Managing Time Series with MongoDB

Paolo Gallo paolo.gallo@uniud.it

May 18, 2020



Outline

Introduction

2 Schema design(s)

③ Impact on requirements and performances



Case Study



The United States is beginning to make its transition to self-driving cars.

For this reason United States Department of Transportation is setting up central service to monitor traffic conditions nationwide

Sensors over the interstate system monitor traffic conditions like: car speeds, pavement and weather conditions, etc.



Interstate Highway System





Traffic Sensors



- 16.000 sensors
- Measure
 - Speed
 - Travel Time
 - Weather, pavement, and traffic condition
- Support desktop, mobile, and car navigation systems



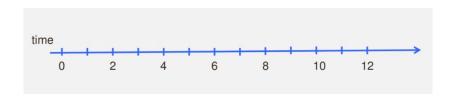
Other Requirements

- Need to keep 3 years of history
- Three data centres
 - New York
 - Chicago
 - Los Angeles
- Need to support 5 millions simultaneous users
 - Peak volume (rush hour)
 - Every minute, each request the 10 minute average speed for 50 sensors



Time Series

A (discrete) *time series* is a sequence of data, measured typically at successive points in time, spaced at uniform time intervals.





Time Series Everywhere











Schema Design Consideration

- Horizontal Scalability
- Store event data
- Support analytical queries
- Find best compromise between:
 - Memory utilisation
 - Write performance
 - Read/Analytical query performance
 - Accomplish with a realistic amount of hardware



Designing for reading, writing, ...

There are different choices in order to store upcoming information with a document DB:

- Document per event
- Document per minute (average)
- Document per minute (second)
- Document per hour
- ..



Document per event

```
segID: "I80_mile34",
speed: 63,
ts: ISODate("2016-11-10T22:56:00.00-0500")
}
```

- Relational centric approach
- Insert driven workload
- Overall performances and resources consumption are same as with a relational DB



Document per minute (average)

```
{
   segID: "I80_mile34",
   speed_num: 18,
   speed_sum: 1256,
   ts: ISODate("2016-11-10T22:56:00.00-0500")
}
```

- Pre-aggregate to compute average per minute easily
- Update driven workload
- Resolution at the minute level



Document per minute (second)

```
segID: "I80_mile34",
speed: {0:63, 1:23, 2:45, ..., 59:65},
ts: ISODate("2016-11-10T22:56:00.00-0500")
}
```

- Store per-second data at minute level
- Update driven workload
- Pre allocate structure to avoid document move



{

Document per hour

```
segID: "I80_mile34",
speed: {0:63, 1:23, 2:45, ..., 3599:65},
ts: ISODate("2016-11-10T22:00:00.00-0500")
}
```

- Store per-second data at hourly level
- Update driven workload
- Pre allocate structure to avoid document move
- Updating last second requires 3599 steps



Document per hour (by second)

- Store per-second data at hourly level with nesting
- Update driven workload
- Pre allocate structure to avoid document move
- Updating last second requires 59 + 59 steps



Characterizing write differences

- Example: data generated every second
- For one minute:
 - Document Per Event \longrightarrow 60 Writes
 - *Document Per Minute* →1 Write, 59 updates
- Transition from write driven to update driven
 - individual writes are smaller
 - performance and concurrency benefits



Characterizing read differences

- Example: data generated every second
- Reading data for a single hour requires:
 - *Document Per Event* → 3600 reads
 - *Document Per Minute* \longrightarrow 60 reads
- Read performance is greatly improved:
 - optimal with tuned block and read ahead
 - fewer disks seeks



Characterizing memory differences

- _id index for 1 billion events:
 - Document Per Event → 32 Gb
 - Document Per Minute → 0.5 Gb
- _id index plus segId and ts index:
 - Document Per Event → 100Gb
 - Document Per Minute→ 2 Gb
- memory requirements significantly reduced:
 - fewer shards
 - lower capacity servers



Quick analysis

- Writes:
 - 16.000 sensors, 1 update per minute
 - 16.000 / 60 = 267 updates per second

- Reads:
 - 5 millions simultaneous users
 - Each request data for 50 sensors per minute



Reads: impacts of alternative schemas

Query: Find the average speed over the last ten minutes

10 minute average query			
Schema	1 sensor	50 sensors	
1 doc per event	10	500	
1 doc per 10 min	1.9	95	
1 doc per hour	1.3	65	

10 minute average query with 5M users			
Schema	ops/sec		
1 doc per event	42M		
1 doc per 10 min	8M		
1 doc per hour	5.4M		



Writes: impacts of alternative schemas

Schema	Inserts	Updates
doc/event	60	0
doc/10 min	6	54
doc/hour	1	59
	0 Sensors – 1	
1600 Schema	0 Sensors – 1 Inserts	
		Day Updates 0

.38M

doc/hour

22.7M



Memory: impacts of alternative schemas

1	1 Sensor - 1 Hour				
Schema	# of Documents	Index Size (bytes)			
doc/event	60	4200			
doc/10 min	6	420			
doc/hour	1	70			

16000 Sensors - 1 Day				
Schema	# of Documents	Index Size		
doc/event	23M	1.3 GB		
doc/10 min	2.3M	131 MB		
doc/hour	.38M	1.4 MB		



Document structure



Query: two indexes required



A Smart Sample Index



Sample index: range query

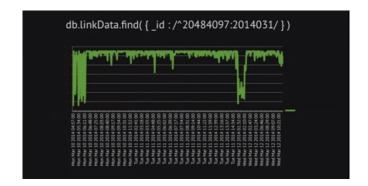


Preallocate data

```
{ _id: "900006:140312",
                                   Pre-allocated,
   { speed: NaN, time: NaN }, •
                                   60 element array of
                                   per-minute data
 conditions: {
   status: "Snow / Ice Conditions".
   pavement: "Icy Spots",
   weather: "Light Snow"
```



Chart

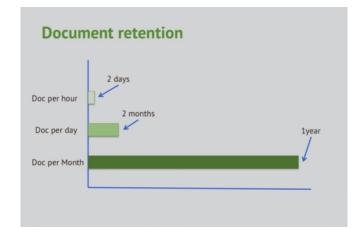




Rollup



Document Retention





What is the average speed for a segment?

{ id:"20484097", ave:"47.3665646546"}