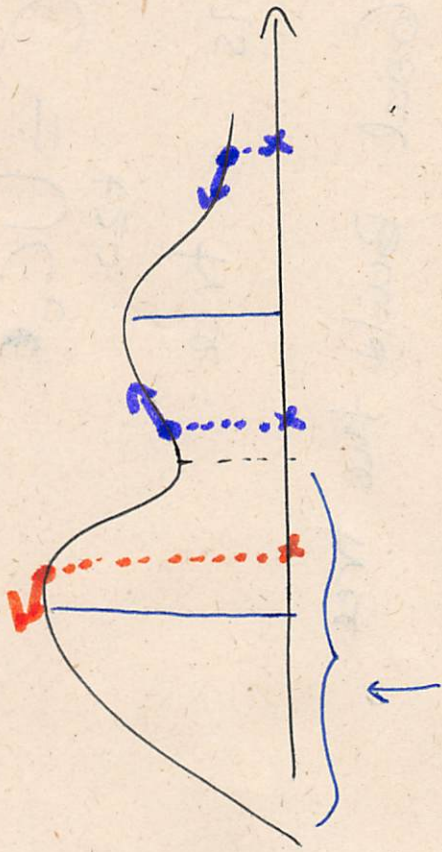


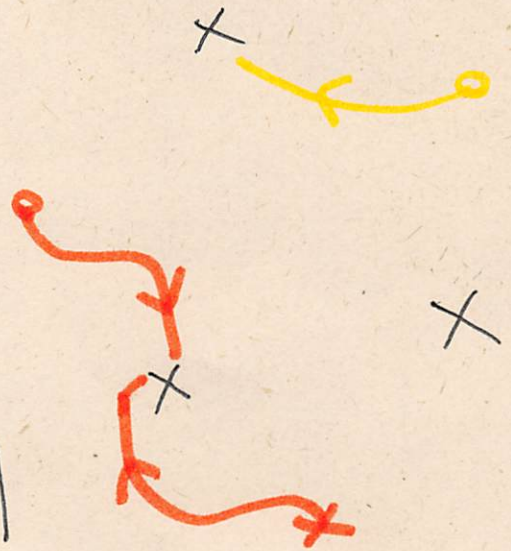
Mode = local max.

1D:



basin of attraction

2D:



Define Paths:

$$\pi_x(t)$$

$$\pi_x(0) = 0$$

$$\pi'_x(t) = \nabla_p(\pi_x(t))$$

$$\text{dest}(x) = \lim_{t \rightarrow \infty} \pi_x(t) \in \text{set of modes}$$

MS-Algorithm

① Pick any point $a = a^{(0)}$.

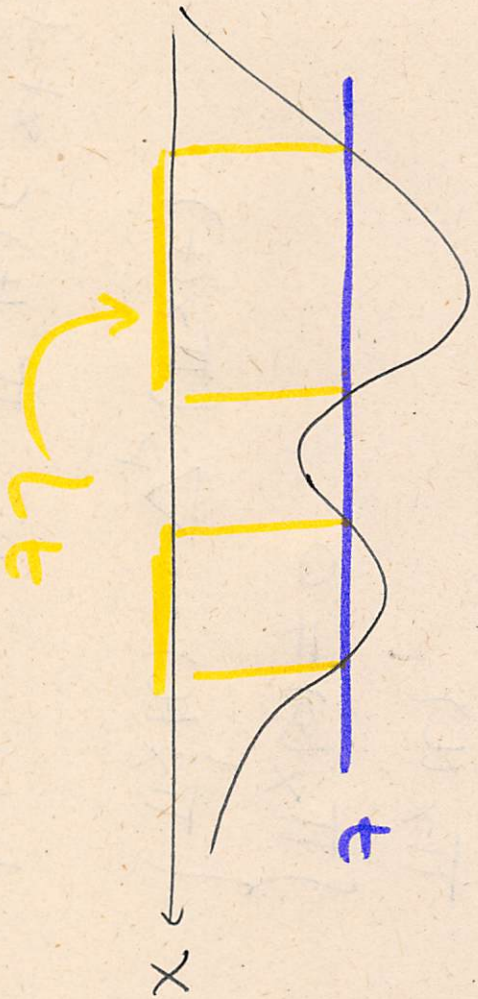
$$d^{(1)} \leftarrow \frac{\sum_{i=1}^n x_i k\left(\frac{x_i - a_k^{(0)}}{u}\right)}{\sum_{i=1}^n k\left(\frac{x_i - a_k^{(0)}}{u}\right)}$$

Density tree

$$C = \bigcup_{t \geq 0} C_t$$

is a tree.

Goal: Build the tree.



$$L_t := \{x : f(x) > t\} = \bigcup_i C_i$$

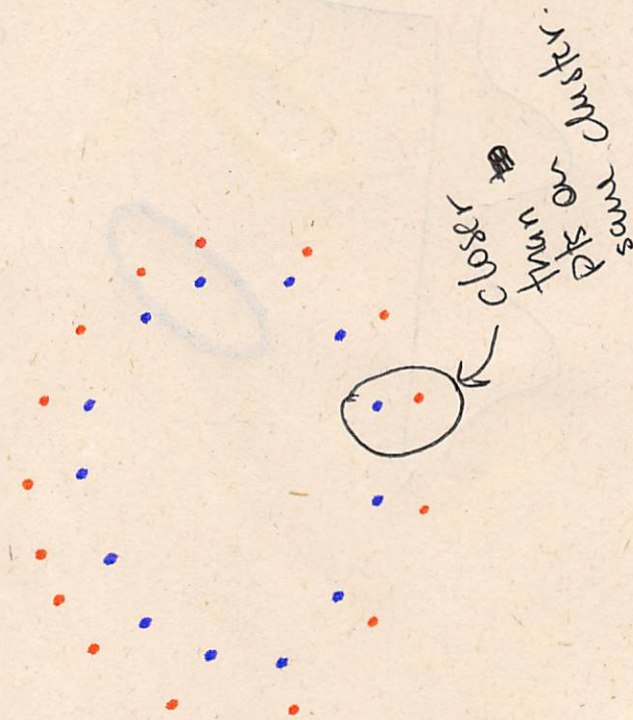
- High density region is an intuitive notion of cluster

$$X \sim \mathcal{N}(0, I); \quad X \in \mathbb{R}^d$$

Not hard to show that:

$$\|X\|^2 = d$$

i.e. points lie on rings.



One fix:
Modify the distance

$$\rho(x, y) := \frac{1}{n-2} \sum_{z \neq x, y} \left| \|x - z\| - \|y - z\| \right|$$

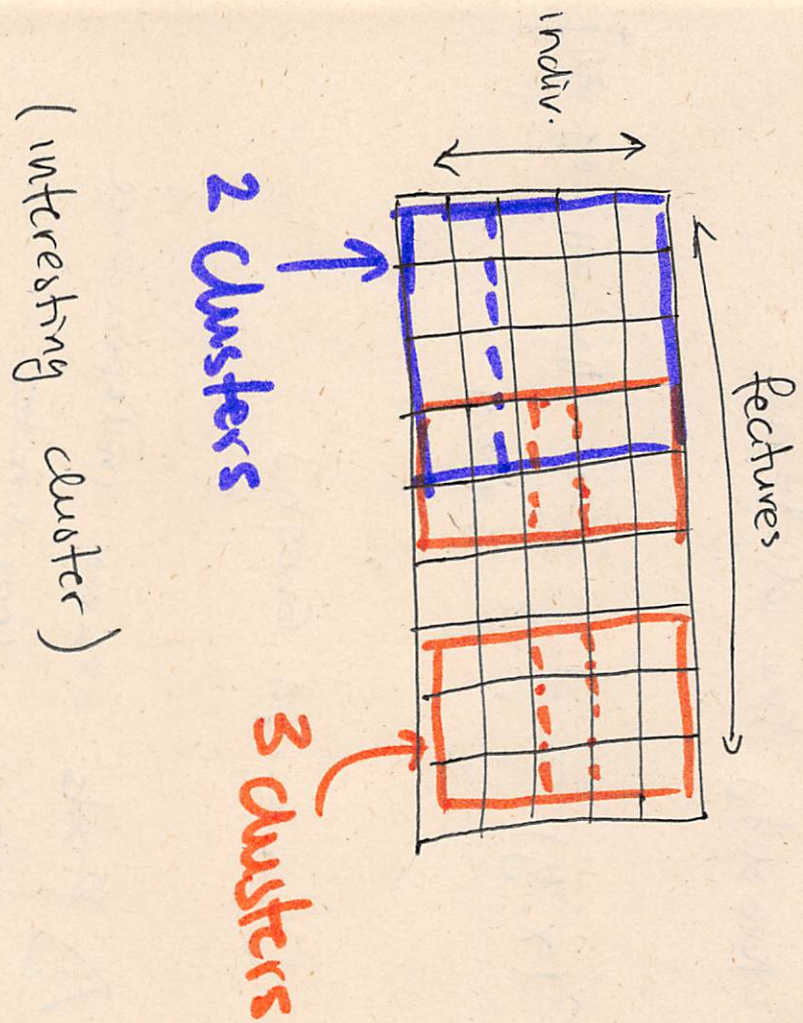
[Sarkar, & Ghosh]

⚠ Proofs assume independence between coordinates...

Open problem:

Adjust k-means to deal with high dim. effects

Find Features that define clusters.



(interesting cluster)

A First Idea:

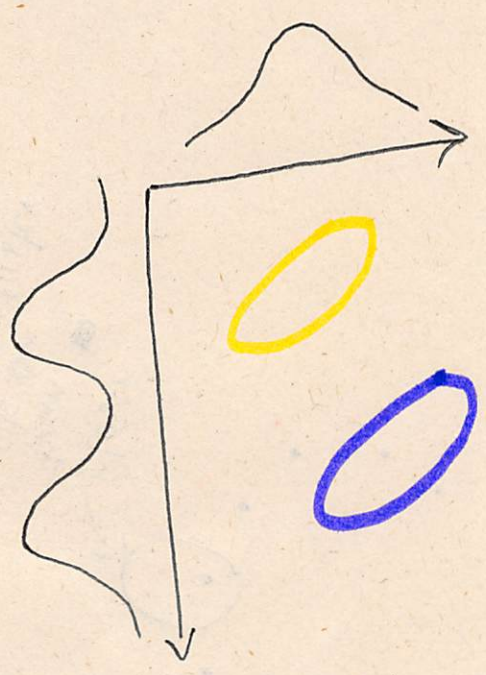
Screening: Are there marginal clusters?

$$X = (X[1], \dots, X[n])$$

$$X_1[1], \dots, X_m[n] \sim F$$

H_0 : F is unimodal.

H_1 : F is not.



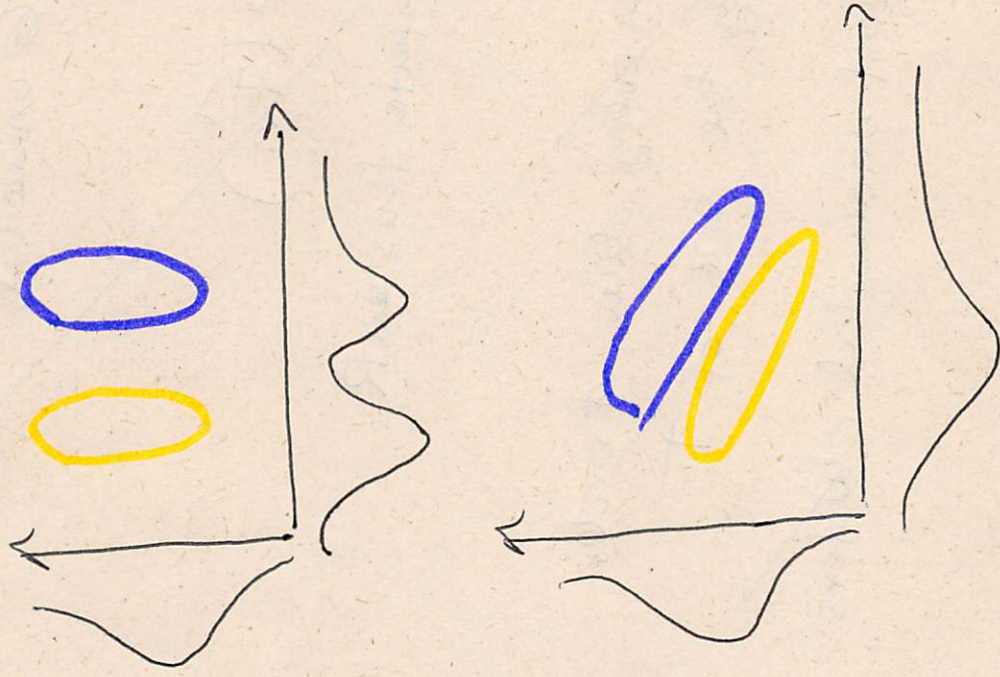
$$\text{DIP}(F) = \inf_{\text{G unimodal}} \sup_x |F(x) - G(x)|$$

$$T := \text{DIP}(\hat{F}_n)$$

↑
can compute this in \mathbb{R}

• Do this in every draw & order the T 's.

• Select least unimodal?



$$\text{DIP}(F) := \inf_{\mathcal{C} \text{ unimodal}} \sup_x |F(x) - G(x)|$$

Sparse Alternating Clustering

$$T = \text{DIP}(\hat{F}_n)$$

can compute this in R .

- Do this in every dim. & order them by the size of T .
- Select least unimodal clusters.

$$R_n = \frac{1}{n} \sum_{j=1}^n \min_c \|x_j - c\|^2$$

$$= \dots = \sum_{j=1}^K \frac{1}{|C_j|} \sum_{i,j} \|x_i - x_j\|^2$$

Choose $|S| = L$.

$a = 1, \dots, d$ index of features

$$\delta_a(i, j) := (x_i[a] - x_j[a])^2$$

Rescale so that $\sum_{i,j} \delta(i, j) = 1$

Do k-means with an alternative step:

- Find the centers.

- Find the features



minimize R_n to find the features.