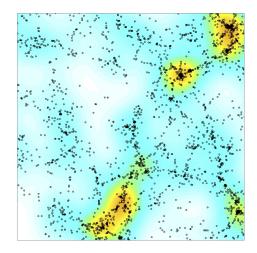


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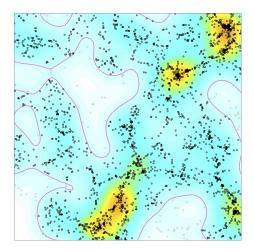
Rawdata



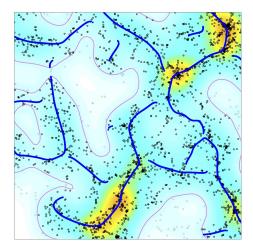
- Rawdata
- 2 Density Reconstruction

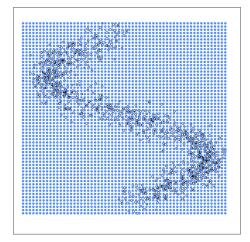


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- Thresholding



- Rawdata
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- Indege Recovery

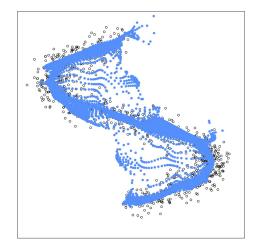




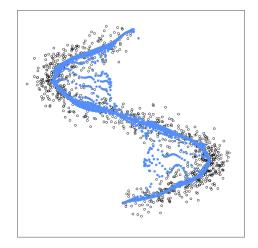
This is proposed in Ozertem and Erdogmus (2011).



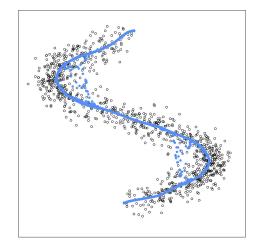
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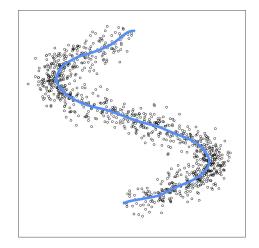
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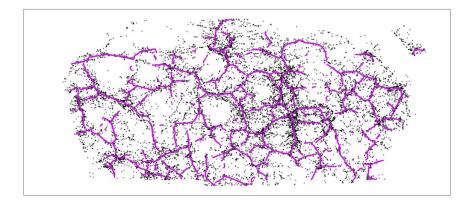


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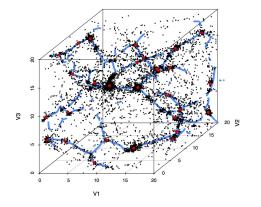
### Density Ridges on an Example



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### MassiveBlack-II Simulation

• Method: smoothed particle hydrodynamics.



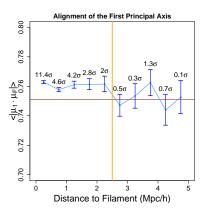
This is a joint work with Ananth Tenneti, Rachel Mandelbaum, Rupert Croft, and Tiziana DiMatteo.

Yen-Chi Chen (CMU-Stats)

Cosmic Web Reconstruction

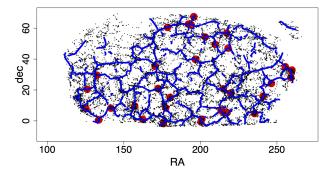
### Galaxy Alignment to Filaments

- Principal axes for a galaxy  $(\mu_1)$ .
- Orientation of the nearest filament  $(\mu_F)$ .
- Distance to the nearest filament  $(d_F)$ .



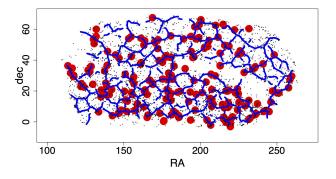
### Sloan Digital Sky Survey

- Data: the Sloan Digital Sky Survey, data release 12.
- We take 2-D slices of the Universe to detect filaments ( $\Delta z = 0.005$ ).
- Blue: filaments. Red: galaxy clusters (redMaPPer).



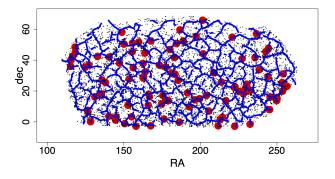
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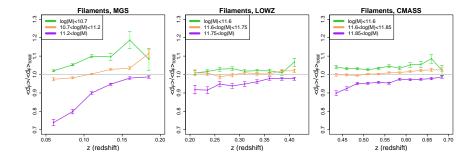
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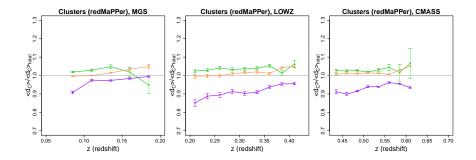


- Mass from Flexible Stellar Population Synthesis method [Conroy, Gunn, and White 2009].
- We partition galaxies into three mass-bins according to their mass.
- We compare the average distance to filaments from each bin.

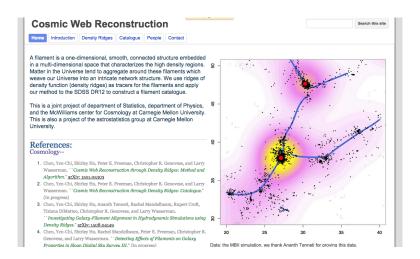
### SDSS: Stellar Mass of Galaxies



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### A Filament Catalogue for the SDSS



#### https://sites.google.com/site/yenchicr/

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Thank you!

- Chen, Yen-Chi, Shirley Ho, Peter E. Freeman, Christopher R. Genovese, and Larry Wasserman. "Cosmic Web Reconstruction through Density Ridges: Method and Algorithm." To appear in Monthly Notices of the Royal Astronomical Society..
- Chen, Yen-Chi, et al. "Investigating Galaxy-Filament Alignments in Hydrodynamic Simulations using Density Ridges." arXiv preprint arXiv:1508.04149 (2015).
- Chen, Yen-Chi, Christopher R. Genovese, and Larry Wasserman. "Asymptotic theory for density ridges." The Annals of Statistics 43.5 (2015): 1896-1928.
- Conroy, Charlie, James E. Gunn, and Martin White. "The propagation of uncertainties in stellar population synthesis modeling. I. The relevance of uncertain aspects of stellar evolution and the initial mass function to the derived physical properties of galaxies." The Astrophysical Journal 609.1 (2009): 486.
- 5. Eberly, David. Ridges in image and data analysis. Vol. 7. Springer Science & Business Media, 1996.
- 6. Genovese, Christopher R., et al. "Nonparametric ridge estimation." The Annals of Statistics 42.4 (2014): 1511-1545.
- 7. Ozertem, Umut, and Deniz Erdogmus. "Locally defined principal curves and surfaces." The Journal of Machine

Learning Research 12 (2011): 1249-1286.

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Local modes:

$$\mathsf{Mode}(p) = \{x : \nabla p(x) = 0, \lambda_1(x) < 0\}.$$

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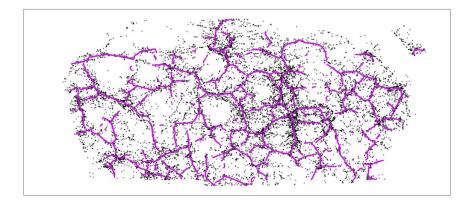
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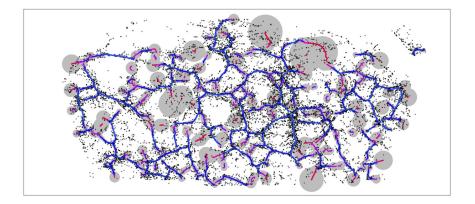
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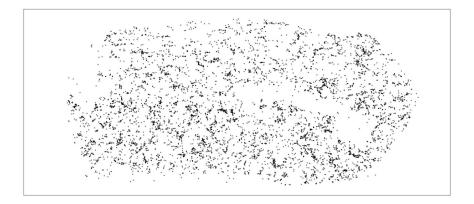
 →Subspace Constrained Mean Shift Algorithm [Ozertem and Erdogmus 2011].

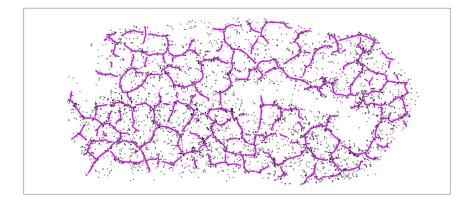
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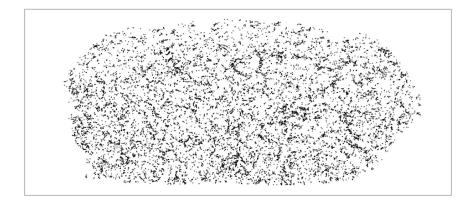


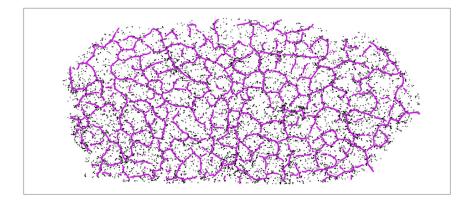
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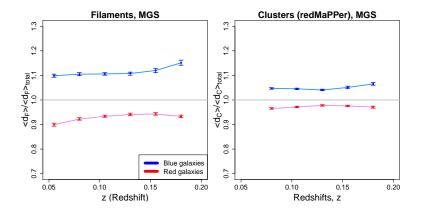






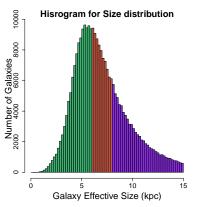
- Redshift range: 0.05 < z < 0.20 (main sample galaxy).
- Color cut: (g r) = 0.8.

# SDSS: Red and Blue Galaxies

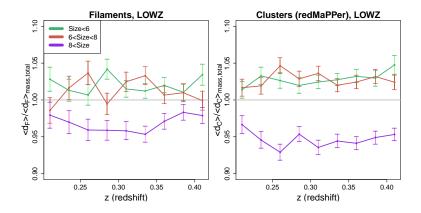


# SDSS: Size for Galaxies

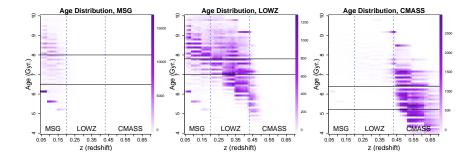
- Size: Effective Radii.
- 2 Data: LOWZ (0.20 < z < 0.43)</p>
- Partitioning galaxies into three groups according to their size.



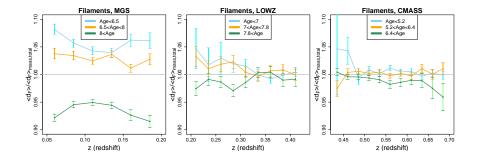
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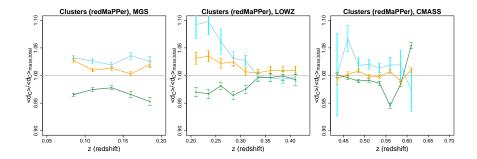
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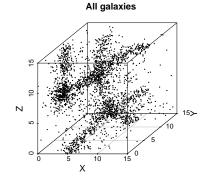


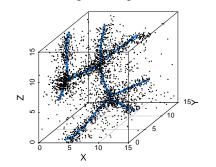
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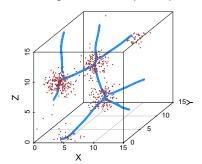
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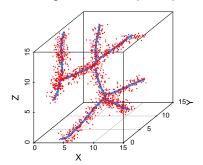




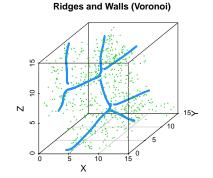
**Ridges and all galaxies** 

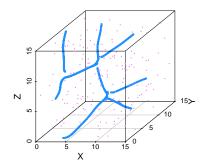


**Ridges and Clusters (Voronoi)** 



**Ridges and Filaments (Voronoi)** 





Ridges and Voids (Voronoi)