Big Data in the Entreprise: Lesson Learned

(in french speaking Switzerland)

Alexandre Masselot OCTO Technology Switzerland @OCTOSuisse @alex_mass Geneva, septembre 14, 2017



Big Data in the Entreprise: Lesson Learned

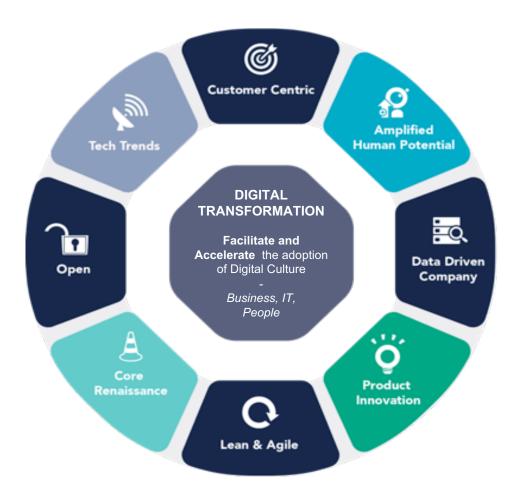
(in french speaking Switzerland)

Alexandre Masselot OCTO Technology Switzerland @OCTOSuisse @alex_mass Geneva, septembre 14, 2017



WE BELIEVE THAT *information technology* TRANSFORMS COMPANIES WE KNOW THAT THE greatest archievements ARE THE RESULT OF **shared** KN@WLEDGE AND THE JOY OF WORKING +OGETHER WE ARE ALWAYS on the lookout FOR BETTER WAYS OF WORKING

OCTO, DIGITAL TRANSFORMATION ACCELERATOR





BIG DATA @ OCTO : THE NUMBERS

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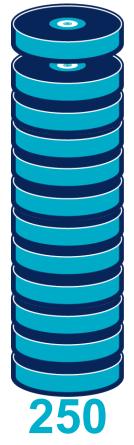
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TB, the biggest volume of data analyzed by **OCTO's data** scientists

Is the number of **Big Data projects at** OCTO in the past 12 months

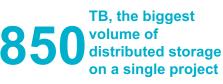
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The number of active partnerships with major Big Data actors

BIG DATA @ OCTO: PUBLICATIONS



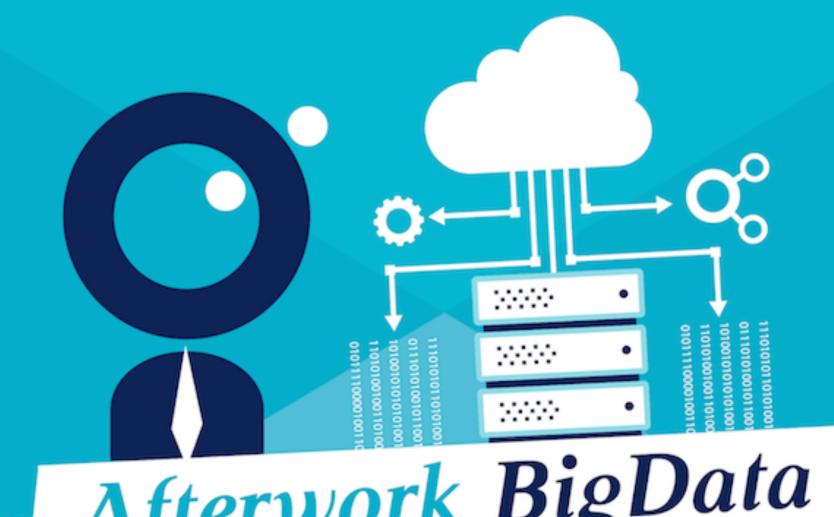


OCTO Folks Work Hard, Play Hard

1 st	۲	Caisse de dépôts - score de délivrance d'un brevet européen	datascience .net
2&4	۲	Argus - prédiction du prix de vente de véhicules d'occasion	datascience .net
3 rd	۲	SNCF - prédiction de la fréquentation des gares en lle de France	datascience .net
6 th	۲	Imperial College London - Loan Default Prediction	kaggle
13 th	۲	Allstate – purchase prediction challenge	kaggle
2 nd	۲	Tradeshift – Text classification	kaggle
5 th	۲	Microsoft - Malware classification	kaggle

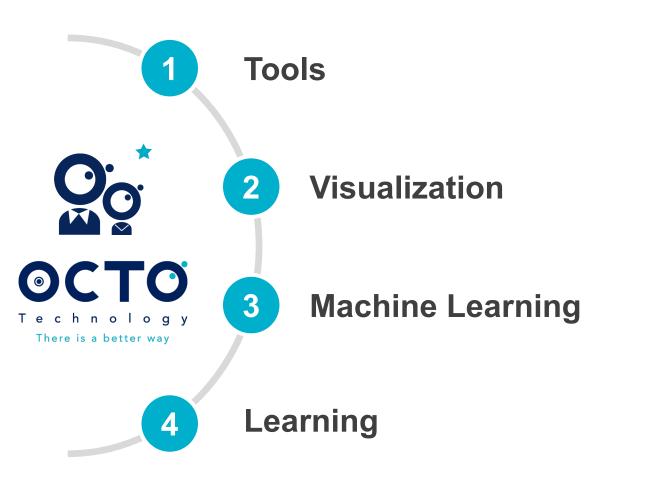
OCTO, there is a better way to learn, recruit and have fun!



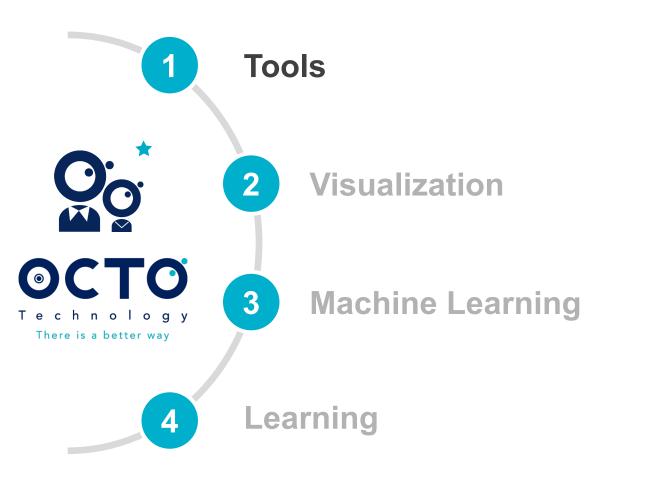


Afterwork BigData

FOUR PILLARS OF BIG DATA

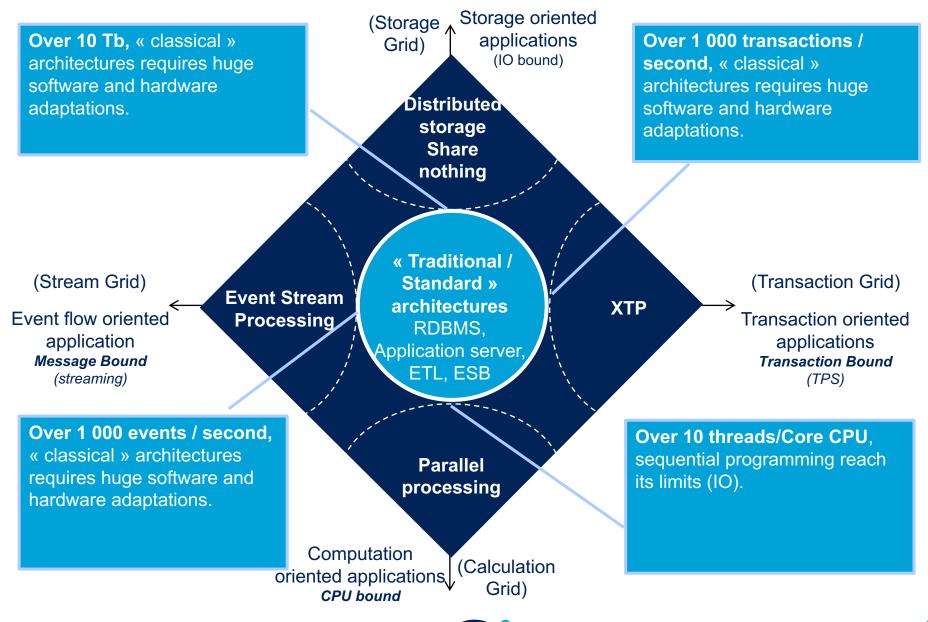


FOUR PILLARS OF BIG DATA



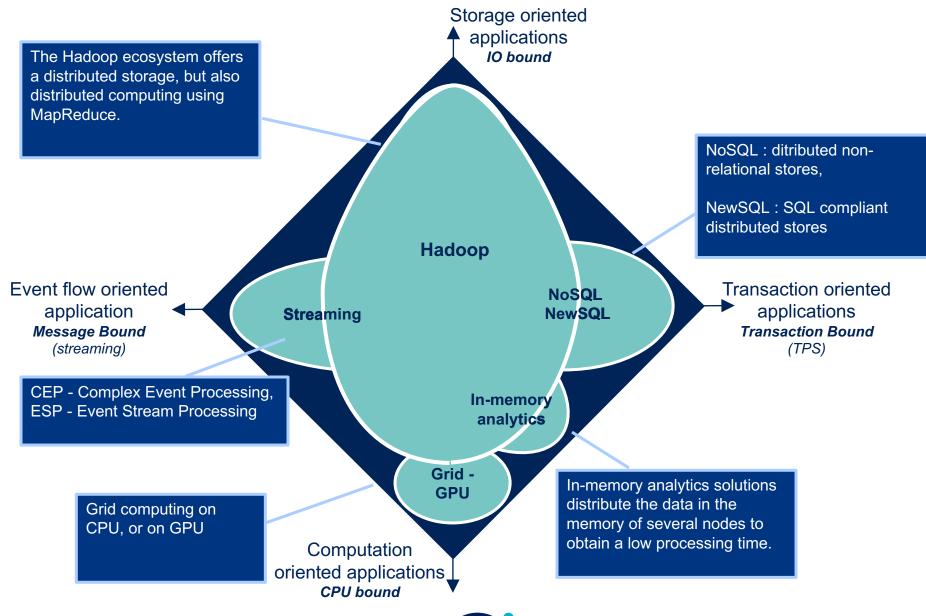


LIMITATIONS OF TRADITIONAL ARCHITECTURES

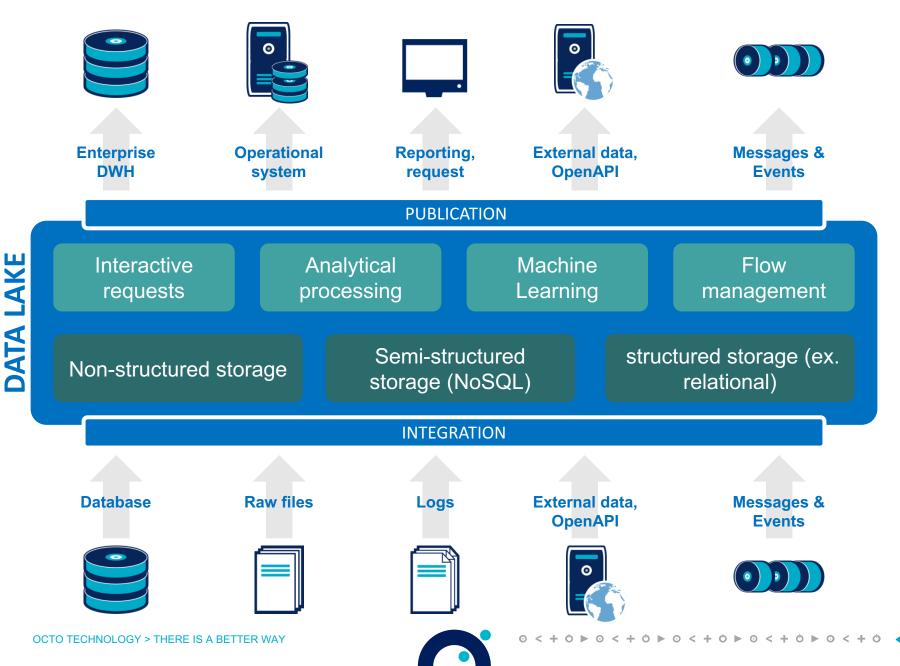




BIG DATA - EMERGING FAMILIES



ANTI-PATTERN #1: "I want a Data Lake"



ANTI-PATTERN #1: "I want a Data Lake"



You do not need to store/compute petabyte of data...

Business

JUL 1, 2016 @ 10:00 AM 921 VIEWS

Is Your Data Lake Destined To Be Useless?

TERADATA Big

TeradataVoice Big Data to Data-Driven Insights FULL BIO V



By Stephen Brobst

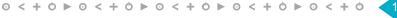
Gartner has predicted that through 2018, 90% of deployed data lakes will be useless. That word "useless" should grab your attention. It's worse than a "failed



From Datalake...



... to Dataswamp



ANTI PATTERN #2: THE LOTTERY

BIG DATA LANDSCAPE 2017

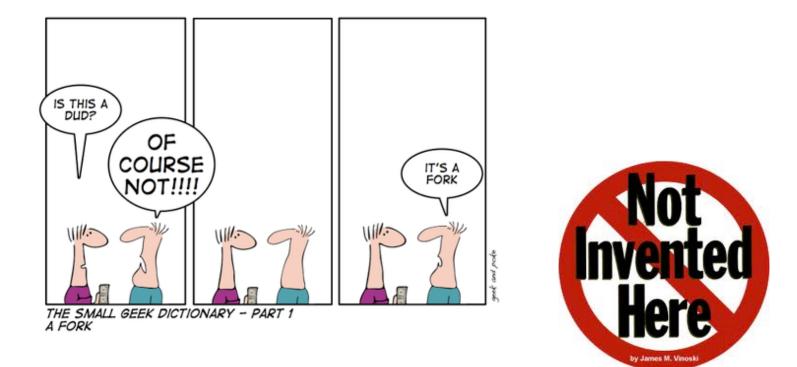


V2 - Last updated 5/3/2017

© Matt Turck (@mattturck), Jim Hao (@jimrhao), & FirstMark (@firstmarkcap) mattturck.com/bigdata2017

FIRSTMARK 📂

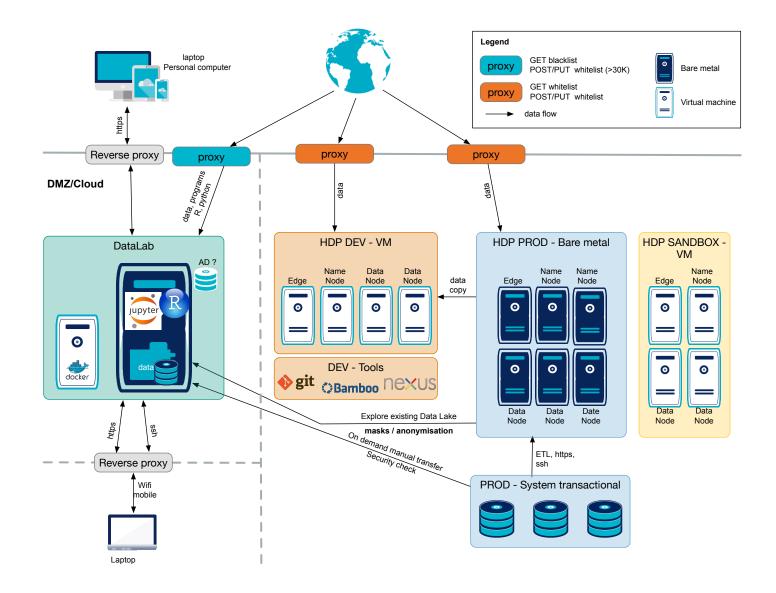
ANTI PATTERN #3: FORKING THE SOLUTION



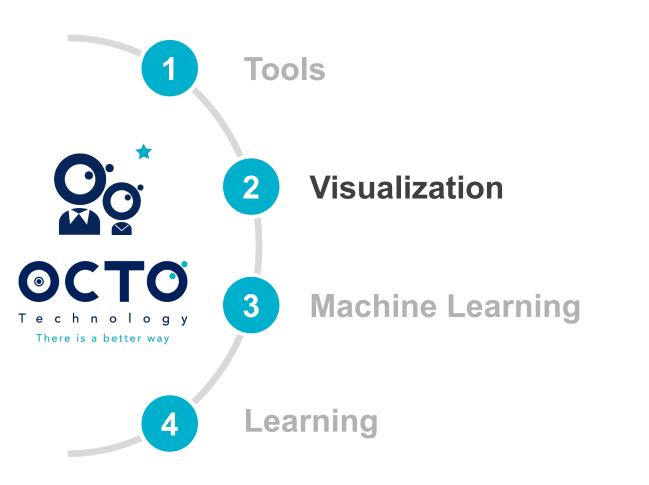
"The NIH syndrome (Not Invented Here) is a disease." Linus Torvalds



ANTI PATTERN #4: OVER ARCHITECTING FROM DAY 1



FOUR PILLARS OF BIG DATA





ANOTHER PERSPECTIVE ON VISUALIZATION

Who said that? When?

"There is danger in giving too much information to executives of small brain capacity."

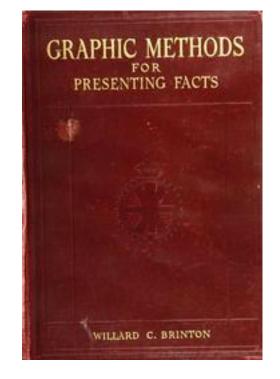
"As a cathedral is to its foundations, so is an effective presentation of the fact to the data."

"The answer is that the executive of the future will be forced on the analysis of facts which have been collected and arranged for his instantaneous and continuous use."



ANOTHER PERSPECTIVE ON VISUALIZATION





Willard C. Brinton

1914

100yrsofbrinton.tumblr.com

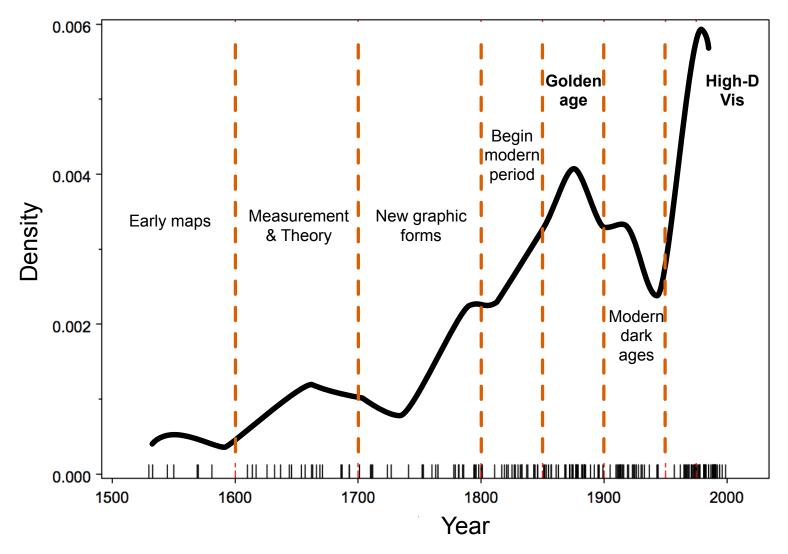
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Graphics Milestones: Time course of developments



The distribution of milestone items over time, shown by a rug plot and density estimate. Michael Friendly et Daniel J. Denis. https://www.researchgate.net/publication/221649568



ANOTHER PERSPECTIVE ON VISUALIZATION



Photo by the International News Service

Fig. 238. Statistical Exhibits in the Municipal Parade by the Employees of the City of New York, May 17, 1913

Many very large charts, curves and other statistical displays were mounted on wagons in such manner that interpretation was possible from either side of the street. The Health Department, in particular, made excellent use of graphic methods, showing in most convincing manner how the death rate is being reduced by modern methods of sanitation and nursing

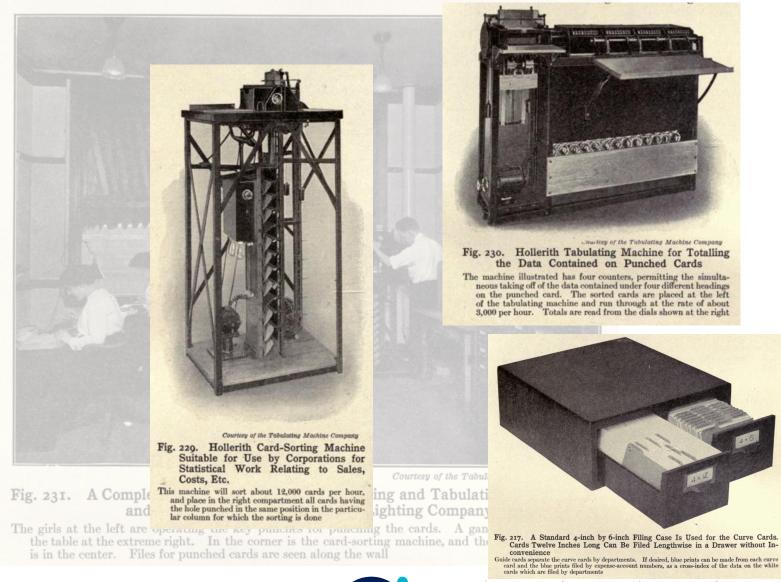
100yrsofbrinton.tumblr.com



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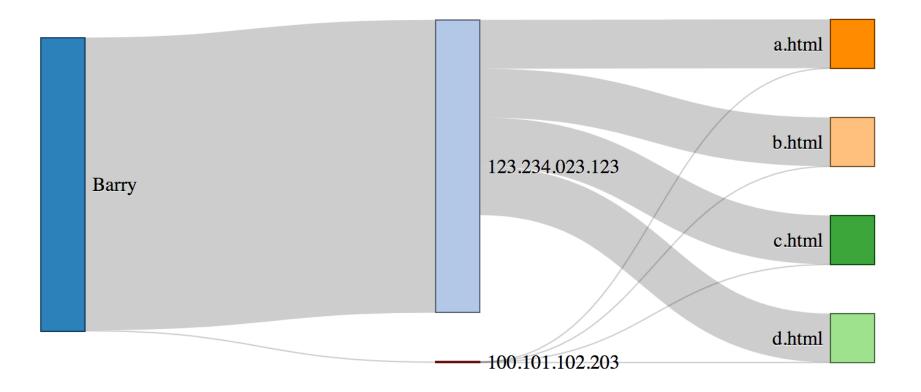
THE PREVIOUS BIG DATA REVOLUTION (END 1800s)





ANTI PATTERN #1: USING THE WRONG GRAPH TYPE

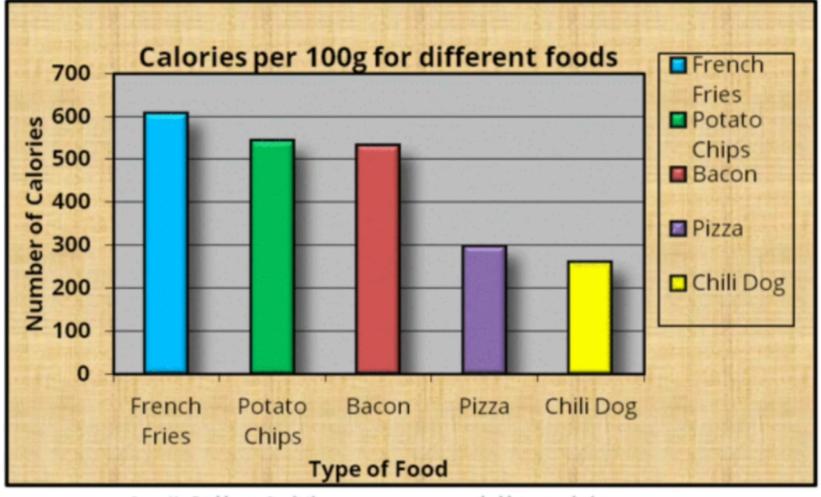
Violating all principles





ANTI PATTERN #2: OVERLOADING

Remove backgrounds



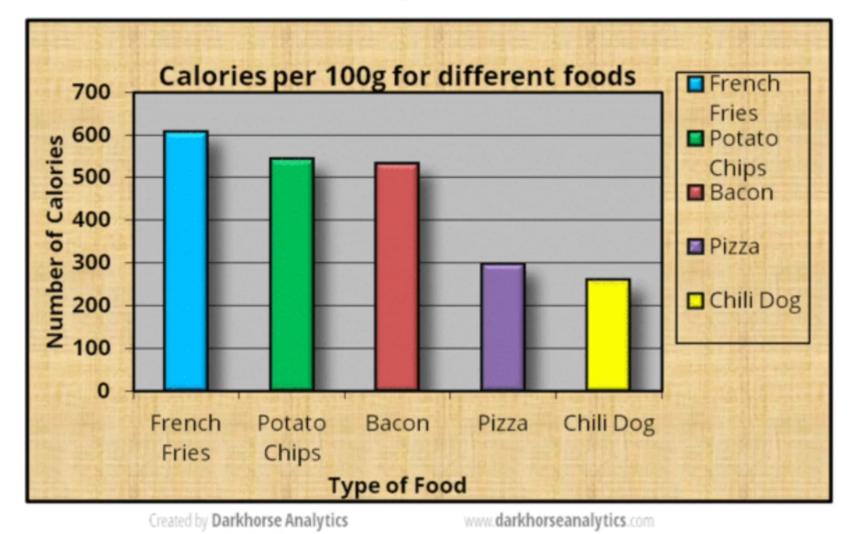
Created by Darkhorse Analytics

www.darkhorseanalytics.com



ANTI PATTERN #2: OVERLOADING

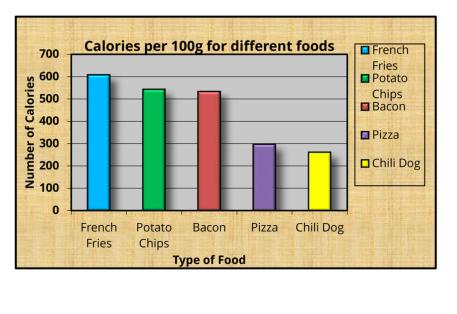
Remove backgrounds

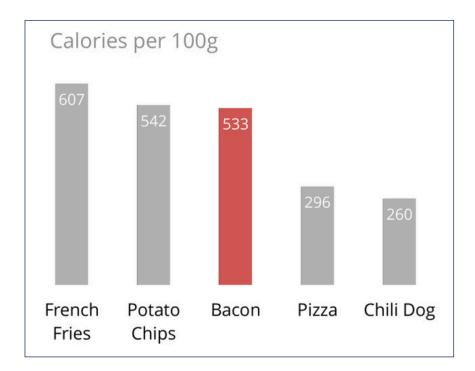


OCTO TECHNOLOGY > THERE IS A BETTER WAY

ANTI PATTERN #2: OVERLOADING

Data/Ink Ratio

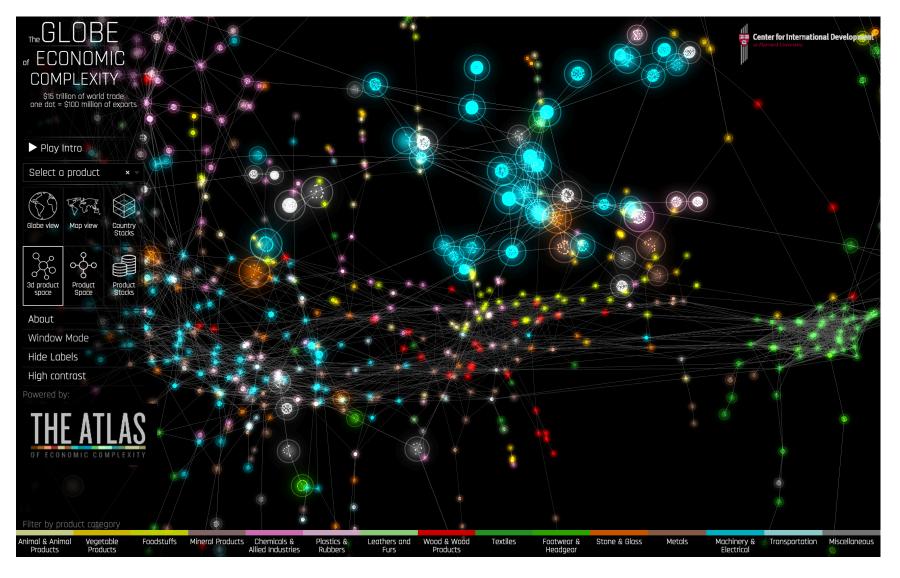




"Perfection is achieved not when there is nothing more to add, but when there is nothing left to take away" Antoine de St Exupéry Terre des Hommes, 1939



Three.js

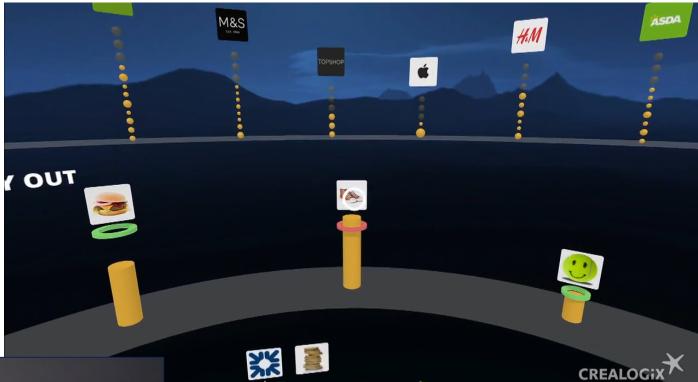


Interactive display wall

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http://earlymodernconversions.com/activity/history-visualization-lab/

Virtual reality







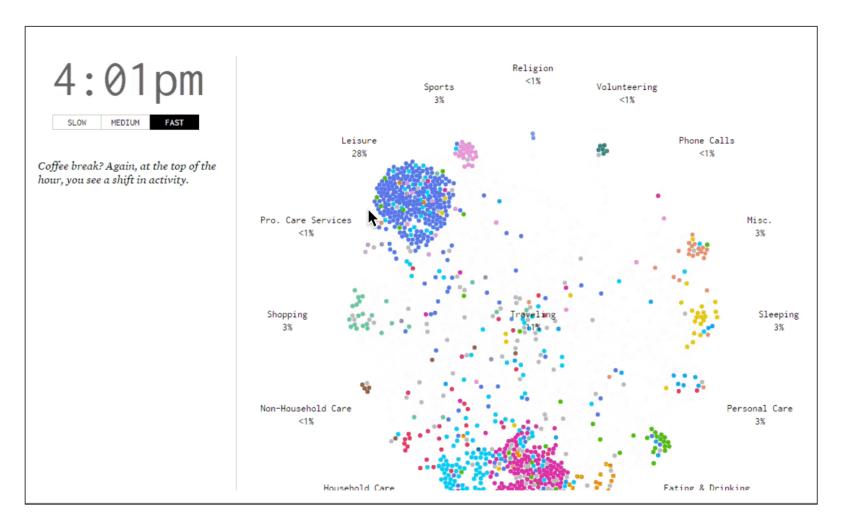
Data sonification



http://earlymodernconversions.com/activity/history-visualization-lab/



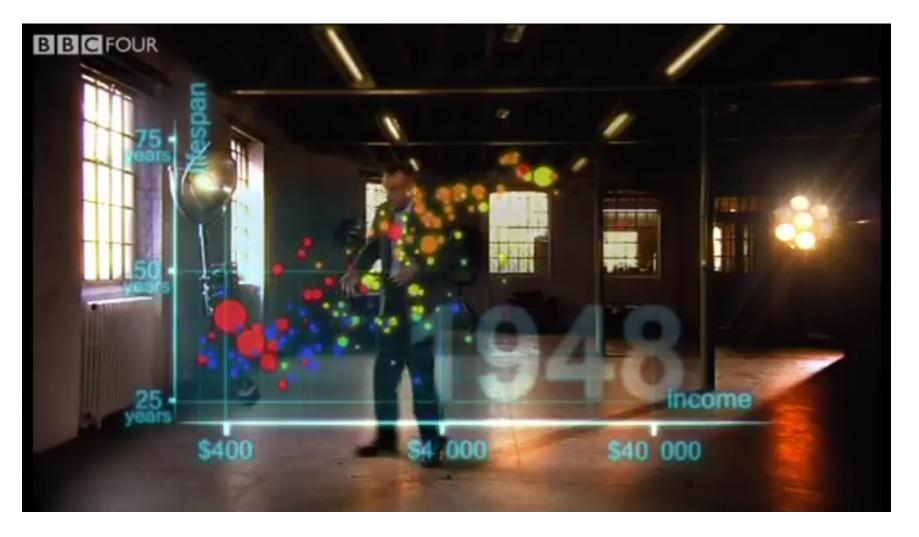
Animation



A Day in the Life of Americans – Nathan Yau

EVEN THOUGH IT CAN BE GREAT, BY HANS ROSLING

Animation



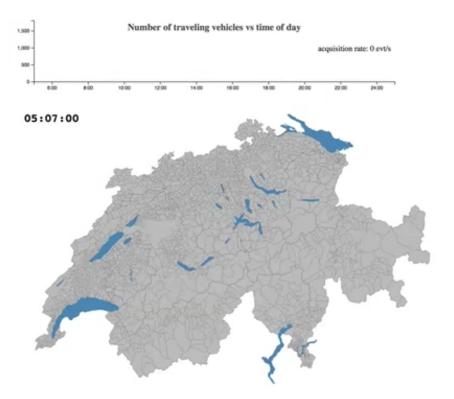
Hans Rosling... The Revolutionary

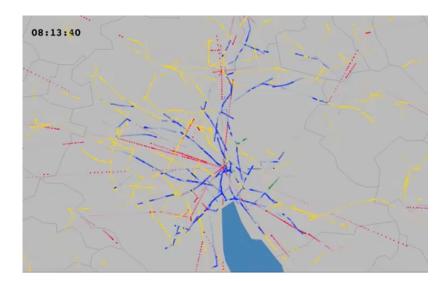


ANTI PATTERN #4: UNDERESTIMATE YOUR BROWSER

Browser power with high throughput data

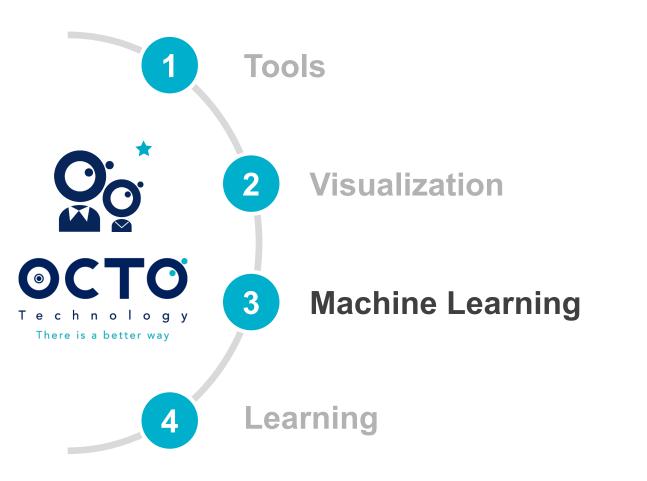
Up to 10'000 events per second







FOUR PILLARS OF BIG DATA



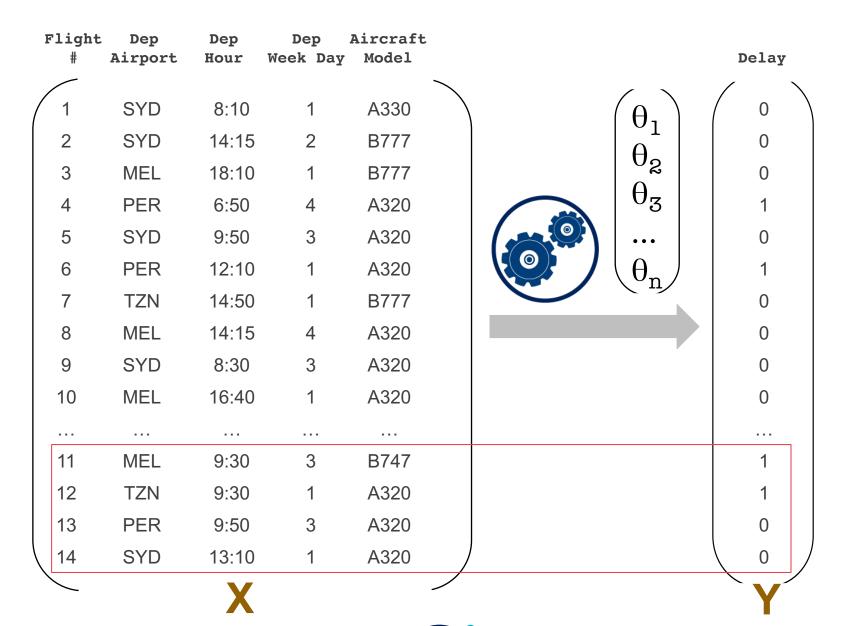
"Machine learning explores the study and construction of **algorithms** that can **learn** from and make **predictions** on **data**"

https://en.wikipedia.org/wiki/Machine_learning

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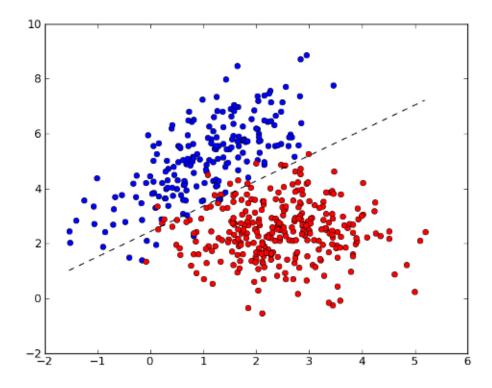


BUILD A MODEL



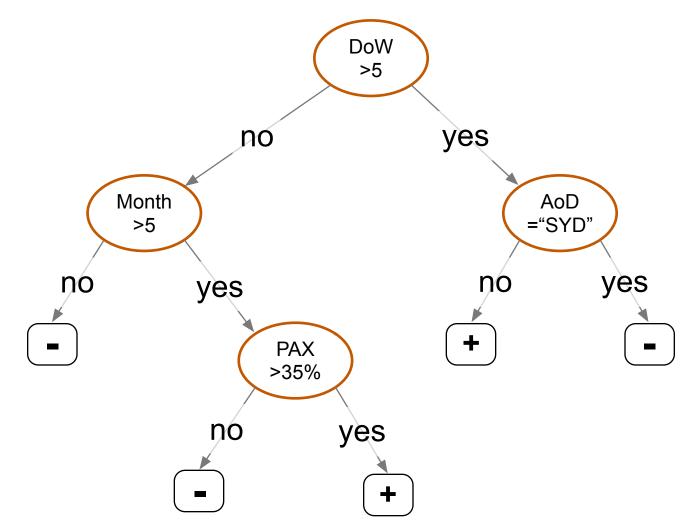
LOGISTIC REGRESSION

Classification algorithm



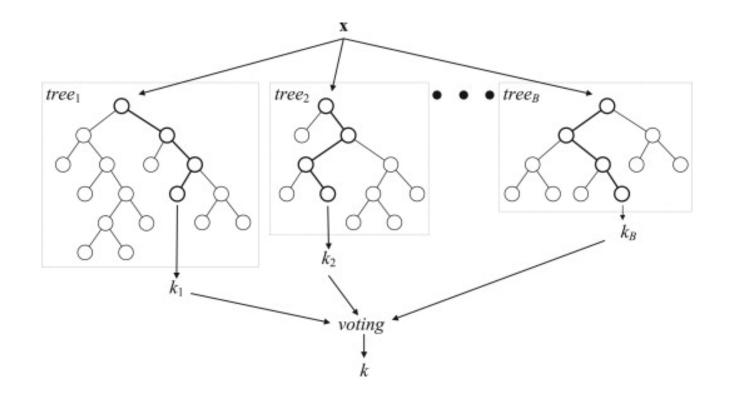
DECISION TREE

Classification algorithm



RANDOM FOREST

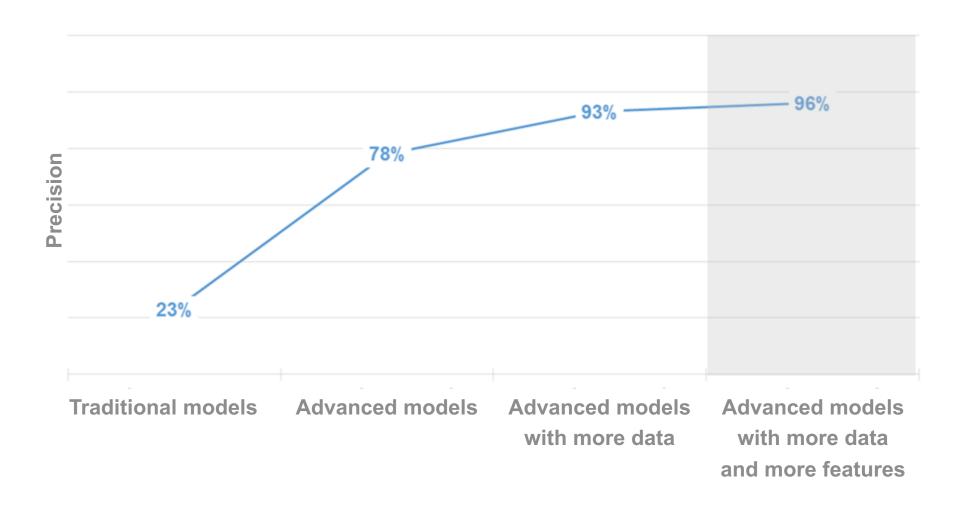
Classification algorithm



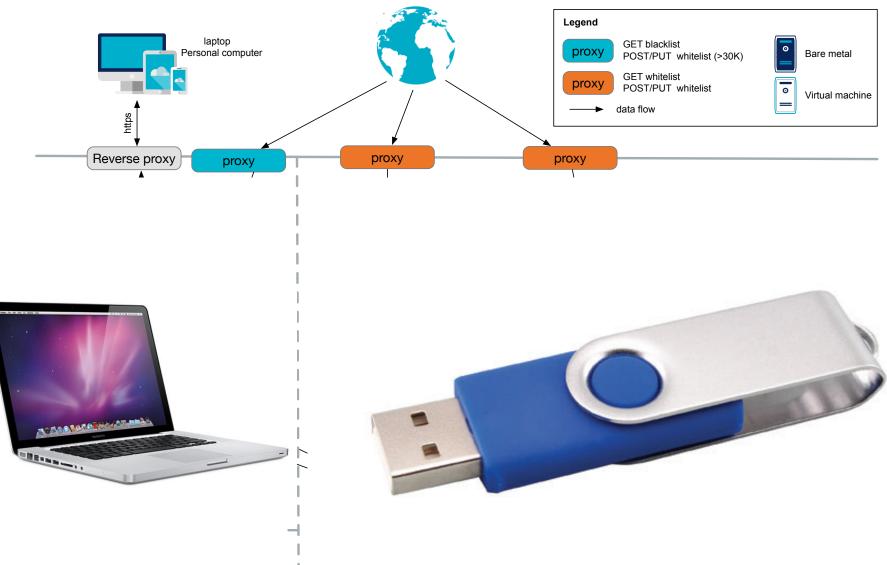


MODELS & DATA

Precision score for the TOP 20%



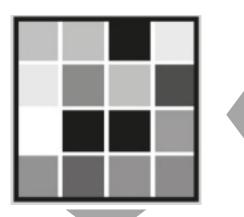
ANTI PATTERN #1: WAIT FOR THE FULL STACK BEFORE STARTING



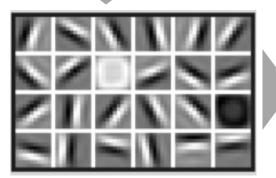
ANTI PATTERN #2: START TOO COMPLEX

input layer input

Identify pixels







Identify edges and simple shape



Identify complex shapes and object

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Identify which shape to be used to define a human face



ANTI PATTERN #3: FORGET THE LIFE AFTER THE NOTEBOOK

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		In June 1	▶ 🖿 bdacore.egg-info 1	
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	# import des différents modèles (librairie scikit-learn)		🕨 🖿 dist	L2 pairs =
	<pre>from sklearn.linear model import LinearRegression, Ridge from sklearn.ensemble import RandomForestRegressor, ExtraTreesRegressor, GradientBoostingRegressor</pre>		▶ b docs	
	from sklearn.svm import SVR		reports	
	from sklearn.neighbors import KNeighborsRegressor		▶ m src	
In [37]:	# initialisation des différents modèles		tests	
	<pre>lr = LinearRegression() rf = RandomForestRegressor(n_estimators=100, max_depth=10, n_jobs=-1)</pre>		🛃 .bumpversion.cfg	l9 setup(
	<pre>et = ExtraTreesRegressor(n estimators=100, max_depth=10, n_jobs=-1) gb = GradientBoostingRegressor()</pre>		🛃 .cookiecutterrc 🛛 💈	
	svr = SVR() knn = KHeighborsRegressor(5)			version=_ver
	NIII = Nezzinovi snogi osavi (3)			22 author="BDA 1 23 author_email:
	Séparation du dataset en 2 : train (2009) & test (2010)			url="https:/
In [38]:	X_train = X[X.CalYear == 2009] y_train = y[X.CalYear == 2009]		a Aomoraliae	
	X test = X[X.CalYear == 2010]		GHANGELOG.rst	27 package_dir= 28 package_data=
	y_test = y[X.CalYear == 2010]		GONTRIBUTING.rst	9 description=
				long_descript
	Apprentissage du modèle sur le dataset d'apprentissage			
In [39]:	lr.fit(X train, y train)			
Out[39]:	LinearRegression(copy_X=True, fit_intercept=True, n_jobs=1, normalize=False)		🖬 Makefile	
	rf.fit(X train, y train)			
OUT[40]:	RandomForestRegressor(bootstrap=True, criterion='mse', max_depth=10, max_features='auto', max_leaf_nodes=None, min_samples_leaf=1,		Enterterior	
	<pre>min_samples split=2, min_weight fraction leaf=0.0, n_estimators=100, n_jobs=-1, oob score=False, random state=None,</pre>		🔮 setup.cfg	
	verbose=0, warm_start=False)		🖡 setup.py	
In [41]:	knn.fit(X_train, y_train)			
Out[41]:	<pre>KNeighborsRegressor(algorithm='auto', leaf_size=30, metric='minkowski',</pre>			
	<pre>metric_params=None, n_jobs=1, n_neighbors=5, p=2, weights='uniform')</pre>		P de jupycer indexeensions	
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	Evaluation (test) du modèle sur le set de test		► ad ► Madoc	
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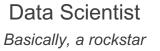
🖿 DD 👌 🖿 bdacore 👌 💑 setup.py

🔕 meta.yaml

ANTI PATTERN #3: FORGET THE LIFE AFTER THE NOTEBOOK

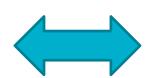
Competencies







- Artificial Intelligence
- Bleeding edge algorithm





OPS

• Run things



ANTI PATTERN #3: FORGET THE LIFE AFTER THE NOTEBOOK

Competencies



- (Machine Learning)
- (Statistics)

But overall...

- Software Craftsmanship
- A software development culture
 - > Production?
 - > Architecture
 - > Agility principles?



ANTI PATTERN #4: MAKING A POC INSTEAD OF A PILOT PROJECT

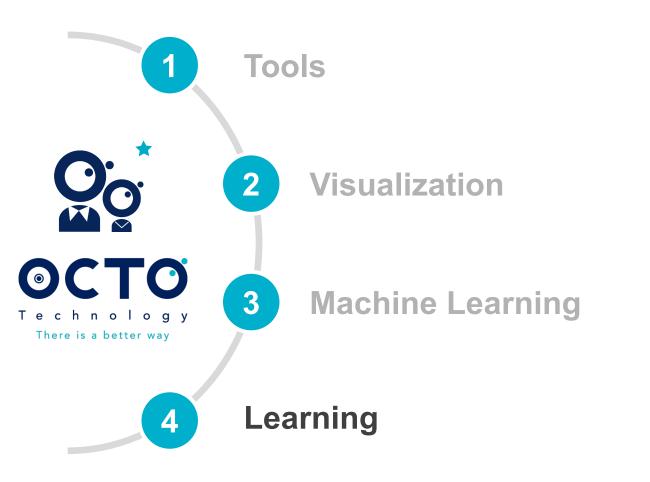
• POC : Check the model works

• Pilot project : Check it is useful

Customer next be	est action					
Extract more value from you 1. Choose a customer 2. Examine attributes im 3. Get next best action		:				
1 Select your custor	ner			2 Select	actionable featu	Ires
1	Get insight!			nb_cont_cou	ur_ent,ass_fact,age_tr	Get next action!
Q Informations		Insights		A Next be	est action	
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Name score_id score_id id prediction crc_isnan prea darty	Value 12.0 1.0 0.0687926006625 0.0 1.0 0.0	Variable name nb_cont_cour_ent ass_fact age_tr_pred_fmt lb_prof_payr cel_fmt crc puls_sous	Contribution 16 12 12 11 11 11 11 11 11 11 11 11 11 11	% nb_cont_cour_er % ass_fact % age_tr_pred_fmt % Score % First % Gest score % Gest score	Current Curren	6 1 5 0.2121057026 Next Last 2.



FOUR PILLARS OF BIG DATA







Learn To Change

AGILE & LEAN

- BEST Découvrir les démarches agiles et la culture agile
- NEW Certification Scrum Master
- NEW Adopter les bonnes pratiques de gestion de projet agile
- BEST Facilitation graphique par la pratique
- Animer une rétrospective projet
- Le rôle de Product Owner en pratique
- BEST Certification Leading SAFe au Scaled Framework Séb...
- Conception logicielle : écrire des cas d'utilisation efficaces
- Kanban : mieux travailler en maitrisant son flux de production
- Lean Startup en entreprise
- Lean IT : optimiser ses flux plutôt que ses ressources
- EXCLU Devenir Coach Agile ou Scrum Master

CULTURE CHANGE

- CHANGE ET TRANSFORMATION
- EXCLU L'atelier du changement
- EXCLU Théorie U
- Exclu Explorer et partager la démarche entrepreneuriale aut...

FACILITATION

- Excuu Techniques et méthodes de facilitation de groupe : niv...
- EXCLU Facilitation d'un plan stratégique participatif : niveau 2
- Exclu Enrichir et approfondir sa posture de facilitateur dans ...

MANAGEMENT ET LEADERSHIP

- Excuu Optimiser son temps et ses priorités Ismaël Héry
- Parole performante et communication impactante
- BEST Donner et recevoir des feedbacks efficaces
- BEST Management collaboratif et agile : there is a better way !

BIG DATA

DATA SCIENCE

- BEST Fondamentaux de la Data Science
- BEST Data Science : niveau avancé

HADOOP HORTONWORKS

- Les fondamentaux d'Hadoop
- BEST Administrer la plateforme Hadoop 2.X Hortonworks : ni...
- EXCLU Administrer la plateforme Hadoop 2.X Hortonworks : n...
- EXCLU Administrer la plateforme Hadoop 2.X Hortonworks : s...
- BEST Analyse de données pour Hadoop 2.X Hortonworks av...
- EXCLU Développer des applications pour Apache Spark avec ...

HADOOP CLOUDERA

- BEST Développer des applications pour Spark avec Hadoop C...
- NEW Administrer la plateforme Hadoop Cloudera
- NEW Utiliser Pig, Hive et Impala avec Hadoop Cloudera pour ...

SPARK DATABRICKS

EXCLU Programmer avec Apache Spark de Databricks

NOSQL

- BEST NoSQL : découverte des solutions et architecture de la ...
- Déployer et gérer un cluster Couchbase
- Requêtes, modélisation de données, optimisation et migration...
- Gérer efficacement ses logs avec la stack ELK
- NEW Concevoir un moteur de recherche avec Elasticsearch

Inspirer dans la durée

Favoriser l'innovation Accompagner les transformations

academy.octo.ch



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