

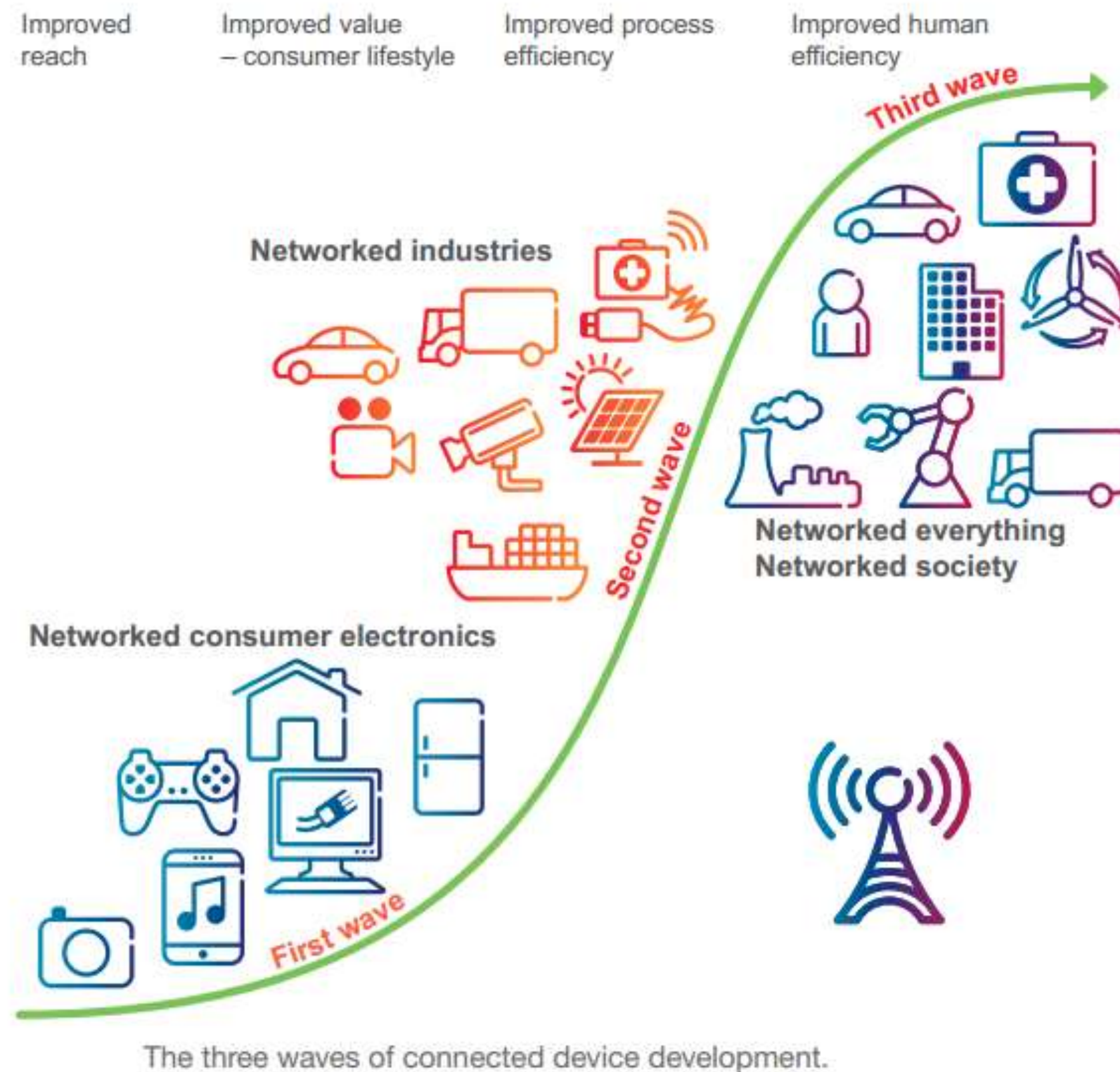
ENERO 31, 2023

FUNDAMENTOS DE BIG DATA

Qué es BigData

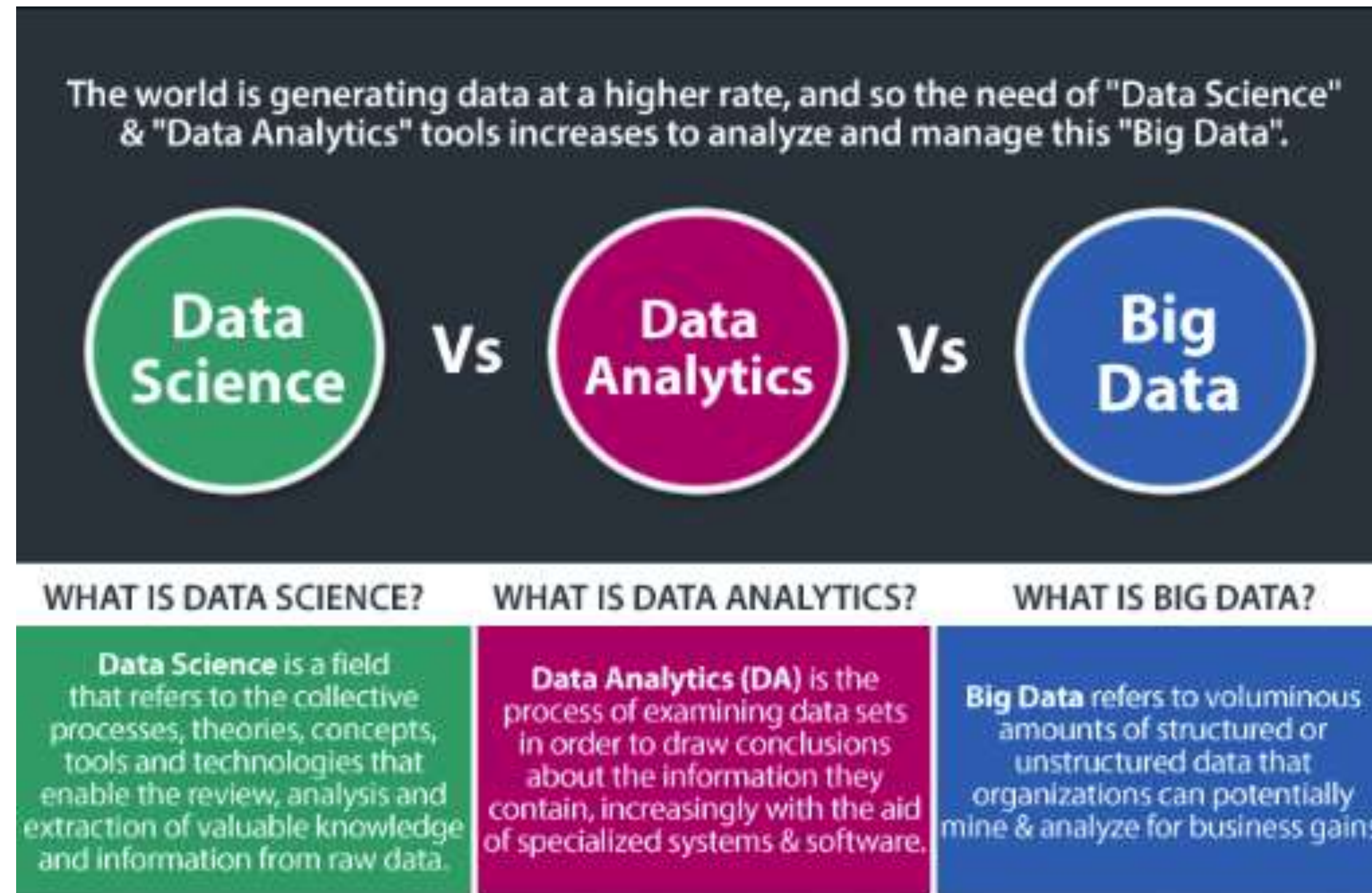
Nuevo enfoque para entender los datos, dar valor y tomar decisiones, a partir de la descripción de datos (estructurados, no estructurados o semi estructurados) analizados desde un punto de vista no relacional (costo, tiempo, recursos), utilizando herramientas de supercómputo nativo o en la nube.

La relación del Big Data con otras tecnologías

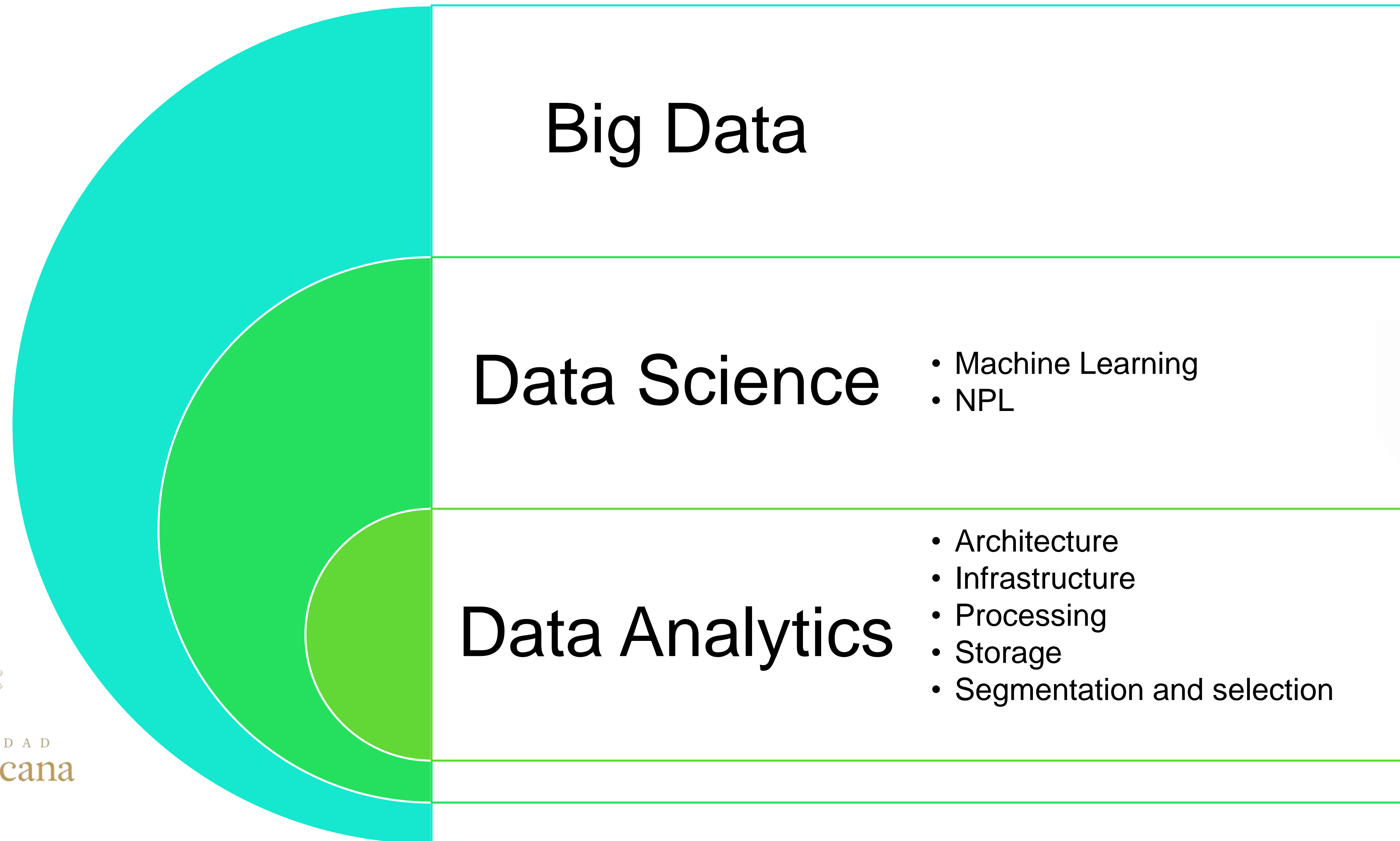


- **Gran masa**
- **Diversidad**
- **Almacenamiento**
- **Seguridad**
- **Tomar decisiones y responder**

Qué es BigData



Qué es BigData



Qué es BigData, 10 V's



Qué es Big Data



Google BigQuery



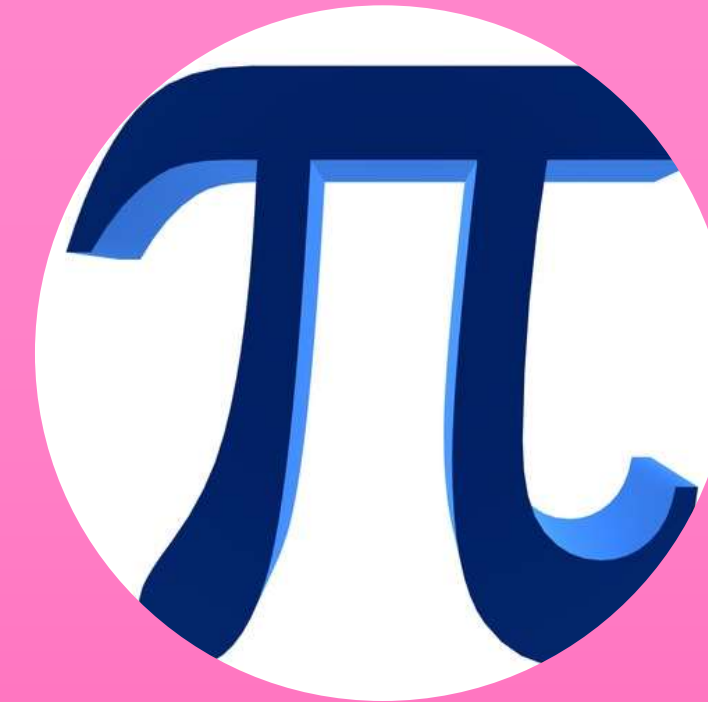
Qué es necesario



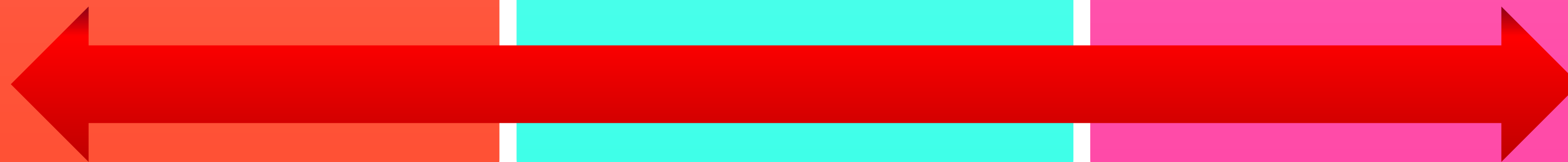
Infraestructura



Fuente(s) de
datos



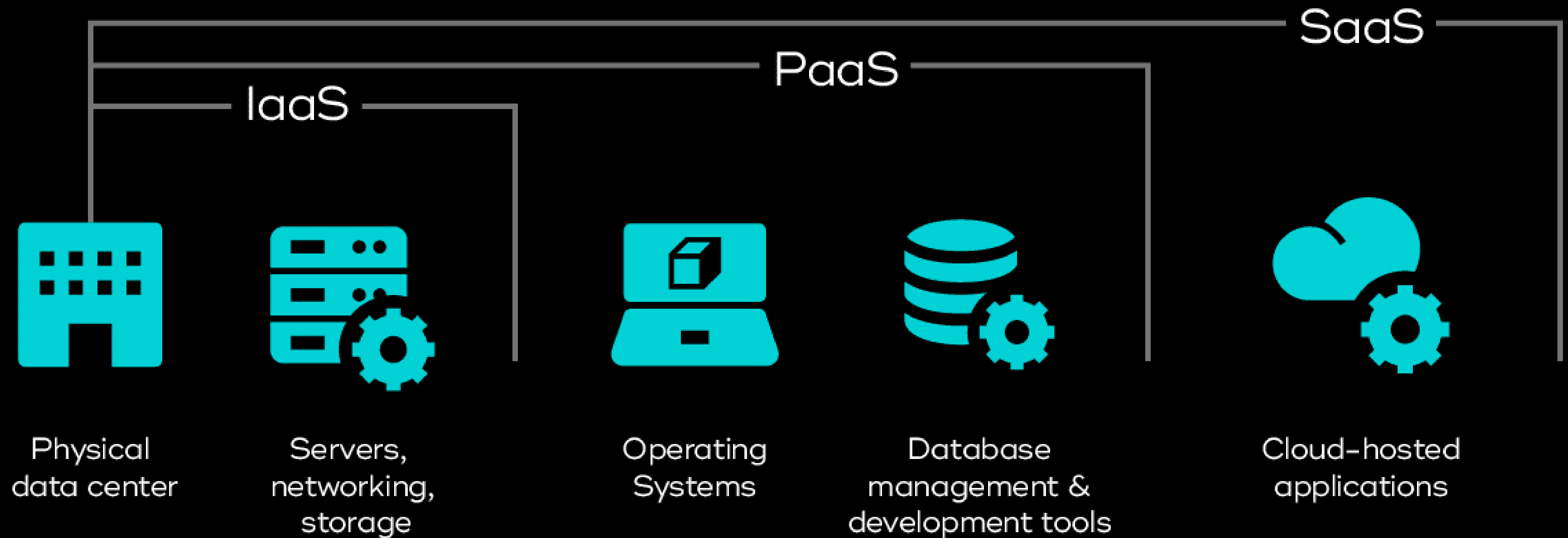
Modelo
matemático



Big Data, generalidades

- Los motores de BD tradicionales no pueden tratar altas cantidades de datos, ni reaccionar inmediatamente a ellos. Almacena **datos no relacionales y relacionales**.
- Esquemas de consulta:
 1. **Key-value**
 2. **Columnal**
 3. **Documental**
 4. **Gráfos**

Cloud Computing Overview



IaaS, Infrastructure as a Service
PaaS, Platform as a Service
SaaS, Software as a Service

Escenarios de cómputo en la nube

On-site	IaaS	PaaS	SaaS
Applications	Applications	Applications	Applications
Data	Data	Data	Data
Runtime	Runtime	Runtime	Runtime
Middleware	Middleware	Middleware	Middleware
O/S	O/S	O/S	O/S
Virtualization	Virtualization	Virtualization	Virtualization
Servers	Servers	Servers	Servers
Storage	Storage	Storage	Storage
Networking	Networking	Networking	Networking



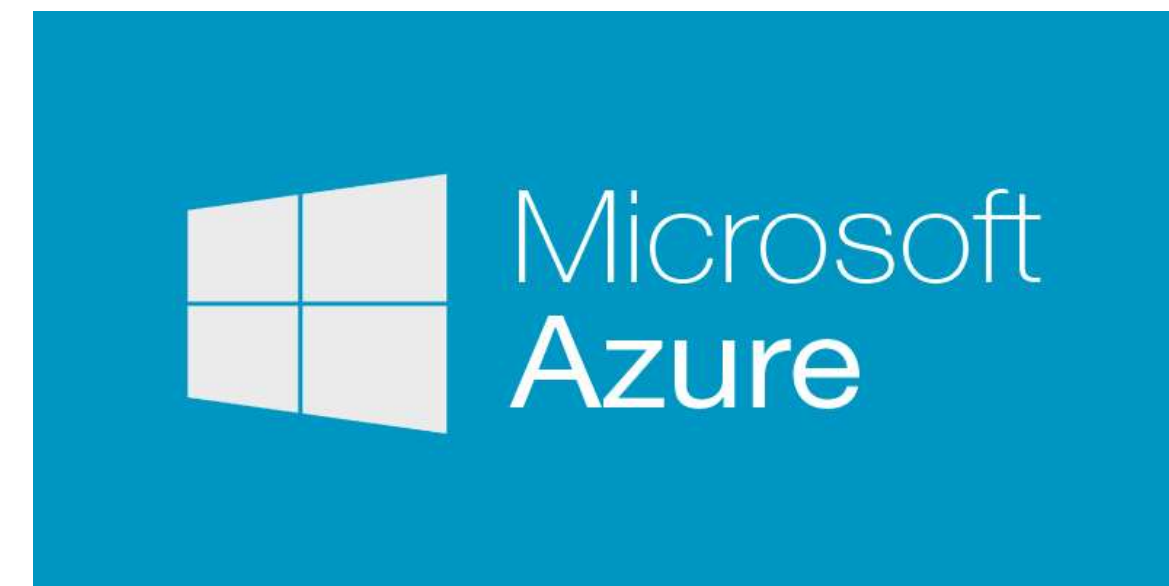
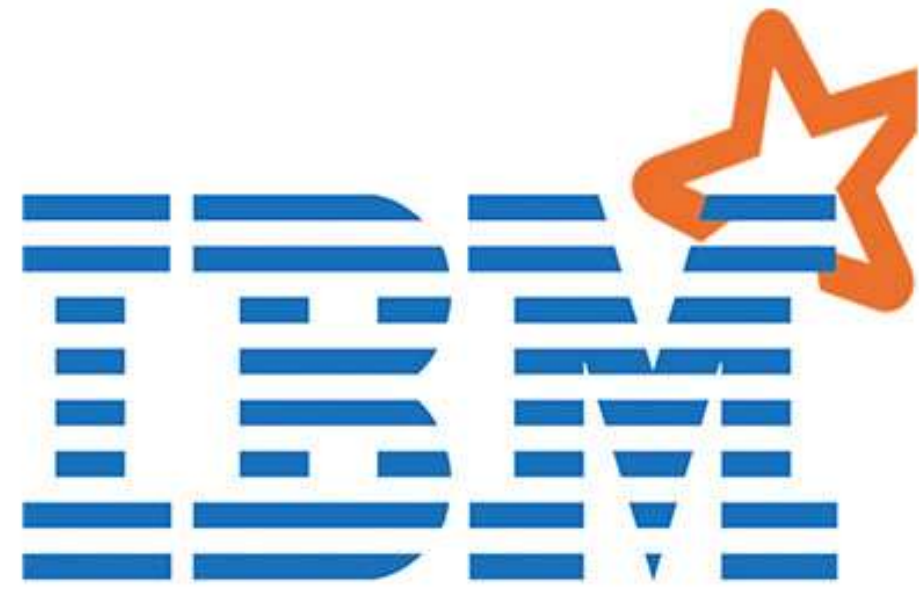
You manage



Service provider manages

Tecnologías líderes Big Data engine

ORACLE®
NOSQL DATABASE



Qué se puede hacer con BigData

La nube consumía software, la nube ahora consume datos e integrando software desde donde se puede obtener información inteligente:

Análisis retrospectivo

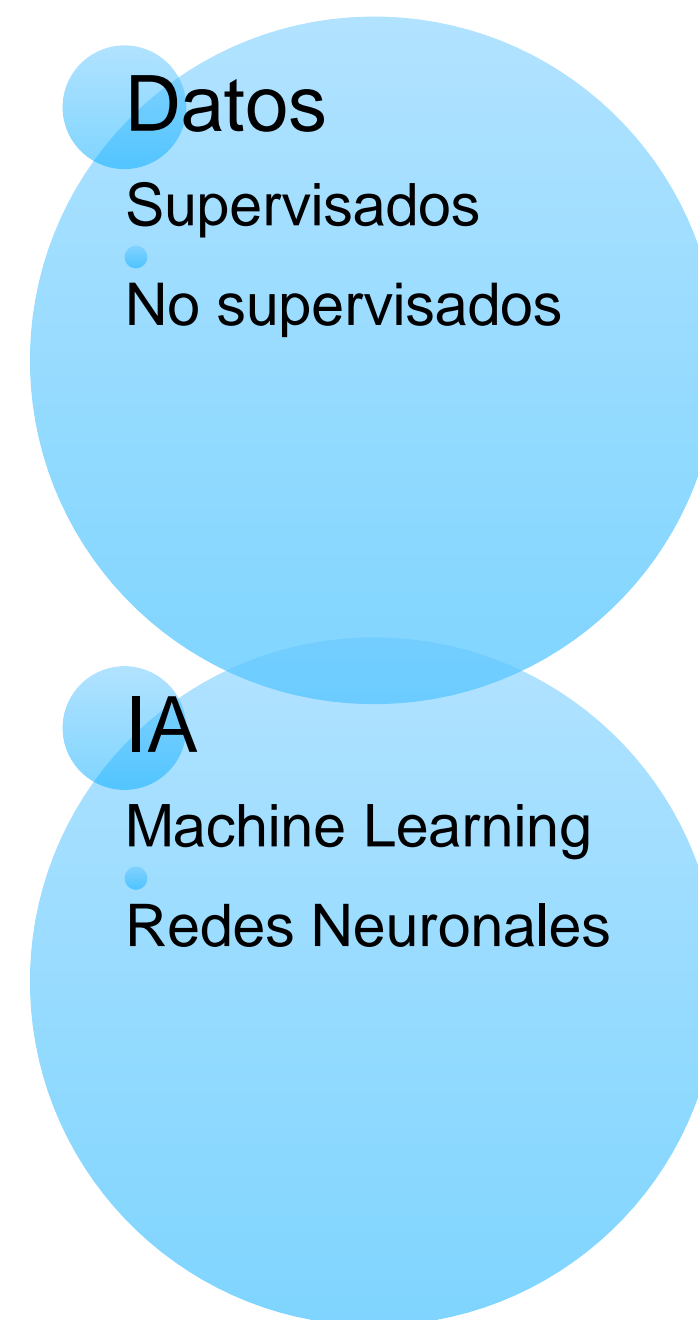
Análisis en tiempo real

Análisis predictivo

App Inteligentes SaaS



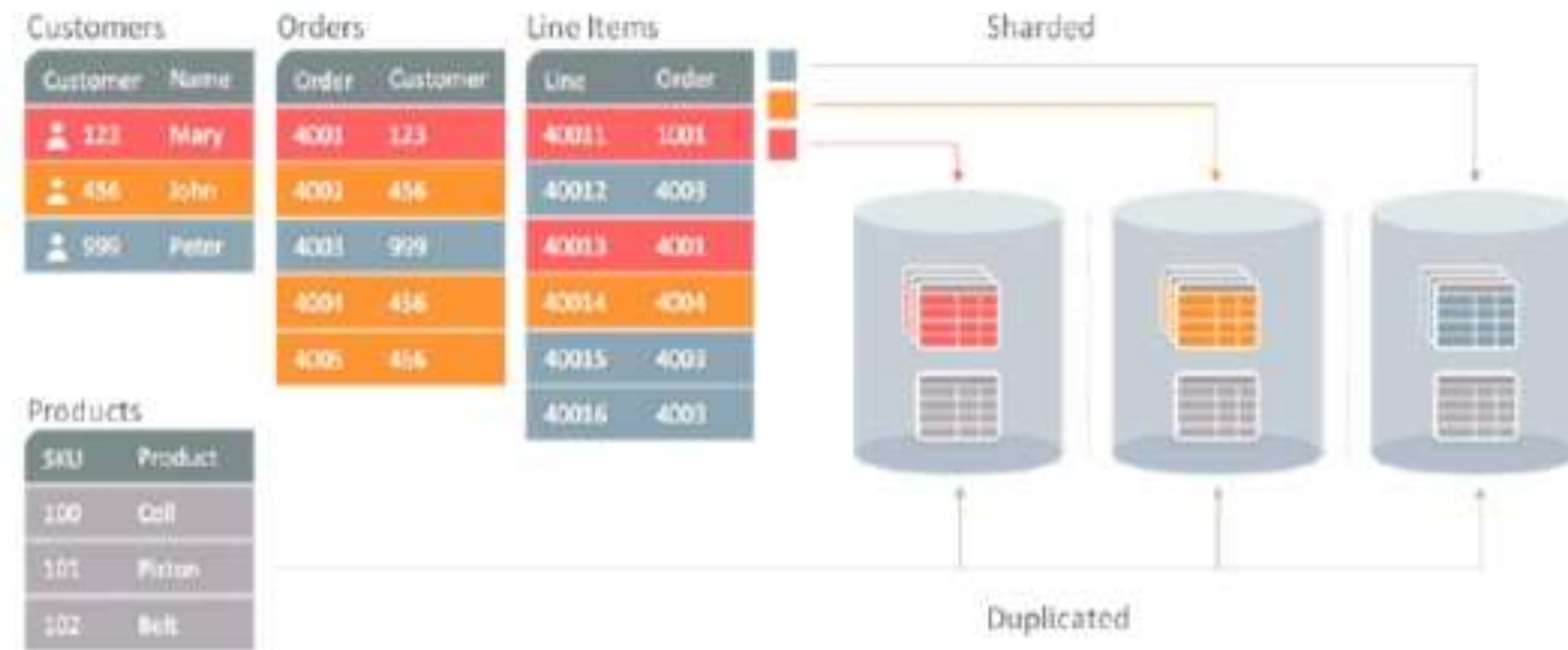
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Queue adicional y backup

- Sharding

ORACLE SHARDING USE CASE



Oracle Sharding is implemented based on the Oracle Database partitioning feature.

Oracle Sharding is "Distribute Partitioning".

Algo de historia

- Big Data fue precedido por Google (mapreduce), Amazon (Dynamo) y Software Open Source (Hadoop, MongoDB, Cassandra, RabbitMQ, etc)
- RDBMS tienen funciones específicas donde las relaciones entre los datos toman valor único

Propiedades deseadas de un BD

1. Robustes y tolerancia a fallas
2. Baja latencia y actualizaciones
3. Escalabilidad
4. Generalización
5. Extensabilidad
6. Queries ad-hoc
7. Mantenimiento mínimo
8. Debugabilidad
9. Problemas de Arq. Incremental (nuevas funciones)

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ESSENTIAL DIGITAL HEADLINES

OVERVIEW OF THE ADOPTION AND USE OF CONNECTED DEVICES AND SERVICES



TOTAL
POPULATION



7.91
BILLION

URBANISATION

57.0%

UNIQUE MOBILE
PHONE USERS



5.31
BILLION

vs. POPULATION

67.1%

INTERNET
USERS



4.95
BILLION

vs. POPULATION

62.5%

ACTIVE SOCIAL
MEDIA USERS



4.62
BILLION

vs. POPULATION

58.4%

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9

SOURCES: UNITED NATIONS; U.S. CENSUS BUREAU; GOVERNMENT BODIES; GSMA INTELLIGENCE; ITU; GWI; EUROSTAT; CNNIC; ABIJ; CIA WORLD FACTBOOK; COMPANY ADVERTISING RESOURCES AND EARNINGS REPORTS; OECD; TECHRASA; KEPIOS ANALYSIS. **ADVISORY:** SOCIAL MEDIA USERS MAY NOT REPRESENT UNIQUE INDIVIDUALS. **COMPARABILITY:** SOURCE AND BASE CHANGES.

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POPULATION ESSENTIALS

DEMOGRAPHICS AND OTHER KEY INDICATORS



TOTAL
POPULATION



7.91
BILLION



FEMALE
POPULATION



49.6%



MALE
POPULATION



50.4%



YEAR-ON-YEAR CHANGE
IN TOTAL POPULATION



+1.0%



MEDIAN AGE OF
THE POPULATION



31.4

URBAN
POPULATION



57.0%



POPULATION DENSITY
(PEOPLE PER KM²)



60.8



OVERALL LITERACY
(ADULTS AGED 15+)



86.7%



FEMALE LITERACY
(ADULTS AGED 15+)



83.3%



MALE LITERACY
(ADULTS AGED 15+)



90.1%



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DEVICE OWNERSHIP

PERCENTAGE OF INTERNET USERS AGED 16 TO 64 WHO OWN EACH KIND OF DEVICE



ANY KIND OF
MOBILE PHONE



GWl.

96.6%

YEAR-ON-YEAR CHANGE
-0.5% (-50 BPS)

SMART
PHONE



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96.2%

YEAR-ON-YEAR CHANGE
-0.4% (-40 BPS)

FEATURE
PHONE



GWl.

8.8%

YEAR-ON-YEAR CHANGE
-2.2% (-20 BPS)

LAPTOP OR
DESKTOP COMPUTER



63.1%

YEAR-ON-YEAR CHANGE
-2.0% (-130 BPS)

TABLET
DEVICE



34.8%

YEAR-ON-YEAR CHANGE
+1.5% (+50 BPS)

GAMES
CONSOLE



20.3%

YEAR-ON-YEAR CHANGE
-5.1% (-110 BPS)

SMART WATCH OR
SMART WRISTBAND



GWl.

27.4%

YEAR-ON-YEAR CHANGE
+17.6% (+410 BPS)

TV STREAMING
DEVICE



15.5%

YEAR-ON-YEAR CHANGE
+7.6% (+110 BPS)

SMART HOME
DEVICE



GWl.

14.1%

YEAR-ON-YEAR CHANGE
+14.6% (+180 BPS)

VIRTUAL REALITY
DEVICE



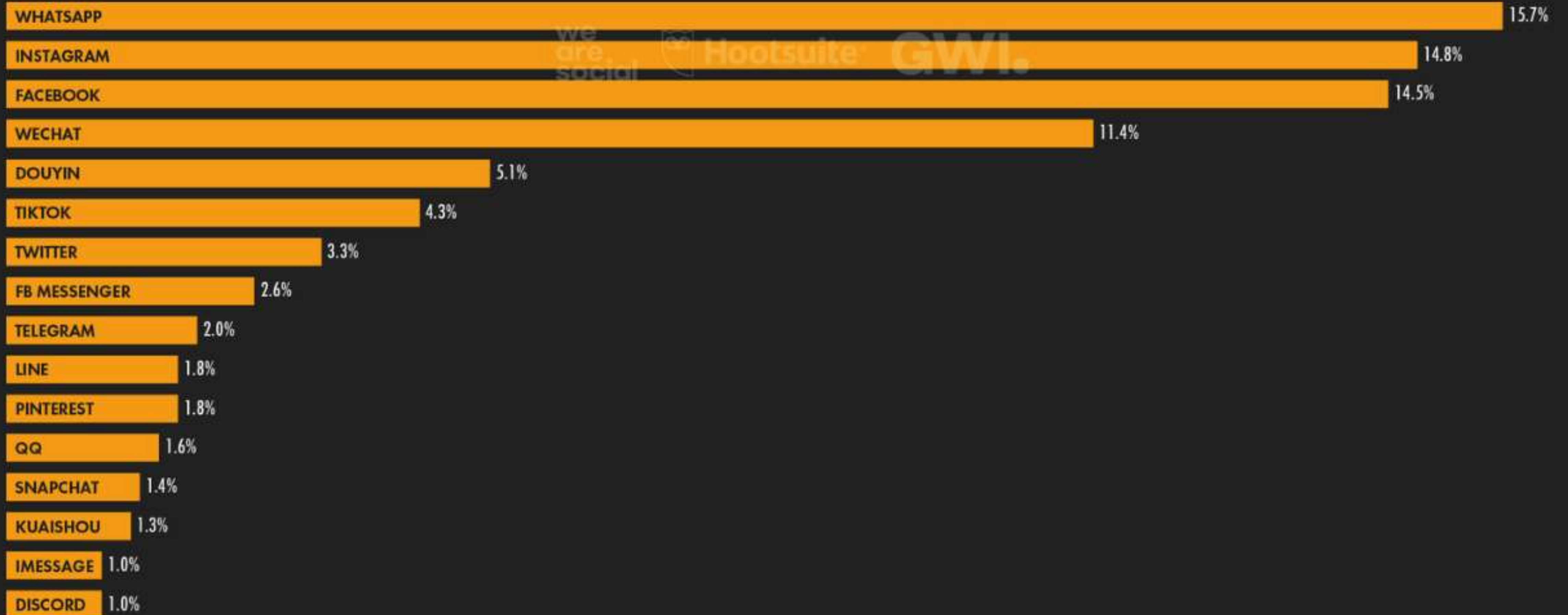
4.8%

YEAR-ON-YEAR CHANGE
+9.1% (+40 BPS)

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FAVOURITE SOCIAL MEDIA PLATFORMS

PERCENTAGE OF INTERNET USERS AGED 16 TO 64 WHO SAY THAT EACH OPTION IS THEIR "FAVOURITE" SOCIAL MEDIA PLATFORM



103

SOURCE: GWI (Q3 2021). SEE [GWI.COM](https://www.gwi.com) FOR FULL DETAILS. **NOTES:** ONLY INCLUDES USERS AGED 16 TO 64. SURVEY RESPONDENTS COULD CHOOSE FROM OTHER OPTIONS NOT SHOWN ON THIS CHART, SO VALUES MAY NOT SUM TO 100%. YOUTUBE IS NOT AVAILABLE AS AN ANSWER FOR THIS QUESTION IN GWI'S SURVEY; WE REPORT GWI'S VALUES FOR TIKTOK IN CHINA SEPARATELY AS DOUYIN, AS PER BYTEDANCE'S CORPORATE REPORTING. **COMPARABILITY:** VERSIONS OF THIS CHART THAT FEATURED IN OUR PREVIOUS REPORTS DID NOT INCLUDE DATA FOR CHINA, SO VALUES ARE **NOT COMPARABLE**

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TIME SPENT WITH SOCIAL MEDIA APPS

AVERAGE TIME PER MONTH THAT USERS SPEND USING EACH PLATFORM'S ANDROID APP, RANKED BY CUMULATIVE TIME ACROSS ALL ANDROID USERS



01: YOUTUBE



23.7
HOURS / MONTH



02: FACEBOOK



19.6
HOURS / MONTH



03: WHATSAPP



18.6
HOURS / MONTH



04: INSTAGRAM



11.2
HOURS / MONTH



05: TIKTOK



19.6
HOURS / MONTH

06: FACEBOOK MESSENGER



3.0
HOURS / MONTH



07: TWITTER



5.1
HOURS / MONTH



08: TELEGRAM



3.0
HOURS / MONTH



09: LINE



11.6
HOURS / MONTH



10: SNAPCHAT



3.0
HOURS / MONTH



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INSTAGRAM REELS AUDIENCE OVERVIEW

THE POTENTIAL AUDIENCE THAT MARKETERS CAN REACH WITH AD PLACEMENTS IN INSTAGRAM REELS

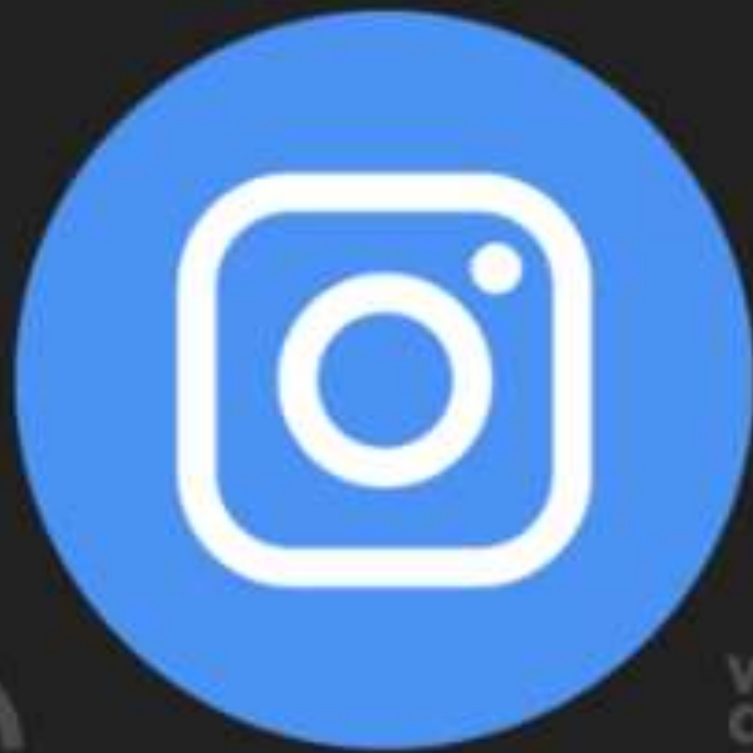


POTENTIAL AUDIENCE
THAT META REPORTS CAN
BE REACHED WITH ADS
IN INSTAGRAM REELS



675.3
MILLION

INSTAGRAM REELS AD
REACH AS A PERCENTAGE
OF INSTAGRAM'S TOTAL
ADVERTISING REACH



45.7%

INSTAGRAM REELS
ADVERTISING REACH AS
A PERCENTAGE OF TOTAL
POPULATION AGED 13+



10.9%

PERCENTAGE OF THE
INSTAGRAM REELS
AD AUDIENCE THAT
META REPORTS IS FEMALE



46.1%

PERCENTAGE OF THE
INSTAGRAM REELS
AD AUDIENCE THAT
META REPORTS IS MALE



53.9%

151

SOURCE: META'S ADVERTISING RESOURCES. **ADVISORY:** AUDIENCE FIGURES MAY NOT REPRESENT UNIQUE INDIVIDUALS, AND MAY NOT MATCH EQUIVALENT FIGURES FOR THE TOTAL ACTIVE USER BASE.
NOTES: FIGURES USE MIDPOINT OF PUBLISHED RANGES. META'S ADVERTISING RESOURCES ONLY PUBLISH GENDER DATA FOR "FEMALE" AND "MALE".

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Necesidades Globales de datos:

1. Crecimiento de logs exponencial
2. Crecimiento de demanda
3. Crecimiento de popularidad
4. Mayor inteligencia



¿Existe una correlación entre el nivel de vida de un país medido por el ingreso económico de su población, edad y el acceso a internet?

Data sources types in Big Data

1. Structured

2. None structured

3. Semi-structured



Structured Data



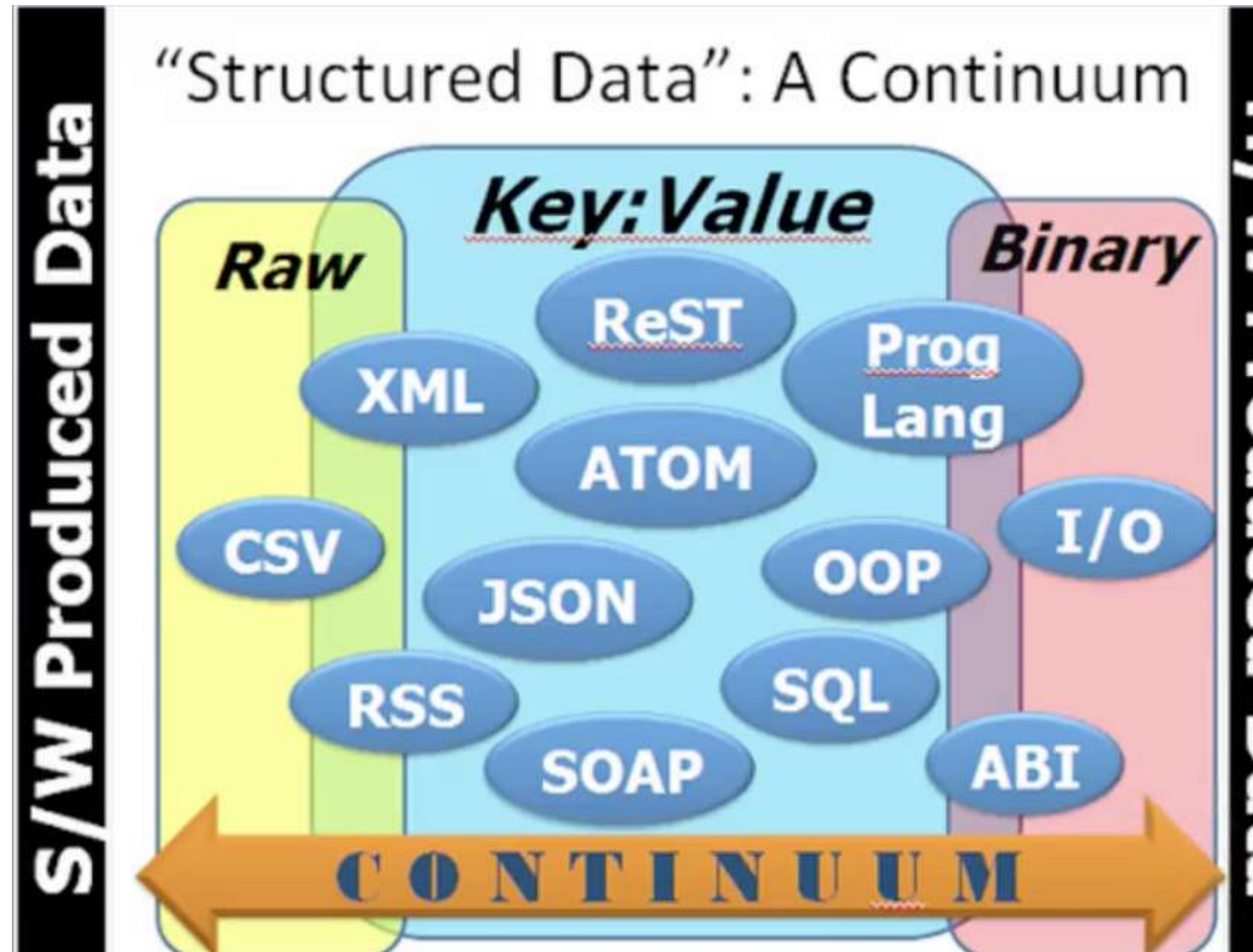
- **Big mass**
- **Diversity**
- **Storage**
- **Security**
- **Decision Support Systems and Answers Recovery**

None Structured Data, Organizations



- **80 -90%** world wide data
- **Images, videos, email, searches, text, pdfs, etc.**
- $V = \frac{\Delta X}{\Delta t}$

Data schemas, samples



Structured Data, Organizations

Past stored in SILOS

- Unconnected stored islands
- Hindered stored and connect silos for pattern recognition
- Outdated and unsynchronized

RDMS (Highly structured data) – SILOS → Value

Structured Data, Organizations

**Using pattern recognition a organizations
can use it for:**

- 1.- Detect correlated products**
- 2.- Estimated demand**
- 3.- Dapture fraudulent actions**

**Commerce+Open Data+Analytics→Better
predictions (Business Intelligence)**

Semi- Structured Data

16 PB data per year
1 mile per driver route
optimized, savings 50 mdd



250 millions clients, 10000
stores

2.5 PB per hour

New products, customize

recommendations,

predictive support



None-Structured Data, work path

Data acquisition

Storage

Retrieval

Cleaning

Processing



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For all data workload

ETL data
(Extract, Transform and Load)

To integrate different sources and loaded it into:
Data Warehouse-Data Lake-Data Mart

4 key properties of a data transaction

Atomicity

All changes to data are performed as if they are a single operation

Consistency

Data is in a consistent state when a transaction starts and when it ends.

Isolation

The intermediate state of a transaction is invisible to other transactions. As a result, transactions that run concurrently appear to be serialized.

Durability

After a transaction successfully completes, changes to data persist and are not undone, even in the event of a system failure.

None Structured Data, NoSQL

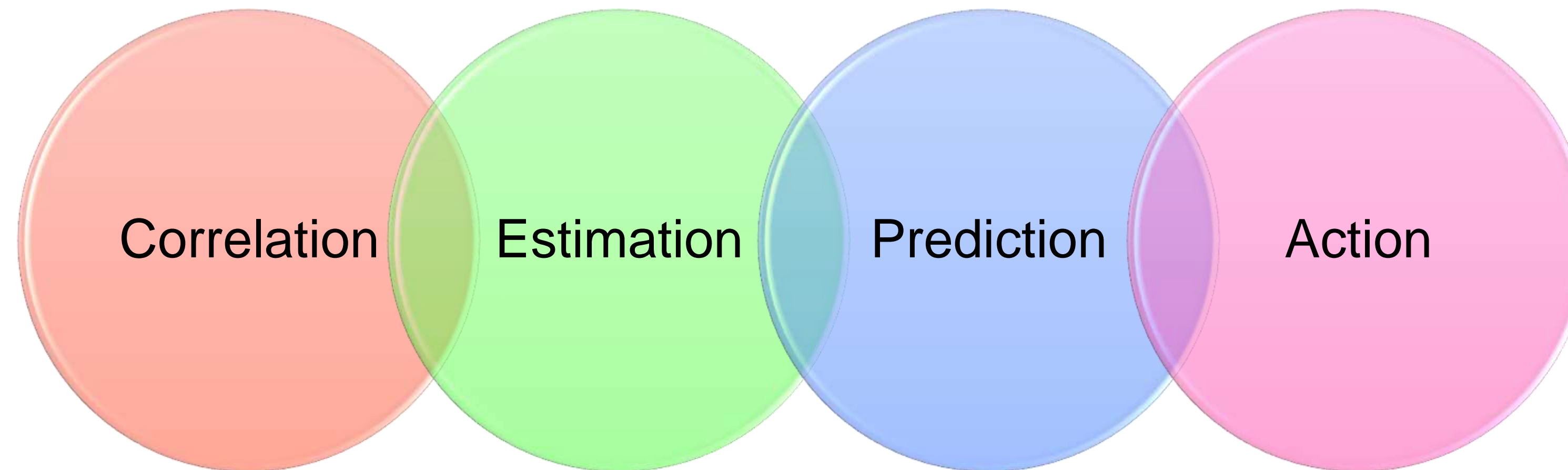
1. **Graph Data Base.**— used to find connections between data sets (Neo4j)
2. **Key Value Pairs.**— access and process data with key value pairs (Cassandra)
 - Layers for Value Big Data
 - Retrieval and Storage
 - Pre-processing
 - Analysis

12 TB/day representative data to sentiment analysis for a product or service (crisis mappers)

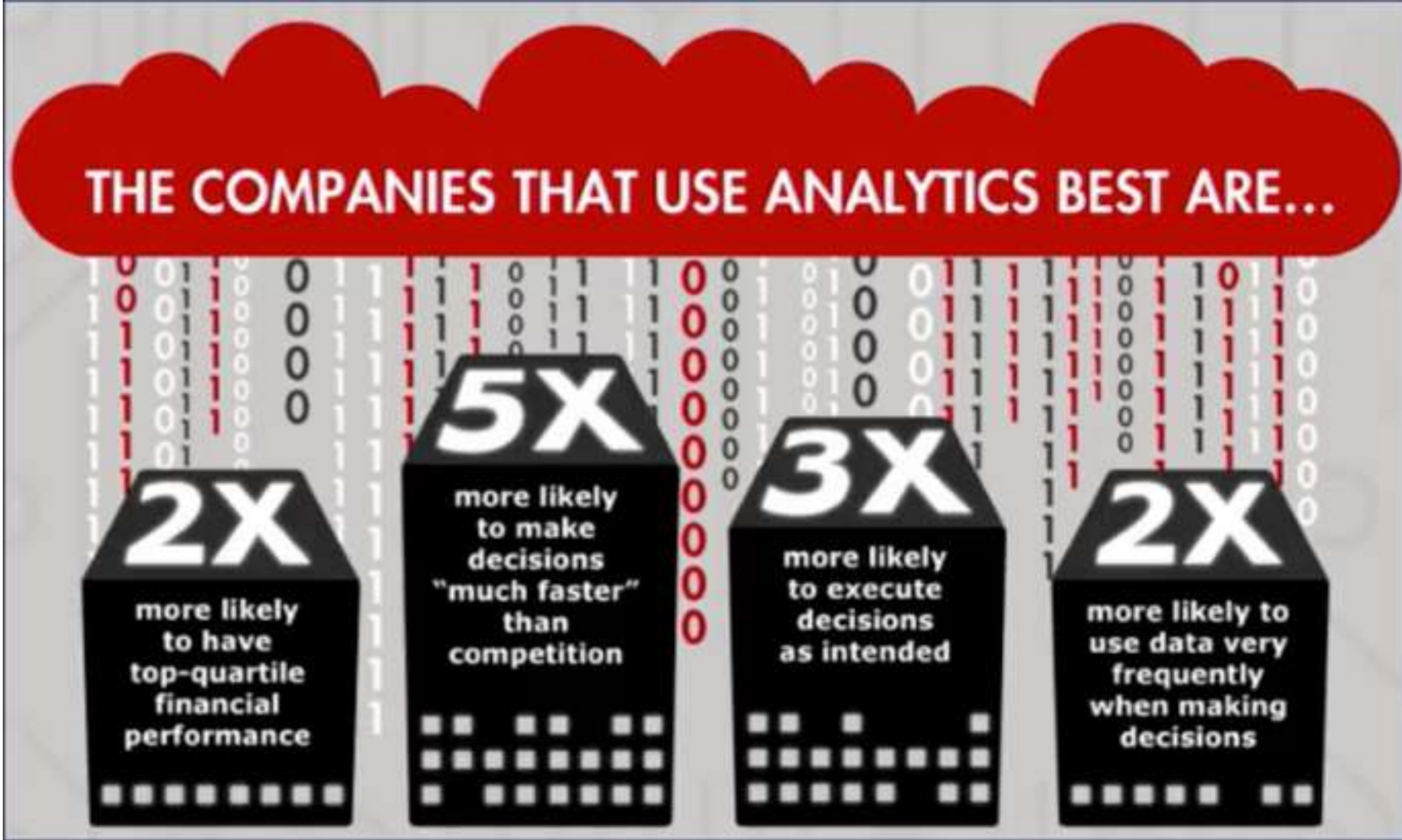


None Structured Data, NoSQL

Flow decision making



Strategical Big Data growth, new economical system approach



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OUTLINE OF A CIRCULAR ECONOMY

PRINCIPLE

1

Preserve and enhance natural capital by controlling finite stocks and balancing renewable resource flows
ReSOLVE levers: regenerate, virtualise, exchange

PRINCIPLE

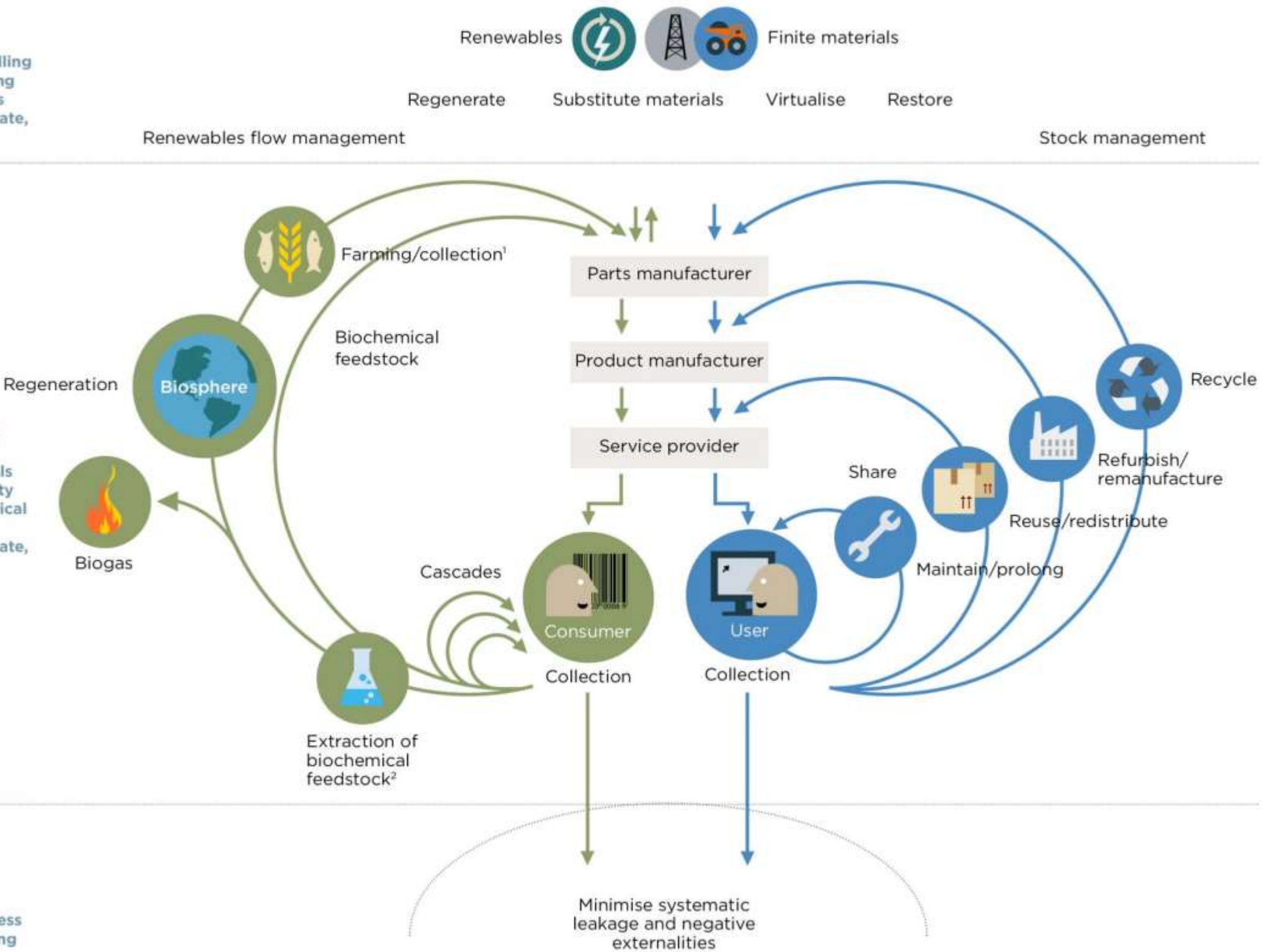
2

Optimise resource yields by circulating products, components and materials in use at the highest utility at all times in both technical and biological cycles
ReSOLVE levers: regenerate, share, optimise, loop

PRINCIPLE

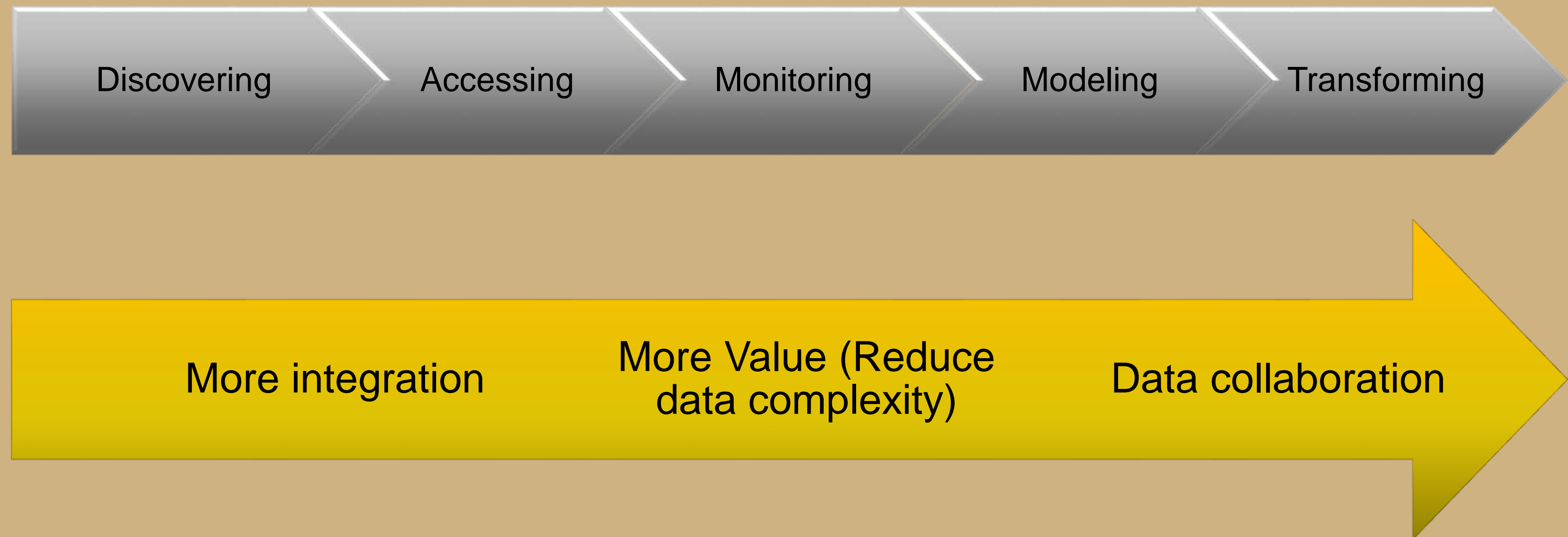
3

Foster system effectiveness by revealing and designing out negative externalities
All ReSOLVE levers



1. Hunting and fishing
2. Can take both post-harvest and post-consumer waste as an input
Source: Ellen MacArthur Foundation, SUN, and McKinsey Center for Business and Environment; Drawing from Braungart & McDonough, Cradle to Cradle (C2C).

Data integration Process



Activity 2:

1. Read the appropriate article according the assigned team.
2. Answer the questiones assigned to the article.
3. Every team must prepare and present a 5 minute resume

Team 1

- [Link](#)

Team 2

- [Link](#)

Team 3

- [Link](#)

Team 4

- [Link](#)

Team 5

- [Link](#)

Questions Team 1:

1. Does the Reels are changing the consumer experiences and it reflect a new data Vs value?
2. Wich one's and how?
3. What trends (3) do you identify after Covid-19 over data user production?
4. Are those data commonly Structured, NoStructured or SemiStructured? Explain, why?

Questions Team 2:

1. What are the relations between Data Analytics and Business Intelligence (3)?
2. What are the relations between Data Science and Business Intelligence (3)?
3. How do you explain this for a new user or Company who wants to be involved in BI?
4. How could they start with it?

Questions Team 3:

1. For the 12 data keypoint, how many of them are affected by AI?
2. In conclusion, the D&A should be focused on people or metaverse? Explain, why?
3. How do you resume this article in one paragraph?
4. Does geopolitics are changing data production and consumption?

Questions Team 4:

1. How do you represent graphically the relation between Data Science-ML-AI?
2. How do you explain the role of Big Data in Data Science, ML and AI?
3. Describe three agrees, and three disagrees about the article's content
4. According the study case: "Self-driving car". Do you agree with the ML, AI and Data Science descriptions? Explain, What is the Big Data role in this case according 10 V's?

Questions Team 5:

1. Does Data Scientist and Data Engineer would do the same job? Explain, why?
2. Define five key skills that make difference between Data Scientist and Data Engineer
3. Does DASA is leading changing the employment bases for Data Scientist and Data Engineers?
4. How has been involved the Big Data in this new industrial revolution called 4.0?