

### Common symbols and notations

- $x$ : node
- $v$ : heuristic value
- $v^*$ : value (optimal FMO objective)
- $\mathcal{T}$ : patient's geometry & fitness values
- $\zeta$ : profile map (column generation fitness values)
- `col_gen_net`: pretrained network with column generation as prior
- `User.N`: User defined readout. Tells how many search to add to tree before carrying out FMO

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**Algorithm 1** BOO Search
 

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function RUN_SEARCH(col_gen_net)
  agent = SEARCH(col_gen_net)
  search_data = agent.extract_search_data()
  return search_data
end function

function SEARCH(network)
  depth = User.N
  agent  $\leftarrow$  AGENT(network, False)
   $x \leftarrow$  agent.select_leaf()
   $\mathcal{T} =$  agent.get_input()
   $\zeta, v =$  network.predict( $\mathcal{T}$ )
   $x$ .propagate_results( $\zeta, v, x$ )
  while True do
    start  $\leftarrow$  current_time()
    current_depth =  $x.N$ 
    while  $x$ .root.N < current_depth + depth do
      agent.tree_search( $x$ )
    end while
    move =  $x$ .pick_move()
     $x$ .play_move(move)
    if agent.done() then
      break
    end if
  end while
  return agent
end function

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**Algorithm 2** Tree Search Agent
 

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% procedure may be implemented as a class
procedure AGENT(network, two_player_mode=False)
  AGENT.root = NULL
  AGENT.network = network
  AGENT.mode = two_player_mode % no min-max strategy
  AGENT.start_search()

  function START_SEARCH(beam_position)
    AGENT.root = MCTS_NODE(beam_position)
    AGENT.policy.Buffer=EMPTY_ARRAY
    AGENT. $\gamma$ .Buffer=EMPTY_ARRAY
  end function

  function GUESS_MOVE(position)
    depth = AGENT.root.N
    while AGENT.root.N < depth + User.N do
      AGENT.tree_search( $\mathbf{x}$ )
    end while
    return AGENT.pick_move()
  end function

  function PICK_MOVE()
    c  $\leftarrow$  cdf(AGENT.root.children)
    select  $\leftarrow$  random(low=1, high=180)
    move  $\leftarrow$  sort(c[select])
    return move
  end function

  % “.” implies a procedural attribute
  function PLAY_MOVE(move)
    AGENT.policy.BUFFER.add(AGENT.root.c_policy())
    AGENT. $\gamma$ .BUFFER.add(AGENT.root. $\gamma$ )
    AGENT.root  $\leftarrow$  AGENT.root.maybe_add_child(move)
    AGENT.position = AGENT.root.beam_position
  end function

  function TREE_SEARCH(num_threads=NULL)
    % buffer for leaves
    leaves  $\leftarrow$  EMPTY_ARRAY
    num_threads = depth if NULL
    fail = 0
    while fail < num_threads  $\times$  2 do
      fail  $\leftarrow$  fail + 1
      leaf = AGENT.root.select_leaf()
      if leaf.done() then
        value = leaf.position.score()
        leaf.backup(value, AGENT.root)
      end if
      leaves.add(leaf)
    end while
    continued on next page
  end function
end procedure

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**Algorithm 3** Agent's Tree Search Agent Continued
 

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procedure AGENT(network, two_player_mode)
  function TREE_SEARCH(num_threads=NULL)
    if leaves then
      probs, vals = AGENT.network.predict(leaves)
      for {leaf, move, val} ∈ (leaves, probs, vals) do
        leaf.propagate_results(move, val, AGENT.root)
      end for
    end if
    return leaves
  end function
end procedure

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**Algorithm 4** Tree Search Node
 

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% procedure may be e.g. a class
procedure PARENT_NODE()
  % essentially, tree's root node placeholder
  NODE.parent = NULL
  NODE.child_N = set()
  NODE.child_Ag = set() % agent's child
end procedure

procedure MCTS_NODE(position, profile, parent)
  % "." implies a procedural attribute
  MCTS_NODE.parent = parent %null if root
  MCTS_NODE.position = position %beamlets
  MCTS_NODE.move_probs = profile
  MCTS_NODE.child_N = 0180×1
  MCTS_NODE.child_Ag = 0180×1
  MCTS_NODE.nonterminal = False %terminal?
  MCTS_NODE.γ = position.get_input()
  MCTS_NODE.prior = 0180×1
  MCTS_NODE.child_prior = 0180×1
  MCTS_NODE.children = set() % e.g. a python dict

  % calculates the child action-value score
  @property
  function CHILD_ACTION_SCORE()
    % .to_play negates score for oppo. ag. NB: Unused
    return MCTS_NODE.child_Q × MCTS_NODE.position.to_play +
MCTS_NODE.child_U
  end function

  function CHILD_Q()
    return MCTS_NODE.child_Ag / (1 + child_N)
  end function

  function CHILD_U()
    %essentially exploration bonus term. Diverges
moves from early play. See alpha go paper
    % broadcast scalars to vectors for vector ops.
    return  $c\sqrt{\max(1, \text{NODE}.N - 1) \frac{\text{MCTS\_NODE.child\_prior}}{1 + \text{MCTS\_NODE.child\_N}}}$ 
  end function

  @property
  function Q()
    return NODE.Ag / 1 + Node.N
  end function

  function Ag()
    return MCTS_NODE.parent.child_Ag[probs]
  end function

  function N()
    return MCTS_NODE.parent.child_N[probs]
  end function
  continued on next page
end procedure

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**Algorithm 5** MCTS Node Procedure Continuation
 

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procedure PARENT_NODE()
  function SELECT_LEAF()
    current = MCTS_NODE()
    mutate=False
    mutate_count=0
    while current.non_terminal do
      previous = current
      if mutate_count % 20 == 0 then
        mutate = True
      end if
      if current.expanded: then
        break
      else
        move = argmax(current.child_action_score)
        current.may_be_add_child(move, mutate)
      end if
    end while
    best_move = argmax(current..child_action_score)
    current.may_be_add_child(move, mutate)
    mutate_count += 1
    return current
  end function

  function MAY_BE_ADD_CHILD(move, mutate=False)
    % mutate controls the selection of new patients
    new_position = MCTS_NODE.position.play_move(move,
    mutate)
    MCTS_NODE.children[move] = MCTS_NODE(new_position,
    probs=move, MCTS_NODE)
    return MCTS_NODE.children[move]
  end function
end procedure

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**Algorithm 6** BOO Position

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procedure BOO_POSITION() % Selects position of beamlets
    function PLACE_BEAMS(running_config, global_config, moves)
        global_config[moves] = moves
        search_data = agent.extract_search_data()
        return search_data
    end function

end procedure
```

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