

Lemma 33.1. For $0 \leq r \leq 1$,

$$\inf_{0 \leq \lambda \leq 1} e^{\frac{1}{4}(1-\lambda)^2} r^{-\lambda} \leq 2 - r.$$

Proof. Taking \log , we need to show

$$\inf_{0 \leq \lambda \leq 1} \left(\frac{1}{4}(1-\lambda)^2 - \lambda \log r - \log(2-r) \right) \leq 0.$$

Taking derivative with respect to λ ,

$$-\frac{1}{2}(1-\lambda) - \log r = 0$$

$$\lambda = 1 + 2 \log r \leq 1$$

$$0 \leq \lambda = 1 + 2 \log r$$

Hence,

$$e^{-1/2} \leq r.$$

Take

$$\lambda = \begin{cases} 1 + 2 \log r & e^{-1/2} \leq r \\ 0 & e^{-1/2} \geq r \end{cases}$$

Case a): $r \leq e^{-1/2}$, $\lambda = 0$

$$\frac{1}{4} - \log(2-r) \leq 0 \Leftrightarrow r \leq 2 - e^{\frac{1}{4}}. \quad e^{-1/2} \leq 2 - e^{\frac{1}{4}}.$$

Case a): $r \geq e^{-1/2}$, $\lambda = 1 + 2 \log r$

$$(\log r)^2 - \log r - 2(\log r)^2 - \log(2-r) \leq 0$$

Let

$$f(r) = \log(2-r) + \log r + (\log r)^2.$$

Is $f(r) \geq 0$? Enough to prove $f'(r) \leq 0$. Is

$$f'(r) = -\frac{1}{2-r} + \frac{1}{r} + 2 \log r \cdot \frac{1}{r} \leq 0.$$

$$r f'(r) = -\frac{r}{2-r} + 1 + 2 \log r \leq 0.$$

Enough to show $(r f'(r))' \geq 0$:

$$(r f'(r))' = \frac{2}{r} - \frac{2-r+r}{(2-r)^2} = \frac{2}{r} - \frac{2}{(2-r)^2}.$$

□

Let \mathcal{X} be a set (space of examples) and P a probability measure on \mathcal{X} . Let x_1, \dots, x_n be i.i.d., $(x_1, \dots, x_n) \in \mathcal{X}^n$, $P^n = P \times \dots \times P$.

Consider a subset $A \in \mathcal{X}^n$. How can we define a distance from $x \in \mathcal{X}^n$ to A ? Example: hamming distance between two points $d(x, y) = \sum I(x_i \neq y_i)$.

We now define *convex hull distance*.

Definition 33.1. Define $V(A, x)$, $U(A, x)$, and $d(A, x)$ as follows:

$$(1) V(A, x) = \{(s_1, \dots, s_n) : s_i \in \{0, 1\}, \exists y \in A \text{ s.t. if } s_i = 0 \text{ then } x_i = y_i\}$$

$$\begin{aligned} x &= (x_1, x_2, \dots, x_n) \\ &= \neq \dots = \\ y &= (y_1, y_2, \dots, y_n) \\ s &= (0, 1, \dots, 0) \end{aligned}$$

Note that it can happen that $x_i = y_i$ but $s_i \neq 0$.

$$(2) U(A, x) = \text{conv } V(A, x) = \{\sum \lambda_i u^i, u^i = (u_1^i, \dots, u_n^i) \in V(A, x), \lambda_i \geq 0, \sum \lambda_i = 1\}$$

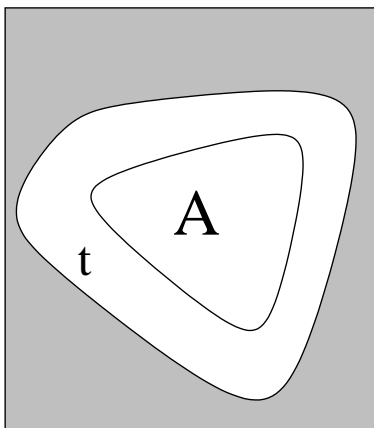
$$(3) d(A, x) = \min_{u \in U(A, x)} |u|^2 = \min_{u \in U(A, x)} \sum u_i^2$$

Theorem 33.1.

$$\mathbb{E} e^{\frac{1}{4}d(A, x)} = \int e^{\frac{1}{4}d(A, x)} dP^n(x) \leq \frac{1}{P^n(A)}$$

and

$$P^n(d(A, x) \geq t) \leq \frac{1}{P^n(A)} e^{-t/4}.$$



Proof. Proof is by induction on n .

n = 1 :

$$d(A, x) = \begin{cases} 0, & x \in A \\ 1, & x \notin A \end{cases}$$

Hence,

$$\int e^{\frac{1}{4}d(A,x)} dP^n(x) = P(A) \cdot 1 + (1 - P(A))e^{\frac{1}{4}} \leq \frac{1}{P(A)}$$

because

$$e^{\frac{1}{4}} \leq \frac{1 + P(A)}{P(A)}.$$

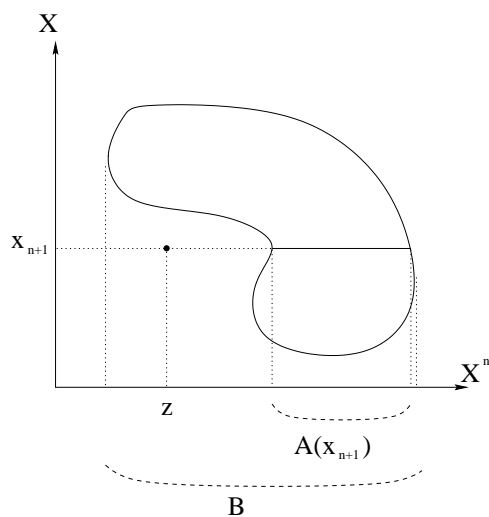
$\mathbf{n} \rightarrow \mathbf{n} + 1$:

Let $x = (x_1, \dots, x_n, x_{n+1}) = (z, x_{n+1})$. Define

$$A(x_{n+1}) = \{(y_1, \dots, y_n) : (y_1, \dots, y_n, x_{n+1}) \in A\}$$

and

$$B = \{(y_1, \dots, y_n) : \exists y_{n+1}, (y_1, \dots, y_n, y_{n+1}) \in A\}$$



One can verify that

$$s \in U(A(x_{n+1}, z)) \Rightarrow (s, 0) \in U(A, (z, x_{n+1}))$$

and

$$t \in U(B, z) \Rightarrow (t, 1) \in U(A, (z, x_{n+1})).$$

Take $0 \leq \lambda \leq 1$. Then

$$\lambda(s, 0) + (1 - \lambda)(t, 1) \in U(A, (z, x_{n+1}))$$

since $U(A, (z, x_{n+1}))$ is convex. Hence,

$$\begin{aligned} d(A, (z, x_{n+1})) &= d(A, x) \leq |\lambda(s, 0) + (1 - \lambda)(t, 1)|^2 \\ &= \sum_{i=1}^n (\lambda s_i + (1 - \lambda)t_i)^2 + (1 - \lambda)^2 \\ &\leq \lambda \sum s_i^2 + (1 - \lambda) \sum t_i^2 + (1 - \lambda)^2 \end{aligned}$$

So,

$$d(A, x) \leq \lambda d(A(x_{n+1}), z) + (1 - \lambda)d(B, z) + (1 - \lambda)^2.$$

Now we can use induction:

$$\int e^{\frac{1}{4}d(A, x)} dP^{n+1}(x) = \int_{\mathcal{X}} \int_{\mathcal{X}^n} e^{\frac{1}{4}d(A, (z, x_{n+1}))} dP^n(z) dP(x_{n+1}).$$

Then inner integral is

$$\begin{aligned} \int_{\mathcal{X}^n} e^{\frac{1}{4}d(A, (z, x_{n+1}))} dP^n(z) &\leq \int_{\mathcal{X}^n} e^{\frac{1}{4}(\lambda d(A(x_{n+1}), z) + (1 - \lambda)d(B, z) + (1 - \lambda)^2)} dP^n(z) \\ &= e^{\frac{1}{4}(1 - \lambda)^2} \int e^{(\frac{1}{4}d(A(x_{n+1}), z))\lambda + (\frac{1}{4}d(B, z))(1 - \lambda)} dP^n(z) \end{aligned}$$

We now use Hölder's inequality:

$$\begin{aligned} \int fg dP &\leq \left(\int f^p dP \right)^{1/p} \left(\int g^q dP \right)^{1/q} \quad \text{where } \frac{1}{p} + \frac{1}{q} = 1 \\ e^{\frac{1}{4}(1 - \lambda)^2} \int e^{(\frac{1}{4}d(A(x_{n+1}), z))\lambda + (\frac{1}{4}d(B, z))(1 - \lambda)} dP^n(z) \\ &\leq e^{\frac{1}{4}(1 - \lambda)^2} \left(\int e^{\frac{1}{4}d(A(x_{n+1}), z)} dP^n(z) \right)^\lambda \left(\int e^{\frac{1}{4}d(B, z)} dP^n(z) \right)^{1 - \lambda} \\ &\leq (\text{by ind. hypoth.}) e^{\frac{1}{4}(1 - \lambda)^2} \left(\frac{1}{P^n(A(x_{n+1}))} \right)^\lambda \left(\frac{1}{P^n(B)} \right)^{1 - \lambda} \\ &= \frac{1}{P^n(B)} e^{\frac{1}{4}(1 - \lambda)^2} \left(\frac{P^n(A(x_{n+1}))}{P^n(B)} \right)^{-\lambda} \end{aligned}$$

Optimizing over $\lambda \in [0, 1]$, we use the Lemma proved in the beginning of the lecture with

$$0 \leq r = \frac{P^n(A(x_{n+1}))}{P^n(B)} \leq 1.$$

Thus,

$$\frac{1}{P^n(B)} e^{\frac{1}{4}(1 - \lambda)^2} \left(\frac{P^n(A(x_{n+1}))}{P^n(B)} \right)^{-\lambda} \leq \frac{1}{P^n(B)} \left(2 - \frac{P^n(A(x_{n+1}))}{P^n(B)} \right).$$

Now, integrate over the last coordinate. When averaging over x_{n+1} , we get measure of A .

$$\begin{aligned}
 \int e^{\frac{1}{4}d(A,x)} dP^{n+1}(x) &= \int_{\mathcal{X}} \int_{\mathcal{X}^n} e^{\frac{1}{4}d(A,(z,x_{n+1}))} dP^n(z) dP(x_{n+1}) \\
 &\leq \int_{\mathcal{X}} \frac{1}{P^n(B)} \left(2 - \frac{P^n(A(x_{n+1}))}{P^n(B)} \right) dP(x_{n+1}) \\
 &= \frac{1}{P^n(B)} \left(2 - \frac{P^{n+1}(A)}{P^n(B)} \right) \\
 &= \frac{1}{P^{n+1}(A)} \frac{P^{n+1}(A)}{P^n(B)} \left(2 - \frac{P^{n+1}(A)}{P^n(B)} \right) \\
 &\leq \frac{1}{P^{n+1}(A)}
 \end{aligned}$$

because $x(2-x) \leq 1$ for $0 \leq x \leq 1$.

□