# Geometry and topology: Applications to cosmological datasets

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# Outline

- Background Geometry and Topology
  - Minkowski functionals (Euler characteristic/genus)
  - ≻Homology
  - ≻Hierarchical homology (persistent homology)
- Topological characteristics of CMB fluctuations
   ➤Temperature
   ➤Full sky
   ➤Hemispheres
   ➤Polarization (full sky)
- Conclusion
- (Structure identification through topology)

#### **Minkowski Functionals**

• Predominantly Geometric quantities : include the notion of Volume, surface area, contour length of a manifold *M*.

• (D+1) quantifiers for *D*-dimensional sets

• Go by various names and orderings: quermassintegrals, Dehn and Steiner functionals, curvature integrals, intrinsic volumes, Minkowski functionals, and Lipschitz-Killing curvatures.

•We need only MFs and LKCs for our purpose, which when properly defined are related by

$$Q_j(M) = j! \,\omega_j \mathcal{L}_{D-j}(M), \qquad j = 0, \dots, D,$$
$$\omega_k = \pi^{k/2} \,\Gamma\left(\frac{1+k}{2}\right): \text{volume of } k - \text{dim. unit ball}$$

Pranav et al, MNRAS, 485 (3), 4167-4208

#### **Minkowski Functionals**

- A useful way of defining these quantities is via the *Steiner formula* or *Weyl's tube formula* :  $V_D(\{x \in \mathbb{R}^D : \min \| x - y \|\} \le \rho, y \in M) = \sum_{j=0}^{D} \frac{\rho^j}{j!} Q_j(M) = \sum_{j=0}^{D} \omega_{D-j} \rho^{D-j} \mathcal{L}_j(M)$
- The set in the LHS is known as the "tube around M", and  $\rho$  is small.
- Trivial to check  $Q_0$  and  $\mathcal{L}_D$  measure *D*-dimensional volume (set  $\rho = 0$  above).
- $Q_1$  and  $2\mathcal{L}_2$  measure surface area.
- Other functionals are harder to define, but always a true and deep result that:

$$\chi(M) = \mathcal{L}_0(M) = \frac{1}{D! \,\omega_D} Q_D(M).$$

• In 3D, this only leaves  $Q_2$  and  $\mathcal{L}_1$ . If the manifold *M* is convex,  $\mathcal{L}_1(M) = Q_2(M)/2\pi$  is twice the *caliper diameter* of *M*.

#### Geometry and Topology

- Theorema Egrerium (latin for remarkable theorem) of Gauss states that the Gaussian curvature of a surface is an *intrinsic invariant*, meaning it is a constant irrespective of how the surface is bent (or twisted) in space
- Leads to the Gauss-Bonnet theorem

$$\int_M K \ dA + \int_{\partial M} k_g \ ds = 2\pi \chi(M)$$

- K : Gaussian curvature of M,  $k_g$ : geodesic curvature of the boundary of M
- The theorem is remarkable because it links and proves that a topological invariant (EC) can be computed purely from geometrical properties.

# **Euler characteristic**

•Originally defined for polyhedra

$$\chi = V - E + F$$



## Genus

- For a connected, orientable surface, the Genus has a linear relationship with the maximal number of independent simple closed curves that can be drawn on the surface without rendering it disconnected
- Number of handles attached to a surface



#### Why the Euler characteristic?

#### SIMPLICIAL TOPOLOGY

Simplices, complexes, cycles, numbers of simplices, Betti numbers

 $\sum_{k} (-1)^{k} \# \{k \text{-dimensional simplices}\}$ 

 $\sum_{k} (-1)^k \beta_k$ 

ALGEBRAIC TOPOLOGY Homology, homotopy, dimensions of groups, Betti numbers, persistence

#### INTEGRAL GEOMETRY

Convexity, convex ring kinematic formulae Minkowski functionals

 $\mathcal{M}_k(M) = c_{dk} \int_{\mathrm{Graff}(d,d-k)} \chi(M \cap V) \, d\mu_{d-k}^d(V)$ 

 $\sum_{k} (-1)^{k} \# \{ \text{critical points of index } k \}$ 

 $\int_M \operatorname{Tr}(R^{m/2}) \operatorname{Vol}_g$ 

DIFFERENTIAL TOPOLOGY Curvature, forms, Betti numbers, Morse theory, integration, Lipschitz-Killing curvatures

#### **Beyond the Genus, EC and MFs**

• The genus is defined only for connected and closed 2-dimensional surfaces, and has no generalizations in higher or lower dimensions.

• The Euler-Poincare formula states that the Euler characteristic is an alternating sum of another topological invariant called the *Betti numbers* 

• A formalism capable of expressing topology in a hierarchical fashion would present an interesting and powerful extension for describing the hierarchical structures in the cosmos.

# Genus, Euler & Betti

• Euler – Poincare formula

Relationship between Betti Numbers & Euler Characteristic  $\chi$ :



# Homology

Topology:

Study of connectivity and spatial relations that remain invariant under homeomorphisms (= continuous mapping between two topological objects)

Homology:

- Description of topology of a space in terms of cycles/boundaries.
- Fundamental lemma : Boundary of a boundary is necessarily empty.



## Topological cycles and holes

• intuitive interpretation



0 dimensional holes : gaps between connected objects 1 dimensional holes : loops/tunnels

2 dimensional holes : voids

# Critical points and filtration

birth and death of topological cycles

- Study the change in topology of a manifold w.r.t. the growing excursion sets of the function f
- Topology only changes at critical points of the function
- Addition of a critical point with index k, either creates a k-dimensional hole, or it destroys a (k-1)-dimensional hole





 $\emptyset = M_0 \subseteq M_1 \subseteq \dots \subseteq M$ 

#### Filtrations

- Let K be a simplicial complex,
  - $f: K \to \mathbb{R}$  a monotonic function, i.e.  $f(\sigma) \leq f(\tau)$  when  $\sigma$  is a face of  $\tau$ ,

Let  $a_1 < \ldots < a_n$  be the values of f on the simplices of K.

 We get n + 1 different subcomplexes K<sub>i</sub> = f<sup>-1</sup>(-∞, a<sub>i</sub>] of K which give a filtration of f:

$$\emptyset = K_0 \subseteq K_1 \subseteq \ldots \subseteq K_n = K.$$

For i ≤ j, the inclusion map K<sub>i</sub> → K<sub>j</sub> induces the homomorphism f<sup>i,j</sup><sub>p</sub> : H<sub>p</sub>(K<sub>i</sub>) → H<sub>p</sub>(K<sub>j</sub>) for each dimension p.
 Sequence of homology groups, for each dimension p:

$$0 = H_p(K_0) \to H_p(K_1) \to \ldots \to H_p(K_n) = H_p(K)$$

#### p-th persistent homology groups

The *p*-th persistent homology groups are

$$H_p^{i,j} = \operatorname{im}(f_p^{i,j})$$
 for  $0 \le i \le j \le n$ .

 $H_p^{i,j}$  consists of the homology classes of  $K_i$  that are still alive at  $K_j$ :

$$H_p^{i,j} = Z_p(K_i)/(B_p(K_j) \cap Z_p(K_i)).$$

The *p*-th persistent Betti numbers are  $\beta_p^{i,j} = \operatorname{rank} H_p^{i,j}$ .



Persistence:  $pers(\gamma) = a_j - a_i$ .

#### Birth, death and life-time(persistence): hierarchical topology



- Representation of multi-scale topology
- Dots in the diagram record birth and death
- A diagram for each ambient dimension of the manifold
  - 0-dimensional diagram: representation of merger of isolated objects (merger trees)
  - 1-dimensional diagrams: formation and filling up of loops (percolation)
  - 2-dimensional diagrams: formation and destruction of topological voids (voids)

Pranav et al, MNRAS, 465 (4), 4281-4310, 2017

# Computation

(3)

 $\langle 4 
angle = (1,2)$ 

(3)



## **Topology and Geometry of 3D Gaussian fields**

Pranav et. al. MNRAS, 485(3), 4167 (2019)

### **Gaussian Random fields**

•Given a spatial location s, a Gaussian random field is a random function X(s) on R<sup>3</sup> such that, when restricted to any finite set, one has a multivariate normal distribution.

• m-point joint distribution function:

$$P[f_1,...,f_m]df_1...df_m = \frac{1}{(2\pi)^N (\det M)^{1/2}} \cdot \exp(-\frac{\sum \Delta f_i(M^{-1})_{ij} \Delta f_j}{2}) df_1...df_m$$

- Mean:  $\Delta f_i = f_i \langle f_i \rangle$
- Covariance matrix:

$$M_{ij} = \langle \Delta f_i \Delta f_j \rangle$$

## Gaussian fields: Minkowski Functionals



## Gaussian fields: Betti Topology



- Shape of Betti numbers dependent on power spectrum; EC and MF are not
- Gott et. al. (1986): Sponge-like topology of the LSS

Pranav et. al. MNRAS, 485(3), 4167 (2019)



# Topology of the CMB



Figure 1: Left: A blue excursion set on the sphere consisting of an upper left component with a hole, an upper right component, and a lower component. Its Betti numbers are  $\beta_0 = 3$ ,  $\beta_1 = 1$ ,  $\beta_2 = 0$ , and its Euler characteristic is EC = 3 - 1 + 0 = 2. Middle: A pink mask in which the data is not reliable. It covers part of the upper left component and hole, its hole is fully contained in the upper right component, and it overlaps the lower component in two disconnected pieces. Right: A visualization of the relative homology groups obtained by shrinking the mask to a point and pulling the excursion set with it. We have  $b_0 = 0$  because all three components connect to the shrunken mask,  $b_1 = 2$  because the loop in the upper left component is preserved and a new loop in the lower component is formed, and  $b_2 = 1$  because the upper right component takes on the shape of sphere. The (relative) Euler characteristic is therefore EC<sub>rel</sub> = 0 - 2 + 1 = -1.

#### **Planck Data**

- Specified on S2, as the deviation from the background average (HEALPIX format)
- Measurement unreliable in some parts: foregrounds
- Unreliable parts masked
- Field converted to N (0,1) using unmasked pixels only









**Temperature** Full Sky

#### Masked degraded maps (multi-scale analysis)



- Maps degraded to N\_side = 2048, 1024, 512, 256, 128, 64, 32 and 16 (not shown), corresponding to FWHM = 5', 10', 20', 40', 80', 160', 320', 640'
- Binary Mask degraded similarly (converts it to non-binary) reconverted to binary by setting the threshold 0.9 (as done by Planck coll.)



Isolated objects (betti 0)

P.Pranav, A&A 659, A115 (2022)

2.0

3.0 0.0

Degrade = 256

Degrade = 16

1.0

2.0

3.0



Loops (betti 1)

P.Pranav, A&A 659, A115 (2022)

Degrade = 256

Degrade = 16

-2.0

-1.0

0.0

0.0 -3.0

# CMB Loops





# Statistical tests

- The data consists of topological summaries (b0, b1, EC) obtained from simulations, and observed CMB field
- Goal: estimate the probability that the physical model that produced the simulations would produce
- quantities consistent with those from the observed CMB field
  x<sub>i</sub> ∈ ℝ<sup>m</sup>, i = 1,...,n,
  Let y ∈ ℝ<sup>m</sup> ip, pi=1, ib a sample of i.i.d. \$m\$-dimensional vectors, drawn from a distribution F. Let y ∈ ℝ<sup>m</sup> ip, pi=1, ib another sample point, assumed to be drawn from a distribution G.
- Test the (null) hypothesis that F=G
- p-values compute the probability that **y** is `consistent' with this hypothesis.
- Two methods :
  - Mahalanobis Distance or chi<sup>2</sup> test : parametric (Prasanta Chandra Mahalanobis, 1936)
  - Tukey depth : non-parametric (John Tukey, 1974)

# Mahalanobis distance or $\chi 2$ -test: • Mean : $\bar{\mathbf{x}} = \sum_{i=1}^{n} \mathbf{x}_i / n$ • Variance: $\mathbf{S} = \sum_{i=1}^{n} (\mathbf{x}_i - \bar{\mathbf{x}}) (\mathbf{x}_i - \bar{\mathbf{x}})^T / (n-1)$

- Mahalanobis distance:  $d_{\text{Mahal}}^2(\mathbf{y}) = (\mathbf{y} - \bar{\mathbf{x}})^{\mathrm{T}} \mathbf{S}^{-1} (\mathbf{y} - \bar{\mathbf{x}}).$
- If **F** is assumed to be Gaussian and **n** is large, then under the hypothesis that **G=F** the squared Mahalanobis distance is approximately distributed as a **chi^2** distribution with **m** degrees of freedom. Thus the corresponding p-value is

$$p_{\text{Mahal}}(\mathbf{y}) = P[\chi_m^2 > d_{\text{Mahal}}^2(\mathbf{y})].$$

# Tukey depth

- **F** may not always conform to elliptical contours and therefore may not be Gaussian. In such a setting, \$p\$-values computed using the Mahalanobis distance may not be reliable.
- The Tukey half-space depth provides a general metric for identifying outliers in a flexible manner and in a non-parametric setting.
- Take  $x_i$ , i=1,...,n and y as before, making no assumptions on the structure of F and G, and let Z be any point in  $\mathbb{R}^m$ .
- Half-space depth d<sub>dep</sub>(x<sub>j</sub>; x<sub>1</sub>,..., x<sub>n</sub>) of z within the sample of the x<sub>i</sub> is the smallest fraction of the n points \$x<sub>1</sub>,...,x<sub>n</sub>\$ to either side of any hyperplane passing through Z
- Points that have the same depth constitute a non-parametric estimate of the isolevel contour of the distribution  $\mathbf{F}$ .
- To evaluate a p-value for y, compute  $d_j = d_{dep}(x_j; x_1, \dots, x_n)$  for every point  $x_j$ ,  $(j=1, \dots, n)$ , yielding an empirical distribution of depth. p-value is the proportion of points whose depth is lower than that of y:

 $p_{dep}(\mathbf{y}) = \#\{j \mid d_j > d_{dep}(y)\}/n$ 

A&A	proofs:	manuscript	no. npipe	CMB

Relative homology												
		$\chi^2$ (	theoret	ical)	$\chi^2$	(empiri	cal)	Tukey Depth				
Res	FWHM	$b_0$	$b_1$	$EC_{\mathrm{rel}}$	$b_0$	$b_1$	$EC_{\mathrm{rel}}$	$b_0$	$b_1$	$EC_{\mathrm{rel}}$		
threshold = 0.90												
2048	5	0.708	0.569	0.858	0.676	0.584	0.851	0.630	0.318	0.912		
1024	10	0.649	0.475	0.383	0.628	0.476	0.363	0.672	0.352	0.000		
512	20	0.156	0.417	0.203	0.161	0.408	0.183	0.332	0.308	0.000		
256	40	0.398	0.596	0.772	0.403	0.600	0.756	0.362	0.560	0.642		
128	80	0.356	0.204	0.563	0.383	0.186	0.555	0.000	0.312	0.730		
64	160	0.398	0.494	0.768	0.401	0.468	0.746	0.465	0.718	0.755		
32	320	0.020	0.002	0.001	0.028	0.013	0.001	0.000	0.000	0.000		
16	640	0.974	0.001	0.203	0.981	0.030	0.222	0.983	0.000	0.440		
summary	NA	0.478	0.023	0.045	0.290	0.076	0.052	0.803	0.000	0.000		

Table 1: Table displaying the two-tailed *p*-values for relative homology obtained from parametric (Mahalanobis distance) and nonparametric (Tukey depth) tests, for different resolutions and smoothing scales for the NPIPE dataset. The last entry is the *p*-value for the summary statistic computed across all resolutions. Marked in boldface are *p*-values 0.05 or smaller. Temperature Hemispheres













*b*0 for the temperature maps (NPIPE and FFP10 dataset) Northern (top two rows) and the southern hemisphere (bottom two rows).

•

- The graphs present the normalized differences, and each panel presents the graphs for a range of degradation and smoothing scales.
- PR3 temperature common mask.



-3.5 -3 -2.5 -2.5 -1.5



*b1* for the temperature maps (NPIPE and FFP10 dataset) Northern (top two rows) and the southern hemisphere (bottom two rows).



-2.0

level (v)

(b)

-1.0

-2.0

-1.0

0.0 -3.0

-2.0

-1.0

0.0 -3.0

-2.0

-1.0

0.0

0.0 -3.0

0.0 -3.0

-3.0

-2.0

-1.0

0.0 -3.0

-2.0

-1.0

0.0 -3.0

-2.0

-1.0

(c)



(d)

Relative homology – $\chi^2$ (empirical) – Northern hemisphere						Relative homology – $\chi^2$ (empirical) – Southern hemisphere									
			FFP10			NPIPE				FFP10			NPIPE		
Res	FWHM	<i>b</i> 0	$b_1$	EC <sub>rel</sub>	<i>b</i> 0	$b_1$	EC <sub>rel</sub>	Res	FWHM	<i>b</i> 0	<i>b</i> <sub>1</sub>	EC <sub>rel</sub>	<i>b</i> 0	$b_1$	EC <sub>rel</sub>
	threshold = 0.90						threshold = 0.90								
1024	10'	0.240	0.110	0.133	0.408	0.073	0.148	1024	10'	0.967	0.457	0.467	0.992	0.762	0.887
512	20'	0.433	0.237	0.287	0.463	0.085	0.165	512	20'	1.000	0.857	0.967	0.993	0.818	0.993
256	40'	0.037	0.327	0.273	0.060	0.335	0.477	256	40'	0.090	0.677	0.247	0.350	0.817	0.728
128	80'	0.000	0.047	0.013	0.037	0.037	0.035	128	80'	0.653	0.623	0.847	0.662	0.902	0.915
64	160'	0.063	0.070	0.233	0.043	0.058	0.073	64	160'	0.287	0.260	0.507	0.283	0.722	0.850
32	320'	0.157	0.247	0.343	0.020	0.215	0.102	32	320'	0.853	0.247	0.460	0.597	0.103	0.172
16	640'	0.817	0.540	0.693	0.560	0.290	0.470	16	640'	0.873	0.023	0.037	0.428	0.178	0.268
summary	N/A	0.010	0.087	0.000	0.023	0.053	0.017	summary	N/A	0.543	0.397	0.013	0.577	0.877	0.925

(a)

(b)



• Minkowski functionals and skeleton length (WMAP)

Eriksen et al 2004

### **Polarization**

## Experimental set up

- Use full sky to generate alms from TQU maps (prevents E/B leakage)
- Use the grad and curl-like like elements to synthesize the E and B maps
- Mask : PR3 common mask (+ b)





*E-mode maps* Isolated objects (betti 0)



#### NPIPE : Planck 2018 common Mask



B-mode maps Isolated objects (betti 0)

#### NPIPE : Planck 2018 common Mask



Loops(betti 1)



B-mode maps Isolated objects (betti 0)



*B-mode maps* Loops(betti 1)

Relative homology – NPIPE <i>E</i> -mode							Relative homology – NPIPE <i>B</i> -mode								
		$\chi^2$	(empiri	cal)	$\chi^2$ (	theoreti	ical)			$\chi^2$ (empirical)			$\chi^2$ (theoretical)		
Res	FWHM	$b_0$	$b_1$	$EC_{\mathrm{rel}}$	$b_0$	$b_1$	$EC_{\mathrm{rel}}$	Res	FWHM	$b_0$	$b_1$	$\text{EC}_{\text{rel}}$	$b_0$	$b_1$	$EC_{\mathrm{rel}}$
threshold = $0.90$								threshold = 0.90							
2048	5	0.600	0.665	0.457	0.601	0.671	0.470	2048	5	0.140	0.708	0.473	0.154	0.708	0.470
1024	10	0.498	0.815	0.495	0.504	0.806	0.506	1024	10	0.210	0.377	0.315	0.226	0.381	0.324
512	20	0.173	0.638	0.215	0.180	0.631	0.214	512	20	0.000	0.100	0.008	0.000	0.097	0.006
256	40	0.582	0.187	0.388	0.596	0.190	0.407	256	40	0.000	0.000	0.000	0.000	0.000	0.000
128	80	0.447	0.992	0.705	0.476	0.989	0.699	128	80	0.005	0.062	0.002	0.003	0.053	0.000
64	160	0.953	0.962	0.993	0.936	0.960	0.991	64	160	0.817	0.172	0.777	0.839	0.183	0.775
32	320	0.358	0.027	0.018	0.371	0.016	0.015	32	320	0.587	0.475	0.753	0.594	0.487	0.760
16	640	0.670	0.130	0.088	0.764	0.198	0.058	16	640	0.423	0.565	0.540	0.453	0.683	0.569



*Q maps* Isolated objects (betti 0)





Degrade = 64

-3 -2 -1 0

Degrade = 256



u

100

0 20

-3 -2 -1

ل و 10



u





*Q maps* Loops(betti 1)





*U maps* Isolated objects (betti 0)



u





u



Degrade = 64

u

-2 -1 0



U maps Loops(betti 1)

Relative homology													
		$\chi^2$ (	empiri	cal)	$\chi^2$ (	theoret	ical)	Tukey Depth					
Res	FWHM	$b_0$	$b_1$	$EC_{\mathrm{rel}}$	$b_0$	$b_1$	$EC_{\mathrm{rel}}$	$b_0$	$b_1$	$EC_{\text{rel}}$			
threshold = $0.90$													
2048	5	0.578	0.332	0.583	0.584	0.334	0.574	0.540	0.360	0.485			
1024	10	0.517	0.957	0.838	0.503	0.950	0.830	0.513	0.948	0.870			
512	20	0.480	0.363	0.812	0.468	0.376	0.799	0.487	0.338	0.795			
256	40	0.038	0.102	0.090	0.048	0.097	0.088	0.000	0.000	0.000			
128	80	0.178	0.168	0.430	0.201	0.169	0.427	0.000	0.000	0.462			
64	160	0.103	0.503	0.707	0.102	0.494	0.698	0.000	0.505	0.750			
32	320	0.802	0.587	0.825	0.781	0.617	0.828	0.780	0.558	0.798			
16	640	0.178	0.103	0.228	0.177	0.061	0.217	0.000	0.000	0.000			
summary	NA	0.582	0.017	0.057	0.569	0.011	0.060	0.000	0.000	0.000			

Table 1: Table displaying the two-tailed p-values for relative homology obtained from parametric (Mahalanobis distance) and non-parametric (Tukey depth) tests, for different resolutions and smoothing scales for the NPIPE dataset Q component. The last entry is the p-value for the summary statistic computed across all resolutions. Marked in boldface are p-values 0.05 or smaller.

Relative homology													
		$\chi^2$ (	empiri	cal)	$\chi^2$ (	theoret	ical)	Tukey Depth					
Res	FWHM	$b_0$	$b_1$	$EC_{\mathrm{rel}}$	$b_0$	$b_1$	$EC_{\mathrm{rel}}$	$b_0$	$b_1$	$EC_{rel}$			
threshold = $0.90$													
2048	5	0.045	0.768	0.295	0.049	0.767	0.309	0.000	0.785	0.000			
1024	10	0.593	0.275	0.192	0.620	0.264	0.171	0.625	0.252	0.000			
512	20	0.593	0.150	0.283	0.602	0.145	0.285	0.577	0.000	0.000			
256	40	0.298	0.747	0.932	0.314	0.747	0.925	0.335	0.730	0.887			
128	80	0.933	0.700	0.760	0.936	0.690	0.755	0.953	0.567	0.832			
64	160	0.823	0.040	0.298	0.831	0.043	0.302	0.787	0.000	0.000			
32	320	0.682	0.435	0.447	0.679	0.439	0.444	0.502	0.362	0.000			
16	640	0.312	0.177	0.667	0.320	0.139	0.700	0.255	0.355	0.827			
summary	NA	0.023	0.008	0.010	0.022	0.003	0.011	0.000	0.000	0.000			

Table 2: Table displaying the two-tailed p-values for relative homology obtained from parametric (Mahalanobis distance) and non-parametric (Tukey depth) tests, for different resolutions and smoothing scales for the NPIPE dataset U component. The last entry is the p-value for the summary statistic computed across all resolutions. Marked in boldface are p-values 0.05 or smaller.

#### Signal-to-Noise





Fig. 5. Signal-to-noise ratio for the variance estimator in polarization for Commander (red), NILC (orange), SEVEM (green), and SMICA (blue), obtained by comparing the theoretical variance from the *Planck* FFP10 fiducial model with an MC noise estimate (right-hand term of Eq. (7)). Note that the same colour scheme for distinguishing the four component-separation maps is used throughout this paper.

## Conclusions

- Topology and geometry powerful means of characterizing data.
- CMB data exhibits mild to significant anomaly in both temperature and polarization data.
  - Temperature: full sky (large-scale); hemisphere (Degree-scale)
  - Polarization: full-sky (degree-scale)
- Results are based on legitimate mathematical foundations, and not a case of over-exploitation of data through tailor-made statistics.
- Hints of violation of cosmological principal, or other late time mechanisms at play (including doppler boosting/dipolar modulation)?