

Today

- Link-cut Paths
- Extend to Link-cut Trees
- Running Time Analysis

Application

Finding a blocking flow in capacitated graphs



advance  
retreat  
augment

- Two tasks:
- ① Don't waste work
  - ② Query and update capacities quickly

Link-cut Paths

Assume degree 1

⇒ concatenate and split paths

draw here

Link( $v, w$ ): make  $w$  parent of  $v$   
 $v$  must be a root  
 $w$  must be a leaf

cut( $v$ ): cut  $v$  off from its parent

find-root( $v$ ): returns root of  $v$ 's path

[We will delay discussion of storing capacities until later]

Goal:  $O(\lg n)$  time amort. per op.

[Asks what's a good  $O(\lg n)$  amort. per op. and can store paths?]  
data structure that has

Use splay trees!

Order nodes by depth

link( $v, w$ ):

splay( $v$ )  
splay( $w$ )  
 $\text{left}(v)=w$



[what  $v$ 's splay tree looks like]

cut( $v$ ):

splay( $v$ )  
 $\text{left}(v)=\text{null}$



find-root : ≡ find-min

splay( $v$ )  
walk left to min  
splay the min

Supporting Argument

represent capacities at a node instead of an edge  
each node only has at most 1 parent in the path

Let  $u(v) = u(v, w)$  where  $w$  is  $v$ 's parent in the path

find-min( $v$ ): returns min capacity node in  $v$ 's path

update( $v, x$ ): adds real number  $x$  to all capacities in  $v$ 's path

~~Store capacities~~

~~for each edge in path~~

~~store capacity of edge~~

~~as well as parent information~~

~~so we can find min~~

~~and update~~

~~in O(1) time~~

[Want to take advantage of BST structure.]

~~But this is too costly to do~~

~~because we have to store parents~~

~~and edges~~

~~and capacities~~

~~for each node~~

~~in the path~~

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[Intuition: if splaying something, you can walk down from the root to it and calculate its capacity just by adding up the  $\Delta u$ 's on the nodes in that path]

on root-to- $v$  path in splay tree,

$$u(v) = \sum_{w \in \text{path}} \Delta u(w)$$

store min values as well

Let  $\min(v)$  = value of mincapacity node in subtree rooted at node  $v$

- also too costly by itself

$$\text{Store } \Delta \min(v) = u(v) - \min(v) \text{ instead}$$

$$\geq 0 \forall v$$

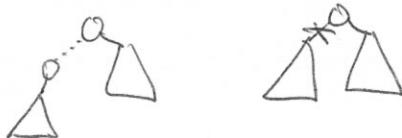
$\Rightarrow$  can calculate  $\min(v)$  with  $\Delta \min(v)$  and  $u(v)$   
stored at node calculated from traversal

$\Delta$  values can be maintained during splays



do simple arithmetic using known values at involved nodes  
 $O(1)$  per splay

also true for link, cut



find-min( $v$ ):

splay( $v$ )  
 $\min(v) = u(v) - \Delta \min(v)$

while  $u(v) \neq \min(v)$ :

$v \leftarrow$  child  $w$  of  $v$  where  $\min(w) = \min(v)$

return  $v$   
(splay  $v$  is optional!)

update( $v, x$ ):

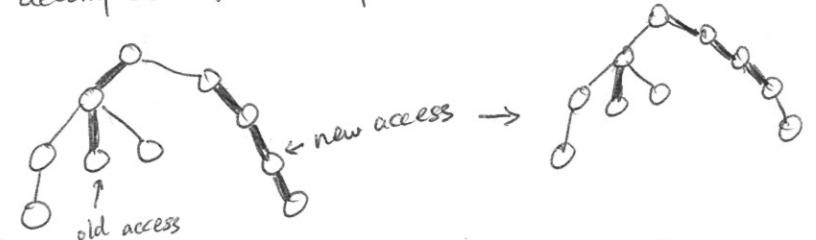
Splay( $v$ )  
 $\Delta u(v) + = x$

[So to support augment, we call find-min to find the min, update using the value, and then call find-min and cut for each node that now has capacity  $\emptyset$ .]

## Link-cut Trees

reuse ideas

decompose a tree into paths

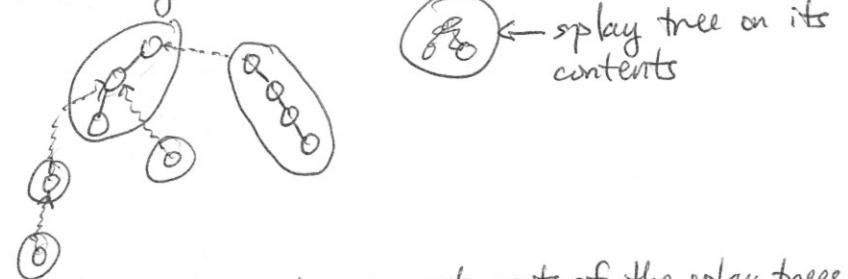


"preferred path": paths connected by preferred edges

a node has  $\leq 1$  preferred child

still store paths as splay trees; must maintain structure of tree

Auxiliary tree:



path-parent pointers connect roots of the splay trees to the actual parent of the path represented

"super splay" needed to bring a node into the root splay tree for link, cut, etc. ops.

[see example at top of page]

preferred path from root always ends at last item accessed

## Access (v)

splay(v) within its splay tree  
 cut off right subtree of v  
 and give it a path-parent ptr to v  
 loop until we get to the root:

w ← path-parent(v)

splay(w)

cut off w's right  
 subtree and have it point to w

right(w) = v

path-parent(v) = null

v = w

splay v

v now root of the aux. tree's root splay tree

link(v, w): v must still be a root of the actual tree  
 w can be anything

v in root splay tree by itself

access(v)

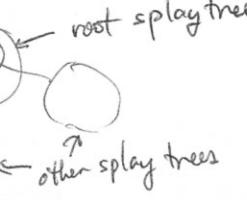
access(w)

left(v) = w

cut(v):

access(v)

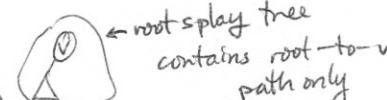
left(v) = null



find-min(v): finds <sup>root</sup> on root-to-v path

access(v)

do find-min in v's splay tree



find-min similar  
 and update

Δ values can still be updated in O(1) more time per op



ptr to v



## Running Time

Page 3

All ops have O(1) calls to access and O(1) calls to splay (outside access)

O(access) + O(splay)

↑

O(lgn)

(# splay trees or  
 path changes to the root) × (splay)

Technique: Heavy-light decomposition

useful for trees of any degree

Let size(v) = # nodes in v's subtree

Call the edge between v and parent(v) heavy  
 if  $\text{size}(v) > \frac{1}{2} \text{size}(\text{parent}(v))$

and light otherwise.

Notes:

≤ 1 heavy child per node

# light edges on root-to-v path [Ask here]

≤ lgn

Bound # preferred child changes

O(lgn) light edges become preferred

heavy edges harder [Ask for ideas]

Amortized analysis:

sequence of m operations ↓

after all heavy edges become preferred,

a heavy edge only gets preferred again if it is unpreferred first

O(lgn) heavy edges become unpreferred

per access (light edges would have to get preferred)

at most n-1 heavy edges (chain)

⇒ O(m lgn) + n-1 heavy edges preferred over m accesses

for m big enough, get O(lgn)

⇒ O(lgn) bound for access [But I promised you O(lgn)!]

## More Amortization

Page 4

Intuition: all those splays in the loop are not that bad

~~whole auxiliary tree~~ is like one big compartmentalized splay tree, and accessing a node costs the same

$\text{size}(v)$  = # nodes under  $v$  in auxiliary tree

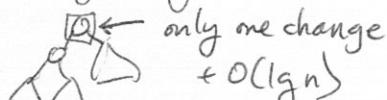
$$\Phi = \sum_v \lg \text{size}(v)$$

Each splay only affects  $v$ 's splay tree; other parts of aux. tree is untouched.

Intermediary splay costs  $O(\lg \text{size}(r) - \lg \text{size}(v))$

Total cost is a telescoping sum, and we get

$$O(\lg n - \lg \text{size}(v)) = O(\lg n)$$

link:  +  $O(\lg n)$

cut: only decreases potential

$\Rightarrow O(\lg n)$  per link-cut operation

$\Rightarrow O(m \lg n)$  per blocking flow

$\Rightarrow O(mn \lg n)$  for max flow