

PAPA*: Path-Aware Parallel A*

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Abstract

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Fancy Stuff

Hello world.

Algorithm 1 $bound(s)$

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

Throughout this paper, we assume all edge costs are bounded below by c_l , that $\epsilon \geq 1$, $\delta \geq 0$ and $0 \leq w \leq \epsilon$. For most applications, we recommend using $\delta = 0$ and $w = \epsilon$. However, as we will see, it may pay to use $w < \epsilon$ when there are a lot of processors available, or $\delta > 0$ when the mean edge cost of paths is known to be much higher than the lower bound c_l .

Definition 1 (TODO: these definitions are never used, so we don't need them). We say a state s is independent of s' if $g(s) \leq g_p(s') + \epsilon h(s', s) + 2\delta$. We say $s \in OPEN$ is safe to expand if s is independent of all $s' \in OPEN \cup BE$.

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 2 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
       $f(s') := g(s') := g_p(s') := \infty$ 
    end if
    if  $s' \notin CLOSED$  then
       $g_p(s') = \min(g_p(s), g_{bound} + \epsilon c(s, s'))$ 
      if  $g(s') > g(s) + c(s, s')$  then
         $g(s') = g(s) + c(s, s')$ 
         $f(s') = g(s') + wh(s')$ 
         $bp(s') = s$ 
        insert/update  $s'$  in(to)  $OPEN$  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. *For all states s , $\text{bound}(s) \leq \epsilon g^*(s) + k(s)\delta$ where $k(s)$ is the least number of edges used in a minimum-cost path to s . Hence, upon expanding s , $g(s) \leq \epsilon g^*(s) + k(s)\delta$.*

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

$$\text{bound}(s) \leq \min_{s' \in \text{OPEN} \cup \text{BE}} g_p(s') + \epsilon h(s', s) + 2\delta.$$

We complete the proof by showing that the latter expression is bounded above by $\epsilon g^*(s) + k(s)\delta$. Fix any optimal path to s , and let s' be the first node on it which is in $\text{OPEN} \cup \text{BE}$. Let s_p be the predecessor of s' on this path. By the induction hypothesis, $g(s_p) \leq \epsilon g^*(s_p) + k(s_p)\delta$. Therefore,

$$\begin{aligned}
\epsilon g^*(s) &= \epsilon (g^*(s_p) + c(s_p, s') + c^*(s', s)) \\
&\geq g(s_p) - k(s_p)\delta + \epsilon c(s_p, s') + \epsilon h(s', s) \\
&\geq g_p(s') - k(s_p)\delta + \epsilon h(s', s) \\
&\geq g_p(s') + \epsilon h(s', s) + 2\delta - k(s)\delta
\end{aligned}$$

□

TODO: clear up initial case where $s = s' = s_{\text{start}}$ and mention that use of δ was inspired by (Klein and Subramanian 1997).

Corollary 1. *If δ is not more than m times the mean cost of the edges along the minimum-cost path to s , then upon expansion, $g(s) \leq (\epsilon + m)g^*(s)$.*

Proof. We assumed $\delta \leq mg^*(s)/k(s)$. It follows from Lemma 2 that $g(s) \leq \epsilon g^*(s) + k(s)\delta \leq (\epsilon + m)g^*(s)$. □

References

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