

# 基于ClickHouse的海量数据交互式分析场景实践

胡甫旺  
哔哩哔哩OLAP平台



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# ClickHouse在B站

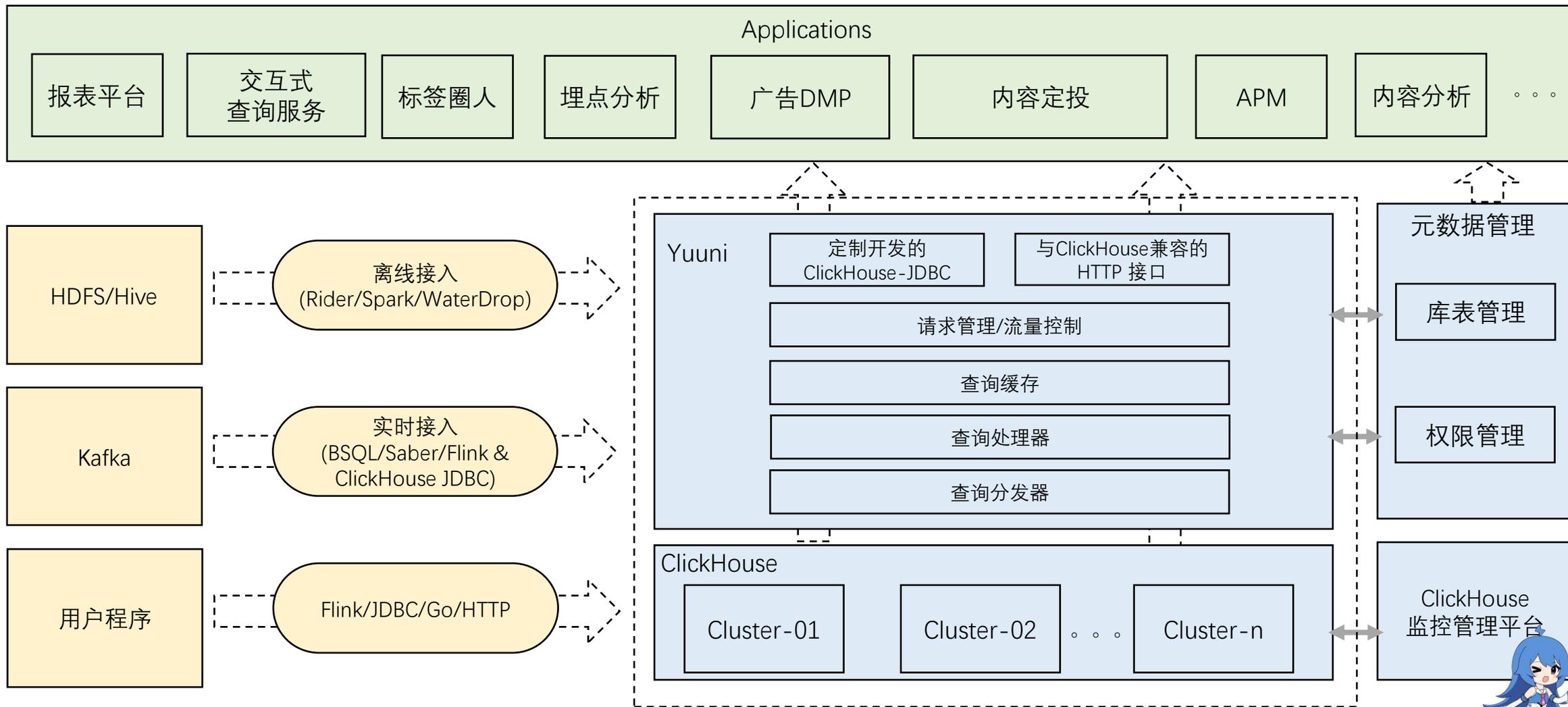


# 站ClickHouse应用概况

- ❖ 近200个节点，20+个集群
- ❖ 日均4000+亿条数据摄入
- ❖ 日均450+万次Select请求
- ❖ 日均2000+万次Insert请求
- ❖ 应用场景包括（不限于）：
  - 用户画像分析
  - 圈人定投
  - 广告DMP（包括统计分析，人群预估）
  - 电商交易分析
  - OGV内容分析
  - 日志/Trace分析
  - APM (Application Performance Management)



# 基于ClickHouse的交互式OLAP技术架构



# ClickHouse 集群规范化管理

## ❖ 集群信息规范化:

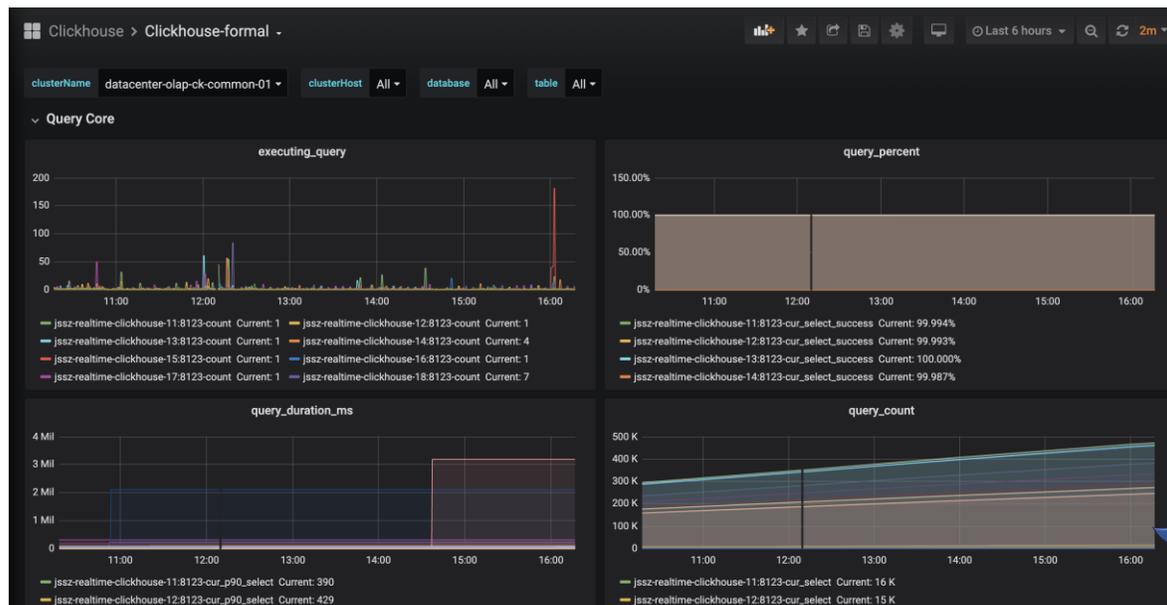
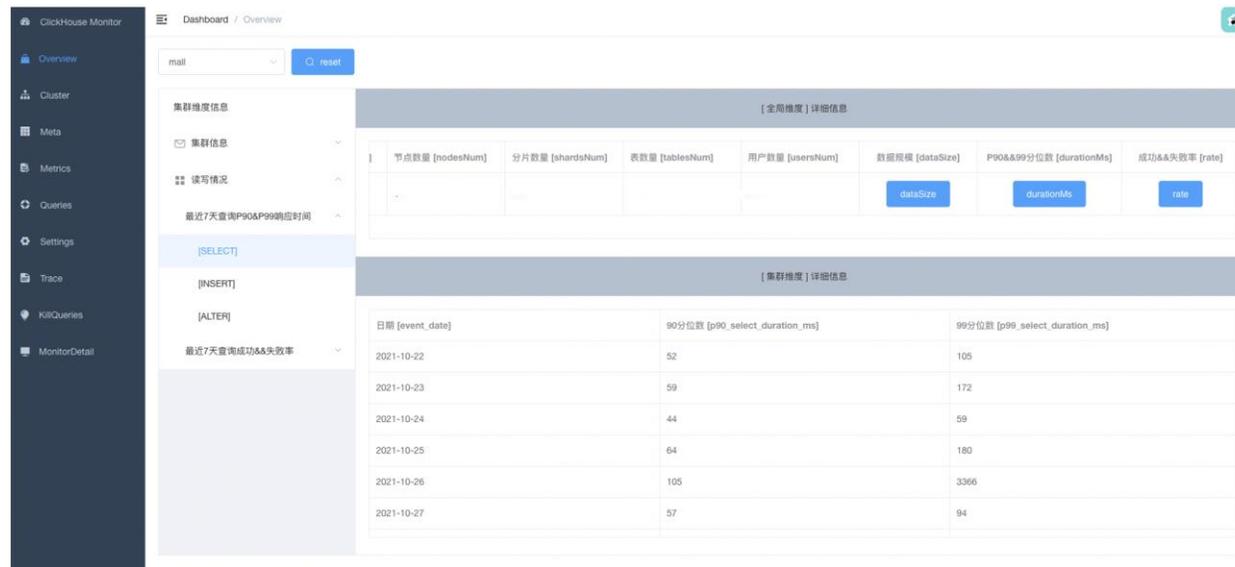
- 服务树规范
- 机器命名规范
- 监控指标统一规范
- 库表命名规范

## ❖ 业务接入规范化:

- 建表
- 数据导入
- 测试流程
- 风险评估

## ❖ 集群操作规范和服务化:

- 新建集群
- 集群扩容
- 数据迁移
- 故障修复



# ClickHouse as Service

## ❖ Berserker数据源管理:

- 建表
- 修改表元数据
- 表元数据管理

## ❖ Yuuni:

- 屏蔽集群信息
- 原生JDBC, HTTP接口
- 读写分离
- 动态查询缓存
- 流量控制

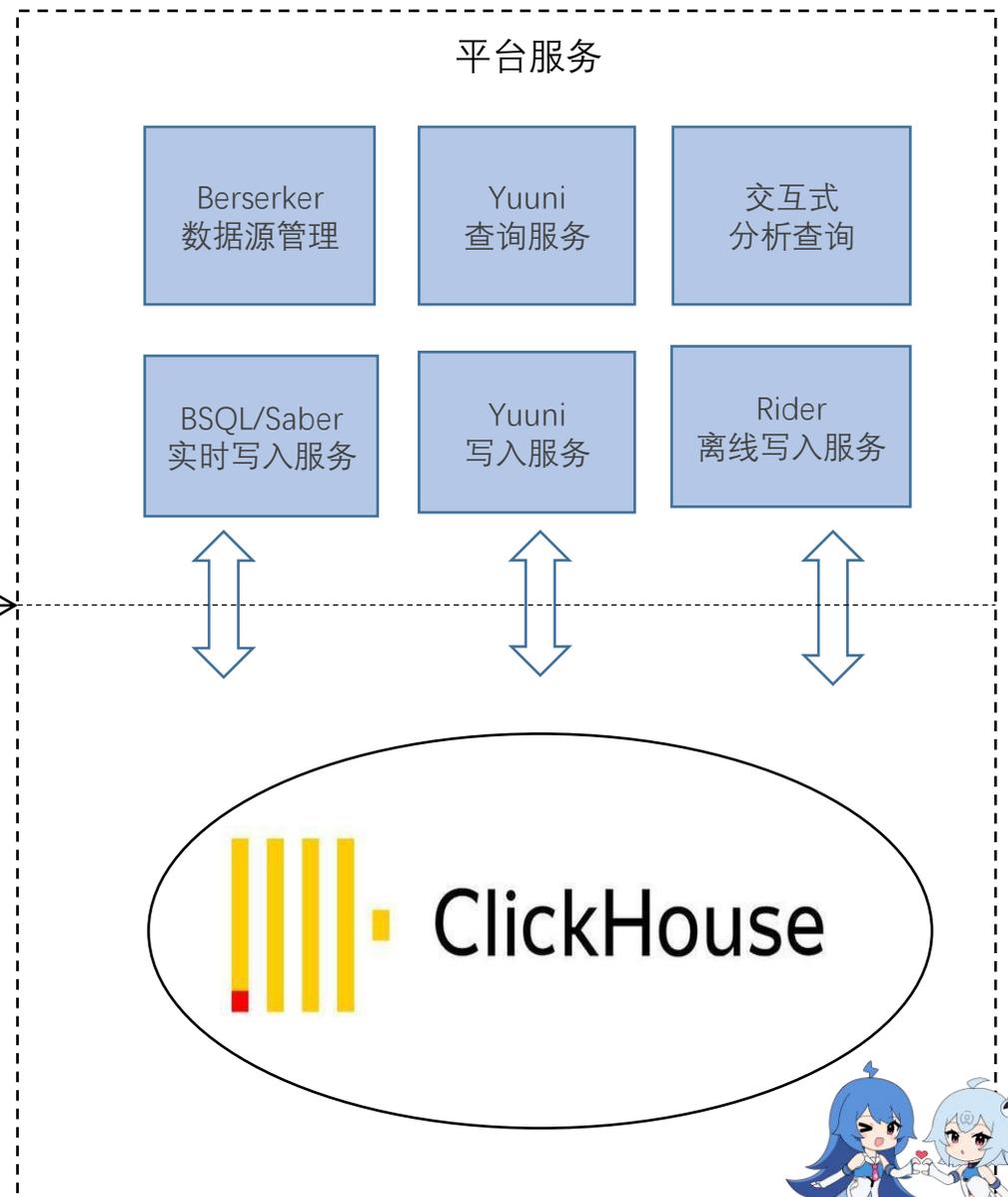
## ❖ 交互式分析查询: Superset提供即时查询能力

## ❖ 离线写入服务 (Rider)

## ❖ 实时写入服务 (BSQL/Saber)



用户



# 场景分析



# 事件分析

- ❖ 海量埋点事件数据，日增数据千亿级。
- ❖ 用户行为事件的多维度分析场景。
- ❖ 事件包含公共属性和私有属性，均可作过滤和聚合维度。
- ❖ 不同事件有不同的私有属性字段。
- ❖ 动态选择的过滤维度和聚合维度。
- ❖ 交互式分析延迟要求 (5秒内)。



# 事件分析

SQL-V1:

```
SELECT log_date, brand, indicator
FROM (
  SELECT log_date, brand, SUM(pv) AS indicator
  FROM user_event_table
  WHERE log_date BETWEEN '20211103' AND '20211103'
    AND event_id = 'event_1'
    AND app_id = 1
    AND extended_field_values[indexOf(extended_field_keys, 'goto')]='new_tunnel'
    AND extended_field_values[indexOf(extended_field_keys, 'sub_goto')]='activity'
  GROUP BY log_date, brand
  UNION ALL
  SELECT " AS log_date, brand, SUM(pv) AS indicator
  FROM user_event_table
  WHERE log_date BETWEEN '20211103' AND '20211103'
    AND event_id = 'event_1'
    AND app_id = 1
    AND extended_field_values[indexOf(extended_field_keys, 'goto')]='new_tunnel'
    AND extended_field_values[indexOf(extended_field_keys, 'sub_goto')]='activity'
  GROUP BY brand
) AA
ORDER BY indicator DESC
LIMIT 500;
```

SQL-V2:

```
SELECT log_date, brand, SUM(pv) AS indicator
FROM user_event_table
WHERE log_date BETWEEN '20211103' AND '20211103'
  AND event_id = 'event_1'
  AND app_id = 1
  AND extended_props['goto']='new_tunnel'
  AND extended_props['sub_goto']='activity'
GROUP BY GROUPING SETS ((log_date, brand), (brand))
ORDER BY SUM(pv) DESC
LIMIT 500;
```



# 漏斗分析

- ❖ 预定义事件漏斗。
- ❖ 支持各个事件单独设置过滤条件。
- ❖ 查询时间跨度最大一个月。
- ❖ 数据按user id做Sharding，查询下推。



# 漏斗分析

原始SQL:

```
SELECT level, uniq(buvid) ③
from (
SELECT ②
  buvid,
  windowFunnel(86400)(time_iso, event_id = 1, event_id = 2, event_id = 3, event_id = 4) AS level
FROM
( ①
SELECT
  time_iso,
  event_id,
  buvid
FROM event_analysis_table_dst
where log_date='20201201'and event_id in (1, 2, 3, 4)
) GROUP BY buvid
) GROUP BY level
```

执行步骤:



# 漏斗分析

优化SQL-V1（数据按buid做sharding）：

```
select level, count(distinct buvid) ②  
from (  
  SELECT ①  
  buvid,  
  windowFunnel(3600)(time_iso, event_id = 1, event_id = 2, event_id = 3, event_id = 4) AS level  
from event_analysis_table_dst  
where log_date='20201201' and event_id in (1, 2, 3, 4)  
GROUP BY buvid  
settings distributed_group_by_no_merge=1  
) group by level
```

优化结果：较原始SQL性能提升**5+**倍

执行步骤：



# 漏斗分析

优化SQL-V2（数据按buvid做sharding）：

```
select level, sum(cnt) as cnt ②
from cluster('cluster_name');

view(
select level, count(distinct buvid) as cnt ①
from (
SELECT
buvid,
windowFunnel(3600)(time_iso, event_id = 1, event_id = 2, event_id = 3, event_id =4) AS level
from event_analysis_table_local
where log_date = 20201201 and event_id in (1, 2, 3, 4)
GROUP BY buvid
) group by level)
) group by level
```

优化结果：较优化SQL-V1性能提升**30+%**

执行步骤:

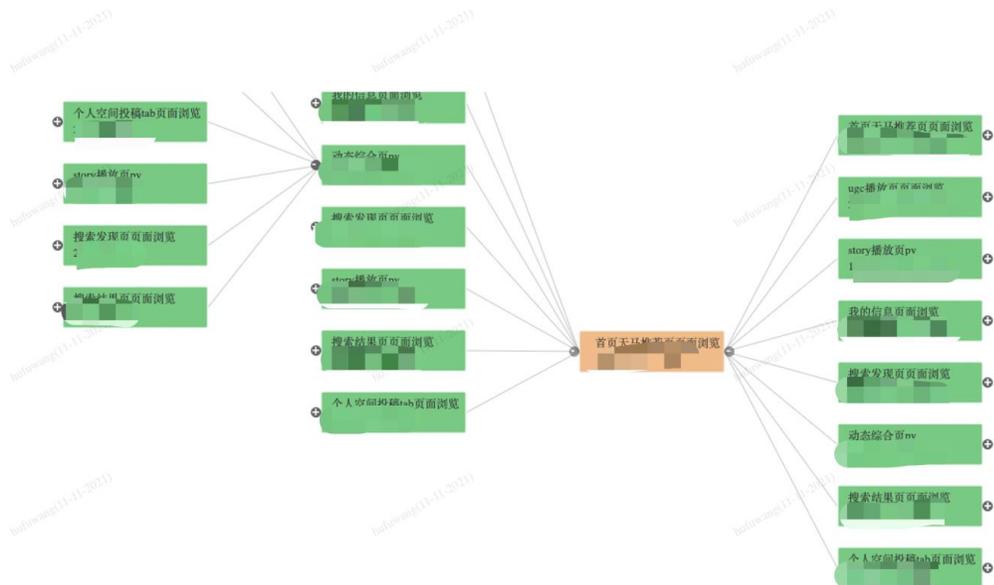


# 路径分析

- ❖ 选定中心事件。
- ❖ 按事件窗口确定上下游事件。
- ❖ 离线Spark与计算出事件路径及相关用户id的RBM。
- ❖ 离线计算结果导入ClickHouse做交互式路径分析。

用户路径图

时间 2021-11-10



# 路径分析

```
CREATE TABLE polaris.polaris_dws_flow_app_path_eventid_i_a_d_v3_local
(
  `event_id` String,
  `event_ids` Array(String),
  `path_direction` Int16,
  `buids_rbm_string` String,
  ...
  `path_lenth` Int32,
  `buids_roaring_bitmap` AggregateFunction(groupBitmap, UInt64)
  MATERIALIZED base64Decode(buids_rbm_string),
  INDEX platform_idx platform TYPE minmax GRANULARITY 3
);
```

```
SELECT log_date, path, path_direction, bitmapCardinality(groupBitmapOrState(buids_roaring_bitmap)) AS uv
FROM
(
  SELECT log_date, path, path_direction, buids_roaring_bitmap
  FROM cluster('polaris_cluster_replica', view(
    SELECT log_date, path_direction, [event_ids[1]] AS path,
           groupBitmapOrState(buids_roaring_bitmap) AS buids_roaring_bitmap
    FROM polaris.polaris_dws_flow_app_path_eventid_i_a_d_v3_local
    WHERE (log_date = '20211105') AND (event_id = 'mall.magic-item-view.0.0.pv')
          AND (app_id = 1) AND (path_lenth = 5)
    GROUP BY log_date, path, path_direction
  ))
)
GROUP BY log_date, path, path_direction
UNION ALL
...
```



# 引擎增强



# ClickHouse引擎内核增强

## Map数据类型

- ❖ 适用于动态列场景，较Array易用性更强。
- ❖ 添加多种类型索引支持，加速map取数。
  - bloom\_filter
  - tokenbf\_v1
  - ngrambf\_v1

```
SELECT
  buvid, XXX
FROM polaris.polaris_event_analysis_table_v2_local
WHERE extended_field_values[indexOf(extended_field_keys, 'goto')] = 'banner'
```

支持Map类型后

```
SELECT
  buvid, XXX
FROM polaris.polaris_event_analysis_table_v2_local
WHERE extended_fields['goto'] = 'banner'
```

```
CREATE TABLE polaris.polaris_event_analysis_table_v2_local
(
  `buvid` String CODEC(ZSTD(15)),
  ...
  extended_field_keys` Array(String),
  extended_field_values` Array(String),
  ...
)
```

支持Map类型后

```
CREATE TABLE polaris.polaris_event_analysis_table_v2_local
(
  `buvid` String CODEC(ZSTD(15)),
  ...
  extended_fields` Map(String, String),
  INDEX idx_extended_fields TYPE tokenbf_v1(256, 3, 0) GRANULARITY 2
  ...
)
```



# ClickHouse引擎内核增强

## Map数据类型

- ❖ 添加自定义map函数，支持map数据过滤。
  - mapContainsKeyLike
  - mapExtractKeyLike
- ❖ 添加map函数对索引的支持，提升map函数性能。

```
SELECT
  buvid,
  mapExtractKeyLike(extended_fields, '%game%')
FROM polaris.polaris_event_analysis_table_v2_local
WHERE mapContainsKeyLike(extended_fields, '%game%')
```



# ClickHouse引擎内核增强

## Grouping Sets

- ❖ 支持多个维度组合做聚合。
- ❖ 简化用户SQL，多个查询合并为单个查询。
- ❖ 大幅减少磁盘读取开销。

```
SELECT
    log_date AS logDate,
    'all' AS event_id,
    CAST(uniq(buvid_I), 'Float64') AS indicator
FROM polaris.polaris_event_analysis_table_local
GROUP BY log_date
```

UNION ALL

```
SELECT
    'all' AS logDate,
    event_id,
    CAST(uniq(buvid_I), 'Float64') AS indicator
FROM polaris.polaris_event_analysis_table_local
GROUP BY event_id
```

支持Grouping Sets后

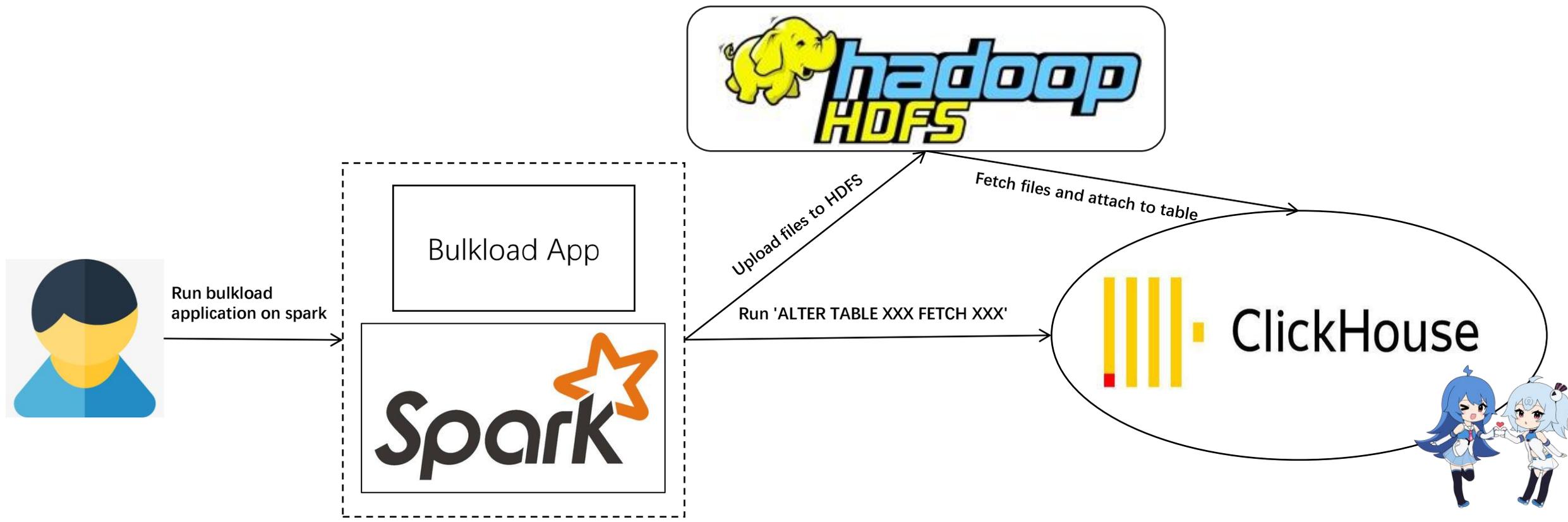
```
SELECT
    if(log_date='', 'all', log_date) AS log_date,
    if(event_id='', 'all', log_date) AS event_id,
    CAST(uniq(buvid_I), 'Float64') AS indicator
FROM polaris.polaris_event_analysis_table_local
GROUP BY GROUPING SETS ((log_date), (event_id))
```



# ClickHouse引擎内核增强

## ClickHouse Bulkload

- ❖ ClickHouse数据文件生成过程的外置到Spark application。
- ❖ ClickHouse支持ALTER TABLE ... FETCH 'hdfs://xxx'
- ❖ 数据导入过程几乎不占用ClickHouse的CPU，内存资源。



# ClickHouse引擎内核增强

## Z-ORDER

- ❖ 支持多字段构造zvalue做主键索引。
- ❖ 配合跳数索引加速多场景单字段过滤查询。
- ❖ 加速多维度范围查询场景。
- ❖ 快速跳过zcurve seam。



# ClickHouse引擎内核增强

## Z-ORDER

```
CREATE TABLE user_profiling
(
  buvid String,
  game_watch_duration_week UInt32,
  science_watch_duration_week UInt32,
  ...
  log_date String
)
PARTITION BY log_date
ORDER BY zCurve(game_watch_duration_week, science_watch_duration_week)
...
```

```
SELECT countDistinct(buvid) as uv
FROM user_profiling
WHERE game_watch_duration_week >= 3600
and game_watch_duration_week <= 10*3600
and science_watch_duration_week > 5*3600
and science_watch_duration_week <= 10*3600
```

```
CREATE TABLE user_profiling
(
  buvid String,
  game_watch_duration_week UInt32,
  science_watch_duration_week UInt32,
  ...
  log_date String,
  INDEX idx_1 game_watch_duration_week TYPE minmax GRANULARITY 2,
  INDEX idx_2 science_watch_duration_week TYPE minmax GRANULARITY 2
)
PARTITION BY log_date
ORDER BY zCurve(game_watch_duration_week, science_watch_duration_week)
...
```

```
SELECT countDistinct(buvid) as uv
FROM user_profiling
WHERE game_watch_duration_week >= 3600
and game_watch_duration_week <= 10*3600
```



# Future Work



# Future Work

- ❖ 实现ClickHouse离线和实时写入的原子性
- ❖ 集群虚拟化：ClickHouse on k8s
- ❖ Map类型进一步改造：用隐式列实现Map类型
- ❖ 存算分离方案实现
- ❖ 更多索引类型支持：如支持Bit-sliced index



# Q&A

