

Introduction to Mathematical Physics

Fall 2010 : Spectral Theory

Main references:

- [RS1] M. Reed, B. Simon: Methods of Modern Mathematical Physics I: Functional Analysis (1980)
- [Rudin, FA] W. Rudin: Functional Analysis (1991)
- [kato] T. Kato: Perturbation Theory for Linear Operators (2008)

For background material (measure theory and complex analysis):

- [Rudin, RCA] W. Rudin: Real and Complex Analysis (1987)

* Course announcements & further references from the course homepage

* Lectures: Tue & Thu 14-16
Exercises: Fri 10-12

* The course can be taken for 10 cu.
Grade consists of $\left\{ \begin{array}{l} 50\% \text{ Final exam} \\ 50\% \text{ Exercises} \end{array} \right.$

* Please remember to register in OOD1 if you wish to take the course.

1. Introduction: Matrices ($\mathbb{C}^{d \times d} \cong \mathcal{B}(\mathbb{C}^d)$)

(2)

Recap of basic notions of linear algebra

* $\mathbb{C}^d, d \in \mathbb{N}_+$, is a complex vector space, with a

scalar product $a \cdot b := \sum_{i=1}^d a_i^* b_i$

($a^* = \operatorname{Re} a - i \operatorname{Im} a =$ complex conjugate of a . Note that here the first argument is antilinear (physics convention).)

* The length of a vector $a \in \mathbb{C}^d$ is

denoted by $|a| := \sqrt{\sum_{i=1}^d |a_i|^2}$.

* "Hat" will be used to denote the direction of $a \in \mathbb{C}^d$:

$$\hat{a} := \frac{a}{|a|}$$

* Consistently, we use "hat" also if we wish to emphasise unit vectors (i.e. $\hat{a} \in \mathbb{C}^d \Rightarrow |\hat{a}| = 1$)

* Same notations are used for $\mathbb{R}^d \subset \mathbb{C}^d$.
(Then $a, b \in \mathbb{R}^d \Rightarrow a \cdot b = \sum_{i=1}^d a_i b_i$)

* Special notation $\hat{e}_\nu, \nu = 1, 2, \dots, d$, is reserved for the Cartesian basis of \mathbb{C}^d :

$$(\hat{e}_\nu)_i := \mathbb{1}(i = \nu); \quad i, \nu = 1, 2, \dots, d$$

* Here $\mathbb{1}(P) = \begin{cases} 1, & \text{if } P \text{ is true} \\ 0, & \text{otherwise} \end{cases}$

* Kronecker delta: $\delta_{ij} = \mathbb{1}(i = j)$ for $i, j \in \mathbb{Z}^d$.

③

* $\{\hat{e}_\nu\}_{\nu=1}^d$ forms an orthonormal basis for \mathbb{C}^d :

a) $\mathbb{C}^d = \text{span}\{\hat{e}_\nu \mid \nu=1, \dots, d\}$

b) $\hat{e}_{\nu'} \cdot \hat{e}_\nu = \delta_{\nu'\nu}$, $\forall \nu', \nu=1, \dots, d$

* A matrix is a "vector" in $\mathbb{C}^{d_1 \times d_2} := (\mathbb{C}^{d_2})^{d_1}$
($\cong \mathbb{C}^{d_1 d_2}$)

(Notational reminder:

* $\bar{X}^I := \{ f: I \rightarrow \bar{X} \}$, for any sets I, \bar{X} ;

I is called the index set, $x_i := f(i)$.

* $\bar{X}^n := \bar{X}^{\{1, 2, \dots, n\}}$, for $n \in \mathbb{N}_+$.

* If $M \in (\bar{X}^J)^I$, $M_{ij} := (M_i)_j \in \bar{X}$
 $\Rightarrow (\bar{X}^J)^I \cong \bar{X}^{I \times J}$

* Row-column shorthand notation of matrices: $M = \begin{pmatrix} M_{11} & M_{12} & M_{13} \\ M_{21} & M_{22} & M_{23} \end{pmatrix} \in \mathbb{C}^{2 \times 3}$

* Matrices can be identified with linear maps: if $T \in \mathcal{B}(\mathbb{C}^{d_2}, \mathbb{C}^{d_1})$ and $M \in \mathbb{C}^{d_1 \times d_2}$

$T \cong M$ if $M_{ij} = \hat{e}_i \cdot T\hat{e}_j$ $\forall i=1, \dots, d_1$
 $j=1, \dots, d_2$

$\Rightarrow a \cdot Tb = \sum_{i=1}^{d_1} \sum_{j=1}^{d_2} a_i^* M_{ij} b_j$ $\forall a \in \mathbb{C}^{d_1}, b \in \mathbb{C}^{d_2}$.

* Here $\mathcal{B}(\bar{X}, \bar{Y}) := \{ T: \bar{X} \rightarrow \bar{Y} \mid T \text{ linear and } \|T\| < \infty \}$

where $\|T\| := \sup_{\substack{a \in \bar{X}, \\ \|a\| \leq 1}} \|Ta\|$, \bar{X}, \bar{Y} are normed spaces.

* This induces the matrix norm on $\mathbb{C}^{d_1 \times d_2}$

$\|M\| := \sup_{\substack{\hat{a} \in \mathbb{C}^{d_1} \\ \hat{b} \in \mathbb{C}^{d_2}, \|\hat{a}\|=\|\hat{b}\|=1}} \left| \sum_{i,j} a_i^* M_{ij} b_j \right| = \max_{b \in \mathbb{C}^{d_2}, b \neq 0} \left| \frac{1}{\|b\|} \sum_{j=1}^{d_2} M_{i,j} b_j \right|$

* Matrix product: $A \in \mathbb{C}^{d_1 \times d_2}, B \in \mathbb{C}^{d_2 \times d_3}$ then

$AB \in \mathbb{C}^{d_1 \times d_3}$ defined by $(AB)_{ij} := \sum_{\nu=1}^{d_2} A_{i\nu} B_{\nu j}$.

$\Rightarrow AB \cong A \circ B$. Also $\|AB\| \leq \|A\| \|B\|$.

* The matrix norm is (numerically) different from the natural norm of $\mathbb{C}^{d_1 \times d_2}$

$$\|M\|_{HS} := \sqrt{\sum_{i,j} |M_{ij}|^2} = \sqrt{\text{Tr}(M^+M)} = \sqrt{\text{Tr}(MM^+)}$$

where $(M^+)_{ij} := (M_{ji})^*$ defines the adjoint matrix of M , ($M^+ \in \mathbb{C}^{d_2 \times d_1}$)

and $\text{Tr } B := \sum_{i=1}^d B_{ii}$ is the trace

of a square matrix $B \in \mathbb{C}^{d \times d}$.

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* Notation: $\mathcal{B}(\mathbb{C}^d) := \mathcal{B}(\mathbb{C}^d, \mathbb{C}^d)$
 $\Rightarrow \mathcal{B}(\mathbb{C}^d) = \mathbb{C}^{d \times d} =$ set of square matrices, = set of operators on \mathbb{C}^d .

* Proof: $\| \begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix} \| = 2 < \sqrt{5} = \| \begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix} \|_{HS}$

* $\|M\|_{HS}$ is called the Hilbert-Schmidt norm of M .

* The main aim of the course is to generalize the following matrix result to operators on infinite-dimensional Hilbert-spaces:

Theorem (diagonalization of self-adjoint matrices)

Suppose $A \in \mathcal{B}(\mathbb{C}^d)$ satisfies $A^+ = A$.
Then there is a unitary matrix $U \in \mathcal{B}(\mathbb{C}^d)$
and a real diagonal matrix $D \in \mathcal{B}(\mathbb{C}^d)$
such that $A = U^+ D U$.

Consequence: Let $\lambda_i = R_{ii} \in \mathbb{R}, \Rightarrow (R^n)_{ij} = \lambda_i^n \delta_{ij}$,

$$(e^{zR})_{ij} = e^{z\lambda_i} \delta_{ij} \quad \& \quad A^n = U^+ R^n U, \quad e^{zA} = U^+ e^{zR} U \quad \forall z \in \mathbb{C}.$$

(Matrix exponential will be defined in the exercises.)

* U unitary means $U \in \mathbb{C}^{d \times d}$ and $U^{-1} = U^\dagger$; they are important since they generate Hilbert space isomorphisms: $Ua \cdot Ub = a \cdot U^\dagger Ub = a \cdot b$.

* Linear isomorphism is called a change of basis and they are in

1-1 correspondence with invertible square matrices: If $\{e_i\}$ are the old and $\{e'_i\}$ the new basis vectors, then $\exists L \in \mathbb{C}^{d \times d}$ s.t. $e'_i = L e_i \forall i$ and $\exists L^{-1}$.

* A change of basis L induces the following change to an operator $B \in \mathbb{C}^{d \times d}$:
 $B \mapsto L B L^{-1}$.

* Thus the theorem can also be summarized by saying that to every self-adjoint matrix, there is a unitary change of basis such that it becomes a real multiplication operator. ($D = UAU^{-1}$)

* $B \in \mathcal{B}(\mathbb{C}^d)$ is a multiplication operator, if there is $\vartheta: \mathbb{C}^d \rightarrow \mathbb{C}$ such that $(Ba)_i = \vartheta_i a_i \forall a \in \mathbb{C}^d$ and $i=1, \dots, d$. Thus multiplication operators are in 1-1 correspondence with diagonal matrices.

* In quantum mechanics, time-evolution of a wavevector $\psi_0 \in \mathcal{H} = \text{Hilbert-space}$ is defined by giving a self-adjoint operator H and defining time-evolved wavevectors by

$$\psi_t := e^{-itH} \psi_0, \quad t \in \mathbb{R}.$$

"Diagonalization" of H will vastly simplify the analysis of properties of ψ_t .

Before going any further, let us recall another useful property of matrices which does not generalize to operators on infinite-dim. Hilbert spaces :

1.1. Jordan decomposition of finite-dim. matrices

a) Differentiation and integration of matrices

Consider a map $t \mapsto M(t)$ which maps an interval $(t_1, t_2) \subset \mathbb{R}$ to $\mathbb{C}^{d_1 \times d_2}$. Is there any natural way to define differentiation and integration of such maps?

* One method is to define it "component-wise" that is define $(M'(t))_{ij} = \frac{d}{dt} M(t)_{ij} \Big|_{t=t_0}$

$$\text{and } \left(\int_{t_1}^{t_2} dt M(t) \right)_{ij} = \int_{t_1}^{t_2} dt M(t)_{ij}$$

whenever $M(t)_{ij}$ is differentiable (integrable) for all i, j .

* For integrals, we use the component-wise definition. This will generalize to what is often called "vector valued integration" in topological vector spaces.

* For derivatives, it is more convenient to think of approximations by linear maps, i.e., of linearizations.

Definition $t \mapsto M(t)$ is said to be differentiable at $t_0 \in (t_1, t_2)$ if $\exists M_0 \in \mathbb{C}^{d_1 \times d_2}$ such that

$$\lim_{h \rightarrow 0} \frac{1}{|h|} \| M(t_0+h) - M(t_0) - M_0 h \| = 0.$$

Then M_0 is called the derivate at t_0 , which we denote by $M'(t_0)$ or $\frac{d}{dt} M(t) \Big|_{t=t_0}$.

* If $M'(t_0)$ exists, it is obviously unique.

b) Block matrices

= convenient shorthand notation

$$M = (B_{ij})_{\substack{i=1, \dots, N_1 \\ j=1, \dots, N_2}} \in \mathbb{C}^{\bar{d}_1 \times \bar{d}_2} \text{ and each}$$

B_{ij} is a matrix, $\in \mathbb{C}^{d_{1i} \times d_{2j}}$;

where the dimensions have to "match", i.e., satisfy

$$\sum_{i=1}^{N_1} d_{1i} = \bar{d}_1 \text{ and } \sum_{j=1}^{N_2} d_{2j} = \bar{d}_2 .$$

Idea is that M acts as "matrix of matrices",

" $(Ma)_i = \sum_{j=1}^{N_2} B_{ij} a_j$ ", which means explicitly that $\forall k' = 1, \dots, \bar{d}_1; k = 1, \dots, \bar{d}_2$

$$M_{k'k} = (B_{ij})_{l'l} \text{ for the unique}$$

$i \in \{1, \dots, N_1\}, j \in \{1, \dots, N_2\}, l' \in \{1, \dots, d_{1i}\}$
 $l \in \{1, \dots, d_{2j}\}$ such that

$$k' = l' + \sum_{i'=1}^{i-1} d_{1i'} \text{ and } k = l + \sum_{j'=1}^{j-1} d_{2j'}$$

Example: If $A \in \mathbb{C}^{d_1 \times d_1}, B \in \mathbb{C}^{d_2 \times d_2}$, and $C \in \mathbb{C}^{d_1 \times d_2}$ then

$$M = \begin{pmatrix} A & C \\ C^+ & B \end{pmatrix} \in \mathbb{C}^{(d_1+d_2) \times (d_1+d_2)}$$

c) Nilpotent matrices

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* If M is diagonalizable $\Rightarrow \forall n \geq 1: M^n = L R^n L^{-1}$.
Thus $M^n = 0 \Rightarrow R^n = 0 \Rightarrow R = 0 \Rightarrow M = 0$.

Definition If $M \in \mathbb{C}^{d \times d}$ and there is $n \in \mathbb{N}_+$ such that $M^n = 0$, M is called nilpotent.

* Since, for instance, $M = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$ satisfies $M^2 = 0$, there are non-zero nilpotent matrices \Rightarrow not every matrix is diagonalizable!

* The following theorem yields a characterization of nilpotent matrices:

Theorem: Let $M \in \mathbb{C}^{d \times d}$ be a nilpotent matrix. Then there is a change of basis $L \in \mathbb{C}^{d \times d}$ which brings it into a block-diagonal form, where each diagonal block is a simple Jordan block matrix.

Explicitly: $\exists L$ s.t. $M = L^{-1} J L$

$$\text{where } J = \begin{pmatrix} J_1 & 0 & 0 & \dots & 0 \\ 0 & J_2 & 0 & \dots & 0 \\ 0 & 0 & J_3 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & J_N \end{pmatrix}, \quad N \geq 1,$$

and for each $n = 1, \dots, N$, $\exists d_n \geq 1$, so that

$$a) \sum_{n=1}^N d_n = d$$

$$b) \begin{aligned} (J_n a)_i &= a_{i+1}, \quad i=1, \dots, d_n-1 \\ (J_n a)_{d_n} &= 0 \end{aligned}$$

$$\text{that is, } J_n = \begin{pmatrix} 0 & 1 & 0 & 0 & \dots & 0 \\ 0 & 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & \dots & 1 \\ 0 & 0 & 0 & \dots & \dots & 0 \end{pmatrix}$$

* Note that $(0) \in \mathbb{C}^{1 \times 1}$ is also a simple Jordan block matrix.

* A convenient formula for the matrix elements of a simple Jordan block matrix $J \in \mathbb{C}^{d \times d}$ is given by

$$J_{ij} = \mathbb{1}(j=i+1) \quad \forall i, j = 1, \dots, d.$$

* Notations: For $M \in \mathbb{C}^{d_1 \times d_2} \equiv \mathcal{B}(\mathbb{C}^{d_2}, \mathbb{C}^{d_1})$

We denote its range by $R(M) := \{Ma \mid a \in \mathbb{C}^{d_2}\}$

and kernel or null space by

$$\ker(M) := \{a \in \mathbb{C}^{d_2} \mid Ma = 0\}.$$

Reminders: $R(M)$ is a subspace of \mathbb{C}^{d_1}
 $\ker(M)$ is a subspace of \mathbb{C}^{d_2}

Proof of the Theorem:

Since M is nilpotent there is

$$m = \min \{n \geq 0 \mid M^{n+1} = 0\}.$$

If $m = 0 \Rightarrow M^1 = M = 0$, and we can choose $I = \mathbb{1}$ (= identity matrix), $N = d$, $d_n = 1 \quad \forall n$ and $J_n = (0) \quad \forall n$. ($\Rightarrow J = 0$).

Assume thus that $m > 0 \Rightarrow M^m \neq 0$. Each $R(M^n)$ is a subspace, and let $r_n = \dim R(M^n)$. Since obviously $R(M^{n+1}) \subset R(M^n)$, we have now $d = r_0 \geq r_1 \geq \dots \geq r_m > 0 = r_n \quad \forall n > m$.

Then $R(M^m)$ has a basis $\{v_j^{(m)}\}_{j=1}^{r_m}$.
 For each $v_j^{(m)}$ we choose $e_j^{(m)} \in \mathbb{C}^d$ s.t.
 $v_j^{(m)} = M^m e_j^{(m)}$.

Define then $N_m := r_m$ and set

$$e_{j,n}^{(m)} := M^n e_j^{(m)} \quad \text{for } j=1, \dots, N_m; n \geq 0.$$

By definition, $e_{j,n}^{(m)} = 0 \quad \forall n > m$ and

$(e_{j,n}^{(m)})$ is a basis for $R(M^m)$.

As an aside: the vectors $e_{j,n}^{(m)}; j=1, \dots, N_m; n=0, \dots, m$

are all linearly independent.

(PF. if $\sum_{j,n} \alpha_{j,n} e_{j,n}^{(m)} = 0 \Rightarrow \forall n' \geq 0:$

$$0 = \sum_{j,n} \alpha_{j,n} M^{n'+n} e_j^{(m)} \stackrel{n''=n'+n}{=} \sum_{j=1}^{N_m} \sum_{n''=n'} \alpha_{j,n''-n'} M^{n''} e_j^{(m)}$$

Thus $n'=m \Rightarrow \alpha_{j,0} = 0 \quad \forall j$. Then an induction in n shows that $\alpha_{j,n} = 0 \quad \forall j, n \geq 0$ (use $n'=m-n$)

We now claim that $\forall n=0, \dots, m \exists N_n \geq 0$ and vectors $e_j^{(n)} \in \mathbb{C}^d, j=1, \dots, N_n$, s.t. with $e_{j,n'}^{(n)} := M^{n'} e_j^{(n)}, n' \geq 0, e_{j,n'}^{(n)} = 0$ for $n' > n$

and $(e_{j,k'}^{(k)})_{\substack{k=n, \dots, m \\ k'=n, \dots, k \\ j=1, \dots, N_k}}$ is a basis for $R(M^n)$.

The construction can be done iteratively starting from $n=m$ down until $n=0$ is reached. Assume the claim is true for values up to $n+1$. If

$$\sum_{k=n+1}^m \sum_{k'=n}^k \sum_{j=1}^{N_k} \alpha_{j,k,k'} M^{k'} e_j^{(k)} = 0$$

By applying M , and using the induction assumption, $\Rightarrow \alpha_{j,k,k'} = 0 \quad \forall k' < k$.

$$\Rightarrow 0 = \sum_{k=n+1}^m \sum_{j=1}^{N_k} \alpha_{j,k,k} \underbrace{M^k e_j^{(k)}}_{\in R(M^{n+1})} \Rightarrow \alpha_{j,k,k} = 0.$$

Thus $e_{j,k'}^{(k)}, k=n+1, \dots, m; k'=n, \dots, k; j=1, \dots, N_k$ form a lin. indep. subset of $R(M^n)$

Let $N_n = r_n - \sum_{k=n+1}^m (k+1-n)N_k \geq 0$.

If $N_n = 0$, the previous vectors already span $R(M^n)$, and the statement holds for n .

If $N_n > 0$, we can add new vectors $\tilde{v}_j^{(n)}$, $j=1, \dots, N_n$, to complete the set of vectors into a basis for $R(M^n)$. Then $M\tilde{v}_j^{(n)} \in R(M^{n+1})$ and thus $\exists \alpha_j(j', k', \epsilon)$ s.t.

$$M\tilde{v}_j^{(n)} = \sum_{k=n+1}^m \sum_{k'=n+1}^k \sum_{j'=1}^{N_k} \alpha_j(j', k', \epsilon) M^{k'} e_{j'}^{(k)}$$

$$= M r_j^{(n)} \text{ where}$$

$$r_j^{(n)} = \sum_{k=n+1}^m \sum_{k'=n}^{k-1} \sum_{j'=1}^{N_k} \alpha_j(j', k'+1, k) M^{k'} e_{j'}^{(k)} \in R(M^n).$$

We define $v_j^{(n)} = \tilde{v}_j^{(n)} - r_j^{(n)} \in R(M^n) \Rightarrow Mv_j^{(n)} = 0$.

It is easy to check that also adding the vectors $v_j^{(n)}$, $j=1, \dots, N_n$, yields a basis for $R(M^n)$. Then we choose $e_j^{(n)}$ s.t. $v_j^{(n)} = M^n e_j^{(n)}$.

This completes the induction step.

Now $(e_{j,k}^{(k)})_{\substack{k=0, \dots, m \\ k'=0, \dots, k \\ j=1, \dots, N_k}}$ is a basis for $R(M^0) = \mathbb{C}^d$,

and thus there is invertible $L \in \mathbb{C}^{d \times d}$ s.t. $L^{-1} \hat{e}_i = e_{j,k}^{(k)}$ where we go through the indices

by starting from $k=m, j=1, k'=k$, then decreasing k' until 0 is reached, after which $j=2$ (if $N_m > 1$) and k' again runs from k to 0. After $j=N_m$ has been completed, we decrease k , but skip over values for which $N_k = 0$. Since $M e_{j,k}^{(k)} = e_{j,k+1}^{(k)}$ inside each block of $k'=0, \dots, k$,

We clearly have $LM^{-1} \hat{e}_i = J \hat{e}_i \forall i$, where J has N_k blocks of size $k+1$, $k=0, \dots, m$.

$$\Rightarrow N = \sum_{k=0}^m N_k. \quad \square$$

d) Jordan normal form

Consider an arbitrary $M \in \mathbb{C}^{d \times d}$.

* $\lambda \in \mathbb{C}$ is an eigenvalue of M if there is $a \in \mathbb{C}^d$ s.t.

$$Ma = \lambda a.$$

* $\sigma(M) = \{ \lambda \in \mathbb{C} \mid \lambda \text{ is an eigenvalue of } M \}$

* "known" : a) $1 \leq |\sigma(M)| \leq d$.

b) $\exists R_\lambda := (\lambda I - M)^{-1}$ iff. $\lambda \notin \sigma(M)$.

Theorem For any $M \in \mathbb{C}^{d \times d}$ we can find a change basis $L \in \mathbb{C}^{d \times d}$ which brings it into a Jordan normal form.

Explicitly: $\exists L$ s.t. $M = L^{-1} J L$,

$J = \text{Block-diagonal}(J_1, J_2, \dots, J_n)$

and each J_n is a Jordan block matrix:

$$J_n = \lambda_n I + S_n, \quad S_n = \text{simple Jordan block.}$$

Here $\lambda_n \in \sigma(M)$ and each eigenvalue appears at least once.

Proof. Maybe later...

* This result does not generalize to $\mathbb{R}(\mathbb{R})$ if $\dim \mathcal{H} = \infty$.

$$* J_n = \begin{pmatrix} \lambda_n & 1 & 0 & 0 & \dots & 0 \\ 0 & \lambda_n & 1 & 0 & \dots & 0 \\ 0 & 0 & \lambda_n & 1 & \ddots & \vdots \\ \vdots & \vdots & \ddots & \ddots & \ddots & 0 \\ 0 & 0 & \dots & \dots & \dots & \lambda_n \end{pmatrix}$$

* The different "blocks" can be expressed in terms of integrals over the resolvent:

$$M = \sum_{\lambda \in \sigma(M)} (\lambda P_\lambda + T_\lambda)$$

where all (P_λ, T_λ) commute, P_λ is a projection and T_λ is nilpotent.

In addition,

$$P_\lambda = \oint_{\Gamma_\lambda} \frac{dz}{2\pi i} R_z \quad \text{and} \quad T_\lambda = \oint_{\Gamma_\lambda} \frac{dz}{2\pi i} (z-\lambda) R_z$$

where Γ_λ denotes a path in \mathbb{C} which goes once around λ and avoids all other points in the spectrum.

