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/*
\ ^{*} For every possible move for the current board state, find the move that has
 * the highest evaluation from evaluate(). Has similar implementation to
 * maxValue(), however, it also has book keeping to record which move had the
 * highest evaluation.
 * Computational complexity analysis:
              Let moves.size() = n
              Let depth_{-} = m
              Complexity: O(n^m)
 * Explanation:
 * minimax() calls minValue which is indirectly recursive with maxValue()
 ^{st} for a total of m times. For each of these calls, moves are evaluated for
 * a total of n times. For example, if we assume that n = 3, and m = 2, then
 * we have three moves for every level of depth, for a total of 6 moves that
 * need to be processed:
 * m_0: n_0, n_1, n_2
 * m_1: n_0, n_1, n_2
 * which is 3^2 = 6, or n^m
 * Therefor, the computational complexity is O(n^m)
 * Pre-condition:
 * - The board is not full.
 * Post-condition:
 \ ^{*} - The board is left unchanged from simulations.
 * @return A valid move.
Move minimax();
Move HardComputer::minimax()
{
       map<float_t, Move> valueMoveMap;
       auto moves = gameAnalyzer ->findAllPossibleMoves(playerColor );
       for (const auto& move : moves)
       {
              const auto pipSum = gameAnalyzer_->sumPipForMove(move);
              board_->setMove(move, pipSum);
              auto value = minValue(depth_ - 1);
              valueMoveMap[value] = move;
              board_->undoMove();
       }
       return valueMoveMap.rbegin()->second;
}
```

```
float_t HardComputer::minValue(const uint32_t depth)
       // Base case: board is full or depth of search reaches 0.
      if (depth == 0 || board_->isBoardFull())
      {
             return evaluate();
      }
       // general case: there are board states to search.
      auto value = INFINITY;
      auto moves = gameAnalyzer ->findAllPossibleMoves(oppositionColor );
      for (const auto& move : moves)
             const auto pipSum = gameAnalyzer_->sumPipForMove(move);
             board_->setMove(move, pipSum);
             value = min(value, maxValue(depth - 1));
             board_->undoMove();
      }
      return value;
}
float_t HardComputer::maxValue(const uint32_t depth)
       // Base case: board is full or depth of search reaches 0.
       if (depth == 0 || board_->isBoardFull())
       {
              return evaluate();
       }
       // general case: there are board states to search.
       auto value = -INFINITY;
       auto moves = gameAnalyzer_->findAllPossibleMoves(playerColor_);
       for (const auto& move : moves)
       {
              const auto pipSum = gameAnalyzer_->sumPipForMove(move);
              board_->setMove(move, pipSum);
              value = max(value, minValue(depth - 1));
              board ->undoMove();
       }
       return value;
}
* Finds the heuristic value for the current board state. That is,
* large values for the maximizing player where that player has more
* occupied cells, and small values for the minimizing player, where
* that player has less occupied cells.
float_t HardComputer::evaluate() const
       const auto whiteCount
= static_cast<float_t>(gameAnalyzer_->countCellsWithColor(WHITE));
       const auto blackCount
= static_cast<float_t>(gameAnalyzer_->countCellsWithColor(BLACK));
       if (playerColor_ == WHITE)
       {
              return whiteCount - blackCount;
       }
       return blackCount - whiteCount;
}
```