

Problem Set 4 Solutions

学号: MP1933008

姓名: 黄睿

1. Problem1

From the definition of Algorithm \mathcal{A} , we have that:

$$\mathbb{E} [\hat{Z}] = \frac{3}{4}Z$$

Denote $Y_i = \begin{cases} 1, & \text{if } \hat{Z}_i \text{ success;} \\ 0, & \text{otherwise.} \end{cases}$, As \mathbf{X} is the median of s independent

trials, $\mathbf{X} \text{ success} \Leftrightarrow \sum_{i=1}^s Y_i \geq \frac{1}{2}s$.

As each trail is performed independently, $\mathbb{E} [Y_i] = 3/4, i = 1, 2, \dots, s$.

Applying linearity of expectations: $\mathbb{E} \left[\sum_{i=1}^s Y_i \right] = \sum_{i=1}^s \mathbb{E} [Y_i] = \frac{3}{4}s$.

Consider $\Pr [X \text{ fail}] < \delta \Leftrightarrow \Pr [X \text{ success}] \geq 1 - \delta$:

$$\begin{aligned}
\Pr [\mathbf{X} \text{ fail}] &= \Pr \left[\sum_{i=1}^s Y_i < \frac{1}{2}s \right] \\
&= \Pr \left[\sum_{i=1}^s Y_i - \frac{3}{4}s < -\frac{1}{4}s \right] \\
&= \Pr \left[\sum_{i=1}^s Y_i - \mathbb{E} \left[\sum_{i=1}^s Y_i \right] < -\frac{1}{4}s \right] \\
&= \Pr \left[\sum_{i=1}^s Y_i - \mathbb{E} \left[\sum_{i=1}^s Y_i \right] < -\frac{1}{3} \mathbb{E} \left[\sum_{i=1}^s Y_i \right] \right] \\
&< \exp \left(-\frac{E \left[\sum_{i=1}^s Y_i \right] \cdot \left(\frac{1}{3} \right)^2}{2} \right) \quad (\text{Chenoff Bound}) \\
&= \exp \left(-\frac{1}{24}s \right) \\
&< \delta
\end{aligned}$$

Above inequalities hold for $s = \lceil -24 \ln \delta \rceil$.

2. Problem2

(1) Q1

Define $|V| = n$ mutually independent random variables :

$X_1, X_2, \dots, X_n \in \{0,1\}$. Where $X_i = \begin{cases} 1, & \text{if vertex } i \in T; \\ 0, & \text{otherwise.} \end{cases}$.Accordingly, we

can define bad events $A_i, (i = 1,2,\dots, |E| = m)$ as: for edge $e_i \in E$, both vertices in e_i are selected thus e_i will be in the transversal. Then we have:

$$\begin{aligned}
\Pr[A_i] &= \Pr[(u, v) \in e_i \text{ is selected}] \\
&= \Pr[u \text{ is selected} \wedge v \text{ is selected}] \\
&= \Pr[X_u = 1 \wedge X_v = 1] \\
&\leq \left(\frac{1}{2e\Delta}\right)^2
\end{aligned}$$

Since each vertex v has degree at most Δ , the dependence degree $d \leq 2\Delta$. We have:

$$\forall i, \Pr[A_i] \cdot e(d+1) \leq \left(\frac{1}{2e\Delta}\right)^2 \cdot e(2\Delta+1) \leq 1.$$

The above inequality holds for all $\Delta > 0$. According to **Lovász Local Lemma** (Lovász 1977), it holds that:

$$\Pr\left[\bigwedge_{i=1}^m \bar{A}_i\right] > 0.$$

Thus there must be an independent transversal of $\{S_1, S_2, \dots, S_r\}$.

(2) Q2

Algorithm can be designed according to **Moser-Tardos** as below:

Sample all X_1, X_2, \dots, X_n ;
While \exists an occurring bad event A_i :
 resample all $X_j \in \text{vbl}(A_i)$;

From the analysis in class, the algorithm will find an independent transversal in expected polynomial time.

3. Problem3

(1) Q1

For simplicity, we give the LP-relaxed dual program as below,:

$$\begin{aligned}
&\text{maximize } \sum_{C \in \mathcal{C}} y_C \\
&\text{subject to } \sum_{C: v \in C} y_C \leq 1, \forall v \in V, \\
&\quad y_C \geq 0, \forall C \in \mathcal{C}.
\end{aligned}$$

(2) Q2

According to the primal-dual schema, we can design the algorithm as below:

Initially $\mathbf{x} = 0, \mathbf{y} = 0$;
while $\mathcal{C} \neq \emptyset$:
 pick a cycle $C \in \mathcal{C}$ and raise y_C to 1;
 set $x_v = 1$ for those $v \in C$ and delete all $C : v \in C$ from \mathcal{C} ;

Each deleted cycle C is incident to a v that $x_v = 1$ and all cycles are eventually deleted, thus we get a feasible solution \mathbf{x} satisfying:

$$\forall C \in \mathcal{C}, \sum_{v \in C} x_v \geq 1$$

For simplicity, we denote the number of vertices in $G(V, E)$ as n , then we obtain:

$$\forall C : \text{either } \sum_{v \in C} x_v \leq n \text{ or } y_C = 0$$

$$\forall v : \text{either } \sum_{C: v \in C} y_C = 1 \text{ or } x_v = 0$$

thus, it holds:

$$SOL = \sum_{v \in V} x_v \leq n \cdot OPT$$

The primal-dual schema gets an approximation ratio of n .

4. Problem4

(1) Q1

From the definition of Ising model, we can directly calculate the marginal distribution:

$$\begin{aligned}
\mu_v \left(+1 \mid \sigma(V \setminus \{v\}) \right) &= \frac{\mu_v \left(v = +1, \sigma(V \setminus \{v\}) \right)}{\mu_v \left(\sigma(V \setminus \{v\}) \right)} \\
&= \frac{\frac{1}{Z} \exp \left(- \sum_{\{u,v\} \in E} \beta \sigma(u) + C \right)}{\frac{1}{Z} \exp \left(- \sum_{\{u,v\} \in E} \beta \sigma(u) \sigma(v) + C \right)}, \quad C = \sum_{\substack{\{w,u\} \in E \\ w \neq v \\ u \neq v}} \beta \sigma(w) \sigma(u) \\
&= \frac{\exp \left(- \sum_{\{u,v\} \in E} \beta \sigma(u) \right)}{\exp \left(- \sum_{\{u,v\} \in E} \beta \sigma(u) \sigma(v) \right)} \\
&= \frac{\exp \left(- \sum_{\{u,v\} \in E} \beta \sigma(u) \right)}{\exp \left(- \sum_{\{u,v\} \in E} \beta \sigma(u) \right) + \exp \left(\sum_{\{u,v\} \in E} \beta \sigma(u) \right)} \\
&= \frac{1}{1 + \exp^2 \left(\sum_{\{u,v\} \in E} \beta \sigma(u) \right)}
\end{aligned}$$

$$\begin{aligned}
\mu_v \left(-1 \mid \sigma(V \setminus \{v\}) \right) &= 1 - \mu_v \left(+1 \mid \sigma(V \setminus \{v\}) \right) \\
&= \frac{\exp^2 \left(\sum_{\{u,v\} \in E} \beta \sigma(u) \right)}{1 + \exp^2 \left(\sum_{\{u,v\} \in E} \beta \sigma(u) \right)}
\end{aligned}$$

(2) Q2

To prove the irreducibility, $\forall \sigma, \tau \in \{-1, +1\}^V$, we need to prove the transition probability $\Pr^t[\tau \mid \sigma] > 0$. Denote

$\sigma = (Y_1, Y_2, \dots, Y_V)$, $\delta = (X_1, X_2, \dots, X_V)$, where $X_i, Y_i \in \{0, 1\}$.

$$\begin{aligned}
\Pr[\tau | \sigma] &= \Pr[X_1 \rightarrow Y_1, \dots, X_V \rightarrow Y_V] \\
&= \Pr[Y_1 | X_2, X_3, \dots, X_V] \\
&\quad \cdot \Pr[Y_2 | Y_1, Y_3, \dots, X_V] \\
&\quad \cdot \Pr[Y_3 | Y_1, Y_2, \dots, X_V] \\
&\quad \vdots \\
&\quad \cdot \Pr[Y_V | Y_1, Y_2, \dots, Y_{V-1}]
\end{aligned}$$

From Q1, we know that all terms of the above equation are positive. Thus, $\Pr[\tau | \delta] > 0$, for another word, the Glauber dynamics for Ising model is irreducible.

The aperiodicity holds when $\forall \sigma \in \{-1, +1\}^V$, $\Pr[\sigma | \sigma] > 0$. It's just a special case of irreducibility.

To prove the reversibility of Glauber dynamics, we should prove the detailed balance equation: $\mu(\sigma) \Pr[\tau | \sigma] = \mu(\tau) \Pr[\sigma | \tau]$.

According to the definition of Glauber dynamics, one step transition can change at most one vertex v . Without loss of generality, let

$\sigma(v) = -1, \tau(v) = +1$, other cases are similar.

$$\mu(\sigma) \Pr[\tau | \sigma] = \mu(\sigma) \mu_v(+1 | \delta(V \setminus \{v\}))$$

$$\begin{aligned}
&= \frac{1}{Z} \exp \left(\sum_{\{u,v\} \in E} \beta \sigma(u) + C \right) \cdot \frac{1}{1 + \exp^2 \left(\sum_{\{u,v\} \in E} \beta \sigma(u) \right)} \\
&= \frac{\exp(C)}{Z} \cdot \frac{\exp \left(\sum_{\{u,v\} \in E} \beta \sigma(u) \right)}{1 + \exp^2 \left(\sum_{\{u,v\} \in E} \beta \sigma(u) \right)}
\end{aligned}$$

$$\tau(\sigma) \Pr[\sigma | \tau] = \mu(\tau) \mu_v(-1 | \delta(V \setminus \{v\}))$$

$$= \frac{1}{Z} \exp \left(- \sum_{\{u,v\} \in E} \beta \sigma(u) + C \right) \cdot \frac{\exp^2 \left(\sum_{\{u,v\} \in E} \beta \sigma(u) \right)}{1 + \exp^2 \left(\sum_{\{u,v\} \in E} \beta \sigma(u) \right)}$$

$$= \frac{\exp(C)}{Z} \cdot \frac{\exp \left(\sum_{\{u,v\} \in E} \beta \sigma(u) \right)}{1 + \exp^2 \left(\sum_{\{u,v\} \in E} \beta \sigma(u) \right)}$$

It's obvious that $\mu(\sigma) \Pr[\tau | \sigma] = \mu(\tau) \Pr[\sigma | \tau]$. This proves the reversibility of Glauber dynamics for Ising model.