

LECTURE 21

Robust

Linear Regression

Setup: $\hat{l}(x) = \underbrace{\langle l, x \rangle}_{\text{---}}$ $\hat{l} \in \mathbb{R}^d$ $\|\hat{l}\| = 1$

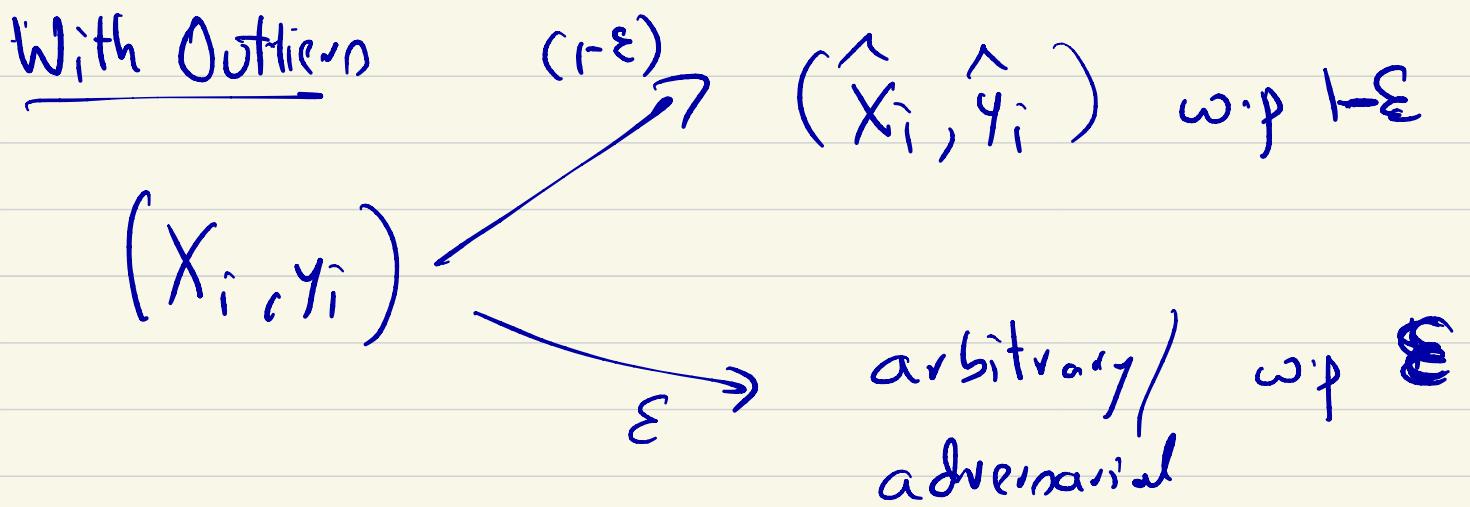
TRUE SAMPLES: $(\hat{x}, \hat{l}(\hat{x}) + \gamma)$
 \uparrow random $\{\pm 1\}$ vector, value of linear
 \hat{x} function

gaussian noise

GOAL: $\{\hat{x}_i, y_i\}$ examples find a linear

function l

$$\underset{\|l\| \leq 1}{\text{Min}} \sum_i (\langle l, x_i \rangle - y_i)^2$$



Goal: Recover $\hat{\theta}$.

Polynomial System (P) $\{x_i, y_i\}_{i=1}^n$ samples

$$\{\underline{w_i}\}_{i=1}^n$$

$$\underline{l} = (l_1 \dots l_d)$$

$$\begin{cases} 1 & \text{if } i \text{ is true sample} \\ 0 & \end{cases}$$

$$1) \quad w_i^2 - w_i = 0 \quad \forall i = 1 \dots n \quad | \quad w_i = \begin{cases} 0 \\ 1 \end{cases}$$

$$2) \quad \sum_{i=1}^n w_i = (1-\varepsilon)n$$

$$3) \quad \|l\|^2 = \sum_i l_i^2 \leq 1 \quad (w \& l)$$

$$\text{Min} \left[\frac{1}{n(1-\varepsilon)} \sum_{i=1}^n w_i (y_i - \langle l, \underline{x}_i \rangle)^2 \right]$$

Identifiability Any solution (w, l) to P

satisfies

$$\sum_{i \in S} (y_i - \langle l, \underline{x}_i \rangle)^2 \leq \Delta$$

1) Solve deg 10 SoS SDP relaxation:

$$\tilde{E}[\omega_1], \dots, \tilde{E}[\omega_n]$$

$$\tilde{E}[\ell_i \ell_j], \quad \tilde{E}[\omega_i \omega_j], \quad \tilde{E}[\omega_i \ell_i \ell_b], \dots$$

(fictitious/pseudo moments of a distribution
over solutions to $P\}$)

2) Output $\ell^* = (\tilde{E}[\ell_1], \dots, \tilde{E}[\ell_n])$

$$\ell^*(x) = \underbrace{\langle \tilde{E}[\ell], x \rangle}_{=}$$

Want $w_i \in \underline{[0, 1]}$ $\Rightarrow \tilde{E}[w_i] \in [0, 1]$

How to prove: $\tilde{E}[w_i] \in \underline{[0, 1]}$??

$$\tilde{E}[w_i] \geq 0 \Rightarrow w_i^2 - w_i = 0$$

$$\tilde{E}[w_i] = \tilde{E}[w_i^2] \geq 0$$



square poly

$$\left\{ w_i \geq 0 \iff w_i = w_i^2 \geq 0 \right\}$$

Using SOS
 $w_i \leq 1$

$$\left. \begin{array}{l} (1-w_i)^2 \geq 0 \\ \parallel \\ 1-2w_i+w_i^2 \parallel (w_i^2-w_i=0) \\ 1-w_i \\ \hline 1-w_i \geq 0 \Rightarrow \underline{w_i \leq 1} \end{array} \right\}$$

$$\omega_i^2 - \omega_i = 0$$

\Downarrow

$$\omega_i(1-\omega_i) = 0$$

\Downarrow

$$\omega_i \in \{0, 1\}$$

\Downarrow

$$\omega_i \leq 1$$

Proof

$$\{\omega_i^2 - \omega_i = 0\} \xrightarrow[2.5\text{os}]{} \omega_i \leq 1$$

$$(1-\omega_i)^2 \geq 0$$

$$1 - 2\omega_i + \omega_i^2 \geq 0$$

$$1 - \omega_i$$

$$\frac{1 - \omega_i}{1 - \omega_i > 0} \Rightarrow \underline{\omega_i \leq 1}$$

$$\tilde{E}[\omega_i] \leq 1$$

\Longleftrightarrow

$$\underbrace{P \Rightarrow a(x) \geq 0, P \Rightarrow b(x) \geq 0}_{\longrightarrow}$$

$$P \Rightarrow a(x) \cdot (\alpha(x)^2 + b(x) \cdot (\beta(x))^2) \geq 0$$

$$P \Rightarrow 1 \geq 0$$

$$P \Rightarrow (\alpha(x))^2 \geq 0$$

Using an SoS proof: show identifiability

"

Any solution (ω, ℓ) to P
satisfies $\sum_{i \in S} (y_i - \langle \ell, x_i \rangle)^2 < \Delta''$

$$\left[P \xrightarrow[\text{deg 8}]{\text{SoS}} \sum_{i \in S} (y_i - \langle \ell, x_i \rangle)^2 \leq \Delta \right]$$

↓

$$\tilde{E} \left[\sum_{i \in S} (y_i - \langle \ell, x_i \rangle)^2 \right] \leq \Delta$$

∨

$$\sum_{i \in S} (y_i - \langle \tilde{E}[\ell], x_i \rangle)^2$$

Thm: Suppose \tilde{E} is a deg 4 p.e.f for
and $\ell^* = \tilde{E}[\ell]$ polynomial system P

$$\frac{1}{|S|} \tilde{E} \sum_{i \in S} (y_i - \langle \ell, x_i \rangle)^2 \leq \frac{1}{|S|} \sum_{i \in S} (y_i - \hat{\langle} \ell, x_i \rangle)^2 + O(\sqrt{\epsilon})$$

$$\hat{w}_i = \begin{cases} 1 & \text{if } i \in S \text{ (true sample)} \\ 0 & \text{otherwise} \end{cases}$$

$$\text{LHS} = \frac{1}{(1-\epsilon)n} \tilde{E} \left[\sum_{i=1}^n \hat{w}_i (y_i - \langle \ell, x_i \rangle)^2 \right]$$

$$= \tilde{E} \sum_{i=1}^n \underbrace{\hat{w}_i}_{\hat{w}_i(1-w_i)} (y_i - \langle \ell, x_i \rangle)^2 \xrightarrow{(1)} \hat{w} = \hat{w}_i(1-w_i) + \hat{w}_j w_j + \dots$$

$$+ \tilde{E} \sum_{i=1}^n \underbrace{\hat{w}_i w_i}_{\hat{w}} (y_i - \langle \ell, x_i \rangle)^2 \xrightarrow{(2)}$$

(2)

$$+ \sum_{i=1}^n \hat{w}_i w_i (y_i - (\ell, x_i))^2$$

$$\leq \tilde{E} \left[\sum_{i=1}^n w_i (y_i - \langle l, x_i \rangle)^2 \right]$$

$\underbrace{\hspace{10em}}$

SDP-OPT

$$\leq \sum_{i \in S} (y_i - \hat{f}(x_i))^2 \quad \left\{ \begin{array}{l} \because \text{SOP} \\ \text{is orientation} \end{array} \right.$$

$$\tilde{E} \sum_{i=1}^n \hat{w}_i^{\downarrow} \underbrace{(-\omega_i)}_{a_i} \left(y_i - \underbrace{\langle \underline{\ell}, \underline{x}_i \rangle}_{\text{bias}} \right)^2$$

\Leftarrow

$$\left[\tilde{E} \sum_{i=1}^n (1 - \omega_i)^2 \right]^{1/2} \quad \left[\tilde{E} \sum_i \hat{w}_i^2 (y_i - \langle \underline{\ell}, \underline{x}_i \rangle)^2 \right]^{1/2}$$

||(Mean) (4) || n n

$$(3) = \tilde{E} \left[\sum (1 - \omega_i)^2 \right]$$

$$= \tilde{E} \left[\sum (1 - 2\omega_i + \omega_i^2) \right] = \tilde{E} \sum_i (1 - \omega_i)$$

$$= n - \tilde{E} \left[\sum_i \omega_i \right]$$

 $\sum n$

$$(4) \leq O(n)$$

$$\begin{aligned} & \tilde{E} \sum_i \hat{w}_i^2 (y_i - \langle l, x_i \rangle)^4 \\ & \leq 8 \tilde{E} \left[\sum_{i=1}^n \hat{w}_i^2 (y_i^4 + \langle l, x_i \rangle^4) \right] \end{aligned}$$

$(a+b)^4 \leq 8a^4 + 8b^4$
(using Sos)

$$\leq 8 \tilde{E} \left[\sum_{i \in S} y_i^4 + \sum_{i \in S} \langle l, x_i \rangle^4 \right]$$

$$\leq 8 [O(n) + \tilde{E} \sum_{i \in S} \langle l, x_i \rangle^4]$$

$$\leq O(n) + 8 \tilde{E} \left[\sum_{q \in S} \sum_{a,b,c,d} l_a l_b l_c l_d \cancel{x_{ia} x_{ib} x_{ic} x_{id}} \right]$$

$$\leq O(n) + 8 |S| \tilde{E} \left[\sum_{q \in S} \left(\sum_{i \in S} x_{ia} x_{ib} x_{ic} x_{id} \right) l_a l_b l_c l_d \right]$$

$$\leq O(n) + 8 O(n) \cdot \tilde{E} (\|l\|^2)^2 \leq O(n) + O(n)$$

Fact:

Univariate poly $p(x) \geq 0$ $\forall x \in \mathbb{R}$



$$P(x) = \sum q_i^2(x)$$

FACT:

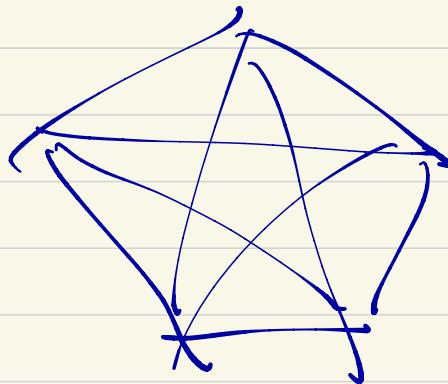
$$\left. \begin{array}{l} x_i^2 = x_i \\ \sum_{i=1}^n x_i = \gamma_2 + \pi \end{array} \right\} \in P \quad \begin{array}{l} x_i \in \{0, 1\} \\ \Rightarrow \end{array}$$

Thm:

$$P \Rightarrow -1 \geq 0 \quad \text{using SOS}$$

need s degree $S(n)$

Input: Clique on $2n-1$ vertices

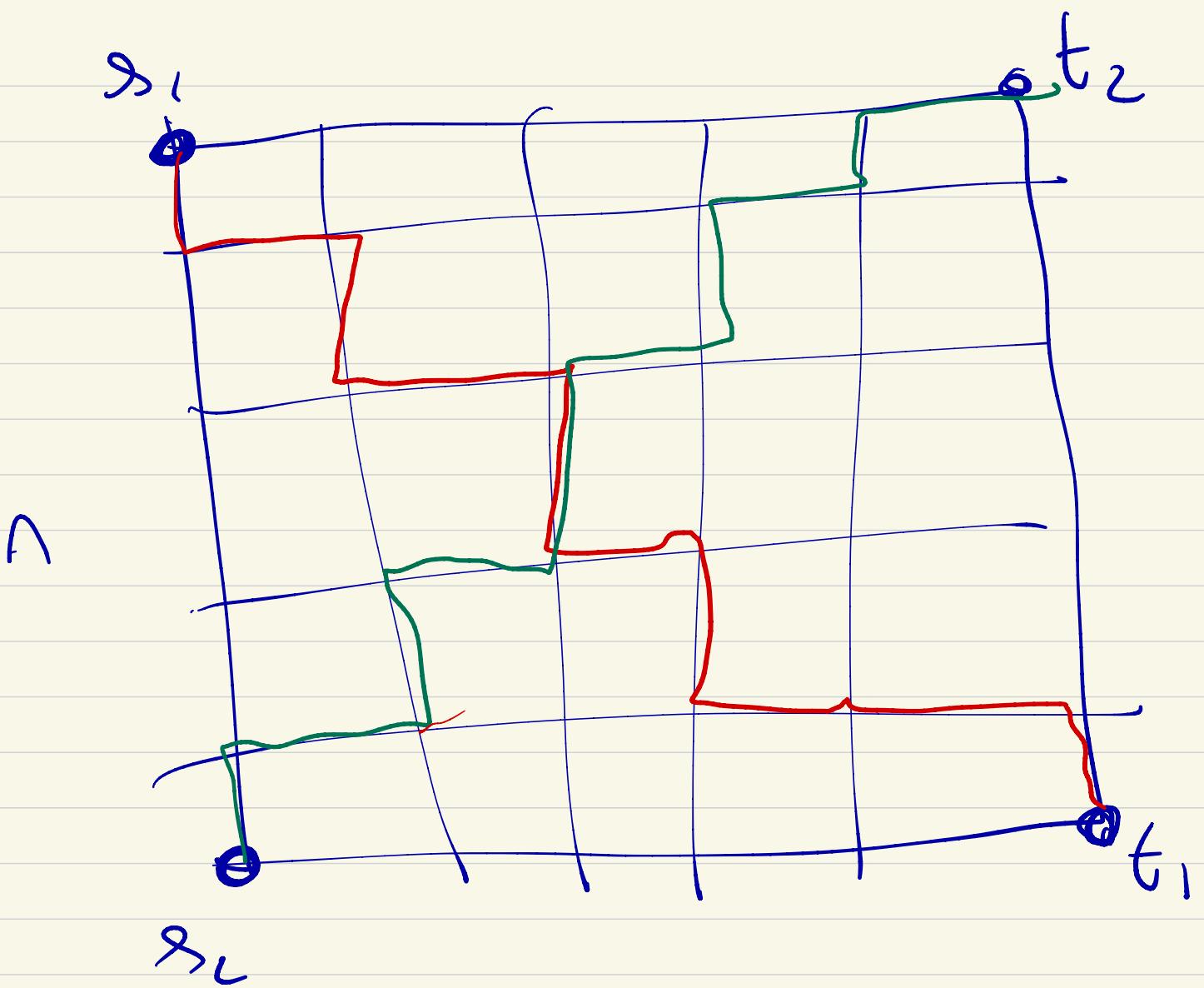


Perfect matching ??

$$x_{ij} = \begin{cases} 1 & \text{if } (i,j) \in \text{Matching} \\ 0 & \text{otherwise} \end{cases}$$

$$\sum_j x_{ij} = 1 \quad \forall i$$
$$x_{ij} = x_{ji}$$
$$x_{ij}^2 = x_{ij} \quad (x_{ij} = \{0, 1\})$$

Needs $S(n)$ degree SoS proof



A
grid

$$\frac{1}{|S|} \sum_{i \in S} \langle l, x_i \rangle^4 \xrightarrow[\substack{\text{sample} \\ \text{dist., but}}]{\substack{\text{various}}} \mathbb{E}_{\substack{x \sim \text{sample} \\ \text{dist., but}}} [\langle l, x \rangle^4]$$

$$= \sum_{a,b,c,d} l_a l_b l_c l_d \mathbb{E}_{\substack{x \sim \text{sample} \\ \text{dist.}}} [x_a x_b x_c x_d]$$

$$\leq 3 \left(\sum_b l_a^2 \right) \left(\sum_b l_b^2 \right)$$

$$a=b \quad c=d$$

$$a=c \quad b=d$$

$$a=d \quad b=c$$

$$\leq 3 \left(\|l\|^2 \right)^2$$

$$\text{[Cauchy-Schwarz]} \quad \sum_{i=1}^n a_i b_i \leq (\sum a_i^2)^{1/2} (\sum b_i^2)^{1/2}$$

$$E\left[\sum_{i=1}^n a_i b_i\right] \leq \left(E\left[\sum_i a_i^2\right]\right)^{1/2} \left(E\left[\sum_i b_i^2\right]\right)^{1/2}$$

Pseudo-expectation (Cauchy-Schwarz)

$$\tilde{E}\left[\sum_{i=1}^n a_i b_i\right] \leq \underbrace{\left[\tilde{E}\left(\sum_i a_i^2\right)\right]^{1/2} \left(\tilde{E}\left[\sum_j b_j^2\right]\right)^{1/2}}$$

$$\left(\sum_{i=1}^n a_i b_i\right)^2 \leq \left(\sum_i a_i^2\right) \left(\sum_j b_j^2\right)$$

$$\text{RHS - LHS} = \sum_{i,j} (a_i b_j - a_j b_i)^2 \geq 0$$