```
// Implementation of the expectimax algorithm.
     // Description: expectimax is an adversarial search algorithm that assumes near-optimal
     play(stochastic 'mistakes' are permitted.)
 3
                     it performs this by utilizing the idea of minimax (shown here with
     //
     alpha-beta pruning heuristic -
     http://inst.eecs.berkeley.edu/~cs61b/fa14/ta-materials/apps/ab tree practice/)
                     but instead of treating the values as absolutes it will average the
     benefit of actions.
 5
                     The importance of this algorithm can be demonstrated by visualizing the
     following...
 6
                         The World is an infinite 1-Dimensional Grid
 7
     //
                         8
    //
                         The Protagonist Needs to move 3 spaces right to reach it's goal
     giving it a score of +9999
 9
     //
                         [ | | |P| | |G| | | | ]
10
    //
                         However an antagonistic agent is between...
11
    //
                         [ | | |P| |A|G| | | | ]
12
                         Each move the current score ticks down by 1, touching the
     antagonist ends the game with the current score.
13
14
     //
                         minimax would assume that A will play perfectly and P will decide
     the most optimal action is to suicide into A.
                         expectimax however doesn't assume that A is perfect and will leave
15
     things up to chance that A will move right twice
16
     //
                         giving P the chance to get a score of +9999
17
     // The Problem: Unlike minimax expectimax doesn't have heuristics that are garunteed to
18
     return the absolute best action (Alpha-Beta Pruning)
                     certain heuristics can provide psuedo-optimal play such as a "Better
19
     Than Last Time" heuristic where it is short-circuited when
20
                     an action would put it in a better state than the previous action did.
     But this ignores the chance for 'big wins' which is a
21
                     strength of expectimax. For this reason parallelization is key to
     improving the performance of expectimax.
22
23
     action* expectimax(state<action> s, int maxDepth);
24
     float expectiminValue(action* currAction, state<action> s, int depth);
25
     float expectimaxValue(action* currAction, state<action> s, int depth);
26
     float expectimaxValueParallel(action* currAction, state<action> s, int depth, int
     remainingThreads);
27
     float expectiminValueParallel(action* currAction, state<action> s, int depth, int
     remainingThreads);
28
     action* expectimaxParallel(state<action> s, int maxDepth, int numThreads);
29
30
     // Serial Implementation
31
     action* expectimax(state<action> s, int maxDepth) {
32
         // Begin with score of negative infinity
33
         float bestScore = -9999;
34
         action* bestAction = NULL;
35
         for(int i = 0; i < NUM ACTIONS; i++) {</pre>
36
             // Get Each Action
37
             action* currAction = (s.Agents[s.CurrentAgent]).Actions[i];
38
             //if(!(currAction->Possible(s))) { continue; }
39
             // Get the utility of each action.
40
            float expmaxVal = expectiminValue(currAction,s, maxDepth);
41
             // Check to see if the current action is better than previous.
42
             if(expmaxVal > bestScore){
43
                 bestScore = expmaxVal;
44
                 bestAction = currAction;
45
             }
46
47
         // return the best action to take.
48
         return bestAction;
49
     }
```

```
50
      float expectiminValue(action* currAction, state<action> s, int depth) {
 51
          // If end has been reached then evaluate the action and terminate.
 52
          if(currAction->Evaluate(s) == 9999 || depth <= 0) return currAction->Evaluate(s);
 53
          // Otherwise initialize the utility value as 0.
 54
          float expVal = 0;
 55
          // Get the next state.
 56
          state<action> sS = currAction->Perform(s);
 57
          for(int i = 0; i < NUM ACTIONS; i++) {</pre>
 58
              // Get each of the successor actions..
 59
              action* nextAction = (sS.Agents[sS.CurrentAgent]).Actions[i];
 60
              // if(!(nextAction->Possible(sS))) continue;
 61
              // get the utility of the resulting states.
 62
              expVal += expectimaxValue(nextAction,sS, depth-1);
 63
 64
          // Average the utility among possible actions.
 65
          return expVal / NUM ACTIONS;
 66
 67
      // Essentially repeats the main function call, the catch is that the score is what is
      cared about, not the action.
 68
      float expectimaxValue(action* currAction, state<action> s, int depth) {
 69
          // If end has been reached then evaluate the action and terminate.
 70
          if(currAction->Evaluate(s) == 9999 || depth <= 0) return currAction->Evaluate(s);
 71
          // Otherwise initialize the best score as negative infinity.
 72
          float bestScore = -9999;
 73
          // Get the next state.
 74
          state<action> sS = currAction->Perform(s);
 75
          for(int i = 0; i < NUM ACTIONS; i++){</pre>
 76
              // Get each of the successor actions
 77
              action* nextAction = (sS.Agents[sS.CurrentAgent]).Actions[i];
 78
              //if(!(nextAction->Possible(sS))) { continue; }
 79
              // Get the utility of each of the successor states.
 80
              float expminVal = expectiminValue(nextAction, sS, depth-1);
 81
              // Whenever a better utility is found use it.
 82
              if(expminVal > bestScore) bestScore = expminVal;
 83
 84
          // return the found utility.
 85
          return bestScore;
 86
      }
 87
 88
 89
      // Parallel Implementation
 90
      int NeededThreads(int numThreads,int partitions){
 91
          int output = (numThreads < NUM ACTIONS) ? numThreads : NUM ACTIONS;</pre>
 92
          if(output <= 0) { output = 1; }
 93
          return output;
 94
 95
     action* expectimaxParallel(state<action> s, int maxDepth, int numThreads){
 96
          action* bestAction = NULL;
 97
          float expmaxVal[NUM ACTIONS];
 98
 99
          /// First parallelizable section. This section provides the most benefit.
100
          /// A linear speedup should be possible until numThreads > NUM ACTIONS
101
          int forThreads = NeededThreads(numThreads, NUM ACTIONS);
102
          #pragma omp parallel for num threads(forThreads)
103
          for(int i = 0; i < NUM ACTIONS; i++){</pre>
104
              // Get Each Action
105
              action* currAction = (s.Agents[s.CurrentAgent]).Actions[i];
106
              // Get the utility of each action.
107
              expmaxVal[i] = expectiminValueParallel(currAction,s, maxDepth, numThreads -
              forThreads);
108
          // Begin with score of negative infinity
109
110
          float bestScore = -9999;
111
          /// Unparallelizable section is moved out. Computation of bestScore depends on
```

```
previous values.
112
          for(int i = 0; i < NUM ACTIONS; i++) {</pre>
113
              // Check to see if the current action is better than previous.
114
              if(expmaxVal[i] > bestScore){
115
                  bestScore = expmaxVal[i];
116
                  bestAction = (s.Agents[s.CurrentAgent]).Actions[i];
117
              }
118
119
          // return the best action to take.
120
          return bestAction;
121
      }
      float expectiminValueParallel(action* currAction, state<action> s, int depth, int
122
      remainingThreads) {
123
          // If end has been reached then evaluate the action and terminate.
124
          if(currAction->Evaluate(s) == 9999 || depth <= 0) return currAction->Evaluate(s);
125
          // Otherwise initialize the utility value as 0.
126
          float expVal = 0;
127
          // Get the next state.
128
          state<action> sS = currAction->Perform(s);
129
          /// Second parallelizable section. Diminshing returns should be observed based on
          tree depth.
130
          /// The explicit split is because of how many instances may be created, the
          overhead of parallel for on 1 thread was noticable.
131
          int forThreads = NeededThreads(remainingThreads, NUM ACTIONS);
132
          if(forThreads > 1) {
              #pragma omp parallel for num threads(forThreads) reduction(+ : expVal)
133
134
              for(int i = 0; i < NUM ACTIONS; i++) {</pre>
135
                  // Get each of the successor actions..
                  action* nextAction = (sS.Agents[sS.CurrentAgent]).Actions[i];
136
137
                  // get the utility of the resulting states.
138
                  expVal += expectimaxValueParallel(nextAction,sS, depth-1, remainingThreads
                  - forThreads);
139
140
          } else {
141
              for(int i = 0; i < NUM ACTIONS; i++){</pre>
142
                  // Get each of the successor actions..
143
                  action* nextAction = (sS.Agents[sS.CurrentAgent]).Actions[i];
144
                  // get the utility of the resulting states.
145
                  expVal += expectimaxValue(nextAction,sS, depth-1);
146
              }
147
148
          // Average the utility among possible actions.
149
          return expVal / NUM ACTIONS;
150
151
      // Essentially repeats the main function call, the catch is that the score is what is
      cared about, not the action.
      float expectimaxValueParallel(action* currAction, state<action> s, int depth, int
152
      remainingThreads) {
153
          float expminVal[NUM ACTIONS];
154
          // If end has been reached then evaluate the action and terminate.
          if(currAction->Evaluate(s) == 9999 || depth <= 0) return currAction->Evaluate(s);
155
156
          // Otherwise initialize the best score as negative infinity.
157
          float bestScore = -9999;
158
          // Get the next state.
159
          state<action> sS = currAction->Perform(s);
          /// Third parallelizable section. Diminishing returns should be observed based on
160
          tree depth.
          /// The explicit split is because of how many instances may be created, the
161
          overhead of parallel for on 1 thread was noticable.
162
          int forThreads = NeededThreads(remainingThreads, NUM ACTIONS);
163
          if(forThreads > 1){
164
              #pragma omp parallel for num threads(forThreads)
165
              for(int i = 0; i < NUM ACTIONS; i++){</pre>
166
                  // Get each of the successor actions
```

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167
                  action* nextAction = (sS.Agents[sS.CurrentAgent]).Actions[i];
                  //if(!(nextAction->Possible(sS))) { continue; }
168
169
                  // Get the utility of each of the successor states.
170
                  expminVal[i] = expectiminValueParallel(nextAction, sS, depth-1,
                  remainingThreads);
171
172
          } else {
173
              for(int i = 0; i < NUM ACTIONS; i++){</pre>
174
                  // Get each of the successor actions
175
                  action* nextAction = (sS.Agents[sS.CurrentAgent]).Actions[i];
176
                  //if(!(nextAction->Possible(sS))) { continue; }
177
                  // Get the utility of each of the successor states.
178
                  expminVal[i] = expectiminValue(nextAction, sS, depth-1);
179
              }
180
          }
181
182
          /// Second section which couldn't be parallelized. It is moved out.
183
          // Whenever a better utility is found use it.
184
          for(int i = 0; i < NUM ACTIONS; i++)</pre>
185
              if(expminVal[i] > bestScore) bestScore = expminVal[i];
186
          // return the found utility.
187
          return bestScore;
188
      }
```