

DYNAMIC PLANAR *POINT LOCATION*
WITH
SUB-LOGARITHMIC *LOCAL UPDATES*

Maarten Löffler

Utrecht University

Joe Simons

University of California, Irvine

Darren Strash

Intel Oregon

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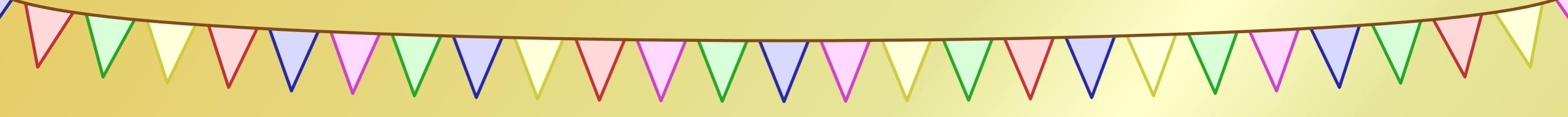
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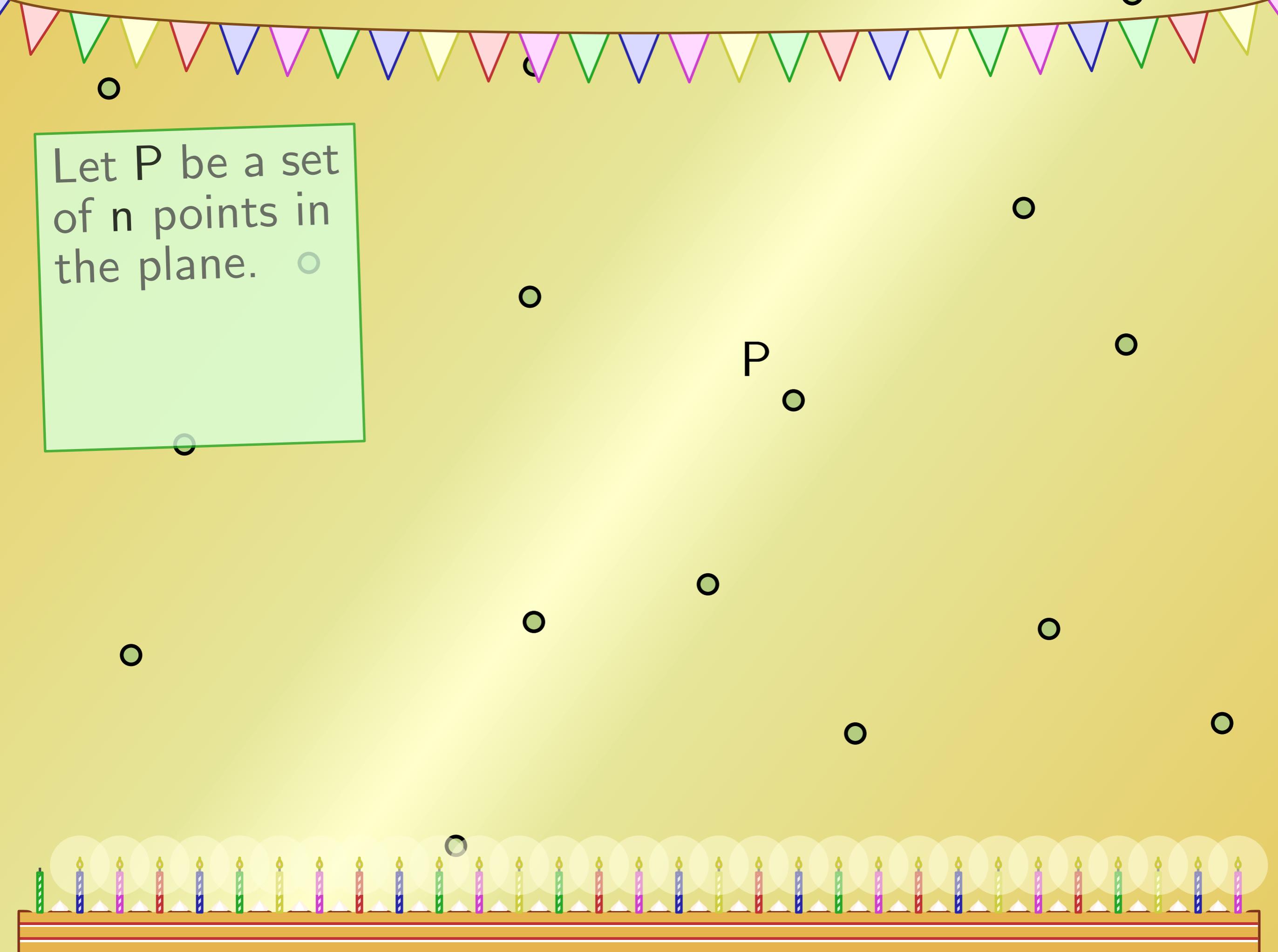


Let P be a set
of n points in
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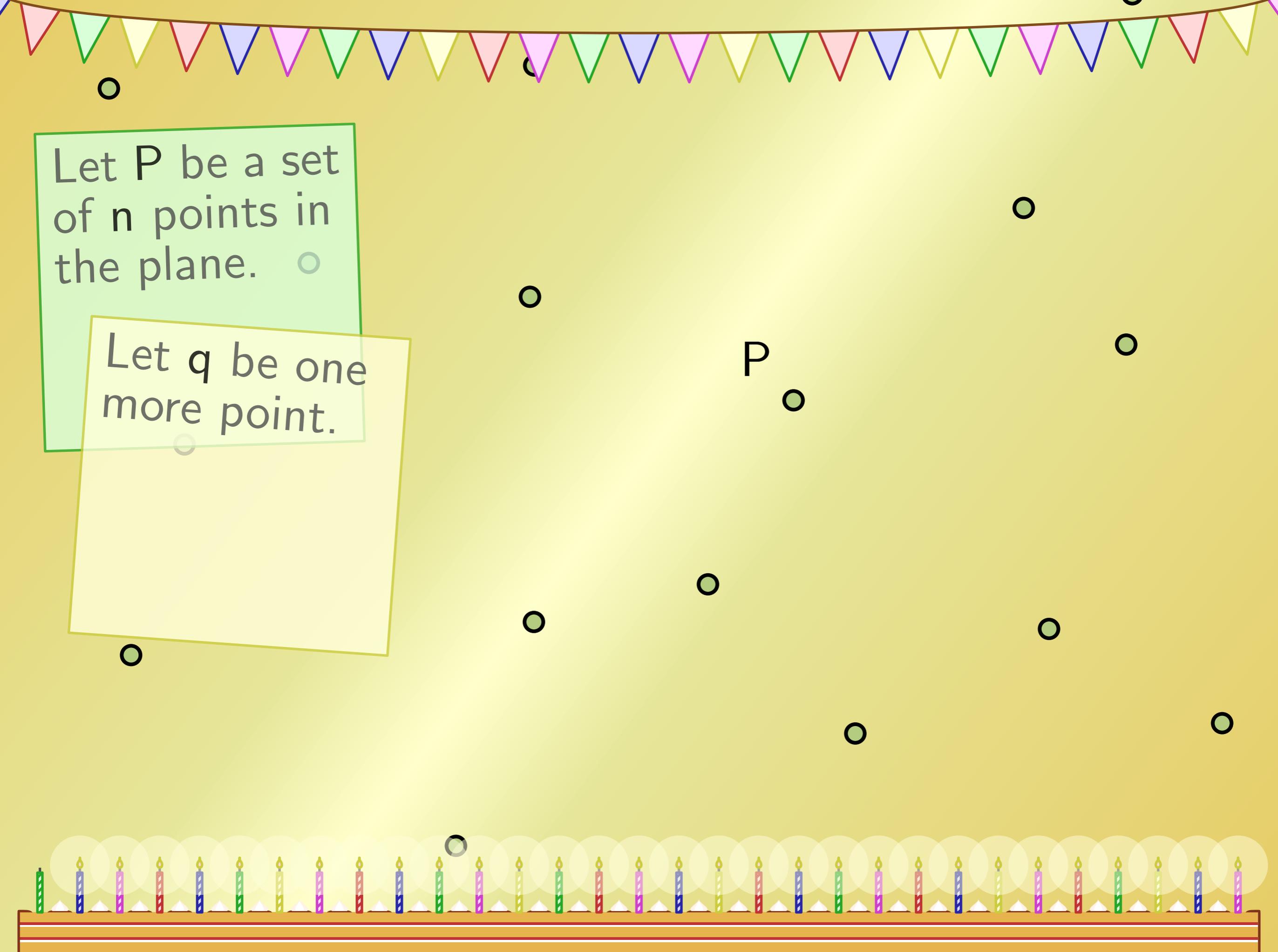
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Let P be a set
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Let q be one
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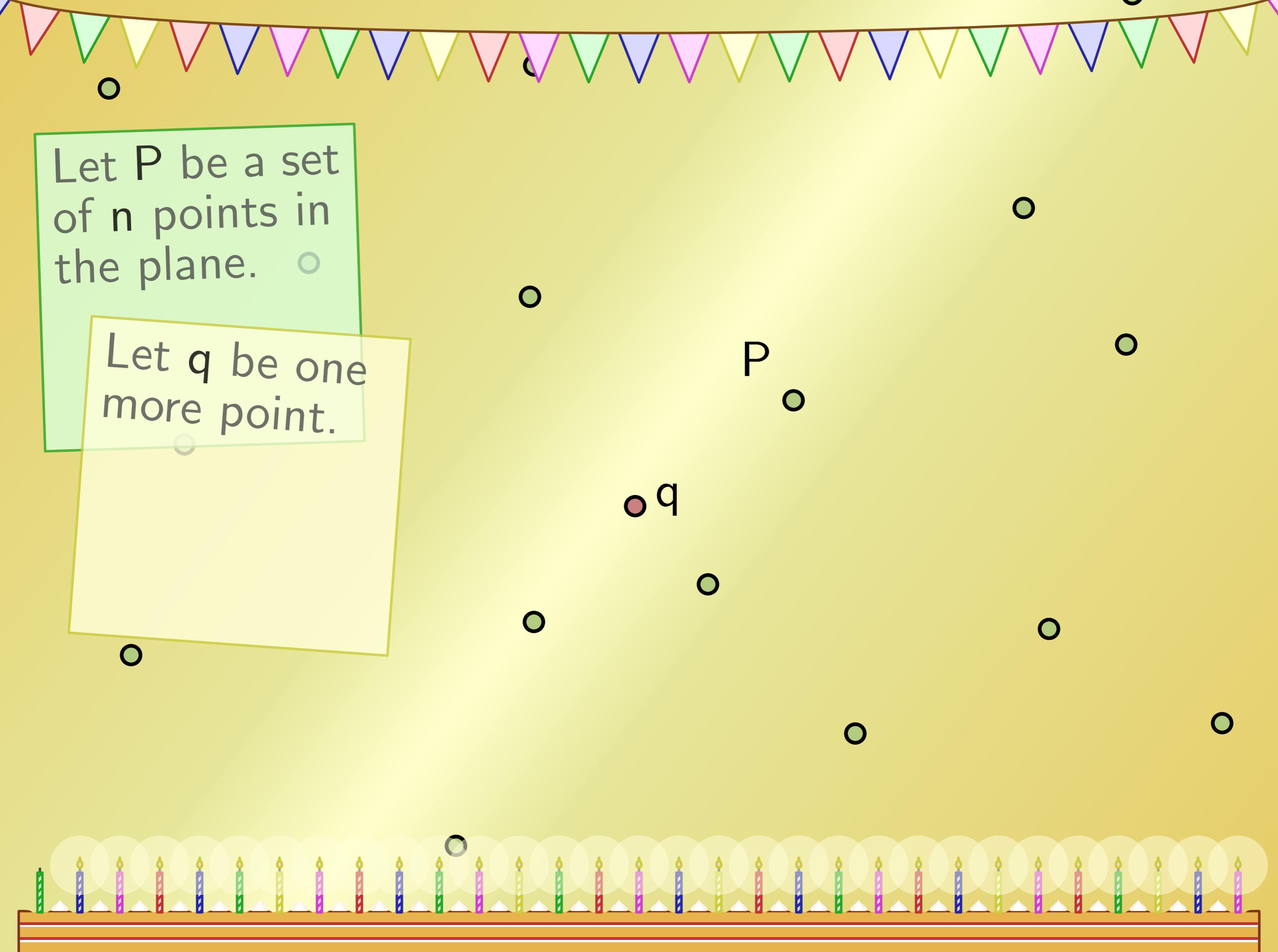


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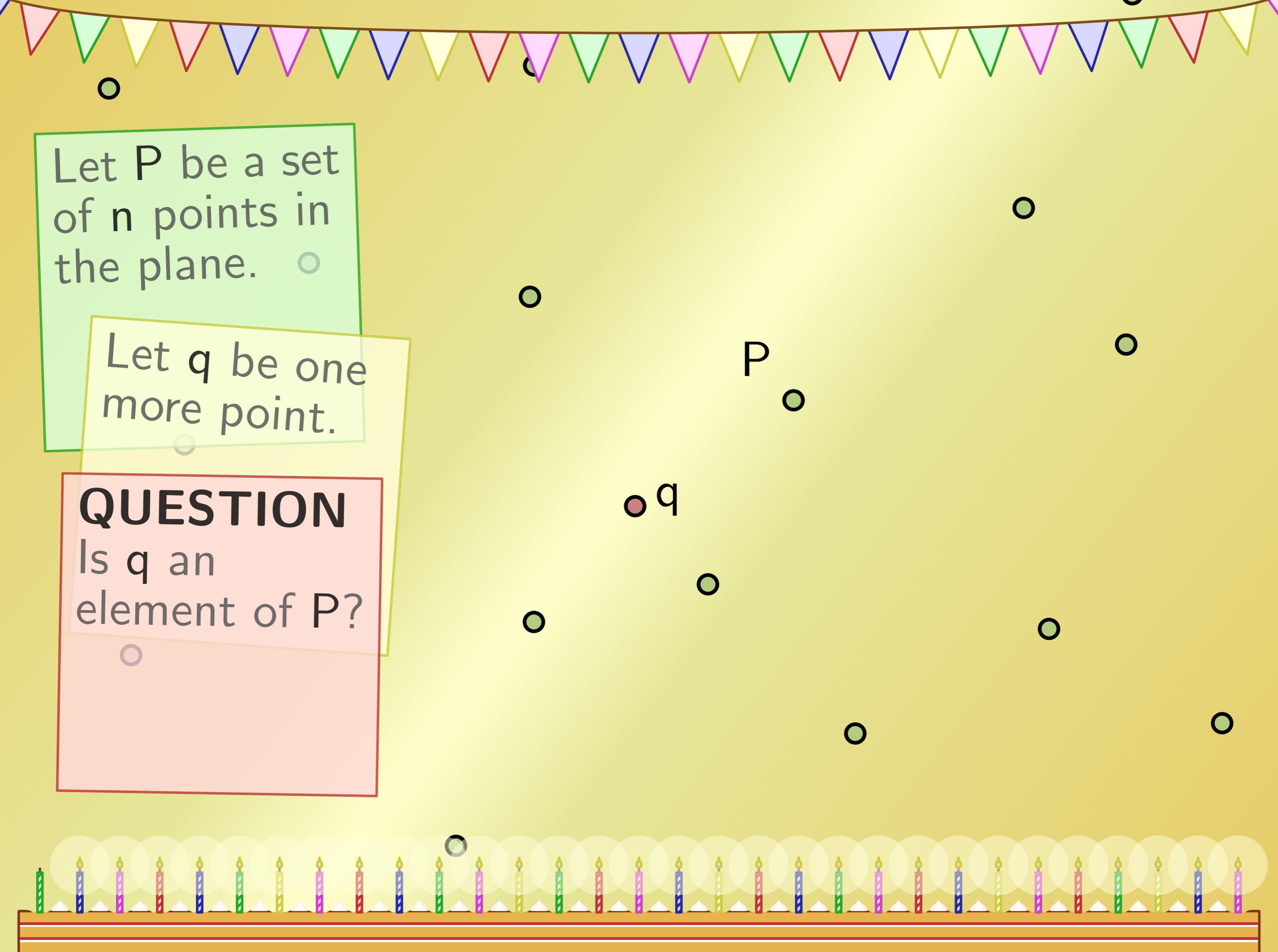
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QUESTION

Is q an element of P ?

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Let P be a set of n points in the plane.

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Two possible answers: *yes* or *no*.



P



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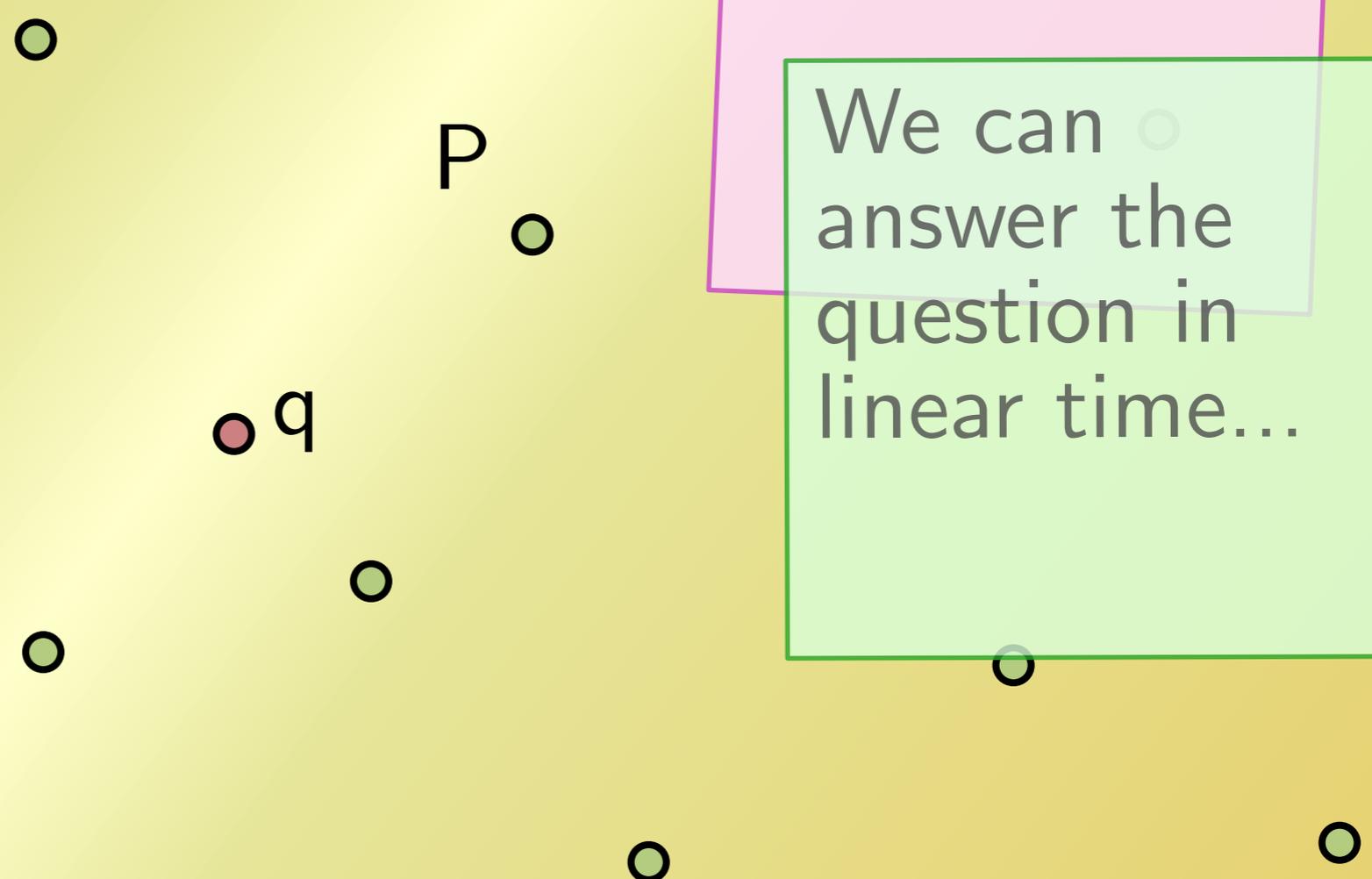
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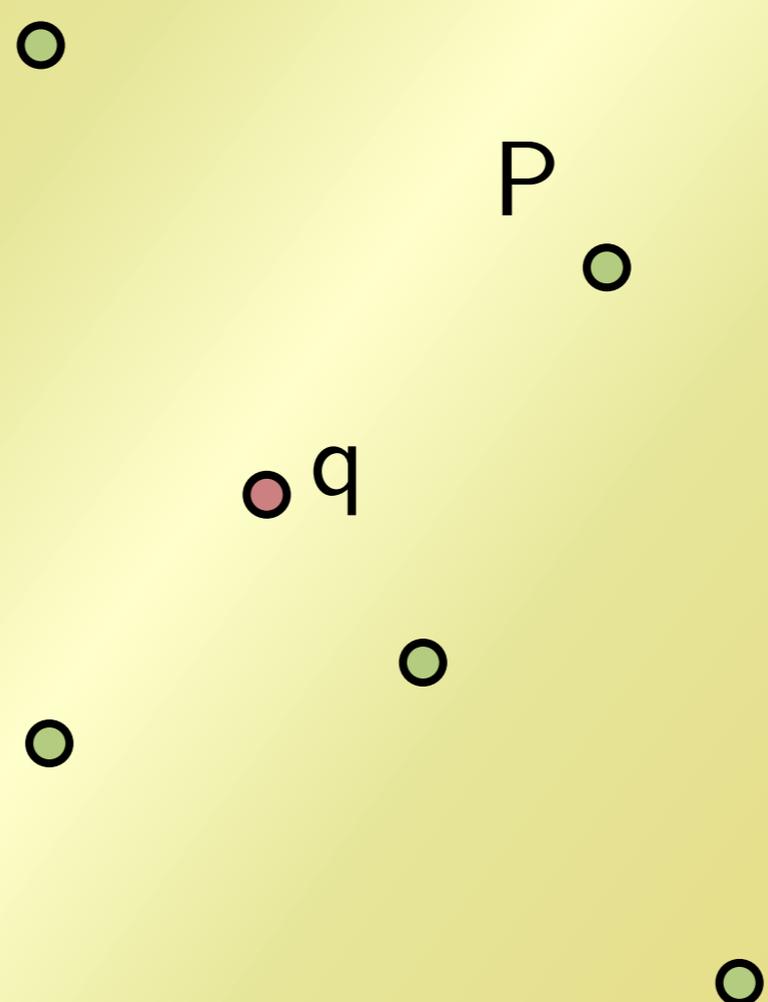
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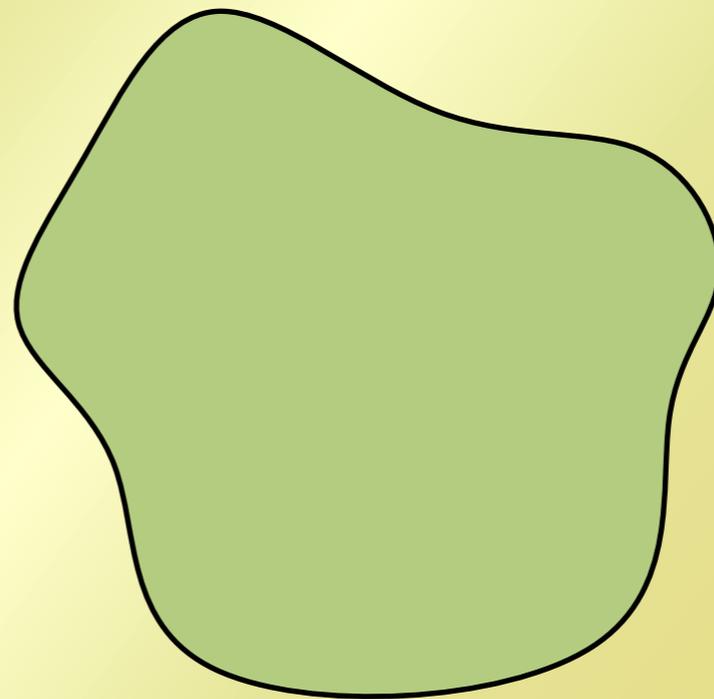
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That is, each
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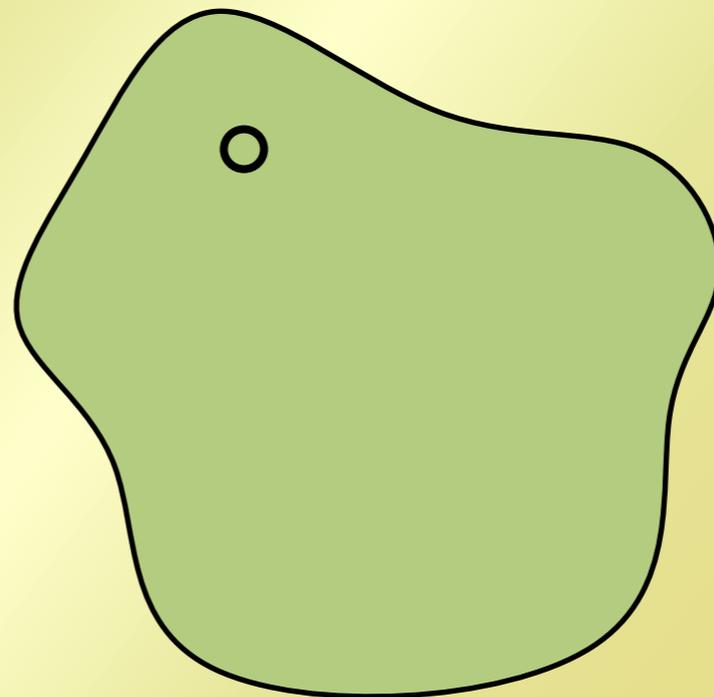
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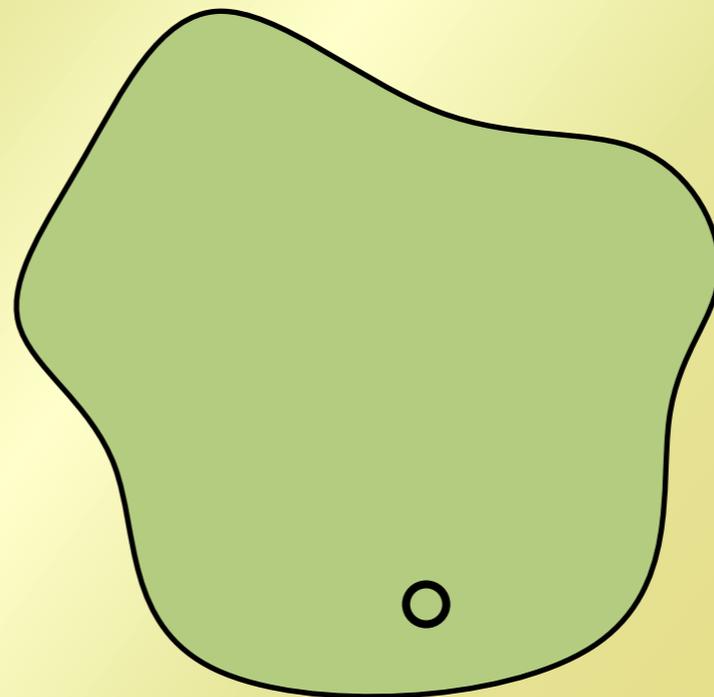
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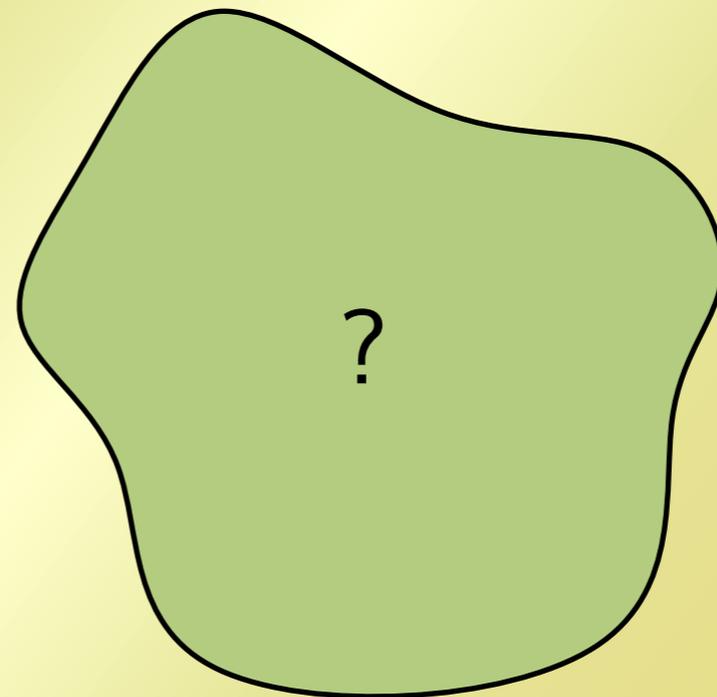
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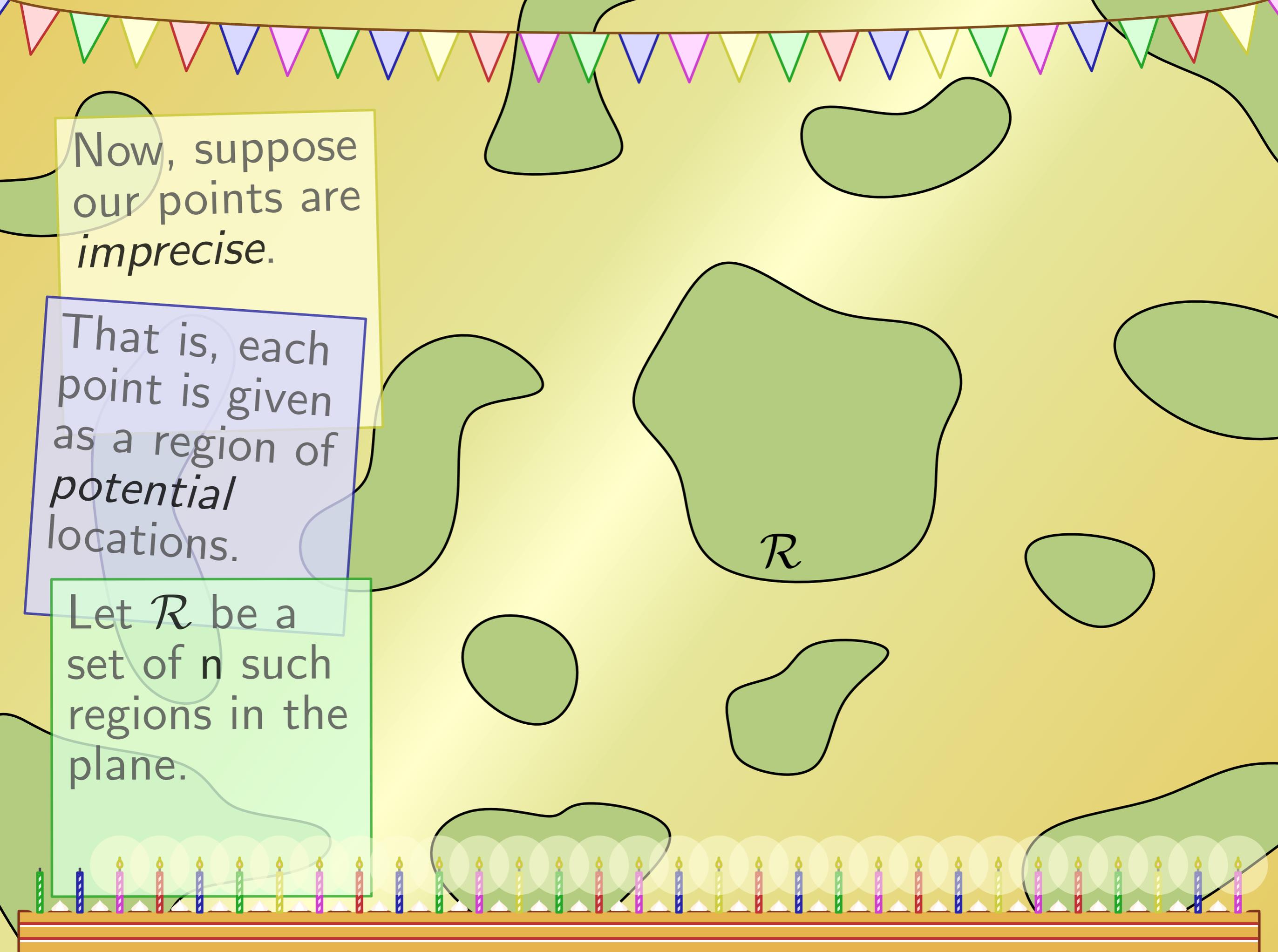
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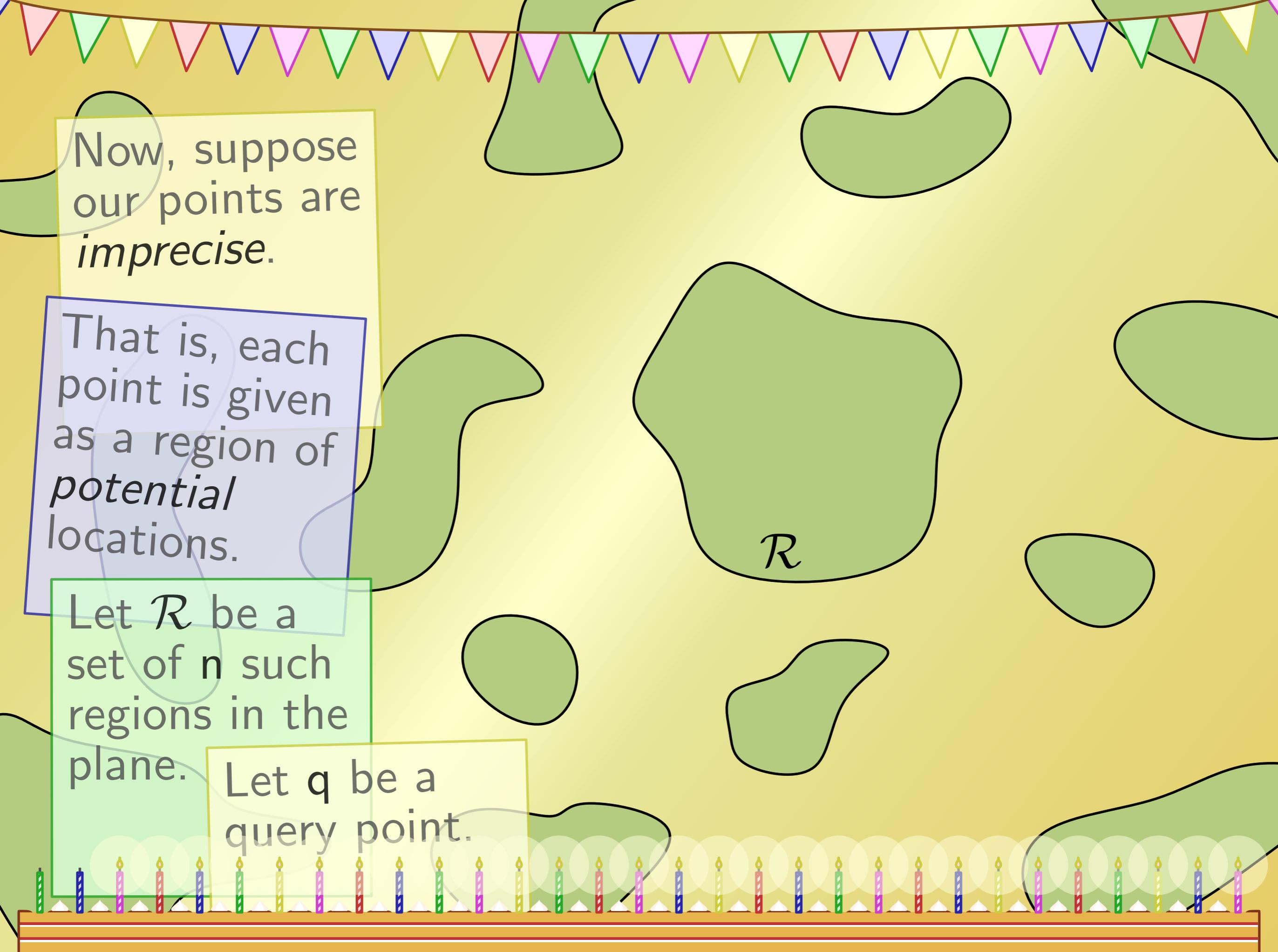


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Let \mathcal{R} be a
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\mathcal{R}



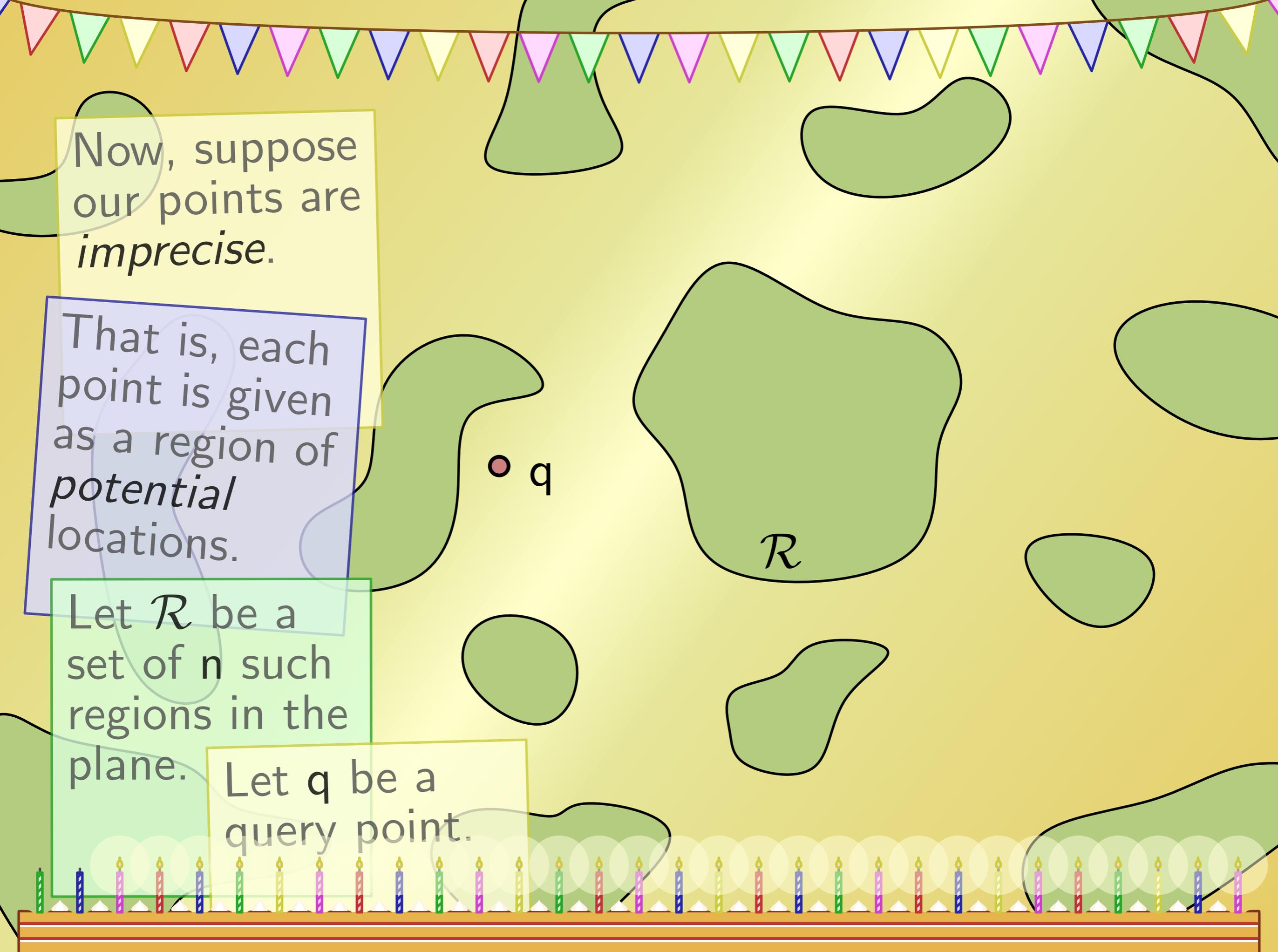
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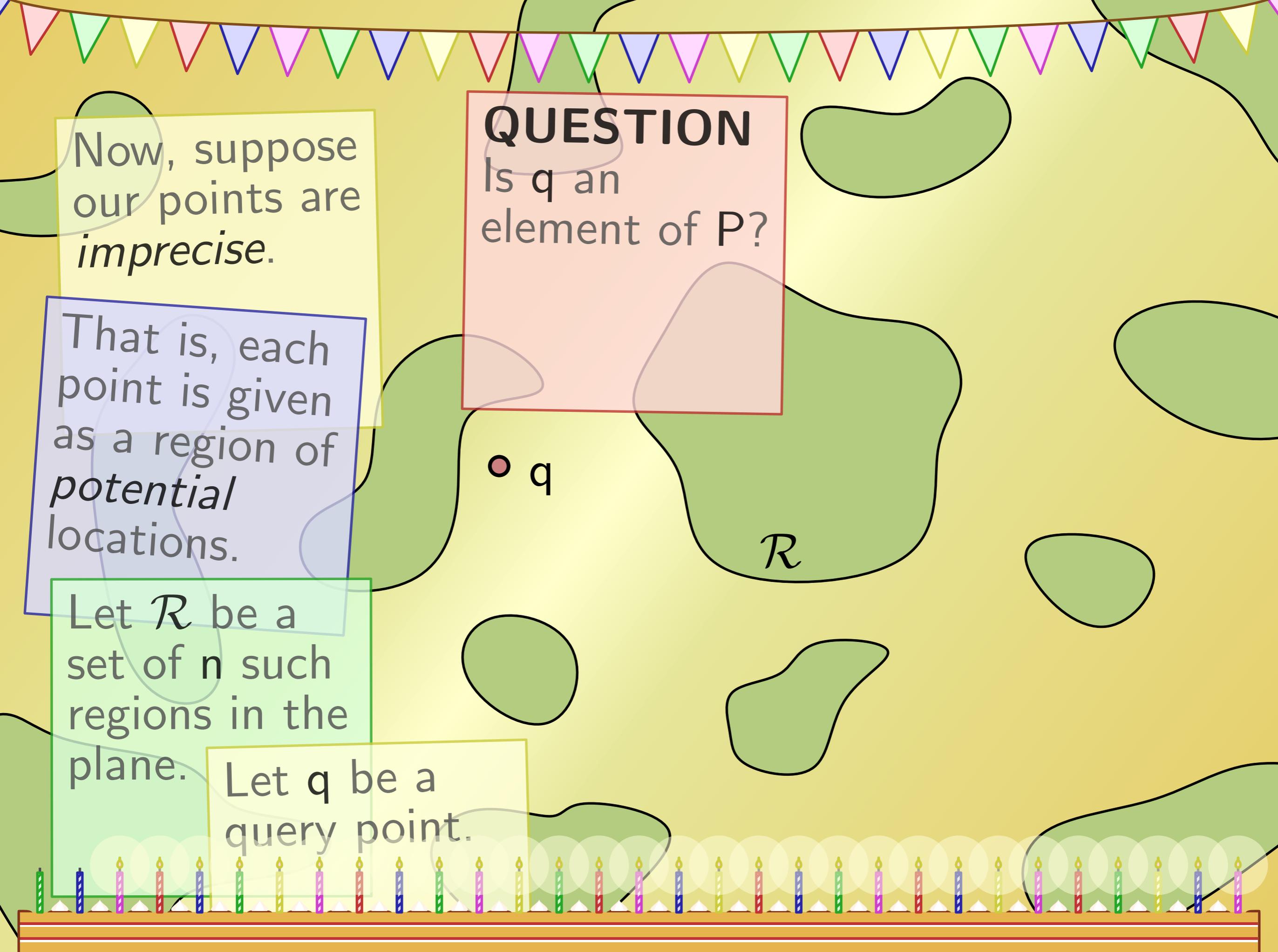
That is, each point is given as a region of *potential* locations.

Let \mathcal{R} be a set of n such regions in the plane.

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q

\mathcal{R}



Now, suppose our points are *imprecise*.

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Two possible answers:
maybe or *no*.

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Again, we can answer the question in logarithmic time after preprocessing





Suppose
furthermore
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Our estimate
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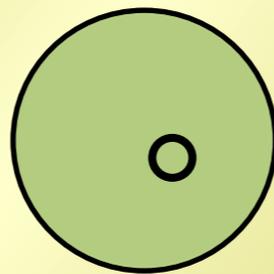
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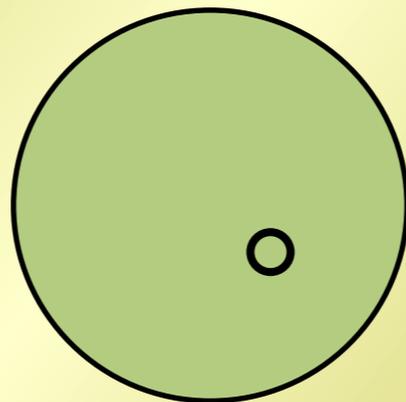
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Suppose
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Our estimate
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... or the true
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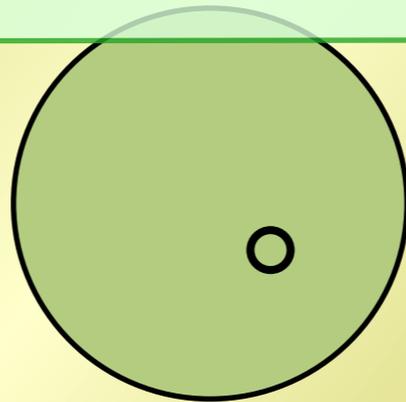


Suppose furthermore that our points are *dynamic*.

Let \mathcal{R} be a set of n dynamic regions in the plane.

Our estimate of a point's location may change...

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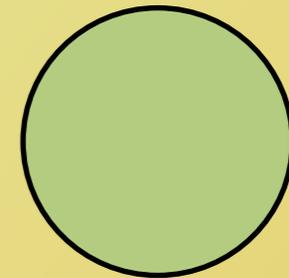
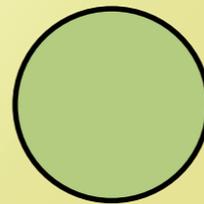
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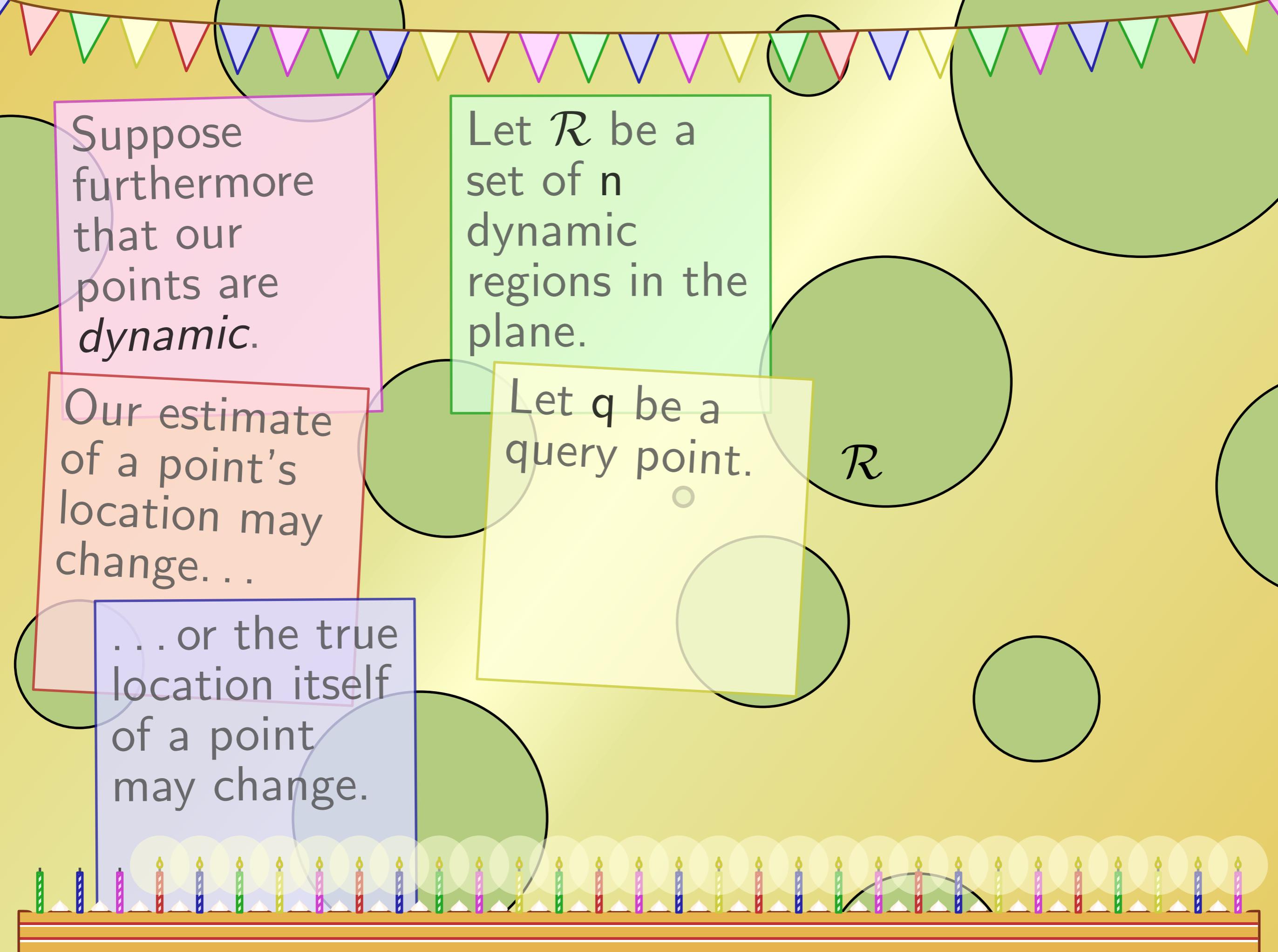
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Suppose furthermore that our points are *dynamic*.

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QUESTION

Is q an element of P ?

... or the true location itself of a point may change.

\bullet q

\circ

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Suppose furthermore that our points are *dynamic*.

Let \mathcal{R} be a set of n dynamic regions in the plane.

We can still answer the question in logarithmic time after preprocessing

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Suppose furthermore that our points are *dynamic*.

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But now we also need to respond to changes in \mathcal{R}

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QUESTION
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We want to also handle *updates* efficiently.

... or the true location itself of a point may change.

\bullet q

\mathcal{R}

\circ





What is known about dynamic planar point location?



$O(\log^2 n)$
queries with
 $O(\log n)$
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[Cheng & Janardan, 1992]

What is
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[Arge *et al.*, 2006]

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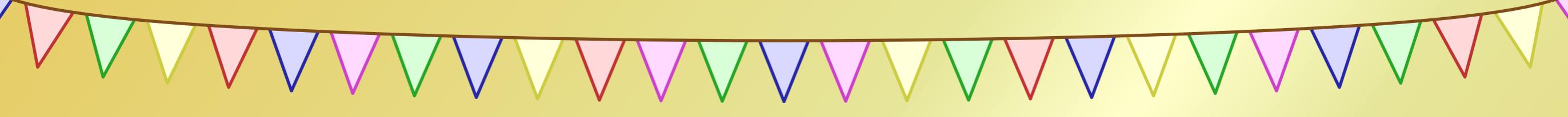
[Cheng & Janardan, 1992]

What is
known about
dynamic
planar point
location?

... or in
rectilinear
subdivisions.

[Blelloch, 2008]
[Giora & Kaplan, 2009]





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QUESTION
Is it possible
to break the
 $\log n$ barrier
in this case?





PROBLEM STATEMENT & RESULTS





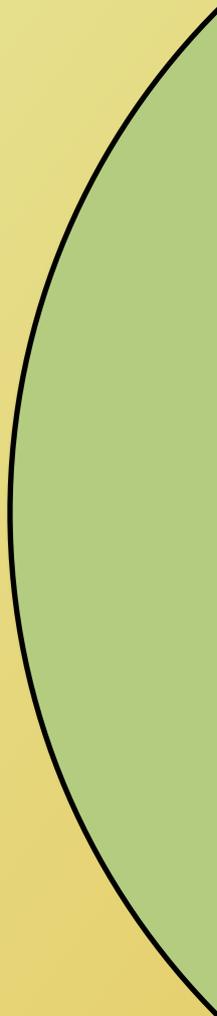
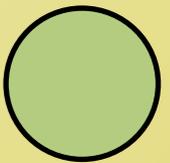
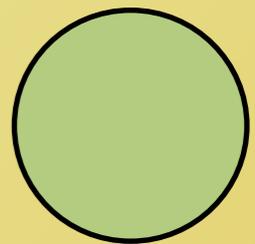
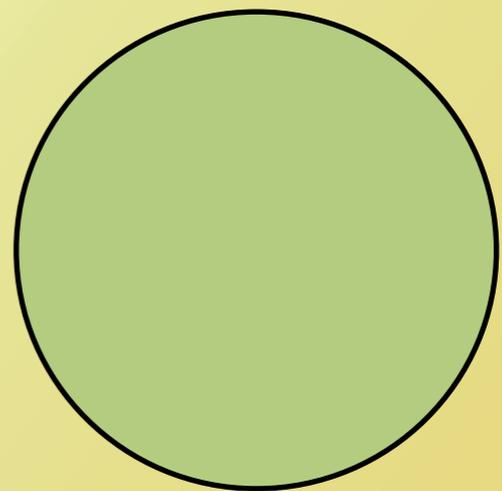
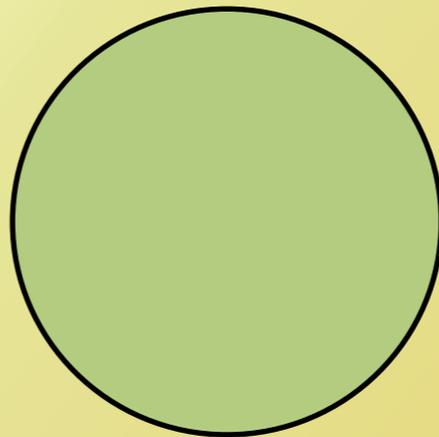
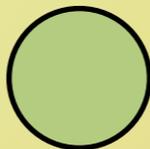
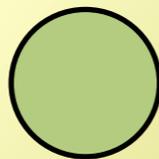
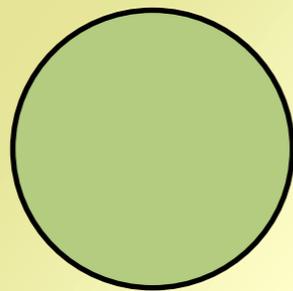
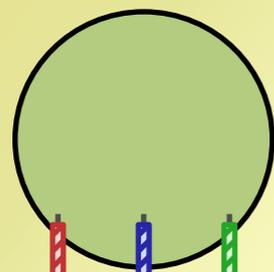
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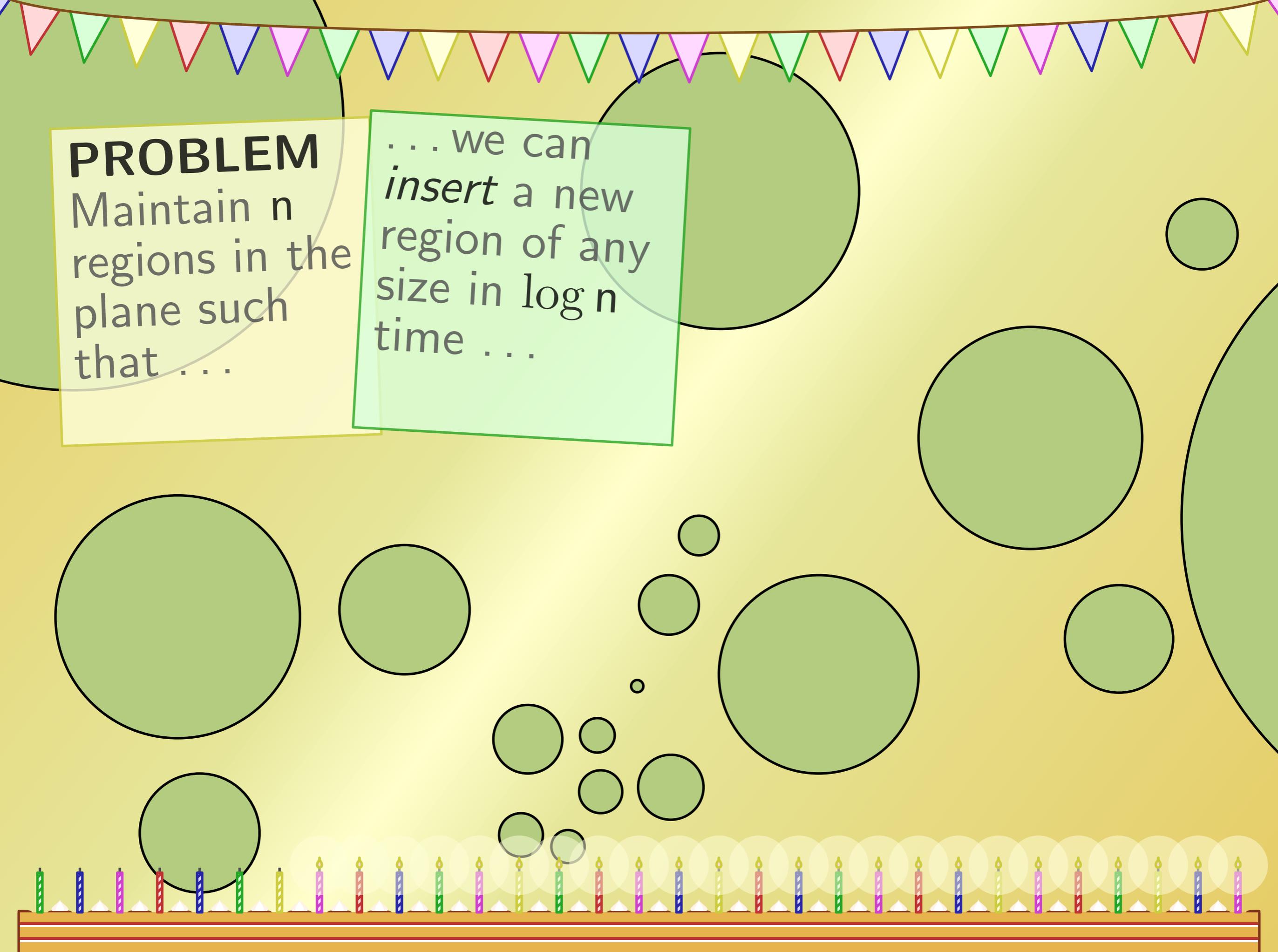




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Maintain n regions in the plane such that ...

... we can *insert* a new region of any size in $\log n$ time ...



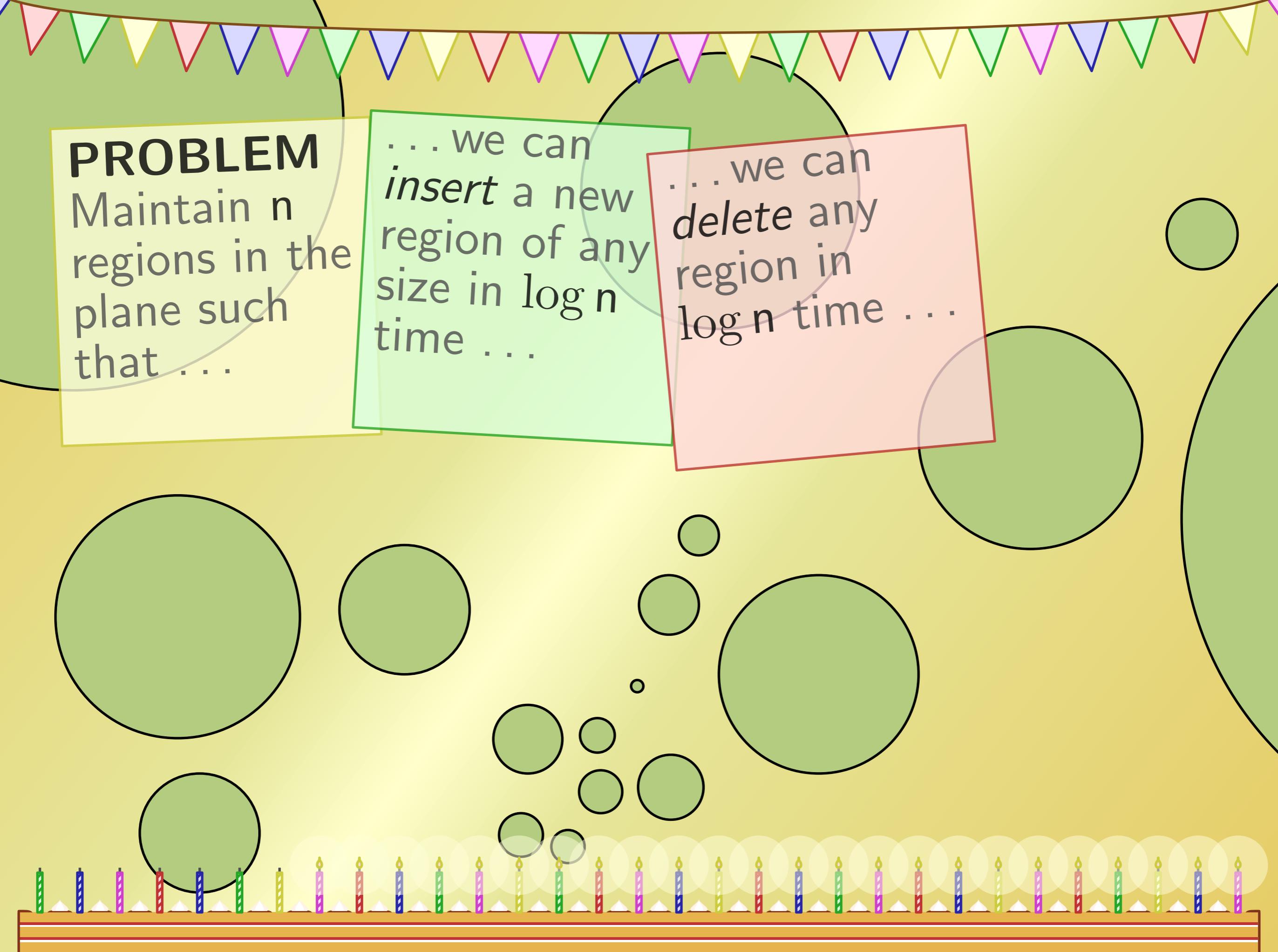


PROBLEM

Maintain n regions in the plane such that

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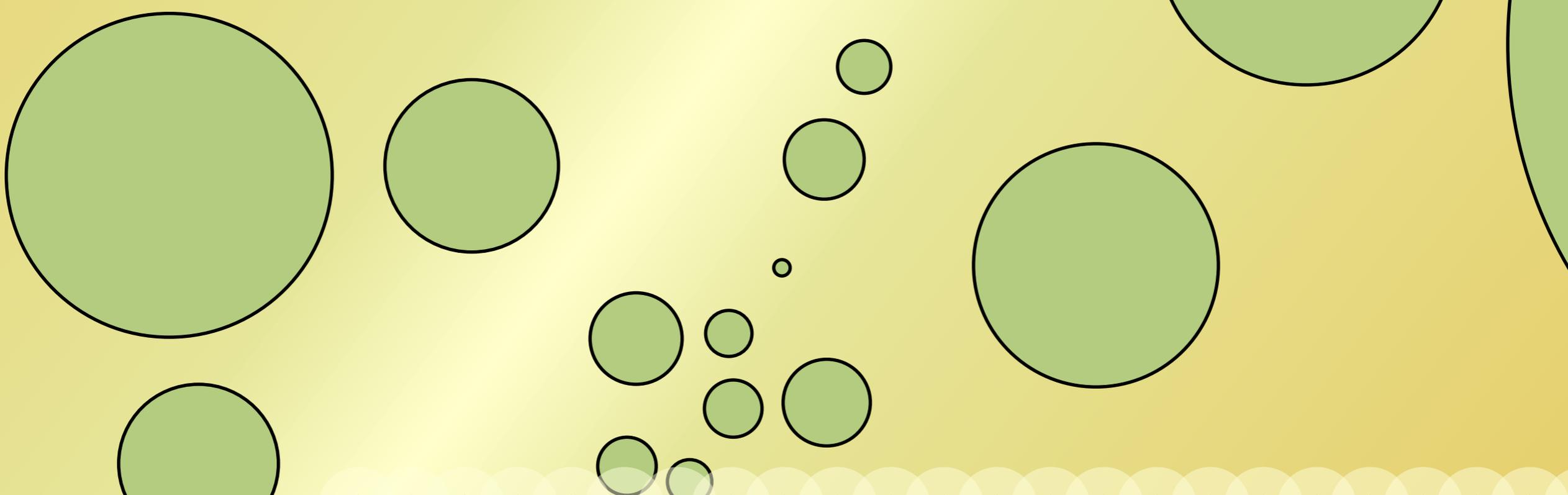
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Maintain n regions in the plane such that ...

... we can *insert* a new region of any size in $\log n$ time ...

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... we can *locally alter, or update,* a region in less than $\log n$ time ...



PROBLEM

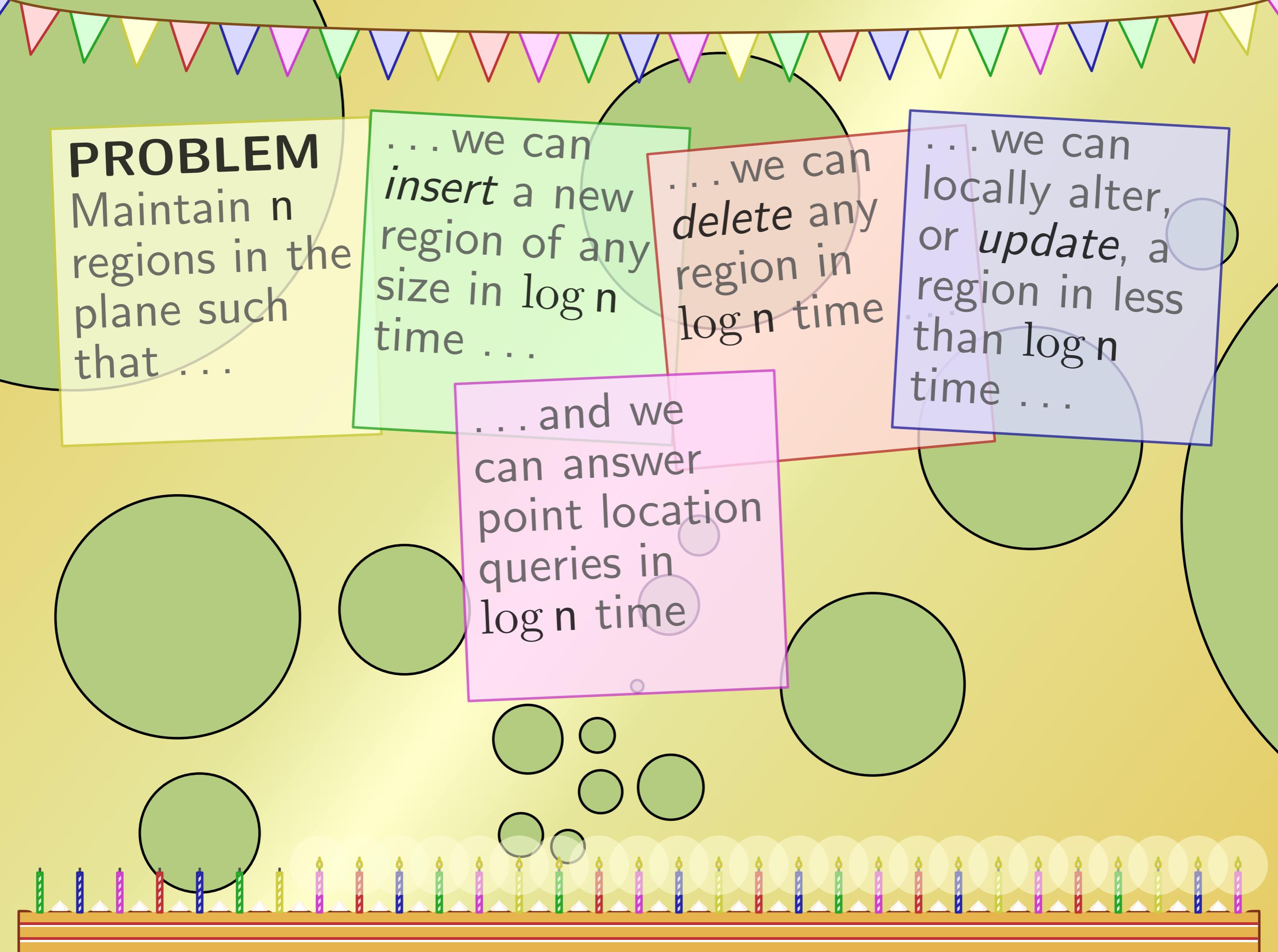
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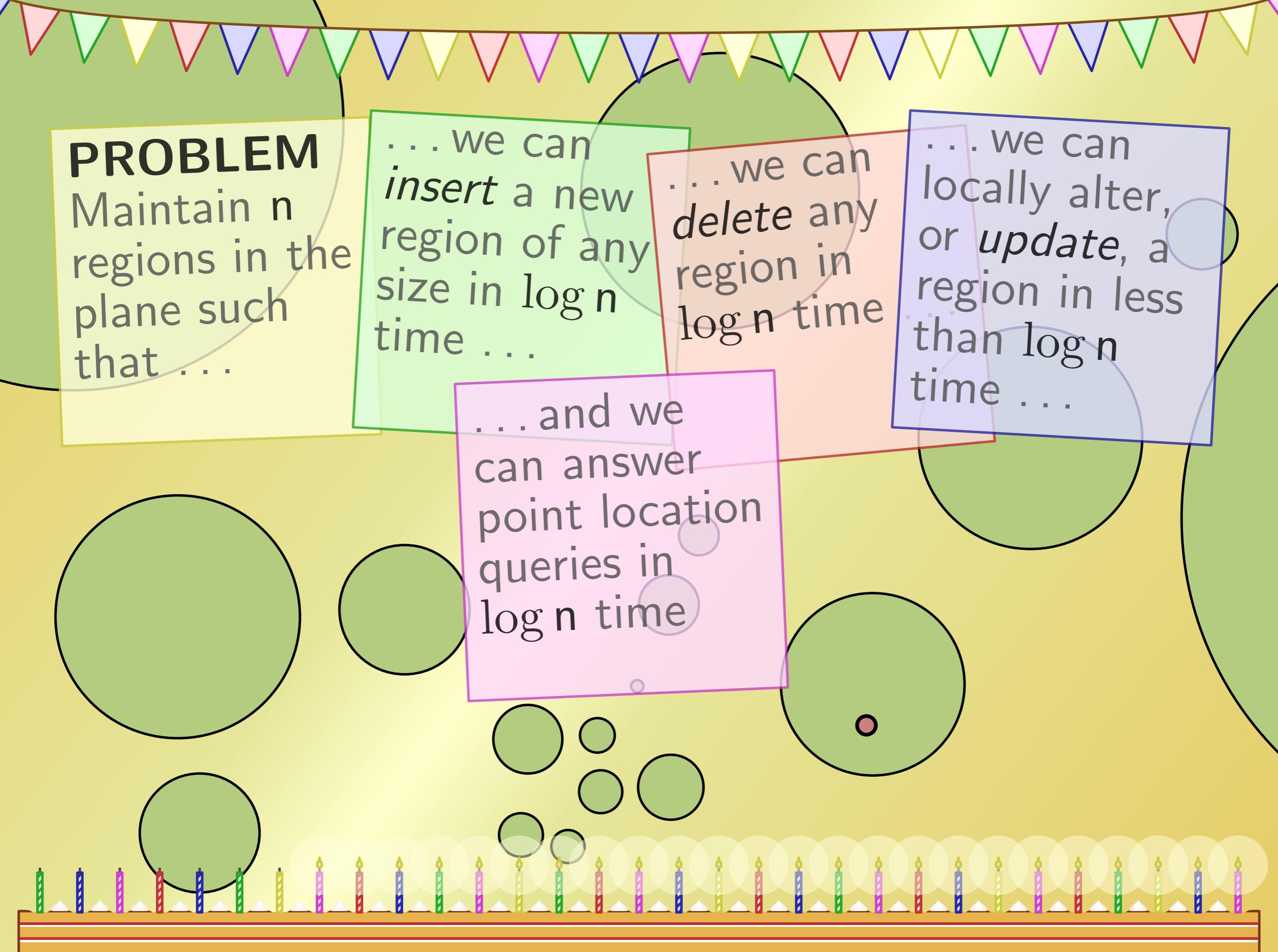
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When is an
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When is an update *local*?

Regions can grow or shrink by at most a constant factor.

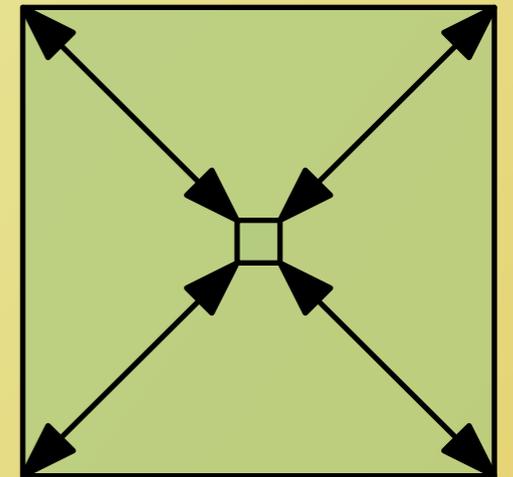
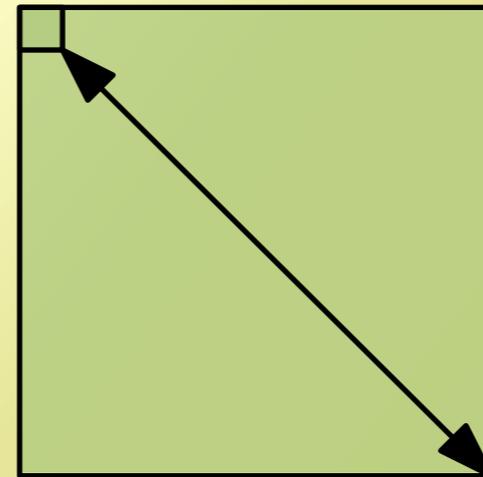
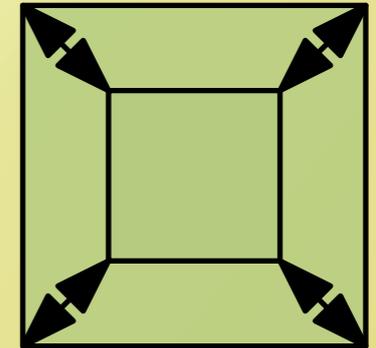
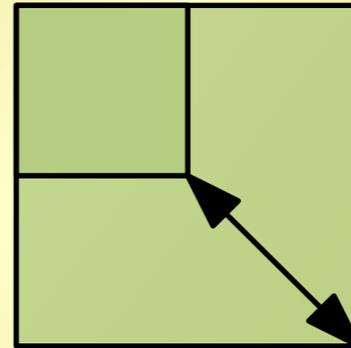


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GOOD



BAD



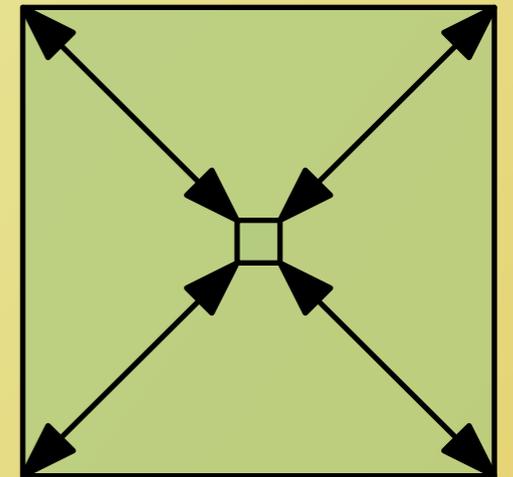
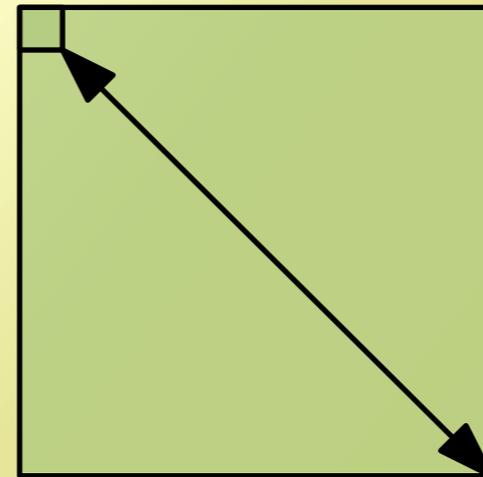
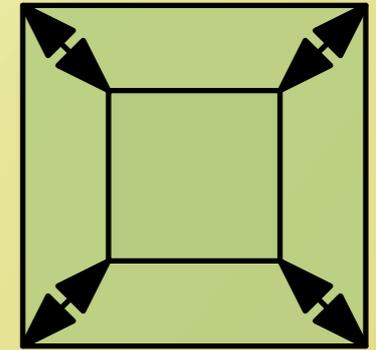
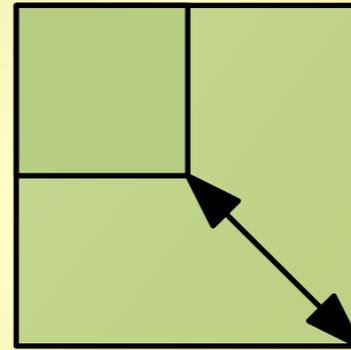
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Regions can move a constant times their current size.

GOOD



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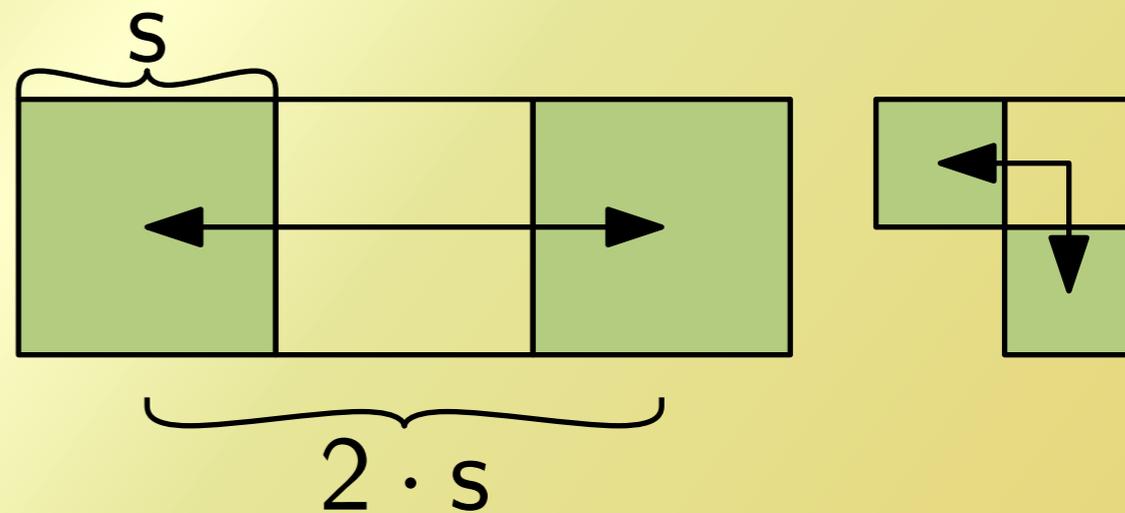
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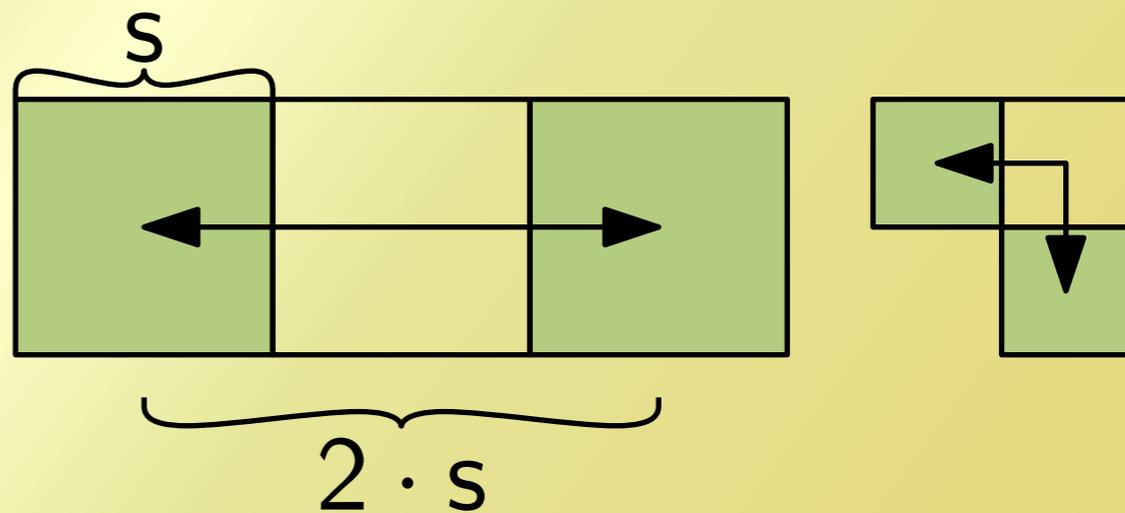
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Regions can grow or shrink by at most a constant factor.

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Regions can change their shape, as long as they stay *fat*.

GOOD



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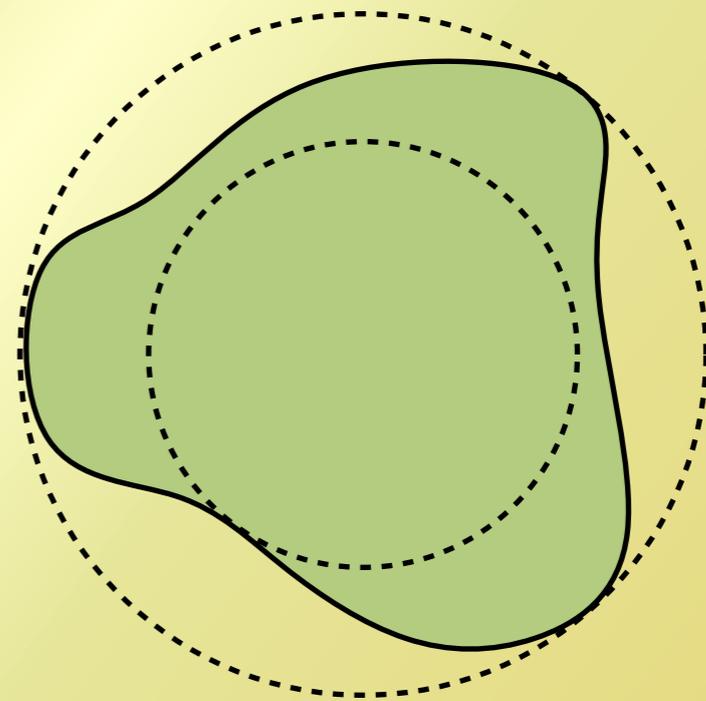
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GOOD



BAD

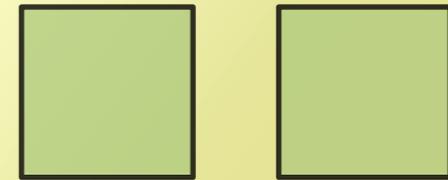


One more
assumption:
the regions
are and stay
disjoint!

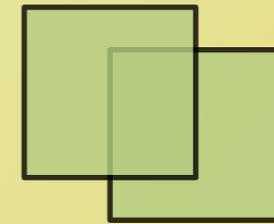


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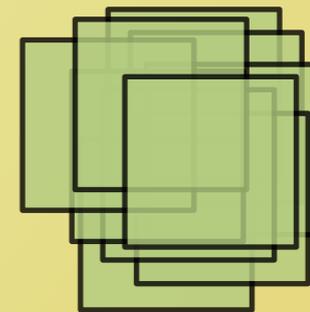
GOOD



BAD



UGLY





And the
results are

...





And the
results are

...

1D:





And the
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...

1D:

Queries: $O(\log n)$ time





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1D: Queries: $O(\log n)$ time
 Insertions and deletions: $O(\log n)$ time





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2D:





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 Local updates: $O(1)$ time

2D: Queries: $O(\log n)$ time





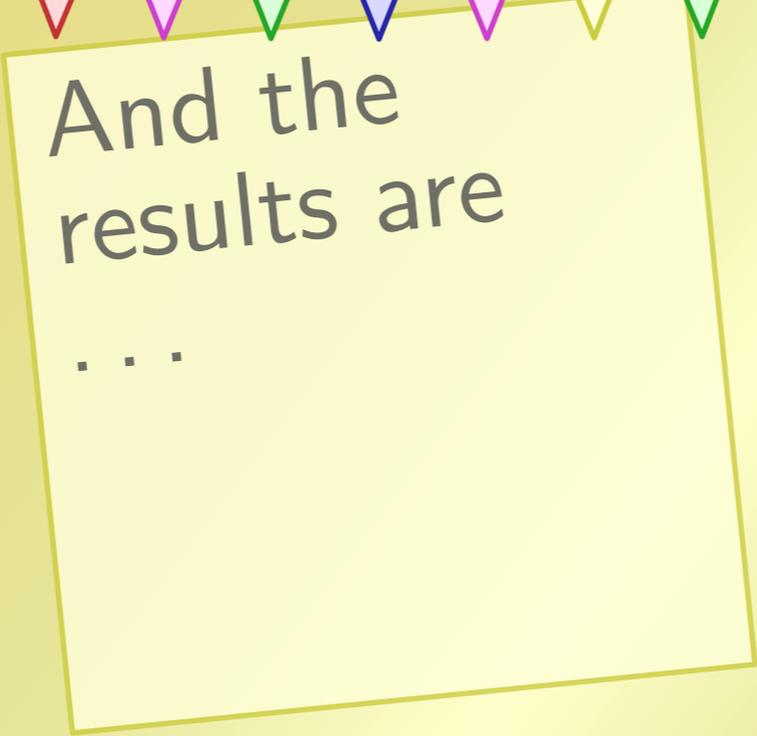
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 Insertions and deletions: $O(\log n)$ time
 Local updates: $O(\log n / \log \log n)$ time



TECHNICAL DETAILS: 1 DIMENSION



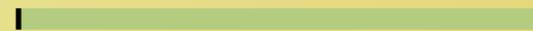
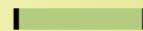
1-dimensional
regions are
intervals.





1-dimensional
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They move
around on a
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intervals are
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ones are slow.





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NOTE

Big intervals
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We need a structure that provides quick access to “similar places” . . .





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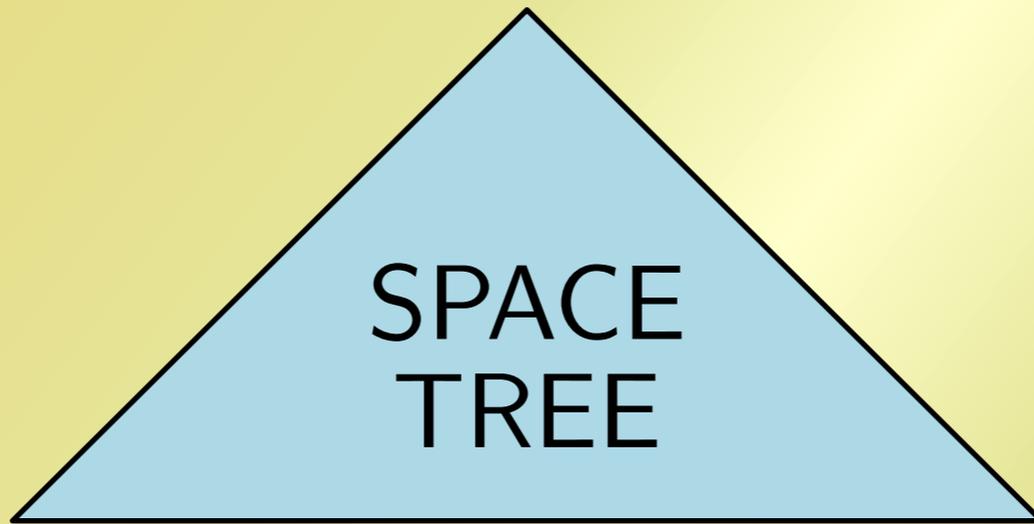
IDEA Let's maintain two trees.



We need a structure that provides quick access to “similar places” ...

... but also supports some sort of binary search.

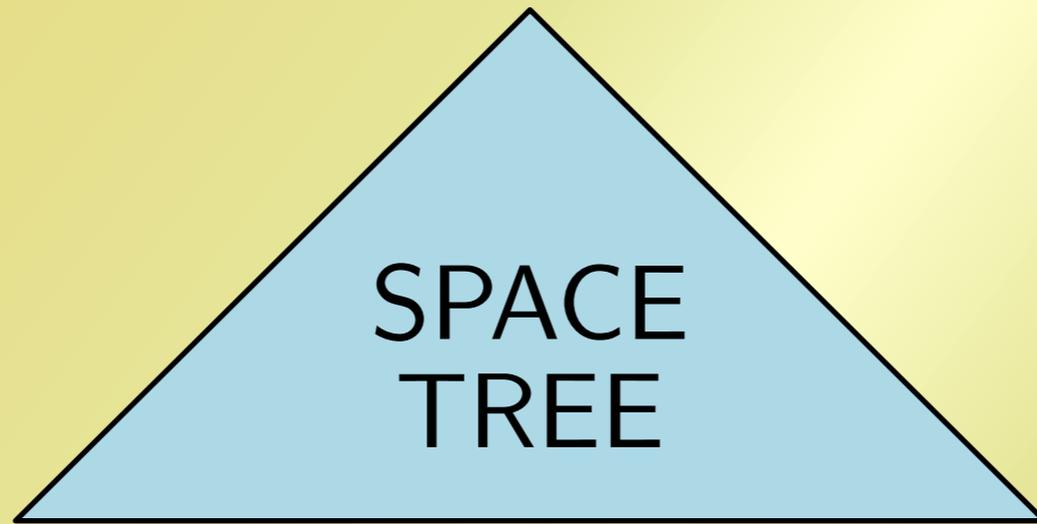
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\mathcal{R}

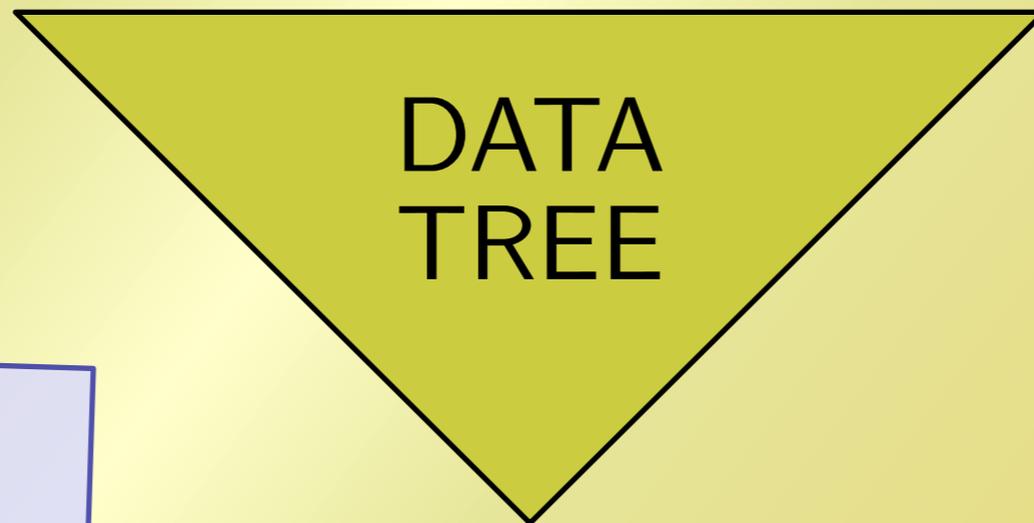
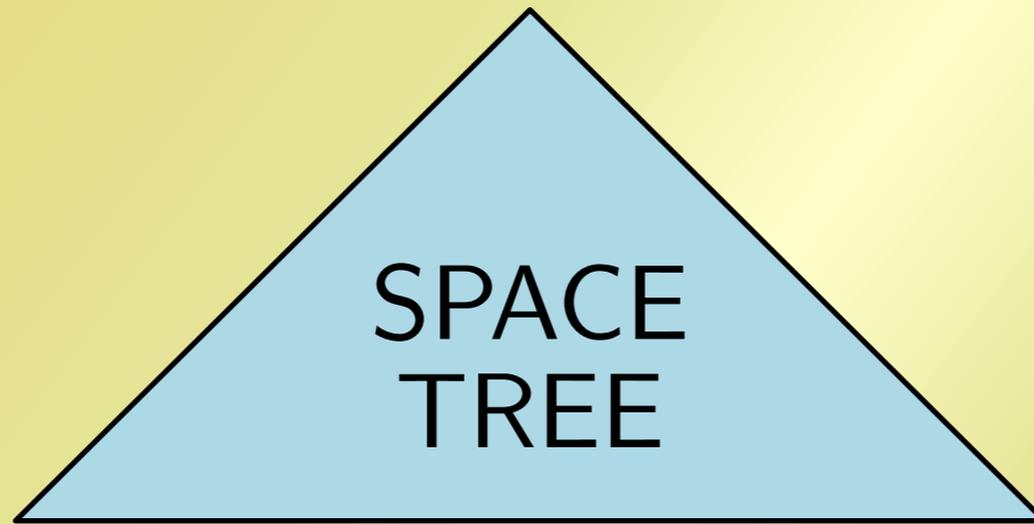


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$O(1)$ Updates



$O(\log n)$ Queries



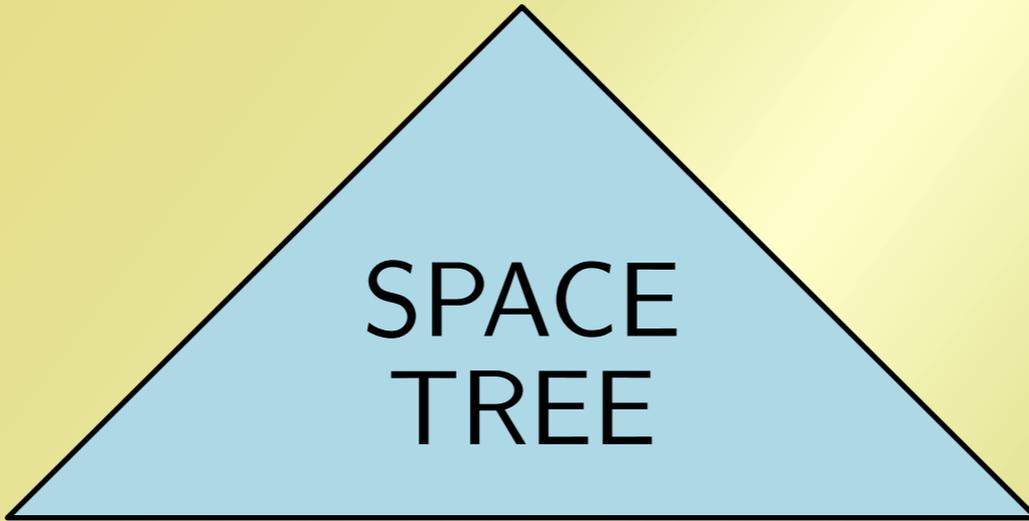


SPACE
TREE

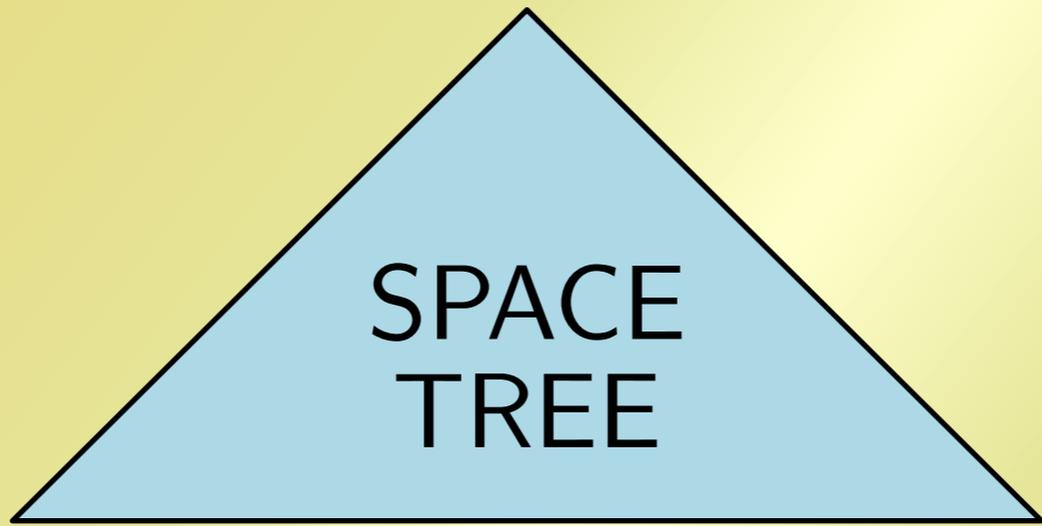




For the space tree we use a *quadtree*.



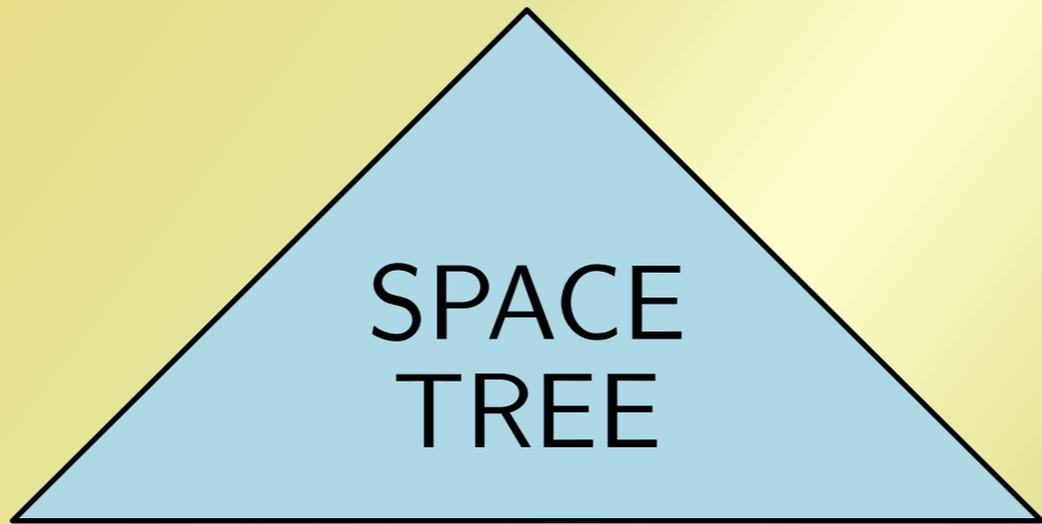
For the space tree we use a *quadtree*.



Consider the set P of midpoints of the intervals.



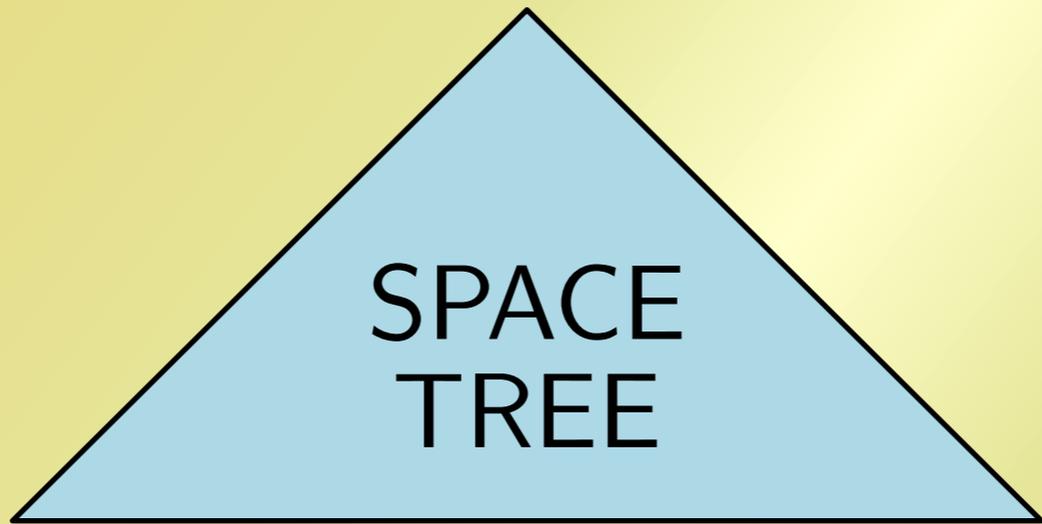
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Consider the set P of midpoints of the intervals.



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Consider the set P of midpoints of the intervals.

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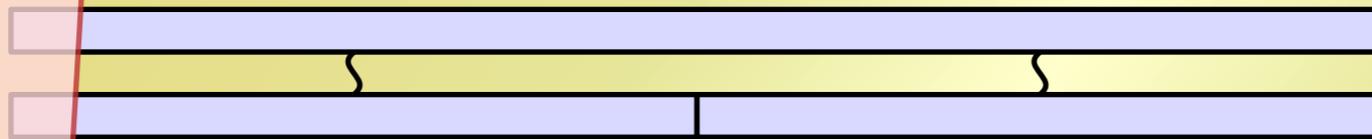
Consider the set P of midpoints of the intervals.

Construct a *root* box containing all points of P .

Recursively split boxes that contain at least 2 points.



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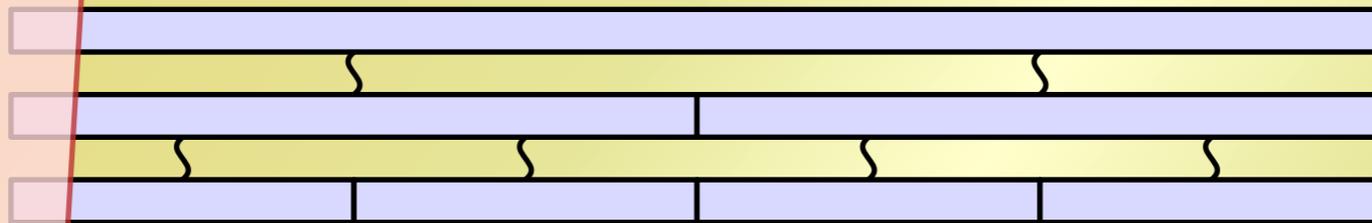
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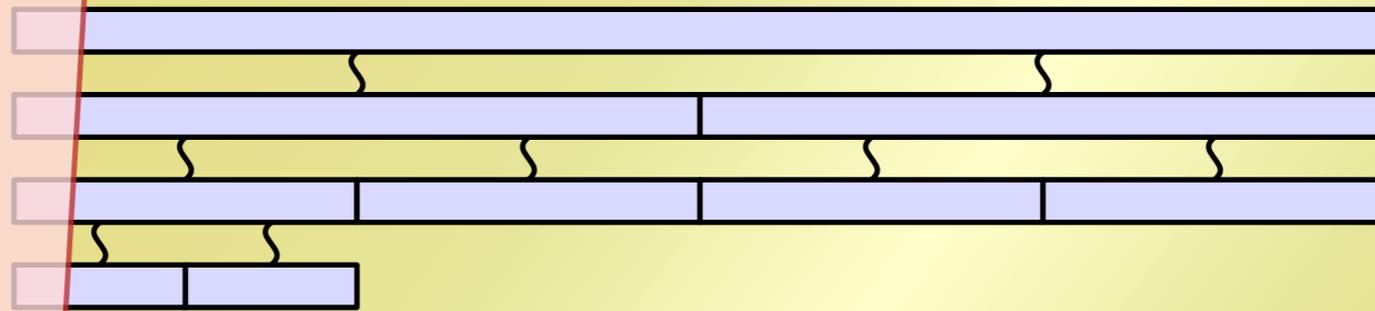
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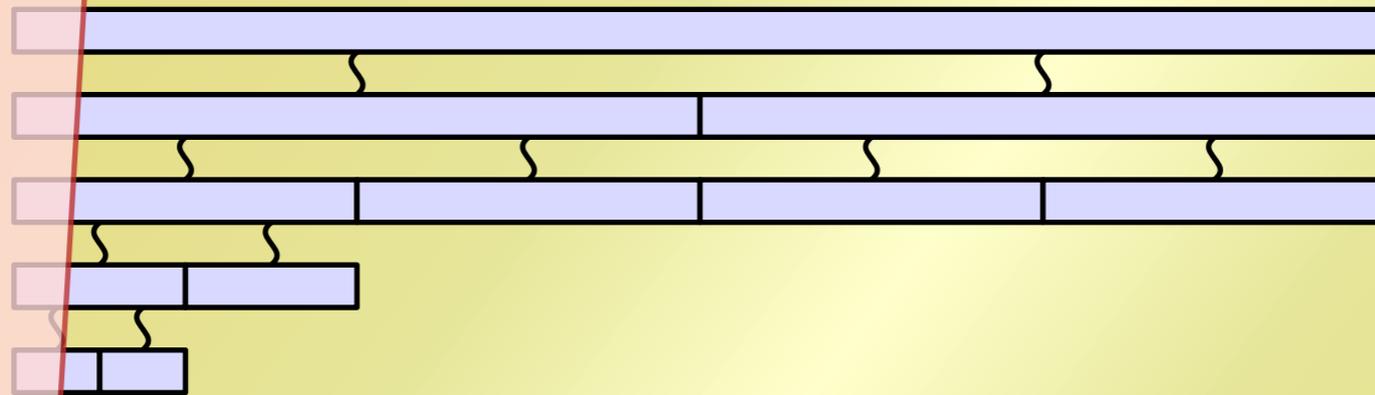
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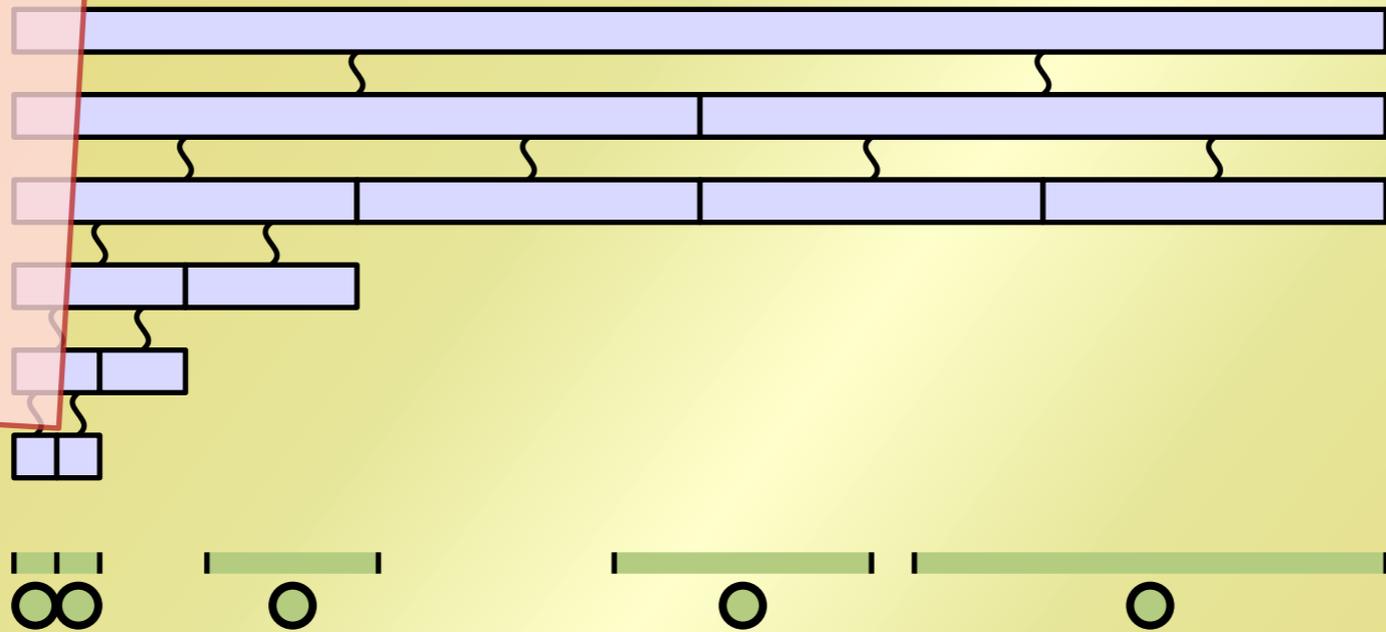
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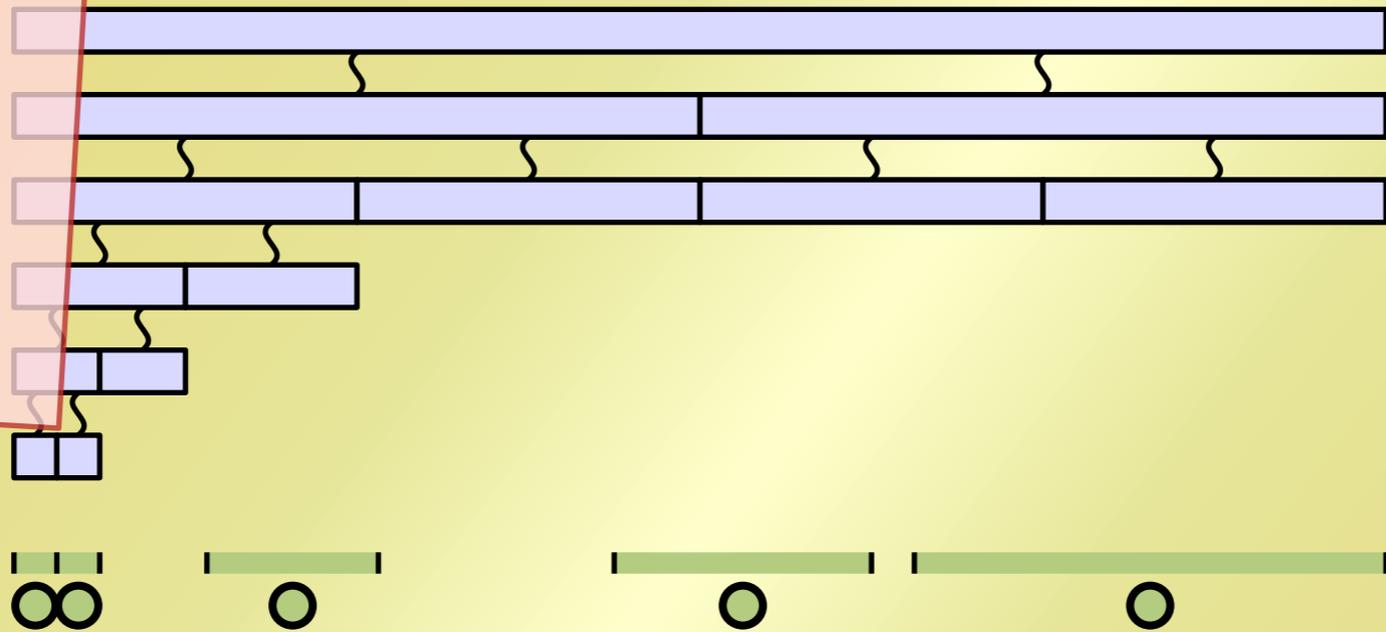
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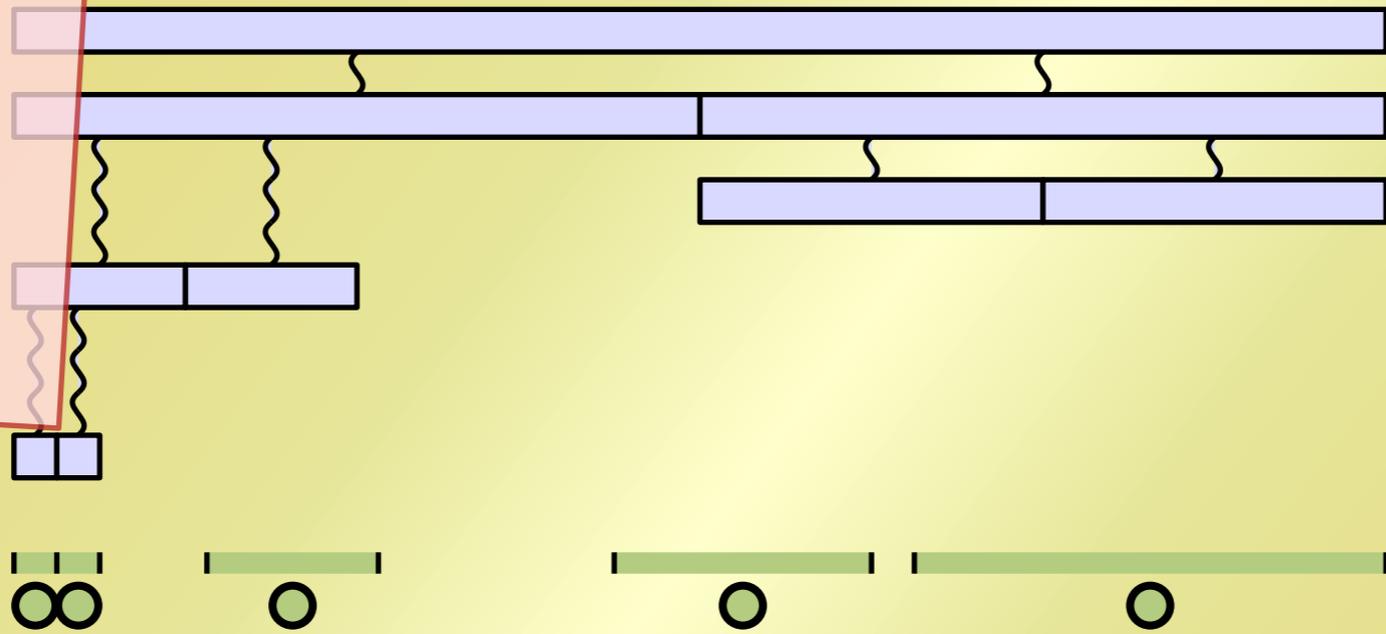
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Compress the tree by deleting long empty paths.



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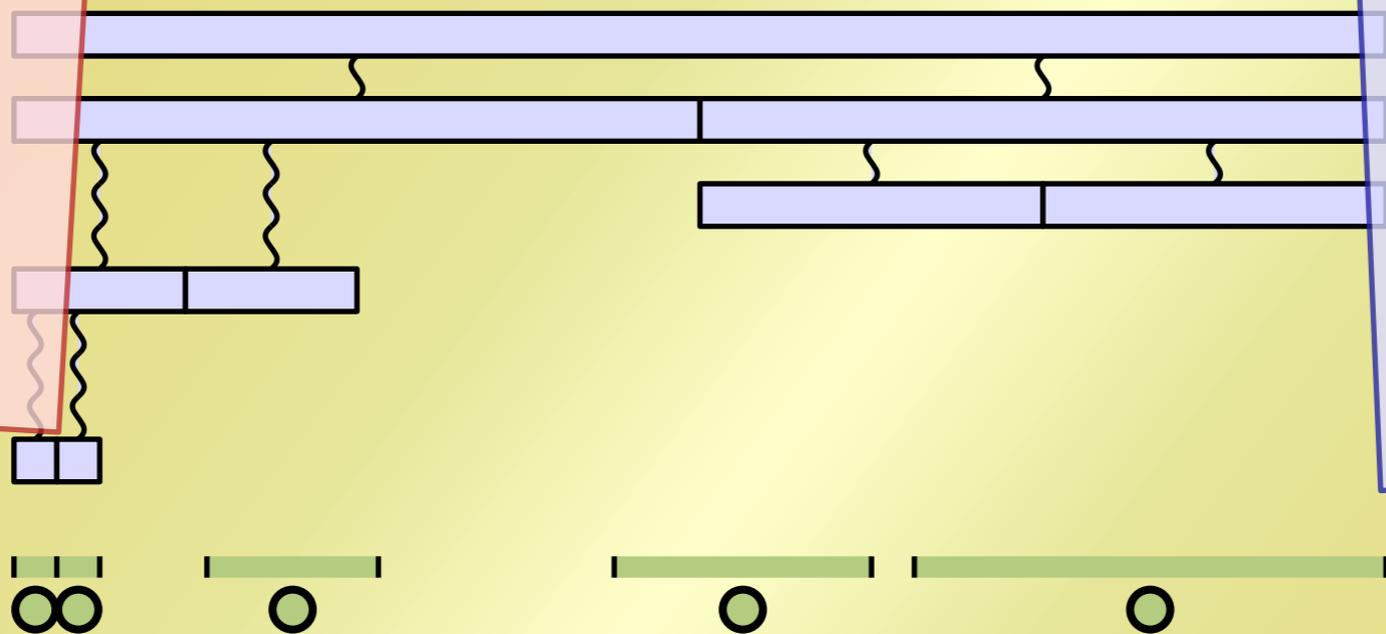
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Finally, add pointers between neighbouring boxes of the same size.

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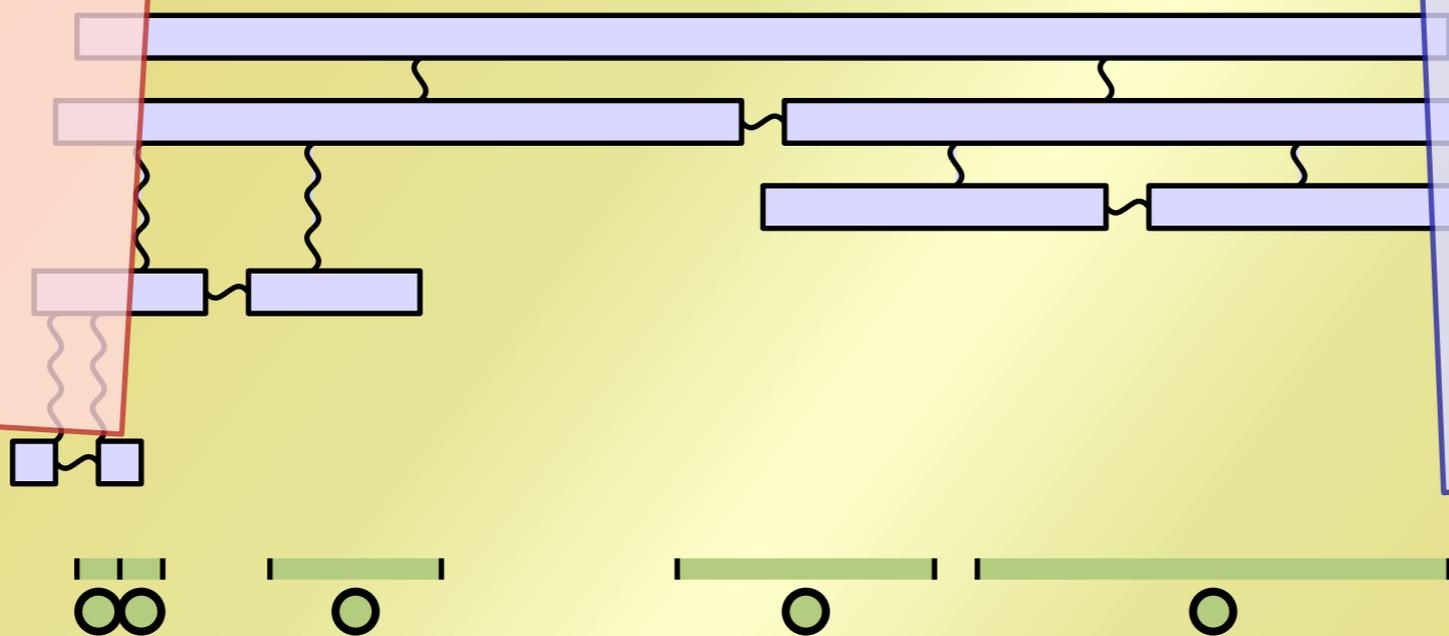
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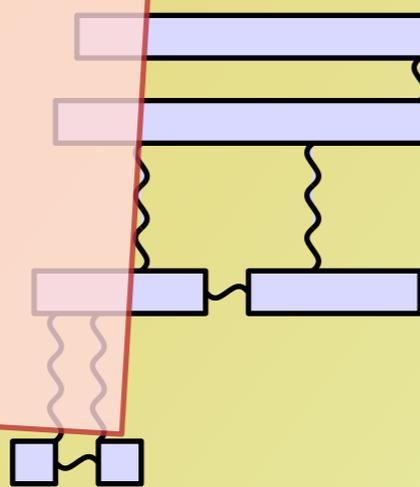
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LEMMA

No leaf is much smaller than the interval it stores.

Finally, add pointers between neighbouring boxes of the same size.

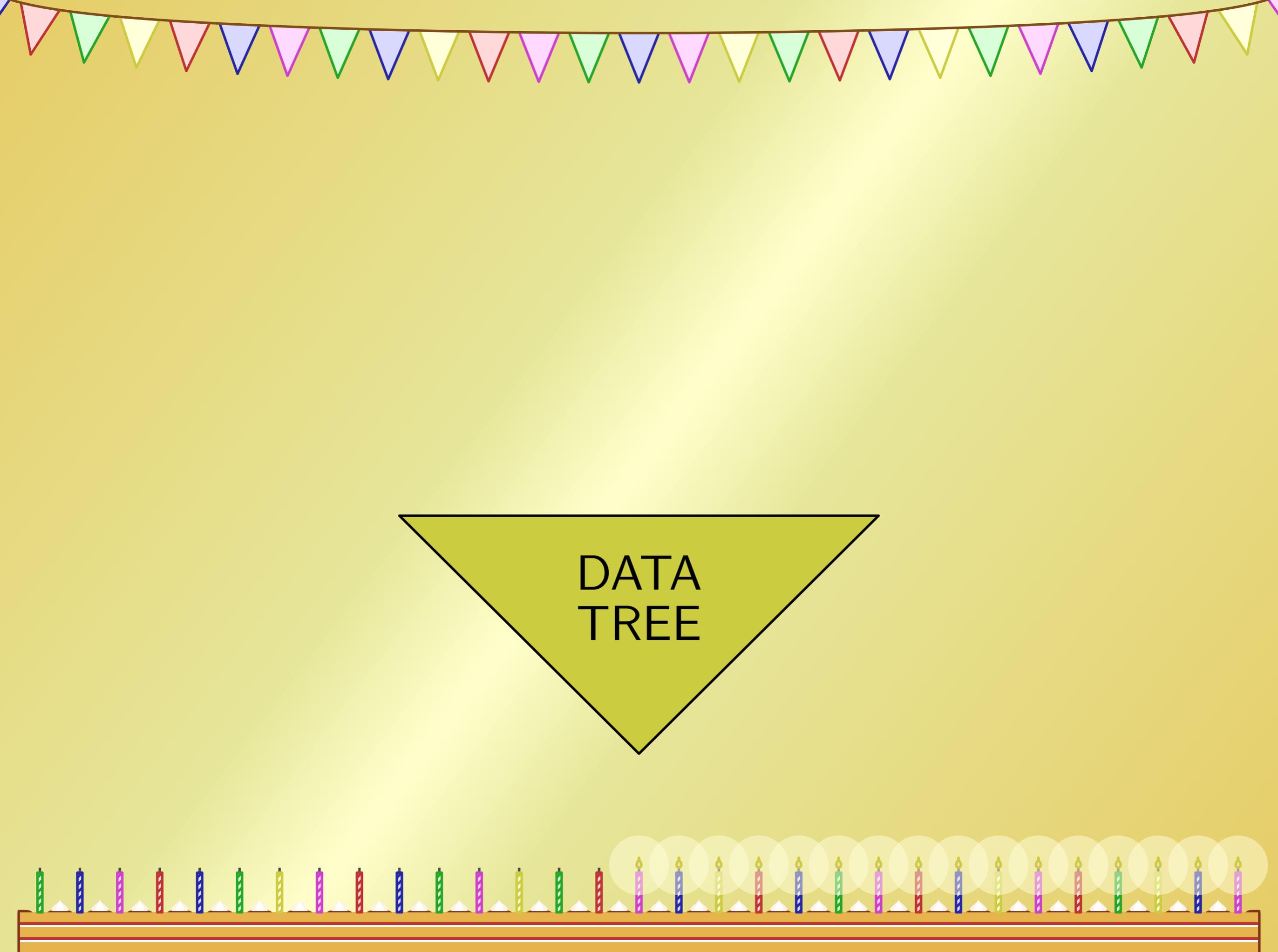
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DATA
TREE

For the data tree we use a dynamic search tree.

DATA
TREE





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Again, consider the midpoints of the intervals.



DATA
TREE



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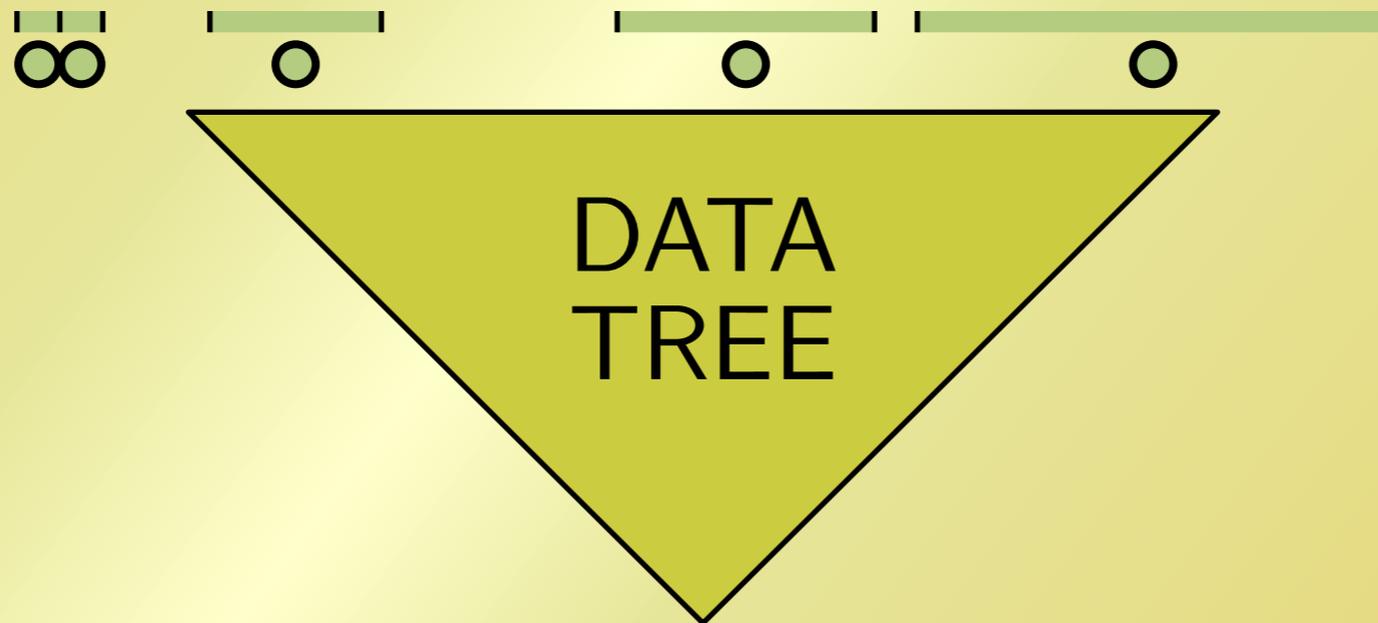
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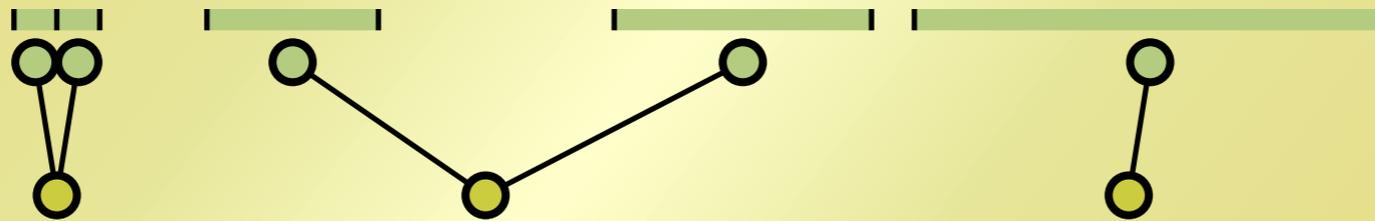
We now only care about their order, and build a *balanced* tree.



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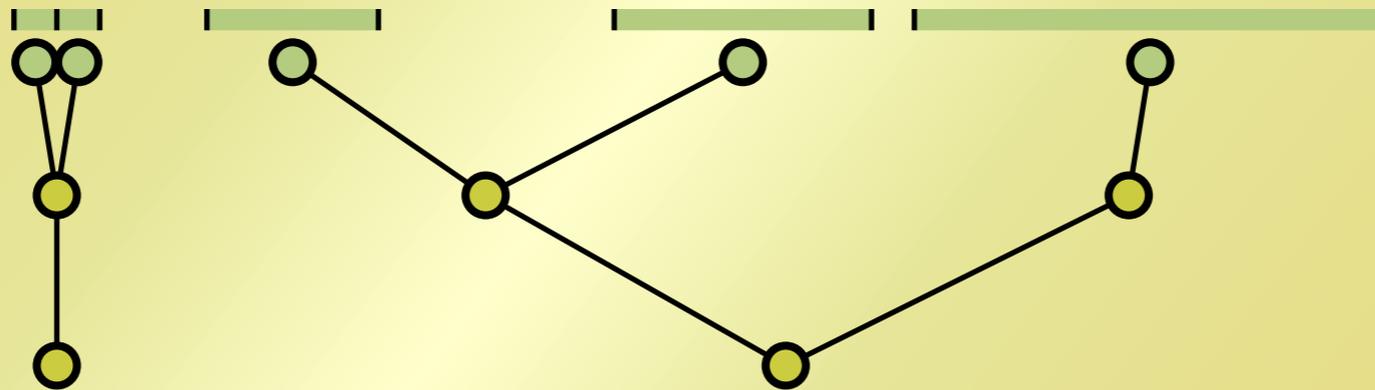
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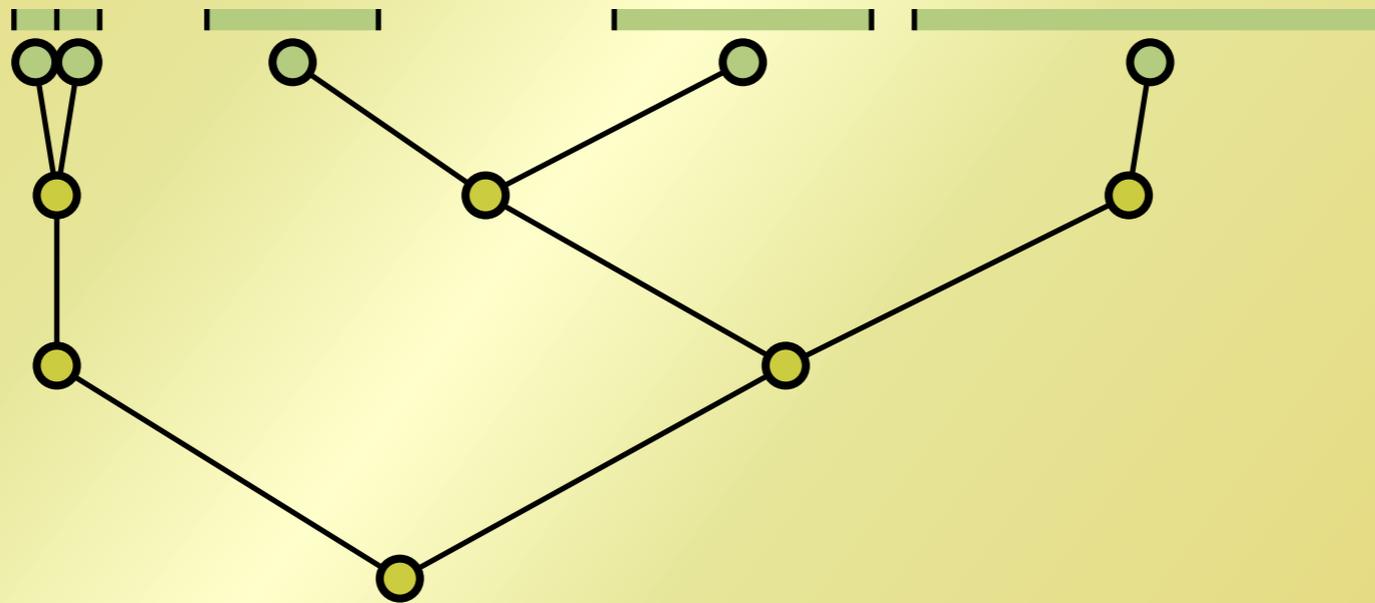
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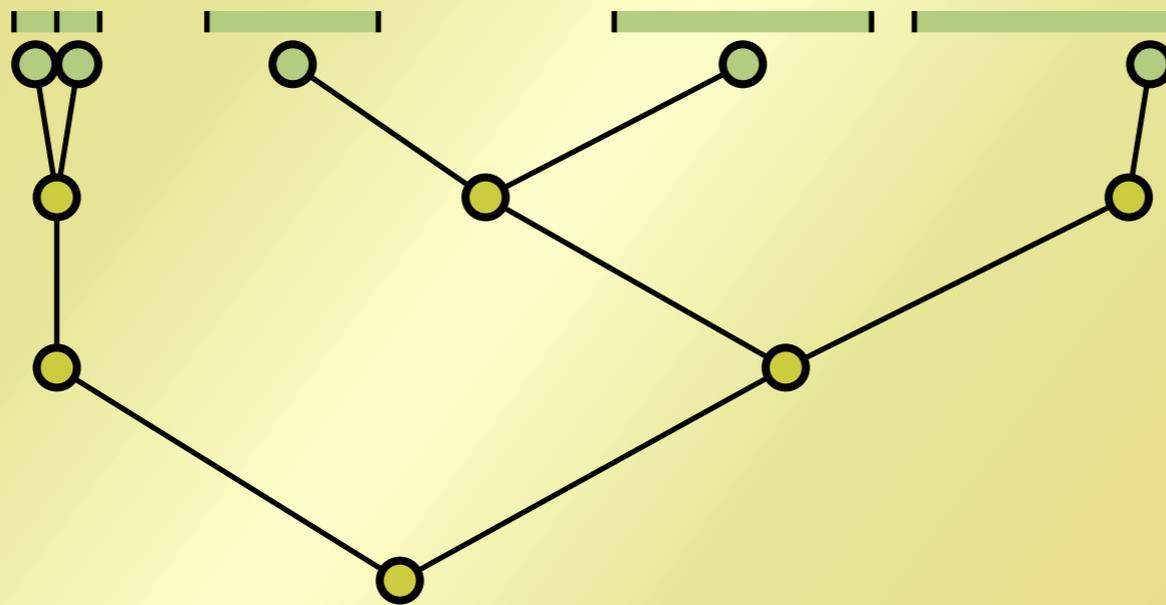
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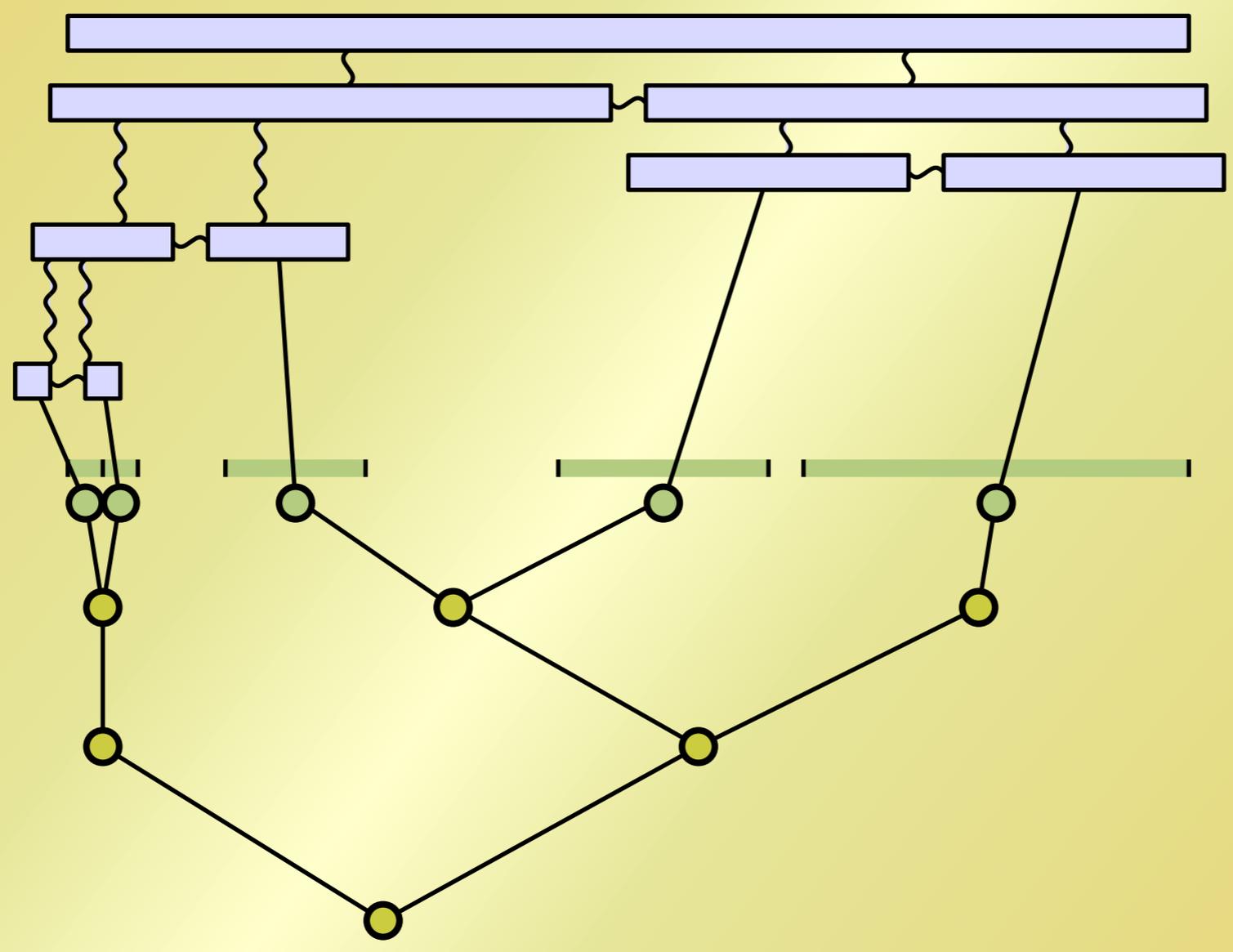
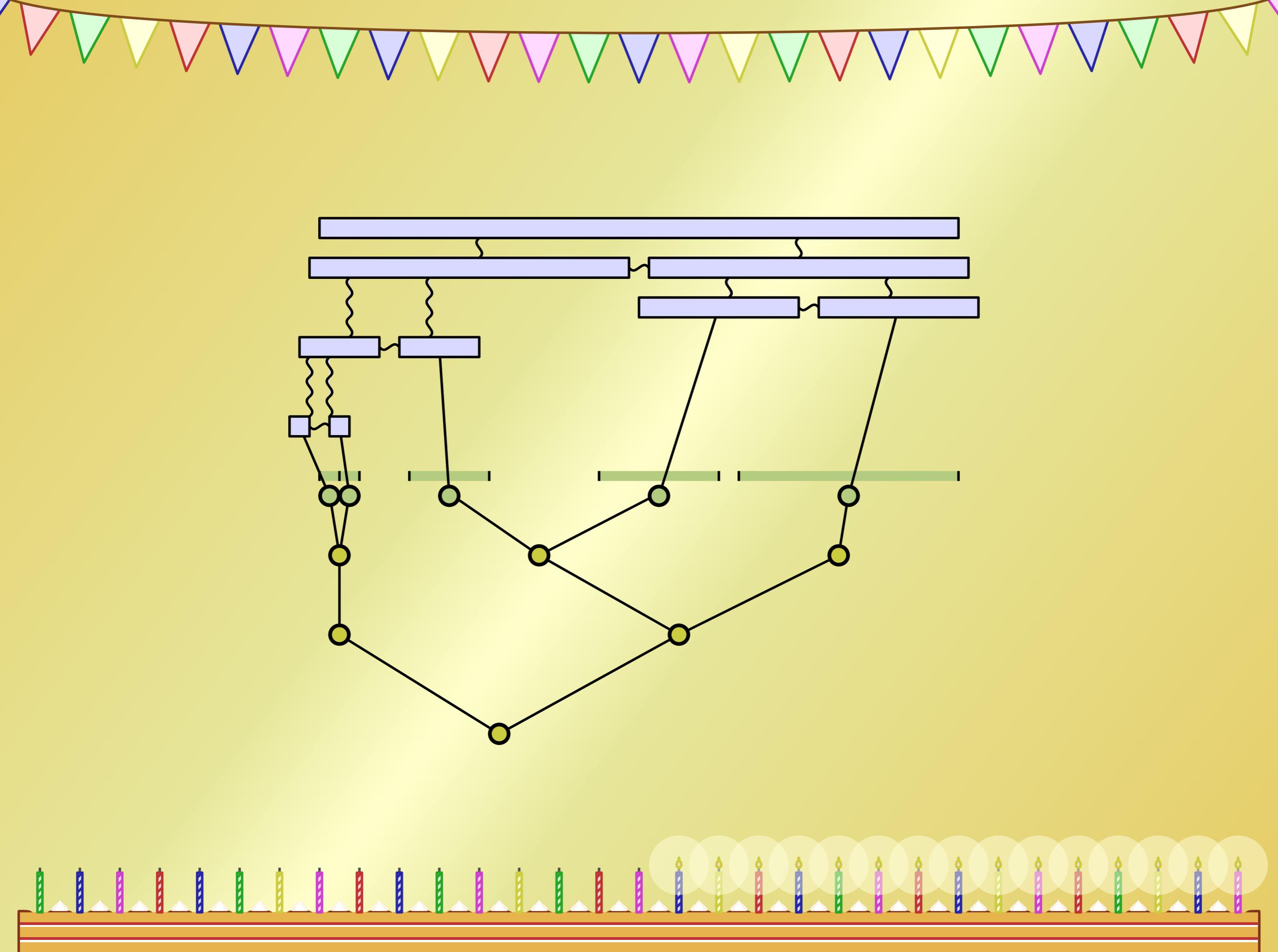
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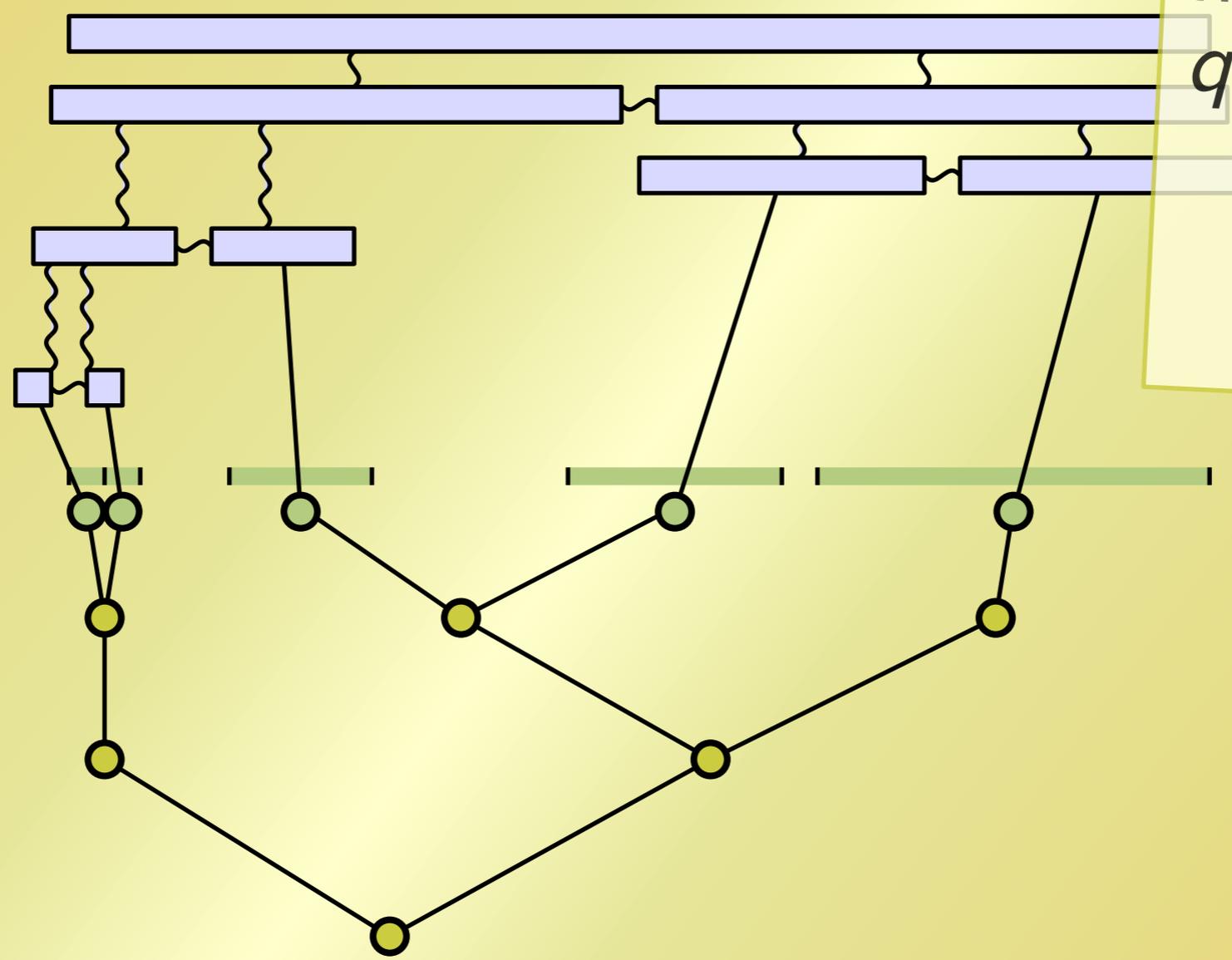
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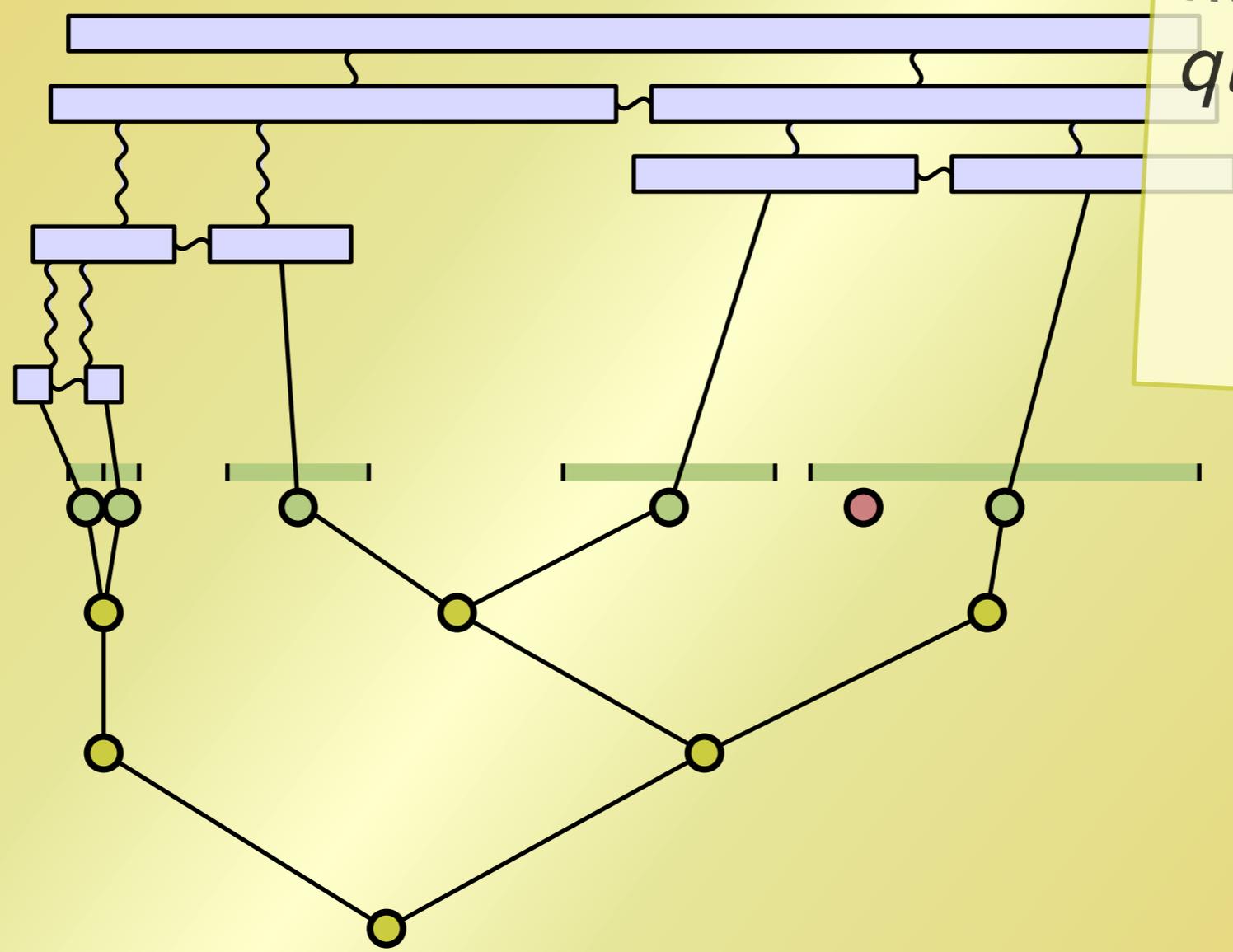
LEMMA
The search tree has logarithmic depth.



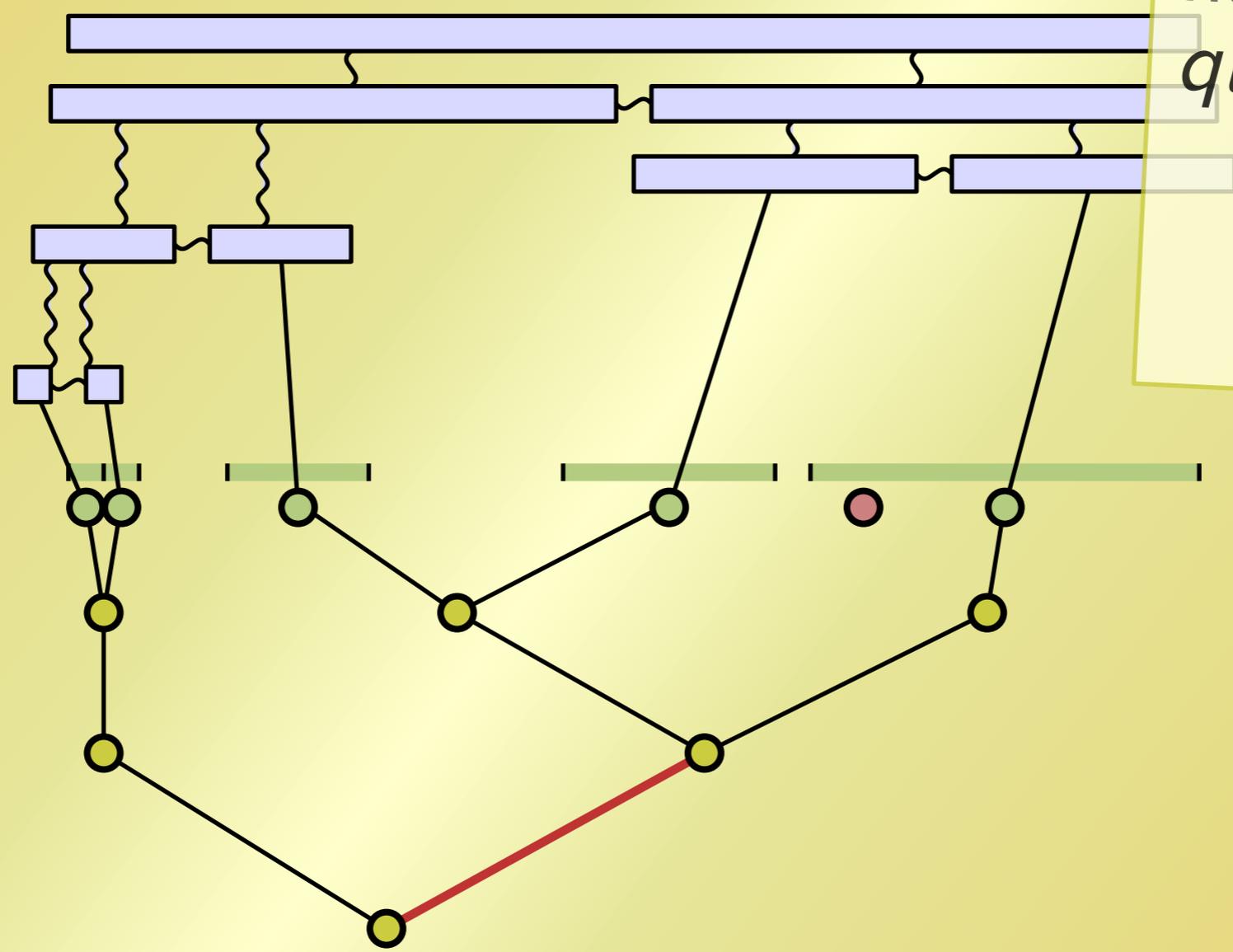
How do we handle a query?



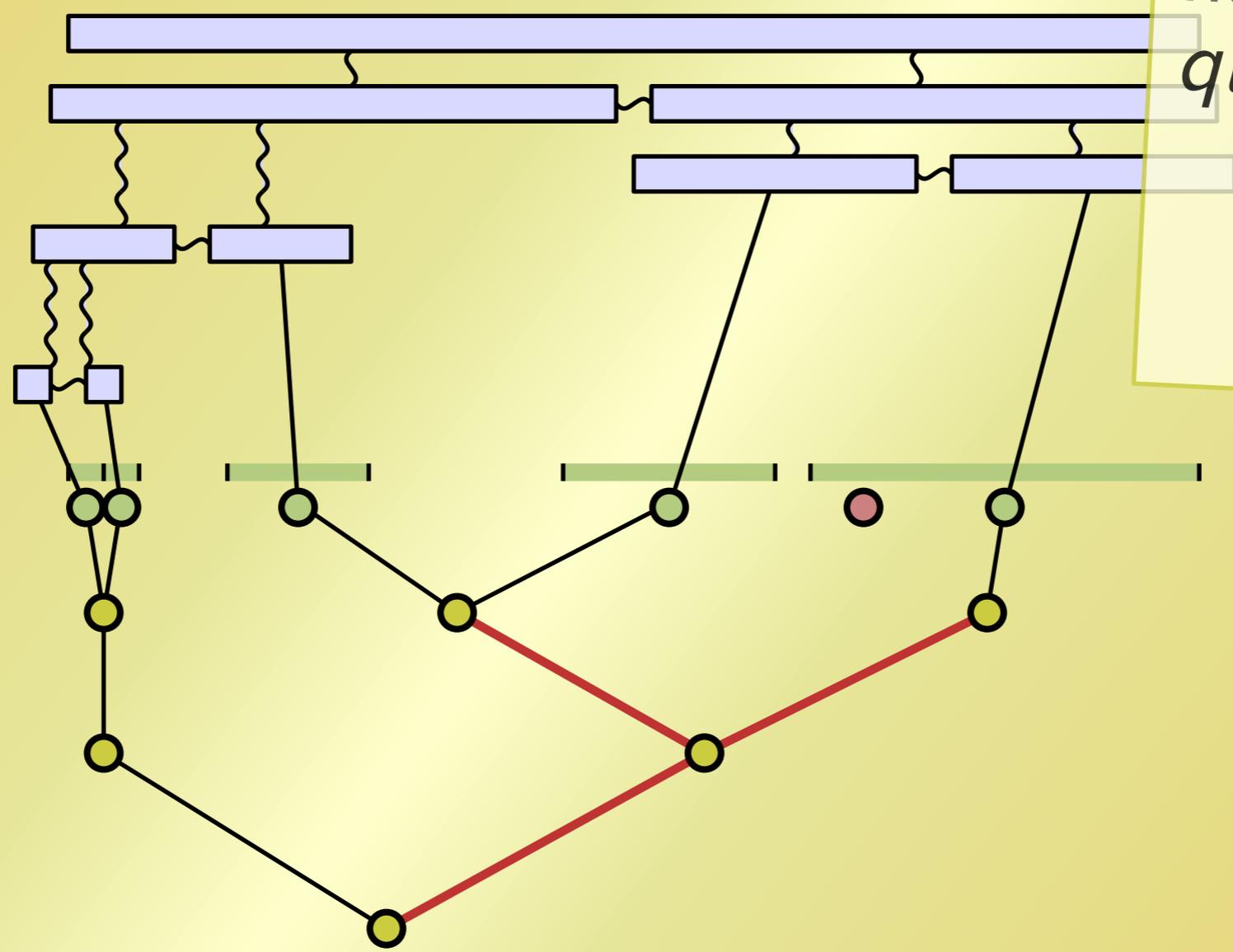
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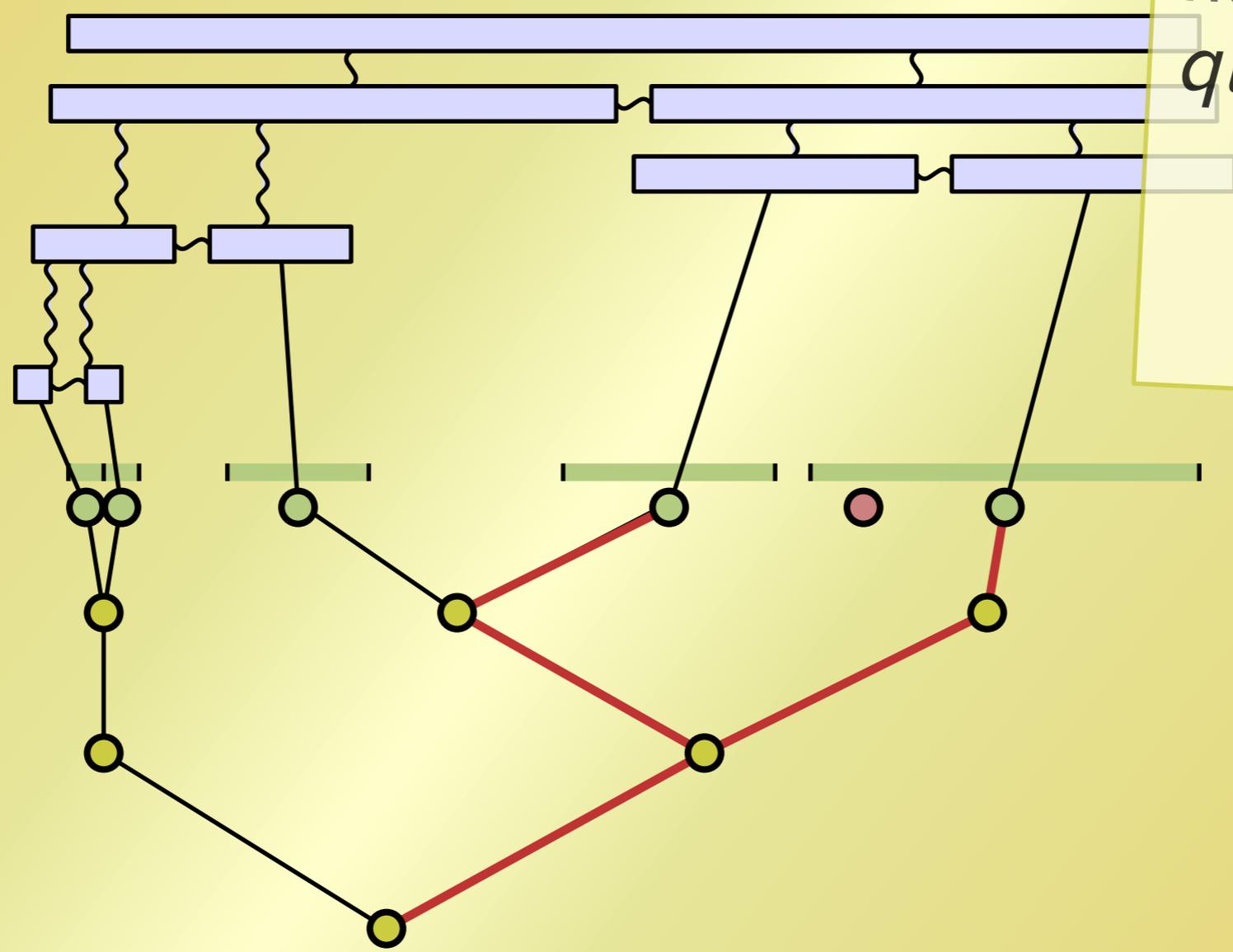


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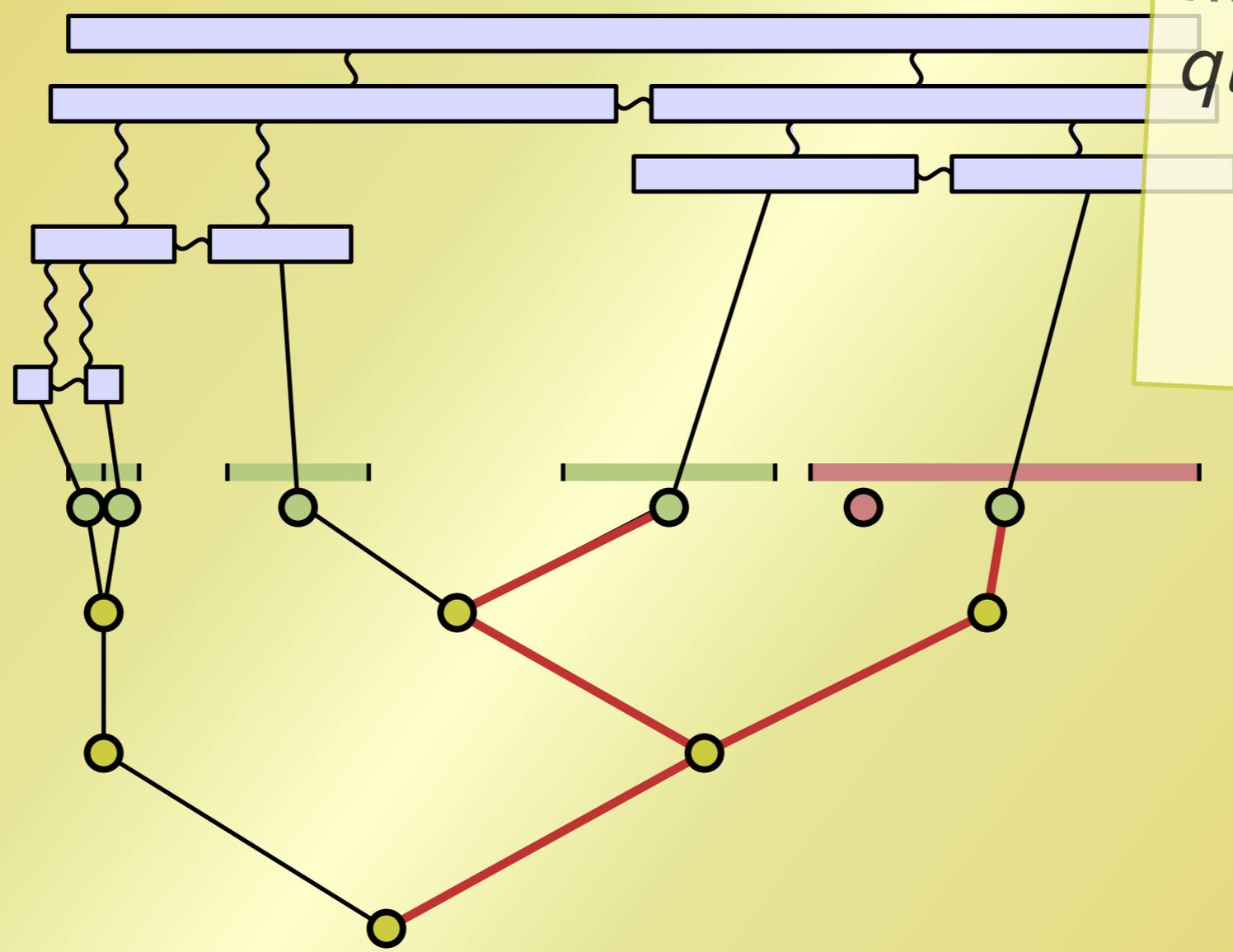




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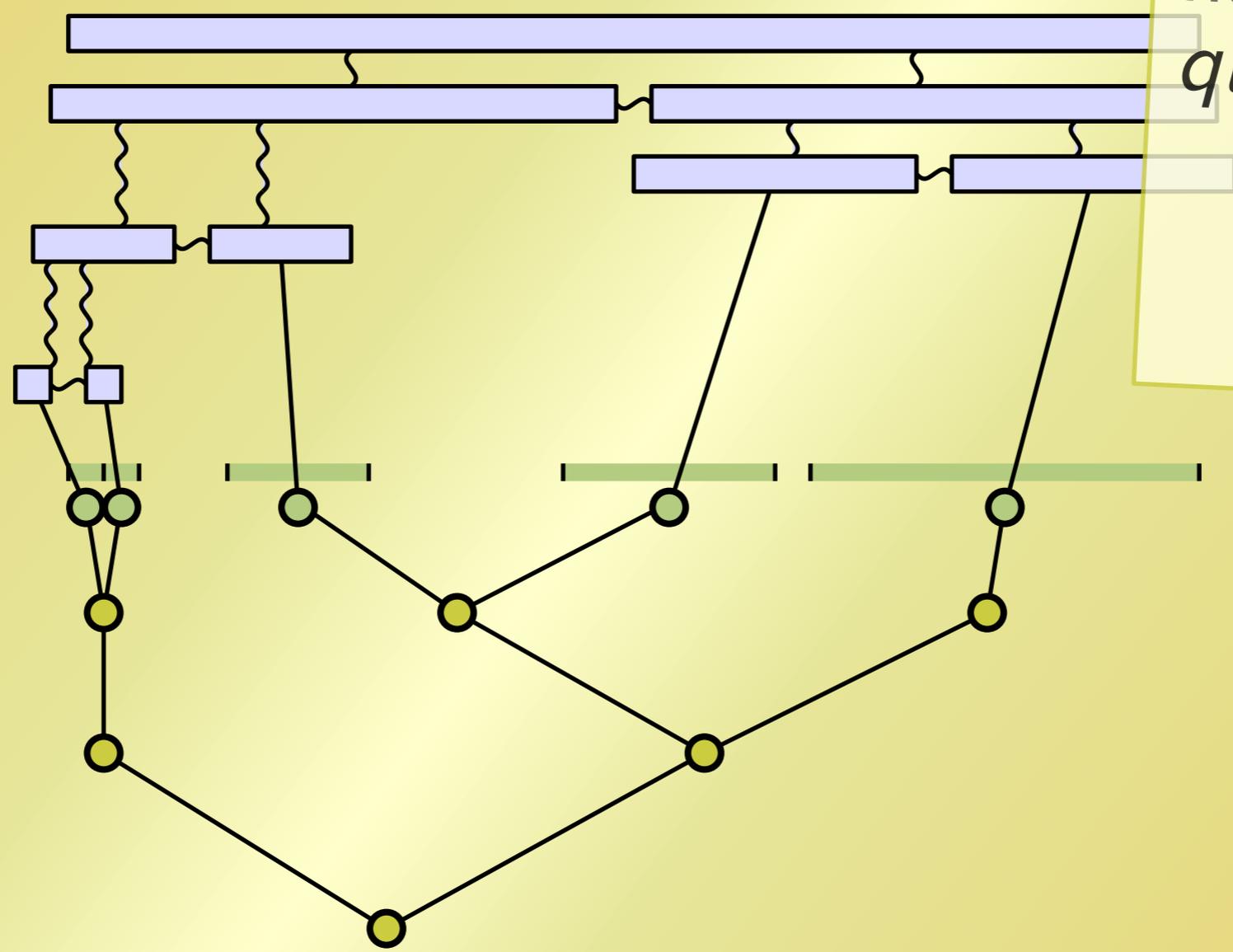


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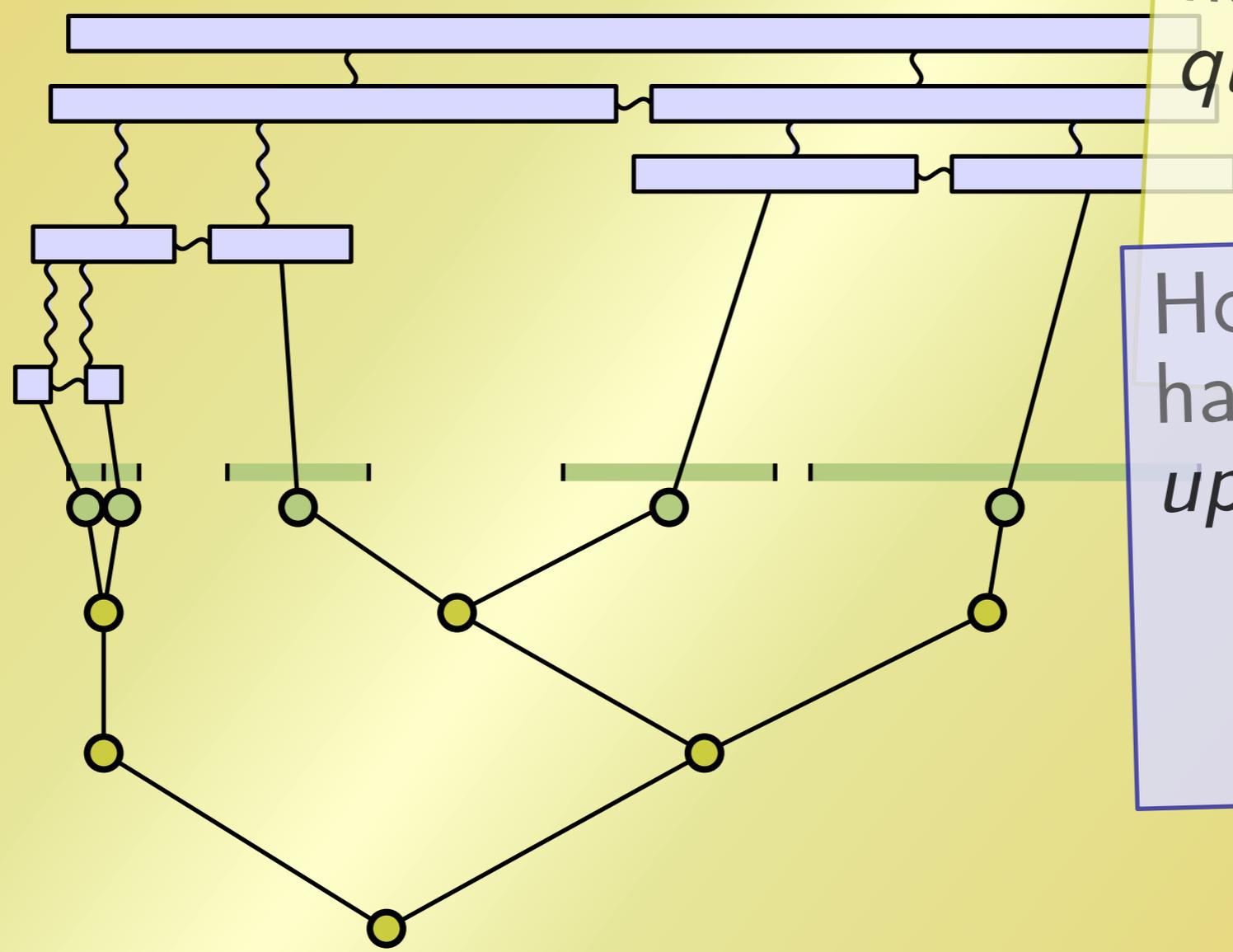
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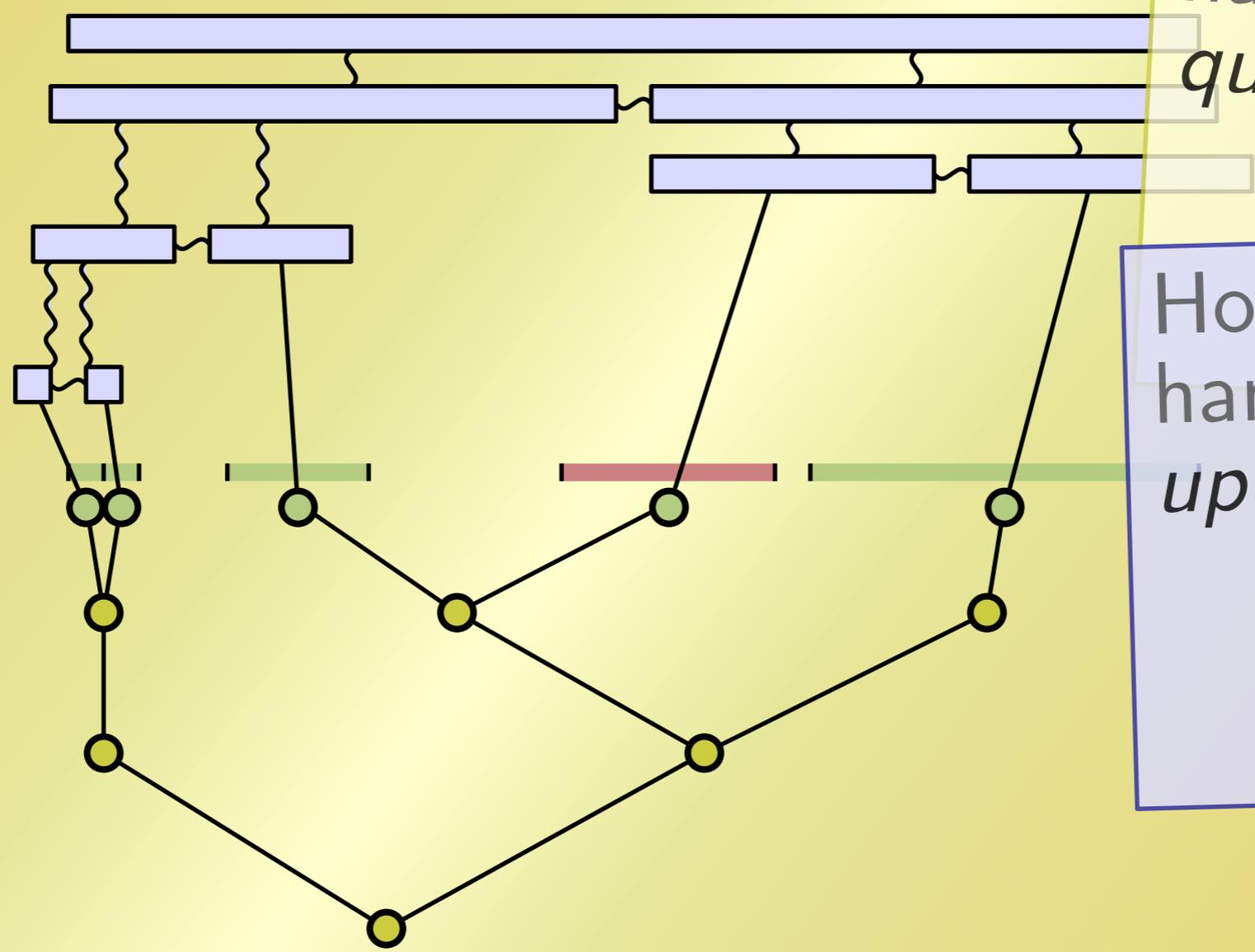
How do we handle an update?





How do we handle a query?

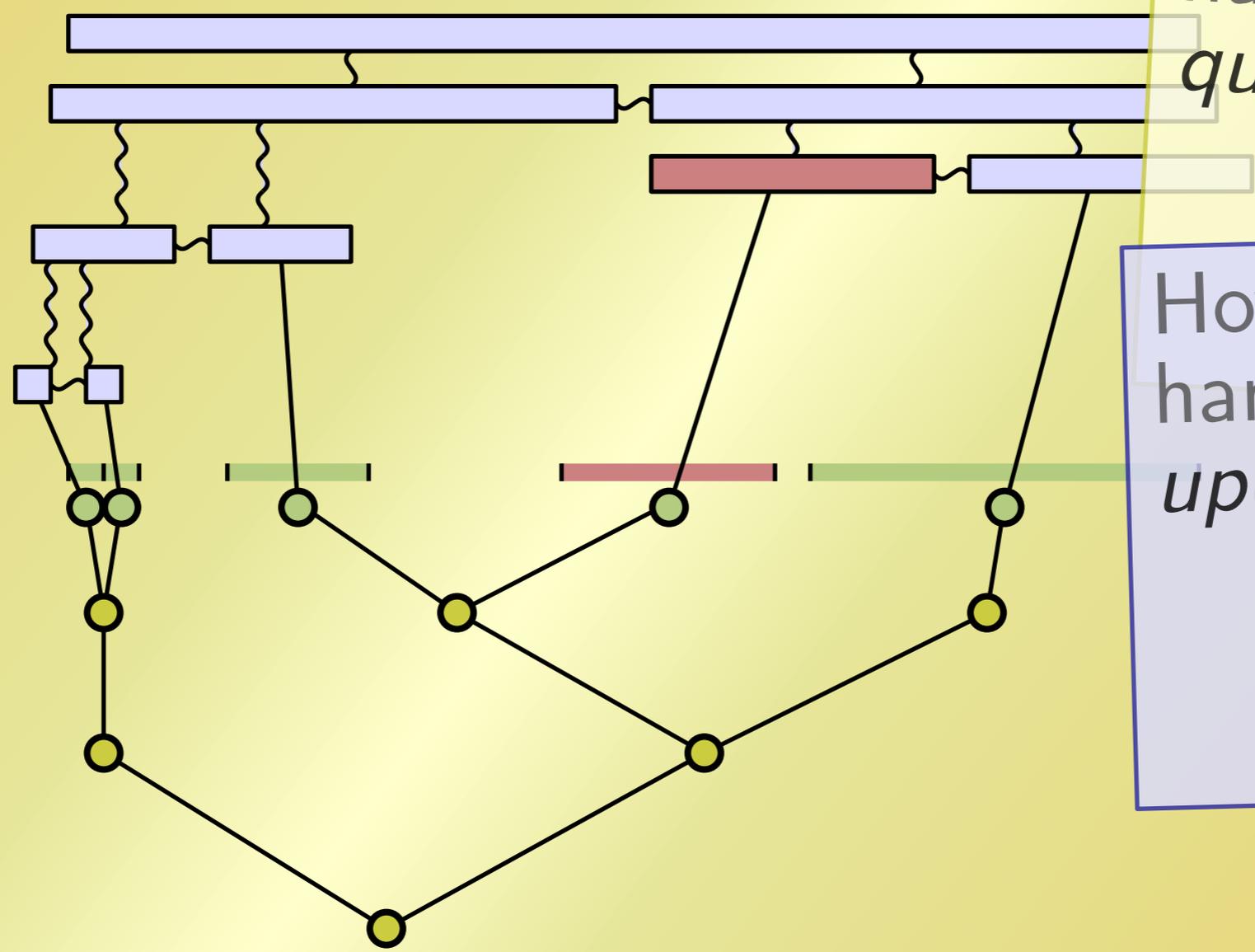
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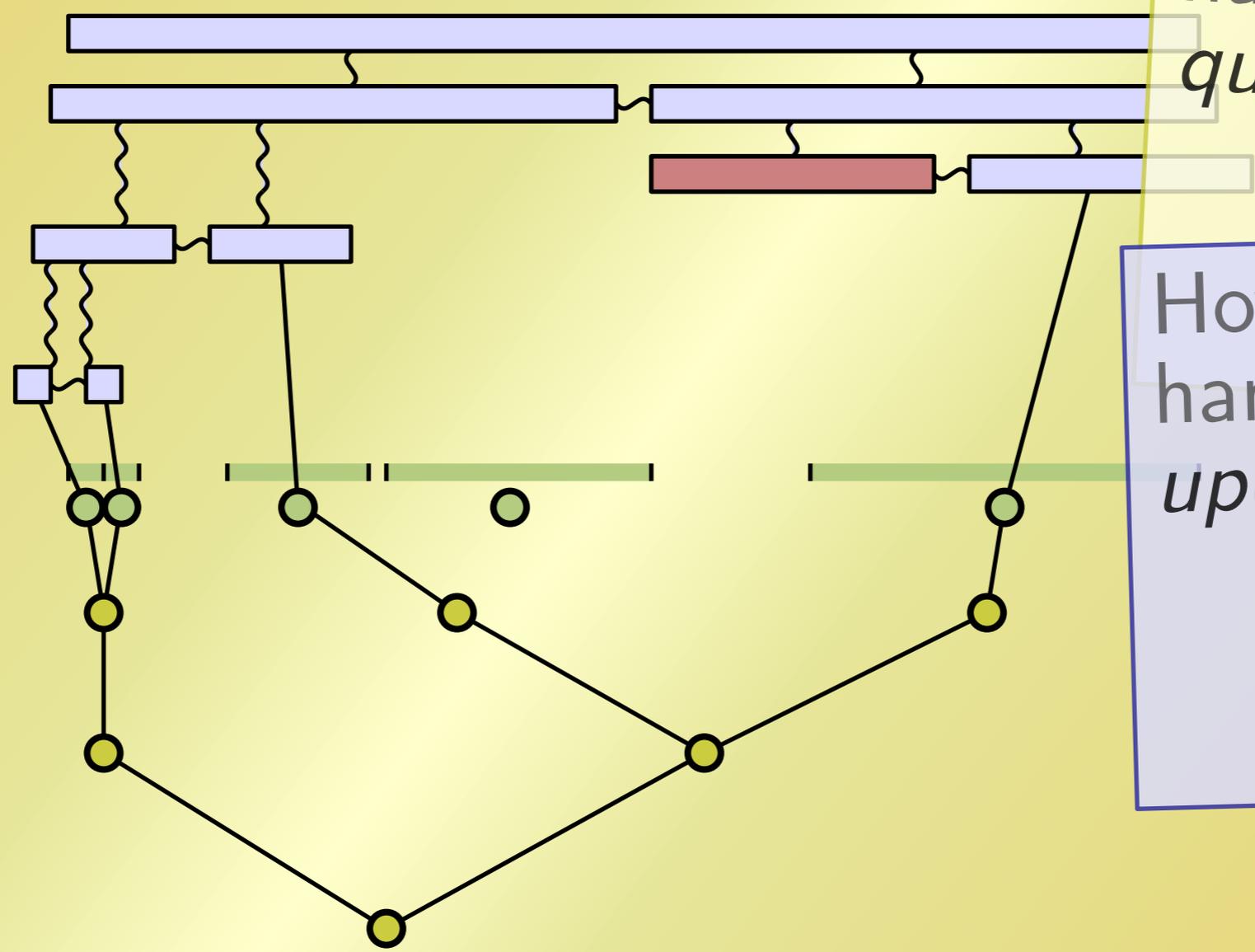
How do we handle an update?





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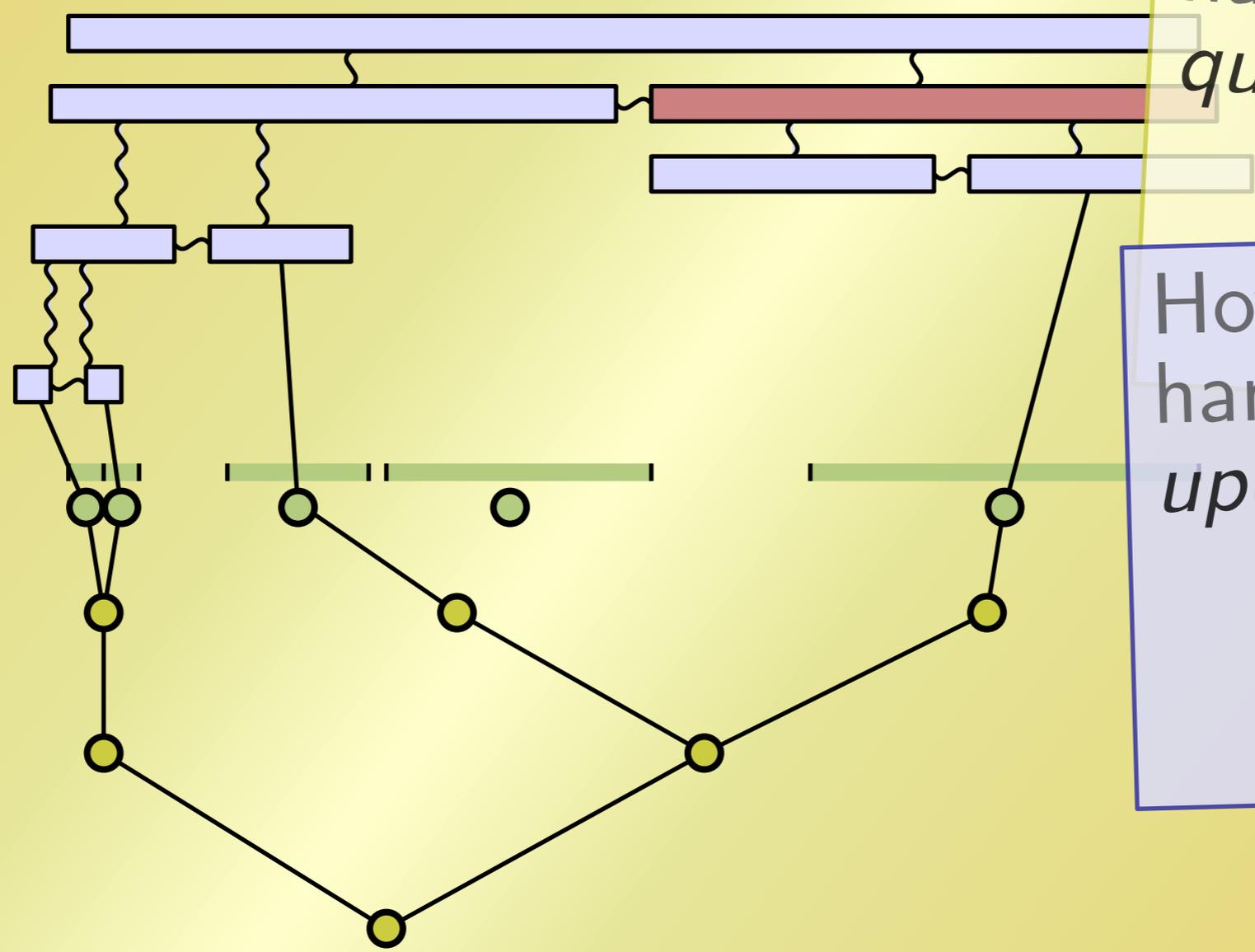
How do we handle an update?





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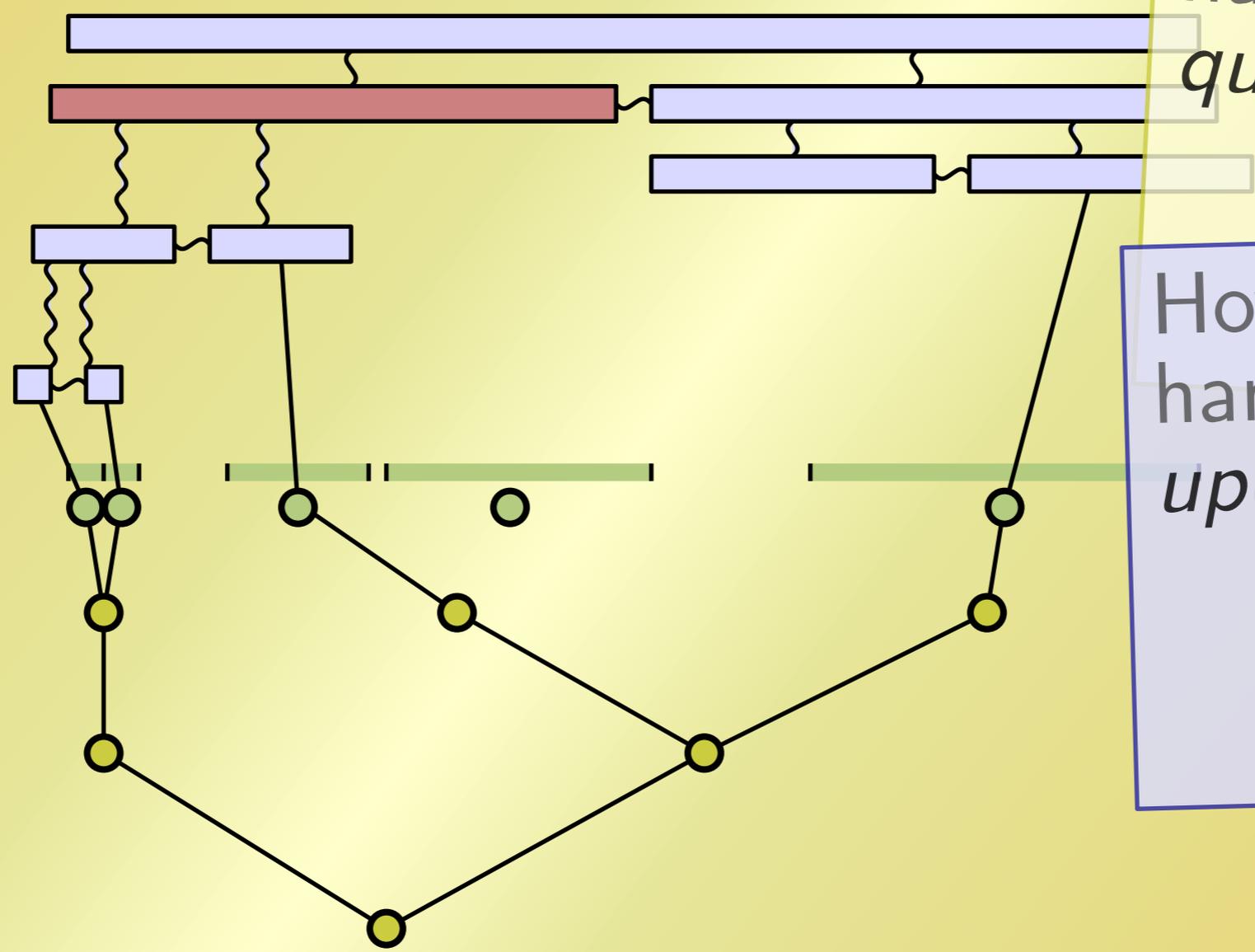
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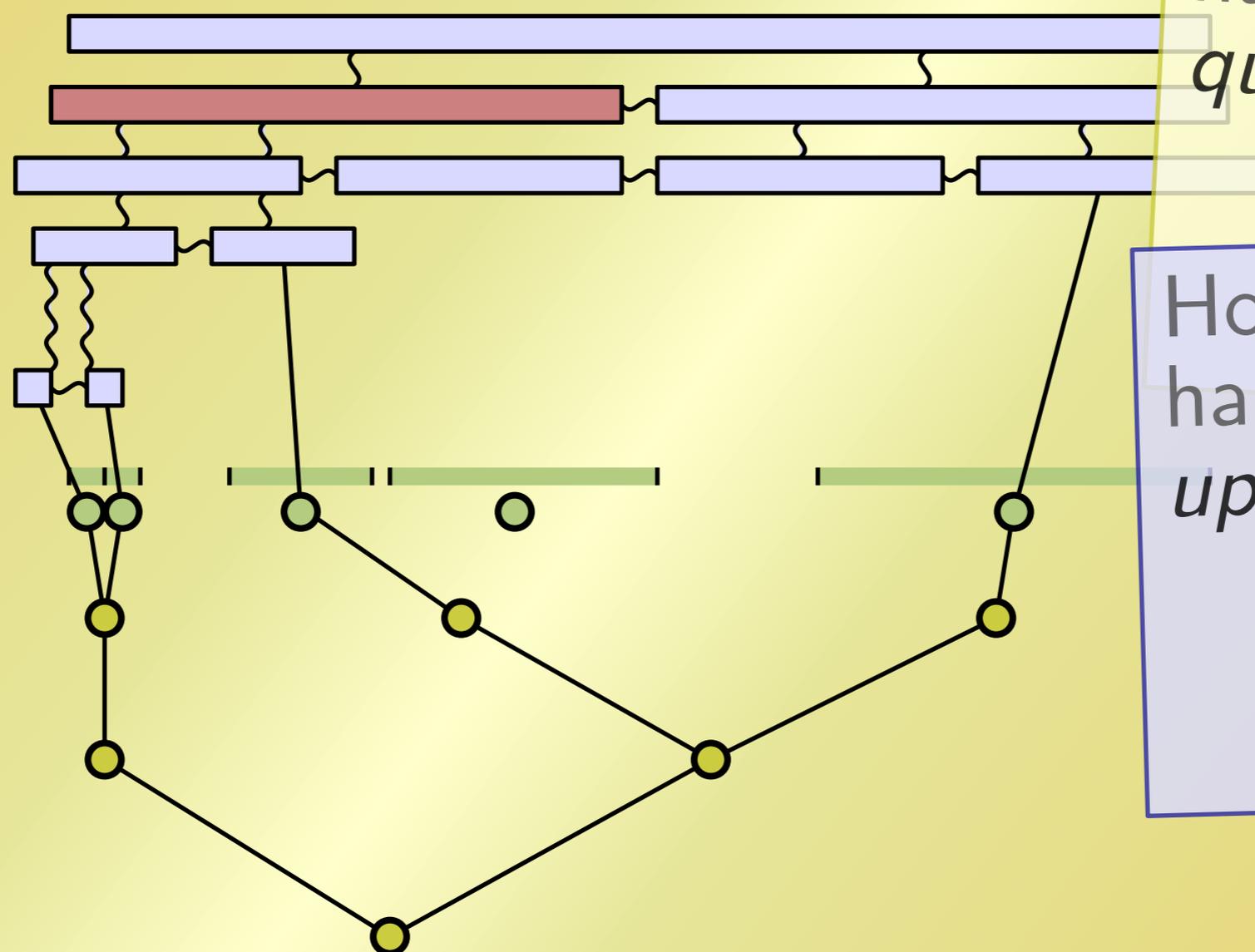




How do we handle a query?

How do we handle an update?





How do we handle a *query*?

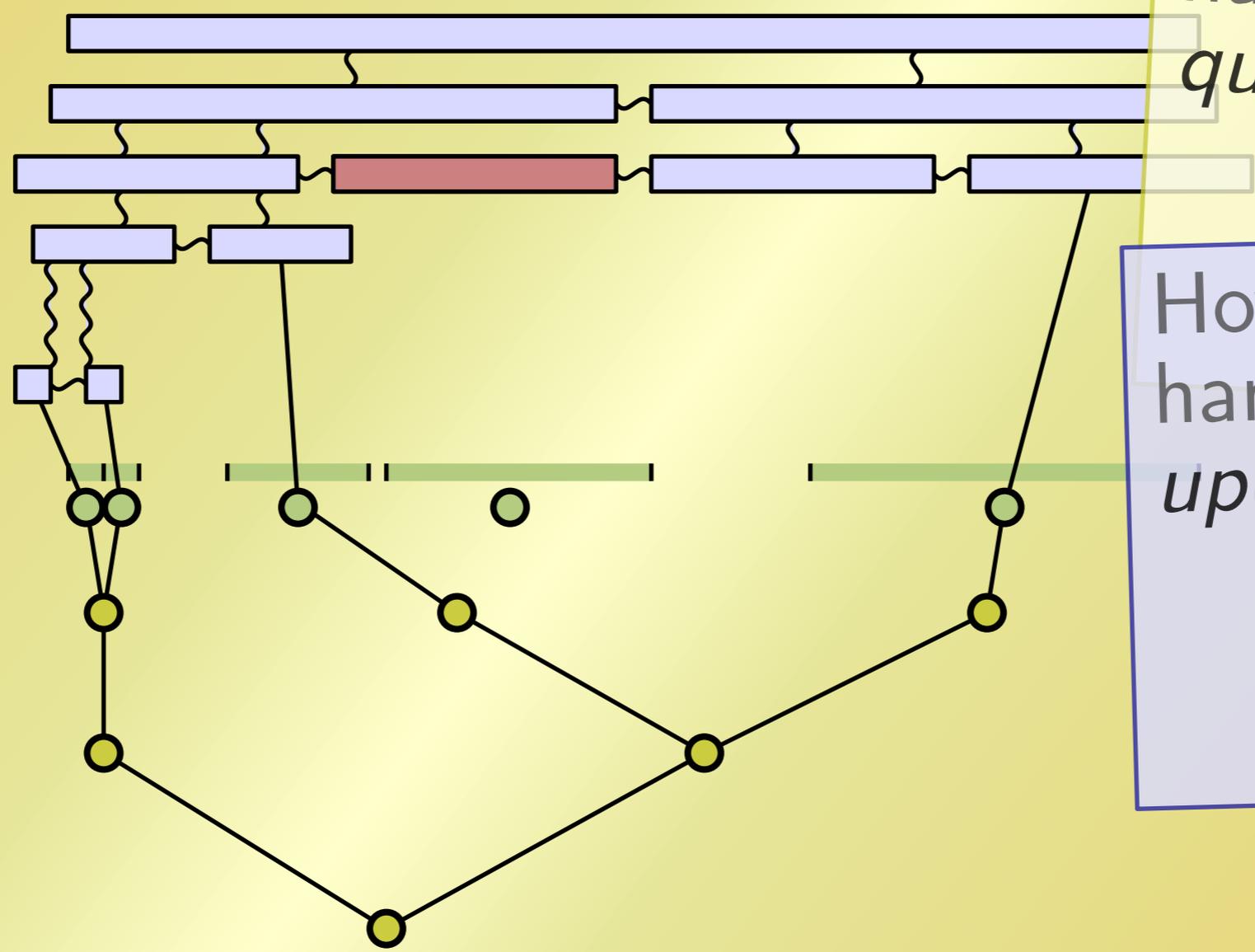
How do we handle an *update*?





How do we handle a *query*?

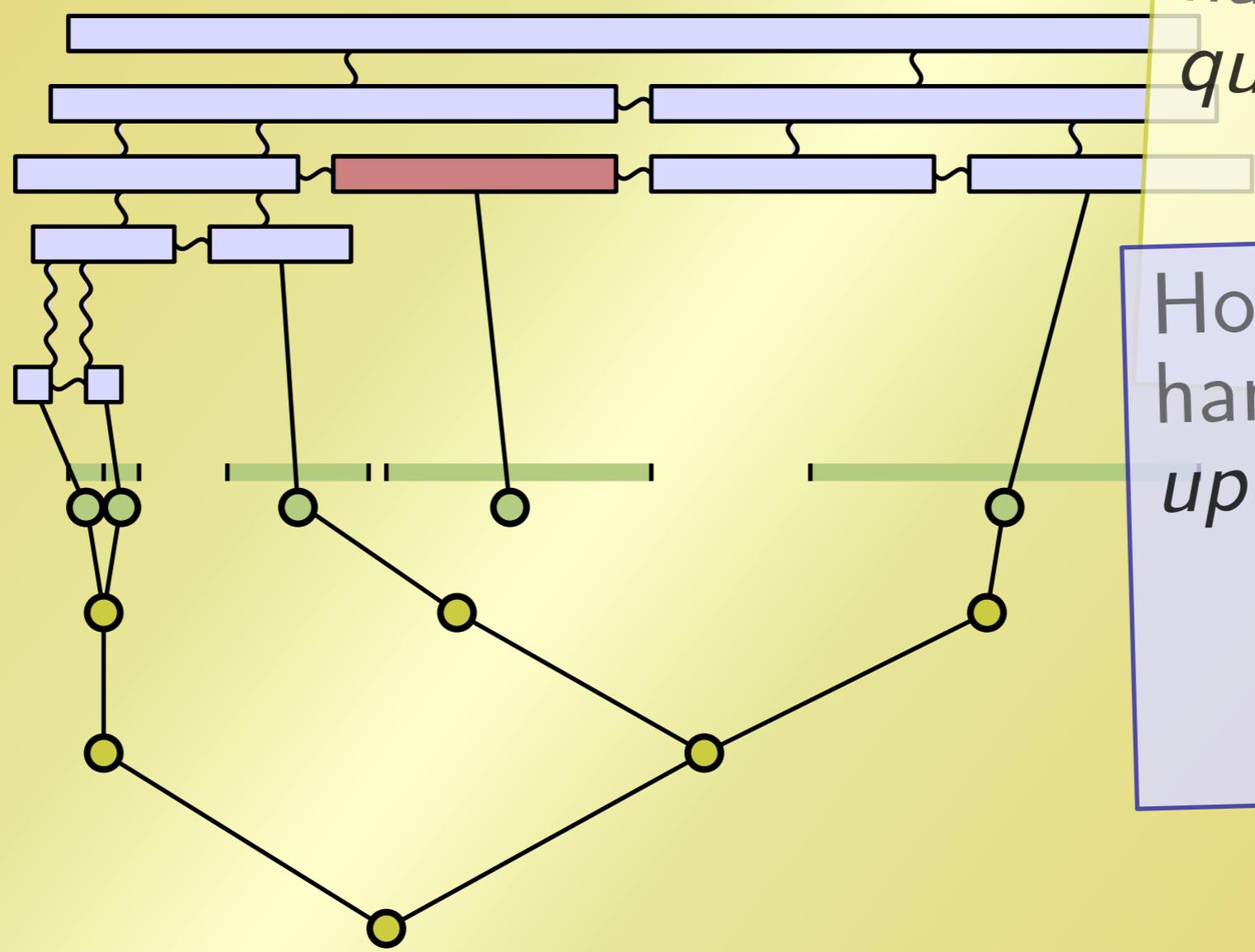
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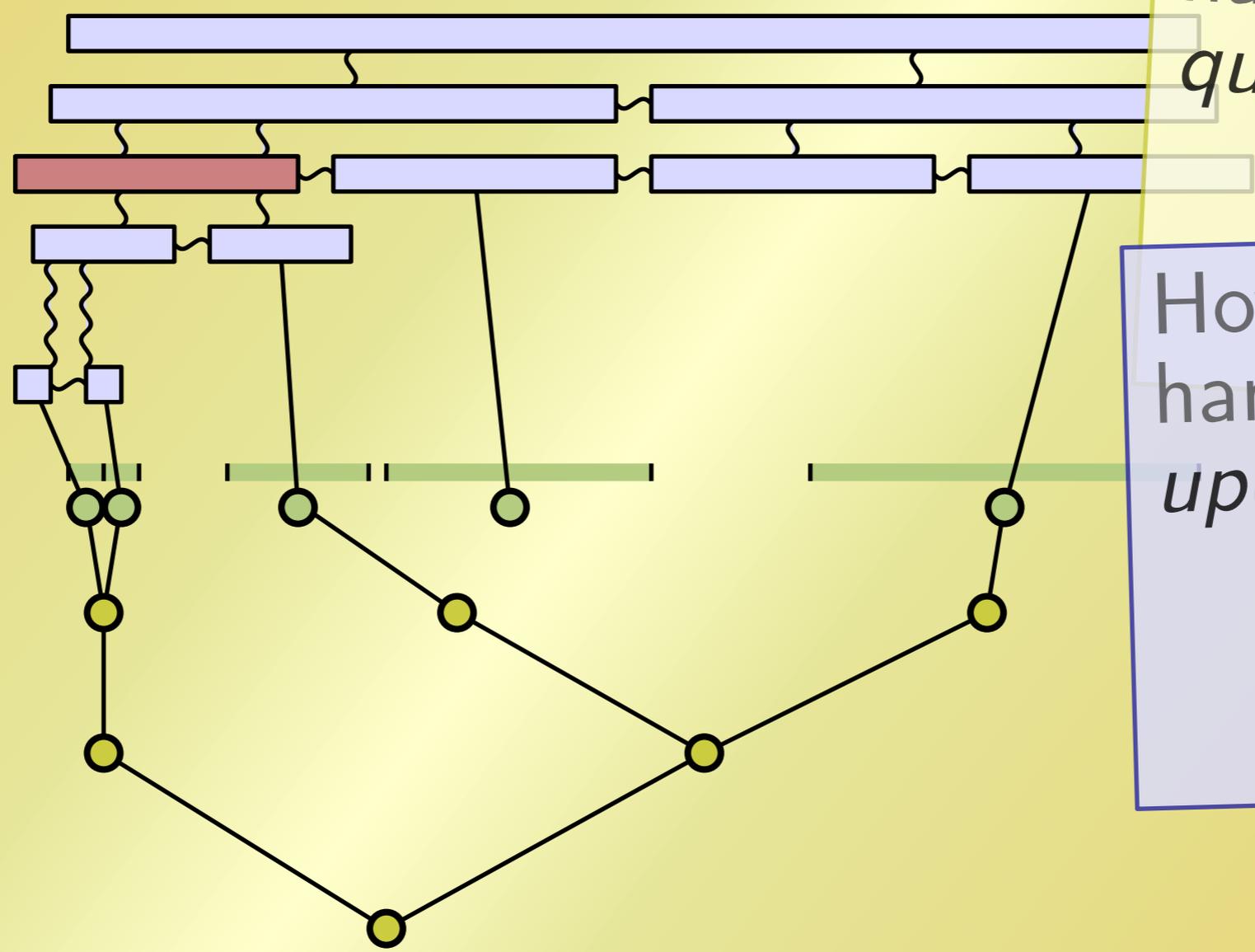
How do we handle an update?

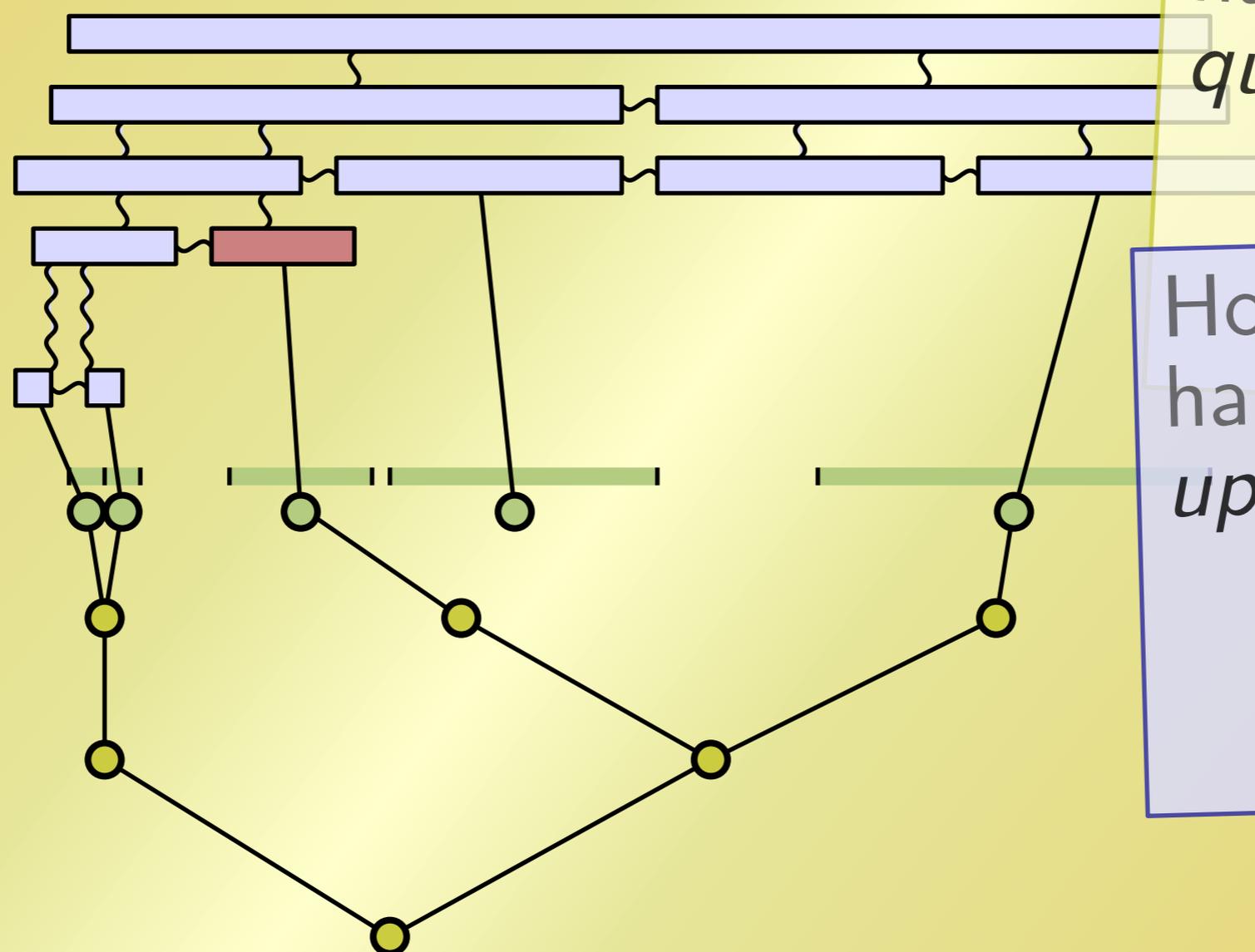




How do we handle a *query*?

How do we handle an *update*?





How do we handle a *query*?

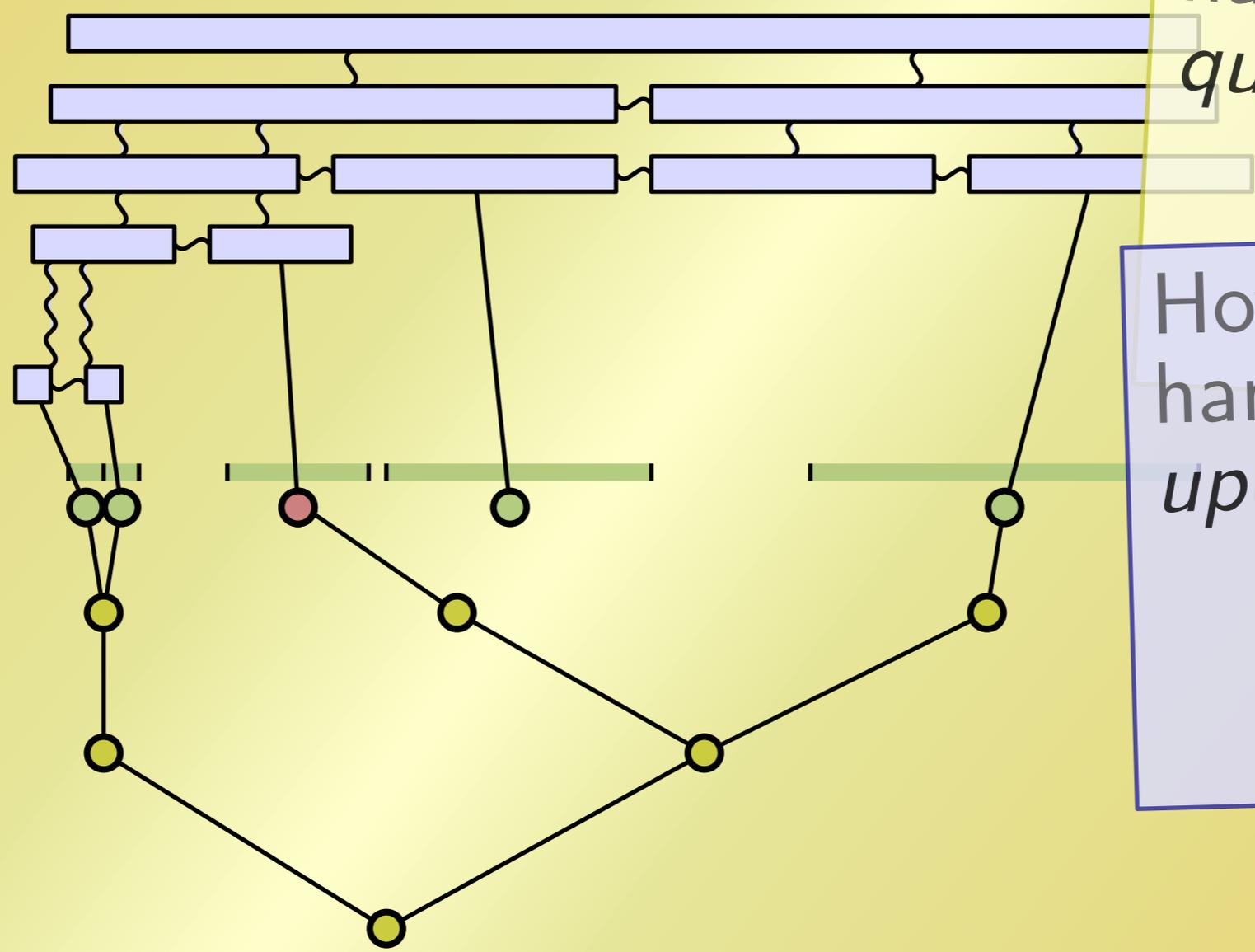
How do we handle an *update*?





How do we handle a *query*?

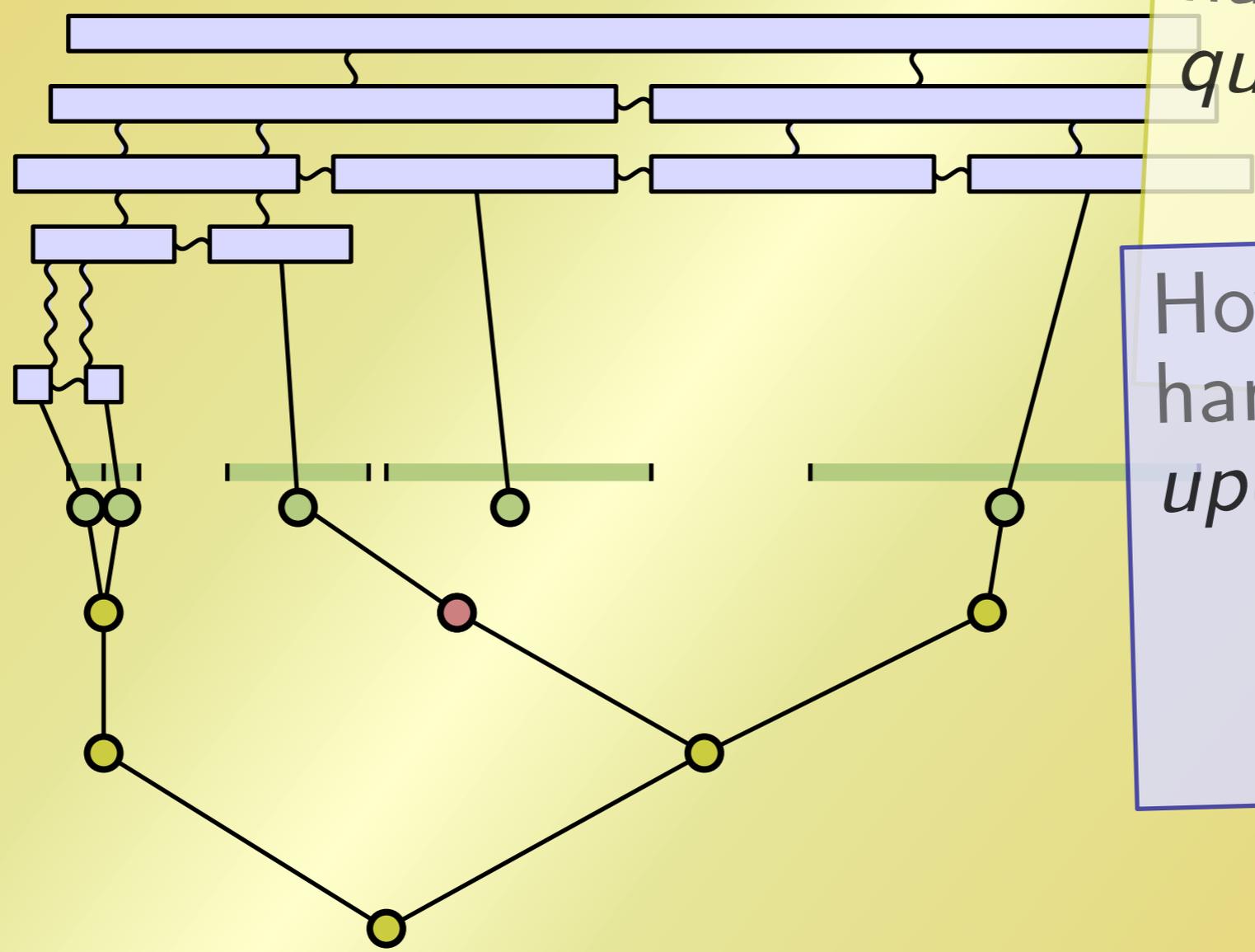
How do we handle an *update*?





How do we handle a *query*?

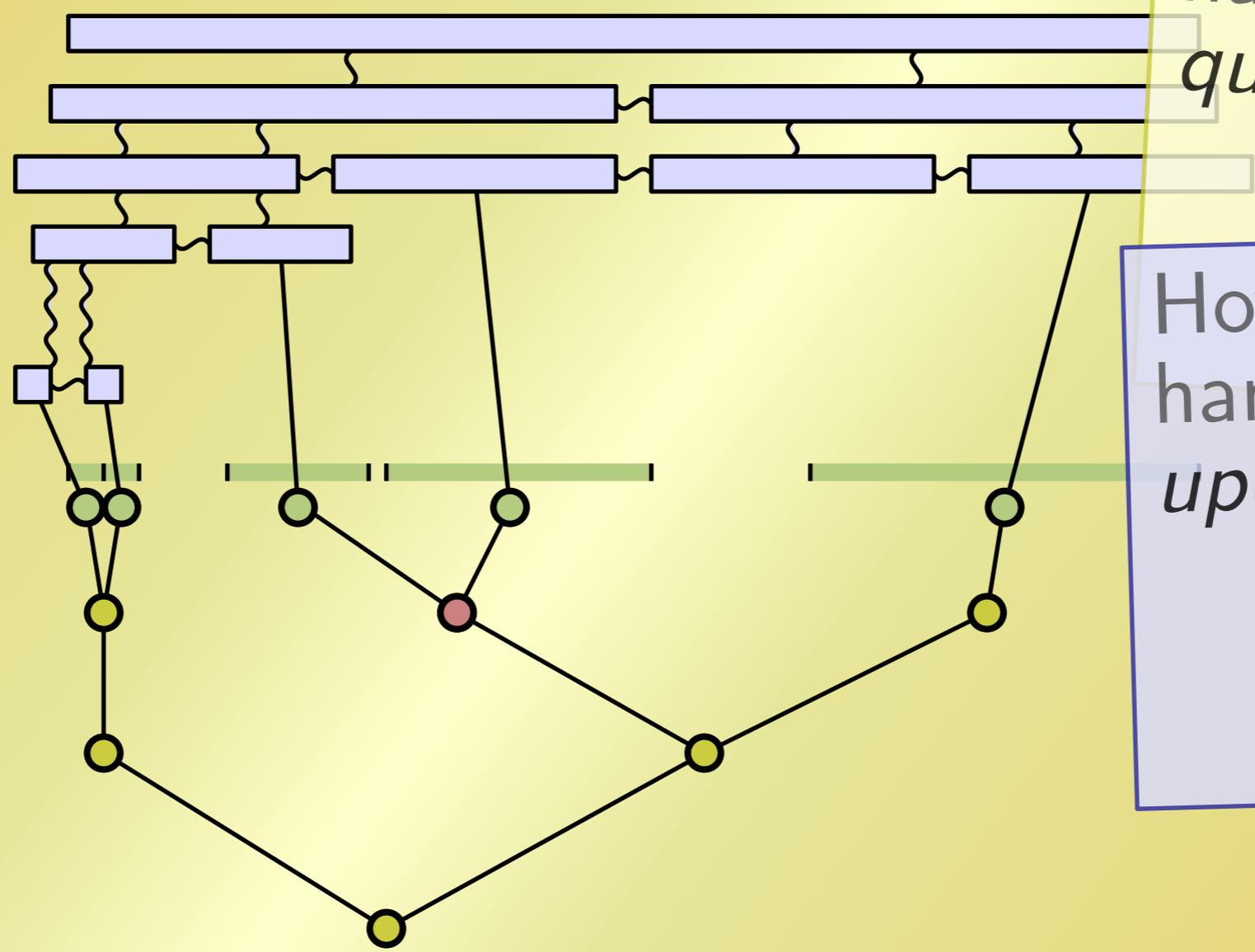
How do we handle an *update*?





How do we handle a *query*?

How do we handle an *update*?





TECHNICAL DETAILS: 2 DIMENSIONS



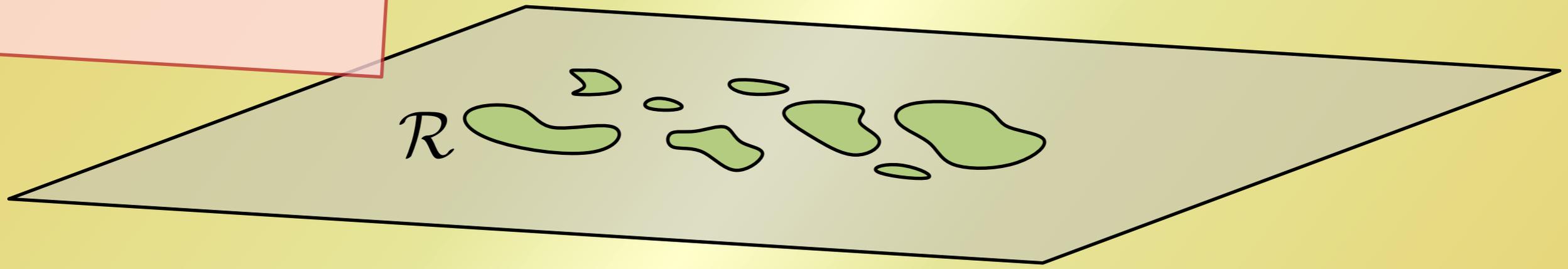


In \mathbb{R}^2 , we
would like to
use a similar
strategy.





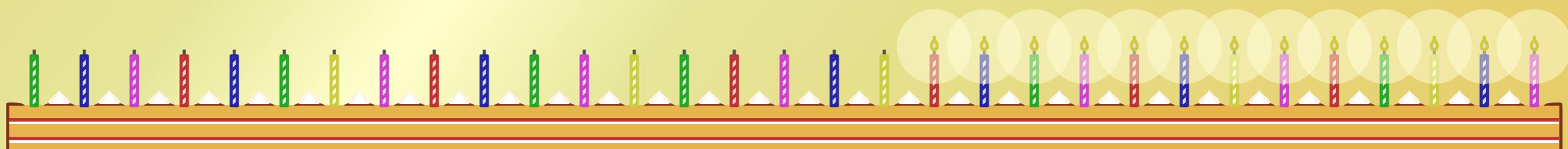
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SPACE TREE

\mathcal{R}



In \mathbb{R}^2 , we would like to use a similar strategy.

$o(\log n)$ Updates

SPACE TREE

\mathcal{R}

DATA TREE

$O(\log n)$ Queries



In \mathbb{R}^2 , we would like to use a similar strategy.

$o(\log n)$ Updates

SPACE
TREE

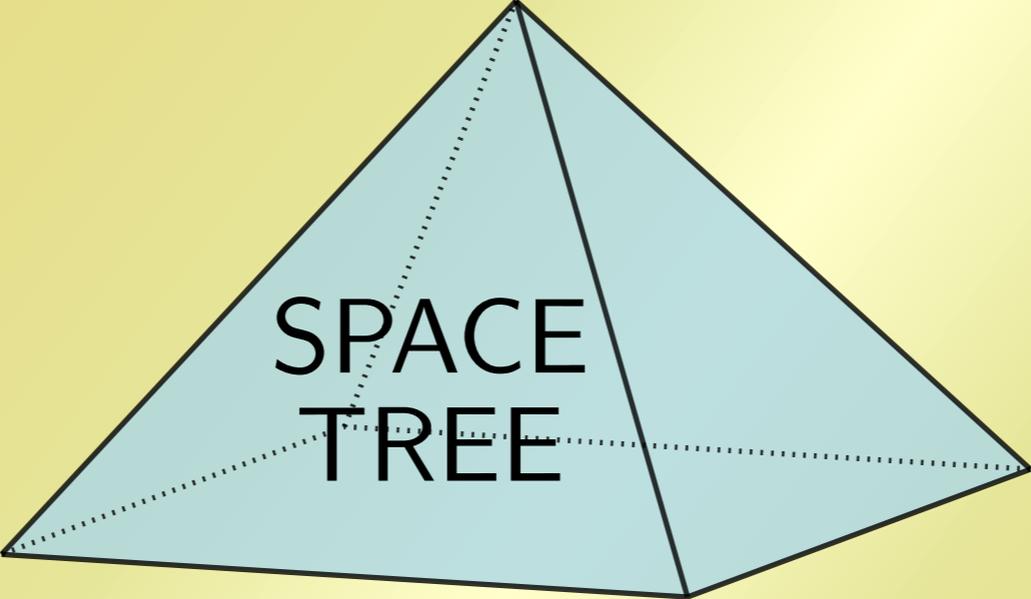
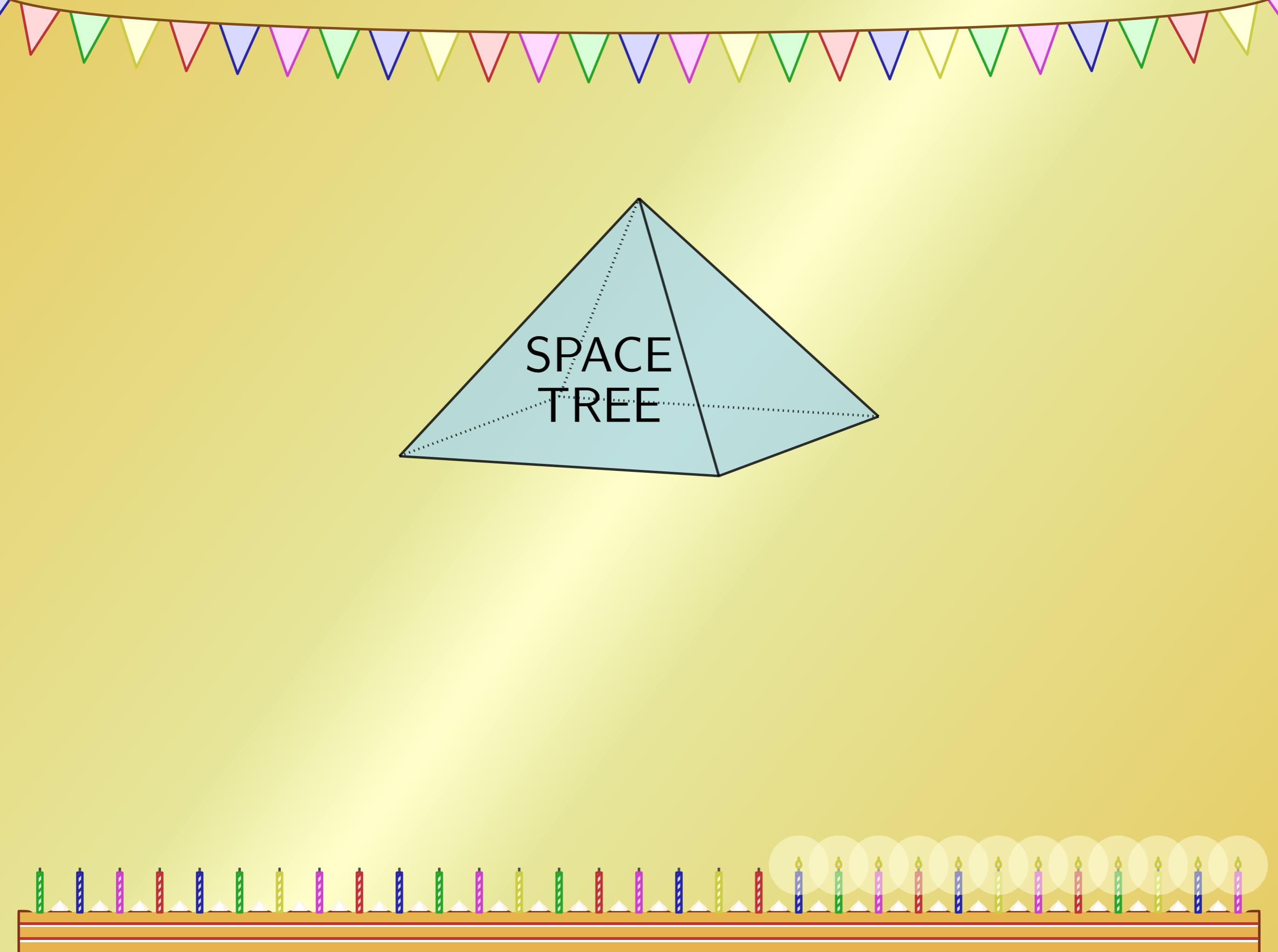
\mathcal{R}

DATA
TREE

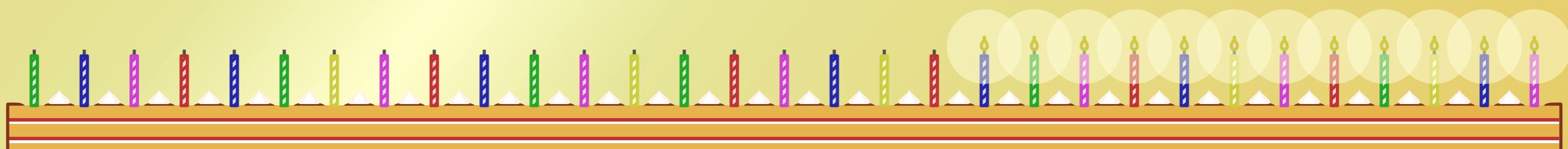
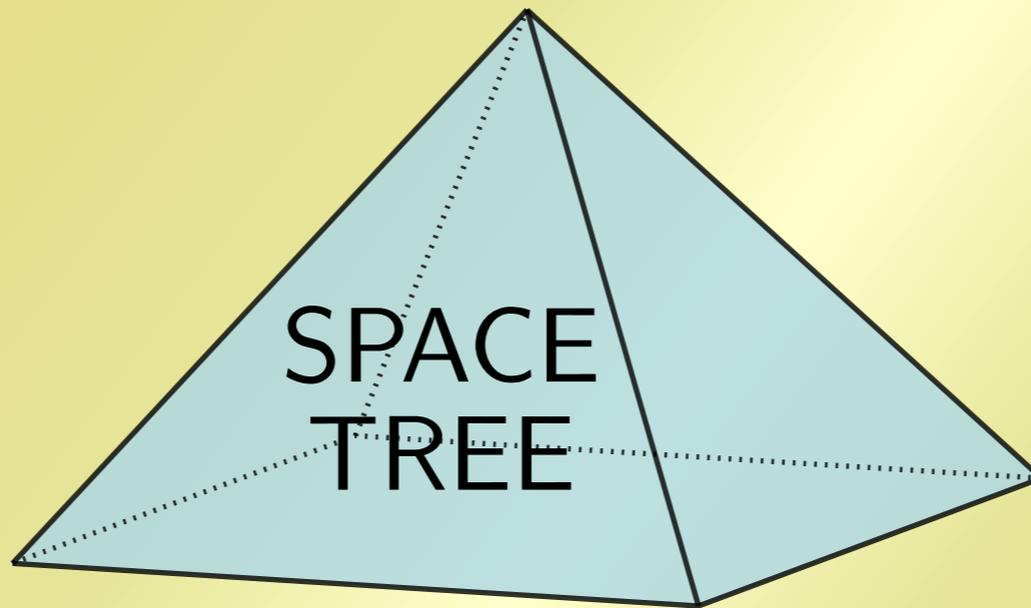
**NAIVE
THOUGHT**
How hard can
it be?

$O(\log n)$ Queries

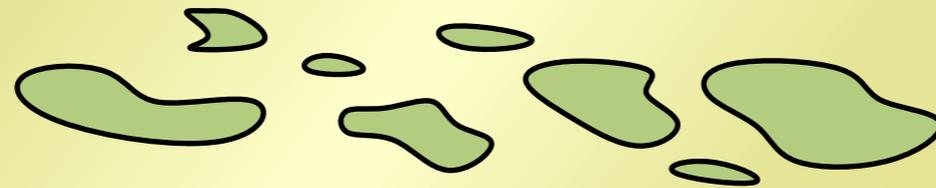




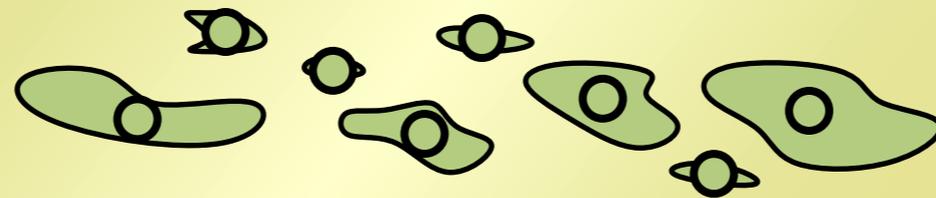
Quadtrees
also exist in 2
dimensions!



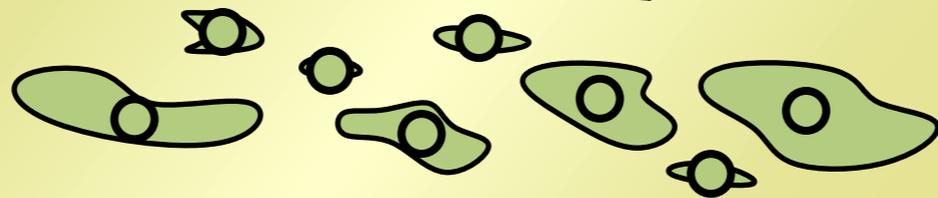
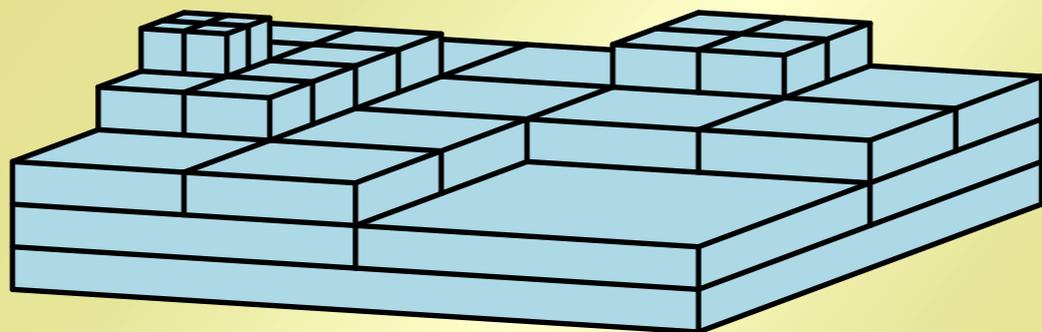
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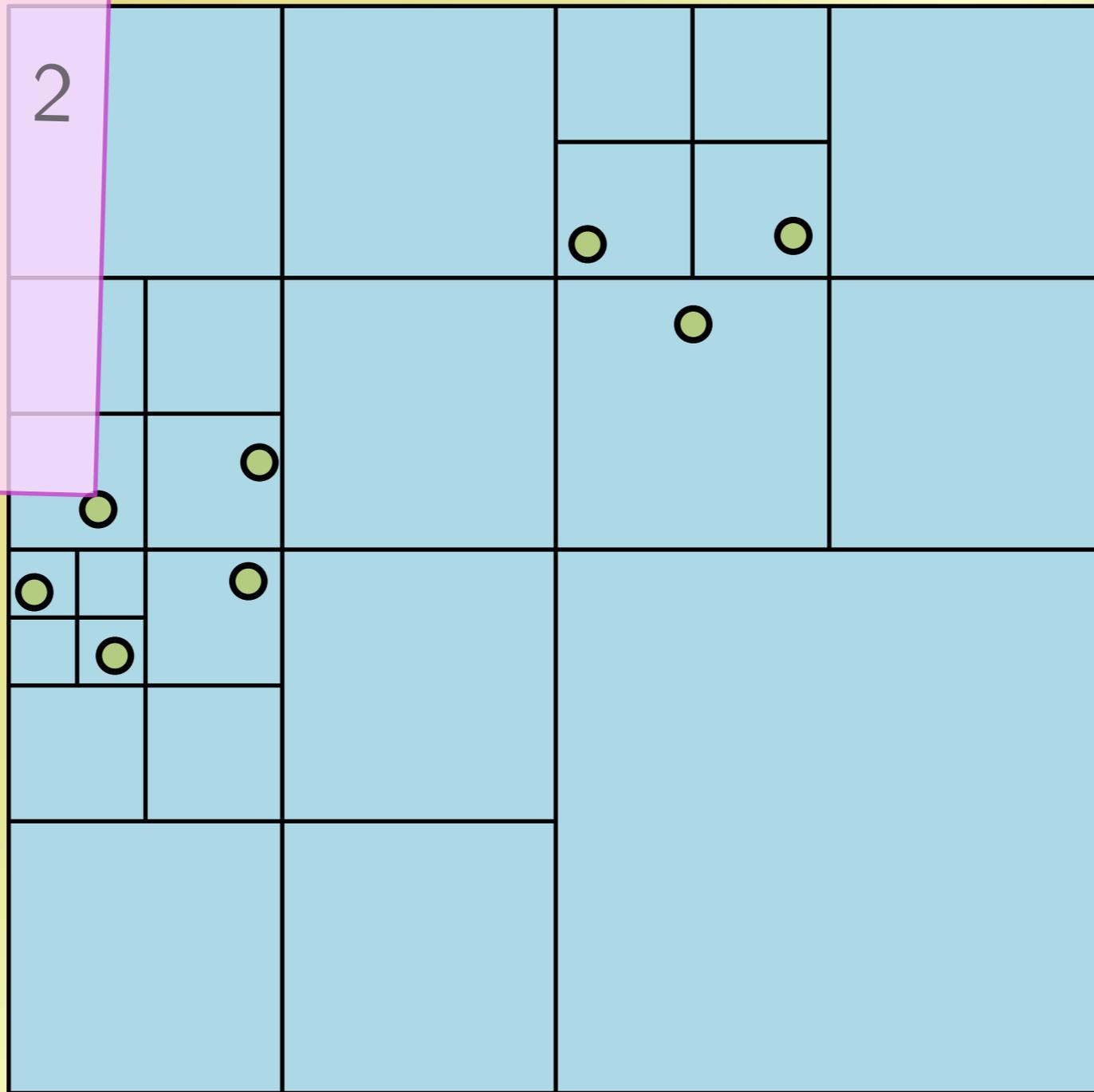
Quadtrees
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dimensions!



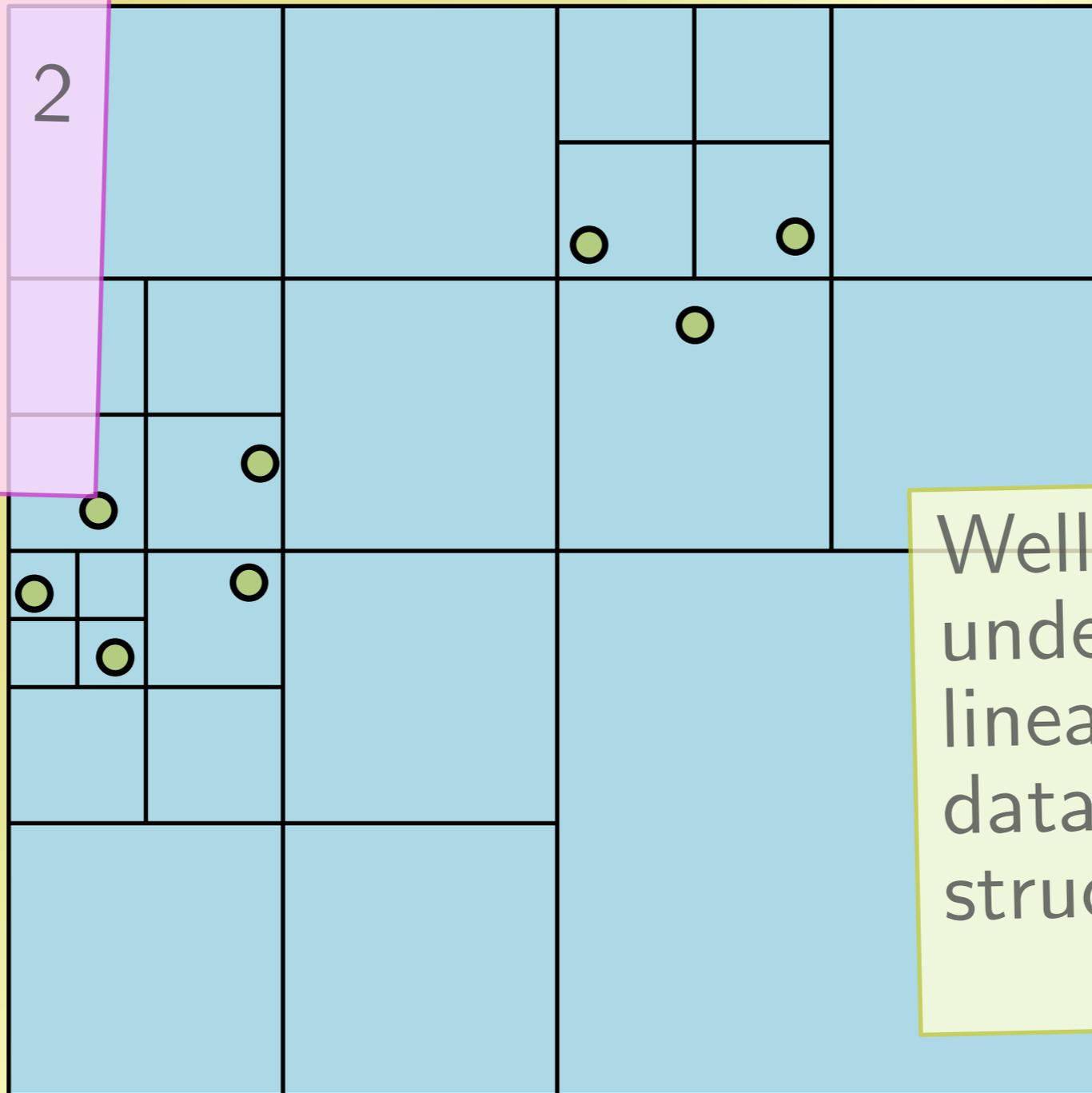
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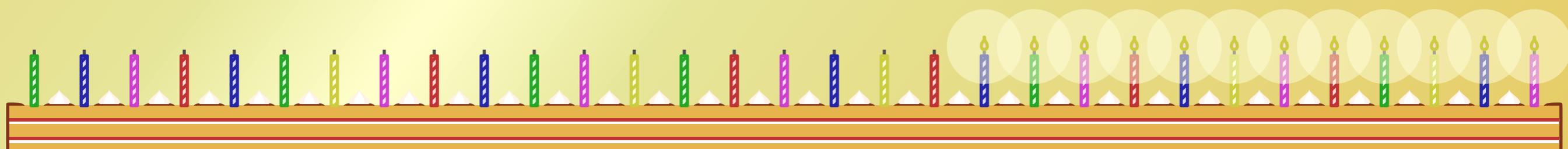
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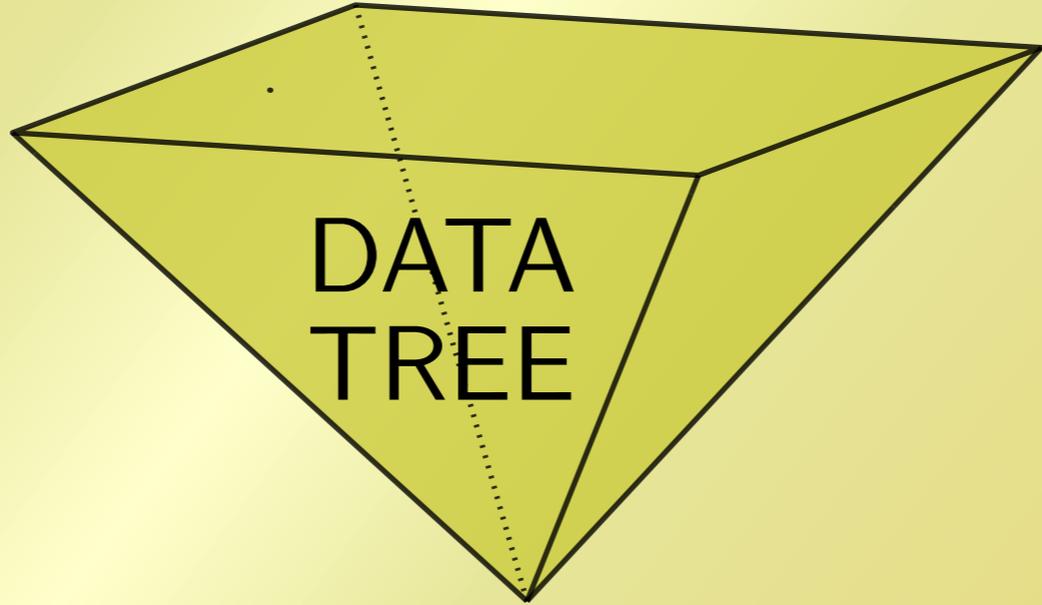
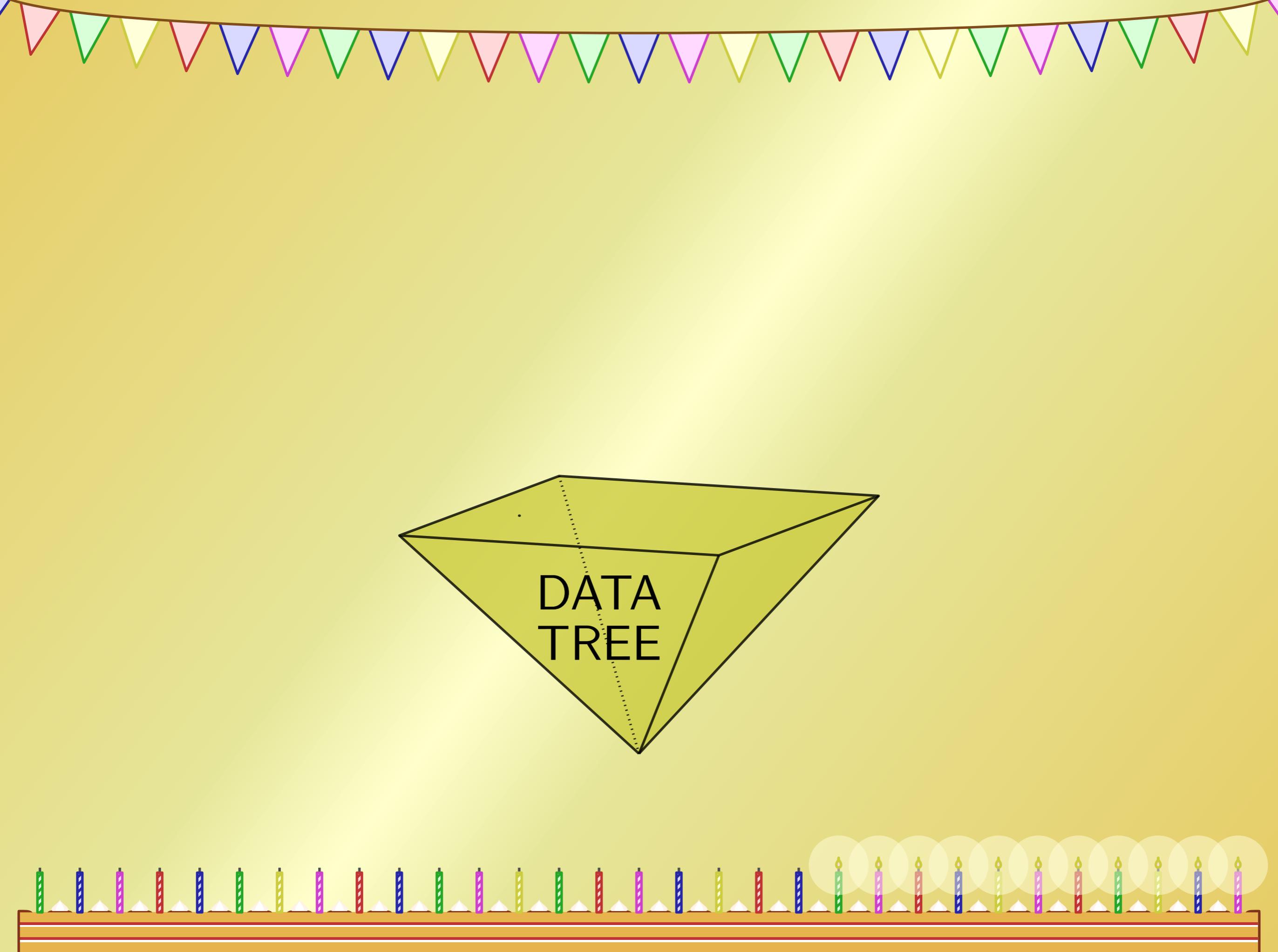


Quadtrees
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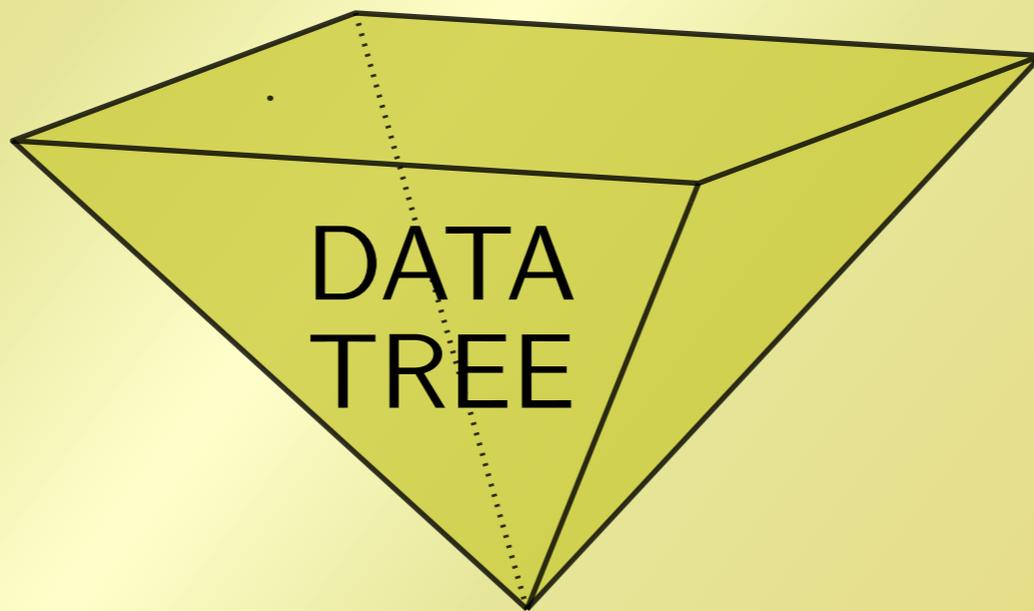
Well
understood,
linear size
data
structure.





DATA
TREE

We still need something for the actual point location.



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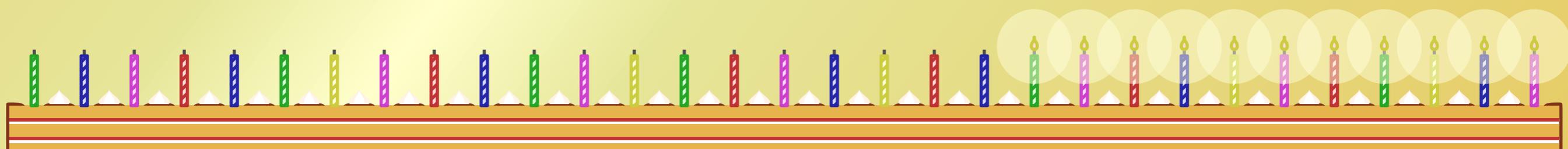
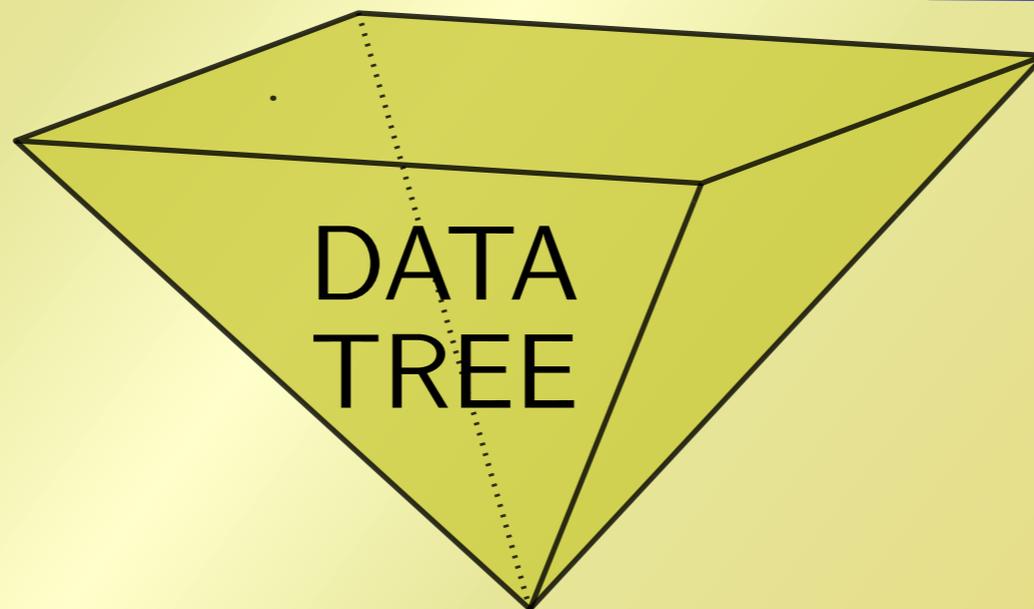
Build existing structure on regions, and use cross pointers as before?



We still need something for the actual point location.

Build existing structure on regions, and use cross pointers as before?

How do regions relate to the quadtree?





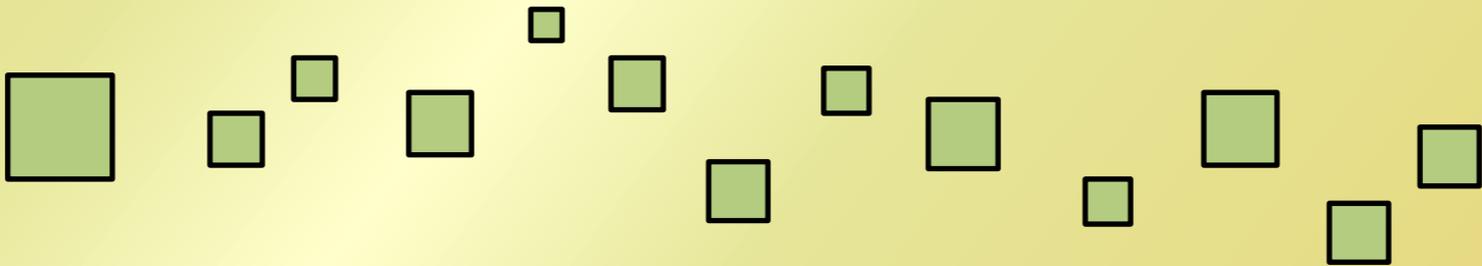
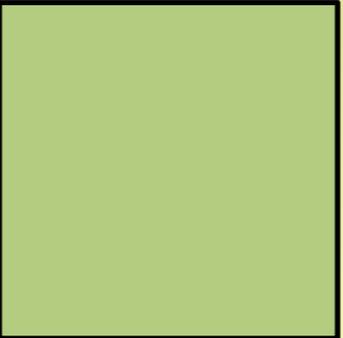
PROBLEM

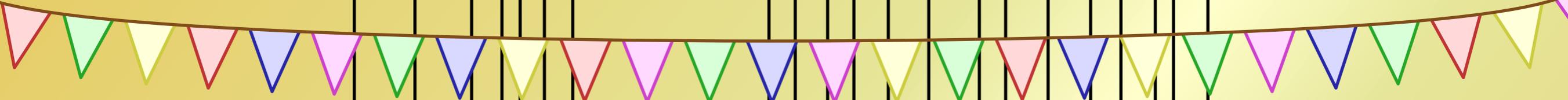
We can't just use any search tree anymore.



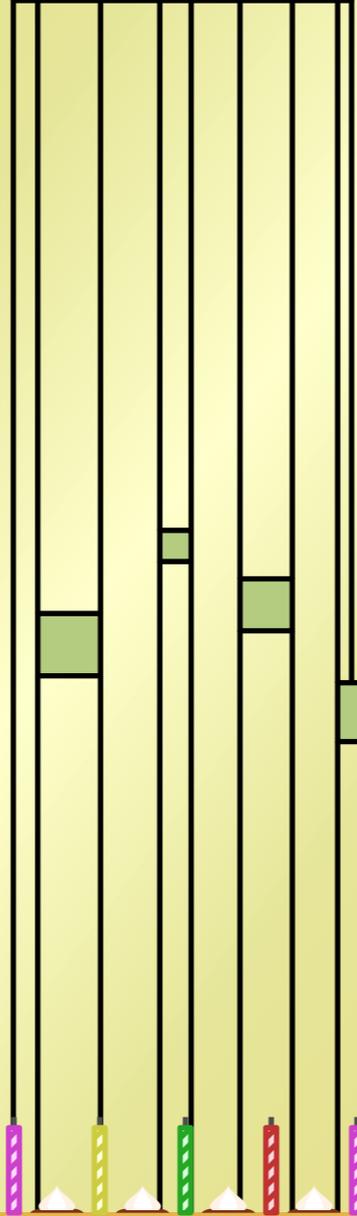
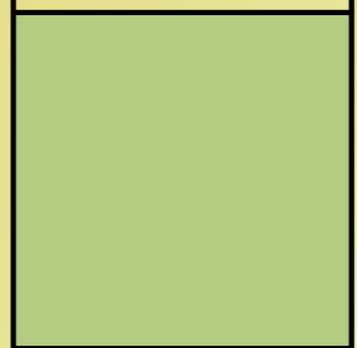


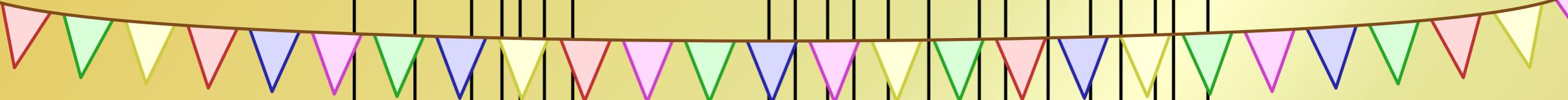
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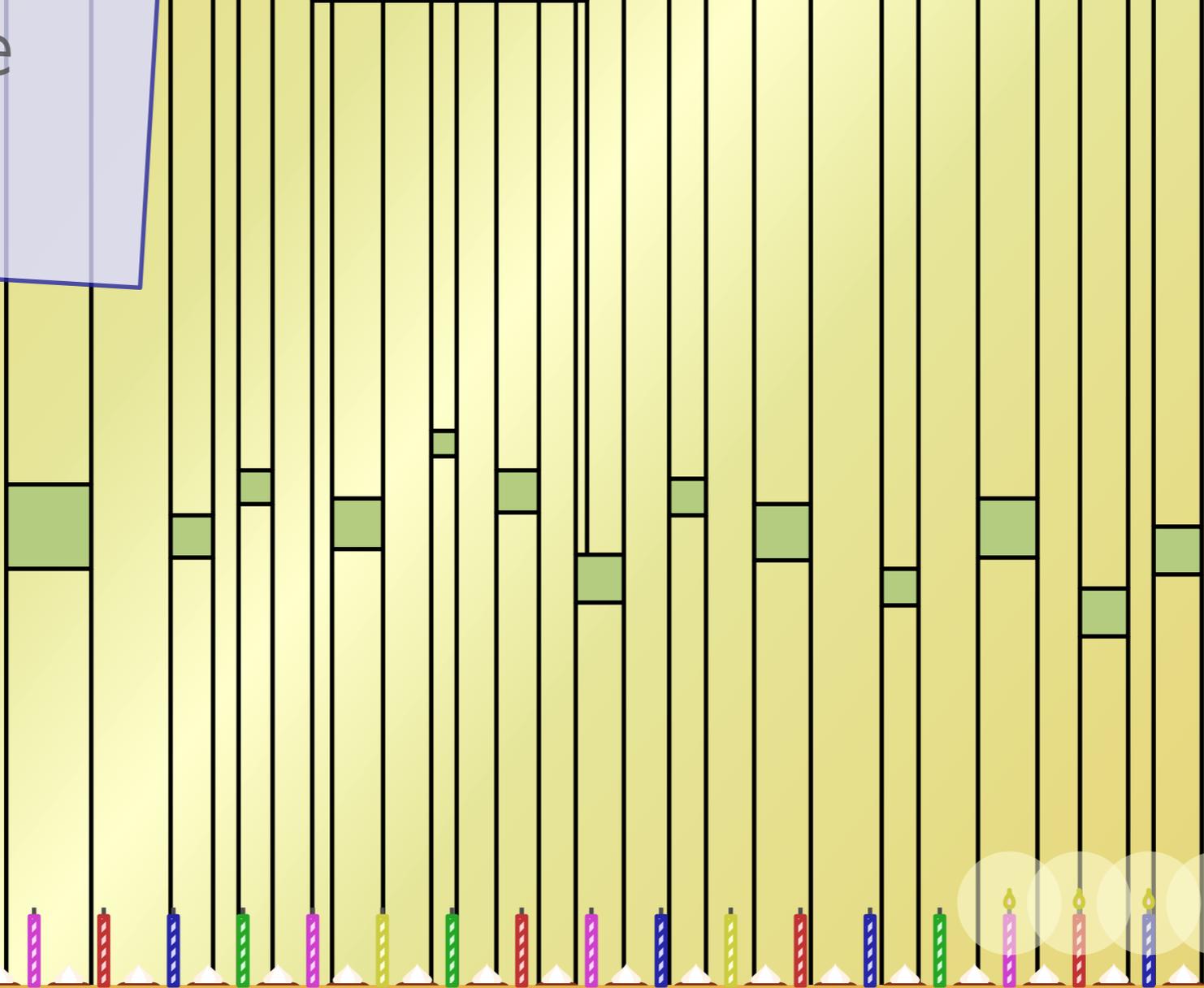
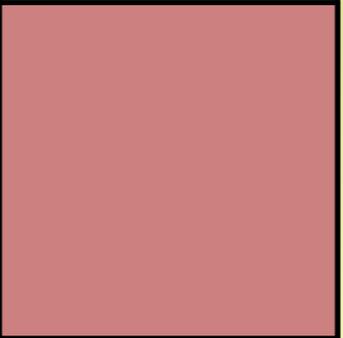


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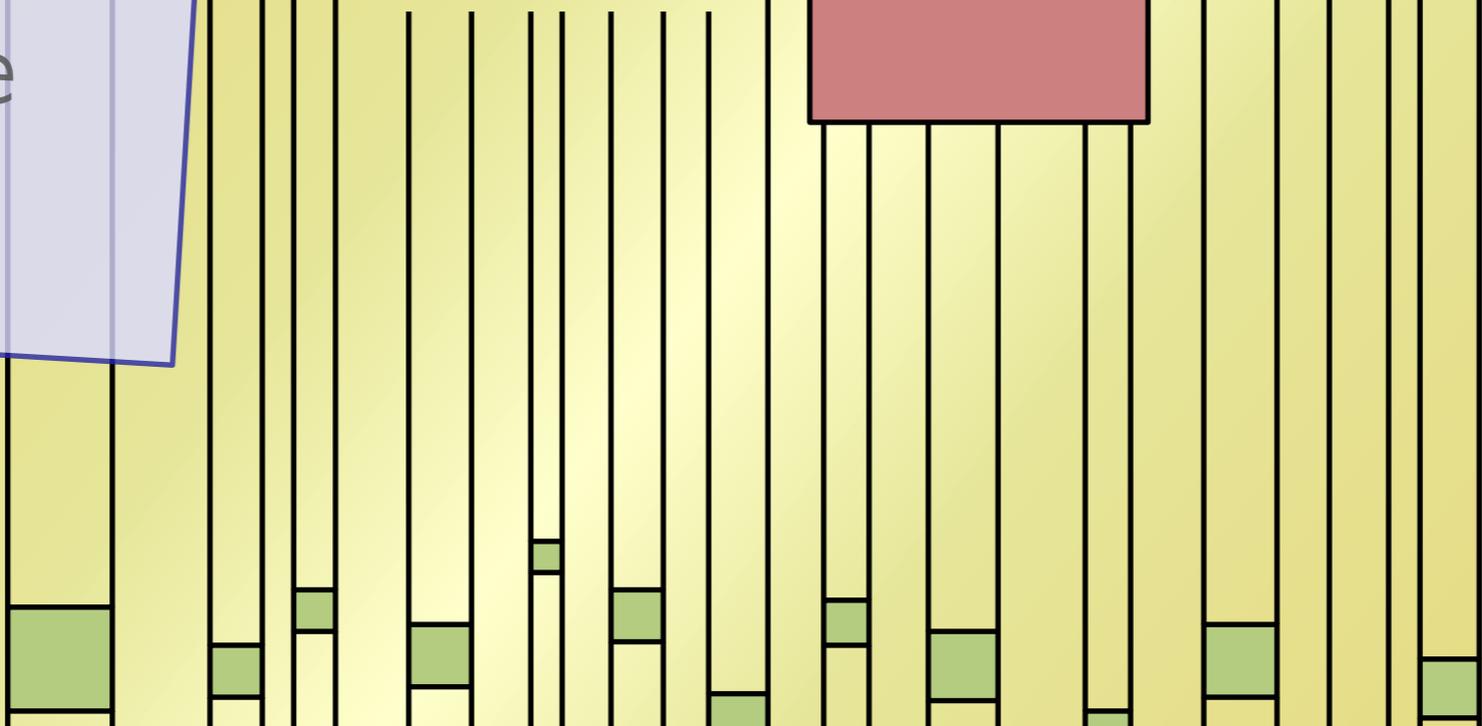
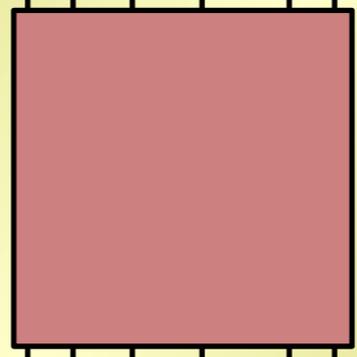


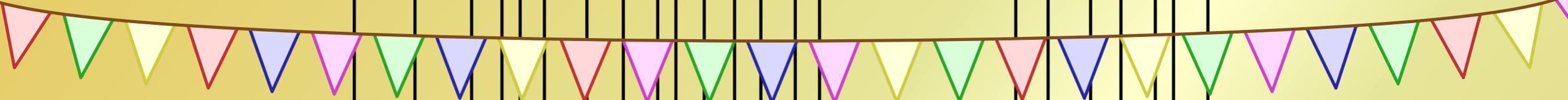
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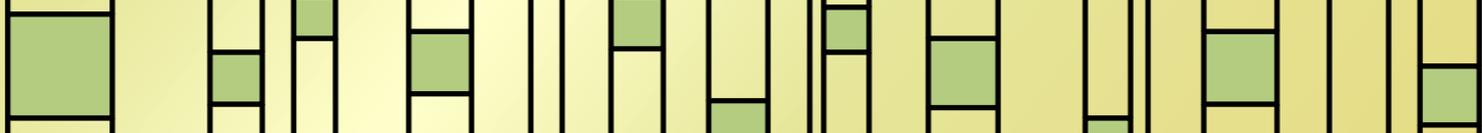
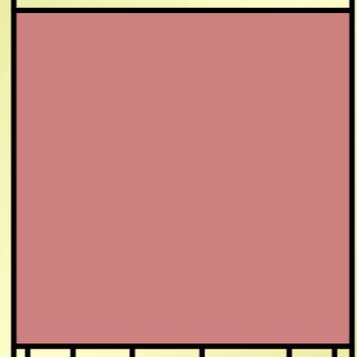


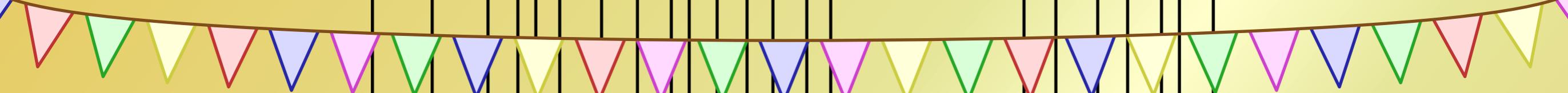
PROBLEM
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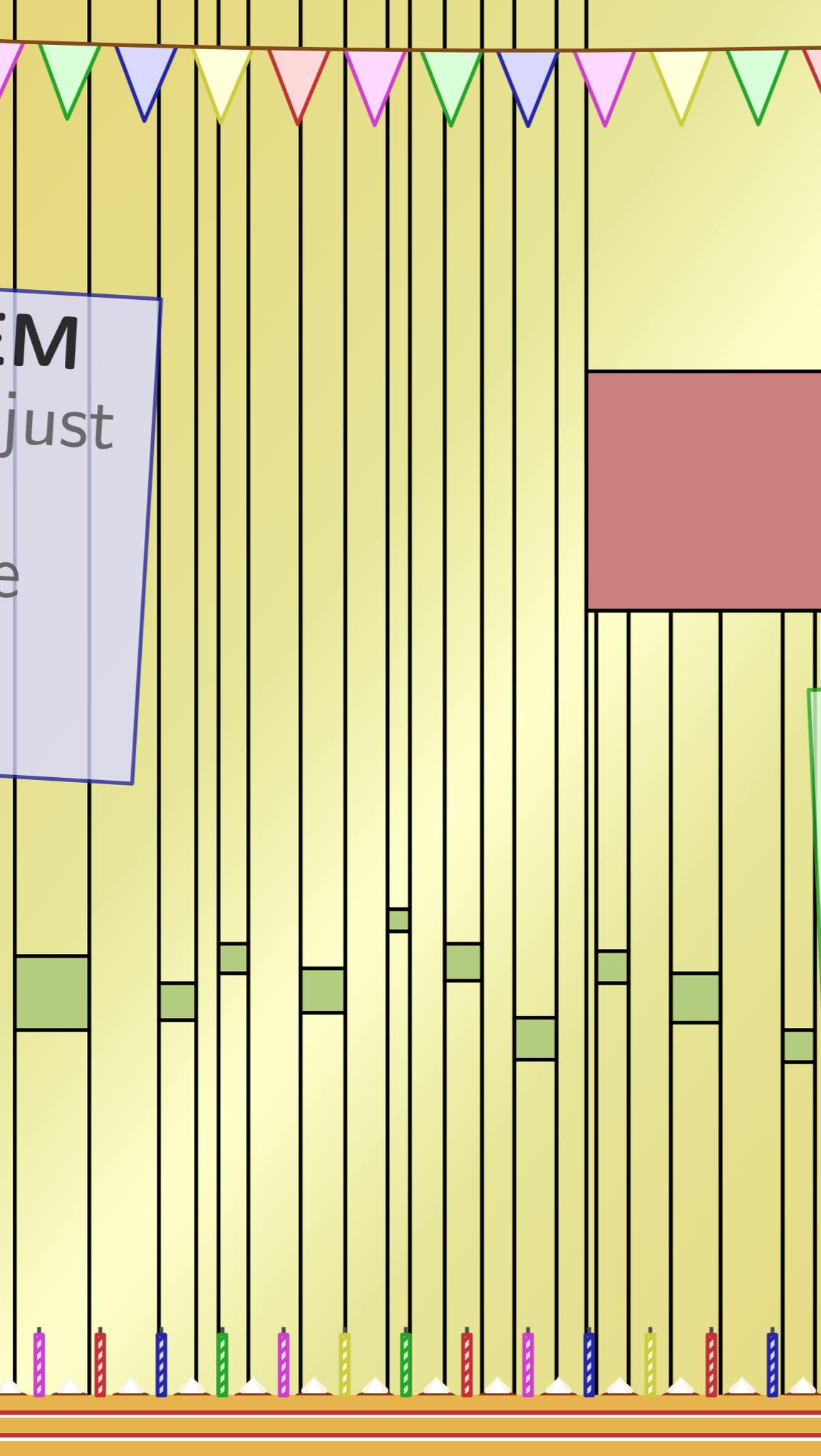
PROBLEM
We can't just use any search tree anymore.





PROBLEM

We can't just use any search tree anymore.



IDEA

Build a search tree *on the quadtree leaves*.

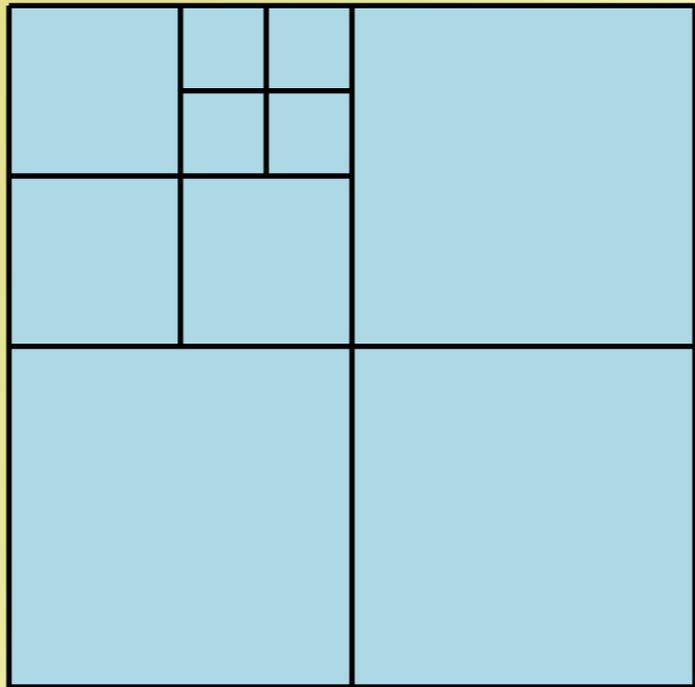




Take another
look at a
quadtree.

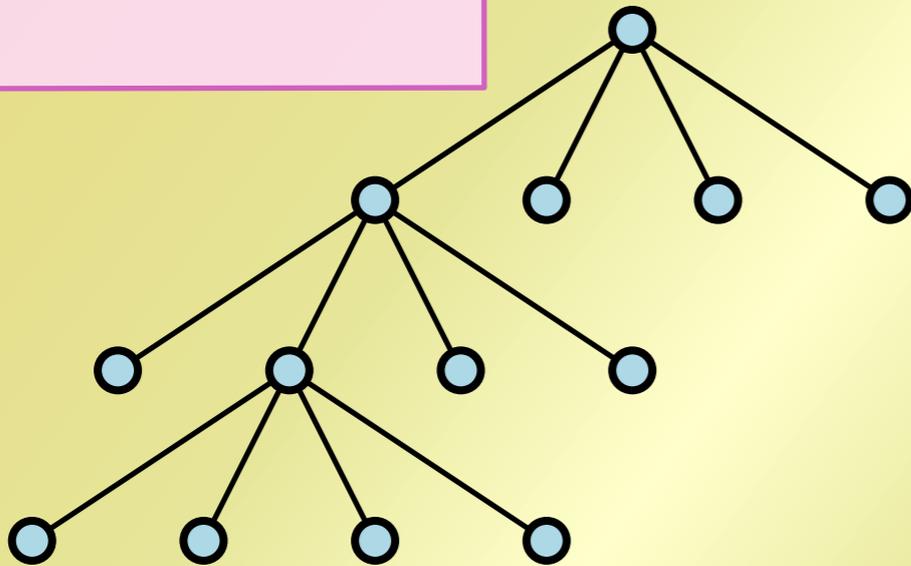
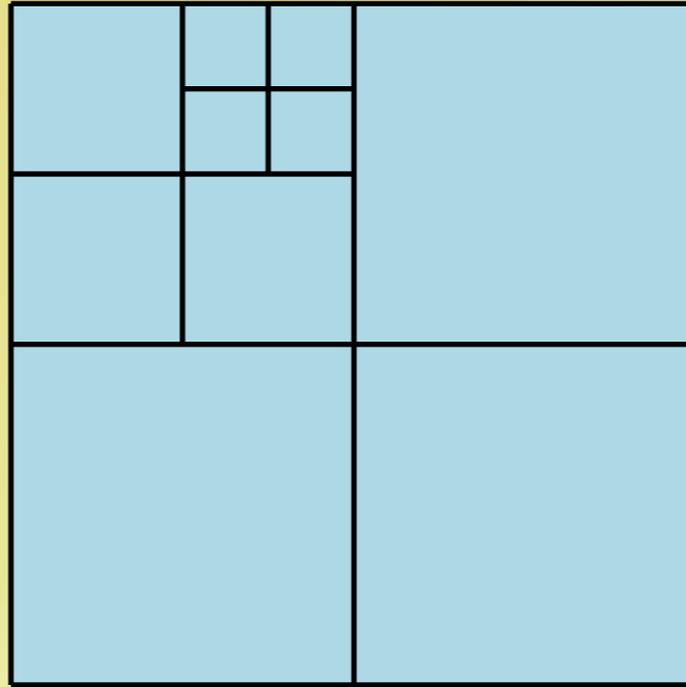


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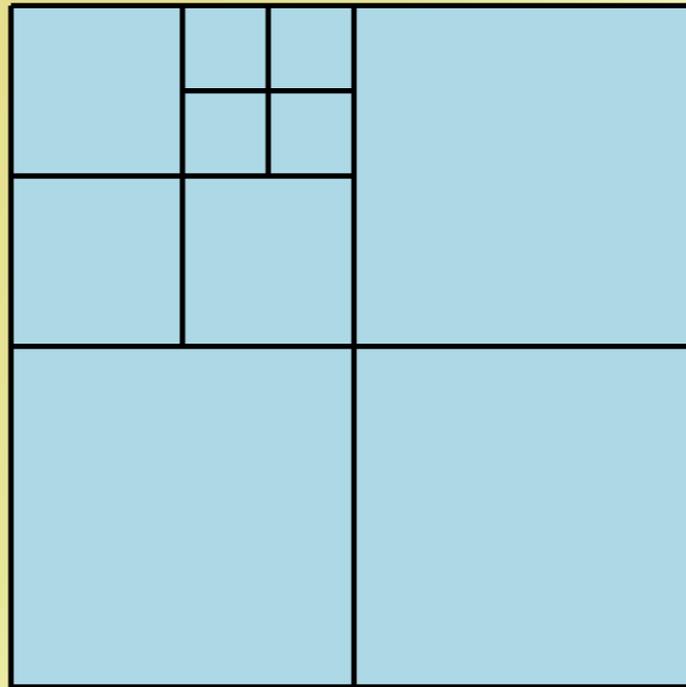
Take another look at a quadtree.

It's a degree 4 tree of potentially linear height.

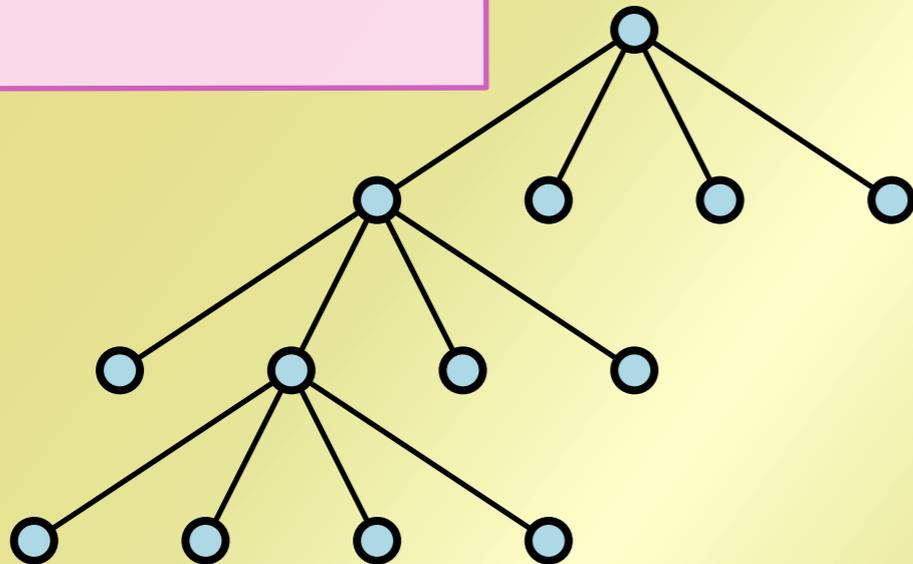


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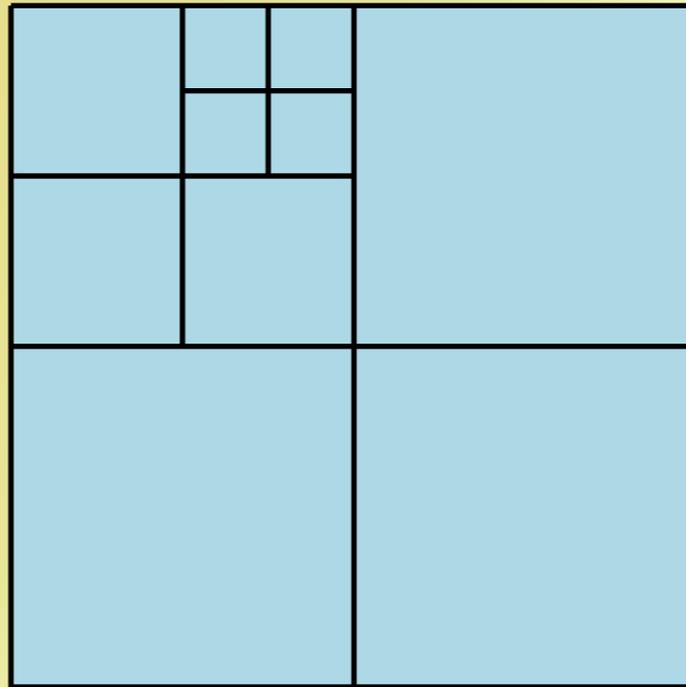


We build another tree, with a vertex for each edge of the quadtree.

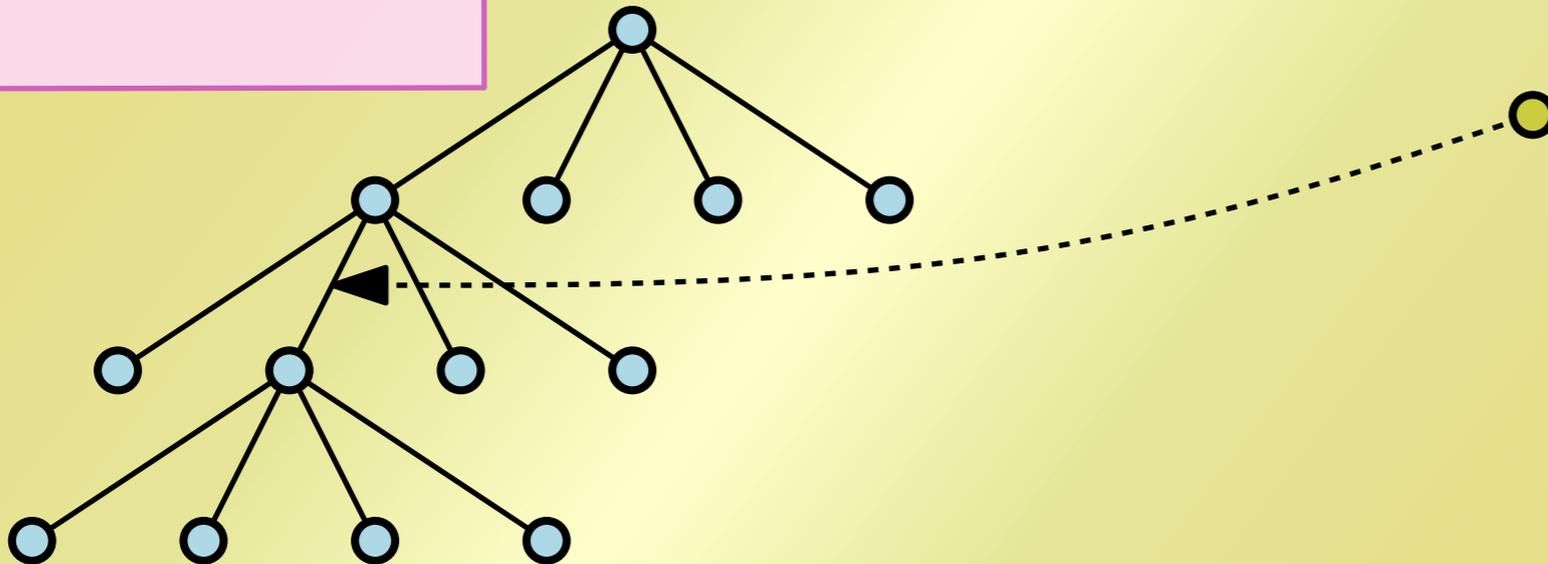


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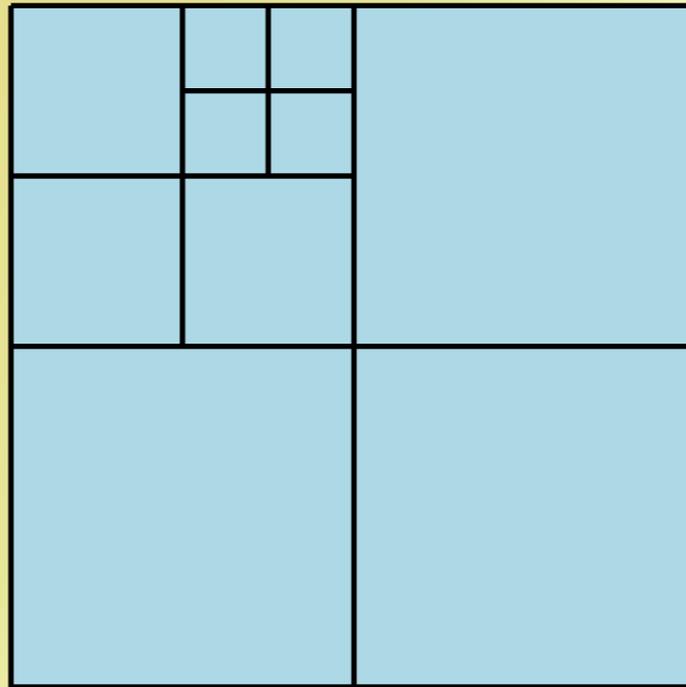


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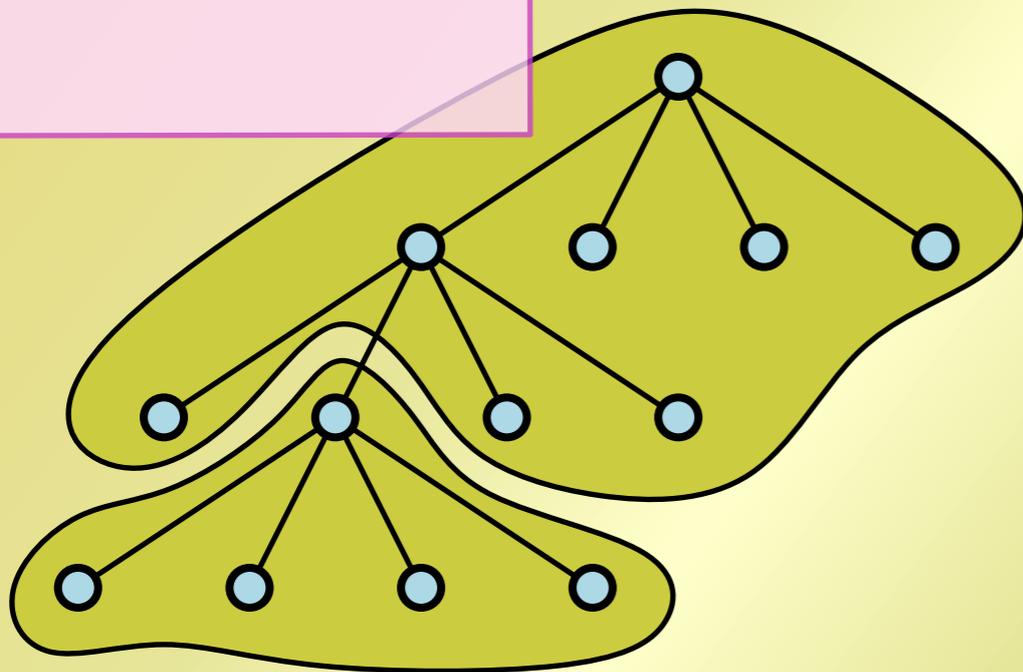


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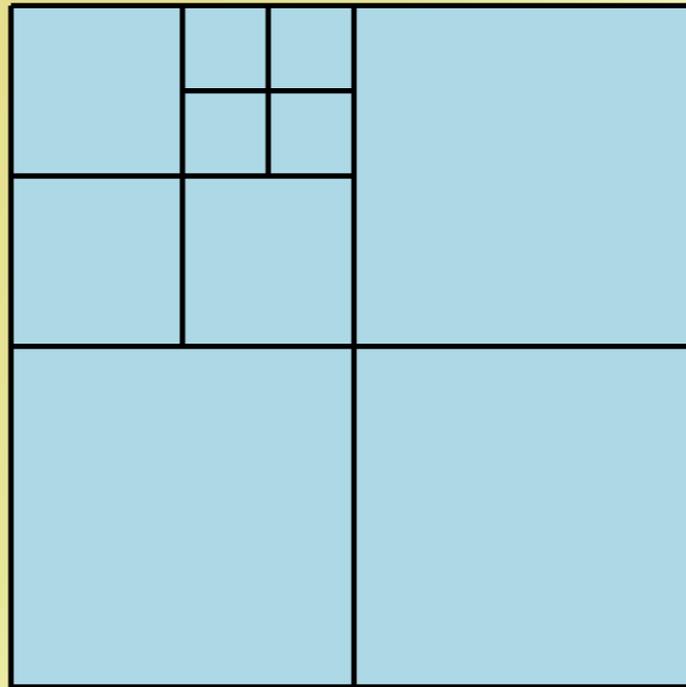


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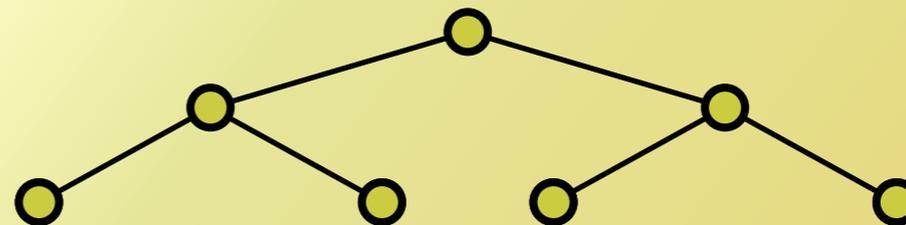
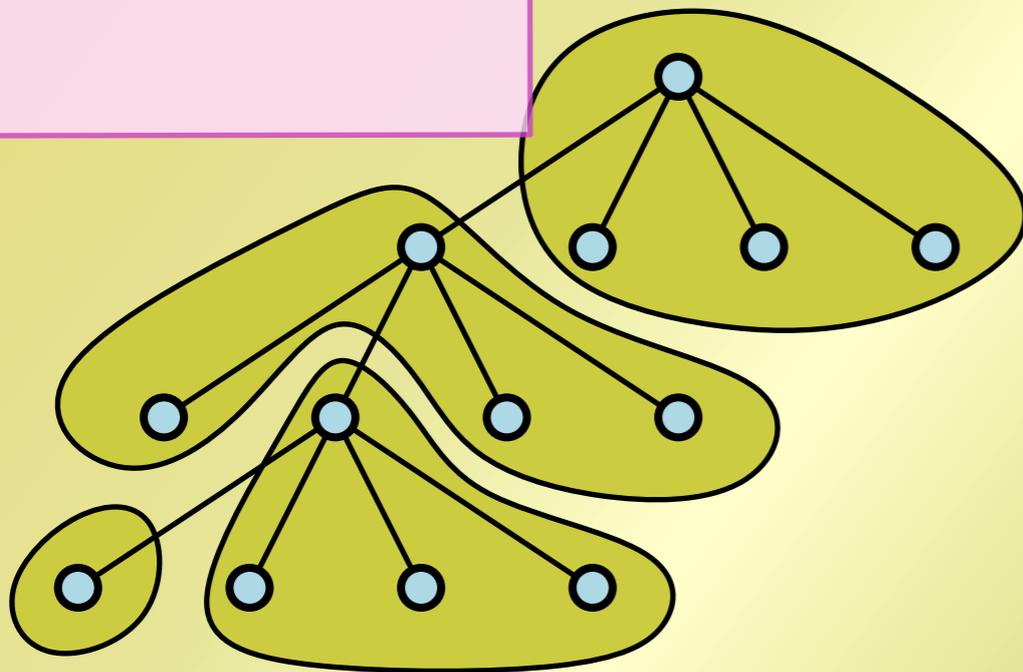


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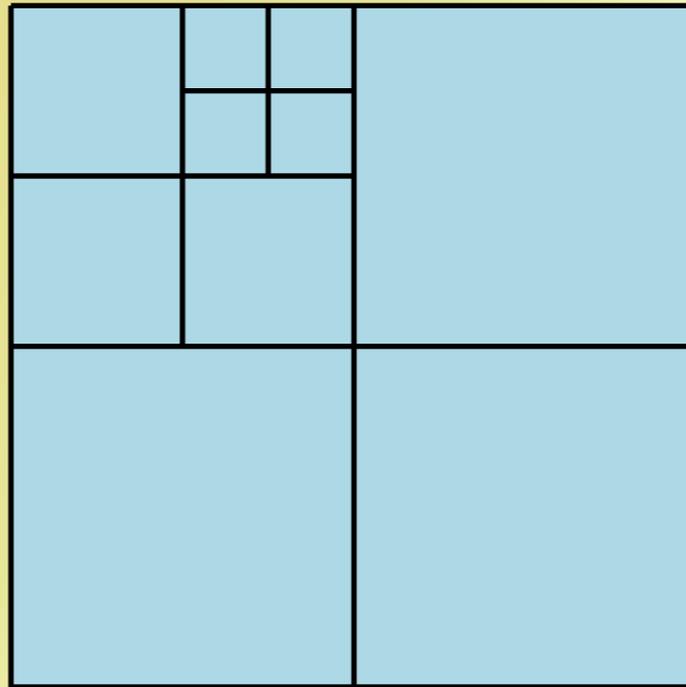


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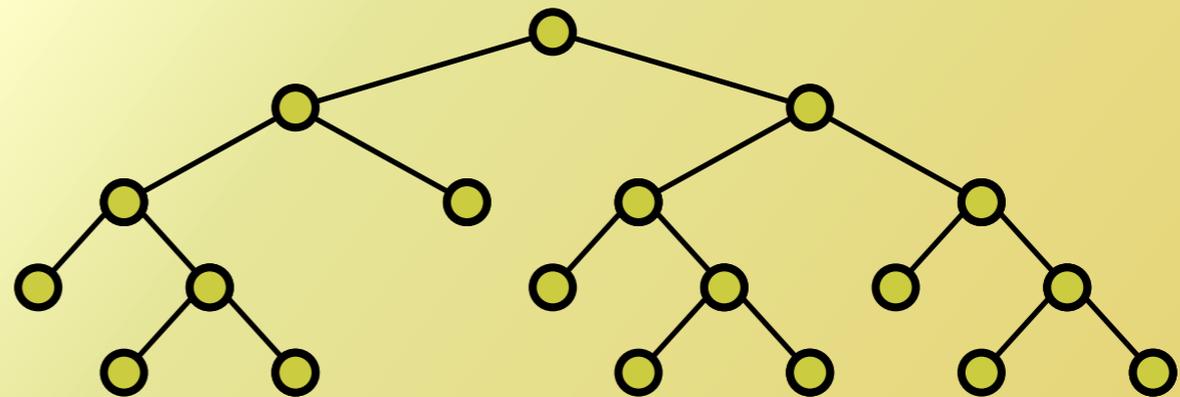
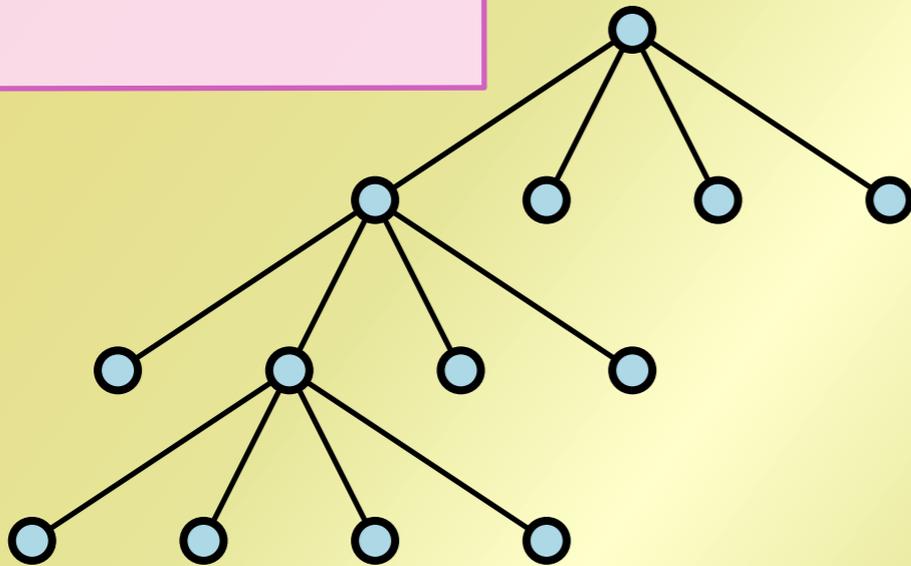


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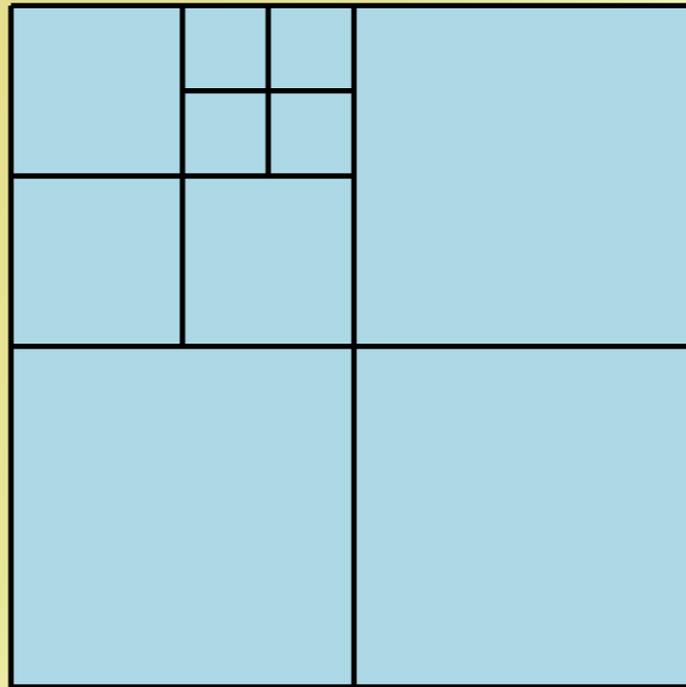


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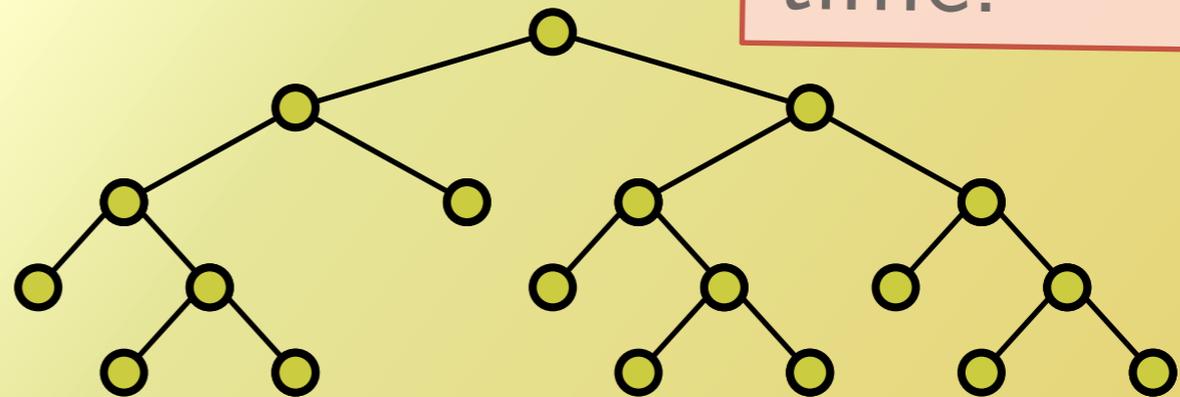
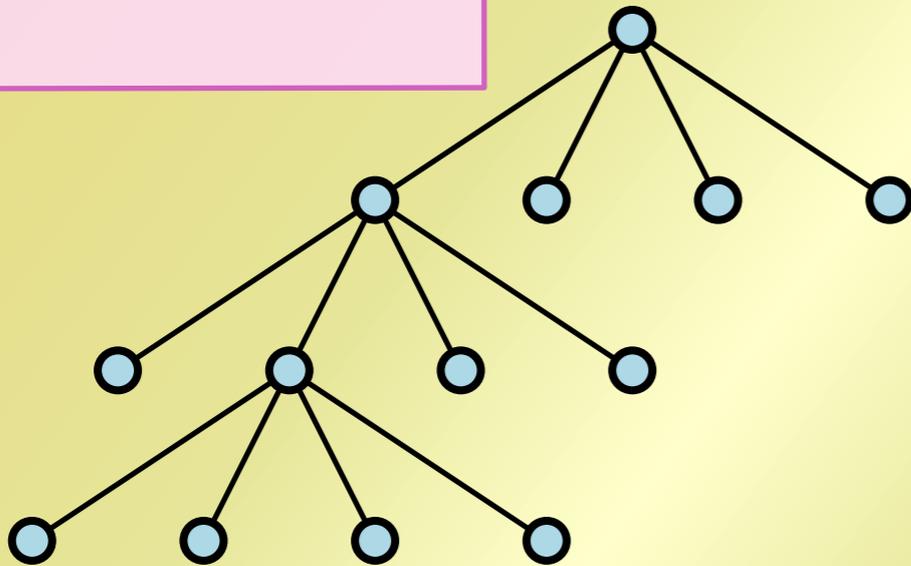
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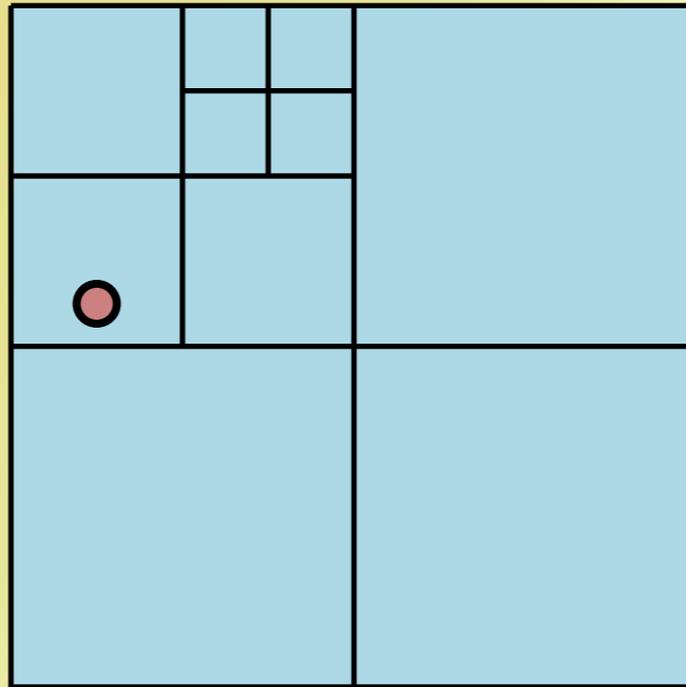
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Now we can locate points in the quadtree in $O(\log n)$ time.



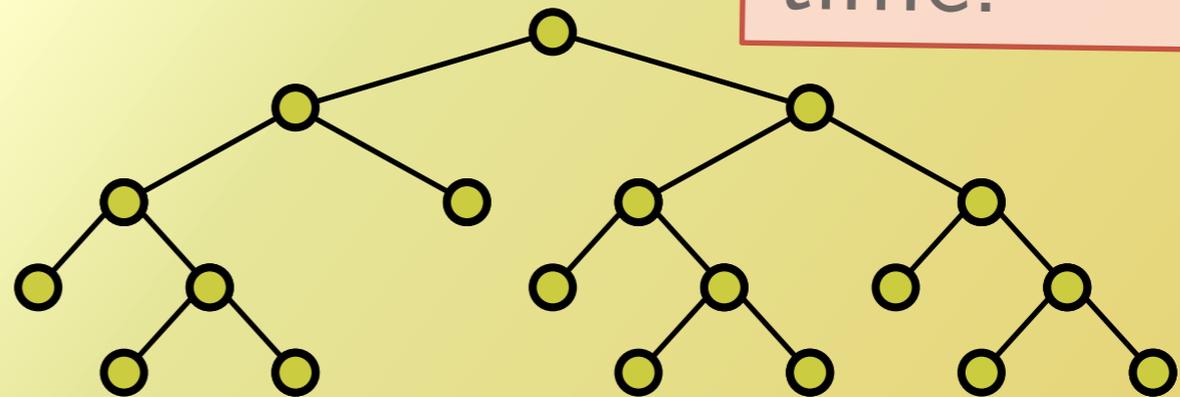
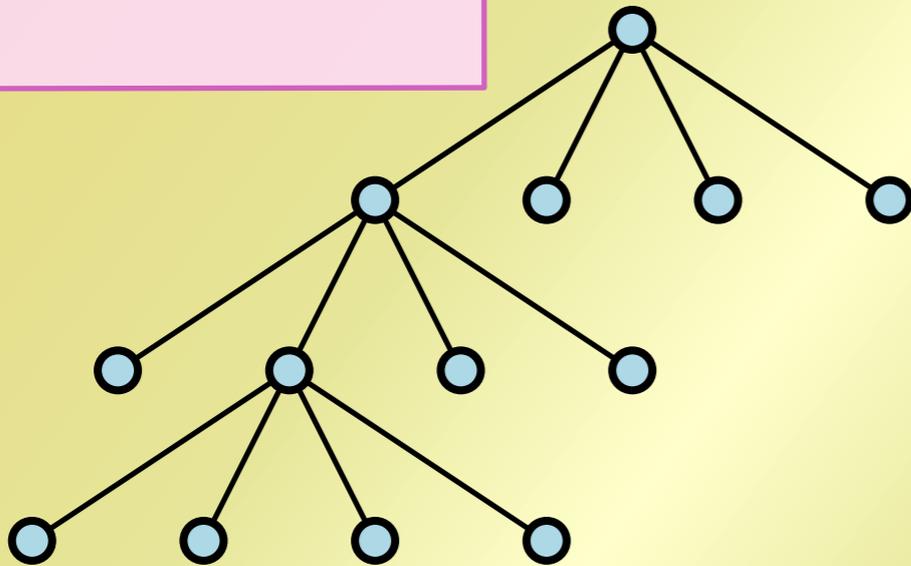
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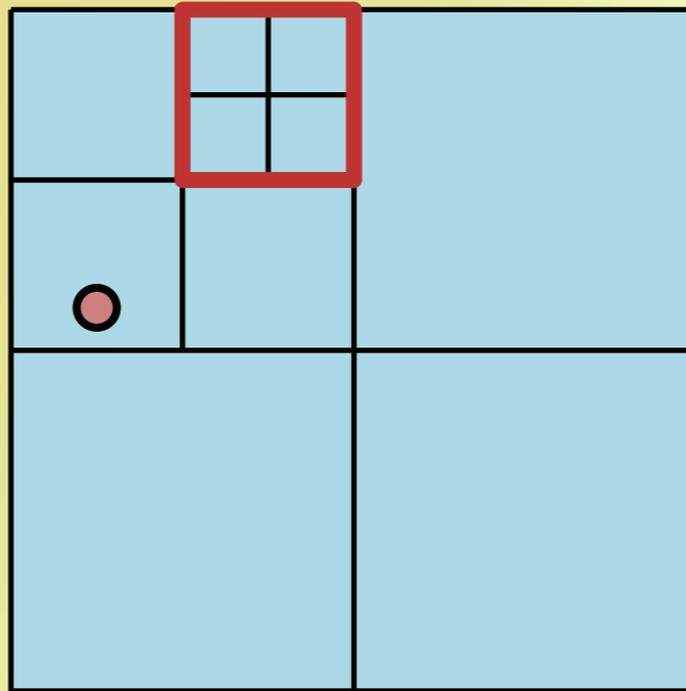
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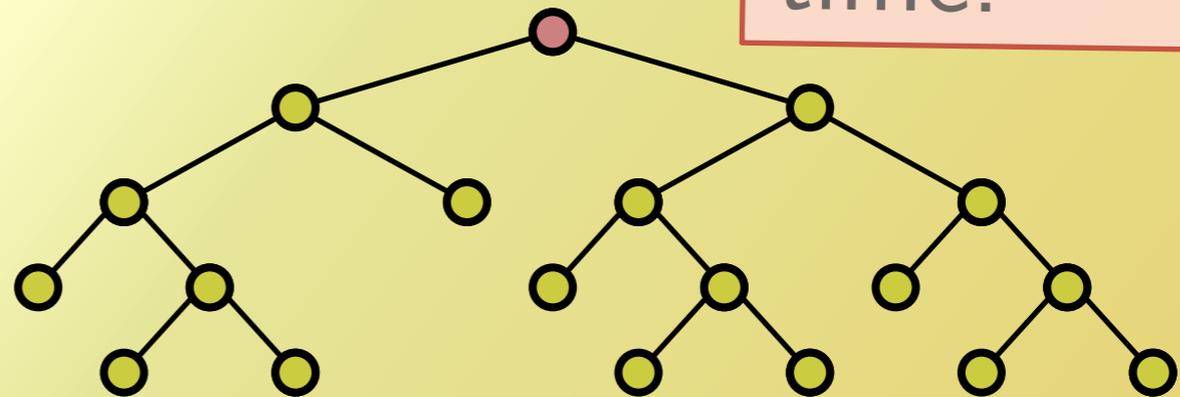
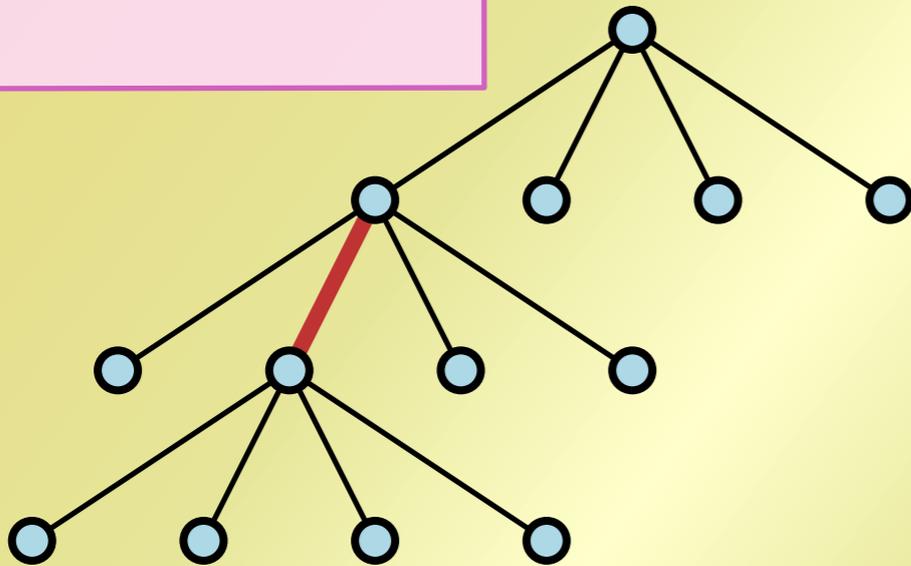
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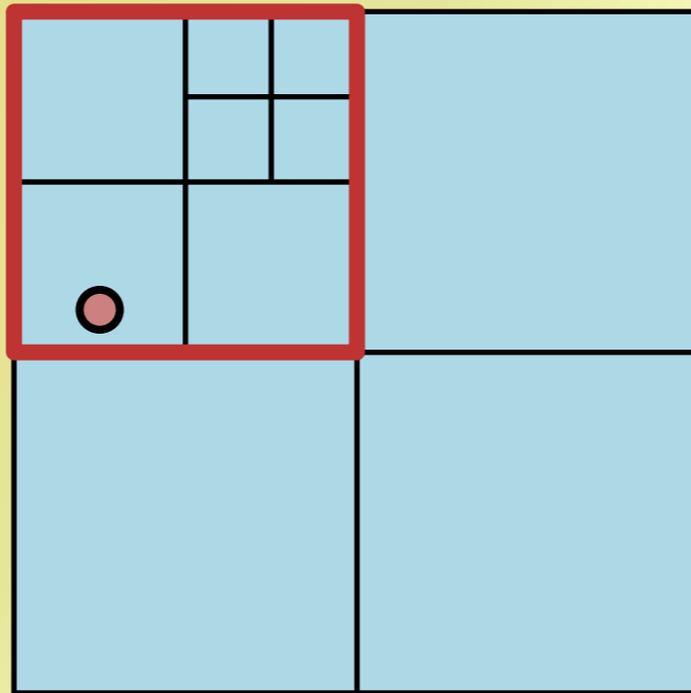
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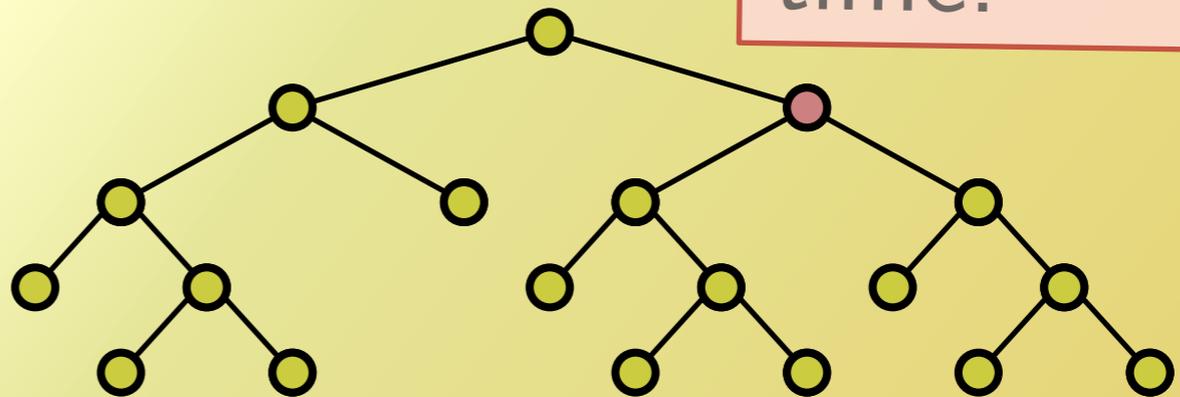
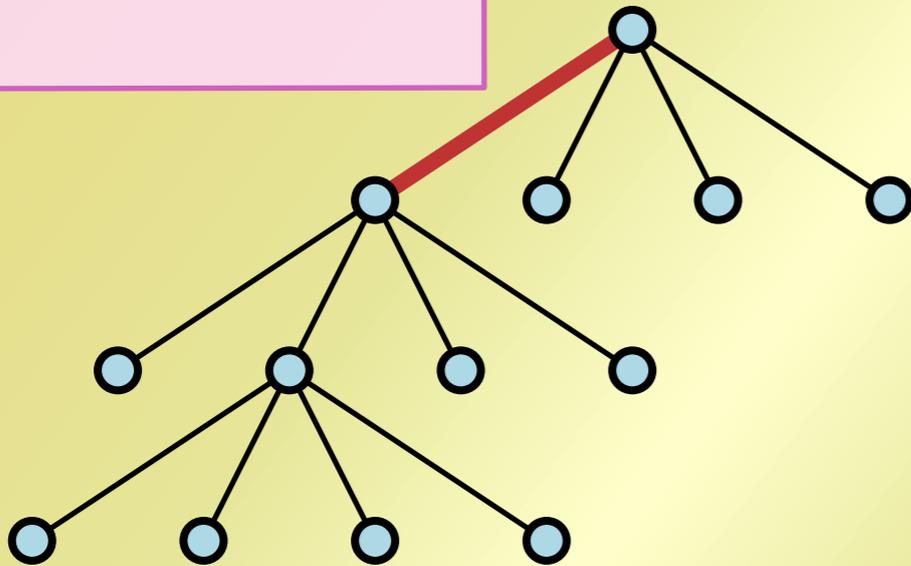
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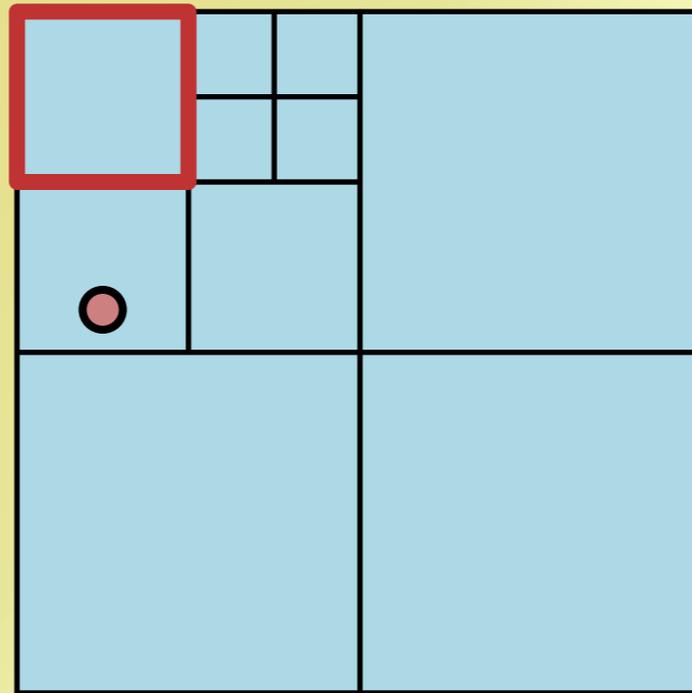
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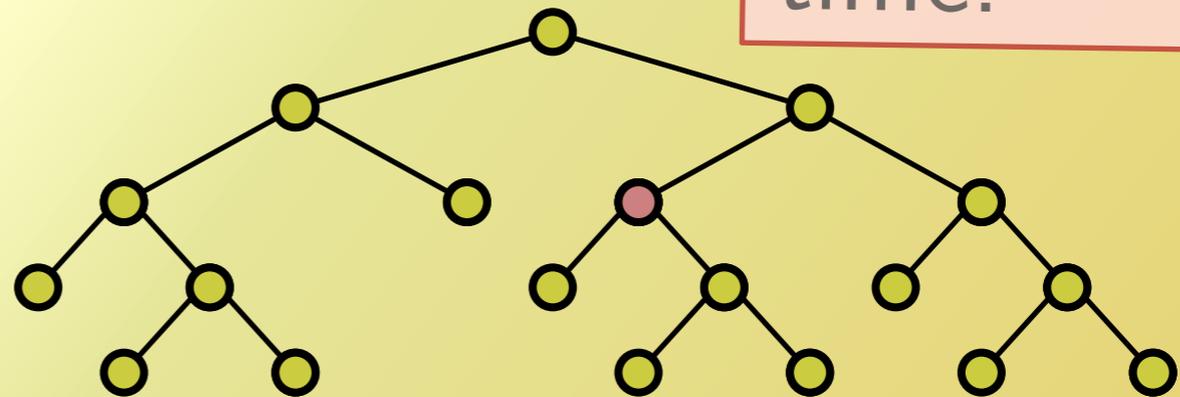
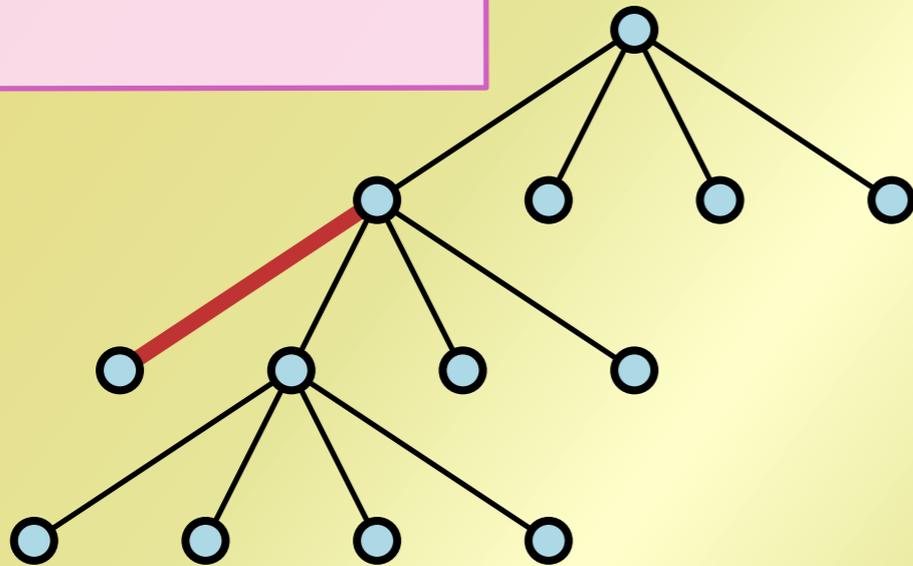
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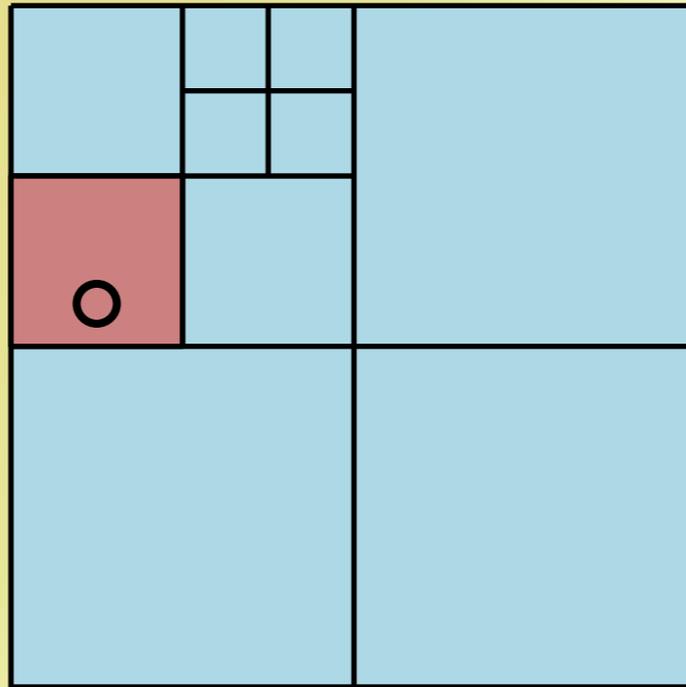
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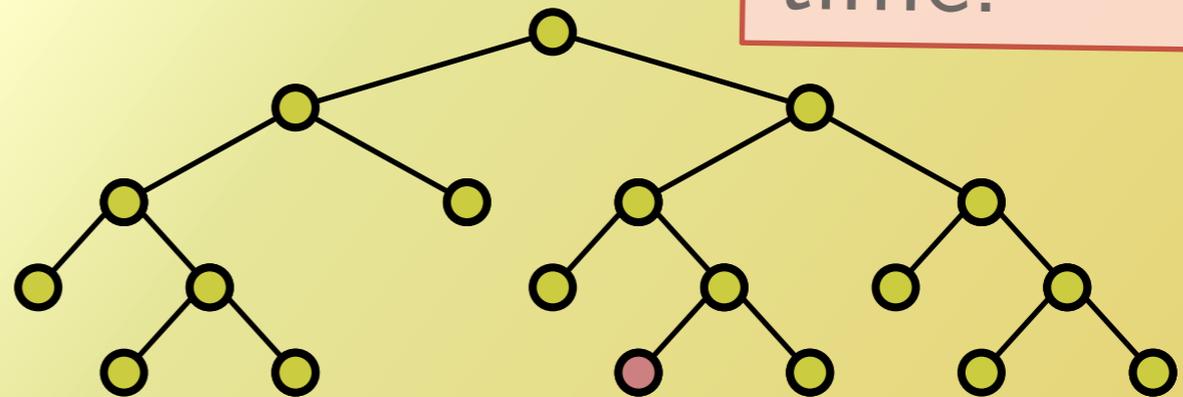
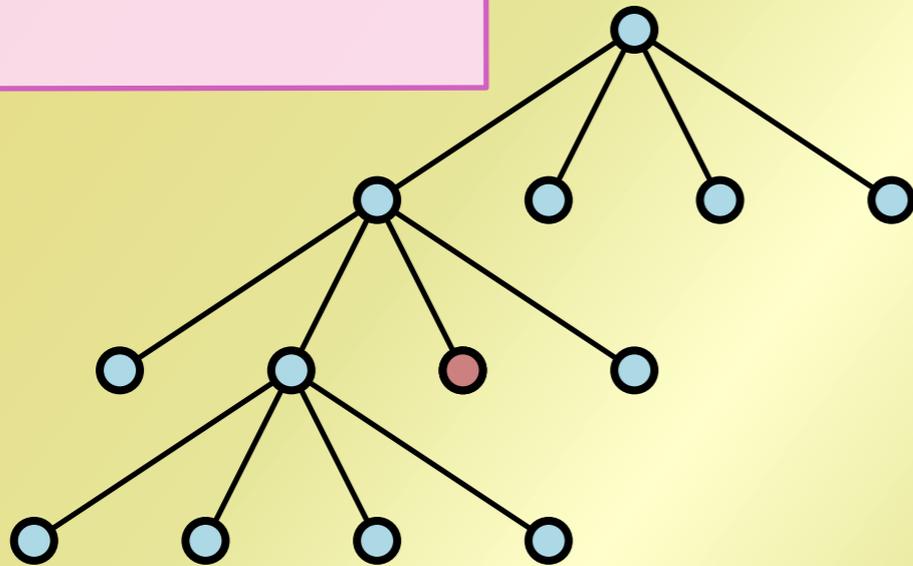
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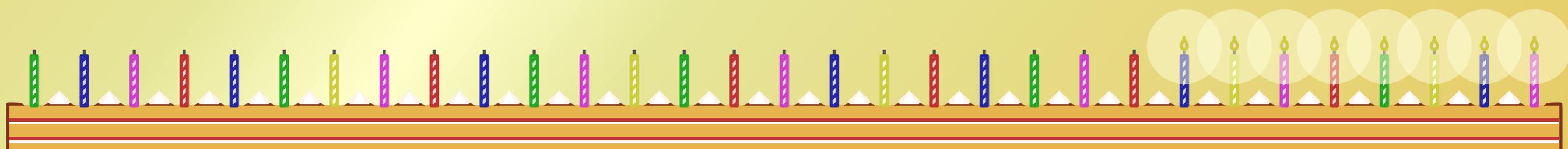
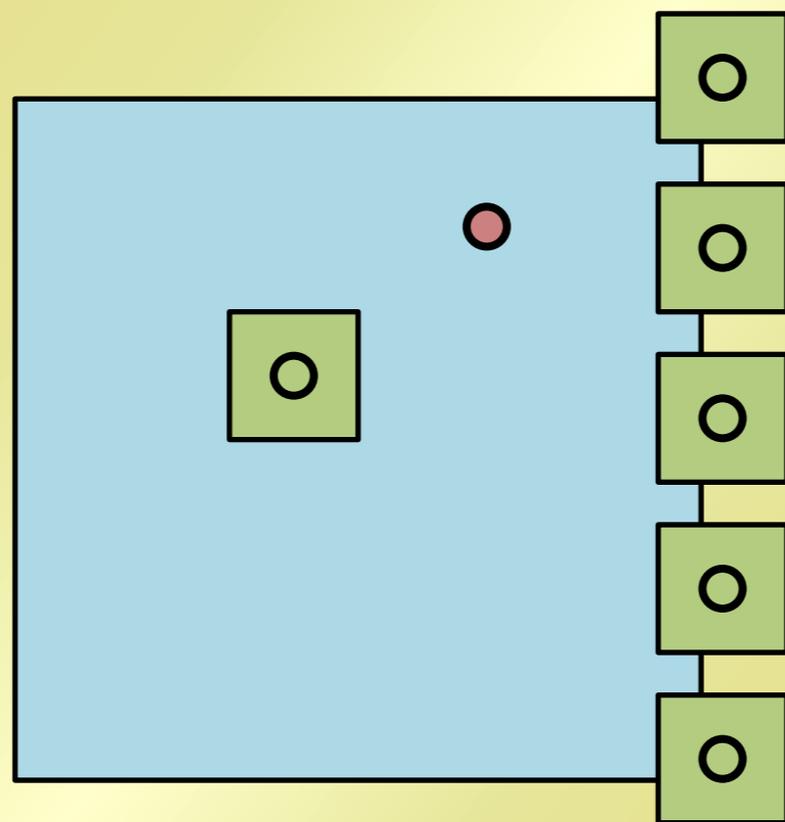
PROBLEM

The number of regions intersecting a quadtree leaf can be linear!



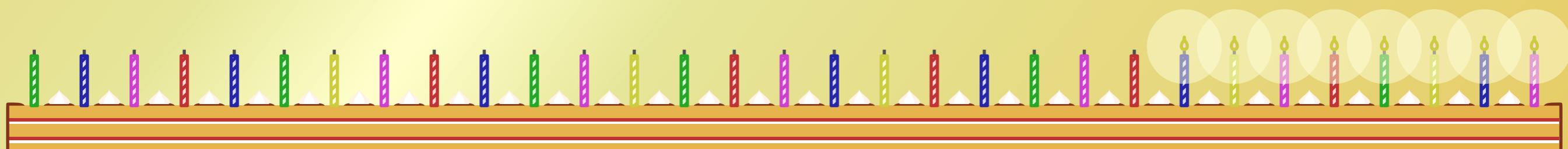
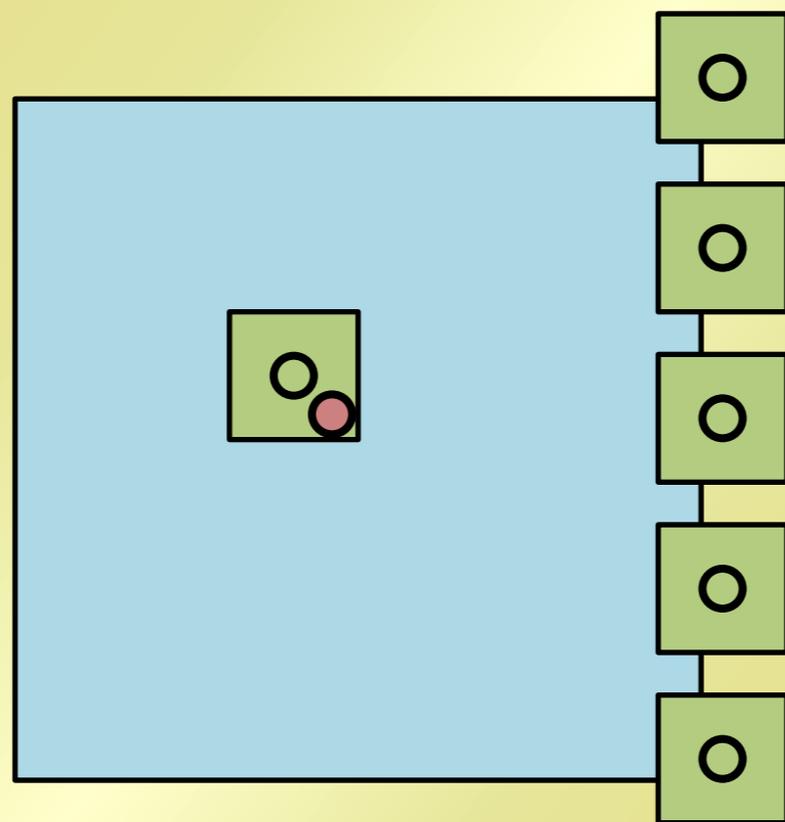
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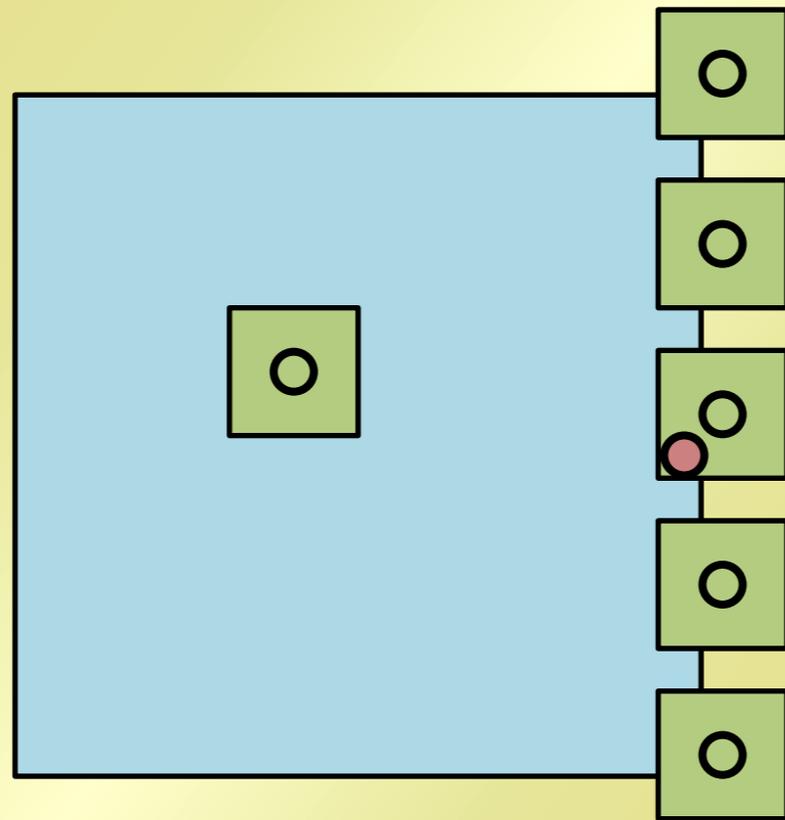
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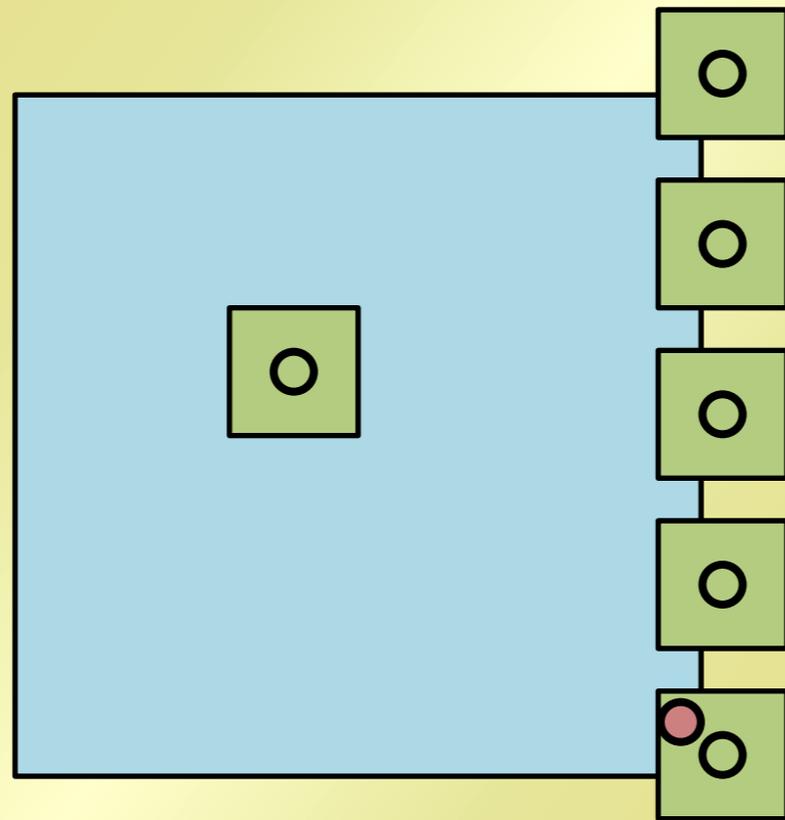
PROBLEM

The number of regions intersecting a quadtree leaf can be linear!



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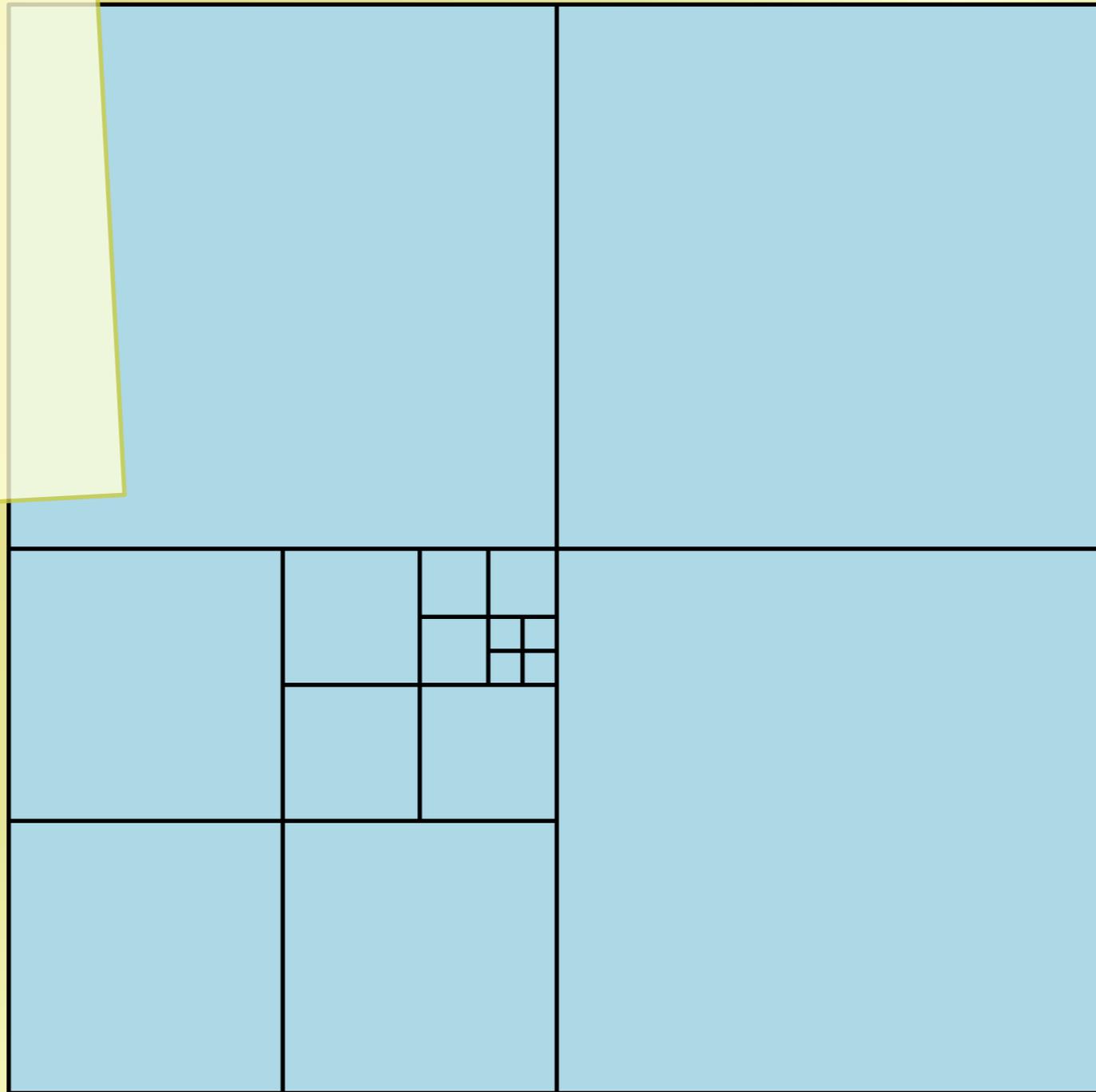




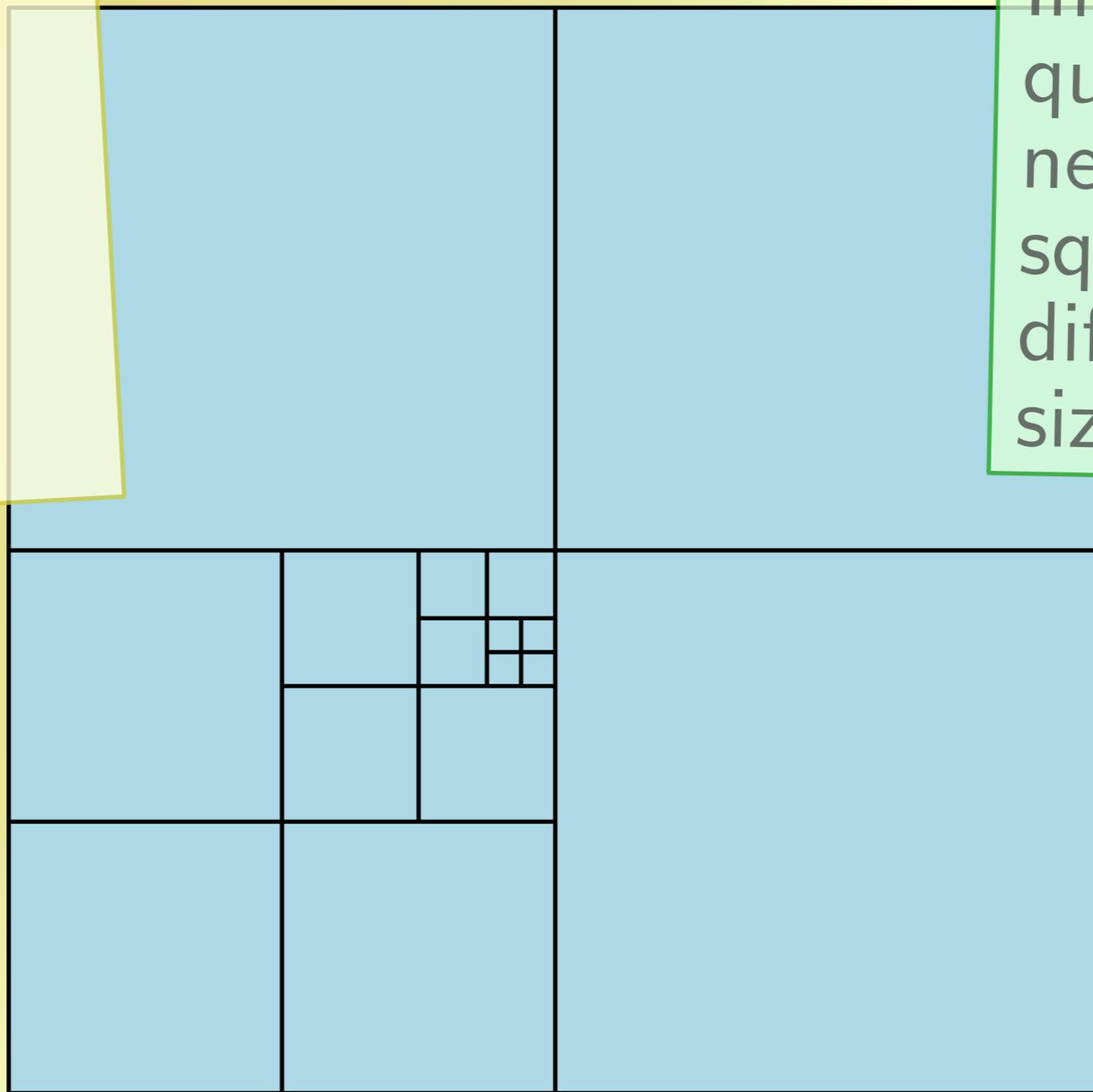
Consider a
quadtree
again.



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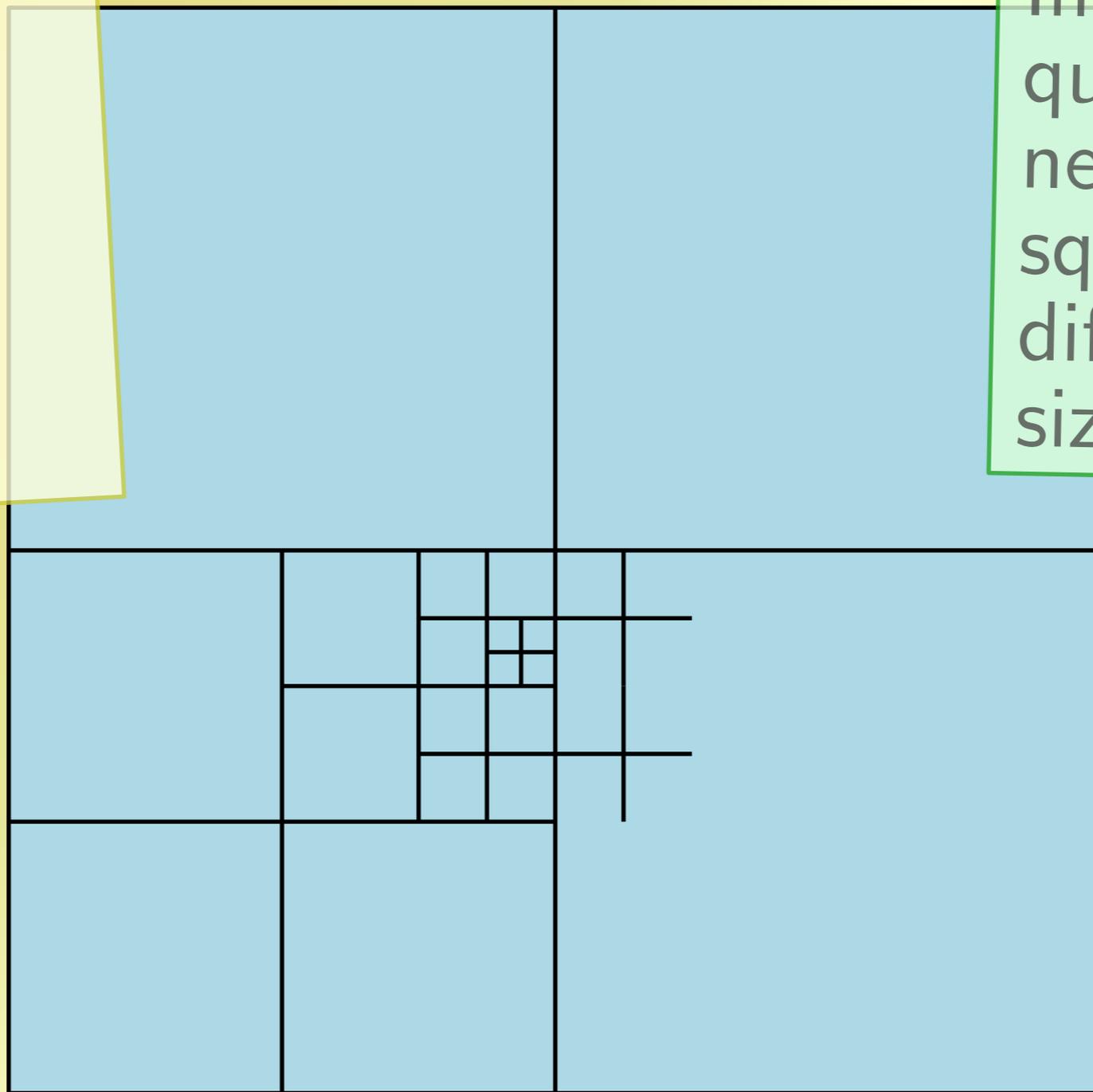
Consider a
quadtree
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In a *balanced*
quadtree,
neighbouring
squares don't
differ much in
size.



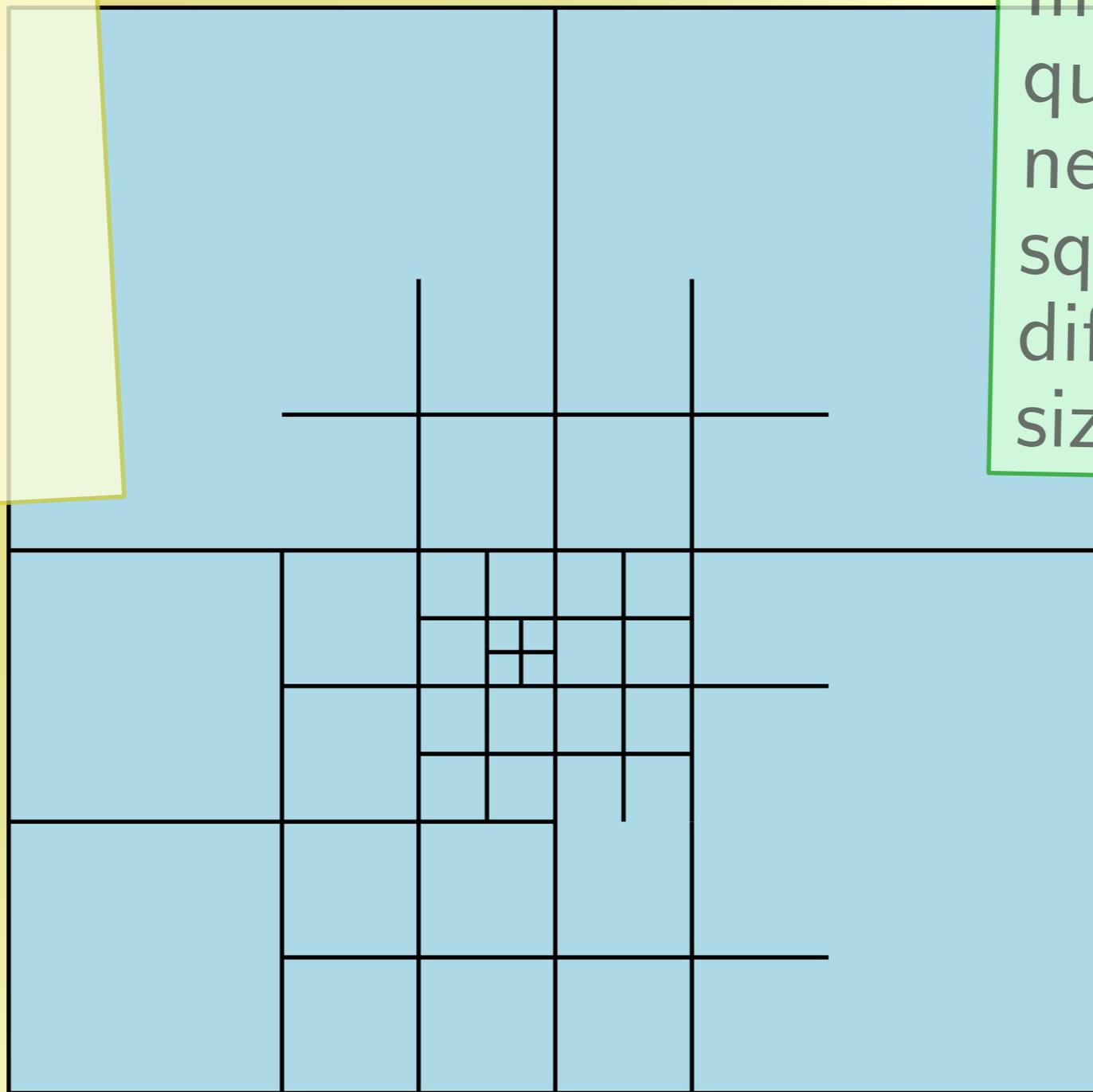
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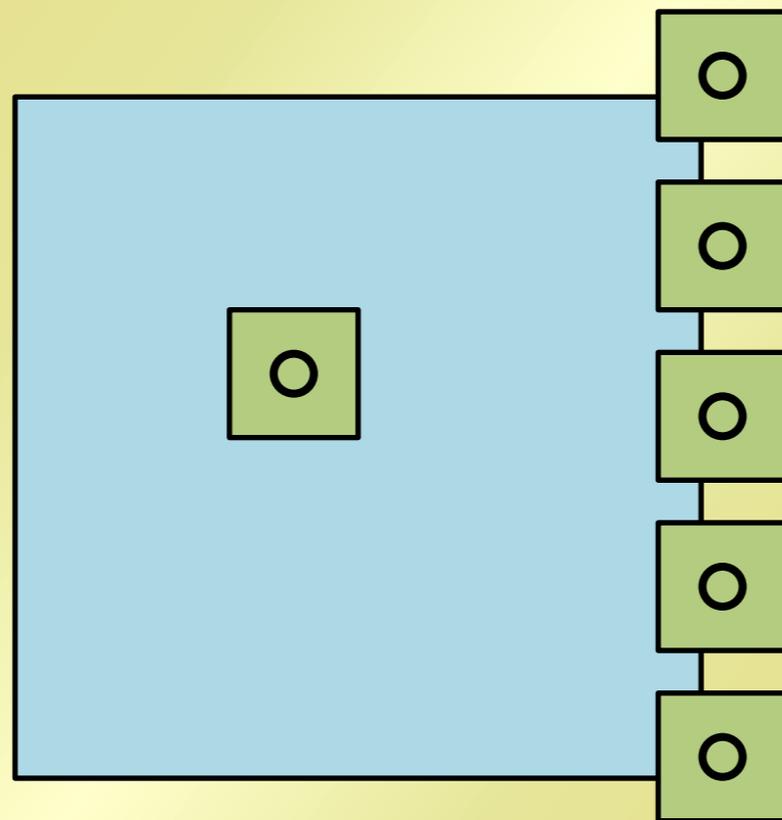
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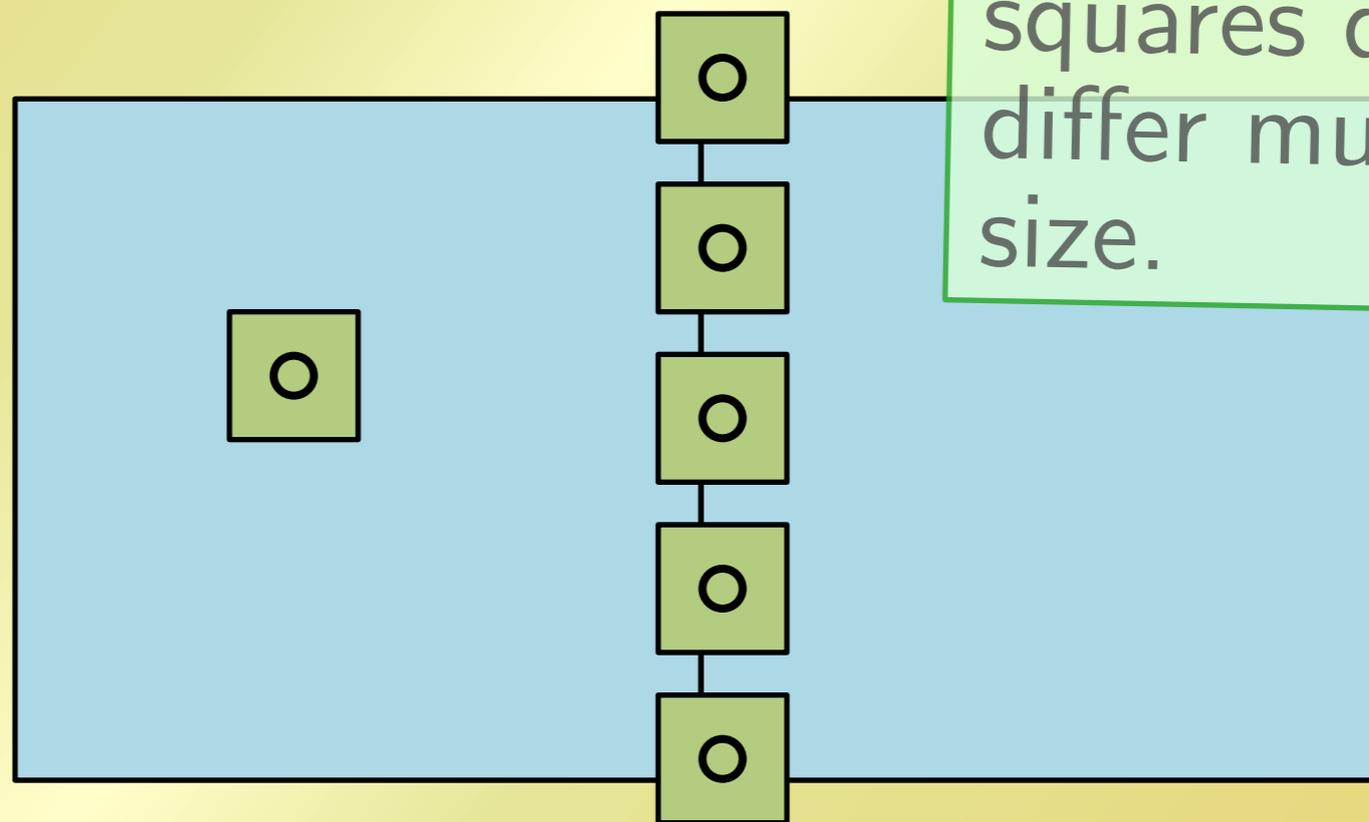
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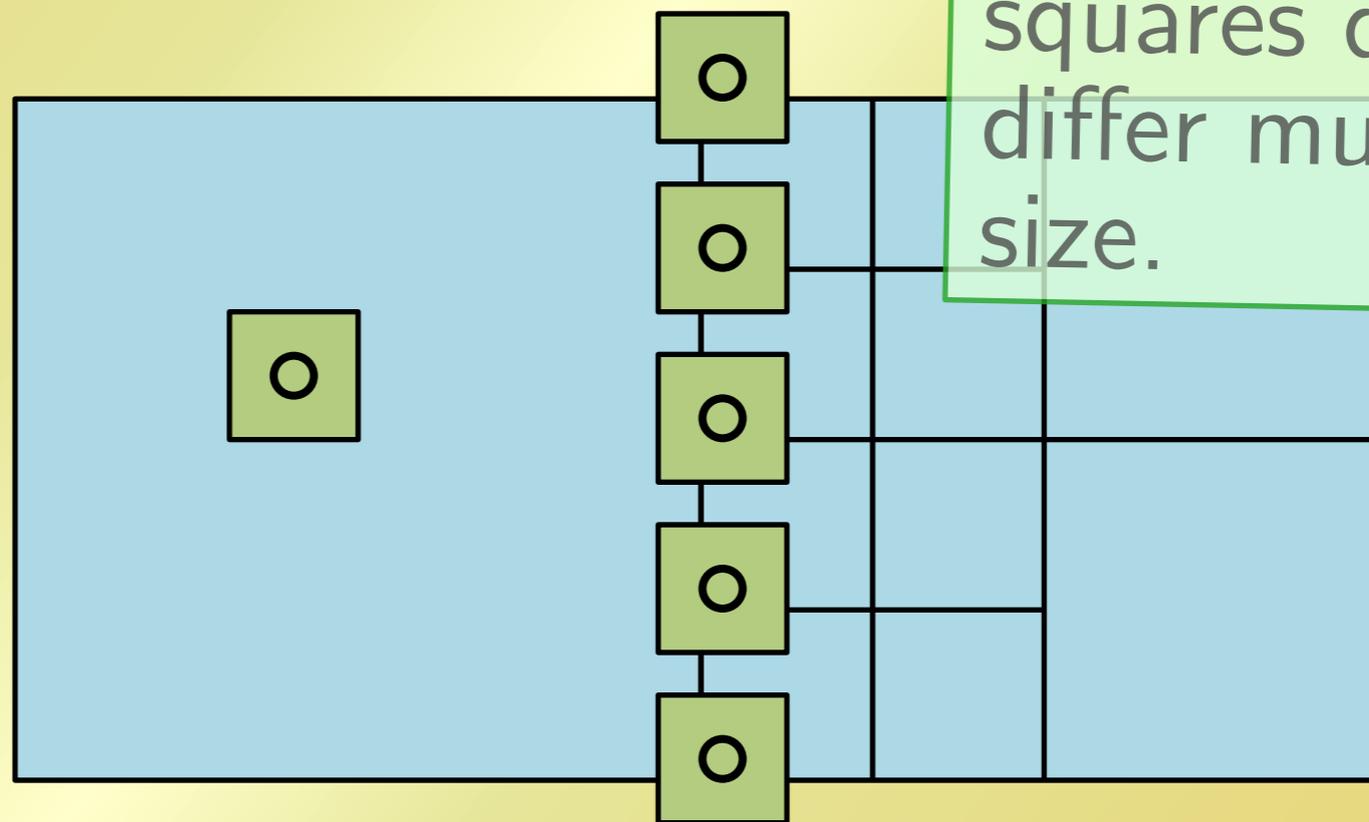
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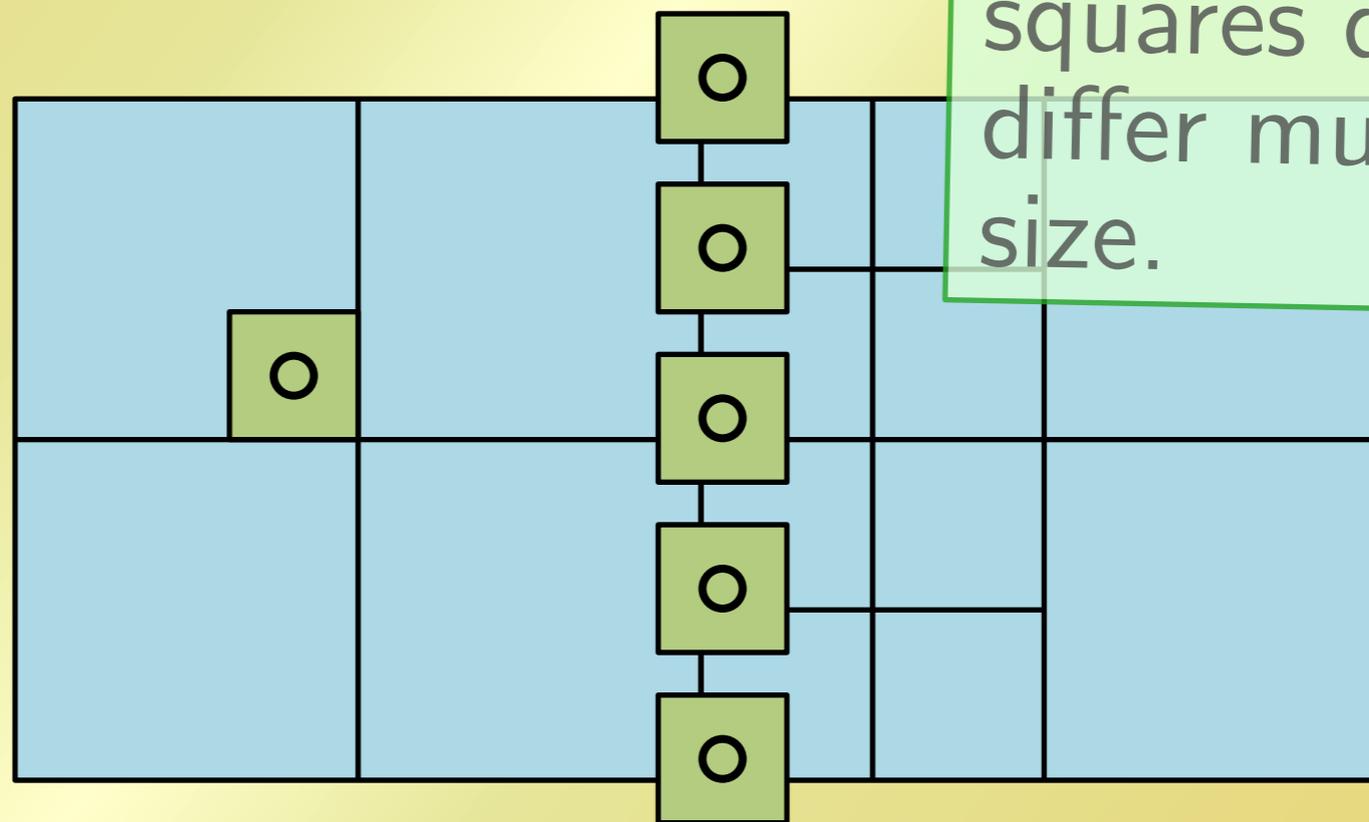
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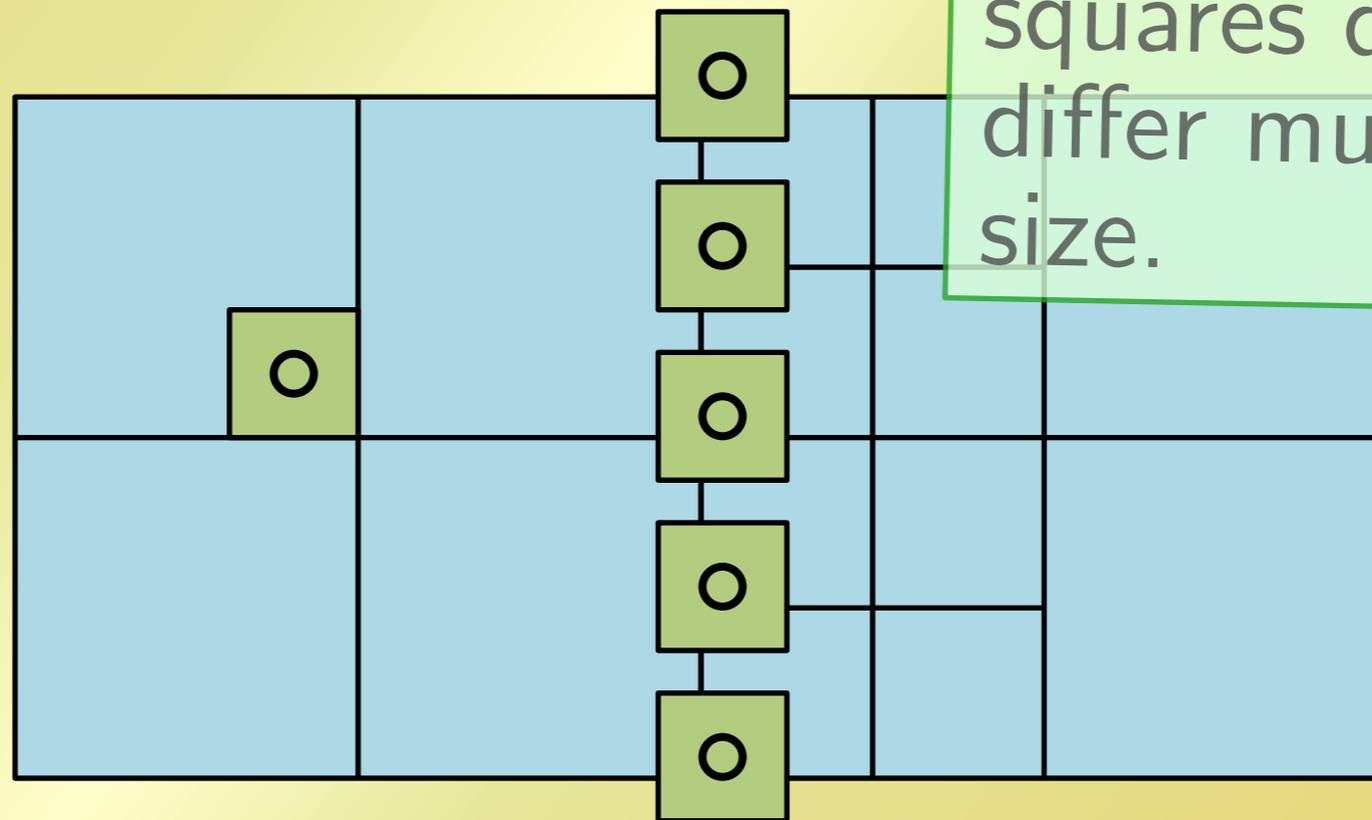


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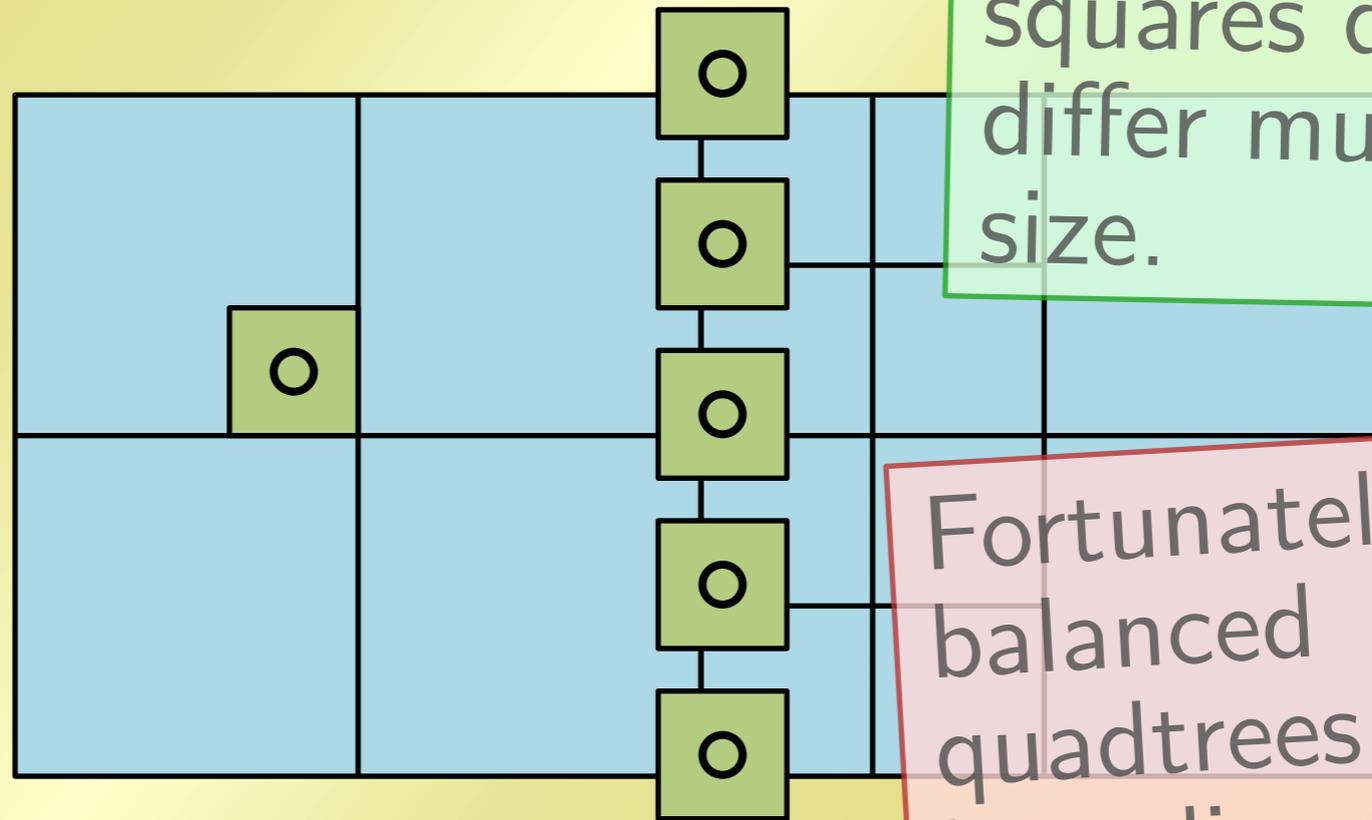
LEMMA
Now each
leaf intersects
at most $O(1)$
regions.



Consider a
quadtree
again.

In a *balanced*
quadtree,
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LEMMA
Now each
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Fortunately,
balanced
quadtrees still
have linear
size.

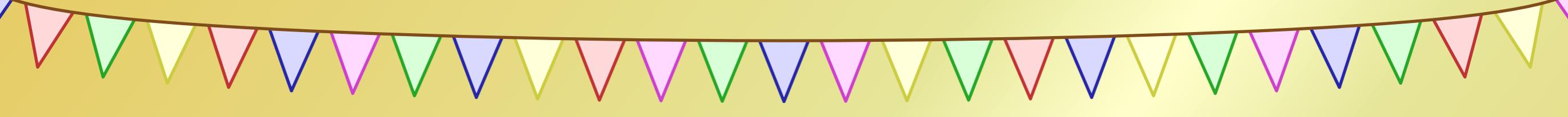




PROBLEM

But now we
can't change
the quadtree
locally in
 $O(1)$ time!





PROBLEM

But now we
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ADVICE

Don't worry,
be happy!





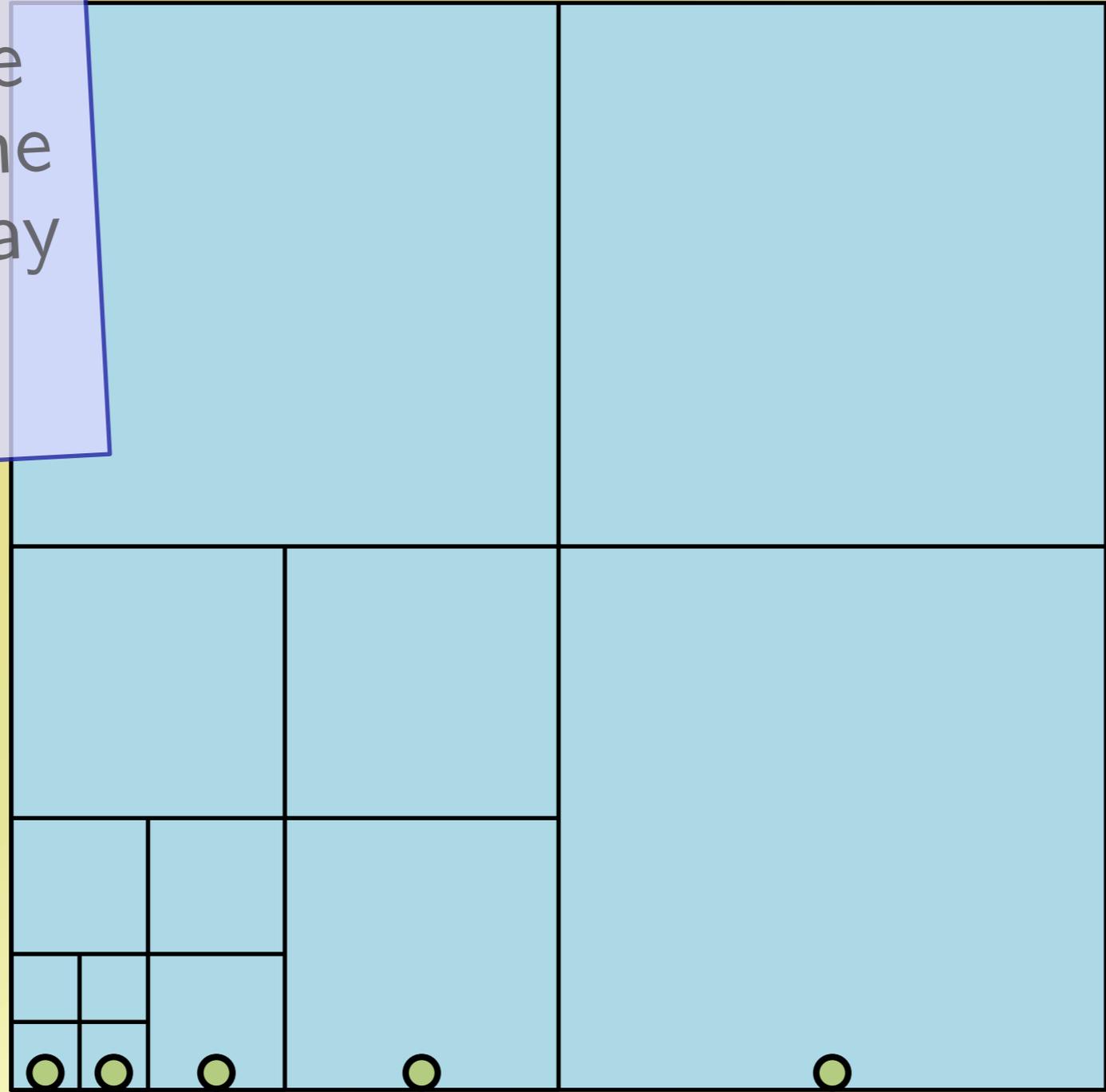
PROBLEM

The distance from q to the right cell may be linear!



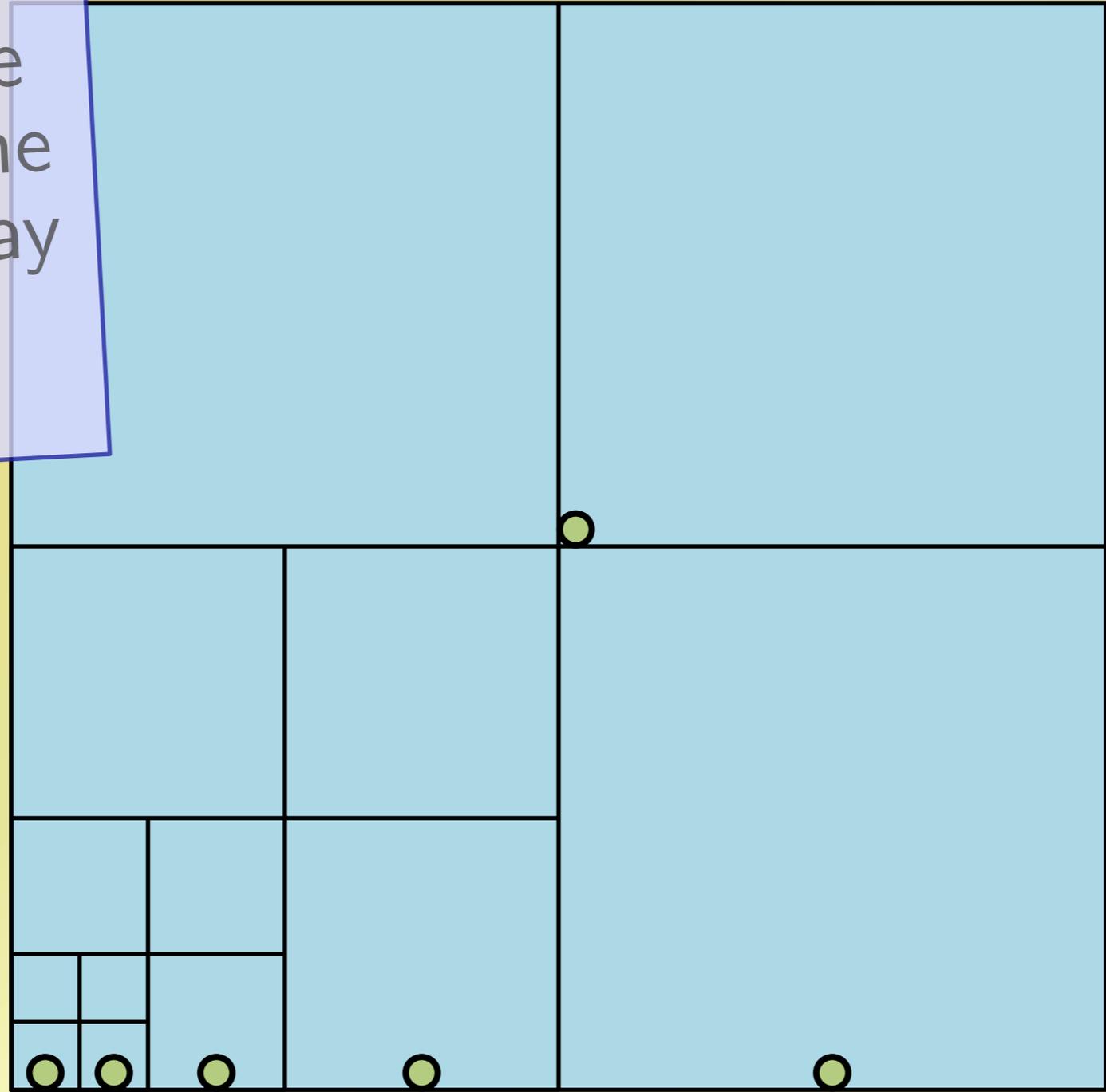
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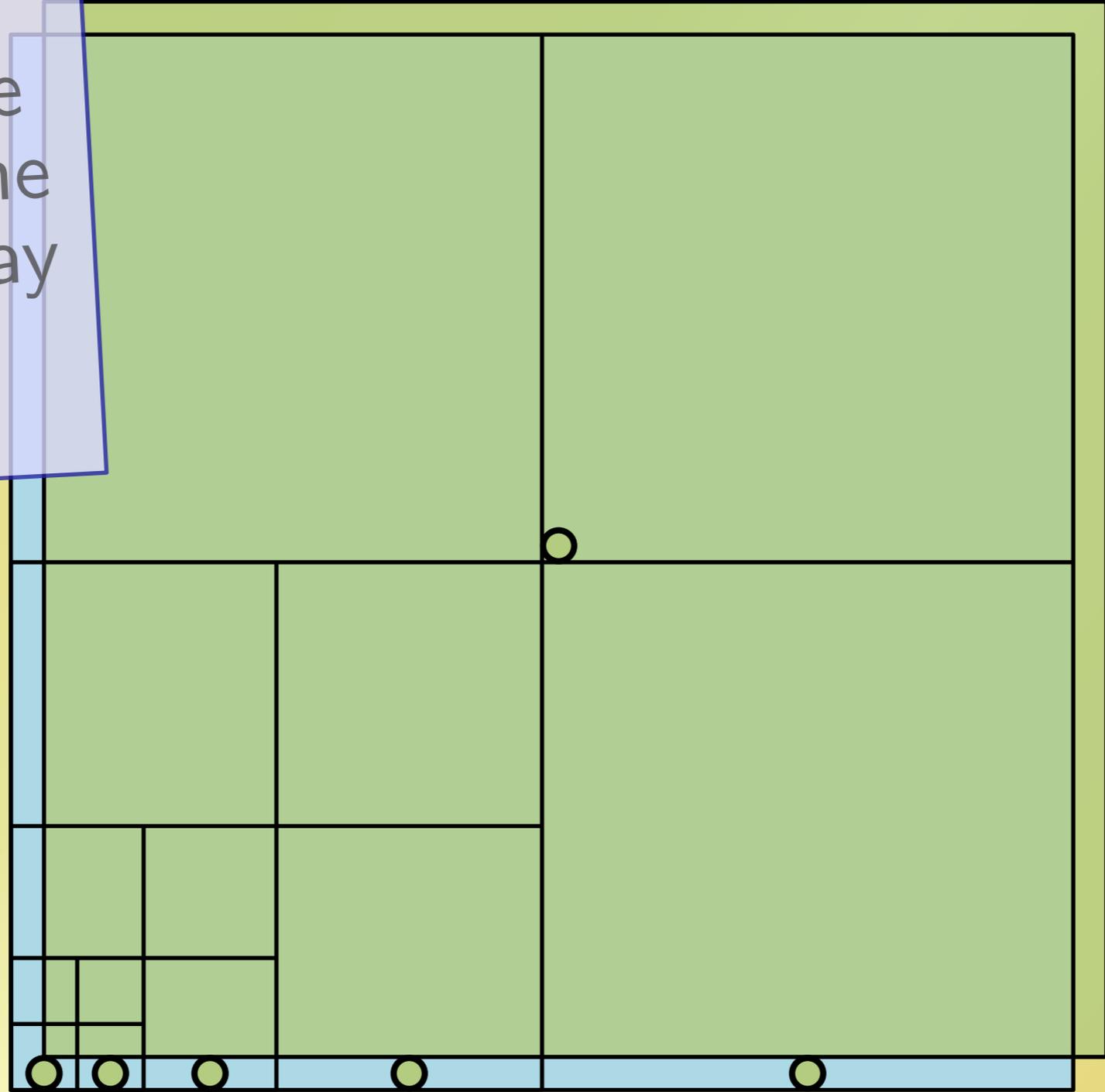
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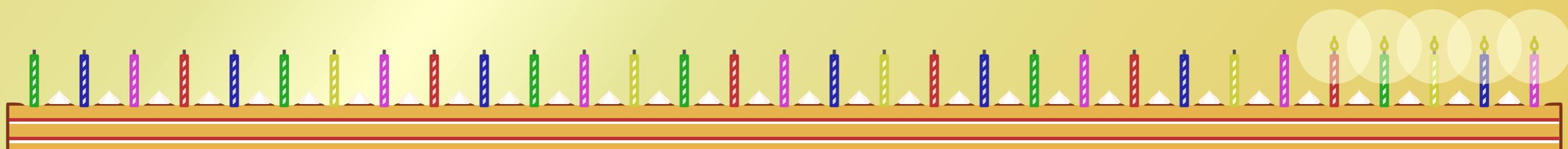
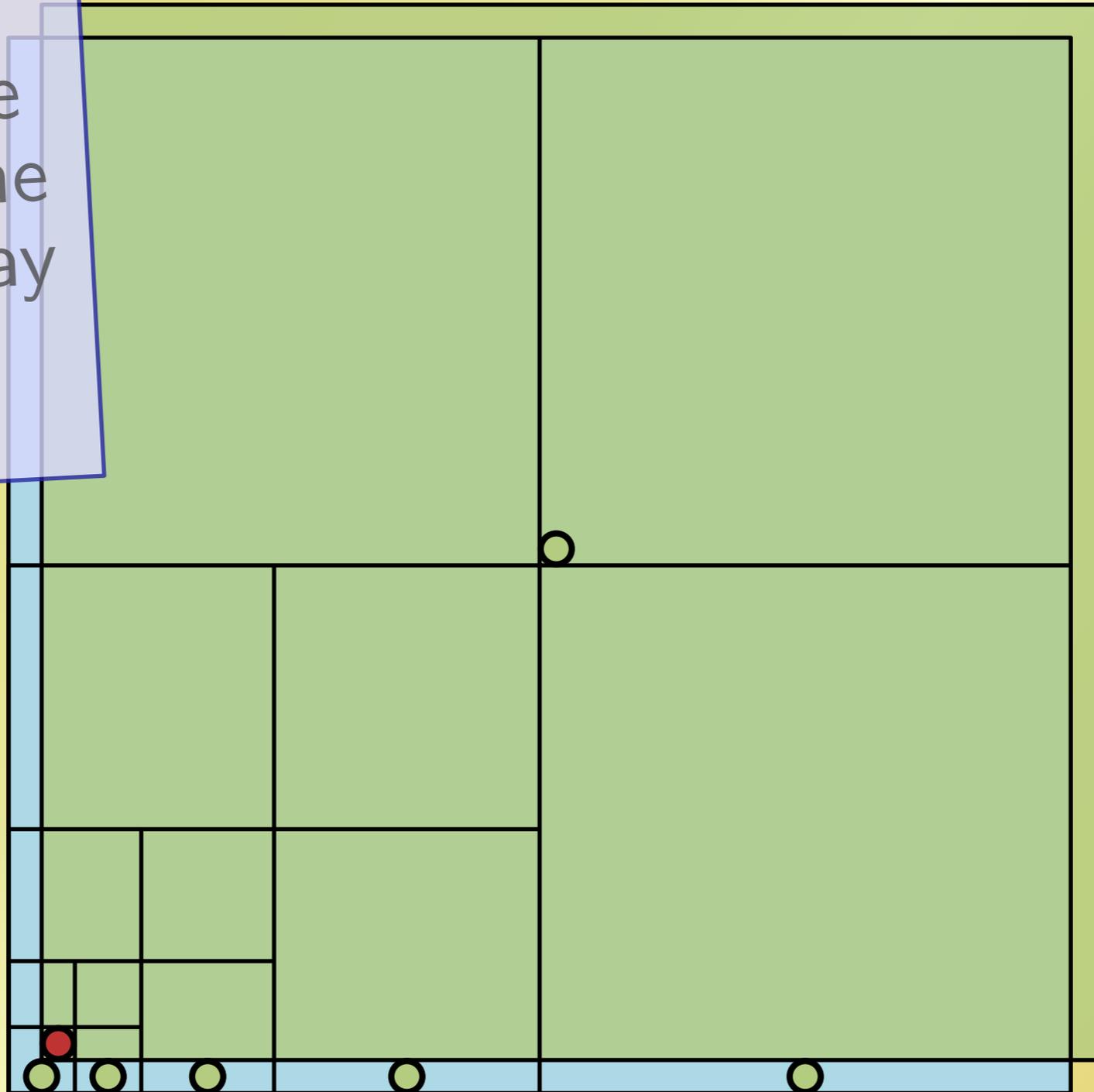
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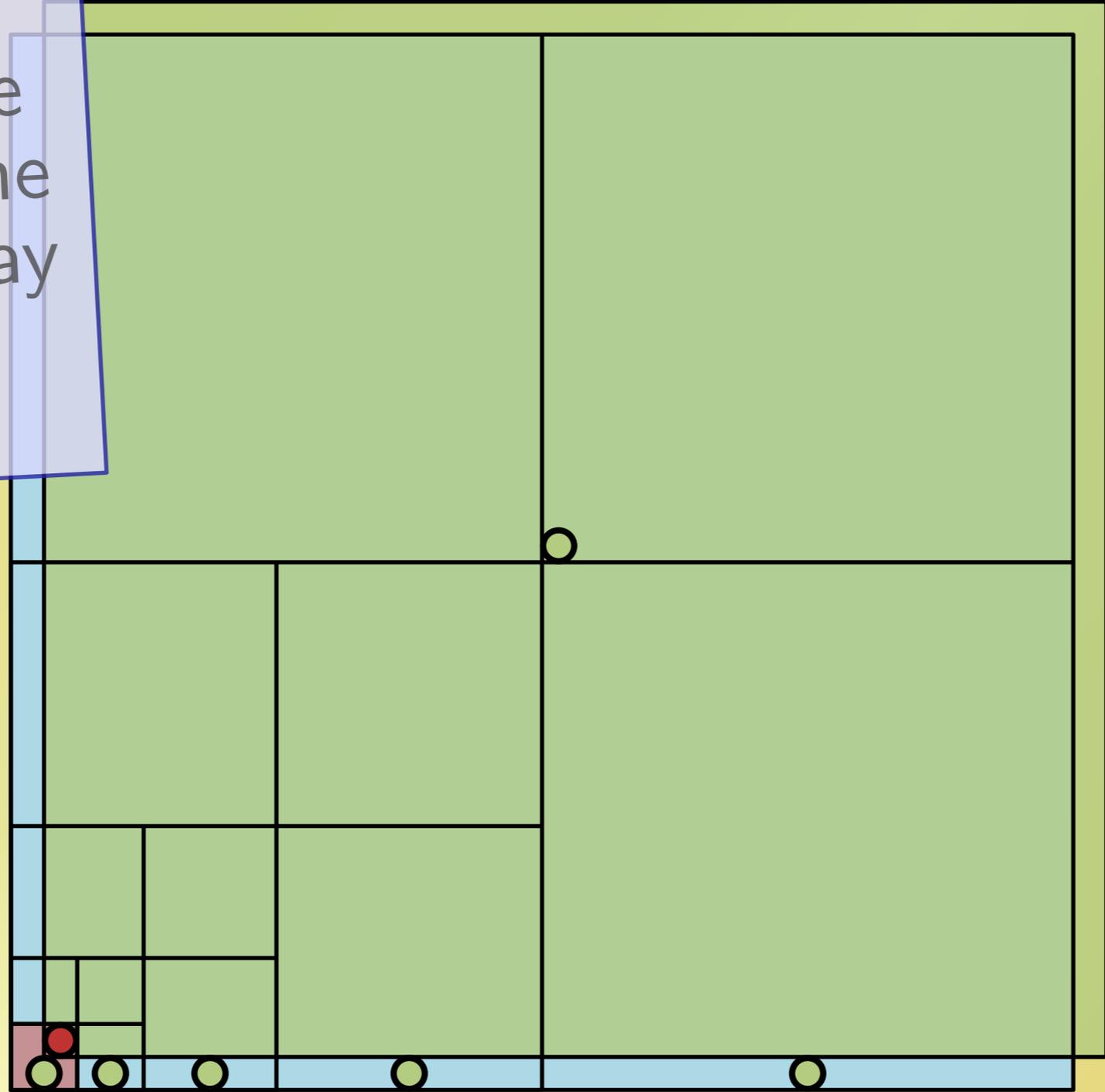
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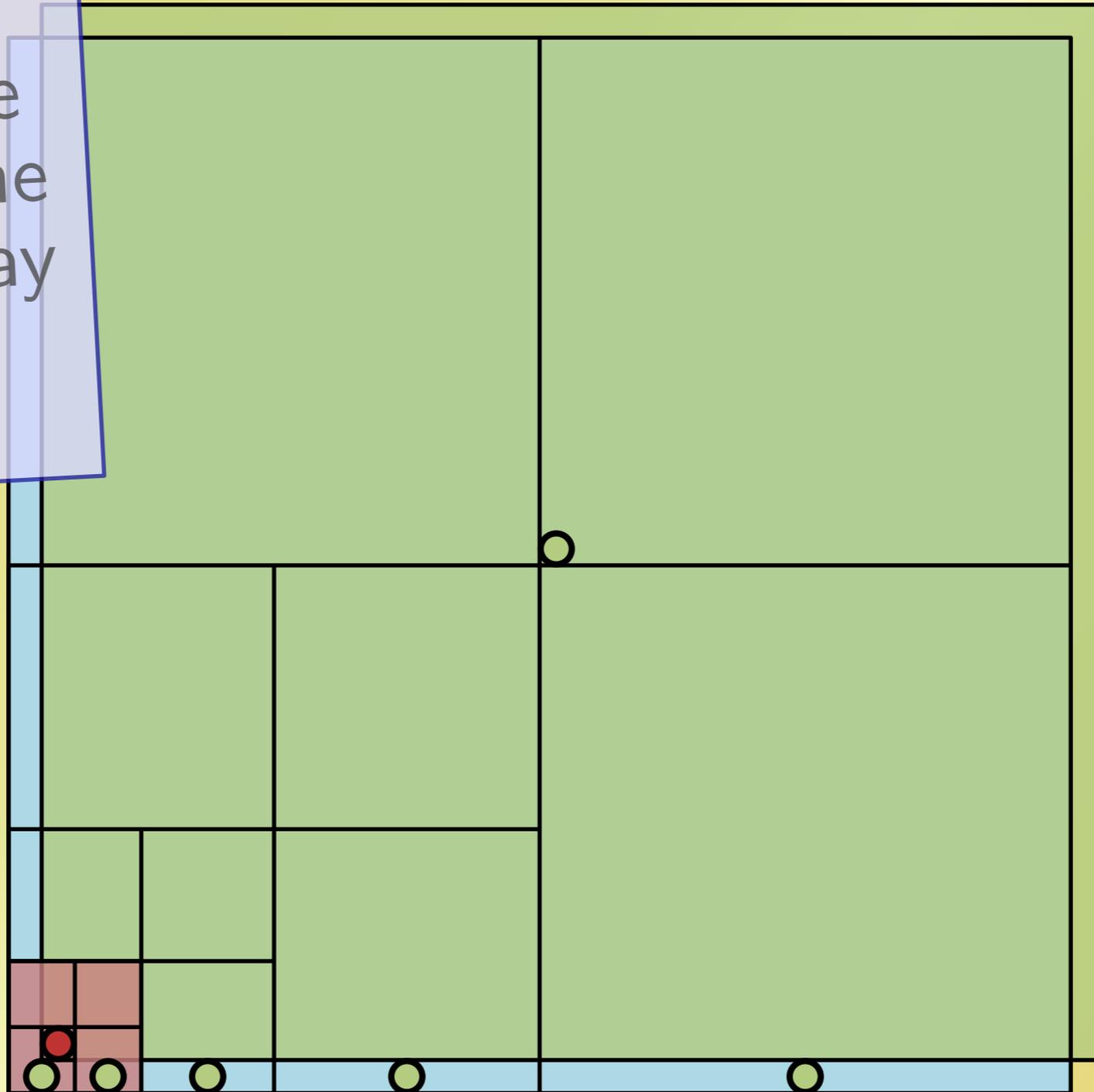
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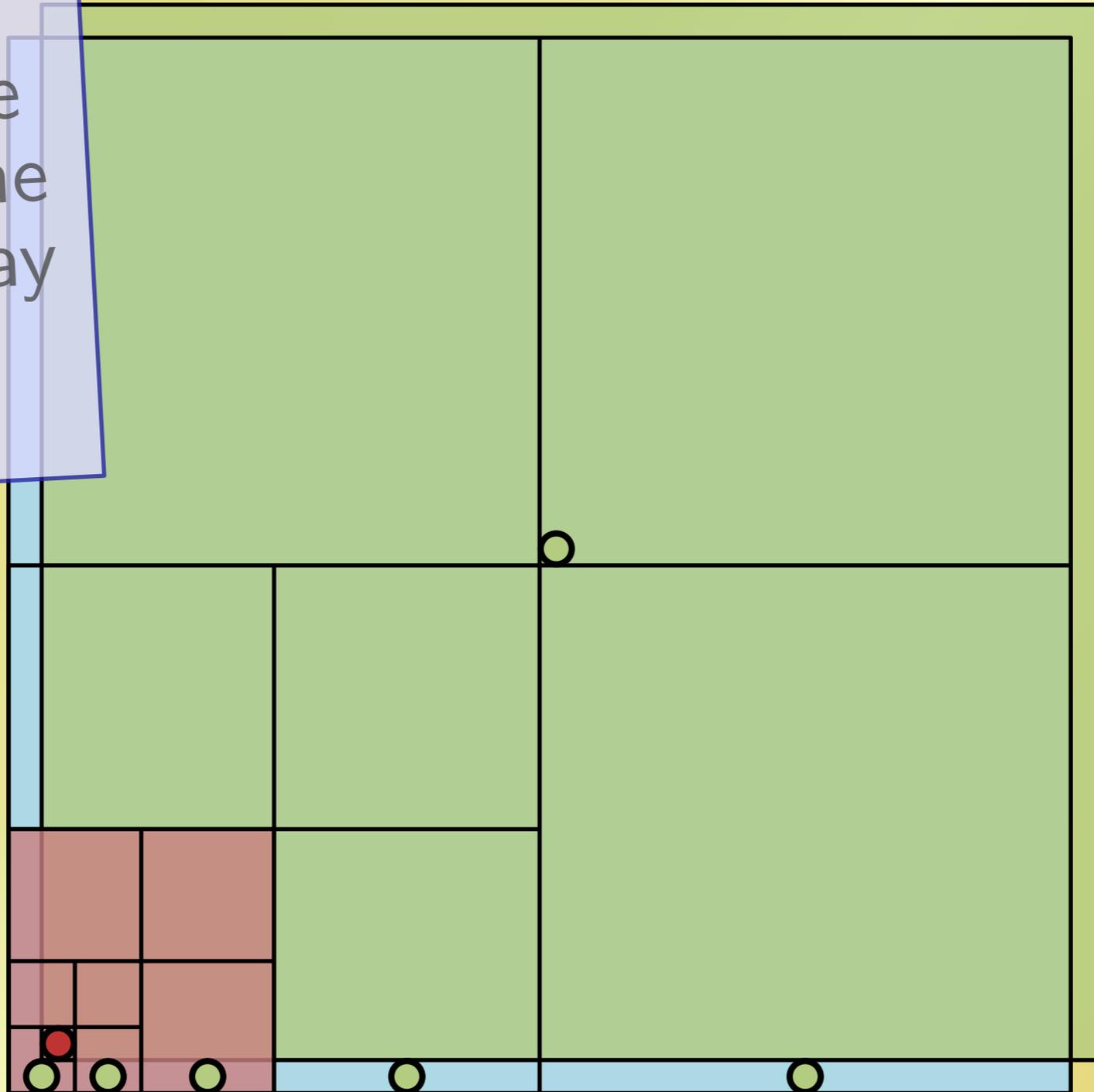
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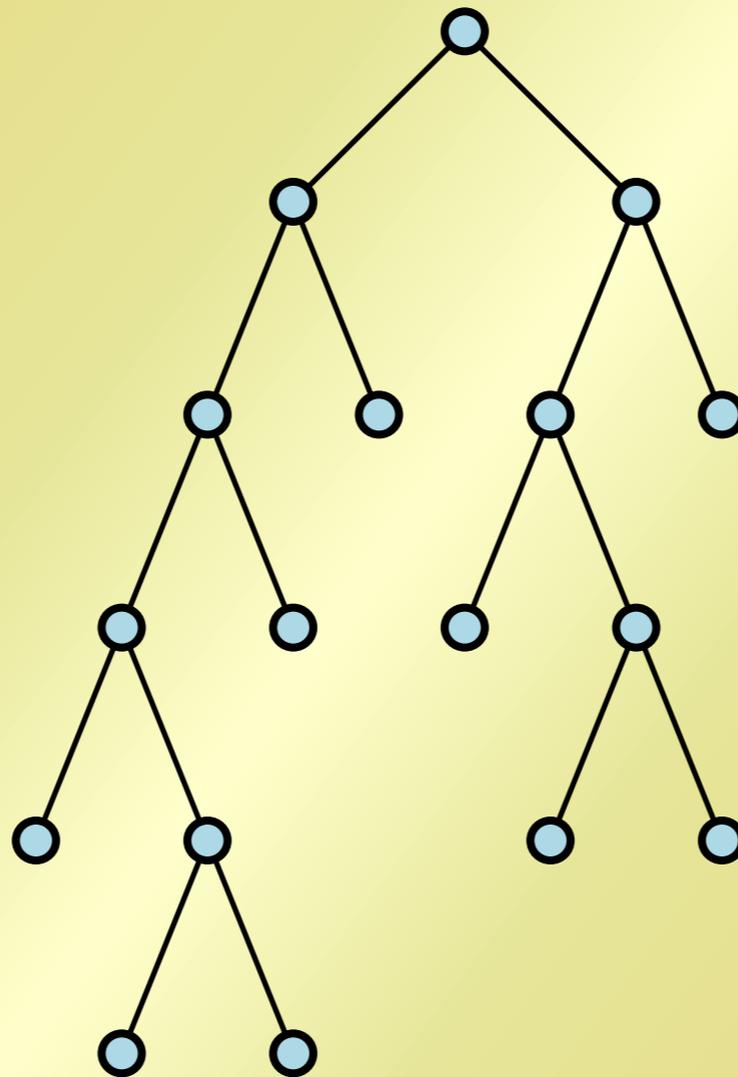
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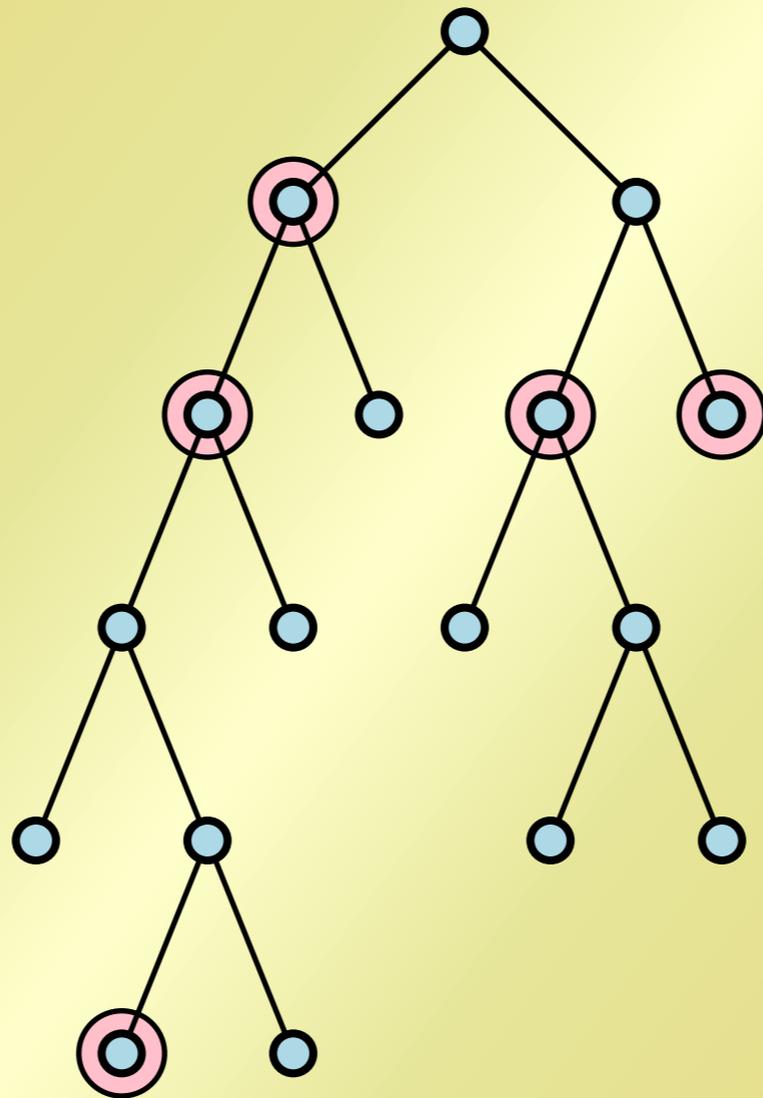
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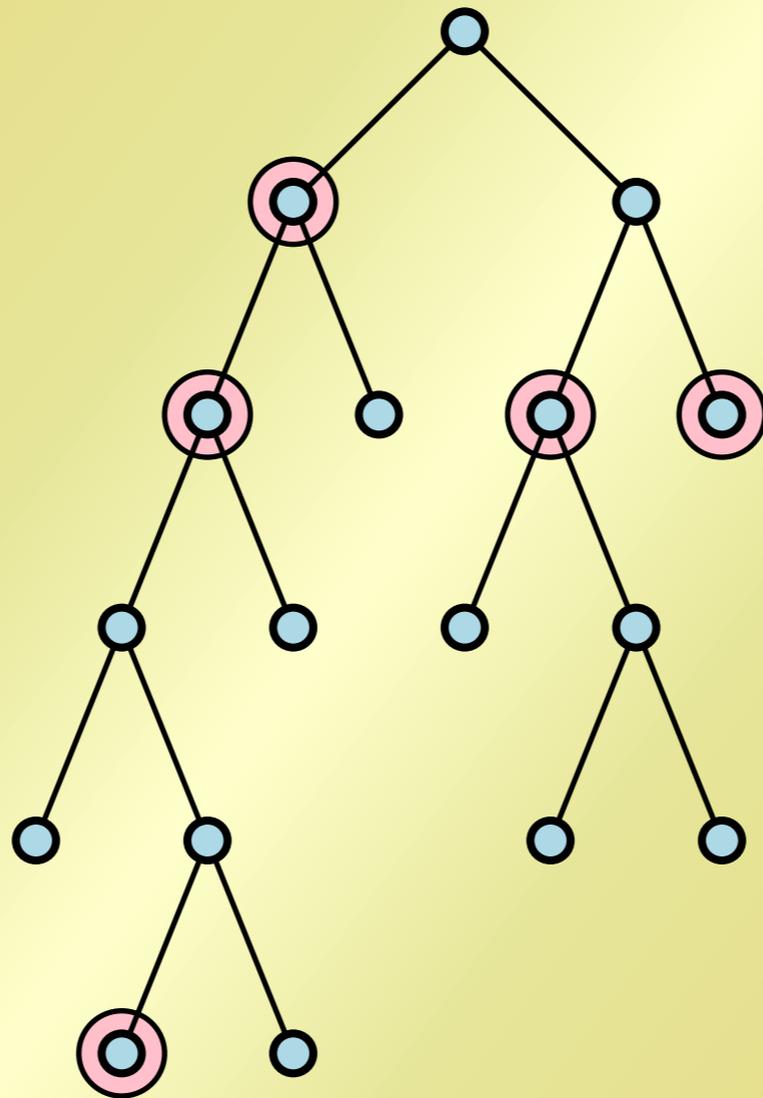
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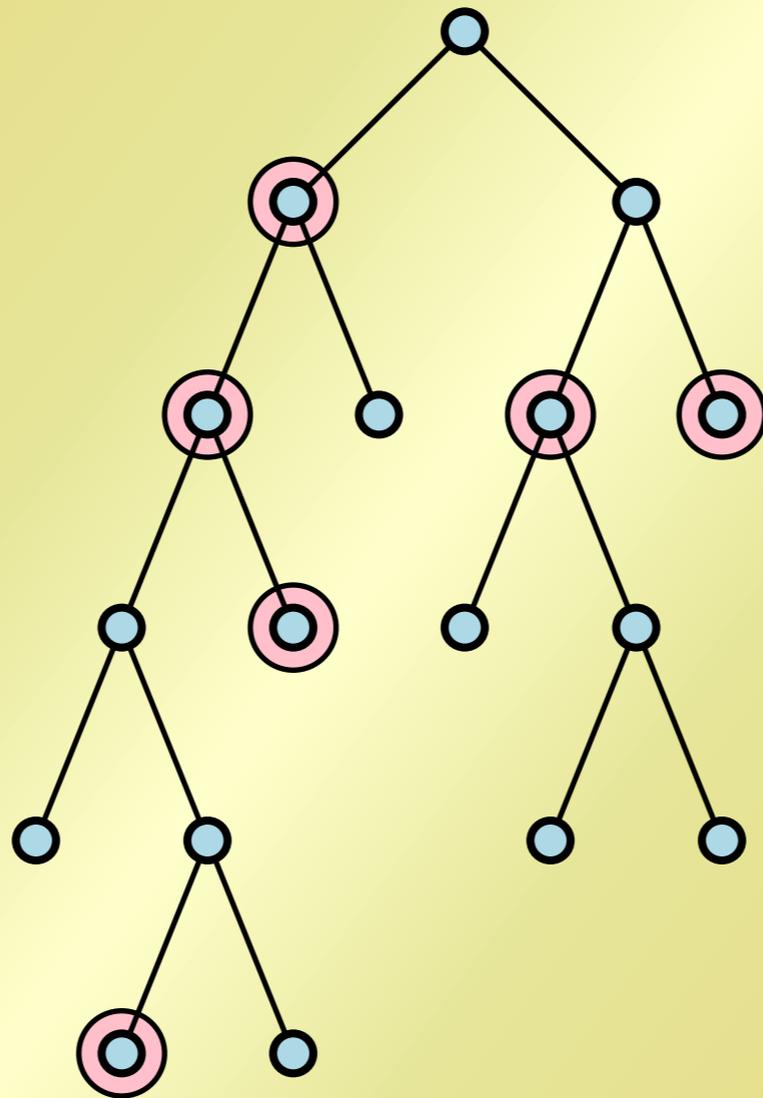
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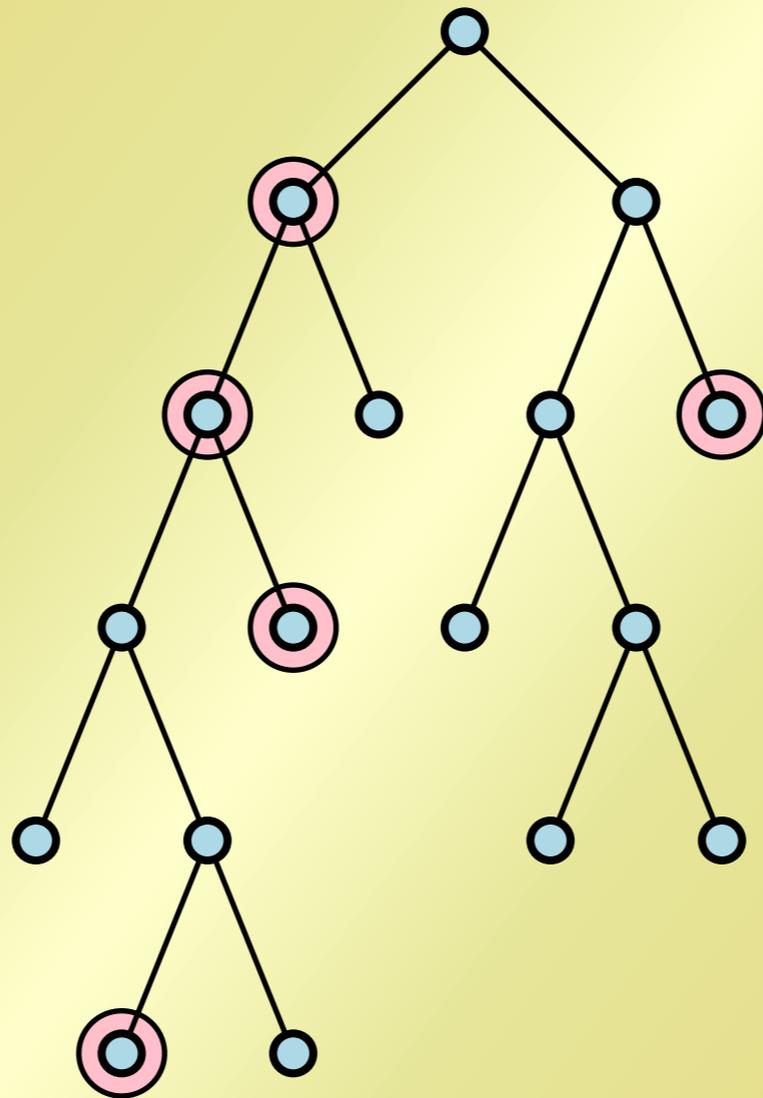
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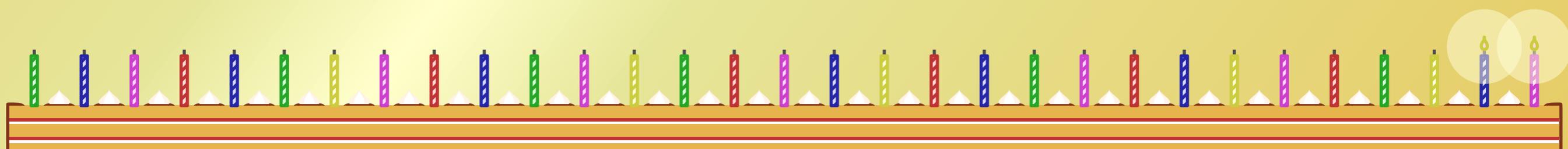
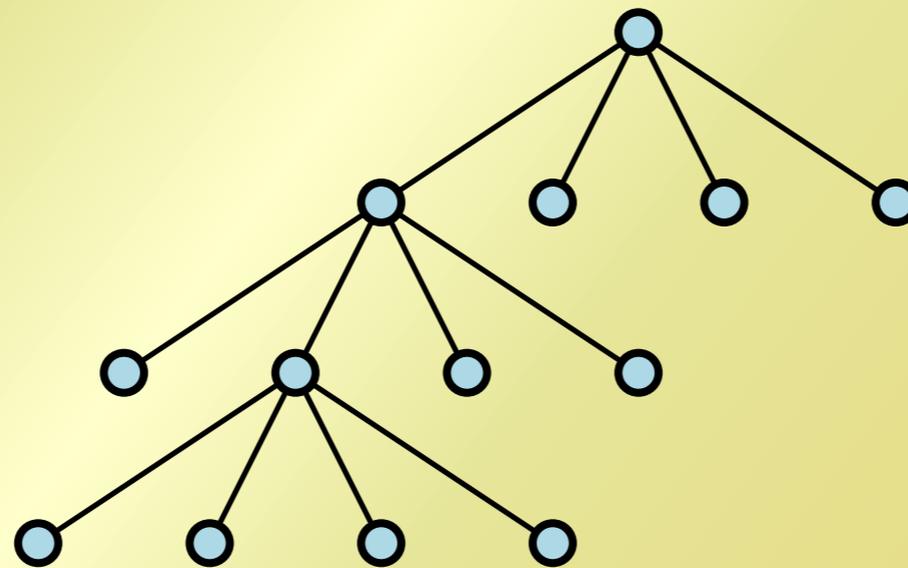
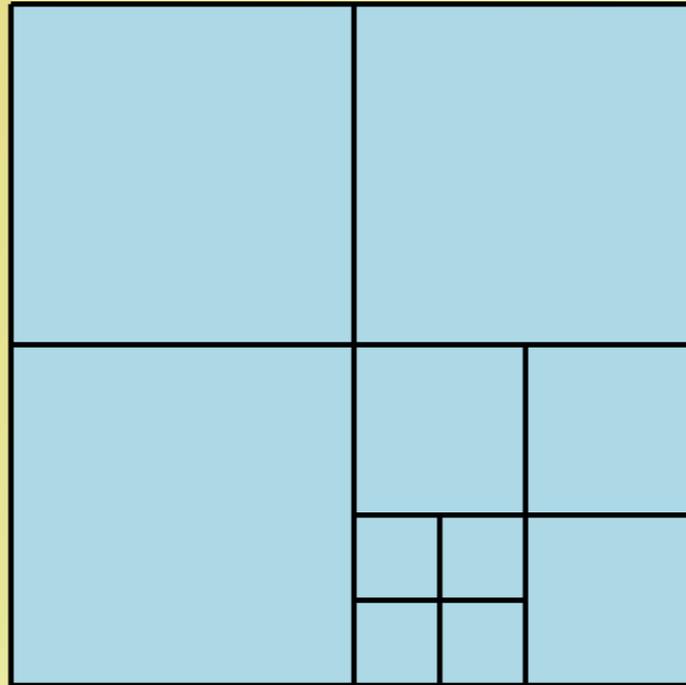




We build 4
MA trees on
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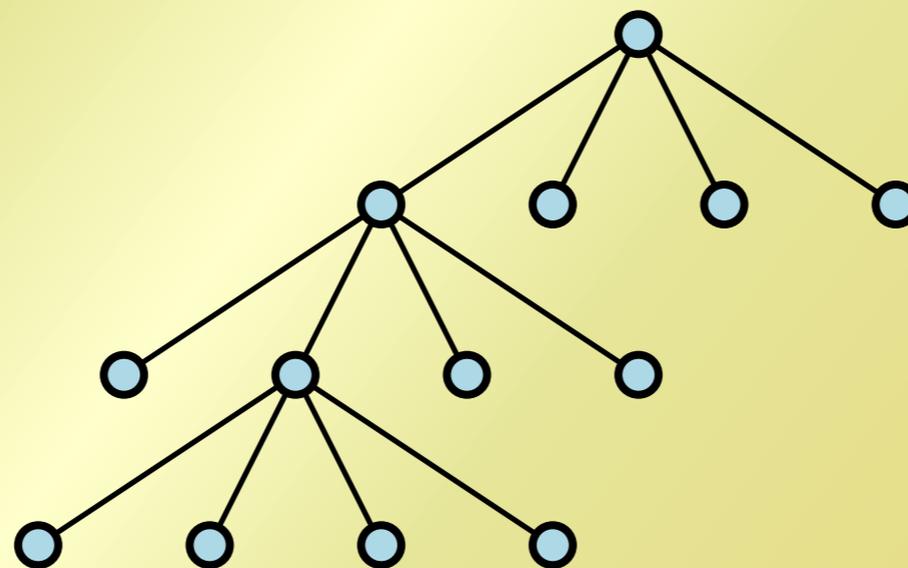
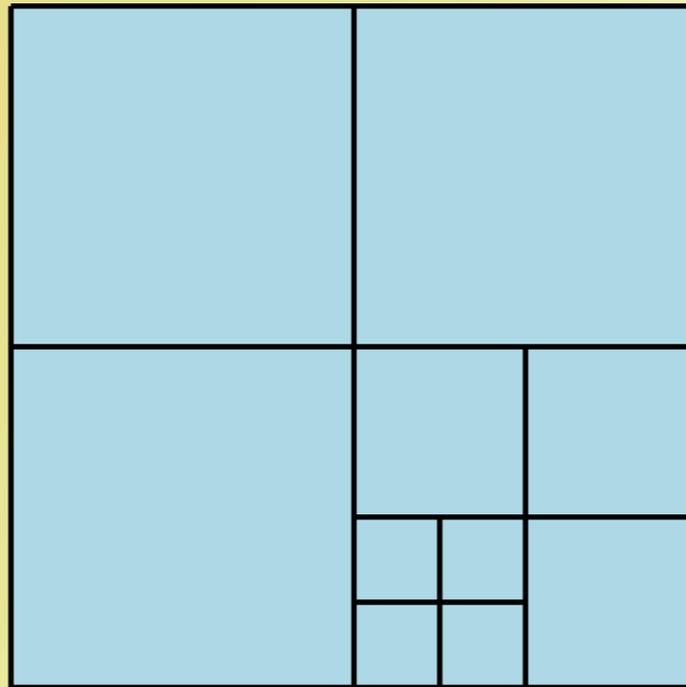


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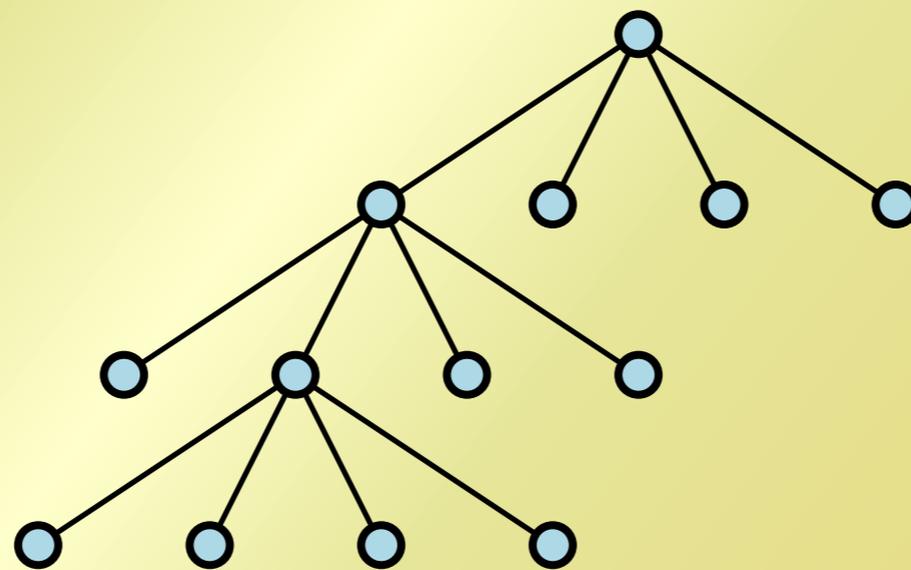
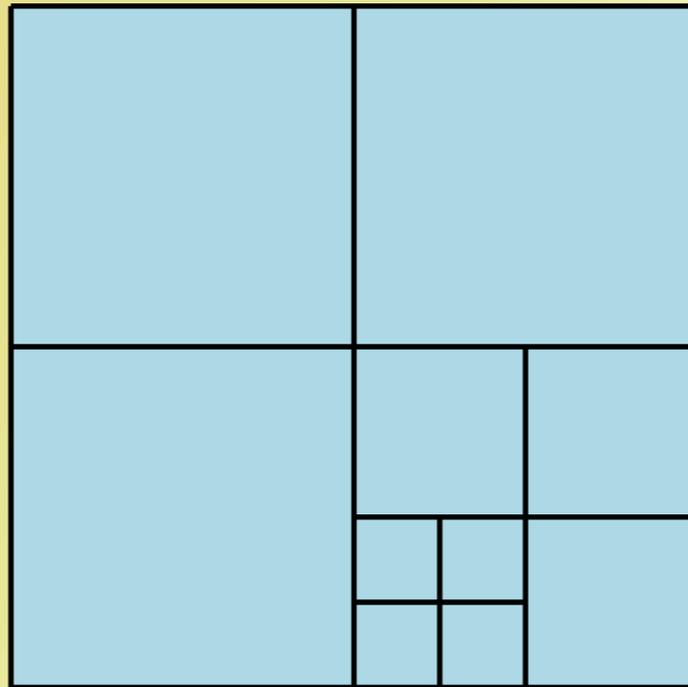
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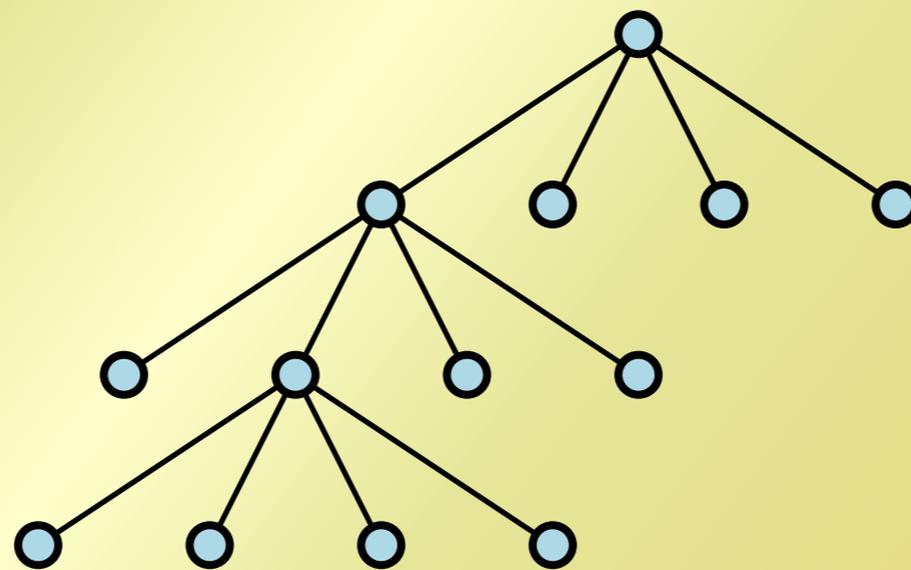
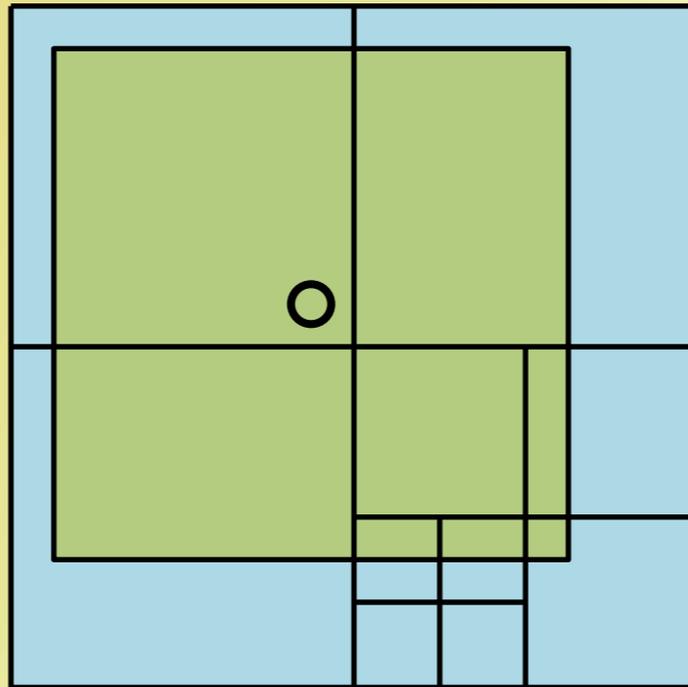
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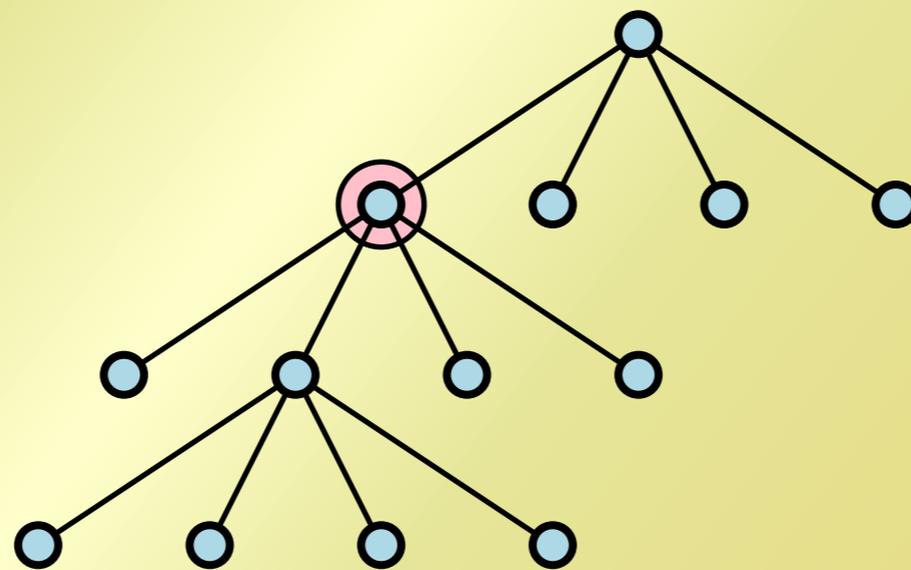
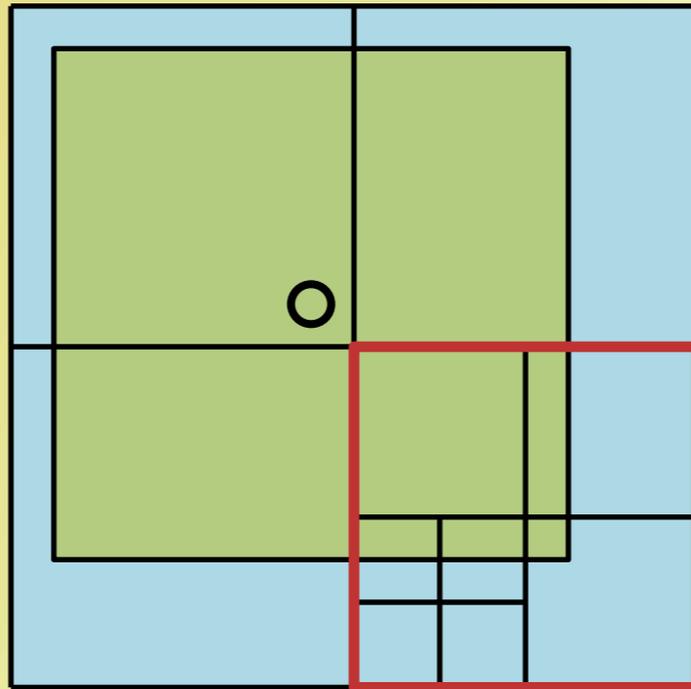
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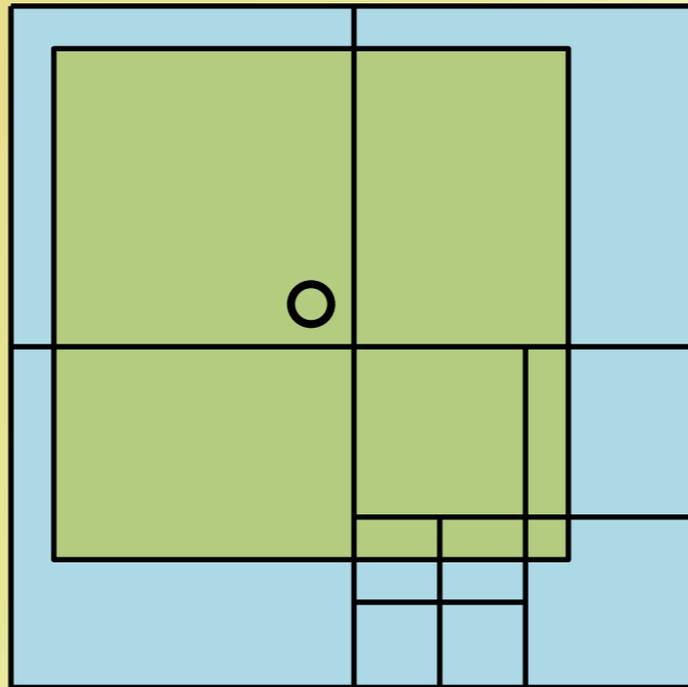
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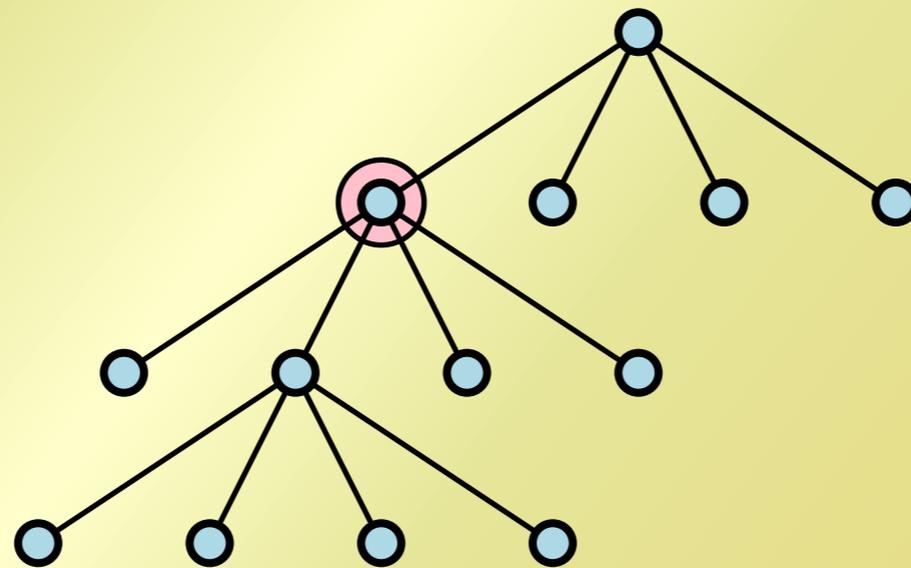
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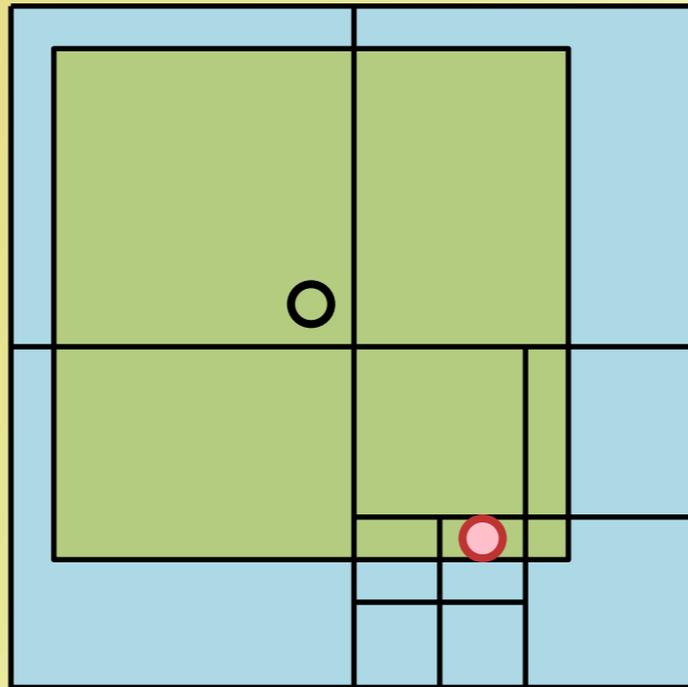
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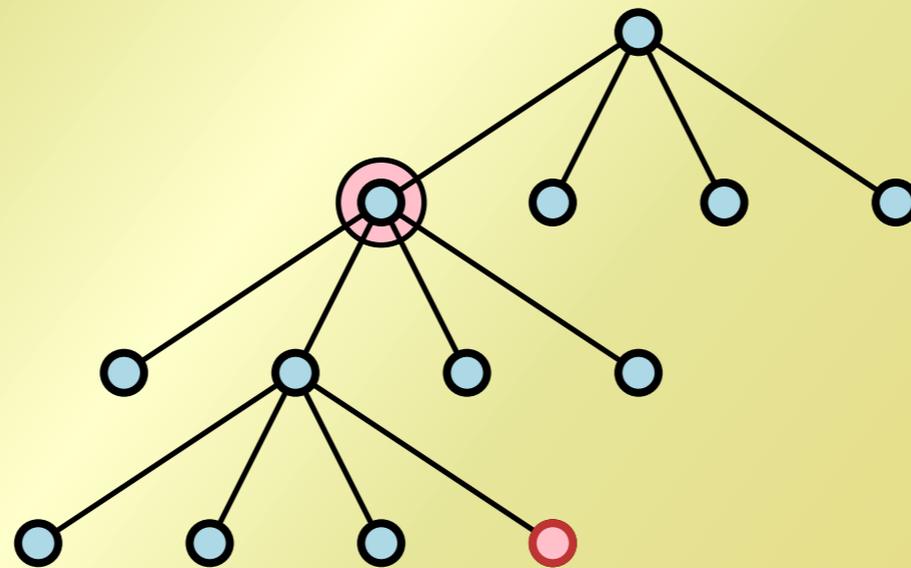
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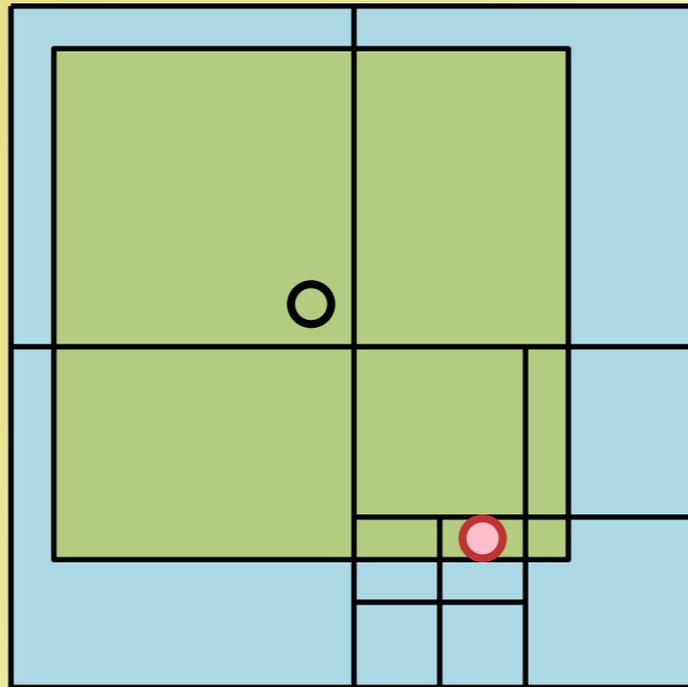
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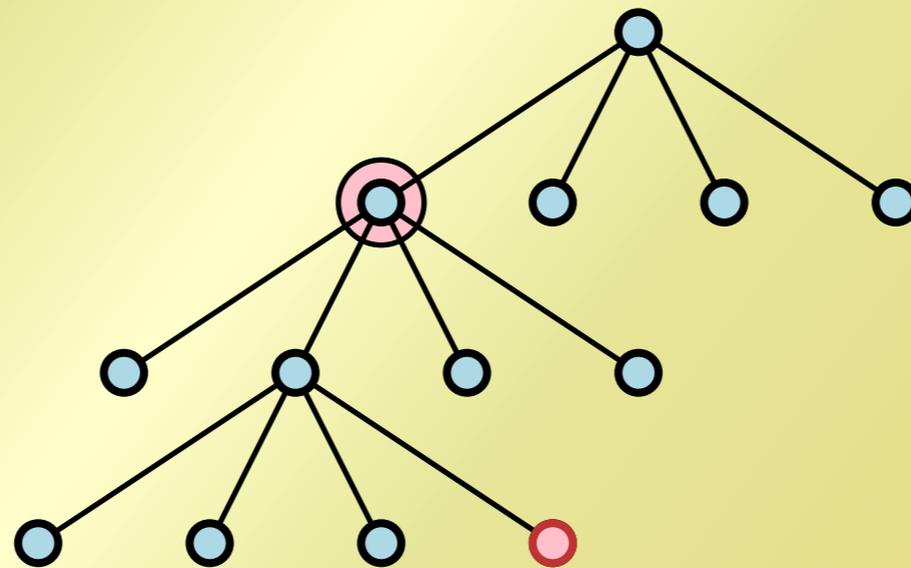
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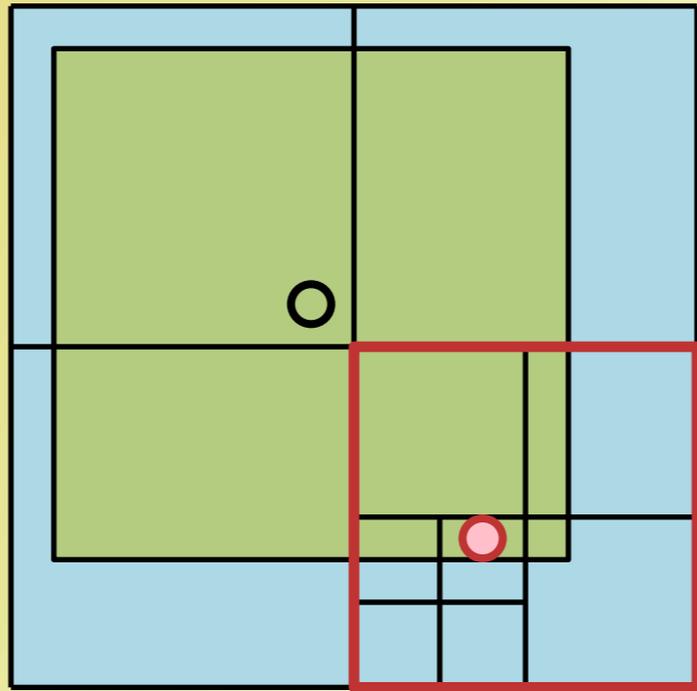
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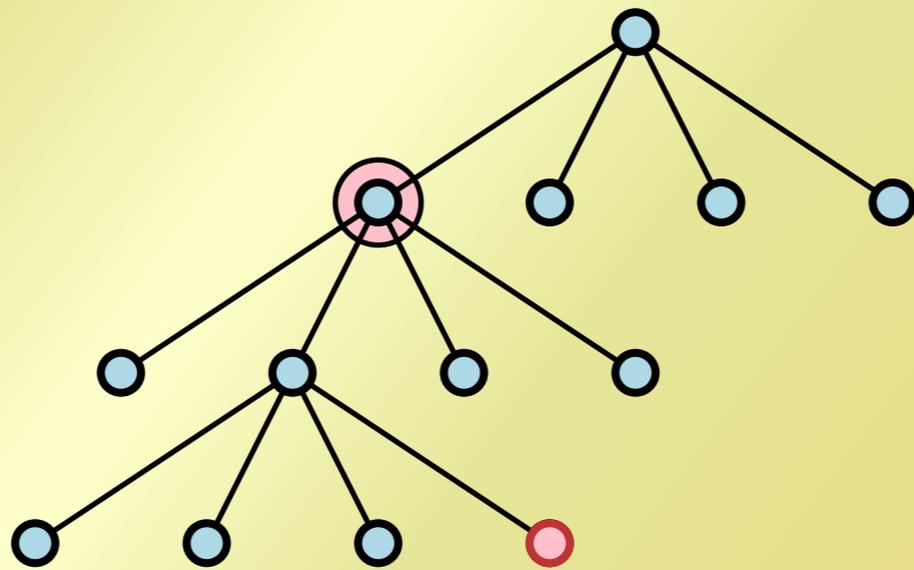
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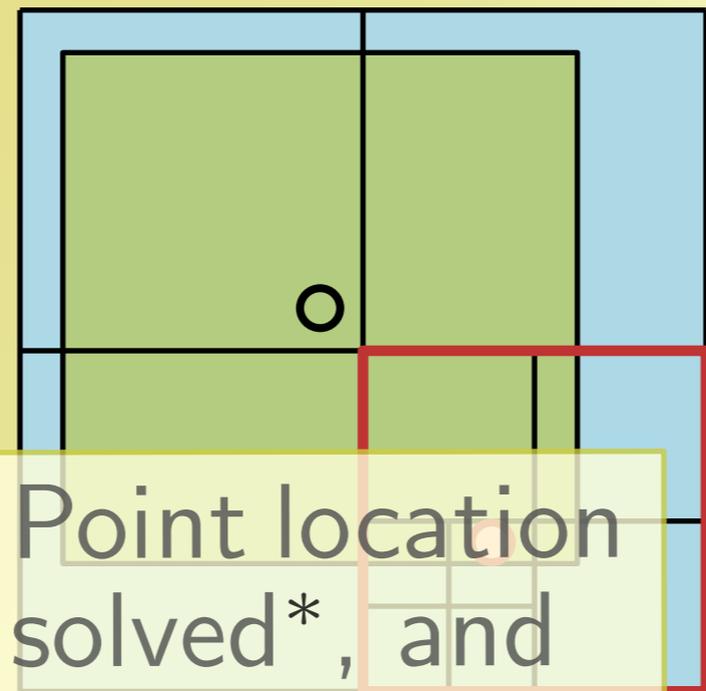
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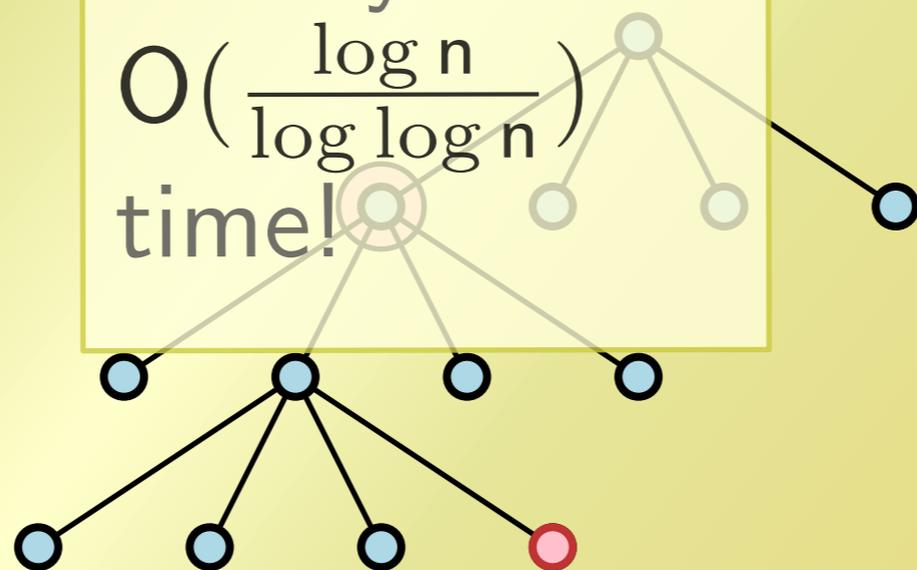
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Point location solved*, and in only $O\left(\frac{\log n}{\log \log n}\right)$ time!



Now, given a query point in a small cell of the quadtree ...

... we can quickly find its first marked ancestor.

* CAUTION! Many details have been swept under the rug. Be extremely careful not to trip when walking on the rug.



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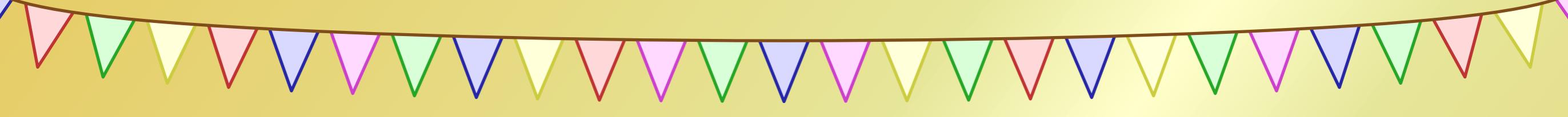
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Do realistic input assumptions help?





THANKS!



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