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# Making Session Stores More Intelligent

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# What is a session store?

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# A session store is...

- An chunk of data that is connected to one "user" of a service
  - "user" can be a simple visitor
  - or proper user with an account
- Often persisted between client and server by a token in a cookie\*
  - Cookie is given by server, stored by browser
  - Client sends that cookie back to the server on subsequent requests
  - Server associates that token with data
- Often the most frequently used data by that user
  - Data that is specific to the user
  - Data that is required for rendering or common use
- Often ephemeral and duplicated



# **Session Storage Uses Cases**

#### Traditional

- Username
- Preferences
- Name
- "Stateful" data

#### Intelligent

- Traditional +
- Notifications
- Past behaviour
  - content surfacing
  - analytical information
  - personalization



# In a simple world

 $\overbrace{}$ 

Internet

Server



# **Good problems**



Internet

Traffic Grows...

Server





## **Good solution**



Internet

Server

Session storage on the server



# More good problems



Internet

Struggling

Server

Session storage on the server



## **Problematic Solutions**



Internet

Server

Session storage on the server



# **Multiple Servers + On-server Sessions?**



Robin

Server



# **Multiple Servers + On-server Sessions?**



Robin

Server



## **Better solution**





# What is Redis?

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### Who We Are

![](_page_13_Picture_1.jpeg)

Open source. The leading **in-memory database platform**, supporting any high performance operational, analytics or hybrid use case.

![](_page_13_Picture_3.jpeg)

The open source home and commercial provider of **Redis Enterprise** technology, platform, products & services.

![](_page_13_Picture_5.jpeg)

# **Redis Top Differentiators**

![](_page_14_Figure_1.jpeg)

**redis**labs

**Bit field** 

Hashes

Lists

**Geospatial Indexes** 

Hyperloglog

**Streams** 

![](_page_14_Picture_2.jpeg)

![](_page_14_Figure_3.jpeg)

![](_page_14_Picture_4.jpeg)

Couchbase

cassandra

DATASTAX

**4EROSPIKE** 

**Performance:** The Most Powerful Database

Highest Throughput at Lowest Latency in High Volume of Writes Scenario

Least Servers Needed to Deliver 1 Million Writes/Sec

350

![](_page_15_Figure_3.jpeg)

![](_page_15_Figure_4.jpeg)

Benchmarks published in the Google blog

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Benchmarks performed by Avalon Consulting Group

16 

1

![](_page_16_Figure_0.jpeg)

![](_page_16_Picture_2.jpeg)

![](_page_17_Picture_0.jpeg)

- Add-ons that use a Redis API to seamlessly support additional use cases and data structures.
- Enjoy Redis' simplicity, super high performance, infinite scalability and high availability.

![](_page_17_Picture_3.jpeg)

- Any C/C++/Go program can become a Module and run on Redis.
- Leverage existing data structures or introduce new ones.
- Can be used by anyone; Redis Enterprise Modules are tested and certified by Redis Labs.
- Turn Redis into a Multi-Model database

![](_page_17_Picture_8.jpeg)

## **Redis Labs Products**

![](_page_18_Figure_1.jpeg)

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# Concepts

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# **Concept:** Bloom Filters (presence)

{x, y, z } 0 1 0 1 1 1 0 0 0 0 0 1 0 1 0 1 0 W

- Probabilistic data structure
- Hash -> sample bits -> set bits
- Properties:
  - False negatives not possible
  - False positives possible, but controllable
  - Bits per item stored
  - Add or check if exists
  - Like the Tardis, it's bigger on the inside than outside
- Availability:
  - Redis Module
  - On top of bitfields

![](_page_20_Picture_13.jpeg)

# Concept: HyperLogLog (cardinality) Probabilistic data structure Hash -> count runs -> store runs Properties:

- Estimates unique items
- Bits per item stored 2<sup>64</sup> unique items in 12kb / error rate 0.81%
- Add, count or merge!
- Like the Tardis, it's bigger on the inside than outside
- Availability:
  - All versions of Redis

![](_page_21_Figure_7.jpeg)

![](_page_21_Picture_8.jpeg)

# **Concept:** Bit counting (time series)

![](_page_22_Figure_1.jpeg)

the n-minute since starting point

- represent activity • Properties:
  - Size relative to length of time (byte round)
  - Count totals or ranges
  - BITOP (AND/XOR/OR/NOT)
- Availability:

• It's just bits!

- All versions of Redis

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# **Group Notifications**

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# **Traditional Group Notification Pattern**

#### Process

- Group of users get notification "Sale on sweaters"
- Insert into central table of notifications
- Insert row in table with each user of group with notification and seen flag
- Each time it is needed, query notifications table where seen flag is false.

# **Traditional Group Notification Pattern**

#### Challenges

- Adding/removing means touching a row for each user in group.
  - Fine for groups of 10 users, what about 1 million?
  - Also multi-step
- Storage is proportional to size of group and notifications
- Constant DB hits, not easily cacheable
- Setting "read" is DB write

# **Modern & Intelligent Group Notification Pattern**

#### Process

- Add notification to single group based structure or table (easily cacheable)
- First *n* notifications are read by all users in group.
- The notifications are checked to see if they are in a session-based Bloom filter or not.
- Mark read by adding to Bloom filter in session store.

![](_page_26_Picture_6.jpeg)

# Modern & Intelligent Group Notification Pattern

#### Advantages

- Adding a notification only writes to a single table, single row.
- Model fits use unread assumed.
- Fast. Checking for read / writing read is unrelated to number of items in the filter. Consistent.
- ~5-*bits* per item, but Bloom filter doesn't always grow.
- Gentle scaling

![](_page_27_Picture_7.jpeg)

## Visual

![](_page_28_Figure_1.jpeg)

![](_page_28_Picture_2.jpeg)

# Fresh Content

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# **Traditional Content Surfacing Pattern (Basic)**

#### Process

- Hand pick and rotate a small number of content/items
- Stored in DB table
- Served out dumbly to users

#### Challenges

- May serve content multiple times
- Freshness is linked to a manual curatorial process

![](_page_30_Picture_8.jpeg)

# **Traditional Content Surfacing Pattern (Advanced)**

#### Process

- Batch process builds content list to surface
   Not Real-time for each user
- List is stored in DB Table
- Served out to user
- Rotated on a schedule

#### Challenges

- May serve content multiple times
- Un-cacheable DB content
- Hard to scale

![](_page_31_Picture_11.jpeg)

# Modern & Intelligent Content Surfacing Pattern

#### Process

- Middleware adds each content read to a Bloom filter stored in the session
- Featured content list is built, can be extensive.
- Featured items are checked vs Bloom filter on-the-fly

#### Advantages

- No DB hits for user
- Featured content is cacheable
- Will not to show content multiple times if read
- Tiny storage requirements even at scale
- Freshness can be achieved with zero/low human input
- Real-time recording of activity immediate impact on fresh content

![](_page_32_Picture_12.jpeg)

![](_page_33_Figure_0.jpeg)

![](_page_33_Picture_1.jpeg)

# Activity Pattern Monitoring & Personalization

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# **Activity Pattern Monitoring & Personalization?**

- Monitor the usage behaviour
  - Content viewed
  - Activity over time
  - Combinations of content history and activity
- Personalize the content based on the behaviour
- Seen as difficult to accomplish
  - Analytics data
    - Stored in another service
    - Anonymized
  - Complicated graph or ML based solutions
    - Inferences
    - Black boxes

![](_page_35_Picture_13.jpeg)

# **Activity Pattern Monitoring & Personalization**

- Record site activity with bit counting
- Unique page views in HyperLogLog
- Leverage the page visit Bloom filter
- Simpler counter for pages consumed
- Create criteria based on session stored analytics
  - New to a page? Bloom filter
  - New to the site? Unique Page view = 1 (HLL) && Previously Visited = false (Bloom)
  - Inactive user? Sum the bit count over the last five records, if = 0 then inactive
  - Been to a cluster of pages (infer interest)? Check cluster of pages vs Bloom filter combo!

![](_page_36_Picture_10.jpeg)

# **Activity Pattern Monitoring & Personalization**

- Why is this suddenly possible?
  - Probabilistic data structures are small/fast
  - Bit counting is small/fast
  - Decoupled from operational database
- What about privacy?
  - Legitimate concern
  - Non-reversible probabilistic structures
  - Siloed from rest of database

![](_page_37_Picture_9.jpeg)

# Questions?

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# Thank you!

Demo source code: https://github.com/stockholmux/qcon-redis-session-store-demo

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