

# Study Group Lecture #2

## Lect #1: GMM, Compression Scheme

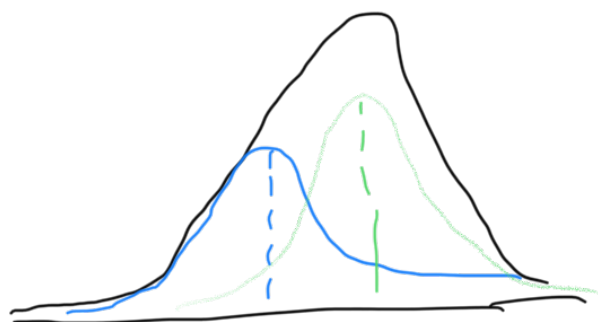
Today: K MV 10, MV 10



$$N(\mu, \Sigma) = \frac{1}{\sqrt{(2\pi)^d \det(\Sigma)}} \exp\left(-\frac{1}{2}(x-\mu)^T \Sigma^{-1}(x-\mu)\right)$$

$$GMM_2(x) = w_1 N(\mu_1, \Sigma_1, x) + (1-w_1) N(\mu_2, \Sigma_2, x)$$

captures heterogeneous populations



Given  $x \sim GMM_K$

Find  $\vec{w}, \vec{\mu}, \vec{\Sigma} \dots$

Pearson (1894) (crabs) 1000 (1dim)

Given  $x_1, \dots, x_m \sim GMM_2$ , can we estim the 5 unknowns?

Method of Moments

$$\vec{\theta} = (\underbrace{w_1, \mu_1, \sigma_1^2, \mu_2, \sigma_2^2}_{\text{5 unknowns}}) \in \mathbb{R}^5.$$

$$\mathbb{E} X = \underbrace{w_1}_{\text{unknown}} \underbrace{\mu_1}_{\text{unknown}} + (1-\underbrace{w_1}_{\text{unknown}}) \underbrace{\mu_2}_{\text{unknown}}$$

$X \sim GMM_2$

$$\mathbb{E} X^2 = w_1 (\mu_1^2 + \sigma_1^2) + (1-w_1) (\mu_2^2 + \sigma_2^2)$$

$$\mathbb{E} X^r = M_r(\vec{\theta})$$

Pearson's 6th Moment

Question: Prove  $M_r(\vec{\theta})$  is  $\text{poly}(\vec{\theta})$

Stein Lemma  $\mathbb{E}[f'(X)] = \mathbb{E}[X f(X)]$

$X \sim N(0,1)$   $X \sim N(0,1)$

$\mathbb{E}[(X-\mu)f] = \mathbb{E}[f'] \sigma^2$

Find  $\vec{\theta}$

$$T = \{x_i\} \sim \text{GMM}_2$$

$$\hat{M}_r = \text{empirical } r\text{-th moment} = \frac{1}{|T|} \sum_{x_i \in T} x_i^r, \quad r=1, \dots, 5$$

$$\text{Solve } \left\{ \hat{M}_r = M_r(\vec{\theta}) \right\}_{r \in \underline{[5]}} \rightarrow \underline{\vec{\theta}_1}, \underline{\vec{\theta}_2} \in \mathbb{R}^5$$

verify which is closer in 6th moment.

$$\text{dim } \# \text{mixt} \\ \underline{d=1, k=2} \quad (1894) \rightarrow 6$$

(a)  $d=1, k \geq 2$  ???

(b) Why? Distr  $\Leftrightarrow$  Moments

Moment Problems (1890++)

Given  $(m_n)_{n \geq 0}$ :  $m_n = \int x^n \underline{d\mu}$ , does there exist pos. Borel measure  $\mu$ ?

If so, is it unique?

$\mu$  has finite support  $\stackrel{\text{by}}{\Rightarrow}$  Linear algebra.

$\mu$  has compact support  $\stackrel{\text{by}}{\Rightarrow}$  Weierstrass thm & Riesz repr.

$X=[0,1]$  (1923, Hausdorff)

$$m_n = \int_0^1 x^n \underline{d\mu} \quad \rightarrow \text{(determinate)}$$

has unique soln

$$(2^k m)_n = \sum_{i=0}^k \binom{k}{i} (-1)^i m_{n+i}$$

$k \wedge k \wedge$

$11 \dots$

$$(-1)^k (\underbrace{\partial m}_n) \geq 0 \quad \forall n, k \geq 0$$

$X = \mathbb{R}$  (1920)

Hamburger moments (H.M.)

$$\underline{m_n = \int_{-\infty}^{\infty} x^n d\mu}$$

(H.M.) is solvable

$$\Leftrightarrow A = \begin{pmatrix} m_0 & m_1 & m_2 & \dots \\ m_1 & m_2 & m_3 & \dots \\ m_2 & m_3 & m_4 & \dots \\ \vdots & \vdots & \vdots & \ddots \end{pmatrix} \succeq 0$$

Hankel matrix

solution space is convex

$$\underbrace{K(x, y)}_{\text{kernel}} : \mathbb{N} \times \mathbb{N} \rightarrow \mathbb{R}$$

$\forall \text{ seq } (v_j) \text{ with finite support}$

$$\forall A v v^* \geq 0$$

sketch

"only if": Say  $\mu$  exists.

$$\sum_{j, k} m_{j+k} v_j \overline{v_k} = \int_{-\infty}^{\infty} \left| \sum_{j \geq 0} v_j x^j \right|^2 d\mu \geq 0$$

$\Leftarrow \dots$

Carleman's condition (1926)

(Suff. condition)

$\mu$  on  $\mathbb{R}$

If  $\sum_{n=1}^{\infty} m_{2n}^{-\frac{1}{2n}} = +\infty$  then (quasianalytic)

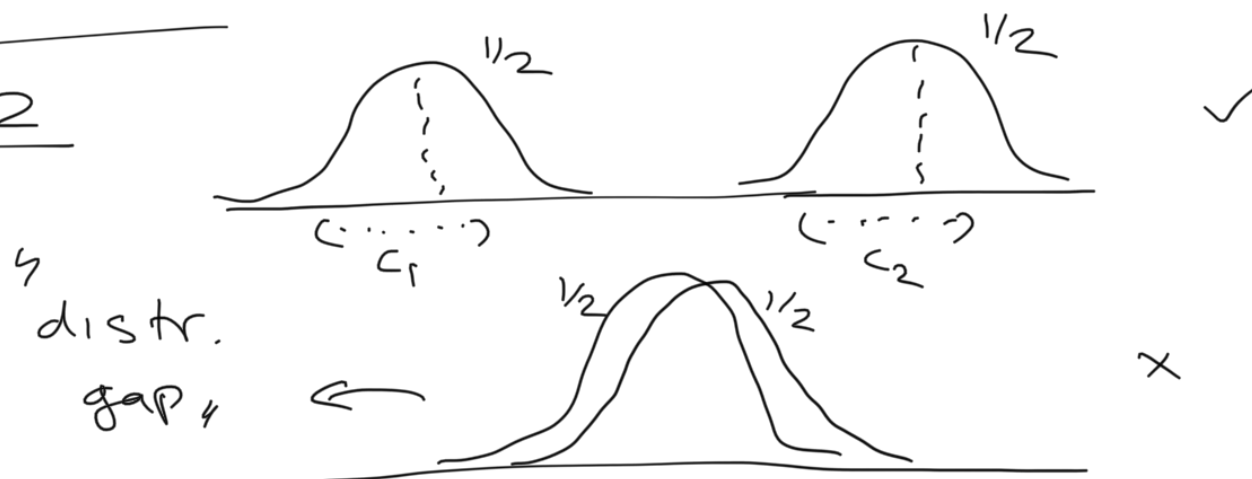
$(m_n)$  is determinate

dim  
 $d=1, k=2$  ( 6 "matched" moments  $\Rightarrow$  "Learning" )

$d=1, k=?$

Is Pearson  
"Robustly Identifiable"

Part 2



① Clustering  $\Rightarrow$  Strong Separability Assumptions.

Today  $\Rightarrow$  NO  $\gg$

② GMM is identifiable (Teicher)  $g: \text{param} \rightarrow \text{density}$   
"1-1"

③  $\underline{X} \sim N(\mu, \Sigma)$   
 $\sqrt{\Sigma}$  ✓

Proj of GMM is one-dim GMM.

(K) upper bound

Output:  $\hat{F} = \sum_{i=1}^{\hat{k}} \hat{w}_i \hat{F}_i$

correctly learn  $\hat{k}$

(v1) & approx  $(w_i, \mu_i, \Sigma_i) \rightarrow$  algo 2 ( $\frac{\text{high}}{d}$ )

(v2)  $\rightarrow TV(P, Q) = \frac{1}{2} \int_{\mathbb{R}^d} |p(x) - q(x)| dx \rightarrow$  algo 3  
 $\in [0, 1]$

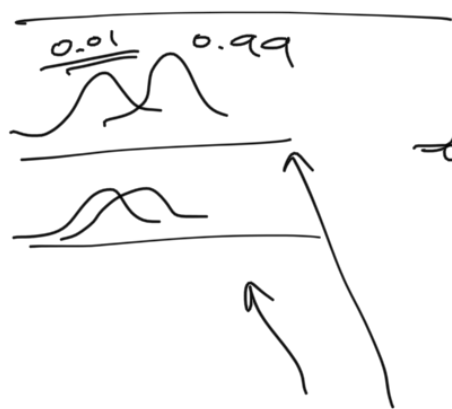
Plan

algo 1 ( $\frac{1}{d}$ )

algo 2 ( $\frac{\text{high}}{d}$ )

algo 3

$$\text{TV}(\cdot, \cdot) < \varepsilon \Rightarrow \|\cdot\|_2, \|\cdot\|_F < O(\varepsilon)$$



How hard?

Inspiration from Lin. Alg

Condition #. (GMM)

$$\kappa(A) = \frac{\lambda_{\max}(A)}{\lambda_{\min}(A)} \geq 1$$

$\kappa \rightarrow \infty$   
bad  
 $Ax = b$   
robust + stable

$$\kappa(F) = \frac{1}{\min \left\{ \{w_i\}_i, \text{TV}(F_i, F_j)_j \right\}}$$

$\in [0, 1] \checkmark$

Any algo requires  $\Omega(\kappa(F))$  samples.

Main Result (Algo 3) (2010)

$$\forall \kappa \geq 1, \exists C_\kappa = C(\kappa) \text{ s.t.}$$

$$\forall d \text{ GMM } F, \delta > 0 \quad n > \left( \frac{\kappa(F) \cdot d}{\varepsilon \delta} \right)^{C_\kappa}$$

the estim. algo 3 finds  $\hat{F}$  s.t.

$$\text{w.p.} \geq 1 - \delta$$

$$\hat{\kappa} = \kappa \quad \exists n \in S_n$$

$$|w_i - \hat{w}_{\pi(i)}| \leq \varepsilon \quad \forall i \in [k]$$

$$\text{TV}(F_i, \hat{F}_{\pi(i)}) \leq \varepsilon$$

Runtime:  $\text{poly}(n)$

Pearson was right!

$$d=1, \kappa \rightarrow \boxed{4\kappa-2}$$

↑↑↑↑↑↑

(Kalai, Moitra, Valiant)  
STOC 2010

Input:  $\{x_i\} \sim 1d \text{ GMM}_\kappa, \varepsilon$

compute  $4\kappa-2$  for  $(w_i, \mu_i, \sigma_i^2)$ :

grid  
search

$$w_i \in [0, 1]$$

$$k_i \in [\min x_j, \max x_j]$$

$$\sigma_i \in [0, \max x_j]$$

$$\text{grid} \in \underline{G_k}$$

Return param set where first  $4k-2$  ...

### Part 3: Robust Identifiability (Finite sample)

Thm (Wanted)

$F, F'$  GMM  $< k$  comp.

with  $k(F), k(F') < \frac{1}{\varepsilon}$

$$\text{If } \forall i \leq \boxed{4k-2} : \underbrace{|M_i(F) - M_i(F')| \leq \varepsilon^{C_k}}_{\Rightarrow}$$

$\Rightarrow$  (1)  $F, F'$  same  $\hat{k}$

(2) param gap  $\leq \varepsilon$  ✓

$$\|\mu - \mu'\|_2, \|\Sigma - \Sigma'\|_F$$


TV is too  
strong  
sometimes

$$\hat{\varepsilon}_n, \hat{\tau}_n$$

$$TV\left(\frac{S_n}{\sqrt{n}}, N\right) = 1 \quad \forall n$$

### Observations:

1. If two distr (bounded)

have  $\downarrow$  TV   
 $f, g$

$\Rightarrow$  Low-order  
moments  
must be  
close.

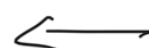
$$\left. \begin{array}{l} \cancel{h} = f - g \\ \rightarrow h \end{array} \right\} \begin{array}{l} \text{oscillating wave} \\ 0 \end{array}$$

$4k-2$

2. Dream 1: Algo 2

Low order  
moment  
"same"

$\Rightarrow$  params  
"same"



Dream 2: Generally:

params  
same,

TV small

(Algo 3)

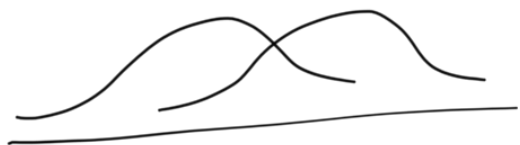
## Deconvolution & Moments

# zero crossings

dictates

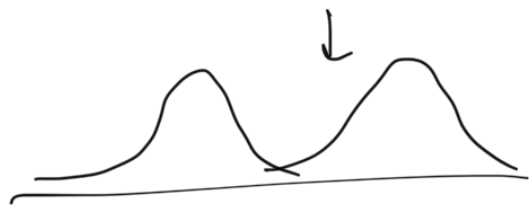
# moments required.

$$GMM_k: f_1, \dots, f_k$$



$\alpha$ -deconvolve

$$N(\mu, \sigma^2) \rightarrow N(\mu, \sigma^2 - \alpha)$$



Convolution

$$f * g$$

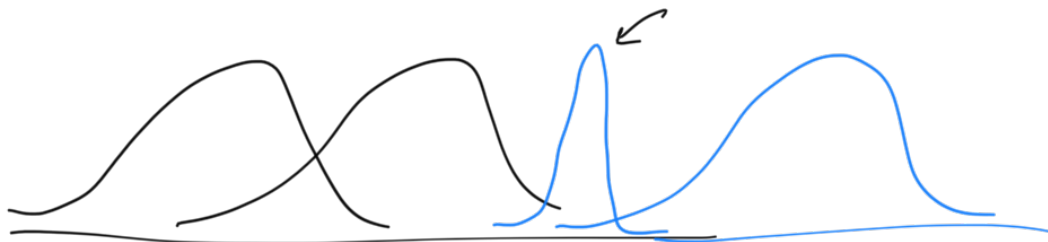
$$\xrightarrow{F}$$

$$\hat{f} \cdot \hat{g}$$

heat equation

Deconvolution

is heat equation reversed.



$$1 \text{ or } F F'$$

$\epsilon$ -standard

"param gap"

Sketch

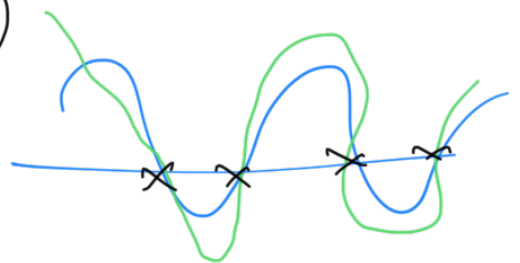
1. Let  $F, F'$   $TV(F, F') > \epsilon$

$$f(x) = \underline{F(x) - F'(x)}$$

$$\int |f(x)| > 0$$

2.  $f$  has at most  $\boxed{6} \cdot \boxed{4k-2}$   
crossings  $(\tau_{\text{hm}})$

3.  $\exists$  poly deg  $\boxed{6} \cdot \boxed{4k-2}$   
 whose sign matches  $f$ .



$$0 < \left| \int \underline{f(x)} p(x) dx \right| = \sum c_i \underline{x_i}$$

$$= \left| \int \sum_{i=1}^6 c_i x^i f(x) dx \right|$$

$$\leq \sum_i |c_i| \left| \int x^i f(x) dx \right|$$

$$\leq \sum_i |c_i| \left| \underline{M_i(F) - M_i(F')} \right|$$

$\exists i$  that moments  
 have gap  $\nabla$

Steps for main thm  $\leftarrow$

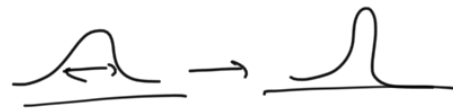
$F, F'$   
 $\epsilon$ -standard  $\rightarrow$  Moment gap



(param gap)  $\Rightarrow$  ...

$$\delta''_{\min} \sum_{i=1}^K \left\{ |w_i - w'_{nc(i)}| + \|\psi_i - \psi'_{nc(i)}\|_2 + \|\Sigma_i - \Sigma'_{nc(i)}\|_F \right\} \geq \varepsilon$$

(1) Strip away common parts.



(2) Deconvolve the min variance part

(3) This part  $\propto$  Dirac

$$\boxed{\mathcal{F}_\alpha}$$

$\Downarrow$   
generates a TV dist. gap in  $\mathcal{F}_\alpha$ -space

(4) TV gap  $\Rightarrow$  Low-order moment gap (in  $\mathcal{F}_\alpha$  space)

$$\exists i : M_i(\mathcal{F}_\alpha(\mathcal{F})) - M_i(\mathcal{F}_\alpha(\mathcal{F}')) \geq \varepsilon$$

(5) Magic Lemma:

Deconvolution preserves low-order moments.

"If, in  $\mathcal{F}_\alpha$  space, has gap

$\Rightarrow$  original space has gap

$\square$

Why?  $f(x): \mathbb{R} \rightarrow \mathbb{R}$

analytic & has  $n$  zeros

$$\Rightarrow g(x) = f(x) * \mathcal{N}(0, \sigma^2, x)$$

has at most  $n$  zeros

$$\sum_{i \in \mathcal{N}} |M_i(\mathcal{F}_\alpha(\mathcal{F})) - M_i(\mathcal{F}_\alpha(\mathcal{F}'))| < \varepsilon$$

$$\leq \frac{(r+1)!}{L^{\lfloor r/2 \rfloor!}} \left| \sum_{i=1}^r M_i(F) - M_i(F') \right|$$

Algo 2

high-dim

$d^2$   $\nearrow$   
2

Algo 3

param  
close

$\Rightarrow TV \ll \epsilon$



$TV = 1$   
measure 0

Lect #3:  $\uparrow \oplus$  Map  $\oplus$  Opt. Trans.

#4 : Representation theory

#5 : DP

#6 : DP2

#7 : LDP