Problem Set 4 Solutions

学号: MP1933008

姓名: 黄睿

1. Problem1

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

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

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 \mathbf{s} independent

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

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

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

Consider $\Pr\left[X \; fail\right] < \delta \Leftrightarrow \Pr\left[X \; success\right] \ge 1 - \delta$:

$$\Pr\left[\mathbf{X} \; fail\right] = \Pr\left[\sum_{i=1}^{s} Y_{i} < \frac{1}{2}s\right]$$

$$= \Pr\left[\sum_{i=1}^{s} Y_{i} - \frac{3}{4}s < -\frac{1}{4}s\right]$$

$$= \Pr\left[\sum_{i=1}^{s} Y_{i} - \mathbb{E}\left[\sum_{i=1}^{s} Y_{i}\right] < -\frac{1}{4}s\right]$$

$$= \Pr\left[\sum_{i=1}^{s} Y_{i} - \mathbb{E}\left[\sum_{i=1}^{s} Y_{i}\right] < -\frac{1}{3}\mathbb{E}\left[\sum_{i=1}^{s} Y_{i}\right]\right]$$

$$< \exp\left[-\frac{E\left[\sum_{i=1}^{s} Y_{i}\right] \cdot \left(\frac{1}{3}\right)^{2}}{2}\right] \text{ (Chenoff Bound)}$$

$$= \exp\left(-\frac{1}{24}s\right)$$

$$< \delta$$

Above inequalities hold for $s = [-24 \ln \delta]$.

2. Problem2

(1) Q1

Define |V| = n mutually independent random variables :

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

can define bad events A_i , $(i=1,2,\cdots,|E|=m)$ 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:

$$\Pr[A_i] = \Pr[(u, v) \in e_i \text{ is selected}]$$

$$= \Pr[u \text{ is selected} \land v \text{ is selected}]$$

$$= \Pr[X_u = 1 \land X_v = 1]$$

$$\leq \left(\frac{1}{2e\Delta}\right)^2$$

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

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

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

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

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
$$X_1, X_2, \cdots, X_n$$
; While \exists an occurring bad event A_i : resample all $X_j \in vbl(A_i)$;

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

3. Problem3

(1) Q1

For simplicity, I'll give the LP-relaxed dual program:

$$\begin{split} \max \min & z e \sum_{C \in \mathscr{C}} y_C \\ \text{subject to } & \sum_{C: v \in C} y_C \leq 1, \forall v \in V, \\ & y_C \geq 0, \forall C \in \mathscr{C} \,. \end{split}$$

(2) Q2

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

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

$$\forall C \in \mathscr{C}, \ \sum_{v \in C} x_v \ge 1$$

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

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

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

thus, it holds:

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

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:

$$\mu_{v}\left(+1 \mid \sigma\left(V \setminus \{v\}\right)\right) = \frac{\mu_{v}\left(v = +1, \sigma\left(V \setminus \{v\}\right)\right)}{\mu_{v}\left(\sigma\left(V \setminus \{v\}\right)\right)}$$

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

$$= \frac{\exp\left(-\sum_{\{u,v\} \in E} \beta \sigma(u)\right)}{\exp\left(-\sum_{\{u,v\} \in E} \beta \sigma(u)\right)}$$

$$= \frac{\exp\left(-\sum_{\{u,v\} \in E} \beta \sigma(u)\right)}{\exp\left(-\sum_{\{u,v\} \in E} \beta \sigma(u)\right) + \exp\left(\sum_{\{u,v\} \in E} \beta \sigma(u)\right)}$$

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

$$\begin{split} \mu_{v}\left(-1\,|\,\sigma\left(V\backslash\{v\}\right)\right) &= 1 - \mu_{v}\left(+1\,|\,\sigma\left(V\backslash\{v\}\right)\right) \\ &= \frac{\exp^{2}\left(\sum_{\{u,v\}\in E}\beta\sigma(u)\right)}{1 + \exp^{2}\left(\sum_{\{u,v\}\in E}\beta\sigma(u)\right)} \end{split}$$

(2) Q2

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

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

$$Pr^{t}[\tau \mid \sigma] = Pr \left[X_{1} \rightarrow Y_{1}, \dots, X_{V} \rightarrow Y_{V} \right]$$

$$= Pr \left[Y_{1} \mid X_{2}, X_{3}, \dots, X_{V} \right]$$

$$\cdot Pr \left[Y_{2} \mid Y_{1}, Y_{3}, \dots, X_{V} \right]$$

$$\cdot Pr \left[Y_{3} \mid Y_{1}, Y_{2}, \dots, X_{V} \right]$$

$$\vdots$$

$$\cdot Pr \left[Y_{V} \mid Y_{1}, Y_{2}, \dots, Y_{V-1} \right]$$

From Q1, we know that all terms of the above equation are positive. Thus, $\Pr^t[\tau \mid \delta] > 0$, 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: $\mu(\sigma) \Pr[\tau \mid \sigma] = \mu(\tau) \Pr[\sigma \mid \tau]$.

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.

$$\mu(\sigma) \Pr[\tau \mid \sigma] = \mu(\sigma) \mu_{\nu}(+1 \mid \sigma(V \setminus \{v\}))$$

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

$$\mu(\tau) \Pr[\sigma \mid \tau] = \mu(\tau) \mu_{\nu}(-1 \mid \sigma(V \setminus \{v\}))$$

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

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

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