Scout and NegaScout

Tsan-sheng Hsu

徐讚昇

tshsu@iis.sinica.edu.tw

`http://www.iis.sinica.edu.tw/~tshsu`

Abstract

- It looks like alpha-beta pruning is the best we can do for an exact 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 by returning an exact value.*
    - ▷ *Is there a way to search a tree by only returning a bound?*
    - ▷ *Is searching with a bound 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.***
      $\Rightarrow$ This is called fails the test.
    - ▷ ***If TRUE, then search T_i.***
      $\Rightarrow$ 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.***
      $\Rightarrow$ This is called fails the test.
    - ▷ ***If TRUE, then search T_j.***
      $\Rightarrow$ This is called passes the test.

How to TEST $> v$

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

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

- **determine the successor positions $p_1, \ldots, p_b$ of p**
- **if $b = 0$, then // terminal**
  - ▷ *if $f(p) > v$ then // f(): evaluation 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 $> v$**

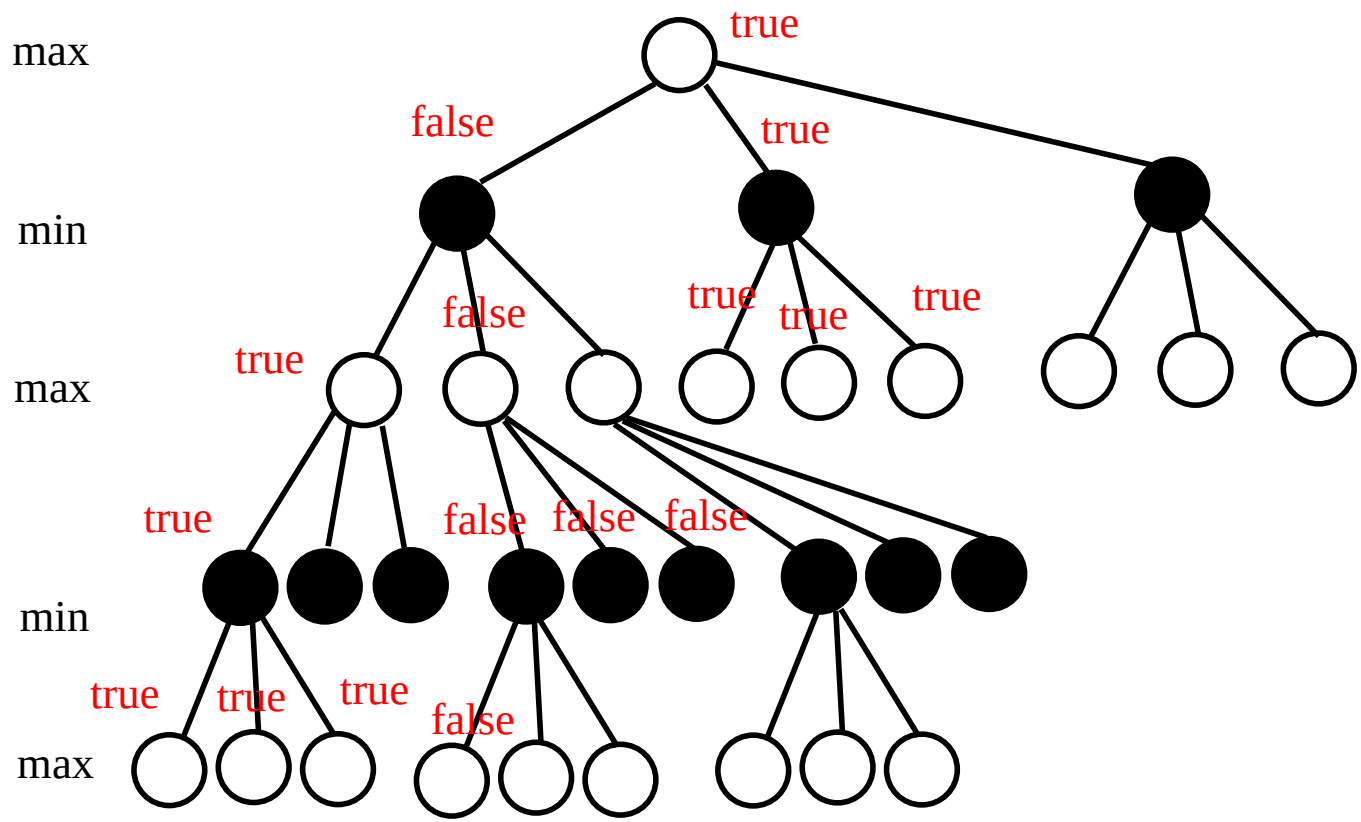
How to TEST $< v$

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

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

- **determine the successor positions $p_1, \ldots, p_b$ of p**
- **if $b = 0$, then // terminal**
  - ▷ *if $f(p) < v$ then // f(): evaluation 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 $\geq v$**

Illustration of TEST$_>$



Short circuit operations for TEST$_>$

- **For a MAX node:**
  - **if a branch is TRUE, then there is no need to do further testing;**
  - **if a branch is FALSE, then we need to do more testing on other branches.**
  - **It is better to test branches with better probabilities of being TRUE first.**
- **For a MIN node:**
  - **if a branch is FALSE, then there is no need to do further testing;**
  - **if a branch is TRUE, then we need to do more testing on other branches.**
  - **It is better to test branches with better probabilities of being FALSE first.**

How to TEST — Discussions

- **Sometimes it may be needed to test for "$\geq v$", or "$\leq v$".**
  - **TEST$_{>}(p,v)$ is TRUE** $\equiv$ **TEST$_{\leq}(p,v)$ is FALSE**
  - **TEST$_{>}(p,v)$ is FALSE** $\equiv$ **TEST$_{\leq}(p,v)$ is TRUE**
  - **TEST$_{<}(p,v)$ is TRUE** $\equiv$ **TEST$_{\geq}(p,v)$ is FALSE**
  - **TEST$_{<}(p,v)$ is FALSE** $\equiv$ **TEST$_{\geq}(p,v)$ is TRUE**
- **Practical consideration:**
  - **Set a depth limit and evaluate the position's value when the limit is reached.**

Main SCOUT procedure

Algorithm SCOUT(position p)

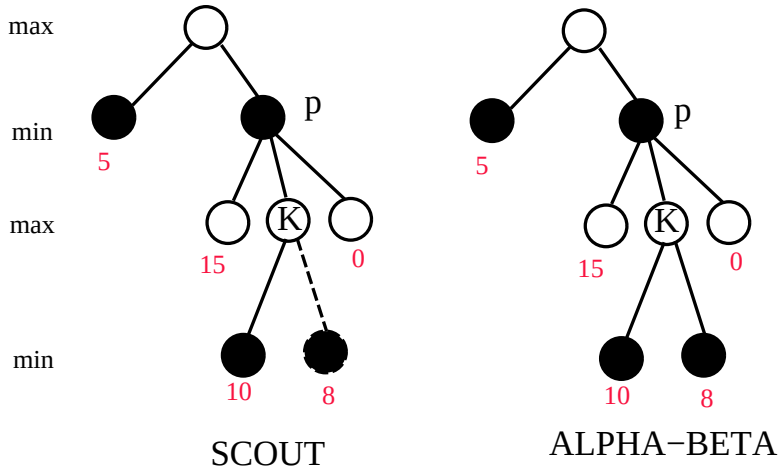
- **determine the successor positions** $p_1, \ldots, p_b$
- **if** $b = 0$**, then return** $f(p)$
- **else** $v = SCOUT(p_1)$ **// SCOUT the first branch**
- **if** p **is a MAX node**
  - **for** $i := 2$ **to** b **do**
    - ▷ ***if*** $TEST_{>}(p_i, v)$ ***is TRUE, // TEST first for the rest of 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

Discussions for SCOUT (1/3)

- **Initially, we use recursive call to find the value v of the first branch.**
- **From now on, 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;***
    - ▷ ***and make this the new v.***
  - **We say that 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;***
    - ▷ ***and make this the new v.***
  - **We say that p_i passes the test if TEST$_<$(p_i, v) is TRUE.**

Discussions for SCOUT (2/3)

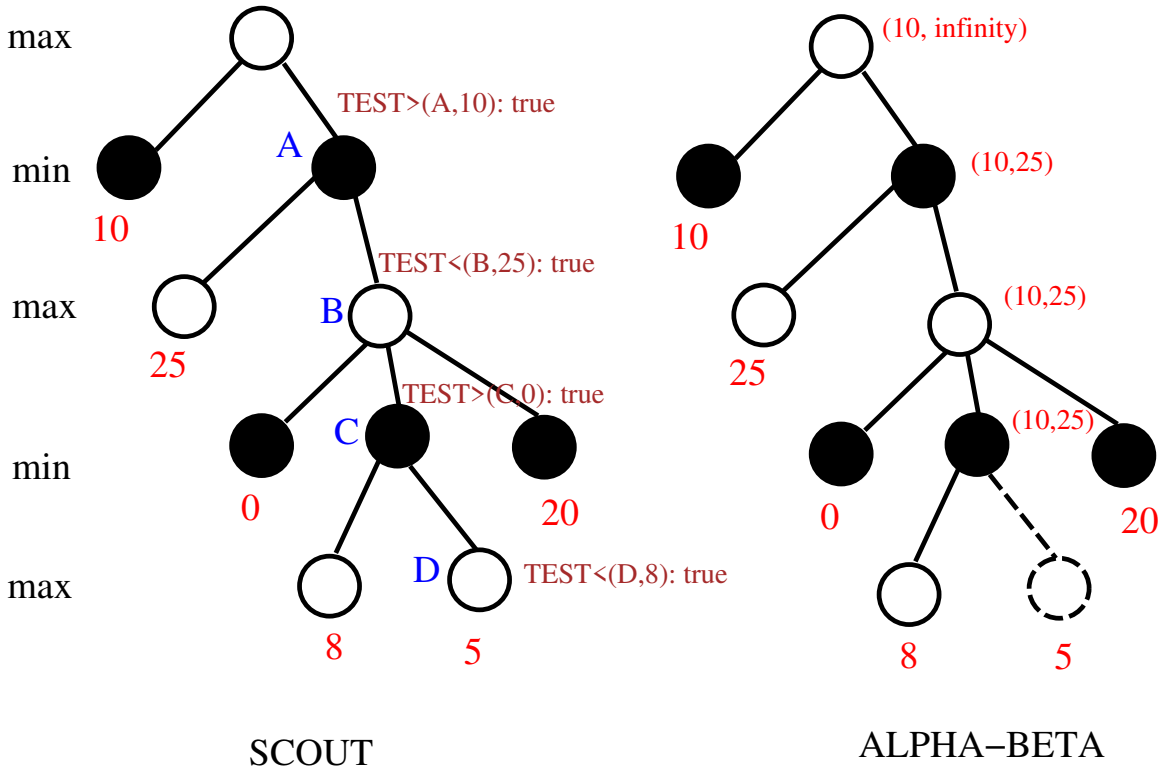
- **TEST which is called by SCOUT may visit less nodes than that of alpha-beta.**



  - **Assume TEST$_>$(p,5) is called by the root after the first branch of the root 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 rooted p.***
  - **Alpha-beta needs to visit K's second branch.**

Discussions for SCOUT (3/3)

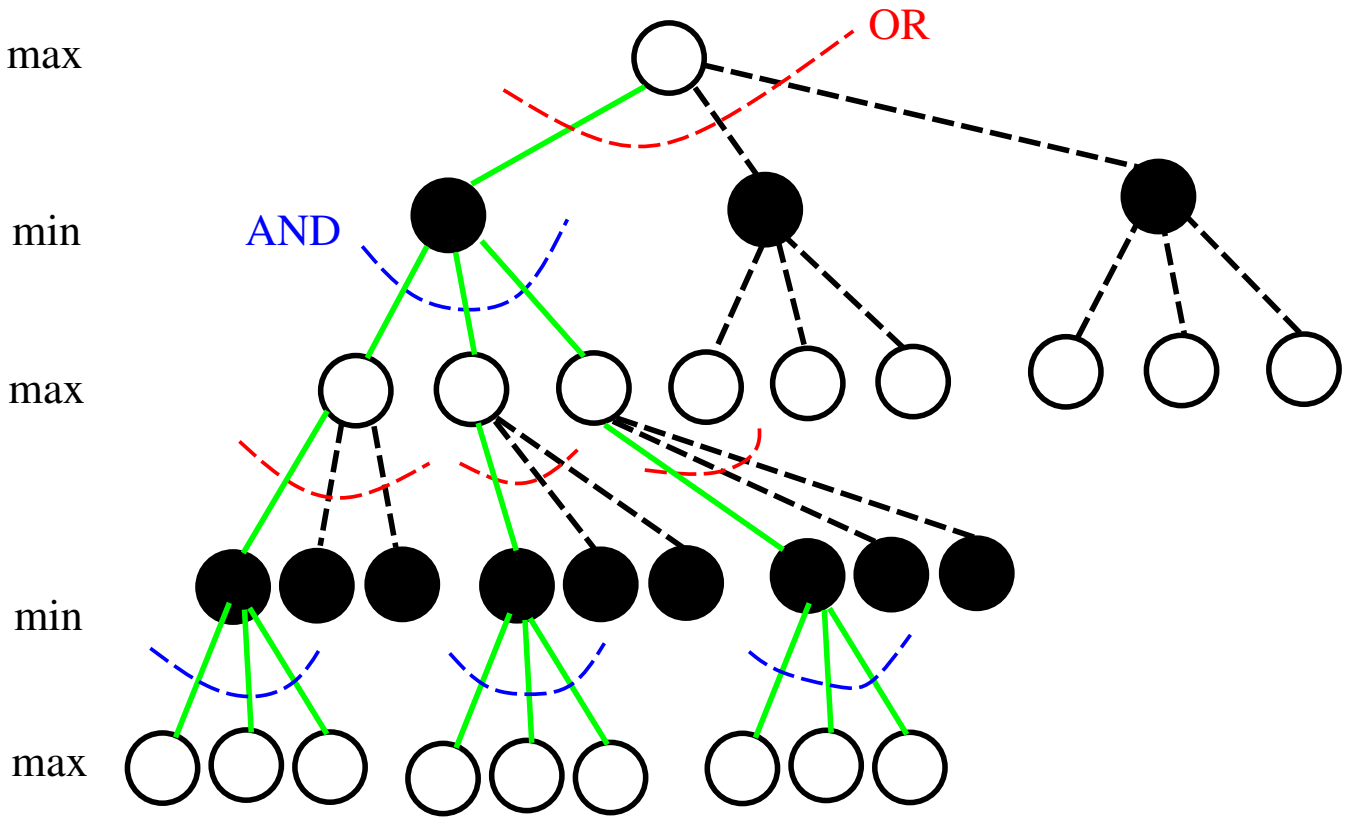
- **SCOUT may pay many visits to a node that is cut off by alpha-beta.**



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)



Number of nodes visited (3/4)

- **Assumptions:**
  - **Assume a full complete b-ary tree with depth ℓ.**
  - **The depth of the root, which is a MAX node, is 0.**
  - **Assume ℓ is even in the analysis.**
- **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.**

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

Number of nodes visited (4/4)

- **Assumptions:**
  - **Assume a full complete b-ary tree with depth ℓ.**
  - **The depth of the root, which is a MAX node, is 0.**
  - **Assume ℓ is even in the analysis.**
- **H_2: the minimum number of nodes visited by alpha-beta.**

$$
\begin{aligned}
H_2 &= \sum_{i=0}^{\ell}(b^{\lceil i/2\rceil} + b^{\lfloor i/2\rfloor} - 1) \\
&= \sum_{i=0}^{\ell} b^{\lceil i/2\rceil} + \sum_{i=0}^{\ell} b^{\lfloor i/2\rfloor} - (\ell+1) \\
&= \sum_{i=0}^{\ell} b^{\lceil i/2\rceil} + H_1 - (\ell+1) \\
&= (1 + b + b + \cdots + b^{\ell/2-1} + b^{\ell/2} + b^{\ell/2}) + H_1 - (\ell+1) \\
&= (H_1 - 1 + b^{\ell/2}) + H_1 - (\ell+1) \\
&= 2 \cdot H_1 + b^{\ell/2} - (\ell+2) \\
&\geq 2 \cdot H_1 \text{ if } b > 3
\end{aligned}
$$

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 ith branch needs to be searched exactly.**
  - **If the value of the first $i-1$ branches of the root is better than the value of ith branch, then we do not have to evaluate exactly for the ith 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 pruning as long as some relative move orderings are "good."**
    - ▷ *The moving orders of your children and the children of your ancestor who is odd level up "together" decide a cut-off.*
  - **The bounds are updated during searching.**
    - ▷ *Sometimes, a deep alpha-beta cut-off occurs because a bound found from your ancestor a distance away.*

Performance of SCOUT (1/3)

- **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.**
  - **For every node v in a branch that is not the first visited child of its parent with a depth[1] of ℓ,**
    - ▷ ***every ancestor of v may initiate a TEST to visit v.***
    - ▷ ***It can be visited ℓ times by TEST.***

[1] The depth of the root is defined to be 0.

Performance of SCOUT (2/3)

- **Show great improvements on $depth > 3$ over brute-force methods for games with small branching factors.**
  - **It traverses most of the nodes without evaluating them preciously.**
  - **Few subtrees remained to be revisited to compute their exact mini-max values.**
- **Show good improvement over alpha-beta on game trees with certain characteristics.**
- **Experimental data on the game of Kalah show [UCLA Tech Rep UCLA-ENG-80-17, A comparison of the Alpha-Beta and SCOUT algorithms using the game of Kalah, Noe 1980]:**
  - **SCOUT favors "skinny" game trees, that are game trees with high depth-to-width ratios.**
    - ▷ ***Q: why?***
  - **On depth = 5, it saves over 40% of time.**
  - **May not be good for games with large branching factors.**
  - **Move ordering is very important.**
    - ▷ ***The first branch, if is good, offers a great chance of pruning further branches.***

Performance of SCOUT (3/3)

- **Comparing alpha-beta pruning and SCOUT [Pearl 1984] on uniform game trees:**
  - **Alpha-beta is always better than SCOUT in the experiments using random game trees.**
    - ***In theory, when both are in their best cases, SCOUT cuts out more, but this rarely happens in practice.***
  - **Let $r_{b,d} = \frac{N_{scout}}{N_{AB}}$ where N_{scout} is the nodes searched using SCOUT and N_{AB} is the nodes searched using alpha-beta on depth-d random-valued game trees with a uniform branching factor of b.**
    - $1 \leq r_{b,d} \leq 1.275$ ***for any positive integers*** b ***and*** d***.***
    - $r_{b_1,d} \geq r_{b_2,d}$ ***if*** $b_1 \leq b_2$***: ratio is closer when the branching factor is larger.***
    - $r_{b,d_1} \geq r_{b,d_2}$ ***if*** $d_1 \leq d_2$***: ratio is closer when the searching depth is larger.***
    - $r_{2,20} \sim 1.04$***.***
    - $r_{b,20} \sim 1$***: after depth*** > 20***, the two are almost the same.***

Comments

- **Q1:**
  - **Currently, we use a "feasible" test to decide whether we need to search this branch or not.**
    - ▷ *If a new branch has a chance of larger than v, then we explore it in details. Otherwise, we skip it.*
  - **How about using the idea of "infeasible" test?**
    - ▷ *If a new branch has no chance of larger than v, then we do not explore it in details. Otherwise, we do.*
  - **How about a hybrid approach?**
    - ▷ *When to use one instead of the other?*
- **Q2: What can we do with regard to the first branch?**
  - **Can some previous values of some previous positions be used?**
  - **When iterative deepening is used, can we use previous results?**

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,$m+1$).**
    - ▷ ***Can never happen in a normal alpha-beta pruning when starts with $(-\infty,\infty)$.***
  - **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.***
- **The above argument works for the shallow fail hard ($F1$), general fail hard ($F2$) and general fail soft ($F3$) versions of the alpha-beta algorithm.**

Behaviors of Null window search

- **When $F2(p, m, m+1, \infty)$ returns $m+1$:**
  - **for the MAX node p, returns immediately after the first child p_i, namely the smallest index i, returning a value $\geq m+1$.**
  - **for the MIN node p_i, every child $p_{i,j}$ returns a value $\geq m+1$**
  - **for each MAX node $p_{i,j}$, returns immediately after the first child $r_{i,j,k}$, namely the smallest index k, returning a value $\geq m+1$.**
  - ...
  - **Remark: $F3(p, m, m+1, \infty)$ returns a value $\geq m+1$ in this case.**
- **Exactly like the OR-AND tree shown in TEST$_>$ when TEST is passed.**
- **We can observe similar behaviors when $F2(p, m, m+1, \infty)$ returns m as if TEST is failed.**
  - **Remark: $F3(p, m, m+1, \infty)$ returns a value $\leq m$ in this case.**

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.
    - *TEST replies only on a value from the first branch.*
    - *Using the values from a recursively maintained search window for TEST can do more cutting.*
  - 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 – Mini-Max (1/2)

- **Algorithm** $F4'$**(position** p**, value** $alpha$**, value** $beta$**, integer** $depth$**)**
  - **determine the successor positions** $p_1, \ldots, p_b$
  - **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 ***is max possible or*** $m \geq beta$ ***then return***(m) // ***beta cut off***

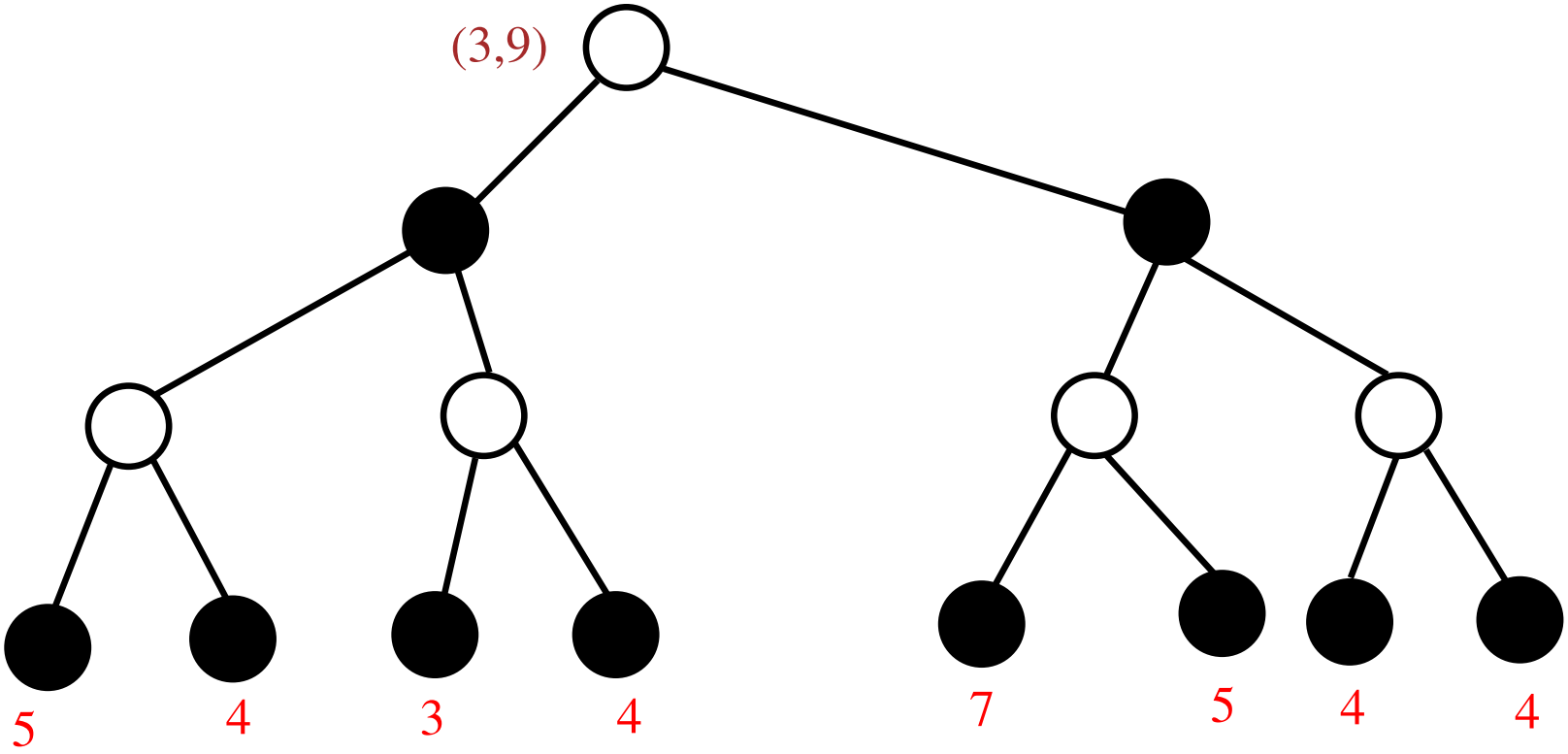
    end
  - **return** m

The NegaScout Algorithm – Mini-Max (2/2)

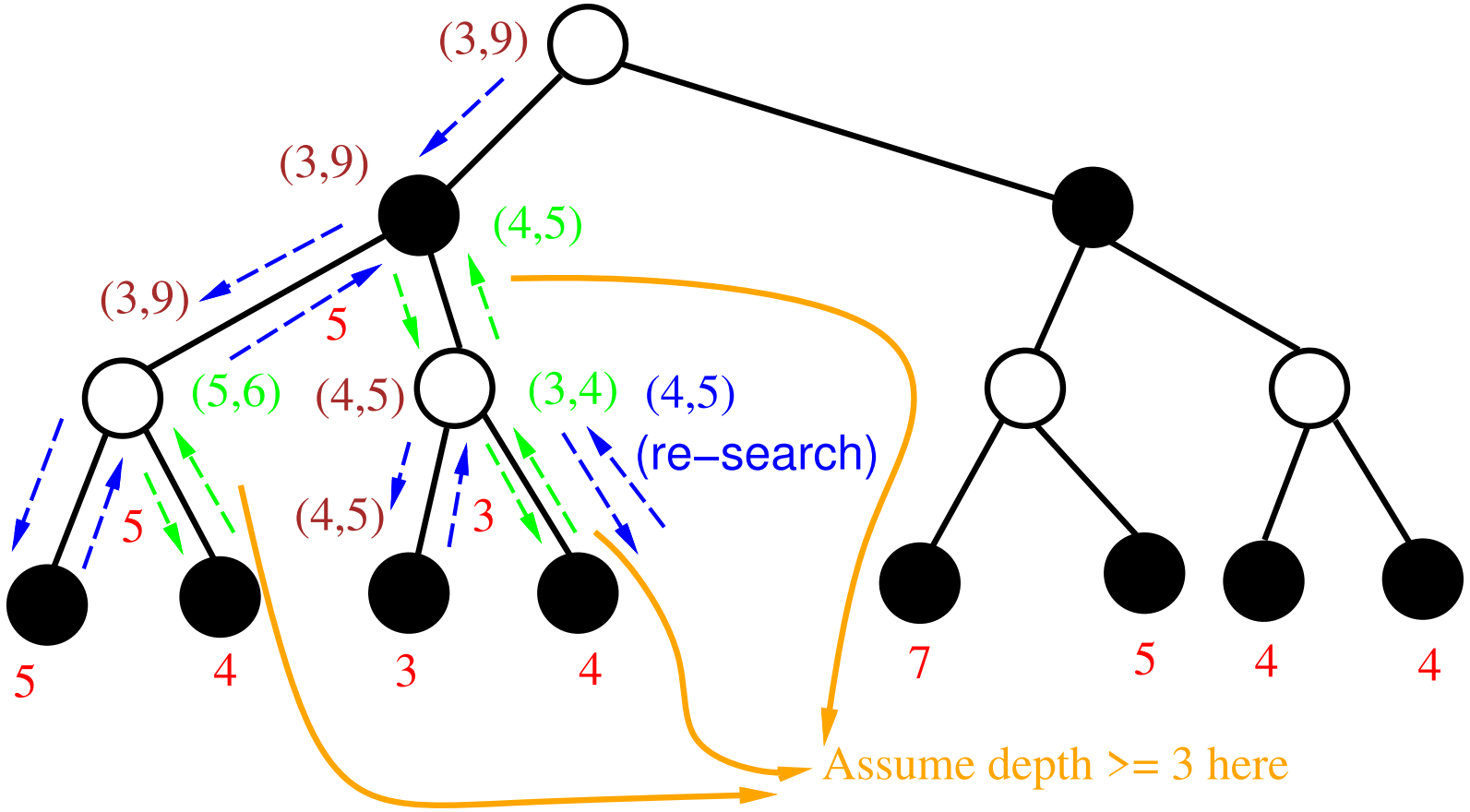
- **Algorithm** $G4'$**(position** p**, value** $alpha$**, value** $beta$**, integer** $depth$**)**
  - **determine the successor positions** $p_1, \ldots, p_b$
  - **if** $b = 0$ // **a terminal node**
    or $depth = 0$ // $depth$ **is the remaining depth to search**
    or time is running up // **from 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 ***is min possible or*** $m \leq alpha$ ***then return***(m)// ***alpha cut off***

    end
  - **return** m

NegaScout – Mini-Max version (1/2)



NegaScout – Mini-Max version (2/2)



The NegaScout Algorithm

- **Use Nega-MAX format.**
- **Algorithm** $F4$**(position** p**, value** $alpha$**, value** $beta$**, integer** $depth$**)**
  - **determine the successor positions** $p_1, \ldots, p_b$
  - **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** $h(p)$ **else**
    - $m := -\infty$ *// the current lower bound; fail soft*
    - $n := beta$ *// the current upper bound*
    - ***for*** $i := 1$ ***to*** b ***do***
    - ***9:*** $t := -F4(p_i, -n, -max\{alpha, m\}, depth - 1)$
    - ***10:*** ***if*** $t > m$ ***then***
      11: ***if*** $(n = beta$ ***or*** $depth < 3$ ***or*** $t \geq beta)$
      12: ***then*** $m := t$
      13: ***else*** $m := -F4(p_i, -beta, -t, depth - 1)$ *// re-search*
    - ***14:*** ***if*** m ***is max possible or*** $m \geq beta$ ***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.**
- **Search the first child p_1 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, equivalent to***

9: $t := -F4(p_i, -beta, -alpha, depth - 1)$

searching with the normal window $(alpha, beta)$

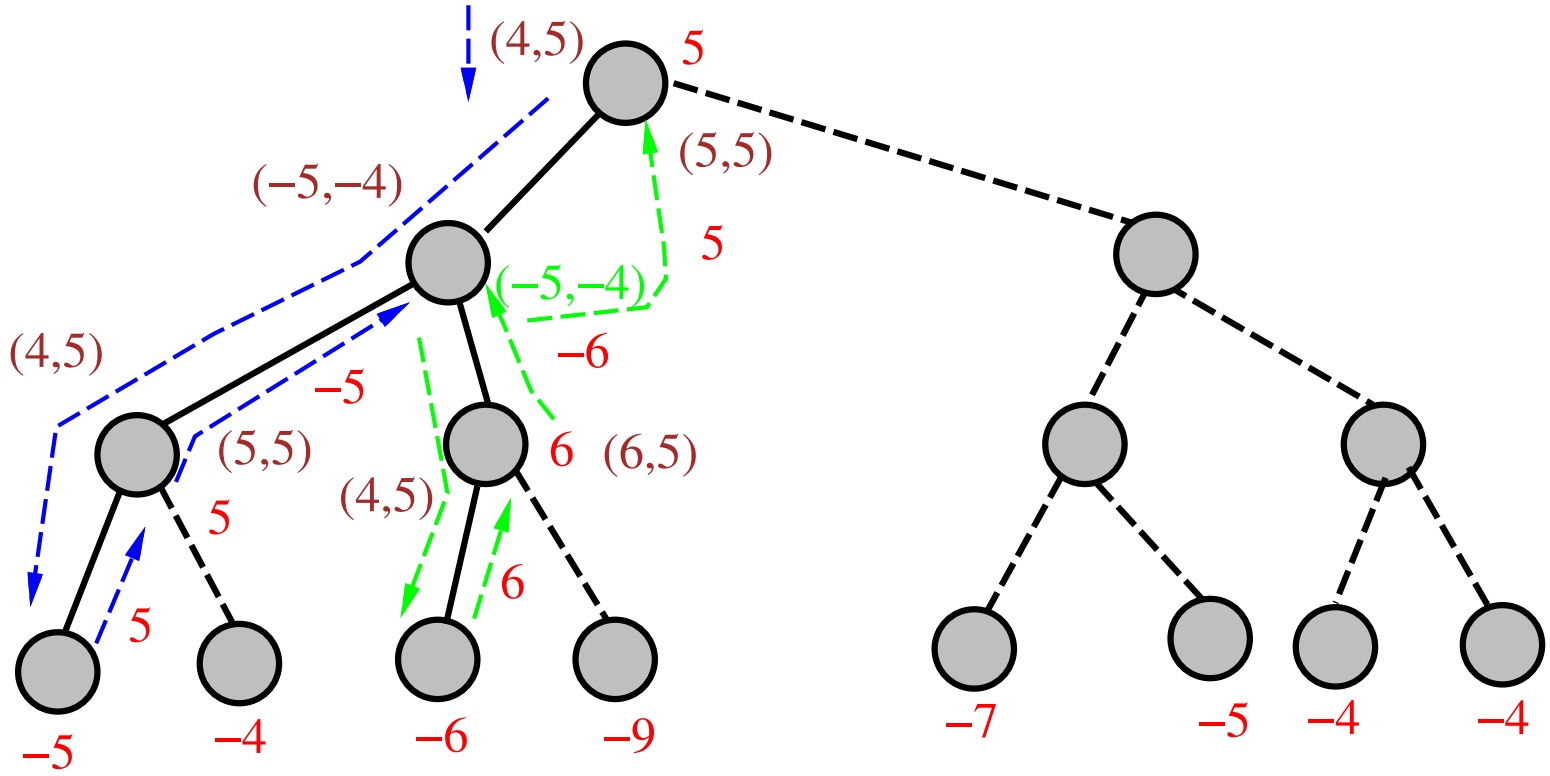
Search behaviors (2/3)

- **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***
    - ▷ ***$alpha$ is the previous lower bound***
    - ▷ ***m's value will be first set at line 12 because $n = beta$***
    - ▷ ***The value of $n = max\{alpha, m\} + 1$ is set at line 15.***
  - **line 11:**
    - ▷ ***If $n = beta$, we are at the first iteration.***
    - ▷ ***If $depth < 3$, we are on a smaller depth subtree, i.e., depth at most 2, NegaScout always returns the best value.***
    - ▷ ***If $t \geq beta$, 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 some form of a transposition table which will be introduced later.
  - Order the moves in a priority list.
  - Reduce the number of re-searching's.

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 reused.**
- **Using TEST in SCOUT to do the first search because it has a chance to visit less nodes than that of ALPHA-BETA.**

References and further readings

* J. Pearl. Asymptotic properties of minimax trees and game-searching procedures. *Artificial Intelligence*, 14(2):113–138, 1980.
* A. Reinefeld. An improvement of the scout tree search algorithm. *ICCA Journal*, 6(4):4–14, 1983.
- Noe, T. A comparison of the Alpha-Beta and SCOUT algorithms using the game of Kalah Technical Report UCLA-ENG-80-17, Cognitive Systems Laboratory, University of California, Los Angeles, 1980.
- Pearl, Judea. Heuristics: intelligent search strategies for computer problem solving. Addison-Wesley Longman Publishing Co., Inc., 1984.