



COSMO HUB on Hadoop

<https://cosmohub.pic.es>

A web portal to analyze and
distribute cosmology data

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Outlook

- What is CosmoHub
- Motivation
- Drawbacks
- Hadoop solution
- Comparison
- Demo
- Conclusions & future work



Build your own Universe

Real-time data analysis of massive cosmological data without any SQL knowledge



Hundreds of millions
of observed and
simulated galaxies



Superfast queries
means superfast
results



Features to make
you work faster and
easier



Online plotting
preview and data
download

CosmoHub on PostgreSQL

- CosmoHub was first thought as a way to share data within two closed related projects, the Physics of the Accelerating Universe (PAU) and the Marenostrum Institut de Ciències de l'Espai simulations (MICE)
- It was built on top of a PostgreSQL relational database
- It was developed by people from the Institut de Ciències de l'Espai (ICE), the Port d'Informació Científica (PIC - www.pic.es), CIEMAT and IFAE
- It was hosted and operated at PIC

Some numbers

- CosmoHub is currently supporting four different cosmology projects:



Marenostrum Institut
de Ciències de l'Espai
Simulations



DARK ENERGY
SURVEY



- ~ 400 users
- ~ 1300 custom catalogs
- ~ 250 prebuilt downloads
- ~ 3 TiB hosted data
- > 10^9 objects

Already available features

- Custom catalogs without any SQL knowledge (CSV.BZ2 only)
- Plot & preview tool: small sample of data using a scatter plot or generate a 1D-histogram (**query time limited to < 2'**)
- Value-Added-Data ready to be downloaded

What happened?

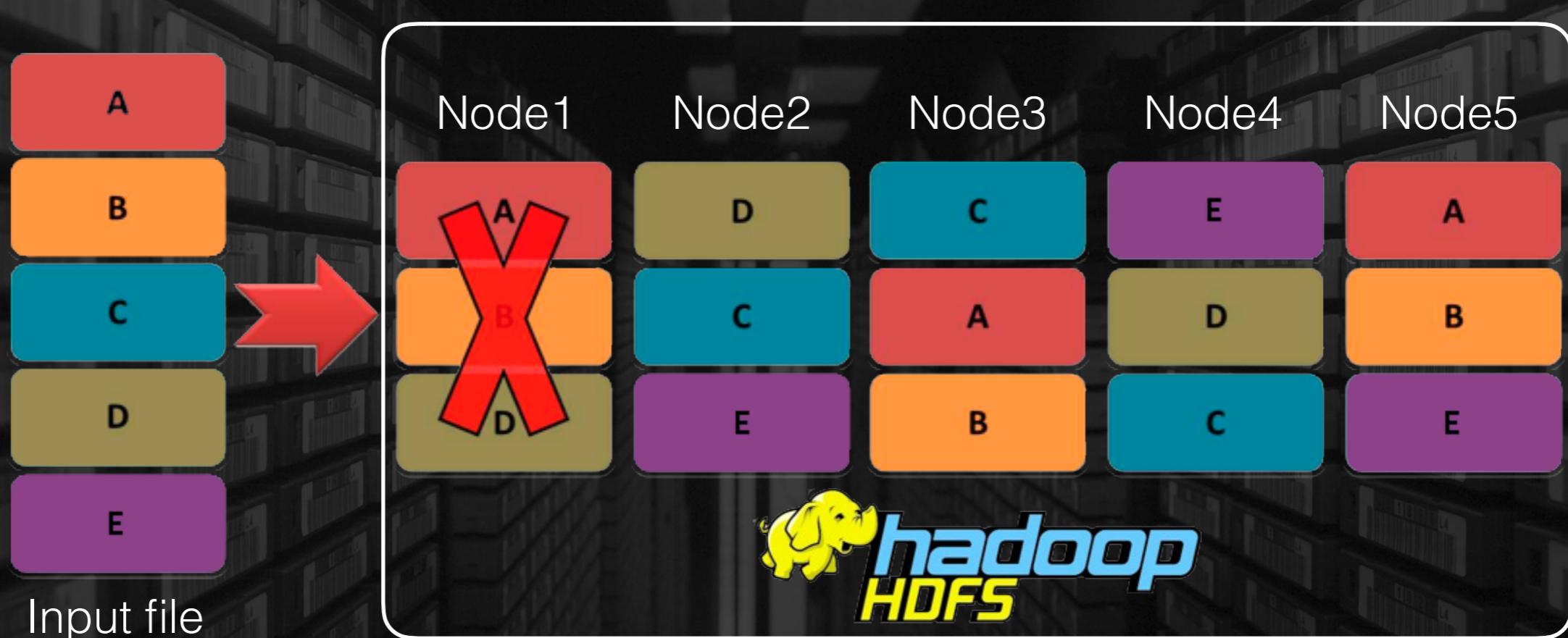
- MICECATv2.0 catalog contains about 500M entries with more than 120 fields
- Managing large volume of data in PostgreSQL had some drawbacks:
 - Indices are not used for large datasets
 - Most custom catalogs lasted several hours
 - Changing the schema was very slow
 - Removing large subsets of data is very inefficient
- Future galaxy catalogs will contain a few 10^9 entries

Apache Hadoop & Hive

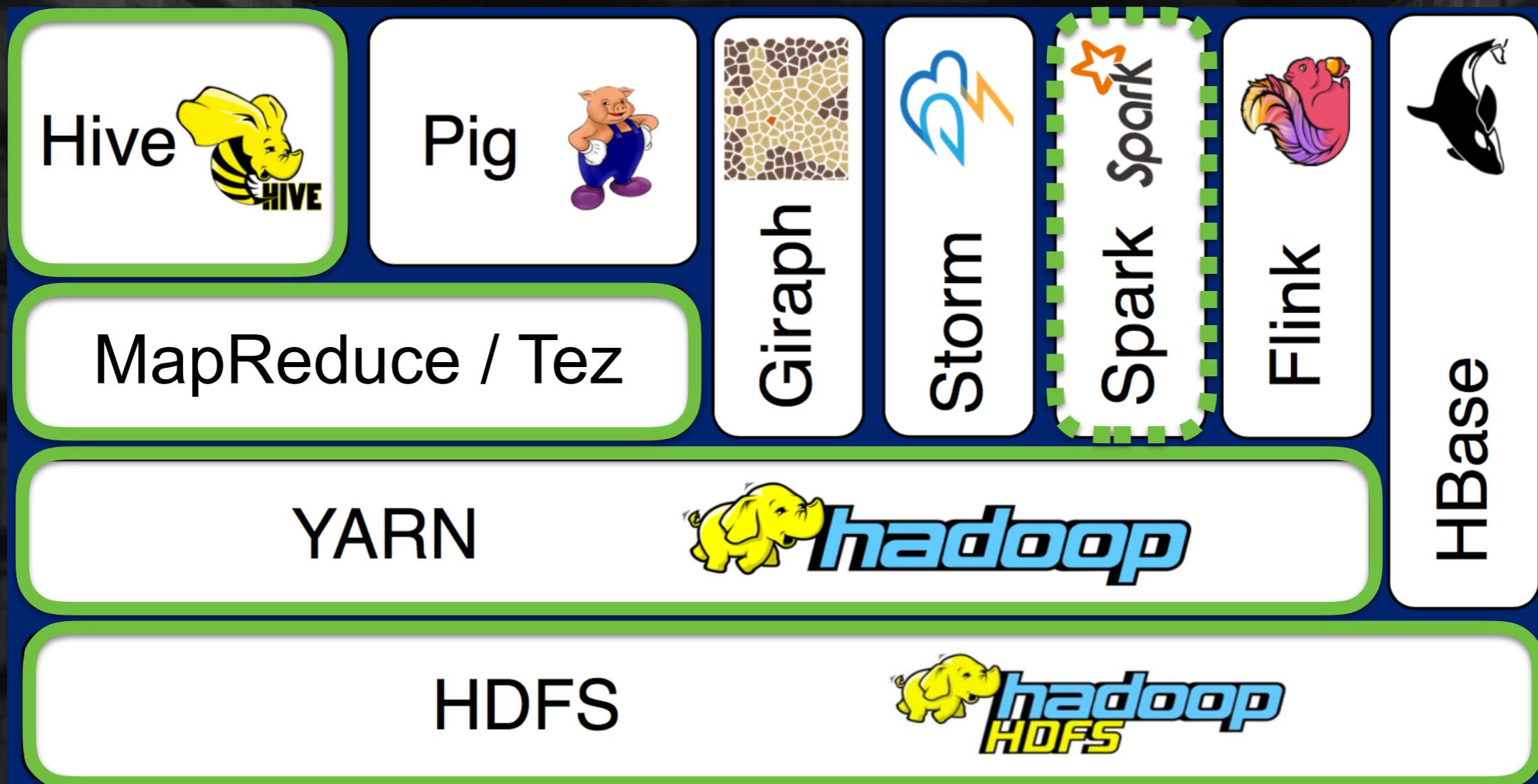
- Apache Hadoop:
 - one of the most popular Big Data platform
 - open-source software
 - based on commodity computer clusters
 - distributed storage and distributed processing
 - scalable from dozens up to even thousands of nodes
 - failure tolerance
- Apache Hive: query over massive data volumes



Hadoop basics



Hadoop stack



Hadoop vs. PostgreSQL

- Larger relative gain in execution time for increasing complexity in datasets and/or as queries request larger data volumes

(Comparisons are odious. It is very likely to unjust to one or other of them)

- Nodes: 15
 - Cores: 12 (Intel Xeon X5650 @ 2.67 GHz) [180]
 - RAM: 24 GiB [360 GiB]
 - DISK: 1 TiB [15 TiB raw; ~5 TiB net]
 - Network: 1 GbE

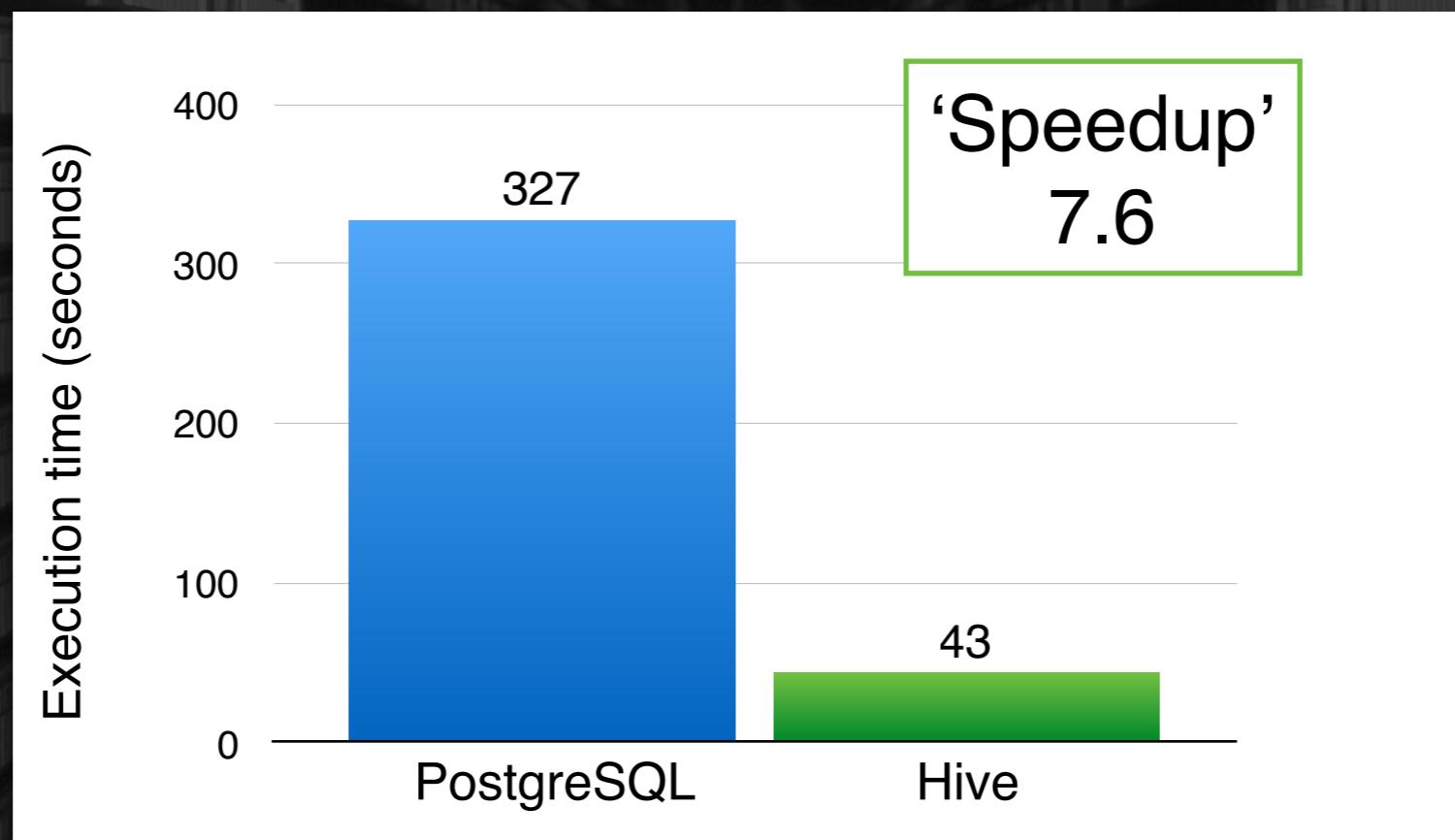
Hadoop

- Hardware:
 - Cores: 24 (Intel Xeon X5675 @ 3.07 GHz)
 - RAM: 96 GiB
 - DISK: 600 GB HDD x 8 (in RAID 6) ~ 3.6 TB net
 - Network 1GbE
- Software:
 - Scientific Linux 6.1
 - PostgreSQL 9.1

PostgreSQL

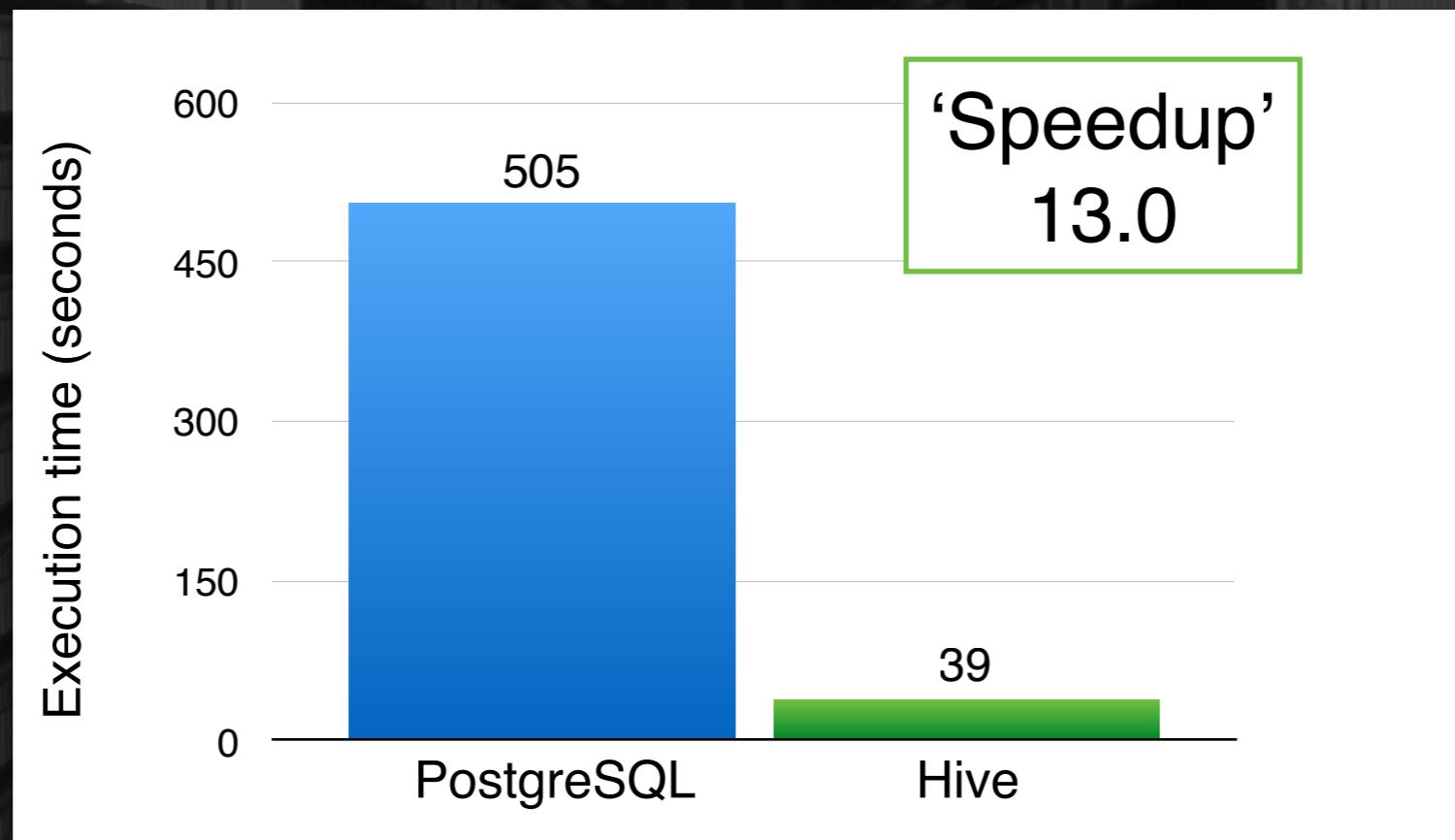
Hadoop vs. PostgreSQL

```
SELECT ra, dec, z, z_v, x_c, y_c, z_c FROM  
micecatv1 WHERE x_c < 700 AND y_c < 700 AND  
z_c < 700; (~5.8M out of ~205M rows)
```



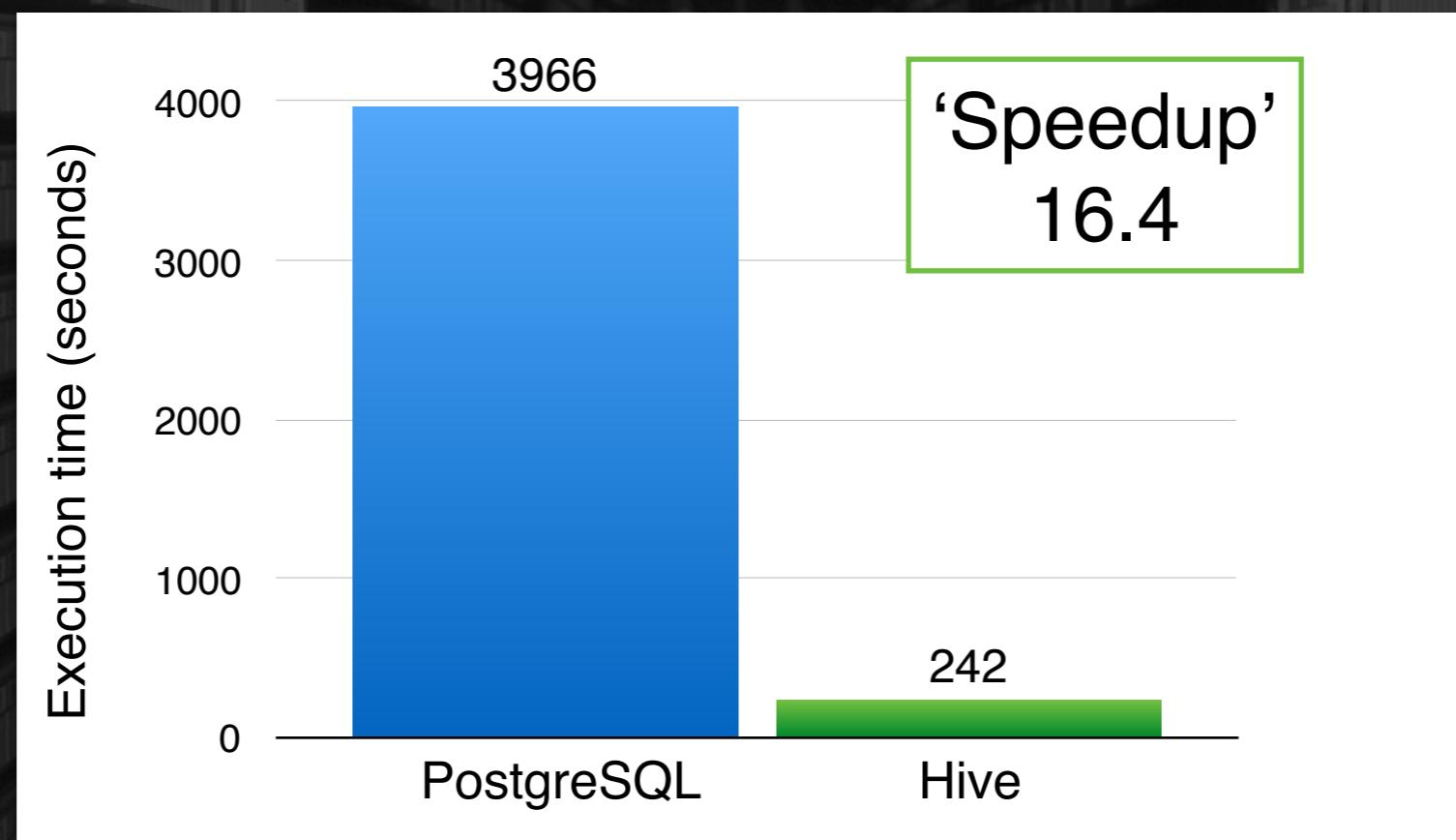
Hadoop vs. PostgreSQL

`SELECT x_c, y_c, z_c FROM micecatv1 WHERE x_c < 1e3 AND y_c < 1e3 AND z_c < 1e3; (~16.5M out of ~205M rows)`



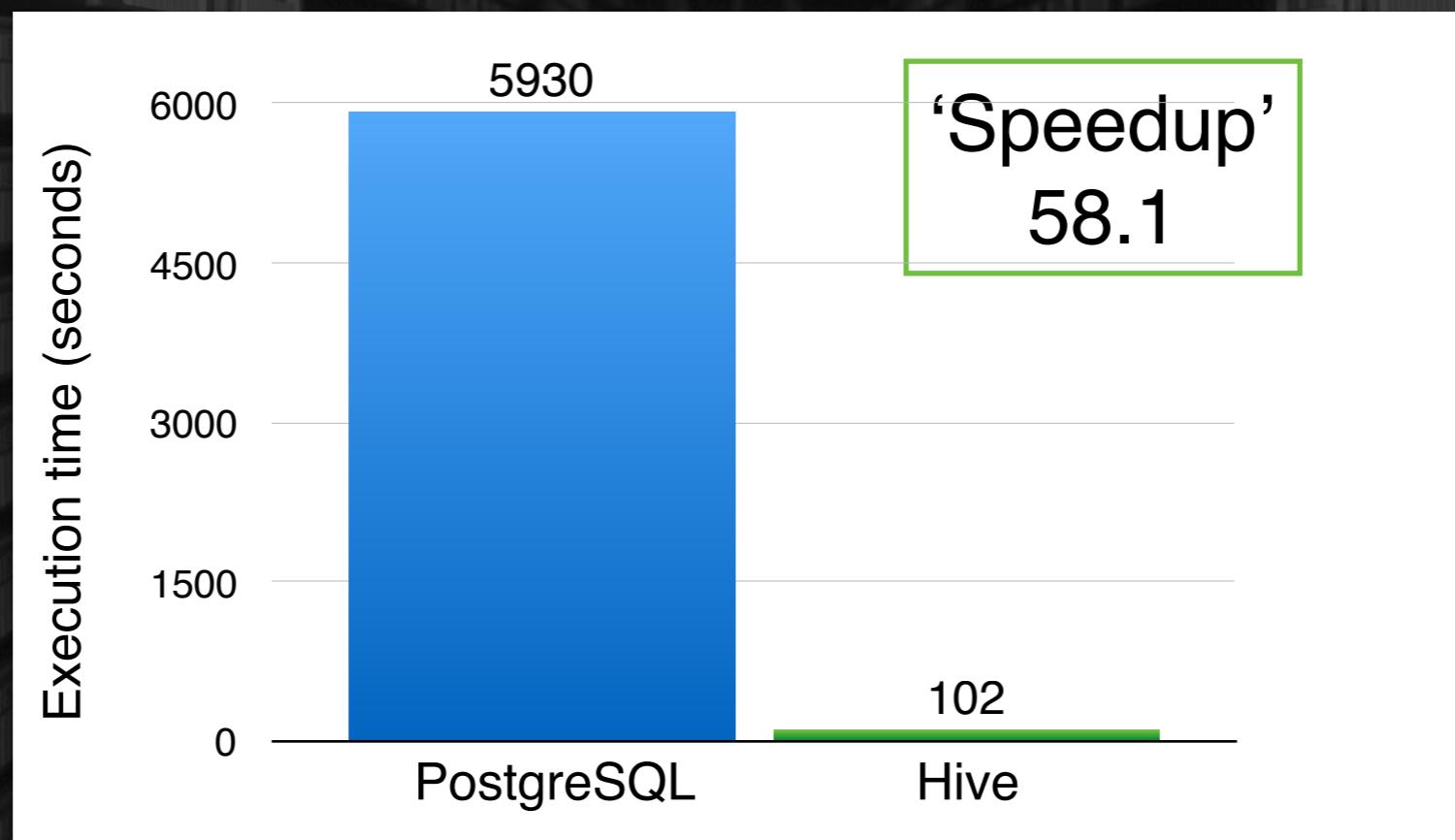
Hadoop vs. PostgreSQL

```
SELECT coadd_objects_id, ra, dec, mag_auto_i,  
magerr_auto_i desdm_zp, mean_z_bpz, z_mc_bpz  
FROM des_y1a1 WHERE modest_class = 1 AND  
flags_gold = 0 AND flags_badregion = 0;  
(~81.9M out of ~137M rows)
```



Hadoop vs. PostgreSQL

```
SELECT ra_gal, dec_gal, kappa, gamma1, gamma2  
FROM micecatv2 WHERE lmhalo >= 12.16 AND  
flag_central = 0 AND z_cgall > 0.4 AND z_cgall  
< 0.6; (~25.9M out of ~500M rows)
```

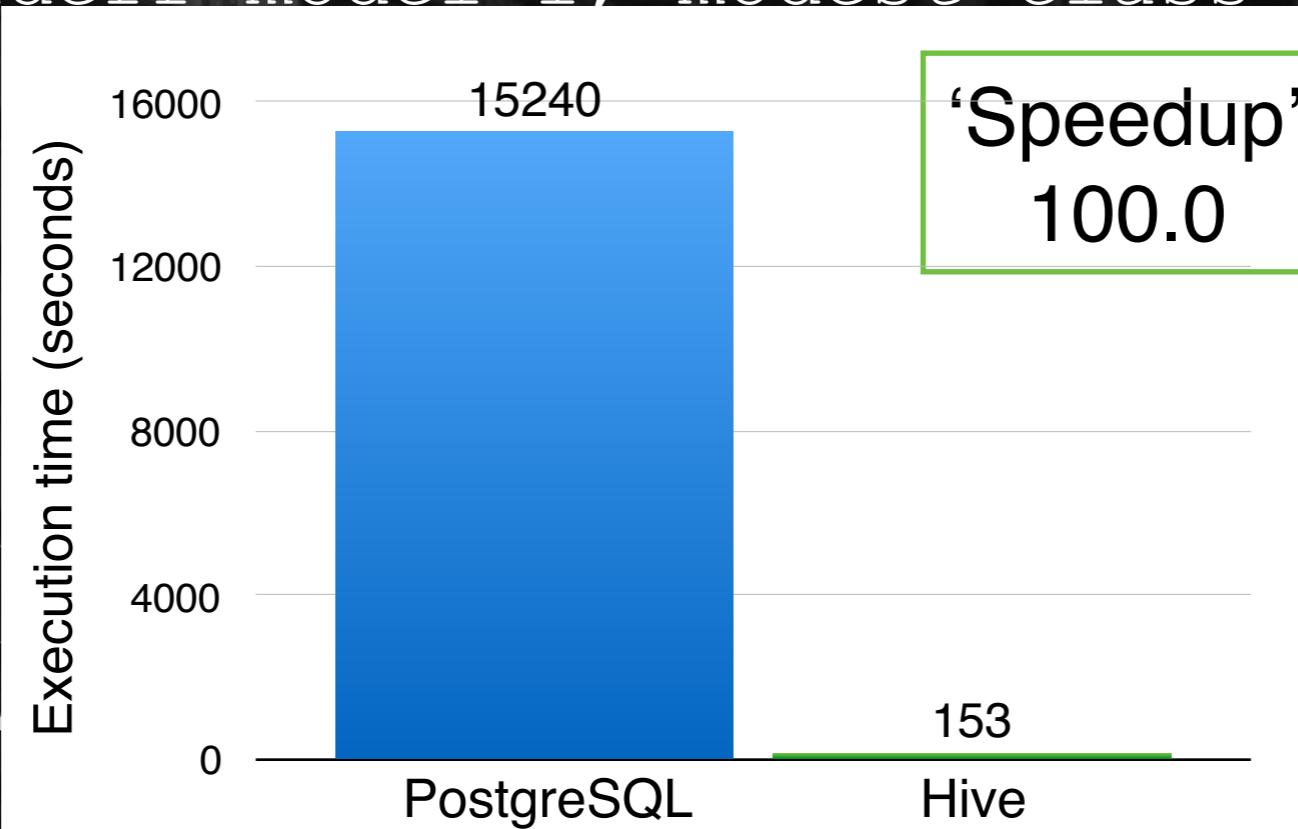


Hadoop vs. PostgreSQL

```
SELECT coadd_objects_id, ra, dec, mag_auto_g,  
mag_auto_r, mag_auto_i, mag_auto_z,  
mean_z_bpz, mode_z_bpz, median_z_bpz,  
z_mc_bpz, t_b, spread_model_i,  
spreaderr_model_i, modest_class FROM des_y1al  
WHERE mag_auto_i > 17.5 AND mag_auto_i < 22  
AND (flags_badregion <= 3 and flags_gold = 0)  
AND ((mag_auto_g - mag_auto_r) BETWEEN -1.  
and 3.) AND ((mag_auto_r - mag_auto_i)  
BETWEEN -1. and 2.5) AND ((mag_auto_i -  
mag_auto_z) BETWEEN -1. and 2.) AND (ra < 15  
or ra > 290 or dec < -35); (~34.8M out of  
~137M rows)
```

Hadoop vs. PostgreSQL

```
SELECT coadd_objects_id, ra, dec, mag_auto_g,  
mag_auto_r, mag_auto_i, mag_auto_z,  
mean_z_bpz, mode_z_bpz, median_z_bpz,  
z_mc_bpz, t_b, spread_model_i,  
spreaderr model i, modest class FROM des_y1al  
WHERE  
AND ( i < 22  
AND ( gold = 0 )  
EN -1 .  
i )  
i -  
(ra < 15  
out of  
~137M  
15240  
'Speedup'  
100.0
```

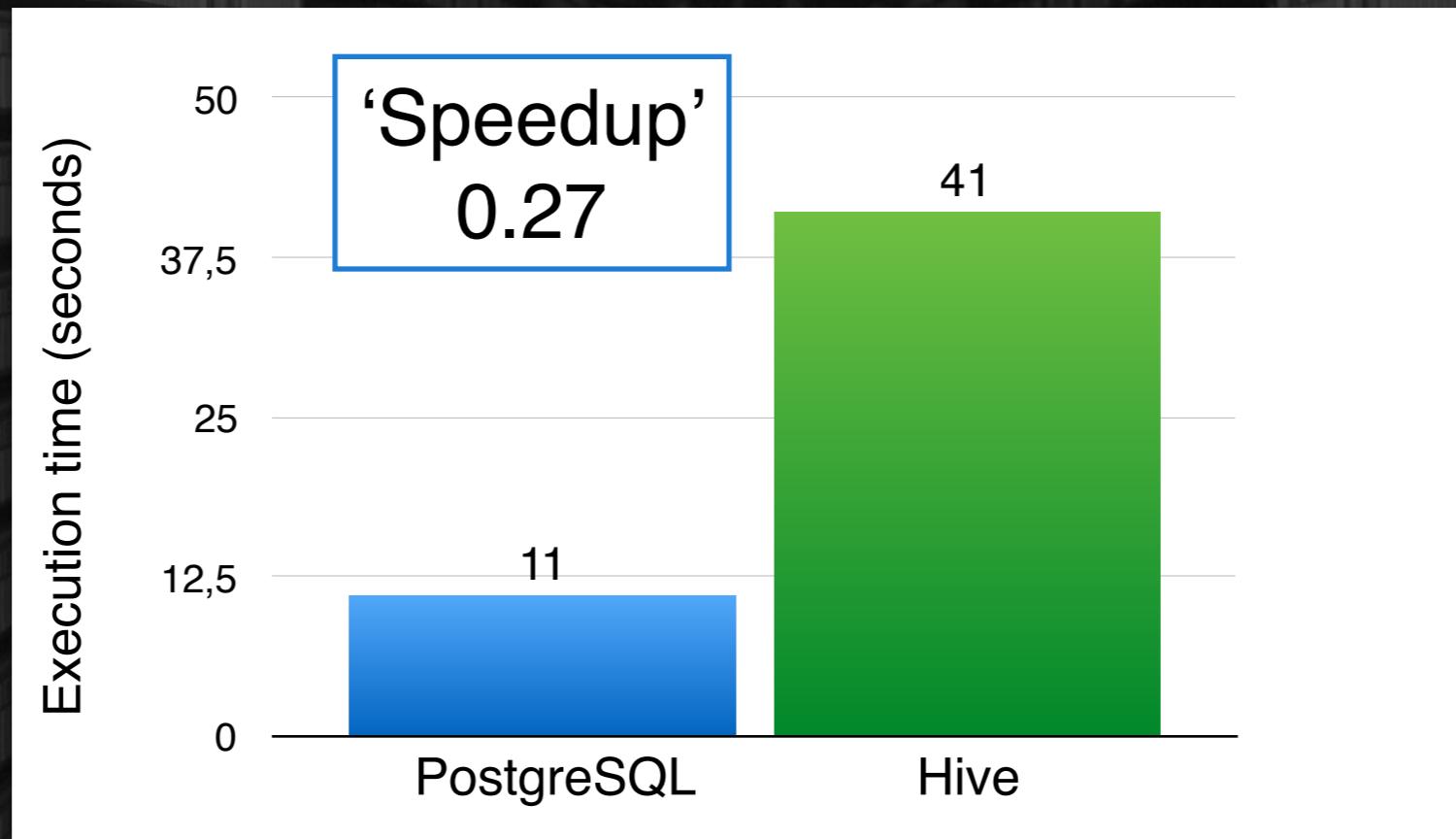


A bar chart titled 'Speedup' comparing the execution time of two systems. The y-axis is labeled 'Execution time (seconds)' and ranges from 0 to 16,000. The x-axis has two categories: 'PostgreSQL' and 'Hive'. The PostgreSQL bar is blue and reaches approximately 15,240 seconds. The Hive bar is green and reaches approximately 153 seconds. A callout box highlights the 'Speedup' factor of 100.0.

System	Execution Time (seconds)
PostgreSQL	15240
Hive	153

Hadoop vs. PostgreSQL

```
SELECT z, log_m FROM micecatv1 WHERE z < .25  
AND z > .23 AND ra < 20 AND dec < 20; (~52K  
out of ~205M rows)
```



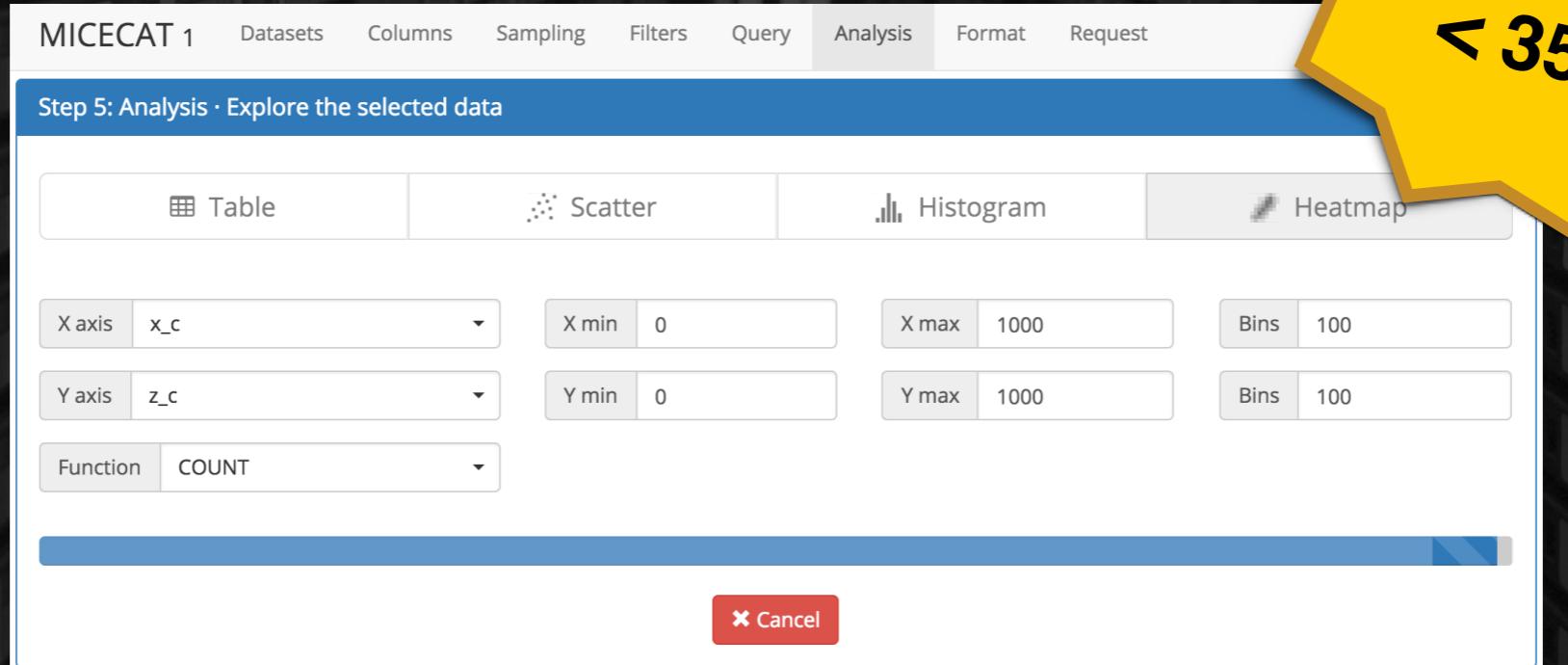
Properly using indices and a very small amount of data requested!

CosmoHub on Hadoop

- CosmoHub is a portal for real-time analysis and distribution of massive cosmology data without any SQL knowledge
- It is built on top of Hadoop and uses the Apache Hive infrastructure
- It is fully developed, hosted and operated at PIC

New features

- Real time analysis (no time constraint)
- Sampling: select a random subset of the catalog to get faster results when exploring the data
- Heatmap plot
- 2 more file formats to download the selected data: FITS and ASDF



MICECAT 1 Datasets Columns Sampling Filters Query Analysis Format Request

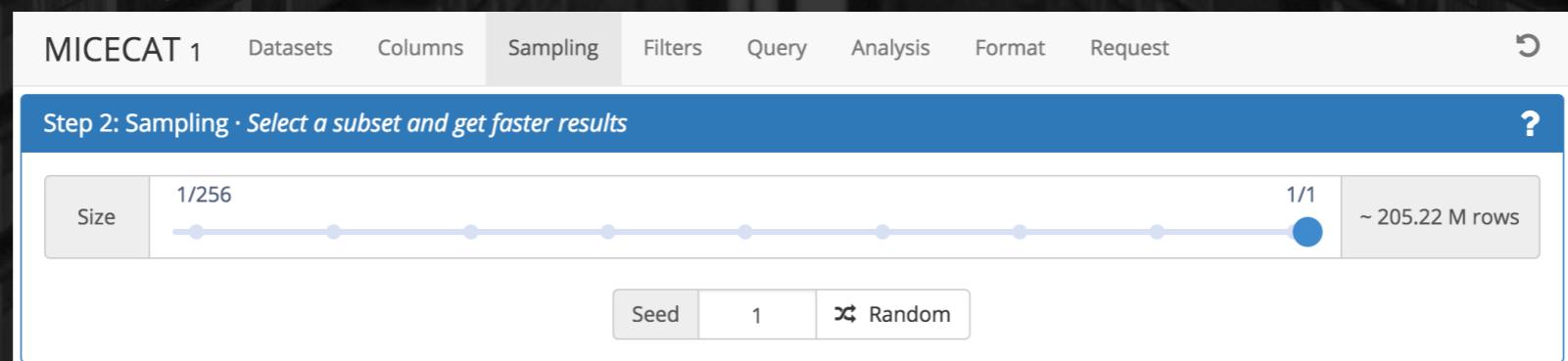
Step 5: Analysis · Explore the selected data

Table Scatter Histogram Heatmap

X axis: x_c X min: 0 X max: 1000 Bins: 100
Y axis: z_c Y min: 0 Y max: 1000 Bins: 100
Function: COUNT

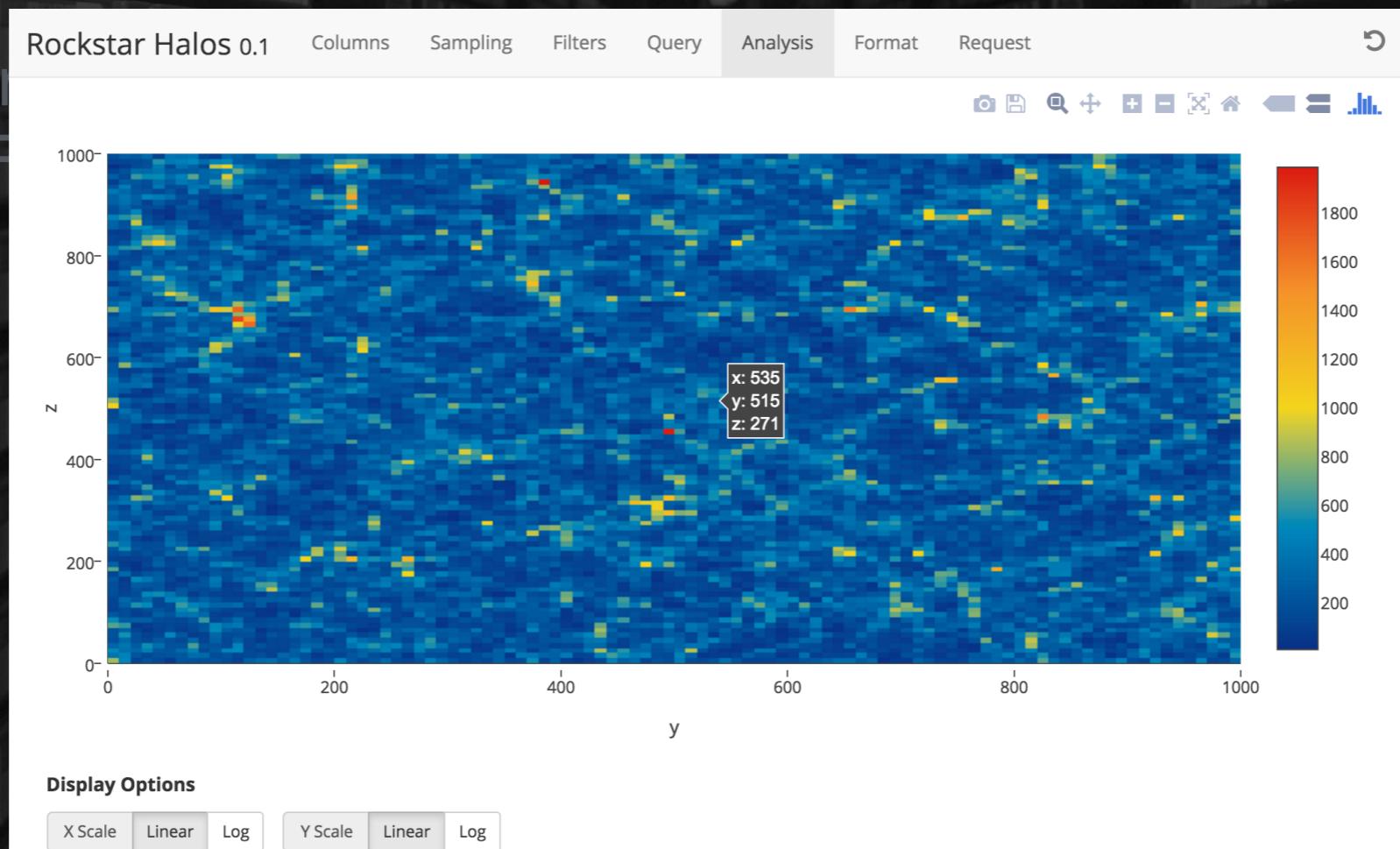
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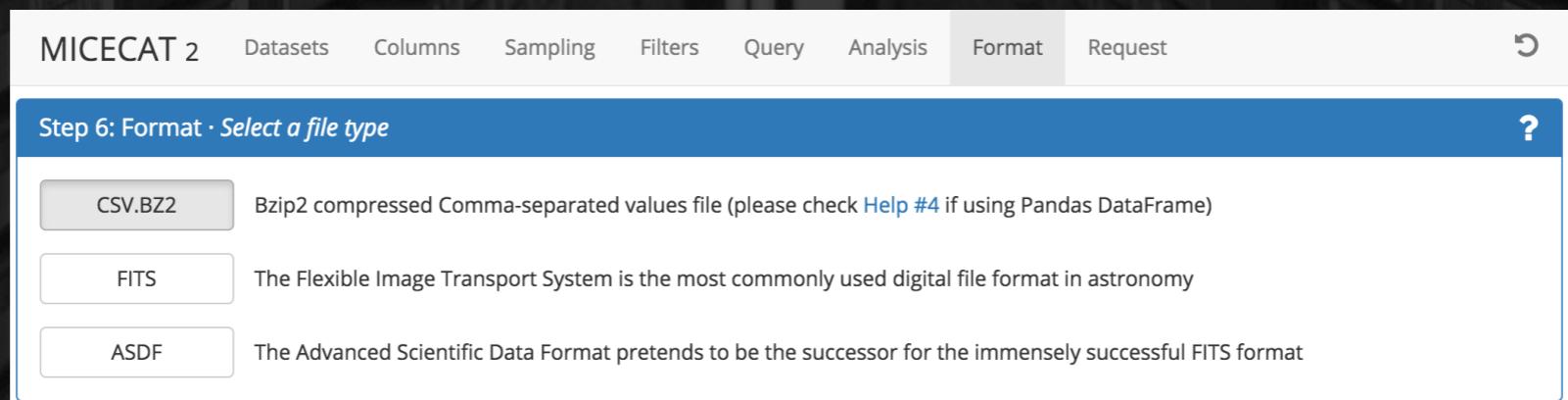
New features

- Real time analysis (no time constraint)
- Sampling: select a random subset of the catalog to get faster results when exploring the data
- Heatmap plot
- 2 more datasets
ASDF



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The screenshot shows the MICECAT 2 interface with a navigation bar at the top: MICECAT 2, Datasets, Columns, Sampling, Filters, Query, Analysis, Format, Request, and a help icon. The main area is titled "Step 6: Format · Select a file type". It displays three options: "CSV.BZ2" (selected), "FITS", and "ASDF". Each option has a description below it.

File Type	Description
CSV.BZ2	Bzip2 compressed Comma-separated values file (please check Help #4 if using Pandas DataFrame)
FITS	The Flexible Image Transport System is the most commonly used digital file format in astronomy
ASDF	The Advanced Scientific Data Format pretends to be the successor for the immensely successful FITS format

Demo

Conclusions & future work

- Great improvement in response time
- New release is more reliable
- Still exploring the vast Hadoop ecosystem

Conclusions

- New plot types and analysis
- Collaboration with more experiments
 - More data, more catalogs, more users
- Other use cases (other than Cosmology)

Future work



<https://cosmohub.pic.es>

Thanks for your attention!

Backup slides

Hive tuning

- We have set the platform so that queries over large tables are really fast:
 - Apache Tez execution engine instead of the venerable Map-reduce engine
 - ORCfile: a new table (column based) storage format
 - Vectorized query technique: batches of 1024 rows at once



Load balancing

- Set up two different queues given the two different profiles:
 - ‘Interactive’: real-time analysis (low latency)
 - ‘Batch’: custom catalogs (high latency)
- Configure queue shares and preemption:
 - batch jobs take idle resources to maximize efficiency (10-90)
 - interactive jobs can take resources from batch queue (90-100)

Backend

- ReST API powered by Flask:
 - flask-restful - ReST framework
 - sqlalchemy - database ORM
 - websockets - bidirectional communications
 - gevent - asynchronous framework
 - pyhive - hive connection library
 - pyhdfs - hdfs bindings



Frontend

- Responsive Web interface powered by:
 - Angular JS - web app oriented HTML framework
 - Bootstrap - responsive frontend framework
 - Plot.ly for plotting
 - Wordpress as backend to edit "static" content



Demo

CosmoHub YouTube channel