# Vapnik-Červonenkis (VC) Classes and Uniform Entropy

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### Background

Recall the uniform entropy condition in the Donsker theorem

$$\int_0^\infty \sup_Q \sqrt{\log N(\epsilon \|F\|_{Q,2}, \mathcal{F}, L_2(Q))} \, d\epsilon < \infty.$$
 (1)

• In particular, (1) holds if for some  $\delta > 0$ ,

$$\sup_{Q} \log N(\epsilon \|F\|_{Q,2}, \mathcal{F}, L_2(Q)) \leq K\left(\frac{1}{\epsilon}\right)^{2-\delta}, \quad 0 < \epsilon < 1.$$

• A much stronger condition is that for some number V,

$$\sup_{Q} N(\epsilon \|F\|_{Q,2}, \mathcal{F}, L_2(Q)) \le K\left(\frac{1}{\epsilon}\right)^V, \quad 0 < \epsilon < 1.$$
 (2)

• Vapnik-Červonenkis (VC) classes satisfy (2).

### Outline



- 2 VC Classes of Functions
- Convex Hulls and VC Hull Classes
- 4 Examples and Properties of VC Classes

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### VC classes of sets

Consider an arbitrary collection of *n* points  $\{x_1, \ldots, x_n\}$  in a set  $\mathcal{X}$  and a collection  $\mathcal{C}$  of subsets of  $\mathcal{X}$ .

- We say that C *picks out* a certain subset A of  $\{x_1, \ldots, x_n\}$  if  $A = C \cap \{x_1, \ldots, x_n\}$  for some  $C \in C$ .
- We say that C shatters {x<sub>1</sub>,..., x<sub>n</sub>} if all of the 2<sup>n</sup> possible subsets of {x<sub>1</sub>,..., x<sub>n</sub>} are picked out by the sets in C.
- The *VC index V*(*C*) of the class *C* is the smallest *n* for which no set of size *n* is shattered by *C*.
- More formally, VC index is defined through

$$\Delta_n(\mathcal{C}; x_1, \ldots, x_n) = \Big| \Big\{ \mathcal{C} \cap \{x_1, \ldots, x_n\} : \mathcal{C} \in \mathcal{C} \Big\} \Big|,$$
  
$$V(\mathcal{C}) = \inf \Big\{ n : \max_{x_1, \ldots, x_n \in \mathcal{X}} \Delta_n(\mathcal{C}; x_1, \ldots, x_n) < 2^n \Big\}.$$

## VC classes of sets (cont.)

Some books define V(C) as the largest n such that some set of size n is shattered by C. (i.e., V(C) − 1 in our notation).

- If C shatters sets of arbitrarily large size, we set  $V(C) = \infty$ .
- Clearly, the more refined C is, the higher the VC index.

• We say that C is a VC class if  $V(C) < \infty$ .

## Example 1

- Let  $\mathcal{X} = \mathbb{R}$  and define the collection of sets  $\mathcal{C} = \{(-\infty, \mathbf{c}] : \mathbf{c} \in \mathbb{R}\}.$
- C shatters no two-point set {x<sub>1</sub>, x<sub>2</sub>}, because it fails to pick out the largest of the two points.



- Thus  $V(\mathcal{C}) = 2$  and  $\mathcal{C}$  is a VC class.
- When extended to X = ℝ<sup>d</sup>, the VC index of the same type of sets is d + 1.

## Example 2

- Let X = ℝ. Now we consider C = {(a, b] : -∞ ≤ a < b ≤ ∞}. This collection shatters every two-point set.
- For any set of three points, C cannot pick out the subset consisting of the smallest and largest points.

$$- \left( \begin{array}{c} \bullet \\ \bullet \\ x_1 \end{array} \right) \left( \begin{array}{c} \bullet \\ x_2 \end{array} \right) \left( \begin{array}{c} \bullet \\ \bullet \\ x_3 \end{array} \right) \rightarrow$$

- Thus  $V(\mathcal{C}) = 3$  and  $\mathcal{C}$  is a VC class.
- With more effort, it can be seen that the VC index of the same type of sets in ℝ<sup>d</sup> is 2d + 1.

### A combinatorial result

Recall that we previously defined  $\Delta_n(\mathcal{C}; x_1, \ldots, x_n)$  to be the number of subsets of  $\{x_1, \ldots, x_n\}$  picked out by  $\mathcal{C}$ . The following lemma provides an upper bound for  $\Delta_n(\mathcal{C}; x_1, \ldots, x_n)$ .

#### Sauer's lemma

For a VC class of sets C, one has

$$\max_{x_1,\ldots,x_n\in\mathcal{X}}\Delta_n\left(\mathcal{C};x_1,\ldots,x_n\right)\leq\sum_{j=0}^{V(\mathcal{C})-1}\binom{n}{j}$$

Since the RHS is bounded by  $V(\mathcal{C})n^{V(\mathcal{C})-1}$ , the LHS grows polynomially of order at most  $O(n^{V(\mathcal{C})-1})$ .

## Proof of Sauer's lemma

We need to use the following lemma<sup>1</sup>:

#### Lemma

For any set of *n* points  $\{x_1, \ldots, x_n\}$  and any collection of sets C,  $\Delta_n(C; x_1, \ldots, x_n)$  is bounded above by the number of subsets of  $\{x_1, \ldots, x_n\}$  shattered by C.

Sauer's lemma follows immediately from the above lemma, since the size of any shattered set is at most V(C) - 1.

<sup>&</sup>lt;sup>1</sup> See Lemma 2.6.2 of VW (pp. 135-136) for the proof.

## Bound on covering number

Let  $1\{C\}$  denote the collection of all indicator functions of sets in the class C. The following theorem<sup>2</sup> gives an upper bound on the  $L_r$  covering numbers of  $1\{C\}$ :

#### Theorem 1

There exists a universal constant K such that for any VC class of sets C, any probability measure Q, any  $r \ge 1$ , and any  $0 < \epsilon < 1$ ,

$$N(\epsilon, \mathbf{1}\{\mathcal{C}\}, L_r(\mathbf{Q})) \leq KV(\mathcal{C})(4e)^{V(\mathcal{C})} \left(\frac{1}{\epsilon}\right)^{r(V(\mathcal{C})-1)}$$

<sup>&</sup>lt;sup>2</sup> See Theorem 2.6.4 of VW (pp. 136-139) for the proof.

### Remarks

Since  $F \equiv 1$  serves as an envelope for  $1\{C\}$ , it follows immediately from the preceding theorem that the uniform entropy integral

$$\int_0^\infty \sup_Q \sqrt{\log N\left(\epsilon \|F\|_{Q,2}, 1\{\mathcal{C}\}, L_2(Q)\right)} \, d\epsilon$$
$$\lesssim \int_0^1 \sqrt{\log(1/\epsilon)} \, d\epsilon = \int_0^\infty u^{1/2} e^{-u} \, du \le 1.$$

Thus, for any VC class C,  $1\{C\}$  is GC and Donsker, provided the requisite measurability conditions hold.

### Outline



### 2 VC Classes of Functions

- Convex Hulls and VC Hull Classes
- 4 Examples and Properties of VC Classes

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## Subgraph

The subgraph of a function  $f : \mathcal{X} \to \mathbb{R}$  is the subset of  $\mathcal{X} \times \mathbb{R}$  given by

 $\{(x, t): t < f(x)\}.$ 



## VC classes of functions

- A class *F* of measurable real-valued functions on the sample space *X* is called a *VC subgraph class*, or just a *VC class*, if the collection of all subgraphs of the functions in *F* forms a VC class of sets (in *X* × ℝ).
- Let  $V(\mathcal{F})$  denote the VC index of the set of subgraphs of  $\mathcal{F}$ .
- Just as for sets, the covering numbers of VC classes of functions grow at a polynomial rate.

## Bound on covering number

This is more precisely stated in the following theorem:

#### Theorem 2

For a VC class of functions  $\mathcal{F}$  with measurable envelope function F and  $r \ge 1$ , one has for any probability measure Q with  $||F||_{Q,r} > 0$ ,

$$N\left(\epsilon \|F\|_{Q,r}, \mathcal{F}, L_r(Q)
ight) \leq KV(\mathcal{F})(4e)^{V(\mathcal{F})}\left(rac{2}{\epsilon}
ight)^{r(V(\mathcal{F})-1)}$$

for a universal constant K and  $0 < \epsilon < 1$ .

Thus a VC class of functions easily satisfies the uniform entropy condition. Hence, a suitably measurable VC class is Donsker, provided its envelope has a weak second moment.

## Proof for r = 1

- Let C be the set of all subgraphs  $C_f$  of functions  $f \in \mathcal{F}$ . By Fubini's theorem,  $Q|f - g| = (Q \times \lambda) (C_f \Delta C_g)$ , where  $\lambda$  is Lebesgue measure on the real line, and  $A \Delta B = A \cup B - A \cap B$ for any two sets A and B.
- We renormalize  $Q \times \lambda$  to a probability measure on the set  $\{(x, t) : |t| \le F(x)\}$  by defining  $P = (Q \times \lambda)/(2QF)$ .
- For any  $f, g \in \mathcal{F}$ ,  $||f g||_{Q,1} = 2QF||1\{C_f\} 1\{C_g\}||_{P,1}$ .
- By the result for sets in Theorem 1, there exists a universal constant *K* such that

$$\begin{split} N(\epsilon 2QF, \mathcal{F}, L_1(Q)) &= N(\epsilon, 1\{\mathcal{C}\}, L_1(P)) \\ &\leq KV(\mathcal{F})(4e)^{V(\mathcal{F})} \left(\frac{1}{\epsilon}\right)^{V(\mathcal{F})-1} \end{split}$$

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### Proof for r > 1

• For r > 1, define a probability measure *R* with density  $F^{r-1}/QF^{r-1}$  with respect to *Q*, so that  $Rf = Q\{fF^{r-1}\}/QF^{r-1}$ .

$$\begin{aligned} &Q|f-g|^{r} \leq Q\left\{|f-g|(2F)^{r-1}\right\} = 2^{r-1}R|f-g|QF^{r-1}\\ \Rightarrow \|f-g\|_{Q,r} \leq 2^{1-1/r}(QF^{r-1})^{1/r}\|f-g\|_{R,1}^{1/r}\\ \Rightarrow \frac{\|f-g\|_{Q,r}}{2\|F\|_{Q,r}} \leq \left(\frac{QF^{r-1}}{2QF^{r}}\right)^{1/r}\|f-g\|_{R,1}^{1/r} = \left(\frac{\|f-g\|_{R,1}}{2RF}\right)^{1/r}.\end{aligned}$$

Hence, elementary manipulations yield

$$egin{aligned} & N\left(\epsilon 2\|F\|_{\mathcal{Q},r},\mathcal{F},L_r(\mathcal{Q})
ight) \leq N\left(\epsilon^r 2RF,\mathcal{F},L_1(R)
ight) \ & \leq KV(\mathcal{F})(4e)^{V(\mathcal{F})}\left(rac{1}{\epsilon}
ight)^{r(V(\mathcal{F})-1)}. \end{aligned}$$

## Outline





- Convex Hulls and VC Hull Classes
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## Convex hulls and VC hull classes

The symmetric convex hull of a class of functions F is defined by

sconv 
$$\mathcal{F} := \left\{ \sum_{i=1}^m \alpha_i f_i : \sum_{i=1}^m |\alpha_i| \le 1, f_i \in \mathcal{F} \right\}.$$

Similarly, the convex hull of F is defined by

$$\operatorname{conv} \mathcal{F} := \left\{ \sum_{i=1}^m \alpha_i f_i : \ \alpha_i > \mathbf{0}, \ \sum_{i=1}^m |\alpha_i| \le 1, \ f_i \in \mathcal{F} \right\}.$$

- We use conv F and sconv F to denote the pointwise closures of conv F and sconv F, respectively.
- A class of measurable functions  $\mathcal{F}$  is called a *VC hull class* if  $\mathcal{F} = \overline{\text{sconv}}\mathcal{G}$  for some VC class  $\mathcal{G}$ .

## Bound on entropy

The following theorem<sup>3</sup> gives an upper bound on the entropy of a VC hull class:

#### Theorem 3

Let Q be a probability measure on  $(\mathcal{X}, \mathcal{A})$ , and let  $\mathcal{F}$  be a class of measurable functions with measurable square integrable envelope F such that  $QF^2 < \infty$  and

$$N\left(\epsilon \|F\|_{Q,2}, \mathcal{F}, L_2(Q)
ight) \leq C\left(rac{1}{\epsilon}
ight)^V, \quad 0<\epsilon<1.$$

Then there exists a constant K depending only on C and V such that

$$\log N\left(\epsilon \|F\|_{Q,2}, \overline{\operatorname{conv}}\mathcal{F}, L_2(Q)\right) \leq K\left(\frac{1}{\epsilon}\right)^{2V/(V+2)}$$

<sup>&</sup>lt;sup>3</sup> See Theorem 2.6.9 of VW (pp. 142-144) for the proof.

### Remarks

- The preceding theorem shows that the entropy of the convex hull of any polynomial class is of lower order than  $(1/\epsilon)^r$  for some r < 2, which is just enough to ensure that the uniform entropy condition holds.
- Since sconv *F* is contained in the convex hull of *F* ∪ {−*F*} ∪ {0}, and the covering number of *F* ∪ {−*F*} ∪ {0} is at most twice the covering number of *F* plus 1, the bound in Theorem 3 is valid for sconv*F* as well.
- Thus, any VC hull class satisfies the uniform entropy condition.
- However, VC hull classes can be considerably larger than VC classes, so we do not have similar results for covering numbers.

## Bound for VC hull classes

Finally, we have an easy corollary that gives precise bounds for entropy numbers of VC hull classes:

Corollary 4

For any VC hull class  $\mathcal{F}$  of measurable functions and any probability measure Q,

$$\log N\left(\epsilon \|F\|_{Q,2}, \mathcal{F}, L_2(Q)\right) \leq K\left(\frac{1}{\epsilon}\right)^{2-2/V_m(\mathcal{F})}, \quad 0 < \epsilon < 1,$$

for a constant K that depends only on the VC index  $V_m(\mathcal{F})$  of the VC subgraph class associated with  $\mathcal{F}$ .

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## Proof of the corollary

#### Proof.

Let  $\mathcal{G}$  be the VC class associated with  $\mathcal{F}$ , i.e.,  $\mathcal{F} = \overline{\text{sconv}}\mathcal{G}$ . Since  $\mathcal{G}$  is contained in  $\mathcal{F}$ , F is also an envelope function for  $\mathcal{G}$ . By Theorem 2,

$$N\left(\epsilon \|F\|_{Q,2}, \mathcal{G}, L_2(Q)\right) \leq C\left(\frac{1}{\epsilon}\right)^{2(V_m(\mathcal{F})-1)}, \quad 0 < \epsilon < 1$$

The desired bounds follow immediately from Theorem 3 with  $V = 2(V_m(\mathcal{F}) - 1)$ .

## Outline

VC Classes of Sets

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## Example: vector space of functions

#### Lemma 5

Any finite-dimensional vector space  $\mathcal{F}$  of measurable functions  $f: \mathcal{X} \to \mathbb{R}$  is VC-subgraph with  $V(\mathcal{F}) \leq dim(\mathcal{F}) + 2$ .

#### Proof.

Suppose  $V(\mathcal{F}) > dim(\mathcal{F}) + 2$ , then there exists a collection of  $n = dim(\mathcal{F}) + 2$  points  $(x_1, t_1), \ldots, (x_n, t_n)$  in  $\mathcal{X} \times \mathbb{R}$  that can be shattered by the subgraphs of  $\mathcal{F}$ . By assumption, the vectors  $(f(x_1) - t_1), \ldots, (f(x_n) - t_n)^T$ , as *f* ranges over  $\mathcal{F}$ , are contained in a  $dim(\mathcal{F}) + 1 = (n - 1)$ -dimensional subspace of  $\mathbb{R}^n$ .

## Proof of Lemma 5 (cont.)

#### Proof.

There exists a vector *a* with at least one strictly positive coordinate that is orthogonal to this subspace. Thus,

$$\sum_{i:a_i>0} a_i \left(f\left(x_i\right) - t_i\right) = \sum_{i:a_i<0} \left(-a_i\right) \left(f\left(x_i\right) - t_i\right), \quad \text{for every } f \in \mathcal{F}.$$
 (3)

Consider the subset  $A = \{(x_i, t_i) : a_i > 0\}$  and its complement  $A^c = \{(x_i, t_i) : a_i \le 0\}$ . Since *A* is picked out by the subgraphs of  $\mathcal{F}$ , it must be contained in the subgraph of some  $f \in \mathcal{F}$ , while  $A^c$  must be outside the subgraph of this *f*. Then the LHS of (3) is strictly positive while the RHS is nonpositive (contradiction!)

## Example: translates of monotone function

#### Lemma 6

The set of all translates { $\psi(x - h) : h \in \mathbb{R}$ } of a fixed monotone function  $\psi : \mathbb{R} \to \mathbb{R}$  is VC-subgraph of index 2.

#### Proof.

Without loss of generality, we assume  $\psi$  is nondescreasing. For any  $h_1 > h_2$ , the subgraph of  $x \mapsto \psi(x - h_1)$  is contained in the subgraph of  $x \mapsto \psi(x - h_2)$ . Any collection of sets with this property shatters no two-point set, thus has VC index 2.

## Example: monotone stochastic process

#### Lemma 7

Let  $\{X(t) : t \in T\}$  be a monotone increasing stochastic process, where  $T \subset \mathbb{R}$ . Then X is VC-subgraph of index 2.

#### Proof.

Let  $\mathcal{X}$  be the set of all increasing functions mapping T to  $\mathbb{R}$ . For any  $t \in T$ , define function  $f_t : \mathcal{X} \to \mathbb{R}$  with  $f_t(x) = x(t)$ . We only need to show that the class of functions  $\mathcal{F} = \{f_t : t \in T\}$  is VC of index 2. For any  $t_1 < t_2$ , the subgraph of  $f_{t_1}$  is contained in the subgraph of  $f_{t_2}$ .

### Build VC classes from basic VC classes of sets

#### Lemma 8

Let C and D be VC classes of sets in a set X and  $\mathcal{E}$  a VC class of sets in  $\mathcal{Y}$ . Also, let  $\phi : X \to \mathcal{Y}$  and  $\psi : Z \to X$  be fixed functions. Then<sup>a</sup> (i)  $C^c = \{C^c : C \in C\}$  is VC with  $V(C^c) = V(C)$ ; (ii)  $C \sqcap D = \{C \cap D : C \in C, D \in D\}$  is VC with index  $\leq V(C) + V(D) - 1$ ; (iii)  $C \sqcup D = \{C \cup D : C \in C, D \in D\}$  is VC with index  $\leq V(C) + V(D) - 1$ ; (iv)  $D \times \mathcal{E}$  is VC in  $X \times \mathcal{Y}$  with VC index  $\leq V(D) + V(\mathcal{E}) - 1$ ; (v)  $\phi(C)$  is VC with index V(C) if  $\phi$  is one-to-one; (vi)  $\psi^{-1}(C)$  is VC with index  $\leq V(C)$ .

<sup>a</sup>See Lemma 9.7 in Kosorok (pp. 159-160) for the proof.

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## Build VC classes from basic VC classes of functions

#### Lemma 9

Let  $\mathcal{F}$  and  $\mathcal{G}$  be VC subgraph classes of functions on a set  $\mathcal{X}$  and  $g: \mathcal{X} \to \mathbb{R}, \phi: \mathbb{R} \to \mathbb{R}$ , and  $\psi: \mathcal{Z} \to \mathcal{X}$  fixed functions. Then<sup>a</sup>

- (i)  $\mathcal{F} \wedge \mathcal{G} = \{f \wedge g : f \in \mathcal{F}, g \in \mathcal{G}\}$  is VC-subgraph with index  $\leq V(\mathcal{F}) + V(\mathcal{G}) 1;$
- (ii)  $\mathcal{F} \lor \mathcal{G} = \{f \lor g : f \in \mathcal{F}, g \in \mathcal{G}\}$  is VC-subgraph with index  $\leq V(\mathcal{F}) + V(\mathcal{G}) 1;$
- (iii)  $\{\mathcal{F} > 0\} = \{\{f > 0\} : f \in \mathcal{F}\}$  is VC with index V( $\mathcal{F}$ );
- (iv)  $-\mathcal{F}$  is VC-subgraph with index  $V(\mathcal{F})$ ;

(v)  $\mathcal{F} + g = \{f + g : f \in \mathcal{F}\}$  is VC-subgraph with index  $V(\mathcal{F})$ ;

(vi)  $\mathcal{F} \cdot g = \{ fg : f \in \mathcal{F} \}$  is VC-subgraph with index  $\leq 2V(\mathcal{F}) - 1$ ;

(vii)  $\mathcal{F} \circ \psi = \{f(\psi) : f \in \mathcal{F}\}$  is VC-subgraph with index  $\leq V(\mathcal{F})$ ;

(viii)  $\phi \circ \mathcal{F}$  is VC-subgraph with index  $\leq V(\mathcal{F})$  for monotone  $\phi$ .

<sup>&</sup>lt;sup>a</sup>See Proof of Lemma 9.9 in Kosorok (pp. 174-175) for the proof.

## Relation between C and $1\{C\}$

#### Lemma 10

 $1\{C\}$  is VC-subgraph if and only if C is a VC class of sets. Moreover, their respective VC indices are equal.

#### Proof.

Let sub(*C*) denote the subgraph of the indicator function  $1\{C\}$  for any  $C \in C$ . Let D denote the collection of all sub(*C*). We can easily verify that for any  $x \in \mathcal{X}$  and  $C \in C$ ,

$$x \in C \iff (x, 0) \in \operatorname{sub}(C).$$

Thus, for any n > 0 and any  $x_1, \ldots, x_n \in \mathcal{X}$ ,  $\{x_1, \ldots, x_n\}$  can be shattered by  $\mathcal{C}$  if and only if  $\{(x_1, 0), \ldots, (x_n, 0)\}$  can be shattered by  $\mathcal{D}$ . The conclusion then follows.

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## Result of monotone functions within [0, 1]

#### Lemma 11

The set  $\mathcal F$  of all monotone functions  $f:\,\mathbb R\to[0,1]$  satisfies

$$\log N(\epsilon, \mathcal{F}, L_2(Q)) \leq \frac{K}{\epsilon}, \quad 0 < \epsilon < 1$$

for a universal constant K and any probability measure Q.

#### Proof.

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We first consider the set  $\mathcal{F}_+$  of all monotone increasing functions  $f : \mathbb{R} \to [0, 1]$ . Any  $f \in \mathcal{F}_+$  is the pointwise limit of the sequence

$$f_m = \sum_{i=1}^m \frac{1}{m} \operatorname{1}\left\{f > \frac{i}{m}\right\}.$$

## Proof of Lemma 11 (cont.)

#### Proof.

Thus,  $\mathcal{F}_{+} = \overline{\text{conv}} 1{\mathcal{G}}$  for some class  $\mathcal{G}$  of sets of the form  $\{f > t\}$ , with f ranging over  $\mathcal{F}_{+}$  and t over  $\mathbb{R}$ . Since f is increasing,  $\mathcal{G}$  is contained in the collection of intervals  $\{(t, +\infty) : t \in \mathbb{R}\} \cup \{[t, +\infty) : t \in \mathbb{R}\}, \text{ which is VC of index 2. Thus, } \mathcal{G}$  is also VC of index 2. By Corollary 4,

$$\log N(\epsilon, \mathcal{F}_+, L_2(Q)) \leq \frac{K_0}{\epsilon}, \quad 0 < \epsilon < 1.$$

Any monotone decreasing function  $g : \mathbb{R} \to [0, 1]$  can be written as 1 - f for some  $f \in \mathcal{F}_+$ . Thus, for  $0 < \epsilon < 1$ ,

$$\log N(\epsilon, \mathcal{F}, L_2(\mathcal{Q})) = \log(2) + \log N(\epsilon, \mathcal{F}_+, L_2(\mathcal{Q})) \leq rac{K}{\epsilon}.$$