

High-Dimensional Measures and Geometry

Lecture Notes from Jan 28, 2010

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3 Measure Concentration in boolean cubes

3.1.1 Definition. The boolean cube is the set $I_n = \{0, 1\}^n$ endowed with distance $d(x, y)$ between $x = (x_1, \dots, x_n)$ and $y = (y_1, \dots, y_n)$ given by

$$d(x, y) = \sum_{i=1}^n |x_i - y_i|$$

A natural probability measure on I_n is the counting probability measure μ_n defined by

$$\mu_n(\{x\}) = \frac{1}{2^n} \text{ for any } x \in I_n$$

Notation: We will denote $\frac{1}{2^n} \sum_{x \in I_n} f(x)$ by $\int_{I_n} f d\mu_n$ for any function f on I_n .

We will see that “almost all” points are “close” to any sufficiently large subset $A \subset I_n$. This will help us show concentration of measure on the cube. Again, we use the Laplace Transform.

3.1.2 Theorem. Let $A \subset I_n$ be a non-empty, let $f : I_n \rightarrow \mathbb{Z}^+$, $f(x) = d(x, A)$ be the distance from A . For any $t \geq 0$, we then have

$$\int_{I_n} e^{tf} d\mu_n \leq \frac{1}{\mu_n(A)} \left(\frac{1}{2} + \frac{e^t + e^{-t}}{4} \right)^n$$

We delay the proof of this theorem and first derive a consequence.

3.1.3 Corollary. For $\epsilon > 0$,

$$\mu_n \left(\{x \in I_n : d(x, A) \geq \epsilon \sqrt{(n)}\} \right) \leq e^{-\epsilon^2 / \mu_n(A)}.$$

Proof of Corollary. By Taylor expansion of both sides, we get

$$\frac{1}{2} + \frac{e^t + e^{-t}}{4} \leq e^{t^2/4}$$

Using Laplace Transform, we obtain

$$\begin{aligned} \mu_n \left(\{x \in I_n : d(x, A) \geq \epsilon\sqrt{(n)}\} \right) &\leq e^{-t\epsilon\sqrt{n}} \int_{I_n} e^{td(x,A)} d\mu_n \\ &\leq e^{-t\epsilon\sqrt{n}} \frac{1}{\mu_n(A)} \left(\frac{1}{2} + \frac{e^t + e^{-t}}{4} \right)^n \\ &\text{[By preceding Theorem]} \end{aligned}$$

Now, using inequality from the Taylor expansion, we get

$$\mu_n \left(\{x \in I_n : d(x, A) \geq \epsilon\sqrt{(n)}\} \right) \leq e^{-t\epsilon\sqrt{n}} \frac{1}{\mu_n(A)} e^{t^2 n/4}$$

Now choosing $t = 2\epsilon/\sqrt{n}$, we get

$$\mu_n \left(\{x \in I_n : d(x, A) \geq \epsilon\sqrt{(n)}\} \right) \leq e^{-\epsilon^2} \frac{1}{\mu_n(A)} e^{\epsilon^2} \leq \frac{e^{-\epsilon^2}}{\mu_n(A)}$$

□

Proof of Theorem. We abbreviate

$$\left(\frac{1}{2} + \frac{e^t + e^{-t}}{4} \right) = c(t).$$

We use induction over n to prove this theorem. For $n = 1$, we have either $|A| = 1$ or $|A| = 2$. When $|A| = 1$, we have $\mu_1(A) = 1/2$ and

$$\begin{aligned} \int_{I_1} e^{tf} d\mu_1 &= \frac{1}{2} + \frac{e^t}{2} \\ &\leq 2c(t) \\ &\text{[since } \mu_1(A) = 1/2\text{]} \end{aligned}$$

When $|A| = 2$, we have

$$\begin{aligned} \int_{I_1} e^{tf} d\mu_1 &= \frac{1}{2} + \frac{1}{2} \\ &= 1 \leq c(t) \\ &\text{[since } c(t) \geq 1\text{]} \end{aligned}$$

Now, assume that the result is true on I_{n-1} . Consider

$$I_n^1 = \{x \in I_n : x_n = 1\} \text{ and } I_n^0 = \{x \in I_n : x_n = 0\}.$$

Also, we let

$$A_0 = \{x \in I_n^0 : x \in A\} \text{ and } A_1 = \{x \in I_n^1 : x \in A\}.$$

Define μ_{n-1} on I_n^1 and I_n^0 . Then, by definition, we have

$$\mu_n(A) = \frac{\mu_{n-1}(A_1) + \mu_{n-1}(A_0)}{2}.$$

Now, we pick $x \in I_n$ and let $x' = (x_1, \dots, x_{n-1}) \in I_{n-1}$. Since $x \in I_n^1$ or $x \in I_n^0$, we have

$$d(x, A) = \min\{d(x', A_1), d(x', A_0) + 1\} \text{ if } x \in I_n^1$$

or

$$d(x, A) = \min\{d(x', A_0), d(x', A_1) + 1\} \text{ if } x \in I_n^0,$$

because a closest point is in $A_1 \subset I_n^1$ or in $A_0 \subset I_n^0$.

Define

$$f_0(x) = d(x, A_0) \text{ and } f_1(x) = d(x, A_1).$$

Now, we have

$$\begin{aligned} \int_{I_n} e^{tf} d\mu_n &= \int_{I_n^1} e^{tf} d\mu_n + \int_{I_n^0} e^{tf} d\mu_n \\ &= \int_{I_n^1} e^{t \min\{d(x', A_1), d(x', A_0) + 1\}} d\mu_n(x') + \int_{I_n^0} e^{t \min\{d(x', A_0), d(x', A_1) + 1\}} d\mu_n(x') \\ &= \int_{I_n^1} \min\{e^{td(x', A_1)}, e^t e^{d(x', A_0)}\} d\mu_n(x') + \int_{I_n^0} \min\{e^{td(x', A_0)}, e^t e^{d(x', A_1)}\} d\mu_n(x') \\ &\leq \min \left\{ \int_{I_n^1} e^{td(x', A_1)} d\mu_n(x'), \int_{I_n^1} e^t e^{d(x', A_0)} d\mu_n(x') \right\} \\ &\quad + \min \left\{ \int_{I_n^0} e^{td(x', A_0)} d\mu_n(x'), \int_{I_n^0} e^t e^{d(x', A_1)} d\mu_n(x') \right\} \\ &= \min \left\{ \frac{1}{2} \int_{I_n^1} e^{td(x', A_1)} d\mu_{n-1}(x'), \frac{1}{2} \int_{I_n^1} e^t e^{d(x', A_0)} d\mu_{n-1}(x') \right\} \\ &\quad + \min \left\{ \frac{1}{2} \int_{I_n^0} e^{td(x', A_0)} d\mu_{n-1}(x'), \frac{1}{2} \int_{I_n^0} e^t e^{d(x', A_1)} d\mu_{n-1}(x') \right\} \end{aligned}$$

[Using induction assumption, we have]

$$\begin{aligned} &\leq \frac{1}{2} \min \left\{ \frac{c^{n-1}(t)}{\mu_{n-1}(A_1)}, e^t \frac{c^{n-1}(t)}{\mu_{n-1}(A_0)} \right\} + \frac{1}{2} \min \left\{ \frac{c^{n-1}(t)}{\mu_{n-1}(A_0)}, e^t \frac{c^{n-1}(t)}{\mu_{n-1}(A_1)} \right\} \\ &= \frac{1}{2} \frac{c^{n-1}(t)}{\mu_n(A)} \left[\min \left\{ \frac{\mu_n(A)}{\mu_{n-1}(A_1)}, \frac{e^t \mu_n(A)}{\mu_{n-1}(A_0)} \right\} + \min \left\{ \frac{\mu_n(A)}{\mu_{n-1}(A_0)}, \frac{e^t \mu_n(A)}{\mu_{n-1}(A_1)} \right\} \right] \quad (*) \end{aligned}$$

We abbreviate

$$a_0 = \frac{\mu_{n-1}(A_0)}{\mu_n(A)} \text{ and } a_1 = \frac{\mu_{n-1}(A_1)}{\mu_n(A)}.$$

So, we have $a_0 + a_1 = 2$. The R. H. S. of the inequality (*) is maximal if we choose a_0, a_1 to minimize the expression $\min\{a_1^{-1}, e^t a_0^{-1}\} + \min\{a_0^{-1}, e^t a_1^{-1}\}$.

WLOG, we may assume $a_0 \geq a_1$. Now, at the end points, we have

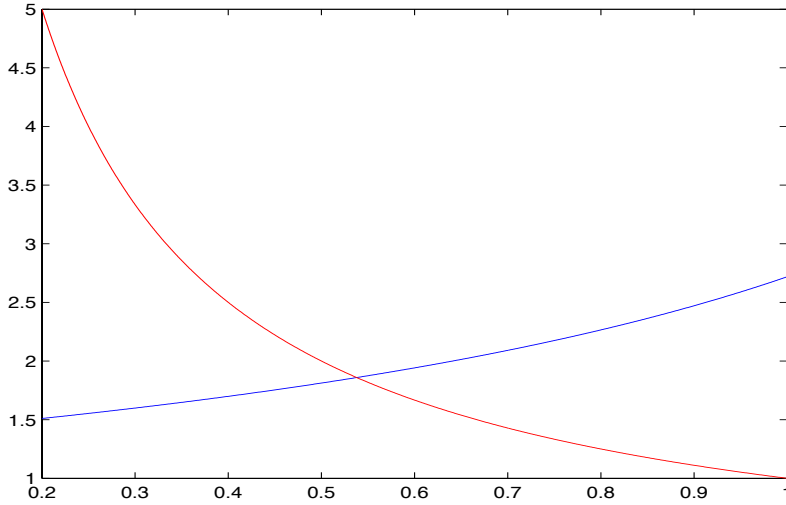
1. $a_0 = 2, a_1 = 0$ and therefore

$$\text{R. H. S.} \leq \frac{1}{2} \frac{c^{n-1}(t)}{\mu_n(A)} \left(\frac{e^t}{2} + \frac{1}{2} \right) = \frac{c^{n-1}(t)}{\mu_n(A)} \left(\frac{1}{4} + \frac{e^t}{4} \right)$$

2. $a_0 = 1 = a_1$ and then

$$\text{R. H. S.} \leq \frac{1}{2} \frac{c^{n-1}(t)}{\mu_n(A)} (\min\{1, e^t\} + \min\{1, e^t\}) = \frac{c^{n-1}(t)}{\mu_n(A)} \leq \frac{c^n(t)}{\mu_n(A)}$$

Now, when $a_0 > a_1$, $a_0^{-1} < a_1^{-1} < e^t a_1^{-1}$. So, the minimum is relevant in the first term only. The following is the sketch of the graph of a_1^{-1} [in red] and $e^t(2 - a_1)^{-1}$ [in blue, for $t = 1$] for $a_1 \in [0, 1]$.



One is increasing and the other is decreasing. Therefore, $\min\{a_1^{-1}, e^t a_0^{-1}\}$ is maximized when $a_1^{-1} = e^t(2 - a_1^{-1}) \Rightarrow \frac{2}{a_1} - 1 = e^t \Rightarrow a_1 = \frac{2}{1+e^t}$ and $a_0 = \frac{2e^t}{1+e^t}$. Inserting these gives us the maximum of $\min\{a_1^{-1}, e^t a_0^{-1}\} + \min\{a_0^{-1}, e^t a_1^{-1}\}$

$$= \frac{1 + e^t}{2} + \frac{1 + e^t}{2e^t} = 1 + \frac{1}{2}(e^t + e^{-t}) = 2c(t).$$

Therefore, from (*), we get

$$\int_{I_n} e^{tf} d\mu_n \leq \frac{1}{\mu_n(A)} \left(\frac{1}{2} + \frac{e^t + e^{-t}}{4} \right)^n.$$

□