



mydataball

Performance Knowledge Discovery



VISUAL DATA ANALYTICS COLLABORATIVE SOLUTION
DECISION-MAKING VALUE CHAIN



23 january 2020

- To quickly access the extended business digital heritage
- To accelerate the analytical capacity
- To synthesize the detection of business diagnostics by your AI
- To meet the challenges of communicating figures
- To capitalize on the professional knowledge of employees

Data to accomplish business strategies



Les principaux objectifs des décideurs interrogés en termes de gestion analytique et décisionnelle - France, 2012 (multi-réponses, liste suggérée)



Echantillon : 164 décideurs dont l'entreprise est ouverte aux solutions analytiques et décisionnelles

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- "What dashboards should I see first? "
- "How long to read and interpret my dashboards? "

GROWING NUMBER OF INSTRUMENT PANELS vs NEED TO REDUCE INTERPRETATION TIMES



- Recurrence and acceleration of the need for analysis
- Data extracts and analyzes are scattered (Excel)



The existing BI and analysis tools are no longer sufficient:

- ✓ impose frozen tables
- ✓ offer only a reduced number of manually chosen axes of analysis
 - ❑ limits the field of exploration of diagnostics therefore productivity

The Data Complexity : $\sum_{i=0}^N C_i^N$

IA MyDataBall

- ❑ **MyDataBall** is a methodology for building AI on dashboards
- ❑ It is no longer a user who goes to metrics but the **performance metric** which goes to the right people
- ❑ Through a question and answer game, a **machine learning** integrates user needs to predict and prevent high ROI rules

Strategy / Communication / Risk Management

Compilation of trees



MyDataBall plateforme data analytics



- ⇒ Find deep & high ROI diagnostics
- ⇒ Detect causality in the world of correlations
- ⇒ Accelerate the capacity for collaborative business analysis

Démonstration

Uses Case

MyDataBall généralist & vertical solution

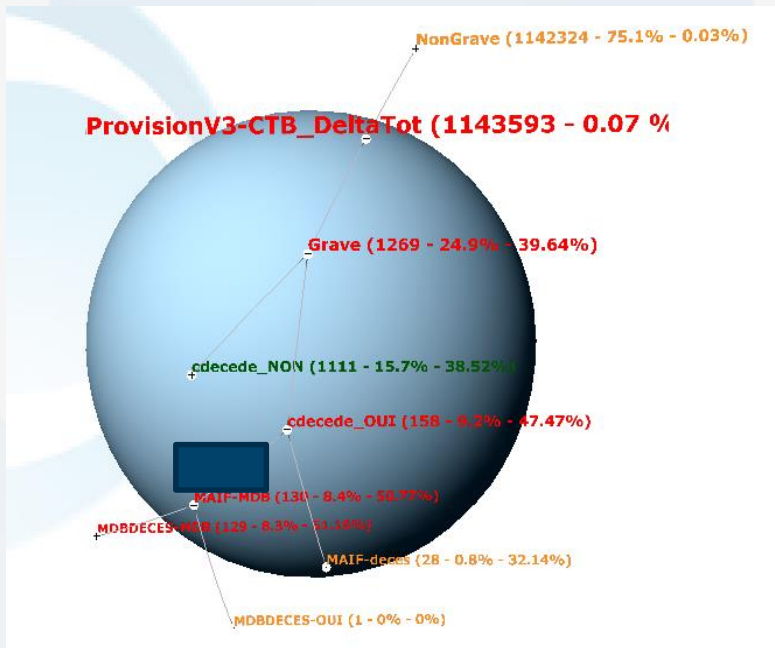
- Distribution network:** where are the most favorable locations?
- Marketing:** what is the profile of my loyal customers? How to refine customer segmentation?
- Commercial:** what are the operational instructions to give to my teams?
- Commercial:** what is the state of my sales networks?
- Communication:** what information to communicate to my partners?
- Risk Management:** compliance management in Basel 3 and Solvency 2.
- Finance:** what are the profiles of short, medium and long term risk products?
- Credit risk:** what is the quality of my client portfolio and what are the explanatory elements of my risk?
- Accounting:** what are the typologies of fraud and money laundering?
- After-sales service:** what are my levers to limit its cost and its recovery?
- Lean management:** what are the factors that make my productivity fall?
- Institutional communication:** what are the priority messages?
- R&D:** what is the resistance of my materials in changing contexts?
- Pharma:** what are the trends in the exclusion criteria?
- RH:** where are the charges derived from? How satisfied are my employees? Reduction in absenteeism.
- IoT:** alerts on malfunction forecast, optimization of efficiency and yields.
- CSR:** diagnostics and outlook for strategies, performance lever for water consumption.
- Provident:** carers / caregivers, personal services

Finance / DAF : optimised provisioning



Claims Analysis

- Objective: to detect and share the assumptions regarding claims provisioning
- Project carried out with SAS over 2 years by 3 departments involving 20 people, no results
- MyDataBall project: 2 months of realization, results shared by everyone, collective detection and validation of hypotheses



▼ 2013

CTB_Ratio_2013	ctb_2013_Rechargement
depense_2013	ctb_2013
TxDep_2013	Dep_dif_2013
CTB_dif_2013	ctb_2013_Degagement
IPP01_10_2013	IPP01_10_dif_2013
IPP11_20_2013	IPP11_20_dif_2013
IPP21_30_2013	IPP21_30_dif_2013
IPP31_40_2013	IPP31_40_dif_2013
IPP41_50_2013	IPP41_50_dif_2013
IPP51_60_2013	IPP51_60_dif_2013
IPP61_70_2013	IPP61_70_dif_2013
IPP71_80_2013	IPP71_80_dif_2013
IPP81_90_2013	IPP81_90_dif_2013
IPP91_100_2013	IPP91_100_dif_2013
IPP_Global_2013	IPP_Global_dif_2013
Trauma_2013	Trauma_dif_2013
amp_2013	amp_dif_2013
cdecdecde_2013	cdecdecde_dif_2013
cnbvic_2013	cnbvic_dif_2013
handicap_2013	handicap_dif_2013
para_2013	para_dif_2013
poly_2013	poly_dif_2013
polytc_2013	polytc_dif_2013
tetra_2013	tetra_dif_2013
ctb_2013_chargement	

Alert message, recurring rules with high ROI

- Prediction / Prescription

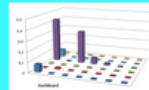
Vous voulez diagnostiquer les raisons pour lesquelles l'indicateur **NbTrame** (*) est supérieur à 20.

Les statistiques de votre indicateur

- Vous avez utilisé 7 variables pour comprendre l'indicateur **NbTrame**
- 72.93 % des sondes-passerelles ont une valeur supérieure à 20 de **NbTrame**
- La variable MANUFAC est la plus discriminante de **NbTrame**

Les règles explicatives apparues (**)

- Règle 1 ([dashboard](#), proba = 0.93 ; pureté = 0.03) :
 - a. [Day] = 20191106
 - b. [MANUFAC] = SAP
 - c. [TypeFournisseur] = Diehl Metering
 - d. [Type_Data] = Gaz Fioul Lourd



Les règles prévues pour votre indicateur NbTrame

- Demain règle identique à aujourd'hui à 0.76 % :
 - a. [Day] = 20191106
 - b. [MANUFAC] = SAP
 - c. [TypeFournisseur] = Diehl Metering
 - d. [Type_Data] = Gaz Fioul Lourd
- Prédiction 1 Proba (0.95) :
 - a. [Module_type] = LORA_ST
- Prédiction 2 Proba (0.95) :
 - a. [Sonde] = Presence Sonde
 - b. [Pulse_ALARME] = D

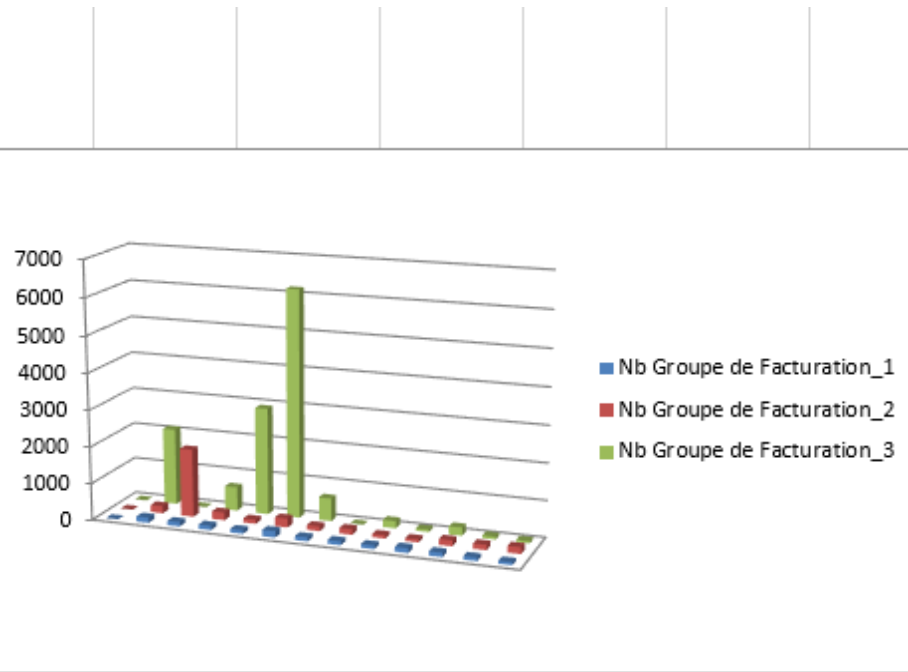
(*) L'indicateur **NbTrame** mesure le nombre de données par jour pour une sonde-passerelle -- Seuil bas 2 Seuil haut 20

(**) Lecture des règles : la conjonction des items décrits expliquent ce que vous voulez comprendre selon une probabilité de vraisemblance et une pureté (1 proche d'une raison causale).

Use in trades services

Detection of the best dashboards among a multitude

- Prediction / Prescription



	Moyenne des indices de delta	Indice_Conso_6_Temperature	Nb Groupe de Facturation_1	Nb Groupe de Facturation_2	Nb Groupe de Facturation_3
Moyenne des	Indice_Cons	0,0011	0	0	
Moyenne des	Indice_Cons	143,722	195,167	2094,95	
Moyenne des	Indice_Cons	120,796	1857,53	0	
Moyenne des	Indice_Cons	118,09	211,892	651,378	
Moyenne des	Indice_Cons	99,835	111,602	2918,4	
Moyenne des	Indice_Cons	172,467	239,662	6181,97	
Moyenne des	Indice_Cons	105,677	129,317	623,767	
Moyenne des	Indice_Cons	104,43	152,333	0	
Moyenne des	Indice_Cons	94,9734	89,7512	196	
Moyenne des	Indice_Cons	128,644	79,9091	64	
Moyenne des	Indice_Cons	114,214	158,806	221,101	
Moyenne des	Indice_Cons	90,058	142,125	75,1009	
Moyenne des	Indice_Cons	76,4733	180,461	56,5506	

Use in trades services



Optimization of filters on the tools of the classic dataviz market

- Prédiction / Prescription

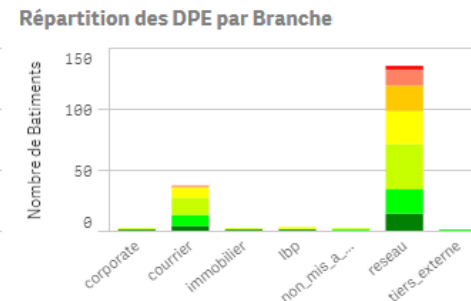
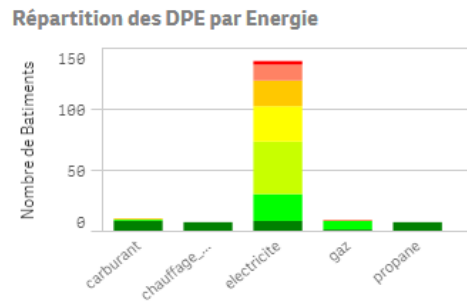
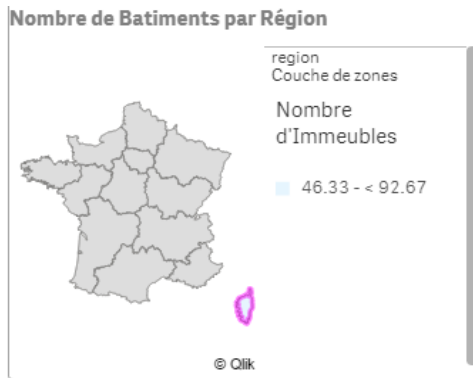
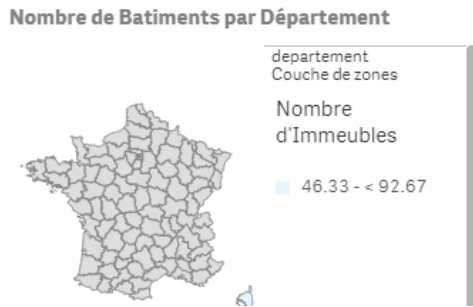
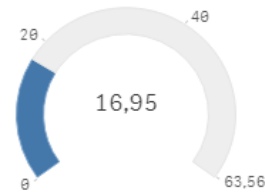


Tableau de Données

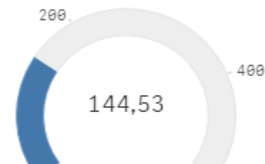
Immeuble	Energie	Branche	Conso en kWh/m ²	Coût en €
AFA GA	electricite	reseau	149,47	
AGOSTA	electricite	reseau	203,86	
AJACCIO CENTRE DE TRI POSTAL	electricite	courrier	104,21	
AJACCIO CENTRE DE TRI POSTAL	electricite	reseau	104,21	
AJACCIO FRANCHINI	electricite	reseau	88,37	
AJACCIO HOTEL DES POSTES	electricite	courrier	51,82	
AJACCIO HOTEL DES POSTES	electricite	reseau	51,82	
AJACCIO HOTEL DES POSTES	electricite	tiers_externe	51,82	



Conso moyenne €/m²



Conso moyenne kWh/m²



Use in trades services

Detection of priority areas

- Prediction / Prescription



Use in trades services



Detecting categories based on text

- Prediction / Prescription

2 - declaration of loss --> activity ▼

Enter a sentence : j'ai peignais mes volets et je suis tombé

Get response

Results

Contract type	Likelihood
Activité_domestique	0.8913174867630005
Autre_activité	0.05152815207839012

Go back

Our Customers



MyDataBall created from analytical collaboration



What our customers say

- "A tool that makes it easier to use data without multiplying point clouds"
- "A tool that accelerates the implementation of our reporting appendices and the sharing of our studies"
- "A tool that guides us towards the best explanatory axes"
- "Saving time in the analysis, sharing and understanding of our activity, enhancement of the digital heritage to improve the management of decisions"
- "I was finally able to exchange with the directors of the agencies and obtain the real performance levers and the validation of good practices"
- "With MyDataBall, we will point directly to the points to be dug, rather than exploring point by point while waiting to come across something that could hit the target."



The 3 stages of making an application

From data to knowledge

1

Data Preparation & Management



- Connectors that assemble the heterogeneity of data in a very large mass.
- Digital heritage augmented with simple and complex indicators
- Industrial and universal modeling by granularity

2

Data Analytics & Knowledge Discovery



- Multidimensional interrogation of data by trades and restitution in the form of trees
- Optimized Visualization of Dashboards compatible with market tools
- Aggregation and synthesis of questions and answers to generate a machine learning machine.

3

Knowledge management & IA



- Predictive and prescriptive model on data flows
- Fast deeplearning reducing map user interface and simulator
- Risk Management & Auditability of the generated AI



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*Turn data into knowledge,
knowledge into insights and insights into business outcomes*

