## From Data Warehouse to Data Lake

—Telecom Big Data Architecture evolution

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#### The TMN/FCAPS Model

Current mobile network architecture



F C A P S
Fault Management
Configuration Management
Accounting Management
Performance Management
Security Management





## Enable technology Technique Insight

Massive network data

Most structured data

**Variety** 

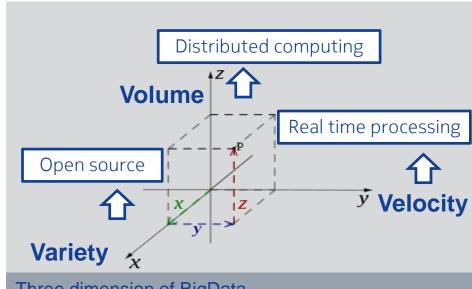


Every second in real-time



**Velocity** 

Real-time measurement, analytics, decision and action



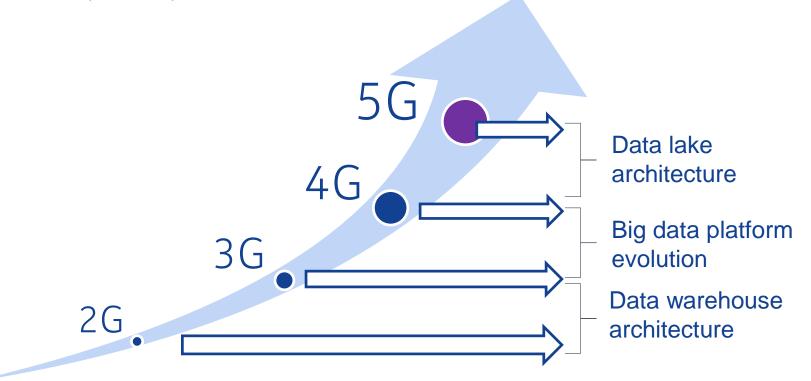
#### Three dimension of BigData

- The streaming paradigm to deal with the velocity of the data
- Distributed computing to deal with the volume of the data
- open source for the variety



## Stepwise approach

Backwards compatibility, evolution with reasonable steps

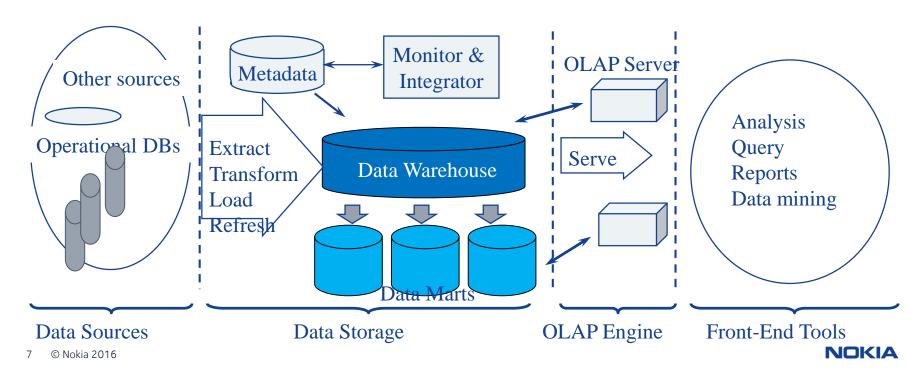




#### What is Data Warehouse?

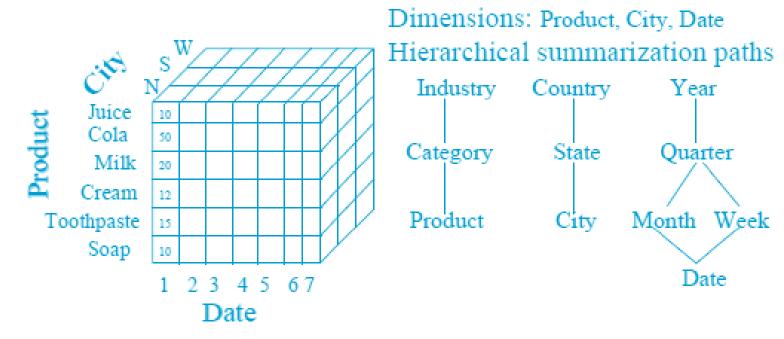
Typical data warehousing architecture : A Multi-Tiered Architecture

"A data warehouse is a subject-oriented, integrated, time-variant, and nonvolatile collection of data in support of management's decision-making process." —W. H. Inmon



#### Multidimensional Data Data cube

A dimension is a schema object that defines hierarchical relationships between columns or column sets.



# **Typical OLAP Operations**Typical OLAP Operations

Computer home entertainment item (types)

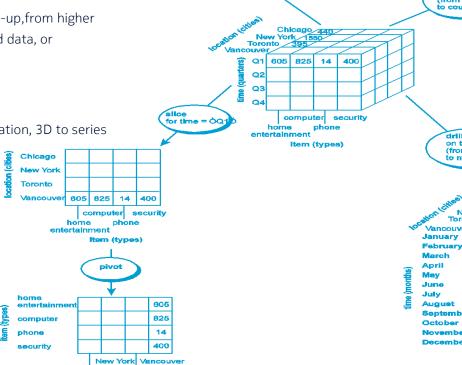
Roll up (drill-up): summarize data by climbing up hierarchy or by dimension reduction

Drill down (roll down): reverse of roll-up, from higher level summary to lower level summary or detailed data, or introducing new dimensions

Slice and dice: project and select

Pivot (rotate): reorient the cube, visualization, 3D to series

of 2D planes



Chicago Toronto

location (cities)

(location = ÒTorontoÓ or ÒVancouverÓ)

and (time = ÒQ1Ó or ÒQ2Ó) and (item = Òhome entertainmentÓ or Òcomput computer se

Item (types)

150 100

150

entertainment

roll-up on location (from cities to countries

> drill-dowr on time

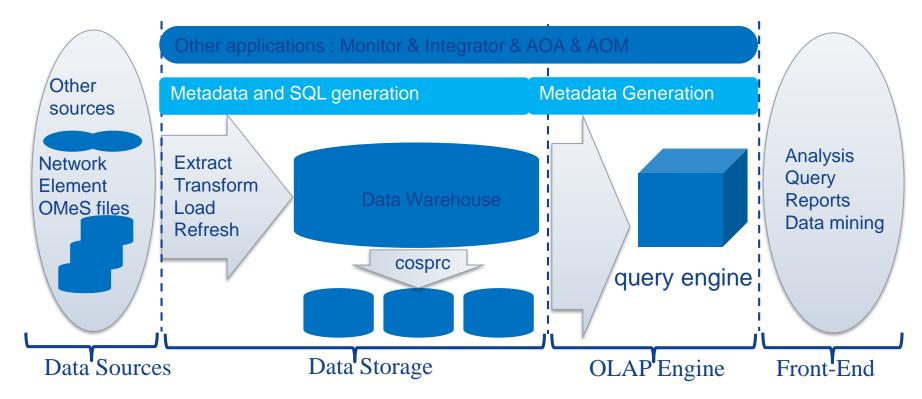
to months)

(from quarters

Chicago

entertainment

## Reference architecture Meta Data Driven Architecture Design



### Pain Point of current data warehouse architecture From Oracle RAC to Big Data stack

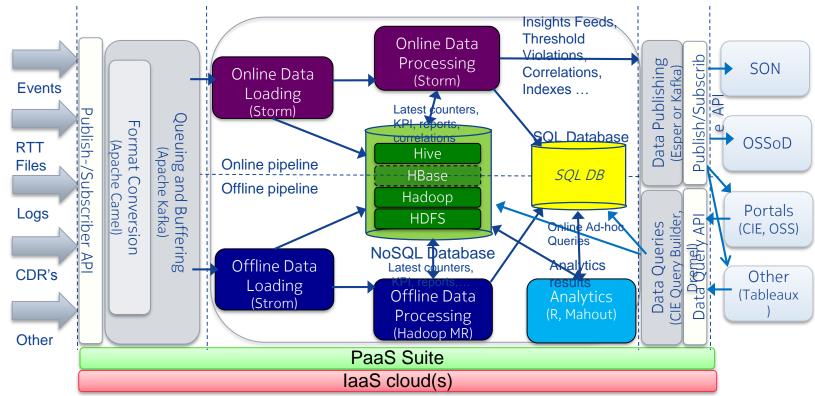
- Data warehouse is currently based on Oracle relational database
- DB scalability is handled via Oracle RAC solution, whenever vertical scalability is no longer possible (due to HW limitations)
- Oracle RAC is not linear scalable and introduces several challenges wrt existing deployments (e.g., administration overhead, Customer CoDe, ...)
- Storage solution for proper tuned I/O sub-system is quite expensive and storing raw/hourly data for extended periods is also cost prohibitive





#### Big Data Platform Architecture

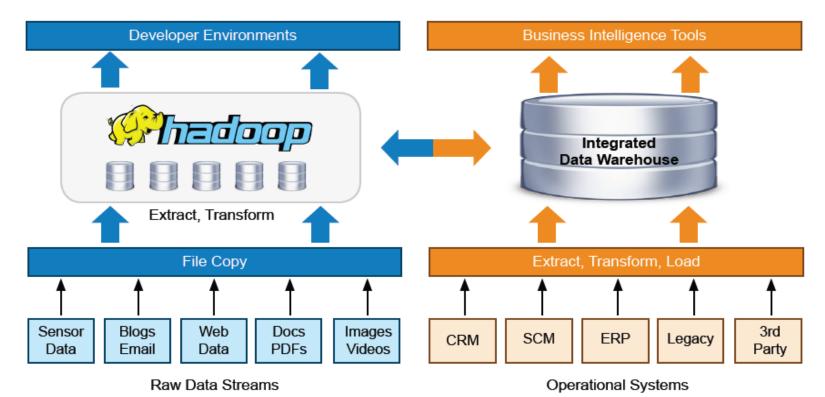
Online & Offline





#### **Architecture Compare**

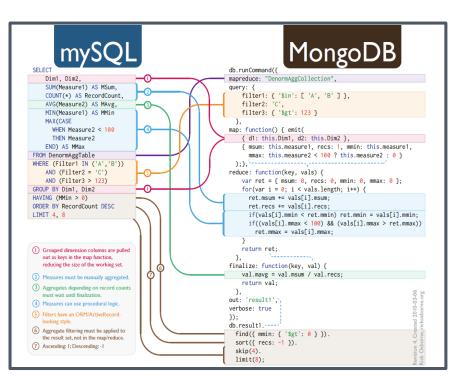
#### Same architecture for Data warehouse





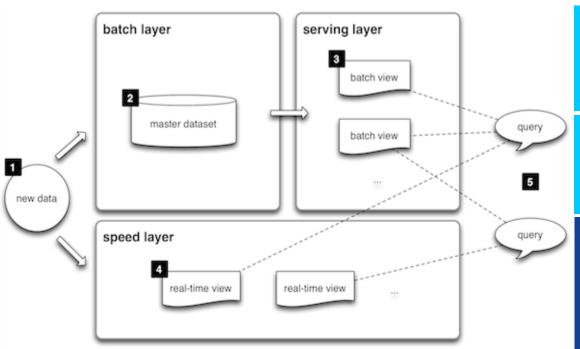
#### **Architecture Compare**

## Group-By-Aggregation and Map-Reduce



- The Map-Reduce programming model provides a good abstraction of group-by-aggregation operations over a cluster of machines.
- The programmer provides a map function that performs grouping and a reduce function that performs aggregation.
- The underlying run-time system achieves parallelism by partitioning the data and processing different partitions concurrently using multiple machines.

## Reference architecture Lambda architecture is popular



## speed layer

- (i) compensates for the high latency of updates to the serving layer
- (ii) deals with recent data only

#### serving layer

- (i) indexes the batch views
- (ii) Can be queried in low-latency, ad-hoc way

## batch layer

- (i) managing the master dataset (an immutable, append-only set of raw data),
- (ii) pre-compute the batch views





Flat operations architecture

Data Lake Motivation

Some pain point for the hierarchy architecture:

Slow response times;

Poor scaling;

Slow development;

Overlapping development.

Flat operations architecture



Data Lake



# Distributed analytics Data Lake Motivation

Distributed analytics

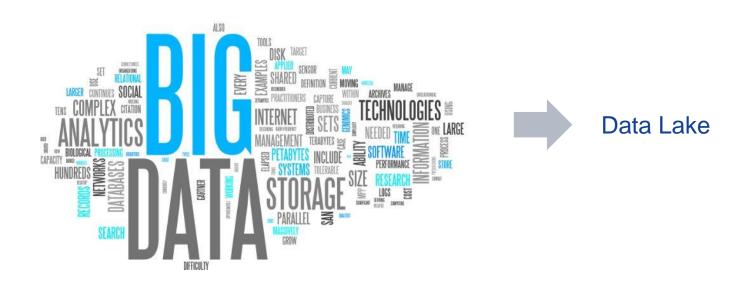




#### 5G Impact

Data Lake Motivation

5G → BigData (Semi-Structured and Unstructured) → Modern Data Architecture for Enterprise → Data Lake Storage Architecture → Data Lake

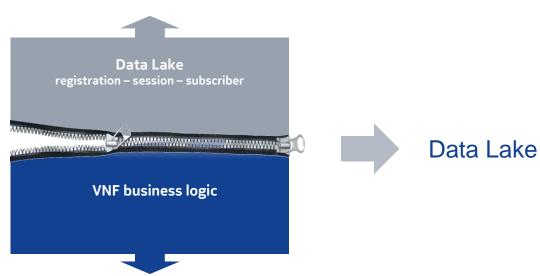




## Cloud transformation Data Lake Motivation

Cloud → Network Function Cloudification → Network Function Virtualization → stateless VNF→ Distributed Sharing Storage → Data Lake

Non-breakable, Open, Ultra-fast, Service-logic agnostic, Multi-tenant capable





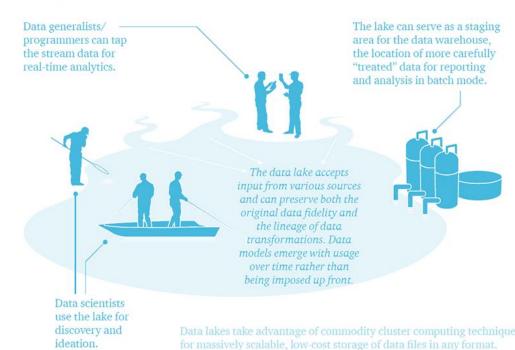
## What is Data Lake? PWC definition

"Size and low cost"

"Fidelity: Hadoop data lakes preserve data in its original form"

"Ease of accessibility: Accessibility is easy in the data lake"

"Late binding: Hadoop lends itself to flexible, task-oriented structuring and does not require up-front data models" A repository for large quantities and varieties of data, both structured and unstructured.



#### What is Data Lake?

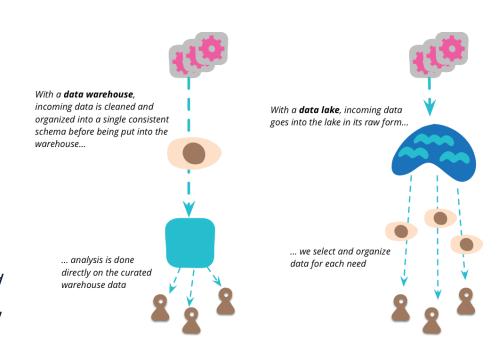
#### Martin fowler 's view

#### Data warehouse:

- It represents an abstracted picture of the business organized by subject area.
- It is highly transformed and structured.
- Data is not loaded to the data warehouse until the use for it has been defined.
- It generally follows a methodology such as those defined by <u>Ralph Kimball</u> and <u>Bill Inmon</u>.

#### Data lake:

- All data is loaded from source systems. No data is turned away.
- Data is stored at the leaf level in an untransformed or nearly untransformed state.
- Data is transformed and schema is applied to fulfill the needs of analysis.



http://martinfowler.com/bliki/DataLake.html



# What is Data Lake? Comparing the Enterprise Data Warehouse and the Data Lake

Dimension	Enterprise Data Warehouse	Data Lake
Workload	Hundreds to thousands of concurrent users performing interactive analytics	Batch processing of data at scale.
Schema	Typically schema is defined before data is stored. <b>Schema on write</b> means required data is identified and modeled in advance.	Typically schema is defined after data is stored. <b>Schema on read</b> means data must be captured in code for each program accessing the data.
Data	Cleansed	Raw
Complexity	Complex joins	Complex processing
Scale	Can scale to large data volumes at moderate cost	Can scale to extreme data volumes at low cost
SQL	ANSI SQL, ACID compliant	Flexible programming, evolving SQL

