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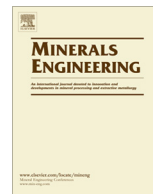
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Designing gold extraction processes: Performance study of a case-based reasoning system



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ABSTRACT

This paper presents a method for externalising and formalising knowledge involving the selection of hydrometallurgical process flowsheets for gold extraction from ores. A case-based reasoning (CBR) system was built using an open source software myCBR 3.0. The aim of the systems is to recommend flow-sheet alternatives for processing a potential gold ore deposit. Nine attributes: *Ore type*, *Gold ore grade*, *Gold distribution*, *Gold grain size*, *Sulfide present*, *Arsenic sulfide*, *Copper sulfide*, *Iron sulfide* and *Clay present* were modelled and several literature sources of actual gold mines and processes were used for acquiring cases for the system. After preliminary testing, functional evaluation of the built CBR system was carried out by using five real mining projects as test cases. Additionally, human experts in the field of gold hydrometallurgy were interviewed to demonstrate the benefits of the CBR system as it holds no human biases towards any processing techniques. It was found that the suggestions of the CBR system provided useful information and direction for further process design and performed well compared to the interviewed human experts, thus confirming that the system is of practical relevance to the process engineer designing an industrial gold processing plant. The current model was found to be a functioning basis for further development through additional attributes, adjusted attribute weighting and increased number of cases.

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1. Introduction

The governing method for gold ore processing has been cyanide leaching since the late 19th century (Marsden and House, 2006). After decades of active development of the process for various types of ores and concentrates, there are several different hydrometallurgical flowsheets for cyanide leaching. As ores differ greatly, the flowsheet needs to be tailored for the deposit in question. The process design is initially based on existing knowledge and then on experimental results. The amount of information available in journal articles and industry reports concerning the processing of gold ores is large and increases continuously. Therefore, the challenge is not the task of acquiring knowledge, but

rather the task of managing, classifying and performing comparative analysis of the available information. Efficient exploitation of the existing information aids the professional in defining the needed experiments for developing a process flowsheet for an ore of interest, and in consequence of that, achieve bench and pilot scale experiments sooner. Additionally, rapid financial analysis and cost evaluation of possible flowsheets can be made more attainable through effective comparison techniques. It is well known that ore mineralogies and composition often change within the same deposit. If these variations are known before planning the initial processing plant, comparing possible processes for the different mineralogies in the deposit can lead to a compromise that remains more feasible over time.

Modelling all facets of a processing plant with a vast number of straightforward rules and deterministic equations is highly challenging, as the available data is often incomplete and fuzzy (Rintala et al., 2012, 2015). Instead, the target of this study is to develop a software system that is able to give starting points for gold ore process design by helping the user to remember and compare previously successfully applied processing options on similar

Abbreviation: CBR, case-based reasoning.

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mining sites (Sauer et al., 2013, 2014). To develop such a decision support method, systematic knowledge formalisation is required (Kolodner, 1992).

The three most prominent reasoning methodologies available to create a decision support system are rule-based, case-based and model-based reasoning. Of these three, only case-based reasoning (CBR) is able to handle incomplete and fuzzy knowledge in a way suitable for recommending hydrometallurgical process alternatives (Rintala et al., 2011), and was therefore chosen as the reasoning methodology for this study. CBR has already been applied in various fields of engineering and process design. To name a few examples, Vong et al. (2002) have utilised CBR to support hydraulic production machine design, and Seuranen et al. (2005) have studied how to develop a method for recommending feasible separation process sequences and a separation process structure in chemical technology.

CBR uses the knowledge of past problems, cases, and predicts the likely outcome or applicable solution to a current problem. It performs this prediction based on the knowledge stored in previous cases which are gathered in a case base (Aamodt and Plaza, 1994; Richter, 1998). The knowledge is stored in the case's various attributes, such as pH, chemical formula, price, location, symptom, colour, etc. The current problem is formulated into a case by defining its attribute values and is referred to as the query case.

When using the CBR system, a user makes a query by entering values for each attribute and then the system retrieves cases from the case base organised by their similarity with the query. These similarity measures get values between 0 and 1, the former denoting that the query case and retrieved case are completely dissimilar and the latter indicating that they are identical. The total similarity (global similarity) between case and query is a result of the combination of attribute specific similarities (local similarities) by applying a suitable amalgamation function. When a case consists of n attributes, the global similarity, $Sim(q, c)$, between query q and case c in the case base is calculated as the weighted sum of the attribute specific local similarities according to Eq. (1) (Stahl and Roth-Berghofer, 2008):

$$Sim(q, c) = \sum_{i=1}^n \omega_i \cdot sim_i(q_i, c_i) \quad (1)$$

Here sim_i and ω_i denote the local similarity measure and the weight of attribute i .

The aim of this research is to construct and study the functionality of a CBR system, designed to recommend possible processing flowsheets for a gold ore of interest. The CBR methodology is applied to compare and rank process alternatives based on similarities between ore properties as defined by the selected nine attributes. Additionally, the constructed CBR system is tested through preliminary retrieval tests and its functionality is evaluated against the expertise of senior level hydrometallurgical experts.

2. Methods

This section describes the construction of the CBR system, methodology of the retrieval tests and interviewing techniques applied during knowledge acquisition.

2.1. Knowledge formalisation

The knowledge formalisation described in this paper was performed using the open source similarity-based retrieval tool myCBR in its latest version 3.0 (myCBR, 2012). The myCBR tool offers a set of graphical user interphases (GUIs) called myCBR workbench, which can be employed for rapid knowledge modelling and prototyping of CBR systems (Stahl and Roth-Berghofer,

2008). This specific CBR tool was selected due to its various useful functionalities such as the possibility to model several local similarity measures for one attribute and then select which one is used in the retrieval step.

2.1.1. Defining case attributes

At the beginning of knowledge formalisation, the relevant entities in the domain need to be identified, as well as their relationships with each other. In this study, the relevant entities were the mineralogical properties of gold ores. Marsden and House (2006) have suggested that after determining the gold mineral type, the ore composition, especially the concentration of gold, other valuable minerals, and minerals detrimental to processing, must be determined prior to gold process design. They also discuss the importance of gold grain size distribution and liberation characteristics of valuable minerals. In this study, nine attributes: *Ore type*, *Gold ore grade*, *Gold distribution*, *Gold grain size*, *Sulfide present*, *Arsenic sulfide*, *Copper sulfide*, *Iron sulfide* and *Clay present* were modelled.

Gold mineral type, referring to the most general description of the ore, such as “Free milling” or “Silver rich”, and gold concentration, or *Gold ore grade*, were relatively straightforward to model into attributes. Other valuable minerals, such as silver, were not seen being as characterising as gold with regards to process design and profitability. Overall mineralogical composition is also important, but significantly more complicated to model into attributes. However, some minerals are more influential than others. The flowsheet design is significantly different for sulfidic gold ores compared to other types, such as free milling ores, because sulfides consume cyanide during leaching. Therefore, three mineral attributes were selected to describe the sulfidic mineralogy of the ore: *Arsenic sulfide*, *Copper sulfide*, and *Iron sulfide*. Additionally, a simple attribute stating the presence of sulfides, without determining the kind of sulfidic mineral was included in the model. Another aspect of ore composition that affects the process design is the presence of clay; hence, an attribute *Clay present* was included in the system. Clay minerals reduce the gold dissolution rates, whether directly associated with the gold, or just present in the ore. Clays tend to hinder the cyanidation process for example by forming impermeable coatings over the surface of the gold which develop after grinding (Gasparrini, 1993). Gold grain size distribution is often described rather vaguely in literature with terms such as “Fine grains”. It was however included in the model, despite the possible loss of information related to its modelling. The liberation characteristics of all valuable minerals affect the processing methods, but gold was seen as the most defining. Therefore, the attribute *Gold distribution* was formulated to model the mode of gold occurrence as either “Free” or “Enclosed in mineral”.

In conclusion, the following attributes were selected to be modelled in the first version of the CBR system: *Ore type*, *Gold ore grade*, *Gold distribution*, *Gold grain size*, *Sulfide present*, *Arsenic sulfide*, *Copper sulfide*, *Iron sulfide*, and *Clay present*.

2.1.2. Case representation

Attribute-value pairs were selected for case representation, describing the mineralogy of an industrially utilised gold ore/concentrate. In myCBR the user can select from several attribute data types, which indicate the nature of the attribute. Examples of data types are numerical values and symbolic values, such as names of substances. The attribute types employed in the built system were symbols, Boolean, and floating point numbers. The attributes and their respective data types are presented in Table 1.

The flowsheets related to the ores in the case base were also gathered to be used as starting points for process design for the ore of interest i.e. the queried ore. The flowsheets were formalised into a separate data base, where the user can examine them. Sim-

Table 1

Case representation including the data types of attributes.

Attribute	Possible values	Data type
Ore type	Carbonaceous, Copper rich, Free milling, Refractory Arsenopyritic, Refractory antimony sulfide, Refractory iron sulfide, Silver rich, Telluride	Symbol
Gold ore grade	[g/t]	Floating point number
Gold distribution	Free, Grain enclosed in mineral	Symbol
Gold grain size	Coarse, Fine, Micronized, Sub-micronized	Symbol
Sulfide present	Yes, No	Boolean
Arsenic sulfide	Arsenic sulfide, Arsenopyrite, Orpiment, Realgar, Any, None	Symbol
Copper sulfide	Bornite, Chalcopyrite, Copper sulfide, Covellite, Digenite, Any, None	Symbol
Iron sulfide	Iron sulfide, Marcasite, Pyrite, Pyrrhotite, None	Symbol
Clay present	Yes, No	Boolean

Table 2

Structure of flowsheet formalisation demonstrated by two examples; East Driefontein and Sao Bento.

Process step	East Driefontein	Sao Bento
Ore type	Free milling	Ref. arsenopyritic
Gold ore grade [g/t]	8	7.2
Comminution	78% < 75 µm	75% < 75 µm
Enrichment	TH	FL, G, TH
Pretreatment 1	Alkaline preaeration	Acidic pretreatment
Pretreatment 2		Acidic PO
Leaching	Agitated CL	Agitated CL
Recovery 1	CIP	CIP/CIL
Recovery 2	Carbon elution	AARL elution
Recovery 3	EW	EW
Refining	Smelting	Smelting
Product	Bullion	Ag-Au bullion

AARL = Anglo American Research Laboratory method, CL = Cyanide leaching, CIP = Carbon-in-pulp, EW = Electrowinning, FL = Flotation, G = Grinding, PO = Pressure oxidation, TH = Thickening.

ilarity modelling of the flowsheets was not in the scope of this research. Table 2 presents the structure of the flowsheet formalisation by two example cases: East Driefontein and Sao Bento. The latter utilises gravity concentration and then the concentrate and tailings are processed in separate streams.

2.1.3. Similarity measures

The local similarity of the attribute *Gold ore grade* was modelled by a symmetric distance function illustrated in Fig. 1 and defined by Eq. (2).

$$y = -\left| \frac{1}{71}x \right| + 1 \quad (2)$$

Here y expresses the local similarity value and x indicates the remainder of the two *Gold ore grade* values in the queried ore and the case. The denominator in Eq. (2) is defined by the maximum difference between *Gold ore grade* values in the case base. For example, if two values for this attribute are compared using this function: Value A:50 and Value B:40 (both expressed in g/t), then $x = 10$, which calculates to a similarity of B being 0.86 “similar” to A.

The attributes *Clay present* and *Sulfide present* are Boolean attributes, and can therefore have only the values Yes or No, and the local similarity between them is exactly 0 or 1. Other selected attributes, *Ore type*, *Gold distribution*, *Gold grain size* and *Arsenic*, *Copper* and *Iron sulfide*, were modelled by comparative similarity tables. The 10th edition of the Nickel–Strunz classification of minerals (Mindat, 2014) was used as a reference for local similarity models of minerals. Each mineral in the Nickel–Strunz classification has an assigned class code depending on the elements and ions within it. The general format of the code is two numbers, two letters, and two numbers with a possible additional letter, if there are more than one mineral having the same classification code: XX.YY.XX (y). There are some exceptions to the general format, but none of

these were included in the case base. Only the attribute values that were used in the case base were modelled.

The following rules for local similarities of minerals were formulated based on the Nickel–Strunz classification:

- If the first numbers of the mineral class code are equal, local similarity is 0.5.
- If the number and the first letter of the mineral class code are equal, local similarity is 0.7.
- If the number and both letters of the mineral class code are equal, local similarity is 0.8.

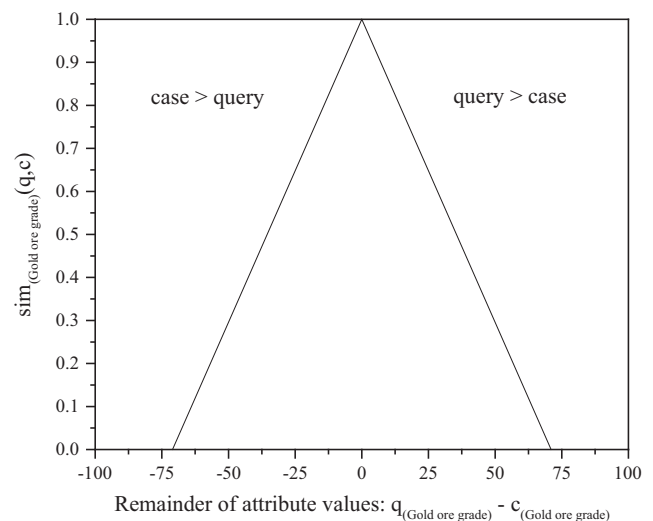


Fig. 1. Difference-based local similarity measure for the attribute *Gold ore grade* between the query q and case c .

Table 3The local similarity table for the symbol type attribute *Arsenic sulfide*.

Query value	Case value				
	Arsenopyrite	Any	None	Realgar	Orpiment
Arsenopyrite	1	0.5	0	0.5	0.5
Any	0.5	1	0	0.5	0.5
None	0	0	1	0	0
Realgar	0.5	0.5	0	1	0.8
Orpiment	0.5	0.5	0	0.8	1

- If the main class code is equal, but the additional letter is different, local similarity is 0.9.
- If the class code is identical, local similarity is 1.00.

In the instance of the attribute *Arsenic Sulfide*, five different values were modelled; three different minerals and the values of “None” and “Any”. The Nickel-Strunz classification codes for arsenopyrite (FeAsS), orpiment (As₂S₃) and realgar (α-As₄S₄) are 02.EB.20, 02.FA.30 and 02.FA.15a respectively and the local similarity model in Table 3 was created according to the formulated rules. The rules for the “None” and “Any” values were the following: the value “Any” is 0.5 similar to any mineral in the similarity table of a specific mineral group, here *Arsenic sulfide*, and the value “None” is similar only with itself.

2.2. Case construction

After the case structure was formalised, a case base could be compiled. The cases were constructed by extracting the mentioned nine ore attributes from the case sources and transferring them into the attribute-value tuples presented in Table 1. At the moment, it is not possible to use several ore descriptions for one case. However, such a deposit that is a combination of more than one ore type can be added in the case base as two cases.

2.3. Retrieval tests

All the retrieval tests were performed using the built-in retrieval tool of the myCBR software. The preliminary retrieval tests were performed using one attribute-value pair at a time and then increasing the number of attribute-value pairs used, while verifying the validity of the calculations. These hypothetical queries enabled the investigation of the quality of local similarity models, the calculation of global similarities according to Eq. (1), and the consideration of possible weighting factors.

The flowsheet recommendation ability of the built CBR system was tested using mineralogy and process information of actual gold mines (Marsden and House, 2006) that were not included in the case base. These five test queries T1–T5 are presented in Table 4. The objective of this test phase was to investigate how similar

process flowsheets are between the query case and the best matching cases. Additionally, these results could later be compared with process flowsheet proposals by human experts.

Process flowsheets of the actual gold processing plants applied for the test queries T1–T5 (Table 4.) are presented in Table 5. All of the queried ores are processed in two separate streams in their original flowsheets.

2.4. Interviewing technique

Interviewing techniques can be divided into three categories based on the predefined questions and their control over the course of the interview; structured i.e. a form, semi-structured or theme interview and unstructured or open interview (Preece et al., 2002). In this study experts were interviewed with a semi-structured interviewing technique where the questions were predefined, asked in the same order and no additional questions were asked. Because the aim of the interviews was not quantitative analysis of the results, but to simply compare the qualitative differences of human experts and the CBR system, the number of interviews was set at three.

The interviews were conducted with experienced hydrometallurgical experts, Interviewees I, II and III, who have worked with gold processing for several years. Their experience level can be described as follows:

- I: 30 years of metallurgical experience in industry of which 11 years in gold processing,
- II: 26 years of experience in extractive metallurgy of which 18 years in academia focusing on non-ferrous metals, especially gold,
- III: 33 years of non-ferrous process engineering, R&D and metallurgical plant design with the latest 5 years exclusively dedicated to gold hydrometallurgy.

They were presented with the original descriptions of the ores that were used to construct the test queries T1–T5 in Table 7. An example of such a description is the way Marsden and House (2006) described the ore mined in Homestake Lead, operated by Barrick Gold Corporation until shutdown in 2000. This description was used for test case T1 and was presented to the interviewees in its original form, seen here:

“Fine free gold and minor silver are associated with a predominantly chloritic-quartzite gangue. Small quantities of pyrrhotite and other minor sulfides occur in the ore”
[Marsden and House, 2006]

Also the gold concentration [g/t] and the location of the deposit were given to the interviewees. They were then asked to draw a block flowsheet they would propose for the described ore. During the interviewing situation, the interviewer was to not interfere with the interviewees’ answering technique after the instructions

Table 4

The queries T1–T5 formed based on the five actual gold mines (Marsden and House, 2006). The attributes that were not used in the query are marked by z.

Attribute	Query				
	T1	T2	T3	T4	T5
Ore type	Free milling	Free milling	Refractory arsenopyritic	Refractory iron sulfide	Carbonaceous
Gold ore grade [g/t]	4.7	6.0	2.4	2.2	8.0
Primary gold distribution	Free	Free	Grain enclosed in mineral	Grain enclosed in mineral	Free
Primary gold grain size	Fine	Fine	Fine	Micronized	Fine
Sulfide present	Yes	Yes	Yes	Yes	Yes
Arsenic sulfide	z	z	Arsenopyrite	Arsenopyrite	z
Iron sulfide	Pyrrhotite	z	Pyrite	Pyrite	z
Clay present	No	Yes	No	No	No

attribute and the knowledge loss is smaller. Though, this approach shifts the challenge to the retrieval phase. An additional layer must be added into the similarity calculation, so that the retrieval finds also the cases where the values are similar to the query, but the order of them is vice versa.

3.2. Retrieval tests

Preliminary testing of the functionality of the calculation algorithms were performed on a simplified case base of 25 cases and the final retrieval tests, based on actual gold processes, were carried out using the full case base of 48 cases.

3.2.1. Results of the preliminary retrieval tests

The similarity values of the retrieved cases were compared to the manually calculated similarities of the queried attribute values as they should be equal according to Eq. (1). The results of the preliminary retrieval tests showed that all the global similarity values matched the manually calculated ones, and were therefore correct. However, the retrieval results indicated a need for further development of modelling the following attributes:

1. *Gold ore grade*: Increased impact of larger differences in the attribute value. An advanced function, being modelled in the myCBR tool, using its function editors, could be utilised as the local similarity measure, instead of the simplified polynomial function.
2. *Gold grain size*: Lower weighting of the attribute should be used to reduce the role of this attribute.
3. *Clay present*: As the presence of clay has a great impact on processing requirements, the weighting of the attribute could be used to emphasise its role in the global similarity. Besides weighting of the attribute in the global similarity calculation, it is possible to use certain attribute values to exclude cases from the retrieval results. In the matter of *Clay present*, the cases that are not equivalent to the queried value for *Clay present* could be excluded.

As an example of the global similarity distribution for one hypothetical set of queried attribute values, Fig 2 shows all the cases

within the simplified case base, from most similar to least similar compared to the query.

It was shown that the built CBR system is able to retrieve cases from the case base using a set of selected attributes and rank them. The cases with the highest global similarities would then be used as starting points for further gold ore process flowsheet development for the ore of interest. It is up to the user to decide what global similarity level is used as acceptance level. In the query results shown in Fig. 2 the user might select e.g. three, five or seven most similar cases.

3.2.2. Retrieval tests using actual mineralogies of gold mining projects

The five best matching and two worst matching cases retrieved and their similarities with the queries T1-T5 are presented in Table 6.

The best matching case for Query T1 is the East Driefontein deposit, with a global similarity value of 0.86. The ores in both cases are relatively similar, both ores are free milling, in which the gold is distributed as fine free grains and the ores contain sulfides, of which pyrrhotite is mentioned. The differences of the cases are that in the query case there is no clay present, but in the description of the East Driefontein the presence of clay is undefined. The process flowsheets of T1 and East Driefontein are fairly similar as well. Both ores are crushed and ground to somewhat similar size. In both cases alkaline preaeration is utilised before agitated cyanide leaching. The main difference in the flowsheets is that in the case T1 the sand fraction of the ore is treated by vat cyanide leaching, whereas in East Driefontein the whole ore fraction is treated in a cyanide leaching reactor.

The best matching cases for Query T2 are the Kidston and Pine Creek cases, with global similarity values of 1.0. All the ores are free milling ores with fine gold particles and clay present. In the case source (Marsden and House, 2006) the ore of the case T2 is described as follows: "Epithermal deposit of free-milling, oxidised ore overlie refractory sulfides. Major gangue minerals are limestone, dolomite, and sandstones. Orebody contains heavily silicified regions and between 10% and 20% clay. Fine free gold and electrum, and minor mercury (1–20 g/t) occur." The ore of the Kidston case is described as follows in the case source:

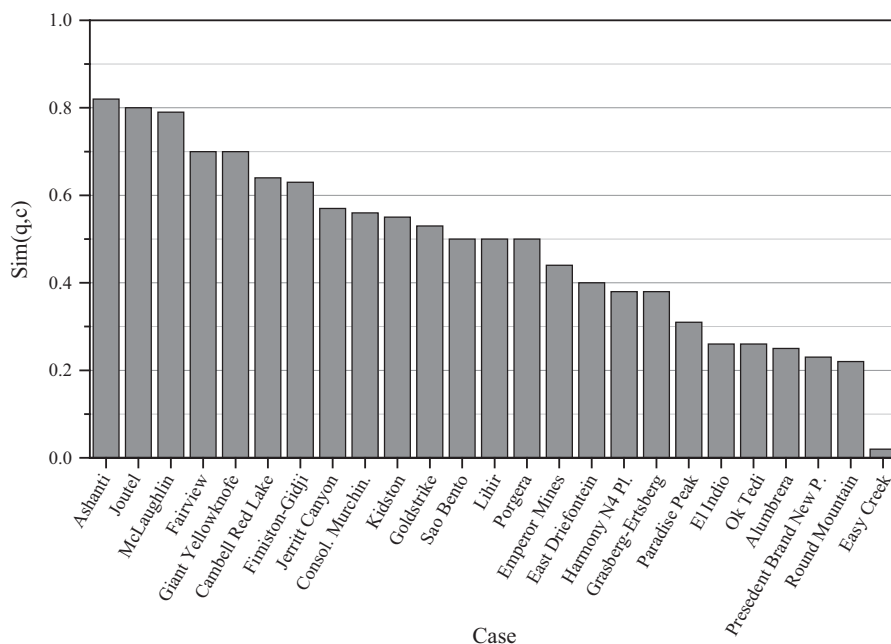


Fig. 2. Distribution of the global similarity values for one hypothetical query performed on the simplified case base of 25 cases.

Table 6

The most similar (1st–5th) and dissimilar (47th–48th) cases for the test queries T1–T5, global similarity was calculated by Eq. (1) and has a value between 0 and 1.

Paired Cases	Query T1		Query T2		Query T3		Query T4		Query T5	
	Case	Sim	Case	Sim	Case	Sim	Case	Sim	Case	Sim
1st	East Driefontein	0.86	Kidston	1.00	Ashanti	1.00	Barneys Canyon	0.88	Grasberg-Ertsberg	0.83
2nd	Fort Knox	0.71	Pine Creek	1.00	Cambell Red Lake	0.88	Barneys Canyon	0.88	Fort Knox	0.67
3rd	Fort Knox	0.71	Fort Knox	0.83	Fairview	0.88	Joutel	0.88	Fort Knox	0.67
4th	Yanacocha	0.71	Fort Knox	0.83	Giant Yellowknife	0.88	Ashanti	0.75	East Driefontein	0.67
5th	Yellowknife	0.71	East Driefontein	0.83	Joutel	0.88	Cambell Red Lake	0.75	Yanacocha	0.67
47th	Lihir2	0.29	Easy Creek	0.33	Ok Tedi	0.25	Harmony N4 Plant	0.13	Porgera2	0.33
48th	Porgera2	0.29	Porgera2	0.33	President Brand New Plant	0.13	President Brand New Plant	0.13	Barneys Canyon	0.17

Table 7

Interviewing results from the three senior experts.

Query	Interview	Enrichment	Pretreatments	Leaching technique
T1	I	Flash flotation Knelson gravity separation	Pretreatment	Agitated cyanide leaching
	IIa		Pressure oxidation	CIL
	IIb		Atmospheric pre-oxidation	Intensive cyanidation
	IIc		Pretreatment	Gold room (melting)
	III			Agitated cyanide leaching
T2	I	Flotation	Roasting	CIL
	II		Alkaline pressure oxidation	CIL
	IIIa		Roasting/Pressure oxidation	Agitated cyanide leaching
	IIIb			Agitated cyanide leaching
T3	I	Flotation	Pressure oxidation	CIL
	II		Alkaline pressure oxidation	Thiosulfate leaching with ion exchange for recovery
	IIIa		Fine grind	CIL
	IIIb			CIL
T4	I	Flotation	Pressure oxidation	CIL
	II		Partial (60–80 %) acidic pressure oxidation	CIL
	III		Regrind	CIL
T5	I	Flotation	Roasting	CIL
	II		Roasting	CIL
	III		Roasting	Agitated cyanide leaching

“A predominantly free-milling ore with oxidised, transition, and sulfide zones were present in volcanic breccia. Gangue was mainly quartz, muscovite, chlorite, and carbonates with moderate clay (kaolinite) content. Some pyrite (<2%) was present. Fine free gold was present, some intimately associated with pyrite. Copper mineralisation was variable. Free gold was liberated at approximately 53 μm .” These descriptions show the extent of the knowledge loss, when only a few attributes are used to describe the ore in the knowledge formalisation. However, the attributes employed are appropriate as the three cases are notably similar and the process flowsheets of the cases are rather similar as well. No pretreatment is employed and all ores are treated with agitated cyanide leaching followed by either carbon-in-pulp, carbon-in-column or both. The main difference in the flowsheets is that in the case T2 low-grade material is treated by heap cyanide leaching, whereas Pine Creek and Kidston apply only reactor leaching.

The best matching case for Query T3 is the Ashanti plant, with a global similarity value of 1.0. All the queried values are similar in these two cases. However, the process flowsheets are slightly different. The refractory sulfide ore (Au 2.0 g/t) of the case T3 is pretreated by nonacidic pressure oxidation before leaching by carbon-in-leach. The oxidised ore of the case T3 is treated by carbon-in-leach followed by pressure zinc precipitation. In the Ashanti case no pretreatments, such as pressure oxidation, are used and agitated cyanide leaching is employed as the leaching technique. This shows that the most similar case does not necessarily present an identical flowsheet with the query. Nevertheless, the idea of the CBR tool is that the system user can utilise the cases with highest global similarities as starting points for further

process development. By observing the four next similar cases to case T3 (Cambell Red Lake, Fairview, Giant Yellowknife and Joutel with global similarity values of 0.88), it is observed that in all these next similar cases a pretreatment is used prior to cyanide leaching. This indicates to the system user that the option of using a pretreatment is likely to be beneficial, although not applied in the most similar case (Ashanti). In one of the next best matching cases (Giant Yellowknife) carbon-in-leach is utilised as in the case T3. To illustrate the global similarity distribution of the entire case base, all the cases and their similarities with T3 are illustrated in Fig 3.

The best matching cases for Query T4 are the two Barneys Canyon cases and the Joutel project, with global similarity values of 0.88. The ores in all cases are refractory iron sulfides, in which the gold grains are enclosed in mineral. The ores contain sulfides, of which pyrite and arsenopyrite are mentioned. Joutel differs from T4 in gold grain size and the Barneys Canyon ores contain clay, unlike T4. However, the process flowsheet of Joutel is much more similar to T4 than the Barneys Canyon cases. In T4 and Joutel the ores are pretreated, by pressure oxidation and preaeration respectively, before cyanide leaching. In the dissimilar Barneys Canyon cases the ore is heap leached without any pretreatment. This underlines the benefit of CBR, as it provides several options for the user to compare.

The best matching case for Query T5 is the Grasberg-Ertsberg mining project, with a global similarity value 0.83. The process flowsheet industrially applied in the case T5 is very different from the process flowsheet of Grasberg-Ertsberg. In the case T5 majority of the fine carbonaceous material is processed by conditioning with kerosene prior to resin-in-pulp treatment. The sulfide

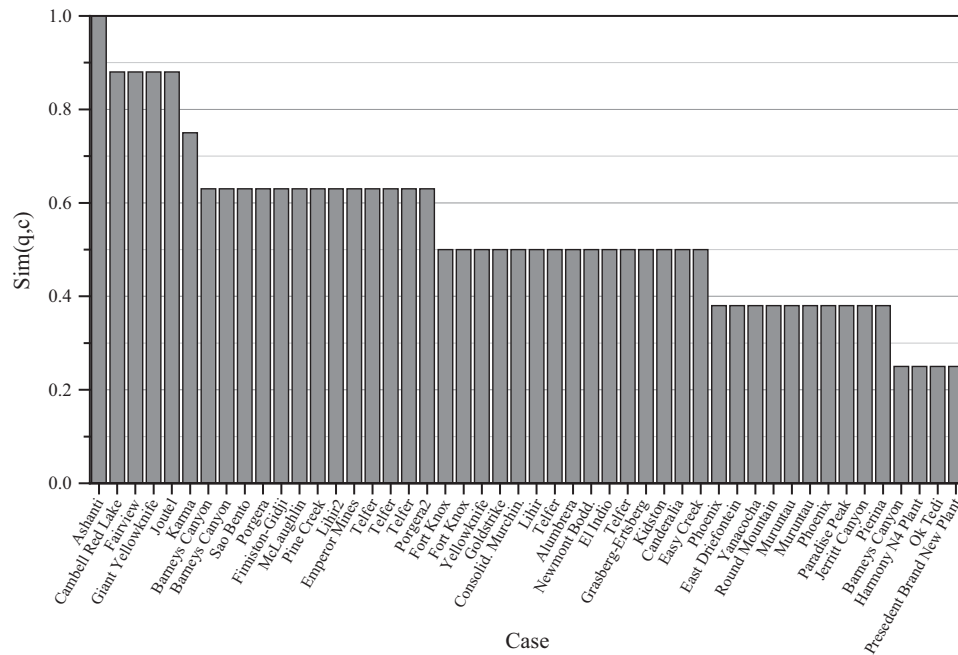


Fig. 3. The distribution of the global similarity values for Query T3 performed on the full case base of 48 cases.

concentrate of the case T5 is treated by using intensive cyanide leaching. In the Grasberg-Ertsberg case the process employed produces merely Cu-Au concentrate by flotation, which is further processed in copper smelters. Actually none of the process flowsheets of the best matching cases are similar to the process flowsheet industrially applied in the test case T5. Even though more attributes were used in the query, the system could not find similar raw material that have similar process flowsheet, as at the moment there are no similar process flowsheets applied for any of the cases in the case base.

The retrieval results for all of the queries demonstrate the inherent capability of CBR to handle fuzzy knowledge. Overall, the results were at least satisfactory for T1–T4 providing similar process flowsheets as either the most similar case (T1, T2 and T4) or in the five most similar cases (T4). The CBR system provides results not only in the form of the best matching case but also a number of next best matching cases, sorted by their global similarity to the query. Thus CBR is able to provide results even when no exact match is found. This capability combined with the approach of similarity based retrieval allows the CBR system to provide possible solutions even on only sparse, incomplete or fuzzy queries.

3.3. Human expert interviews

Three hydrometallurgical experts, here Interviewee I, II and III, answered the interviewing questions. As can be seen from Table 7, their answers show some similarities, but also differ in many ways. For clarity, the table only holds the answers concerning enrichment, pretreatment and leaching techniques. All interviewees implied comminution before these steps and basic recovery techniques thereafter. In some cases the interviewee suggested several parallel process paths for the ore and in these cases the different paths are indicated as Ia, Ib, etc.

In T1, the answers are quite similar, since all would choose a pretreatment for oxidising sulfides, followed by cyanide leaching (Interviewee I did not specify the pretreatment further). For T2, Interviewee I would suggest roasting, Interviewee II's suggestion

is alkaline pressure oxidation and Interviewee III suggests one or the other for the refractory part of the ore. The answers differ most regarding T3 where Interviewee I would float and pressure oxidise the concentrate, then apply CIL (carbon-in-leach), but Interviewee II suggests applying alkaline pressure oxidation followed by thiosulfate leaching due to the ore's preg-robbing characteristics. Interviewee III however would divide the ore into oxidic and sulfidic streams, only leach the oxidic part, while applying flotation, fine grinding and CIL for the sulfidic stream. All interviewees would suggest pressure oxidation for T4 and Roasting for T5. Two out of three interviewees would apply flotation for T4 and T5.

Generally, it can be said that interviewees I and II prefer CIL and possibly would recommend it, unless there was an obvious reason for concluding otherwise. Whereas interviewee III clearly prefers agitated cyanide leaching, unless the ore is preg-robbing. The interviewees agree on the pretreatment methods on none of the test cases, even though the given information is identical and relatively straightforward. This indicates that their assumptions of the ores, past experiences etc. play a role in their decision making. Certain patterns can be seen in their answers. For example, it is possible that Interviewee I has had experience with using flotation before pressure oxidation (T3 and T4), but not before roasting (T2 and T5). Interviewee II might have had lots of experiences with pressure oxidation (T1–T4) and less with roasting T5. Interviewee III seems to be more familiar with regrinding the ore during the process than the other two experts (T3 and T4).

Compared to the actual processes utilised for the ores in queries T1–T5 the only question where the interviewees proposed similar flowsheet was T4. On other questions, the process flowsheets they suggested were distinctly dissimilar from the original processes. This merely comes to show that making process suggestions based on lacking knowledge of the ore is very challenging. Humans naturally use their past experiences in situations where rapid decisions need to be made based on lacking information, as was the case in the interviewing situation. However, these presumptions can then

follow the decision maker further into the decision making process, even after sufficient information and resources are available. Automated decision support systems, such as CBR, can lessen these human biases by providing objective analysis based on greater amounts of information.

4. Conclusions

The evaluation showed that the CBR methodology and the modelled attributes resulted in a system that was able to retrieve similar raw materials that have relatively similar process flowsheets compared to the test cases (T1–T5). The results of the CBR retrieval tests were compared with the interview results of human experts (I–III). It was shown that the constructed CBR system could retrieve more similar process flowsheets compared to the human experts. When the objective is to quickly assess different processing methods for a gold ore deposit of interest, the CBR system is able to present a variety of options within a few seconds without human biases. These options can then be thoroughly evaluated with geological and metallurgical experiments prior to process design.

To enable a more accurate retrieval of cases, in the next phase of this research the number and accuracy of attributes will be improved to cover other relevant mineralogical properties of the ore. For example, *Gold occurrence* was modelled in a limited manner; the attribute was described by two stages only: “Free” or “Enclosed in mineral”. As also other factors, such as surface exposure and the type of mineral that has enclosed the gold, are relevant in the behaviour of the leaching process, the modelling of the *Gold occurrence* attribute will be studied in more detail. Additional attributes could include, for example, the size of the ore body, which has a great impact on the feasibility of capital investments.

The size of the case base will be extended further. This will increase the probability of a similar ore being present in the case base when a query is made. During the construction of the case base, it was noted that the CBR software needs to be modified in a way that one attribute can have more than one value. For example, if two different iron sulfides are present in the ore, the user should be able to enter this into the query. In addition, observations were made about the system's functionality related to similarity calculations. It was suggested that the impact of large differences in the value of *Gold ore grade* should be emphasised through an advanced local similarity function. The impact of the attribute *Gold grain size* needs to be decreased and that of the attribute *Clay present* needs to be increased by applying appropriate weighting factors.

Eventually, the constructed CBR tool should serve gold processing experts in the initial stages of flowsheet design by providing new ideas and possibly a second opinion. The real benefit of the automated decision support system is that it does not exclude any processing methods merely due to human biases, such as familiarity with another processing method. With continued development of the system, its functionality in providing starting points for gold process design will increase further.

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Appendix A . Case base

Case	Ore type	Gold ore grade [g/t]	Gold distribution	Gold grain size	Sulfide present	Arsenic sulfide	Copper sulfide	Iron sulfide	Clay present	Case source
East Driefontein	Free milling	8.0	Free	Fine	Yes	None	None	Pyrrhotite	_undefined_	Marsden and House (2006), Chapter 12.2.2.1
Pierina	Free milling	2.1	Free	Sub-micronized	Yes	Any	Any	Pyrite	_undefined_	Marsden and House (2006), Chapter 12.2.2.12
Harmony No4Plant	Free milling	2.4	Free	Fine	No	None	None	None	_undefined_	Marsden and House (2006), Chapter 12.2.2.2
President Brand New Plant	Free milling	5.0	Free	_undefined_	No	None	None	None	_undefined_	Marsden and House (2006), Chapter 12.2.2.3
Round Mountain	Free milling	1.1	Free	_undefined_	Yes	_undefined_	_undefined_	Pyrite	_undefined_	Marsden and House (2006), Chapter 12.2.2.6
Kidston	Free milling	2.0	Free	Fine	Yes	_undefined_	_undefined_	Pyrite	Yes	Marsden and House (2006), Chapter 12.2.2.7
Pine Creek	Free milling	2.4	Free	Fine	Yes	Arsenopyrite	_undefined_	Pyrite	No	Marsden and House (2006), Chapter 12.2.2.8
Muruntau	Free milling	2.5	Free	_undefined_	Yes	Arsenopyrite	_undefined_	_undefined_	_undefined_	Marsden and House (2006), Chapter 12.2.2.9
Muruntau	Free milling	2.5	Free	_undefined_	Yes	Arsenopyrite	_undefined_	_undefined_	_undefined_	Marsden and House (2006), Chapter 12.2.2.9

Appendix A (continued)

Case	Ore type	Gold ore grade [g/t]	Gold distribution	Gold grain size	Sulfide present	Arsenic sulfide	Copper sulfide	Iron sulfide	Clay present	Case source
Fort Knox	Free milling	0.9	Free	Fine	Yes	None	None	Pyrite	_undefined_	Marsden and House (2006), Chapter 12.2.2.10
Fort Knox	Free milling	0.9	Free	Fine	Yes	None	None	Pyrite	_undefined_	Marsden and House (2006), Chapter 12.2.2.10
Yanacocha	Free milling	1.5	Free	Fine	Yes	None	None	None	_undefined_	Marsden and House (2006), Chapter 12.2.2.11
Giant Yellowknife	Ref. arsenopyritic	8.0	Grain encl. in min.	Sub-micronized	Yes	Arsenopyrite	Any	Pyrite	No	Marsden and House (2006), Chapter 12.2.5.3
Sao Bento	Ref. arsenopyritic	7.2	_undefined_	_undefined_	Yes	Arsenopyrite	None	Pyrrhotite	No	Marsden and House (2006), Chapter 12.2.5.5
Fairview	Ref. arsenopyritic	7.0	Grain encl. in min.	Sub-micronized	Yes	Arsenopyrite	Any	Pyrite	No	Marsden and House (2006), Chapter 12.2.5.6
Ashanti refractory sulfide plant	Ref. arsenopyritic	12.0	Grain encl. in min.	Fine	Yes	Arsenopyrite	None	Pyrite	No	Marsden and House (2006), Chapter 12.2.5.1
Cambell Red Lake	Ref. arsenopyritic	20.0	Grain encl. in min.	_undefined_	Yes	Arsenopyrite	Chalcopyrite	Pyrite	No	Marsden and House (2006), Chapter 12.2.5.2
Joutel	Ref. iron sulfide	5.8	Grain encl. in min.	Fine	Yes	Arsenopyrite	None	Pyrite	No	Marsden and House (2006), Chapter 12.2.4.1
Fimiston-Gidji	Ref. iron sulfide	3.8	Grain encl. in min.	Fine	Yes	None	None	_undefined_	No	Marsden and House (2006), Chapter 12.2.4.2
McLaughlin	Ref. iron sulfide	4.7	Grain encl. in min.	Fine	Yes	None	Chalcopyrite	Pyrite	Yes	Marsden and House (2006), Chapter 12.2.4.3
Lihir	Ref. iron sulfide	5.0	Grain encl. in min.	Sub-micronized	Yes	_undefined_	_undefined_	Pyrite	_undefined_	Marsden and House (2006), Chapter 12.2.4.4
Porgera	Ref. iron sulfide	5.0	Grain encl. in min.	Sub-micronized	Yes	None	Chalcopyrite	Pyrite	No	Marsden and House (2006), Chapter 12.2.4.5
Goldstrike	Ref. iron sulfide	8.5	Grain encl. in min.	Sub-micronized	Yes	None	None	Pyrite	Yes	Marsden and House (2006), Chapter 12.2.4.6
Barneys Canyon	Ref. iron sulfide	3.0	Grain encl. in min.	Micronized	Yes	Arsenopyrite	_undefined_	Pyrite	Yes	Marsden and House (2006), Chapter 12.2.4.8
Barneys Canyon	Ref. iron sulfide	3.0	Grain encl. in min.	Micronized	Yes	Arsenopyrite	_undefined_	Pyrite	Yes	Marsden and House (2006), Chapter 12.2.4.8
Barneys Canyon	Free milling	1.5	Grain encl. in min.	Micronized	No	None	None	None	Yes	Marsden and House (2006), Chapter 12.2.4.8
Consolidated Murchison	Ref. antimony sulfide	3.0	Free	Coarse	Yes	Arsenopyrite	_undefined_	_undefined_	No	Marsden and House (2006), Chapter 12.2.7.1
Paradise Peak	Silver rich	3.1	Free	_undefined_	Yes	Orpiment	None	None	No	Marsden and House (2006), Chapter 12.2.3.1
Easy Creek	Silver rich	70.0	_undefined_	_undefined_	Yes	None	Chalcopyrite	Pyrite	No	Marsden and House (2006), Chapter 12.2.3.2
Grasberg-Ertsberg	Copper rich	2.0	Free	Fine	Yes	None	Chalcopyrite	None	No	Marsden and House (2006), Chapter 12.2.6.1

(continued on next page)

Case	Ore type	Gold ore grade [g/t]	Gold distribution	Gold grain size	Sulfide present	Arsenic sulfide	Copper sulfide	Iron sulfide	Clay present	Case source
OkTedi	Copper rich	2.8	Free	Coarse	No	None	None	None	No	Marsden and House (2006), Chapter 12.2.6.2
ElIndio	Copper rich	8.4	Free	Coarse	Yes	None	None	Pyrite	No	Marsden and House (2006), Chapter 12.2.6.3
Alumbrera	Copper rich	0.8	Free	Coarse	Yes	None	Chalcopyrite	Pyrite	No	Marsden and House (2006), Chapter 12.2.6.4
Phoenix	Copper rich	1.1	_undefined_	_undefined_	Yes	_undefined_	Chalcopyrite	Pyrite	_undefined_	Marsden and House (2006), Chapter 12.2.6.5
Phoenix	Copper rich	1.1	_undefined_	_undefined_	Yes	_undefined_	Chalcopyrite	Pyrite	_undefined_	Marsden and House (2006), Chapter 12.2.6.5
Canderalia	Copper rich	0.3	Free	Fine	Yes	_undefined_	Chalcopyrite	Pyrite	_undefined_	Marsden and House (2006), Chapter 12.2.6.6
Telfer	Copper rich	1.5	Grain encl. in min.	Fine	Yes	Any	Chalcopyrite	Pyrite	_undefined_	Marsden and House (2006), Chapter 12.2.6.7
Telfer	Copper rich	1.5	Free	Fine	Yes	Any	Chalcopyrite	Pyrite	_undefined_	Marsden and House (2006), Chapter 12.2.6.7
Telfer	Copper rich	1.5	Grain encl. in min.	Fine	Yes	Any	Chalcopyrite	Pyrite	_undefined_	Marsden and House (2006), Chapter 12.2.6.7
Telfer	Copper rich	1.5	Grain encl. in min.	Fine	Yes	Any	Chalcopyrite	Pyrite	_undefined_	Marsden and House (2006), Chapter 12.2.6.7
Telfer	Copper rich	1.5	Free	Fine	Yes	Any	Chalcopyrite	Pyrite	_undefined_	Marsden and House (2006), Chapter 12.2.6.7
EmperorMines	Telluride	25.0	Free	_undefined_	Yes	Arsenopyrite	Any	Pyrite	No	Marsden and House (2006), Chapter 12.2.8.1
JerrittCanyon	Carbonaceous	6.7	Free	_undefined_	Yes	Realgar	None	Pyrite	Yes	Marsden and House (2006), Chapter 12.2.9.1
Yellowknife	Silver rich	5.1	Free	Fine	Yes	Arsenopyrite	None	Pyrrhotite	_undefined_	Tyhee Development Group (2010)
Newmont Boddington	Copper rich	0.8	Free	Micronized	Yes	Arsenopyrite	Chalcopyrite	Pyrite	_undefined_	The AusIMM Bulletin (2015)
Lihir2	Ref. iron sulfide	2.2	Grain encl. in min.	_undefined_	Yes	Arsenopyrite	Chalcopyrite	Pyrite	Yes	Newcrest Mining (2014)
Karma	Ref. arsenopyritic	1.1	_undefined_	Fine	Yes	Arsenopyrite	Chalcopyrite	Pyrite	_undefined_	True Gold Mining (2014)
Porgera2	Ref. iron sulfide	3.9	Grain encl. in min.	Sub-micronized	Yes	Arsenopyrite	None	Pyrite	_undefined_	Mining-technology (2005)

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