

Learning mixtures of spherical Gaussians.

$$X = \sum_{i=1}^K w_i \mathcal{N}(\mu_i, I) \subset \mathbb{R}^n.$$

Given samples from  $X$  and you want to learn  $X$ .

(either: learn  $\hat{X}$  s.t.  $d_{TV}(X, \hat{X}) < \epsilon$ ,

learn  $\hat{\mu}_i$  s.t. (up to a perm.)  $|\mu_i - \hat{\mu}_i| < \epsilon$ )

Note: Learning Non-spherical Gaussians has  
 $n^{\Omega(k)}$  lower bound in the  
SQ model.

Natural approach is method of moments.



[as long as  $\mu$  is not too big]

easy approximate 1st  
d moments of  $X$  for  
any constant  $d$ .

0

0

Question What are these moments?

$$X = D * G$$

→  
discrete distr.  
over mms.

↑  
 $N(0, I)$

$$\mathbb{E}[X^{\otimes d}]$$

$$= \sum_i (\text{sgn}(D^{\otimes d})_{ii} \otimes G^{\otimes d})$$

Not hard to deconvolve.  
approximate  $\mathbb{E}[D^{\otimes d}]$

Q Can we learn  $D$  from its low order moments?

A No. Consider 1-d. problem.

first ~~is~~  $d$  moments,  $d+1$  parameters

but  $X$  has  $2k$  parameters.

$\Rightarrow$  Unless  $d \geq 2k$ , moments don't determine  $D$ .

$D, D'$  match moments, (even up to small distance)

$N \cdot D, N \cdot D'$  also match moments.

embed into  $n$  dimensions

$$V^d = G_1 x^{d-1} + \dots + G_d \leq r_i(t) \text{ for } t \leq d$$

Sym poly of deg  $2d$ .  $(\sqrt{r_i})$   
 $(r_i)$   
 $\dots$

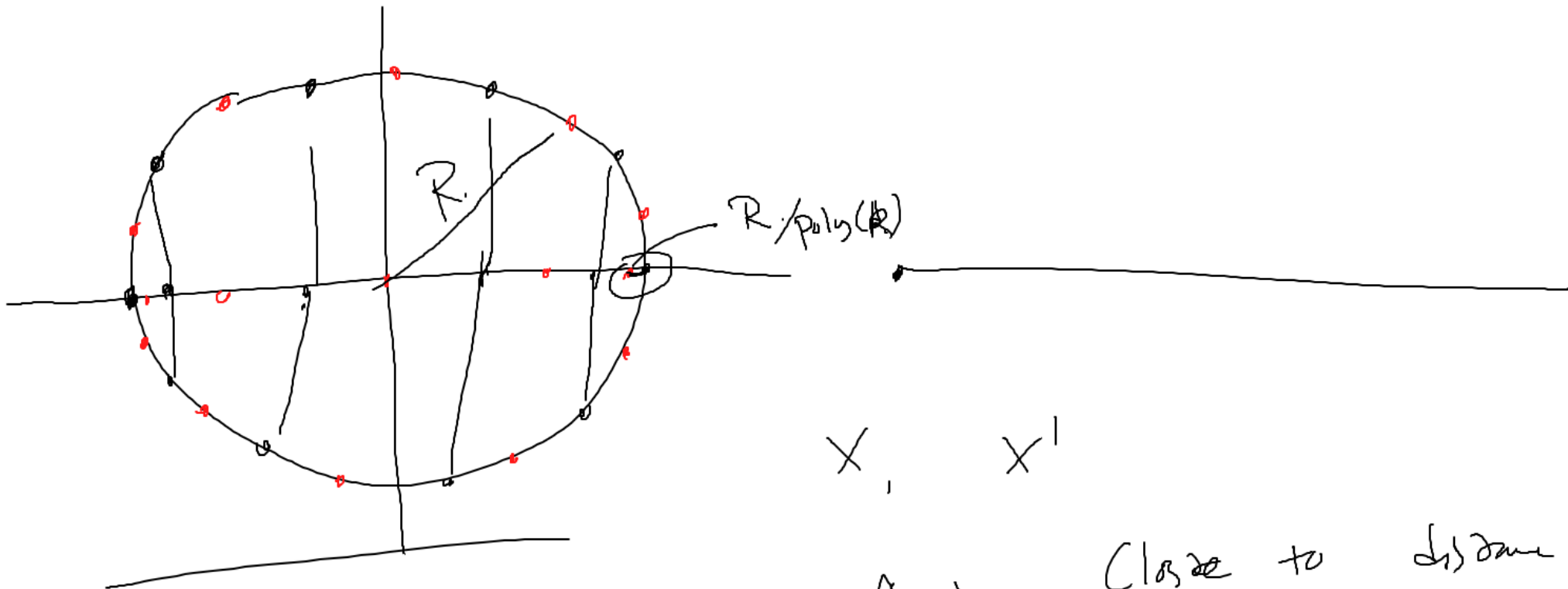
(coeff of  $p$   
 $=$  Sym poly  
 in roots)

Coef of  $\delta'$   
 $= \dots$   
 $= \dots$

Pick  $P, P'$  two degree  $d$  polys that  
 differ by a constant.

(w/  $d$  real roots).

Let  $\alpha$  let  $\alpha$  be roots of  $P$   
 $\alpha'$  be roots of  $P'$



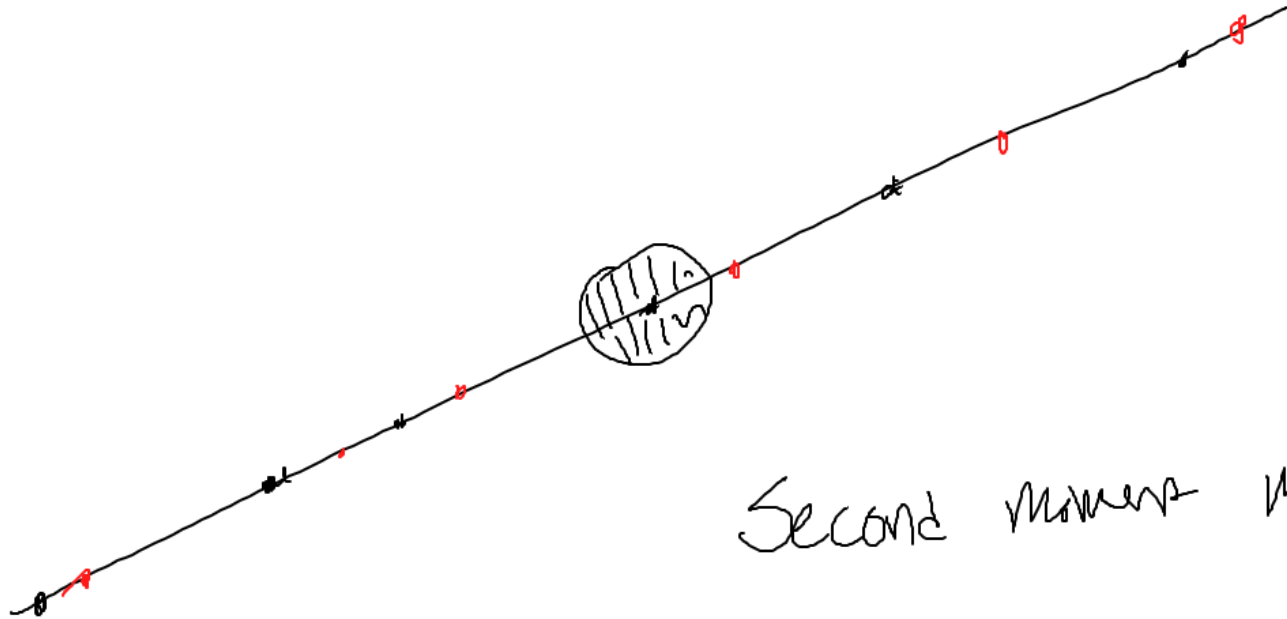
$X, X'$

~~Arb.~~

Close to distance 1  
in dTV.

but first ~~two~~ moments match.

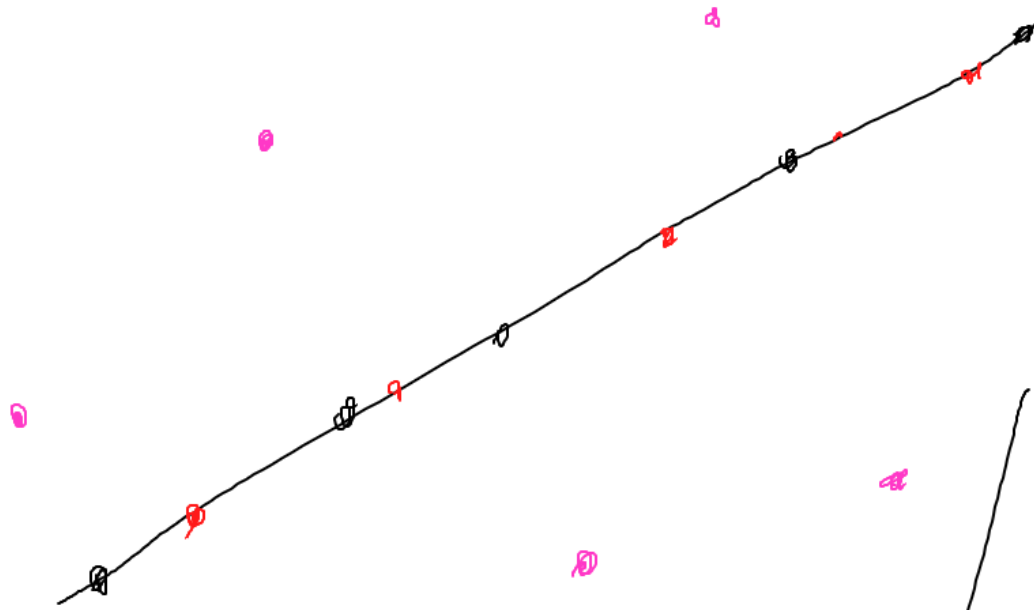
any 1-d  $X$  approximated  
by empirical  
distribution  $*N(0, \Sigma)$



Second moment matrix of  $(X)$

$$= I + \text{Second moments of } (D)$$

$\swarrow$  Suppose on a ~~span~~  $k$  Subspace of dim.  $k$   
~~Span~~  $\text{rank } k$   
 $\text{Span} \subset \text{Span of } \mu_i$ 's.



reduce to  
k-dim  
problem.

to prove lower bounds  
(say in SA)

Want is measures that  
are spatially symmetric.

Impossible

$$X = D * G.$$

low deg moments of  $X \rightarrow$  low deg moments of  $D$

$$\rightarrow \sum_{i=1}^R w_i \mu_i \quad \text{for any low degree poly. } q$$

$$q(x) = p^2(x) \quad \text{for some } x$$

$$= \sum w_i p^2(\mu_i)$$

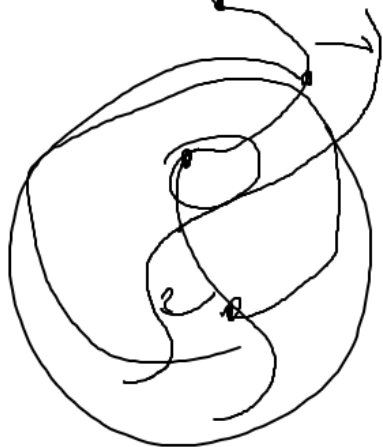
$$= 0 \quad \text{iff} \quad \boxed{p(\mu_i) = 0} \quad \text{for all } i.$$



Are the low degree polys  $\in$  S.A

$P(\mu_i) \Rightarrow$  for all  $i$ ?

Yes



$p=0$

Linear transformation  $\begin{pmatrix} n+d \\ d \end{pmatrix} \rightarrow \bigcup \mathbb{R}^k$   
 $\left\{ \begin{matrix} \deg \leq d \\ \text{poly} \end{matrix} \right\} \rightarrow \frac{(P(\mu_1), P(\mu_2), \dots, P(\mu_n))}{\dim k}$

Kernel of  $J$ .

Kernel has  $\dim \leq k$ .

$\begin{bmatrix} \cdot \\ \cdot \\ \cdot \end{bmatrix} \mathbb{R}^{n'}$

$\begin{pmatrix} n'+d \\ d \end{pmatrix} > k$  if  $\underline{n'} > \underline{d + k}$ .

Compose neurons  $\rightarrow$  (can)  $\rightarrow$  Compose space of low degree  
 Vanishing Polynomials  $\mathbb{P}_d$   
 (exactly).

$V =$  Variety defined by these polynomials

$=$  Set of pts for which these polys vanish.

$M_i \in V$ .

intuition  $V$  small

Proof  $\dim(\text{deg-}d \text{ polys on } V) \leq \text{Codim of space of defining polys.}$   
 $\leq R$ .

$\binom{\dim(V) + d}{d}$

$\Rightarrow \underline{\dim(V) \leq d \cdot R/d.}$

$\mathbb{R}$

Idea "project" onto  $V$ .

do something exhaustive within  $\exp(dn(U))$

$$d = \lg(K)$$

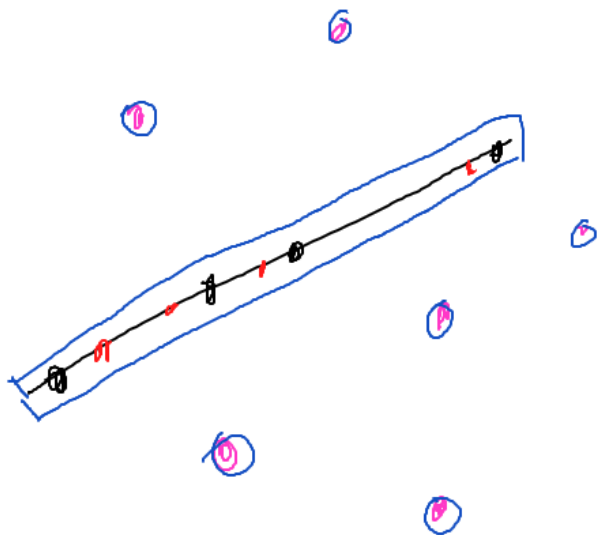
Complexity of answer data

$(nd)$  time.

reduce to  $d^{1/d} \approx d$ . dimension varies

exhaustive then takes  $\approx 2^d$  time.

quasi-poly alg



## Technical problems

- ①  $V$  is low dimension but is it "simple"?
- ② How do we project onto  $V$ ?
- ③ How do we compute on  $V$ ?
- ④ How do I deal with approx. defining polynomials?

Idea ~~if~~ replace  $V$  w/ a point cloud.

~~Set~~ Vector space  $U$  of degree  $d$  polys  $\mathbb{R}$   
 $\text{Codim}(U) \leq k$ .

$$S = \{ x \in \mathbb{R}^n : |x| < R, \quad |p(x)| < \underbrace{\delta}_{\text{we'll get back to this}} |p| \text{ for all } p \in U \}$$

Want for  $\varepsilon > 0$  to find a small  $\varepsilon$ -cover of  $S$ .

~~Thm~~ Thm If  $\varepsilon > \delta^{\frac{1}{2d}} \text{poly}(R \text{ and } k)$  then  
 there exists an  $\varepsilon$ -cover of size at most  
 $(2(R/\varepsilon) \cdot dk) O(d^2 k^{1/d})$

expect lower side should be  $\geq$

$$O(R/\varepsilon)^{\dim(V)} \leftarrow$$

$$\underline{\underline{dk' \varepsilon}}$$

we get  $\underline{\underline{d^2 k' \varepsilon}}$

$$\boxed{\varepsilon > \delta^{1/2} (\dots)}$$

$$\bigcirc \quad \frac{1}{\varepsilon}$$

## Technicalities

①  $U \subset$  hom. degree- $d$  polynomials.

②  $\delta < (\epsilon / R^d)^{1/d}$

$$p(x) = A \cdot x^{\otimes d}$$

$$|p| := |A|_2.$$

$|p|$  is spherically symmetric.

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Define  $f(\underline{\epsilon}, \underline{R}, \underline{d}, \underline{k}, \underline{n}, \underline{\delta}) =$  ~~the~~ least case can size.

Prove a recursive upper bound on  $f$ .

$$\mathbb{R}^n = \mathbb{R}^{n'} \times \mathbb{R}^{n-n'}$$

$\downarrow$                        $\downarrow$   
 $(x, y)$

$n'$  as small as

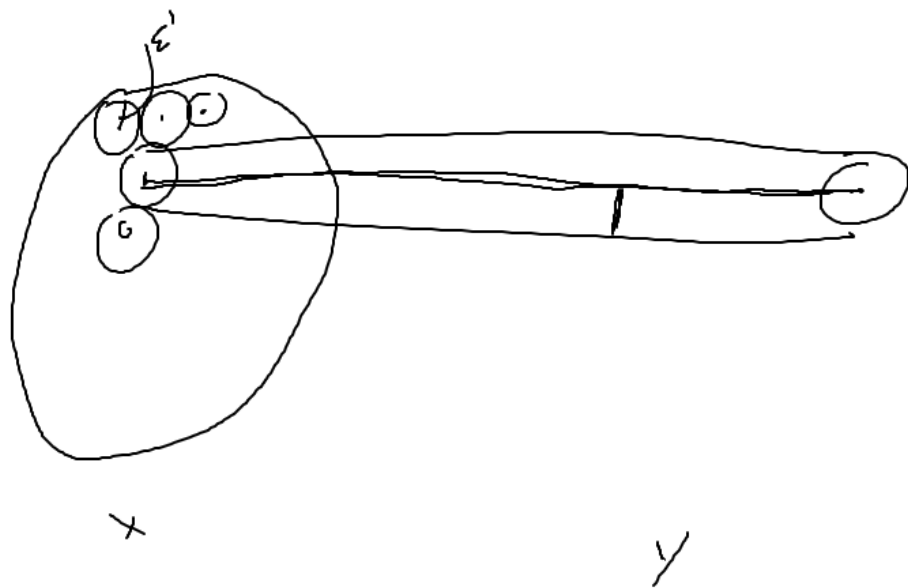
$W = \{ \text{polys in } U \text{ that are degree } \leq d \text{ in } x \text{ \& degree } \leq d-1 \text{ in } y \}$

$$\text{Codim}(W) \leq k.$$

if  $U$  is  $\text{Codim } k$  in  $V$ .

then  $U \cap W$  is  $\text{Codim } \leq k$  in  $V \cap W$ .





$\epsilon'$  very small

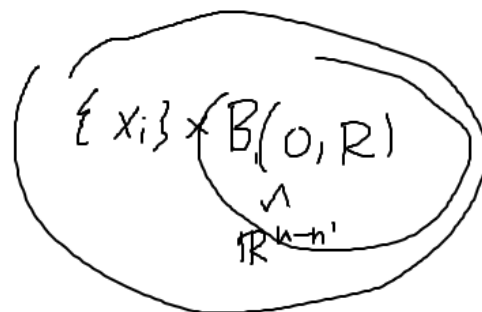
Small enough

If I change  $\delta$  by a bit...

It is enough to cover

this center line.  $\{x_0\} \times \mathbb{R}^{n-1}$

$\{x_i\}$



$$A_{x_i} : W \longrightarrow \{ \text{new deg } d-1 \text{ polys, in } y \}$$

$$P(x, y) \longrightarrow P(x_i, y)$$

deg 1, deg d-1  
polys, in U.

$$S_{x_i} = \{ \text{new vanishing pts w/ } x = x_i \}$$

not too small!

$$n' + (-1)$$

$$\text{If } q = A_{x_i} P.$$

$$\cancel{1/q} |q| > \eta |p|.$$

then  $S_{x_i}$  must nearly vanish on  $q$ .

$(x_i, y)$   
nearly vanishes

$U_{x_i} = \text{Span of the singular vectors of } A_{x_i} \text{ w/}$   
singular value  $\geq \eta$

$$\delta |p| = (\delta/\eta) |q|.$$

$$S_{x_i} \subset \{ \text{pts that nearly vanish on } U_{x_i} \}$$

recursive version of Drignani problem.

Say  $x_i$  is good if  $\text{codim}(U_{x_i}) \leq R'$

If  $x_i$  is good then the cylinder can be covered by  
a set of size.

$$f(d-1, R', n-n', R, \tilde{\epsilon}, \tilde{\delta}/\eta)$$

What  
about  
the rest  
good pts

Then I can  
cover all of the  
cylinders using  
all of the  
good

$$\delta \leq d \dots$$

$$\eta \approx \text{poly}(\epsilon)$$

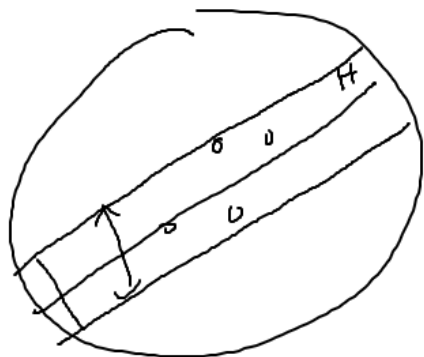
$$f(\dots) \leq (\dots)^{d^2 R^{d-1}}$$

$$R' \approx R^{\frac{d-1}{d}}$$

points w/ an appropriate  
# of covers.

Prop There exists a subspace  $H \subset \mathbb{R}^{n'}$   
 of dimension  $\leq 2k/k!$

s.t. all of the bad pts are  
close to  $H$



Cover  $H \times \mathbb{R}^{h-n'}$  as close as  
 $n' \gg k/k! \approx k^{1/d}$ .  
 Subspace of dim  $\mathbb{R}^{h-n' + 2k/k!}$  Then I  
 should be fine.

$\rightarrow f(d, k, R, \tilde{\epsilon}, \delta, \underline{n-n' + 2k/k!})$