

Lecture 5: The Lovasz Local Lemma (Beyond the Union Bound).

Until now: sps. want to show something good happens. → nothing bad.

$$\text{Show: } \sum_{i: \text{all bad events}} \Pr(\text{bad event } i) < 1 \Rightarrow \Pr(\bigcup_i \text{Bad event } i) < 1$$

unim bound.

$$\Rightarrow \text{Pr}(\text{nothing bad happens}) > 0.$$

E.g.: k -satisfiability.

$$\varphi = (x_1 \vee \bar{x}_5 \vee x_9) \wedge (x_2 \vee \bar{x}_3 \vee \bar{x}_{11}) \wedge \dots \dots \dots \quad k=3.$$

m clauses, n variables, k -CNF formula

Want to understand when φ must have satisfying assignment.

Pick a random assignment: $x \in \{T, F\}^n$ unif. at random.

B_i = event that clause i is unsatisfied.

$$\Pr(B_i) = 2^{-k} \quad (\text{all vars set "wrong"}).$$

$$\Rightarrow \text{if } m \cdot 2^{-k} < 1 \Rightarrow \Pr(\bigcup_i B_i) < 1 \Rightarrow \Pr(\varphi \text{ satisfied by random assignment}) > 0$$

unim bound

$\Rightarrow$ $\exists$ a satisfying assignment for φ .
if $m \leq 2^k$.

Also: cannot do ^{any} ~~much~~ better.

if $m = 2^k$ could have all 2^k clauses on some k variables.
one must be false in every assignment

$\Rightarrow$ could be unsatisfiable.

Suppose we know more facts:

• Suppose φ is k -CNF formula (m clauses)
 n vars

and each variable occurs on only L clauses. from the "small" L
 $\hookrightarrow$ "locality" constraint.

Then can we say that φ is satisfied with many true clauses?

Thm: Sps. each var appears in $< 2^{k-2}/k$ clauses.

then φ is satisfiable indep of values of m, n .

Just depends on the "locality" value.

Follows from an important tool: the Lovász Local Lemma
(in a paper of Erdős and Lovász).

Needs some buildup.

Consider some probability space.

• Define a collection of good events

• for any subset $S \subseteq [m]$, let

$$G = (G_1, G_2, \dots, G_m)$$

$\bigcap_{i \in S} G_i$ be written
as G_S

dependency
 • Define a graph on these events $H = ([m], E)$

such that $\forall i$, event G_i is independent of all events not in its neighborhood.

i and its neighborhood

Formally: for all S, T disjoint subsets of $[m] \setminus (\partial_H(i) \cup i)$

$$\Pr(G_i) = \Pr(G_i \mid \left(\bigcap_{j \in S} G_j \right) \cap \left(\bigcap_{j \in T} \bar{G}_j \right))$$

$\uparrow$ all events in S happened $\uparrow$ all events in T did not happen

Example: bad event B_i = clause i not satisfied by random x .

good event $G_i = \bar{B}_i$ = clause i satisfied by x .

then can define $E = \{(i, j) \mid \text{clause } i \text{ and clause } j \text{ do not share variables}\}$.

Now: $\Pr(\text{clause } i \text{ sat}) = \Pr(\text{clause } i \text{ sat} \mid \text{any settings of vars not in } i)$

$$= \Pr(G_i \mid \text{any events for clauses not in } i \cup \partial_H(i))$$

Note: these dependency graph was simple to check.

Should be careful that we check that G_i is independent of all subsets S, T etc.

OK. Now the LLL:

Lovász Local Lemma (1975) Symmetric Version.

Let $G_1, G_2 \dots G_m$ be events. Let $H = ([m], E)$ be dependency graph for them.

Suppose $\Pr(\bar{G}_i) \leq p \quad \forall i$
 $\leftarrow \Pr(\text{bad event}) \leq \text{small}$

And suppose $\max \text{degree of } H \leq d.$

$\Rightarrow \Pr(\cap G_i) > 0$ as long as $p d \leq 1/4.$

$\uparrow$
degree of
dependency
small

$\uparrow$
 $\Pr(\text{all good events happen})$
 $= 1 - \Pr(\text{some bad event happens})$

$\uparrow$
much better than
 $p \cdot m < 1$
by union bound!!

let's prove Thm 1 (using the LLL)

$G_i = \{\text{clause } i \text{ sat'd by random } x\} \Rightarrow \Pr(\bar{G}_i) = 1/2^k.$

also $H = \{(i, j) \mid G_i \text{ and } G_j \text{ share vars}\}.$

~~also~~ Now: G_i contains k vars

Each belongs to $\leq 2^{k-2}/k$ clauses

$\Rightarrow G_i$ adjacent to $\leq 2^{k-2}$ other G_j in $H.$

$\Rightarrow d \leq 2^{k-2}.$

$\Rightarrow p \cdot d \leq 2^{-k} \cdot 2^{k-2} = 1/4$ as desired

$\Rightarrow \Pr(\cap G_i) = \Pr(\text{all clauses sat'd}) > 0.$

$\Rightarrow \exists$ a satisfying assignment 😊.

probabilistic
method!

Great: Next steps.

- ① Prove LLL.
- ② Prove it again (via "witness" tree) — get an algo from it.
- ③ Hint of yet another proof (maybe). — using entropy.
- ④ Application to the Beck-Fiala Problem.

— X —

① Original proof of LLL was "not algorithmic".

Pf: define $G_S = \bigcap_{i \in S} G_i$.

Claim: $\exists S \subseteq [m]$ s.t. $\Pr(G_S) > 0$. for $i \notin S$, $\Pr(\bar{G}_i | G_S) \leq 2p$.

Pf: ~~Induction on |S|~~ Induction on |S|. $|S|=0$ is trivial, since $G_\emptyset = \Omega$ and $\Pr(\bar{G}_i) \leq p$ by the assumption.
for $|S| \neq \emptyset$, let T be nbrs of G_i , $U \subseteq S$ be non neighbors of G_i .

$$\Pr(\bar{G}_i | G_T \cap G_U) = \frac{\Pr(\bar{G}_i \cap G_T \cap G_U)}{\Pr(G_T \cap G_U)} = \frac{\Pr(\bar{G}_i \cap G_T | G_U) \cancel{\Pr(G_U)}}{\Pr(G_T | G_U) \cancel{\Pr(G_U)}}$$

$$\leq \frac{\Pr(\bar{G}_i | G_U)}{\Pr(G_T | G_U)} = \frac{\Pr(\bar{G}_i)}{\Pr(G_T | G_U)} \left[\begin{array}{l} \uparrow \text{smaller event} \\ \uparrow \text{SP} \\ \uparrow \Pr(G_U) \neq 0 \Rightarrow \Pr(G_U) \neq 0. \end{array} \right]$$

union bound

$$1 - \prod_{j \in T} \Pr(G_j | G_U) \geq 1 - |T| \cdot 2p \geq \frac{1}{2}.$$

$\uparrow$ if $\exists j \in T \Rightarrow |U| < |S| \Rightarrow$ induction !!

$$\Rightarrow \Pr(\bar{G}_i | G_S) \leq \frac{p}{1/2} = 2p.$$

Now: $Pr(G_{[m]}) = \prod_{i=1}^m Pr(G_i | G_{(j \leq i)}) \geq \prod_{i=1}^m (1-2p) > 0.$

$pd \leq 1/4 \Rightarrow p \leq 1/4$
 $\square$

————— x —————

OK:

Can improve LLL condition to say:

if $p(d+1) \leq 1/e \Rightarrow Pr(G_i) > 0.$

Similar proof
see books

And this is best possible.

————— x —————

OK. Great. Show these amazing existential results (like for k-SAT)
and Packet routing
and Beck Fiala....

E.g.:

Routing: Suppose graph G .

s_i, t_i source-sink pairs

P_i : ^{fixed.} path from source s_i to sink t_i .

Want to send 1 packet _{each} from s_i to t_i

Only one packet per edge per time.

Like hypercube routing, now fixed paths.

How long?

Must take: Dilation $D = \max_i |P_i|$ max path length.

Congestion $C = \max_e |\{i : P_i \text{ uses edge } e\}|$

Thm: [Leighton Morsg Rao]

$\exists$ a routing algorithm that sends packets, and completes

in $O(C+D)$ steps!! (For any graphs!)
(Any paths!)

Section!

So far: existential results, $\exists$ a good outcome ~~st~~ b/c $\Pr(\bigcap_{i \in [m]} G_i) > 0$.

But only showed $\Pr(\bigcap G_i) \geq (1-2p)^m$. Could be tiny!

Not algorithm that's efficient.

Let's fix that problem... (quite spectacularly)

————— X —————

Focus on k -SAT, same idea works in general.

A Trivial (?) Algorithm

- (i) Start with a uniformly random truth assignment x .
- (ii) while $\exists$ a clause C unsatisfied:
 re flip all vars in C .

Does this even stop? (Even if each var in $\leq 2^{k-2}/k$ clauses).

Thm 2: Trivial Algo stops in expected $O(m)$ time !!!

[Moser
- Tardos 2010]

Before we prove thm...

Same idea for many LL applications. [see Moser-Tardos, ^{other people's} Lecture notes ...] on webpage.

Sps each event G_i depends on some set $\text{vars}(i)$ of underlying rvs.

And $E = \{\{i, j\} \mid \text{vars}(i) \cap \text{vars}(j) \neq \emptyset\}$ be dependency graph

Then trivial algo says: -

[while $\exists i$ st. G_i not satisfied:

 re randomize vars in $\text{vars}(i)$.]

Pf (for Thm 2)

Proceeds by building 'witness' trees. Suppose $C_1, C_2, C_3, \dots, C_t, \dots$
be clauses refuted by algo. (a clause may appear many times).

$\forall t$: build a witness tree T_t as follows.

- C_t is root

- for $i = t-1, t-2, \dots, 1$.

if clause C_i has no vars in common with clauses in current tree, discard.

Else add as child of deepest node it shares vars with.

Claim 1: $\Pr(\text{some tree } T_i \text{ appears}) \leq (2^{-k})^{|T_i|}$.

Pf: peel off some clause that is a leaf. It was refuted.

$\Rightarrow$ not sat'd by orig assignment (with prob 2^{-k}).

remove it, look at next leaf (in peeled tree).

Either leaf in orig tree (so unsat in orig assignment, w.p. 2^{-k}).

or was unsat after its children were flipped.

But each var is again conditionally indep and unbiased
 $\Rightarrow$ unsat w.p. 2^{-k} .

This happens for each of $|T_i|$ clauses $\Rightarrow \Pr(\text{size } s \text{ tree}) \leq (2^{-k})^s$.

■

Claim 2: # of possible trees of size $s \leq m \cdot \binom{(d+1)s}{s-1} \leq m(e(d+1))^s$

Pf: Suppose clauses are called $K_1, K_2, \dots, K_m$ (in this order).
"klausess"

How to specify tree:-

- ① write down root (m possible options)
- ② Now traverse this tree in some order (say preorder)
for each node, write its children, smartly.

Note: by the way we built tree, each ~~tree~~ node (clauses)
children are some subset of its neighbors (or itself)
(no repeats). (d+1) choices

So write down a (d+1)-bit vector indicating which are children.

(apart from root)

$\Rightarrow$ tree^y is given by bit string of length $(s-1)(d+1)$
with $(s-1)$ 1s in it.

000...0
 $\Rightarrow$ no children

$\uparrow$ total of $s-1$ children (since 1 root)

$$\Rightarrow \# \text{ trees} \leq m \cdot \binom{s(d+1)}{s-1} \leq m(e(d+1))^s \text{ trees.}$$

$\Rightarrow$ Proof of LLL: each tree has prob $\leq (2^{-k})^s = p^s$.

and at most $m(e(d+1))^s$ trees.

$$\begin{aligned} \Rightarrow E(\# \text{ of trees seen by algo}) &\leq \sum_s p^s \cdot m(e(d+1))^s \\ &= m \cdot \sum_s (e \cdot p \cdot (d+1))^s = O(m) \end{aligned}$$

if $ep(d+1) < 1$.

This kind of witness tree argument
also used for 2-choice proofs, etc.

Powerful technique.



The Beck-Fiala Conjecture & Discrepancy

Suppose subsets $S_1, S_2, \dots, S_m \subseteq [n]$.

We want to partition the elements of $[n]$ into two groups Red/Blue

such that $\max_i |(\text{Red} \cap S_i) - (\text{Blue} \cap S_i)|$ is minimized

discrepancy of coloring

coloring

Why? { Pick a subset of half the people from the room s.t.
people with each characteristic (gender / education / income)
are fairly preserved.

Can repeat (or use similar ideas) to try and get a representative subset of size $s, \dots$ (see work on "Distortion")

— x —

OK: In Exercises from HW2,

show that if random coloring, then

$$\text{discrepancy} \leq \sqrt{n \log m} \quad \leftarrow \begin{array}{l} \text{whp.} \\ \# \text{ subsets} \end{array}$$

$\uparrow$
elements

• Suppose we know that each person belongs to $\leq K$ subsets

⇒ Beck Fiala Conjecture says:-

$$\text{discrepancy} \leq O(\sqrt{K})$$

Open!

Best results show: $O(K) \sim O(\sqrt{K \log \log n})$

→ [Beck Fiala]

→ In fact, Bansal-Jiang recently!
Solve the case $K \geq \log^2 n$
Bansal Jiang
FOCS 25

However, suppose

(a) Each person belongs to $\leq k$ subsets

(b) Each subset contains $\leq k$ people ← Extra condition

$\Rightarrow$ can show discrepancy $\leq O(\sqrt{k \log k})$ using the LLL.

Pf: (via LLL).

① Do a random coloring with each person being $\overset{+1}{R}/\overset{-1}{B}$ w.p. $\frac{1}{2}$ indep.

② $\Pr(\text{Set } S_i \text{ has discrepancy} \geq c\sqrt{k \log k})$

$$= \Pr\left(\underbrace{\left|\sum x_{ij}\right|}_{\text{Radomachors}} \geq c\sqrt{k \log k}\right) \leq \exp\left(-\frac{c^2(k \log k)}{2 \cdot k}\right) \approx \frac{1}{K^{10}}$$

say if $c \geq \sqrt{20}$.

③ Define G_i = event that S_i has discrepancy $\leq c\sqrt{k \log k}$
 $\Rightarrow \Pr(\overline{G_i}) \leq 1/K^{10} = p$ "balanced"

Moreover

define dependency graph

define $E = \{(i, j) \mid S_i \cap S_j \neq \emptyset\}$

then degree of this graph $\leq |S_i|$, #sets containing any element
vertex i $\leq k$, $K = k^2$.

$$\Rightarrow d \leq K^2$$

$$\Rightarrow p \cdot d \leq \frac{1}{4} \text{ for } k \geq 2$$

$\Rightarrow$ LLL says $\exists$ solution where all sets "balanced".

Glimpse of Entropy-Compression Proof

Robin Moxer's Stoc 2010 talk

+ Terry Tao + Mary Woollens
blog notes.

Focus on k -SAT:

Important: if clause violated, every literal in it evals to False.

So by telling you the name of ^{violated} clause, I give you k bits of info.

But naively, need $\log m$ bits to say name.

let's name things
smartly, using low degree.

— $\times$ —

- Let's run algo, and suppose it runs forever. (say $\geq T$ steps)

$\Rightarrow$ use $n + kT$ random bits.

- Let's try to write random bits another way:-

Trivial algo:

Pick n bits randomly for assignment

Let $K_1, K_2 \dots K_j$ be violated
for these violated clauses $i=1 \dots j$

Print "fixing clause K_i "

Fix(K_i)

output final assignment.

Fix(K_i)

rerandomize bits in K_i

Suppose its children are

$K_i = K_{i1}, K_{i2}, K_{i3} \dots K_{id}$

for violated children K_{ij}

print " j th child"

Fix(K_{ij})

output "wrapping up"

we perform

if T fixes: - output

$\cdot O(m \log m)$ bits in "fixing clause XXXX"

$\cdot O(\log d)$ bits in " XX^{th} child"

$\cdot n$ bits at end.

$\times T$ steps
or "wrapping up"

observe:
can recover
the $Tk + n$
bits!

$\Rightarrow n \log m + n + T \log d$ bits.

But cannot "compress randomness"
- needs formalisation!!

Great:

→ we can time out after some 10 steps.

So suppose we run for T steps. then.

We've taken $n + TK$ bits of randomness

And output $m \log m + n + T(\log d + O(1))$ bits of transcript

And: it's possible to infer the $n + TK$ bits from this transcript !

Since $n + TK$ bits were truly random,
"cannot" compress them".

(formally, any compression scheme must use $\geq L - c$ bits for all but 2^{-c} fraction of them)

So morally: (and we can formalize this, see notes on webpage): -

$$\underline{n + TK} \leq \underline{m \log m + n + T(\log d + c)} \text{ bits}$$

← constant

⇒ if $d \neq 2^{k-c-1}$ then

$$n + TK \leq m \log m + n + T(k-1)$$

$$\Rightarrow T \leq m \log m. \text{ steps.}$$

Proof #2 (the Algorithmic proof) was one ^{other} way to capture this idea.

See notes for slightly different viewpoints. 2 approaches.

To wrap up:

LLL: collection of "bad events" $\{B_i\}_{i=1}^m$
(and associated good events $G_i = \overline{B_i}$)

$$\text{s.t. } \Pr(B_i) \leq p$$

$$\text{and } \deg_H(i) \leq d \quad \forall i$$

$$\text{then if } pd \leq 1/4 \quad (\text{or } p(d+1) \leq 1/e)$$

$$\Rightarrow \Pr\left(\bigcap_{i=1}^m G_i\right) > 0.$$

dependency graph

$$H = ([m], E)$$

st B_i is indep of

$$[m] \setminus (N(i) \cup \{i\})$$

- Better than union bound when applicable
- Elegant entropic / witness tree proof to show algorithmic versions
- Used in
 - Packet Routing (see upcoming HW)
 - Beck's lemma
 - k-sat Algorithms
 - and other surprising applications....