What can be sampled locally?

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Preliminaries

- The LOCAL model
- Markov random fields (MRFs) and local CSP
- Local sampling problem
- Mixing rate

The LOCAL model

- G(V, E) represents the processor network
- Δ is the maximum degree
- n = |V|

Markov random field

Given a graph G(V, E) and a finite domain [q] = {1, 2, ..., q},
 The probability measure of each configuration σ is proportional to the weight:

$$w(\sigma) := \prod_{e=uv \in E} A_e(\sigma_u, \sigma_v) \prod_{v \in V} b_v(\sigma_v),$$

Markov random field

Given a graph G(V, E) and a finite domain [q] = {1, 2, ..., q},
 The probability measure of each configuration σ is proportional to the weight:

$$w(\sigma) = \prod_{c=(f_c, S_c) \in \mathcal{C}} f_c(\sigma|_{S_c}),$$

Local sampling

Measured by total variation distance:

$$d_{\text{TV}}(\mu, \nu) = \sum_{\sigma \in \Omega} \frac{1}{2} |\mu(\sigma) - \nu(\sigma)| = \max_{A \subseteq \Omega} |\mu(A) - \nu(A)|.$$

Mixing rate

- Ergodicity: Irreducible + Aperiodic
- Mixing rate is defined as

$$\tau(\epsilon) = \max_{\sigma \in \Omega} \min \Big\{ t : d_{\text{TV}} \left(\pi_{\sigma}^{(t)}, \pi \right) \leq \epsilon \Big\},$$

Luby's algorithm for maximal independent set (MIS)

Initialize I to an empty set.

While V is not empty:

- Sample an independent set S in V.
- 2. Add the set S to I.
- 3. Remove from V the set S and all the neighbours of nodes in S.

Return I.

LubyGlauber algorithm

6 return X_v;

$$\forall c \in [q], \quad \mu_v(c \mid X_{\Gamma(v)}) = \frac{b_v(c) \prod_{u \in \Gamma(v)} A_{uv}(c, X_u)}{\sum_{a \in [q]} b_v(a) \prod_{u \in \Gamma(v)} A_{uv}(a, X_u)}.$$

LubyGlauber algorithm

 Assumption: Marginal distribution is well defined, chain is irreducible among all feasible configurations

Proposition 3.1. The Markov chain LubyGlauber is reversible and has stationary distribution μ . Furthermore, under the above assumption, $d_{TV}(\mu_{LG}, \mu)$ converges to 0 as $T \to \infty$.

Theorem 3.2. Under the same assumption as Proposition 3.1, if the total influence $\alpha < 1$, then the mixing rate of the LubyGlauber chain is $\tau(\epsilon) = O\left(\frac{\Delta}{1-\alpha}\log\left(\frac{n}{\epsilon}\right)\right)$.

Theorem 1.1. If $q \ge \alpha \Delta$ for an arbitrary constant $\alpha > 2$, there is an algorithm which samples a uniform proper q-coloring within total variation distance $\epsilon > 0$ in $O\left(\Delta \log\left(\frac{n}{\epsilon}\right)\right)$ rounds of communications on any graph G(V, E) with n = |V| vertices and maximum degree $\Delta = \Delta(n)$.

LocalMetropolis algorithm

10 each $v \in V$ returns X_v ;

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Algorithm 2: Pseudocode for the LocalMetropolis algorithm
  Input: Each vertex v \in V receives \{A_{uv}\}_{u \in \Gamma(v)} and b_v as input.
1 each v \in V initializes X_v to an arbitrary value in [q];
2 for t = 1 through T do
      for each v \in V do
        propose a random \sigma_v \in [q] with probability b_v(\sigma_v) / \sum_{c \in [q]} b_v(c);
4
      for each e = (u, v) \in E do
           pass the check independently with probability
            A_e(\sigma_u, \sigma_v)A_e(X_u, \sigma_v)A_e(\sigma_u, X_v)/\left(\max_{i,j \in [q]} A_e(i,j)\right)^3;
      for each v \in V do
          if all edges e incident with v pass the checks then X_v \leftarrow \sigma_v;
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LocalMetropolis algorithm

Theorem 4.1. The Markov chain LocalMetropolis is reversible and has stationary distribution μ . Furthermore, under above assumptions, $d_{TV}(\mu_{LM}, \mu)$ converges to 0 as $T \to \infty$.

Theorem 4.2. If $q \ge \alpha \Delta$ for a constant $\alpha > 2 + \sqrt{2}$, the mixing rate of the LocalMetropolis chain for proper q-coloring on graphs with maximum degree at most $\Delta = \Delta(n) \ge 9$ is $\tau(\epsilon) = O(\log(\frac{n}{\epsilon}))$, where the constant factor in $O(\cdot)$ depends only on α but not on the maximum degree Δ .

Lower bounds

Theorem 1.3. For $\Delta \geq 6$, there exist infinitely many graphs G(V, E) with maximum degree Δ and diameter $\operatorname{diam}(G) = \widetilde{\Omega}(\sqrt{|V|})$ such that any algorithm that samples uniform independent set in G within sufficiently small constant total variation distance ϵ requires $\Omega(\operatorname{diam}(G))$ rounds of communications, even assuming the vertices $v \in V$ to be aware of G.

Thanks!