

# Scout, NegaScout and Proof-Number Search

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# Introduction

- It looks like alpha-beta pruning is the best we can do for a **generic searching procedure**.
  - What else can be done generically?
  - Alpha-beta pruning follows basically the “intelligent” searching behaviors used by human when domain knowledge is not involved.
  - Can we find some other “intelligent” behaviors used by human during searching?
- Intuition: **MAX node**.
  - Suppose we know currently we have a way to gain at least 300 points at the first branch.
  - If there is an efficient way to know the second branch is at most gaining 300 points, then there is no need to search the second branch in detail.
    - ▷ *Is there a way to search a tree approximately?*
    - ▷ *Is searching approximately faster than searching exactly?*
- Similar intuition holds for a **MIN node**.

# SCOUT procedure

- Invented by Judea Pearl in 1980.
- It may be possible to verify whether the value of a branch is greater than a value  $v$  or not in a way that is faster than knowing its exact value.
- High level idea:
  - While searching a branch  $T_b$  of a MAX node, if we have already obtained a lower bound  $v_\ell$ .
    - ▷ *First TEST whether it is possible for  $T_b$  to return something greater than  $v_\ell$ .*
    - ▷ *If FALSE, then there is no need to search  $T_b$ . This is called **fails the test**.*
    - ▷ *If TRUE, then search  $T_b$ . This is called **passes the test**.*
  - While searching a branch  $T_c$  of a MIN node, if we have already obtained an upper bound  $v_u$ .
    - ▷ *First TEST whether it is possible for  $T_c$  to return something smaller than  $v_u$ .*
    - ▷ *If FALSE, then there is no need to search  $T_c$ . This is called **fails the test**.*
    - ▷ *If TRUE, then search  $T_c$ . This is called **passes the test**.*

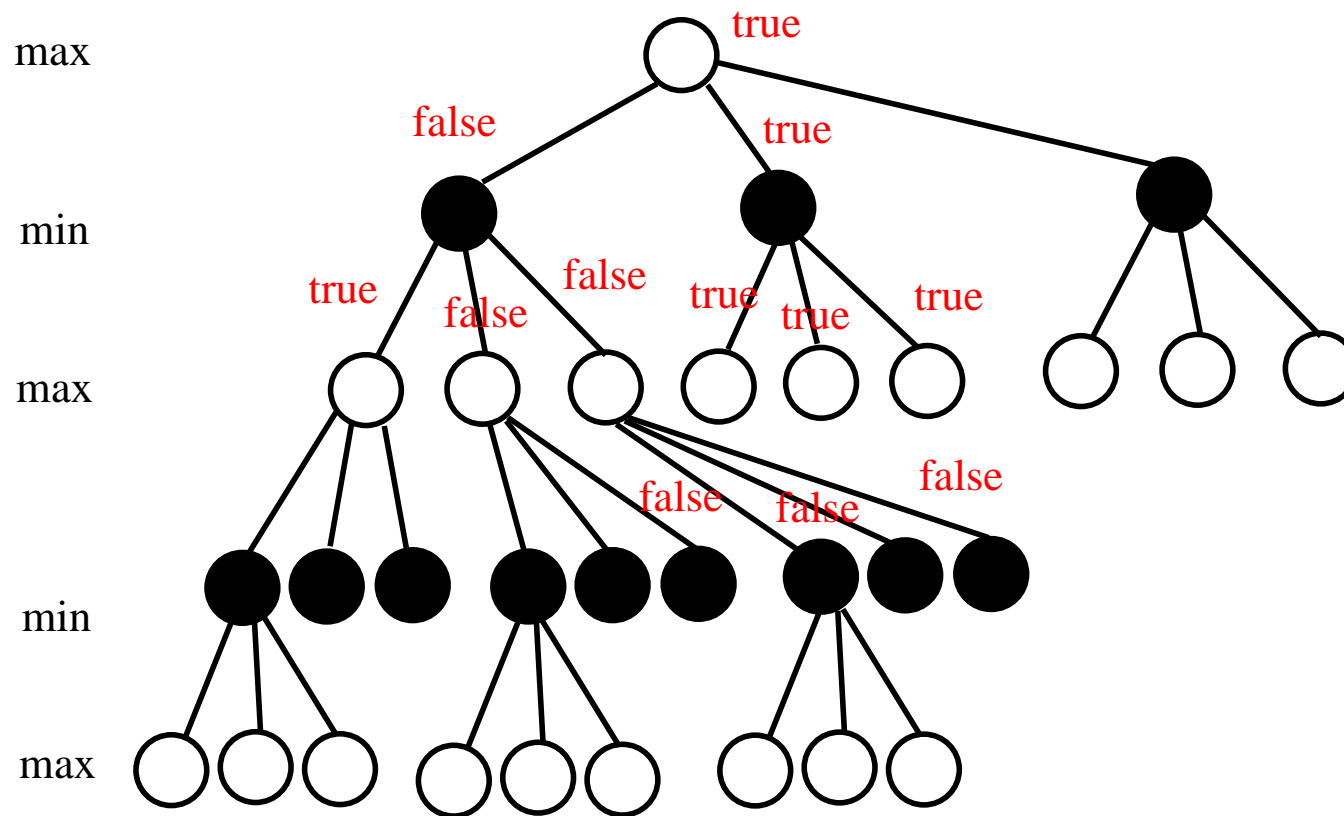
# How to TEST $> v$

procedure TEST(position  $p$ , value  $v$ , condition  $>$ )

*// test whether the value of the branch at  $p$  is  $> v$*

- determine the successor positions  $p_1, \dots, p_d$  of  $p$
- if  $d = 0$ , then *// terminal*
  - ▷ if  $f(p) > v$  then *// f(): evaluating function*
  - ▷ return TRUE
  - ▷ else return FALSE
- if  $p$  is a MAX node, then
  - for  $i := 1$  to  $d$  do
    - ▷ if TEST( $p_i, v, >$ ) is TRUE, then  
return TRUE *// succeed if a branch is  $> v$*
  - return FALSE *// fail only if all branches  $\leq v$*
- if  $p$  is a MIN node, then
  - for  $i := 1$  to  $d$  do
    - ▷ if TEST( $p_i, v, >$ ) is FALSE, then  
return FALSE *// fail if a branch is  $\leq v$*
  - return TRUE *// succeed only if all branches are  $> v$*

# Illustration of TEST



# How to TEST — Discussions

- Condition can be stated as  $<$  by properly revising the algorithm.

- $\text{TEST}(p, v, >) \text{ is TRUE} \equiv \text{TEST}(p, v, \leq) \text{ is FALSE}$
- $\text{TEST}(p, v, >) \text{ is FALSE} \equiv \text{TEST}(p, v, \leq) \text{ is TRUE}$
- $\text{TEST}(p, v, <) \text{ is TRUE} \equiv \text{TEST}(p, v, \geq) \text{ is FALSE}$
- $\text{TEST}(p, v, <) \text{ is FALSE} \equiv \text{TEST}(p, v, \geq) \text{ is TRUE}$

- **Practical consideration:**

- Set a depth limit and evaluate the position's value when the limit is reached.

# How to TEST $< v$

procedure TEST(position  $p$ , value  $v$ , condition  $<$ )

*// test whether the value of the branch at  $p$  is  $< v$*

- determine the successor positions  $p_1, \dots, p_d$  of  $p$
- if  $d = 0$ , then *// terminal*
  - ▷ if  $f(p) < v$  then *// f(): evaluating function*
  - ▷ return TRUE
  - ▷ else return FALSE
- if  $p$  is a MAX node, then
  - for  $i := 1$  to  $d$  do
    - ▷ if TEST( $p_i, v, <$ ) is FALSE, then  
return FALSE *// succeed if a branch is  $\geq v$*
  - return TRUE *// succeed only if all branches  $< v$*
- if  $p$  is a MIN node, then
  - for  $i := 1$  to  $d$  do
    - ▷ if TEST( $p_i, v, <$ ) is TRUE, then  
return TRUE *// succeed if a branch is  $< v$*
  - return FALSE *// fail only if all branches are  $\geq v$*

# Main SCOUT procedure (1/2)

## Algorithm SCOUT(position $p$ )

- determine the successor positions  $p_1, \dots, p_d$
- if  $d = 0$ , then return  $f(p)$
- else  $v = SCOUT(p_1)$  // **SCOUT the first branch**
- if  $p$  is a **MAX** node
  - for  $i := 2$  to  $d$  do
    - ▷ if  $TEST(p_i, v, >)$  is **TRUE**, // **TEST first for the rest of the branches**  
then  $v = SCOUT(p_i)$  // **find the value of this branch if it can be  $> v$**
- if  $p$  is a **MIN** node
  - for  $i := 2$  to  $d$  do
    - ▷ if  $TEST(p_i, v, <)$  is **TRUE**, // **TEST first for the rest of the branches**  
then  $v = SCOUT(p_i)$  // **find the value of this branch if it can be  $< v$**
- return  $v$

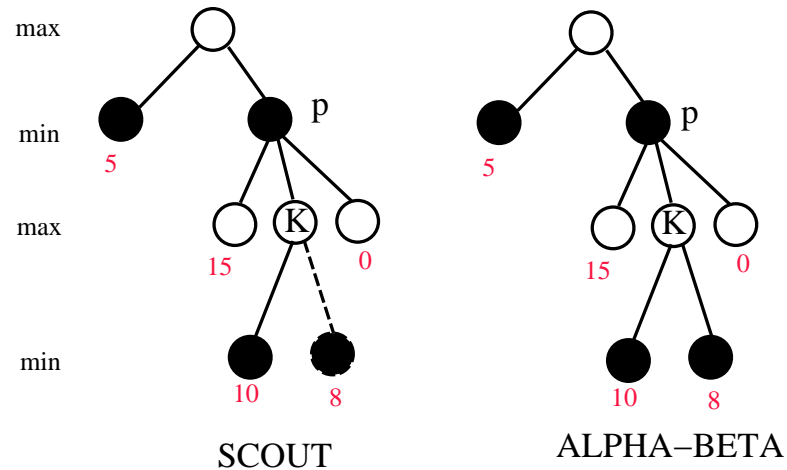


# Main SCOUT procedure (2/2)

- Note that  $v$  is the current best value at any moment.
- MAX node:
  - For any  $i > 1$ , if **TEST**( $p_i, v, >$ ) is TRUE,
    - ▷ then the value returned by *SCOUT*( $p_i$ ) must be greater than  $v$ .
  - We say the  $p_i$  **passes the test** if **TEST**( $p_i, v, >$ ) is TRUE.
- MIN node:
  - For any  $i > 1$ , if **TEST**( $p_i, v, <$ ) is TRUE,
    - ▷ then the value returned by *SCOUT*( $p_i$ ) must be smaller than  $v$ .
  - We say the  $p_i$  **passes the test** if **TEST**( $p_i, v, <$ ) is TRUE.

# Discussions for SCOUT (1/2)

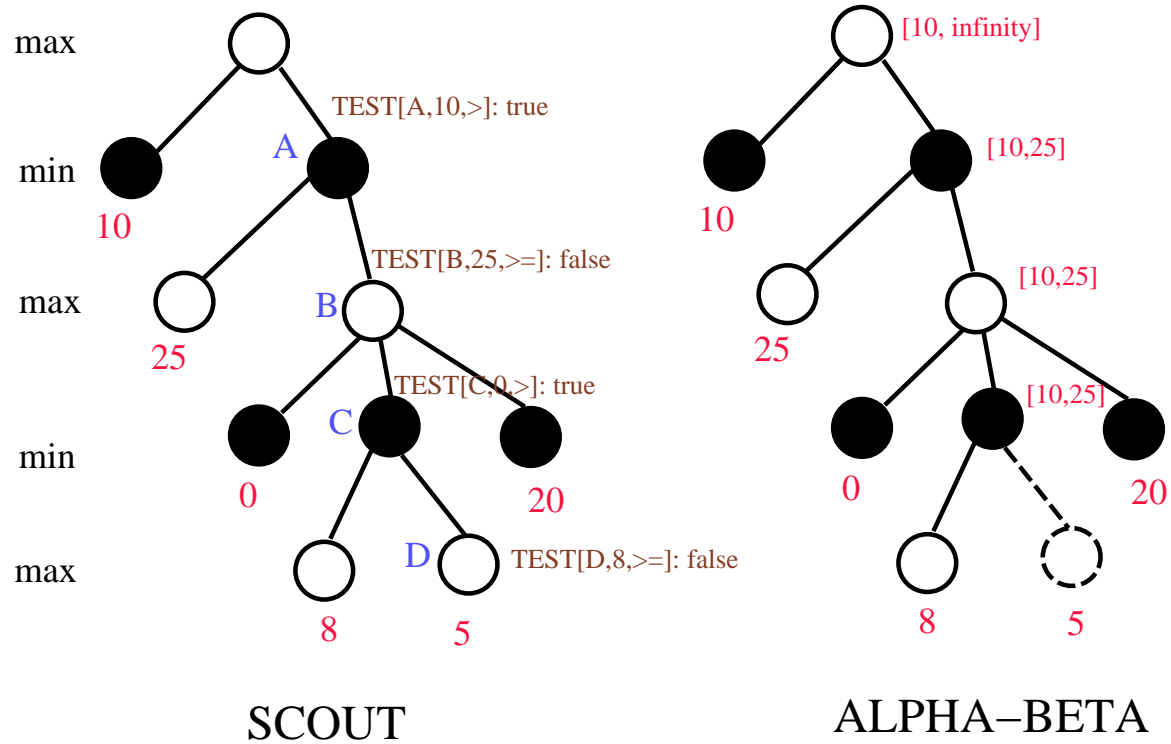
- TEST who is called by SCOUT may visit less nodes than alpha-beta.



- Assume  $TEST(p, 5, >)$  is called by the root after the first branch is evaluated.
  - ▷ It calls  $TEST(K, 5, >)$  which skips  $K$ 's second branch.
  - ▷  $TEST(p, 5, >)$  is **FALSE**, i.e., fails the test, after returning from the 3rd branch.
  - ▷ No need to do SCOUT for the branch  $p$ .
- Alpha-beta needs to visit  $K$ 's second branch.

# Discussions for SCOUT (2/2)

- SCOUT may pay many visits to a node that is cut off by alpha-beta.



# Number of nodes visited (1/3)

- **For TEST to return TRUE for a subtree  $T$ , it needs to evaluate at least**
  - ▷ *one child for a MAX node in  $T$ , and*
  - ▷ *and all of the children for a MIN node in  $T$ .*
  - ▷ *If  $T$  has a fixed branching factor  $b$  and uniform depth  $d$ , the number of nodes evaluated is  $\Omega(b^{d/2})$ .*
- **For TEST to return FALSE for a subtree  $T$ , it needs to evaluate at least**
  - ▷ *one child for a MIN node in  $T$ , and*
  - ▷ *and all of the children for a MAX node in  $T$ .*
  - ▷ *If  $T$  has a fixed branching factor  $b$  and uniform depth  $d$ , the number of nodes evaluated is  $\Omega(b^{d/2})$ .*

# Number of nodes visited (2/3)

## ■ Assumptions:

- Assume a full complete  $d$ -ary tree with depth  $\ell$  where  $\ell$  is even.
- The depth of the root, which is a MAX node, is 0.

■ The total number of nodes in the tree is  $\frac{d^{\ell+1}-1}{d-1}$ .

■ The minimum number of nodes visited by **TEST** when it returns **TRUE**.

$$= 1 + 1 + d + d + d^2 + d^2 + d^3 + d^3 + \dots + d^{\ell/2-1} + d^{\ell/2-1} + d^{\ell/2}$$

$$= 2 \cdot (d^0 + d^1 + \dots + d^{\ell/2}) - d^{\ell/2}$$

$$= 2 \cdot \frac{d^{\ell/2+1}-1}{d-1} - d^{\ell/2}$$

■ The minimum number of nodes visited by alpha-beta.

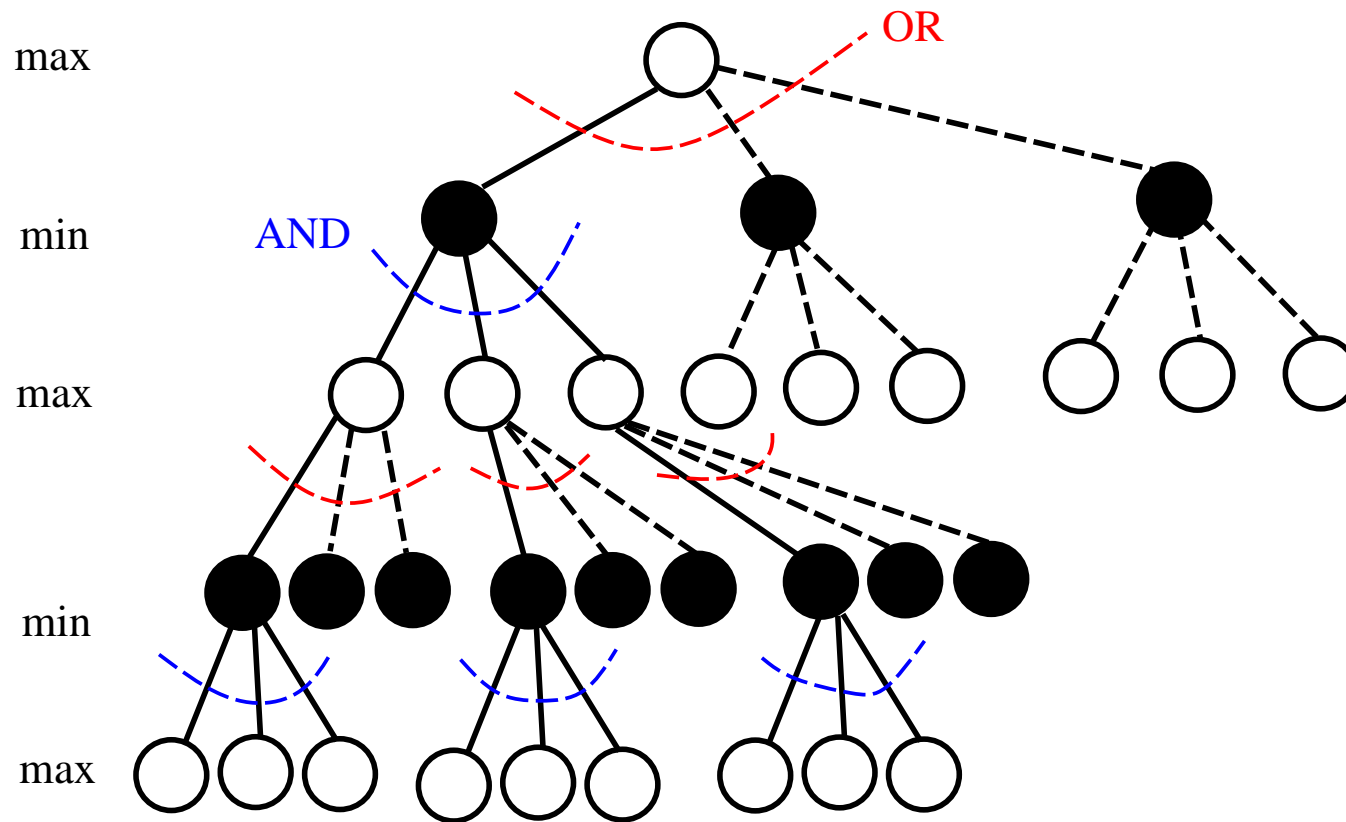
$$= \sum_{i=0}^{\ell} d^{\lceil i/2 \rceil} + d^{\lfloor i/2 \rfloor} - 1$$

$$= \sum_{i=0}^{\ell} d^{\lceil i/2 \rceil} + \sum_{i=0}^{\ell} d^{\lfloor i/2 \rfloor} - (\ell + 1)$$

$$= (1 + d + d + \dots + d^{\ell/2} + d^{\ell/2}) +$$

$$(1 + 1 + d + d + \dots + d^{\ell/2-1} + d^{\ell/2-1} + d^{\ell/2}) - (\ell + 1)$$

# Number of nodes visited (3/3)



# Comparisons

- **When the first branch of a node has the best value, then TEST scans the tree fast.**
  - The best value of the first  $i - 1$  branches is used to test whether the  $i$ th branch needs to be searched exactly.
  - If the value of the first  $i - 1$  branches of the root is better than the value of  $i$ th branch, then we do not have to evaluate exactly for the  $i$ th branch.
- **Compared to alpha-beta pruning whose cut off comes from bounds of search windows.**
  - It is possible to have some cut-off for alpha-beta as long as there are some relative move orderings are “good.”
    - ▷ *The moving orders of your children and the children of your ancestor who is odd level up decide a cut-off.*
  - The search bound is updated during the searching.
    - ▷ *Sometimes, a deep alpha-beta cut-off occurs because a bound found from your ancestor a distance away.*

# Performance of SCOUT (1/2)

- **A node may be visited more than once.**
  - First visit is to TEST.
  - Second visit is to SCOUT.
    - ▷ *During SCOUT, it may be TESTed with a different value.*
  - Q: Can information obtained in the first search be used in the second search?
- **SCOUT is a recursive procedure.**
  - A node in a branch that is not the first child of a node with a depth of  $\ell$ .
    - ▷ *Note that the depth of the root is defined to be 0.*
    - ▷ *Every ancestor of you may initiate a TEST to visit you.*
    - ▷ *It can be visited  $\ell$  times by TEST.*



# Performance of SCOUT (2/2)

- Show great improvements on  $depth > 3$  for games with small branching factors.
  - It traverses most of the nodes without evaluating them precisely.
  - Few subtrees remained to be revisited to compute their exact mini-max values.
- Experimental data on the game of Kalah show [UCLA Tech Rep UCLA-ENG-80-17, Noe 1980]:
  - SCOUT favors “skinny” game trees, that are game trees with high depth-to-width ratios.
  - On depth = 5, it saves over 40% of time.
  - Maybe bad for games with a large branching factor.
  - Move ordering is very important.
    - ▷ *The first branch, if is good, offers a great chance of pruning further branches.*

# Alpha-beta revisited

- In an alpha-beta search with a window  $[alpha, beta]$ :
  - **Failed-high** means it returns a value that is larger than its upper bound  $beta$ .
  - **Failed-low** means it returns a value that is smaller than its lower bound  $alpha$ .
- **Null or Zero window search:**
  - Using alpha-beta search with the window  $[m, m + 1]$ .
  - The result can be either failed-high or failed-low.
  - Failed-high means the return value is at least  $m + 1$ .
    - ▷ *Equivalent to  $TEST(p, m, >)$  is true.*
  - Failed-low means the return value is at most  $m$ .
    - ▷ *Equivalent to  $TEST(p, m, >)$  is false.*

# Alpha-Beta + Scout

## ■ Intuition:

- Try to incooperate SCOUT and alpha-beta together.
- The searching window of alpha-beta if properly set can be used as TEST in SCOUT.
- Using a searching window is better than using a single bound as in SCOUT.
- Can also apply alpha-beta cut if it applies.

## ■ Modifications to the SCOUT algorithm:

- Traverse the tree with two bounds as the alpha-beta procedure does.
  - ▷ *A searching window.*
  - ▷ *Use the current best bound to guide the TEST value.*
- Use a fail soft version to get a better result when the returned value is out of the window.

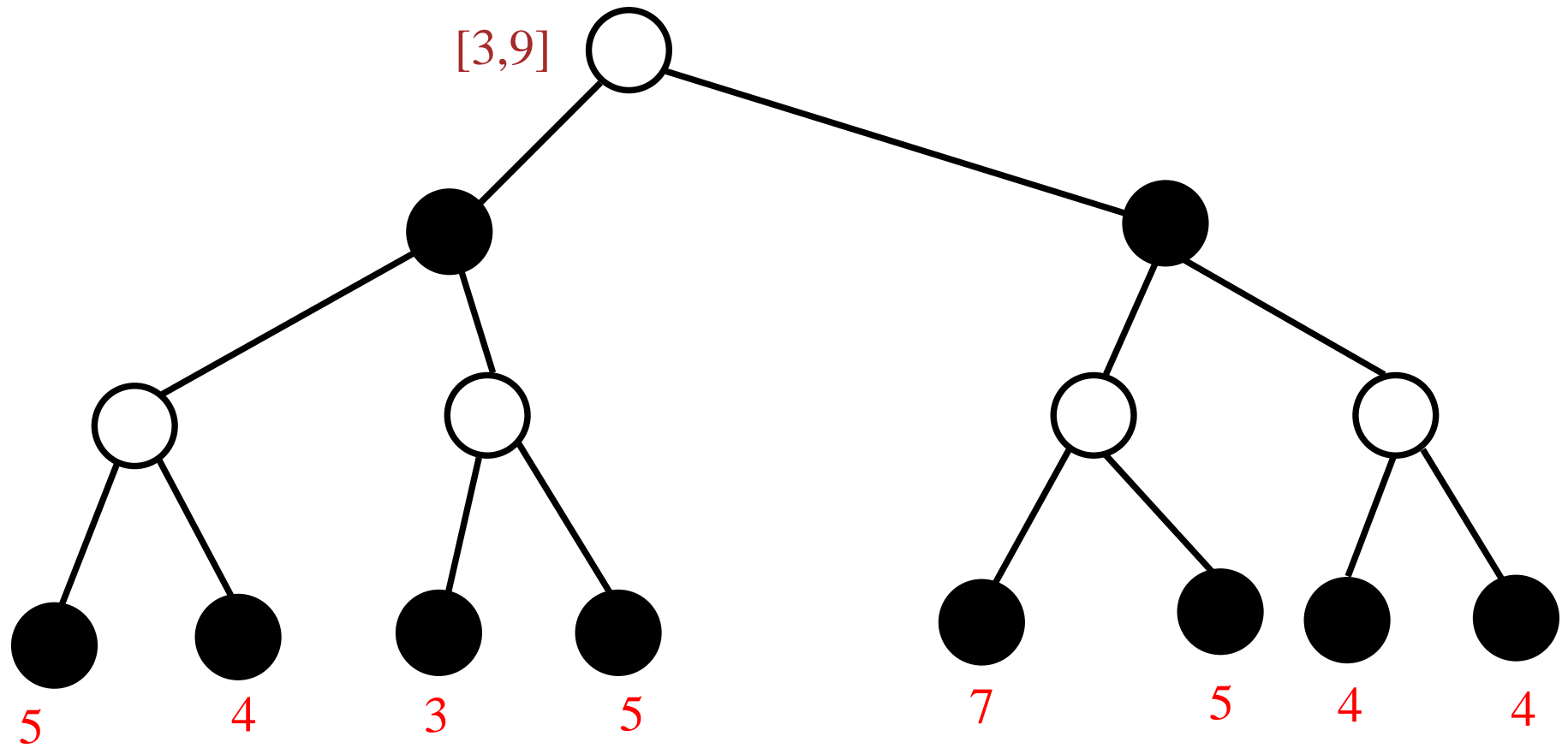
# The NegaScout Algorithm – MiniMax (1/2)

- Algorithm  $F4'$ (position  $p$ , value  $alpha$ , value  $beta$ , integer  $depth$ )
  - determine the successor positions  $p_1, \dots, p_d$
  - if  $d = 0$  // a terminal node  
or  $depth = 0$  //  $depth$  is the remaining depth to search  
or time is running up // from timing control  
or some other constraints are met // apply heuristic here
  - then return  $f(p)$  else  
begin
    - ▷  $m := -\infty$  //  $m$  is the current best lower bound; fail soft  
 $m := \max\{m, G4'(p_1, alpha, beta, depth - 1)\}$  // the first branch  
if  $m \geq beta$  then return( $m$ ) // beta cut off
    - ▷ for  $i := 2$  to  $d$  do
    - ▷ 9:  $t := G4'(p_i, m, m + 1, depth - 1)$  // null window search
    - ▷ 10: if  $t > m$  then // failed-high
    - 11: if ( $depth < 3$  or  $t \geq beta$ )
    - 12: then  $m := t$
    - 13: else  $m := G4'(p_i, t, beta, depth - 1)$  // re-search
    - ▷ 14: if  $m \geq beta$  then return( $m$ ) // beta cut off
  - end
  - return  $m$

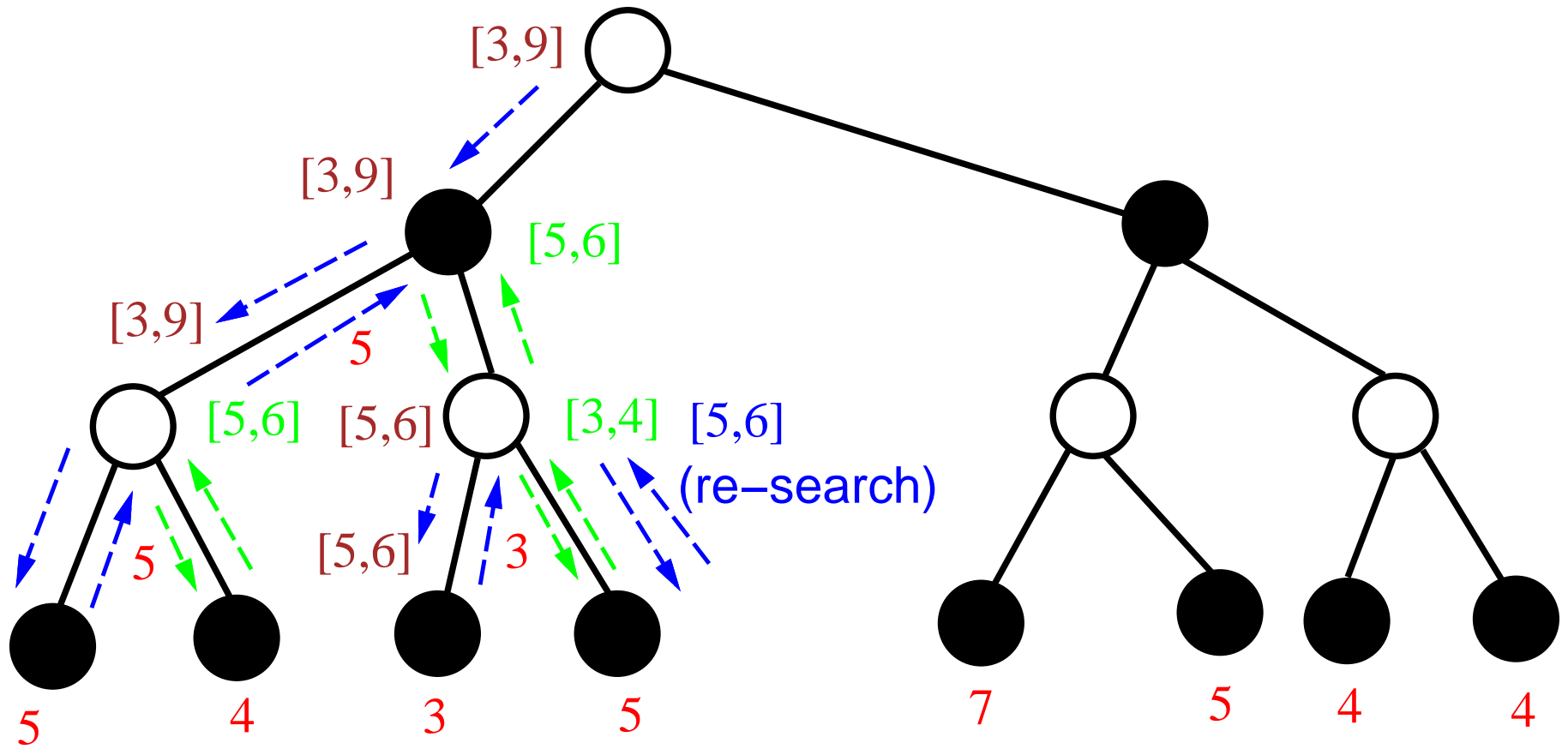
# The NegaScout Algorithm – MiniMax (2/2)

- Algorithm  $G4'$ (position  $p$ , value  $alpha$ , value  $beta$ , integer  $depth$ )
  - determine the successor positions  $p_1, \dots, p_d$
  - if  $d = 0$  // a terminal node  
or  $depth = 0$  //  $depth$  is the remaining depth to search  
or time is running up // from timing control  
or some other constraints are met // apply heuristic here
  - then return  $f(p)$  else  
begin
    - ▷  $m = \infty$  //  $m$  is the current best upper bound; fail soft  
 $m := \min\{m, F4'(p_1, alpha, beta, depth - 1)\}$  // the first branch  
if  $m \leq alpha$  then return( $m$ ) // alpha cut off
    - ▷ for  $i := 2$  to  $d$  do
    - ▷ 9:  $t := F4'(p_i, m, m + 1, depth - 1)$  // null window search
    - ▷ 10: if  $t \leq m$  then // failed-low
    - 11: if ( $depth < 3$  or  $t \leq alpha$ )
    - 12: then  $m := t$
    - 13: else  $m := F4'(p_i, alpha, t, depth - 1)$  // re-search
    - ▷ 14: if  $m \leq alpha$  then return( $m$ ) // alpha cut off
  - end
  - return  $m$

# NegaScout – MiniMax version (1/2)



# NegaScout – MiniMax version (2/2)



# The NegaScout Algorithm

- Use Nega-MAX format.
- Algorithm  $F4(\text{position } p, \text{value } \alpha, \text{value } \beta, \text{integer } \text{depth})$ 
  - determine the successor positions  $p_1, \dots, p_d$
  - if  $d = 0$  // a terminal node  
or  $\text{depth} = 0$  //  $\text{depth}$  is the remaining depth to search  
or time is running up // from timing control  
or some other constraints are met // apply heuristic here
  - then return  $h(p)$  else
    - ▷  $m := -\infty$  // the current lower bound; fail soft
    - ▷  $n := \beta$  // the current upper bound
    - ▷ for  $i := 1$  to  $d$  do
    - ▷ 9:  $t := -F4(p_i, -n, -\max\{\alpha, m\}, \text{depth} - 1)$
    - ▷ 10: if  $t > m$  then
    - 11: if  $(n = \beta \text{ or } \text{depth} < 3 \text{ or } t \geq \beta)$
    - 12: then  $m := t$
    - 13: else  $m := -F4(p_i, -\beta, -t, \text{depth} - 1)$  // re-search
    - ▷ 14: if  $m \geq \beta$  then return( $m$ ) // cut off
    - ▷ 15:  $n := \max\{\alpha, m\} + 1$  // set up a null window
  - return  $m$



# Search behaviors (1/3)

- If the depth is enough or it is a terminal position, then stop searching further.

- Return  $h(p)$  as the value computed by an evaluation function.
- Note:

$$h(p) = \begin{cases} f(p) & \text{if depth of } p \text{ is 0 or even} \\ -f(p) & \text{if depth of } p \text{ is odd} \end{cases}$$

- Fail soft version.
- For the first child  $p_1$ , search using the normal alpha beta window..
  - line 9: normal window for the first child
  - the initial value of  $m$  is  $-\infty$ , hence  $-\max\{\alpha, m\} = -\alpha$ 
    - ▷  $m$  is the current best value
  - that is, searching with the normal window  $[\alpha, \beta]$

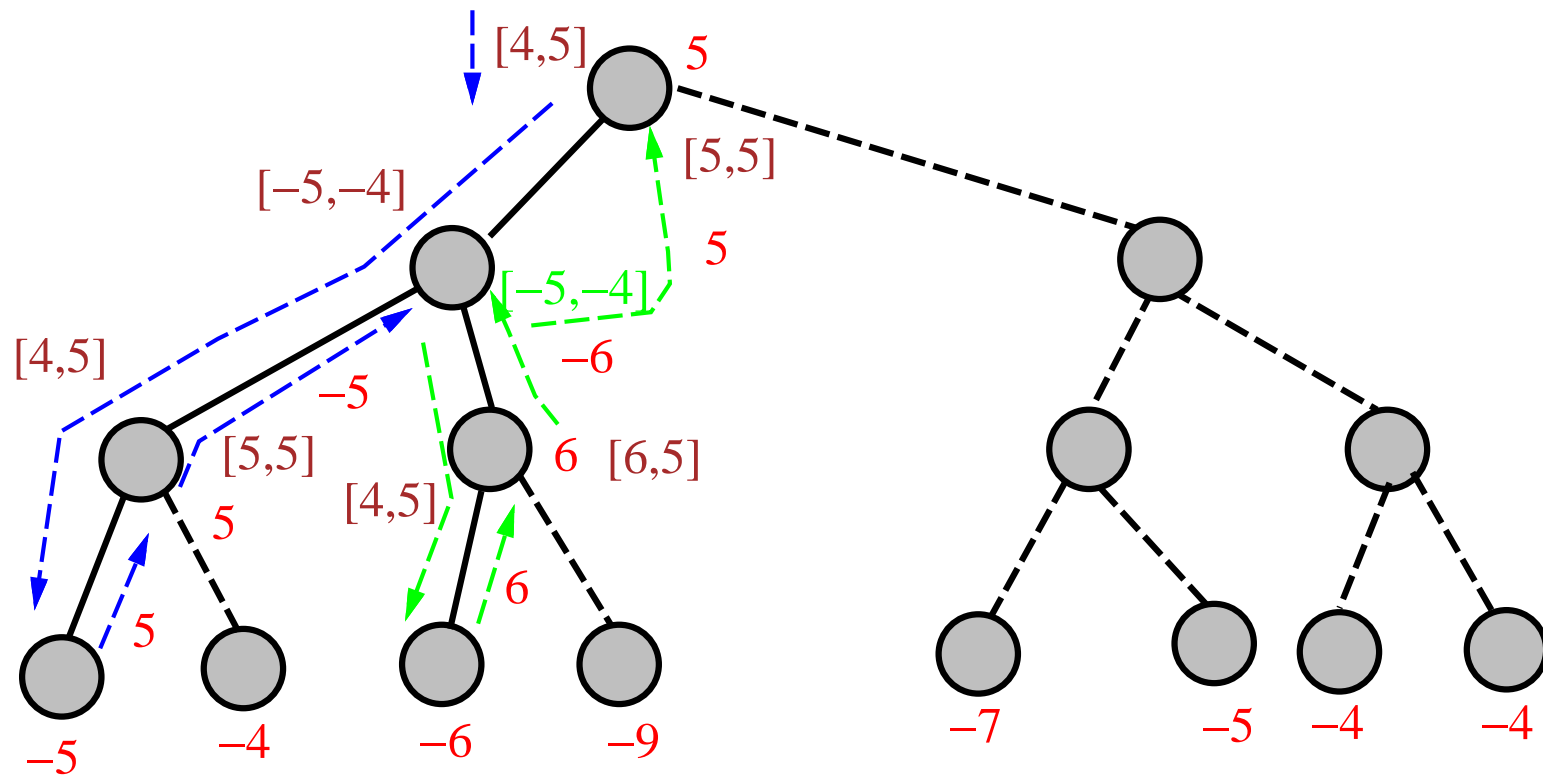
# Search behaviors (2/3)

- For the second child and beyond  $p_i$ ,  $i > 1$ , first perform a null window search for testing whether  $m$  is the answer.
  - line 9: a null-window of  $[m, m + 1]$  searches for the second child and beyond.
    - ▷  $m$  is best value obtained so far
    - ▷  $m$ 's value will be first set at line 12 because  $n = \text{beta}$
    - ▷ The null window is set at line 15.
  - line 11:
    - ▷  $n = \text{beta}$ : we are at first iteration.
    - ▷  $\text{depth} < 3$ : on a smaller depth subtree, i.e., depth at most 2, NegaScout always returns the best value.
    - ▷  $t \geq \text{beta}$ : we have obtained a good enough value from the failed-soft version to guarantee a beta cut.

# Search behaviors (3/3)

- For the second child and beyond  $p_i$ ,  $i > 1$ , first perform a null window search for testing whether  $m$  is the answer.
  - line 11: on a smaller depth subtree, i.e., depth at most 2, NegaScout always returns the best value.
    - ▷ *Normally, no need to do alpha-beta or any enhancement on very small subtrees.*
    - ▷ *The overhead is too large on small subtrees.*
  - line 13: **re-search** when the null window search fails high.
    - ▷ *The value of this subtree is at least  $t$ .*
    - ▷ *This means the best value in this subtree is more than  $m$ , the current best value.*
    - ▷ *This subtree must be re-searched with the the window  $[t, beta]$ .*
  - line 14: the normal pruning from alpha-beta.

# Example for NegaScout



# Refinements

- When a subtree is re-searched, it is best to use information on the previous search to speed up the current search.
  - Restart from the position that the value  $t$  is returned.
- Maybe want to re-search using the normal alpha-beta procedure.
- $F4$  runs much better with a good move ordering and transposition tables.
  - Order the moves in a priority list.
  - Reduce the number of re-searches.

# Performances

- Experiments done on a uniform random game tree [Reinefeld 1983].
  - Normally superior to alpha-beta when searching game trees with branching factors from 20 to 60.
  - Shows about 10 to 20% of improvement.

# Comments

- Incooperating both SCOUT and alpha-beta.
- Used in state-of-the-art game search engines.
- The first search, though maybe unsuccessful, can provide useful information in the second search.
  - Information can be stored and then be reused.

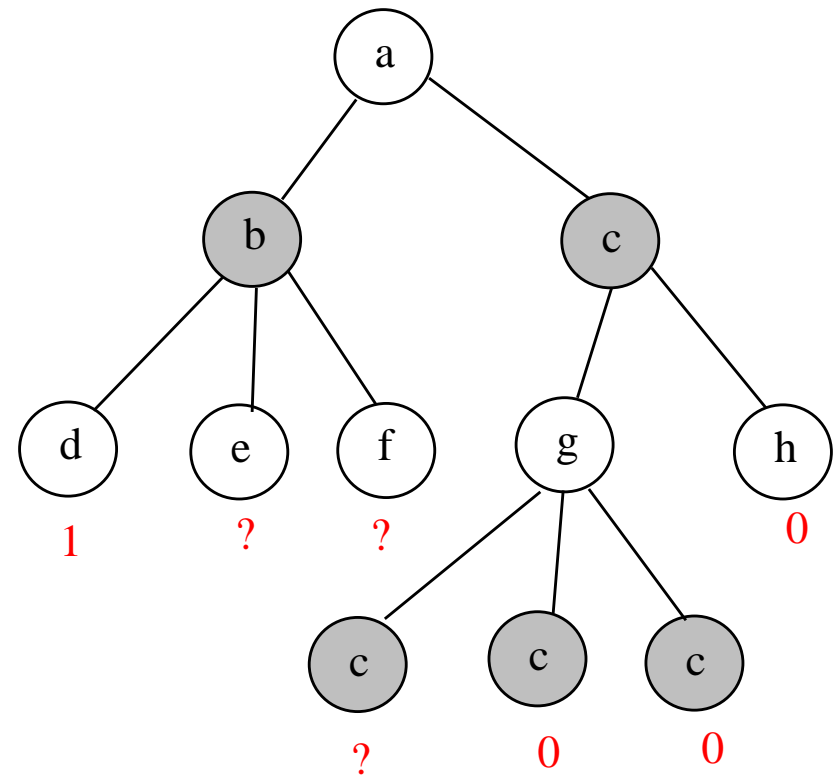
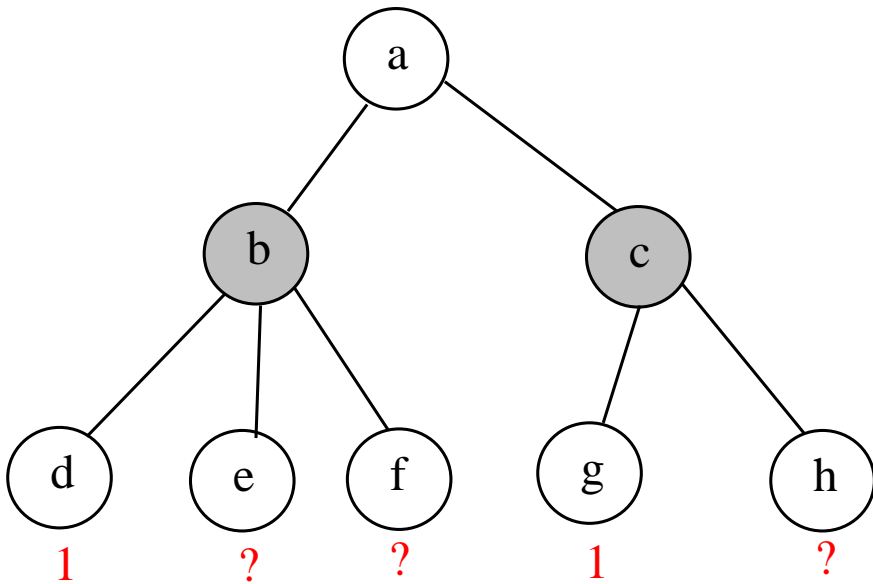
# Ideas for new search methods

- Consider the case of a 2-player game tree with either 0 or 1 on the leaves.
  - win, or not win which is lose or draw;
  - lose, or not lose which is win or draw;
  - Call this a **binary valued game tree**.
- If the game tree is known as well as the values of some leaves are known, can you make use of this information to search this game tree faster?
  - The value of the root is either 0 or 1.
  - If a branch of the root returns 1, then we know for sure the value of the root is 1.
  - The value of the root is 0 only when all branches of the root returns 0.
  - An AND-OR game tree search.



# Which node to search next?

- A most proving node for a node  $u$ : a node if its value is 1, then the value of  $u$  is 1.
- A most disproving node for a node  $u$ : a node if its value is 0, then the value of  $u$  is 0.



# Proof or Disproof Number

- Assign a **proof number** and a **disproof number** to each node  $u$  in a binary valued game tree.
  - $proof(u)$ : the minimum number of leaves needed to visited in order for the value of  $u$  to be 1.
  - $disproof(u)$ : the minimum number of leaves needed to visited in order for the value of  $u$  to be 0.

# Proof Number: Definition

- $u$  is a leaf:
  - If  $value(u)$  is unknown, then  $proof_v(u)$  is the cost of evaluating  $u$ .
  - If  $value(u)$  is 1, then  $proof(u) = 0$ .
  - If  $value(u)$  is 0, then  $proof(u) = \infty$ .
- $u$  is an internal node with children  $u_1, \dots, u_k$ :
  - if  $u$  is a MAX node,

$$proof(u) = \min_{i=1}^{i=k} proof(u_i);$$

- if  $u$  is a MIN node,

$$proof(u) = \sum_{i=1}^{i=k} proof(u_i).$$

# Disproof Number: Definition

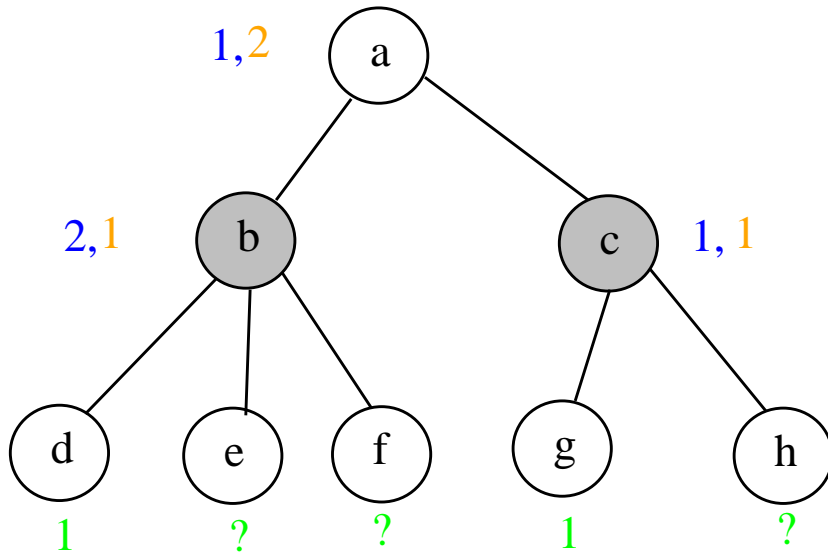
- $u$  is a leaf:
  - If  $value(u)$  is unknown, then  $proof_v(u)$  is cost of evaluating  $u$ .
  - If  $value(u)$  is 1, then  $disproof(u) = \infty$ .
  - If  $value(u)$  is 0, then  $disproof(u) = 0$ .
- $u$  is an internal node with children  $u_1, \dots, u_k$ :
  - if  $u$  is a MAX node,

$$disproof(u) = \sum_{i=1}^{i=k} disproof(u_i);$$

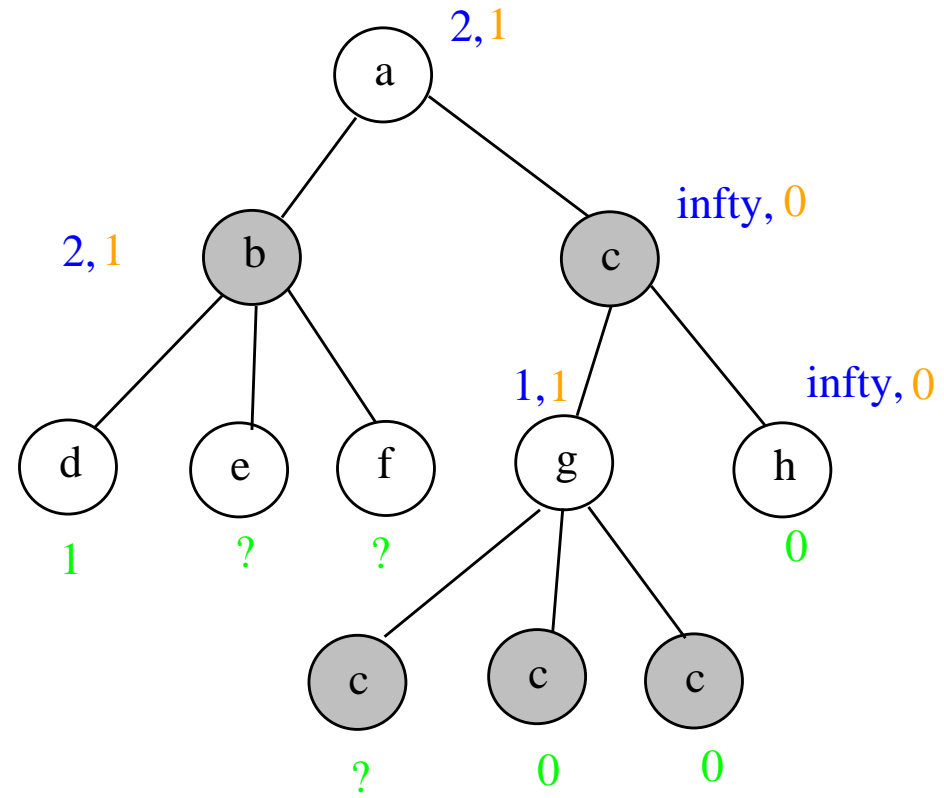
- if  $u$  is a MIN node,

$$disproof(u) = \min_{i=1}^{i=k} disproof(u_i).$$

# Illustrations



proof number, disproof number



proof number, disproof number

# How to Use these Numbers

- If the numbers are known in advance, then from the root, we search a child  $u$  with the value equals to  $\min\{proof(root), disproof(root)\}$ .
  - Then we find a path from the root towards a leaf recursively as follows,
    - ▷ if we try to prove it, then pick a child with the least proof number for a MAX node, and pick any node that has a chance to be proved for a MIN node.
    - ▷ if we try to disprove it, then pick a child with the least disproof number for a MIN node, and pick any node that has a chance to be disproved for a MAX node.
- Assume each leaf takes a lot of time to evaluate.
  - For example, the game tree represents an open game tree or an endgame tree.
  - Depends on the results we have so far, pick the next leaf to prove or disprove.
- Need to be able to update these numbers on the fly.

# PN-search: algorithm

- *loop*: **Compute or update proof and disproof numbers for each node in a bottom up fashion.**
  - If  $proof(root) = 0$  or  $disproof(root) = 0$ , then we are done, otherwise
    - ▷  $proof(root) \leq disproof(root)$ : we try to prove it.
    - ▷  $proof(root) > disproof(root)$ : we try to disprove it.
- $u \leftarrow root$ ; **{\* find the leaf to prove or disprove \*}**
  - if we try to prove, then
    - ▷ while  $u$  is not a leaf do
    - ▷ if  $u$  is a MAX node, then  
 $u \leftarrow$  leftmost child of  $u$  with the smallest non-zero proof number;
    - ▷ if current is a MIN node, then  
 $u \leftarrow$  leftmost child of  $u$  with a non-zero proof number;
  - if we try to disprove, then
    - ▷ while  $u$  is not a leaf do
    - ▷ if  $u$  is a MAX node, then  
 $u \leftarrow$  leftmost child of  $u$  with a non-zero disproof number;
    - ▷ if current is a MIN node, then  
 $u \leftarrow$  leftmost child of  $u$  with the smallest non-zero disproof number;
- **Prove or disprove  $u$ ; go to *loop*;**

# Multi-Valued game Tree

- The values of the leaves may not be binary.
  - Assume the values are non-negative integers.
  - Note: it can be in any finite countable domain.
- Revision of the proof and disproof numbers.
  - $proof_v(u)$ : the minimum number of leaves needed to visited in order for the value of  $u$  to  $\geq v$ .
    - ▷  $proof(u) = proof_1(u)$ .
  - $disproof_v(u)$ : the minimum number of leaves needed to visited in order for the value of  $u$  to  $< v$ .
    - ▷  $disproof(u) = disproof_1(u)$ .



# Multi-Valued Proof Number

- $u$  is a leaf:
  - If  $value(u)$  is unknown, then  $proof_v(u)$  is cost of evaluating  $u$ .
  - If  $value(u) \geq v$ , then  $proof_v(u) = 0$ .
  - If  $value(u) < v$ , then  $proof_v(u) = \infty$ .
- $u$  is an internal node with children  $u_1, \dots, u_k$ :
  - if  $u$  is a MAX node,

$$proof_v(u) = \min_{i=1}^{i=k} proof_v(u_i);$$

- if  $u$  is a MIN node,

$$proof_v(u) = \sum_{i=1}^{i=k} proof_v(u_i).$$

# Multi-valued Disproof Number

- $u$  is a leaf:
  - If  $value(u)$  is unknown, then  $proof_v(u)$  is cost of evaluating  $u$ .
  - If  $value(u) \geq v$  is 1, then  $disproof_v(u) = \infty$ .
  - If  $value(u) < v$  is 0, then  $disproof_v(u) = 0$ .
- $u$  is an internal node with children  $u_1, \dots, u_k$ :
  - if  $u$  is a MAX node,

$$disproof_v(u) = \sum_{i=1}^{i=k} disproof_v(u_i);$$

- if  $u$  is a MIN node,

$$disproof_v(u) = \min_{i=1}^{i=k} disproof_v(u_i).$$

# Revised PN-search( $v$ ): algorithm

- **loop: Compute or update  $proof_v$  and  $disproof_v$  numbers for each node in a bottom up fashion.**
  - **If  $proof_v(root) = 0$  or  $disproof_v(root) = 0$ , then we are done, otherwise**
    - ▷  $proof_v(root) \leq disproof_v(root)$ : we try to prove it.
    - ▷  $proof_v(root) > disproof_v(root)$ : we try to disprove it.
- **$u \leftarrow root$ ; { \* find the leaf to prove or disprove \* }**
  - **if we try to prove, then**
    - ▷ while  $u$  is not a leaf do
    - ▷ if  $u$  is a MAX node, then  
 $u \leftarrow$  leftmost child of  $u$  with the smallest non-zero  $proof_v$  number;
    - ▷ if current is a MIN node, then  
 $u \leftarrow$  leftmost child of  $u$  with a non-zero  $proof_v$  number;
  - **if we try to disprove, then**
    - ▷ while  $u$  is not a leaf do
    - ▷ if  $u$  is a MAX node, then  
 $u \leftarrow$  leftmost child of  $u$  with a non-zero  $disproof_v$  number ;
    - ▷ if current is a MIN node, then  
 $u \leftarrow$  leftmost child of  $u$  with the smallest non-zero  $disproof_v$  number;
- **Prove or disprove  $u$ ; go to loop;**

# Multi-valued PN-search: algorithm

- When the values of the leaves are not binary, use an open value binary search to find an upper bound of the value.
  - Set the initial value of  $v$  to be 1.
  - *loop*:  $\text{PN-search}(v)$ 
    - ▷ *Prove the value of the search tree is  $\geq v$  or disprove it by showing it is  $< v$ .*
  - If it is proved, then double the value of  $v$  and go to *loop* again.
  - If it is disproved, then the true value of the tree is between  $\lfloor v/2 \rfloor$  and  $v - 1$ .
  - **{\* Use a binary search to find the exact returned value of the tree. \*}**
  - $low \leftarrow \lfloor v/2 \rfloor$ ;  $high \leftarrow v - 1$ ;
  - **while**  $low \leq high$  **do**
    - ▷ *if  $low = high$ , then return  $low$  as the tree value*
    - ▷  $mid \leftarrow \lfloor (low + high)/2 \rfloor$
    - ▷  $\text{PN-search}(mid)$
    - ▷ *if it is disproved, then  $high \leftarrow mid - 1$*
    - ▷ *else if it is proved, then  $low \leftarrow mid$*

# Comments

- **Appears to be good for certain searching certain game trees.**
  - Find the easiest way to prove or disprove a conjecture.
  - A dynamic strategy depends on work has been done so far.
- **Take into consideration the fact that some nodes may need more time to process than the other nodes.**

# References and further readings

- \* J. Pearl. Asymptotic properties of minimax trees and game-searching procedures. *Artificial Intelligence*, 14(2):113–138, 1980.
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