

Minimum sum multicoloring on the edges of trees

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Minimum sum multicoloring

- **Given:** a graph $G(V, E)$, and demand function $x: V \rightarrow \mathbb{N}$
- **Find:** an assignment of $x(v)$ colors (integers) to every vertex v , such that neighbors receive disjoint sets

Finish time: $f(v)$ of vertex v is the largest color assigned to it in the coloring.

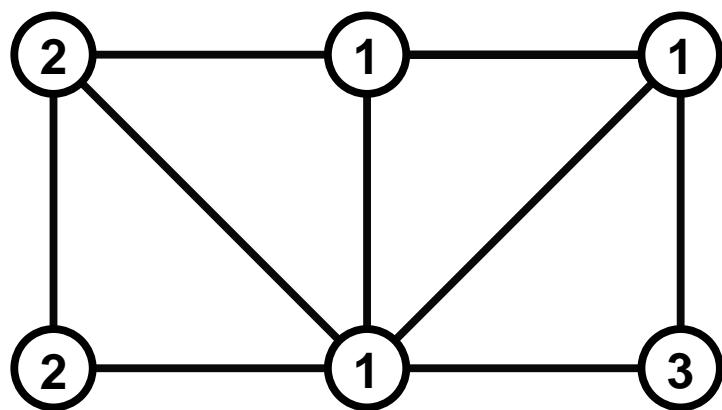
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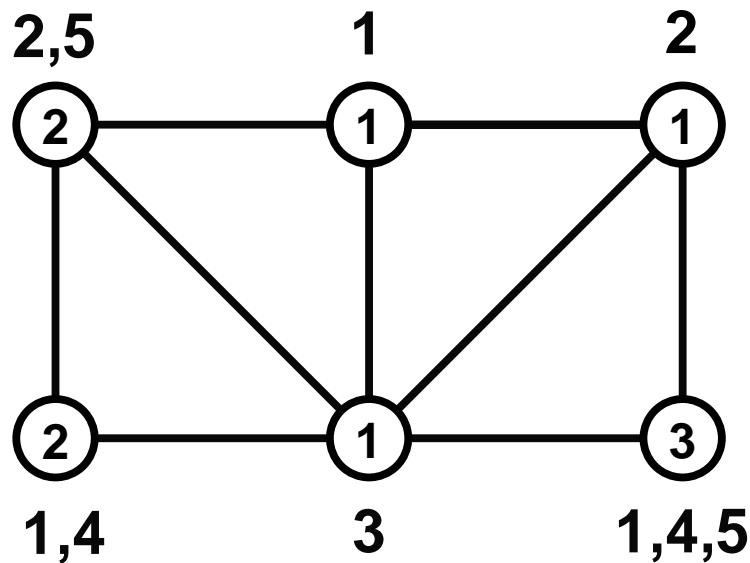


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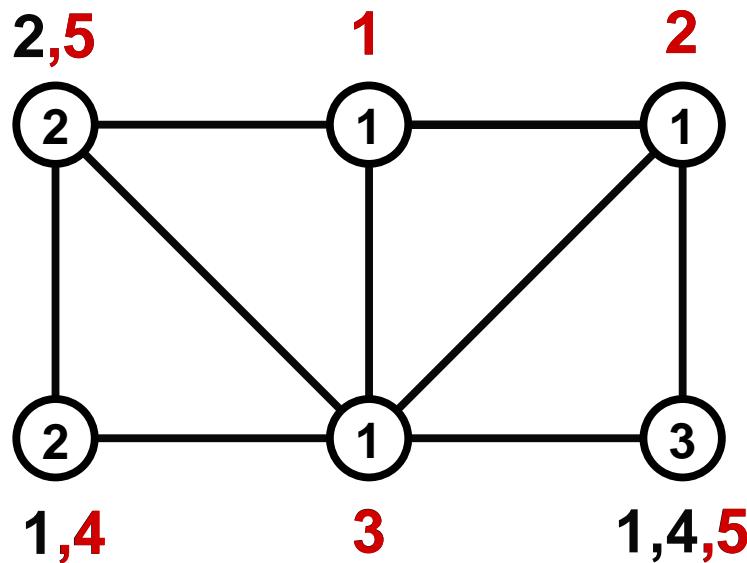


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Sum of the coloring:
 $5 + 1 + 2 + 4 + 3 + 5 = 20$

Known results

Special case: the chromatic sum problem: $x(v) = 1, \forall v \in V$

- **Trees:**

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 - ★ polynomial time solvable if every demand is 1 [Kubicka, 1989],
 - ★ sum multicoloring is **NP**-hard for binary trees [Marx, 2002]
 - ★ $(1 + \varepsilon)$ -approximation for sum multicoloring [Halldórsson et al., 1999]

- **Partial k -trees:**

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- **Bipartite graphs:**

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 - ★ **APX**-hard, even if every demand is 1 [Bar-Noy and Kortsarz, 1998]
 - ★ 1.5-approximation for sum multicoloring [Bar-Noy et al., 1998]

Edge coloring version

Assign $x(e)$ colors to each edge e , minimize the sum of finish times of the edges.
Each color can appear at most once at a vertex.

Application: scheduling dedicated biprocessor tasks

Each task requires the simultaneous work of two preassigned processors for a given number of time slots. Goal: minimize the sum of completion times.

vertices	\longleftrightarrow	processors
edges	\longleftrightarrow	jobs
demand	\longleftrightarrow	length of job
colors	\longleftrightarrow	time slots

Preemptive scheduling: jobs can be interrupted and continued later

Bipartite graphs: processors are divided into clients and servers

Edge coloring results

- **Known results:**

- ★ Polynomial time solvable on trees with demand 1 [Giaro and Kubale 2000]
- ★ **NP-hard** on bipartite graphs even if every demand is 1 [Giaro and Kubale 2000]
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- **Our results:**

- ★ **NP-hard** on trees even if every demand is 1 or 2
- ★ $(1 + \varepsilon)$ -approximation on trees with arbitrary demand

Scaling the demand

Theorem: If the graph is a tree, then multiplying the demand of each edge by integer q multiplies the minimum sum by *exactly* q .

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- ⇒ The problem can be solved in polynomial time on trees if every edge has the same demand
- ⇒ Increasing the demand to the next power of $(1 + \varepsilon)$ increases the sum by at most a factor of $(1 + \varepsilon)$ ⇒ we can assume that each demand is of the form $(1 + \varepsilon)^i$

Bounded degree trees

Theorem: Minimum sum edge coloring admits a linear time PTAS in bounded degree trees.

Method:

The line graph of a tree with max degree d is a partial $(d - 1)$ -tree, hence the PTAS of Halldórsson and Kortsarz can be used.

In a partial k -tree we can compute a polynomial number of color sets for each vertex such that there is a good approximate solution using only these sets \Rightarrow PTAS with standard dynamic programming

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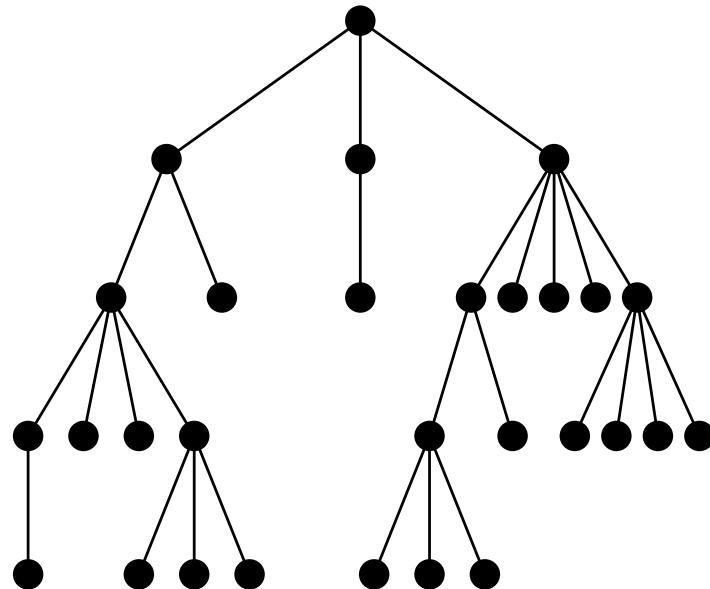
Bounded degree trees: In edge coloring bounded degree trees, a *constant* number of color sets is sufficient for each edge \Rightarrow linear time PTAS

Almost bounded degree trees: trees that have bounded degree after deleting the degree 1 nodes. Algorithm works for such trees as well.

General case

Theorem: Linear time PTAS for general trees.

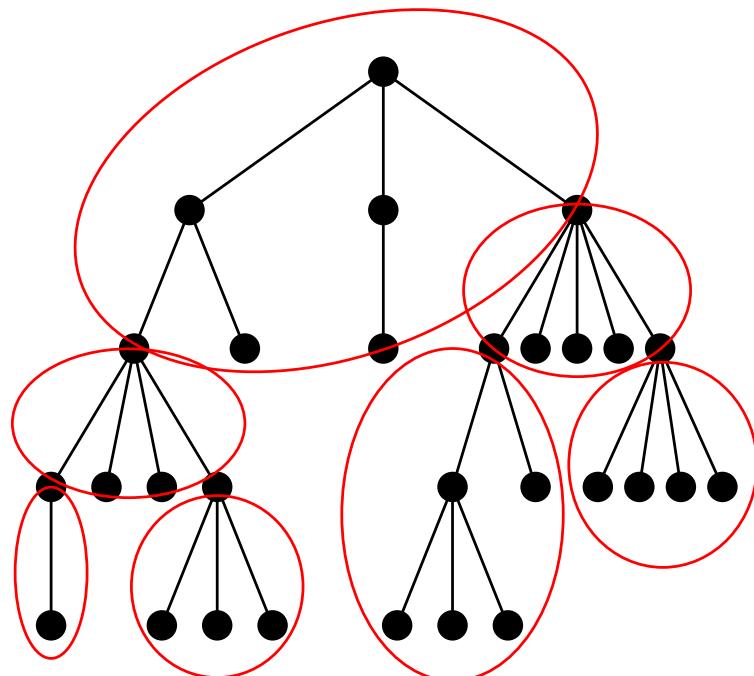
- Partition the tree into almost bounded degree subtrees
- Use the PTAS for subtrees
- Merge the colorings of the subtrees to a coloring of the whole tree



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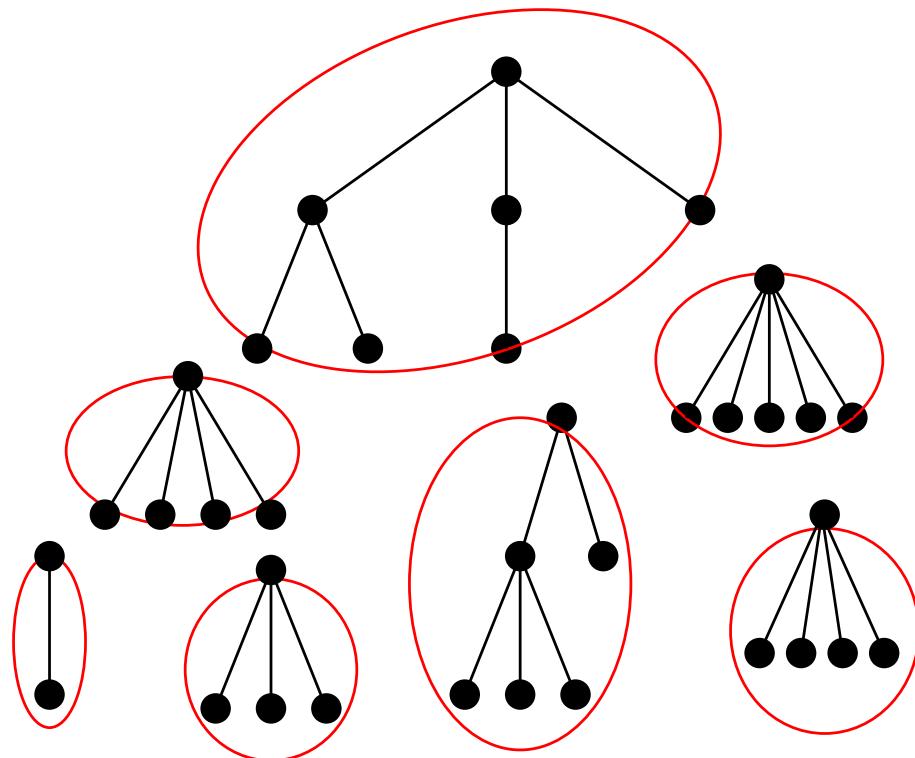
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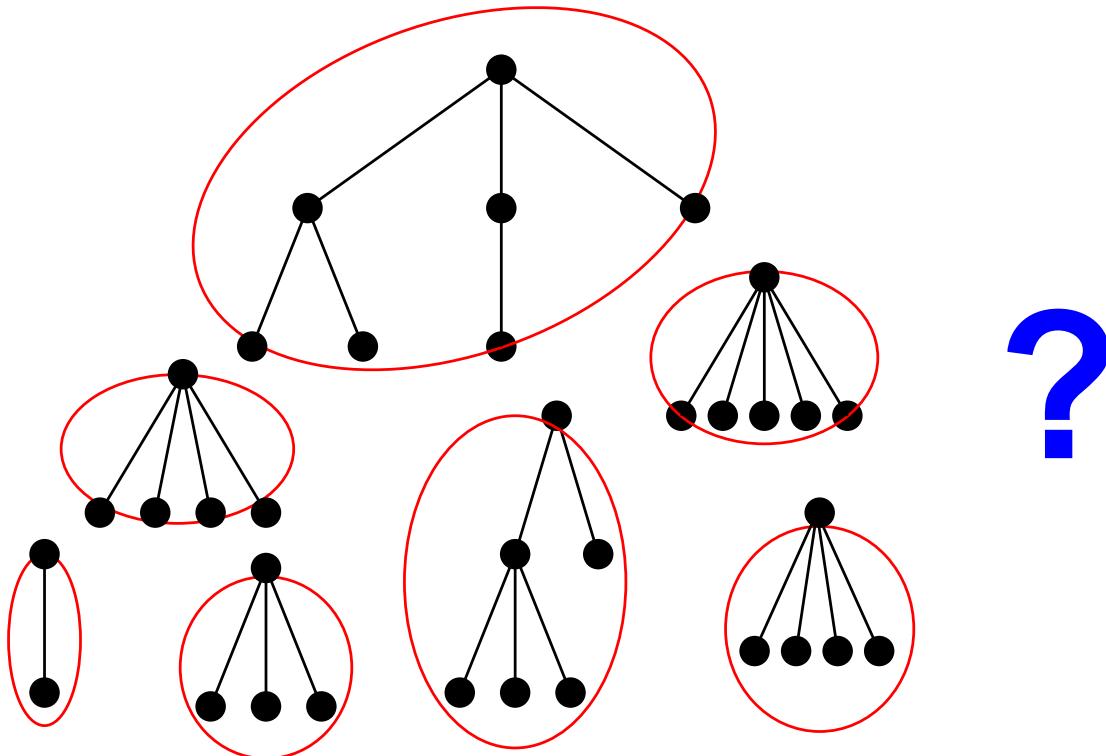
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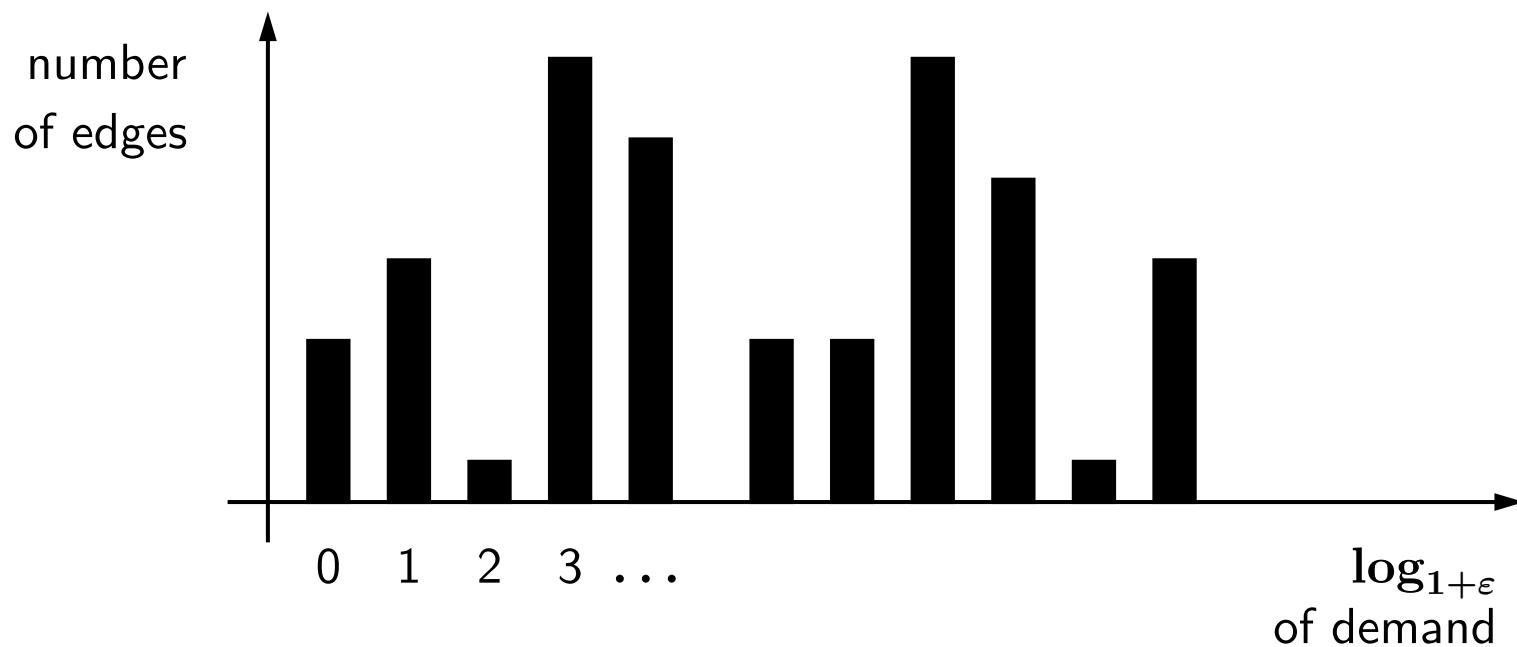
How to partition the tree?
How to resolve the conflicts
when merging the colorings?

The Small, the Large, and the Frequent

The child edges of a given node are divided into **small**, **large**, and **frequent** edges.

Every demand is of the form $(1 + \varepsilon)^i$.

Number of edges for each demand size:

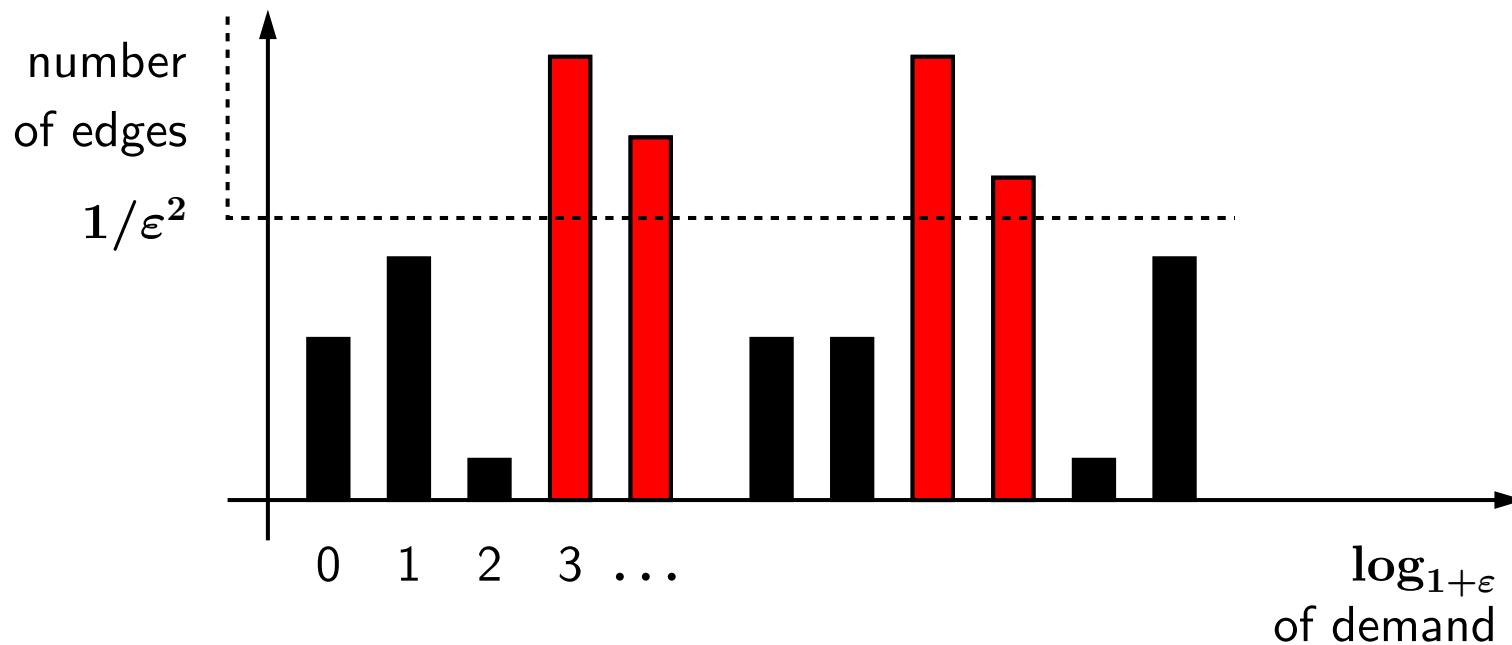


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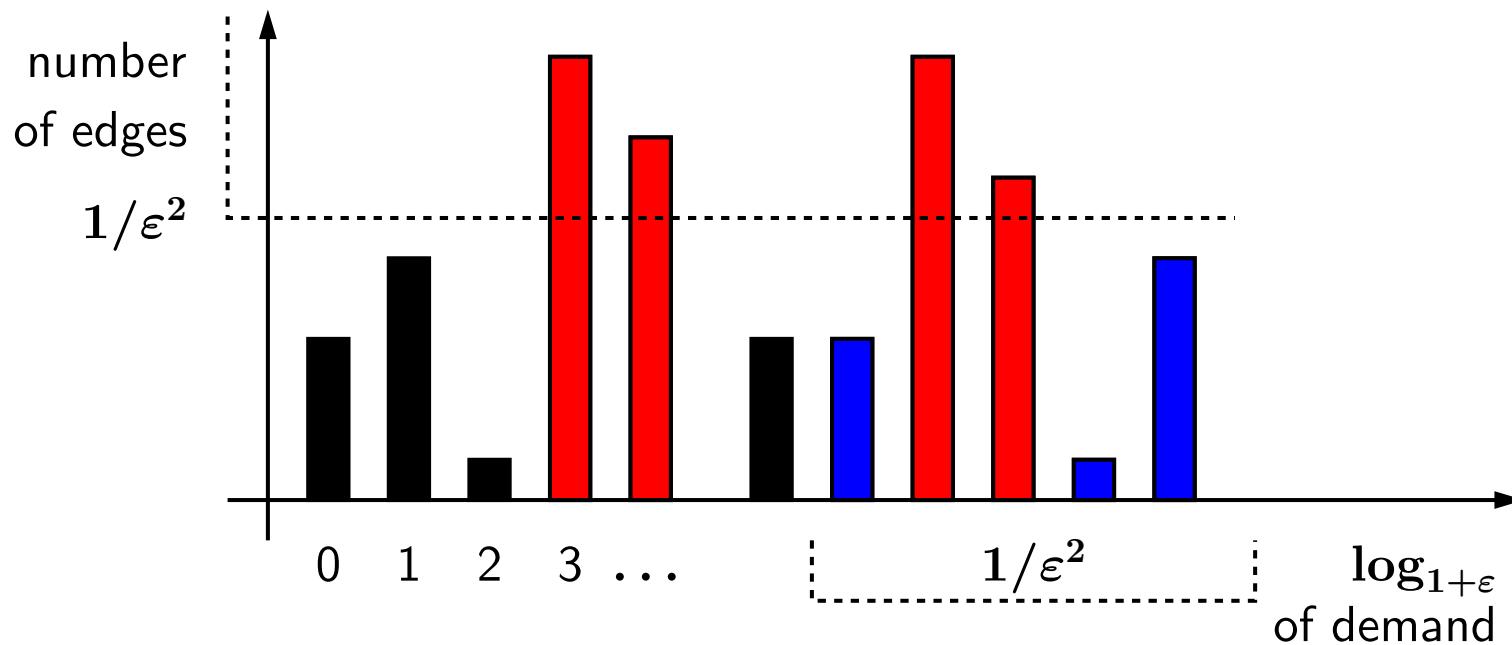


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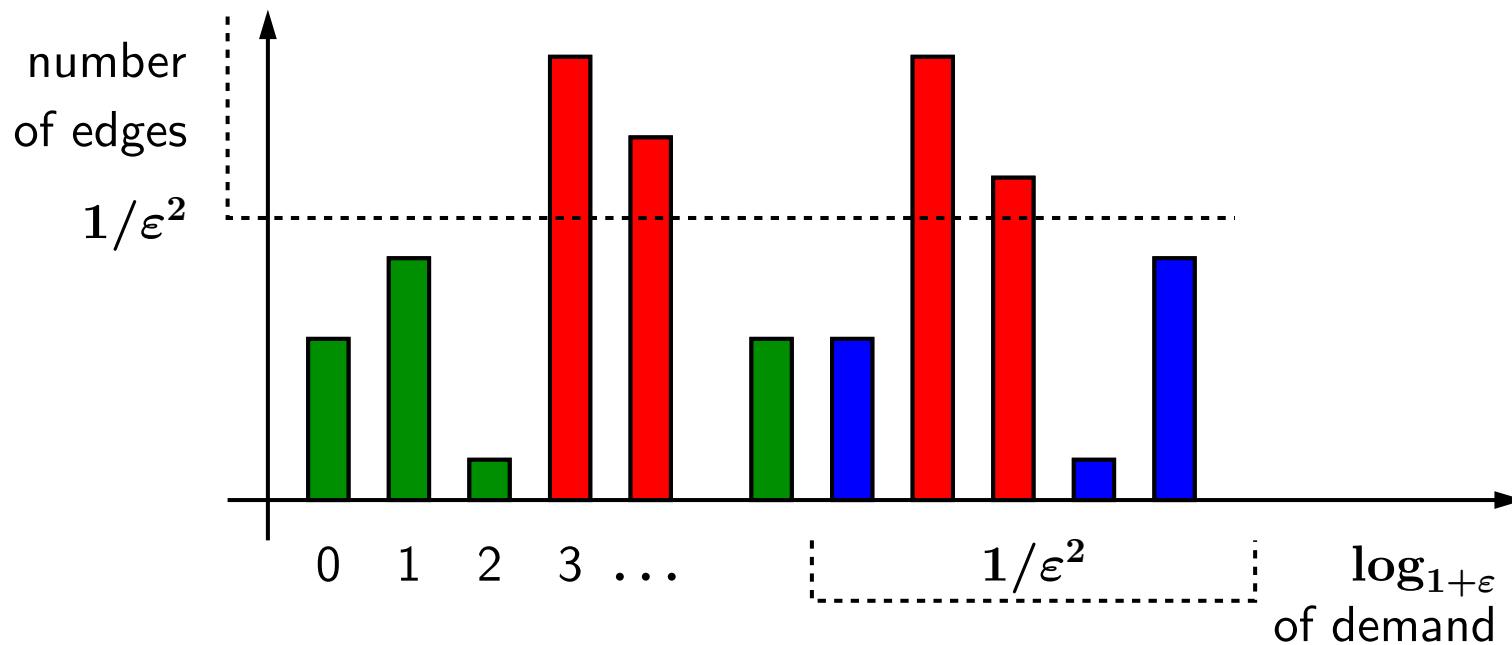


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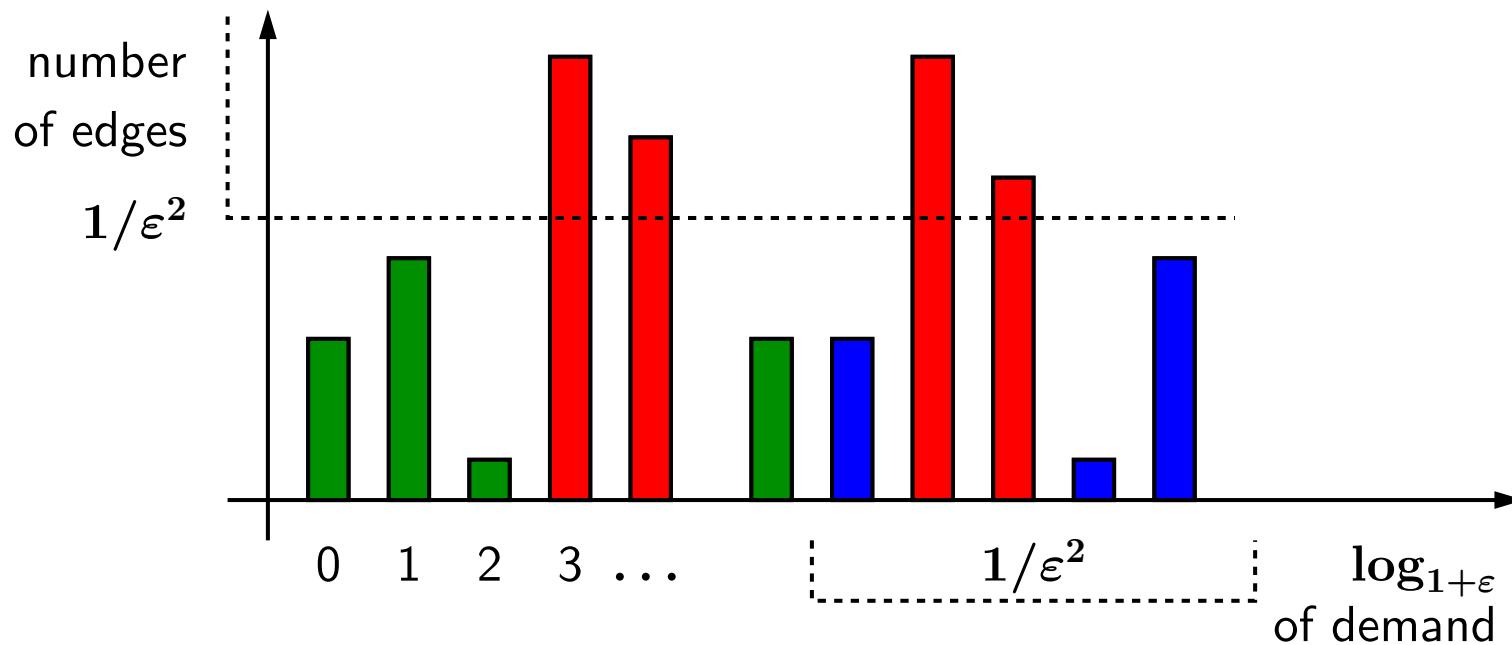


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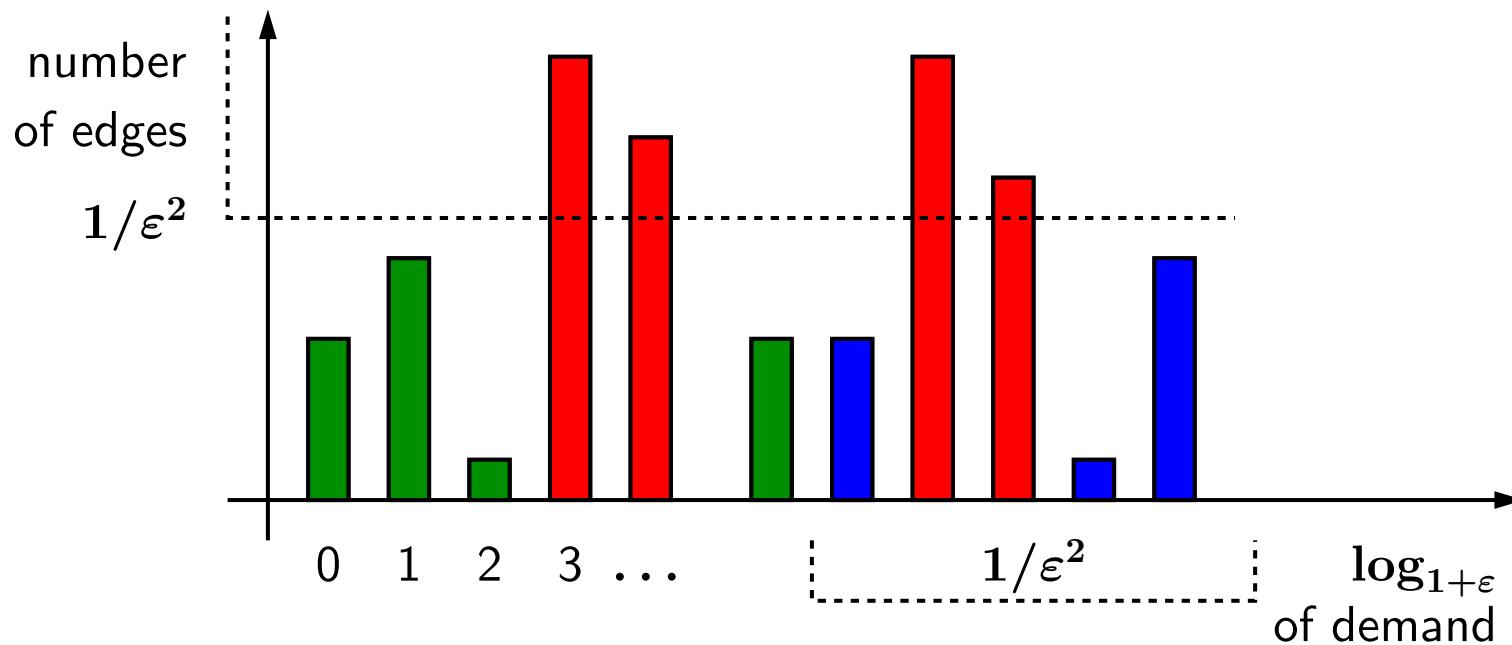
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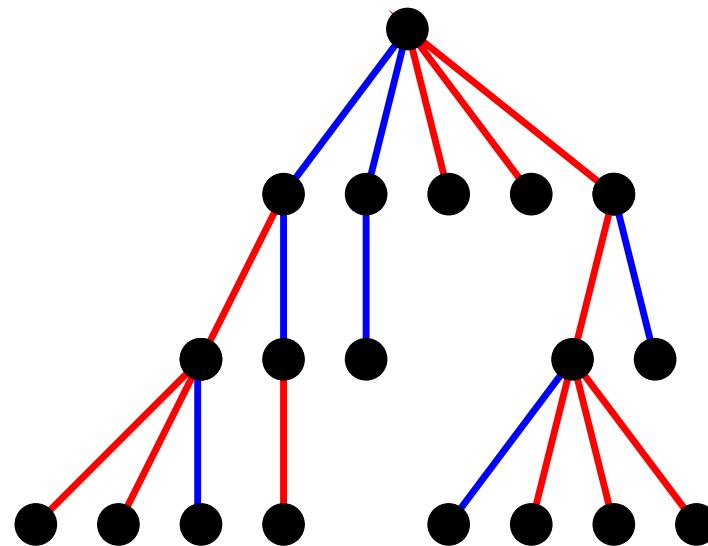


Total demand of the **small** edges is very small, they can be thrown away.

Each node has at most a constant number ($\leq 1/\varepsilon^4$) of **large** child edges.

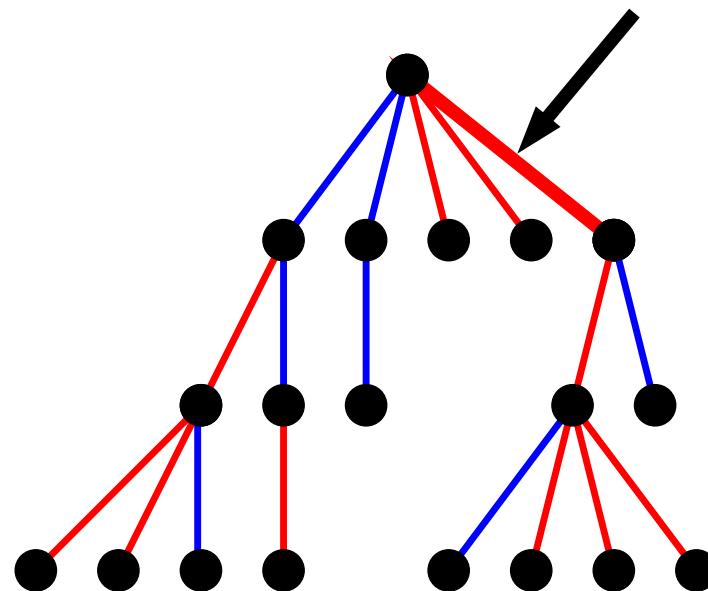
Partitioning the tree

The tree is split at the **frequent** edges:



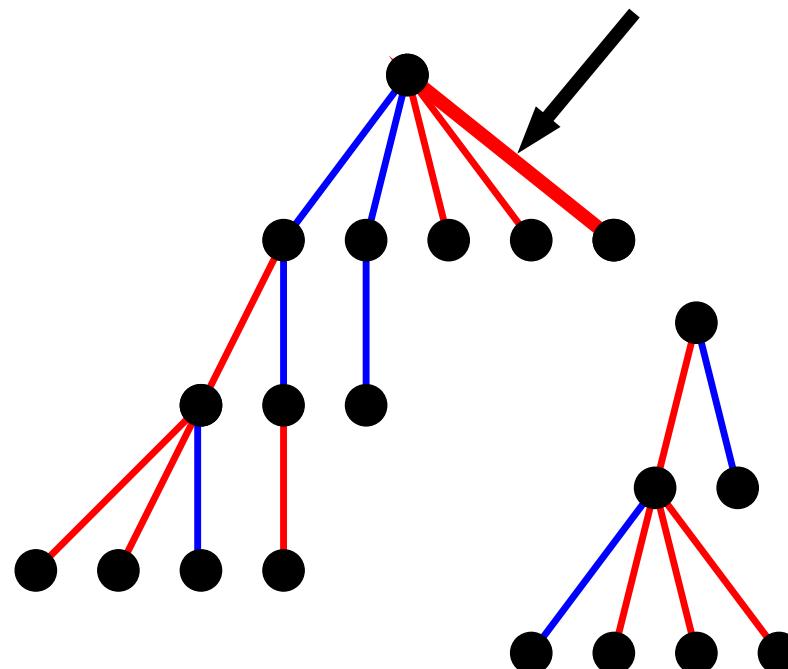
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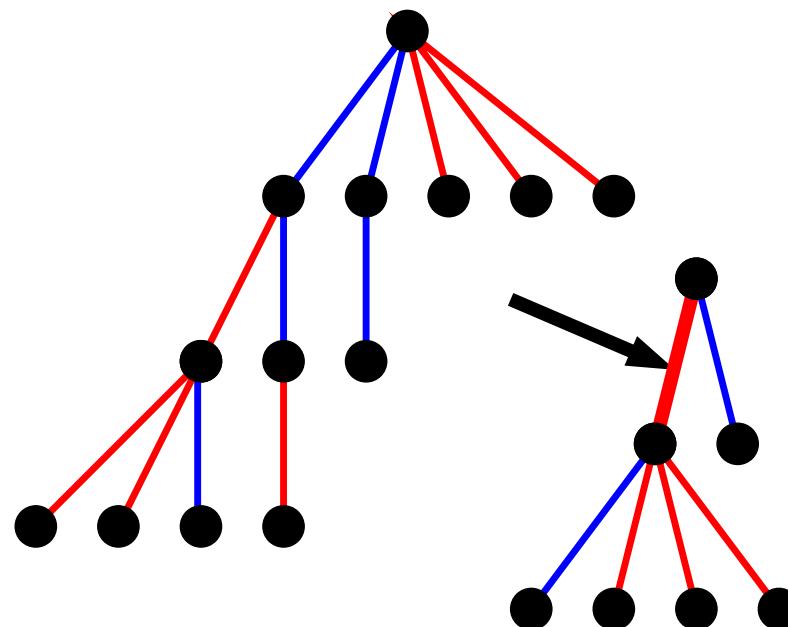
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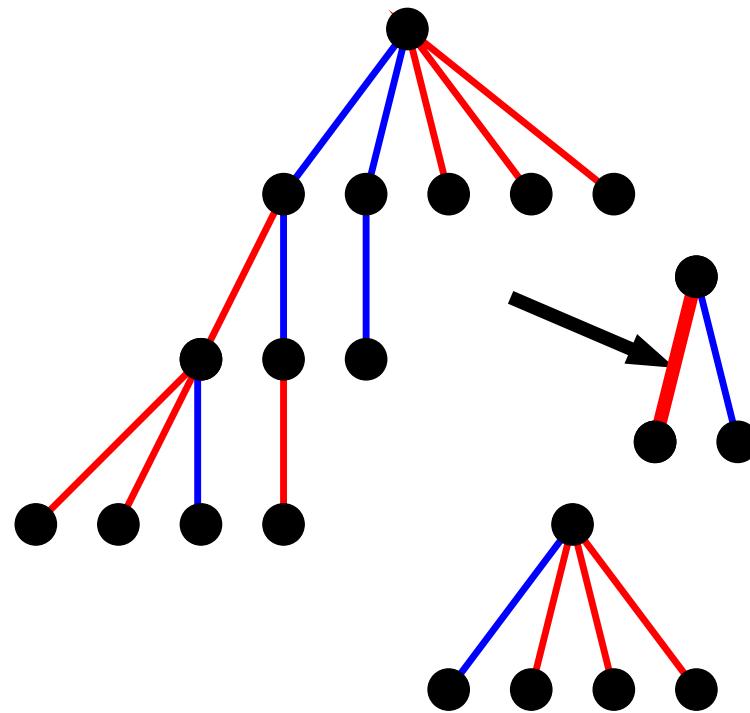
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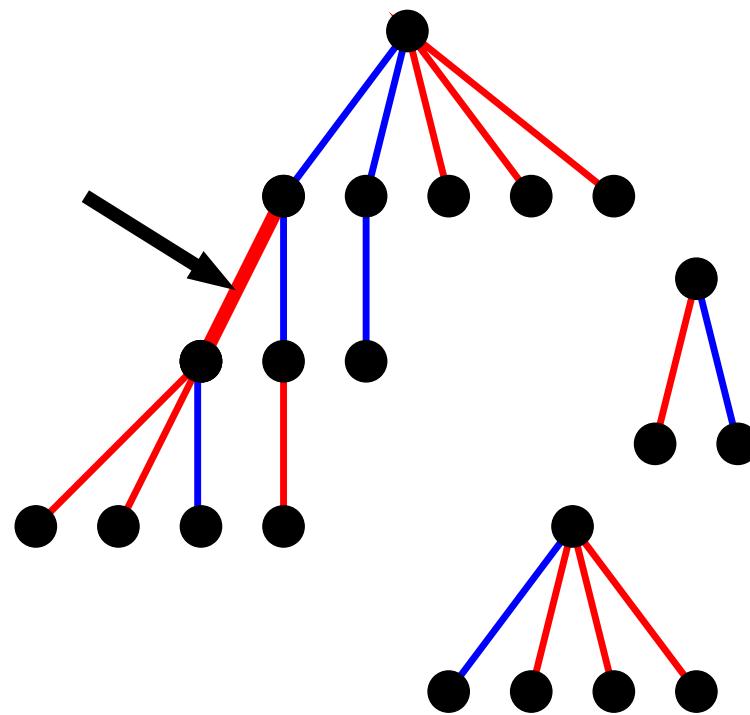
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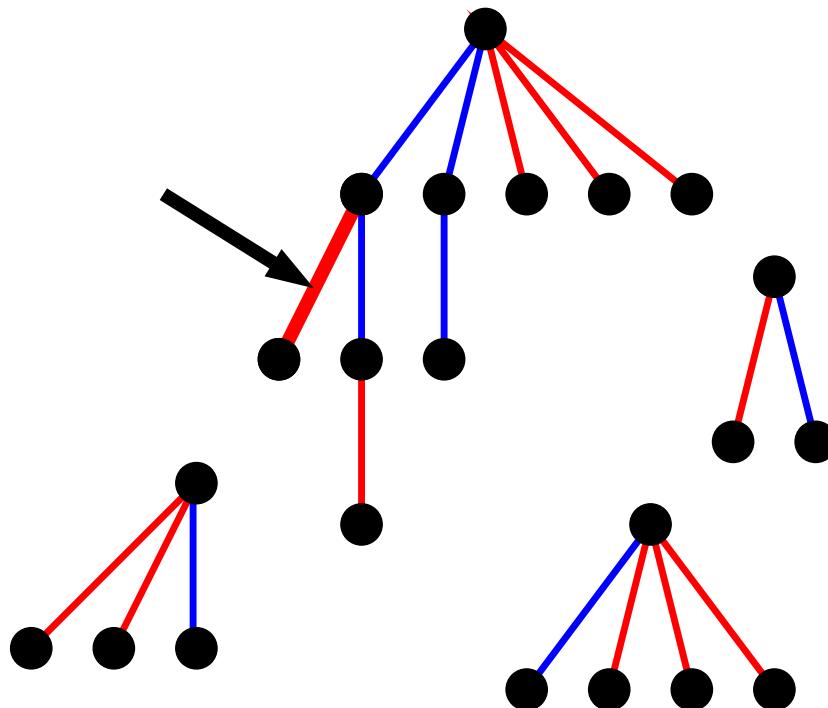
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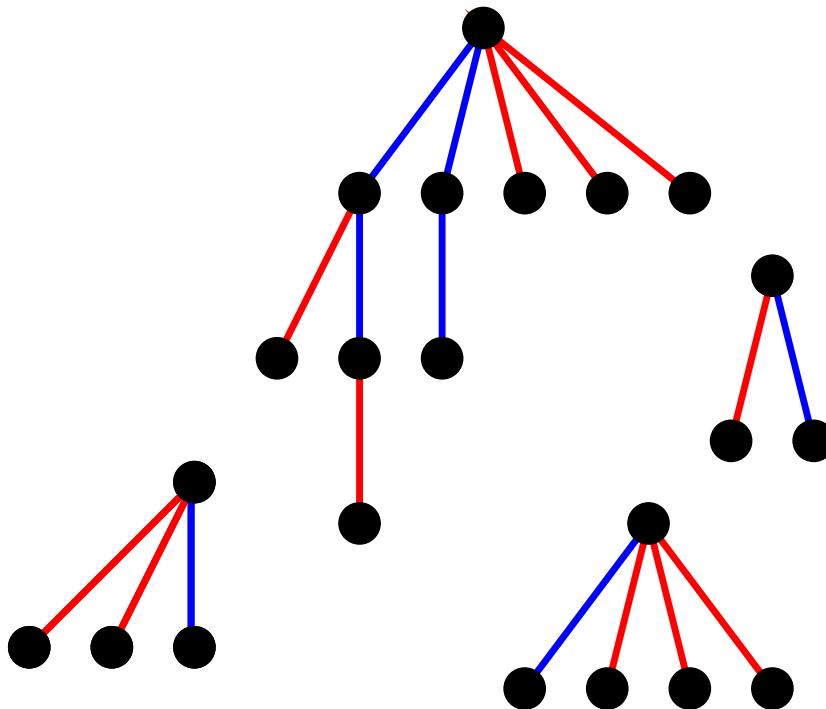
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Claim: Each subtree is an almost bounded degree tree \Rightarrow PTAS can be used

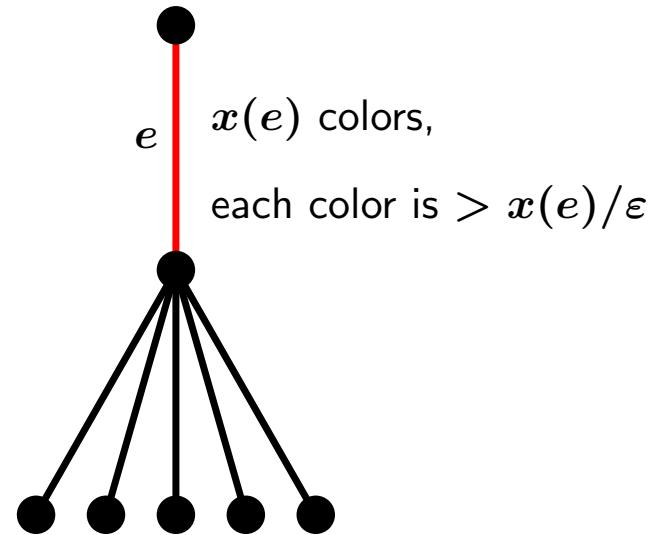
Proof:

- Deleting the degree 1 nodes deletes every **frequent** edge
- Only the **large** edges remain
- Each node has at most a constant number of **large** child edges

How to merge the colorings?

Shifting the frequent edges: We modify the coloring such that each frequent edge e uses only colors above $x(e)/\varepsilon$. This can be done with only a small increase of the sum.

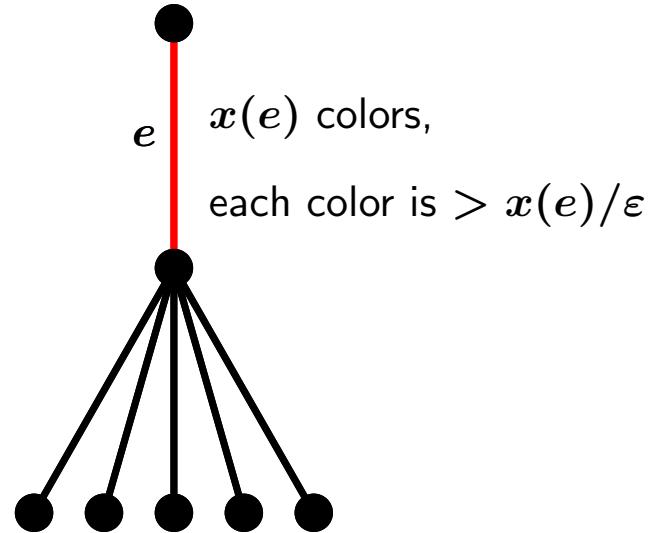
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We remove at most $x(e)$ colors, each of them is greater than $x(e)/\varepsilon$.

To replace these colors, it is easy to find $x(e)$ unused colors below $x(e)/\varepsilon$.

Conclusions

- Problem: edge coloring version of minimum sum multicoloring on trees.
- PTAS for vertex coloring partial k -trees implies a PTAS for edge coloring bounded degree trees. Linear time PTAS with additional techniques.
- Linear time PTAS for general trees uses the algorithm for bounded degree trees as a subroutine
- Minimum sum edge multicoloring is **NP-hard** on trees.