

A PRACTICAL INTRODUCTION TO HANDLING LOG DATA IN CLICKHOUSE

with Robert Hodges

Introduction to Presenter





30+ years on DBMS plus virtualization and security.

ClickHouse is DBMS #20



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Leading software and services provider for ClickHouse

Major committer and community sponsor in US and Western Europe



Introduction to ClickHouse

Understands SQL

Runs on bare metal to cloud

Shared nothing architecture

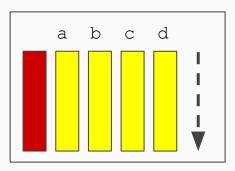
Uses column storage

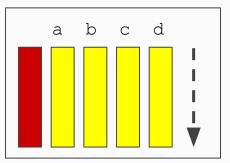
Parallel and vectorized execution

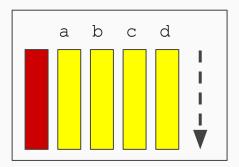
Scales to many petabytes

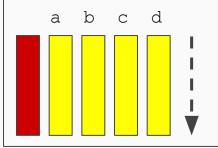
Is Open source (Apache 2.0)

And it's really fast!

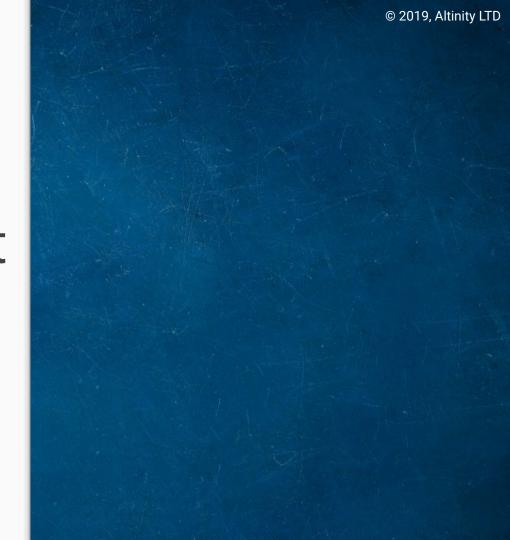






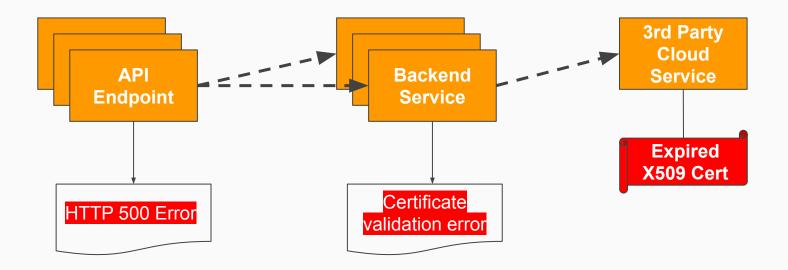


Introduction to log management

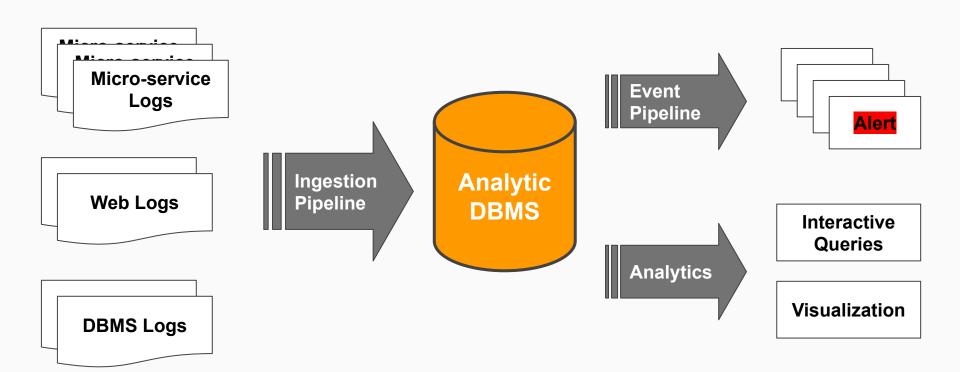


Why is log management interesting?

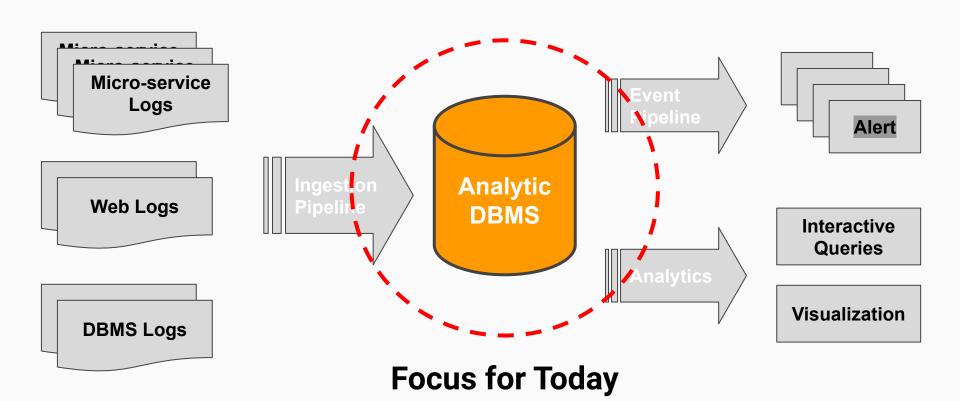
- Logs are one of the biggest sources of data about operational systems
- Service-oriented architectures require ability to scan many logs, not just one
- Timely access to log data can improve service "-ilities"



Log management architecture



Log management architecture

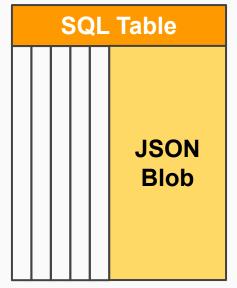


Data layout for web logs

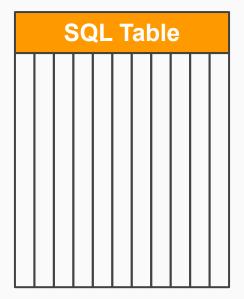
Nginx access log in JSON format:

```
"time iso8601": "2019-12-12T17:25:08-08:00",
"remote addr": "127.0.0.1",
"remote user": "",
"request": "GET /1 HTTP/1.1",
"status": "404",
"body bytes sent": "178",
"request time": "0.000",
"request length": "74",
"connection": "1",
"http referrer": "",
"http user agent": "curl/7.58.0"
```

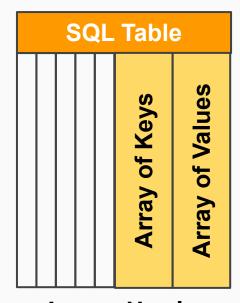
How to model log data in ClickHouse



JSON: Header values with JSON string ("blob")



Tabular: every value is a scalar column



Arrays: Header values with key-value pairs

Managing logs as JSON blobs



Simplest table design stores JSON as text

```
CREATE TABLE log row
    `file date` Date,
    `file timestamp` DateTime,
    `file name` String,
    'row' String
ENGINE = MergeTree
PARTITION BY file date
ORDER BY file name
SETTINGS index granularity = 512
```

Write a Python script to create CSV rows

```
#!/usr/bin/env python3
import sys, os.path, datetime
FILE = sys.argv[1]
with open (FILE) as f:
   for line in f:
       dt = datetime.datetime.fromtimestamp(os.path.getctime(FILE))
       print("'{0}', '{1}', '{2}', '{3}'".format(
               dt.date().isoformat(),
               dt.isoformat(timespec='seconds'),
               FILE,
               line
```

Now load the log files

```
query="INSERT INTO logs.log_row FORMAT CSV"
for log_file in /var/log/nginx/access.log*
do
    ./ingest-file.py $log_file | \
    clickhouse-client --query="$query"
done
```

How big is the data??

```
SELECT
  table, name,
  sum(data_compressed_bytes) AS compressed,
  sum(data_uncompressed_bytes) AS uncompressed,
  floor((compressed / uncompressed) * 100, 4) AS percent
FROM system.columns WHERE database = currentDatabase()
GROUP BY table, name
ORDER BY table ASC, name ASC
```

—table——	name	extstyle ext	extstyle ext	—percent—
log_row	file_date	83	2576	3.222
log_row	file_name	257	33662	0.7634
log_row	file_timestamp	97	5152	1.8827
log_row	row	3485	353410	0.9861

Use JSON* functions to get JSON

```
-- Get a JSON value
SELECT JSONExtractString(row, 'status')
FROM log row LIMIT 3
-- Get it with proper type.
SELECT toInt16(JSONExtractString(row, 'status')) AS status
FROM log row LIMIT 3
-- Use the value to select rows
SELECT
   toInt16(JSONExtractString(row, 'status')) AS status
FROM log row WHERE status >= 400 LIMIT 3
```

JSON* vs visitParam functions

```
-- Get using JSON function

SELECT JSONExtractString(row, 'status')

FROM log_row LIMIT 3

-- Get it with proper type.

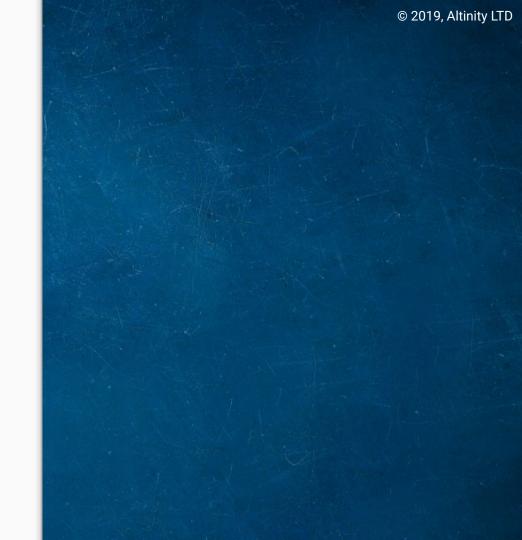
SELECT visitParamExtractString(row, 'status')

FROM log_row LIMIT 3
```

<u>FASTER</u>

Incomplete JSON parser may fail on some data

Moving from JSON to tabular schema



Tabular form is easy for queries

```
SELECT remote addr, status, count() FROM
GROUP BY remote addr, status
ORDER BY remote addr, status
```

...But converting from JSON is awkward

```
SELECT remote addr, status, count() FROM (
  SELECT
    parseDateTimeBestEffort(JSONExtractString(row, 'time iso8601'))
    AS time iso8601,
    JSONExtractString(row, 'remote addr') AS remote addr,
    JSONExtractString(row, 'remote user') AS remote user,
    JSONExtractString(row, 'request') AS request,
    toInt16(JSONExtractString(row, 'status')) AS status,
    toInt32(JSONExtractString(row, 'body bytes sent')) AS body bytes sent,
    toFloat32(JSONExtractString(row, 'request time')) AS request time,
    toInt32(JSONExtractString(row, 'request length')) AS request length,
    toInt32(JSONExtractString(row, 'connection')) AS connection,
    JSONExtractString(row, 'referrer') AS referrer,
    JSONExtractString(row, 'http user agent') AS http user agent
 FROM log row
GROUP BY remote addr, status ORDER BY remote addr, status
```

Materialized columns make life easier!

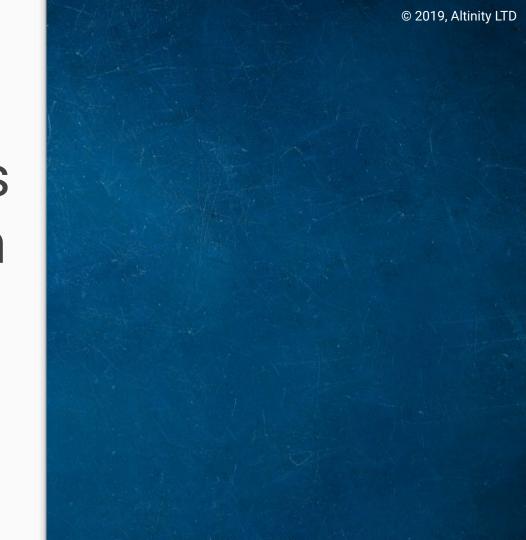
```
Materializes data on
ALTER TABLE log row
  ADD COLUMN
                                                    any merge
   status Int16 DEFAULT
      toInt16(JSONExtractString(row, 'status'))
ALTER TABLE log row
   UPDATE status = status WHERE 1 = 1
                                                   Materializes data on
                                                    INSERT or
ALTER TABLE log row
                                                    OPTIMIZE TABLE*
  ADD COLUMN
   status Int16 MATERIALIZED
      toInt16(JSONExtractString(row, 'status'))
OPTIMIZE TABLE log row FINAL
                                                              *Expensive
```

Checking the size of materialized columns

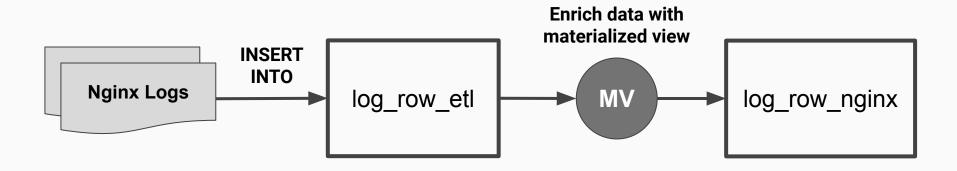
```
SELECT table, name,
   sum(data_compressed_bytes) compressed,
   sum(data_uncompressed_bytes) uncompressed,
   floor(compressed/uncompressed*100, 4)
      as percent
FROM system.columns
WHERE database = currentDatabase()
GROUP BY table, name ORDER BY table, name
```

-name	-compressed $-$	-uncompressed $-$	percent
file_date	83	2576	3.222
file_name	257	33662	0.7634
file_timestamp	97	5152	1.8827
row	3485	353410	0.9861
status	95	2576	3.6878
	<pre>file_date file_name file_timestamp row</pre>	file_date 83 file_name 257 file_timestamp 97 row 3485	file_date 83 2576 file_name 257 33662 file_timestamp 97 5152 row 3485 353410

Using mat views for log ingestion



Introducing ETL pipelines for logs



Create the base ETL table

```
CREATE TABLE log_row_etl (
   file_date Date,
   file_timestamp DateTime,
   file_name String,
   row String
) ENGINE = Null
```

Does not store data but triggers materialized views

Create the target table for NGINX logs

```
CREATE TABLE log row nginx (
    file date Date, file timestamp DateTime,
    file name String, time iso8601 Datetime,
    remote addr IPv4, remote user String,
    request String, status Int16,
   body bytes sent Int32, request_time Float32,
    request_length Int32, connection Int32,
    referrer String, http user agent String
ENGINE = MergeTree
PARTITION BY
ORDER BY (file name, time iso8601)
SETTINGS index granularity = 8196
```

A lower value can make point queries more efficient

Create materialized view for ETL

```
CREATE MATERIALIZED VIEW log row etl mv nginx TO log row nginx AS
  SELECT
    file date, file timestamp, file name,
    parseDateTimeBestEffort(JSONExtractString(row, 'time iso8601'))
    AS time iso8601,
    cast(IPv4StringToNum(JSONExtractString(row, 'remote addr')) AS IPv4)
    AS remote addr,
    JSONExtractString(row, 'remote user') AS remote user,
    JSONExtractString(row, 'request') AS request,
    toInt16(JSONExtractString(row, 'status')) AS status,
    toInt32(JSONExtractString(row, 'body bytes sent')) AS body bytes sent,
    toFloat32(JSONExtractString(row, 'request time')) AS request time,
    toInt32(JSONExtractString(row, 'request length')) AS request length,
    toInt32(JSONExtractString(row, 'connection')) AS connection,
    JSONExtractString(row, 'referrer') AS referrer,
    JSONExtractString(row, 'http user agent') AS http user agent
  FROM log row etl
```

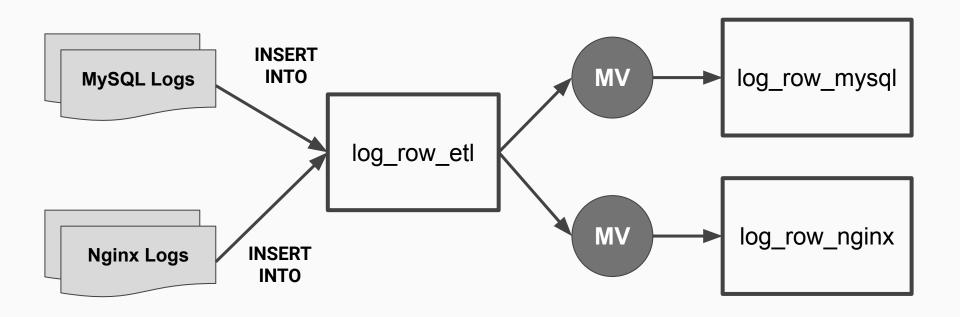
Reload the load files

```
query="INSERT INTO logs.log_row_etl FORMAT CSV"
for log_file in /var/log/nginx/access.log*
do
    ./ingest-file.py $log_file | \
    clickhouse-client --query="$query"
done
```

Now we can query normally

```
SELECT
   remote addr,
   status,
   count(*)
FROM log row nginx
GROUP BY remote addr, status
ORDER BY remote addr, status
  -remote addr——status——count()-
    127.0.0.1
                   200
                              857
    127.0.0.1
                   404
                              431
```

Extending for different log types



Implement branching with WHERE clauses

```
CREATE MATERIALIZED VIEW log_row_etl_mv_nginx TO log_row_nginx AS

SELECT

file_date, file_timestamp, file_name,

parseDateTimeBestEffort(JSONExtractString(row, 'time_iso8601'))

AS time_iso8601,

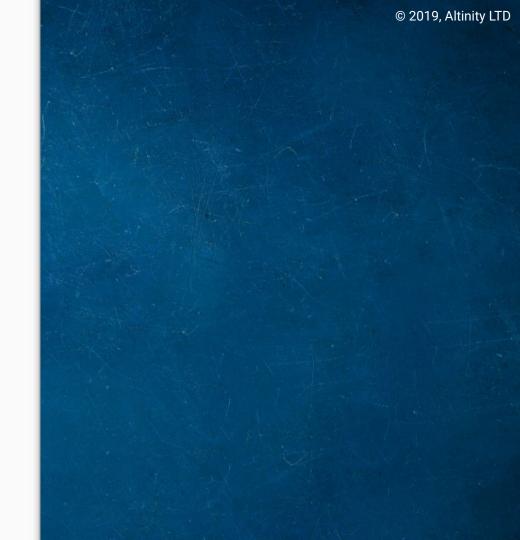
cast(IPv4StringToNum(JSONExtractString(row, 'remote_addr')) AS IPv4)

AS remote_addr,

...

FROM log_row_etl WHERE file_name_like '/var/log/nginx%'
```

Wrap-up



Storing logs in ClickHouse: summary

- Store data as JSON blobs
- Fetch out values using JSON* or visitParam* functions
- Use materialized columns to turn JSON incrementally into table columns
- Use Null table engine and materialized views to convert logs to tables

This is just the beginning

We plan future webinars to explore additional topics related to log management

- Tools for log ingest
- Query and visualization of log contents
- Generating alerts on log events
- Performance tricks for log data

Thank you!

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Contact us for a
1-hour consultation!

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