# Coverage and lacunarity of astronomical datasets

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#### Abstract

We present the use of MOCs (Multi-Order Coverage maps) associated to HiPS datasets (Hierarchical Progressive Surveys) to assess the sky coverage of various datasets (image surveys and source catalogues), but also their lacunarity (presence of holes). The goal is to :

#### **HiPS and MOCs**



See Fernique et al. (2015)

- Based on the HEALPix tesselation of the sphere (equal area)
- Progressive surveys (HiPS) and coverage maps (MOCs) allow to describe complex datasets
- Estimate at best the sky coverage of a dataset
- Find a metric to quantify the scattering of data on the sky (distinguish contiguous datasets from patches of data scattered over large areas, or containing many holes)
- Apply this to source catalogues, without oversampling

## **MOCs of mock data**



- MOCs can be used to estimate the sky coverage, by adding the areas of all HEALPix pixels.
- The total area is overestimating the true area, because pixels spread on the perimeter.
- The highest the order of the MOC, the better the precision.
- The area of a HEALPix pixel at order k is  $\Omega_{\text{pix}} = (\Pi/3) 4^{-k}$ , and
- the estimated area from a MOC at order k is related to the true area by a relation of the form  $S_k = S_{true}(1 + L \cdot 2^{-k})$  where
- L is a parameter related to the perimeter of the MOC.
- More complex MOCs will have larger values of L.
- We define *L* = log(L) as the lacunarity of the dataset : small values will indicate simple and contiguous data distribution, while large values will characterize scattered or irregular data.

The notion of lacunarity was introduced by Mandelbrot to describe the texture of fractals, and was extended (see for example Dong (2000)) to describe the distribution of gap sizes in geometric objects : homogeneous objects have low lacunarity, and heterogeneous objects have high lacunarity.

#### Precision increases with order k

k	$N_{\rm side} = 2^k$	$N_{\rm pix}$	$\theta_{\rm pix}$	$k_{\rm tile,512}$	N <sub>tile,512</sub>	$\theta_{\rm tile,512}$
0	1	12	58°.6			
1	2	48	29.3			
2	4	192	14°.7			
3	8	768	7.33			
4	16	3072	3°.66			
5	32	12288	1°.83			
6	64	49 152	55'.0			
7	128	196 608	27:5			
8	256	786 432	13:7			
9	512	3 145 728	6.'87	0	12	58°.6
10	1024	12 582 912	3:44	1	48	29°.3
11	2048	50 331 648	1:72	2	192	14°.7
12	4096	201 326 592	51".5	3	768	7°.33
13	8192	805 306 368	25''.8	4	3072	3°.66
14	$2^{14}$	$3.22 \times 10^{9}$	12.'9	5	12 288	1°83
15	215	$1.29 \times 10^{10}$	6.44	6	49 152	55:0
16	216	$5.15 \times 10^{10}$	3".22	7	196 608	27:5
17	217	$2.06 \times 10^{11}$	1"61	8	786432	13.7
18	218	$8.25 \times 10^{11}$	0'.'81	9	3 145 728	6:87
19	2 <sup>19</sup>	$3.30 \times 10^{12}$	0.'40	10	12 582 912	3:44
20	$2^{20}$	$1.32 \times 10^{13}$	0.'20	11	50 331 648	1:72
21	$2^{21}$	$5.28 \times 10^{13}$	0.'10	12	201 326 592	51.5
22	222	$2.11 \times 10^{14}$	50.3 mas	13	805 306 368	25".8
23	$2^{23}$	$8.44 \times 10^{14}$	25.1 mas	14	$3.22 \times 10^9$	12".9
24	$2^{24}$	$3.38 \times 10^{15}$	12.6 mas	15	$1.29 \times 10^{10}$	6.44
25	$2^{25}$	$1.35 \times 10^{16}$	6.29 mas	16	$5.15 \times 10^{10}$	3.22
26	$2^{26}$	$5.40 \times 10^{16}$	3.15 mas	17	$2.06 \times 10^{11}$	161







Computing the MOCs at different orders k for a r=15° spherical cap, and fitting the formula over orders k=7 to 10 gives an estimated  $S_{true} = 704.5 deg^2$ , very close to the exact area S=702.8 deg<sup>2</sup>, with  $\pounds = 0.648$ .

For a MOC corresponding to 3 disjoints caps, the fit yields  $S_{true}$ =2109.4deg<sup>2</sup>, very close to the exact area S=2108.4deg<sup>2</sup>, with  $\mathcal{L}$  = 0.715.



### **Application to image HiPS**

For 3 example image HiPS, we compute MOCs from order *k*=3 down to a maximum order, and study the evolution of the MOC area with *k*.



#### **Application to catalogues**

Estimating the sky coverage of a catalogue from the positions of its individual sources is more difficult than for an image survey : if the density of sources per HEALPix pixel becomes too low at high orders k, empty pixels cause an underestimation of the area.

To illustrate this, we study here two VizieR catalogues with a similar number of sources, but very different distributions on the sky.

## **CDS/I/335/table1**: LUT Survey Catalogue Data Release 1 (Men et al. 2016) N=86 467 sources.

When k is too large, there is a single source per HEALPix pixel, and the MOC area becomes  $S_k = N(\Pi/3) 4^{-k}$  (blue dashed line in the graphs).







#### **CDS/I/273A/erlcat**: Extragalactic Reference Link Catalog (de Vegt et al., 2001)

Observatoire گ			astronomique		
		de Strasbourg ObAS			





MOC orde

k=6 to 9 allow to estimate  $S_{true} = 219.5 deg2$  and  $\pounds = 2.6$ . The MOC server gives S=60.39 deg2, using k=11.

**Conclusions**: MOCs can provide a very convenient way to estimate the sky coverage of image surveys or catalogues. The analysis of the variations of the coverage with the MOC order also provide information on the scattering within the dataset, for example using the lacunarity introduced in this study.



