

1. Requirement to calculate unique users per minute, day, week, month, year.

Estimator is appropriate bitmap structure (hash table) available in HyperLogLog and Linear counting algorithms. Idea is to produce estimator every minute (minimum time interval).

Record structure produced directly (online) from Kafka topic (stream reader) can be:

```
{"ts":<timestamp>, "range":<range>, "ec":<ecvalue>, "est":<estimator>}
```

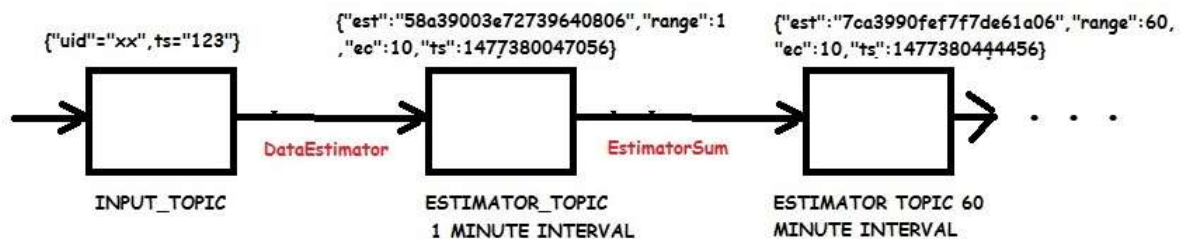
Value <timestamp> unix time.

Value <range> is in minutes (1, 5, 60, 1440, ...) or coded (1 for 1 min, 2 for 5 min, 3 for 60 min etc).

Value <ecvalue> is parameter of HyperLogLog (<=30) or Linear counting (>30).

Value <estimator> is hex dump of serialized object (bitmap).

Output is written to separate resulting Kafka topic every minute.



To sum estimators, "ec" parametar (bitmap number of bytes in the case of Linear counting) should be the same.

Estimator addition is scalable. Cardinality estimation is based on resulting estimator for given time range. Resulting estimator has the same size as any estimator used in calculation.

Proposed granularity (configurable):

- 1 minute interval – last 2 days-+ ($60 \times 24 \times 2 = 2880$ records)
- 5 minute interval – current week ($12 \times 24 \times 7 = 2016$ records)
- 60 minute interval – last 3 months (max $24 \times 31 \times 3 = 2231$ records)
- 1 day interval – last 6 years (max $366 \times 6 = 2196$ records)

Every category is written into dedicated Kafka topic. One criteria is to maintain similar number of records in every topic (in this case ~ 2200 records).

Separate batch processing (cron job) maintains granularity of estimators (addition and purge). Optional backup on permanent media is possible.

2. There can be several estimator records containing same timestamp and range.

Query from application reads all records that match given criteria and performs estimator addition to produce resulting cardinality estimation.

To minimize error, resulting cardinality should be less than expected cardinality (provided as input parametar)

Any batch processing (late packets, packet fixes, restore from backup, etc) produces same way separate estimator records with appropriate time stamps and range according to configured granularity. Another batch job can consolidate estimators performing addition and purge.

Scalability is achieved same way:

Separate Kafka nodes read from one or more Kafka topic and produce estimator records in proposed manner.

3. Expected cardinality

Expected cardinality is the most important parameter of HyperLogLog and Linear counting algorithms.

<http://highscalability.com/blog/2012/4/5/big-data-counting-how-to-count-a-billion-distinct-objects-us.html>

HyperLogLog - an improved version of LogLog that is capable of estimating the cardinality of a set with accuracy = $1.04/\sqrt{m}$ where expected cardinality is $m = 2^b$. So we can control accuracy vs space usage by increasing or decreasing $b \leq 30$.

Linear counting – parameter n is size of bit array in bytes. Expected cardinality is $m = 8*n$ So we can control accuracy vs space usage by increasing or decreasing n .