# Math 255A' Lecture 20 Notes

## Daniel Raban

November 15, 2019

# **1** Positive Operators and Spectral Families

#### **1.1** Positive operators

We want to generalize the following theorem, without the assumption of compactness.

**Theorem 1.1** (Spectral theorem in finite dimensions). Let  $\dim(H) < \infty$ , and let  $T : H \to H$  be a self-adjoint operator with eigenvalues  $a \le \lambda_1 < \lambda_2 < \cdots < \lambda_m = b$ . Then

$$T = \sum_{i=1}^{n} \lambda_i P_{\lambda_i},$$

where  $P_{\lambda_i}$  is the projection onto ker $(T - \lambda_i)$ .

**Example 1.1.** On  $L^2([0,1])$  we have Tf(x) = xf(x), the multiplication operator. Then  $||T||_{\text{op}} \leq 1$ , and

$$\langle Tf,g\rangle = \int_0^1 \overline{x}f(x)\overline{g(x)}\,dx = \langle f,Tg\rangle$$

However, T has no eigenvectors! If  $Tf = \lambda f$ , then  $xf(x) = \lambda f(x)$  for a.e. x. So f = 0 a.e.

Observe that if  $V = \ker(T - \lambda) \neq \{0\}$ , then V is reducing and  $T|_V = \lambda I_V$ . We want to loosen this to  $\mu I_V \leq T|_V \leq \lambda I_V$  for  $\mu < \lambda$ .

**Definition 1.1.**  $T \in \mathcal{B}(H)$  is **positive** (written  $T \ge 0$ ) if T is self-adjoint and  $\langle Tx, x \rangle \ge 0$ . If S, T are self-adjoint, we say  $S \le T$  if  $T - S \ge 0$ .

This defines a partial order on the set of self-adjoint operators. How does this relate to our previous examples?

**Example 1.2.** In the finite dimensional case, for  $\lambda \in \mathbb{R}$ , define

$$E(\lambda) := \sum_{i:\lambda_i \leq \lambda} P_{\lambda_i}, \qquad E(\mu, \lambda) := \sum_{\mu < \lambda_i \leq \lambda} P_{\lambda_i} = E(\lambda) - E(\mu).$$

These all reduce T, and

$$\mu E(\mu, \lambda) \le T E(\mu, \lambda) \le \lambda E(\mu, \lambda)$$

for all  $\mu \leq \lambda$ . If  $\lambda_i$  is the unique element of  $\sigma_p(T) \cap (\mu, \lambda], \lambda_i P_i \leq T P_{\lambda_i} \leq \lambda_i P_i$ .

**Example 1.3.** With the multiplication operator T on  $L^2$ , let  $V(\mu, \lambda) := \{f \in L^2([0,1]) : f = f \mathbb{1}_{(\mu,\lambda]}\}$  for any  $\mu \leq \lambda$ . Then let  $E(\mu, \lambda) = P_{V(\mu,\lambda)}$ . We can check that

$$TE(\mu, \lambda)f(x) = xf\mathbb{1}_{(\mu,\lambda]}(x).$$

Then  $\mu E(\mu, \lambda) \leq TE(\mu, \lambda) \leq \lambda E(\mu, \lambda)$ .

**Lemma 1.1.** Let T be self-adjoint, and let  $a = \inf_{\|x\|=1} \langle Tx, x \rangle$ . and  $b = \sup_{\|x\|=1} \langle Tx, x \rangle$ . Then  $a \leq T \leq b$  and  $\|T\| = \max(|a|, |b|)$ .

*Proof.* If ||x|| = 1, then

$$\langle (T-a)x, x \rangle = \langle Tx, x \rangle = a \ge 0.$$

The upper bound is the same.

We have seen already that  $||T|| = \sup |\langle Tx, x \rangle|$ .

**Corollary 1.1.** If  $S \leq T$  and  $T \leq S$  then S = T.

*Proof.* This implies that  $\langle (S-T)x, x \rangle = 0$  for all x. So the norm is ||S-T|| = 0.

**Lemma 1.2.** For projections P,Q, the following are equivalent:

1.  $P \leq Q$ .

2. 
$$QP = PQ = P$$
.

- 3. Q P is a projection.
- 4.  $||Px|| \le ||Qx||$ .
- 5. ran  $P \subseteq \operatorname{ran} Q$ .

*Proof.* (1)  $\implies$  (5): If (5) is false, then there is some  $x \neq 0$  such that Px = x but  $Qx \neq x$ . Then  $||x||^2 = \langle Px, x \rangle$ , but  $\langle Qx, x \rangle = ||Qx||^2 < ||x||^2$ . This contradicts (1).

(5)  $\implies$  (2): QP = P by the condition of (5), and we get  $(QP)^* = P^*Q^* = PQ$  by self-adjointness.

$$\begin{array}{l} (2) \implies (4): \|Px\| = \|PQx\| \le \|Qx\|. \\ (2) \implies (3): \langle (Q-P)x, x \rangle = \langle Q(1-P)x, x \rangle = \langle Q(1-P)x, Qx \rangle \ge 0. \\ (3) \implies (1): Q-P \text{ is a projection, so } Q-P \ge 0. \end{array}$$

#### **1.2** Spectral families and the spectral theorem

**Definition 1.2.** A spectral family on H is a map  $\lambda \mapsto E(\lambda)$  from  $\mathbb{R} \to \{\text{proj. on } H\}$  such that

- 1. If  $\lambda > \mu$ , then  $E(\lambda) \ge E(\mu)$
- 2. There exist  $a, b \in \mathbb{R}$  such that  $E(\lambda) = 0$  if |lambda < a and  $E(\lambda) = I$  if  $\lambda \ge b$ .
- 3.  $E(\lambda)x \to E(\mu)x$  as  $\lambda \downarrow \mu$  for all  $x \in H$  (convergence in the strong operator topology).

**Theorem 1.2.** Let T be a self-adjoint operator on H. Then there exists a spectral family  $(E(\lambda))_{\lambda \in \mathbb{R}}$  such that

$$a = \inf_{\|x\|=1} \langle Tx, x \rangle, \qquad b = \sup_{\|x\|=1} \langle Tx, x \rangle$$
$$T = \int_{\mathbb{R}} \lambda \, dE(\lambda).$$

This means  $\langle Tx, y \rangle = \int_{[a,b]} \lambda \, d\mu_{x,y}$  for all  $x, y \in H$ , where  $\mu_{x,y}$  is the Lebesgue-Stieltjes measure corresponding to  $F_{x,y}$ .

To interpret this integral, we need the following lemma.

**Lemma 1.3.** If E is a spectral family, then for any  $x, y \in H$ , then function  $F_{x,y} : \lambda \mapsto \langle E(\lambda)x, y \rangle$  is right-continuous and of bounded variation.

*Proof.* Right continuity follows from property (3) of a spectral family. For bounded variation,

Step 1: If y = x, then  $F_{x,x}(\lambda) = ||E(\lambda)x||^2$ , which is increasing with  $\lambda$ . Step 2:

$$F_{x,y}(\lambda) = \langle E(\lambda)x, y \rangle = \frac{\langle E(\lambda)(x+y), x+y \rangle - \langle E(\lambda)x, x \rangle - \langle E(\lambda)y, y \rangle}{2}$$

is a difference of nondecreasing functions, so it is of bounded variation.

**Example 1.4.** In the finite dimensional case,  $E(\lambda)$  is constant, except for finitely many jumps. So the integral becomes a finite sum.

**Example 1.5.** Returning to the multiplication operator on  $L^2$ , if  $f, g \in L^2([0,1])$ , then

$$\langle Tf,g \rangle = \int_0^1 x f(x) \overline{g(x)} \, dx, \qquad dx = d\mu_{f,g}$$

Here,  $E(\lambda)$  is the proejction onto  $\{f = f \mathbb{1}_{[0,\lambda]}\}$ , and  $\langle E(\lambda)f,g \rangle = \int_0^\lambda f \overline{g} \, dx$ .

### **1.3** Functional calculus

How do we find this map  $\lambda \mapsto E(\lambda)$ ? In the finite dimensional case, we have a self-adjoint T with eigenvalues  $a = \lambda_1 < \lambda_2 < \cdots < \lambda_m = b$  and  $T = \sum_i \lambda_i P_{\lambda_i}$ . If  $p(t) = \sum_{j=1}^k x_j t^j \in \mathbb{R}[t]$  is a polynomial, we can write  $p(T) = \sum_{j=1}^k c_j T^j$ . Since  $T^j = \sum_i \lambda_i^j P_i$ , we have  $p(T) = \sum_i p(\lambda_i) P_{\lambda_i}$ .

Choose any  $p_{\lambda} \in \mathbb{R}[t]$  such that

$$p_{\lambda}(t) = \begin{cases} 1 & t = \lambda_i \leq \lambda \\ 0 & t = \lambda_i > \lambda. \end{cases}$$

Then

$$p_{\lambda}(T) = \sum_{\lambda_i \leq \lambda} P_{\lambda_i} = E(\lambda).$$

We need to make this work in infinite dimensions. But  $\mathbb{R}[t]$  is not rich enough. We must extend the map  $\mathbb{R}[T] \to \mathcal{B}(H)$  taking  $p \mapsto p(T)$  to a larger class of functions. After doing so, we get the **functional calculus** of T. In particular, we want to be able to get the function p(T), where  $p(t) = \mathbb{1}_{(-\infty,\lambda]}(t)$ .