

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.
 - ▷ *Alpha-beta cut algorithm is one way to make sure of this exactly.*
 - ▷ *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

- 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 [Judea Pearl 1980].
- High level idea:
 - While searching a branch T_i of a MAX node, if we have already obtained a lower bound v_ℓ .
 - ▷ *First TEST whether it is possible for T_i to return something greater than v_ℓ .*
 - ▷ *If FALSE, then there is no need to search T_i . This is called **fails the test**.*
 - ▷ *If TRUE, then search T_i . This is called **passes the test**.*
 - While searching a branch T_j of a MIN node, if we have already obtained an upper bound v_u .
 - ▷ *First TEST whether it is possible for T_j to return something smaller than v_u .*
 - ▷ *If FALSE, then there is no need to search T_j . This is called **fails the test**.*
 - ▷ *If TRUE, then search T_j . This is called **passes the test**.*

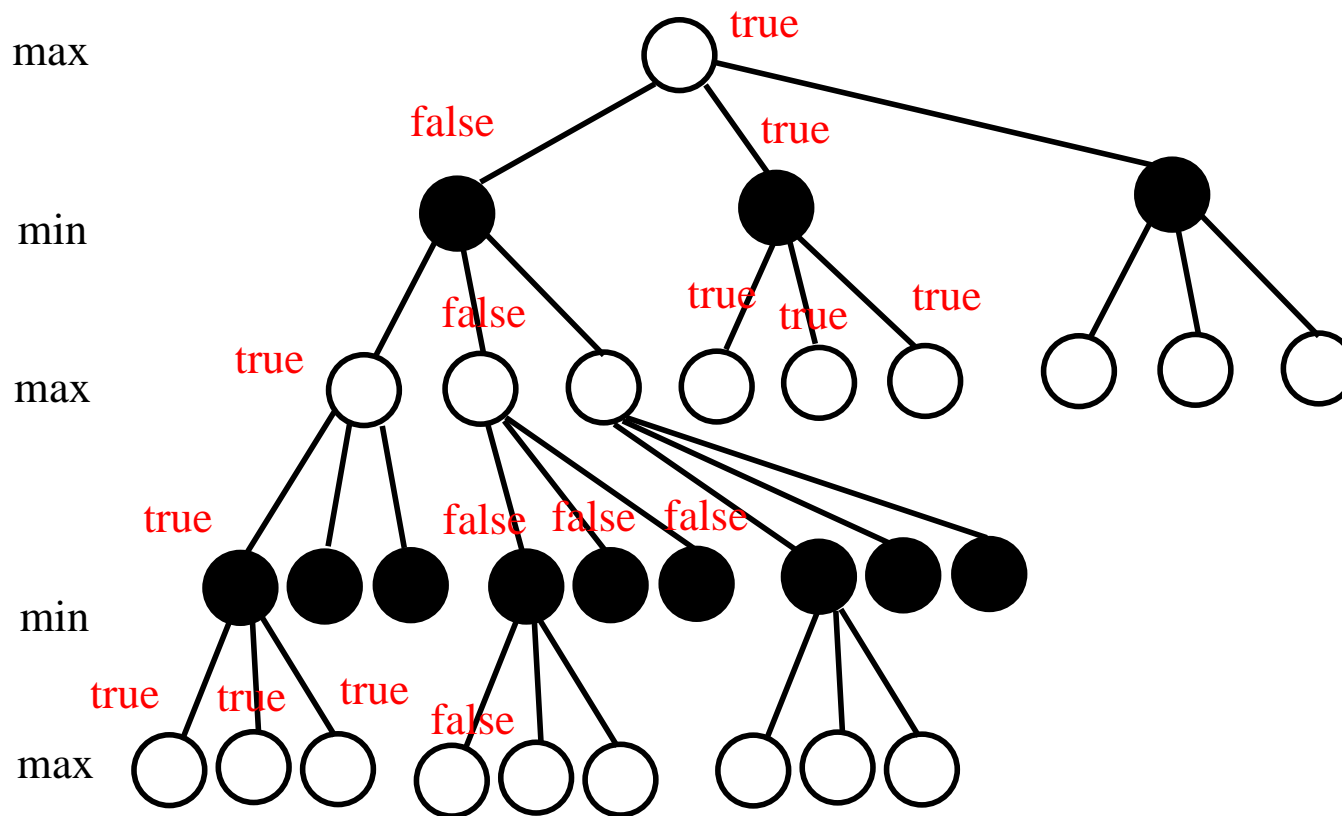
How to TEST $> v$

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

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

- determine the successor positions p_1, \dots, p_b of p
- if $b = 0$, 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 b 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 b 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

- Sometimes it may be needed to test for “ $\geq v$ ”, “ $< v$ ” or “ $\leq v$ ”.

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

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

Main SCOUT procedure

Algorithm SCOUT(position p)

- determine the successor positions p_1, \dots, p_b
- if $b = 0$, then return $f(p)$
- else $v = SCOUT(p_1)$ // **SCOUT the first branch**
- if p is a **MAX** node
 - for $i := 2$ to b 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 b 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

How to TEST $< v$

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

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

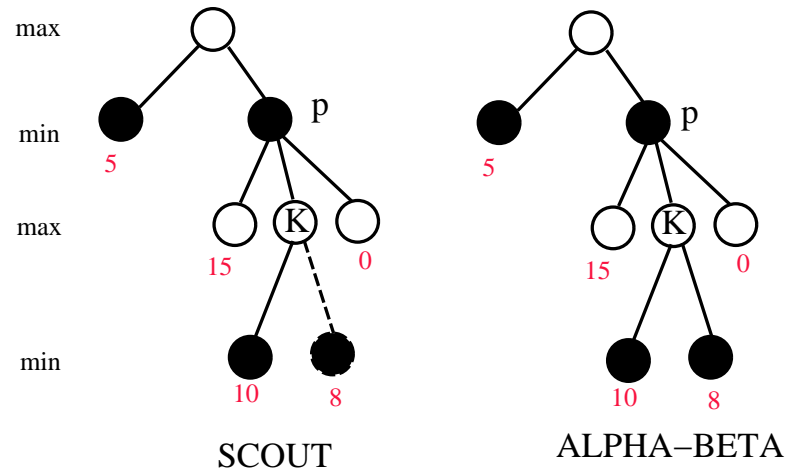
- determine the successor positions p_1, \dots, p_b of p
- if $b = 0$, then *// terminal*
 - ▷ if $f(p) < v$ then *// f(): evaluating function*
 - ▷ return TRUE
 - ▷ else return FALSE
- if p is a MAX node, then
 - for $i := 1$ to b do
 - ▷ if TEST($p_i, <, v$) is FALSE, then
return FALSE *// fail if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

Discussions for SCOUT (1/3)

- 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 (2/3)

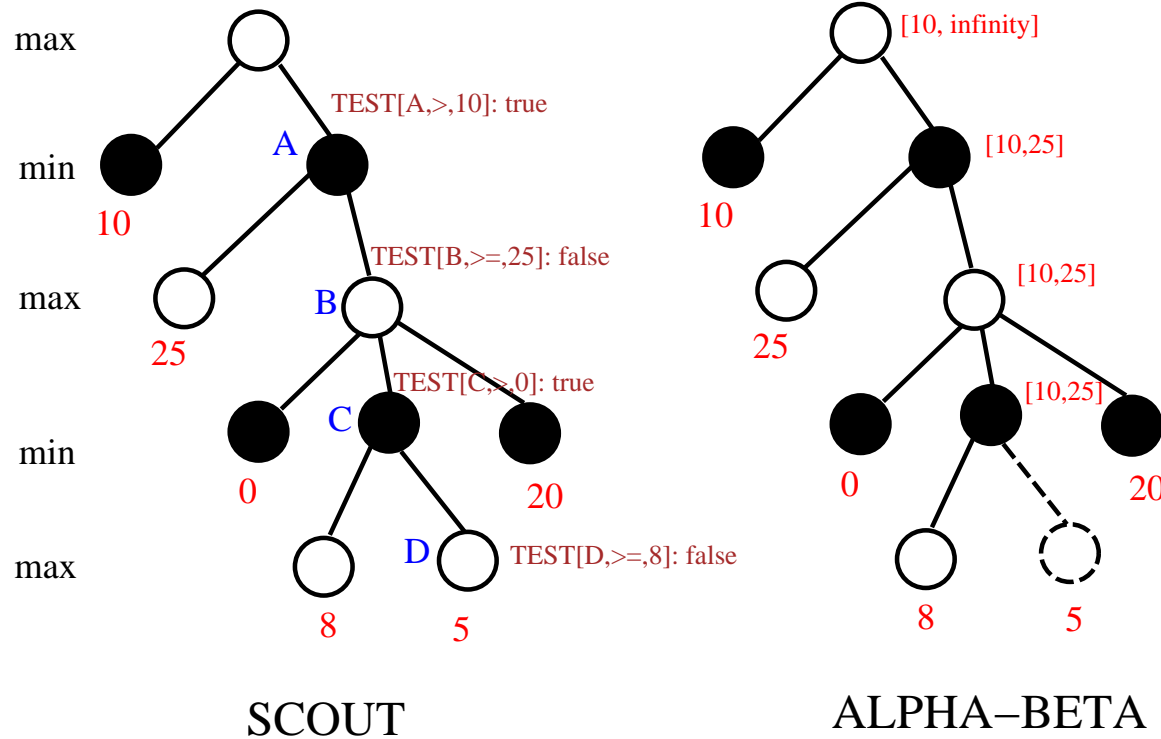
- TEST which is called by SCOUT may visit less nodes than that of alpha-beta.



- 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 (3/3)

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



Number of nodes visited (1/4)

- **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 b , the number of nodes evaluated is $\Omega(b^{\ell/2})$ where ℓ is the depth of the tree.*
- **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 b , the number of nodes evaluated is $\Omega(b^{\ell/2})$.*

Number of nodes visited (2/4)

■ Assumptions:

- Assume a full complete b -ary tree with depth ℓ where ℓ is even.
- The depth of the root, which is a MAX node, is 0.

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

■ H_1 : the minimum number of nodes visited by TEST when it returns TRUE.

$$\begin{aligned} H_1 &= 1 + 1 + b + b + b^2 + b^2 + b^3 + b^3 + \dots + b^{\ell/2-1} + b^{\ell/2-1} + b^{\ell/2} \\ &= 2 \cdot (b^0 + b^1 + \dots + b^{\ell/2}) - b^{\ell/2} \\ &= 2 \cdot \frac{b^{\ell/2+1}-1}{b-1} - b^{\ell/2} \end{aligned}$$

Number of nodes visited (3/4)

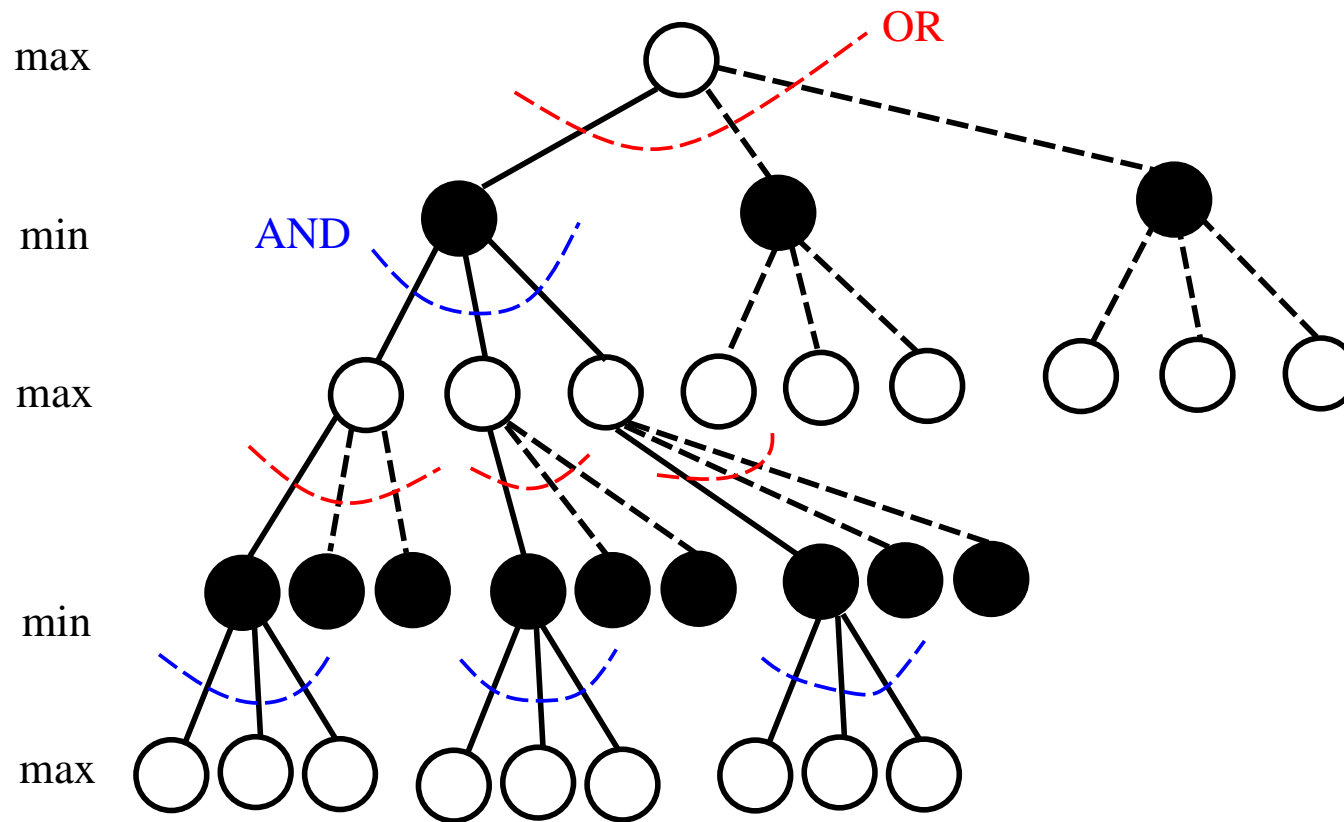
■ Assumptions:

- Assume a full complete b -ary tree with depth ℓ where ℓ is even.
- The depth of the root, which is a MAX node, is 0.

■ H_2 : the minimum number of nodes visited by alpha-beta.

$$\begin{aligned}H_2 &= \sum_{i=0}^{\ell} (b^{\lceil i/2 \rceil} + b^{\lfloor i/2 \rfloor} - 1) \\&= \sum_{i=0}^{\ell} b^{\lceil i/2 \rceil} + \sum_{i=0}^{\ell} b^{\lfloor i/2 \rfloor} - (\ell + 1) \\&= \sum_{i=0}^{\ell} b^{\lceil i/2 \rceil} + H_1 - (\ell + 1) \\&= (1 + b + b + \dots + b^{\ell/2-1} + b^{\ell/2} + b^{\ell/2}) + H_1 - (\ell + 1) \\&= (H_1 - 1 + b^{\ell/2} - b^{\ell/2-1}) + H_1 - (\ell + 1) \\&= 2 \cdot H_1 + b^{\ell/2} - b^{\ell/2-1} - (\ell + 2) \\&\sim (2.x) \cdot H_1\end{aligned}$$

Number of nodes visited (4/4)



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 ℓ .
 - ▷ *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 ℓ 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 or equal to its upper bound $beta$.
 - **Failed-low** means it returns a value that is smaller than or equal to 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 value used in TEST.*
- Use a fail soft version to get a better result when the returned value is out of the window.

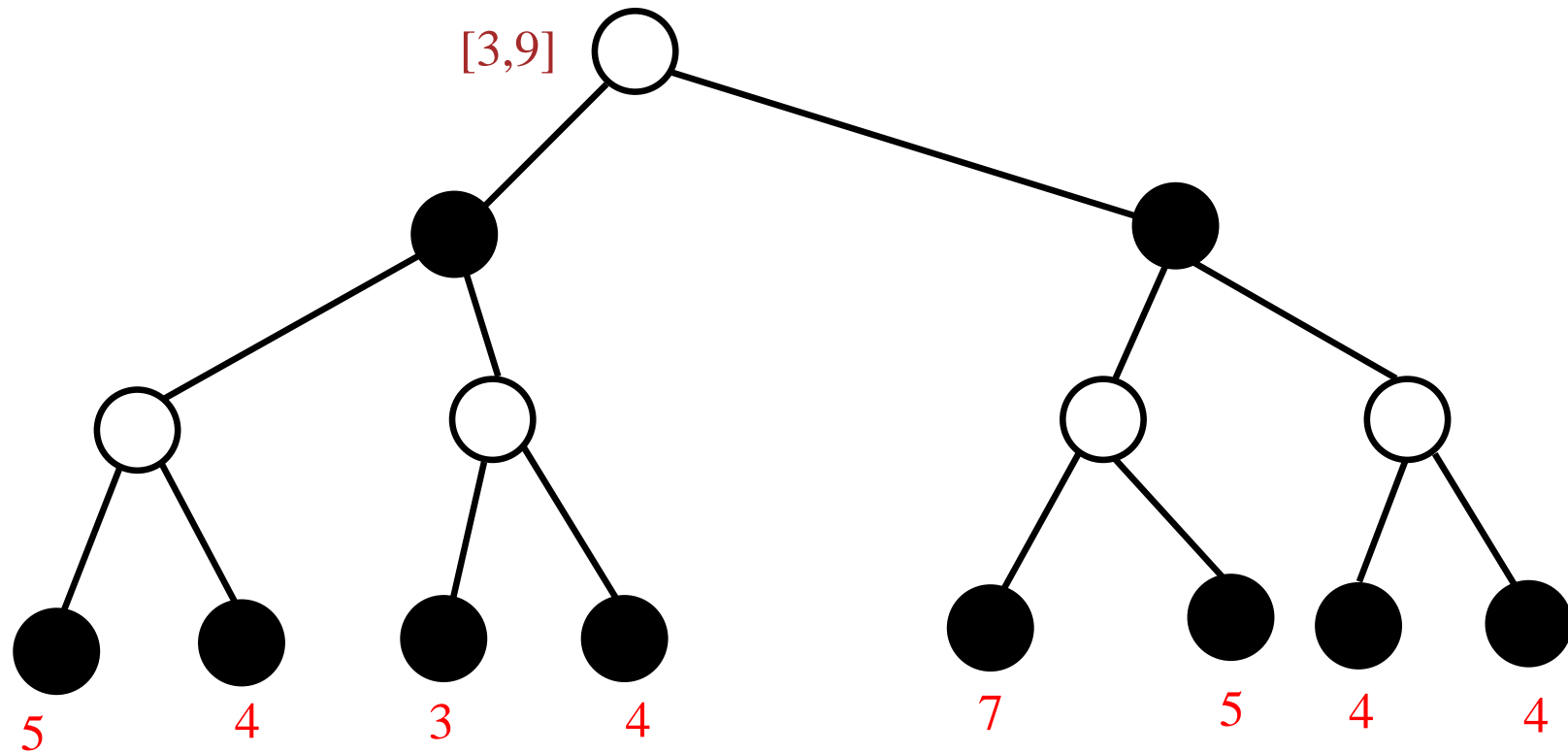
The NegaScout Algorithm – MiniMax (1/2)

- Algorithm $F4'$ (position p , value $alpha$, value $beta$, integer $depth$)
 - determine the successor positions p_1, \dots, p_b
 - if $b = 0$ // 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 b 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

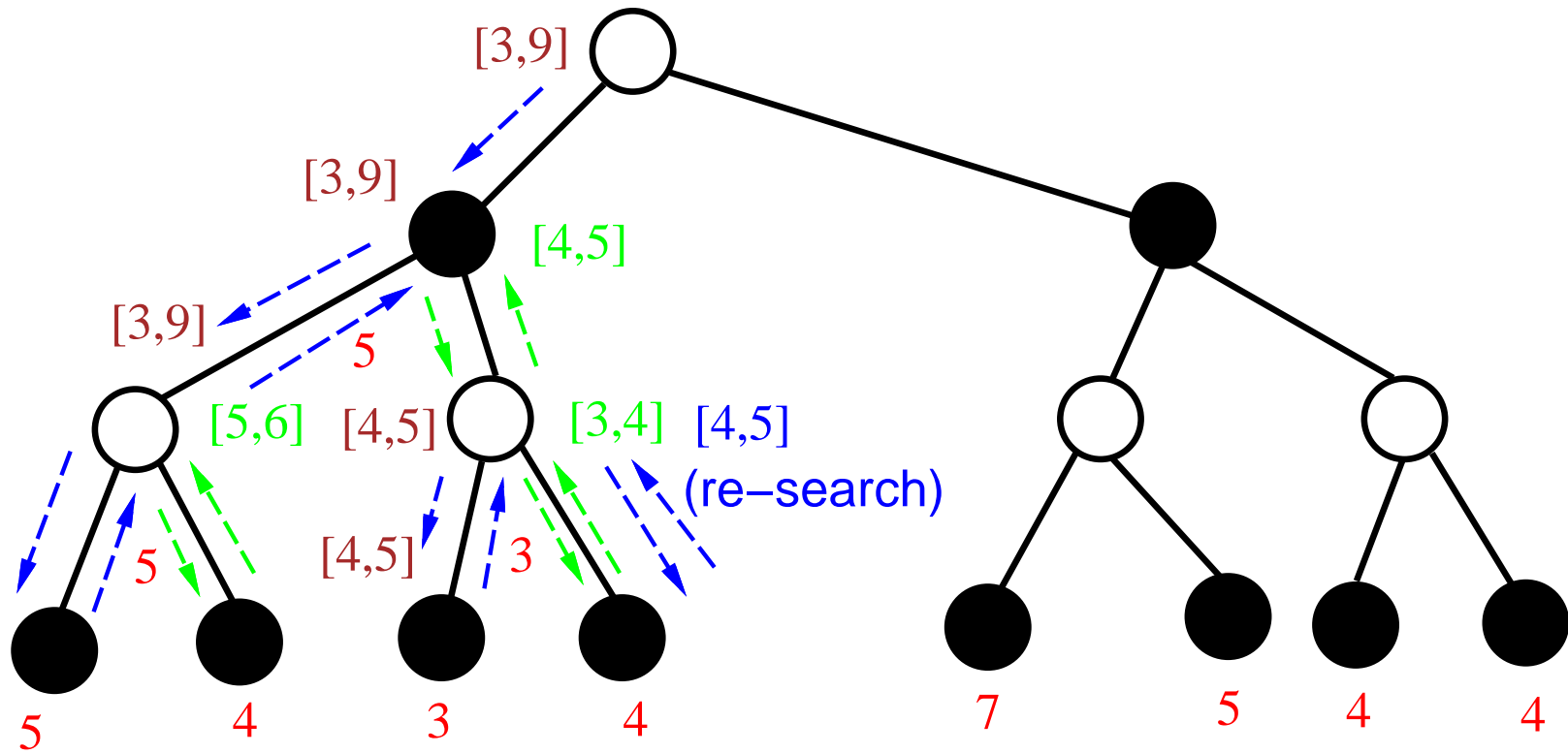
The NegaScout Algorithm – MiniMax (2/2)

- Algorithm $G4'$ (position p , value $alpha$, value $beta$, integer $depth$)
 - determine the successor positions p_1, \dots, p_b
 - if $b = 0$ // 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 b do
 - ▷ 9: $t := F4'(p_i, m - 1, m, depth - 1)$ // null window search
 - ▷ 10: if $t < m$ 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 $F4(\text{position } p, \text{value } \alpha, \text{value } \beta, \text{integer } \textit{depth})$
 - determine the successor positions p_1, \dots, p_b
 - if $b = 0$ // a terminal node
or $\textit{depth} = 0$ // \textit{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 b do
 - ▷ 9: $t := -F4(p_i, -n, -\max\{\alpha, m\}, \textit{depth} - 1)$
 - ▷ 10: if $t > m$ then
 - 11: if ($n = \beta$ or $\textit{depth} < 3$ or $t \geq \beta$)
 - 12: then $m := t$
 - 13: else $m := -F4(p_i, -\beta, -t, \textit{depth} - 1)$ // 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 $[\alpha, \beta]$

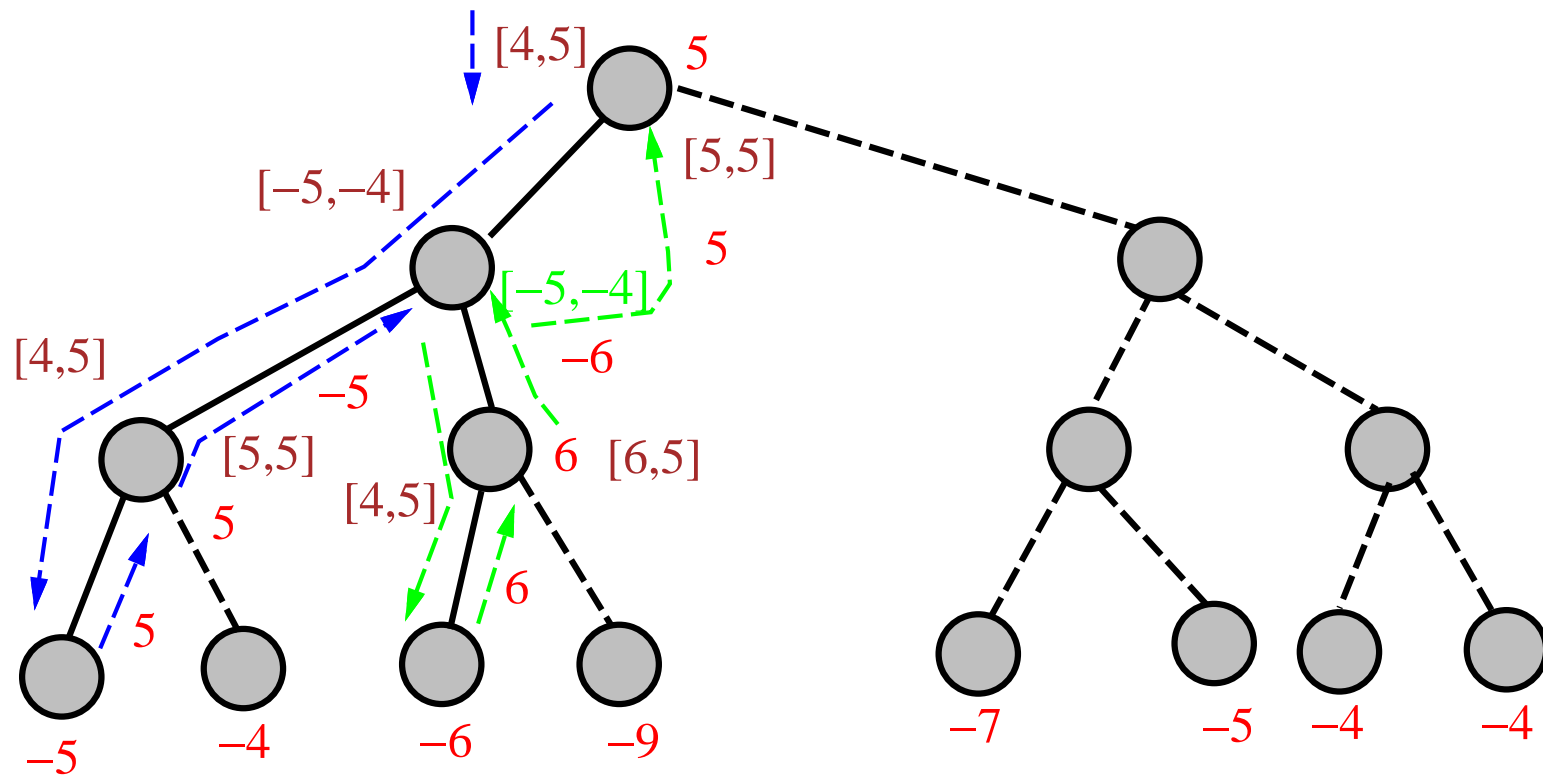
Search behaviors (2/3)

- 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 $[n - 1, n]$ searches for the second child and beyond where $n = \max\{\alpha, m\} + 1$.
 - ▷ m is best value obtained so far
 - ▷ α is the previous lower bound
 - ▷ m 's value will be first set at line 12 because $n = \beta$
 - ▷ The value of n is set at line 15.
 - line 11:
 - ▷ $n = \beta$: 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.
 - ▷ $t \geq \beta$: 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

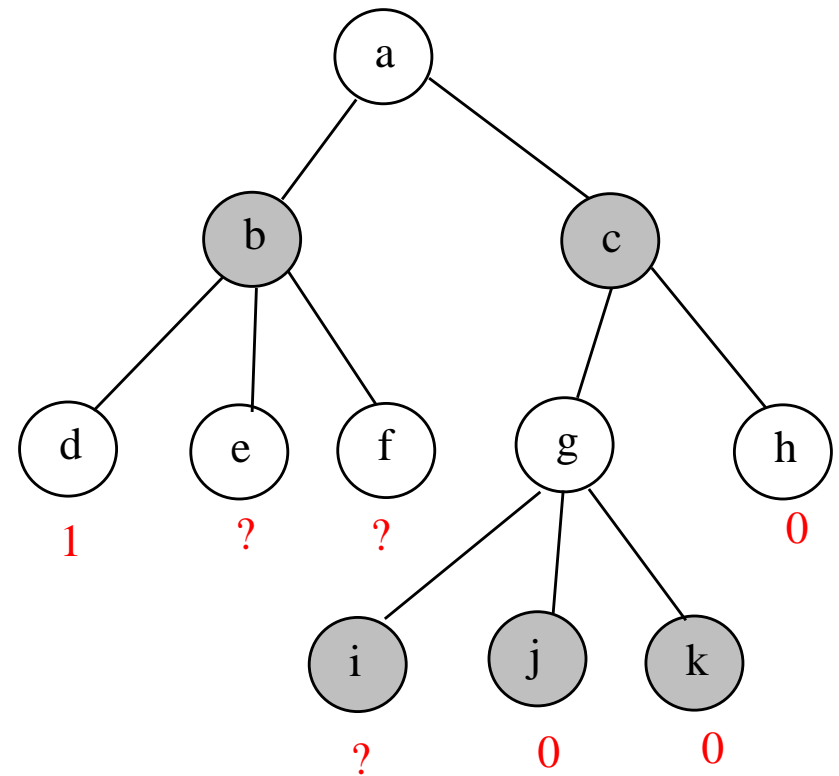
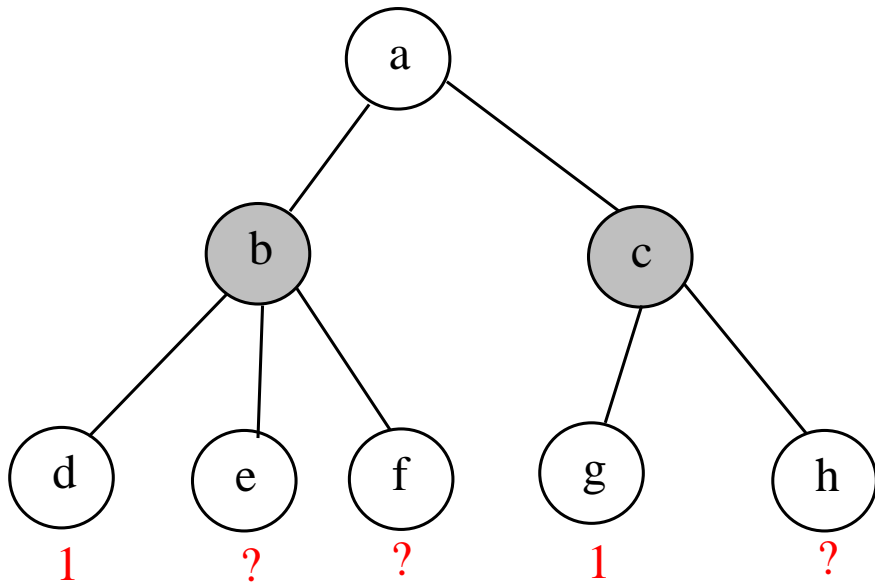
- Incooperating both SCOUT and alpha-beta.
- Used in state-of-the-art game search engines.
- The first search, though maybe unsuccessful, can provide useful information in the second search.
 - Information can be stored and then be reused.

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 $proof(u)$ 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 all of the children u_1, \dots, u_b :
 - if u is a MAX node,

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

- if u is a MIN node,

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

Disproof Number: Definition

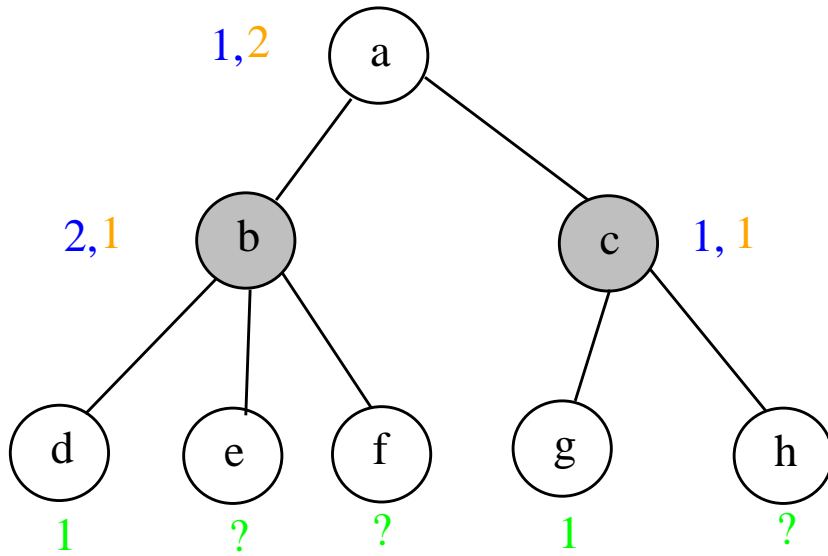
- u is a leaf:
 - If $value(u)$ is unknown, then $disproof(u)$ is cost of evaluating u .
 - If $value(u)$ is 1, then $disproof(u) = \infty$.
 - If $value(u)$ is 0, then $disproof(u) = 0$.
- u is an internal node with all of the children u_1, \dots, u_b :
 - if u is a MAX node,

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

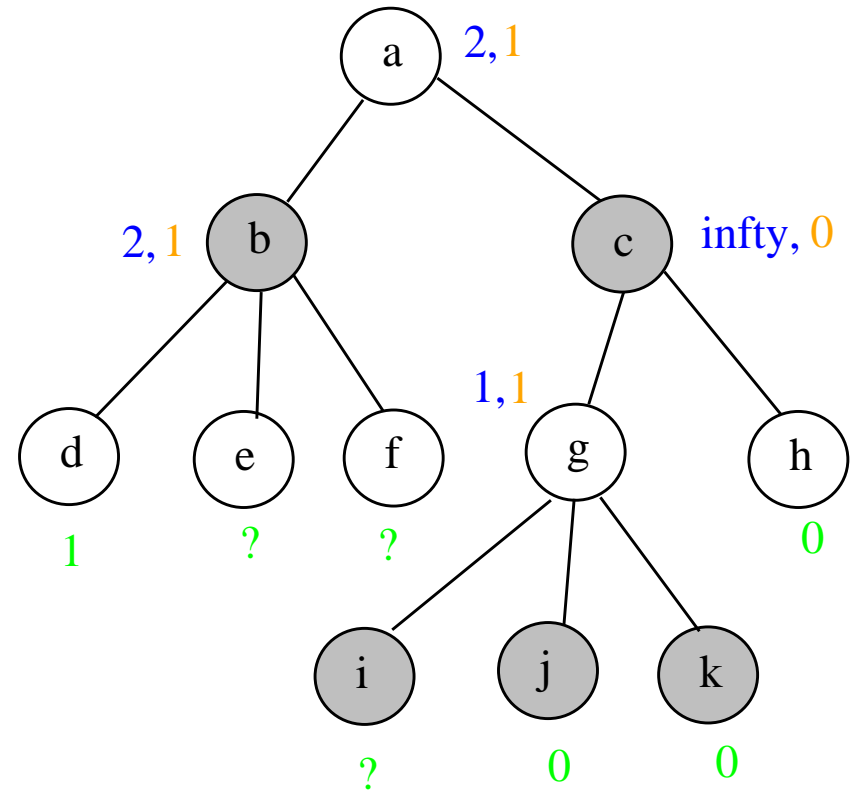
- if u is a MIN node,

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

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.

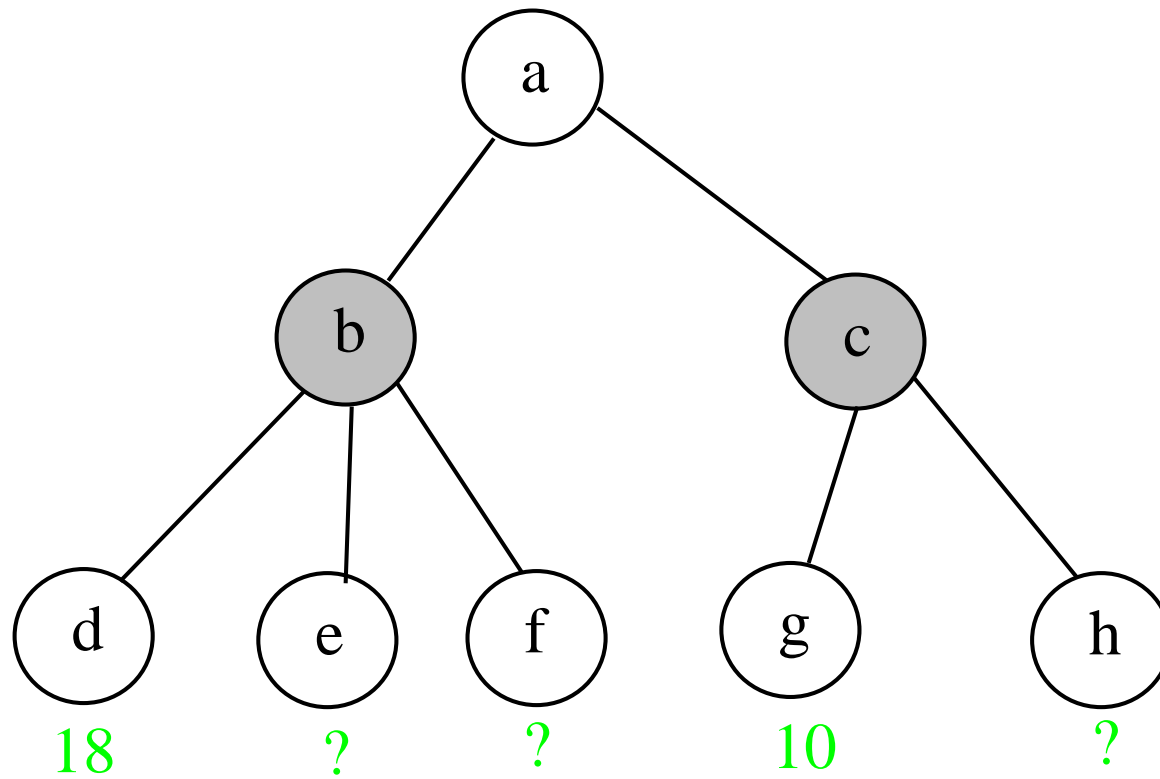
PN-search: algorithm

- *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*;**

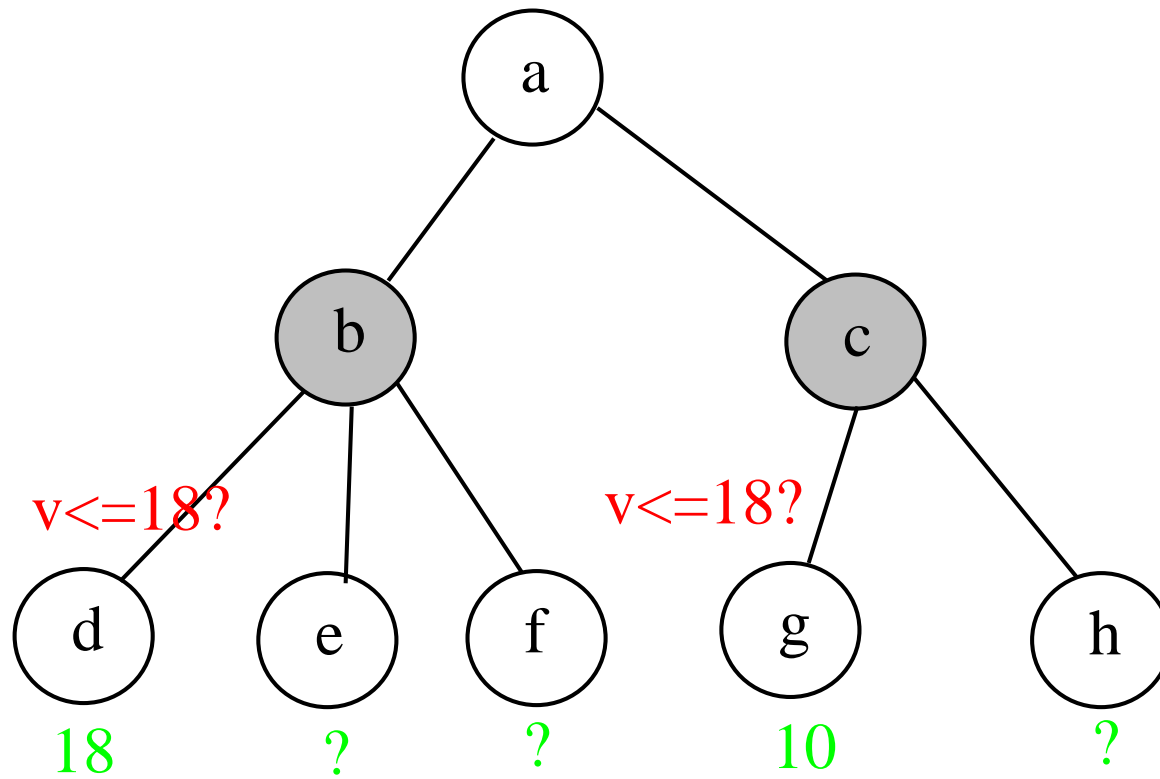
Multi-Valued game Tree

- 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 $\geq v$.
 - ▷ $proof(u) \equiv proof_1(u)$.
 - $disproof_v(u)$: the minimum number of leaves needed to visited in order for the value of u to $< v$.
 - ▷ $disproof(u) \equiv disproof_1(u)$.

Illustration



Illustration



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 all of the children u_1, \dots, u_b :
 - if u is a MAX node,

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

- if u is a MIN node,

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

Multi-valued Disproof Number

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

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

- if u is a MIN node,

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

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*: 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*
 - ▷ *$mid \leftarrow \lfloor (low + high)/2 \rfloor$*
 - ▷ *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 searching certain types of game trees.**
 - Find the easiest way to prove or disprove a conjecture.
 - A dynamic strategy depends on work has been done so far.
- **Performance has nothing to do with move ordering.**
 - Performance of most previous algorithms depends heavily on whether a good move ordering can be found.
- **Searching the “easiest” branch may not give you the best performance.**
 - Performance depends on the value of each internal nodes.
- **Commonly used in verifying conjectures, e.g., first-player win.**
 - Partition the opening moves in a tree-like fashion.
 - Try to the “easiest” way to prove or disprove the given conjecture.
- **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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