PAPA*: Path-Aware Parallel A*

AAAI 2015 Submission X

Abstract

PAPA* is an anytime parallel heuristic search algorithm based on ARA* and PA*SE, which are in turn based on A*.

Fancy Stuff

Keys are always computed by $f(s) = g(s) + wh(s)$, and we assume all edge costs are bounded below by c_l. For most applications, we recommend using $w = \epsilon$. However, our analysis will show that using small w yields strong parallelism guarantees.

Algorithm 1 bound(s)

$g_{front} := \infty$
$s' :=$ first node in $OPEN \cup BE$
$g_{back} := g(s) + f(s') - f(s) + (2\epsilon - w - 1)c_l$
while $g_{back} < g(s) \leq g_{front}$ **do**
  $g_{front} := \min(g_{front},\ g_p(s') + \epsilon h(s', s))$
  $s' :=$ node following s' in $OPEN \cup BE$
  $g_{back} := g(s) + f(s') - f(s) + (2\epsilon - w - 1)c_l$
end while
return $\min(g_{front},\ g_{back})$

Algorithm 2 main()

$g_p(s_{start}) := g(s_{start}) := 0$
$g_p(s_{goal}) := g(s_{goal}) := \infty$
$OPEN := BE := CLOSED := \emptyset$
$FROZEN := \{s_{start}\}$
repeat
  choose $\epsilon \in [1, \infty]$ and $w \in [0, \epsilon]$
  $OPEN := OPEN \cup FROZEN$ with keys $f(s)$
  $CLOSED := FROZEN := \emptyset$
  run PAPA* on multiple threads in parallel
until path is good enough or planning time runs out

Lemma 1. *At all times, for all states s, s':*

$$g(s) + f(s') - f(s) + (2\epsilon - w - 1)c_l \leq g_p(s') + \epsilon h(s', s).$$

Algorithm 3 PAPA*

LOCK
while $g(s_{goal}) > bound(s_{goal})$ **do**
  remove an s from $OPEN$ that has the smallest $f(s)$ among all states in $OPEN$ with $g(s) \leq bound(s)$ and let $g_{bound} := bound(s)$
  if such an s does not exist **then**
    UNLOCK
    wait until $OPEN$ or BE change
    LOCK
    continue
  end if
  insert s into BE
  insert s into $CLOSED$
  UNLOCK
  $S := getSucessors(s)$
  LOCK
  for all $s' \in S$ **do**
    if s' has not been generated yet **then**
      $g_p(s') := g(s') := \infty$
    end if
    $g_p(s') = \min(g_p(s),\ g_{bound} + \epsilon c(s, s'))$
    if $g(s') > g(s) + c(s, s')$ **then**
      $g(s') = g(s) + c(s, s')$
      $bp(s') = s$
      if $s' \in CLOSED$ **then**
        insert s' in $FROZEN$
      else
        insert/update s' in $OPEN$ with key $f(s')$
      end if
    end if
  end for
  remove s from BE
end while

Proof.

$$\begin{aligned} & g(s) + f(s') - f(s) + (2\epsilon - w - 1)c_l \\ = \; & g(s') + w(h(s') - h(s)) + (2\epsilon - w - 1)c_l \\ \leq \; & g(s') + wh(s', s) + (2\epsilon - w - 1)c_l \\ \leq \; & g(s') + \epsilon h(s', s) + (w - \epsilon)c_l + (2\epsilon - w - 1)c_l \\ = \; & g(s') + \epsilon h(s', s) + (\epsilon - 1)c_l \\ \leq \; & g_p(s') + \epsilon h(s', s) \end{aligned}$$

□

Lemma 2. *$bound(s) \leq \min_{s' \in OPEN \cup BE} g_p(s') + \epsilon h(s', s)$. Furthermore, $g(s) \leq bound(s)$ iff $g(s) \leq \min_{s' \in OPEN \cup BE} g_p(s') + \epsilon h(s', s)$.*

Proof. By construction, $bound(s)$ is bounded above by $g_p(s') + \epsilon h(s', s)$ for states s' which are checked in the loop. As for the remaining states $s' \in OPEN \cup BE$, the algorithm ensures that $bound(s) \leq g(s) + f(s') - f(s) + (2\epsilon - w - 1)c_l$ for these by using a minimum representative. By Lemma 1, it follows that

$$bound(s) \leq \min_{s' \in OPEN \cup BE} g_p(s') + \epsilon h(s', s).$$

For the second part, note that the loop in $bound(s)$ terminates under only two conditions. Either $g(s) > g_{front}$, in which case we have $g(s) > g_p(s') + \epsilon h(s', s) \geq bound(s)$ for the s' which began the final iteration; or $g(s) \leq g_{back}$, in which case $g(s) \leq bound(s)$ iff $g(s) \leq g_{front}$ iff $g(s) \leq g_p(s') + \epsilon h(s', s)$ for all $s' \in OPEN \cup BE$. □

Theorem 1. *For all states s, $bound(s) \leq \epsilon g^*(s)$. Hence, upon expanding s, $g(s) \leq \epsilon g^*(s)$.*

Proof. Fix any optimal path to s, and let s' be the first node on it which is in $OPEN \cup BE$. Let s_p be the predecessor of s' on this path. By the induction hypothesis, $g(s_p) \leq \epsilon g^*(s_p)$. Therefore,

$$\begin{aligned} \epsilon g^*(s) &= \epsilon \left(g^*(s_p) + c(s_p, s') + c^*(s', s)\right) \\ &\geq g(s_p) + \epsilon c(s_p, s') + \epsilon h(s', s) \\ &\geq g_p(s') + \epsilon h(s', s) \end{aligned}$$

Therefore by Lemma 2,

$$bound(s) \leq \min_{s' \in OPEN \cup BE} g_p(s') + \epsilon h(s', s) \leq \epsilon g^*(s).$$

□

TODO: clear up initial case where $s = s' = s_{start}$.

Theorem 2. *If $w \leq 1$, the parallel depth of checkless PAPA* is bounded above by*

$$\min\left(\frac{\epsilon g^*(s_{goal})}{(1-w)c_l}, \frac{\left(\epsilon g^*(s_{goal})\right)^2}{(4\epsilon - 2w - 2)c_l^2}\right).$$

Proof. We prove the two bounds separately. For the first, note that if the lowest f-value is f_{min}, every state with f-value up to $f_{min} + (2\epsilon - w - 1)c_l$ can simultaneously be expanded. Since h is consistent, the successors' f-values is at least $f_{min} + (1 - w)c_l$. Therefore, the depth is at most

$$\frac{\epsilon g^*(s_{goal})}{(1-w)c_l}$$

For the other bound, notice that since f-values never decrease along paths, once the minimum f-value in $OPEN$ surpasses f_{min}, from then on all nodes with f-value up to $f_{min} + (2\epsilon - w - 1)c_l$ are always safe to expand. And during each iteration of the simultaneous expansions, the g-value of all such nodes increases by at least c_l. Since g cannot exceed f, this continues for at most $(f_{min} + (2\epsilon - w - 1)c_l)/c_l = f_{min}/c_l + 2\epsilon - w - 1$ iterations, after which every node in $OPEN$ has f-value $>= f_{min} + (2\epsilon - w - 1)c_l$. Continuing this process until f_{min} exceeds $\epsilon g^*(s_{goal})$, a bound on the total iterations is:

$2\epsilon - w - 1 + 2(2\epsilon - w - 1) + 3(2\epsilon - w - 1) + ... + \epsilon g^*(s_{goal})/c_l \approxeq \epsilon g^*(s_{goal})/c_l)^2/(4\epsilon - 2)$. iterations of parallel expansion is enough to find the optimal path. □

Mods

Let $k(s)$ be the least number of edges used in a minimum-cost path to s and fix $\delta > 0$. If g_{front} and g_{back} are each increased by 2δ, then by similar arguments to the proofs earlier in the paper, we find that, upon expanding s, $g(s) \leq \epsilon g^*(s) + \delta k(s)$.

Here's an extension inspired by (Klein and Subramanian 1997): suppose the mean edge cost c_m along the optimal path is known to be much greater than the lower bound c_l. In such a case, the bound in Theorem 2 scales poorly. To remedy the situation, we "grow" the small edges, effectively running PAPA* with $c'_l = c_l + \delta$ and $c'(s, s') = \max(c(s, s'), c'_l)$.

Theorem 3. *If the mean cost of the edges along the minimum-cost path to s is at least c_m, then upon expansion, $g(s) \leq \epsilon(1 + \delta/c_m)g^*(s)$. Therefore, to get the same optimality factor as ϵ, we can set $\delta = (\epsilon - 1)c_m$.*

Proof. We assumed $c_m \leq g^*(s)/k(s)$, so $k(s) \leq g^*(s)/c_m$. It follows from Lemma 1 that $g'(s) \leq \epsilon g'^*(s) \leq \epsilon(g^*(s) + \delta k(s)) \leq \epsilon(1 + \delta/c_m)g^*(s)$. □

Corollary 1. *If $w \leq 1$, the parallel depth of checkless PAPA* can be improved to*

$$\frac{\epsilon g^*(s_{goal})}{(1-w)(c_l + (\epsilon - 1)c_m)}.$$

References

Klein, P. N., and Subramanian, S. 1997. A randomized parallel algorithm for single-source shortest paths. *Journal of Algorithms* 25(2):205–220.