# ALS Iteration / (Non-)Closedness

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## **1 ALS Method for Optimisation Problems**

## **1.1 Formulation of the Problem**

Let

 $\Phi(\mathbf{u}) = \min$ 

be a minimisation problem over the whole tensor space  $\mathbf{u} \in \mathbf{V}.$ 

Approximation: Choose any format  $\mathcal{F} \subset \mathbf{V}$ . Solve

 $\Phi(\mathbf{u}) = \min$  over all  $\mathbf{v} \in \mathcal{F}$ .

This is the minimisation over all parameters in the representation of  $v \in \mathcal{F}$ .

*Difficulty*: While the original problem may be convex, the new problem is not.

**Example:**  $\Phi(\mathbf{u}) = \langle \mathbf{A}\mathbf{u}, \mathbf{u} \rangle - 2 \langle \mathbf{b}, \mathbf{u} \rangle$  for the solution of  $\mathbf{A}\mathbf{u} = \mathbf{b}$  with positive definite matrix  $\mathbf{A}$ .

Example:  $\Phi(\mathbf{u}) = \|\mathbf{v} - \mathbf{u}\|^2$  over all  $\mathbf{u} \in \mathcal{R}_1 = \mathcal{T}_{(1,...,1)}$ .  $\mathbf{v} \in \mathbf{V}$  is arbitrary.

Ansatz:

$$\mathbf{u} = u^{(1)} \otimes u^{(2)} \otimes \ldots \otimes u^{(d)}, \quad u^{(j)} \in V_j = \mathbb{R}^{n_j}$$

Necessary condition:  $\nabla \Phi(\mathbf{u}) = 0$  (multilinear system of equations).

ALS = alternating least-squares method: 1) solve  $\nabla \Phi(u^{(1)} \otimes u^{(2)} \otimes \ldots \otimes u^{(d)}) = 0$  w.r.t.  $u^{(1)} \Rightarrow$  solution:  $\hat{u}^{(1)}$ , 2) solve  $\nabla \Phi(\hat{u}^{(1)} \otimes u^{(2)} \otimes \ldots \otimes u^{(d)}) = 0$  w.r.t.  $u^{(2)} \Rightarrow$  solution:  $\hat{u}^{(2)}$ , : d) solve  $\nabla \Phi(\hat{u}^{(1)} \otimes \ldots \otimes \hat{u}^{(d-1)} \otimes u^{(d)}) = 0$  w.r.t.  $u^{(d)} \Rightarrow$  solution:  $\hat{u}^{(d)}$ All partial steps are linear problems and easy to solve.

One ALS iteration is given by  $\mathbf{u}_0 = u^{(1)} \otimes \ldots \otimes u^{(d)} \mapsto \mathbf{u}_1 = \hat{u}^{(1)} \otimes \ldots \otimes \hat{u}^{(d)}$ . This defines a ALS sequence  $\{\mathbf{u}_m : m \in \mathbb{N}_0\}$ .

Questions: Does  $\mathbf{u}_m$  converge? To what limit? Convergence speed?

### **1.2 First Results**

Mohlenkamp (2013, Linear Algebra Appl. 438):

- The sequence  $\{\mathbf{u}_m : m \in \mathbb{N}_0\}$  is bounded,
- $\|\mathbf{u}_m \mathbf{u}_{m+1}\| \to \mathbf{0},$

• 
$$\sum_{m=0}^{\infty} \|\mathbf{u}_m - \mathbf{u}_{m+1}\|^2 < \infty$$
,

• the set S of accumulation points of  $\{\mathbf{u}_m\}$  is connected and compact.

Conclusion: If S contains an isolated point  $\mathbf{u}^*$ , it follows that  $\mathbf{u}_m \to \mathbf{u}^*$ .

Note that, in general, the limit may depend on the starting value!

### **1.3 Study of Examples**

1.3.1 Case of d = 2

$$\mathbf{v}:=inom{1}{0}\otimesinom{1}{0}+2inom{0}{1}\otimesinom{0}{1},\quad \Phi(\mathbf{u})=\|\mathbf{v}-\mathbf{u}\|^2\,.$$

1)  $\mathbf{u}^{**} = 2\begin{pmatrix} 0\\1 \end{pmatrix} \otimes \begin{pmatrix} 0\\1 \end{pmatrix}$  is the global minimiser and an attractive fixed point. 2)  $\mathbf{u}^* = \begin{pmatrix} 1\\0 \end{pmatrix} \otimes \begin{pmatrix} 1\\0 \end{pmatrix}$  is a fixed point of the ALS iteration:

$$\Phi(\mathbf{u}^* + \delta_1 \otimes {\binom{1}{0}}) = \Phi(\mathbf{u}^*) + \|\delta_1\|^2.$$
  
But  $\Phi\left(\binom{1}{t} \otimes \binom{1}{t}\right) = \Phi(\mathbf{u}^*) - t^2\left(2 - t^2\right)$ 

 $\Rightarrow$  u<sup>\*</sup> is a saddle point and a repulsive fixed point.

Conclusion: Almost all starting values lead to  $\mathbf{u}_m \to \mathbf{u}^{**}$ .

### **1.3.2** Case of $d \ge 3$

For  $a \perp b$  with ||a|| = ||b|| = 1 consider  $\Phi(\mathbf{u}) = ||\mathbf{v} - \mathbf{u}||^2$  with  $\mathbf{v} = \otimes^3 a + 2 \otimes^3 b.$ 

Again  $\mathbf{u}^* = \otimes^3 a$  and  $\mathbf{u}^{**} = 2 \otimes^3 b$  are fixed points,  $\Phi(\mathbf{u}^{**}) < \Phi(\mathbf{u}^*)$ . But now both are local minima (attractive fixed points)! Additional saddle point (repulsive fixed point):  $\mathbf{u}^{***} = c \otimes^3 (a + \frac{1}{2}b)$ .

The sequence  $\{\mathbf{u}_m\}$  corresponding to the starting value

$$\mathbf{u}_{0} = c^{(0)} \left( a + t_{1}^{(0)} b \right) \otimes \left( a + t_{2}^{(0)} b \right) \otimes \left( a + t_{3}^{(0)} b \right)$$

is completely defined by  $t_2^{(0)}$  and  $t_3^{(0)}$ . The characteristic value is

$$au_m := \left| t_2^{(m)} \right|^{lpha} \left| t_3^{(m)} \right|^{eta} \quad ext{with} \quad lpha = 5^{1/2} - 1, \ eta = 2.$$

(A)  $\tau_0 > 2^{-\gamma}$ ,  $\gamma = 5^{1/2} + 1 \Rightarrow \mathbf{u}_m \to \mathbf{u}^{**}$  (global minimiser), (B)  $\tau_0 < 2^{-\gamma} \Rightarrow \mathbf{u}_m \to \mathbf{u}^*$  (local minimiser), (C)  $\tau_0 = 2^{-\gamma} \Rightarrow \mathbf{u}_m \to \mathbf{u}^{***}$  (saddle point, global minimiser on the manifold  $\tau = 2^{-\gamma}$ ). We recall:

Conclusion: If the set of accumulation points of  $\{u_m\}$  contains an isolated point  $u^*$ , it follows that  $u_m \to u^*$ .

Wang–Chu (2014): Global convergence for almost all  $u_0$ .

Uschmajew (2015):

Analysis based on the Łojasiewicz inequality yields: All sequences  $\mathbf{u}_m$  converge to some  $\mathbf{u}^*$  with  $\nabla \Phi(\mathbf{u}^*) = 0$ .

Łojasiewicz (1965, Ensembles semi-analytiques): If  $\Phi$  is analytic,

$$\exists \theta \in (0, 1/2] \quad |\Phi(x) - \Phi(x_*)|^{1-\theta} \leq \|\nabla \Phi(x)\|$$

in some neighbourhood of  $x_*$ .

### Convergence speed?

The proof by the Łojasiewicz inequality is not constructive.

Espig–Khachatryan (2015): Study of sequences for  $\Phi(\mathbf{u}) = \|\mathbf{v} - \mathbf{u}\|^2$  with

$$\mathbf{v} = \otimes^{3} a + \lambda \left( a \otimes a \otimes b + a \otimes b \otimes a + b \otimes a \otimes a \right),$$
  
$$a \perp b, \qquad \|a\| = \|b\| = 1.$$

Depending on the value of  $\lambda$  it is shown that the convergence can be

- sublinear ( $\lambda = 1/2$ ),
- linear ( $\lambda < 1/2$ ).

For  $\mathbf{v} = \otimes^3 a + 2 \otimes^3 b$ ,  $\mathbf{u}_m \to \otimes^3 a$  or  $2 \otimes^3 b$ , we have

• superlinear convergence (of order  $2 + 5^{1/2} > 1$ )

Study of the general case: Gong–Mohlenkamp–Young 2017

# 2 (Non-)Closedness Questions

## 2.1 *r*-Term Format, Rank of a Tensor

K: underlying field ( $\mathbb{R}$  or  $\mathbb{C}$ ).  $V_j$  vector spaces over K. Any algebraic tensor has the form  $\mathbf{v} = \sum_{i=1}^r \bigotimes_{j=1}^d v_i^{(j)}, v_i^{(j)} \in V_j$ , for some  $r \in \mathbb{N}_0$ . Fixing r, we obtain the set

$$\mathcal{R}_r := \left\{ \sum_{i=1}^r \bigotimes_{j=1}^d v_i^{(j)} : v_i^{(j)} \in V_j \right\}$$

of tensors with *representation rank* r. Using the rank

 $\mathsf{rank}(\mathbf{v}) := \min\{m : \mathbf{v} \in \mathcal{R}_m\},\$ 

we may write  $\mathcal{R}_r := \{\mathbf{v} \in \mathbf{V} : \mathsf{rank}(\mathbf{v}) \leq r\}$ .

### The maximal rank of $\mathbf{V}$ is

$$\mu := \sup\{\mathsf{rank}(\mathbf{v}) : \mathbf{v} \in \mathbf{V}\}.$$

 $\mu < \infty$  holds for finite-dimensional  $V_j$  and is equal to min $\{m : \mathcal{R}_{m+1} = \mathcal{R}_m\}$ .

## Properties of $\mathcal{R}_r$ :

- In general, the determination of rank(v) is NP hard (cf. Håstad 1990).
- In general, the maximal rank is not explicitly known. For equal dimensions  $dim(V_j) = n$ :

$$\frac{n^{d-1}}{d} \le r_{\max} \le \frac{d}{2(d-1)}n^{d-1} + O(n^{d-2}).$$

- For *random tensors* there may be more than one tensor rank with positive probability. These ranks are called *typical*.

- Real tensors may have different rank depending on the underlying fields  $\mathbb R$  or  $\mathbb C.$ 

- In general,  $\mathcal{R}_r$  is not closed. Example: a, b linearly independent and

$$\mathbf{v} = a \otimes a \otimes b + a \otimes b \otimes a + b \otimes a \otimes a \in \mathcal{R}_{3} \setminus \mathcal{R}_{2}$$
$$\mathbf{v} = \underbrace{(b+na) \otimes \left(a + \frac{1}{n}b\right) \otimes a + a \otimes a \otimes (b-na)}_{\mathbf{v}_{n} \in \mathcal{R}_{2}} - \frac{1}{n}b \otimes b \otimes a.$$
$$\mathbf{v}_{n} \in \mathcal{R}_{2}$$
- border rank: 
$$\underline{\mathrm{rank}}(\mathbf{v}) := \min\{r \in \mathbb{N}_{0} : \mathbf{v} \in closure(\mathcal{R}_{r})\}.$$

## Numerical Instability

In the previous example, the terms of  $v_n$  grow like O(n), while the result is of size O(1).

This implies *numerical cancellation*:  $\log_2 n$  binary digits of  $\mathbf{v}_n$  are lost.

We say that the sequence  $\{v_n\}$  is unstable.

**Proposition**: Suppose dim $(V_j) < \infty$  and  $\mathbf{v} \in \mathbf{V} = \bigotimes_{j=1}^d V_j$ . A stable sequence  $\mathbf{v}_n \in \mathcal{R}_r$  with  $\lim \mathbf{v}_n = \mathbf{v}$  exists if and only if  $\mathbf{v} \in \mathcal{R}_r$ .

Conclusion: If  $\mathbf{v} = \lim \mathbf{v}_n \notin \mathcal{R}_r$ , the sequence  $\mathbf{v}_n \in \mathcal{R}_r$  is unstable.

*Best approximation problem*: Let  $\mathbf{v}^* \in \mathbf{V}$ . Try to find  $\mathbf{v} \in \mathcal{R}_r$  with

$$\|\mathbf{v}^* - \mathbf{v}\| = \inf\{\|\mathbf{v}^* - \mathbf{w}\| : \mathbf{w} \in \mathcal{R}_r\}.$$

This optimisation problem need not be solvable.

The set of  $v^* \in V$  with inf  $\neq$  min has a positive measure if  $\mathbb{K} = \mathbb{R}$  (De Silva–Lim 2008), but measure zero if  $\mathbb{K} = \mathbb{C}$  (Qi–Michałek–Lim, 2017).

## **3** Strassen's Matrix Multiplication

Standard matrix-matrix multiplication costs  $2n^3$  operations. Strassen 1969:  $4.7n^{\log_2 7} = 4.7n^{2.8074}$ 

Two  $2 \times 2$  block matrices can be multiplied as follows:

$$\begin{bmatrix} a_1 & a_2 \\ a_3 & a_4 \end{bmatrix} \begin{bmatrix} b_1 & b_2 \\ b_3 & b_4 \end{bmatrix} = \begin{bmatrix} c_1 & c_2 \\ c_3 & c_4 \end{bmatrix}, \qquad a_i, b_i, c_i \text{ submatrices with}$$

$$c_1 = m_1 + m_4 - m_5 + m_7, c_2 = m_2 + m_4, c_3 = m_3 + m_5, c_4 = m_1 + m_3 - m_2 + m_6, m_1 = (a_1 + a_4)(b_1 + b_4), m_2 = (a_3 + a_4)b_1, m_3 = a_1(b_2 - b_4), m_4 = a_4(b_3 - b_1), m_5 = (a_1 + a_2)b_4, m_6 = (a_3 - a_1)(b_1 + b_2), m_7 = (a_2 - a_4)(b_3 + b_4).$$

Tensor of the matrix-matrix multiplication  $\begin{vmatrix} a_1 & a_2 \\ a_3 & a_4 \end{vmatrix} \begin{vmatrix} b_1 & b_2 \\ b_3 & b_4 \end{vmatrix} = \begin{vmatrix} c_1 & c_2 \\ c_3 & c_4 \end{vmatrix}$ :

$$c_{
u} = \sum_{\mu,\lambda=1}^{4} \mathbf{v}_{
u\mu\lambda} a_{\mu} b_{\lambda} \qquad (1 \le 
u \le 4).$$

For instance for  $\nu = 1$ , the identity  $c_1 = a_1b_1 + a_2b_3$  shows that  $\mathbf{v}_{111} = \mathbf{v}_{123} = 1$ , and  $\mathbf{v}_{1\mu\lambda} = 0$  otherwise. Assume a representation of  $\mathbf{v}$  by r terms:

$$\mathbf{v} = \sum_{i=1}^r \bigotimes_{j=1}^3 v_i^{(j)} \in \bigotimes_{j=1}^3 \mathbb{K}^4.$$

The insertion into  $c_{\nu} = \sum_{\mu,\lambda=1}^{4} \mathbf{v}_{\nu\mu\lambda} a_{\mu} b_{\lambda}$  yields

$$c_{\nu} = \sum_{i=1}^{r} \sum_{\mu,\lambda=1}^{4} v_i^{(1)}[\nu] \, v_i^{(2)}[\mu] \, v_i^{(3)}[\lambda] \, a_{\mu} \, b_{\lambda}$$
$$= \sum_{i=1}^{r} v_i^{(1)}[\nu] \left( \sum_{\mu=1}^{4} v_i^{(2)}[\mu] \, a_{\mu} \right) \left( \sum_{\lambda=1}^{4} v_i^{(3)}[\lambda] \, b_{\lambda} \right),$$

requiring r multiplications.

Strassen 1969:  $rank(v) \le 7$ , Winograd 1971: rank(v) = 7, Landsberg 2012: rank(v) = 7.

# 4 Matrix-Product (TT) Format, Tensor Networks

The hierarchical tensor format is based on a binary tree. A particular binary tree is  $\begin{array}{c} \scriptstyle \{1,2,3,4,5,6,7\}\\ \scriptstyle \{1,2,3,4,5,6\} \\ \scriptstyle \{1,2,3,4,5\} \\ \scriptstyle \{6\} \\ \scriptstyle \{1,2,3,4\} \\ \scriptstyle \{5\} \\ \scriptstyle \{1,2,3\} \\ \scriptstyle \{4\} \\ \scriptstyle \{1,2\} \\ \scriptstyle \{3\} \\ \scriptstyle \{1\} \\ \scriptstyle \{2\} \end{array}$ Choosing  $U_j := V_j$  for the subspaces at the leaves  $j = 1, \ldots, d$ , one obtains the TT format (Oseledets–Tyrtyshnikov 2005). It coincides with the

description of the matrix product states (Vidal 2003, Verstraete–Cirac 2006) used in physics:

Each component  $\mathbf{v}[i_1,\ldots,i_d]$  of  $\mathbf{v}\in\mathbf{V}=igotimes_{j=1}^d\mathbb{K}^{n_j}$  is expressed by

 $\mathbf{v}[i_1 i_2 \cdots i_d] = V^{(1)}[i_1] \cdot V^{(2)}[i_2] \cdot \ldots \cdot V^{(d-1)}[i_{d-1}] \cdot V^{(d)}[i_d] \in \mathbb{K},$ 

where  $V^{(j)}[i]$  are matrices of size  $r_{j-1} \times r_j$  with  $r_0 = r_d = 1$ . The minimal size of  $r_j$  is rank $\{1,...,j\}$ (v).

To avoid the special roles of the vectors  $V^{(1)}[i_1], V^{(d)}[i_d]$  and to describe periodic situations, the Cyclic Matrix-Product format  $C(d, (r_j))$  is used in physics:

$$\mathbf{v}[i_{1}i_{2}\cdots i_{d}] = \operatorname{trace}\{V^{(1)}[i_{1}] \cdot V^{(2)}[i_{2}] \cdots V^{(d-1)}[i_{d-1}] \cdot V^{(d)}[i_{d}]\} \\ = \sum_{k_{1}=1}^{r_{1}} \cdots \sum_{k_{d}=1}^{r_{d}} V_{k_{d}k_{1}}^{(1)}[i_{1}] \cdot V^{(2)}_{k_{1}k_{2}}[i_{2}] \cdots V^{(d-1)}[i_{d-1}] \cdot V_{k_{d-1}k_{d}}^{(d)}[i_{d}].$$

Tensor Network: tensor representations based on general graphs which are in general not a tree. Here the graph is a cycle with d vertices.

**THEOREM** (Landsberg–Qi–Ye 2012) Formats based on a graph  $\neq$  tree are in general not closed.

Site-independent format  $C_{ind}(d, r)$ :  $V^{(j)}[i] = V[i]$  and  $r_j = r$  for all j.

# 4.1 Example for d = 3, $V = \otimes^3 \mathbb{K}^{2 \times 2}$ , $r_1 = r_2 = r_3 = 2$ by Harris-Michałek-Sertöz 2018

### Let

$$\mathbf{m} := \sum_{k_1=1}^{r_1} \cdots \sum_{k_d=1}^{r_d} E_{k_d,k_1}^{(1)} \otimes E_{k_1k_2}^{(2)} \otimes \ldots \otimes E_{k_{d-2}k_{d-1}}^{(d-1)} \otimes E_{k_{d-1},k_d}^{(d)} \in \bigotimes_{j=1}^d \mathbb{K}^{r_{j-1} \times r_j}$$

$$E_{pq}^{(j)} \text{ is the matrix with entries } E_{pq}^{(j)}[k,\ell] = \delta_{pk}\delta_{q\ell}.$$

$$\{E_{pq}^{(j)}: 1 \le p \le r_{j-1}, 1 \le q \le r_j\} \text{ is the canonical basis of } \mathbb{K}^{r_{j-1} \times r_j}.$$

**LEMMA**. Let  $\mathbf{V} = \bigotimes_{j=1}^{d} V_j$ . The set  $\mathcal{C}(d, (r_j))$  consists of all

$$\mathbf{v} = \mathbf{\Phi}(\mathbf{m})$$
 with  $\mathbf{\Phi} = \bigotimes_{j=1}^d \phi^{(j)}$  and  $\phi^{(j)} \in L(\mathbb{K}^{r_{j-1} imes r_j}, V_j).$ 

In our case, we have  $\phi^{(j)} \in L(\mathbb{K}^{2 \times 2}, \mathbb{K}^{2 \times 2})$ . We first consider the site-independent case  $V^{(j)}[i] = V[i]$  for all  $1 \le j \le d := 3$ . Define  $\psi \in L(\mathbb{K}^{2\times 2}, \mathbb{K}^{2\times 2})$  by  $\psi(E_{12}) = E_{12} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$  and  $\psi(E_{pq}) = 0$  for  $(p,q) \neq (1,2)$ . Together with the identity  $id \in L(\mathbb{K}^{2\times 2}, \mathbb{K}^{2\times 2})$ , define  $\mathbf{v}(t) = \left(\otimes^3(\psi + t \cdot id)\right)$  (m) for  $t \in \mathbb{R}$ ,

where  $\mathbf{m} = \sum_{k_1=1}^2 \sum_{k_2=1}^2 \sum_{k_3=1}^2 E_{k_3k_1} \otimes E_{k_1k_2} \otimes E_{k_2k_3} \in \mathbf{V}$ . Multilinearity yields  $\mathbf{v}(t) = \mathbf{v}_0 + t \cdot \mathbf{v}_1 + t^2 \cdot \mathbf{v}_2 + t^3 \cdot \mathbf{v}_3$  with

 $\mathbf{v}_0 = (\otimes^3 \psi)(\mathbf{m}), \quad \mathbf{v}_1 = [\psi \otimes \psi \otimes id + \psi \otimes id \otimes \psi + id \otimes \psi \otimes \psi](\mathbf{m}), \\ \mathbf{v}_2 = [id \otimes id \otimes \psi + id \otimes \psi \otimes id + \psi \otimes id \otimes id](\mathbf{m}), \quad \mathbf{v}_3 = \mathbf{m}.$ 

Note that  $\psi(E_{ij}) \cdot \psi(E_{k\ell}) = 0$ . Since  $\mathbf{v}_0$  and  $\mathbf{v}_1$  involve three or two  $\psi$  applications,  $\mathbf{v}_0 = \mathbf{v}_1 = 0$  follows. Evaluation of  $\mathbf{v}_2$  yields

$$\mathbf{v}_{2} = E_{21} \otimes E_{11} \otimes E_{12} + E_{22} \otimes E_{21} \otimes E_{12} + E_{11} \otimes E_{12} \otimes E_{21} \\ + E_{21} \otimes E_{12} \otimes E_{22} + E_{12} \otimes E_{21} \otimes E_{11} + E_{12} \otimes E_{22} \otimes E_{21}.$$

 $\mathbf{v}_0 = \mathbf{v}_1 = 0$  allows us to form the limit  $\mathbf{v}_2 = \lim_{t \to 0} t^{-2} \mathbf{v}(t)$ . The Lemma states that  $t^{-2} \mathbf{v}(t) \in \mathcal{C}_{ind}(3,2)$  for t > 0.

The non-closedness of  $C_{ind}(3,2)$  will follow from  $v_2 \notin C_{ind}(3,2)$ .

For an indirect proof assume  $\mathbf{v}_2 \in \mathcal{C}_{ind}(3, 2)$ . The Lemma implies that there is some  $\phi \in L(\mathbb{K}^{2 \times 2}, \mathbb{K}^{2 \times 2})$  with  $\mathbf{v}_2 = (\otimes^3 \phi)(\mathbf{m})$ . It is easy to check that the range of the matricisation  $\mathcal{M}_1((\otimes^3 \phi)(\mathbf{m})) =$ 

 $\phi \mathcal{M}_1(\mathbf{m})(\otimes^2 \phi)^{\mathsf{T}}$  is  $\mathbb{K}^{2 \times 2}$ .

Therefore the map  $\phi$  must be surjective.

Since  $\phi \in L(\mathbb{K}^{2 \times 2}, \mathbb{K}^{2 \times 2})$ , surjectivity implies injectivity. Hence  $\phi : \mathbb{K}^{2 \times 2} \to \mathbb{K}^{2 \times 2}$  is a vector space isomorphism and  $\otimes^{3}\phi : \mathbf{V} \to \mathbf{V}$  a tensor space isomorphisms.  $\mathbf{v}_{2} = (\otimes^{3}\phi)(\mathbf{m}) \Rightarrow \operatorname{rank}(\mathbf{v}_{2}) = \operatorname{rank}(\mathbf{m})$ .

The representation of  $v_2$  yields  $rank(v_2) \le 6$ . On the other hand, rank(m) = 7 holds for the Strassen tensor m. This contradiction proves that  $v_2 \notin C_{ind}(3, 2)$ .

Similarly  $\mathbf{v}_2 \notin \mathcal{C}(3, (2, 2, 2))$  follows (no site-independence).

# 4.2 Example for $\mathbf{V} = \otimes^d \mathbb{C}^2$ , $r_j = 2$

Smallest (nontrivial) dimension:  $V_j = \mathbb{C}^2$ ,

tensor space  $\mathbf{V} = \otimes^d \mathbb{C}^2$ 

Site-independent cyclic format  $C_{ind}(d, 2)$ , i.e.,  $r_j = 2$ 

Result:

d = 3:  $C_{ind}(3, 2)$  is closed (cf. Harris–Michałek–Sertöz 2018)

d > 3:  $C_{ind}(d, 2)$  is not closed (cf. Seynnaeve 2018)

For  $\mathbb{K} = \mathbb{R}$ ,  $d \geq 3$ ,  $C_{ind}(d, 2)$  is not closed (cf. Seynnaeve 2018)

## **5** Minimal Subspaces

### 5.1 Tensor Subspace Format

Set of tensors of multilinear rank  $\leq \mathbf{r} = (r_1, \ldots, r_d) \in \mathbb{N}^d$  is

$$\mathcal{T}_{\mathbf{r}} := \bigcup_{\dim(U_j) \leq r_j} \bigotimes_{j=1}^d U_j.$$

Question: Is  $T_r$  closed?

In the finite-dimensional case, dim  $V_j < \infty$ , compactness arguments show that  $T_r$  is closed.

What happens in the case of infinite-dimensional Banach spaces  $\mathbf{V} = \bigotimes_{j=1}^{d} V_j$ ?

### 5.2 Minimal Subspaces

Let  $\mathbf{v} \in \mathbf{V} = \bigotimes_{j=1}^{d} V_j$  — possibly dim  $V_j = \infty$  — be an *algebraic* tensor. The minimal subspaces  $U_j^{\min}(\mathbf{v})$  are defined by

$$\mathbf{v} \in \bigotimes_{j=1}^{d} U_{j}^{\min}(\mathbf{v}), \text{ and}$$
  
if  $\mathbf{v} \in \bigotimes_{j=1}^{d} U_{j}$  ( $U_{j}$  subspace of  $V_{j}$ ), then  $U_{j}^{\min}(\mathbf{v}) \subset U_{j}$ .

REMARK: (a) dim  $U_j^{\min}(\mathbf{v}) \leq \operatorname{rank}(\mathbf{v}) < \infty$ . (b)  $\left( \bigotimes_{j=1}^d U_j' \right) \cap \left( \bigotimes_{j=1}^d U_j'' \right) = \bigotimes_{j=1}^d \left( U_j' \cap U_j'' \right)$ .

Conclusion:  $U_i^{\min}(\mathbf{v})$  is the subspace of minimal dimension in

$$\mathbf{v} \in U_j \otimes \mathbf{V}_{[j]}$$
 with  $\mathbf{V}_{[j]} := \bigotimes_{k \neq j} V_j$ .

### 5.2.1 Matricisation

The *j*-th matricisation  $\mathcal{M}_j : \mathbf{V} = \bigotimes_{k=1}^d \mathbb{K}^{n_k} \to \mathbb{K}^{n_j \times n_{[j]}}$  defined by  $\mathbf{v} \mapsto M_j := \mathcal{M}_j(\mathbf{v}) \in \mathbb{K}^{n_j \times n_{[j]}}$  with  $n_{[j]} := \prod_{k \neq j} n_k$ ,  $M_j[i_j, \mathbf{i}_{[j]}] := \mathbf{v}[i_1, \dots, i_d], \quad \mathbf{i}_{[j]} := (i_1, \dots, i_{j-1}, i_{j+1}, \dots, i_d).$ **REMARK**. If  $V_j = \mathbb{K}^{n_j}$ , then  $U_j^{\min}(\mathbf{v}) = range(\mathcal{M}_j(\mathbf{v})) = range(M_j).$ 

Consequences:

$$egin{aligned} \mathsf{rank}_j(\mathbf{v}) &:= \mathsf{rank}(\mathcal{M}_j(\mathbf{v})) & ext{ for } 1 \leq j \leq d, \ \mathcal{T}_{\mathbf{r}} &= \{\mathbf{v} \in \mathbf{V} : \mathsf{rank}_j(\mathbf{v}) \leq r_j ext{ for } 1 \leq j \leq d\}. \end{aligned}$$

Generalisation for infinite-dimensional Hilbert spaces is possible  $(n_j = \infty)$ , but not for general Banach spaces.

## **6** Tensor Spaces of Linear Mappings

Let  $V_j, W_j$  be vector spaces defining  $\mathbf{V} := \bigotimes_{j=1}^d V_j$  and  $\mathbf{W} := \bigotimes_{j=1}^d W_j$ . Then the sets of linear maps

$$L_j := L(V_j, W_j)$$

are again vector spaces. They define the tensor space

$$\mathbf{L} := \bigotimes_{j=1}^d L_j.$$

L can be embedded into  $L(\mathbf{V}, \mathbf{W})$ :  $\mathbf{A} = \bigotimes_{j=1}^d A^{(j)} \in \mathbf{L}$  is the linear map defined by

$$\mathbf{A}\bigotimes_{j=1}^{d} v^{(j)} := \bigotimes_{j=1}^{d} \left( A^{(j)} v^{(j)} \right).$$

## 6.1 Functionals

**REMARK**: dim
$$(V_j) = 1 \Rightarrow \mathbf{V} = \bigotimes_{k=1}^d V_k$$
 isomorphic to  
$$\mathbf{V}_{[j]} := V_1 \otimes V_2 \otimes \ldots \otimes V_{j-1} \otimes V_{j+1} \ldots \otimes V_d;$$

in particular

$$\mathbb{K}\otimes\mathbb{K}\otimes\ldots\otimes\mathbb{K}\otimes V_{j}\otimes\mathbb{K}\otimes\ldots\otimes\mathbb{K}\simeq V_{j}.$$

Let functionals  $\varphi_k : V_k \to \mathbb{K}$  be given for all  $k \neq j$ . Then

$$\varphi^{[j]} := \varphi_1 \otimes \ldots \otimes \varphi_{j-1} \otimes id \otimes \varphi_{j+1} \otimes \ldots \otimes \varphi_d \in L(\mathbf{V}, V_j)$$
maps V into  $V_j$ .

We identify 
$$\bigotimes_{k \neq j} \varphi_k \in \mathbf{V}'_{[j]}$$
 with  $\varphi^{[j]} \in L(\mathbf{V}, V_j)$ .

### 6.1.1 Minimal Subspaces

$$U_{j}^{\min}(\mathbf{v}) := \left\{ \varphi(\mathbf{v}) : \varphi \in a \bigotimes_{k \neq j} V_{k}' \right\}$$
$$= \left\{ \varphi(\mathbf{v}) : \varphi \in \left( a \bigotimes_{k \neq j} V_{k} \right)' \right\}.$$

 $V'_k$ : algebraic dual space of  $V_k$ .

In the finite-dimensional case, this statement is equivalent to  $U_j^{\min}(\mathbf{v}) = range(\mathcal{M}_j(\mathbf{v}))$ .

In the infinite-dimensional case, the definition of  $rank_j(v)$  can be extended by

 $\operatorname{rank}_{j}(\mathbf{v}) := \dim(U_{j}^{\min}(\mathbf{v})).$ 

Under rather general assumptions on the norms of  $V_i$  and V we shall prove that

$$\mathbf{v}_n 
ightarrow \mathbf{v} \qquad \Rightarrow \qquad \dim(U_j^{\min}(\mathbf{v})) \leq \liminf_{n \to \infty} \dim(U_j^{\min}(\mathbf{v}_n)).$$

### **Conclusion**:

(1) T<sub>r</sub> is weakly closed.
(2) If V is a reflexive Banach space,

$$\inf_{\mathbf{u} \in \mathcal{T}_{\mathbf{r}}} \|\mathbf{v} - \mathbf{u}\| = \|\mathbf{v} - \mathbf{u}_{\mathsf{best}}\|$$

has a solution  $\mathbf{u}_{\text{best}} \in \mathcal{T}_{\mathbf{r}}.$ 

Why weak convergence? There is a sequence  $\mathbf{u}_n \in \mathcal{T}_{\mathbf{r}}$  with  $\|\mathbf{v} - \mathbf{u}_n\| \rightarrow \inf_{\mathbf{u} \in \mathcal{T}_{\mathbf{r}}} \|\mathbf{v} - \mathbf{u}\|$ . In the reflexive case, there is subsequence such that  $\mathbf{u}_n \rightharpoonup \mathbf{u}_{\text{best}} \in \mathbf{V}$ .

 $\dim(U_j^{\min}(\mathbf{u}_n)) \leq r_j \Rightarrow \dim(U_j^{\min}(\mathbf{u}_{\text{best}})) \leq r_j \Rightarrow \mathbf{u}_{\text{best}} \in \mathcal{T}_{\mathbf{r}}.$ 

# 7 **Topological Tensor Spaces**

## 7.1 Case of Banach Spaces

 $V_j$  ( $1 \le j \le d$ ): normed space with  $\|\cdot\|_j$ , possibly a Banach space (i.e., complete).

 $\mathbf{V}_{\mathsf{alg}} := a \bigotimes_{j=1}^{d} V_j$  is the algebraic tensor space.

 $\|\cdot\|$  chosen norm on  $V_{alg}$ .

Completion of  $\mathbf{V}_{alg}$  w.r.t.  $\|\cdot\|$  yields the topological tensor space (Banach tensor space)

$$\mathbf{V}:=\mathbf{V}_{\mathsf{top}}:={}_{\|\cdot\|}igotimes_{j=1}^dV_j$$
 .

**REMARKS**: (1)  $V_{top}$  depends on the choice of  $\|\cdot\|$ (2)  $\|\cdot\|$  is not fixed by the norms  $\|\cdot\|_j$ .

### 7.2 Crossnorms

A necessary condition for reasonable topological tensor spaces is the continuity of the tensor product, i.e.,

$$\left\|\bigotimes_{j=1}^{d} v^{(j)}\right\| \leq C \prod_{j=1}^{d} \left\|v^{(j)}\right\|_{j}$$

for some  $C < \infty$  and all  $v^{(j)} \in V_j$ .

**DEFINITION**:  $\|\cdot\|$  is called a crossnorm if

$$\left\|\bigotimes_{j=1}^{d} v^{(j)}\right\| = \prod_{j=1}^{d} \left\|v^{(j)}\right\|_{j}$$

**REMARK:** There are different crossnorms  $\|\cdot\|$  for the same  $\|\cdot\|_j$ !

## 7.3 **Projective Norm** $\|\cdot\|_{\wedge}$

The *strongest possible norm* is the projective norm (Schatten, Grothendieck), defined by

$$\begin{aligned} \|\mathbf{v}\|_{\wedge(V_1,\dots,V_d)} &:= \|\mathbf{v}\|_{\wedge} \\ &:= \inf\left\{\sum_{i=1}^m \prod_{j=1}^d \left\|v_i^{(j)}\right\|_j : \mathbf{v} = \sum_{i=1}^m \bigotimes_{j=1}^d v_i^{(j)}\right\} \end{aligned}$$

for 
$$\mathbf{v} \in a \bigotimes_{j=1}^{d} V_j$$
.

•  $\left\|\cdot\right\|_{\wedge}$  is crossnorm.

 $\bullet$  Any norm  $\|\cdot\|$  satisfying the continuity requirement satisfies

 $\left\|\cdot\right\|\lesssim\left\|\cdot\right\|_{\wedge}.$ 

### **7.4** Duals and Injective Norm $\|\cdot\|_{\vee}$

The dual space  $V_j^*$  is the space of the continuous and linear functions on  $V_j$ . We now require:

also the tensor product  $\otimes : \times_{j=1}^{d} V_j^* \to {}_a \otimes_{j=1}^{d} V_j^*$  is continuous, i.e.,

$$\left\|\bigotimes_{j=1}^{d}\varphi_{j}\right\|^{*} \leq C \prod_{j=1}^{d} \left\|\varphi_{j}\right\|_{j}^{*} \text{ for all } \varphi_{j} \in V_{j}^{*}.$$

• For 
$$\mathbf{v} \in {}_{a} \bigotimes_{j=1}^{d} V_{j}$$
 define  $\|\cdot\|_{\vee(V_{1},...,V_{d})}$  by  
 $\|\mathbf{v}\|_{\vee(V_{1},...,V_{d})} := \|\mathbf{v}\|_{\vee} := \sup_{\substack{\mathbf{0} \neq \varphi_{j} \in V_{j}^{*} \\ 1 \leq j \leq d}} \frac{|(\varphi_{1} \otimes \varphi_{2} \otimes \ldots \otimes \varphi_{d})(\mathbf{v})|}{\prod_{j=1}^{d} \|\varphi_{j}\|_{j}^{*}}.$ 

•  $\|\cdot\|_{\vee}$  is a crossnorm.

•  $\|\cdot\|_{\vee}$  is the weakest norm with the continuity condition from above.

### 7.5 Minimal Subspaces, Final Part

ous.

We recall  $U_j^{\min}(\mathbf{v}) := \left\{ \varphi(\mathbf{v}) : \varphi \in a \otimes_{k \neq j} V'_k \right\}$ . Hahn-Banach theorem yields  $U_j^{\min}(\mathbf{v}) = \left\{ \varphi(\mathbf{v}) : \varphi \in a \bigotimes_{k \neq j} V^*_k \right\}$ .  $\varphi = \bigotimes_{k \neq j} \varphi^{(k)} \in a \bigotimes_{k \neq j} V'_k$  induces the map  $\varphi^{[j]} \in L(\mathbf{V}, V_j)$ . (1) If  $\|\cdot\| \gtrsim \|\cdot\|_{\vee}$  then  $\varphi \in a \bigotimes_{k \neq j} V^*_k$  implies that  $\varphi^{[j]} \in \mathcal{L}(\mathbf{V}, V_j)$  is continu-

(2) Weak convergence  $\mathbf{v}_n \rightharpoonup \mathbf{v}$  implies  $\varphi_{[j]}(\mathbf{v}_n) \rightharpoonup \varphi^{[j]}(\mathbf{v})$  in  $V_j$ .

Proof. For any  $\varphi^{(j)} \in V_j^*$  we have  $\varphi^{(j)}(\varphi^{[j]}(\mathbf{v}_n)) = (\bigotimes_k \varphi^{(k)})(\mathbf{v}_n)$ . Since  $\Phi := \bigotimes_k \varphi^{(k)} \in \mathbf{V}^*$ ,  $\mathbf{v}_n \rightharpoonup \mathbf{v}$  yields  $\Phi(\mathbf{v}_n) \rightarrow \Phi(\mathbf{v}) = \varphi^{(j)}(\varphi^{[j]}(\mathbf{v}))$ .

(3) Let the sequences  $(\mathbf{v}_n^{(i)})_{n\in\mathbb{N}}$  for  $1 \leq i \leq N$  converge weakly to linearly independent limits  $\mathbf{v}^{(i)} \in \mathbf{V}$  (i.e.,  $\mathbf{v}_n^{(i)} \rightarrow \mathbf{v}^{(i)}$ ). Then there is an  $n_0$  such that for all  $n \geq n_0$ , the N-tuples  $(\mathbf{v}_n^{(i)} : 1 \leq i \leq N)$  are linearly independent.

Hence  $\mathbf{v}_n \rightarrow \mathbf{v} \Rightarrow \dim(U_j^{\min}(\mathbf{v})) \leq \liminf_{n \rightarrow \infty} \dim(U_j^{\min}(\mathbf{v}_n)).$