



# The multi-stream local Universe

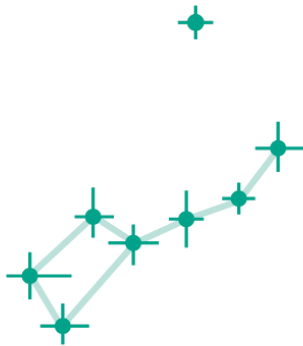
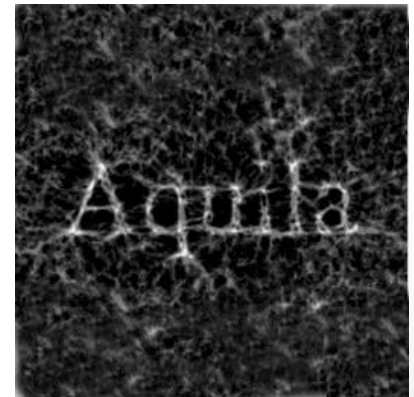
**Florent Leclercq**

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Imperial Centre for Inference and Cosmology  
Imperial College London

Guilhem Lavaux, Jens Jasche,  
Alan Heavens, James Prideaux-Ghee,  
and the Aquila Consortium  
[www.aquila-consortium.org](http://www.aquila-consortium.org)

28 January 2020



# The BORG inference framework

*Bayesian Origin Reconstruction from Galaxies*

- A **Bayesian Hierarchical Model**:

$$\mathcal{P}(\hat{\delta}) \propto \exp \left( -\frac{1}{2} \sum_k \frac{|\hat{\delta}_k|^2}{P_k} \right) \quad \text{initial conditions}$$

$$\rho_{\text{m}} = \mathcal{F}(\delta) \quad \text{total evolved matter density}$$

$$\rho_{\text{g}} = \mathcal{B}(\rho_{\text{m}}) \quad \text{biased galaxy distribution}$$

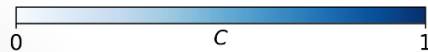
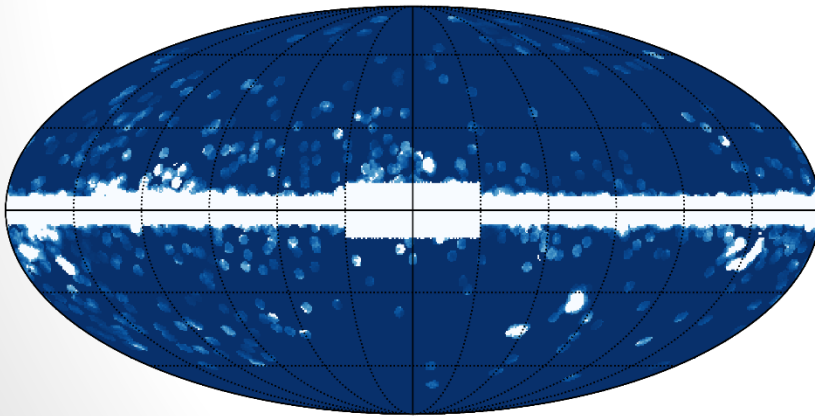
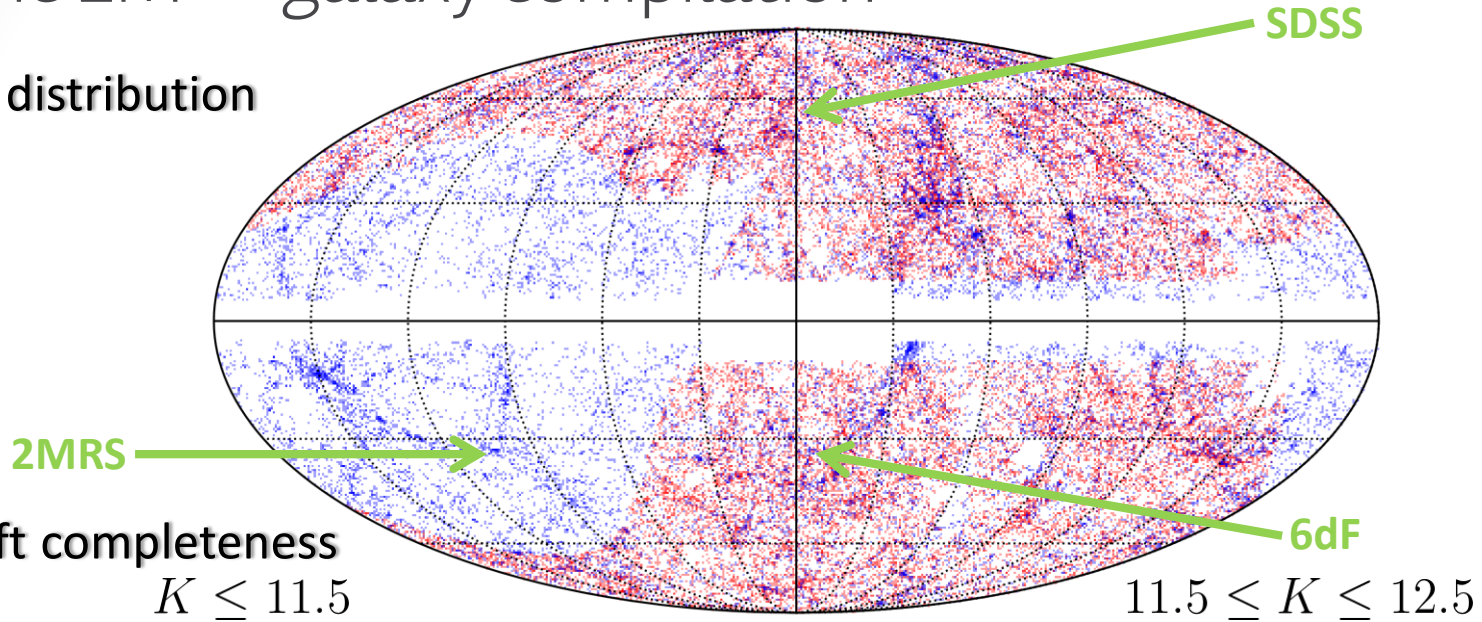
$$\rho_{\text{g}}^{\text{s}}(\vec{x}) = S(\vec{x})\rho_{\text{g}}(\vec{x}) \quad \text{selected sample}$$

$$N_{\text{g}} \curvearrowright \mathcal{P}(N_{\text{g}}|\rho_{\text{g}}^{\text{s}}) \quad \begin{array}{l} \text{galaxy number count:} \\ \text{random extraction (Poisson,} \\ \text{Negative Binomial)} \end{array}$$

- The multi-million dimensional posterior distribution is sampled via **Hamiltonian Monte Carlo**.

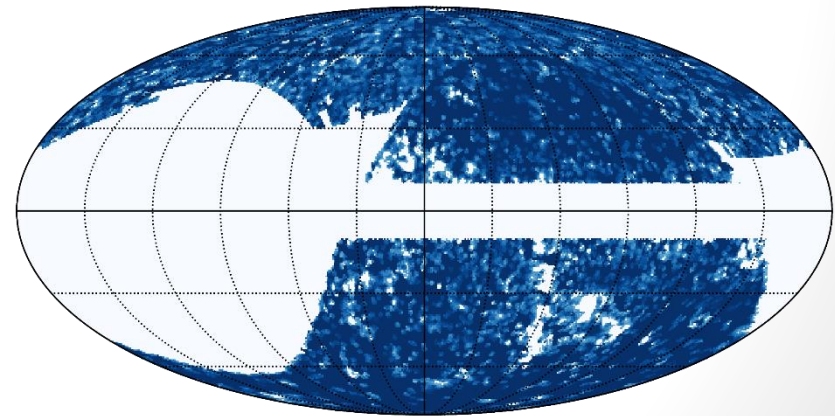
# The 2M++ galaxy compilation

Galaxy distribution



Lavaux & Hudson 2011, 1105.6107

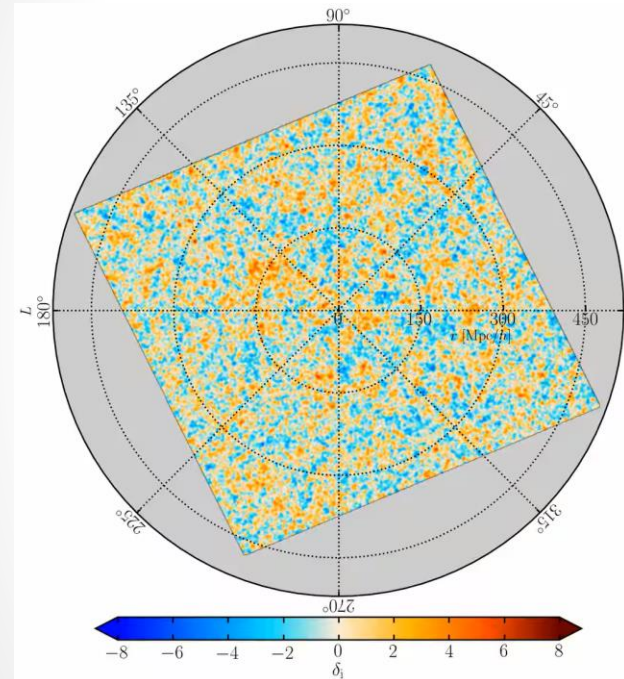
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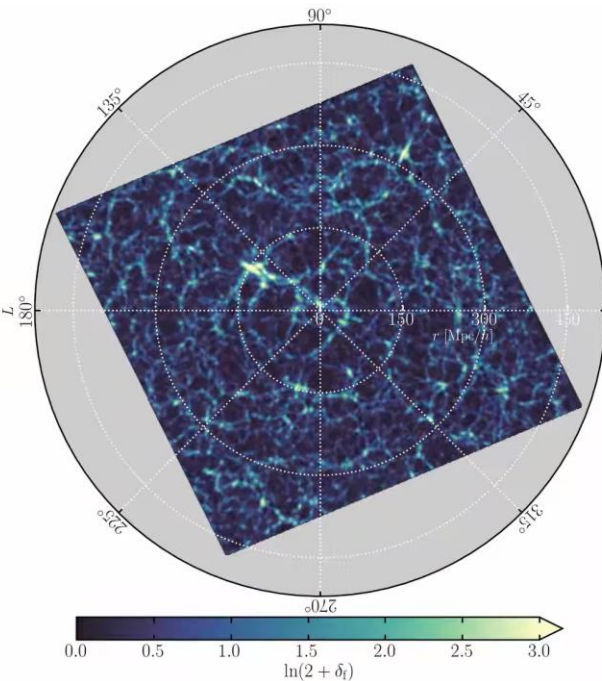
The multi-stream local Universe

# BORG at work: Bayesian chrono-cosmography

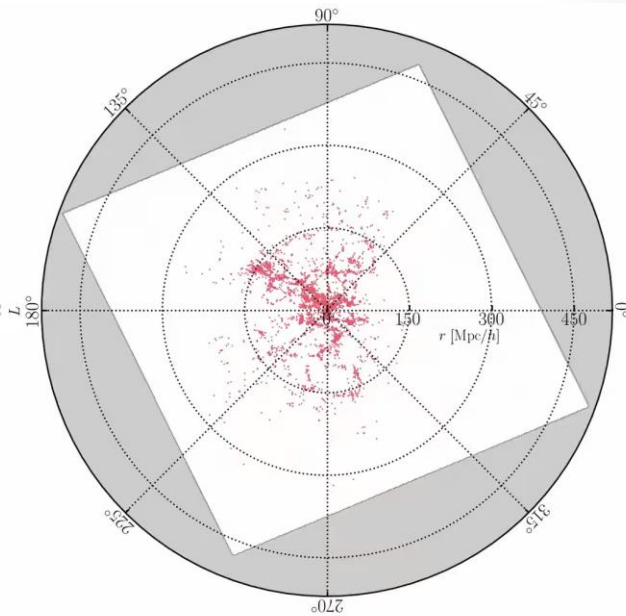
Initial conditions



Final conditions



Observations



Supergalactic plane

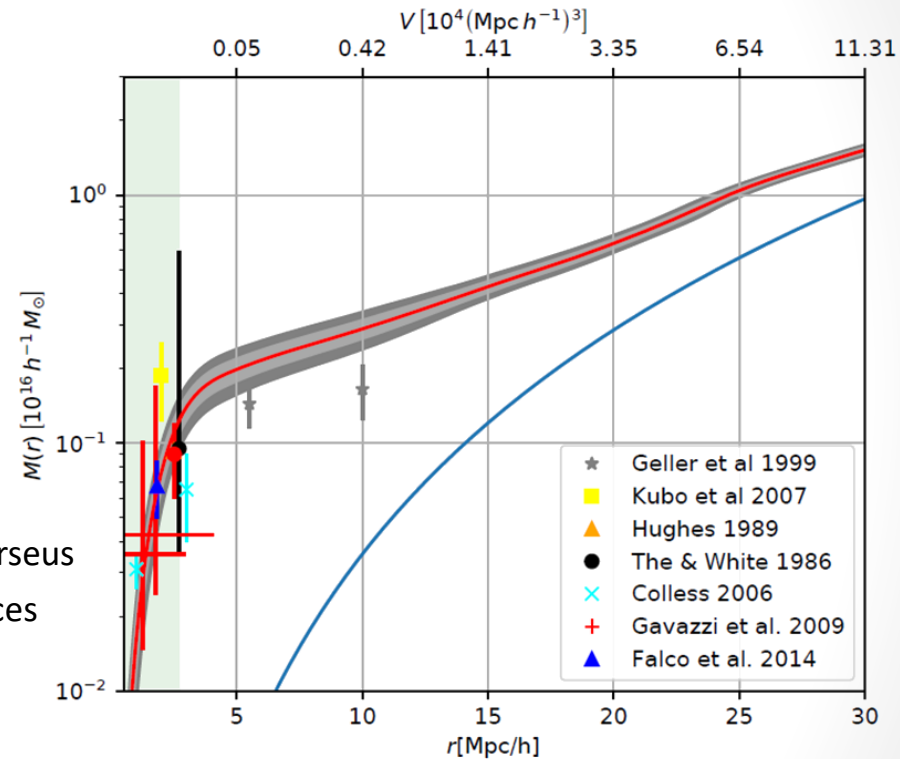
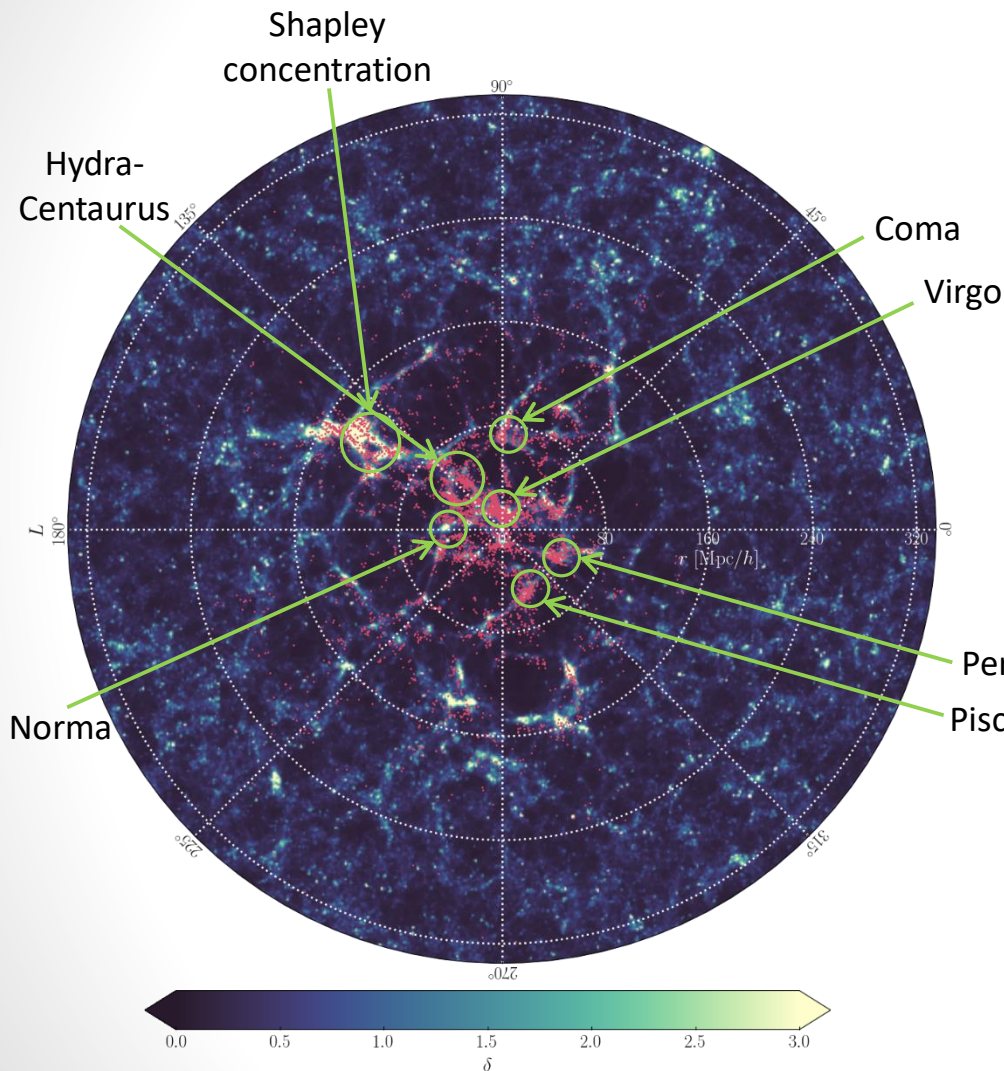
67,224 galaxies,  $\approx 17$  million parameters, 5 TB of primary data products, 10,000 samples,  $\approx 500,000$  forward and adjoint data model evaluations, 1.5 million CPU-hours

Jasche & Lavaux 2019, 1806.11117

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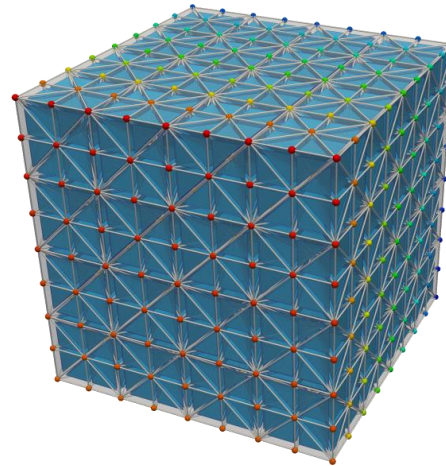
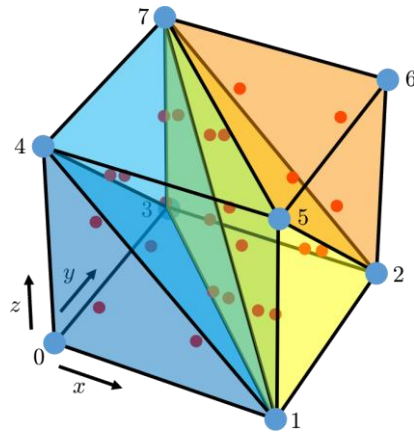
# BORGPM density field: full non-linear dynamics



Mass profile of the **Coma cluster**, in agreement with gravitational lensing and X-ray observations down to a few Mpc.

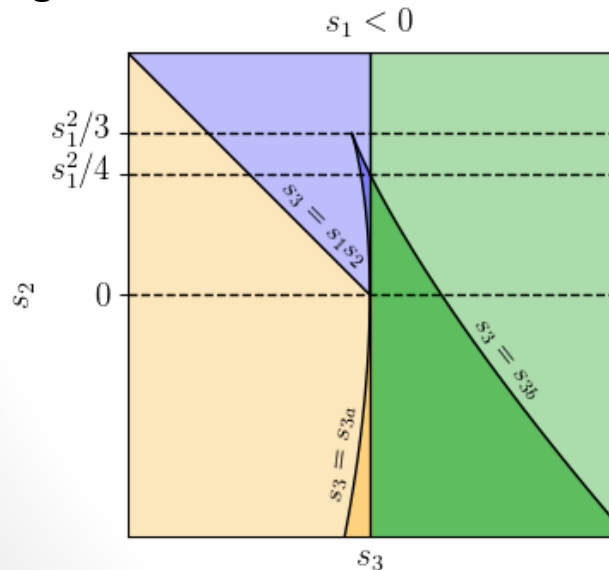
# The phase-space structure of dark matter: tools

Delaunay  
tessellation of  
elementary  
Lagrangian cubes  
(Simplex-In-Cell  
estimator)



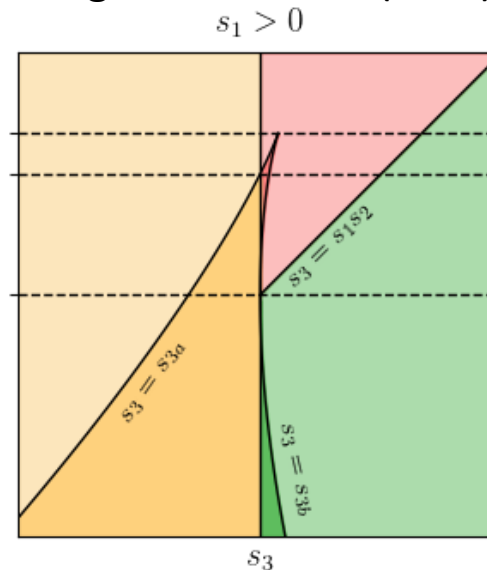
Abel, Hahn & Kaehler 2012, 1111.3944  
Shandarin, Habib & Heitmann 2012, 1111.2366  
Hahn, Abel & Kaehler 2013, 1210.6652  
Hahn & Angulo 2016, 1501.01959  
Sousbie & Colombi 2016, 1509.07720

Lagrangian Invariants Classification of Heterogeneous flows (LICH)



FL, Jasche, Lavaux, Wandelt & Percival 2017, 1601.00093

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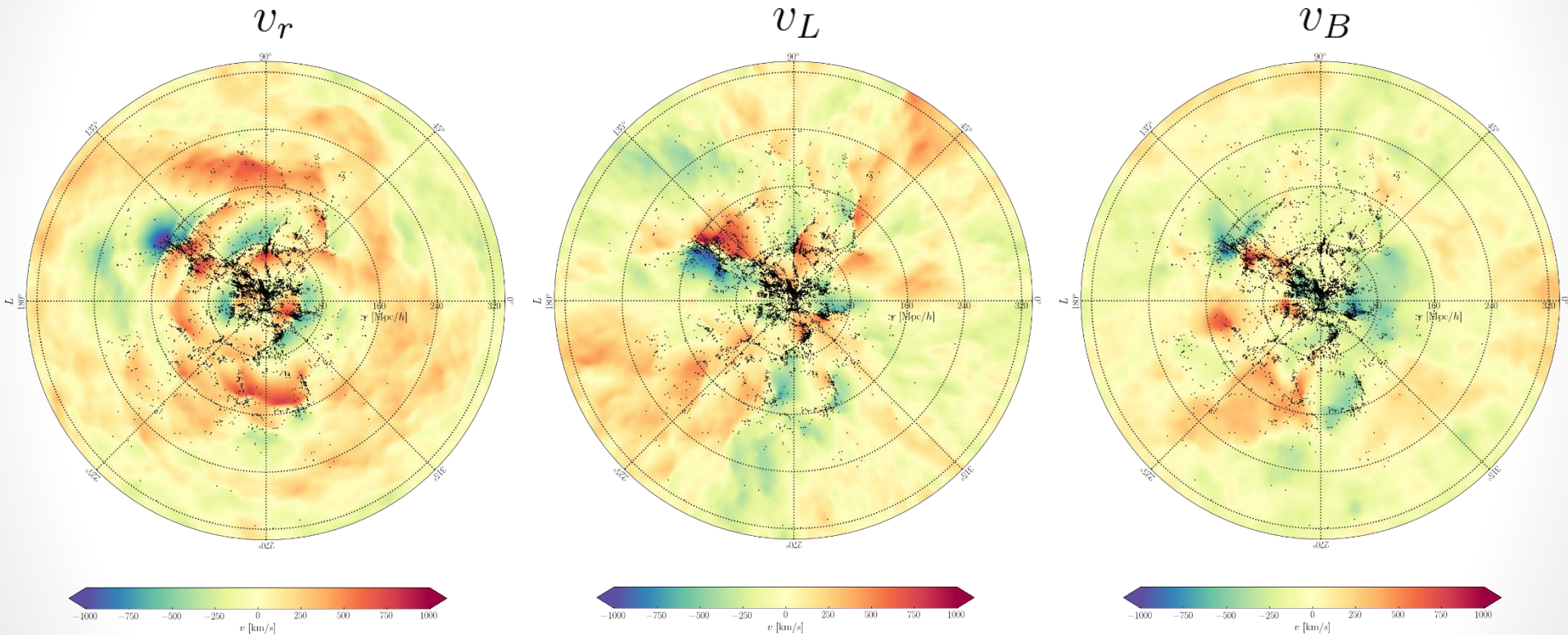
$$\mathcal{R}_{\ell m} \equiv \frac{\partial \Psi_{\ell}}{\partial \mathbf{q}_m}$$

$$\lambda^3 + \boxed{s_1} \lambda^2 + \boxed{s_2} \lambda + \boxed{s_3} = 0$$

- potential clusters
- vortical clusters
- potential filaments
- vortical filaments
- potential sheets
- vortical sheets
- potential voids
- vortical voids

Generalises DIVA,  
Lavaux & Wandelt 2010, 0906.4101

# Velocity field in the supergalactic plane



The **gravitational infall** of known structures can be observed.

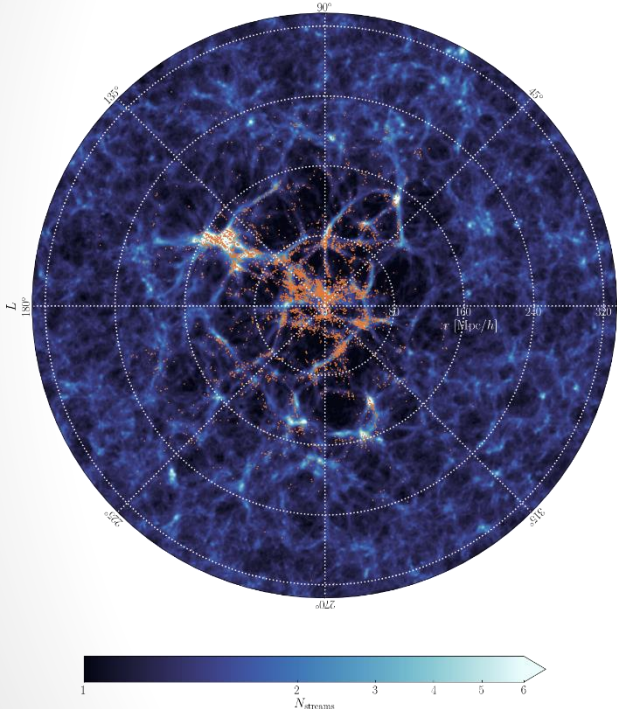
Prideaux-Ghee, FL, Heavens, Lavaux & Jasche, in prep.

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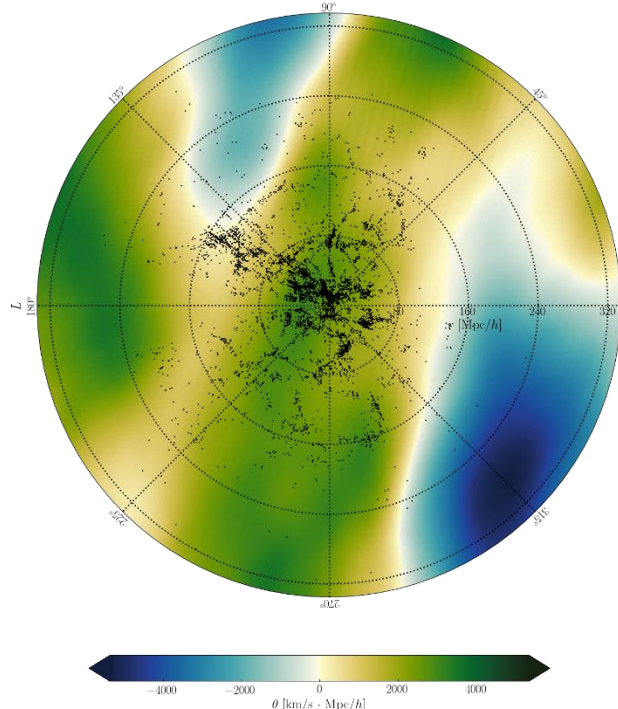
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# Number of streams and vorticity

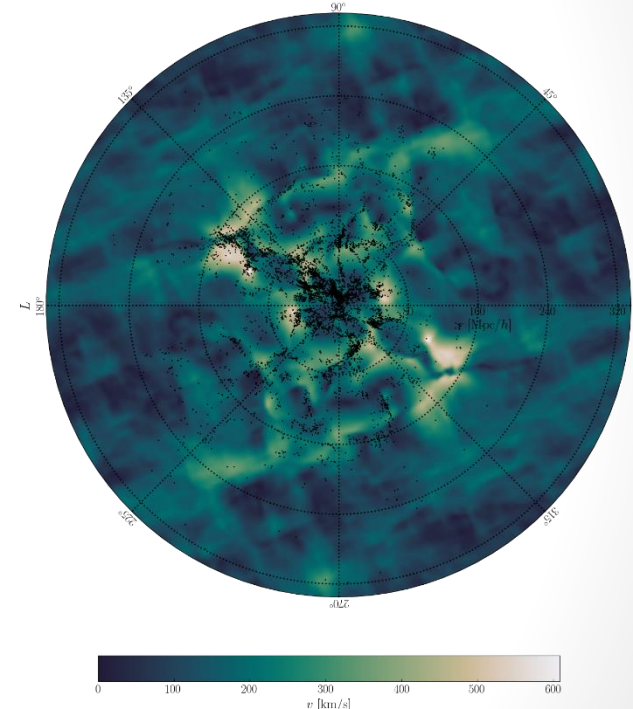
Number of streams



Velocity potential



Norm of vorticity

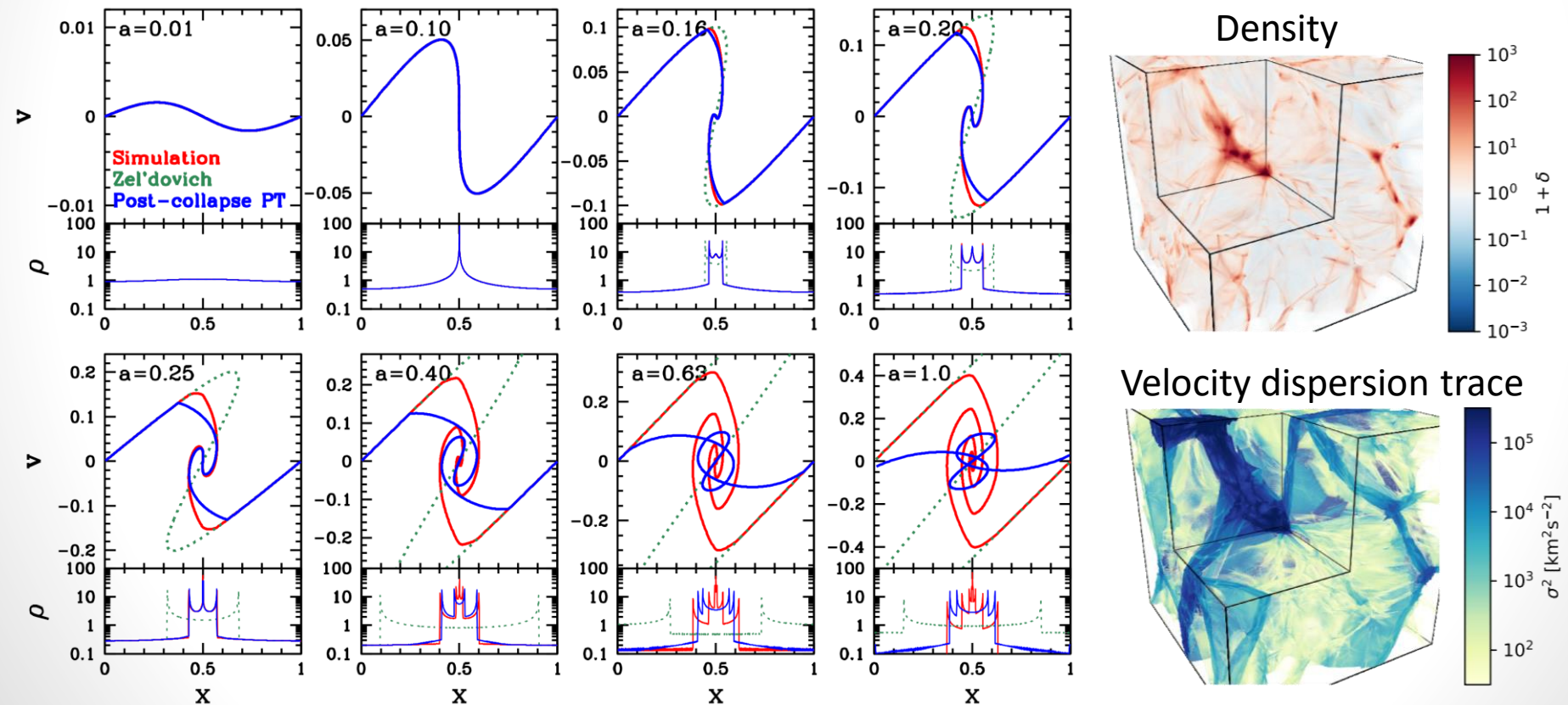


In earlier work ( FL, Jasche, Lavaux, Wandelt & Percival 2017, arXiv:1601.00093 ),  
these were postdictions. Thanks to **BORGPM** (full non-linear dynamics),  
we have now actual **measurements** - with uncertainties.

Prideaux-Ghee, FL, Heavens, Lavaux & Jasche, in prep.

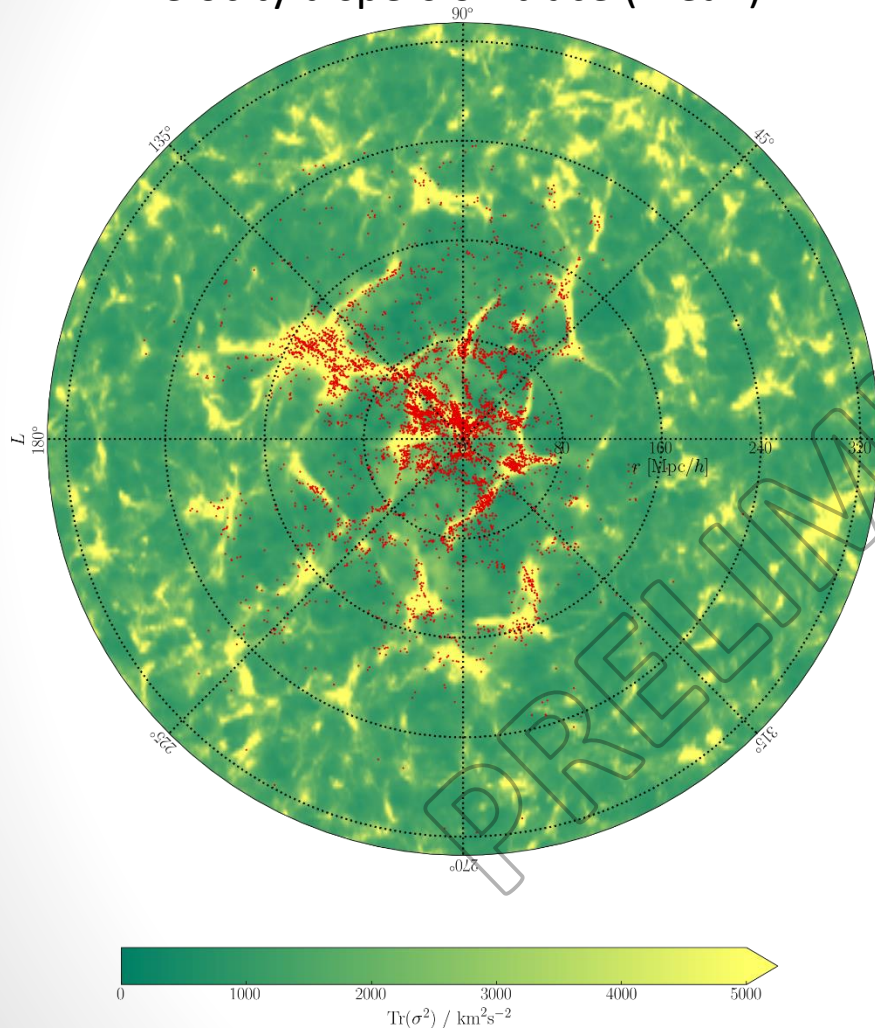
# The multi-stream regime and velocity dispersion

- The breakdown of  $\sigma_{ij} \approx 0$ , describing the generation of velocity dispersion or anisotropic stress due to the multiple-stream regime, is generically known as **shell-crossing**.

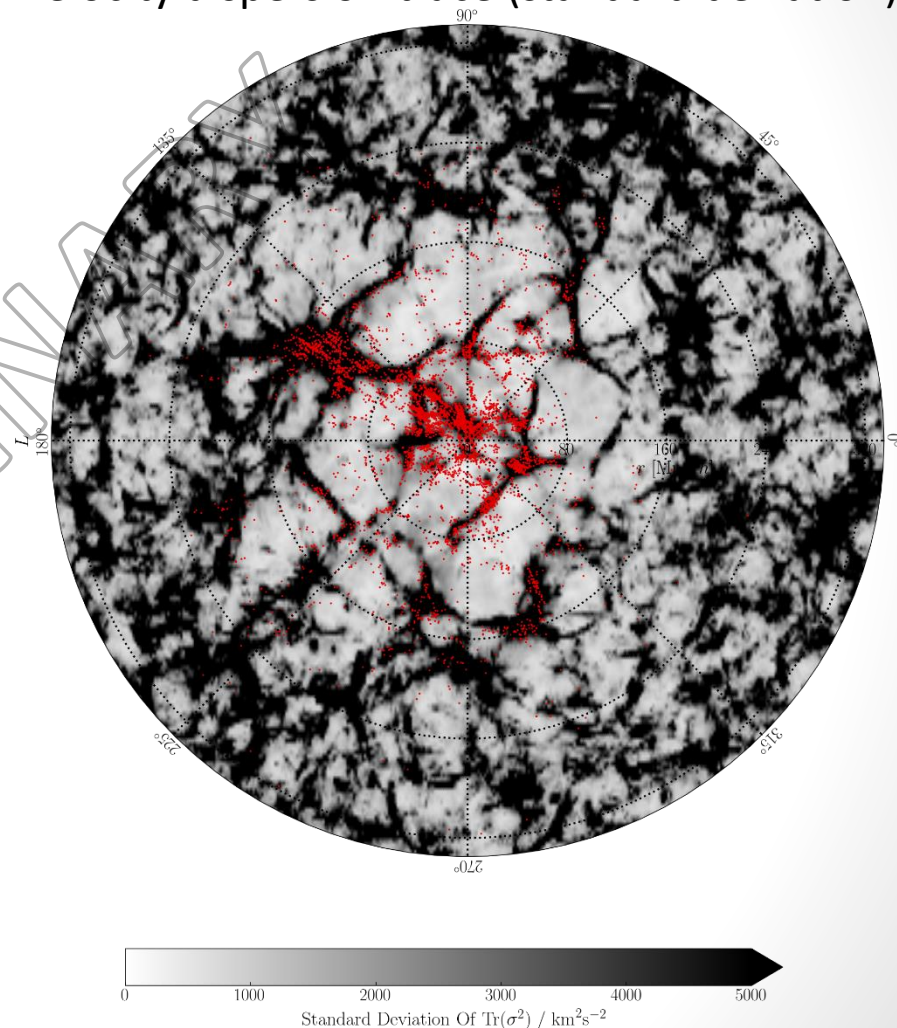


# Velocity dispersion in the local Universe

Velocity dispersion trace (mean)



Velocity dispersion trace (standard deviation)



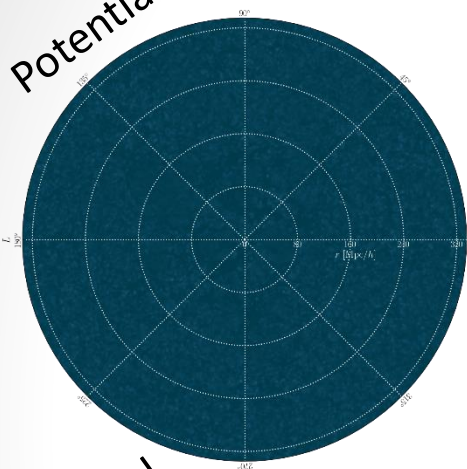
Prideaux-Ghee, FL, Heavens, Lavaux & Jasche, in prep.

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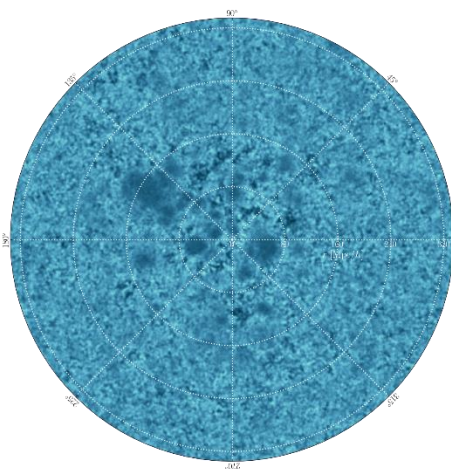
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# LICH initial structures inferred by BORG

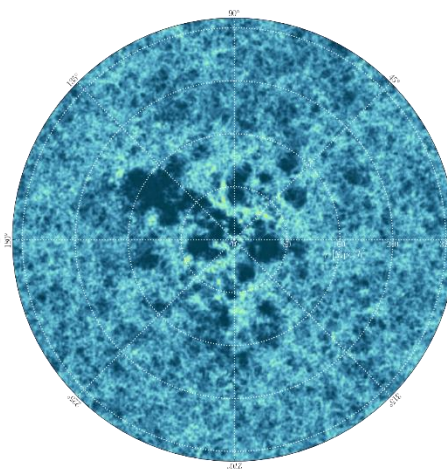
Potential Clusters



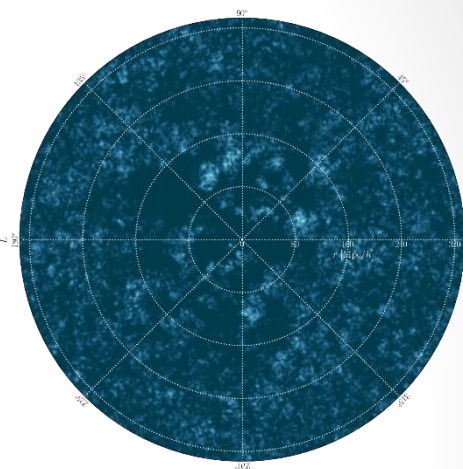
Filaments



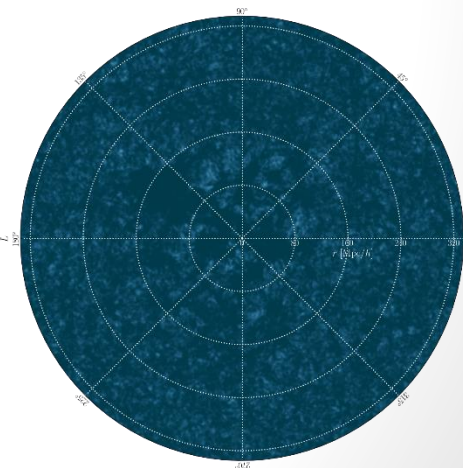
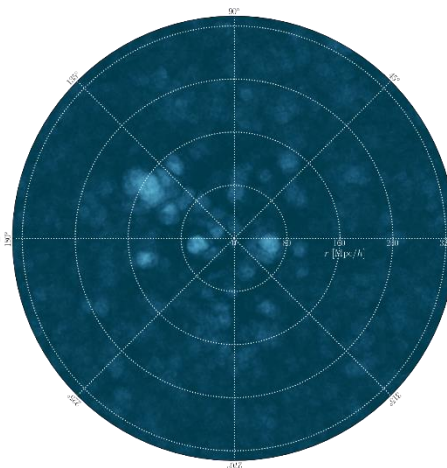
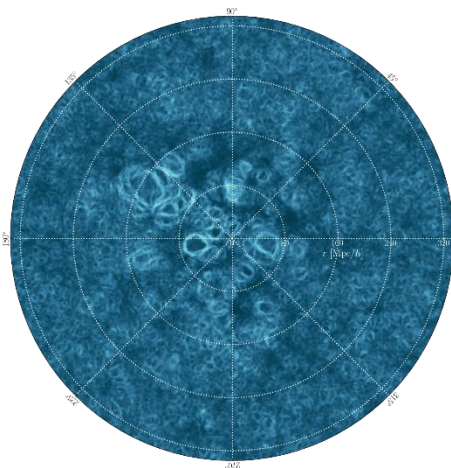
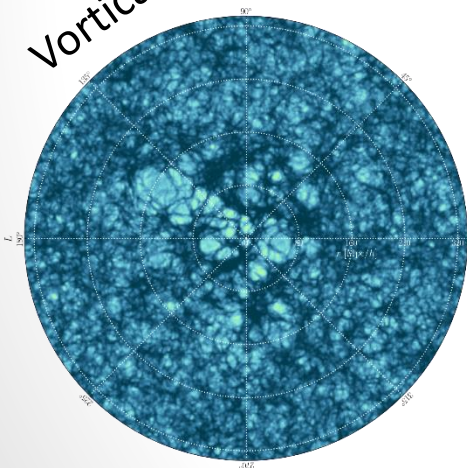
Sheets



Voids



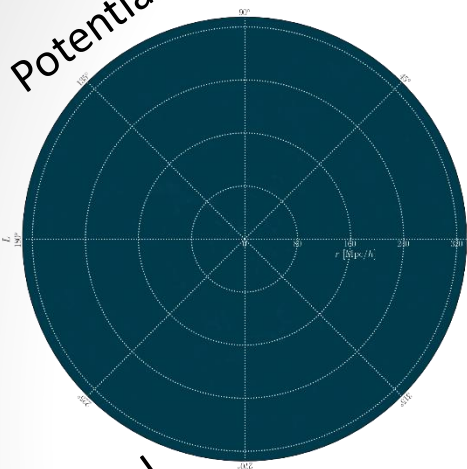
Vortical



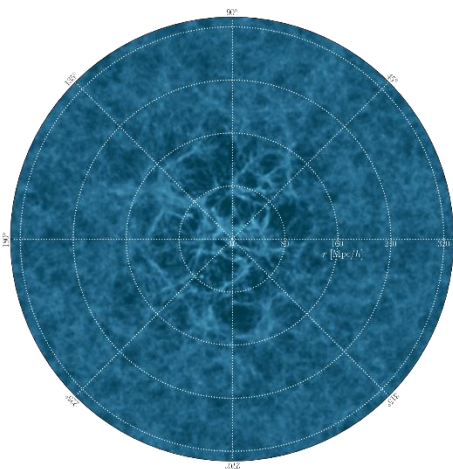
Prideaux-Ghee, FL, Heavens, Lavaux & Jasche, in prep.

# LICH final structures inferred by BORG

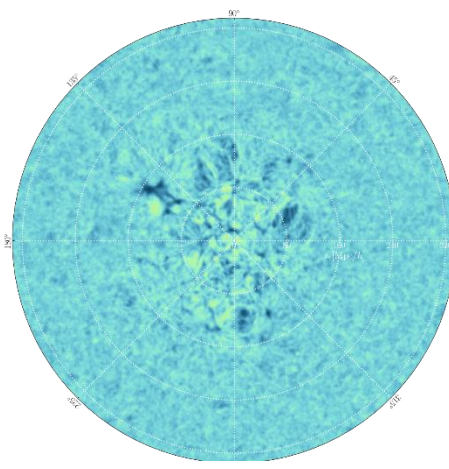
Potential Clusters



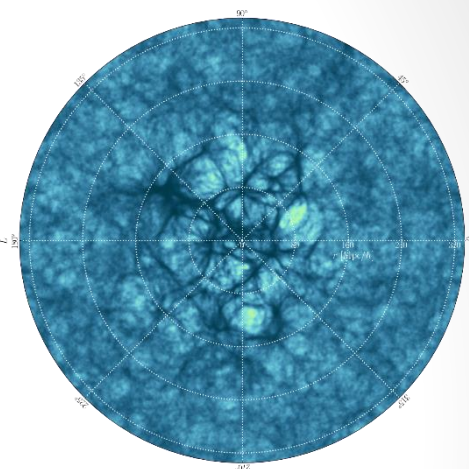
Filaments



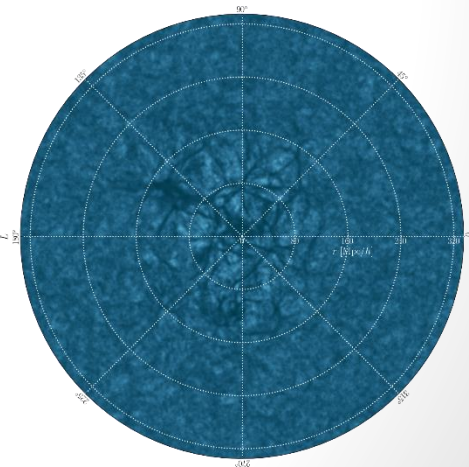
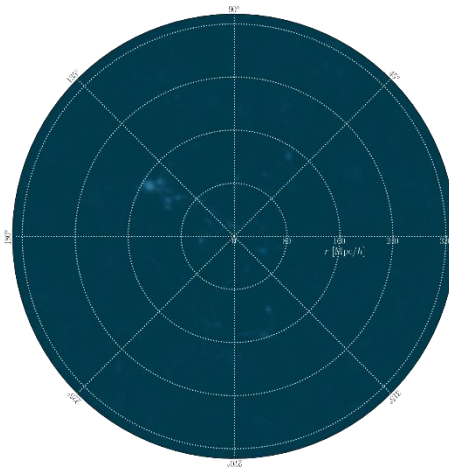
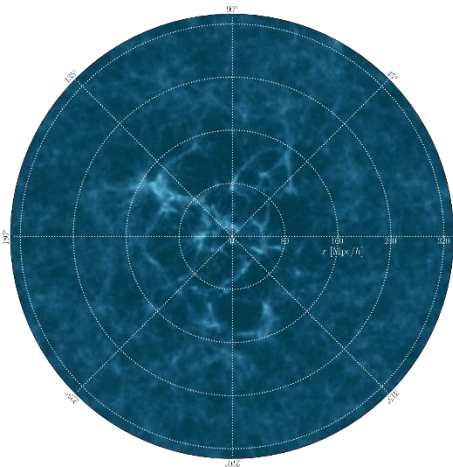
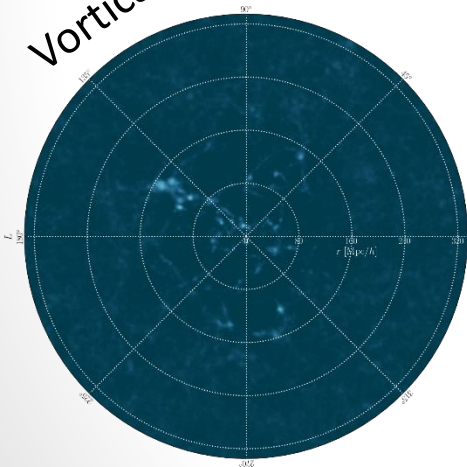
Sheets



Voids



Vortical



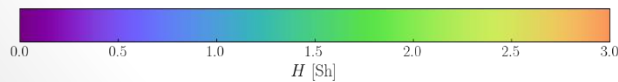
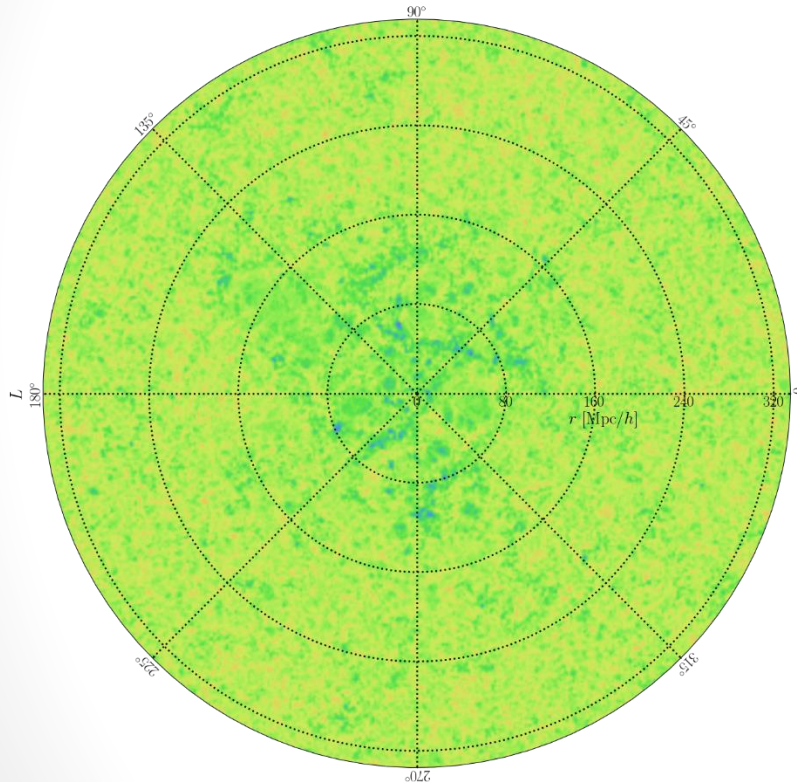
Prideaux-Ghee, FL, Heavens, Lavaux & Jasche, in prep.

# How is information propagated?

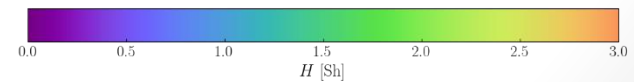
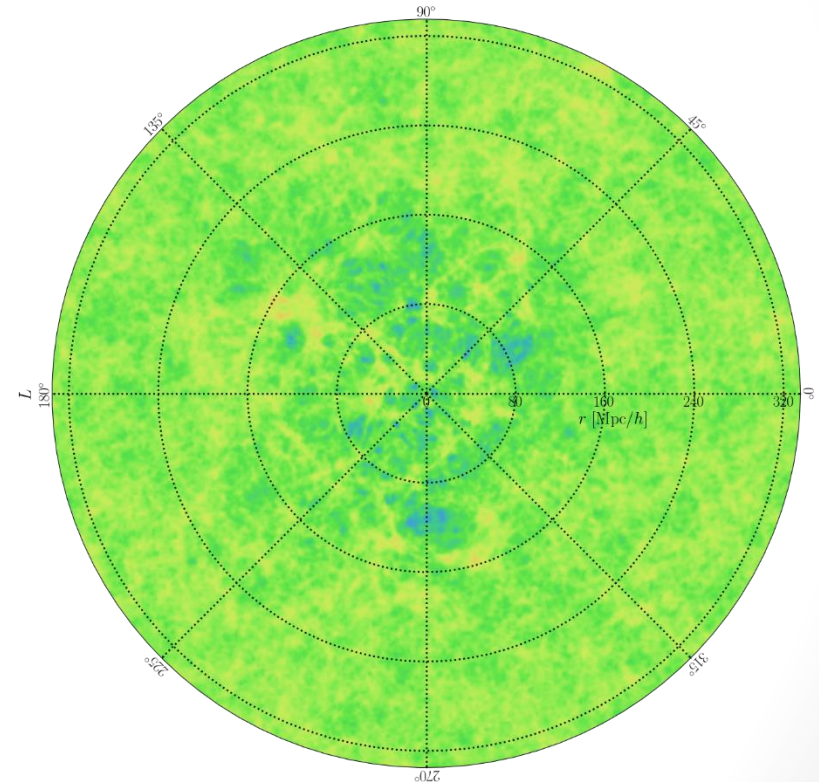
Shannon entropy:

$$H[\mathcal{P}(\mathbf{T}(\vec{x})|d)] \equiv - \sum_{i=0}^7 \mathcal{P}(\mathbf{T}_i(\vec{x})|d) \log_2(\mathcal{P}(\mathbf{T}_i(\vec{x})|d)) \quad \text{in shannons (Sh)}$$

Initial conditions



Final conditions



FL, Lavaux & Jasche, in prep.

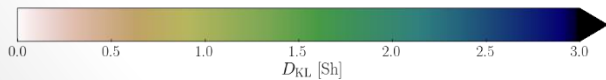
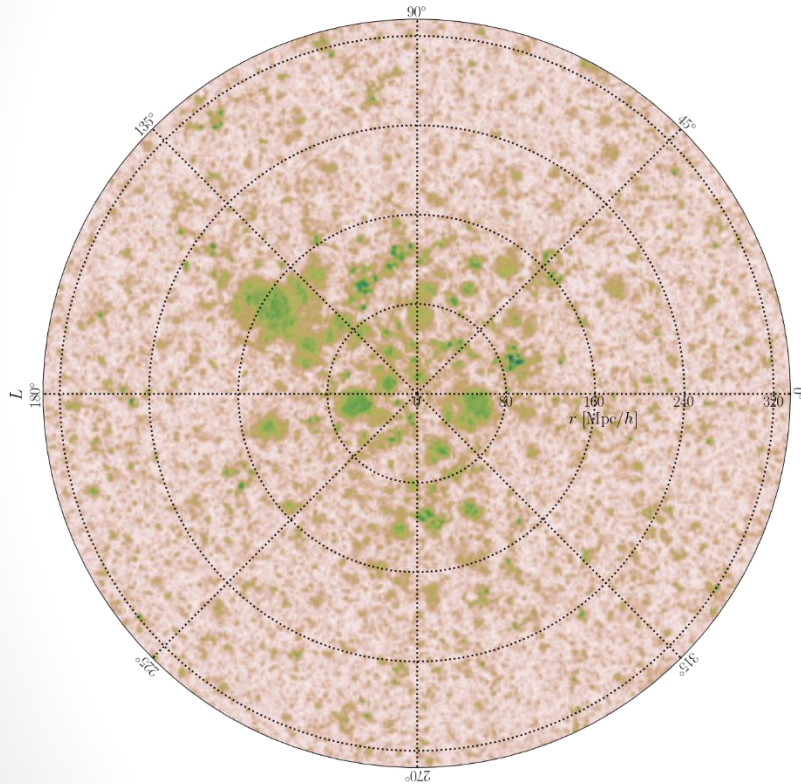
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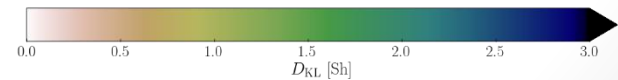
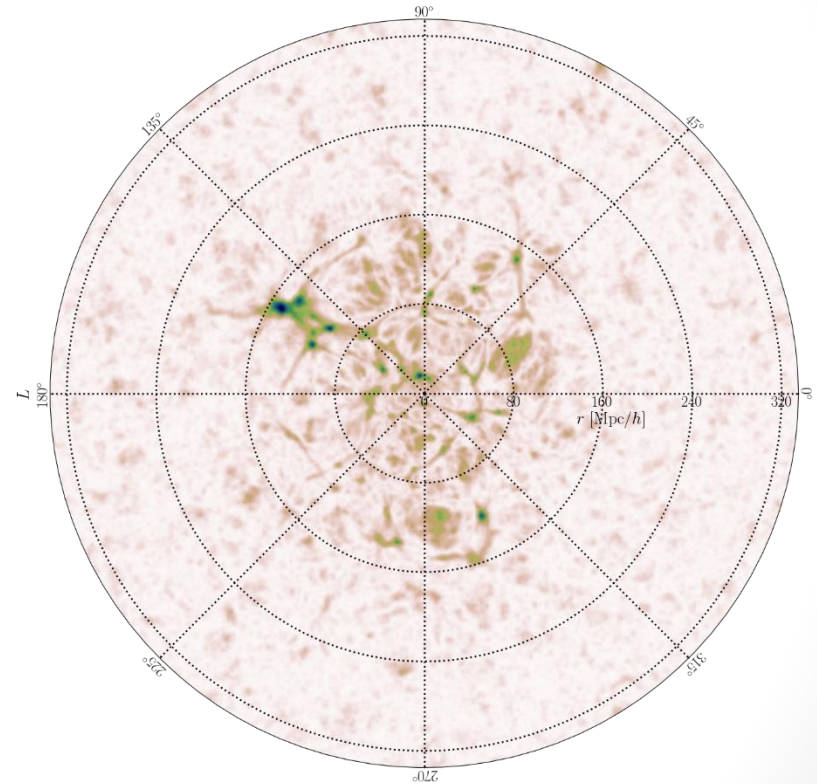
How much did the data surprise us? Information gain:

$$D_{\text{KL}}[\mathcal{P}(\mathbf{T}(\vec{x})|d)||\mathcal{P}(\mathbf{T})] \equiv - \sum_{i=0}^7 \mathcal{P}(\mathbf{T}_i(\vec{x})|d) \log_2 \left( \frac{\mathcal{P}(\mathbf{T}_i(\vec{x})|d)}{\mathcal{P}(\mathbf{T}_i)} \right) \text{ in shannons (Sh)}$$

Initial conditions



Final conditions



FL, Lavaux & Jasche, in prep.

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# A decision rule to build catalogues of objects

- Space of 8 “input features”:

$\{T_0 = \text{potential void}, T_1 = \text{potential sheet}, T_2 = \text{potential filament},$   
 $T_3 = \text{potential cluster}, T_4 = \text{vortical void}, T_5 = \text{vortical sheet},$   
 $T_6 = \text{vortical filament}, T_7 = \text{vortical cluster}\}$

- Space of 9 “actions”:

$\{a_j = \text{“decide structure } T_j\text{” for } 0 \leq j \leq 7,$   
 $a_{-1} = \text{“remain undecided”}\}$

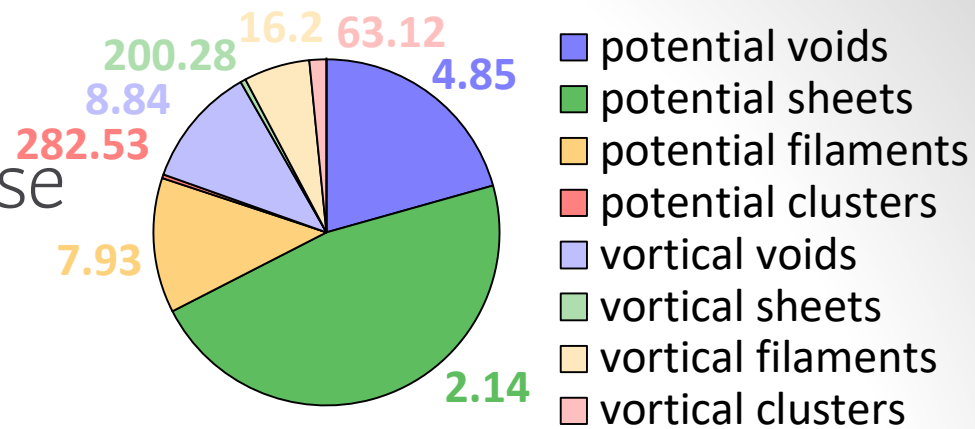
➡ A problem of **Bayesian decision theory**:

one should take the action that maximises the utility

$$U(a_j(\vec{x})|d) = \sum_{i=0}^7 G(a_j|T_i) \mathcal{P}(T_i(\vec{x})|d)$$

- How to write down the gain functions?

# Gambling with the Universe



- One proposal:

$$G(a_j|T_i) = \begin{cases} \frac{1}{\mathcal{P}(T_i)} - \alpha & \text{if } j \in \llbracket 0, 7 \rrbracket \text{ and } i = j & \text{“Winning”} \\ -\alpha & \text{if } j \in \llbracket 0, 7 \rrbracket \text{ and } i \neq j & \text{“Losing”} \\ 0 & \text{if } j = -1. & \text{“Not playing”} \end{cases}$$

- Without data, the expected utility is

$$U(a_j) = 1 - \alpha \quad \text{if } j \neq -1 \quad \text{“Playing the game”}$$

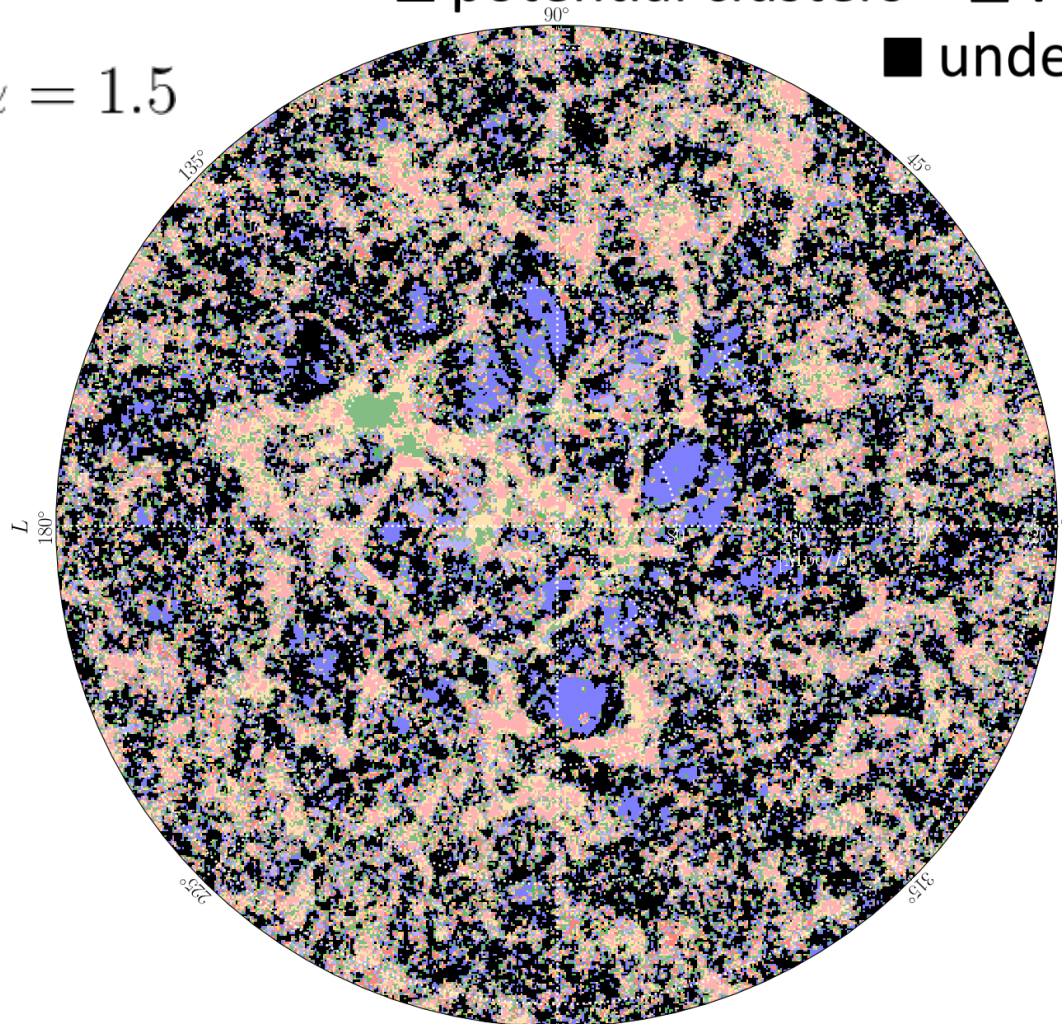
$$U(a_{-1}) = 0 \quad \text{“Not playing the game”}$$

- With  $\alpha = 1$ , it's a *fair game* ➡ always play  
➡ “speculative map” of the LSS
- Values  $\alpha > 1$  represent an *aversion for risk*  
➡ increasingly “conservative maps” of the LSS

Playing the game...

- |   |   |
|---|---|
| <span style="color: blue;">■</span> potential voids       | <span style="color: lightblue;">■</span> vortical voids       |
| <span style="color: green;">■</span> potential sheets     | <span style="color: lightgreen;">■</span> vortical sheets     |
| <span style="color: orange;">■</span> potential filaments | <span style="color: lightorange;">■</span> vortical filaments |
| <span style="color: red;">■</span> potential clusters     | <span style="color: pink;">■</span> vortical clusters         |
| <span style="color: black;">■</span> undecided            |   |

$\alpha = 1.5$



FL, Jasche & Wandelt 2015, 1503.00730 – FL, Lavaux & Jasche, in prep.

# Comparing cosmic web classifiers

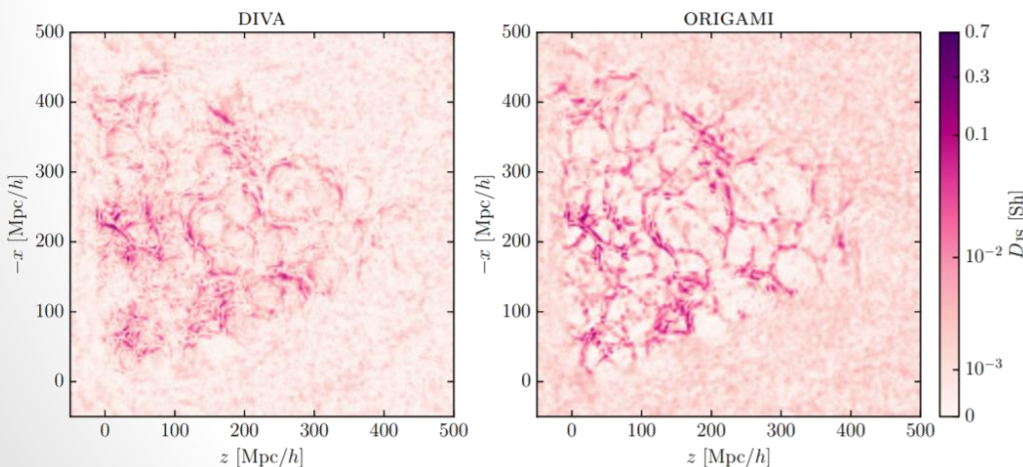
- The **decision problem** can further **extended to the space of classifiers**, with a utility function depending on the desired application:

$$U(\xi) = \int U(\xi, d) p(d|\xi) dd$$

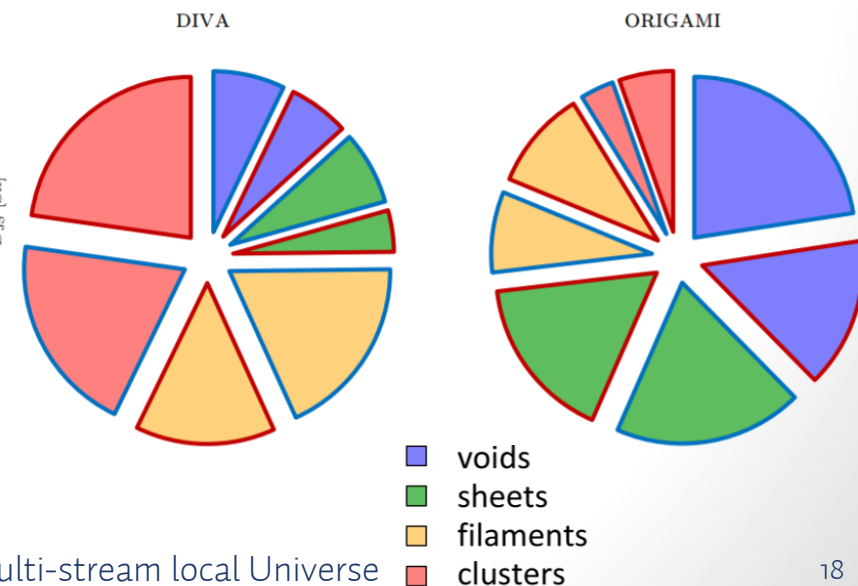
↑  
classifier
↑  
data

- Examples:

## Comparing dark energy models



## Classifying blue/red galaxies as a function of their environment



# Conclusions

- **BORG** is a **Bayesian inference engine** allowing the analysis of the **large-scale structure** and its formation history.
- Thanks to **BORGPM**, it is possible to map the multi-stream local Universe, including the **velocity dispersion** tensor.
- The **cosmic web** can be physically described using **LICH**, a classifier distinguishing potential and vortical flows.
- A probabilistic analysis of the cosmic web yields a data-supported **connection between cosmology and information theory**.
- **Decision theory** offers a framework to **classify structures** in the presence of data constraints and uncertainty.