# Algorithm Theory - Winter Term 2017/2018 Exercise Sheet 2 

Hand in by Thursday 10:15, November 16, 2017

## Exercise 1: Tree Embedding into Grids

A $n \times m$ grid graph is a graph $G=\left(V_{G}, E_{G}\right)$ with nodes $V_{G}:=\{(i, j) \mid i \in\{1, \ldots, n\}, j \in\{1, \ldots, m\}\}$. These nodes are embedded in the Euclidian plane $\mathbb{R}^{2}$ and connected with edges as exemplified by the following $4 \times 15$ grid graph.


An embedding of a tree $T=\left(V_{T}, E_{T}\right)$ into a grid graph $G=\left(V_{G}, E_{G}\right)$ is defined as a one to one mapping of $V_{T}$ to a subset of $V_{G}$, which satisfies the following condition. There exists a set $P_{G}$ of vertex-disjoint paths ${ }^{1}$ in $G$ such that for each $\{u, v\} \in E_{T}$, there is a path $p \in P_{G}$ connecting $u^{\prime}$ to $v^{\prime}$, when $u$ is mapped to $u^{\prime}$, and $v$ is mapped to $v^{\prime}$.
(a) We can embed a complete binary tree with $n$ leaves into a grid, such that the nodes with height $i$ of the tree are placed in the $i^{t h}$ row of the grid. Below you see the embedding of a tree with 8 leaves into a $4 \times 15$ grid as an example.


By this way of embedding, show that we need a grid of $\operatorname{size}^{2} \Theta(n \log n)$ to embed a complete tree with $n$ leaves. To do so, write down the recurrence relations for the width $W(n)$ and the height $H(n)$ of the grid.
(b) Find a more efficient way of embedding a complete binary tree and show that it needs a grid of size $\Theta(n)$, if the tree has $n$ leaves. Write down the recurrence relations for the width $W(n)$ and the height $H(n)$ of the grid.

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## Sample Solution

(a) By doubling the number of the nodes the width of the grid doubles and the height of the grid is increased by one. Therefore, we can have the following recurrence relations for the width and the height of the grid.
$W(n)=2 W(n / 2)+\Theta(1) \Rightarrow W(n)=\Theta(n)$
$H(n)=H(n / 2)+\Theta(1) \Rightarrow H(n)=\Theta(\log n)$
Therefore, size of the grid is $\Theta(n \log n)$.
(b) To provide a more efficient way of embedding, we can avoid to place the nodes with same height in the same row of the grid. You can see below how we can embed a complete binary tree with 16 leaves in a $7 \times 7$ grid.


Here, we have $W(n)=H(n)=2 W(n / 4)+\Theta(1)$. Therefore, $W(n)=H(n)=\Theta(\sqrt{n})$. Hence, the size of the grid is $\Theta(n)$.

## Exercise 2: Polynomial to the power of $k$

Given a polynomial $p(x)$ of degree $n$ in coefficient representation and an integer $k \geq 2$, the goal of this problem is to compute the $k^{t h}$ power $p^{k}(x)$ of $p(x)$ in an efficient way. For simplicity, we assume that $k$ is a power of 2 , that is, $k=2^{\ell}$ for some integer $\ell \geq 1$.
(a) Describe an efficient algorithm to compute $p^{k}(x)$ polynomial using the fast polynomial multiplication procedure from the lecture.
(b) What is the asymptotic runtime of your algorithm in terms of $k$ and $n$ ? Explain your answer.

## Sample Solution

(a) We compute the $k^{t h}$ power of the polynomial $p(x)$ in coefficient representation by iterative multiplication, as

$$
p(x)^{k}=\left(\ldots\left(\left(p(x)^{2}\right)^{2} \ldots\right)^{2}\right.
$$

where square is taken $\log k$ times (assuming $k$ is a power of 2 ) with the procedure using the FFT from the lecture.

```
Algorithm 1 FastPotenciation \((p)\)
    \(q \leftarrow p\)
    for \(i\) from 1 to \(\ell\) do
        \(q \leftarrow\) FastPolynomialMultiplication \((q, q) \quad \triangleright\) procedure from the lecture using FFT
    return \(q\)
```

(b) We know that using the FFT algorithm from the lecture, two polynomials of degree $n$ can be multiplied in $\mathcal{O}(n \log n)$ time. Notice that in every iterative step $i$ of the algorithm, we need to
multiply two polynomials of degree $\left(2^{i-1} \cdot n\right)$ and get a polynomial of degree $\left(2^{i} \cdot n\right)$. Unwrapping the definition of the $\mathcal{O}$-Notation, we know that for large $n$ and for a $c>0$, we can compute the product of two polynomials of degree $\left(2^{i-1} \cdot n\right)$ in $c \cdot 2^{i} \cdot n \cdot \log \left(2^{i} \cdot n\right)$. To compute $p(x)^{k}$ (where $k=2^{\ell}$ for some $\ell$ ), there will be $\log k$ such iterations.
Therefore, the asymptotic running time of the algorithm (i.e., iterative multiplication of two same degree polynomials) would be:

$$
\sum_{i=0}^{\log k-1} c \cdot 2^{i} \cdot n \cdot \log \left(2^{i} \cdot n\right) \quad[\text { where } c \text { is the constant in the } \mathcal{O}(n \log n) \text { bound }]
$$

Let us compute this sum:

$$
\begin{aligned}
& \sum_{i=0}^{\log k-1} c \cdot 2^{i} \cdot n \cdot \log \left(2^{i} \cdot n\right)=c n \cdot \sum_{i=0}^{\log k-1} 2^{i} \cdot(i+\log (n)) \\
& =c n \cdot \sum_{i=0}^{\log k-1} i \cdot 2^{i}+c n \cdot \log (n) \cdot \sum_{i=0}^{\log k-1} 2^{i} \\
& \leq c n \cdot\left((\log (k)-1) \cdot 2^{\log (k)}\right)+c n \cdot \log (n) \cdot\left(2^{\log (k)}-1\right) \\
& \leq c n k \cdot \log (n k)
\end{aligned}
$$

Hence, the asymptotic running time of the algorithm is $\mathcal{O}(n k \cdot \log (n k))$.

## Exercise 3: Greedy Algorithm

In the following, a unit fraction is a fraction where the numerator is 1 and the denominator is some integer larger than 1. For example $1 / 4$ or $1 / 384$ are unit fractions.
It is well-known that every rational number $0<q<1$ can be expressed as a sum of pairwise distinct unit fractions, e.g., we can write $\frac{4}{13}$ as

$$
\frac{4}{13}=\frac{1}{5}+\frac{1}{13}+\frac{1}{32}+\frac{1}{65}
$$

Interestingly such a decomposition into distinct unit fractions can be computed using a simple greedy algorithm.

In the following, assume that you are given two positive integers $a$ and $b$ such that $b>a$. Design a greedy algorithm to compute integers $0<c_{1}<c_{2}<\cdots<c_{k}$ such that

$$
\frac{a}{b}=\frac{1}{c_{1}}+\frac{1}{c_{2}}+\cdots+\frac{1}{c_{k}}
$$

Prove that your greedy algorithm always works and that it decomposes $\frac{a}{b}$ into at most $a$ unit fractions. You can assume that your algorithm can deal with arbitrarily large integer numbers. Note that for the fraction $\frac{4}{13}$, the standard greedy algorithm computes a decomposition which is different from the one given above.

## Sample Solution

## Algorithm Description:

Here we introduce a greedy algorithm which solves the problem in at most $a$ steps for any input fraction $a / b$. In the $i^{t h}$ step, the algorithm uses the remaining fraction from the previous step and outputs $c_{i}$ and a remaining fraction $a_{i} / b_{i}$.

```
Algorithm 2 UnitFracDecomp \((a, b)\)
    \(i \leftarrow 0, a_{0} \leftarrow a, b_{0} \leftarrow b\)
    while \(a_{i} \neq 1\) do
        \(i \leftarrow i+1\)
        Let \(\frac{1}{c_{i}}\) be the largest unit fraction \(\leq \frac{a_{i-1}}{b_{i-1}} \quad \triangleright c_{i}:=\left\lceil\frac{b_{i}}{a_{i}}\right\rceil\) does the trick.
        \(\frac{a_{i}}{b_{i}} \leftarrow \frac{a_{i-1}}{b_{i-1}}-\frac{1}{c_{i}}\)
    return \(\left(\frac{1}{c_{1}}, \ldots, \frac{1}{c_{i}}, \frac{1}{b_{i}}\right)\)
```

To calculate $c_{i}$ and $a_{i} / b_{i}$ in each step the algorithm does the following. If the remaining fraction from the previous step is a unit fraction $\left(a_{i-1}=1\right)$ then $c_{i}:=b_{i-1}$ and the algorithm stops. Otherwise, the algorithm calculates the largest unit fraction $\frac{1}{c_{i}}$ smaller than $\frac{a_{i-1}}{b_{i-1}}$ from the previous step,

$$
\frac{1}{c_{i}}<\frac{a_{i-1}}{b_{i-1}}<\frac{1}{c_{i}-1} .
$$

Then, it calculates the remaining fraction as follows.

$$
\frac{a_{i}}{b_{i}}:=\frac{a_{i-1}}{b_{i-1}}-\frac{1}{c_{i}},
$$

where $c_{i}$ is a positive integer.

## Example:

Let us go through an example to see the algorithm more clearly. Consider the example in the exercise, $a / b=4 / 13$.
$1^{\text {st }}$ step, $a / b=4 / 13$ : The largest unit fraction smaller than $4 / 13$ is $1 / 4$. Then,

$$
c_{1}=4, a_{1} / b_{1}=4 / 13-1 / 4=3 / 52 .
$$

$2^{\text {nd }}$ step, $a_{1} / b_{1}=3 / 52$ : The largest unit fraction smaller than $3 / 52$ is $1 / 18$. Then,

$$
c_{2}=18, a_{2} / b_{2}=3 / 52-1 / 18=1 / 468 .
$$

Since the remaining fraction is a unit fraction, $c_{3}=1 / 468$, the algorithm stops and decomposition into sum of unit fractions is complete:

$$
\frac{4}{13}=\frac{1}{4}+\frac{1}{18}+\frac{1}{468}
$$

## Analysis:

Now, let us show that the above proposed algorithm always stops in a finite number of steps (at most $a$ steps) and works properly. To do this, we mention two claims.

Claim 1. In each step, the numerator of the remaining fraction is positive and strictly smaller than the numerator of the remaining fraction from previous step.
Proof. Fix some step with input fraction of $x / y$ and the generated unit fraction of $1 / c$. Then the remaining fraction is $\frac{x c-y}{y c}$. To prove the claim we should show that $x c-y$ is positive and strictly smaller than $x$. We assumed that $1 / c$ is smaller than $x / y$ and also it is the largest possible unit fraction. Based on the algorithm description we have

$$
\begin{align*}
& \frac{1}{c}<\frac{x}{y} \Rightarrow x c-y>0  \tag{1}\\
& \frac{1}{c-1}>\frac{x}{y} \Rightarrow x c-y<x \tag{2}
\end{align*}
$$

(1) and (2) prove claim 1.

Claim 2. If the algorithm stops after $t \leq a$ steps there does not exist $i \neq j \in\{1, \ldots, t\}$ such that $c_{i}=c_{j}$.
Proof. Fix some step $r$ during the execution and let us assume that for round $r-1$ the remaining fraction is $a_{r-1} / b_{r-1}$ and the calculated integer is $c_{r}$. Then the remaining fraction for round $r$ is

$$
\frac{a_{r}}{b_{r}}=\frac{a_{r-1}}{b_{r-1}}-\frac{1}{c_{r}}=\frac{a_{r-1} c_{r}-b_{r-1}}{b_{r-1} c_{r}}
$$

Since $1 / c_{r}$ is the largest unit fraction smaller than $a_{r-1} / b_{r-1}$ we have

$$
\begin{aligned}
\frac{1}{c_{r}-1}>\frac{a_{r-1}}{b_{r-1}} & \Rightarrow a_{r-1} c_{r}-a_{r-1}<b_{r-1} \\
& \Rightarrow a_{r-1} c_{r}<a_{r-1}+b_{r-1}<2 b_{r-1} \\
& \Rightarrow \frac{a_{r-1}}{b_{r-1}}-\frac{1}{c_{r}}<\frac{1}{c_{r}} \\
& \Rightarrow \frac{a_{r}}{b_{r}}<\frac{1}{c_{r}}
\end{aligned}
$$

Therefore, $c_{r}$ never be selected for any round $r^{\prime} \geq r$.
Based on claim 1, the numerator of the remaining fractions strictly decreases after each step and it is always positive. Therefore, there exists some step with remaining fraction of numerator 1. As a result the algorithm stops after at most $a$ steps and based on claim 2 it generates the desired sequence of integers.

## Exercise 4: Matroids

(a) For a graph $G=(V, E)$, a subset $F \subseteq E$ of the edges is called a forest iff (if and only if) it does not contain a cycle. Let $\mathcal{F}$ be the set of all forests of $G$. Show that $(E, \mathcal{F})$ is a matroid.
Hint: A forest with $k$ edges and $n$ nodes has $n-k$ connected components.
(b) For a matroid $(E, I)$, a maximal independent set $S \in I$ is an independent set that cannot be extended. Thus, for every element $e \in E \backslash S$, the set $S \cup\{e\} \notin I$.
What are the maximal independent sets of the matroid in (a)?

## Sample Solution

(a) We verify the three properties of a matroid $(E, \mathcal{F})$ with ground set $E$ and independent sets $\mathcal{F}$ :
(i) Independence of empty set: $\emptyset \in \mathcal{F}$.
(ii) Hereditary property: For all $F \subseteq E$ and $F^{\prime} \subseteq F: \quad F \in \mathcal{F} \Longrightarrow F^{\prime} \in \mathcal{F}$.
(iii) Augmentation property: $F_{1}, F_{2} \in \mathcal{F}: \quad\left|F_{1}\right|>\left|F_{2}\right| \Longrightarrow \exists e \in F_{1} \backslash F_{2}$ with $F_{2} \cup\{e\} \in \mathcal{F}$.

To (i): $\emptyset$ does not have a cycle and is therefore a forest, i.e. $\emptyset \in \mathcal{F}$.
To (ii): Let $F \subseteq E$ and $F^{\prime} \subseteq F$. This means we obtain $F^{\prime}$ by removing edges from $F$. Given that $F \in \mathcal{F}$ is acyclic and since removing edges does not create any cycles, we have that $F^{\prime}$ is also acyclic, i.e., $F^{\prime} \in \mathcal{F}$.
To (iii): Let $F_{1}, F_{2} \in \mathcal{F}$ with $\left|F_{1}\right|>\left|F_{2}\right|$. From the Hint we know that $F_{1}$ has fewer connected components than $F_{2}$. Hence there exists at least one component $C_{1}$ in $F_{1}$ whose vertices belong to at least two different components of $F_{2}$. Since $C_{1}$ is connected, there exists an edge $e \in F_{1}$ connecting the two aforementioned components of $F_{2}$. Hence adding $e$ to $F_{2}$ does not create any cycle in $F_{2}$ and therefore $F_{2} \cup\{e\} \in \mathcal{F}$.
(b) The maximal independent sets are spanning forests. I.e., a set of spanning trees for all(!) connected components of $G$ is maximal independent.


[^0]:    ${ }^{1}$ We define vertex-disjoint paths in $G$ as paths that may only have common endpoints but are disjoint otherwise.
    ${ }^{2}$ The size of a $n \times m$ grid graph is simply $n \cdot m$.

