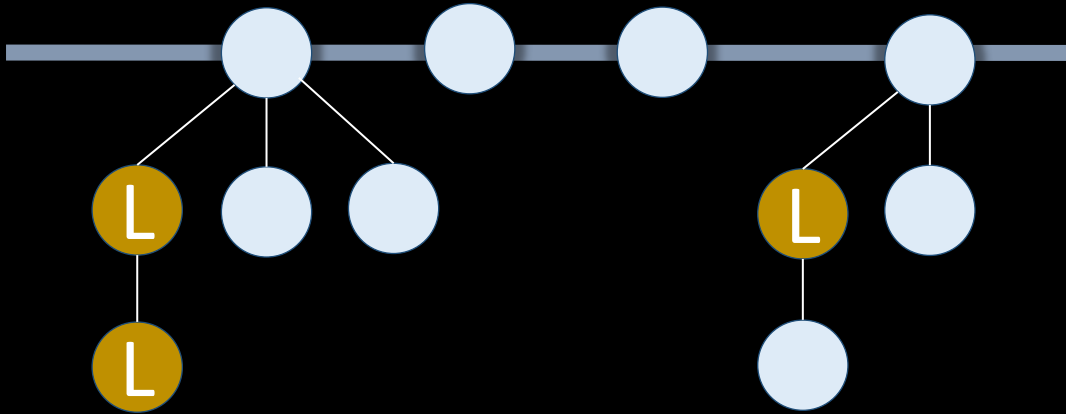
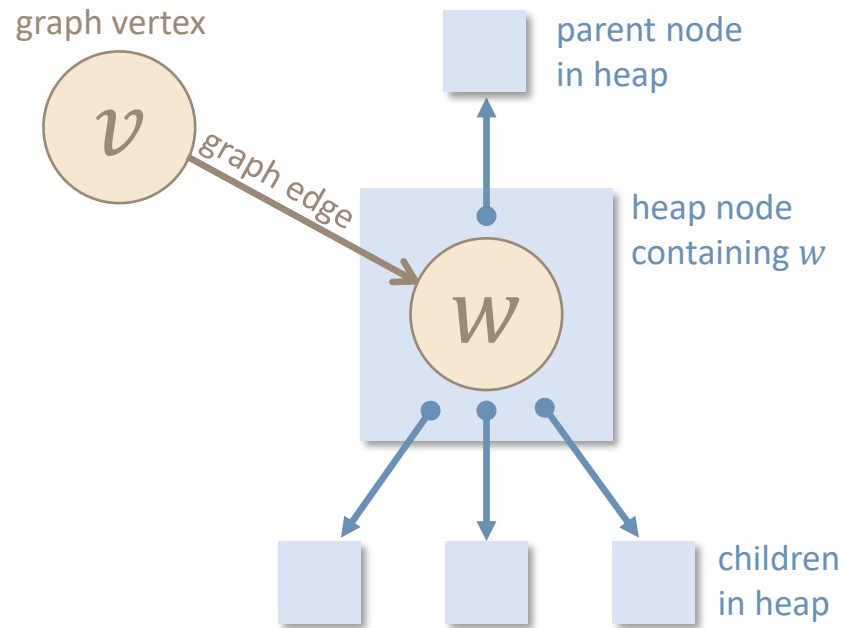


SECTION 7.6

The Fibonacci Heap

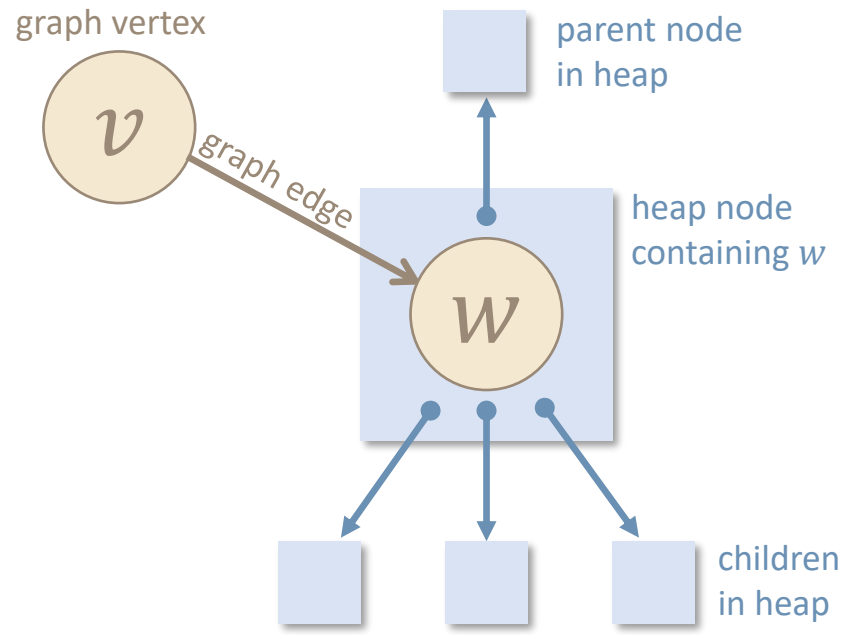


- `push()` — $O(1)$ amortized
Lazy, just adds singleton nodes to the rootlist
- `decreasekey()` — $O(1)$ amortized
Does some work to keep the trees in shape
Adds singleton nodes to the rootlist
- `popmin()` — $O(\log N)$ amortized
Cleans up the rootlist
(at most one tree of any given degree)



```
def dijkstra(g, s):  
    ...  
    toexplore = PriorityQueue()  
    toexplore.push(s, key=0)  
  
    while not toexplore.is_empty():  
        v = toexplore.popmin()  
        for (w,edgcost) in v.neighbours:  
            dist_w = v.distance + edgcost  
            ...  
            toexplore.decreasekey(w, key=dist_w)
```

QUESTION. How can decreasekey be $O(\log N)$?
Doesn't it take $O(N)$ in the first place, to find the heap node that we want to decrease?



```
def dijkstra(g, s):
    ...
    toexplore = PriorityQueue()
    toexplore.push(s, key=0)
    while not toexplore.is_empty():
        v = toexplore.popmin()
        for (w,edgcost) in v.neighbours:
            dist_w = v.distance + edgcost
            ...
            toexplore.decreasekey(w, key=dist_w)
```

Algorithms tick: fib-heap

Fibonacci Heap

In this tick you will implement the Fibonacci Heap. This is an intricate data structure – for some of you, perhaps the most intricate programming you have yet programmed. If you haven't already completed the [dis-set tick](#), that's a good warmup.

Step 1: heap operations

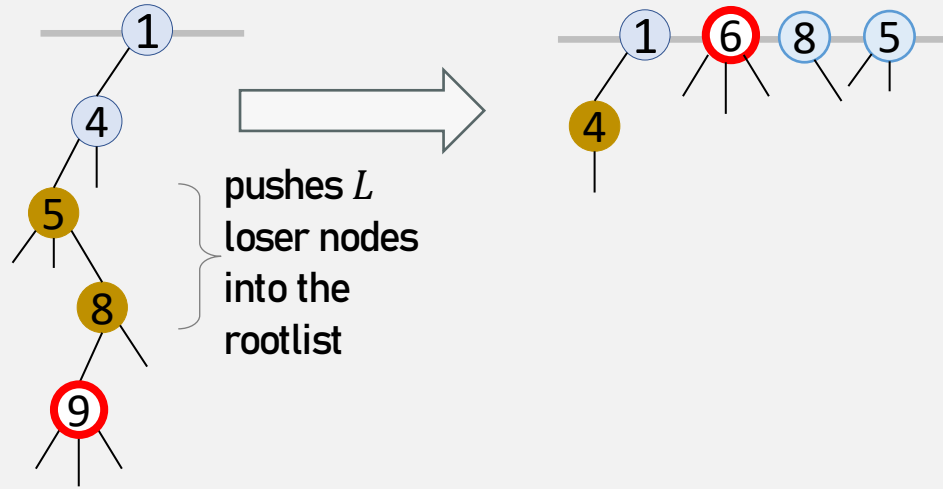
The diagram shows a horizontal line representing the list of root nodes in a Fibonacci heap. Three blue circular nodes are positioned along this line. The leftmost node has three children (blue circles) connected to it by lines. The middle node has one child connected to it. The rightmost node has one child connected to it.

The first step is to implement a `FibNode` class to represent a node in the Fibonacci heap, and a `FibHeap` class to represent the entire heap. Each `FibNode` should store its priority key `k`, and the `FibHeap` should store a list of root nodes as well as the minroot.

SECTION 7.8

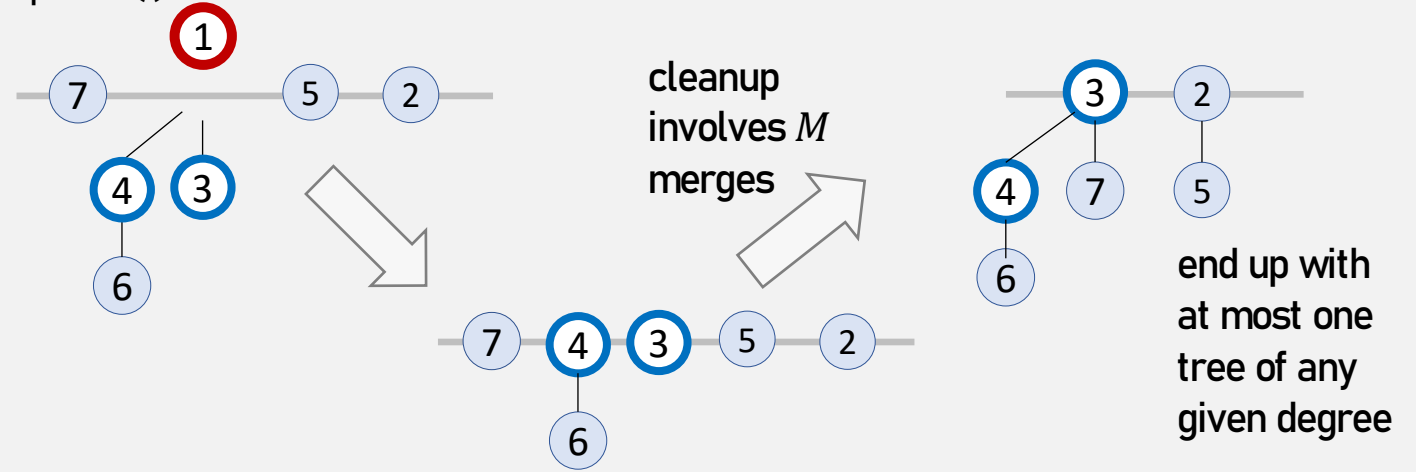
Amortized analysis of the Fibonacci Heap

decreasekey()



decreasekey has true cost $O(L)$
so we want $\Delta\Phi = -L$ to pay for it

popmin()



popmin merges trees in its cleanup phase, true cost $O(M)$
so we want $\Delta\Phi = -M$ to pay for it

$$\Phi = \text{num.roots} + 2 \times \text{num.losers}$$

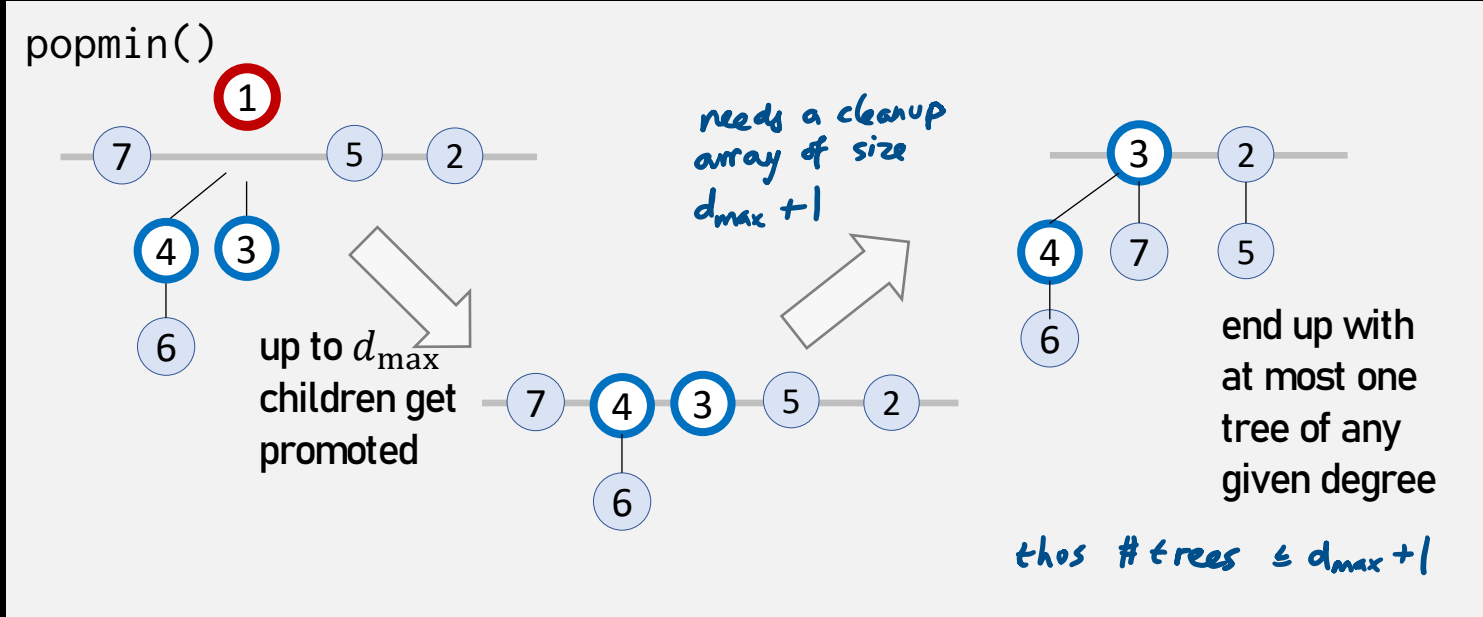
pays in advance for these "uncontrolled" iterations

SECTION 7.8

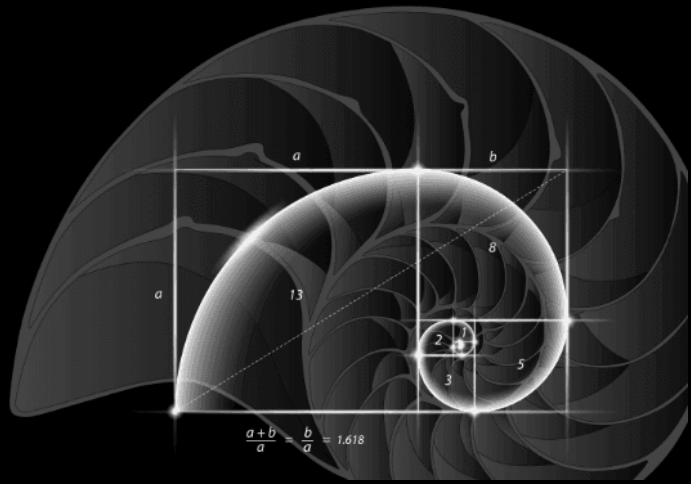
Amortized analysis of the Fibonacci Heap

SHAPE THEOREM

In a Fibonacci heap with N items, every node has degree $\leq \log_{\phi} N$ where ϕ is the golden ratio.



popmin also has to do $O(d_{\max})$ work where d_{\max} is the maximum possible degree in a heap with N items

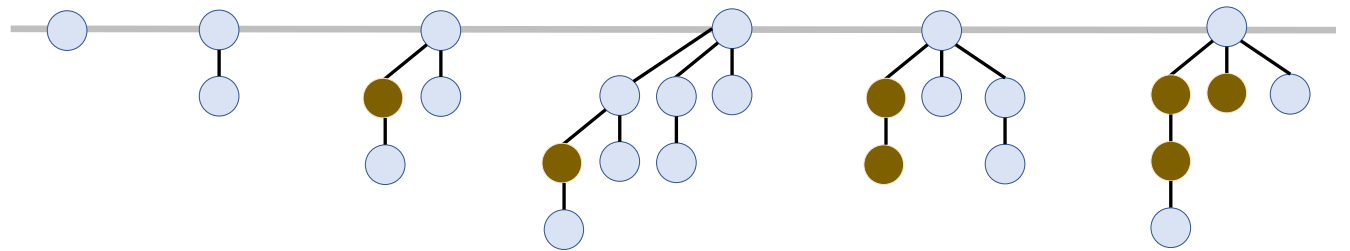


SHAPE THEOREM

In a Fibonacci heap with N items, every node has degree $\leq \log_{\phi} N$

SHAPE LEMMA

Consider a subtree in a Fibonacci heap. If the subtree's root has d children, then the number of nodes in the subtree is $\geq F_{d+2}$ where F_1, F_2, \dots are the Fibonacci numbers



SHAPE THEOREM

In a Fibonacci heap with N items, every node has degree $\leq \log_{\phi} N$

Proof of theorem.

Pick a node with maximum degree, call it d , and consider the subtree rooted at this node.

$$N \geq \text{num.nodes in subtree}$$

$$\geq F_{d+2} \quad \text{by lemma}$$

$$\geq \phi^d \quad \text{linear algebra:}$$

Hence $d \leq \log_{\phi} N$. $F_n = \frac{\phi^n - (-\phi)^n}{\sqrt{5}}$

SHAPE LEMMA

Consider a subtree in a Fibonacci heap. If the subtree's root has d children, then the number of nodes in the subtree is $\geq F_{d+2}$ where F_1, F_2, \dots are the Fibonacci numbers

SHAPE LEMMA

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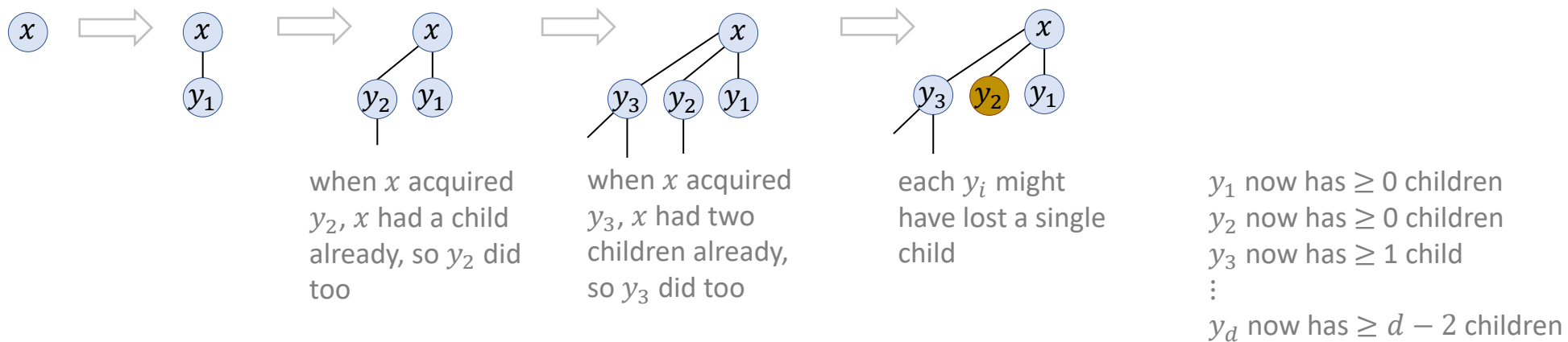
GRANDCHILD RULE

A node x is said to satisfy the grandchild rule if its children can be ordered, call them y_1, \dots, y_d , such that for all $i \in \{1, \dots, d\}$

$$\text{num. grandchildren of } x \text{ via } y_i \geq i - 2$$

ALGORITHMIC CLAIM

In a Fibonacci heap, at every instant in time, every node x satisfies the grandchild rule, when we order its children y_1, \dots, y_d by when they became children of x



SHAPE LEMMA

Consider a subtree in a Fibonacci heap. If the subtree's root has d children, then the number of nodes in the subtree is $\geq F_{d+2}$ where F_1, F_2, \dots are the Fibonacci numbers

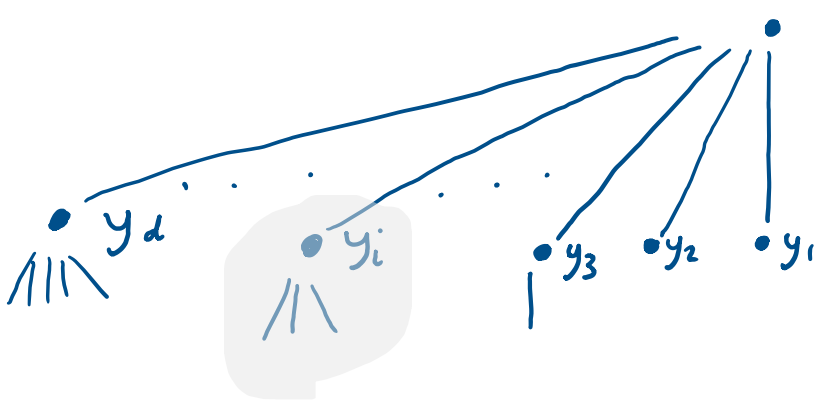
GRANDCHILD RULE

A node x is said to satisfy the grandchild rule if its children can be ordered, call them y_1, \dots, y_d , such that for all $i \in \{1, \dots, d\}$

$$\text{num. grandchildren of } x \text{ via } y_i \geq i - 2$$

MATHEMATICAL CLAIM

Consider a tree where all nodes satisfy the grandchild rule. Let N_d be the smallest number of nodes in a tree whose root has d children. Then $N_d = F_{d+2}$.



$$\text{num. nodes in tree} \geq N_{d-2} + N_{d-3} + \dots + N_1 + N_0 + N_0 + 1$$

$$N_d = N_{d-2} + N_{d-3} + \dots + N_0 + N_0 + 1$$

$$N_{d-1} = N_{d-3} + \dots + N_0 + N_0 + 1$$

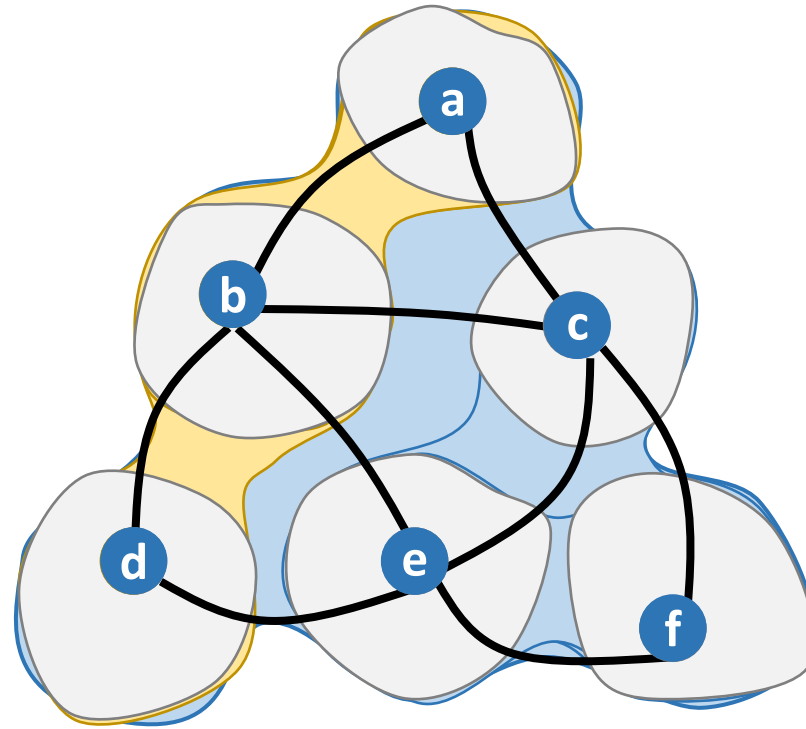
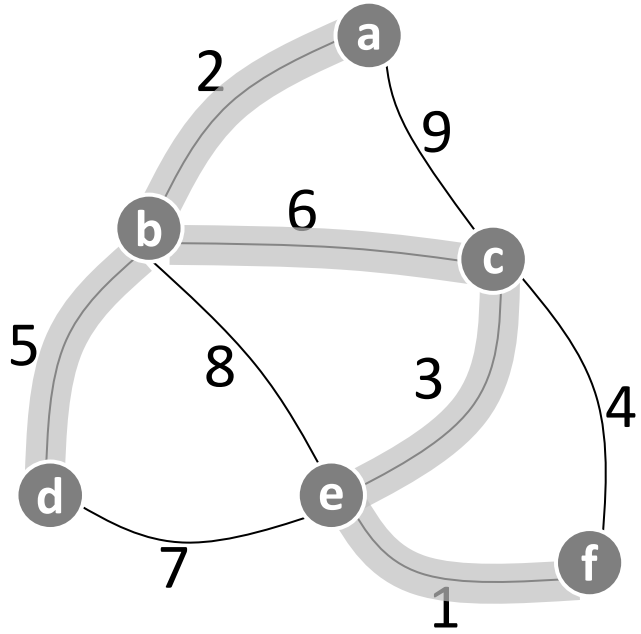
$$\Rightarrow N_d = N_{d-2} + N_{d-1}$$

$\Rightarrow N_d$ is Fibonacci number

child y_i has degree $\geq i - 2$,
so its subtree has $\geq N_{i-2}$ nodes

SECTION 7.9

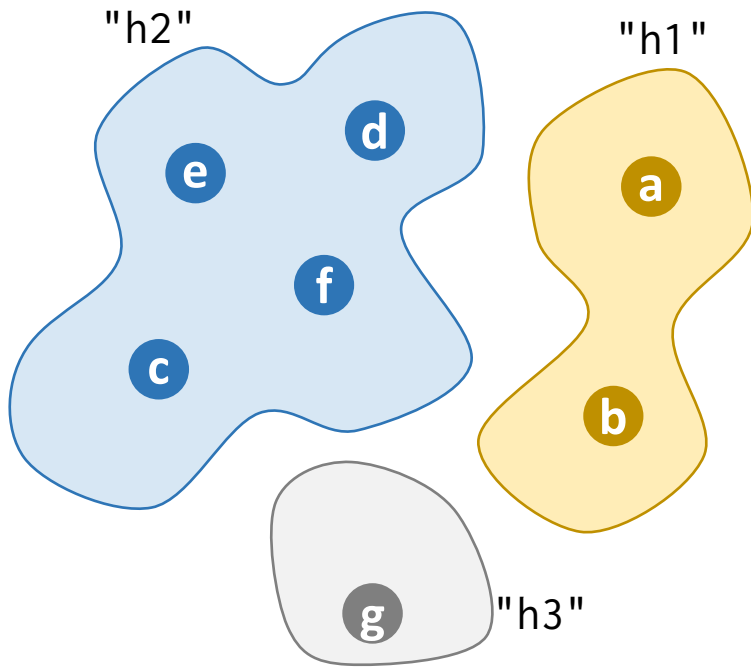
Disjoint sets



```

1 def kruskal(g):
2     tree_edges = []
3     partition = DisjointSet()
4     for v in g.vertices:
5         partition.add_singleton(v)
6     edges = sorted(g.edges, sortkey =  $\lambda(u,v,weight): weight$ )
7
8     for (u,v,edgeweight) in g.edges:
9         p = partition.get_set_with(u)
10        q = partition.get_set_with(v)
11        if p != q:
12            tree_edges.append((u,v))
13            partition.merge(p, q)

```



IMPLEMENTATION 0

```
mysets = {a:"h1", b:"h1", c:"h2", d:"h2", e:"h2", f:"h2", g:"h3"}
```

```
def merge(x,y):  
    for every item in the entire collection:  
        if the item's set is y then update it to be x
```

AbstractDataType DisjointSet:

```
# Holds a dynamic collection of disjoint sets
```

```
# Add a new set consisting of a single item (assuming it's not been added already)
```

```
add_singleton(Item x)
```

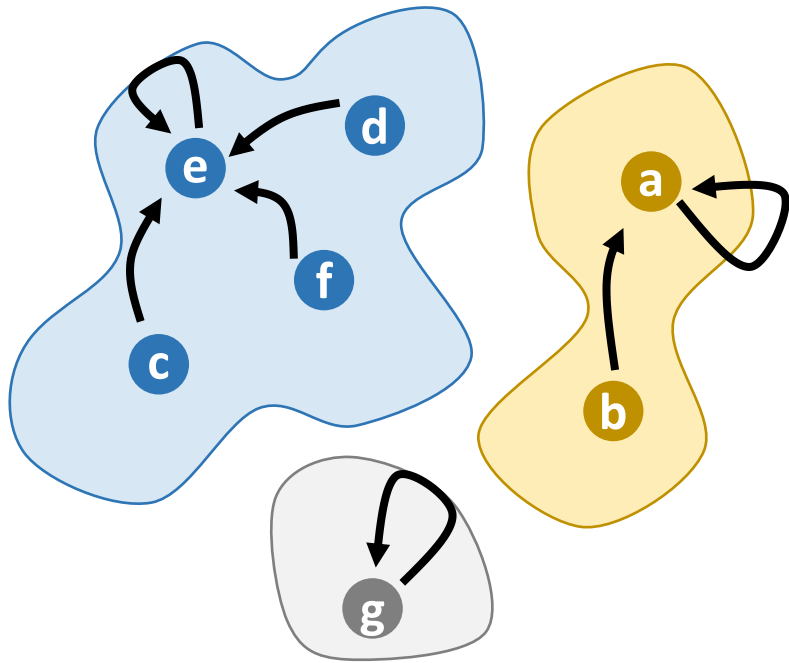
```
# Return a handle to the set containing an item.
```

```
# The handle must be stable, as long as the DisjointSet is not modified.
```

```
Handle get_set_with(Item x)
```

```
# Merge two sets into one
```

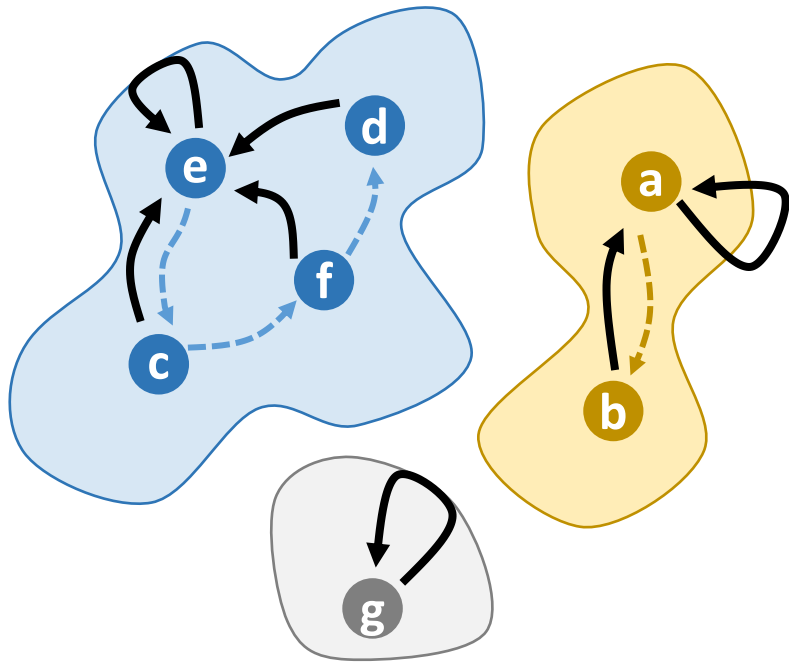
```
merge(Handle x, Handle y)
```



IMPLEMENTATION 0'

Each item points to a representative item for its set

`mysets = {a:a, b:a, c:e, d:e, e:e, f:e, g:g}`

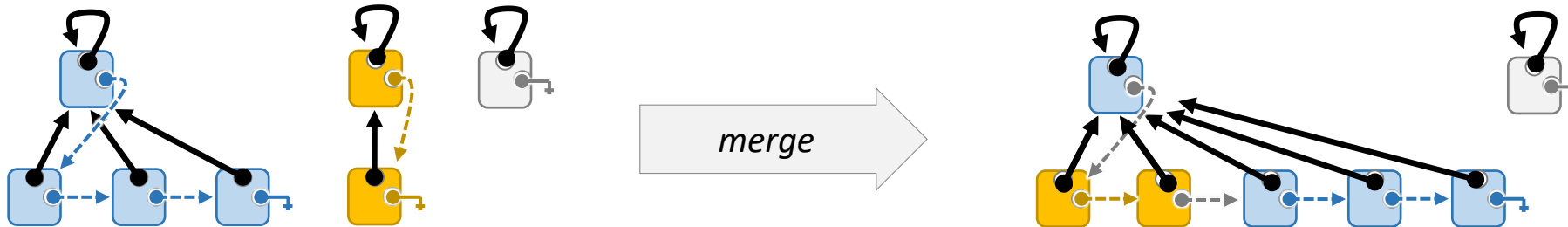


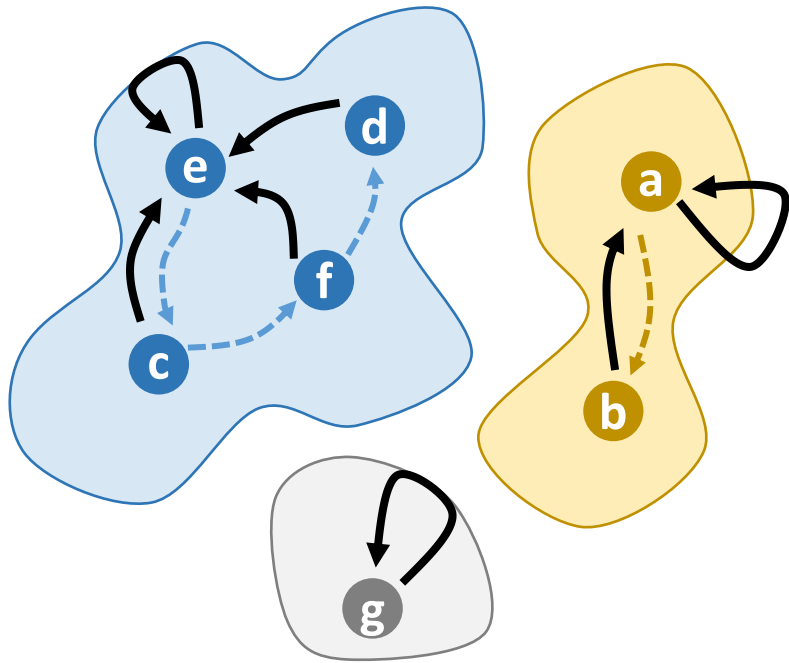
IMPLEMENTATION 1 "FLAT FOREST"

Each item points to a representative item for its set
 Each set has a linked list, starting at its representative

```
def merge(x,y):
  for every item in set y:
    update it to belong to set x
```

```
def get_set_with(x):
  return x's parent
```



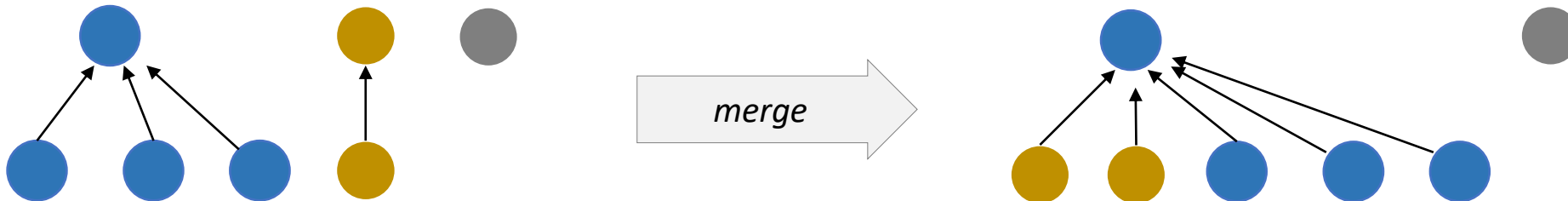


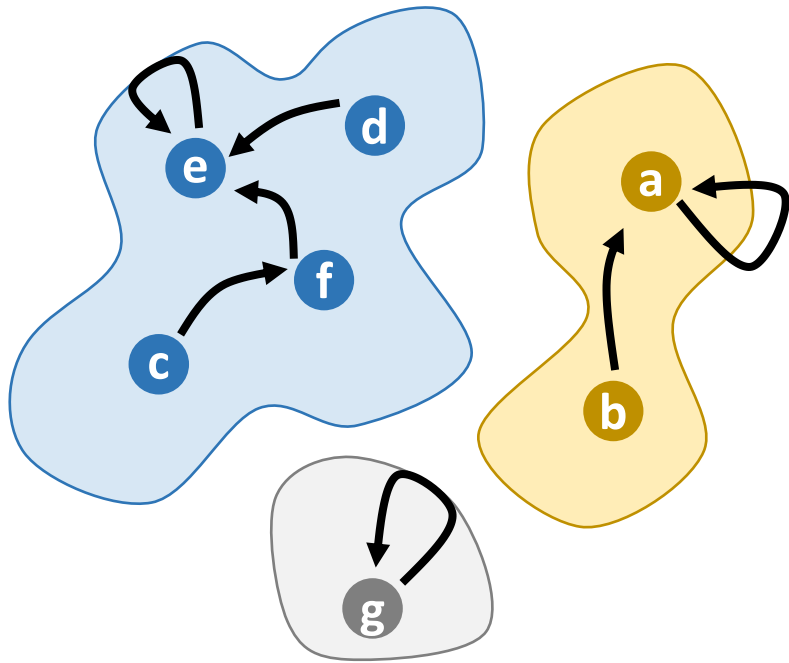
IMPLEMENTATION 1 "FLAT FOREST"

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    return x's parent
```





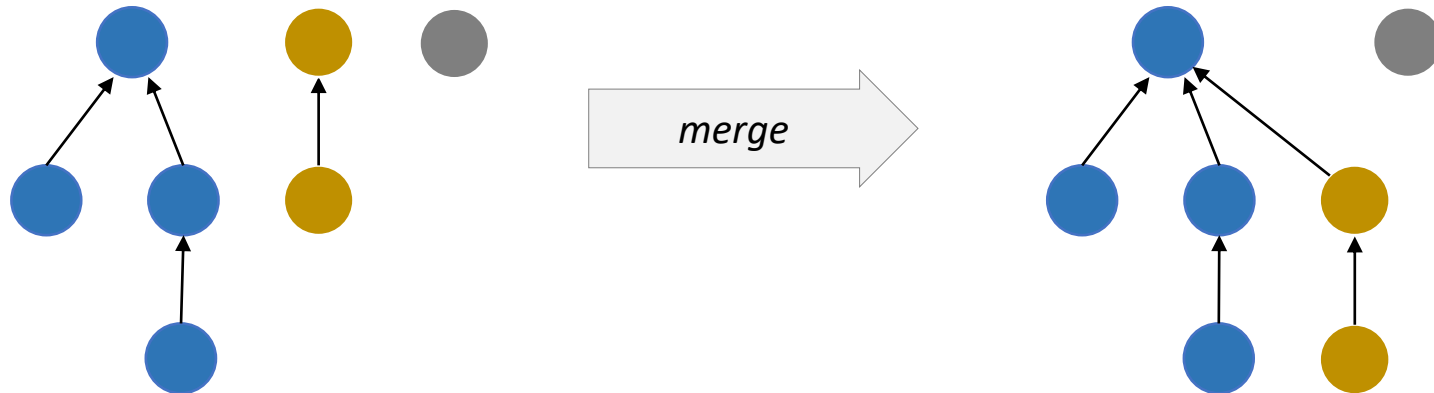
IMPLEMENTATION 2 “DEEP FOREST”

Sets are stored as trees

Use the root item to represent the set

```
def merge(x,y):
    update one of the roots to point to the other
```

```
def get_set_with(x):
    walk up the tree from x to the root
    return this root
```

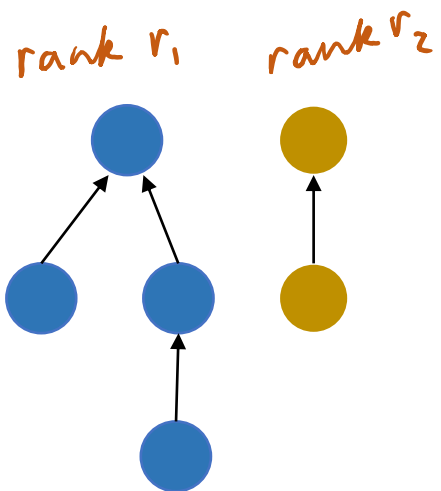


QUESTION. What's a sensible heuristic for merge, to speed up get_set_with?

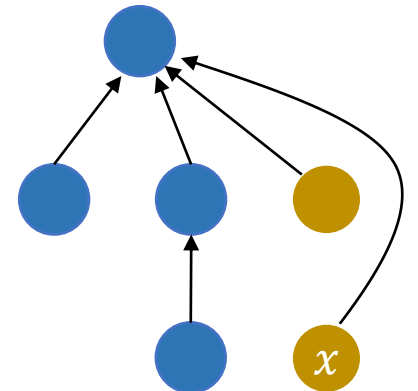
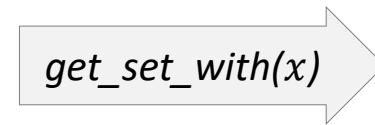
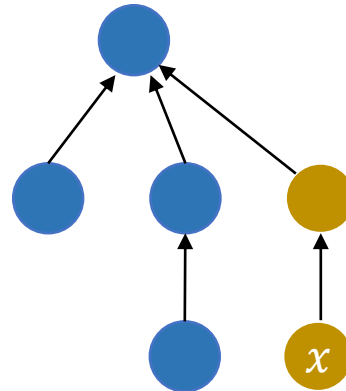
IMPLEMENTATION 3 "LAZY FOREST"

```
def merge(x,y):  
    as before, using the Union by Rank heuristic
```

```
def get_set_with(x):  
    walk up the tree from x to the root  
    walk up again, and make all items point to root  
    return this root
```



$$\text{new rank} = \begin{cases} \max(r_1, r_2) & \text{if } r_1 \neq r_2 \\ r_1 + 1 & \text{if } r_1 = r_2 \end{cases}$$



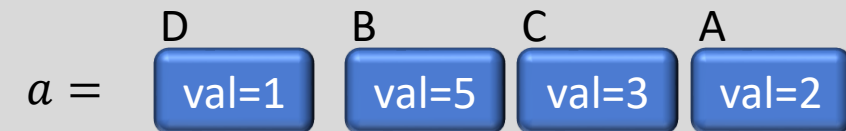
Can we 'manifest' our workings so that subsequent operations benefit?

```
0 def selectSort(a):
1     """BEHAVIOUR: Run the selectsort algorithm on the integer
2     array a, sorting it in place.
3
4     PRECONDITION: array a contains len(a) integer values.
5
6     POSTCONDITION: array a contains the same integer values as before,
7     but now they are sorted in ascending order."""
8
9     for k from 0 included to len(a) excluded:
10        # ASSERT: the array positions before a[k] are already sorted
11
12        # Find the smallest element in a[k:END] and swap it into a[k]
13        iMin = k
14        for j from iMin + 1 included to len(a) excluded:
15            if a[j] < a[iMin]:
16                iMin = j
17        swap(a[k], a[iMin])
```



1. Find the lowest value, and put it at the front

- Is B.val < A.val? No.
- Is C.val < A.val? No.
- Is D.val < A.val? Yes.
- Swap A and D



2. Find the second-lowest in [B,C,A]

we had two useful pieces of information, but we didn't keep them in

Aggregate complexity analysis

Any m operations on up to N items takes

$$O(m + N \log N)$$

[Ex. sheet 6 q. 13]

Flat Forest

(with weighted-union)

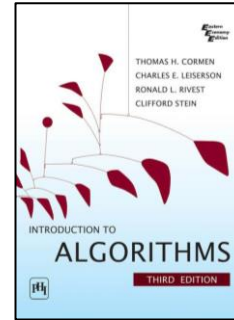
Deep Forest

(with union-by-rank)

Lazy Forest

(with union-by-rank + path compression)

$$O(m \log N)$$



$$O(m \alpha(N))$$

$$\alpha(N) = 0 \quad \text{for } N = 0, 1, 2$$

$$= 1 \quad \text{for } N = 3$$

$$= 2 \quad \text{for } N = 4 \dots 7$$

$$= 3 \quad \text{for } N = 8 \dots 2047$$

$$= 4 \quad \text{for } N = 2048 \dots 10^{80}$$

Aggregate complexity analysis

Any m operations on up to N items takes

$$O(m + N \log N)$$

Flat Forest

(with weighted-union)

$$O(m \log N)$$

Deep Forest

(with union-by-rank)

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(with union-by-rank + path compression)

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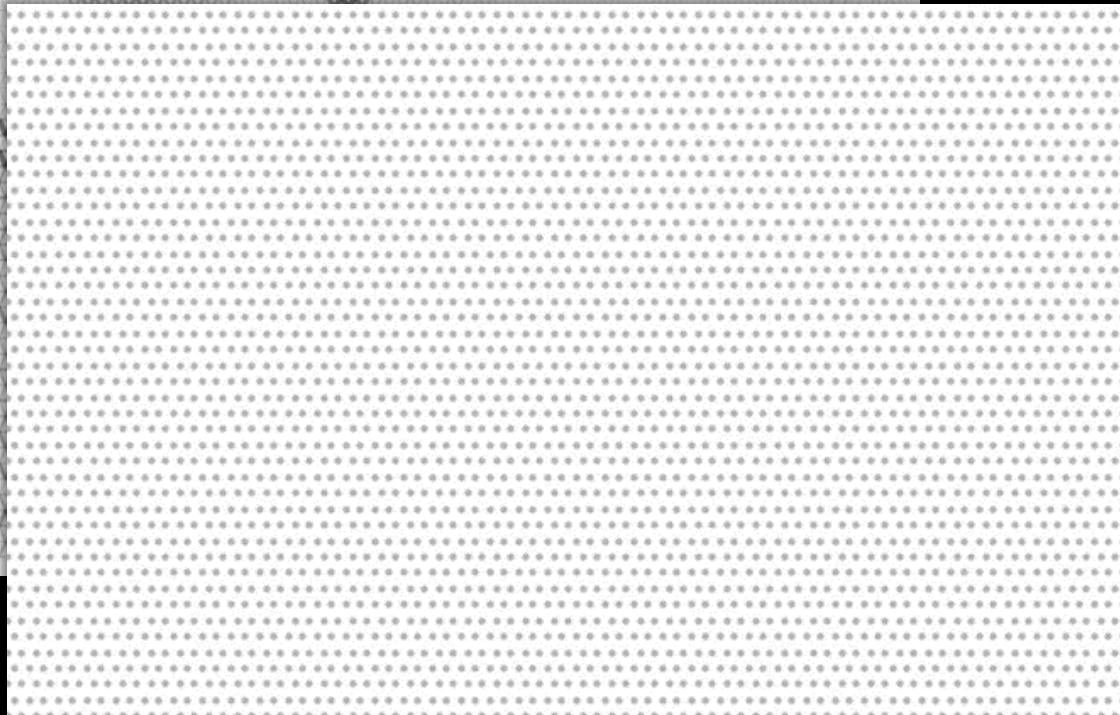
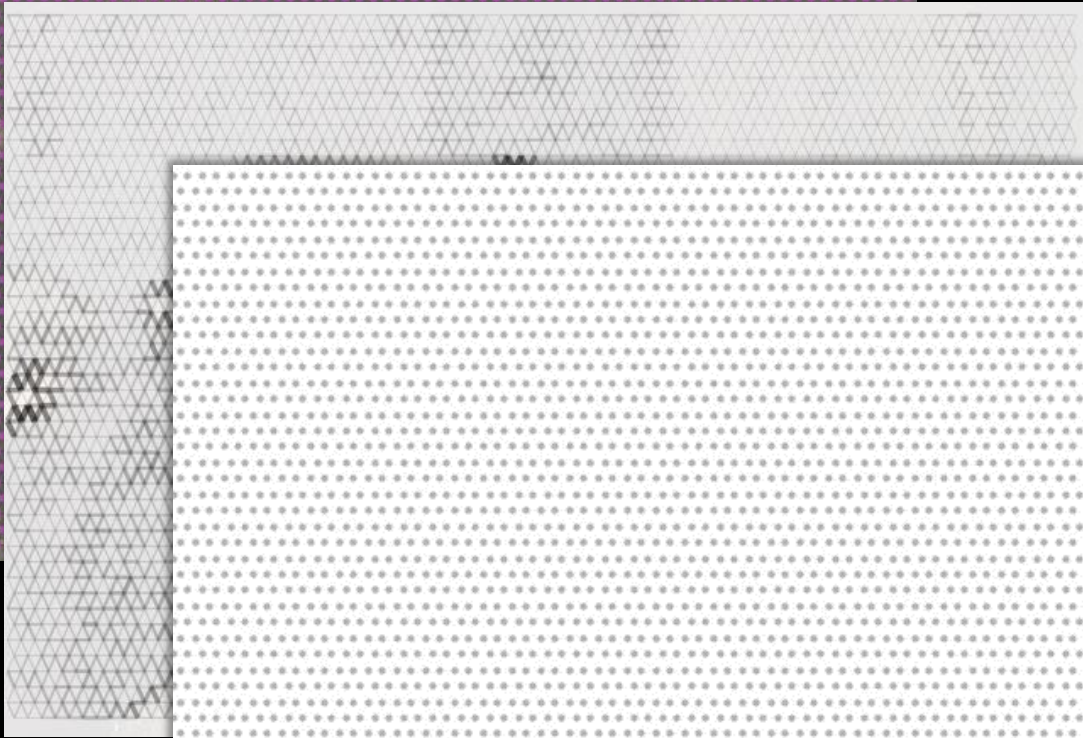
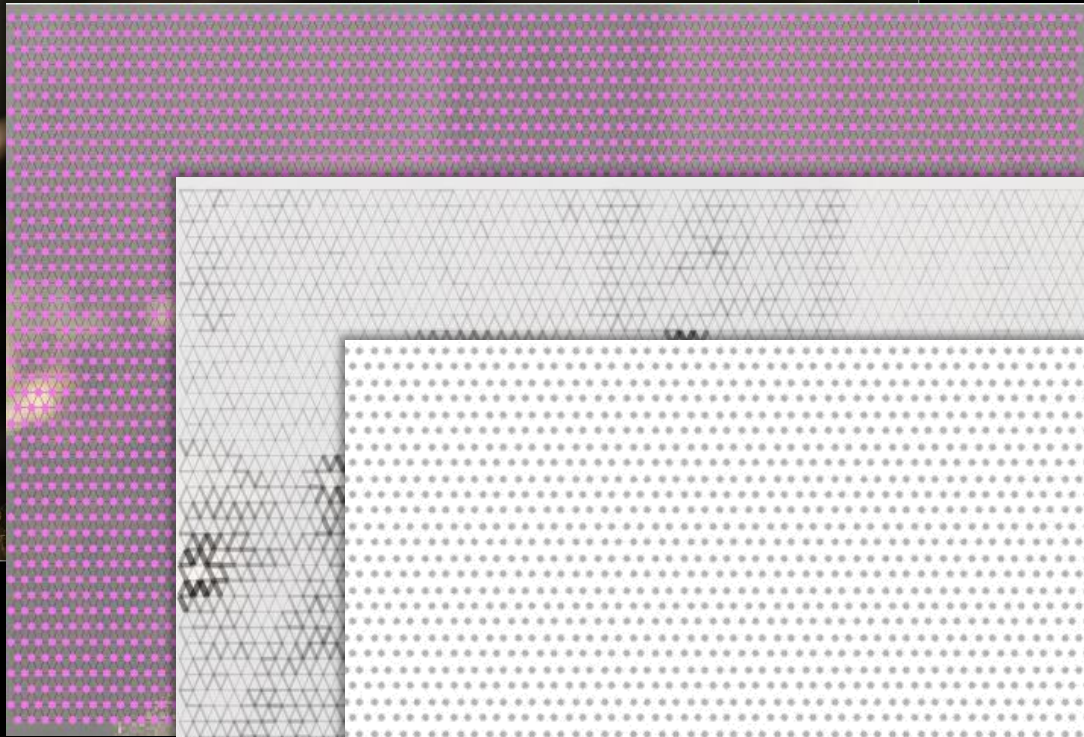
$$= 1 \quad \text{for } N = 3$$

$$= 2 \quad \text{for } N = 4 \dots 7$$

$$= 3 \quad \text{for } N = 8 \dots 2047$$

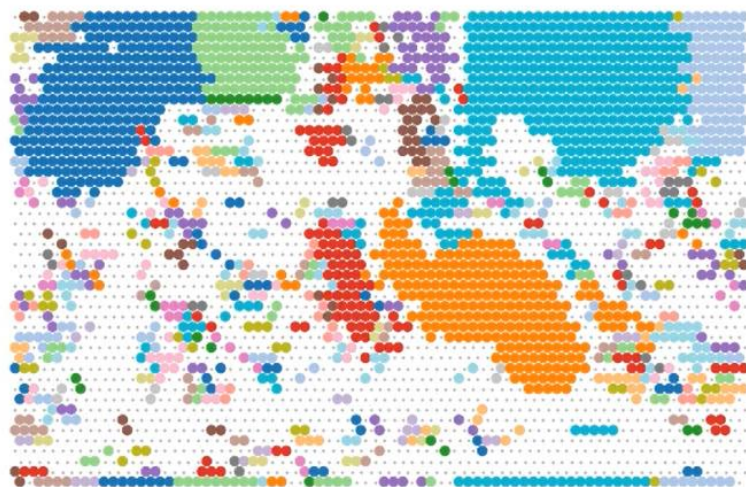
$$= 4 \quad \text{for } N = 2048 \dots 10^{80}$$

1. take a handsome stoat
2. define a graph
vertices on a grid, and edges between adjacent grid cells
3. assign edgeweights
weight=low means vertices have similar colours
4. run Kruskal
and find clusters of similar colour

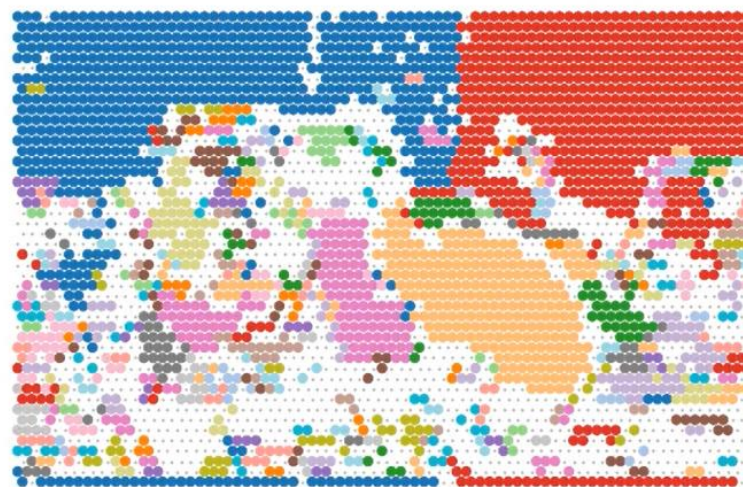




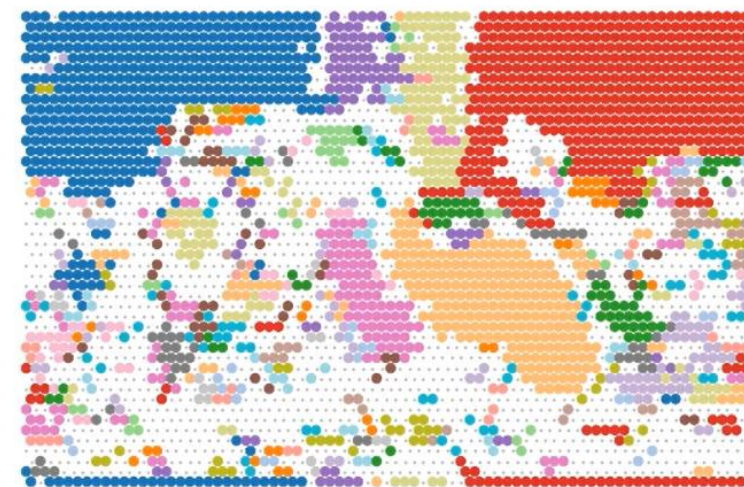
flat



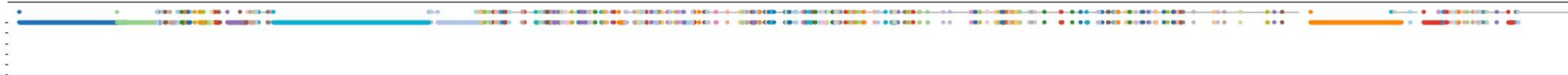
deep



lazy



flat



deep



lazy

