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1 Vector spaces

§1.1 Metric spaces.

$$X, \rho: X \times X \to \mathbb{R}_+$$

Definition. ρ — metric

1.
$$\rho(x,y) \ge 0, = 0 \iff x = y$$

2.
$$\rho(x, y) = \rho(y, x)$$

3.
$$\rho(x,y) \leq \rho(x,z) + \rho(y,z)$$

Definition. (X, ρ) — Metric space.

Definition. $x = \lim x_n \iff \rho(x_n, x) \to 0$

$$X, \tau = \{ G \subset X \}$$

1.
$$\varnothing, X \in \tau$$

2.
$$G_{\alpha} \in \tau, \alpha \in \mathscr{A} \implies \bigcup_{\alpha} G_{\alpha} \in \tau$$

3.
$$G_1, \ldots, G_n \in \tau \implies \bigcap_{j=1}^n G_j \in \tau$$

Definition. (X, τ) — Topological space.

$$x = \lim x_n \quad \forall G \in \tau : x \in G \quad \exists N : \forall n > N \implies x_n \in G$$

$$G$$
 — open in τ

$$F = X \setminus G$$
 — closed

Definition.
$$B_r(a) = \{ x \mid \rho(x, a) < r \}$$
 — open ball

$$\tau = \bigcup B_r(a)$$

Statement 1.1.
$$b \in B_{r_1}(a_1) \cap B_{r_2}(a_2) \implies \exists r_3 > 0 : B_{r_3}(a_3) \subset B_{r_1}(a_1) \cap B_{r_2}(a_2)$$

In this sense metric space is just a special case of topological space.

Example 1.
$$\mathbb{R}, \rho(x, y) = |x - y|, MS$$

Example 2.
$$\bar{x} = (x_1, \dots, x_n) \in \mathbb{R}^n, \rho(\bar{x}, \bar{y}) = \sqrt{\sum_{j=1}^n (x_j - y_j)^2}, MS$$

Example 3.
$$\bar{x} = (x_1, \dots, x_n, \dots) \in \mathbb{R}^{\infty}$$

$$\alpha \bar{x} = (\alpha x_1, \dots, \alpha x_n, \dots)$$

$$\bar{x} + \bar{y} = (x_1 + y_1, \dots, x_n + y_n, \dots)$$

Let's define $\lim \bar{x}_m$

• in
$$\mathbb{R}^n$$
: $\bar{x}_n \to \bar{x} \iff \forall j = 1, \dots, n$ $x_j^{(m)} \xrightarrow[m \to \infty]{} x_j$

• in
$$\mathbb{R}^{\infty}$$
: $\bar{x}_m \to \bar{x} \iff \forall j = 1, 2, 3, \dots \quad x_j^{(m)} \xrightarrow[m \to \infty]{} x_j$

Definition. $\rho(\bar{x}, \bar{y}) \stackrel{\text{def}}{=} \sum_{n=1}^{\infty} \frac{1}{2^n} \underbrace{\frac{|x_n - y_n|}{1 + |x_n - y_n|}}_{\text{Definition}}$ — Urysohn metric.

1

5

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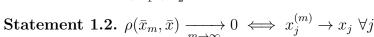
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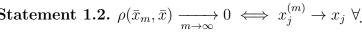
$$\varphi(t) = \frac{t}{1+t}$$

$$\varphi(t_1) + \varphi(t_2) \geqslant \varphi(t_1 + t_2)$$

$$\varphi(t_1) + \varphi(t_2) = \frac{t_1}{1 + t_1} + \frac{t_2}{1 + t_2} \geqslant \frac{t_1}{1 + t_1 + t_2} + \frac{t_2}{1 + t_1 + t_2}$$

$$\varphi(t_1) + \varphi(t_2) \geqslant \frac{t_1 + t_2}{1 + t_1 + t_2} = \varphi(t_1 + t_2)$$





Proof.

$$\overrightarrow{f(|x_k^{(n)} - x_k|)} \leqslant 2^k \rho(x^{(n)}, x)$$
Let $\rho(x^{(n)}, x) \leqslant \frac{\varepsilon}{2^k}$, then $f(|x_k^{(n)} - x_k|) < \varepsilon$

$$|x_k^{(n)} - x_k| = t = \frac{1}{1 - f(t)} - 1$$
, then $t \to 0$

Let's choose
$$k_0$$
 for which $\sum_{k=k_0+1}^{\infty} \frac{1}{2^k} < \varepsilon$

Let's choose n_0 for which $\forall k \leq k_0, n > n_0 : |x_k^{(n)} - x_k| < \varepsilon$.

Then
$$\rho(x^{(n)}, x) < \sum_{k=1}^{k_0} \frac{\varepsilon}{2^k} + \varepsilon < 2\varepsilon$$

Letting $\varepsilon \to 0$, we get what we want

In this way \mathbb{R}^{∞} is a metrizable space.

Example 4. $X, \rho(x,y) \stackrel{\text{def}}{=} \begin{cases} 0, & x=y \\ 1, & x \neq y \end{cases}$ — Discrete metric.

$$x_n \to x, \ \varepsilon = \frac{1}{2}, \ \exists M : \ m > M \implies \rho(x_m, x) < \frac{1}{2} \implies \rho(x_m, x) = 0 \implies x_m = x$$

Definition.

$$(X,\tau); \ \forall A \subset X;$$

$$\operatorname{Int}(A) \stackrel{\text{def}}{=} \bigcup_{G \subset A} G$$
 is open;

$$Cl(A) \stackrel{\text{def}}{=} \bigcap_{A \in C} G$$
 is closed;

$$\operatorname{Fr}(A) \stackrel{\operatorname{def}}{=} \operatorname{Cl}(A) \setminus \operatorname{Int}(A)$$

 (X, ρ) ; Having a metric space one can describe closure of a set.

$$\rho(x, A) \stackrel{\text{def}}{=} \inf_{a \in A} \rho(x, a)$$
$$\rho(A, B) \stackrel{\text{def}}{=} \inf_{\substack{a \in A \\ b \in B}} \rho(a, b)$$
$$\rho(x, A) = f(x), x \in X$$

Statement 1.3. Function f(x) is continuous.

$$\begin{aligned} & \textit{Proof.} \ \, \forall x,y \in X \\ & f(x) = \rho(x,A) \underset{\forall \alpha \in A}{\leqslant} \rho(x,\alpha) \leqslant \rho(x,y) + \rho(y,\alpha) \\ & \forall \varepsilon > 0 \ \exists \alpha_{\varepsilon} \in A : \ \, \rho(y,\alpha_{\varepsilon}) < \rho(y,A) + \varepsilon = f(y) + \varepsilon \\ & f(x) \leqslant f(y) + \varepsilon + \rho(x,y), \ \, \varepsilon \to 0 \\ & \begin{cases} f(x) \leqslant f(y) + \rho(x,y) \\ f(y) \leqslant f(x) + \rho(x,y) \end{cases} \implies |f(x) - f(y)| \leqslant \rho(x,y) \end{aligned}$$

Statement 1.4.
$$x \in Cl(A) \iff \rho(x,A) = 0$$

Let's look at the metric spaces in terms of separation of sets from each other by open sets.

$$x, y$$

$$r = \rho(x, y) > 0$$

$$B_{\frac{r}{3}}(x), B_{\frac{r}{3}}(y)$$

In any metric space separability axiom is true.

Theorem 1.1.

Any metric space is a normal space, i.e. \forall closed disjoint $F_1, F_2 \in X$, \exists open disjoint $G_1, G_2 \colon F_j \in G_j$, j = 1, 2

Proof.
$$g(x) = \frac{\rho(x, F_1)}{\rho(x, F_1) + \rho(x, F_2)}$$
 — continuous on X $x \in F_1$, $Cl(F_1) = F_1$, $\rho(x, F_1) = 0$, $g(x) = 0$ $x \in F_2$, $g(x) = 1$

Let's look at $(-\infty; \frac{1}{3}), (\frac{2}{3}, \infty)$ — by continuity their inverse images under g are open.

$$G_1 = g^{-1}(-\infty; \frac{1}{3})$$

 $G_2 = g^{-1}(\frac{2}{3}; \infty)$

Definition. Metric space is **complete** if $\rho(x_n, x_m) \to 0 \implies \exists x = \lim x_n \mathbb{R}^{\infty}$ – complete (by completeness of the rational numbers). In complete metric spaces the nested balls principle is true.

Theorem 1.2.

X – complete metric space, \overline{V}_{r_n} – system of closed balls.

1.
$$\overline{V}_{r_{n+1}} \subset \overline{V}_{r_n}$$
 – the system is nested.

$$2. r_n \rightarrow 0$$

$$\underline{Then:} \bigcap_{n} \overline{V}_{r_n} = \{a\}$$

Proof. Let b_n be centers of \overline{V}_{r_n} , $m \ge n$, $b_m \in \overline{V}_{r_n}$, $\rho(b_m, b_n) \le r_n \to 0 \ \forall m \ge n$ $\rho(b_m, b_n) \to 0 \Longrightarrow \exists a = \lim b_n \text{ Since the balls are closed } a \in \text{ every ball.}$ $r_n \to 0 \Longrightarrow \text{ there is only one common point.}$

 (X, τ) — topological space $A \subset X, \ \tau_a = \{ G \cap A, G \in \tau \}$ — topology induced on A

Definition. X— metric space, $A \subset X$, Cl(A) = X

Then: $A - \mathbf{dense}$ in X

 $\underline{\text{If}} \operatorname{Int}(\operatorname{Cl}(A)) = \emptyset \text{ A - nowhere dense in X}.$

Note. It is easy to understand, that in metric spaces nowhere density means the following: \forall ball $V \exists V' \subset V \colon V'$ contains no points from A.

Definition. X is called **first Baire category set**, if it can be written as at most countable union of x_n each nowhere dense in X.

Theorem 1.3 (Baire category theorem).

Complete metric space is second Baire category set in itself.

Proof. Let X be first Baire category set.

 $X = \bigcup_{n \in \mathbb{Z}} X_n \quad \forall \overline{V} \ X_1 \text{ is nowhere dense.}$

 $\overline{V}_1 \subset \overline{\overline{V}} \colon \overline{V}_1 \cap X_1 = \emptyset$ $X_2 \text{ is nowhere dense } \overline{V}_2 \subset \overline{V}_1 \colon \overline{V}_2 \cap X_2 = \emptyset$ $r_2 \leqslant \frac{r_1}{2}$

:

$$\overline{\{\overline{V}_n\}}, \ r_n \to 0, \ \bigcap_n \overline{V}_n = \{a\}, \ X = \bigcup X_n, \ \exists n_0 \colon a \in X_{n_0} \\
X_{n_0} \cap \overline{V}_{n_0} = \varnothing \to \leftarrow \ a \in \overline{V}_{n_0}$$

Corollary 1.3.1. Complete metric space without isolated points is uncountable.

Proof. No isolated points are present \implies every point in the set is nowhere dense in it. Let X be countable: $X = \bigcup_n \{X_n\}$, then it is first Baire category set. $\rightarrow \leftarrow$

Definition. K — compact if

1.
$$K = \operatorname{Cl}(K)$$

2.
$$x_n \in K \exists n_1 < n_2 < \dots \ x_{n_j}$$
 – converges in X .

If only 2 is present, the set is called **precompact**.

Theorem 1.4 (Hausdorff).

Let X — metric space, K — closed in X. Then: K — compact $\iff K$ — totally bounded, i.e. $\forall \varepsilon > 0 \; \exists a_1, \ldots, a_p \in X \colon \; \forall b \in K \; \exists a_j \colon \; \rho(a_j, b) < \varepsilon$ $(a_1, \ldots, a_p - finite \; \varepsilon - net)$

Proof.

• Totally bounded \implies compact

K is totally bounded, $x_n \in K$ $n_1 < n_2 < \cdots < n_k < \cdots, x_n$ converges in K

$$\varepsilon_k \downarrow \to 0 \ \varepsilon_1 \quad K \subset \bigcup_{j=1}^p \overline{V}_j, \ rad = \varepsilon_1 \qquad (\varepsilon_1\text{-net})$$

n is finite \implies one ball will contain infinetely many x_n elements.

Let's look at $\overline{V}_{j_0} \cap K$ — totally bounded = K_1 , diam $(K_1) \leq 2\varepsilon_1$ ε_2 $K_1 \subset \bigcup_{j=1}^n \overline{V'}_j$, $rad = \varepsilon_2$.

Then one of $\overline{V'}$ contains infinitely many elements of the sequence contained in K_1 .

 $\overline{V'}_{j_0} \cap K_1 = K_2$, diam $(K_2) \leq 2\varepsilon_2$ and so on.

$$K_n \supset K_{n+1} \supset K_{n+2} \supset \dots$$
, diam $(K_N) \leqslant 2\varepsilon_n \xrightarrow{\text{by compl.}} \bigcap_{n=1}^{\infty} K_n \neq \emptyset$

Take x_{n_1} from K_1 , x_{n_2} from K_2 ...

 \bullet Compact \Longrightarrow totally bounded

K — compact $\forall \varepsilon \exists$ finite ε -net?

By contradiction: $\exists \varepsilon_0 > 0$: finite ε_0 -net is impossible to construct.

 $\forall x_1 \in K \ \exists x_2 \in K \colon \ \rho(x_1, x_2) > \varepsilon_0 \ (\text{or else system of } x_1 - \text{finite } \varepsilon\text{-net})$

 $\{x_1, x_2\}$ - choose $x_3 \in K : \rho(x_3, x_i) > \varepsilon_0, i = 1, 2$ and so on.

 $x_n \in K: n \neq m \ \rho(x_n, x_m) > \varepsilon_0$ — contains no converging subsequence \implies set is not a compact. $\rightarrow \leftarrow$

§1.2 Normed spaces

Definition. X — linear set, x + y, $\alpha \cdot x$, $\alpha \in \mathbb{R}$

The purpose of norm definition, is to construct a topology on X, so that 2 linear operations are continuous on it.

$$\varphi \colon X \to \mathbb{R}$$
:

1.
$$\varphi(x) \geqslant 0$$
, $= 0 \iff x = 0$

2.
$$\varphi(\alpha x) = |\alpha|\varphi(x)$$

3.
$$\varphi(x+y) \leqslant \varphi(x) + \varphi(y)$$

Definition. φ — **norm** on X, $\varphi(x) = ||x||$

$$\rho(x,y) \stackrel{\text{def}}{=} ||x-y||$$
 — metric on X .

Definition.

 $(X, \|\cdot\|)$ — **normed space** — special case of metrical space.

$$x = \lim x_n \stackrel{def}{\Longrightarrow} \rho(x_n, x) \to 0 \iff ||x_n - x|| \to 0$$

Statement 1.5. In the topology of a normed space linear operations are continuous on X.

Proof.

1.
$$x_n \to x, \ y_n \to y; \ \|(x_n + y_n) - (x + y)\| = \|(x_n - x) + (y_n - y)\| \le$$

$$\le \underbrace{\|x_n - x\|}_{\downarrow_0} + \underbrace{\|y_n - y\|}_{\downarrow_0}$$

$$\implies x_n + y_n \to x + y$$

2.
$$\alpha_n \to \alpha$$
, $x_n \to x$; $\|\alpha_n x_n - \alpha x\| = \|(\alpha_n - \alpha)x_n + \alpha(x_n - x)\| \le \underbrace{|\alpha_n - \alpha|}_{\text{bounded}} \cdot \underbrace{\|x_n\|}_{\text{bounded}} + \underbrace{\alpha \|x_n - x\|}_{\text{0}}$

$$x_n \to x \implies \|x_n\| \text{ — bounded.}$$

$$\alpha_x x_n \to \alpha x$$

Statement 1.6. From the triangle inequality $|||x|| - ||y||| \le ||x - y||$ $x_n \to x \implies ||x_n|| \to ||x||$ Norm is continious.

Example 5. \mathbb{R}^n

1.
$$\|\bar{x}\| = \sqrt{\sum_{k=1}^{n} x_k^2}$$

2.
$$\|\bar{x}\|_1 \stackrel{\text{def}}{=} \sum_{k=1}^n |x_k|$$

3.
$$\|\bar{x}\|_2 \stackrel{\text{def}}{=} \max\{|x_1|, \dots, |x_n|\}$$

4.
$$C[a,b]$$
 — functions continuous on $[a,b]$; $||f|| = \max_{x \in [a,b]} |f(x)|$

5.
$$L_p(E) = \left\{ f - \text{measurable, } \int_E |f|^p < +\infty \right\}$$

$$p \geqslant 1, \ ||f||_p = \left(\int_E |f|^p \right)^{\frac{1}{p}}$$

Because the set of points is the same, arises the question about convergence comparison. $\|\cdot\|_1 \sim \|\cdot\|_2, \ x_n \stackrel{\|\cdot\|_1}{\to} x \iff x_n \stackrel{\|\cdot\|_2}{\to} x$

Statement 1.7.

$$\|\cdot\|_1 \sim \|\cdot\|_2 \iff \exists a, b > 0 \colon \forall x \in X \implies a\|x_1\|_1 \leqslant \|x\|_2 \leqslant b\|x\|_1$$

Theorem 1.5 (Riesz).

X, dim $X < +\infty$ — linear set.

Then: Any pair of norms in X are equivalent.

Proof. l_1, \ldots, l_n — linearly independent from X. $\forall x \in X = \sum_{k=1}^n \alpha_k l_K$

$$\bar{x} \leftrightarrow (l_1, \dots, l_n) = \bar{l} \in \mathbb{R}^n$$

Let $\|\cdot\|$ — some norm in X.

$$||x|| \leq \sum_{k=1}^{n} ||l_k|| |\alpha_k| \leq \sum_{\text{Cauchy}} \underbrace{\sqrt{\sum_{k=1}^{n} ||l_k||^2}}_{\text{const}(B), B-basis} \sqrt{\sum_{k=1}^{n} ||\alpha_k||^2} ||\alpha_k||^2$$

$$||x||_1 = \sqrt{\sum_{k=1}^n ||\alpha_k||^2}, \ x = \sum \alpha_k l_k$$

 $||x|| \leqslant b||x||_1$

? $\exists a > 0 \colon a \|x\|_1 \leqslant \|x\| \implies \|\cdot\| \sim \|\cdot\|_1$

Let
$$f(\alpha_1, \dots, \alpha_n) = \left\| \sum_{k=1}^n \alpha_k l_k \right\|$$

$$|f(\bar{\alpha} + \Delta \bar{\alpha}) - f(\bar{\alpha})| = \left\| \sum_{k=1}^{n} \alpha_{k} l_{k} + \sum_{k=1}^{n} \Delta \alpha_{k} l_{K} \right\| - \left\| \sum_{k=1}^{n} \alpha_{k} l_{k} \right\| \leq \left\| \sum_{k=1}^{n} \Delta \alpha_{k} l_{k} \right\| \leq \sum_{\substack{0, \Delta \alpha_{k} \to 0}} \|l_{k}\| |\Delta \alpha_{k}| \implies f \text{ is continuous on } \mathbb{R}^{n}$$

$$S_1 = \left\{ \sum_{k=1}^n |\alpha_k|^2 = 1 \right\} \subset \mathbb{R}^m$$
, f — continuous on S_1 , S_1 — compact, $\bar{\alpha}^* \in S_1$

By Weierstrass theorem there exists a point $\alpha^* \in S_1$ on a sphere, in which function f achieves its minimum $\implies \forall \alpha \in S_1 \ f(\bar{\alpha}^*) \leqslant f(\bar{\alpha})$

If
$$f(\bar{\alpha}^*) = 0$$
, then $\left\| \sum_{k=1}^n \alpha_k^* l_k \right\| = 0 \implies \sum_{k=1}^n \alpha_k^* l_k = 0$, $\bar{\alpha}^* \in S_1$, but $l_1 \dots l_n$ are linearly independent $\to \leftarrow \implies \min_{S_1} f = m > 0$

$$||x|| = \left\| \sum_{k=1}^{n} \alpha_k l_k \right\| = f(\bar{\alpha}) = \sqrt{\sum_{k=1}^{n} \alpha_k^2} \cdot \left\| \sum \left[\frac{\alpha_k}{\sqrt{\sum_{k=1}^{n} \alpha_k^2}} \cdot l_k \right] \right\| \geqslant , \ \bar{\beta} = (\beta_1 \dots \beta_n) \in S_1$$

$$\geqslant m \cdot ||x||_1, \ a = m$$

Corollary 1.5.1.
$$X$$
 — NS , $Y \subset X$, $\dim Y < +\infty \implies Y = \operatorname{Cl}(Y)$

Note. Functional analysis differentiates between linear subset (set of points, closed by addition and scalar multiplication) and linear subspace (closed linear subset).

Proof.
$$Y = \mathcal{L}(l_1, \dots, l_n) = \left\{ \sum_{i=1}^n \alpha_i l_i \mid l_1, \dots, l_n - \text{lin. indep.} \right\}$$

$$y_m \in Y, y_m \to y \text{ in } X \implies y \in Y?$$

$$||y_m - y|| \to 0 \implies ||y_m - y_p|| \to 0, \ m, p \to \infty$$

$$||y||, y \in Y.$$

By Riesz theorem all norm in Y are equivalent.

$$y = \sum_{j=1}^{n} \alpha_{j} l_{J}, \|y\|_{0} = \sqrt{\sum_{j=1}^{n} \alpha_{j}^{2}} \quad \text{some norm (by linear independence)}.$$
By Riesz theorem $\|y\| \sim \|y\|_{0}$

$$\|y_{m} - y_{p}\| \to 0 \implies \|y_{m} - y_{p}\|_{0} \to 0$$

$$\bar{\alpha} = (\alpha_{1}, \dots, \alpha_{n}) \in \mathbb{R}^{n} \ y_{m} = \sum_{i=1}^{n} \alpha_{i}^{(m)} l_{i}$$

$$|\alpha_{i}^{(m)} - \alpha_{i}^{(l)}| \to 0 \ \forall i = 1, \dots, n; \ \bar{\alpha} = (\alpha_{1}^{(m)}, \dots, \alpha_{n}^{(m)}) \to \alpha^{*} = (\alpha_{1}^{*}, \dots, \alpha_{n}^{*})$$

$$y^{*} = \sum_{i=1}^{n} \alpha_{i}^{*} l_{I} \in Y, \ \|y_{m} - y\| \to 0, y = y^{*} \implies y \in Y$$

Definition. If normed space if complete, then it is called **B-space** or **Banach space**.

Example 6. C[a, b] — functions continuous on [a, b].

Example 7. Lebesgue space, $p \ge 1$, $L_p(E) = \left\{ f \text{ is measurable on } E, \int_E |f|^P < +\infty \right\}$.

If
$$X$$
 — Banach space,

$$\sum_{n=1}^{\infty} x_n = \lim_{n \to \infty} \sum_{k=1}^{n} x_k, \sum_{n=1}^{\infty} \|x_n\| < +\infty$$

$$\|S_n - S_m\| = \left\| \sum_{k=m+1}^{n} x_k \right\| \leqslant \sum_{m+1}^{n} \|x_k\| \xrightarrow[n,m \to \infty]{} 0$$

$$\implies \|S_n - S_m\| \to 0 \implies \exists \lim_{n \to \infty} S_n, \sum_{k=1}^{n} x_k \text{— converges.}$$

In Banach spaces works the theory of absolute convergence of numerical series.

Lemma 1 (Riesz's lemma about almost perpendicular). Y — eigen subspace of X — normed space. $\forall \varepsilon \in (0,1) \exists z_{\varepsilon} \in X$:

1.
$$||z_{\varepsilon}|| = 1$$

2.
$$\rho(z_{\varepsilon}, Y) > 1 - \varepsilon$$

Proof.
$$\exists x \notin Y \ d = \rho(x, Y), \ d = 0 \ \exists y_n \in Y \colon \|x - y_n\| < \frac{1}{n}, \ n \to \infty, \ y_n \to x$$

$$Y = \operatorname{Cl}(Y) \implies x \in Y \to \leftarrow x \notin Y, \ d > 0$$

$$\forall \varepsilon \in (0, 1) \ \frac{1}{1 - \varepsilon} > 1 \ \exists y_\varepsilon \in Y \colon \|x - y_\varepsilon\| < \frac{1}{1 - \varepsilon} d$$

$$z_{\varepsilon} = \frac{x - y_{\varepsilon}}{\|x - y_{\varepsilon}\|}, \ \|z_{\varepsilon}\| = 1, \ \forall y \in Y \ \|z_{\varepsilon} - y\| = \left\| \frac{x - y_{\varepsilon}}{\|x - y_{\varepsilon}\|} \right\| = \frac{\|x - \underbrace{(y_{\varepsilon} + \|x - y_{\varepsilon}\| \cdot y)}_{\in Y}\| \ge d}{\|x - y_{\varepsilon}\| < \frac{1}{1 - \varepsilon} d} > \frac{1}{\|x - y_{\varepsilon}\|}$$

 $1 - \varepsilon$ Corollary 1.5.2 dim $Y - +\infty$ S = sphere in Y $x_2 - 1 \int x ||x|| - 1 \rangle \longrightarrow S = not e$

Corollary 1.5.2. dim $X = +\infty$, S — sphere in X, $r_S = 1 \{ x \mid ||x|| = 1 \} \implies S$ — not a compact.

Proof. $\forall x_1 \in S, \ Y_1 = \mathcal{L}\{x_1\}$ — finite dimensional linear set. \implies closed in $X \implies Y_1$ — subspace. dim $X = +\infty \implies Y_1$ — eigen subspace. Then by the Riesz lemma:

$$\exists x_2 \in S \colon ||x_2 - x_1|| > \frac{1}{2}$$
$$Y_2 = \mathcal{L}\{x_1, x_2\} \ \exists x_3 \in S \colon ||x_3 - x_j|| > \frac{1}{2}, \ j = 1, 2$$

Continue by induction. Because $\dim X = +\infty$ the process will never stop. $x_n \in S \colon \|x_n - x_m\| > \frac{1}{2}, \ n \neq m$ — obviously we cannot extract converging subsequence. $\implies S$ — not a compact.

§1.3 Inner product (unitary) spaces

Definition. X — linear space.

$$\varphi \colon X \times X \to \mathbb{R}$$

1.
$$\varphi(x,x) \geqslant 0$$
, $\varphi(x,x) = 0 \iff x = 0$

2.
$$\varphi(x,y) = \varphi(y,x)$$

3.
$$\varphi(\alpha x + \beta y, z) = \alpha \varphi(x, z) + \beta \varphi(y, z)$$

 φ — inner product.

$$\varphi(x,y) = \langle x, y \rangle$$

Definition. $(X, \langle \cdot, \cdot \rangle)$ — inner product space.

Example 8.
$$\mathbb{R}^n, \langle \bar{x}, \bar{y} \rangle = \sum_{j=1}^n x_j y_j$$

Statement 1.8 (Schwarz). $\forall x, y \in X \quad |\langle x, y \rangle| \leq \sqrt{\langle x, x \rangle} \cdot \sqrt{\langle y, y \rangle}$

Proof. $\lambda \in \mathbb{R}$

$$f(\lambda) = \langle \lambda x + y, \lambda x + y \rangle \geqslant 0$$

$$\parallel$$

$$\lambda^{2} \langle x, x \rangle + 2\lambda \langle x, y \rangle + \langle y, y \rangle$$

$$D = 4 \langle x, y \rangle^{2} - 4 \langle x, x \rangle \cdot \langle y, y \rangle \leqslant 0 \quad \blacksquare$$

Corollary 1.5.3 (Cauchy inequality for sums). Consider $X = \mathbb{R}^n$, $||x|| \stackrel{def}{=} \sqrt{\langle x, x \rangle}$. Then

$$||x + y||^2 = \langle x + y, x + y \rangle = ||x||^2 + 2 \cdot \langle x, y \rangle + ||y||^2 \leqslant (||x|| + ||y||)^2$$

Any inner product space is a special case of a normed space. The specifics is that we can measure the angles between points:

$$x \perp y \iff \langle x, y \rangle = 0$$

In this case the Pythagorean theorem takes place:

$$||x + y||^2 = ||x||^2 + ||y||^2$$

In inner product spaces the parallelogram law plays a significant role:

$$||x + y||^2 + ||x - y||^2 = 2||x||^2 + 2||y||^2 \quad \forall x, y \in X$$

In an inner product space norm is determined by inner product: $||x||^2 = \langle x, x \rangle$ It can be proved that if parallelogram law holds, then the norm must be determined by some inner product. Let X be some normed space, $x \in X$, then $\langle \cdot, \cdot \rangle \mapsto ||x|| = \sqrt{\langle x, x \rangle}$. For any norm satisfying the parallelogram law, the inner product generating the norm is unique. **Example 9.** $C_{[a,b]}$, $||f|| = \max_{x \in [a,b]} |f(x)|$, ||f|| doesn't satisfy the parallelogram law and thus is not determined by any inner product. This fact implies that $C_{[a,b]}$ is not an inner product space.

Definition. Orthonormal set — a set of points $\{l_1, l_2, ...\}$ (may be finite):

- 1. $||l_i|| = 1$
- 2. $l_i \perp l_i$, $i \neq j$

Every orthonormal set is linearly independent.

Definition. Let $x \in X$, $\{l_i\}$ — ONS. Then $\langle x, l_j \rangle$ — **Fourier coefficient**, $\sum_i \langle x, l_j \rangle l_j$ — **Fourier series** of point x.

Fourier series is a special case of orthogonal series.

Definition. $\sum_{j} x_{j}$ — orthogonal series $\iff x_{i} \perp x_{j}, i \neq j$

Let
$$\sum_{j=1}^{\infty} x_j$$
, $S_m = \sum_{j=1}^{m} x_j$. Then

$$||S_m||^2 = \langle \sum_{j=1}^m x_j, \sum_{j=1}^m x_j \rangle = \sum_{j=1}^m ||x_j||^2$$

This fact allows us to effectively build the theory of orthogonal series.

An important problem is concerned with Fourier series. Let X is a normed space, Y is a subspace of X,

$$\forall x \in X \quad E_Y(x) = \rho(x, Y) = \inf_{y \in Y} ||x - y||$$

Definition. $E_Y(x)$ — **best approximation** of point x with points of the subspace Y, if $\exists y^* \in Y \quad E_y(x) = ||x - y^*||$ — then y^* is an element of best approximation.

Theorem 1.6 (Borel).

 $\dim Y < +\infty \implies \forall x \in X \quad \exists y^* \in Y - element \ of \ best \ approximation.$

Proof.
$$Y = \mathcal{L}(l_1, l_2, \dots, l_n)$$
lin. indep.

Consider $f(\alpha_1, \ldots, \alpha_n) = ||x - \sum_{k=1}^n \alpha_k l_k|| \to \text{min.}$ By the triangle inequality for norm, $f(\bar{\alpha})$ is continuous on \mathbb{R}^n , $f \geq 0$, $E_Y(x) = \inf f(\bar{\alpha})$. It is easy to find that there always is a ball $B(0,r) \subset \mathbb{R}^n$, outside of which $f > 2E_Y(x)$. So, $E_Y(x)$ is somewhere inside. But f is continuous, the ball is compact, so, by the Weierstrass theorem, the minimum exists and it is located on the sphere S(0,r).

For abstract Fourier series the Borel theorem can be significantly strengthened by specifying the best approximation element.

Theorem 1.7 (extreme quality of Fourier series' partial sums).

$$\{e_j\} - ONS \text{ in } X$$

$$H_n = \mathcal{L}(l1, \dots, l_n)$$

$$E_{H_n}(x), S_n(x) = \sum_{j=1}^n \langle x, l_j \rangle l_j \implies E_{H_n}(x) = ||x - S_n(x)||$$

Proof.
$$y = \sum_{j=1}^{n} \alpha_j l_j \in H_n$$

$$||x - y||^2 = \left\langle x - \sum_{i} \alpha_j l_j, x - \sum_{i} \alpha_j l_j \right\rangle = ||x||^2 - 2 \sum_{i} \alpha_j \langle x, l_j \rangle + \sum_{i} \alpha_j^2 =$$

$$= ||x||^2 + \sum_{i} (\alpha_j - \langle x, l_j \rangle)^2 - \sum_{i} \langle x, l_j \rangle^2 \to \min$$

So, the sum goes to minimum when the second summand is minimal. Obviously, it's minimal when $\forall (\alpha_j - \langle x, l_j \rangle) = 0$. $E_Y(x)$ — Fourier sum.

Corollary 1.7.1 (Bessel's inequality). $\sum_{j} \langle x, l_j \rangle^2 \leq ||x||^2$

Proof.
$$0 \le ||x - y^*||^2 = ||x||^2 - \sum_{i} \langle x, l_i \rangle^2$$

Corollary 1.7.2. The series of Fourier coefficients' squares always converges