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Kaninda Musumbu

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# The Semantics of Kalah Game

Kaninda Musumbu  
LaBRI (UMR 5800 du CNRS),  
Université Bordeaux I, France  
351, cours de la Libération, F-33.405 TALENCE Cedex,  
e-mail: musumbu@labri.fr

## Abstract

*The present work consisted in developing a plateau game. There are the traditional ones (monopoly, cluedo, ect.) but those which interest us leave less place at the chance (luck) than to the strategy such that the chess game. Kallah is an old African game, its rules are simple but the strategies to be used are very complex to implement. Of course, they are based on a strongly mathematical basis as in the film "Rain-Man" where one can see that gambling can be payed with strategies based on mathematical theories. The Artificial Intelligence gives the possibility "of thinking" to a machine and, therefore, allows it to make decisions. In our work, we use it to give the means to the computer choosing its best movement.*

## 1 Introduction

The present work consisted in developing a plateau game. There are the traditional ones (monopoly, cluedo, long course...) but those which interest us leave less place to random than to the strategy. There are the plays of simulation like:

- "The age of the rebirth". Historical reconstitution from 750 to 1350 (of the Middle Ages to the rebirth) highly strategic Play, basing itself on the commercial conquests (and not soldiers). The play advances thanks to discoveries, inventions, etc.
- "Risk". It is a military play of strategy, consisting in conquering the world. The plate represents the chart of the world. Each team must fight, or link itself to remove the opposing armies.
- "Dune". Who doesn't know the Dune planet, with his spice, vital for all the protagonists? Find the environment of Franck Herbbert, while using of diplomacy, strategy and bluff to be the Master of spice.

One finds also sets of rules simple but "effective". In which we can integrate the *Kalah game*. These rules are simple but the strategies to be used are very complex to implement. Of course, they are based on a strongly mathematical basis with calculations of the differences of pawns, empty holes... as in the film "Rain-Man" where we can see that gambling can be played with strategies based on mathematical theories. The *Kalah* game is one of oldest African plays. It is a mathematical play and the most complex versions can be compared with the chess game.

The play of *Kalah* is appeared as table, composed of two lines of six holes plus two called "special" holes *Kalabs*. At the beginning of the game, each hole contains six pawns and the *Kalabs* are empty. The game's goal is to collect more than half of the total number of pawns in its *Kalah*. We have developed the *Kalah* game on the computer with four modes of play:

- Two players on the same machine.
- a player counters the computer.
- Two players in network.
- Two computers.

Paragraph 2 is a general presentation of work with a recall of rules of the game. In paragraph 3, we state the problem posed by the taking into account the modification of the apron following a movement of one of the players. In paragraph 4, we explain the problem arising from the computing time necessary to make play the computer and the solutions brought to really make the play interactive and acceptable the latency.

## 2 General Presentation

For our implementation we decide to put two lines made up of six holes plus two "special" holes called *Kalahs*. Each hole contains six pawns at the beginning of the play and the *Kalahs* are empty. The *Kalahs* are used to store the pawns collected by each player. The player can choose between four modes of play: two computers, a player counters the computer, two players on the same machine, two players in network. The course of the play can be followed using the messages posted to the medium of the screen after each modification made on the table

## 3 Functionalities of Game Engine

This module is used for the direct interaction with all the structures of data necessary for storage and modifications of the states of the play. It manages four modes different of game:

- Two players on the same machine. With each turn, the computer lets to the user click on one of his holes. Once the choice makes the engine carries out the movement( the distribution of pawns) on the structures of definite data and informs with the interface to refresh the screen.
- A player counters the computer. The human player always takes the hand at the beginning of the play. He chooses a hole, the engine will carry out this movement and will give the hand to the computer which will make its movements of continuation with a direct interaction with the Strategies module. Once all the finished movements, the graphic interface will start to post the movement of human then all the movements of the computers for this blow.
- Two players in network. The course of the play is exactly the same one as for two players on the same machine, except that the engine with each movement sending on the distant machine the movement made by the local machine in order to post on two sides the same apron of play.
- Two computers. For each computer the Strategy module will return the best hole for the state of the current apron by using the strategy and the level of currents difficulty. With this hole the engine will make the movement and the storage of the apron and will pass the hand to the other computer or will remain on itself if it replays.

This module uses in the same way the GameLogic class. This class contains all information concerning the rules of the game implemented (including the determination of the winner). The structures used most significant are for example: The BoardState class which represents an apron of play storing all the values of the pawns in the holes and the *kalahs*, data-processing representation of an apron of *Kalah*. We also used a variable containing all the history of a play, with a succession of aprons corresponding to the movements carried out, i.e. the first component represents the initial apron and then each component point out the apron reached according to each movement. This structure of history enabled us to implement the following functionalities easily: To make preceding blows: This option makes it possible to the user to reconsider his last blows. At the level of data structure, it is enough to use before last apron stored in the history as being the current apron. To make following blows: Just as to retrogress, to advance it is enough to seek the aprons which are on the right current and to shift in this direction of the number of following blows.

## 4 Artificial Intelligence

The artificial intelligence is a concept about which, generally, everyone intended to speak. The artificial intelligence gives the possibility "of thinking" of a machine and, therefore, allows him to make decisions. In our work, we used it to be able player against the computer and which it can choose his movement. This intelligence is called artificial because it is based on calculations of the algorithms.

### 4.1 Strategy

The module **Strategy** is the module which manages the artificial intelligence of the computer. It is the most significant part of our work. It gives the possibility "of thinking", i.e., to choose its next movement "grace" mainly with the algorithm **MiniMax** (in concrete terms, an alternative of this algorithm) described below. For recall, we present initially the operation of the **traditional MiniMax** then the variations that we have introduce.

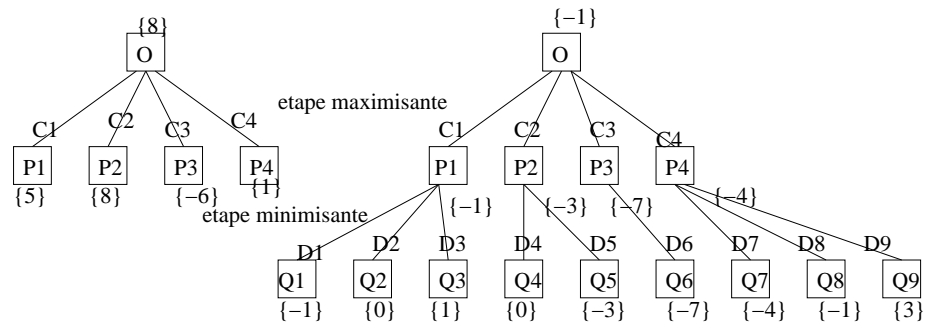
### 4.2 Algorithm MiniMax

The algorithm **MiniMax** is a universal algorithm, which is used in the plays with two players, to decide which is the best movement to be made, at a given time, starting from the current state. In our case one considers the current state as the apron of the game( pawns which are in each hole and in the *kalahs* ), and the player who has the hand. One will use this algorithm preferably to allow the computer to choose his movement or to be able to give to human council in the play of the type *a player against the computer*. Proposed in 1928, by John Von Neumann, this technique leads the computer to review all the possibilities for a limited number of blows and to assign a value to them which takes into account the benefit for the player who has the hand and for his adversary. The best choice being then that which maximizes its benefit (one it calls player *MAX*) while minimizing those of sound adversary(player *MIN*). This algorithm must be able to return a value which will correspond to the movement chosen by the player *MAX*. The basic idea consist to create a tree which will have as many levels as that indicated by a value passed in parameter. Each node of the tree can have more than one number of wire equal to the number of possible movements. For example this number will be six for *Kalah* game with six holes by player. The algorithm **MiniMax** is a depth search algorithm, with a limited depth.

It requires to use:

- a generation function of the legal blows starting from a position
- a evaluation function of a position of play

From a position of the play, the algorithm explores the tree of all legal blows until the required depth. The scores of the tree's sheets are then calculated by the evaluation function. A positive score indicates a good position for the player *MAX* and a negative score a bad position for him, therefore a good position for the player *MIN*. According to one who plays, the passage of a position to another is maximizing (for the player *MAX*) or minimizing for the player *MIN*. The players try to play the most advantageous blows for themselves. By seeking the best blow for *MAX*, the depth search for level 1 will seek to determine the immediate blow which maximizes the score of the new position.



For example, on the figure, the player *MAX* leaves position 0, determines four legal blows, builds these new configurations and evaluates them. Of these scores, its best position is (of score 8). It propagates this value with position 0, indicating speak with this pleasing position in a blow with a new position about score 8 by playing the blow *C2*. in-depth exploration about level 1 is in general not sufficient, because it does not take account of the response of the adversary. That produced of the programs seeking the immediate profit (like the catch of a queen to the chess board), without realizing that the parts are protected or that the position becomes losing (gambit of the queen to make checkmate). An exploration of depth 2 makes it possible to realize by-effect. Figure 2 shows an additional level of development of the tree by taking account of the answer of the player *MIN*. This one to also seek its best blow. For that, the algorithm *MiniMax* will minimize the scores of the nodes of depth 2. The blow who brought to a position immediately score 8 goes, in made, to indeed bring the position of the play to a score of -3. if B plays the blow *D5*, then the score of the position *Q5* is worth -3. We can see that the blow *C1* limits the dégats with a score of -1. It will thus be preferred. In the majority of the plays, it is possible to make lanterns its adversary, by making it play forced blows, with an aim of muddling the situation by hoping that it will make a fault. For that the search for depth 2 is very insufficient for the tactical aspect of the play. The strategic aspect is seldom well exploited by a program because it does not have the vision probable evolution of the position at the end of the part. The difficulty larger depth comes from the combinator explosion. For example, to the failures, the additional exploration 2 depths brings a factor of approximately thousand times of combinations (30\*30). Therefore, if one seeks to calculate a depth of 10, one will obtain approximately 514 position, which is of course too. For that, one tries to prune the tree of research to reduce this complexity.

### 4.3 Alphabeta Pruning

One can note that it is not forcing useful to explore the branch in où measurement the score of this position to depth 1 is already with less good than that found in the branch. In the same way the branch need does not have completely to be explored. As of the calculation of  $Q7$ , we obtain a score lower than that of (always completely explored). Calculates  $Q8$  and  $Q9$  will not be able to improve this situation even if their respective score is better than  $Q7$ . In a minimizing stage, the weakest score went up. Already is known that it will not bring anything again. The alternative alphabeta of the **MiniMax** uses this pruning to decrease the number of branches to be explored. This reduction causes an increase in the performances in time and in space. With the alphabeta pruning, we generate only a number of nodes necessary to decide if each branch will bring us to a better value of that already exist.

### 4.4 Algorithm MiniMax Revisited

The MiniMax algorithm with alphabeta pruning is created for games with two players in which, after each turn the hand changes. By the specificity of the *Kalah* game, there is a rule which allows to keep the hand (replay again), then it was necessary to make modifications on this algorithm. In the traditional exploration tree, one makes the maximization of the wire values to the root (one wants the value the greatest value for the player *MAX* who is, by definition, that calls the algorithm. In the following level, one makes minimization because it is the turn of the adversary (player *MIN* . One continues thus while alternating until the end. In the play of *Kalah*, the problem comes from the fact, that by generating the depth tree of game, it is necessary to store the information which says if the player who made the last movement must replay or not. Alternation is not systematic any more, one looks at initially if that which has just played owes replay. In this case, if the player is the *MAX*, it will again be necessary to maximize in the following level. If not, it will be necessary to minimize. This modification implies that alphabeta pruning cannot be bracket in all the cases. It will be able to apply it only in the following cases:

- One is in a node *MAX* and his/her father is *MIN*
- One is in a node *MIN* and his/her father is *MAX*

The reason is rather obvious. Indeed, let us consider the example of figure 3, First of all, one goes down, in-depth, by the first branch, when we go up, we obtain value 2 out of B, then in A, like temporal value. Now, we go down by the second branch to the sheet, we go up -1, whatever the values which can be obtained they will not exceed -1 is thus will be always lower than 2. We can prune these branches, because they will not bring additional information.

Let us consider the case where the player *MAX* keep the hand, and suppose that the node C is of type *MAX*. A priori it is not known if one of wire of C will not be able to go up a higher value. They will thus have to be go through all. For example in figure 4, the last wire has value 5 and this one is finally the gone up value with A.

With this alternative that appears clear that one loses because each time that a player keep the hand, our algorithm generates more nodes and consequently, more time. But the nature of *Kalah* game does not make it possible to optimize more than what we did.

### 4.5 Levels of difficulty

In the program, we introduced four levels of difficulties which can be chosen by the user when it asks a play against the computer. More the selected level is more high, the difficulty increases and

more the possibilities of gaining decreases. There is a direct correspondence between each one of the levels and the size of the exploration tree generated by the algorithm **MiniMax**. The easy level corresponds to a tree of depth 2, the mean level with 4, the difficult level with 6 and the very difficult level with 8. We used a parameter which is used as coefficient. In our case this parameter is worth 2. We obtain levels of depth 2,4, 6 and 8. It can be changed, for example into 3, and in this case one obtains depths 3, 6, 9 and 12. However, it should not be forgotten that the number of nodes generated by the algorithm believes in an exponential way in each increase of a unit of level. For level 8, with 6 branches(game with 6 holes), the complete tree contains more than two billion nodes. For reason of response time of the machine, we decided not to go beyond. With eight levels, we obtain an acceptable response time.

#### 4.6 Comparison MiniMax and our algorithm

We measure the number of nodes generated by the exploration tree, instead of time, because, in this way, we will obtain a measurement which is always independent on the computer which serves the tests. For the algorithm without pruning alphabeta, the number of generated nodes is equal  $\sum_{i=0}^n 6^i$ . We tested with various states of the apron and various levels. The computed value is the number of nodes generated on average. To measure the improvement, we give the percentage of nodes generated by our algorithm compared to the traditional one. The improvement is very considerable especially with regard to the highest levels.

Level	traditional Minimax	Our algorithm	Percentage
2	43	17,2	40%
4	1555	226,4	15%
6	55987	3402,7	6%
8	2015539	44471,1	2%

## 5 Conclusion

We used a universal algorithm, the MiniMax, that one uses for the games with two players. It is a question of creating a tree of a limited depth containing all possible blows for this depth. In exploring the tree, the machine can decide its next movement. The computing time to generate this tree posed a serious problem. To reduce this computing time we implemented a more efficiency version of this algorithm. We limited ourselves to the presentation of our MiniMax algorithm without speaking about various implemented strategies.

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