

Enabling End-User Datawarehouse Mining Contract No. IST-1999-11993 Deliverable No. D17.2B

Call Center Case

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Abstract

This document presents the Call Center Case developed by National Institute of Telecommunications. This case has been developed for Mining-Mart system and is focusing on selecting the group of prospect clients for targeting marketing campaign. The call center is used as a communication channel with the clients. We are presenting a business problem, the data sources and preprocessing steps.

Chapter 1

Introduction

Dynamic growth of the telecommunications market forces telecommunications operators to take up necessary actions, leading to improve their competitiveness in the market. The business departments are supported by various information systems. Customer Relationship Management (CRM) systems become one of the most important tools of the modern telecommunications company. The data mining module is one of the essential component of such system.

The CRM process consists of four main phases:

- Acquisition and analysis of clients data
- Marketing planning
- Communication with client
- Analysis of the marketing action

In the above process the correct data is a solid base for efficiency and effectiveness of the CRM process. The data preparation is one of the most time consuming task and needs dedicated tools. MiningMart project goal was to provide software tools for advanced pre-processing of data. This paper presents a case that has been prepared for verification of MiningMart system as well as one of the examples of using MiningMart, stored in the case base of MininMart system.

We restrict our case to one communication channel (the call center) with the clients and we focus on selling services by this channel. The process of selling services by call center is as follows:

• a group of potential service subscribers is selected manually by the telecommunications specialists (*primary selection* - figure 1.1); the selection is based e.g. on invoice data but other rules are possible;

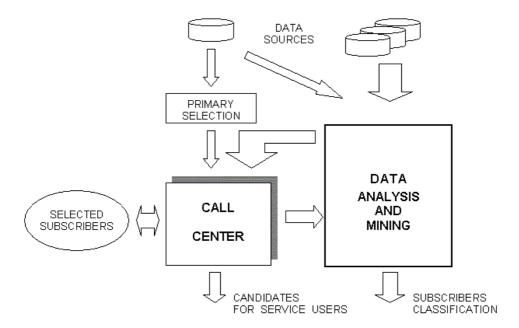


Figure 1.1: Data flow in the classification system

- these candidates are contacted by call center agents; the answers of the potential clients are stored in the data base;
- data stored in the call center, the list of real service users and CDR (Call Detail Records) are the data sources that are used for building data mining table; this table is an input for data mining algorithms;
- results of mining the list of potential candidates for buying the service
 is sent to the call center;
- the clients from the new list are contacted by an agents and a new responses data are stored
- the new data are used once again for building a data mining table and modeling; and this steps can be repeated as long as the list of candidates is not empty.

Simplified model of the call center is shown on the figure 1.2. We can distinguish there the input list of a clients that will be called. The call center agent calls potential client and asks him whether he is interested in a service subscription. The answer is coded as an integer number and kept in the database for further analysis:

- client is not interested in service subscription (code -1)
- client is already service subscriber (code 1)

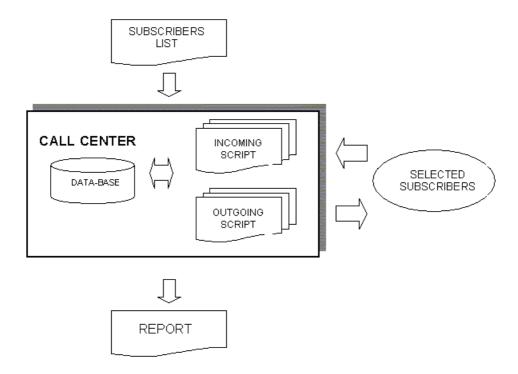


Figure 1.2: Simplified model of call center

• client agrees on service activation (code 2)

Additionally the company is storing the list of people being subscribers of the service. This source is even more reliable than the results of the call center review. Sometimes it happens that questioned people do not deliver real information. Call center questioning results will serve as description of the telecom company's clients preferences on offered service. In the case of the person which cannot be found in any of these data sets, we assign code 0, that means ,,unknown preference".

Each client's profile is created on the basis of data acquired from company's billing system. It is very typical problem of transformation clients activities in past (here we have Call Detail Records (CDR)) into their behavioral description. There are many CDR forms, but all of them contain such information on single call as:

- caller number
- called number
- timestamp
- length of the call

• tariff units per call

We use also class attribute, indicating type of call (e.g. local, long distance, etc.) and disconnection state (may be one of: properly disconnected, improperly disconnected, not answered).

Taking into consideration different tariffications between 8 a.m. and 18 p.m. and between 18 p.m. and 8 a.m. we distinguish calls performed within peak hours and within non-peak hours.

In our model, a client is characterized by the following set of attributes:

- number of calls
- number of calls within peak hours
- number of calls within non-peak hours
- number of different called parties
- number of specific type connections
 - local
 - long distance
 - abroad
 - cellular
 - special numbers prefixed with 0700 (party lines, etc.)
 - special numbers prefixed with 080 (info lines)
 - internet provider (dial-up modem connections)
- number of specific type connections (as above) within peak hours
- number of specific type connections (as above) within non-peak hours
- total length of calls
- total length of calls within peak hours
- total length of calls within non-peak hours
- sum of tariff units
- sum of tariff units for peak hours calls
- sum of tariff units for non-peak hours calls
- number of properly disconnected calls
- number of improperly disconnected calls
- number of not answered calls

Chapter 2

The data model

Three concepts are introduced at the beginning. CallData corresponds to data set containing CDRs. ActualClients has underlying database table holding the list of actual service subscribers. The results of questioning the clients by call center are present in the model as CCExamination concept.

At the beginning we present the class diagrams presenting data associations. Notation is based on that proposed in [1] (there are some extensions to standard UML class diagrams). We will use:

- (\langle base)\rangle stereotype for concept having direct correspondence at the database level,
- (\(\langle \text{aggregate}\)\)\) stereotype for concepts holding summaries of groups in the data (like \(sum, \count, \text{etc.}\)\).
- (\summarize)\rangle stereotype to illustrate the dependency between concept holding summarized data and the concepts the data that is to be summarized come from,
- ((construct)) stereotype for visualizing associations linking concepts used to construct new feature and concept holding that feature

Concepts corresponding to data sources are depicted in figure 2.1. There is one class for each concept.

Person and Client do not have their column sets, but we introduce them to show how the initial concepts are associated. The attribute that is used to identify particular client of the company is his phone number (ClientNumber or PhoneNumber or Caller in fig. 2.1).

Attribute CallClass of CallData allows us to distinguish between seven main types of calls. In the figure 2.2 one can find class connected with each type of call (LocalCallDetails, LDstCallDetails, etc.). There are also three classes for properly, improperly and not answered calls. Attributes of class Statistics are summaries of features of all these classes. Caller is group by

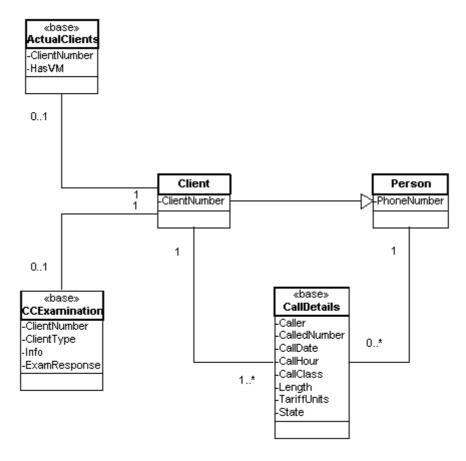


Figure 2.1: Initial concepts

attribute of *Statistics* class and as the consequence appropriate statistics are computed separately for subsets connected with clients.

The price of the tariffication unit differs in time intervals 8-18 (denoted as peak hours) and 18-8 (non-peak hours). CallDetailsPeak and CallDetailsNonPeak are derived from CallDetails. Additional client statistics features are computed by the summarizations (figures 2.3, 2.4). Statistics states complete clients' profiles description.

Decision attribute, that is client preference on the offered service, is constructed basing on the list of current service subscribers (Actual Clients) and an outcome of the call center review (CCExamination) - figure 2.5. The list of subscribers is more reliable source of an information. In the case the data is available in both of them, the value from this list must be taken.

All the statistics - clients profiles merged together with the decision attribute values - form data mining data set.

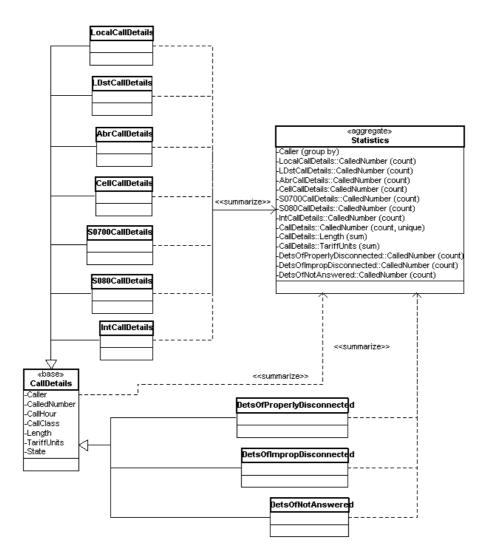


Figure 2.2: Summarization of features of CallData (1)

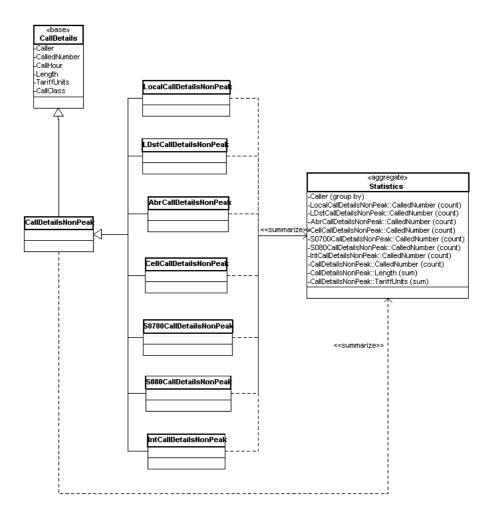


Figure 2.3: Summarization of features of CallData (2)

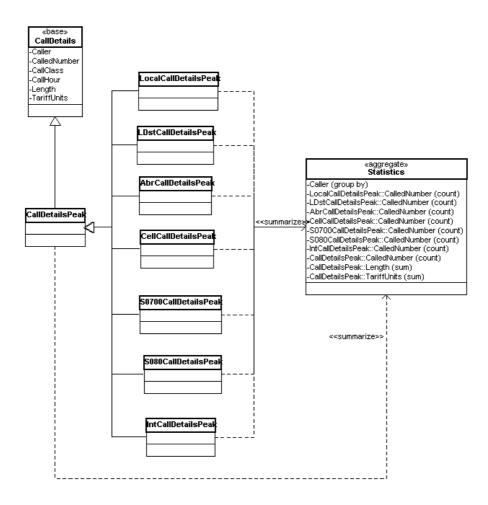


Figure 2.4: Summarization of features of CallData (3)

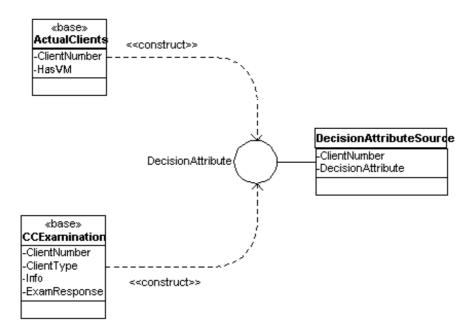


Figure 2.5: Decision attribute sources

Chapter 3

Case modeling by MiningMart

In this chapter we will present a process of building the pre-processing steps with the help of Human Computer Interface. We are expressing our data transformations in terminology of steps, concepts and features, that is specific to MiningMart modeling.

At the conceptual level we focus on just single client for calculating clients attributes. To deal with each client (or *Caller*) separately, *SegmentationStratified* operator is applied at the beginning (figure 3.1).

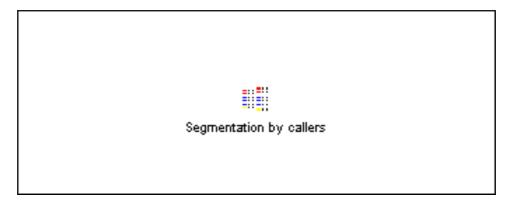


Figure 3.1: Segmentation by Callers

After that the new binary attribute is constructed (according to Call-Hour) that have value "peak" for peak hours calls and "nonpeak" otherwise. We use here TimeIntervalManualDiscretization operator. Subsets of peak hours and non-peak hours calls are selected (figure 3.2) by RowSelectionBy-Query operator.

Three possible values of the *state* attribute indicate three states of the call disconnection, that may be:

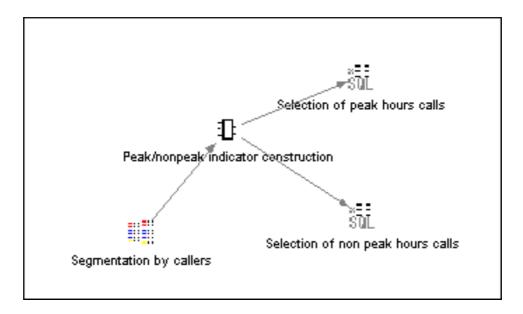


Figure 3.2: Selection of peak/non-peak hours calls

- disconnected properly,
- improperly disconnected broken because of e.g. technical reasons
- not answered by the called party.

Three subsets of calls differing in disconnection state are selected.

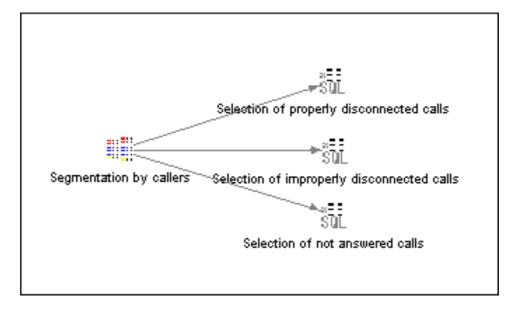


Figure 3.3: Subsets of calls selected according to disconnection state

Peak and non-peak hours calls are again divided into 7 subsets according to classes of calls (local, long-distance, abroad, cellular, 0700, 080, internet provider), see figure 3.4.

The steps introduced above, effects in creation of some new concepts, that enhance available information on client. We have concepts with details of client's calls within peak hours and non peak hours, specific type calls details (like local calls details), details of calls disconnected properly, improperly, not answered and so on. But still we have no client profile that we need. Interesting client's features may be now calculated with the help of SpecifiedStatistics operator. It may count calls, sum up durations of calls, compute total amount of tariff units, count number of different called numbers. In the figures 3.4, 3.5 one can see an application of SpecifiedStatistics. As the input concept we take information, mentioned above. The output concept of the step embedding the operator holds interesting, from our point of view, features of the client.

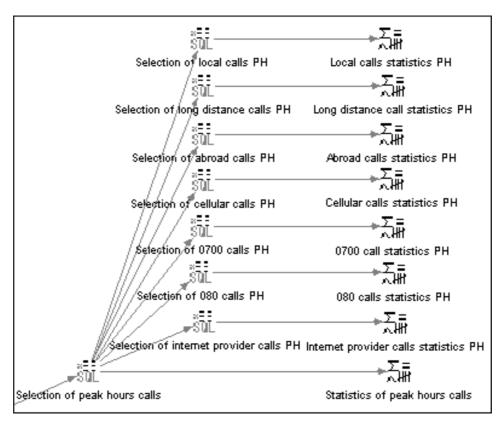


Figure 3.4: Statistics of peak hours calls

The first step we introduced while modeling the case was Segmentation-Stratified operator. From that point, data set underlying to the input concept CallDetails, was split into subsets, one for each client. Then identical

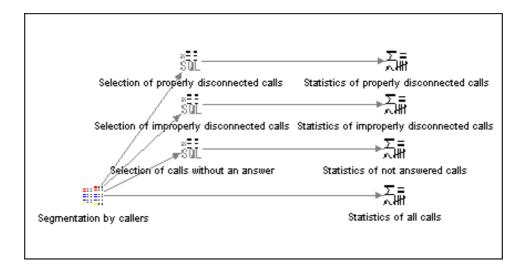


Figure 3.5: More statistics

operations in subsequent steps were performed on these subsets independently and we could simply focus on single client information processing. Now we would like to have the concept corresponding to features of all the clients. To realize this task we use *Unsegment* operator, that is inverse to any operator of type *Segmentation* (figure 3.6). Our focus changes back from one client to all the clients found in initial data set.

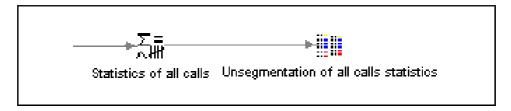


Figure 3.6: Unsegment - inverse of the segmentation

The *Unsegment* steps give as the output the total number of 20 concepts. Features of these concepts, assembled together, constitute almost complete profiles of the clients we want to achieve. *JoinByKey* operator is the best way to create a new concept with all currently available clients' features (see figure 3.7).

In order to build full profiles we need additionally attributes describing e.g. number of local calls or number of all calls, without differentiation of peak hours and non-peak hours calls. The goal is achieved here by summing up these numbers computed by SpecifiedStatistics operator for peak and non-peak hours calls. GenericFeatureConstruction is applied for this purpose (figure 3.8).

We have prepared clients' profiles. Now we can construct the decision

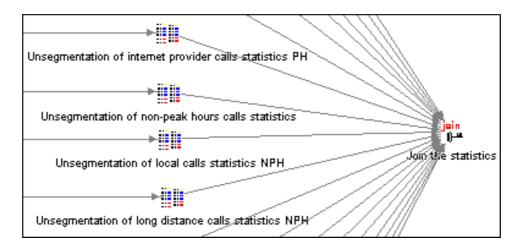


Figure 3.7: All the statistics gathered

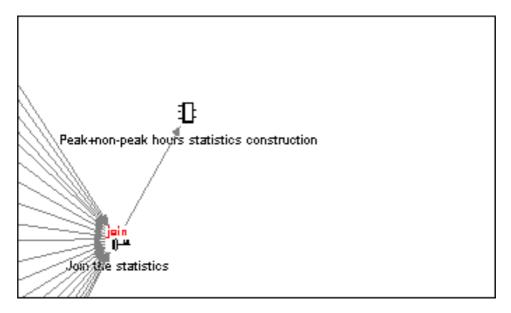


Figure 3.8: Construction of new features

attribute (figure 3.9). Actual Clients and CCExamination are the available sources for the decision attribute. With the use of UnionByKey operator we join both the sources. The operator JoinByKey joins data underlying to its concepts according to key attribute values. The output column set of the operator contains only records marked with common for all its input column sets subset of key attribute values (intersection). Here we would like to have a tuple in the output columnset for every key attribute value found in any input columnset (union of keys rather than intersection).

Two attributes of the output concept UnifiedDecAttrSources of this Union-

ByKey step are now considered, namely ExamResponse (from CCExamination) and HasVM (from ActualClients). As mentioned before, the value present in the clients list delivered by the company is more reliable source than outcome of the call center review.

Having this on mind, order of actions to take looks as follows:

For every key attribute value found in *UnifiedDecAttrSources* concept:

- 1. Look for *NOT NULL* value of *HasVM* attribute. If found there, assume it to be the newly constructed decision attribute value.
- 2. If the value of HasVM for the particular key is NULL, take ExamResponse value.

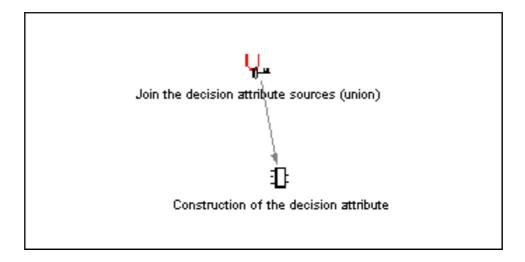


Figure 3.9: Decision attribute construction

The last use of UnionByKey attaches the decision attribute to profiles created before (see figure 3.10).

There are many of clients for whom no decision attribute have been found, neither in actual clients list, nor in the response given to the call center review. We fill (figure 3.11) these missing values of the decision attribute with the default value of 0 (which means ,,clients relation to offered service is unknown").

To improve the efficiency of mining algorithms, some of the attributes are discretized manually. During profile data analysis, we noticed that for majority of callers some of the attributes, like for example number of calls to party-lines, take value 0. We decided to create a binary attribute, reflecting the fact that the client did or did not make such a type of call (figure 3.11).

Figure 3.12 gives an overall look at all the preprocessing steps.

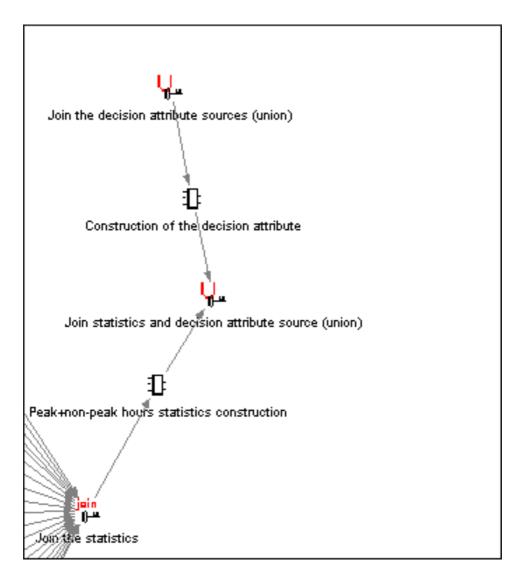


Figure 3.10: Adding decision attribute to clients' profiles

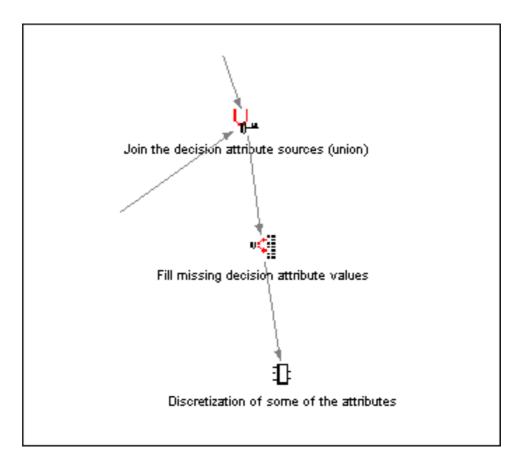


Figure 3.11: Final steps

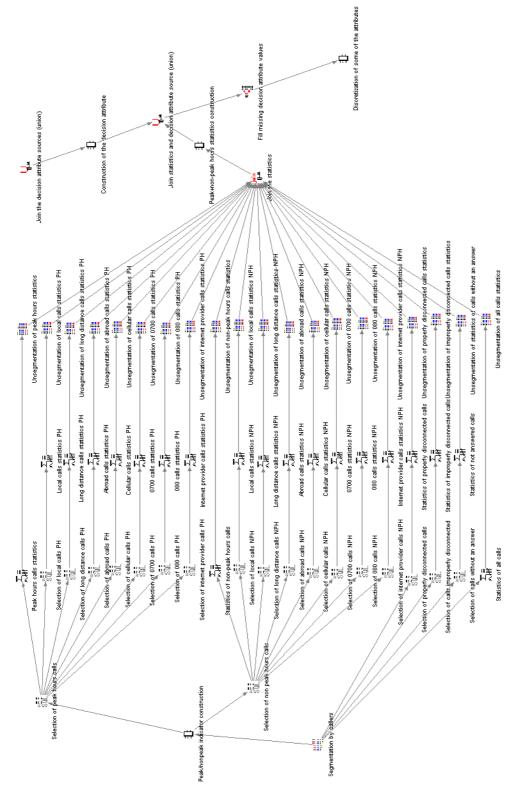


Figure 3.12: Whole preprocessing schema in HCI

Bibliography

[1] R. Bathoorn N. Brandt M. de Haas O. Rem. Problem modeling. MiningMart Deliverable D19 IST research project, 2001.