

# 深入Fast-Cubing及TopN

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### Apache Kylin v1.5 New Features

- Plug-in Architecture
- New MR Cube Engine with fast cubing (1.5x faster)



- New HBase Storage with parallel scan (2x faster)
- Near real-time analysis (experimental)
- User defined aggregations
- Approximate Top-N Measure
- Excel / PowerBI / Zeppelin integration

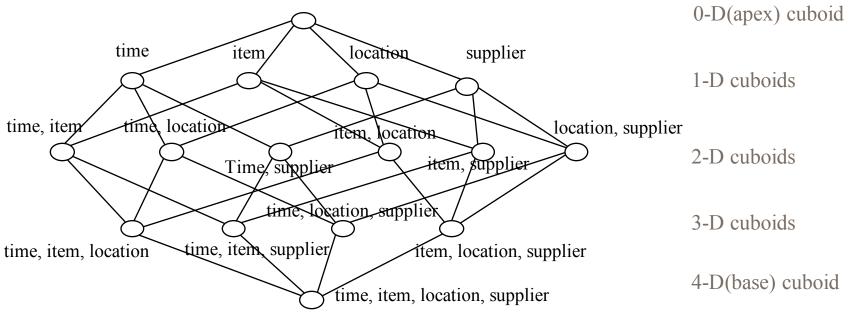




# Fast Cubing Algorithm

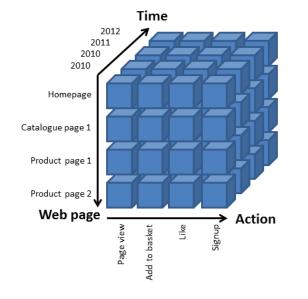


### Cube - Balance Between Space and Time



### (apex) cuboid OLAP Cube

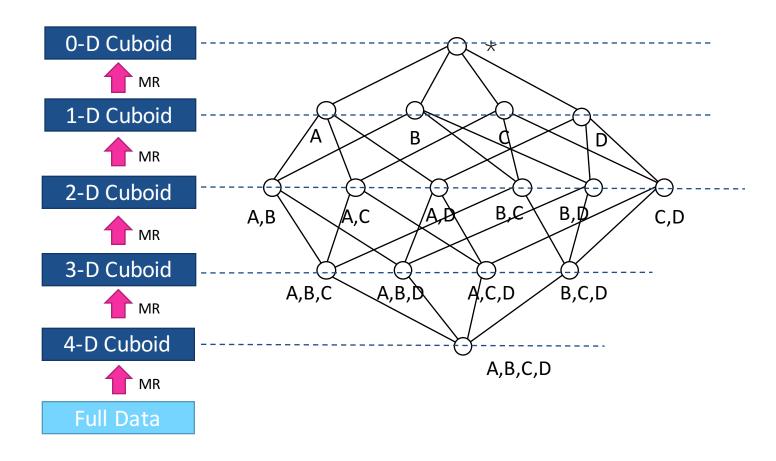
Cuboid = one combination of dimensions Cube = all combination of dimensions (all cuboids)



- Base vs. aggregate cells; ancestor vs. descendant cells; parent vs. child cells
- 1. (9/15, milk, Urbana, Dairy\_land) <time, item, location, supplier>
- 2. (9/15, milk, Urbana, \*) <time, item, location>
- 3. (\*, milk, Urbana, \*) <item, location>
- 4. (\*, milk, Chicago, \*) <item, location>
- 5. (\*, milk, \*, \*) <item>



# Layered Cubing





# Layered Cubing Summary

### Pros

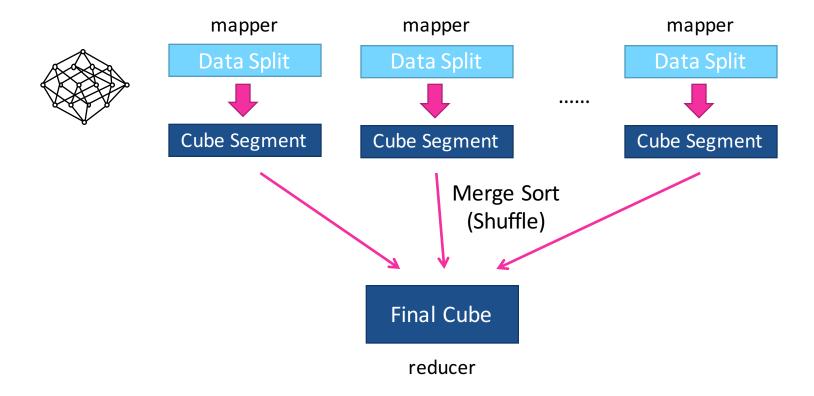
- Depends on MR to sort and aggregate -> Simple
- Low mapper CPU/Memory consumption -> Stable

#### Cons

- High network IO
- Job submission overhead
- Total duration is long



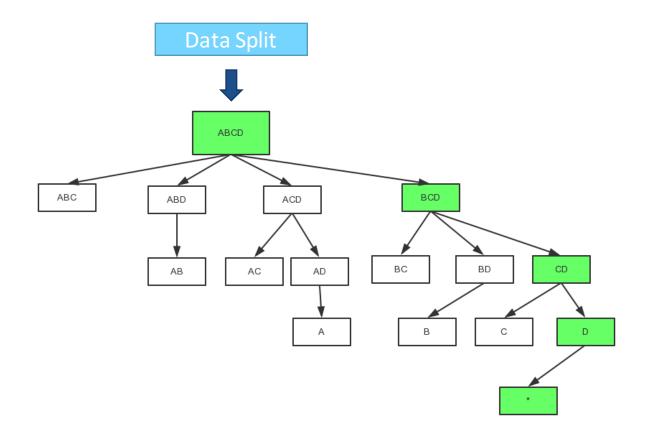
# Fast (in-mem) Cubing





# Inside Mapper

- 深度优先遍历
- 多路并发





### Fast Cubing Summary

#### Pros

- Lesser network pressure
- Independent cubing algorithm that can be reused by Streaming, Spark.
- 30%-50% faster than layered cubing

#### Cons

- Code complexity
- High mapper CPU/Mem consumption
- May fail at reduce phase in extreme case

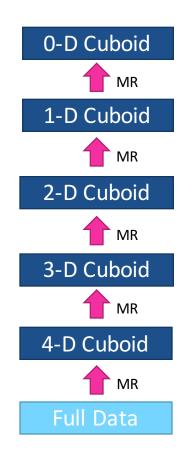


### Make fast cubing faster

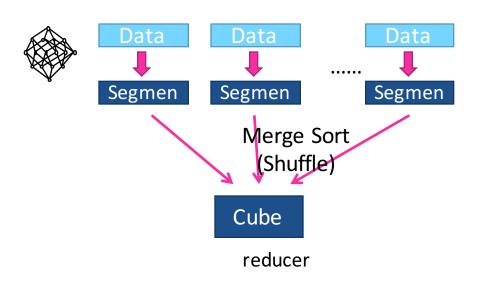
- Give more Mem to Mappers
  - By default Kylin requests 3Gb mem for mapper; Configure more if possible with "kylin.job.cubing.inmem.mrjob\_conf\_override"
- Balance between parallel computing and map side aggregation
  - By default Kylin request 1 mapper per 32 Mb source data;
  - If cube expansion is low, feed more data to a mapper can gain better overall performance;
     vice versa.
  - Customize "dfs.block.size" in kylin\_hive\_conf.xml



### Which is the Best?



OR





## Let Kylin Select!

### Smartly detect before cubing

- Estimate the micro segment size of each mapper
- Estimate the size after reduce
- Get the key overlap ratio

### Decide cubing algorithm

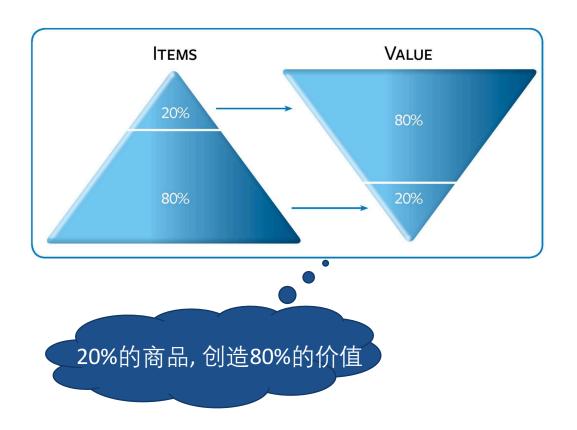
- High key overlap: by layer cubing
- Low key overlap: fast cubing
- Threshold is configurable with "kylin.cube.algorithm.auto.threshold"

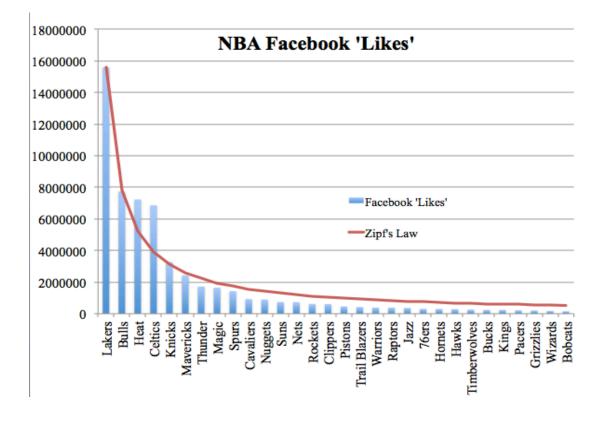


## Approximate Top-N Measure



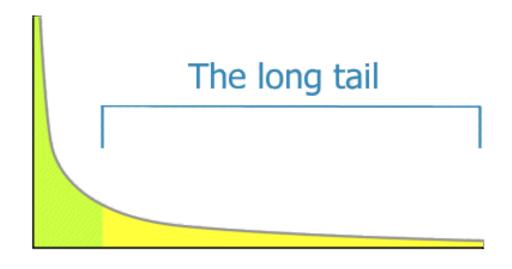
### 20/80 Rule & Zipf Law







# Long tail





UHC维度使得Cube膨胀率奇高,分析数据的难度增大

Long tail -> Ultra High Cardinality Dimension



### Sample Top-N query

```
select dt, loc, item, sum(gmv)
from kylin_sales
where dt='2015-10-01' and loc='BJ'
group by dt, loc, item
order by sum(gmv) desc
limit 100
```



### Cube without Top-N

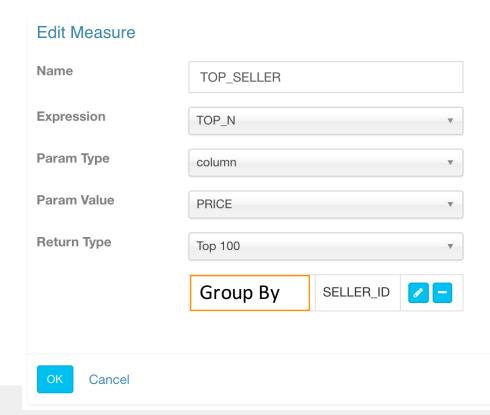
- All "group by" column need be dimension;
- Need fetch all items before sorting and return top records;

DT	LOC	ITEM	Measure Sum(gmv)
2015-10-1	ВЈ	Item-A	300
2015-10-1	ВЈ	Item-B	400
2015-10-1	ВЈ	Item-C	100
2015-10-1	ВЈ	Item-D	200
2015-10-1	ВЈ	Item-E	50
2015-10-1	ВЈ		
2015-10-1	BJ	Item-XXXXXXXXX	200



### Approximate Top-N Measure

- Define Top-N as a measure;
- Calculate Top records during cube build;
- Answer Top-N queries directly from pre-calculation.





### Approximate Top-N Algorithm

### SpaceSaving Top-N

 Ahmed Metwally, et al. "Efficient computation of frequent and top-k elements in data streams". Proceeding ICDT'05 Proceedings of the 10th international conference on Database Theory, 2005.

### A parallel SpaceSaving

• Massimo Cafaro, et al. "A parallel space saving algorithm for frequent items and the Hurwitz zeta distribution". Proceeding arXiv: 1401.0702v12 [cs.DS] 19 Setp 2015.



### Cube with Top-N

- Move High the cardinality column to Measure, together with metric.
- Records are ordered, tail records are truncated.

DT	LOC	Measure Top-N
2015-10-1	BJ	Item-1,10000
		Item-2,9000
		Item-5000,200
2015-10-1	SH	Item-1,10000
		Item-2,9000
		Item-5000,200



# Top-N Merge

DT,LOC	TopN
2015-10-1,BJ	Item A, \$500 Item B, \$490 Item C, \$400 
2015-10-1,SH	Item A, \$800 Item C, \$700 Item B, \$600 

agg

DT	TopN
2015-10-1	Item A, \$1300 Item C, \$1100 Item B, \$1090 

••••



### Top-N Summary

#### Pros

- Saving Space, cube size can decrease 90%
- Avoid massive scan, query performance can improve more than 10 times

#### Cons

- Approximate calculation
- Error accumulates at high level aggregation
- Error ratio depends on data distribution, space allocated, aggregation level, etc.



### **About Kylin**

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