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Lecture 6 : SoS and Robust Regression

$$(x_i, y_i)_{i=1}^N : \text{Corrupted dataset} \quad x_i \in \mathbb{R}^d$$

$$(x_i^*, y_i^*)_{i=1}^N : \text{uncorrupted dataset} \quad y_i \in \mathbb{R}$$

$$a_i^* = \begin{cases} 1 & \text{if } i \text{ is clean} \\ 0 & \text{o.w.} \end{cases}$$

so when $a_i^* = 1$, $(x_i^*, y_i^*) = (x_i, y_i)$

$$y_i^* = \langle w^*, x_i \rangle + \xi_i \sim N(0, \sigma^2)$$

SoS Program

Variables: w (d -dimensional)
 a_1, \dots, a_N (1 -dimensional)

Constraints:

- 1) $a_i^2 = a_i$ (Boolean indicators)
- 2) $\frac{1}{N} \sum a_i \geq 1 - \gamma$ (γ fraction corruptions)

$$\underline{\text{Obj}}: \min \mathbb{E} \left[\frac{1}{N} \sum_{i=1}^N a_i (y_i - \langle w, x_i \rangle)^2 \right]$$

Theorem [Klivans-Kothari-Meka '18]:

Suppose x_i 's are drawn from a mean-0 distribution over \mathbb{R}^d which is certifiably 4 -hypercontractive (see below for defn.). Then for any degree- k pseudodistribution \mathbb{E} over solutions to the program minimizing the objective above,

$$\text{Clean MSE}(\mathbb{E}[w]) \leq (1 + O(\gamma^{1/2})) (\text{DPT} + O(\gamma^{1/2} \sigma))$$

Claim: Suffices to give a deg- k SOS proof

that $\frac{1}{N} \sum_{i=1}^N (y_i^* - \langle w, x_i^* \rangle)^2 \leq \text{RHS}$

Pf: Note $\frac{1}{N} \sum_{i=1}^N a_i^* (y_i^* - \langle w, x_i^* \rangle)^2 \leq \frac{1}{N} \sum_{i=1}^N (y_i^* - \langle w, x_i^* \rangle)^2$

because $x_i^* = x_i^*$ and $y_i^* = y_i^*$ when $a_i^* = 1$. Claim

then follows by following fact about pseudodistribution.

Fact: For any polynomials p, q & $\deg \leq k/2$ in \vec{x}

and any deg- k pseudodistribution $\hat{\mathbb{E}}$ over x ,

$$\hat{\mathbb{E}}[p(x)q(x)] \leq \hat{\mathbb{E}}[p(x)^2]^{1/2} \hat{\mathbb{E}}[q(x)^2]^{1/2}$$

Pf: Suppose $\hat{\mathbb{E}}[p(x)^2], \hat{\mathbb{E}}[q(x)^2] > 0$. By linearity, can assume wlog that $\hat{\mathbb{E}}[p(x)^2] = \hat{\mathbb{E}}[q(x)^2] = 1$. Note $p \cdot q \leq \frac{1}{2}(p^2 + q^2)$ in deg- k SOS, so $\hat{\mathbb{E}}[pq] \leq 1$ as desired.

Suppose next that $\hat{\mathbb{E}}[p(x)^2] = 0$. Then $\forall \alpha > 0$, $2\alpha p \cdot q \leq p^2 + \alpha^2 q^2$ in deg- k SOS, so $\hat{\mathbb{E}}[pq] \leq \alpha^2 \hat{\mathbb{E}}[q^2]$. Send $\alpha \rightarrow 0$ to conclude. \square

Let's try to decompose the polynomial

$$\frac{1}{N} \sum_{i=1}^N (y_i^* - \langle w, x_i^* \rangle)^2 \quad (*)$$

Note:

$$1 = \underbrace{a_i^* a_i^*}_{\substack{\text{points we} \\ \text{correctly identified} \\ \text{as clean}}} + \underbrace{a_i^* (1-a_i^*)}_{\substack{\text{points we} \\ \text{incorrectly identified} \\ \text{as clean}}} + \underbrace{(1-a_i^*)}_{\substack{\text{points we} \\ \text{identified as} \\ \text{corrupted}}}$$

So

$$(\star) = \frac{1}{N} \sum_i a_i a_i^* (y_i^* - \langle w, x_i^* \rangle)^2 \quad (1)$$

$$+ \frac{1}{N} \sum_i a_i (1-a_i^*) (y_i^* - \langle w, x_i^* \rangle)^2 \quad (2)$$

$$+ \frac{1}{N} \sum_i (1-a_i) (y_i^* - \langle w, x_i^* \rangle)^2 \quad (3)$$

(1): when $a_i^* = 1$, $y_i = y_i^*$ and $x_i = x_i^*$

$$(1) = \frac{1}{N} \sum_i a_i a_i^* (y_i - \langle w, x_i \rangle)^2$$

$$\stackrel{(a_i^* \leq 1)}{\leq} \frac{1}{N} \sum_i a_i (y_i - \langle w, x_i \rangle)^2 \leftarrow \text{objective value}$$

(after taking expectation)
OPT

(objective value achieved by w^* , i.e. clean MSE)

Note: $\{a_i^*\}, w^*, \{x_i^*\}, \{y_i^*\}$ is feasible solution to SoS program

$$(2): \frac{1}{N} \sum_i a_i (1-a_i^*) (y_i^* - \langle w, x_i^* \rangle)^2$$

(Cauchy-Schwarz)

$$\leq \left(\frac{1}{N} \sum_i (1-a_i^*)^2 \right)^{1/2} \cdot \left(\frac{1}{N} \sum_i \underbrace{\frac{a_i^2}{\leq 1}}_{\leq \gamma^{1/2}} (y_i^* - \langle w, x_i^* \rangle)^4 \right)^{1/2}$$

$$\leq \gamma^{1/2} \cdot \left(\frac{1}{N} \sum_i (y_i^* - \langle w, x_i^* \rangle)^4 \right)^{1/2}$$

$$③: \frac{1}{N} \sum_i (1-a_i) (\underbrace{y_i^* - \langle w, x_i^* \rangle}_\text{in green})^2$$

(Cauchy-Schwarz)

$$\leq \left(\frac{1}{N} \sum_i (1-a_i)^2 \right)^{1/2} \cdot \left(\frac{1}{N} \sum_{i=1}^N (y_i^* - \langle w, x_i^* \rangle)^4 \right)^{1/2}$$

$$\text{Note: } \frac{1}{N} \sum_i (1-a_i)^2 \stackrel{\text{(Booleanity)}}{=} \frac{1}{N} \sum_i (1-a_i) \stackrel{\text{(\gamma fraction corruptions)}}{\leq} \gamma$$

$$\leq \gamma^{1/2} \cdot \left(\frac{1}{N} \sum_{i=1}^N (y_i^* - \langle w, x_i^* \rangle)^4 \right)^{1/2}$$

How to bound $\frac{1}{N} \sum_{i=1}^N (y_i^* - \langle w, x_i^* \rangle)^4$?

Recall $y_i^* = \langle w^*, x_i^* \rangle + \xi_i^* \stackrel{N(0, \sigma^2)}{\sim}$ so

$$\frac{1}{N} \sum_{i=1}^N (y_i^* - \langle w, x_i^* \rangle)^4 = \frac{1}{N} \sum_{i=1}^N (\langle w^* - w, x_i^* \rangle + \xi_i^*)^4$$

Note elementary inequality $(a+b)^4 \leq 8(a^4 + b^4)$,
 So the above

$$\begin{aligned} &\leq \boxed{\frac{8}{N} \sum_{i=1}^N \langle w^* - w, x_i^* \rangle^4} + \underbrace{\frac{8}{N} \sum_{i=1}^N \xi_i^4}_{\approx 8 \mathbb{E}_{\xi \sim N(0, \sigma^2)} [\xi^4]} \\ &= 24 \sigma^4 = O(\sigma^4) \end{aligned}$$

To summarize:

$$\begin{aligned} &\frac{1}{N} \sum_i (y_i^* - \langle w, x_i^* \rangle)^2 \\ &= \textcircled{1} + \textcircled{2} + \textcircled{3} \\ &= \text{OPT} + 2\gamma^{1/2} \cdot \left(\frac{1}{N} \sum_{i=1}^N (y_i^* - \langle w, x_i^* \rangle)^4 \right)^{1/2} \\ &\leq \text{OPT} + 2\gamma^{1/2} \cdot \left(\frac{8}{N} \sum_i \langle w^* - w, x_i^* \rangle^4 + O(\sigma^4) \right)^{1/2} \\ &\leq \text{OPT} + O(\gamma^{1/2}) \cdot \left[\left(\frac{1}{N} \sum_i \langle w^* - w, x_i^* \rangle^4 \right)^{1/2} + \sigma^2 \right] \end{aligned}$$

• Technically not legit in SOS b/c of fractional power ($1/2$),
 But this step can be made rigorous by writing as
 $(\text{clean MSE} - \text{OPT})^2 \leq O(\gamma) \cdot [\frac{1}{N} \sum_i \langle w^* - w, x_i^* \rangle^4 + \sigma^4]$

To bound $\frac{1}{N} \sum_i \langle w^* - w, x_i^* \rangle^4$, need assumption on distribution:

Def: q is 4-hypercontractive if

$$\mathbb{E}_{x \sim q} [\langle v, x \rangle^4] \leq \left(C \cdot \mathbb{E}_{x \sim q} [\langle v, x \rangle^2] \right)^2 \quad (\#)$$

for all $v \in \mathbb{R}^d$, for some $C = O(1)$.

q is certifiably 4-hypercontractive if $(\#)$ has an SOS proof.

proof at the end

Example: Any rotation of a product distribution (e.g. $N(\mu, \Sigma)$) is certifiably 4-hypercontractive.

Lemma: Let $N \geq \Omega(d^4 \lg^2(d/\delta))$. w.p. at least $1-\delta$, uniform distribution over N i.i.d. draws from a certifiably 4-hypercontractive dist. is certifiably 4-hypercontractive.

So for $v = w^* - w$, we get

$$\left(\frac{1}{N} \sum_{i=1}^N \langle w^* - w, x_i^* \rangle^4 \right)^{1/2} \leq \sqrt{\frac{1}{N} \sum_{i=1}^N \langle w^* - w, x_i^* \rangle^2}$$

$$\begin{aligned}
&= \frac{C}{N} \sum_{i=1}^N (y_i^* - \langle w, x_i^* \rangle - \bar{\epsilon}_i)^2 \\
&\stackrel{(a+b)^2 \leq 2a^2 + 2b^2}{\leq} \frac{2C}{N} \sum_i (y_i^* - \langle w, x_i^* \rangle)^2 + \frac{2C}{N} \sum_i \bar{\epsilon}_i^2
\end{aligned}$$

$$\leq \frac{2C}{N} \sum_i (y_i^* - \langle w, x_i^* \rangle)^2 + O(C\sigma^2)$$

So

$$\begin{aligned}
\text{clean MSE} &\leq \overbrace{\frac{1}{N} \sum_i (y_i^* - \langle w, x_i^* \rangle)^2}^{(1)} \\
&\leq \text{OPT} + O(\gamma^{1/2}) \left[\overbrace{\frac{2C}{N} \sum_i (y_i^* - \langle w, x_i^* \rangle)^2}^{(2)} + O(C\sigma^2) \right]
\end{aligned}$$

Rearranging, we get

$$\begin{aligned}
&\left(1 - O(C\gamma^{1/2})\right) \frac{1}{N} \sum_i (y_i^* - \langle w, x_i^* \rangle)^2 \\
&\leq \text{OPT} + O(C\gamma^{1/2}\sigma^2),
\end{aligned}$$

so if $C\gamma^{1/2}$ sufficiently small,

$$\text{clean MSE} \leq (\text{OPT} + O(C\gamma^{1/2}\sigma^2))$$

Proof IOU:

Claim: Any product distribution $q = q_1 \otimes \dots \otimes q_d$

s.t. $\mathbb{E}_{z \sim q_i} [z] = 0$, $\mathbb{E}_{z \sim q_i} [z^2] = 1$, and

$\mathbb{E}_{z \sim q_i} [z^4] \leq C^2$ $\forall i$ is certifiably 4-hypercontractive.

$$\begin{aligned} \text{PF: } \mathbb{E}_{x \sim q} \langle v, x \rangle^4 &= \sum_i v_i^4 \mathbb{E}[x_i^4] + 6 \sum_{i \neq j} v_i^2 v_j^2 \mathbb{E}[x_i^2 x_j^2] \\ &\leq \sum_i C^2 v_i^4 + 6 \sum_{i \neq j} v_i^2 v_j^2 \end{aligned}$$

$$\begin{aligned} \xrightarrow{\substack{\text{The only} \\ \text{SW step}}} &= \max(3, C^2) \left(\sum_i v_i^2 \right)^2 - \sum_i |C^2 - 3| \cdot v_i^4 \\ &\leq \max(3, C^2) \cdot \left(\sum_i v_i^2 \right)^2 \end{aligned}$$

$$\mathbb{E}_{x \sim q} \langle v, x \rangle^2 = \sum_i v_i^2 \mathbb{E}[x_i^2] = \sum_i v_i^2. \quad \square$$

(note that definition of certifiable hypercontractivity is invariant under linear transformations of q)