

Agenda

- Big Data
- > Data Science
- > Analytics
- > Predictive Analytics
- Data Science/Python
- > Python Code



What Happens in an Internet Minute?



2016 INTERNET MINUTE?



2019 This Is What Happens In An Internet Minute



2017 This Is What Happens In An Internet Minute



2020 This Is What Happens In An Internet Minute



2018 This Is What Happens In An Internet Minute

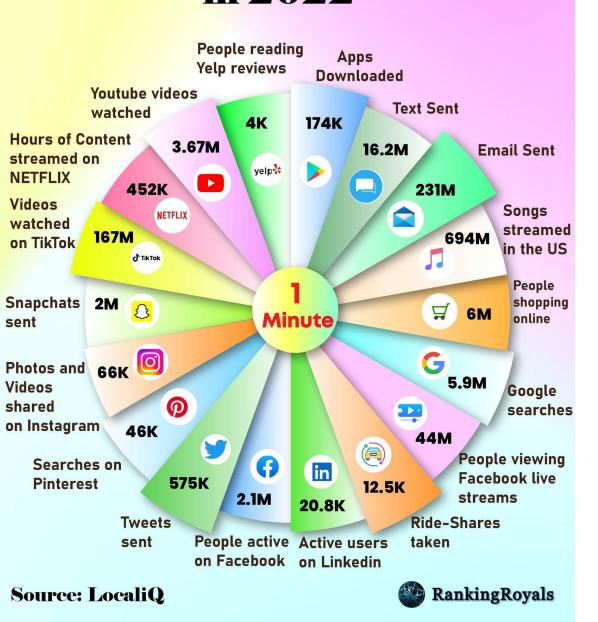


2021 This Is What Happens In An Internet Minute





A Minute on the Internet in 2022

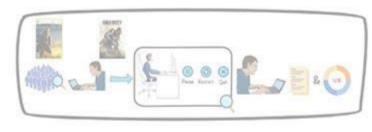


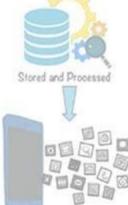
THE INTERNET IN 2023 EVERY MINUTE



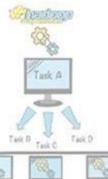












Parallel Processing









Hurricane Sandy in 2012





BIG DATA Key Activities

Store

Process

Access



Structured

Types of Big Data







VOLUME

- Amount of data generated
- Online & offline transactions
- In kilobytes or terabytes
- Saved in records, tables, files



VELOCITY

- Speed of generating data
- Generated in real-time
- Online and offline data
- In Streams, batch or bits

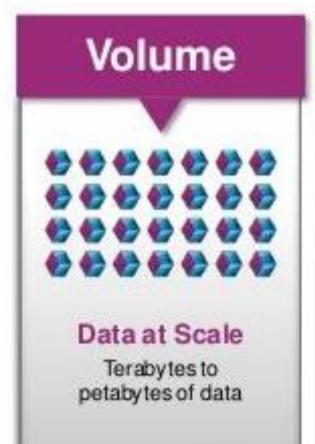
VARIETY

- Structured & unstructured
- Online images & videos
- Human generated texts
- Machine generated readings



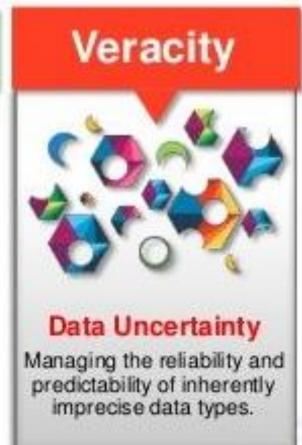














VOLUME VELOCITY **3 3 3** 0000 Data in Data at

Rest

Motion



VARIETY





Data in Doubt



?	?	?	?	?
?	?	?	?	?
?	?	?	?	?
?	?	?	?	?
?	?	?	?	?
?	?	?	?	?
?	?	?	?	?
?	?	?	?	?



Data in Limbo





VOLUME

VARIETY

Structured
Unstructured
Multi-factor
Probabilistic
Linked/dynamic

(Giga)Terabytes Records/archive Tables, files Distributed

VELOCITY

Batch Real/near-time Streams Transactions/ processes

Generic big data requirements:

- · variety,
- volume,
- velocity.

6Vs of big data

Changing data Changing model
Linkage
Integration

Trustworthiness
Authenticity
Origin/reputation
Accountability
Availability

Correlation
Simulation
Prediction
Hypothetical
Events

VALUE

Acquired big data requirements (after entering system):

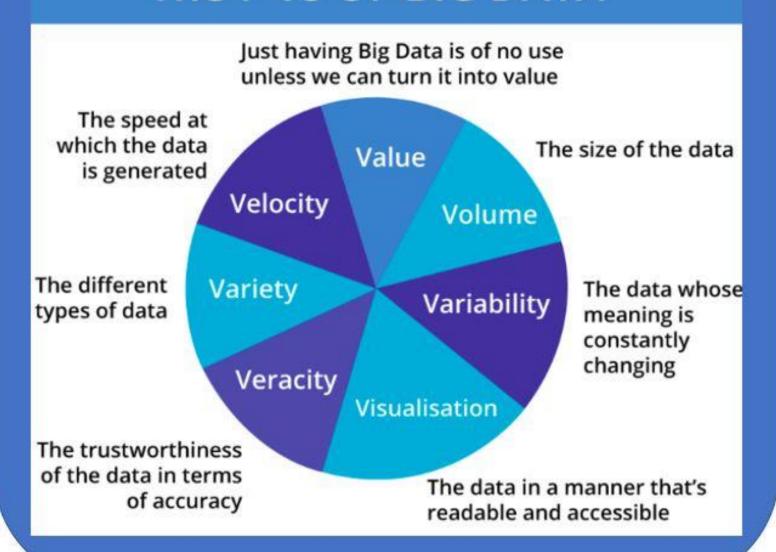
- variability,
- veracity,
- value.

VARIABILITY

VERACITY



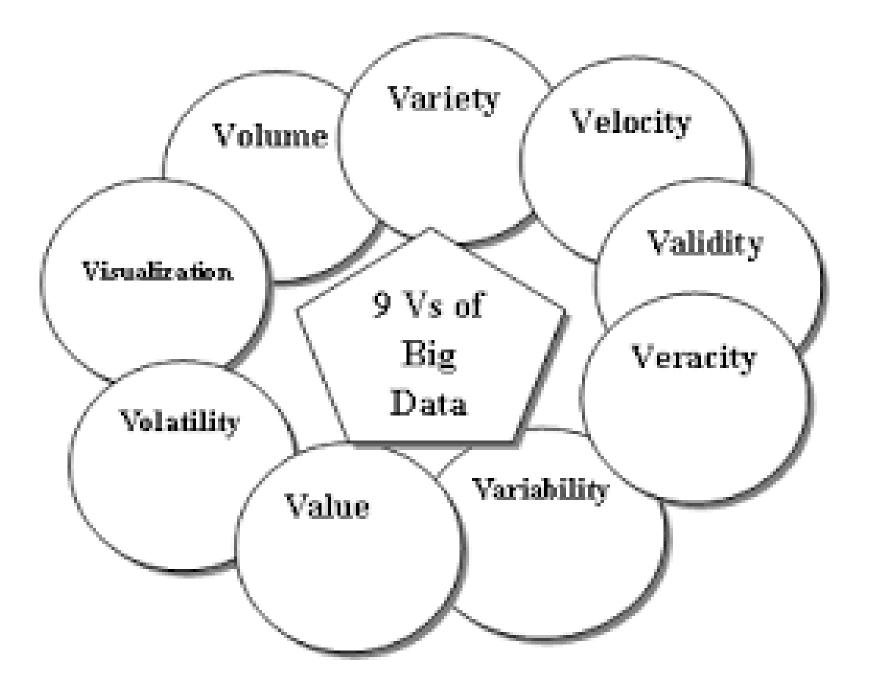
The 7 Vs OF BIG DATA













Value

Can you find the information you are looking for?



Virality

Aha to - go? Does it convey a message that can be pasted into a presentation or instagrammed?



Distributed Heterogeneous Data from Multiple Platforms



Variability

Dynamic, Evolving Behaviour in Data Source



Can you find it when you most need it?



Big Data

With 10 V's



Viscosity

Does it stick with you? Does it call for action?

Veracity

Are you dealing with information or disinformation?



Wisualisation

Can you make sense at a glance? Does it trigger a decision?



information gains momentum and crises & opportunities enolve in real time Hoe is outlook for today?



Variety

is a picture worth a thousand words 102 languages? IS your information balance 302





Retail POS



Machine & Sensor





Bank/Credit card



DATA



Video & Preference

- Social Data
- Machine-Generated Data
- Transactional Data



Transactional



Mobile



Social

DATA PREPARATION

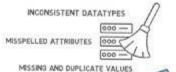
EXPLORATORY DATA ANALYSIS

DATA MODELING

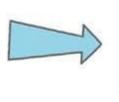


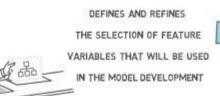
DATA CLEANING

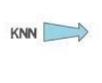
TRANSFORMATION

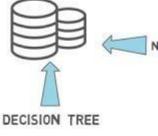














DATA ACQUISITION

- WEB SERVERS

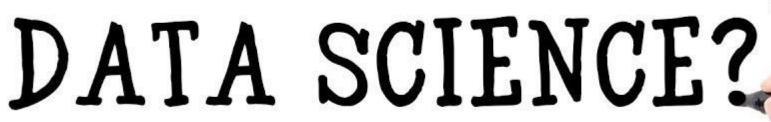


- DATABASES

- APIS

- ONLINE REPOSITORIES

WHATIS



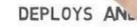


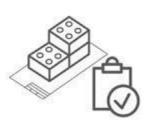
WHY?....WHY?....WHY?....









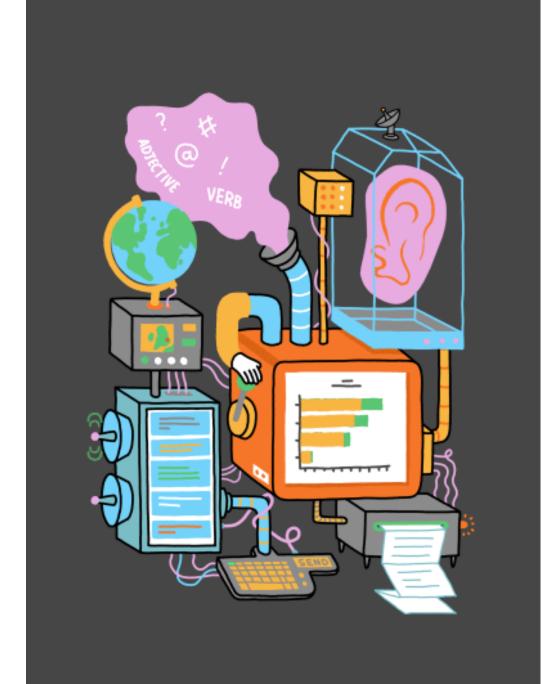






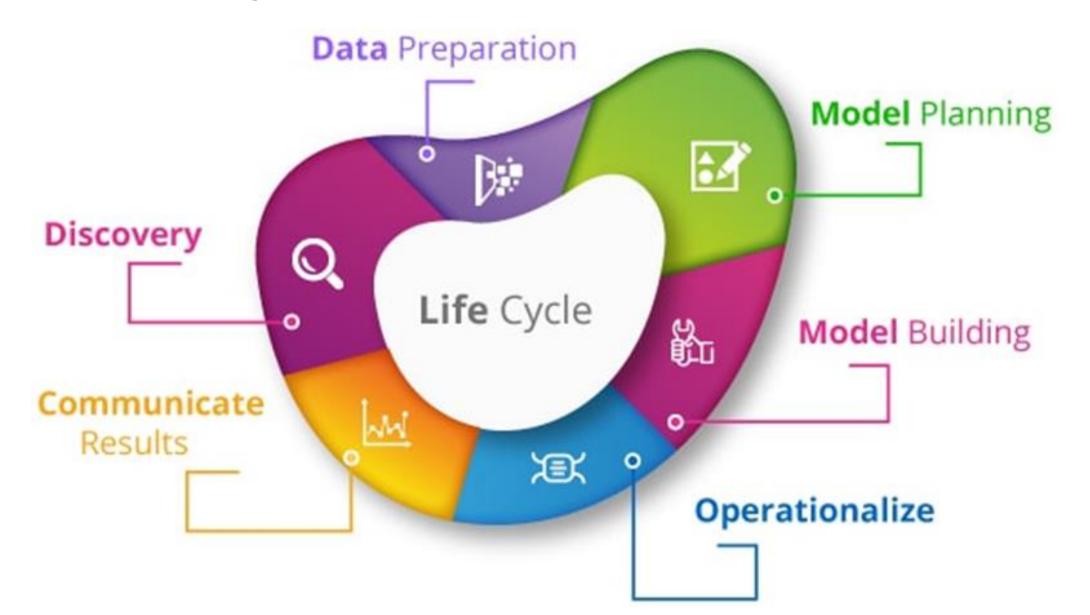








The lifecycle of Data Science

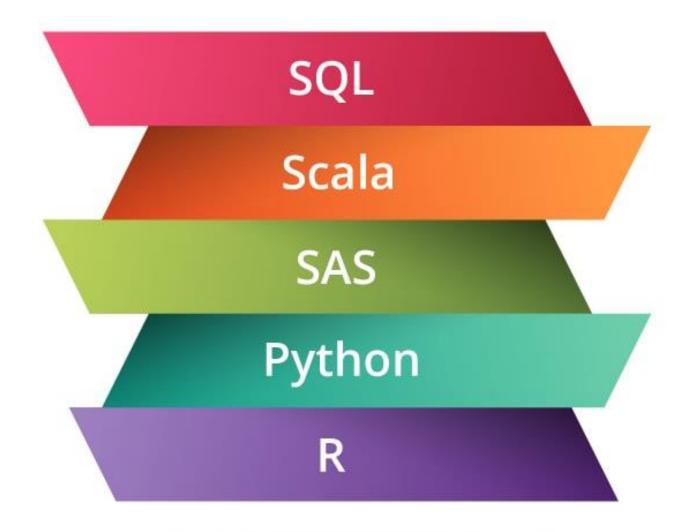




Advantage of Data Science



Top 5 programming languages in Data Science





Basis	Big Data	Data Science
Meaning	revolves around the huge volumes of data which cannot be handled using the conventional data analysis method	skewed towards the scientific approach of interpreting the data and retrieves the information from a given data set
Concept	scientific techniques to process data lextract	obtained with big data is heterogeneous that indicates a diversified data set which has to be pre-cleaned and sorted before running analytics on them
Formation	data filtering, preparation, and analysis	Internet users/ traffic, live feeds, and data generated from system logs
Application		Internet search, digital advertisements, text-to-speech recognition, risk detection, and other activities
Approach	used by businesses to track their presence in the market which helps them develop agility and gain a competitive advantage over others	uses mathematics and statistics extensively along with programming skills to develop a model to test the hypothesis and make decisions in the business



Big Data Vs Data Science

Factors

Concept

Responsibility

Industry

Tools

Big Data

Handling large data

Process huge volumes of data and generate insights

E-commerce, security services, telecommunication

Hadoop, Spark, Flink

Data Science

Analyzing data

Understand pattern within data and make decisions

Sales, image recognition, advertisement, risk analytics

SAS, R, Python



What is Data Analytics?





What Is Big Data Analytics?





WHAT IS DATA SCIENCE? WHAT IS DATA ANALYTICS? WHAT IS BIG DATA? Data Science is a field Data Analytics (DA) is the that refers to the collective Big Data refers to voluminous process of examining data sets amounts of structured or processes, theories, concepts, in order to draw conclusions tools and technologies that unstructured data that about the information they enable the review, analysis and organizations can potentially contain, increasingly with the aid extraction of valuable knowledge mine & analyze for business gains. of specialized systems & software. and information from raw data. APPLICATION AREAS 1. Digital advertisements 1. Communication 1. Gaming 2. Internet Research 2. Travel 2. Retail 3. Recommender System 3. Financial services 3. Energy Management 4. Image/Speech Recognition 4. Healthcare 4. Education **TOOLS & LANGUAGES** 1. Python 1. R 1. Hadoop 2. SAS 2. Tableau Public 2. NoSQL 3. SQL 3. Apache Spark 3. Hive



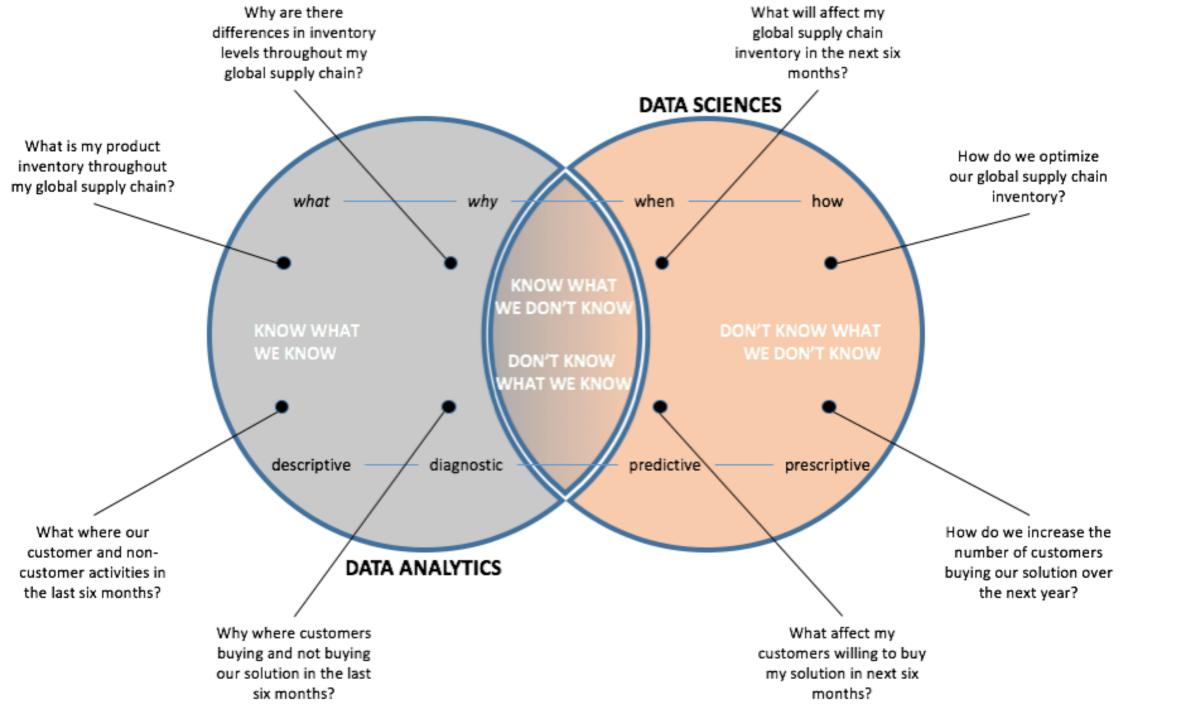
Differentiation	Big Data	Data Science	Data Analytics
Definition	Unprocessed data sets of humongous volumes	Science of cleaning, preparing and aligning the data for analysis using statistical and mathematical models	It is related to examining raw data which is required to provide conclusive information
	Financial services	Delivery of better search results on the internet	Gaining efficiency in the Healthcare
Applications	Fraud analytics	Digital advertisements from display banners to finding the appropriate prospects	Optimization of buying experience through mobile and social media data analysis.
	Communication industry to retain and expand the consumer base	The recommender system to help in the user experience	Collection of data in the gaming industry
	Brick-and-mortar and online retailer for better customer service		Energy management
	In order to become a big data professional, the following skills are required:	A data scientist must highlight a profile that has the following skills:	Following skills are necessary to become a data analyst:
	Analytical skills	Education with a Master's degree in either data analysis, statistics, or mathematics	Programming skills
Skill requirements	Creativity	Knowledge of SAS or R or SPSS	Statistical and mathematics skills
-	Mathematics and statistical skills	Knowledge in coding on Python and Hadoop	Machine learning skills and certificates
	Basic computation knowledge	Working efficiency with unstructured data	Data visualization skills
	Computer science and business skills		Analytical skills
			Data wrangling skills



Analysis looks backwards over time.

Analytics look forward to model the future or predict a result







Data Science vs Data Analytics

	Data Science	Data Analytics
SKILLSET	 Data Modelling Predictive Analytics Advanced Statistics Engineering/Programming 	 BI Tools Intermediate Statistics Solid Programming Skills Regular Expression (SQL)
SCOPE		Micro
EXPLORATION	 Search Engine Exploration Machine Learning Artificial Intelligence Big data - Often Unstructured 	 Data Visualization Techniques Designing Principles Big Data - Mostly Structured
Discover New Questions to Drive Innovation		Use Existing Information to Uncover Actionable Data



Use of Big Data in Data Analytics



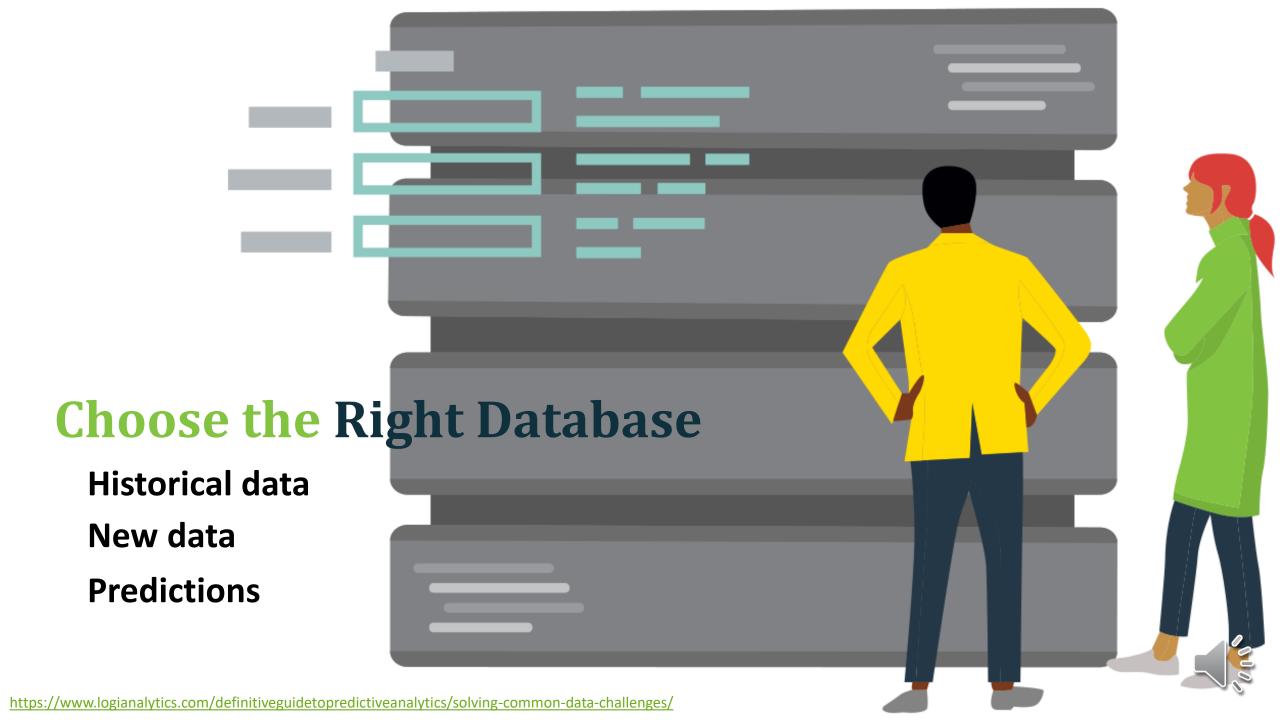


Solving Common Data Challenges

Find the Data You Need **Choose the Right Database Practice** Database Hygiene **Cleanse Your Data Avoid Bias in Your Data and Models** Validate your model is working and establish a performance baseline **Know When It's Time to Refresh Boost Predictive Performance Over Time How to Adapt to Changing Business Requirements**









Cleanse Your Data

1. Missing values

Row no	State	Salary	Yrs of Experience
1	NY	57400	Mid
2	TX	1	Entry
3	NJ	90000	High
4	VT /	36900	Entry
5	TX /	1	Mid
6	CA /	76600	High
7	NY /	85000	High
8	CA //	,	Entry
9	ст / /	45000	Entry
	111		



- 1. Delete the rows
- 2. Replace with mean value
- 3. Predict the value using other columns

2. Outliers



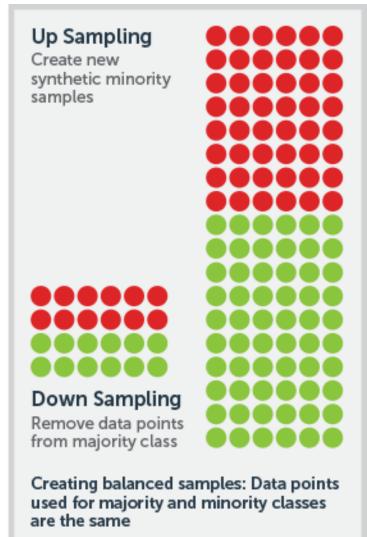
Outliers: Remove or clip it

Component 2

Avoid Bias in Your Data and Models

- 1. Data Bias
- 2. Selection bias









Validate your model is working and establish a performance baseline

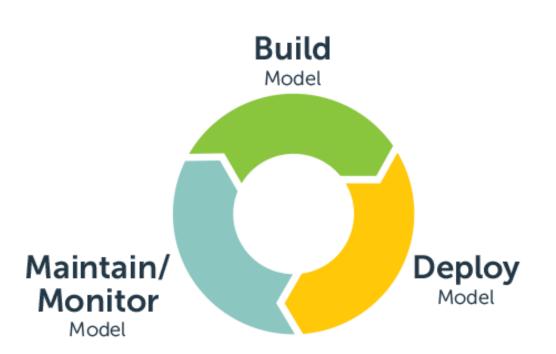
Accuracy
Watch for Imbalanced Data



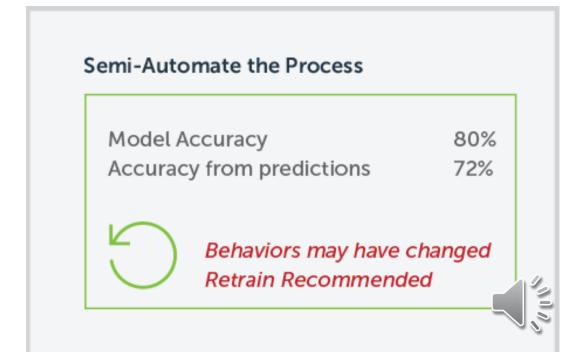


Know When It's Time to Refresh

Seasonal.
Measurement-based.







Boost Predictive Performance Over Time

Data Variety
Model Refinement Based on
Additional Performance Metrics



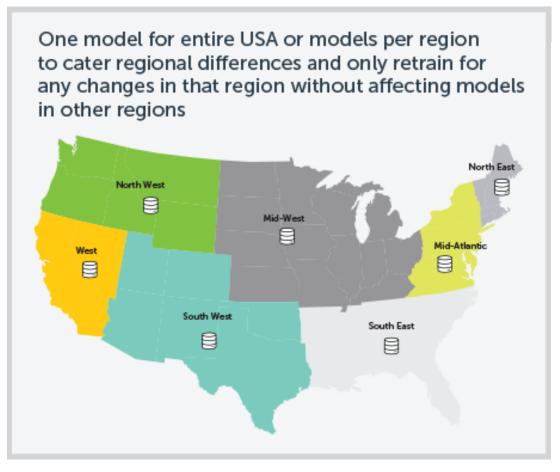


Identify Target Buyer?





How to Adapt to Changing Business Requirements





Data Analytics: Overview

Qualitative and quantitative data Structured and unstructured data

The data analysis process
Types of data analytics
Data analytics trends



Qualitative and quantitative data

What is qualitative data? This bookcase...

- Is made of wood
- Was built in Italy
- Is deep brown
- Has golden knobs
- Smells like oak
- Has a smooth finish

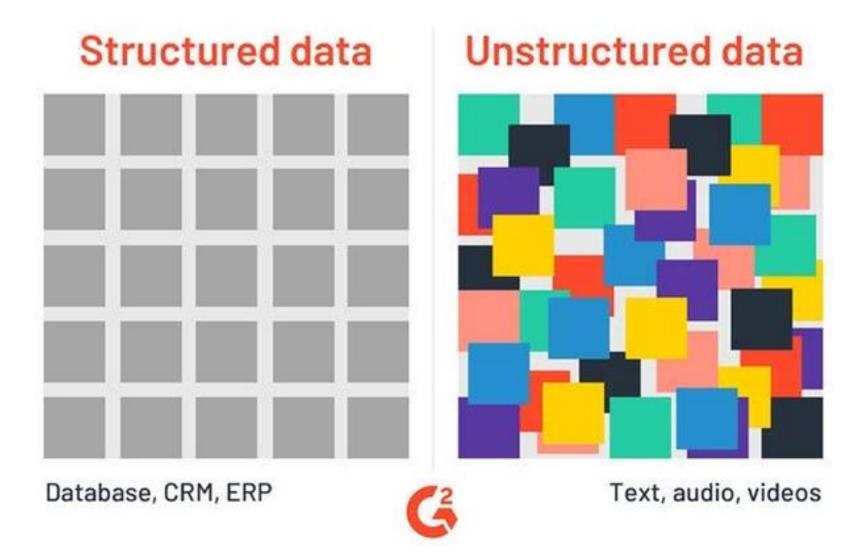


What is quantitative data? This bookcase...

- Is 3 feet tall
- Weighs 100 pounds
- Has 15 books on it
- Has 3 shelves
- Has 2 cabinets
- Sells for \$1500



Structured and unstructured data



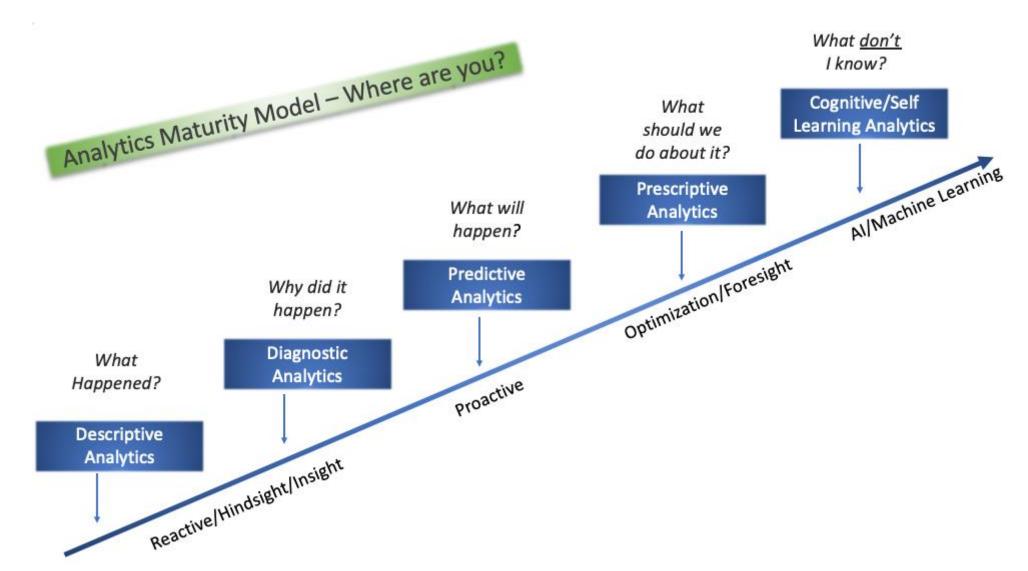


The data analysis process

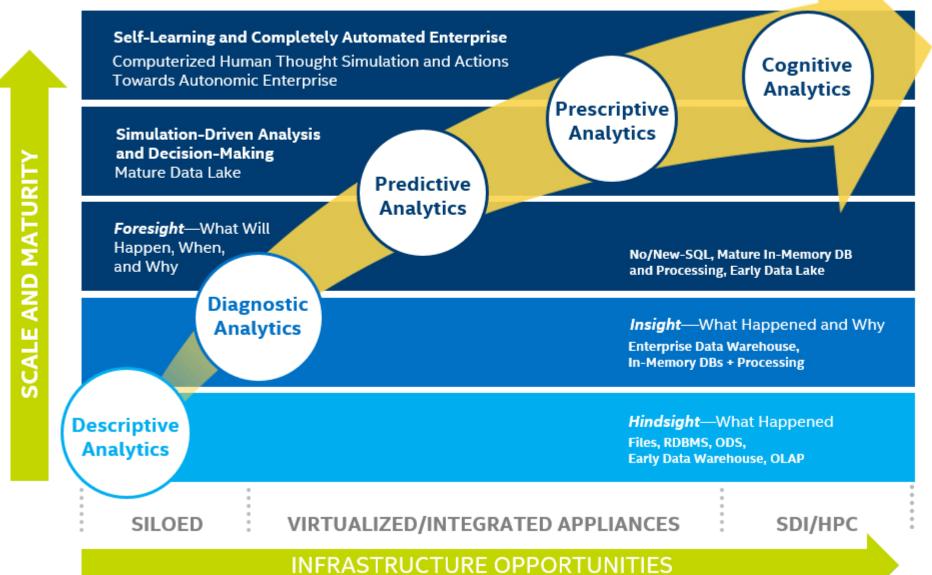


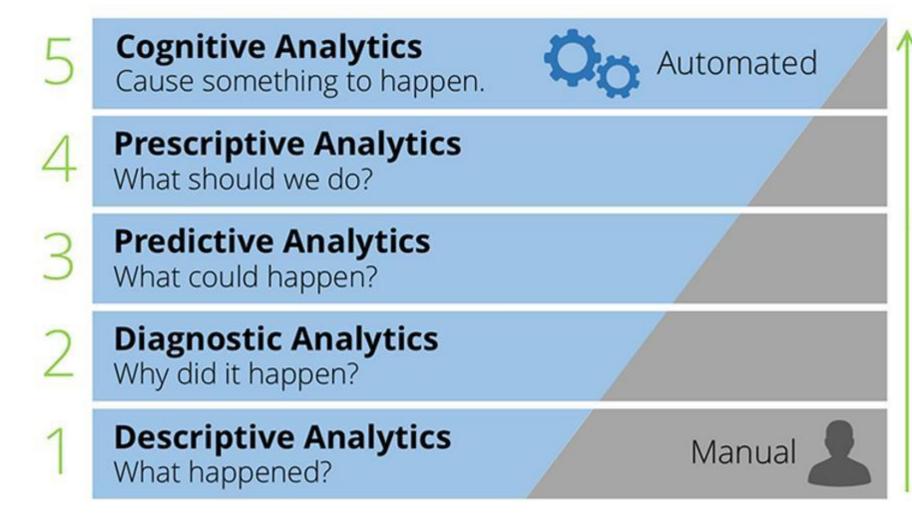


Types of data analytics











Predictive Analytics

- what the future holds (to a certain degree)
- show a variety of possible outcomes

Harvard

A Predictive Analytics Primer

Application

by Thomas H. Davenport











No one has the ability to capture and analyze data from the future. However, there is a way to predict the future using data from the past. It's called predictive analytics, and organizations do it every day.

Why is predictive analytics important?

- Lead generation
- Enhanced marketing efforts, targeted to specific customers
- Identification of future trends
- Identification of growth opportunities
- Reduction of customer churn
- Improved content marketing and distribution



LEAD GENERATION PROMOTION CONSUMER CHANNEL STRATEGY TRAFFIC POTENTIAL INFUENCE



Targeted Profiling of Customers





Improved Lead Scoring





Segmentation for Nurture Campaigns





Improved Content Distribution

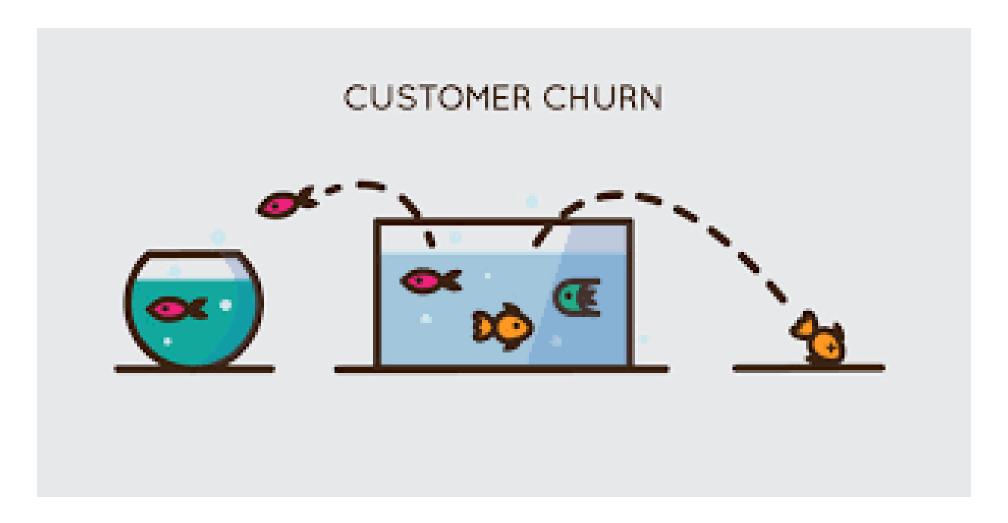




Accurate Prediction of Lifetime Value



More Insight to Reduce Churn





Enhanced Upsell/cross-sell Opportunities





Improved Determination of Product Fit

Determining the Optimal Campaign Channels & Content

Identify New Trends and Growth Opportunities

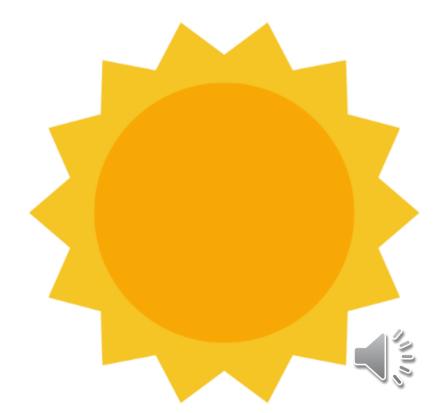


Examples of predictive analysis applications









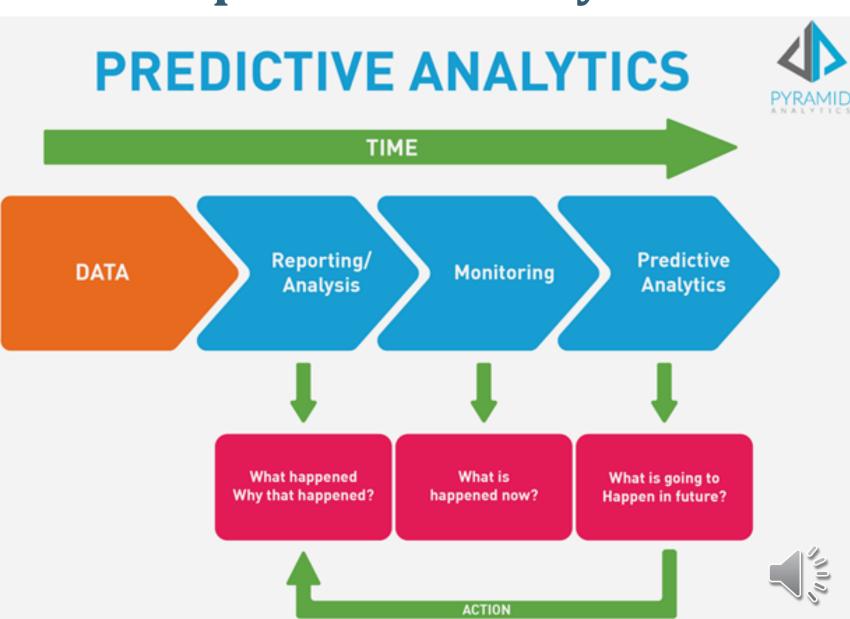
Process of predictive analytics

- 1. Project Definition
- 2. Data Collection
- 3. Data Analysis and Statistics
- 4. Modeling
- 5. Deployment/Integration
- 6. Model Monitoring



Methodologies used in predictive analytics

Logistic Regression
Decision Trees
Time Series Analysis
Text Analytics



Five Industry Examples of Predictive Analytics

Healthcare



Manufacturing



Finance



Insurance



SaaS





Improving Patient Outcomes

Problem:

Benefits:

Data to Analyze:

Actions to Take:

Healthcare





Predictive Maintenance

Problem:

Benefits:

Data to Analyze:

Actions to Take:

Manufacturing



Sensor 1	Sensor 2	Sensor 3	Sensor 4	Sensor 5	Sensor 6		State
0.0051	0.035	0.001750	0.004	1.575	0.173250		HEALTHY
0.0051	0.035	0.001750	0.004	1.575	0.173250		OK



Predicting Late Payments

Problem:

Benefits:

Data to Analyze:

Actions to Take:

Finance





Preventing Fraud

Problem:

Benefits:

Data to Analyze:

Actions to Take:

Insurance





Reducing Customer Churn

Problem:

Benefits:

Data to Analyze:

Actions to Take:

SaaS





Seven Steps to Start Your PA Project

- 1. Identify a Problem to Solve
- 2. Select and Prepare Your Data
- 3. Involve Others
- 4. Run Your Predictive Analytics Models
- 5. Close the Gap Between Insights and Actions
- 6. Build Prototypes
- 7. Iterate Regularly



Identify a Problem to Solve

PADS (Performance Analytics Decision Support) Framework

Preventing Problems:

Assisting Humans:

Detecting Problems:

Streamlining Services:

Requirements for Predictive Analytics Problems

Loss Prevention
Increase Happiness
Improve Processes



Select & Prepare Your Data

Data Volume:

Data Scalability:



Involve Others





Run Your Predictive Analytics Models

- 1. Classification Model:
- 2. Clustering Model:
- 3. Forecast Model:
- 4. Outliers Model:
- 5. Time Series Model:



Close the Gap Between Insights & Actions

Build Prototypes

Iterate Regularly



Data Analysis Techniques

- 1. Techniques based on Mathematics and Statistics
- 2. Techniques based on Artificial Intelligence and Machine Learning
- 3. Techniques based on Visualization and Graphs



Techniques based on Mathematics and Statistics

Descriptive Analysis

Dispersion Analysis:

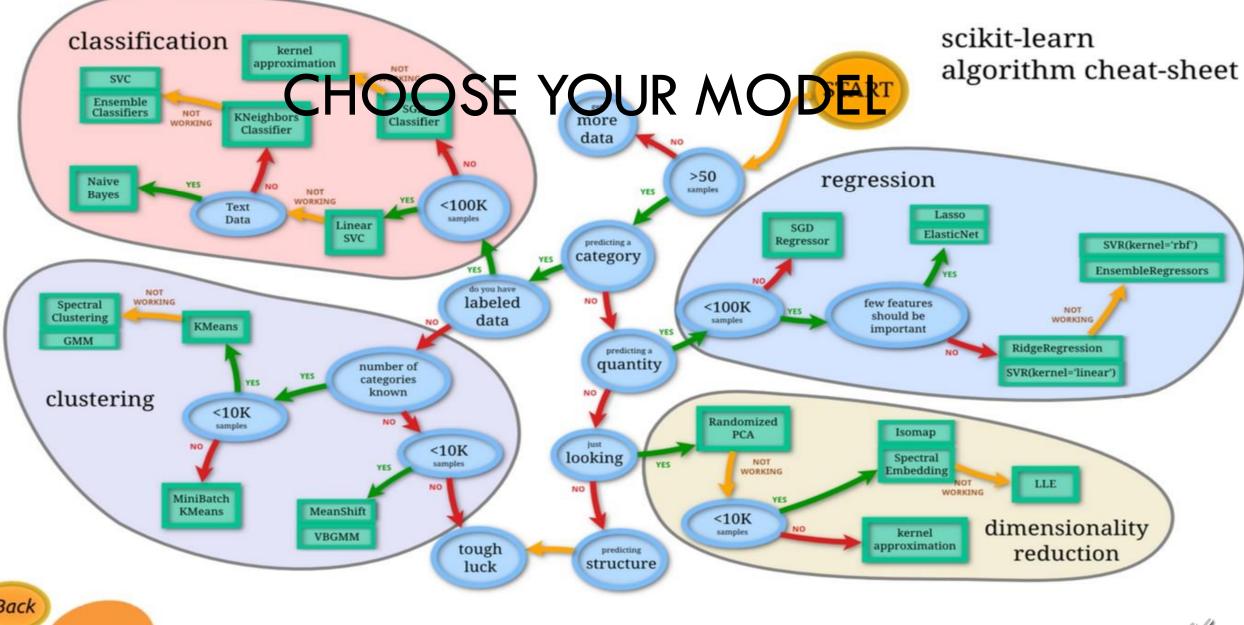
Regression Analysis:

Factor Analysis:

Discriminant Analysis:

Time Series Analysis:









Techniques based on AI and ML

Artificial Neural Networks:

Decision Trees:

Evolutionary Programming:

Fuzzy Logic:



Techniques based on Visualization and Graphs

Column Chart, Bar Chart:

Line Chart:

Area Chart:

Pie Chart:

Funnel Chart:

Word Cloud Chart:

Gantt Chart:

Radar Chart:

Scatter Plot:

Bubble Chart:

Gauge:

Frame Diagram:

Rectangular Tree Diagram:

Map

Regional Map:

Point Map:

Flow Map:

Heat Map:



Data Analysis Tools







- 1. Excel
- 2. Tableau
- 3. Power BI
- 4. Fine Report
- 5. R & Python
- 6. SAS



















Python Packages for Data Analysis









Data Visualization in Python







Python Code of the predictive modeling tasks

Load dataset

```
import pandas as pd
import numpy as np
from sklearn.preprocessing import LabelEncoder
import random
from sklearn.ensemble import RandomForestClassifier
from sklearn.ensemble import GradientBoostingClassifier
```

Identifying missing values

fullData.isnull().any()#Will return the feature with True or False, True means have missing value else Fa



```
#Impute numerical missing values with mean
fullData[num_cols] = fullData[num_cols].fillna(fullData[num_cols].mean(),inplace=True)
```

Impute missing values

```
#Impute categorical missing values with -9999
fullData[cat_cols] = fullData[cat_cols].fillna(value = -9999)
```

Checking correlation and visualization

Training and test data split

```
from sklearn.cross_validation import train_test_split

train, test = train_test_split(df1, test_size = 0.4)
train = train.reset_index(drop=True)

test = test.reset_index(drop=True)

features_train = train[list(vif['Features'])]
label_train = train['target']
features_test = test[list(vif['Features'])]
label_test = test['target']
```



Predictive models on training data

```
from sklearn.ensemble import RandomForestClassifier
clf = RandomForestClassifier()
clf.fit(features train, label train)
pred train = clf.predict(features train)
pred test = clf.predict(features test)
from sklearn.metrics import accuracy score
accuracy train = accuracy score(pred train, label train)
accuracy test = accuracy score(pred test, label test)
from sklearn import metrics
fpr, tpr, = metrics.roc curve(np.array(label train),
clf.predict proba(features train)[:,1])
auc train = metrics.auc(fpr,tpr)
fpr, tpr, = metrics.roc curve(np.array(label test),
clf.predict proba(features test)[:,1])
auc test = metrics.auc(fpr,tpr)
```

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Sometimes questions are more important than answers

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