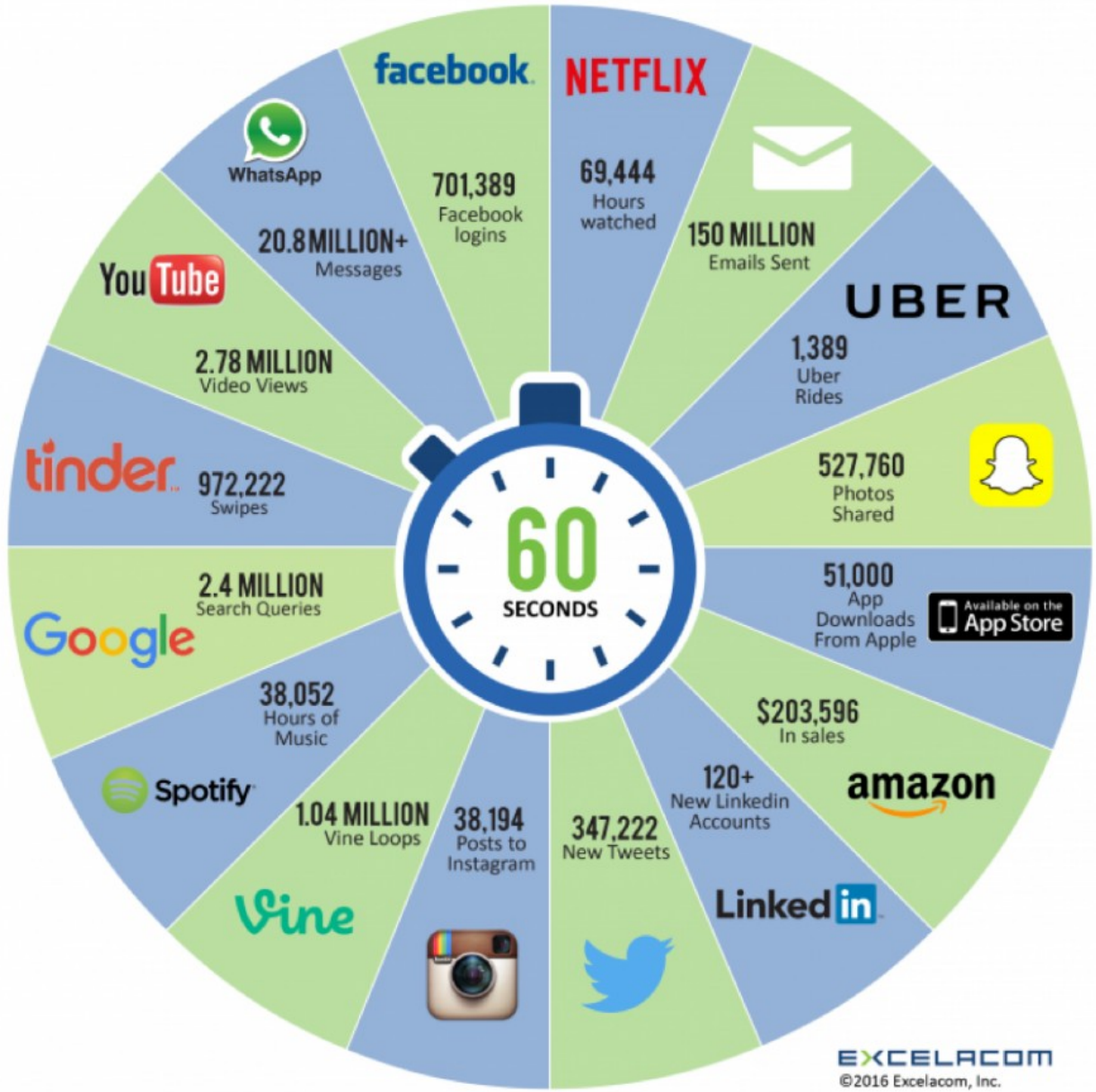
Arquitectura Big Data

Angela Devia
ing.angela.devia@gmail.com
Científicas de Datos
www.cientificasdedatos.com

2016 What happens in an INTERNET MINUTE?



EXCELACOM

The FOUR V's of Big Data

From traffic patterns and music downloads to web history and medical records, data is recorded, stored, and analyzed to enable the technology and services that the world relies on every day. But what exactly is big data, and how can these massive amounts of data be used?

As a leader in the sector, IBM data scientists break big data into four dimensions: **Volume, Velocity, Variety and Veracity**

Depending on the industry and organization, big data encompasses information from multiple internal and external sources such as transactions, social media, enterprise content, sensors and mobile devices. Companies can leverage data to adapt their products and services to better meet customer needs, optimize operations and infrastructure, and find new sources of revenue.

By 2015

4.4 MILLION IT JOBS

will be created globally to support big data, with 1.9 million in the United States

Volume

SCALE OF DATA

40 ZETTABYTES

[43 TRILLION GIGABYTES]

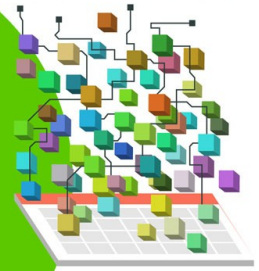
of data will be created by 2020, an increase of 300 times from 2005

It's estimated that

2.5 QUINTILLION BYTES

[2.3 TRILLION GIGABYTES]

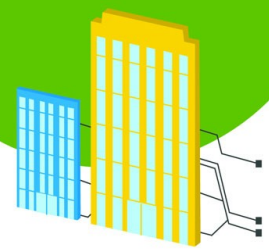
of data are created each day



6 BILLION PEOPLE

have cell phones

WORLD POPULATION: 7 BILLION



Most companies in the U.S. have at least

100 TERABYTES

[100,000 GIGABYTES]

of data stored

Variety

DIFFERENT FORMS OF DATA

As of 2011, the global size of data in healthcare was estimated to be

150 EXABYTES

[161 BILLION GIGABYTES]

30 BILLION PIECES OF CONTENT

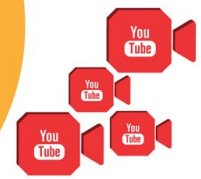
are shared on Facebook every month

By 2014, it's anticipated there will be

420 MILLION WEARABLE, WIRELESS HEALTH MONITORS

4 BILLION+ HOURS OF VIDEO

are watched on YouTube each month



400 MILLION TWEETS

are sent per day by about 200 million monthly active users

Velocity

ANALYSIS OF STREAMING DATA

The New York Stock Exchange captures

1 TB OF TRADE INFORMATION

during each trading session

Modern cars have close to

100 SENSORS

that monitor items such as fuel level and tire pressure

By 2016, it is projected there will be

18.9 BILLION NETWORK CONNECTIONS

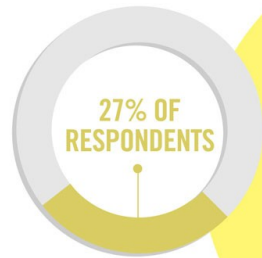
– almost 2.5 connections per person on earth

Veracity

UNCERTAINTY OF DATA

1 IN 3 BUSINESS LEADERS

don't trust the information they use to make decisions



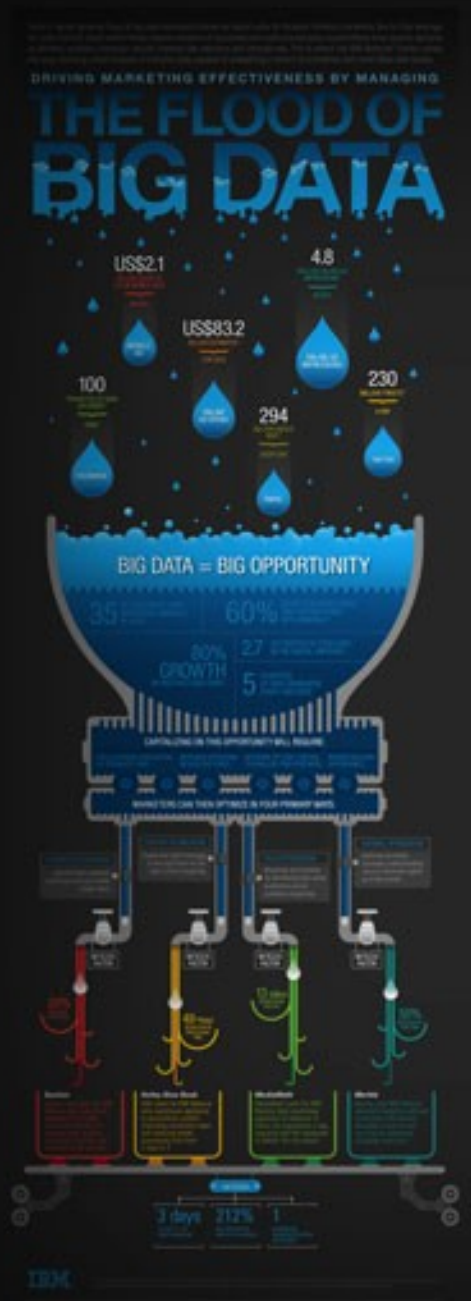
in one survey were unsure of how much of their data was inaccurate

Poor data quality costs the US economy around

$3.1 TRILLION A YEAR

Sources: McKinsey Global Institute, Twitter, Cisco, Gartner, EMC, SAS, IBM, MEPTEC, QAS

IBM



FACEBOOK

EMAIL

BIG DATA = BIG OPPORTUNITY

35 ZETTABYTES OF DATA GENERATED ANNUALLY BY 2020[7]

60% GROWTH IN STRUCTURED AND UNSTRUCTURED DATA ANNUALLY[8]

80% GROWTH IN UNSTRUCTURED DATA[10]

2.7 ZETTABYTES OF DATA EXIST IN THE DIGITAL UNIVERSE[9]

5 EXABYTES OF DATA GENERATED EVERY TWO DAYS[11]

CAPITALIZING ON THIS OPPORTUNITY WILL REQUIRE:

- RULES-DRIVEN INTEGRATION OF DISPARATE DATA
- IMPROVED OPERATING INFRASTRUCTURES
- NETWORK OF DATA-CENTRIC TECHNOLOGY AND PARTNERS
- MARKETING DATA GOVERNANCE

BIG DATA LANDSCAPE 2017

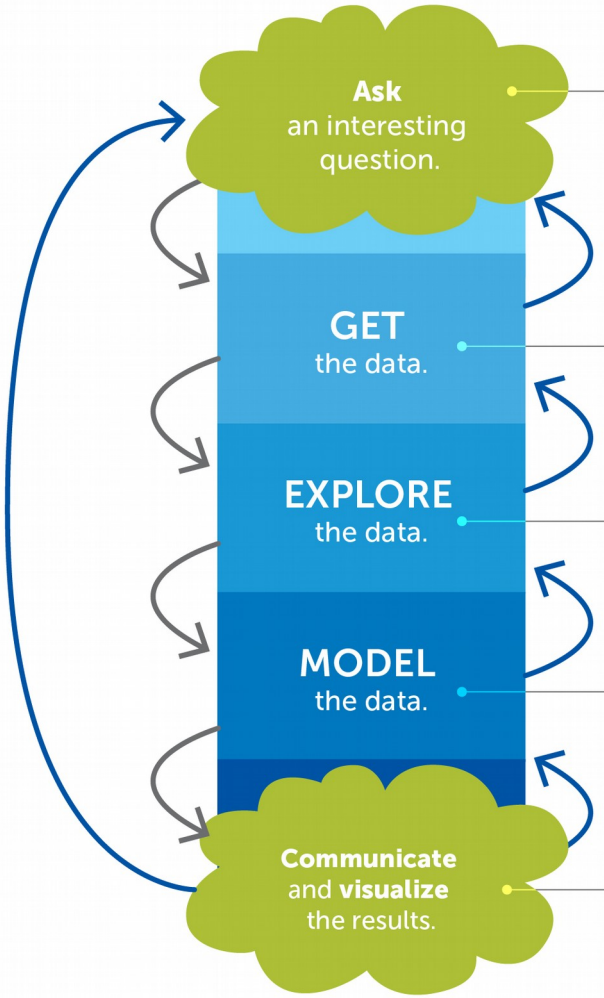
Last updated 4/5/2017

 mattturck.com/bigdata2017

FIRSTMARK
EARLY STAGE VENTURE CAPITAL

Extended Relational Reference Architecture

The **Data Science** Process



What is the scientific **goal?**
What would you do if you had all the **data?**
What do you want to **predict** or **estimate?**

How were the data **sampled?**
Which data are **relevant?**
Are there **privacy** issues?

Plot the data.
Are there **anomalies?**
Are there **patterns?**

Build a model.
Fit the model.
Validate the model.

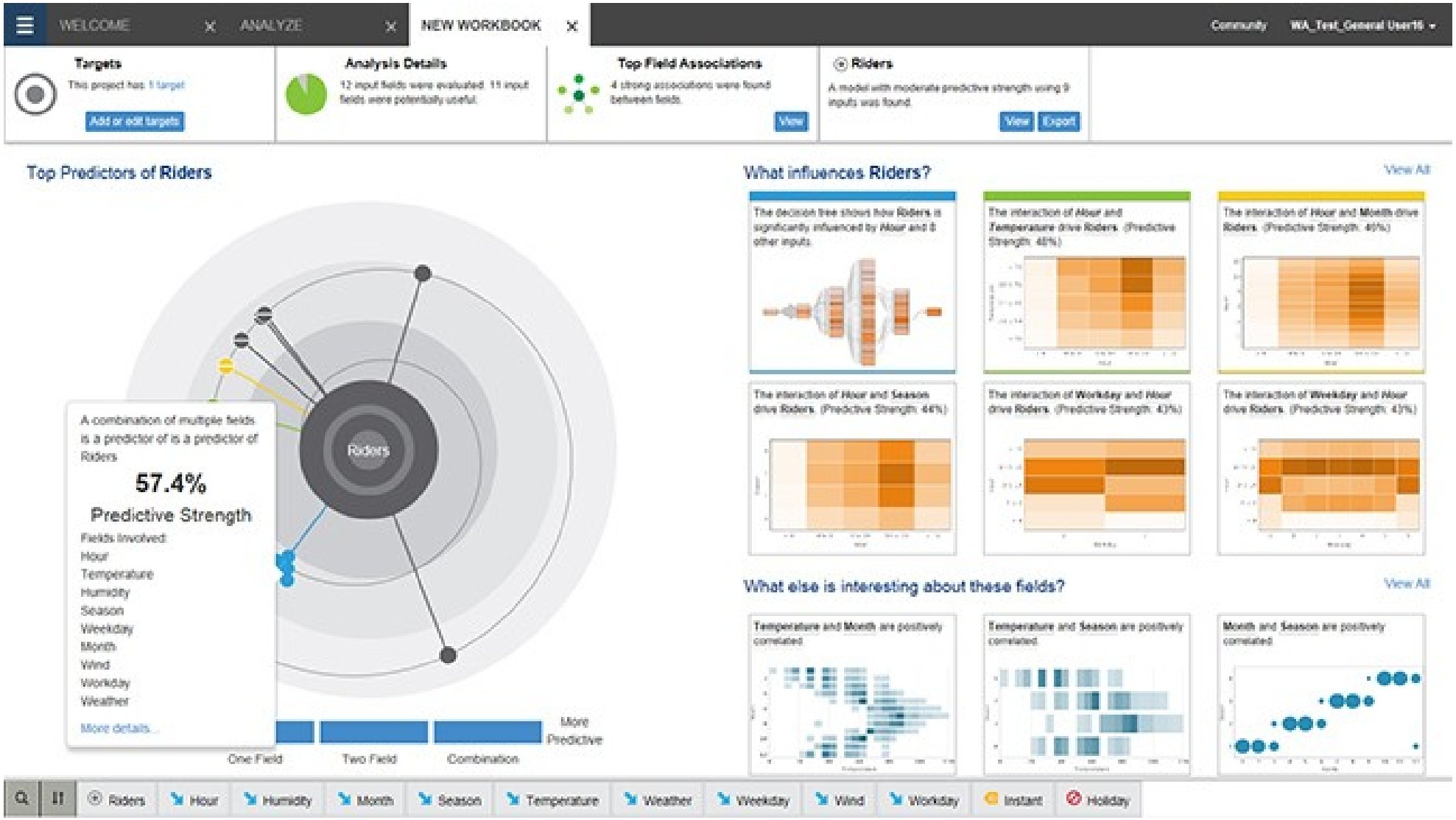
What did we **learn?**
Do the results make **sense?**
Can we tell a **story?**

Derived from the work of Joe Blitzstein and Hanspeter Pfister,
originally created for the Harvard data science course http://cs109.org/.

2016 Magic Quadrant for Advanced Analytics Platforms

2017 Magic Quadrant for Data Science Platforms

IBM Watson


WELCOME
ANALYZE
NEW WORKBOOK
Community
WA_Test_General User16
Targets
This project has 1 target
Add or edit targets
Analysis Details
12 input fields were evaluated. 11 input fields were potentially useful.
Top Field Associations
4 strong associations were found between fields.
View
Riders
A model with moderate predictive strength using 9 inputs was found.
View
Export
Top Predictors of Riders
A combination of multiple fields is a predictor of is a predictor of Riders
57.4%
Predictive Strength
Fields Involved:
Hour
Temperature
Humidity
Season
Weekday
Month
Wind
Workday
Weather
More details...
Riders
One Field
Two Field
Combination
More Predictive
What influences Riders?
View All
The decision tree shows how Riders is significantly influenced by Hour and 8 other inputs.
The interaction of Hour and Temperature drive Riders. (Predictive Strength: 46%)
The interaction of Hour and Month drive Riders. (Predictive Strength: 45%)
The interaction of Hour and Season drive Riders. (Predictive Strength: 44%)
The interaction of Workday and Hour drive Riders. (Predictive Strength: 43%)
The interaction of Weekday and Hour drive Riders. (Predictive Strength: 43%)
What else is interesting about these fields?
View All
Temperature and Month are positively correlated
Temperature and Season are positively correlated
Month and Season are positively correlated
Riders
Hour
Humidity
Month
Season
Temperature
Weather
Weekday
Wind
Workday
Instant
Holiday

TARGETS

This project has **1 target**

Edit

ANALYSIS DETAIL

131 input fields were evaluated.
120 were potentially useful.

120

TOP FIELD ASSOCIATIONS

62 strong associations were found between fields.

View

SALES

A model with high predictive strength using **20 inputs** was found.

20

FAVORITES

One insights has been marked as a favorite.

What predicts **Purchase Made**?

2 strong predictors and **1 moderate predictor** have been found and are shown below.

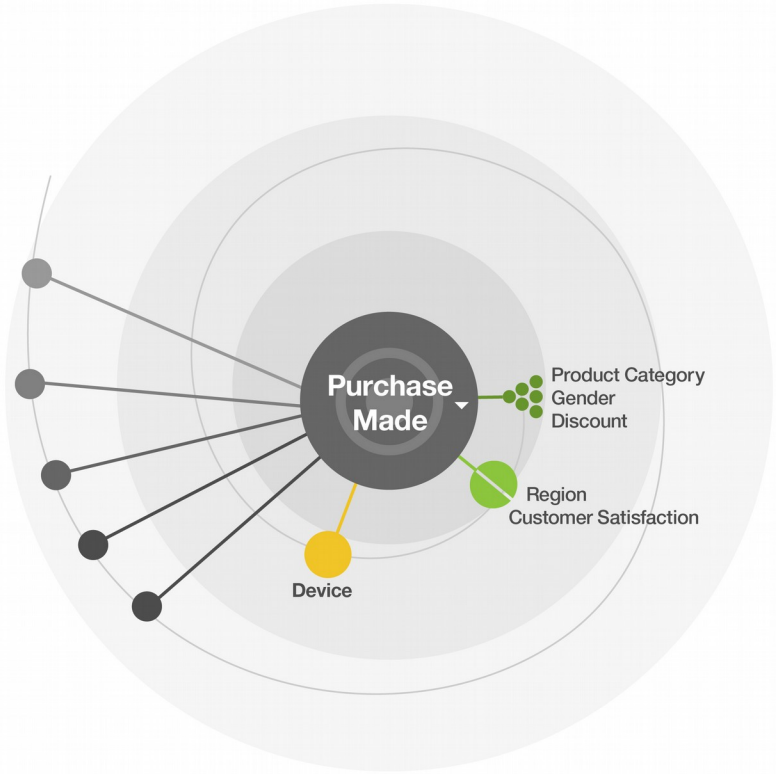
More **Predictive**

Combination

Two field

One field

Easier to **Understand**



What influences **Purchase Made**?

1 ★

Product Category, Gender, and **Discount** predict **Purchase Made**

Product Category
Gender
Discount
Time
Sentiment
Device
Region

2 ☆

Region and **Customer Satisfaction** predict **Purchase Made**

3 ☆

Device predicts **Purchase Made**

What else is interesting about this?

Average Sale Amount differs across **Time** ☆

High Fashion and **Region** are strongly associated. ☆

Discount and **Time on Site** are associated. ☆

Web Sales Data

Date	Time	Source	Device	Product_Category	Purchase Made?	Sale Amount	Country	Discount

IBM resources

- https://www.ibm.com/analytics/us/en/industry/government/

- **IBM Certified Data Architect - Big Data**

http://www-03.ibm.com/certify/certs/50001701.shtml

- https://datascience.ibm.com/

Felicitaciones por tu interés en los datos y cómo usarlos

Ángela Devia - ing.angela.devia@gmail.com

Gracias