CSIS0351/8601: Randomized Algorithms
Lecture 9: Lovasz Local Lemma, Job Shop Scheduling
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Date: 15 Nov 2010

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## 1 Limited Dependency

We have seen how the union bound is used in probabilistic method. Suppose $A_{0}, A_{1}, \ldots, A_{n-1}$ are $b a d$ events (not necessarily independent) such that each $\operatorname{Pr}\left[A_{i}\right] \leq p$. If $n p<1$, then by union bound, we conclude that $\operatorname{Pr}\left[\cup_{i} A_{i}\right] \leq n p<1$, and hence, with positive probability, none of the bad events occur. We see that if the events have limited dependency, we can have the same conclusion under a weaker condition.

Definition 1.1 (Dependency Graph) Suppose $A_{0}, A_{1}, \ldots, A_{n-1}$ are events in some probability space. A dependency graph $H=(V, E)$ is a graph with vertex set $V=[n]$ such that for each $i \in[n]$, if $J:=\{j:\{i, j\} \in E\}$ is the set of neighbors of $i$, then the event $A_{i}$ is independent of all the events $\left\{A_{j}: j \notin J\right\}$.
Formally, for any disjoint subsets $J_{1}, J_{2} \subseteq[n] \backslash J$,
$\operatorname{Pr}\left[A_{i}\right]=\operatorname{Pr}\left[A_{i} \mid\left(\cap_{j \in J_{1}} A_{j}\right) \cap\left(\cap_{j \in J_{2}} \overline{A_{j}}\right)\right]$.
Remark 1.2 Observe that the dependency graph is not unique. The complete graph is trivially a dependency graph, but not a very useful one.

### 1.1 Example: Monochromatic Subsets

Recall from the first lecture that $S_{1}, S_{2}, \ldots, S_{m}$ are $l$-subsets of $U$. We show that if $m<2^{l-1}$, then it is possible to color each element of $U$ BLUE or RED such that none of $S_{i}$ is monochromatic. We show that if the subsets have limited intersection, then $l$ does not have to depend on $m$.
Theorem 1.3 (Lovasz Local Lemma) Suppose the collection $\left\{A_{i}: i \in[n]\right\}$ of events has a dependency graph with maximum degree $D \geq 1$. Suppose further that for each $i, \operatorname{Pr}\left[A_{i}\right] \leq p$, and $4 p D \leq 1$. Then, $\operatorname{Pr}\left[\cup_{i} A_{i}\right]<1$, i.e., with positive probability, none of the events $A_{i}$ happens.

Claim 1.4 Suppose each subset $S_{i}$ intersects at most $2^{l-3}$ other subsets. Then, it is possible to color each element of $U$ BLUE or RED such that none of the subsets $S_{i}$ is monochromatic.

Proof: For each element in $U$, we pick a color uniformly at random. Let $A_{i}$ be the event that the subset $S_{i}$ is monochromatic. Then, $p:=\operatorname{Pr}\left[A_{i}\right]=\frac{1}{2^{l-1}}$.
Observe that the event $A_{i}$ is independent of all events $A_{j}$ 's such that $S_{i} \cap S_{j}=\emptyset$. Hence, in the dependency graph $H=([n], E),\{i, j\} \in E$ iff $S_{i} \cap S_{j} \neq \emptyset$. The maximum degree is $D \leq 2^{l-3}$.
Hence, $4 p D \leq 4 \cdot \frac{1}{2^{l-1}} \cdot 2^{l-3} \leq 1$. By Lovasz Local Lemma, $\operatorname{Pr}\left[\cup_{i} A_{i}\right]<1$.

## 2 Proof of Lovasz Local Lemma

We shall prove the following claim.
Claim 2.1 If $S \subseteq[n]$ and $i \notin S$, then $\operatorname{Pr}\left[A_{i} \mid \cap_{j \in S} \overline{A_{j}}\right]<\frac{1}{2 D}$.
The result follows from the claim because
$\operatorname{Pr}\left[\cap_{i} \overline{A_{i}}\right]=\prod_{i} \operatorname{Pr}\left[\overline{A_{i}} \mid \cap_{j<i} \overline{A_{j}}\right]>\left(1-\frac{1}{2 D}\right)^{n}>0$.
We next prove the claim by induction on the size of $S$.
Base Case. $|S|=0$. In this case, $\operatorname{Pr}\left[A_{i} \mid \cap_{j \in S} \overline{A_{j}}\right]=\operatorname{Pr}\left[A_{i}\right] \leq p \leq \frac{1}{4 D}<\frac{1}{2 D}$.
Inductive Step. Suppose the result holds for all $S$ such that $|S|<r$, for some $r \geq 1$. We now consider $|S|=r$.
Suppose $i \notin S$. Consider decomposition of $S$ into two sets:
(1) $S_{1}:=\{j \in S:\{i, j\} \in E\}$;
(2) $S_{2}:=\{j \in S:\{i, j\} \notin E\}$.

If $S_{1}:=\emptyset$, then $\operatorname{Pr}\left[A_{i} \mid \cap_{j \in S} \overline{A_{j}}\right]=\operatorname{Pr}\left[A_{i} \mid \cap_{j \in S_{2}} \overline{A_{j}}\right]$. By the dependency assumption, the latter quantity equals $\operatorname{Pr}\left[A_{i}\right] \leq p<\frac{1}{2 D}$.
We next consider $S_{1} \neq \emptyset$. Hence, $\left|S_{2}\right|<r$.
Observer that

$$
\operatorname{Pr}\left[A_{i} \mid \cap_{j \in S} \overline{A_{j}}\right]=\frac{\operatorname{Pr}\left[A_{i} \cap\left(\cap_{j \in S_{1}} \overline{A_{j}}\right) \mid \cap_{j \in S_{2}} \overline{A_{j}}\right]}{\operatorname{Pr}\left[\cap_{j \in S_{1}} \overline{A_{j}} \mid \cap_{j \in S_{2}} \overline{A_{j}}\right]} .
$$

We first consider the numerator. First, $\operatorname{Pr}\left[A_{i} \cap\left(\cap_{j \in S_{1}} \overline{A_{j}}\right) \mid \cap_{j \in S_{2}} \overline{A_{j}}\right] \leq \operatorname{Pr}\left[A_{i} \mid \cap_{j \in S_{2}} \overline{A_{j}}\right]=\operatorname{Pr}\left[A_{i}\right] \leq$ $p$, where the equality in the middle follows from the dependency assumption.
We next consider the denominator. For $j \in S_{1}$, we have $\operatorname{Pr}\left[A_{j} \mid \cap_{j \in S_{2}} \overline{A_{j}}\right]<\frac{1}{2 D}$. We can apply the induction hypothesis because $j \notin S_{2}$ and $\left|S_{2}\right|<r$. By the union bound, we conclude that $\operatorname{Pr}\left[\cup_{j \in S_{1}} A_{j} \mid \cap_{j \in S_{2}} \overline{A_{j}}\right]<\frac{\left|S_{1}\right|}{2 D} \leq \frac{1}{2}$. Hence, $\operatorname{Pr}\left[\cap_{j \in S_{1}} \overline{A_{j}} \mid \cap_{j \in S_{2}} \overline{A_{j}}\right]>\frac{1}{2}$.
Therefore, we have $\operatorname{Pr}\left[A_{i} \mid \cap_{j \in S} \overline{A_{j}}\right]<\frac{p}{1 / 2}=2 p \leq \frac{1}{2 D}$, finishing the inductive step of the proof.

## 3 Job Shop Scheduling

Problem Instance. We are given $m$ jobs $J_{1}, J_{2}, \ldots J_{m}$ and $n$ machines $M_{0}, M_{2}, \ldots, M_{n-1}$ with the following rules.

1. Each job $J_{j}$ must be processed by some subset of machines in a specific given order. Each job is processed by a particular machine at most once. For example, job $J_{1}$ has to be processed by the machines in the order $M_{6}, M_{1}, M_{5}, M_{3}$.
2. It takes one unit of time for a machine to process a job during its turn; this is the same over all machines and jobs. A machine can only process at most 1 job at the same time.

Goal. Schedule the jobs among the machines so that the makespan, which is the time for the last job to be finished, is minimized.

## Some Easy Lower Bounds.

In a problem instance, let $C$ be the number of jobs performed by the machine processing the most number of jobs. Since each machine can only process at most one job in one time step, it cannot finish before time $C$.

On the other hand, let $L$ be the number of machines required by the job having the longest machine sequence. Since each machine takes one time-step to perform a job, the job cannot finish before time $L$. Hence, $T:=\max \{C, L\}$ is a lower bound on the makespan of the optimal schedule.

### 3.1 Random Delay

We will use randomness and Lovasz Local Lemma to show there exists a schedule whose makespan approaches the lower bound asymptotically.
Theorem 3.1 There exists a schedule with makespan $2^{O\left(\log ^{*} T\right)} T$.
Remark 3.2 Given a positive integer $n, \log ^{*} n$ is the smallest non-negative integer $i$ such that $\log _{2}^{(i)} n<2$, i.e., the number of times one can take logarithms before the number drops below 2 .
The function log* grows very slowly, and in practice it can be considered as a constant. For instance, $\log ^{*} 2^{65536}=5$, where $2^{65536}$ has more than 19,000 digits in base 10 .
Relax Assumption. Suppose we relax the assumption and allow each machine to process more than 1 job at the same time. However, a job still takes one time-step to be processed by a machine in the required sequence. Then, there is a relaxed schedule $S_{0}$ with makespan $L$. We show how to convert this infeasible schedule to one that is feasible.

Definition 3.3 In a schedule (not necessarily feasible), the relative congestion of a machine in some time window is the ratio of number of jobs performed in the window divided by the number of time steps in the window.
Hence, it follows that if we consider the relaxed schedule $S_{0}$, the relative congestion of each machine in the time window of size $T$ is at most 1 . Moreover, a schedule is feasible if the relative congestion of each machine in all time windows with 1 time step is at most 1 .
We define $T_{0}:=T$, and have the following invariant. In schedule $S_{i}$, each machine in any time window of size $T_{i}$ or more has relative congestion at most 1 . The goal is to decrease $T_{i}$ at each step, at the cost in increasing the makespan.

Scheduling by Random Delay. We convert the schedule $S_{0}$ into $S_{1}$ in the following way. For each job $J_{j}$, pick an integer $x_{j}$ uniformly at random from $\{0,1,2, \ldots, 2 T-1\}$ independently. Each job $J_{j}$ delays for $x_{j}$ time steps before starting as before. As before, we still allow machines to work on more than 1 job at the same time.
We next show that with positive probability, for some $T_{1}<T$, all windows of size $T_{1}$ or more for each machine under schedule $S_{1}$ have relative congestion at most 1 .

### 3.2 Applying Lovasz Local Lemma

Lemma 3.4 There is some function $f: \mathbb{N} \rightarrow \mathbb{N}$ where $f(n)=\Theta(\log n)$ such that with $T_{1}=f(T)$, there is a positive probability that, under schedule $S_{1}$, for every machine, every window of size $T_{1}$ or more has relative congestion at most 1.
Proof: For each machine $M_{i}$, define $A_{i}$ to be the event that there is some window with size at least $T_{1}$ for machine $M_{i}$ that has relative congestion larger than 1 . We specify the exact value of $T_{1}$ later. For the time being, think of $T_{1}=O(\log T)$.

We next form a dependency graph $H=([n], E)$ such that $\{u, v\} \in E$ iff both machines $M_{u}$ and $M_{v}$ process the same job. Observe that $A_{i}$ is independent of all the $A_{j}$ 's for which $M_{i}$ and $M_{j}$ do not process any common job.
We estimate the maximum degree of $H$. Consider machine $M_{i}$. Observe that it can process at most $C \leq T$ jobs. Each of those jobs can go through at most $L \leq T$ machines. Hence, the maximum degree of $H$ is $D \leq T^{2}$.
We next give an upper bound on $\operatorname{Pr}\left[A_{i}\right]$. Consider a fixed window $W$ of size $\tau \geq T_{1}$ for machine $M_{i}$. For each job $J_{j}$ that is being processed by machine $M_{i}$, we define $X_{j}$ to be the indicator random variable that takes value 1 if job $J_{j}$ falls into the window $W$ for machine $M_{i}$, and 0 otherwise.

Observe that $X_{j}$ 's are independent, because the random delays are picked independently. Moreover, $E\left[X_{j}\right]=\operatorname{Pr}\left[X_{j}=1\right] \leq \frac{\tau}{2 T}$.
Define $Y$ to be the number of jobs that fall into the window $W$ for machine $M_{i}$. Then, $Y$ is the sum of $X_{j}$ 's for the jobs $J_{j}$ that are performed by machine $M_{i}$. Note that $Y$ is a sum of at most $T$ independent $\{0,1\}$-independent random variables, each of which has expectation at most $\frac{\tau}{2 T}$.
Introducing Dominating Random Variable $Z$. We define $Z$ to be a sum of $T$ independent $\{0,1\}$-independent random variables, each of which has expectation exactly $\frac{\tau}{2 T}$. Intuitively, $Z$ is more likely to be larger than $Y$. (This can be proved formally. We will talk about this in details in the next lecture.) Observe that $E[Z]=\frac{\tau}{2}$.
Hence, $\operatorname{Pr}[Y>\tau] \leq \operatorname{Pr}[Z>\tau]=\operatorname{Pr}[Z>2 E[Z]]$. By Chernoff Bound, this is at most $\exp \left(-\frac{E[Z]}{3}\right)=\exp \left(-\frac{\tau}{6}\right) \leq \exp \left(-\frac{T_{1}}{6}\right)$. Observe that if we had not used $Z$ to analyze $Y$, then since $E[Y] \leq \frac{\tau}{2}$, we would have obtained $\exp \left(-\frac{E[Y]}{3}\right) \geq \exp \left(-\frac{\tau}{6}\right)$, i.e., the direction of the inequality is not what we want.
Note that there are trivially at most $(3 T)^{2}$ windows. Hence, using union bound, we have $\operatorname{Pr}\left[A_{i}\right] \leq$ $9 T^{2} \cdot \exp \left(-\frac{T_{1}}{6}\right)=: p$.
Hence, in order to use Lovasz Local Lemma, we need $4 p D \leq 1$. Therefore, it is enough to have $36 T^{2} \exp \left(-\frac{T_{1}}{6}\right) \cdot T^{2} \leq 1$. We set $T_{1}:=6 \ln \left(36 T^{4}\right)=\Theta(\log T)$.
By the Lovasz Local Lemma, $\operatorname{Pr}\left[\cap_{i} \overline{A_{i}}\right]>0$. Hence, the result follows.
Conclusion. We begin with a schedule $S_{0}$ of makespan at most $P_{0}:=T$ such that every window of size $T_{0}:=T$ or more for each machine has relative congestion at most 1 . After the transformation, we obtain a schedule $S_{1}$ of make span at most $P_{1}=3 P_{0}$ such that every window of size $T_{1}=$ $6 \ln \left(4 T_{0}^{4}\right)$ or larger for each machine has relative congestion at most 1 .

### 3.3 Recursive Transformation

Observe that we can apply the same transformation to schedule $S_{1}$. In particular, we divide the total time into windows of size $T_{1}$, and apply the same transformation separately for each window to obtain schedule $S_{2}$, with makespan at most $P_{2}=3 P_{1}$ such that for each machine, every window of size $T_{2}=f\left(T_{1}\right)$ or more has relative congestion at most 1 . Here, the function $f$ comes from Lemma 3.4.

Hence, we have the series $T_{0}:=T, P_{0}:=T$, and $T_{i+1}:=f\left(T_{i}\right)$ and $P_{i+1}:=3 P_{i}$. This process can continue as long as $f\left(T_{i}\right)<T_{i}$.
The process stops when $f\left(T_{k}\right) \geq T_{k}$, at which point $T_{k}$ is at most some constant $K_{f}$, which depends only on the function $f$. This means in schedule $S_{k}$, in each time step, each machine has to deal with at most $K_{f}$ of jobs. Hence, it is easy to increase the makespan by a further $K_{f}$ factor to make the schedule feasible. It follows that we have a feasible schedule with makespan at most $K_{f} \cdot P_{k}=O\left(3^{k} \cdot T\right)$.
It remains to bound the value of $k \leq \min \left\{i: f^{(i)}(T) \leq K_{f}\right\}$. Since $f(T)=\Theta(\log T)$, it follows that $k=O\left(\log ^{*} T\right)$. Hence, we have a feasible schedule with makespan $2^{O\left(\log ^{*} T\right)} T$, proving Theorem 3.1.

## 4 Homework Preview

1. Calculation Involving $\log ^{*}$. In this question, you are asked to complete the details of some calculations.
(a) Deriving $K_{f}$. Suppose $f(t):=6 \ln \left(36 t^{4}\right)$. Derive a constant $K_{f}>0$ such that $f(t) \geq t$ implies that $t \leq K_{f}$.
(b) Suppose $k:=\min \left\{i: f^{(i)}(T) \leq K_{f}\right\}$. Prove that $k=O\left(\log ^{*} T\right)$.
(Hint: Make use of big-O notation carefully and avoid messy calculations.)
2. Packet Routing in a Graph. We describe a problem that is closely related to job shop scheduling.

Problem Instance. Suppose $G=(V, E)$ is a directed graph. We are given $m$ sourcesink pairs $\left\{\left(s_{j}, t_{j}\right): j \in[m]\right\}$. We wish to send one data packet from each source to its corresponding sink.
(a) For each $j \in[m]$, a packet must be sent from $s_{j}$ to $t_{j}$ via some specific path $P_{j}$. Each path $P_{j}$ is simple: this means each (directed) edge appears at most once in $P_{j}$.
(b) It takes one unit of time for a data packet to be sent through a directed edge. An edge can only allow at most 1 data packet to be sent at any time.

Goal. Schedule the packets to be sent in the graph so that the makespan, which is the time for the last packet to arrive at its sink, is minimized.
Show that the packet routing problem can be reduced to the job shop scheduling problem. In particular, given an instance of the packet routing problem, construct an instance of the job
shop scheduling problem such that there exists a packet schedule with makespan $T$ iff there exists a job shop schedule with makespan $T$.

