DMIF, University of Udine

Data Management for Big Data

The Gap Srlu Case

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Outline

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Introduction: The Contact Center Domain



Introduction

Multi-channel contact centers are an important component of today's business world.

They serve as a primary customer-facing channel for firms in many different industries, and employ millions of agents across the globe.

During their operation, they generate vast amounts of heterogeneous data, ranging from structured automatically registered logs to semi-structured hand-written notes and unstructured raw voice recordings.



Inbound, Outbound and Backoffice Ops.

Inbound call centers handle incoming traffic, which means that they answer to calls received from the customers, as in the case of help-desks.

Outbound call centers handle outgoing calls, which are initiated from the call center. Such calls may be associated with surveys or telemarketing initiatives, and they typically follow a predefined script.

Backoffice operations may also be carried out, as in the case of data preparation and data analysis tasks.

All operations are carried out within the context of a *service* (e.g., an airline toll-free number), which can be composed of many different *activities* (e.g., ticket booking, or car rental).

Gap Srlu Company



The Company

Gap Srlu is a multi-channel and multi-service Business Process Outsourcer, specialized in contact center activities.

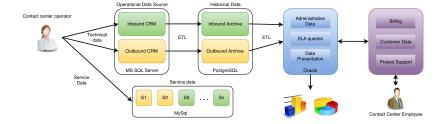
It is active since the early 2000s and, over time, it has experienced a continuous expansion concerning both its business model, and its information system infrastructure.

Nowadays, other than the traditional contact center tasks, it is capable of offering advanced services such as third-party data management and analysis, based on several machine learning technologies.

More info at: https://www.gapitalia.com/?lang=en



The Initial Situation





What are the Issues Here?

Several problems:

- heterogeneous systems require ad-hoc solutions for reading and writing data
- different databases adopt different conventions for storing the data
- possibly (and probably) replicated and inconsistent information
- difficult to perform queries and analyses involving more than one data repository
- the whole architecture is complex, and hard to maintain and update

Warehouse

Development of the Data



Why a Data Warehouse

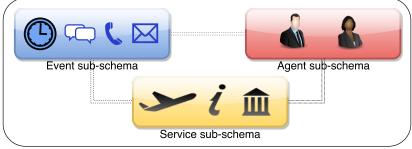
All kind of monitoring and analysis tasks start from the data.

Thus, there is the necessity of having a clear and uniform view over all the company information.

Moreover, a unique, central data repository simplifies the overall infrastructure.



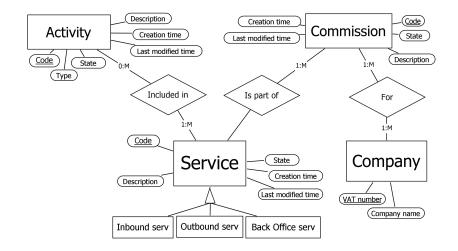
Data Warehouse Overall Design



Data warehouse relational schema

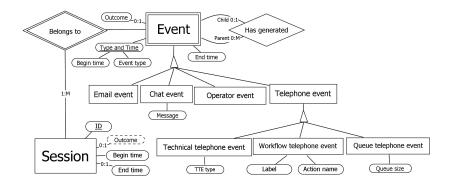


Service Sub-schema



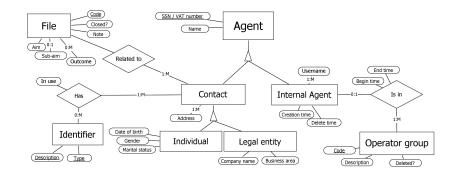


Event Sub-schema



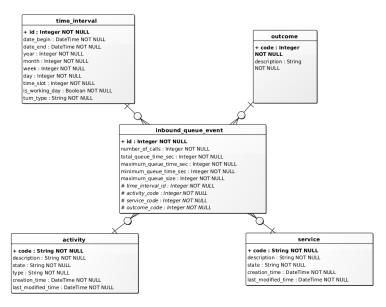


Agent Sub-schema





The Analysis Layer / Data Marts



Analysis Tasks



Operator Performance Assessment

Tracking the performance of agents is a primary issue in contact centers, as it allows, for example:

- the best match to be taken between service and agent
- the recognition of unsatisfactory agent behaviours, due for example to a lack of proper training
- the prediction of future trends, based on the history of observations

A function has been designed, which is capable of assigning a score to each operator-service couple.



Operator Performance Assessment Some of the Considered Information

	Average conversation time
	Average postcall time
Inbound	Generic call notes compiled per session
	Percentage of correctly filled script fields
	Purpose of the call
	Outcome of the call
	Average conversation time
0-411	Average postcall time
Outbound	Amount of surveys over calls
	Number of answered calls per hour
	Number of different kinds of services managed by an operator
C 1	Degree of interleaving between services
General	Respect of work schedule
	Turn flexibility
	-



Operator Performance Assessment Detail of the User Interface





Analysis of Written Notes

As a part of the agent performance evaluation framework, Gap automatically assesses the characteristics of written notes taken by the agents during phone calls:

- how often / in which way does an agent record notes regarding an inbound call?
- compare single agent behaviour with service average values

How to evaluate written notes?

- extract summarizing features from the text
- identify groups of similar notes
- devise a methodology to assign a generic new note to one of the previously identified groups



Analysis of Written Notes Extracted Features

For each note, we calculate:

- numbers of words and characters
- Gulpease readability index value
- fractions of articles and conjunctions over words
- fractions of verbs and adverbs over words
- fraction of adjectives over words
- fraction of prepositions over words
- fraction of quantifiers over words
- fraction of (pro)nouns over words
- fraction of numeric codes over words
- fraction of proper nouns over words
- fraction of words/abbreviations found in Italian dictionary
- fraction of words found in **service-specific** domain
- fraction of unrecognized words



Analysis of Written Notes Identify Groups of Similar Notes

- Random sampling of 1000 notes
- application of a clustering algorithm to the selected notes (E-M algorithm)
- 6 clusters emerged:
 - articulated notes
 - non-articulated notes
 - abbreviated notes
 - domain-specific notes
 - nonsense notes
 - hybrid notes



Analysis of Written Notes Classify a New Note

- Attach a new feature to each of the clustered notes: cluster label
- apply a decision tree learning algorithm (J48), with the goal of predicting the label (94.7% accuracy)
- the tree can then be used to classify new notes

```
riconosciuti abbr su parole <= 0.142857
  riconosciuti_dominio_su_parole <= 0.133333
    preposizioni su parole <= 0
      non_riconosciuti_su_parole <= 0.157895
         congiunzioni su parole <= 0.025
      I I articoli su parole <= 0.071429; non articulated notes
        I articoli su parole > 0.071429: articulated notes
      I congiunzioni_su_parole > 0.025: articulated_notes
      non_riconosciuti_su_parole > 0.157895
      I non riconosciuti su parole <= 0.333333
           articoli_su_parole <= 0.083333: hybrid_notes
         I articoli su parole > 0.083333: articulated notes
    I I non_riconosciuti_su_parole > 0.333333: non_sense_notes
    preposizioni su parole > 0
      indice_gulp <= 129.833333: articulated_notes
      indice_gulp > 129.833333
      I non_riconosciuti_su_parole <= 0.0625: non_articulated_notes
I I I non_riconosciuti_su_parole > 0.0625: hybrid_notes
```



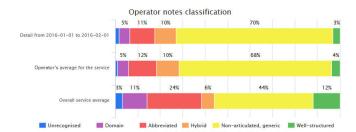
Analysis of Written Notes Example – 1

	valore text	gruppo_nota text
1	info voltura	hybrid
2	invio del f24	articulated
3	informazioni per appunt sub e comunica dati catastali	articulated
4	info posizione pagamenti mensa scolastica	hybrid
5	NON RISPONDE	non-articulated
6	Info	abbreviated
7	VIA MQ 37 C'È SCRITTO 43 BOLLETIAZIONE SBAGLIATA. DEVE PASSARE AGLI SPORTELLI PER RETTIFICA DI METRATURA CON PIANTINA SCALA 1:100. RIFERISCO. C	articulated
8	SIGNORA CHIAMA PER SAPERE SE È STATA APPLICATA LA DETRAZIONE DI 25 euro per figlio sul calcol	articulated
9	la signora avea chiamato il 23/05 per una verifica posizione per la TARES: ha un locale comme	articulated
10	chiede quanto deve pagare per la tassa. Parlato con: deve pagare 61 euro.	articulated
11	info boll	abbreviated
12	rimborso ud	non-articulated
13	tasi	domain-specific
14	INFO GENERICHE IMU, TASI	domain-specific
15	info su avv sosp	hybrid
16	chiede se può rateizzare l'importo da versare per la mensa. Riferito che deve fare richiesta	articulated
17	invio copia boll	hybrid
18	chiede il saldo mensa. Riferito che abbiamo problemi tecnici tecnici al server	articulated



Analysis of Written Notes Example – 2

Agent-service notes class distribution, with respect to the overall distribution for the service.





Anomalous Call Outcomes Detection

Outbound calls follow a pre-defined script, which allows one to predict, to a certain extent, their outcome based just on *dialling*, *conversation*, and *postcall* times.

This allows to detect contact center operators who systematically annotate wrong call outcomes, either by mistake or to simulate surveys which did not take place.

A decision tree model has been developed that, based on *dialling*, *conversation*, and *postcall* times of a phone conversation, derives its most likely outcome, with an accuracy above 93%.



Anomalous Call Outcomes Detection The Developed Model

```
conversation time <= 7
   conversation time <= 0
         dialling time <= 11: busy or nonexistent
            dialling time <= 14: busy or nonexistent
            dialling time > 14: no answer
      dialling time > 30: no answer
   conversation time > 0
      postcall time <= 1
         dialling time <= 29: fax or answermachine
         dialling time > 29
            conversation time <= 1: no answer
            conversation time > 1: fax or answermachine
      postcall time > 1
         conversation time <= 4: fax or answermachine
         conversation time > 4: spoken no survey
conversation time > 7
   conversation time <= 76
      conversation time <= 11
         postcall time <= 1
            conversation time <= 9
               dialling time <= 22
                  conversation time <= 8: fax or answermachine
                  conversation time > 8: spoken no survey
               dialling time > 22: fax or answermachine
            conversation time > 9: spoken no survey
         postcall time > 1: spoken no survey
      conversation time > 11: spoken no survey
      conversation time <= 87
         postcall time <= 0: spoken no survey
         postcall time > 0: survey made
      conversation time > 87: survey made
```



Analysis of Phone Conversation Recordings

The ability to analyze conversational data plays a major role in contact centers, where the core part of the business still focuses on the management of oral interactions.

Several actors already provide speech analytics solutions, e.g., Google or Amazon. However, they come with a price.

Is it possible to develop an in-house effective speech analytics framework in a cost-effective manner?



Analysis of Phone Conversation Recordings The Overall Framework

The focus is on agent voice recordings generated within an outbound survey.

The content of the recordings is typically not too heterogeneous (due to the presence of a script).





Analysis of Phone Conversation Recordings Transcription Phase

An in-house speech-to-text model has been developed, based on the framework Kaldi (https://kaldi-asr.org/) and the following corpora.

Corpus name	# uttera	nces	Recording time		
	training	\mathbf{test}	training	\mathbf{test}	
CLIPS	1025	-	2h 30m	-	
QALL-ME	1208	-	2h 20m	-	
Proprietary (read)	3467	-	4h 28m	-	
Proprietary (spontaneous)	201	339	$30 \mathrm{m}$	35m	

A word error rate of 28.77% is achieved, compared to 18.70% which can be obtained relying on Google Cloud Speech API. This is enough to perform some analyses over the transcripts.



Analysis of Phone Conversation Recordings Analysis of the Transcripts

Several kinds of analysis tasks may be performed over the obtained textual data.

For instance, it is possible to determine whether the agent has pronounced all the parts required by the script.

The overall idea is that of attaching tags to the transcribed phrases, in order to track the presence of different script parts.

This can be done based on user-defined regular expressions, or using some more advanced machine learning approaches.



Analysis of Phone Conversation Recordings Performance Figures

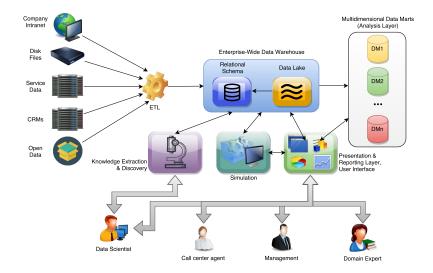
Performance obtained by several approaches, on the task of tag identification in the call transcripts.

	Accu	ıracy	Prec	ision	Re	call	\mathbf{T}	VR.
Keyword	\mathbf{K}	\mathbf{G}	\mathbf{K}	\mathbf{G}	\mathbf{K}	\mathbf{G}	\mathbf{K}	\mathbf{G}
Regular expressions	0.966	0.942	0.912	0.928	0.763	0.575	0.990	0.992
Logistic, unigram	0.972	0.973	0.903	0.916	0.839	0.870	0.989	0.973
Logistic, bigram	0.961	0.966	0.917	0.923	0.691	0.789	0.992	0.980
Logistic, trigram	0.940	0.951	-	0.910	0.494	0.666	0.995	0.895
Hybrid	0.974	0.973	0.886	0.894	0.886	0.896	0.985	0.985

The Overall Novel Infrastructure

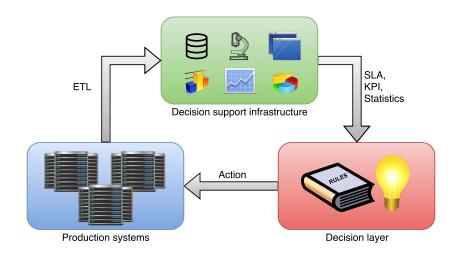


Decision Support System





Decision Management System





A. Brunello, P. Gallo, E. Marzano, A. Montanari, N. Vitacolonna, *An event-based data warehouse to support decisions in multi-channel, multi-service contact centers*, 2019.

A. Brunello, E. Marzano, A. Montanari, G. Sciavicco, *A combined approach to the analysis of speech conversations in a contact center domain*, in review.