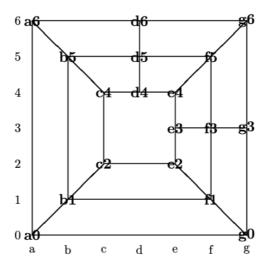
MORRIS GAME

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	
a0	g0	b1	f1	c2	e2	e3	f3	g3	c4	d4	e4	b5	d5	f5	a6	d6	g6	



This variant of Morris Game has 18 positions which are denoted by the array shown above. There are 8 moves for white and black pieces in the game. The pieces with less than 3 on the board loses the game.

The positions are estimated using MiniMax Algorithm and α - β Pruning Algorithm

• The MiniMax program written for opening phase of the game is **MiniMaxOpening.java**This program is recursive defined MiniMax algorithm in the opening phase of the game with a depth value of the tree where the leaf values are calculated using the static estimation function in the opening phase.

The **input** tested: xWxxxWxxBxxxxBxxx

The output:

Board Position: xWxWxWxxxxxxxBxxx Positions evaluated by static estimation: 197

MINIMAX estimate: 1

depth: 2

The **input** tested: xxxxxxWxxxxxxBxxxx

The output:

Board Position: WxxxxxWxxxxxBxxxx Positions evaluated by static estimation: 240

MINIMAX estimate: 0

depth: 2

The MiniMax program written for mid / ending phase of the game is MiniMaxGame.java

This program is recursive defined MiniMax algorithm in the mid/endgame phase of the game with a depth value of the tree where the leaf values are calculated using the static estimation function in the midgame / endgame phase.

The input tested:

WWBBBBxWxxxxWBWxWx

The output:

Board Position: WWBBBBWxxxxxWBWxWx Positions evaluated by static estimation: 38

MINIMAX estimate: 994

depth: 2

The **input** tested: xBBxxxWxxWxxBxBxxW

The **output**:

Board Position: xBBxxxxxxWxxBWBxxW Positions evaluated by static estimation: 442

MINIMAX estimate: -1011

depth: 2

The MiniMax program written for opening phase of the game for black is MiniMaxOpeningBlack.java
 This program is recursive defined MiniMax algorithm for the black moves in the opening
 phase of the game with a depth value of the tree where the leaf values are calculated using
 the static estimation function in the opening phase.

The **input** tested: xWxxxWxxBxxxxBxxx

The output:

Board Position: xWxBxWxxBxxxxxBxxx Positions evaluated by static estimation: 208

MINIMAX estimate: 0

depth: 2

The **input** tested: The **output**:

Board Position: BxxxxxWxxxxxxBxxxx

Positions evaluated by static estimation: 240

MINIMAX estimate: 0

depth: 2

The MiniMax program written for mid / ending phase of the game is MiniMaxGameBlack.java
 This program is recursive defined MiniMax algorithm for the black moves in the
 mid/endgame phase of the game with a depth value of the tree where the leaf values are

calculated using the static estimation function in the midgame / endgame phase.

The **input** tested:

WWBBBBxWxxxxWBWxWx

The output:

Board Position: WWBBxBxWxBxxWBWxWx Positions evaluated by static estimation: 35

MINIMAX estimate: -1021

depth: 2

The **input** tested: xBBxxxWxxWxxBxBxxW

The **output**:

Board Position: xBxxxxxxxWxxBBBxxW Positions evaluated by static estimation: 377

MINIMAX estimate: 10000

depth: 2

• The α - β Pruning program written for opening phase of the game is **ABOpening.java**

This program is recursive defined α - β Pruning algorithm in the opening phase of the game with a depth value of the tree where the leaf values are calculated using the static estimation function in the opening phase.

The input tested:

xWxxxWxxBxxxxxBxxx

The output:

Positions evaluated by static estimation: 40

 $\alpha\beta$ estimate: 1 depth: 2

The same input when done with MiniMax had to evaluate 197 positions which is 40 for α - β . This clearly shows that not all the leaf nodes are evaluated.

The **input** tested:

xxxxxxWxxxxxxBxxxx

The output:

Board Position: WxxxxxWxxxxxBxxxx Positions evaluated by static estimation: 30

 $\alpha\beta$ estimate: 0 depth: 2

• The α - β Pruning program written for mid / ending phase of the game is **ABGame.java**

This program is recursive defined α - β Pruning algorithm in the mid/endgame phase of the game with a depth value of the tree where the leaf values are calculated using the static estimation function in the midgame / endgame phase.

The **input** tested:

WWBBBBxWxxxxWBWxWx

The **output**:

Board Position: WWBBBBWxxxxxWBWxWx Positions evaluated by static estimation: 15

 $\alpha\beta$ estimate: 994

depth: 2

The same input when done with MiniMax had to evaluate 38 positions which is 15 for α - β . This clearly shows that not all the leaf nodes are evaluated.

The **input** tested:

xBBxxxWxxWxxBxBxxW

The output:

Board Position: xBBxxxxxxWxxBWBxxW Positions evaluated by static estimation: 102 $\alpha\beta$ estimate: -1011

depth: 2

The same input when done with MiniMax had to evaluate 442 positions which is 102 for α - β . This clearly shows that not all the leaf nodes are evaluated.

 The MiniMax program written for opening phase of the game with improved static estimation function is MiniMaxOpeningImproved.java

This program is recursive defined MiniMax algorithm in the opening phase of the game with a depth value of the tree where the leaf values are calculated using the improved static estimation function in the opening phase.

The improved static estimation function included the possible Mill counts for white pieces on the board which are potential win positions in the future this increases the value of the leaf nodes in MiniMax calculation and possibly a better winning move for white.

The **input** tested:

xWxxxWxxBxxxxxBxxx

The output:

Board Position: xWxWxWxxxxxxxBxxx Positions evaluated by static estimation: 197

MINIMAX estimate: 4

depth: 2

The MiniMax estimate for the same input was 1 compared to 4 when used with improved static estimation function

The **input** tested: xxxxxxWxxxxxxBxxxx

The output:

Board Position: WxxxxxWxxxxxBxxxx Positions evaluated by static estimation: 240

MINIMAX estimate: 1

depth: 2

The MiniMax estimate for the same input was 0 compared to 1 when used with improved static estimation function

• The MiniMax program written for mid/ending phase of the game with improved static estimation function is **MiniMaxGameImproved.java**

This program is recursive defined MiniMax algorithm in the mid/endgame phase of the game with a depth value of the tree where the leaf values are calculated using the improved static estimation function in the midgame / endgame phase.

The improved static estimation function included the possible Mill counts for white pieces on the board which are potential win positions in the future this increases the value of the leaf nodes in MiniMax calculation and possibly a better winning move for white.

The **input** tested:

 ${\sf WWBBBBxWxxxxWBWxWx}$

The output:

Board Position: WWBBBBxWxWxxxBWxWx Positions evaluated by static estimation: 38

MINIMAX estimate: 2994

depth: 2

The MiniMax estimate for the same input was 994 compared to 2994 when used with improved static estimation function. The board position generated by MinMax was **WWBBBBWxxxxxWBWxWx** the position generated with improved MiniMax is **WWBBBbxWxWxxxBWxWx** which is clearly a better move when mill positions are considered.

The **input** tested: xBBxxxWxxWxxBxBxxW

The **output**:

Board Position: xBBxxxxxxWxxBWBxxW Positions evaluated by static estimation: 442

MINIMAX estimate: -11

depth: 2

The MiniMax estimate for the same input was -1011 compared to -11 when used with improved static estimation function.