

Max SAT

input: Boolean formula ϕ over vars x_1, \dots, x_n in CNF
plus weight w_j for each clause C_j ($1 \leq j \leq m$)

goal: Find a truth assignment that max the total weight of satisfied clauses.

special case: $w_j = 1 \rightarrow$ max # of satisfied clauses.

Max- k -SAT: every clause has up to k literals.

Max-E- k -SAT: " " exactly k " .

Theorem: Max- k -SAT is NP-hard for any $k \geq 2$.

(Note that 2SAT is in P).

Simple Rand Alg for Max-SAT

This is perhaps the most obvious Rand Alg: flip a fair coin for each var x_i and set it to T/F with prob $\frac{1}{2}$.

Theorem (Johnson '74) This is a $\frac{1}{2}$ -approx for max-SAT

proof: let τ be the rand truth assignment and define

$$Y_j = \begin{cases} 1 & \text{if } C_j \text{ is sat} \\ 0 & \text{otherwise} \end{cases} \quad \text{let } W = \sum_{j=1}^m w_j \cdot Y_j$$

$$E[W] = \sum_{j=1}^m w_j E[Y_j] = \sum_{j=1}^m w_j \cdot \Pr[C_j \text{ is sat}] = \sum_j w_j \left(1 - \left(\frac{1}{2}\right)^{|C_j|}\right) \geq \frac{1}{2} \sum_j w_j \geq \frac{1}{2} \text{OPT}$$

Note that if $|C_j| \geq k$ for all j then

this is a $(1 - \frac{1}{2^k})$ -approx. \rightarrow $\frac{7}{8}$ -approx for Max-E-3SAT

Theorem (Håstad '99): There is no $(\frac{7}{8} - \epsilon)$ -approx for Max-E-3SAT

Unless $P=NP$.

Derandomizing using method of cond. expectation.

lemma: suppose we have assigned values $a_1 \dots a_i$ to var's $x_1 \dots x_i$. Then we can compute the expected value of the sol if the rest of the var's are assigned rand.

proof: let f' be the formula on var's $x_{i+1} \dots x_n$ obtained from f by substituting the values of $x_1 \dots x_i$ and simplifying f . Clearly the exp value of sol for any truth assignment to f' can be computed in polytime. Add this to the weight of clauses that are already satisfied by $x_1 \dots x_i$ & we obtain the exp. value of f given $x_1 \dots x_i$ are fixed.

- This suggests a simple deterministic alg:

consider x_1 : for each of T/F assignment to x_1 compute the exp value $E[W|x_1=T]$ & $E[W|x_1=F]$

whichever is bigger assign x_1 accordingly. do this for all var's iteratively. say this value is \bar{v}_{x_1}

$$\begin{aligned} E[W] &= E[W|x_1=T] \cdot \Pr[x_1=T] + E[W|x_1=F] \cdot \Pr[x_1=F] \\ &= \frac{1}{2} (E[W|x_1=T] + E[W|x_1=F]) \end{aligned}$$

if we set x_i as in above then $E[W | x_i = v] \geq E[W] \geq \frac{1}{2} \text{opt}$

Rand. alg with biased coin

First assume we have an instance where all clauses of length 1 have non-negated literals (notly $\bar{x}_3 = C_j$)

We set each var $x_i = T$ with prob P ($P > \frac{1}{2}$ TBD).

For any C_j :

- if $|C_j| = 1$ then $\Pr[C_j \text{ is sat}] = P$

- if $|C_j| \geq 2$ then $\Pr[C_j \text{ is sat}] = 1 - \underbrace{P^\alpha (1-P)^\beta}_{C_j = F} \geq 1 - P^{\alpha+\beta} \geq 1 - P^2$
 α : # of neg. literals
 β : # of non-neg literals
because $P > (1-P)$

lemma: $\Pr[C_j \text{ is sat}] \geq \min\{P, 1 - P^2\}$

$$\text{Set } P = 1 - P^2 \rightarrow P = \frac{\sqrt{5} + 1}{2} \approx 0.618$$

$$E[W] = \sum_j w_j \cdot \Pr[C_j \text{ is sat}] \geq P \sum_j w_j \geq P \cdot \text{opt}$$

Theorem: If all clauses of length 1 have positive literals then this is a P -approx $P = \frac{\sqrt{5} + 1}{2}$

what if there are 1-clauses with neg. literals? e.g. $C_j = \bar{x}_i$
we can easily replace \bar{x}_i with x_i and x_i with \bar{x}_i .

The only prob. is when both x_i & \bar{x}_i appear as a 1-clause. e.g. $C_j = x_i$ & $C_l = \bar{x}_i$

WLOG assum $w_j \geq w_l$ define $w'_i = w_l$ then

$$\text{opt} \leq \sum_j w_j - \sum_{i: \substack{\text{1-clause with neg} \\ \text{literal \& positive} \\ \text{1-clause}}} w'_i$$

U : set of indices of clauses excluding those with neg. literal

We can still show:

$$\begin{aligned} E[W] &= \sum_j w_j \cdot \Pr[C_j \text{ is set}] \geq \sum_{j \in U} w_j \cdot \Pr[C_j \text{ is set}] \\ &\geq p \sum_{j \in U} w_j \\ &\geq p \cdot \text{opt} \end{aligned}$$

Theorem: Biased coin implies a p -approx for max-SAT.

LP-rounding for Max-SAT

let $P_j (N_j)$ be the set of var's that appear in clause C_j in positive (negative) form. and $b_j = |P_j| + |N_j|$. for any x_i

we have $y_i \in \{0, 1\}$: $y_i = 1$ iff $x_i = T$ (F).

Also for each C_j we have $Z_j = 1$ iff C_j is satisfied.

$$\max \sum_j w_j \cdot Z_j$$

$$\sum_{i \in P_j} y_i + \sum_{i \in N_j} (1 - y_i) \geq Z_j \quad \forall C_j \in \mathcal{C}$$

Rand. Rounding for max-SAT

- Solve the LP. let (y^*, z^*) be the opt sol

D ... set it to

$$\sum_{i \in P_j} y_i + \sum_{i \in N_j} (1 - y_i) \geq z_j \quad \forall j \in C$$

$$0 \leq z_j, y_i \leq 1$$

be the opt sol

— for each x_i set it to T with prob y_i

— let \hat{x} be the int sol obtained.

Then (Goemans/Williamson '94):

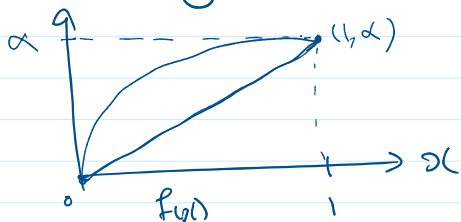
RR is a $(1 - \frac{1}{e})$ -approx for max SAT
 ≈ 0.632

Proof: We use the following two facts

Fact 1: Arithm/Geom mean inequality: non neg #s a_1, \dots, a_k :

$$\frac{a_1 + \dots + a_k}{k} \geq \sqrt[k]{a_1 a_2 \dots a_k} \quad \left(\frac{\sum a_i}{k} \right)^k \geq \prod a_i$$

Fact 2: if $f(x)$ is concave over $[0, 1]$ (i.e. $f''(x) \leq 0$) and $f(0) = 0$ and $f(1) = \alpha$ then the func. f is lower bounded by the line going thru $(0, 0)$ and $(1, \alpha)$



let w_j be the weight contributed by C_j to the total

Lemma: $E[W_j] \geq \underbrace{\left[1 - \left(1 - \frac{1}{k}\right)^k \right]}_{\geq 1 - \frac{1}{e}} w_j \cdot z_j^*$ for any clause C_j with $l_j = k$.

Proof:


$\geq 1 - \frac{1}{e}$
 C_j not Sat

$$\begin{aligned} \Pr[G_j \text{ is sat}] &= 1 - \prod_{i \in P_j} (1 - y_i^*) \prod_{i \in N_j} y_i^* \\ &\geq 1 - \left(\frac{\sum_{i \in P_j} (1 - y_i^*) + \sum_{i \in N_j} y_i^*}{k} \right)^k \quad \text{use A.G.M} \\ &\geq 1 - \left(1 - \frac{\sum_{i \in P_j} y_i^* + \sum_{i \in N_j} (1 - y_i^*)}{k} \right)^k \end{aligned}$$

by const. for $G_j \geq 1 - \left(1 - \frac{z_j^*}{k} \right)^k$

Consider func. $g(z) = 1 - \left(1 - \frac{z}{k} \right)^k$ $g(0) = 0$
 $g(1) = 1 - \left(1 - \frac{1}{k} \right)^k$
 and $g(\cdot)$ is concave in $[0, 1]$.

by fact 2 $\rightarrow g(z) \geq \left(1 - \left(1 - \frac{1}{k} \right)^k \right) \cdot z$

So $\Pr[G_j \text{ is sat}] \geq \left[1 - \left(1 - \frac{1}{k} \right)^k \right] z_j^*$ 

$$\begin{aligned} E[W] &= \sum_i w_j \cdot \Pr[G_j \text{ is sat}] \\ &\geq \left[1 - \left(1 - \frac{1}{k} \right)^k \right] \sum w_j \cdot z_j^* \geq \left(1 - \frac{1}{e} \right) \text{opt} \end{aligned}$$

Since $k \rightarrow \infty \left(1 - \frac{1}{k} \right)^k \rightarrow \frac{1}{e}$

Remark: This alg performs better than $1 - \frac{1}{e}$ if G_j 's are small

Remark 1 we can also Derand. this.

A $\frac{3}{4}$ -approx for max-SAT

Take the better of:

- Rand. Alg.

- Rand Rounding alg. (LP)

Suppose we set $a=0/1$ with

\leftarrow $\underline{a=0}$ prob $\frac{1}{2}$ and

\leftarrow if $a=1$

let Z^* be the opt value of LP and W_j be the cont of clause C_j in our alg.

Lemma: For each C_j : $E[W_j] \geq \frac{3}{4} w_j \cdot Z_j^*$

Proof: Suppose C_j has k literals, and define $\alpha_k = 1 - \frac{1}{2^k}$
and $\beta_k = 1 - (1 - \frac{1}{k})^k$

$$E[W_j | a=0] \geq (1 - \frac{1}{2^k}) w_j \geq \alpha_k w_j \cdot Z_j^* \quad (\text{since } Z_j^* \leq 1)$$

$$E[W_j | a=1] \geq [1 - (1 - \frac{1}{k})^k] w_j \cdot Z_j^* = \beta_k \cdot w_j \cdot Z_j^*$$

$$\begin{aligned} E[W_j] &= E[W_j | a=0] \cdot \underbrace{P_0[a=0]}_{\frac{1}{2}} + E[W_j | a=1] \cdot \underbrace{P_0[a=1]}_{\frac{1}{2}} \\ &\geq \frac{1}{2} [\alpha_k + \beta_k] w_j \cdot Z_j^* \end{aligned}$$

$$\alpha_1 + \beta_1 = \alpha_2 + \beta_2 = \frac{3}{2} \quad \text{for } k \geq 3 \quad \alpha_k + \beta_k \geq \frac{7}{8} + (1 - \frac{1}{e}) \geq \frac{3}{2}$$

$$\text{So } E[W] = \sum_j E[W_j] \geq \frac{3}{4} \sum w_j \cdot Z_j^* \geq \frac{3}{4} \text{opt}$$

We can also derandomize this alg:

Derand $\frac{3}{4}$ -approx for max-SAT

- Use the deterministic of rand alg to find Sol τ_1
- " " " " " " LP-rand " " " τ_2
- output the better of τ_1 & τ_2

Thm: This is a deterministic $\frac{3}{4}$ -approx for max-sat.

Tight example for LP (integrality gap):

$$(x_1, v, x_2) \wedge (\bar{x}_1, v, x_2) \wedge (x_1, v, \bar{x}_2) \wedge (\bar{x}_1, v, \bar{x}_2) \quad \text{all } v_j = 1$$

opt = 3 but if we set $f_i = \frac{1}{2}$ and $z_j = 1$ we have a feasible LP with value 4.

→ integrality gap = $\frac{4}{3}$

Best known ratio for Max-SAT ≈ 0.7846

based on a plausible conj $\xrightarrow{\text{might be}}$ 0.8331 -approx