## 1.3 Hilbert space methods

One of the most often used theorems of functional analysis is the Riesz representation theorem.

Theorem 1.30 (Riesz representation theorem). If  $x^*$  is a continuous linear functional on a Hilbert space H, then there is exactly one element  $y \in H$ such that

$$
\langle x^*, x \rangle = (x, y). \tag{1.32}
$$

### 1.3.1 To identify or not to identify–the Gelfand triple

Riesz theorem shows that there is a canonical isometry between a Hilbert space H and its dual  $H^*$ . It is therefore natural to identify H and  $H^*$  and is done so in most applications. There are, however, situations when it cannot be done.

Assume that H is a Hilbert space equipped with a scalar product  $(\cdot, \cdot)_H$ and that  $V \subset H$  is a subspace of H which is a Hilbert space in its own right, endowed with a scalar product  $(\cdot, \cdot)_V$ . Assume that V is densely and continuously embedded in H that is  $\overline{V} = H$  and  $||x||_H \le c||x||_V$ ,  $x \in V$ , for some constant c. There is a canonical map  $T: H^* \to V^*$  which is given by restriction to V of any  $h^* \in H^*$ :

$$
\langle Th^*, v \rangle_{V^* \times V} = \langle h^*, v \rangle_{H^* \times H}, \quad v \in V.
$$

We easily see that

$$
||Th^*||_{V^*} \le C||h^*||_{H^*}.
$$

Indeed

$$
||Th^*||_{V^*} = \sup_{||v||_V \le 1} |\langle Th^*, v \rangle_{V^* \times V}| = \sup_{||v||_V \le 1} |\langle h^*, v \rangle_{H^* \times H}|
$$
  

$$
\le ||h^*||_{H^*} \sup_{||v||_V \le 1} ||v||_H \le c||h^*||_{H^*}.
$$

Further, T is injective. For, if  $Th_1^* = Th_2^*$ , then

$$
0=_{V^*\times V}=_{H^*\times H}
$$

for all  $v \in V$  and the statement follows from density of V in H. Finally, the image of  $TH^*$  is dense in  $V^*$ . Indeed, let  $v \in V^{**}$  be such that  $\langle v, Th^* \rangle = 0$ for all  $h^* \in H^*$ . Then, by reflexivity,

$$
0=_{V^{**}\times V^*}=_{V^*\times V}=_{H^*\times H},\quad h^*\in H^*
$$

implies  $v = 0$ .

Now, if we identify  $H^*$  with H by the Riesz theorem and using T as the canonical embedding from  $H^*$  into  $V^*$ , one writes

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$$
V\subset H\simeq H^*\subset V^*
$$

and the injections are dense and continuous. In such a case we say that  $H$  is the pivot space. Note that the scalar product in  $H$  coincides with the duality pairing  $\langle \cdot, \cdot \rangle_{V^* \times V}$ :

$$
(f,g)_H = _{V^* \times V}, \quad f \in H, g \in V.
$$

Remembering now that V is a Hilbert space with scalar product  $(\cdot, \cdot)_V$  we see that identifying also  $V$  with  $V^*$  would lead to an absurd – we would have  $V = H = H^* = V^*$ . Thus, we cannot identify simultaneously both pairs. In such situations it is common to identify the pivot space  $H$  with its dual  $H^*$  bur to leave V and  $V^*$  as separate spaces with duality pairing being an extension of the scalar product in H.

An instructive example is  $H = L_2([0,1], dx)$  (real) with scalar product

$$
(u,v) = \int_{0}^{1} u(x)v(x)dx
$$

and  $V = L_2([0, 1], w dx)$  with scalar product

$$
(u, v) = \int_{0}^{1} u(x)v(x)w(x)dx,
$$

where  $w$  is a nonnegative bounded measurable function. Then it is useful to identify  $V^* = L_2([0, 1], w^{-1}dx)$  and

$$
_{V^* \times V} = \int_{0}^{1} f(x)g(x)dx \leq \int_{0}^{1} f(x)\sqrt{w(x)} \frac{g(x)}{\sqrt{w(x)}} dx \leq ||f||_{V} ||g||_{V^*}.
$$

## 1.3.2 The Radon-Nikodym theorem

Let  $\mu$  and  $\nu$  be finite nonnegative measures on the same  $\sigma$ -algebra in  $\Omega$ . We say that  $\nu$  is absolutely continuous with respect to  $\mu$  if every set that has  $\mu$ -measure 0 also has  $\nu$  measure 0.

**Theorem 1.31.** If  $\nu$  is absolutely continuous with respect to  $\mu$  then there is an integrable function g such that

$$
\nu(E) = \int_{E} g d\mu,\tag{1.33}
$$

for any  $\mu$ -measurable set  $E \subset \Omega$ .

*Proof.* Assume for simplicity that  $\mu(\Omega), \nu(\Omega) < \infty$ . Let  $H = L_2(\Omega, d\mu + d\nu)$ on the field of reals. Schwarz inequality shows that if  $f \in H$ , then  $f \in L_1(d\mu +$  $d\nu$ ). The linear functional

$$
\langle x^*, f \rangle = \int_{\Omega} f d\mu
$$

To be completed

#### 1.3.3 Projection on a convex set

Corollary 1.32. Let K be a closed convex subset of a Hilbert space H. For any  $x \in H$  there is a unique  $y \in K$  such that

$$
||x - y|| = \inf_{z \in K} ||x - z||. \tag{1.34}
$$

Moreover,  $y \in K$  is a unique solution to the variational inequality

$$
(x - y, z - y) \le 0 \tag{1.35}
$$

for any  $z \in K$ .

$$
\parallel x-z_{n}\parallel \leq d+\frac{1}{4}
$$

real

**Proof.** Let  $d = \inf_{z \in K} ||x - z||$ . We can assume  $\chi \notin K$  and so  $d > 0$ . Consider  $f(z) = ||x - z||, z \in K$  and consider a minimizing sequence  $(z_n)_{n \in \mathbb{N}}, z_n \in K$ such that  $d \le f(z_n) \le d + 1/n$ . By the definition of  $f, (z_n)_{n \in \mathbb{N}}$  is bounded and thus it contains a weakly convergent subsequence, say  $(\zeta_n)_{n\in\mathbb{N}}$ . Since K is closed and convex, by Corollary 1.24,  $\zeta_n \rightharpoonup y \in K$ . Further we have  $|(h, x-y)| = \lim_{n \to \infty} |(h, x-\zeta_n)| \le ||h|| \liminf_{n \to \infty} ||x-\zeta_n|| \le ||h|| \liminf_{n \to \infty} d + \frac{1}{n}$  $\frac{1}{n} = ||h||d$ 

for any  $h \in H$  and thus, taking supremum over  $||h|| \leq 1$ , we get  $f(y) \leq d$ which gives existence of a minimizer.

To prove equivalence of (1.35) and (1.34) assume first that  $y \in K$  satisfies (1.34) and let  $z \in K$ . Then, from convexity,  $v = (1-t)y + tz \in \overline{K}$  for  $t \in [0,1]$ and thus

$$
||x - y|| \le ||x - ((1 - t)y + tz)|| = ||(x - y) - t(z - y)||
$$

$$
= ((x-1)-(5-1)^\prime (x-1)-f(x-1))
$$

and thus

$$
\|\mathbf{x} \leq y\|^2 \leq \|\mathbf{x} \leq y\|^2 - 2t(x - y, z - y) + t^2\|z - y\|^2.
$$

Hence

$$
t||z - y||^2 \ge 2(x - y, z - y)
$$

for any  $t \in (0, 1]$  and thus, passing with  $t \to 0$ ,  $(x - y, z - y) \leq 0$ . Conversely, assume (1.35) is satisfied and consider

28 1 Basic Facts from Functional Analysis and Banach Lattices<br>  $(x, y) - (y, y) - (y, y) - (y, y) - (y, y)$ 

$$
||x - y||^{2} - ||x - z||^{2} = (x - y, x - y) - (x - z, x - z) \quad - (2,2)
$$
  
\n
$$
\sum_{n=1}^{\infty} \frac{2(x, z) - 2(x, y) + 2(y, y) - 2(y, z) + 2(y, z) - (y, y)}{(y - z, y - z)} \quad (y, y)
$$
  
\nhence  
\n
$$
2 (x, z) - 2(x, y) - 2(y, z) + 2(y, z) - (y, y)
$$
  
\nhence  
\n
$$
||x - y|| \le ||x - z|| \quad - (z, z) + 2(y, z)
$$
  
\nfor any  $z \in K$ .  
\nFor uniqueness, let  $y_1, y_2$  satisfy  
\n
$$
(x - y_1, y_2 - y_3) \in O
$$

For uniqueness, let  $y_1, y_2$  satisfy

$$
(x - y_1, z - y_1) \le 0, \qquad (x - y_2, z - y_2) \le 0, \qquad z \in H.
$$

Choosing  $z = y_2$  in the first inequality and  $z = y_1$  in the second and adding them, we get  $||y_1 - y_2||^2 \leq 0$  which implies  $y_1 = y_2$ .

We call the operator assigning to any  $x \in K$  the element  $y \in K$  satisfying (1.34) the projection onto K and denote it by  $P_{\mathbf{k}}$ .

**Proposition 1.33.** Let  $K$  be a nonempty closed and convex set. Then  $P_K$  is non expansive mapping.

*Proof.* Let  $y_i = P_K x_i$ ,  $i = 1, 2$ . We have

$$
(x_1 - y_1, z - y_1) \le 0,
$$
  $(x_2 - y_2, z - y_2) \le 0,$   $z \in H$ 

so choosing, as before,  $z = y_2$  in the first and  $z = y_1$  in the second inequality and adding them together we obtain

$$
\|y_1 - y_2\|^2 \le (x_1 - x_2, y_1 - y_2),
$$
\n
$$
\|y_1 - y_2\|^2 \le (x_1 - x_2, y_1 - y_2),
$$
\n
$$
\|P_K x_1 - P_K x_2\| \le \|x_1 - x_2\|
$$
\n
$$
\|P_K x_2 - P_K x_2\| \le \|x_1 - x_2\|
$$
\n
$$
\|P_K x_3 - P_K x_3\|^2 \le (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2 + (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2
$$
\n
$$
\le (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2 + (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2
$$
\n
$$
\le (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2 + (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2
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\n
$$
\le (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2 + (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2
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\n
$$
\le (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2 + (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2
$$
\n
$$
\le (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2 + (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2
$$
\n
$$
\le (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2 + (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2
$$
\n
$$
\le (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2 + (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2
$$
\n
$$
\le (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2 + (|x_1 - x_1|)^2 \cdot (|x_1 - x_1|)^2
$$
\n
$$
\le
$$

Consider the Dirichlet problem for the Laplace equation in  $\Omega \subset \mathbb{R}^n$ 

$$
-\Delta u = f \qquad \text{in} \quad \Omega,\tag{1.36}
$$

$$
u|_{\partial\Omega} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \tag{1.37}
$$

Assume that there is a solution  $u \in C^2(\Omega) \cap C(\overline{\Omega})$ . If we multiply  $(1.3\overline{\text{S}})$  by a test function  $\phi \in C_0^{\infty}(\Omega)$  and integrate by parts, then we obtain the problem

$$
\int_{\mathcal{R}} \Delta u \, \rho \, h \, \bar{f} \quad \int_{\mathcal{R}} \varphi \, \nabla u \cdot \nabla \, d \, \bar{g} \quad - \int_{\Omega} \nabla u \, \nabla \varphi \, d \, \bar{f} = \int_{\Omega} f \phi \, dx. \tag{1.38}
$$

$$
\int (u) = k \int \sqrt{1 + p} \sqrt{1} du
$$

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Conversely, if u satisfies  $(1.38)$ , then it is a distributional solution to  $(1.37)$ .

Moreover, if we consider the minimization problem for  $\int_{\mathcal{A}} \mathcal{A}_{\mathbf{L}}$   $\int_{\mathcal{A}} \mathcal{A}_{\mathbf{L}}$ 

$$
J(u) = \frac{1}{2} \int\limits_{\Omega} |\nabla u|^2 dx - \int\limits_{\Omega} f u dx
$$

over  $K = \{u \in C^2(\Omega); u|_{\partial \Omega} = 0\}$  and if u is a solution to this problem then for any  $\epsilon \leqslant \mathbb{R}$  and  $C_0^{\infty}(\Omega)$  we have

 $J(u + \epsilon \phi) \geq J(u),$ 

then we also obtain (1.38). The question is how to obtain the solution.

In a similar way, we consider the obstacle problem, to minimize  $J$  over  $K = \{u \in C^2(\Omega); u|_{\partial \Omega} = 0, u \geq g\}$  over some continuous function g satisfying  $g|_{\partial\Omega}$  < 0. Note that K is convex. Again, if  $u \in K$  is a solution then for any  $\epsilon > 0$  and  $\phi \in K$  we obtain that  $u + \epsilon(\phi - u) = (1 - \epsilon)u + \epsilon \phi$  is in K and therefore

$$
J(u + \epsilon(\phi - u)) \ge J(u).
$$

Here, we obtain only

$$
J(u + \epsilon(\phi - u)) \ge J(u).
$$
\n
$$
\int_{\Omega} \nabla u \cdot \nabla (\phi - u) dx \ge \int_{\Omega} f(\phi - u) dx.
$$
\n
$$
\int_{\Omega} \frac{\partial u}{\partial u} \cdot \nabla u = \int_{\Omega} \frac{\partial u}{\partial u} \cdot \frac{\partial u}{\partial u} \cdot \frac{\partial u}{\partial u}.
$$

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for any  $\phi \in K$ . For twice differentiable u we obtain

$$
\sum_{\Omega} \Delta u(\phi - u) dx \ge \int_{\Omega} f(\phi - u) dx
$$

and choosing  $\phi = u + \psi$ ,  $\psi \in C_0^{\infty}(\Omega)$  we get

$$
\begin{cases}\n\sqrt{2} & -\Delta u \ge f\n\end{cases}
$$

almost everywhere on  $\Omega$ . As u is continuous, the set  $N = \{x \in \Omega; u(x) >$  $g(x)$  is open. Thus, taking  $\psi \in C_0^{\infty}(N)$ , we see that for sufficiently small  $\epsilon > 0$ ,  $u \pm \epsilon \phi \in K$ . Then, on N

$$
-\Delta u = f
$$

Summarizing, for regular solutions the minimizer satisfies

$$
-\Delta u \ge f
$$

$$
u \ge g
$$

$$
(\Delta u + f)(u - g) = 0
$$

on Ω.

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#### Hilbert space theory

We begin with the following definition.

**Definition 1.34.** Let H be a Hilbert space. A bilinear form  $a: H \times H \to \mathbb{R}$ is said to be

 $(i)$  continuous of there is a constant C such that

$$
|a(x, y)| \le C ||x|| ||y||, \qquad x, y \in H;
$$

coercive if there is a constant  $\alpha > 0$  such that

$$
a(x,x) \ge \alpha ||x||^2.
$$

Note that in the complex case, coercivity means  $|a(x, x)| \ge \alpha ||x||^2$ .

 $S+$  $\omega$ peccl $\omega$   $\alpha$ <br>Theorem 1.35. Assume that  $a(\cdot, \cdot)$  is a continuous coercive bilinear form on a Hilbert space  $H$ . Let  $K$  be a nonempty closed and convex subset of  $H$ . Then, given any  $\phi \in \{H^*\}$  there exists a unique element  $x \in K$  such that for any  $y \in K$ 

$$
a(x, y-x) \ge \langle \phi, y-x \rangle_H_{* \times H}
$$
\n(1.40)

Moreover, if a is symmetric, then  $x$  is characterized by the property

$$
x \in K \quad \text{and} \quad \frac{1}{2}a(x,x) - \langle \phi, x \rangle_{H^* \times H} = \min_{y \in K} \frac{1}{2}a(y,y) - \langle \phi, y \rangle_{H^* \times H}.
$$
\n(1.41)

*Proof.* First we note that from Riesz theorem, there is  $f \in H$  such that  $\langle \phi, y \rangle_{H^* \times H} = (f, y)$  for all  $y \in H$ . Now, if we fix  $x \in H$ , then  $y \to a(x, y)$  is a continuous linear functional on  $H$ . Thus, again by the Riesz theorem, there is an operator  $A : H \to H$  satisfying  $a(x, y) = (Ax, y)$ . Clearly, A is linear and satisfies

$$
||Ax|| \le C||x||,\tag{1.42}
$$

$$
\mathbf{e}(x, \mathbf{y}) = (Ax, x) \ge \alpha ||x||^2 / \sum_{\mathbf{z} \text{ sup } |\mathbf{e}(x, \mathbf{y})|} (1.43)
$$

Indeed,

$$
||Ax|| = \sup_{||y||=1} |(Ax, y)| \le C ||x|| \sup_{||y||=1} ||y||,
$$

and (1.43) is obvious.

Problem (1.40) amounts to finding  $x \in K$  satisfying, for all  $y \in K$ ,

$$
\mathcal{S}(Ax, y-x) \geq f(f, y-x). \tag{1.44}
$$
\n
$$
\mathcal{S}(Ax, y-x) \geq f(f, y-x). \tag{1.45}
$$

Let us fix a constant  $\rho$  to be determined later. Then, multiplying both sides of (1.44) by  $\rho$  and moving to one side, we find that (1.44) is equivalent to

 $(\mu - x, y - x)$   $\frac{1}{3}$  Hilbert space methods 31

$$
\bigvee_{y \in K} (\widehat{\rho f} - \rho Ax + x - x, y - x) \le 0. \tag{1.45}
$$

Here we recognize the equivalent formulation of the projection problem (1.35), that is, we can write

$$
x = P_K(\rho f - \rho Ax + x) \tag{1.46}
$$

This is a fixed point problem for x in K. Denote  $Sy = P<sub>K</sub>(\rho f - \rho Ay + y)$ . Clearly  $S: K \to K$  as it is a projection onto K and K, being closed, is a complete metric space in the metric induced from  $H$ . Since  $P_K$  is nonexpansive, we have  $||S_{21} - S_{22}|| < ||(y_1 - y_2) - \rho(Ay_1 - Ay_2)||$ 

and thus

$$
||Sy_1 - Sy_2|| \le ||(y_1 - y_2) - \rho(Ay_1 - Ay_2)||
$$
  

$$
(\mathcal{A}(y_1 - y_1), y_2 - y_1) \ge \alpha ||y_1 - y_1||^2
$$

$$
||Sy_1 - Sy_2||^2 = ||y_1 - y_2||^2 \le 2\rho(Ay_1 - Ay_2, y_1 - y_2) + \rho^2||Ay_1 - Ay_2||^2
$$
  
 
$$
\le ||y_1 - y_2||^2(1 - 2\rho\alpha + \rho^2C^2) \quad \text{or} \quad \text{for} \quad \text{if} \quad \text{
$$

We can choose  $\rho$  in such a way that  $k^2 = \cancel{1} - 2\rho\alpha + \rho^2C^2 < \cancel{1}$  we see that S has a unique fixed point in K.

Assume now that a is symmetric. Then  $(x, y)_1 = a(x, y)$  defines a new scalar product which defines an equivalent norm  $||x||_1 = \sqrt{a(x,x)}$  on H. Indeed, by continuity and coerciveness

$$
||x||_1 = \sqrt{a(x,x)} \le \sqrt{C} ||x||
$$

and

$$
||x||_{\mathbf{1}} = \sqrt{a(x,x)} \ge \sqrt{\alpha} ||x||.
$$

Using again Riesz theorem, we find  $q \in H$  such that

$$
\langle \phi, y \rangle_{H^* \times H} = a(g, y)
$$

and then (1.40) amounts to finding  $x \in K$  such that

$$
(g-x, g-x)_q = a(g-x, y-x) \le 0
$$

for all  $y \in K$  but this is nothing else but finding projection x onto K with respect to the new scalar product. Thus, there is a unique  $x \in K$ 

$$
\sqrt{a(g-x,g-x)} = \min_{y \in K} \sqrt{a(g-y,g-y)}.
$$
  
 
$$
\text{If } g - x \text{ if } g \text{ is a } y \text{ if } g \text{ is a } y.
$$

However, expanding, this is the same as finding minimum of the function

$$
y \to a(g-y, g-y) = a(g, g) + a(y, y) - 2a(g, y) = a(y, y) - 2 < \phi, y >_{H^* \times H} + a(g, g).
$$

Taking into account that  $a(g, g)$  is a constant, we see that x is the unique minimizer of

$$
y \to \frac{1}{2}a(y, y) - \langle \phi, y \rangle_{H^* \times H} .
$$

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**Corollary 1.36.** Assume that  $a(\cdot, \cdot)$  is a continuous coercive bilinear form on a Hilbert space H. Then, given any  $\phi \in H^*$ , there exists a unique element  $x \in H$  such that for any  $y \in H$  $a(x, y) = \langle \phi, y \rangle_{H^* \times H}$  (1.47)

Moreover, if a is symmetric, then  $x$  is characterized by the property

$$
x \in H \quad \text{and} \quad \frac{1}{2}a(x,x) - \langle \phi, x \rangle_{H^* \times H} = \min_{y \in H} \frac{1}{2}a(y,y) - \langle \phi, y \rangle_{H^* \times H}.
$$
\n(1.48)

*Proof.* We use the Stampacchia theorem with  $K = H$ . Then there is a unique element  $x \in H$  satisfying  $|| \nparallel \mathcal{L} \times || \geq \infty || \times ||$ element  $x \in H$  satisfying

Using linearity, this must hold also for  
\n
$$
a(x, y - x) \geq <\phi, y - x>_{H^* \times H}
$$
\n
$$
\downarrow \phi \in H
$$
\n
$$
\downarrow e(\kappa, y) = \langle \phi, y \rangle
$$
\n
$$
\downarrow e(\kappa, y) \leq C\|\kappa\|_{H^1}
$$

$$
a(x, ty - x) \ge \langle \phi, ty - x \rangle_{H^* \times H}.
$$

for any  $t \in R$ ,  $\mathbf{w} \in H$ . Factoring out  $t$ , we find

$$
a(x, ty-x) \ge \phi, ty-x >_{H^* \times H}
$$
\n
$$
\in H.
$$
 Factoring out *t*, we find\n
$$
ta(x, y \neq x^{-1}) \geq y' < \phi, y \neq x^{-1} >_{H^* \times H}
$$
\n
$$
\begin{array}{c}\n\left(\frac{Ax}{X}, y\right) = \left(\frac{y}{Y}\right) \\
\downarrow \frac{y}{X} \parallel \frac{y}{X} \parallel \frac{y}{Y} \\
\downarrow \frac{y}{X} \parallel \frac{y}{X} \parallel \frac{y}{Y} \\
\downarrow \frac{y}{X} \parallel \frac{y}{Y} \parallel \frac{y}{X} \parallel \frac{y}{Y} \\
\downarrow \frac{y}{X} \parallel \frac{y}{Y} \parallel \frac{y}{X} \parallel \frac{y}{Y} \parallel \frac{y}{X} \parallel \frac{y}{Y} \\
\downarrow \frac{y}{X} \parallel \frac{y}{Y} \parallel \frac{y}{X} \parallel \frac{
$$

and passing with  $t \to \pm \infty$ , we obtain<br> $\sim \mathcal{L} \rightarrow \mathbb{R}$ 

$$
\begin{array}{ccc}\n\mathbf{a} & \mathbf{b} & \mathbf{c} & \mathbf{c} & \mathbf{c} \\
\mathbf{c} & \mathbf{b} & \mathbf{c} & \mathbf{c} \\
\mathbf{c} & \mathbf{b} & \mathbf{c} & \mathbf{c} \\
\mathbf{c} & \mathbf{b} & \mathbf{c} & \mathbf{c} \\
\mathbf{d} & \mathbf{c} & \mathbf{c} & \mathbf{c} \\
\mathbf{e} & \mathbf{c} & \mathbf{c} & \mathbf{c} \\
\mathbf{e} & \mathbf{c} & \mathbf{c} & \mathbf{c} \\
\mathbf{e} & \mathbf{c} & \mathbf{c} & \mathbf{c} \\
\mathbf{f} & \mathbf{c} & \mathbf{c} & \math
$$

An important role in functional analysis is played by the operation of taking operator adjoint. If  $A \in \mathcal{L}(X, Y)$ , then the adjoint operator  $A^*$  is defined as

$$
\langle y^*, Ax \rangle = \langle A^* y^*, x \rangle \tag{1.49}
$$

and it can be proved that it belongs to  $\mathcal{L}(Y^*, X^*)$  with  $||A^*|| = ||A||$ . If A is an unbounded operator, then the situation is more complicated. In general, A<sup>∗</sup> may not exist as a single-valued operator. In other words, there may be many operators  $B$  satisfying

$$
\langle y^*, Ax \rangle = \langle By^*, x \rangle, \qquad x \in D(A), \ y^* \in D(B). \tag{1.50}
$$

Operators  $A$  and  $B$  satisfying  $(1.50)$  are called *adjoint to each other.* 

However, if  $D(A)$  is dense in X, then there is a unique maximal operator  $A^*$  adjoint to A; that is, any other B such that A and B are adjoint to each other, must satisfy  $B \subset A^*$ . This  $A^*$  is called the *adjoint operator* to A. It can be constructed in the following way. The domain  $D(A^*)$  consists of all elements  $y^*$  of  $Y^*$  for which there exists  $f^* \in X^*$  with the property