## APPROXIMATE CONSTRAINT SATISFACTION REQUIRES LARGE LP RELAXATIONS

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**Definition** (basic definitions).  $f: \{\pm 1\}^n \to \mathbb{R}$ , we write  $\mathbb{E}f = 2^{-n} \sum_x \in \{\pm 1\}^n f(x)$ .  $\langle f, g \rangle = \mathbb{E}[fg]$ .

Any such f can be written uniquely in the Fourier basis as  $f = \sum_{\alpha \subseteq [n]} \langle f, \chi_{\alpha} \rangle \chi_{\alpha}$ , where  $\chi_{\alpha} = \prod_{i \in \alpha} x_i$ .

**Definition** (d-junta).  $f: \{\pm 1\}^n \to \mathbb{R}$  is called d-junta for  $d \in [n]$  if f depends only a subset  $S \subseteq [n]$  of coordinates with  $|S| \le d$ . In other words, f can be written as  $f = \sum_{\alpha \subset S} \langle f, \chi_{\alpha} \rangle \chi_{\alpha}$ .

**Definition** (density). We say that f is a density if it is non-negative and satisfies  $\mathbb{E}f = 1$ . For such an f, we let  $\mu_f$  denote the corresponding probability measure on  $\{\pm 1\}^n$ . Observer that for any  $g: \{\pm 1\}^n \to \mathbb{R}$ , we have  $\mathbb{E}_{x \sim \mu_f}[g(x)] = \langle f, g \rangle$ .

**Definition** ((c, s)-approx.). We say that a linear programming relaxation  $\mathcal{L}$  for MAX- $\Pi_n$  achieves a (c, s)-approximation if  $\mathcal{L}(\mathcal{I}) \leq c$  for all instances  $\mathcal{I} \in \text{MAX-}\Pi_n$  with  $\text{opt}(\mathcal{I}) \leq s$ .

**Theorem** (2.2). There exists an LP relaxation of size R that achieves a (c,s)-approximation for MAX- $\Pi_n$  if and only if there exist non-negative functions  $q_1, \ldots, q_R : \{\pm 1\}^n \to \mathbb{R}_{\geq 0}$  such that for every instance  $\mathcal{I} \in \text{MAX-}\Pi_n$  with  $opt(\mathcal{I}) \leq s$ , the function  $c - \mathcal{I}$  is a nonnegative combination of  $q_1, \ldots, q_R$ , i.e.

$$c - \mathcal{I} \in \{ \sum_{i} \lambda_i q_i \mid \lambda_i \ge 0 \}.$$

**Lemma** (2.3). In order to show that (c,s) – MAX- $\Pi_n$  requires LP relaxations of size greater than R, it is sufficient to prove the following: For every collection of densities  $q_1, \ldots, q_R : \{\pm 1\}^n \to \mathbb{R}_{\geq 0}$ , there is  $\epsilon > 0$ , a function  $H : \{\pm 1\}^n \to \mathbb{R}$  and a MAX- $\Pi_n$  instance  $\mathcal{I}$  such that

1. 
$$\langle H, c - \mathcal{I} \rangle < -\epsilon$$

2. 
$$\langle H, q_i \rangle \geq -\epsilon$$

**Lemma** (2.4). Suppose that  $f: \{\pm 1\}^n \to \mathbb{R}$  depends only on a subset of at most d coordinates  $S \subseteq [n]$ , then

$$\langle H, f \rangle = \mathbb{E}_{x \sim \mu_S}[f(x)]$$

for some probability measure  $\mu_S$  on  $\{\pm 1\}^n$ .

**Theorem** (Main, 3.1). Fix a positive number  $d \in \mathbb{N}$ . Suppose that the d-round Sherali-Adams relaxation cannot achieve a (c,s)-approximation for MAX- $\Pi_n$  for every n. Then no sequence of LP relaxations of size at most  $n^{d/2}$  can achieve a (c,s)-approximation for MAX- $\Pi_n$  for every n.

**Theorem** (3.2). Consider a function  $f: \mathbb{N} \to \mathbb{N}$ . Suppose that the f(n)-round Sherali-Adams relaxation cannot achieve a (c,s)-approximation for MAX- $\Pi_n$ . Then for all sufficiently large n, no LP relaxation of size at most  $n^{f(n^2)}$  can achieve a (c,s)-approximation for MAX- $\Pi_N$  where  $N \leq n^{10f(n)}$ .

**Lemma** (3.3). For all  $1 \le d$ ,  $t \le n$  and  $\beta > 0$ , the following holds. If  $\mu$  has entropy  $\ge n - t$ , there exists a set  $J \subseteq [n]$  of at most  $\frac{td}{\beta}$  coordinates such that for all subsets  $A \nsubseteq J$  with  $|A| \le d$ , we have

$$\max_{v \in A} H(X_v \mid X_{A \setminus v}) \ge 1 - \beta$$

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**Definition** (KL-divergence).

$$D(\mu||\nu) = \mathbb{E}_{\mu}[\log_2 \frac{\mu(x)}{\nu(x)}].$$

**Lemma** (3.5). Let  $\mu$  be a distribution as in the statement of Lemma 3.3, and let  $J \subseteq [n]$  be the corresponding set of coordinates. If  $A \subseteq [n]$  satisfies  $|A| \leq d$  and  $A \not\subseteq J$ , then

$$|\mathbb{E}_{\mu}[\chi_A(x)]| \le \sqrt{(\ln 4)\beta}.$$

**Lemma** (3.7, random restrictions). For any  $d \in \mathbb{N}$ , the following holds. Let Q be a collection of densities  $q: \{\pm 1\}^n \to \mathbb{R}_{\geq 0}$  such that the corresponding measures  $\mu_q$  have entropy at least n-t. If  $|Q| \leq n^{d/2}$ , then for all integers m with  $3 \leq m \leq n/4$ , there exists a set  $S \subseteq [n]$  such that: 1. |S| = m

2. For each  $q \in Q$ , there is a set of at most d coordinates  $J(q) \subseteq S$  such that under the distribution  $\mu_q$ , all d-wise correlations in S - J(q) are small. Quantitatively, we have

$$|\hat{q}(\alpha)| \leq \left(\frac{32mtd}{\sqrt{n}}\right)^{1/2} \quad \forall \ \alpha \subseteq S, \alpha \not\subseteq J(q), |\alpha| \leq d$$