

Growth function: For fixed $x_1, \dots, x_n \in \mathcal{X}$, and a collection of subsets of $\mathcal{X}: \mathcal{D}$, we define

$$\Delta^{\mathcal{D}}(x_1, \dots, x_n) = |\mathcal{D} \cap \{x_1, \dots, x_n\}|$$

Example If $\mathcal{X} = \mathbb{R}$, $\mathcal{D} = \{(-\infty, t] : t \in \mathbb{R}\}$, then $\Delta^{\mathcal{D}}(x_1, \dots, x_n) \leq n+1$, $\forall \{x_1, \dots, x_n\} \subset \mathbb{R}$

Connection between $\Delta^{\mathcal{D}}(x_1, \dots, x_n)$ and covering numbers.

For a collection \mathcal{D} and $x_1, \dots, x_n \stackrel{iid}{\sim} (\mathcal{X}, \Sigma, P)$, define $\mathcal{J}_0 = \{\mathbf{1}_{\mathcal{D}} : D \in \mathcal{D}\}$ and $N := N(\epsilon, \mathcal{J}_0, d_n^{(\infty)})$, where $d_n^{(\infty)}(f, g) = \max_i |f(x_i) - g(x_i)|$, then $\Delta^{\mathcal{D}}(x_1, \dots, x_n) = N$, $\forall 0 < \epsilon < 1$

Theorem $\frac{1}{n} \log \Delta^{\mathcal{D}}(x_1, \dots, x_n) = o_p(1) \rightarrow \mathcal{J}_0$ is P -GC

Pf: It's sufficient to prove

$$N(\epsilon, \mathcal{J}_0, d_n^{(\infty)}) \geq N(\epsilon, \mathcal{J}_0, d_n^{(1)})$$

Since $d_n^{(1)}(f, g) = \frac{1}{n} \sum_{i=1}^n |f(x_i) - g(x_i)| \leq d_n^{(\infty)}(f, g)$

$\forall f \in \bar{B}^{(\infty)}(g, \epsilon)$, $f \in \bar{B}^{(1)}(g, \epsilon)$

$$\Rightarrow \bar{B}^{(\infty)}(g, \epsilon) \subset \bar{B}^{(1)}(g, \epsilon) \quad \square$$

Def $\Delta^{\mathcal{D}}(n) = \sup_{x_1, x_2, \dots, x_n \in \mathcal{X}} \Delta^{\mathcal{D}}(x_1, \dots, x_n)$, if \exists

$\forall \delta > 0$ s.t. $\Delta^{\mathcal{D}}(n) \leq n^\nu$, we call \mathcal{D} is Vapnik-Chervonenkis class.

Remark: If \mathcal{D} is VC-class, $\mathcal{J}_0 = \{\mathbf{1}_{\mathcal{D}} : D \in \mathcal{D}\}$ is P -GC class, \forall law P . We call the set \mathcal{J}_0 is a

uniform GL class.

Example If $D = \{(-\infty, t], t \in \mathbb{R}\}$, D is VC class.

And the instant result is that

$$\sup_{t \in \mathbb{R}} |\hat{F}_n(t) - F(t)| = o_p(1)$$

Example $D = \{\text{all convex set of } \mathbb{R}^2\}$ is not.

Def VC dimension

$V(D) = \max \{n, \Delta^D(n) = 2^n\}$, which is a break point s.t. $\forall D \in \mathcal{D}$, if $|D| \leq V(D)$, D is shattered.

Remark: If $V(D) < \infty$, $\sum_{k=0}^{V(D)} \binom{n}{k}$ is the number of the subset $A \subset \{x_1, \dots, x_n\}$ which is shattered.

Theorem Sauer's Lemma

D is VC-class iff $V(D) < \infty$

Pf: " \Rightarrow " If $V(D) = \infty$, which means that

$\Delta^D(n) = 2^n, \forall n$, contradiction.

" \Leftarrow " Supposed that $\Delta^D(n) \leq \sum_{k=0}^{V(D)} \binom{n}{k}, n > V(D)$

$\forall \{x_1, \dots, x_n, x_{n+1}\} \subset \mathcal{X}$, define

$D_1 = \{D \in \mathcal{D}; x_{n+1} \in D\}, D_2 = D / D_1$. then

$$\Delta^D(x_1, \dots, x_{n+1}) = |\{x_1, \dots, x_n, x_{n+1}\} \cap D|$$

$$\leq |\{x_1, \dots, x_{n+1}\} \cap D_1| + |\{x_1, \dots, x_{n+1}\} \cap D_2|$$

$$= |\{x_1, \dots, x_n\} \cap (D_1/x_{n+1})| + |\{x_1, \dots, x_n\} \cap D_2|$$

$$\leq \sum_{k=0}^{\text{VC}(D_1)} \binom{n}{k} + \sum_{k=0}^{\text{VC}(D_2)} \binom{n}{k} \quad \textcircled{1}$$

We claim that: $\leq \sum_{k=0}^{\text{VC}(D)} \binom{n+1}{k}$

Notice that $D = |A_1| + |A_2|$, $A_i = \{L \subset \{x_1, \dots, x_n\}; L$
is shattered by $D_i\}$, $A = \{L \subset \{x_1, \dots, x_{n+1}\}; L$ is shattered by $D\}$

If $L \in A_1 \Delta A_2$, $L \in A$

If $L \in A_1 \cap A_2$, $L \in A$, but L is double counted.

Let $L_1 = L \cup \{x_{n+1}\}$, then $L_1 \notin A_1 \cup A_2$, but

$L_1 \in A$ since that: L is shattered by both D_1 and D_2 .

$\Rightarrow \forall \bar{L} \subset L, \exists B_i \in D_i$ s.t.

$$\bar{L} = L \cap B_i$$

then $\forall \bar{L} \subset L_1, \exists \bar{L} \subset L$ s.t. $\bar{L}_1 = \{x_{n+1}\} \cup \bar{L}$
or \bar{L}

$$\Rightarrow \bar{L}_1 = \bar{L} = L_1 \cap B_2 \text{ or}$$

$$= \{x_{n+1}\} \cup \bar{L} = L_1 \cap B_1. \quad \square$$

Remark: Let $V = \text{VC}(D)$

$$\sum_{k=0}^V \binom{n}{k} \leq \sum_{k=0}^V \binom{n}{k} \left(\frac{n}{V}\right)^k / \left(\frac{n}{V}\right)^V$$

$$\leq \left(\frac{n}{V} + 1\right)^V / \left(\frac{n}{V}\right)^V$$

$$= \left(1 + \frac{n}{V}\right)^V$$

$$\leq \left(\frac{en}{V}\right)^V$$

VC stability

D_1 is a collection of X_1 , if $\phi: D_1 \rightarrow D_2$ is a surjection, then $\Delta^{D_1}(n) \geq \Delta^{D_2}(n)$. $V(D_1) \geq V(D_2)$.

Property

- ① D is VC class $\Leftrightarrow D^c$ is VC class.
- ② $\Delta^{D_1 \cup D_2}(n) \leq \Delta^{D_1}(n) \Delta^{D_2}(n)$, if $X_1 = X_2$
- ③ $\Delta^{D_1 \times D_2}(n) \leq \Delta^{D_1}(n) \Delta^{D_2}(n)$

Example $D = \{(-\infty, t], t \in \mathbb{R}^d\}$, then D is VC class and $\Delta^D(n) \leq (n+1)^d$.

Uniform bound for covering number

We focus on a VC class: D and $\mathcal{F} = \{f_D; D \in D\}$, $d^{(2)}(f, g) = (\mathbb{P}|f - g|^2)^{\frac{1}{2}}$
 $d_n^{(2)}(f, g) = (\mathbb{P}_n|f - g|^2)^{\frac{1}{2}}$. Note that:
 $N(\varepsilon, \mathcal{F}, d^{(2)}) \leq N(\varepsilon, \mathcal{F}, d^{(1)})$
Let $v = VCD < \infty$, $0 \leq \varepsilon \leq 1$.

We assume that \mathcal{F} is totally bounded, $\{f_i\}$ is the maximal ε -packing, i.e.
 $d(f_i, f_j) > \varepsilon$,
and $N := |\{f_i\}| \geq N(\varepsilon; \mathcal{F}, d^{(2)})$

Lemma $\exists \varepsilon_0 < \varepsilon$, if n is large enough,

$$P(\{i \neq j \mid d_n^{(2)}(f_i, f_j) > \varepsilon_0\}) > 0$$

Pf: Define $h(x) = (f_i(x) - f_j(x))^2$, for fixed i, j

$$\Rightarrow P(|(d_n^{(2)}(f_i, f_j))^2 - (d^{(2)}(f_i, f_j))^2| > \varepsilon_0) \forall \varepsilon_0 > 0$$

$$= P\left|\frac{1}{n} \sum_{i=1}^n h(x_i) - \mathbb{E} h(x_i)\right| > \varepsilon_0$$

$$= P(|\bar{h}(x_i) - \mathbb{E} h(x_i)| > n\varepsilon_0) \quad ①$$

Note that $|\bar{h}(x_i) - \mathbb{E} h(x_i)|$ is bounded.

$$\Rightarrow ① \leq 2e^{-c\frac{n^2\varepsilon_0^2}{n}} = 2e^{-cn\varepsilon_0^2}$$

$$\Rightarrow P(|(d_n^{(2)}(f_i, f_j))^2 - (d^{(2)}(f_i, f_j))^2| \leq \varepsilon_0) \geq 1 - 2e^{-cn\varepsilon_0^2}$$

And $d^{(2)}(f_i, f_j) > \varepsilon \Rightarrow$

$$P(d_n^{(2)}(f_i, f_j) \geq \sqrt{\varepsilon^2 - \varepsilon_0}) \geq 1 - 2e^{-cn\varepsilon_0^2}$$

Let $\varepsilon_0 = \sqrt{\varepsilon^2 - \varepsilon}$, $\varepsilon_0 \leq \varepsilon^2$, then

$$P(\{i \neq j \mid d_n^{(2)}(f_i, f_j) > \varepsilon_0\}) \geq 1 - 2Ne^{-cn\varepsilon^4} \quad \square$$

Remark: $A = \{i \neq j \mid d_n^{(2)}(f_i, f_j) > \varepsilon_0\}$

is not empty, take $w \in A$,

$$x_i = X_i(w),$$

we define a uniform law on $\{x_i\}$:

$$P^{(n)}$$

Now we mark d , to emphasize the influence of P on metric $d^{(2)}$.

Theorem 1 $\sup_p N(\delta; \mathcal{F}, d_p) \leq (\frac{1}{\delta})^{cv}$, $0 \leq \delta \leq 1$

Pf: Note that $\{f_i\}_{i=1}^{\infty}$ is ϵ_0 -packing of \mathcal{F} with metric $d_{p^{cn}}$

then

$$\begin{aligned} N &= P(\delta; \mathcal{F}, d_p) \leq P(\delta_0; \mathcal{F}, d_{p^{cn}}) \\ &\leq N(\frac{\delta_0}{2}; \mathcal{F}, d_{p^{cn}}) \end{aligned}$$

Define $d_{p^{cn}}^\infty$:

$$d_{p^{cn}}^\infty(f, g) = \max_i |f(x_i) - g(x_i)|, f, g \in \mathcal{F}$$

Then $d_{p^{cn}}^\infty(f, g) \geq d_{p^{cn}}(f, g)$

$$\leq \underbrace{N(\frac{\delta_0}{2}; \mathcal{F}, d_{p^{cn}}^\infty)}_{\text{the growth function of } \mathcal{F} \text{ given } x_1, \dots, x_n} \quad (\delta_0 < 2)$$

the growth function of \mathcal{F} given x_1, \dots, x_n

$$\leq \left(\frac{en}{v}\right)^v, \quad v \text{ is VC-dim of } \mathcal{F}.$$

$$\text{Let } 1 - 2N^2 e^{-cn\delta^4} = 0 \Rightarrow n = \frac{2 \log N - \log \frac{1}{2}}{c\delta^4}$$

$$\text{Take } n = \lceil \frac{2 \log N - \log \frac{1}{2}}{c\delta^4} \rceil \leq \delta^{-4} \log N.$$

$$\begin{aligned} \textcircled{1} \Rightarrow N &\leq \left(\frac{\log N}{v\delta^4}\right)^v \Rightarrow N^{\frac{1}{v}} \leq \frac{2}{\delta^4} \log N^{\frac{1}{v}} \\ \Rightarrow N^{\frac{1}{v}} &\leq \frac{1}{\delta^4} \cdot N^{\frac{1}{2v}} \Rightarrow N \leq \left(\frac{1}{\delta^4}\right)^{2v} = \left(\frac{1}{\delta}\right)^{cv} \end{aligned}$$

VC Class of functions

Lemma

Let \mathcal{H} is a vector space of function: $X \rightarrow \mathbb{R}$
 $d = \dim(\mathcal{H}) < \infty$, then $D = \{x; f(x) \leq 0\}, f \in \mathcal{H}\}$ is a
 VC class and $VC(D) \leq \dim(\mathcal{H})$

Pf: If $VC(D) > \dim(\mathcal{H})$, which means that

$$\forall \{x_1, \dots, x_d, x_{d+1}\} \subset X, \forall I \subset \{1, \dots, d+1\}$$

$\exists f \in \mathcal{H}$ s.t.

$$f(x_i) \leq 0, i \in I, f(x_i) > 0, i \notin I \quad (1)$$

Define $T: \mathcal{H} \mapsto \mathbb{R}^{d+1}, T f = (f(x_1), \dots, f(x_{d+1}))$

Since $\dim(T\mathcal{H}) \leq d$, then $\exists H$ s.t. $\dim(H) \geq 1$

$$\mathbb{R}^{d+1} = T\mathcal{H} \oplus H_{d+1}$$

$$\Rightarrow \exists (y_1, \dots, y_{d+1}) \neq 0, \sum_{i=1}^{d+1} y_i f(x_i) = 0, \forall f \in \mathcal{H}$$

$$\Rightarrow \sum_{\{i; y_i \leq 0\}} y_i f(x_i) = - \sum_{\{i; y_i > 0\}} y_i f(x_i)$$

Let $I = \{i; y_i \geq 0\}$, from (1), contradiction.

Example Half space $\{x \in \mathbb{R}^d; \theta^T x \geq t\}, (\begin{smallmatrix} \theta \\ t \end{smallmatrix}) \in \mathbb{R}^{d+1}$
 $VC(\text{Half Space}) \leq d+1$

Example (Sphere in \mathbb{R}^d)

$$\{ \overline{B}(c; \rho), (\begin{smallmatrix} a \\ \rho \end{smallmatrix}) \in \mathbb{R}^{d+1} \}$$

$$\{x \in \mathbb{R}^d; \|x-a\|^2 - \rho^2 \leq 0\}$$



$$\{x \in \mathbb{R}^d; \|x-a\|^2 - \rho^2 \leq 0\} \supseteq \{x \in \mathbb{R}^d; \|x-a\|^2 \leq \rho^2\}$$

$$\|x\|^2 - 2\langle x, a \rangle + \|a\|^2 - p \rightarrow (x, \|a\|^2 - p)$$

$$\Rightarrow \|x\|^2 - 2\langle x, d \rangle + d \stackrel{\text{linear}}{\Leftrightarrow} c(c, d) \in \mathbb{R}^{d+1}$$

\Rightarrow embedding in $\mathbb{R}^{d+1} \rightarrow V(\text{Sphere}) \leq d+1$

Def

\mathcal{F} is a function space: $X \mapsto \mathbb{R}$. \mathcal{F} is called VC class if the subgraph

$D_{\mathcal{F}} = \{ (x, t); f(x) \geq t \}, f \in \mathcal{F} \}$ is VC class.

And $V(\mathcal{F}) \triangleq V(D_{\mathcal{F}})$

Example D is VC class $\Leftrightarrow \mathcal{F} = \{ \mathbf{1}_D; D \in \mathcal{D} \}$ is VC

Pf: $\forall D \in \mathcal{D}$,

$$\{ (x, t); \mathbf{1}_D(x) \geq t \} = (X \times (-\infty, 0]) \cup (D \times [0, 1]) \\ \cup C \phi \times [1, +\infty)$$

$$\begin{array}{c} \phi \\ \Leftrightarrow \\ D \end{array}$$

Since ϕ is bijection, claim holds.

Example \mathcal{F} is a vector space of function: $X \mapsto \mathbb{R}$, $\dim(\mathcal{F}) < \infty$, then \mathcal{F} is VC and $V(\mathcal{F}) \leq \dim(\mathcal{F}) + 1$

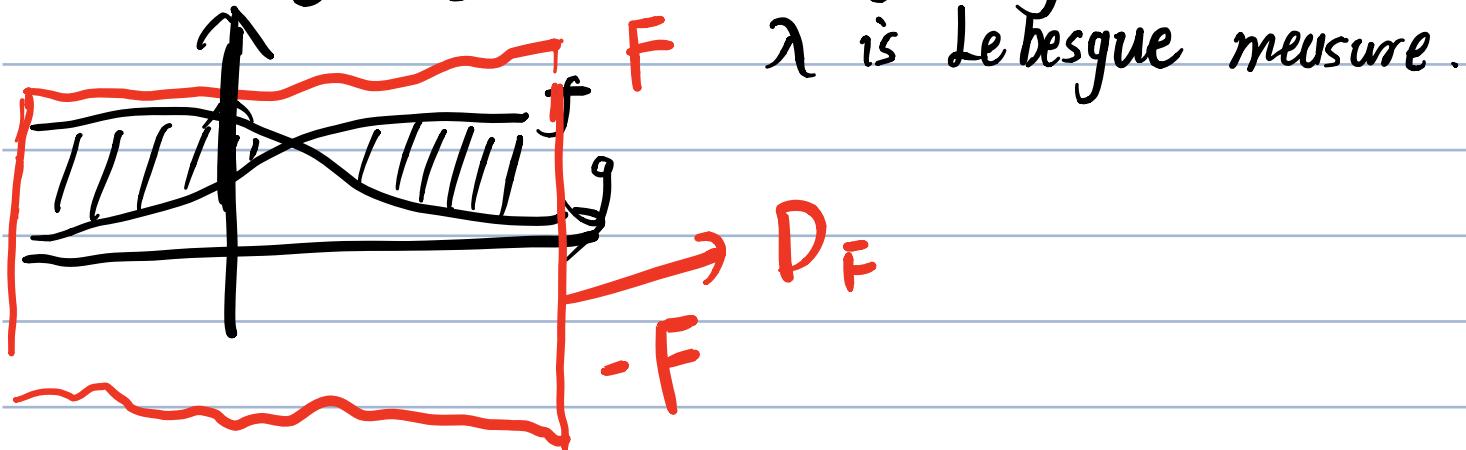
Pf: $\{ (x, t); f(x) \geq t \} = \{ (x, t); t - f(x) \leq 0 \}$
 $t - f(x) \stackrel{\text{linear}}{\Leftrightarrow} (f, t) \Rightarrow V(\mathcal{F}) \leq \dim(\mathcal{F}) + 1$

Theorem

\mathcal{F} is a VC-class of function with $v = V(\mathcal{F})$ and F is the envelope function of \mathcal{F} ($|f(x)| \leq F(x)$, $\forall f \in \mathcal{F}$) s.t. $R = (\mathbb{E} F(X)^2)^{\frac{1}{2}} < \infty$, $X \sim (\mathcal{X}, \bar{\Sigma}, P)$

$$\text{supp } NC \leq R ; \mathcal{F}, d_P \leq (\frac{1}{R})^{Cv}$$

Pf: $\forall f, g \in \mathcal{F}$, let $D_f = \{x, t\}; f(x) \geq t\}$
 then $\mathbb{E} |f(x) - g(x)| = P \times \lambda(D_f \Delta D_g)$,



$$\begin{aligned} \text{Then } \mathbb{E}(f(x) - g(x))^2 &\leq 2 \mathbb{E}|f(x) - g(x)| F(x) \\ &= 2 \int |f(x) - g(x)| F(x) dP(x) \\ &= 2 \mathbb{E}_{x \in \mathcal{X}} |f(x) - g(x)| \left(\frac{dQ}{dP} = F \right) \\ &= 2 Q \times \lambda(D_f \Delta D_g) \end{aligned}$$

Note that $D_f \Delta D_g \subset \{(x, t); |t| \leq F(x)\} \stackrel{\Delta}{=} D_F$
 Define new probability measure on D_F :

$$A \in \sigma(D_F), T(A) := \frac{Q \times \lambda(A)}{Q \times \lambda(D_F)} = \frac{Q \times \lambda(A)}{2 \mathbb{E}_Q F(X)}$$

$$\text{Then } d(f, g) \leq \sqrt{2Q \times \lambda(D_f \Delta D_g)} = \frac{Q \times \lambda(A)}{2 R^2}$$

$$= \sqrt{2T(D_f \Delta D_g)} \cdot \sqrt{2R^2}$$

$$= \sqrt{\mathbb{E}_T (\mathbf{1}_{D_f} - \mathbf{1}_{D_g})^2} \cdot 2R$$

$$\Rightarrow \frac{d_{\text{ptf}, g}}{2R} \leq d_T(\mathbf{1}_{D_f}, \mathbf{1}_{D_g}),$$

Let $\mathcal{Y} := \{\mathbf{1}_{D_f}; f \in \mathcal{F}\}$, then \mathcal{Y} is a class of functions: $X \times \mathbb{R} \rightarrow \mathbb{R}$ and $V(\mathcal{Y}) = V(\mathcal{F}) = V$, then

$$N(\delta; \mathcal{F}, d_p/2R) = N(R\delta; \mathcal{F}, d_p) \\ \leq N(\delta, \mathcal{Y}, d_T) \asymp (\frac{1}{\delta})^{CV}. \quad \square$$