# Alloc ating Servers in Infostations for On-Demand Communications * 

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#### Abstract

Given a set of servic e requests, each char acterize d by a temporal interval and a category, an integer $k$, and an integer $h_{c}$ for each categoryc, the Server A llo cation with Bounded Simultaneous Requests problem consists in assigning a server to each request in such a way that at most $k$ mutually simultaneous requests are assigned to the same server at the same time, out of which at most $h_{c}$ are of category $c$, and the minimum number of servers is used. Sinc ethis problem is computationally intractable, a 2-appr oximationonline algorithm is exhibite dwhich asymptotically gives a $\left(2-\frac{h}{k}\right)$-appr oximation, where $h=\min \left\{h_{c}\right\}$. Generalizations of the problem are considered, where each requestr is also char acterized by a bandwidth rate $w_{r}$, and the sum of the bandwidth rates of the simultaneous requests is bounded, and where each request is characterize dalso by a gender bandwidth. Such gener alizations contain Bin-Packing and Multiprocessor T ask sche duling as special cases, and they admit on-line algorithms providing constant approximations.


## 1 Introduction

An infostation is an isolated pock et area with small coverage (about a h undred of meters) of high bandwidth connectivity (at least a megabit per second) that collects information requests of mobile users and delivers data while users are going through the coverage area. The av ailable bandwidth depends on the distance between the mobile user and the center of the coverage area, increasing with decreasing distance $[4,8,9]$. Infostations could be located along roadways, at airports, in campuses, and they can provide access ports to Internet and/or access to services managed locally.

The infostation system may retriev e the requested data from remote gateways, may provide Internet access, and may home local services (such as building access, credit card transactions, and map downloads). The mobile user connection starts when it first senses the infostation's presence and finishes when it leaves

[^0]the coverage area. Depending on the mobility options, three kinds of users are characterized: drive-through, walk-through, and sit-through. According to the mobility options, the bit-rate connection is high variable for drive-through, low variable for walk-through, and fixed for sit-through. In addition to radio broadcast communication, infostations create opportunities to deliver new wireless information services dedicated to singleusers, which could be supported for example by infrared technologies.

Each mobile user going through the infostation may require a data service out of a finite set of possible service categories available. The admission control, i.e., the task of deciding whether or not a certain request will be admitted, is essential. In fact, a user going through an infostation to obtain a (toll) service is not disposed to have its request delay ed or refused. Hence, the service dropping probability must be kept as low as possible. F or this purpose, many admission control and bandwidth allocation schemes for infostations maintain a pool of servers so that when a request arrives it is immediately and irrevocably assigned to a server thus clearing the service dropping probability. Precisely, once a request is admitted, the infostation assigns a temporal interv al and a proper bandwidth for serving the request, depending on the service category, on the size of the data required and on the mobility kind of the user, as shown in T able 1 for a sample of requests with their actual parameters. Moreover, the infostation decides whether the request may be served locally or through a remote gateway. In both cases, a server (either in the infostation or in the gateway) is allocated on demand to the request during the assigned temporal interval. The request is immediately assigned to its serv er without knowing the future, namely with no knowledge of the next request. The server, selected out of the predefined server pool, serves the requests on-line, that is in an ongoing manner as they become a vailable. Moreover, each server may serve more than one request simultaneously but it is subject to some arc hitecture constraints. F or example, no more than $k$ requests could be served simultaneously by a local server supporting $k$ infrared channels or by a gateway server connected to $k$ infostations. Similarly, no more than $h$ services of the same category cadelie ered simultaneously due to access constraints on the original

| Category | Kbps | Seconds |  |
| :---: | :---: | :---: | :---: |
|  |  | Low rate | High rate |
| FTP download | 10000 | 100 | 10 |
| Video streams | 5000 | 50 | 5 |
| Audio streams <br> E-mail attachments | 512 | 5 | .5 |
| E-mails <br> eb Bro wsing | 64 | .6 | .06 |

Table 1. Example of actual time in tervalsrequired to serve different kinds of requests.
data, such as softw are licenses, limited on line subscriptions and private access.

In this paper, a particular problem arising in the design of infostation systems is faced which consists in finding scheduling algorithms for allocating the minimum number of servers to the user requests in such a w ay that the temporal, architectural and data constrain ts are satisfied. In details, a service request $r$ will be modeled by a service cate gory $c_{r}$ and a temporal interval $I_{r}=\left[s_{r}, e_{r}\right)$ with starting time $s_{r}$ and ending time $e_{r}$. Two requests are simultaneous if their temporal in tervals o verlapThe input of the problem consists of a set $R$ of service requests, a bound $k$ on the number of mutually simultaneous requests to be served by the same server at the same time, and a set $C$ of service categories with each category $c$ characterized by a bound $h_{c}$. The output is a mapping from the requests in $R$ to the servers that uses the minimum possible number of servers to assign all the requests in $R$ subject to the constraints that the same server receives at most $k$ mutually simultaneous requests at the same time, out of which at most $h_{c}$ are of category $c$. Such a problem is called in this paper Server Allocation with Bounded Simultaneous Requests.

It is worth y to note that, equating servers with bins, and requests with items, the above problem is similar to a generalization of Bin-Packing, known as Dynamic Bin-Packing [1], where in addition to size constraints on the bins the items are characterized by an arrival and a departure time, and repacking of already pack ed items is allow edeach time a new item arriv es. The problem considered in this paper, in contrast, does not allo w repac kingand has capacity constraints also on the bin size for each category. F urthermore, equating servers with processors and requests with tasks, the above problem becomes a generalization of deterministic multiprocessor scheduling with task release times and deadlines [6], where in addition each processor can execute more than one task at the same time.

In Section 2, it is sho wn that Server Allocation with Bounded Simultaneous Requests is computationally intractable. Section 3 sho ws that the problem
cannot be $\alpha$-approximated with $\alpha<\frac{4}{3}$. Moreover, a 2-approximation on-line algorithm is exhibited which asymptotically gives a $\left(2-\frac{h}{k}\right)$-approximation, where $h=\min _{c \in C} h_{c}$. In Section 4, a generalization of the problem is considered where each request $r$ is also characterized by an integer bandwidth rate $w_{r}$, and the bounds on the number of simultaneous requests to be served b ythe same server are replaced by bounds on the sum of the bandwidth rates of the simultaneous requests assigned to the same server. F or this problem, on-line and off-line algorithms are proposed which give a constant approximation. Other tw ogeneralizations are proposed in Section 5 in which each request is characterized either by a multi-dimensional bandwidth rate or $\mathrm{b} y$ both a bandwidth rate and a gender bandwidth associated to the category of the requests.

## 2 Computational Intractability

The Server Allocation with Bounded Simultaneous Requests problem on a set $R=\left\{r_{1}, \ldots, r_{n}\right\}$ of requests can be formulated as a coloring problem on the corresponding set $I=\left\{I_{1}, \ldots, I_{n}\right\}$ of temporal intervals.
Problem 1. (Interval Coloring with Bounded Overlapping) Given a set I of intervals each belonging to a cate gory, an integer $k$, and an integer $h_{c}$ for each $c$ ategory c, assign a color to each interval in such a way that at most $k$ mutually overlapping intervals receive the same color, at most $h_{c}$ mutually overlapping intervals all having category $c$ receive the same color, and the minimum number of colors is used.

In order to prove that Problem 1 is computationally in tractable, the following simplified decisional formulation is considered, where there is a bound $h_{c}=1$ for each category $c$.
Problem 2. (Interval Coloring with Unit Bounded Categories) Given a set I of intervals each belonging to a category, and two integers $k$ and $b$, decide whether $b$ colors are enough to assign a color to each interval in such a way that at most $k$ mutually overlapping intervals receive the same color and no two overlapping intervals with the same category receive the same color.

The graph coloring problem below is well-known to be $N P$-complete [2], even if $b \geq 3$.
Problem 3. (Chromatic Number) Given an integer $b$ and an undirected graph $G=(V, E)$, decide whether the nodes in $V$ can be color edwith $b$ colors in such $a$ way that adjacent nodes receive different colors.

Recall that a stable set of a graph $G=(V, E)$ is a subset $S$ of nodes in $V$ such that no two nodes in $S$ are adjacent.

Problem 4. (Balanced Coloring) Given two integers b and $k$, and an undirected graph $G=(V, E)$ of $k b$ nodes, decide whether $V$ can be partitioned into $b$ stable sets each of size $k$.

Lemma 2.1. Balanced Coloring is NP-complete, even if $b \geq 3$.
$\operatorname{Pr} \oplus f:$ Chromatic Number is reduced in polynomial time to Balanced Coloring as follows. Given an instance of Chromatic Number, namely $b$ and $G=$ $(V, E)$, let $k=|V|$ and $G^{\prime}=\left(V^{\prime}, E^{\prime}\right)$ be the graph obtained by considering $b$ disjoint copies of $G$. Clearly, $G$ can be colored with $b$ colors if and only if $V^{\prime}$ can be partitioned into $b$ stable sets each of size $k$.

Theorem 2.2. Interval Coloring with Unit Bounded Categories is NP-complete, even if $b \geq 3$.
$\operatorname{Pr} \oplus f:$ Given an instance of Balanced Coloring, that is $b, k$, and $G=(V, E)$, an instance of Problem 2 is constructed in such a way that there are as many categories as there are nodes in $V$, a subset of in tervals correponds to each node in $V$ so that all such intervals are forced to receive the same color, and some pairs of intersecting intervals belong to the same category if and only if their corresponding nodes in $V$ are adjacent.

Specifically, let $k b=|V|$ and $m=|E|$. Take the set of categories as $C=\{1,2, \ldots, k b\}$. Construct $2 k b+2 m$ intervals as follows. For each generic node $v \in V$, let $v_{1}, \ldots, v_{\ell}$ be the neighbours of $v$ (indexed, for the sake of simplicity, so that $\left.v_{i}<v_{i+1}\right)$. The following $\ell+2$ intervals correspond to node $v$ :

- $I_{v}^{0}=[0,2 v-1)$ with category $c_{v}^{0}=v$,
- $I_{v}^{v}=[2 v, 2 k b+1)$ with category $c_{v}^{v}=v$,
- $I_{v}^{v_{i}}=\left[2 v-1+\frac{i-1}{\ell}, 2 v-1+\frac{i}{\ell}\right)$ with category $c_{v}^{v_{i}}=v_{i}$, for $i=1, \ldots, \ell$.

If a node $v$ is isolated, that is it has no neighbour, then a single interval $I_{v}=[0,2 k b+1)$ with category $c_{v}=v$ corresponds to it.

Note that $[v, u] \in E$, with $v<u$, if and only if both the following conditions hold:

1. $I_{v}^{u} \cap I_{u}^{0} \neq \emptyset$ and $c_{v}^{u}=c_{u}^{0}=u$, and
2. $I_{u}^{v} \cap I_{v}^{v} \neq \emptyset$ and $c_{u}^{v}=c_{v}^{v}=v$.

Thus, adjacency between two nodes $u$ and $v$ in $G$ is coded bytw opairs of overlapping irtervals in $I$ with categories $u$ and $v$, which cannot be colored the same.

As an example, Figure 1 depicts all the $2 k b+2 m$ intervals of $I$ corresponding to a graph $G=(V, E)$ with $k b=6$ nodes and $m=10$ edges (in such a figure,


Figure 1. Example of reduction from Balanced Coloring to Interval Coloring with Unit Bounded Categories (categories are represen ted by num bers abo ve the thervals).
$k=2$ and $b=3$ ). Assume that $G$ can be colored with $b$ colors. Assign to all the intervals $I_{v}^{0}, I_{v}^{v}, I_{v}^{v_{1}}, \ldots, I_{v}^{v_{\ell}}$ the same color that node $v$ has in $G$. Clearly, if nodes $v$ and $u$ in $G$ are colored the same, then they are not adjacent. Thus all the intervals corresponding to $v$ and $u$ can receive the same color. Moreover, since the same color appears in $G$ exactly $k$ times, exactly $k$ mutually overlapping intervals of $I$ receiv e the same color.

Conversely, assume that all the intervals in $I$ can be colored with $b$ colors. Observe that for eac $\mathrm{h} v$, all the intervals $I_{v}^{0}, I_{v}^{v}, I_{v}^{v_{1}}, \ldots, I_{v}^{v_{\ell}}$ must be colored the same (otherwise, by construction, at least $b+1$ colors would be required). Therefore, assign such a color also to the node $v$ of $G$. If $I_{v}^{0}, I_{v}^{v}, I_{v}^{v_{1}}, \ldots, I_{v}^{v_{\ell}}$ and $I_{u}^{0}, I_{u}^{u}, I_{u}^{u_{1}}, \ldots, I_{u}^{u_{p}}$ ha ve the same color, then nodesv and $u$ are not adjacent in $G$, and thus they can receive the same color. Moreover, since $k$ mutually overlapping intervals are colored the same, the same color is used in $G$ exactly $k$ times.

## 3 Interval Coloring with Bounded Overlapping

An $\alpha$-approximation algorithm for a minimization problem is a polynomial-time algorithm producing a solution of value $\operatorname{appr}(x)$ on input $x$ such that, for all the inputs $x$,

$$
\operatorname{appr}(x) \leq \alpha \cdot o p t(x)
$$

where $\operatorname{opt}(x)$ is the value of the optimal solution on $x$. In other words, the approximate solution is guaran teed to never be greater than $\alpha$ times the optimal
solution [2]. F or the sake of simplicity, from no won, $\operatorname{appr}(x)$ and $\operatorname{opt}(x)$ will be simply denoted by appr and opt, respectively.

Corollary 3.1. The optimization version of Interval Coloring with Unit Bounded Categories admits no $\alpha$ approximation algorithm for $\alpha<\frac{4}{3}$.
$\operatorname{Pr} \oplus f:$ By the reduction in the proof of Theorem 2.2, if there is an $\alpha$-approximation algorithm with $\alpha<\frac{4}{3}$, then there is a decision algorithm for Balanced Coloring with $b=3$.

Assume that the intervals in $I$ arriv e one by one, and are indexed by non-decreasing starting times. When an interval $I_{i}$ arriv es, it is immediately and irrewocably colored, and the next interv al $I_{i+1}$ becomes known only after $I_{i}$ has been colored. An algorithm that works in such an ongoing manner is said on-line [5]. On-line algorithms are opposed to off-line algorithms, where the in tervals are not colored as they become ailable, but they are all colored only after the entire sequence $I$ of intervals is known.

A simple polynomial-time on-line algorithm for the more general Interv al Coloring with Bounded Overlapping can be designed based on the following greedy strategy:

```
Algorithm 1 Greedy \(\left(I_{i}\right)\)
- To color interv al \(I_{i}\) use, if possible, a color already used for previous intervals, otherwise use a brand new color.
```

Theorem 3.2. A lgorithm Gredy provides a 2approximation and, asymptotically, a $\quad\left(2-\frac{h}{k}\right)$ approximation, where $h=\min _{c \in C} h_{c}$, for Interval Coloring with Bounded Overlapping.
$\operatorname{Pr} \oplus f:$ Let $a p p r=\phi$ be the solution given by the algorithm and assume the colors $1, \ldots, \phi$ ha vebeen in troduced in this order. Let $I_{r}=\left[s_{r}, e_{r}\right)$ be the first interval colored $\phi$. Let $\Omega_{1}$ be the set of intervals in $I$ containing $s_{r}$ and let $\Omega_{2}$ be the set of intervals in $I$ containing $s_{r}$ and with category $c_{r}$. Hence, $\Omega_{2} \subseteq \Omega_{1}$. Let $\omega_{1}=\left|\Omega_{1}\right|$ and $\omega_{2}=\left|\Omega_{2}\right|$. Clearly, opt $\geq\left\lceil\frac{\omega_{1}}{k}\right\rceil$ and opt $\geq\left\lceil\frac{\omega_{2}}{h_{c_{r}}}\right\rceil$. Color $\phi$ was introduced to color $I_{r}$ because, for every $\gamma \in\{1, \ldots, \phi-1\}$, at least one of the following tw o conditions held:

1. at least $k$ intervals in $\Omega_{1}$ ha ve color $\gamma$;
2. at least $h_{c_{r}}$ intervals in $\Omega_{2}$ ha ve color $\gamma$.

For $i=1,2$, let $n_{i}$ be the number of colors in $\{1, \ldots, \phi-1\}$ for which condition $i$ holds (if for a color both conditions hold, then choose one of them arbitrarily). Hence, $n_{1}+n_{2}=\phi-1$ or, equivalently, appr $=\phi=n_{1}+n_{2}+1$. Clearly, $\omega_{1} \geq k n_{1}+h_{c_{r}} n_{2}+1$ and $\omega_{2} \geq h_{c_{r}} n_{2}+1$. Therefore:

$$
\begin{gathered}
\text { opt } \geq \max \left\{\left\lceil\frac{\omega_{1}}{k}\right\rceil ;\left\lceil\frac{\omega_{2}}{h_{c_{r}}}\right\rceil\right\} \geq \\
\max \left\{\left\lceil\frac{k n_{1}+h_{c_{r}} n_{2}+1}{k}\right\rceil ;\left\lceil\frac{h_{c_{r}} n_{2}+1}{h_{c_{r}}}\right\rceil\right\} \geq \\
\max \left\{n_{1}+\left\lceil\frac{h_{c_{r}} n_{2}+1}{k}\right\rceil ; n_{2}+1\right\} \geq \\
\max \left\{n_{1}+\frac{h}{k} n_{2} ; n_{2}+1\right\}
\end{gathered}
$$

where $h=\min _{c \in C} h_{c}$.
If $n_{2}+1 \geq n_{1}+\frac{h}{k} n_{2}$, then:

$$
\begin{gathered}
\frac{a p p r}{o p t} \leq \frac{n_{1}+n_{2}+1}{n_{2}+1} \leq \\
\frac{n_{2}\left(1-\frac{h}{k}\right)+1+n_{2}+1}{n_{2}+1}=2-\frac{h}{k} \frac{n_{2}}{n_{2}+1} \leq 2
\end{gathered}
$$

If $n_{2}+1 \leq n_{1}+\frac{h}{k} n_{2}$, then:

$$
\frac{a p p r}{o p t} \leq \frac{n_{1}+n_{2}+1}{n_{1}+\frac{h}{k} n_{2}} \leq
$$

$$
\frac{n_{1}+n_{1}+\frac{h}{k} n_{2}}{n_{1}+\frac{h}{k} n_{2}}=1+\frac{n_{1}}{n_{1}+\frac{h}{k} n_{2}} \leq 2
$$

Therefore, Algorithm 1 gives a 2-approximation.
T oachiev e the asymptotic apprximation, first observe that opt $\geq \max \left\{n_{1}+\frac{h}{k} n_{2} ; n_{2}\right\}$. Moreover, when opt $\rightarrow \infty$, also $\phi \rightarrow \infty$, and hence $\phi^{\prime}=\phi-1 \rightarrow \infty$, too. The ratio $\frac{a p p r}{o p t}$ is maximum when opt is minimum, that is, since opt $\geq n_{2}$ and opt $\geq n_{1}+\frac{h}{k} n_{2}$, for $o p t^{\prime}=n_{2}=n_{1}+\frac{h}{k} n_{2}$. Thus,

$$
\frac{a p p r}{o p t} \rightarrow \frac{\phi^{\prime}}{o p t^{\prime}}=\frac{n_{1}+n_{2}}{n_{2}}=\frac{\left(1-\frac{h}{k}\right) n_{2}+n_{2}}{n_{2}}=2-\frac{h}{k}
$$

Hence, asymptotically, Algorithm 1 gives a $\left(2-\frac{h}{k}\right)$ approximation.

The following Corollary 3.3 shows that the $\left(2-\frac{h}{k}\right)$ approximation is the best possible for Algorithm 1, ev en in the case that $h=1, k=2$, and no interval properly contains another interval.

Corollary 3.3. A lgorithm Gredy admits no $\alpha$ approximation with $\alpha<2-\frac{1}{k}$ for Interval Coloring with Unit Bounded Categories.
$\operatorname{Pr} \oplus f:$ It is sho wn that there is an instance for whid the $\left(2-\frac{1}{k}\right)$-approximation is achiev able. Consider the particular input instance consisting of the $k^{2}$ mutually overlapping irtervals $I_{1}, I_{2} \ldots, I_{k^{2}}$, where the $i$-th interval $I_{i}=\left[i, i+k^{2}\right)$ has category

$$
c_{i}=\left\{\begin{array}{lll}
i & \text { if } \quad 1 \leq i \leq k^{2}-k \\
k^{2}-k+1 & \text { if } \quad k^{2}-k+1 \leq i \leq k^{2}
\end{array}\right.
$$

The Greedy algorithm colors the interv al $I_{i}$ as soon as it becomes available, that is at time $i$, thus assigning color 1 to $I_{1}, \ldots, I_{k}$, color 2 to $I_{k+1}, \ldots, I_{2 k}$, and so on. In particular, color $j$ is assigned to $I_{(j-1) k+1}, \ldots, I_{j k}$, and color $k-1$ is given to $I_{(k-2) k+1}, \ldots, I_{(k-1) k}$. Moreover, for the remaining intervals $I_{k^{2}-k+1}, \ldots, I_{k^{2}}, k$ additional colors are employed, one for each interval. Overall, $2 k-1$ colors are used. How ever, an optimal off-line algorithm, that kno wsin advance the en tire sequence of intervals, uses $k$ colors assigning color 1 to intervals $I_{1}, \ldots, I_{k-1}$ and interval $I_{k^{2}-k+1}$, color 2 to intervals $I_{k}, \ldots, I_{2(k-1)}$ and $I_{k^{2}-k+2}$, and so on. In particular, color $j$ is assigned to intervals $I_{(j-1)(k-1)+1}, \ldots, I_{j(k-1)}$ and $I_{k^{2}-k+j}$, while color $k$ is given to $I_{(k-1)(k-1)+1}, \ldots, I_{k(k-1)}$ and $I_{k^{2}}$. Therefore, $\frac{a p p r}{o p t}=\frac{2 k-1}{k}=2-\frac{1}{k}$.

## 4 Interval Coloring with Weighted Overlapping

Consider now a generalization of Server Allocation with Bounded Simultaneous Requests, where each request $r$ is also characterized by an integer bandwidth rate $w_{r}$, and the bounds on the number of simultaneous requests to be served by the same server are replaced by bounds on the sum of the bandwidth rates of the simultaneous requests assigned to the same server. Suc h a problem can be formulated as a weigh ted generalization of Problem 1 as follows.

Problem 5. (In tervalColoring with Weighted Overlapping) Given a set I of intervals, with each interval $I_{r}$ char acterized by a category $c_{r}$ and an integer weight $w_{r}$, an integer $k$, and an inte ger $h_{c}$ for each category c, assign a color to each interval in such a way that the sum of the weights for mutually overlapping intervals receiving the same color is at most $k$, the sum of the weights for mutually overlapping intervals of category c receiving the same color is at most $h_{c}$, and the minimum number of colors is used.

More formally, denote by

- $I[t]$ the intervals active at instant $t$, that is, $I[t]=$ $\left\{I_{r} \in I: s_{r} \leq t \leq e_{r}\right\} ;$
- $I[c]$ the intervals belonging to the same category $c$, that is $I[c]=\left\{I_{r} \in I: c_{r}=c\right\}$; and
- $I(\gamma)$ the set of intervals colored $\gamma$.

Moreover, let $I(\gamma)[t]=I(\gamma) \cap I[t]$ be the intervals colored $\gamma$ and activ e at instant $t$. Finally, let $I(\gamma)[t][c]=I(\gamma)[t] \cap I[c]$ be the intervals of category $c$, colored $\gamma$, and active at instant $t$.

The constraints on the sum of the weights for mutually ov erlapping intervals receiving the same color can be stated as follows:

$$
\begin{gather*}
\sum_{I_{r} \in I(\gamma)[t]} w_{r} \leq k \quad \forall \gamma, \forall t  \tag{1}\\
\sum_{I_{r} \in I(\gamma)[t][c]} w_{r} \leq h_{c} \quad \forall \gamma, \forall t, \forall c \tag{2}
\end{gather*}
$$

An approximation on-line algorithm for Problem 5 (which contains Bin-Packing [1] as a special case) is presented below.

```
Algorithm 2 First-Color ( \(I_{i}\) )
- To color interval \(I_{i}\) use the low est possible indexed color among those already used for previous intervals. If no such color exists, use a brand new color.
```

Theorem 4.1. A lgorithm First-Color asymptotically provides a constant approximation for Interval Coloring with Weighted Overlapping.
$\operatorname{Pr} \oplus f:$ Assume the on-line First-Color algorithm emplo ys $\phi$ colors. Consider the first interval $I_{r}=\left[s_{r}, e_{r}\right)$ which is colored $\phi$. A ttime $s_{r}, I_{r}$ cannot be colored with any color in $\{1, \ldots, \phi-1\}$ since otherwise at least one of the constraints (1) and (2) would be violated. Two cases may occur, depending on whether $w_{r}$ is smaller or larger than $\frac{h}{2}$, where $h=\min _{c \in C} h_{c}$. Case 1: Suppose $w_{r} \leq \frac{h}{2}$. Let $\Omega_{1}$ be the set of intervals in $I$ con tainings $s_{r}$ and let $\Omega_{2}$ be the set of intervals in $I$ containing $s_{r}$ and having category $c_{r}$. Let $w\left(\Omega_{1}\right)$ (resp., $w\left(\Omega_{2}\right)$ ) be the sum of the weights of the intervals in $\Omega_{1}$ (resp., $\Omega_{2}$ ). Clearly, opt $\geq\left\lceil\frac{w\left(\Omega_{1}\right)}{k}\right\rceil$ and opt $\geq\left\lceil\frac{w\left(\Omega_{2}\right)}{h_{c_{r}}}\right\rceil$.

Color $\phi$ was used for $I_{r}$ because, for every $\gamma \in$ $\{1, \ldots, \phi-1\}$, at least one of the following two conditions held:

1. $w\left(\Omega_{1}(\gamma)\right) \geq \frac{k}{2}$, where $\Omega_{1}(\gamma)$ are the interv als in $\Omega_{1}$ already colored $\gamma$,
2. $w\left(\Omega_{2}(\gamma)\right) \geq \frac{h_{c_{r}}}{2}$, where $\Omega_{2}(\gamma)$ are the intervals in $\Omega_{2}$ with category $c_{r}$ already colored $\gamma$.

For $i=1,2$, let $n_{i}$ be the number of colors in $\{1, \ldots, \phi-1\}$ for which $i$ holds (if for a color both conditions hold, then choose one of them arbitrarily). Hence, $n_{1}+n_{2}=\phi-1$. Clearly, $w\left(\Omega_{1}\right) \geq n_{1} \frac{k}{2}+n_{2} \frac{h_{c_{r}}}{2}$ and $w\left(\Omega_{2}\right) \geq n_{2} \frac{h_{c_{r}}}{2}$. Therefore,

$$
\begin{gathered}
o p t \geq \max \left\{\left\lceil\frac{w\left(\Omega_{1}\right)}{k}\right\rceil ;\left\lceil\frac{w\left(\Omega_{2}\right)}{h_{c_{r}}}\right\rceil\right\} \geq \\
\quad \max \left\{\left\lceil\frac{n_{1} k+n_{2} h}{2 k}\right\rceil ;\left\lceil\frac{n_{2}}{2}\right\rceil\right\}
\end{gathered}
$$

If $\frac{n_{2}}{2} \geq \frac{n_{1} k+n_{2} h}{2 k}$, then:

$$
\begin{gathered}
\frac{a p p r}{o p t} \leq 2 \frac{n_{1}+n_{2}+1}{n_{2}} \leq \\
2 \frac{n_{2}\left(2-\frac{h}{k}\right)+1}{n_{2}} \rightarrow 2\left(2-\frac{h}{k}\right)
\end{gathered}
$$

$$
\text { If } \frac{n_{2}}{2} \leq \frac{n_{1} k+n_{2} h}{2 k} \text {, then: }
$$

$$
\begin{gathered}
\frac{a p p r}{o p t} \leq \frac{n_{1}+n_{2}+1}{\frac{n_{1} k+n_{2} h}{2 k}}= \\
\frac{2\left(n_{1}+n_{2}+1\right)}{n_{1}+\frac{h}{k} n_{2}} \leq 4
\end{gathered}
$$

by the bound proved in Theorem 3.2.
Case 2: Suppose $w_{r}>\frac{h}{2}$. Two further subcases may come up.

Case 2.1: For each color $\gamma \in\left\{\left\lceil\frac{\phi}{2}\right\rceil, \ldots, \phi-1\right\}$ at least one of the two following conditions holds:

$$
\begin{gather*}
\sum_{I_{r} \in I(\gamma)\left[s_{r}\right]} w_{r} \geq \frac{k}{2}  \tag{3}\\
\sum_{I_{r} \in I(\gamma)\left[s_{r}\right][c]} w_{r} \geq \frac{h_{c}}{2} \quad \text { for some } c \tag{4}
\end{gather*}
$$

Therefore, the sum of the weights of the intervals active at time $s_{r}$ is at least $\frac{h}{2}\left\lfloor\frac{\phi}{2}\right\rfloor$. Thus,

$$
\text { opt } \geq\left\lceil\frac{\frac{h}{2}\left\lfloor\frac{\phi}{2}\right\rfloor}{k}\right\rceil \quad \text { and } \quad \frac{a p p r}{o p t} \leq \frac{\phi}{\frac{\frac{h}{2}\left\lfloor\frac{\phi}{2}\right\rfloor}{k}} \rightarrow \frac{5 k}{h}
$$

Case 2.2: There is a color $\bar{\gamma} \in\left\{\left\lceil\frac{\phi}{2}\right\rceil, \ldots, \phi-1\right\}$ for which both conditions (3) and (4) do not hold. Thus, there is an interval $I_{\bar{r}}=\left[s_{\bar{r}}, e_{\bar{r}}\right)$, colored $\bar{\gamma}$, of weight $w_{\bar{r}}<\frac{h}{2}$. When $I_{\bar{r}}$ was colored, $\bar{\gamma} \mathrm{w}$ as the low est possible indexed color. Therefore, this subcase reduces to Case 1 above loosing a factor of 2 .

It is worthy to note that in the case there are no constraints on the total weight of mutually overlapping intervals of the same category, the above algorithm yields a 4-approximation. This can be easily check ed assuming $h=k$ in the proof of Theorem 4.1 and observing that, in the Case 2.1, only condition (3) holds for all the colors $\gamma$.

Moreover, note that the w orstapproximation constant of Algorithm 2 is given by $\frac{5 k}{h}$ when $\frac{k}{h}>\frac{8}{5}$, and by 8 otherwise. How eve, an off-line algorithm for Problem 5 can be proposed which guarantees an 8approximation even in the case that $\frac{k}{h}>\frac{8}{5}$. The algorithm runs three passes over the entire input sequence $I$. Each pass scans the interv als in $I$ by non decreasing starting times and delivers a new set of colors. Hence, denoted by $\phi_{i}$ the number of colors delivered in pass $i$, the total number of colors employ ed is $\phi=\phi_{1}+\phi_{2}+\phi_{3}$.

```
Algorithm 3 Three-Pass-Greedy \((I)\)
PASS 1: Color all intervals \(I_{r}\) with \(w_{r}>\frac{k}{2}\);
Pass 2: Color all the intervals \(I_{r}\) with \(w_{r}>\frac{1}{2} h_{c_{r}}\);
Pass 3: Color all the remaining intervals.
```

Note that, in the first pass, the same color cannot be assigned to two overlapping intervals, and hence the problem reduces to coloring an interval graph, which can be done optimally in polynomial time [3]. Therefore, $\phi_{1} \leq o p t$. In the second and third passes, instead, Algorithm 1 is employed. For the second pass alone, an approximation of 3 can be shown by modifying the proof of Theorem 3.2 obtaining $\phi_{2} \leq 3 \cdot o p t$. F or the third pass alone, the analysis on the approximation guarantee of Theorem 3.2 can again be easily adapted by loosing a factor of 2 , thus having $\phi_{3} \leq 4 \cdot$ opt. As a consequence, appr $=\phi_{1}+\phi_{2}+\phi_{3} \leq 8 \cdot o p t$.

## 5 F urtherGeneralizations

This section considers tw o generalizations of the Server Allocation with Bounded Simultaneous Requests problem, where each request $r$ is characterized by real bandwidths, normalized in $[0,1]$ for analogy with the Bin-Pac king problem [1].

In the first generalization, which con tains MultiDimensional Bin-Packing as a special case, each request $r$ is characterized by a $k$-dimensional bandwidth rate $\mathbf{w}_{r}=\left(w_{r}^{(1)}, \ldots, w_{r}^{(k)}\right)$, where the $c$-th component specifies the bandwidth needed for the $c$-th category and $k$ is the number of categories, i.e. $k=|C|$. The overall sum of the bandwidth rates of the simultaneous requests of the same category assigned to the same server at the same time is bounded by 1 , which implies that the total sum of the bandwidth rates over all the categories is bounded by $k$.

Problem 6. (Interval Coloring with MultiDimensional Weigh ted Overlapping) Given a set $I$ of intervals, with each interval $I_{r}$ char acterize dby a $k$-dimensional weight $\mathbf{w}_{r}=\left(w_{r}^{(1)}, \ldots, w_{r}^{(k)}\right)$, where $w_{r}^{(c)} \in[0,1]$, for $1 \leq c \leq k$, assign a color to each interval in such a way that the overall sum of the weights of the same category for mutually overlapping intervals receiving the same color is b oundd by 1.

More formally, according to the notations introduced in Section 4, the constraints on the sum of the w eigh ts of the same category for motually overlapping intervals receiving the same color can be stated as follows:

$$
\begin{equation*}
\sum_{I_{r} \in I(\gamma)[t][c]} w_{r}^{(c)} \leq 1 \quad \forall \gamma, \forall t, \forall c \tag{5}
\end{equation*}
$$

Note that the above constraints, added up over all the categories in $C$, imply the follo wing redundant constrain ts :

$$
\begin{equation*}
\sum_{c=1}^{k} \sum_{I_{r} \in I(\gamma)[t][c]} w_{r}^{(c)} \leq k \quad \forall \gamma, \forall t \tag{6}
\end{equation*}
$$

Problem 6 can be solved on-line by Algorithm 2, introduced in the previous section.

Theorem 5.1. The First-Color algorithm provides a $4 k$-appr oximation for Interval Coloring with MultiDimensional Weighted Overlapping.
$\operatorname{Pr} \oplus f:$ Assume the on-line algorithm employs $\phi$ colors. Consider the first interv al $I_{\bar{r}}=\left[s_{\bar{r}}, e_{\bar{r}}\right)$ which is colored $\phi$. At time $s_{\bar{r}}, I_{\bar{r}}$ cannot be colored with any color in $\{1, \ldots, \phi-1\}$.

Consider tw o cases.
Case 1: There is a color $\bar{\gamma}$ among $\left\{\left\lceil\frac{\phi}{2}\right\rceil, \ldots, \phi-1\right\}$ such that for ead component $c$, with $1 \leq c \leq k$ :

$$
\sum_{I_{r} \in I(\bar{\gamma})\left[s_{\vec{r}}\right][c]} w_{r}^{(c)} \leq \frac{1}{2}
$$

Let $I_{r^{\prime}}$ be any interval in $I(\bar{\gamma})\left[s_{\bar{r}}\right][c]$. Clearly, $w_{r^{\prime}}^{(c)} \leq$ $\frac{1}{2}$, for all $c$. Consider no winstant $s_{r^{\prime}}$, when interval $I_{r^{\prime}} \mathrm{w}$ as colored $\bar{\gamma} \geq\left\lceil\frac{\phi}{2}\right\rceil$. Since $I_{r^{\prime}}$ cannot be colored with any color in $\left\{1, \ldots,\left\lceil\frac{\phi}{2}\right\rceil-1\right\}$, then for every $\gamma \in$ $\left\{1, \ldots,\left\lceil\frac{\phi}{2}\right\rceil-1\right\}$ and for every $c$, with $1 \leq c \leq k$ :

$$
\sum_{I_{r} \in I(\gamma)\left[s_{r^{\prime}}\right]\left[c_{r^{\prime}}\right]} w_{r}^{(c)}>\frac{1}{2}
$$

Hence,

$$
\begin{gathered}
o p t \geq \sum_{I_{r} \in I\left[s_{r^{\prime}}\right]} \frac{1}{k} \sum_{c=1}^{k} w_{r}^{(c)} \geq \\
\frac{1}{k} \sum_{\gamma=1}^{\left\lceil\frac{\phi}{2}\right\rceil-1} \sum_{I_{r} \in I(\gamma)\left[s_{r^{\prime}}\right]} \sum_{c=1}^{k} w_{r}^{(c)} \geq \\
\frac{1}{k} \sum_{\gamma=1}^{\left\lceil\frac{\phi}{2}\right\rceil-1} \frac{1}{2} \geq \frac{1}{k} \frac{\phi}{2} \frac{1}{2} \geq \frac{\phi}{4 k}
\end{gathered}
$$

Case 2: For every color $\gamma$ in $\left\{\left\lceil\frac{\phi}{2}\right\rceil, \ldots, \phi-1\right\}$, there is a category $c$, with $1 \leq c \leq k$, such that

$$
\sum_{I_{r} \in I(\gamma)\left[s_{r}\right][c]} w_{r}^{(c)}>\frac{1}{2}
$$

By a reasoning analogous to Case 1, it follows that:

$$
\begin{gathered}
o p t \geq \sum_{I_{r} \in I\left[s_{s^{\prime}}\right]} \frac{1}{k} \sum_{c=1}^{k} w_{r}^{(c)} \geq \\
\frac{1}{k} \sum_{\gamma=\left\lceil\frac{\phi}{2}\right\rceil}^{\phi-1} \sum_{I_{r} \in I(\gamma)\left[s_{r^{\prime}}\right]} \sum_{c=1}^{k} w_{r}^{(c)} \geq \\
\frac{1}{k} \sum_{\gamma=\left\lceil\frac{\phi}{2}\right\rceil}^{\phi-1} \frac{1}{2} \geq \frac{1}{k} \frac{\phi}{2} \frac{1}{2} \geq \frac{\phi}{4 k}
\end{gathered}
$$

Since appr $=\phi$, an approximation of $4 k$ holds.
The above problem, when considered as an off-line problem, is APX-hard since it contains a multidimensional bin-packing as a special case. Multidimensional bin-packing is APX-hard [7] already for $|C|=2$.

In the second generalization, each request $r$ is characterized by a gender bandwidth rate $g_{r, c_{r}}$ associated to the category $c_{r}$ and by a bandwidth rate $w_{r}$. The overall sum of the bandwidth rates of the simultaneous requests assigned to the same server at the same time is bounded by 1 , as well as the ov erall sum of the gender bandwidth rates of the simultaneous requests of the same category assigned to the same server at the same time, which is also bounded by 1.

Problem 7. (In terval Coloring with Double Weighted Overlapping) Given a set I of intervals, with each interval $I_{r}$ char acterize $d$ by a gender weighty $r_{r, c_{r}} \in(0,1]$ asso ciated to the cate gory $c_{r}$ and by a bandwidth weight $w_{r} \in(0,1]$, assign a color to $e$ ach interval in such a way that the overall sum of the gender weights for mutually overlapping intervals of the same cate gory $r$ eceiving the same color is boundd by 1 , the overall sum of the bandwidth weights for mutually overlapping intervals receiving the same color is bounded by 1, and the minimum number of colors is used.

F ormally, the constraints of Problem 7 are given below:

$$
\begin{gather*}
\sum_{I_{r} \in I(\gamma)[t]} w_{r} \leq 1 \quad \forall \gamma, \forall t  \tag{7}\\
\sum_{I_{r} \in I(\gamma)[t][c]} g_{r, c} \leq 1 \quad \forall \gamma, \forall t, \forall c \tag{8}
\end{gather*}
$$

Problem 7 can again be solved on-line by Algorithm 2, in troduced in the previous section.

Theorem 5.2. The First-Color algorithm provides a 10-appr oximation for Interval Coloring with Double Weighted Overlapping.

## 6 Conclusions

This paper has considered sev eral on-line approximation algorithms for problems arising in infostations, where a set of requests characterized by categories and temporal intervals ha veto be assigned to servers in sucha w aythat a bounded number of simultaneous requests are assigned to the same server and the number of servers is minimized. How eer, several questions still remain open. For instance, one could low erthe approximation bounds deriv edfor Problems 5, 6 and 7. Moreover, one could consider the scenario in which the n umber of servers is given in input, each request has a deadline, and the goal is to minimize the overall completion time for all the requests.

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