

# The Notion of Approximate Tensorization

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## 1 Background

Recently, a novel notation called **Approximate Tensorization** was used to build the rapid mixing of Glauber dynamics for general spin systems in the uniqueness region[CLV20]. They use this notation to establish a (constant factor) connection between the modified log-sobolev constant of the block dynamics and the single site dynamics on general spin systems.

For spin system, it turns out that this notation is equivalent to the well-known poincaré inequality and log-sobolev inequality, while it contains more intuition for spin systems such that one could use the inner properties of spin systems more easily.

For more information of this notation, one could refer to [CMT15, CP20].

## 2 Approximate Tensorization of Variance

Lets start this notation from the easier part, namely, variance and the corresponding constant i.e. spectra gap  $1 - \lambda_2$ .

**Fact 2.1** (Variance). *For  $\forall f : \Omega \rightarrow \mathbb{R}$ , we have*

$$\text{Var}_\mu f = \frac{1}{2} \sum_{\sigma, \tau \in \Omega} \mu(\sigma) \mu(\tau) (f(\sigma) - f(\tau))^2$$

**Fact 2.2** (Dirichlet Form). *For  $\forall f : \Omega \rightarrow \mathbb{R}$ , we have*

$$\langle f, (I - P)f \rangle_\mu = \frac{1}{2} \sum_{\sigma, \tau \in \Omega} \mu(\sigma) P(\sigma, \tau) (f(\sigma) - f(\tau))^2$$

### 2.1 Single Site Dynamics

The equivalence between approximate tensorization and the poincaré inequality relies on a careful manipulation on Dirichlet form, which is stated as follows.

$$\begin{aligned}
\langle f, (I - P)f \rangle_\mu &= \frac{1}{2} \sum_{\sigma, \tau \in \Omega} \mu(\sigma) P(\sigma, \tau) (f(\sigma) - f(\tau))^2 \\
&= \frac{1}{2} \sum_{\sigma \in \Omega} \mu(\sigma) \left( \sum_{u \in V} \sum_{\substack{\tau_u \in \Omega_u^{\sigma_{V \setminus u}} \\ \tau_{V \setminus u} = \sigma_{V \setminus u}}} \frac{1}{n} \mu_u^{\sigma_{V \setminus u}}(\tau_u) \right) (f(\sigma) - f(\tau))^2 \\
&= \frac{1}{n} \sum_{u \in V} \sum_{\gamma \in \Omega_{V \setminus u}} \mu_{V \setminus u}(\gamma) \cdot \frac{1}{2} \sum_{x \in \Omega_u^\gamma} \mu_u^\gamma(x) \sum_{y \in \Omega_u^\gamma} \mu_u^\gamma(y) (f_\gamma(x) - f_\gamma(y))^2 \\
&= \frac{1}{n} \sum_{u \in V} \sum_{\gamma \in \Omega_{V \setminus u}} \mu_{V \setminus u}(\gamma) \cdot \text{Var}_{\mu_u^\gamma}(f_\gamma) \\
&= \frac{1}{n} \sum_{u \in V} \mu[\text{Var}_u(f)]
\end{aligned}$$

**Fact 2.3** (Equivalence for Single Site Dynamics). *Recall that the poincaré inequality is*

$$\forall f, (1 - \lambda_2) \cdot \text{Var}_\mu(f) \leq \langle f, (I - P)f \rangle_\mu$$

*And we could restate it as follows*

$$\forall f, (1 - \lambda_2) \cdot \text{Var}_\mu(f) \leq \frac{1}{n} \sum_u \mu[\text{Var}_u(f)]$$

## 2.2 Block Dynamics

Let  $P_{n, n-\ell}^\vee$  denote the block dynamics where we choose  $\ell$  vertices uniformly at random, and then sample a configuration on this  $\ell$  vertices according to the conditional distribution.

Note that the above approach for single site dynamics could also be applied to block dynamics.

$$\begin{aligned}
\langle f, (I - P_{n, n-\ell}^\vee)f \rangle_\mu &= \frac{1}{2} \sum_{\sigma, \tau \in \Omega} \mu(\sigma) P_{n, n-\ell}^\vee(\sigma, \tau) (f(\sigma) - f(\tau))^2 \\
&= \frac{1}{2} \sum_{\sigma \in \Omega} \mu(\sigma) \left( \sum_{S \in \binom{V}{\ell}} \sum_{\substack{\tau_S \in \Omega_S^{\sigma_{V \setminus S}} \\ \tau_{V \setminus S} = \sigma_{V \setminus S}}} \frac{1}{\binom{n}{\ell}} \mu_S^{\sigma_{V \setminus S}}(\tau_S) \right) (f(\sigma) - f(\tau))^2 \\
&= \frac{1}{\binom{n}{\ell}} \sum_{S \in \binom{V}{\ell}} \sum_{\gamma \in \Omega_{V \setminus S}} \mu_{V \setminus S}(\gamma) \cdot \frac{1}{2} \sum_{\alpha \in \Omega_S^\gamma} \mu_S^\gamma(\alpha) \sum_{\beta \in \Omega_S^\gamma} \mu_S^\gamma(\beta) (f_\gamma(\alpha) - f_\gamma(\beta))^2 \\
&= \frac{1}{\binom{n}{\ell}} \sum_{S \in \binom{V}{\ell}} \sum_{\gamma \in \Omega_{V \setminus S}} \mu_{V \setminus S}(\gamma) \cdot \text{Var}_{\mu_S^\gamma}(f_\gamma) \\
&= \frac{1}{\binom{n}{\ell}} \sum_{S \in \binom{V}{\ell}} \mu[\text{Var}_S(f)]
\end{aligned}$$

**Fact 2.4** (Equivalence for Block Dynamics). *Recall that the poincaré inequality for the block dynamics is*

$$\forall f, (1 - \lambda_2) \text{Var}_\mu f \leq \langle f, (I - P_{n, n-\ell}^\vee)f \rangle_\mu$$

*And we could restate it as follows*

$$\forall f, (1 - \lambda_2) \text{Var}_\mu f \leq \frac{1}{\binom{n}{\ell}} \sum_{S \in \binom{V}{\ell}} \mu[\text{Var}_S(f)]$$

In the block case, the **Approximate Tensorization** is some times called **Uniform Block Factorization** (see [CLV20]) and it is a special case of **Block Factorization** given in [CP20].

### 2.3 Connection with Local-to-Global Argument

**Definition 2.1.** For any function  $f^{(s)} : X_s \rightarrow \mathbb{R}$ , and  $r < s$ , we could define another function as  $f^{(r)} \triangleq P_r^\dagger P_{r+1}^\dagger \dots P_{s-1}^\dagger f^{(s)}$  Which could also be denoted as

$$f^{(r)} = [\pi_r \leftrightarrow \pi_{r+1}] \dots [\pi_{s-1} \leftrightarrow \pi_s] f^{(s)}$$

**Fact 2.5.**  $\pi_s f^{(s)} = \pi_r f^{(r)}$

**Definition 2.2.** Let  $f^{(s)}$  be a function defined on  $X_s$ . Let  $\gamma \in X_r$ ,  $\alpha \in X_{s-r}^\gamma$ , then we denote  $f^{(s)}(\gamma \cup \alpha)$  as  $f_\gamma^{(s-r)}(\alpha)$

**Fact 2.6.**  $\mathbb{E}_{\pi_\ell^\gamma}[f_\gamma^{(\ell)}] = f^{(\ell)}(\gamma)$

**Fact 2.7.**

$$\frac{1}{\binom{n}{\ell}} \sum_{s \in \binom{V}{\ell}} \mu[\text{Var}_s(f)] = \text{Var}_{\pi_n}(f^{(n)}) - \text{Var}_{\pi_{n-\ell}}(f^{(n-\ell)})$$

## References

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