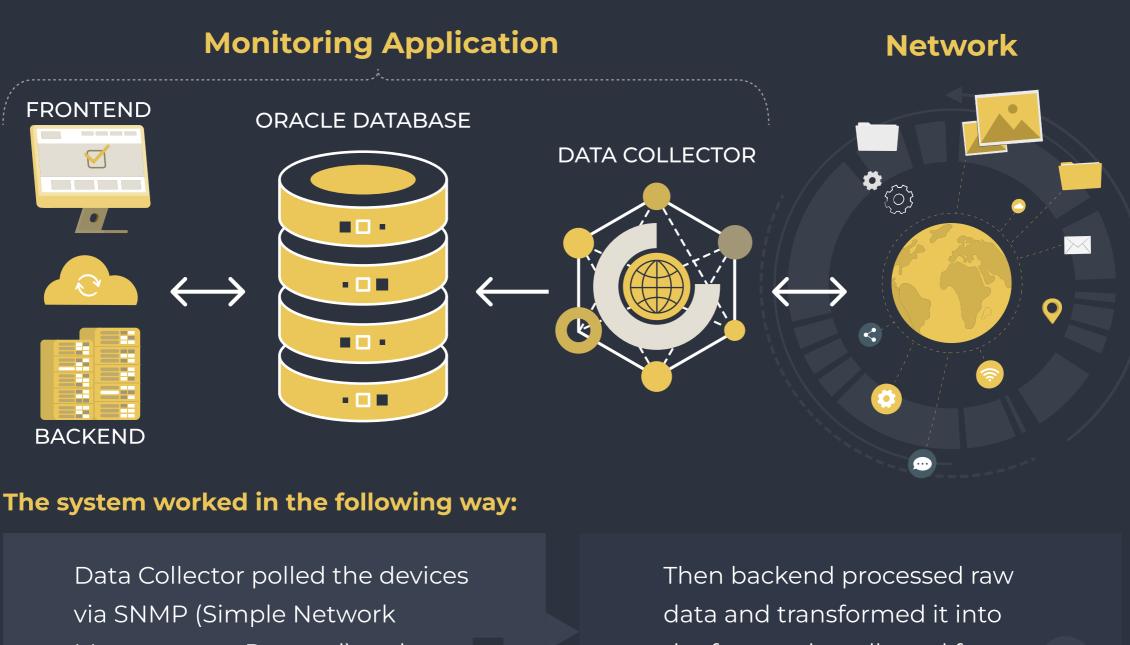


Best Practices Of Big Data Implementation

Part 1.

How we struggled with big data implementation.

The whole story about big data implementation started with an ongoing project. Our team was working on a project for monitoring a range of devices: switches, routers, computers and more. The main goal of this system was to provide businesses with advanced real-time performance reporting by collecting and analyzing KPI across IT infrastructure. The project was based on classic old-school architecture, with Oracle Database used as a data storage.



Management Protocol) and stored raw data in Oracle DB. the format that allowed for better big data analytics.

As a result, the customer was able to discover devices in the network using IP range, view statistic for selected devices for any period, build reports, charts, set up alarms and manage multi-vendor networks, systems and databases.

Challenge

Although the solution worked well for most users, some major customers had several millions of devices. The system was launched in 1998, so it had to store the detailed statistic for at least 10 years to compare results and make predictions.

Each device produced around **10 KB** of data. Some new devices produced much more information, around 100-300 KB, but let's take the average of 10 KB for calculation.

Each device was polled every 1 minute, which made up 10 KB x 60 x 24 =14 MB per day, or 14 MB *365 = **5 GB** per year.

Since this solution had no capabilities for horizontal scaling, it was impossible to work with such amount of data. To provide customer with the detailed statistics, we introduced some limitations on the number of devices and days. However, this solution was inconvenient for the customer.

Assuming the customer had 10 million of devices, his devices produced around 48 PB of data per year.

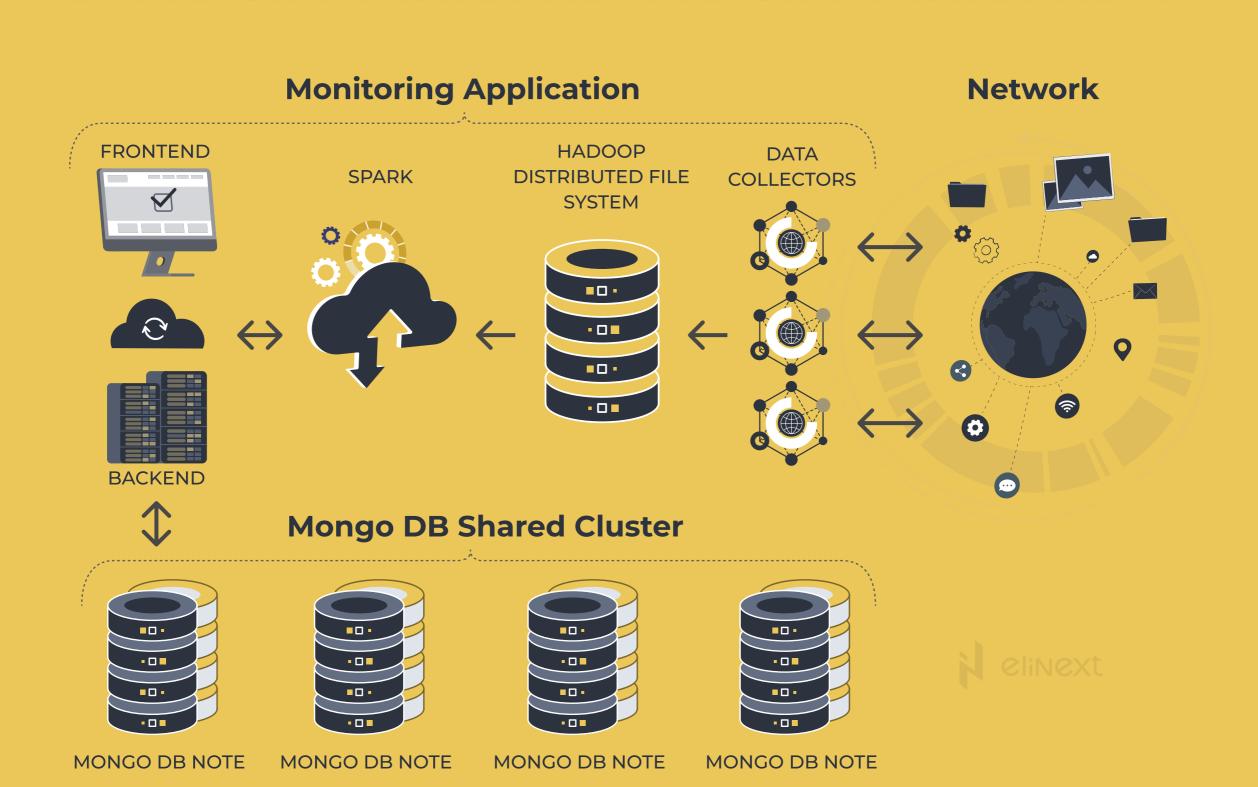
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To improve our current implementation of data storing and processing, we experimented a bit with big data technologies:

First, we decided to use MongoDB shared cluster instead of single Oracle database. This allowed client to store much more data, but the performance was almost the same, because data was collected by single data collector and processed by single backend. That's why we also decided to implement Hadoop Distributed File System (HFDS) for raw data storage and Hadoop's MapReduce for data processing to store it in MongoDB.

After tool research on the market, Hadoop MapReduce was replaced with Apache Spark, as it was easier to use and allowed running programs faster.

As a result, we received a system with multiple data collectors installed on several segments of the network, HDFS, Spark, and MongoDB. We emulated the monitoring of 4 million of devices and haven't detected any problem neither in the performance of the system nor in data storage capabilities. As a comparison, the initial implementation without Big Data was able to monitor up to 100,000 devices only, so we can talk about significant system improvement.



According to our implementation plan, the next step was to bring Kafka in our project and add a search engine (we were thinking about either Elasticsearch or Solr). What is more, since the project scaled significantly, we were considering to replace MongoDB with Cassandra for its column-oriented database.

However, our big data experiment was stopped, as the client decided to put the project on the shelf and engage Elinext dedicated team that worked on this project in building the product of a similar kind, but with more functionality.



Top questions you should ask before starting big

A lot of questions may arise when coming up with a big data strategy. The answers depend on the amount and type of data and the goals you want to achieve with its exploration.



and technologies stack that we suggest.

running applications on large cluster built of commodity hardware. Indeed, Hadoop is the standard for big data processing. The product we decided to exploit in our project of is Hadoop

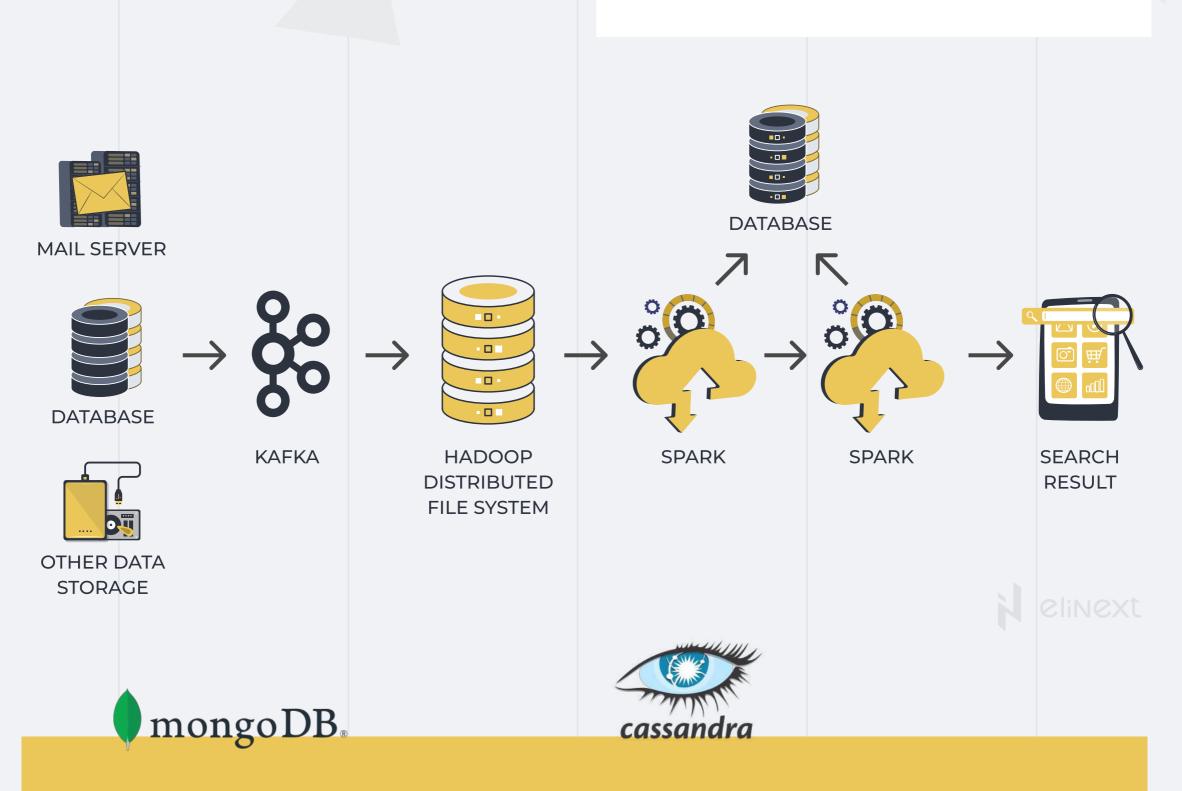


Apache Kafka is a distributed system used for building real-time data pipelines and streaming apps used for processing logs, metrics, and collections. We have chosen this technology for its horizontal scalability, fault-tolerance, and high speed. Within our project, Kafka was used for data retrieval from the mail servers. The data received was consequently put into Hadoop distributed file system for storage.

distributed file system (HDFS) that stores data on the compute nodes, providing very high aggregate bandwidth across the cluster.



Apache Spark is a fast and general engine for large-scale data processing which is all-compatible with Hadoop. We preferred Spark to MapReduce for our project's purposes because of higher speed of big data processing and ease of deployment. Moreover, Spark allows for iterative machine learning algorithms with each iteration contributing to the overall results.



Our choice of MongoDB was determined by its well-supported integration with Hadoop and its aggregation framework. It helped us store data in convenient form for processing.

Cassandra caught our interest as a possible replacement of MongoDB within our project because of its support for dynamic columns and distributed counters. We also were impressed by its scalability and reasonable costs of ownership.





Both search platforms are built on top of Apache Java library Lucene, so many of their functionalities and their search capabilities are relatively similar. However, **Elasticsearch** is considered to be easier to use and better for analytical queries and distributed indexing, while Solr is well-oriented in everything connected with text search and is popular for being consistently documented.



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