# Big Data & Al for Industry 4.0 – Feedback and Lesson Learned



**Rafiqul Haque, Ph.D.** Chief Technology Officer, Cognitus





# Cognitus – In a Nutshell





Cognitus is a research-driven innovative solution and service provider headquartered in Paris.



It was founded by a group of computer scientists and industry experts.



Cognitus has a vibrant development team of experienced and young data scientists, Big Data architect, solution architect, developers etc..



Cognitus is a Young Innovative Enterprises affiliated by the ministry of research and innovation in France.



The core activities of Cognitus include research, solution development, training and consulting.



Cognitus has a specialist data science problem solving team consisting of scientists from top quality universities within Europe and North America.

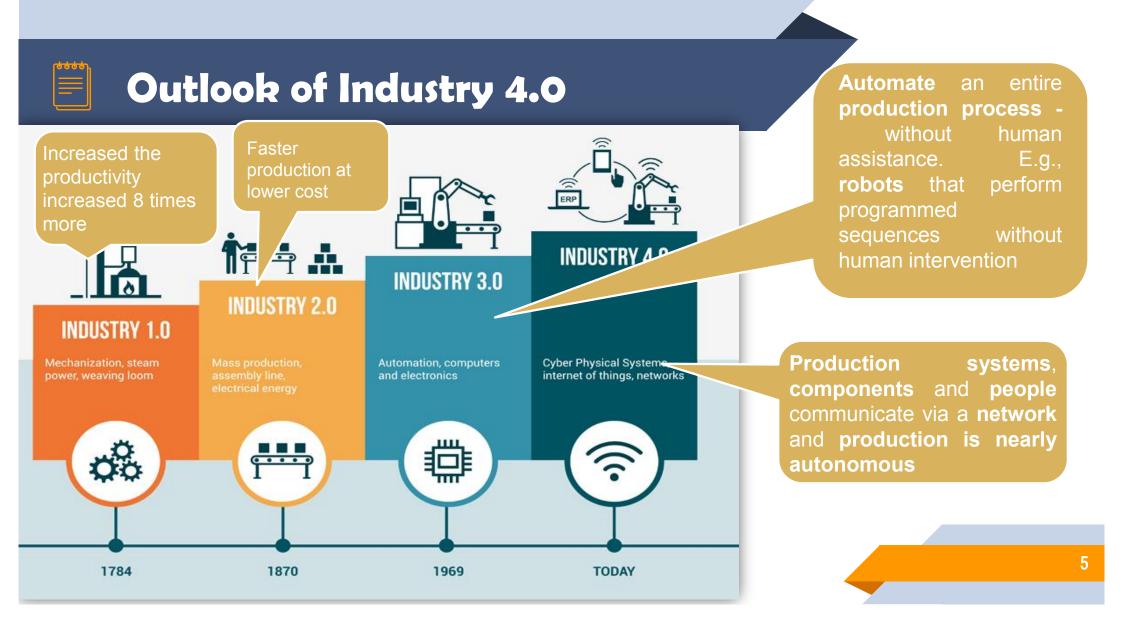


How Industry 4.0 Looks Like Today?



Intelligencia





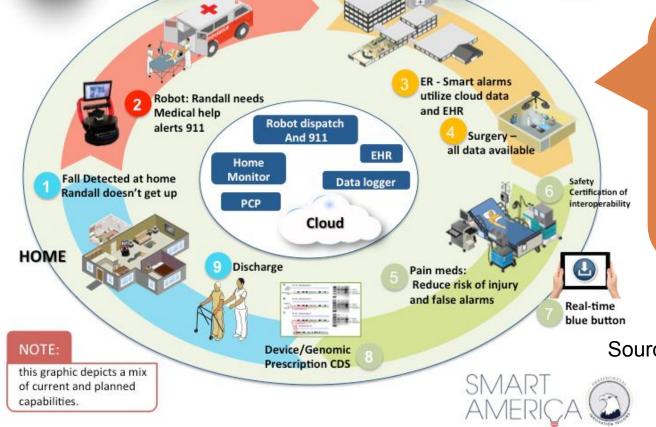


Marty, as it's called, is a tall robotic assistant that will be introduced to





### Closed Loop HealthCare Team: Home to Hospital to Home



### Healthcare 4.0:

SMART America is a research project aimed at building a connected ecosystem for healthcare services. The project consortium included the big players including NIST, Intel, Harvard University etc.









### Now & Tomorrow

Network & Services Quality of Experience, Capacity, Capability, Scalability, etc...



### Telecom 4.0:

Telecom industry has survived disruption including Industry 4.0 have pioneered many of the technologies and high connectivity (e.g., 5G), Al Chatbot) at its core, making innovation possible,









Social Network



Diverse Services, Applications, Devices, Spectrum & Deployment

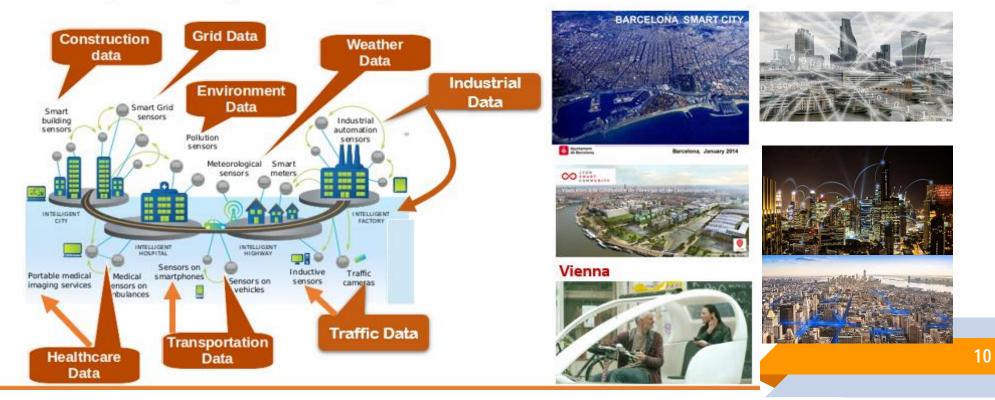








• Internet of Things is leading to a massively connected smart urban life.







2

# What is the Core of Industry 4.0

Foundation of Smart Ecosystem

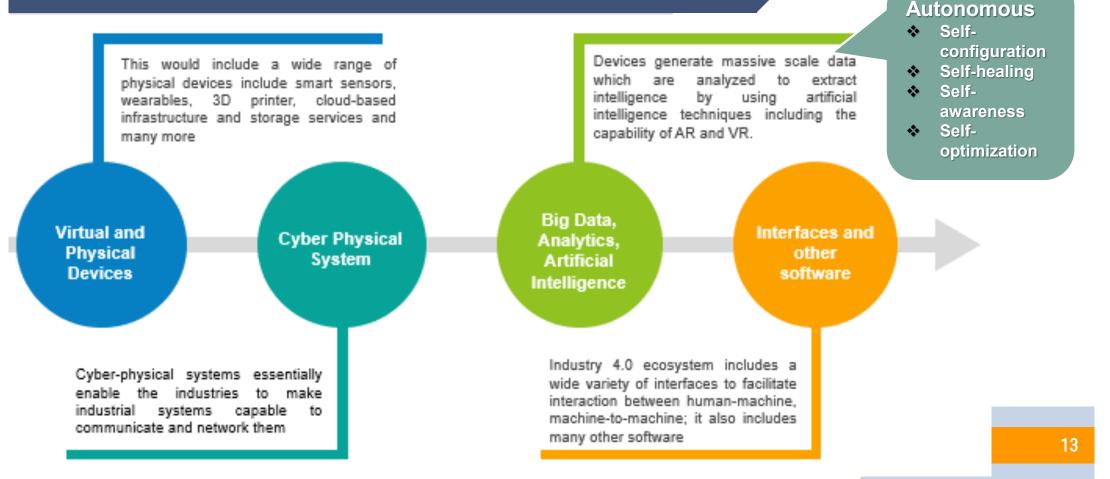




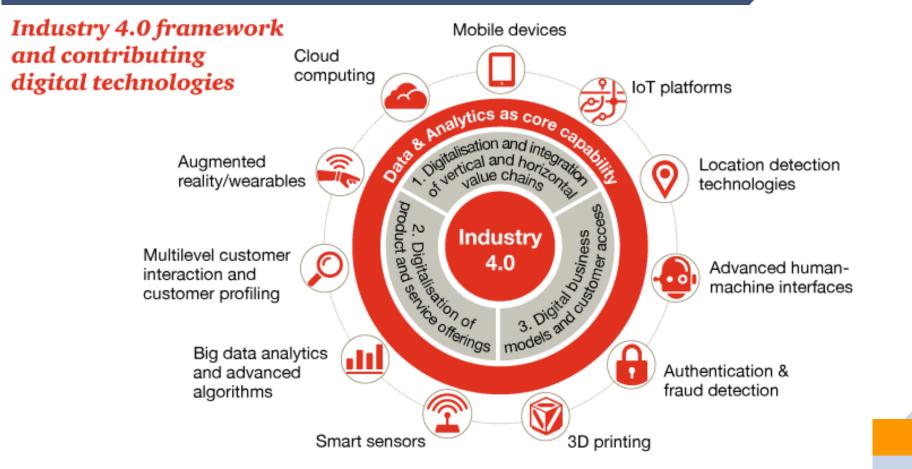
12



### The Core of Industry 4.0







14

Source: Industry 4.0: Building the digital enterprise, 2016 Global Industry 4.0 Survey, PwC, Engineering & Construction, 2016

3

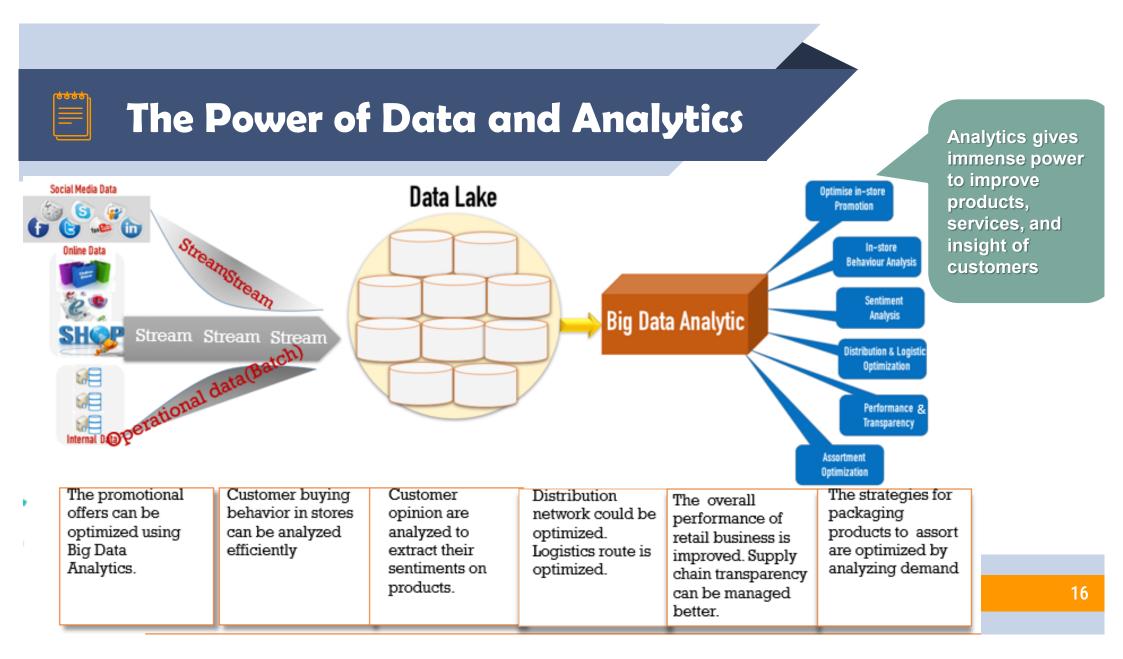
# The Power of Data and Analytics for Industries

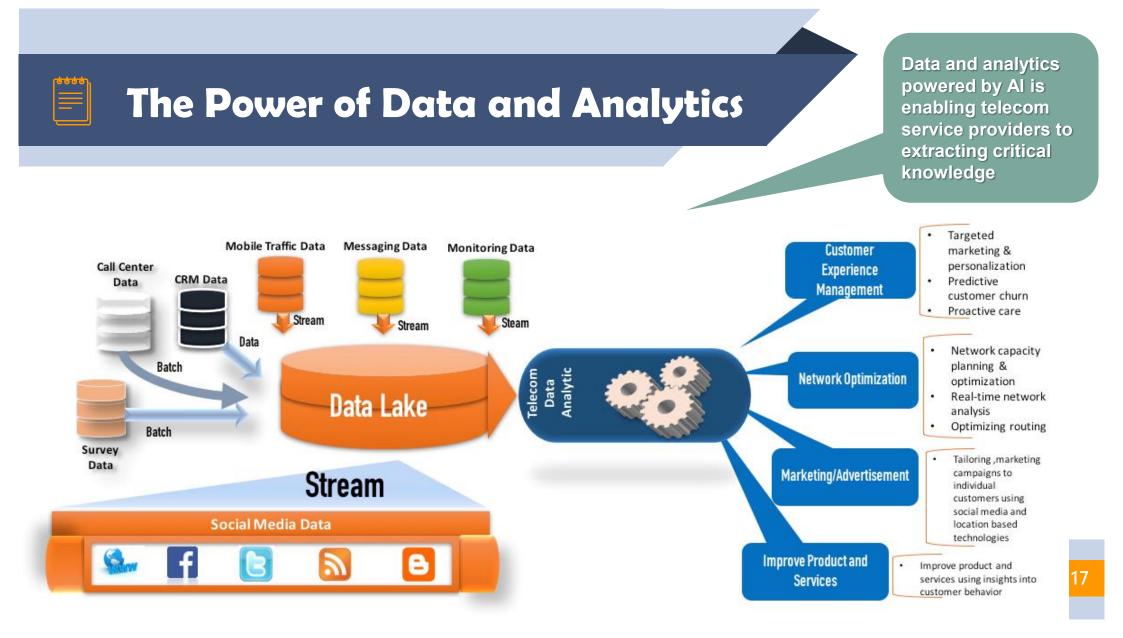
Data keeps intelligence inside

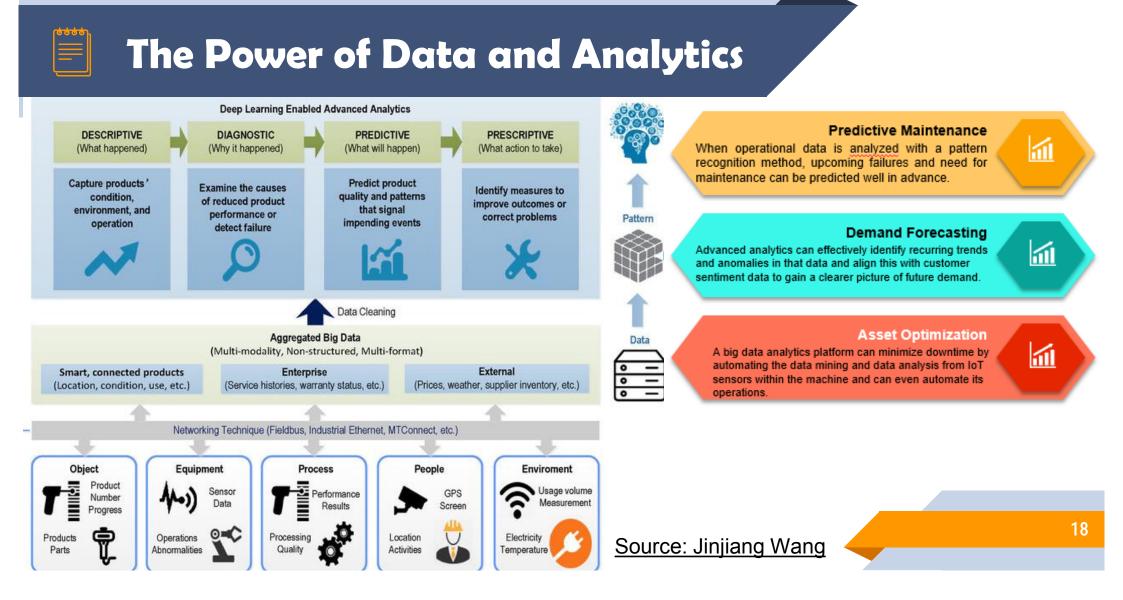












# The Power of Data and Analytics



Big data analytics powered by Al gives criminal investigators an ability to identify suspects, predict crime, realtime alert

# The Power of Data and Analytics

### **City Traffic analytics**

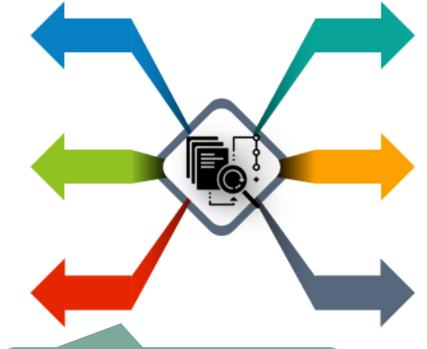
Through the analysis of data collected from transport authorities, you can study the patterns that will result in decreased traffic congestion and help transport authorities come up with intelligent ways to manage and monitor transport within the city.

### Urban Planning

The effective use of data can help in identifying areas that need improvement and upgrading.

### **Budgeting and Spending**

Using big data analytics, data collected in a smart city can suggest the majorly impacted areas and what type of upgrades are needed. Through proper analysis, investments can be made accordingly in the required fields.



Big data analytics powered by Al gives criminal investigators an ability to identify suspects, predict crime, realtime alert.

### **Future Proofing**

- Data collected from various sources can be utilized to provide a sustainable environment with higher energy efficiency and less wastage of resources.
- Through predictive analysis, it's possible to analyze the growth of current infrastructure and plan for future needs of the city

### **Public Security**

Predictive analysis can be of help to study historical and geographical data to recognize when and where crimes are likely to happen. A significant amount of improvement will be seen when the desired data turns a city into a much safer place.

### Quality of Life

With better efficient work, services and living models, smart cities will have better Quality of life. The result of location and living/work spaces, more transportation for better and faster services and enough availability of information to make decision.

# 3

# Some of our key projects

Extract value from your data

# Customer 360 Analytics

-	-					-				_
1	Cu	sto	mer l	Profile Analysis		<b>@</b> ``	FULLY	NTEGRATE	ANALYSIS	6
THE NAME I	Lett Name 1	April	Marital	jan i	Residence 1	Contact I	Active 1			_
hri.	Smith	24	Single	Therapic, schapetone	K20 Levels Summer	42,879,958	1	Filters		
ten.	Smith	40	Single	Accountant, chartered	478 Schulz Drive Subr 111	87.542.433				
hn	Smith	45	Single	Adult nume	1234 Clark Prairie	38.345.554		Viller Commission, Na	State True Yarra	
5yn	2040	47	Warred	Technical lates engineer	151 Perinargitan Trakkapi, \$74	76374,533		Salart		~
ten ten	Aguine	41	Deprind	Sel designer Product designer	00715 Roberts Creek Suite K32 585 Robert Springer, Apt. 580	\$5.021.915	-			-
NO	Agurre Marris	55	Decried	Product Bengher Ruthfunk dengher	SBS Rubert Spregs Apt. SBS SSS17 George Marce	3.091.723	-			Section.
nen.	marin	- 56	Single	Toranial plannet	125 Milegfvin Route Apt. 824	82,792,127				
hati	Abbett	47	Denied	Mantenance engineer	CMP1 Meurism Lancton Apr. 205	47,246,329	1	Filter California No.	their Loci Norice	
het	Alarra	-	Tingle	Patheries officer	22753 multifalts	16.671.642		telet		
htud	Lopey	36	Married	Tourist information centre manager	NUM Landa Renty	31,776,042	1			-
lichael .	Loper	45	Married	Plannabil, heightst	188 Marris Corner Salte 835	\$3,044.738	4			1
ichael .	Laper	42	Dupried	Hospital disclor	RE2 Trave Squares	1174LND	1			
(chard)	Dart	24	Married	Engineer, building services	0213 pow stat bute N22	63,348,547		Titles Californian by	the electron and	
Actual .	Devit		<b>Decried</b>	Optimizing of	299 Copeland Mountains Apr. 518	15.636.737	4	later.		~
Not performent	menuruha	38	Incred	Commercial residential surveyor	8054 Campbell Lodge Apt. (IN6	24/35/852	3			-
shiel	Hernardes	29	Marred	Physiological sciences	241 Juni Bridge Suite MIL	\$7,345,859				11111111
1.0 eff	Smut	40	Owned	Adviz worker	9070 Mary Lodge	41305314				
6ch peli	Smart	97	married	righer education careers adviser	0517 Patrik Hiel	BEARLEY	.1	The General Ig	the brider	
ACT AND	Adams	28	Decented.	Research science (Hattis)	Root Halt Faids	7.577.695	1	Select-		
aviti .	Adams	42	blarted	Parring and development surveyor	2918 Kayle Turinel Suite 814	85.836.376	3			
and .	Areford	42	Desired	Voketteer coordinator	26809 Sydney Light Suite 379	13.776.211			Calvet Compta	AUXINE
and .	Berritt.	44	Single Married	Architectural technologist	51521 Ravet Parks	1.945.230	-			
and .	Berner	57	Traja	Engineer, petrolikum Comptroliker	40407 Becky sacks Suite 700 4040 Store Island	84,273,318		Time Laboration In	Call Starting Street, Spice	
ubari.	Anderson .	-	Decent of	Patarate	483 Zachary valley Apt, 117	25,205,514				
ubert.	Anderson .	45	Married	Inster mail	728 brandi Rest	36.112.291	1			
abart	Automate	-	Seals	During general graphics	and in Date Sector	85156281				Auto 174747
(Bell	Littler	-	Married	Abovulunt	URABLE Dana Crest	74.636.617				-
tine's	side	47	Marraid	Teacher: Erghit as a foreign language	Salid Michael Overpant Sole \$73	14465311	4	Film Company in	the local dataset	
deri .	Inhedon	40	Married	Designet textile	605 Owyerne Camer Apt. 917	34,886,417		Sale 2		10
iken .	Robertson	- 91	Invited	Perestrementant manager	SARTI Struct Troughusy Sully 482	9476414				
diert.	Balar	34	Segn.	Intelligence analyst	10171 Rule Mountain Apr. 088	85.840.852	1			Attent
dert.	Brook	49	Matted	Psychologist, simplement	148 Cessendra Likop	8.147,457	4.1			Concernance of Concernance
flam	Millel	48	Single	Event organizar	04285 Variesia Esteriouria Sulle 719	8,835,625		Piler Lameters In		
tige.	sechel	52	Marred	Sports theraped	AD1 Peters Purt	94747,854		Viter Commence in		
than	Peterseti	47	Single	Programmer, cytterio	1381 drywn Miladow	95,306,402	3			
than	Petersen	45	Desired	Provide Science	6211 Nuthule Perty	01473431				
rilan	Tutlet	- 67	Detroited	Minerals surveyor	20005 Mercley Camp	47,722,585	1			Charlenge of the local division of the local
than	Tucker	67	formed.	Litrarian, public	0582 Angela Locks Apt, 871	12,545.425	1			
than	<i>Abspies</i>	17	Married	Surveyor, sural practice Pharmacent Assurat	5057 jenária Ports. A4104 Aloha Hillo	37,298,526		Pillar Cylinamia Ny	040 pta 9310	
lian .	Anderson.	-	Married	Pharmacitt, horgital Textile designer	adribit Abbria Petto 2013 Argenia Externation Apri, 415	45.039.392	-	•		
and the second	Read	-	Decree .	Textus designer Playing and development surveyor	203 Argena Extension Apr. 815 39130 Schaefer Perry	28.653.927	-			
heldinghar	Berry .		Seeps	Playing and development surveyor Aronal numbered	29100 Schweler Perry 87544 Merch Rapid	44,435,355	-			Antiliant
in the second	Bearing Braddhaw	37	Sarred	Caturnational	875ad Marith Rapid 87583 Paters Point Apr. 371	4213.632	-			
and a second	Bridges	-	Married .	Catometrial Proemacologia	107242 Perses Punct Apr. 371 262 Masew Common	4103.001	-	The Loneners is	their age	
noter	Garca	-	Single	HURSA HERITATI	1543 Miller Turrgile Suite 442	BAUTARIS		•		
notes .	Garca	1	lingle	Visal inclusion	B43 mines Brooks	33,490,310	-			
ereder .	Librar.	÷.	Decent.	Sensitive of Sets Sensitive	at2 from field	24.801.616				Acres Server
				Margaret Margaret	http://www.intension.com	21.212.214				-



<b>6</b> 0	redit	Risk	Ar	alysi	is				<b>⊘</b> _^	FULLY	INTEGRATE	D ANALYSIS	6
Viji hurbat of tadamet of	v rogʻi Laur	a 198 Av 769 T		1,7	67	46	9				The College of St	is there that have	
and the state of the local state of the	1000										-		*
<b>0</b> 1	30-25 days	49-49-6155 Table 1	Ni deg	n Dabela	ie Crelitineskoti	ers Referre	aandriins Kanbe	O'Dependential Mar	-	out Level			and the set
01100-055-055-089- 435/54/561	3	1	1	- 14	÷			3.0					
401/54/104/55/4/5 (021/857/4/104/75/54/5 (922/52/104/75	1			1.21	.9			1952			Phar Coloma v A	their control basiss	
ADVECTOR (2014) 1141- ADVECTOR (2014) 1141-	1	2		0.24	12			4.19			defined as		
COLUMN ATTACASE. allos.				6.16	12			1.00					Assessment
425462154024 00073151 6400-4678-4703	3			2.9			4	1.04					
(0030ad 4757 423a 8dta-				1.71			4	-			other California		
R10736A3		*		- 10							Pher Lamonari I.	the beau	5
00123000-0007-4890-01984				1.29				249					-
Annual States													in the second
Year in sector	-	10	-					1.0					
MSTADA543 (SCIALSE TRAT 4004-Res)	-	2	-	1.07		-		4.00			The Louise is		
ASCHATISTICS (Delarith Tube abarchitt)	-		-			<u>.</u>		10			see.		
25cs4402CMS 00ex0403 7360-4555 57ea	-		-	141	u.			3.56					En la compañía de la
Decision and access of	<u>.</u>		-		4								
957521985640											10x Lummer		
erenetiko betrikogi 4021. attosogiatzee	*		1	11.42			10	6.00			Alter Compton of	-	
011125-03-0423-4440-4075- 4010024010e	1		*	1.64	10	3		.19					
01(20082-0540-4456-5275 55059011cab	20	A						6.00					(Accessed)
010092782418.H05	5	3		0.61	18	- Const		1.00					
ferme(1291					- maging - m	otcato	,381	46	an court of	Owept Se	86	Number of Costs	ung ben
					- 944 Te 11878	437/54 002100 404x20 003000 405/27	0.4155-415-5465- 44500 10.5451-8545- 10.54 10.5455-4555	ittatury i (1.4 15.7	4-10-94-00 4-10-94-00 10 23 0 0		Liens (	Problems 1	Contract of an Contract of 1



# **Customer 360 Analytics**

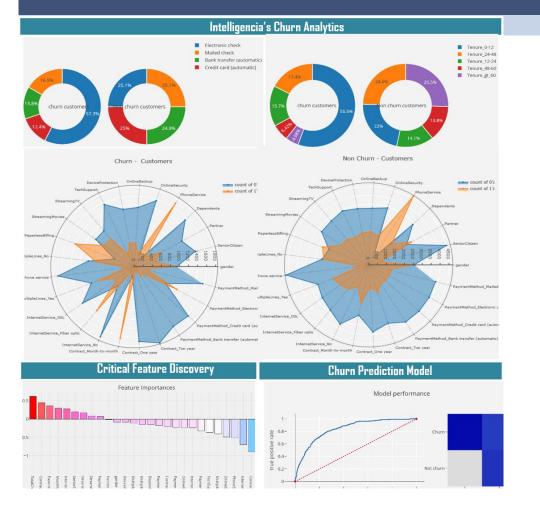
Customer	Complaints Analysis	INTEGRATED ANALYSIS	Loan Activities	` <b>@</b> ```	FULLY INTEGRATED ANALYSIS
San	And	Files Files Persona years Persona	And leads to the lead     And many     And many     And many       And leads to the lead     And many     And many     And many       Ministration down and many     And many     And many     And many       Ministration down and many     And many     And many     And many       Ministration down and many     And many     And many     And many       Ministration down and many     And many     And many     And many       Ministration down and many     And many     And many     And many       Ministration down and many     And many     And many     And many       Ministration down and many     And many     And many     And many       Ministration down and many     And many     And many     And many       Ministration down and many     And many     And many     And many       Ministration down and many     And many     And many     And many       Ministration down and many     And many     And many     And many	Gratter - Grand Lange Control - Research Control -	mandalis Inst. v Inst. v Card the Card darge Analytic The Card the C
Dosed with explanation - C	4 996 3350 42 representation of the second	BBI Dashboard BBI Dashboard Cathornes Sentiment Analy Cathornes Sentim	Firm		

- Deep Integration of Data: The Data Integrator • combines and enriches data about the customer from social channels, customer feedback, customer service centers, and others.
- Deep and Wide Analysis Customer Behavior: • Hawk-I provides and predicts customers buying or consumption behavior.
- Customer Interaction: Hawk-I visualizes ٠ through different customers interactions channels.

Customer preference through market-basket analysis: Hawk-I extracts customer preference between products and services, a pair of products, a pair of services, etc.



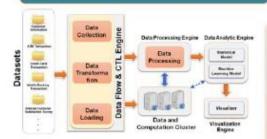
# Customer Churn Analytics



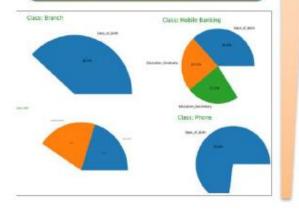
- Data Manipulation: Manipulates data such as dealing with missing values, transformation, separating numerical and non-numerical values, and separating churn and non-churn users.
- Data Preprocessing: Enables pre-processing data such as level encoding, scaling numerical data, merging scaled values for numerical data, duplicating columns, etc.
- Data Preparation: Enables performing data preparation tasks such as dimensionality reduction.
- Exploratory Analysis: Enables users to explore data such as variable (e.g., senior citizen, dependent, partner, phone service, internet service) distribution and discovering correlation matrix.
  - Churn Prediction: Discover the essential features of products, services, and customers and identify all potential churn.

# Customer Journey Analytics

### Banking Use Case of Intelligencia's Customer Journey Analytics



### The customer journey analytics solution finds hidden factors that influences customer to use specific channels .

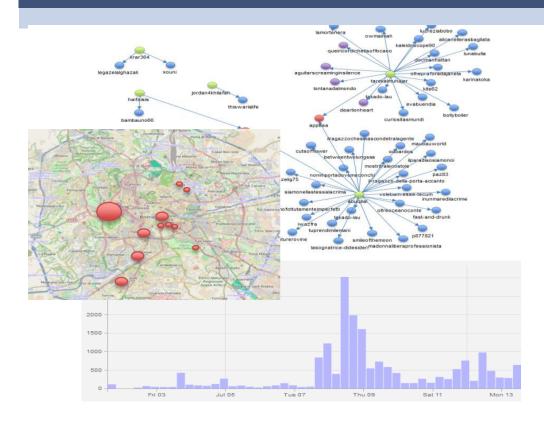


# Our customer journey analytics answers various questions (shown in the figure bolow)



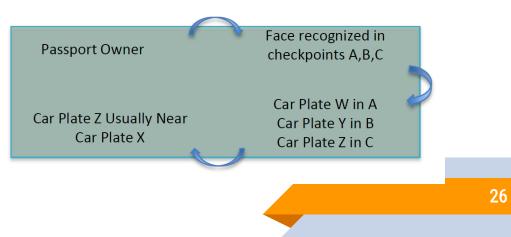
- **Powerful Data Processing Engine**: The data processing engine can aggregate, cleanse, wrangle structured and unstructured data.
- **Statistical Scoring Model**: The analytics visualizes a percentage of customers using specific touchpoints/channels.
- **Customer Segmentation**: It can segment customers over the usage of touchpoints.
- **Causal Model**: It enables users to perform causal analysis precisely why a customer uses a touchpoint.
- **Comprehensive Visualization**: A powerful visualization engine is integrated to communicate with results comprehensively.
- **Scalable**: It is a highly scalable solution and hence can accommodate and analyze any volume of data.
- **Customizable**: It can be customized based on customer requirements and also based on target application domains such as Banking and Telecommunication.

# **Deep Link Analytics**



Powerful correlation analysis

- Geographic relations
- Time based relations
- Relations among different objects (Vehicles>Owners>Faces)
- Geographic Anomalies
- Time Anomalies





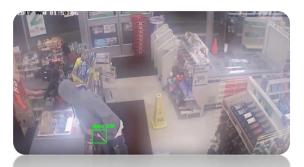
# **Criminal Intelligence 4.0**



Smart Glass: Facialrecognition glasses that can identify suspects within milliseconds.



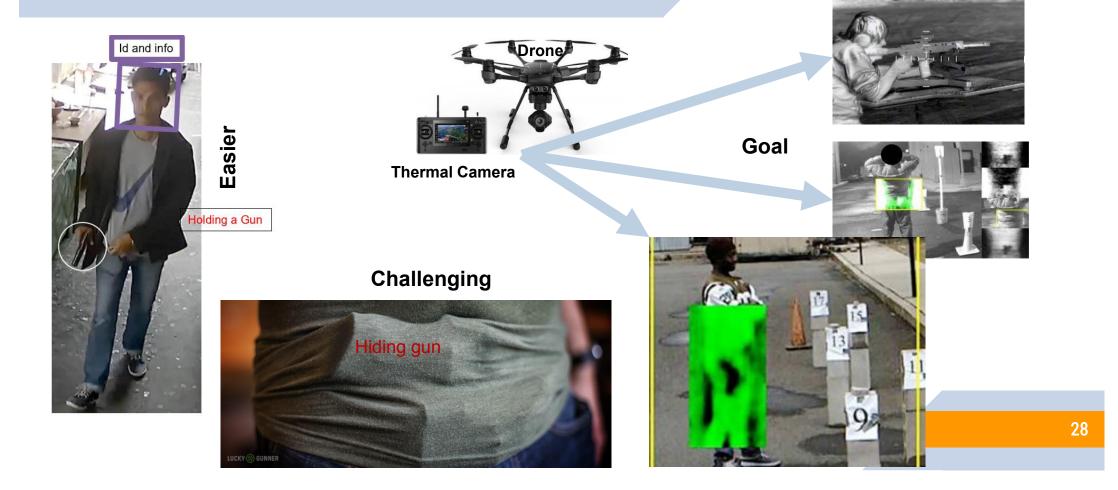
Real-timelicensenumberandplatedetection:It can detectthelicenseplaceandrecognizethenumber.



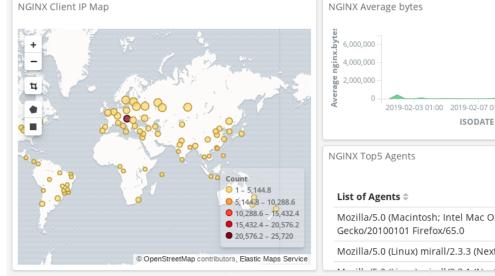
Real-time hazardous object detection: It can identify person for access to a building, detect a weapon as it's being wielded



# Criminal Intelligence 4.0



# WebServer Log Analysis



43,238 19,611 16,644 7,960 4,772

2.699

Missing

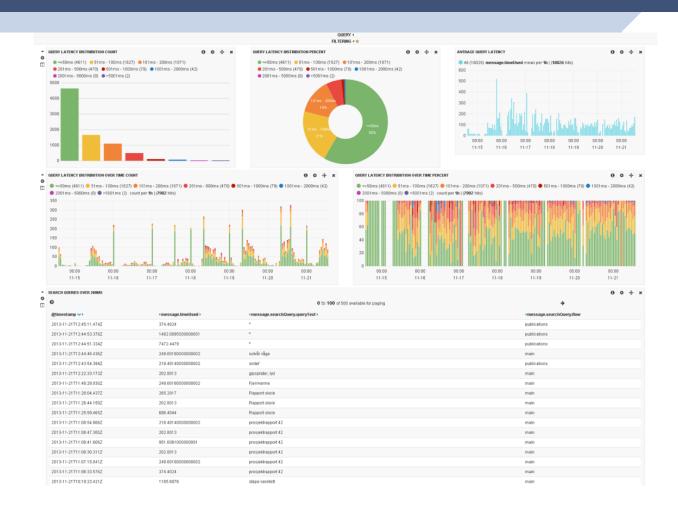
	NGINX Average byt	alysis		NGINX TOP 10 Response codes
	Average agin, by the second se	03 01:00 2019-02-07 01:00 2019-02-11 01:00 ISODATE per 12 hours	3	
ount 1 - 5,144.8 5,144.8 - 10,288.6	NGINX Top5 Agents	5	Count ¢	
10,288.6 - 15,432.4 15,432.4 - 20,576.2 20,576.2 - 25,720	Mozilla/5.0 (Maci Gecko/20100101	ntosh; Intel Mac OS X 10.14; rv:65.0) Firefox/65.0	)) 13,791	
s, Elastic Maps Service	Mozilla/5.0 (Linux	() mirall/2.3.3 (Nextcloud)	13,166	
NGINX Top10 HTT	Puser	NGINX Top5 Requested URLs		
Username ≑	Count ≑	Requested URLs 🗢		Count ≑

NGINX Top5 404 resp. / IP	
Client IP ≑	404s ≑
172.20.0.1	497
111.231.	356
123.207.	356
47.98.	355
106.13	355

1	Requested URLs ≑	Count ‡
	/oc/remote.php/dav/files.	7,012
	/oc/ocs/v2.php/apps/notifications/api/v2/notifications	5,959
	/oc/remote.php/dav/files/	4,603
	/oc/remote.php/dav/files/	4,161
	/oc/remote.php/dav/files/	2,389



# Query Log Analysis





4

# The Reality About Big Data & Al Projects

Let's Face It

# The Reality About Big Data Projects

### • The Reality About Big Data Analytics Projects

- July 2019: VentureBeat AI reports 87% of data science projects never make it into production
- Jan 2019: NewVantage survey reports 77% of businesses report that "business adoption" of big data and AI initiatives continues to represent a big challenge for business. That means 3/4 of the software being built is apparently collecting dust. Ouch.
- Jan 2019: Gartner says 80% of analytics insights will not deliver business outcomes through 2022 and 80% of AI projects will "remain alchemy, run by wizards" through 2020.



# The Reality About Big Data Projects

- The Reality About Big Data Analytics Projects
  - **Nov. 2017**: Gartner says 60% of #bigdata projects fail to move past preliminary stages. Oops, they meant 85% actually.
  - Nov. 2017: CIO.com lists 7 sure-fire ways to fail at analytics.
     "The biggest problem in the analysis process is having no idea what you are looking for in the data," says Tom Davenport, a senior advisor at Deloitte Analytics (source)
  - May 2017: Cisco reports only 26% of survey respondents are successful with IOT initiatives (74% failure rate) (source)



ſ	U	U	o	U	h
	=			=	
	Ξ				

# Key Reasons Of Failure

### Poor Communication

Poor communication is the primary contributor to project failure one third of the time, and can have a negative impact on project success more than half the time

### Leadership troubles

Harvard Business Review indicates that a data strategy helps organisations "clarify the primary purpose of their data and guides them in strategic data management." Astoundingly, according to management consultants McKinsey, 30% of organizations have no data strategy"



### Lack of Skills

The lack of skills in organisations contributes 30% of the failure. This affects or takes effect on several level:

- not having the digital leadership mindset to drive strategy
- Line managers not understanding the data they have within them
- Rest of the company, not understanding the lingo of analytics

### Ambitious intentions

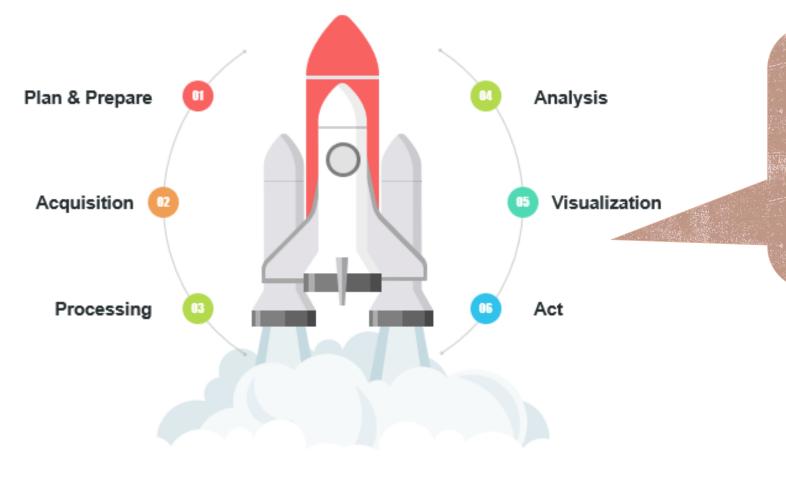
Nearly all companies that embark on becoming data-driven organisations or digital transformation initiatives are too ambitious. They either spend millions of dollars on infrastructure or claim a framework for analytic or digital transformation that might not be wholly sustainable or stable



5

# What Lessons We Learned? The Truth

# What Lessons We Learned?



You continuously learn about each phase of the Big Data analytics journey driven by the data science and Al spaceship.



# What Lessons We Learned?

### Plan & Prepare

Planning and Preparation are Critical!!

- Understand and Prepare Business Goal
- Prepare Good Business Cases
- · Plan your platform

### Communication/Collaboration

Strategic Alliance with Executive Stakeholders Data analysis done right is not about technology. It's about business. Before you start any big data analytics project, you first need to secure the support of the company's executive stakeholders.

### Acquisition

Focus on Relevance Rather Than Quantity Don't bother with huge datasets simply to indulge your clients. Use relevant data samples instead – results will be the same, while costs will be much lower.



### Processing

### Focus on quality

Data that's inaccurate, poorly formatted, or obtained from dodgy sources can kill even the most carefully planned big data project

### Analysis

Choose Your AI Horse Carefully You can simply impress your audience and add a unique zing and appeal to your Presentations. Easy to change colors, photos and Text.

### Visualization

Choose the right pattern for comprehensive Communication

You can simply impress your audience and add a unique zing and appeal to your Presentations. Easy to change colors, photos and Text.

# What Lessons We Learned?

### Data quality - non negotiable

"[The classic phrase is] s\*\*\* in = s\*\*\* out, but with artificial intelligence it is even stronger because it's s\*\*\* in = total mess out

### Manage expectations

"You must manage expectations. Many people talk about artificial intelligence without really understanding what's behind it, It's a very wide domain – you have image recognition, natural language processing, machine learning etc..

### Always understand your artificial intelligence

"You must [always] understand your artificial intelligence. I mentioned the black box – if a company is promising you some magic with a black box, don't believe them. You are going into a dangerous area because you have really to be in control of the [intelligence].



### Both humans and machines are needed to deliver the best result

Your AI always gets smarter with the help of human intelligence.

### Don't fall into the one tool to rule them all trap

If you buy a hammer, you want everything to be a nail. Companies want to buy a single tool for the job.

### Transfer learning can kickstart machine learning efforts in organizations

Transfer Learning is here, and it's fantastic. Companies can shortcut the process of developing algorithms by using a model that was trained for a specific task as the starting point for developing a new model for a different job.