THE ALGEBRAIC GEOMETRY ALGEBRAIC TOPOLOGY DUMP

ERNEST YEUNG ERNESTYALUMNI@GMAIL.COM

From the beginning of 2016, I decided to cease all explicit crowdfunding for any of my materials on physics, math. I failed to raise any funds from previous crowdfunding efforts. I decided that if I was going to live in abundance, I must lose a scarcity attitude. I am committed to keeping all of my material **open-sourced**. I give all my stuff for free.

In the beginning of 2017, I received a very generous donation from a reader from Norway who found these notes useful, through PayPal. If you find these notes useful, feel free to donate directly and easily through PayPal, which won't go through a 3rd. party such as indiegogo, kickstarter, patreon. Otherwise, under the open-source MIT license, feel free to copy, edit, paste, make your own versions, share, use as you wish.

gmail: ernestyalumni

linkedin: ernestyalumni twitter: ernestyalumni

Contents

Part 1.	Reading notes	on Cox, Li	ittle, O'Sh	ea's <i>Ideals</i> ,	Varieties,	and	Algorithms:	An	Introduction	tc
Computational Algebraic Geometry and Commutative Algebra										

- 1. Geometry, Algebra, and Algorithms
- 2. Groebner Bases
- 3. Elimination Theory
- 4. The Algebra-Geometry Dictionary
- 5. Polynomial and Rational Functions on a Variety
- 6. Robotics and Automatic Geometric Theorem Proving

Part 2. Reading notes on Cox, Little, O'Shea's Using Algebraic Geometry

- 7. Introduction
- 8. Solving Polynomial Equations
- 9. Resultants
- 10. Computation in Local Rings
- 11.
- 12.
- 13. Polytopes, Resultants, and Equations
- 14. Polyhedral Regions and Polynomials
- 15. Algebraic Coding Theory
- 16. The Berlekamp-Massey-Sakata Decoding Algorithm

Part 3. Algebraic Topology

17. Simplicial Complexes

Part 4. Graphs, Finite Graphs

18. Graphs, Finite Graphs, Trees

Date: 5 mars 2017.

Key words and phrases. Algebraic Geometry, Algebraic Topology.

	Part 5. Tensors, Tensor networks; Singular Value Decomposition, QR decomposition, Density Matrix Renormalization Group (DMRG), Matrix Product states (MPS)	(
	19. Introductions to Tensor Networks							
	20. Density Matrix Renormalization Group; Matrix Product States (MPS)	10						
1	21. Matrix Product States (MPS)	1						
1	References	10						
1 1 1	Abstract. Everything about Algebraic Geometry, Algebraic Topology							
1	Part 1. Reading notes on Cox, Little, O'Shea's Ideals, Varieties, and Algorithms: An Introduction to Computational Algebraic Geometry and Commutative Algebra							
1	1. Geometry, Algebra, and Algorithms							
2 3	1.1. Polynomials and Affine Space. fields are important is that linear algebra works over any field							
4	Definition 1 (2). set of all polynomials in x_1, \ldots, x_n with coefficients in k , denoted $k[x_1, \ldots, x_n]$							
4 4 4 4	polynomial f divides polynomial g provided $g = fh$ for some $h \in k[x_1, \dots, x_n]$ $k[x_1, \dots, x_n]$ satisfies all field axioms except for existence of multiplicative inverses; commutative ring, $k[x_1, \dots, x_n]$ polynomial f ring							
4 5 5	Exercises for 1. Exercise 1. \mathbb{F}_2 commutative ring since it's an abelian group under addition, commutative in multiplicative identity exists, namely 1. It is a field since for $1 \neq 0$, the multiplicative identity is 1.	ation						
0	Exercise 2.							
5	(a) (b)							

1

(c)

1.2. Affine Varieties.

6 1.3. Parametrizations of Affine Varieties.

ERNEST YEUNG ERNESTYALUMNI@GMAIL.COM

- 1.4. Ideals.
- 1.5. Polynomials of One Variable.
- 2. Groebner Bases

- 2.1. Introduction.
- 2.2. Orderings on the Monomials in $k[x_1, \ldots, x_n]$.
- 2.3. A Division Algorithm in $k[x_1, \ldots, x_n]$.
- 2.4. Monomial Ideals and Dickson's Lemma.
- 2.5. The Hilbert Basis Theorem and Groebner Bases.
- 2.6. Properties of Groebner Bases.
- 2.7. Buchberger's Algorithm.

- 3. Elimination Theory
- 3.1. The Elimination and Extension Theorems.
- 3.2. The Geometry of Elimination.
- 4. The Algebra-Geometry Dictionary
- 4.1. Hilbert's Nullstellensatz.
- 4.2. Radical Ideals and the Ideal-Variety Correspondence.
 - 5. Polynomial and Rational Functions on a Variety
- 5.1. Polynomial Mappings.
 - 6. Robotics and Automatic Geometric Theorem Proving
- 6.1. Geometric Description of Robots.
- Part 2. Reading notes on Cox, Little, O'Shea's Using Algebraic Geometry

Using Algebraic Geometry. David A. Cox. John Little. Donal O'Shea. Second Edition. Springer. 2005. ISBN 0-387-20706-6 QA564.C6883 2004

7. Introduction

7.1. Polynomials and Ideals. monomial

(1) $(1.1) x_1^{\alpha_1} \dots x_n^{\alpha_n}$

total degree of x^{α} is $\alpha_1 + \cdots + \alpha_n \equiv |\alpha|$

field $k, k[x_1 \dots x_n]$ collection of all polynomials in $x_1 \dots x_n$ with coefficients k.

polynomials in $k[x_1...x_n]$ can be added and multiplied as usual, so $k[x_1...x_n]$ has structure of commutative ring (with identity)

however, only nonzero constant polynomials have multiplicative inverses in $k[x_1 \dots x_n]$, so $k[x_1 \dots x_n]$ not a field however set of rational functions $\{f/g|f,g\in k[x_1\dots x_n],g\neq 0\}$ is a field, denoted $k(x_1\dots x_n)$

so

$$f = \sum_{\alpha} c_{\alpha} x^{\alpha}$$

where $c_{\alpha} \in k$

$$f \in k[x_1 \dots x_n] = \{f | f = \sum_{\alpha} c_{\alpha} x^{\alpha}, x^{\alpha} = x_1^{\alpha_1} \dots x_n^{\alpha_n}, c_{\alpha} \in k\}$$

f homogeneous if all monomials have same total degrees polynomial f is homogeneous if all monomials have the same total degree

Given a collection of polynomials $f_1 \dots f_s \in k[x_1 \dots x_n]$, we can consider all polynomials which can be built up from these by multiplication by arbitrary polynomials and by taking sums

Definition 2 (1.3). Let
$$f_1 ... f_s \in k[x_1 ... x_n]$$

Let $\langle f_1 ... f_s \rangle = \{p_1 f_1 + \dots + p_s f_s | p_i \in k[x_1 ... x_n] \text{ for } i = 1 ... s\}$

Exercise 1.

(a)
$$x^2 = x \cdot (x - y^2) + y \cdot (xy)$$

(b)

$$p \cdot (x - y^2) = px - py^2$$

and for pxy = (py)x

(c)

$$p(y)(x - y^2) = p(y)x - p(y)y^2 \notin \langle x^2, xy \rangle$$

Exercise 2

$$\sum_{i=1}^{s} p_i f_i + \sum_{j=1}^{s} q_j f_j = \sum_{i=1}^{s} (p_i + q_i) f_i, \quad p_i + q_i \in k[x_1 \dots x_n]$$

 $\langle f_1 \dots f_s \rangle$ closed under sums in $k[x_1 \dots x_n]$

If
$$f \in \langle f_1 \dots f_s \rangle$$
, $p \in k[x_1 \dots x_n]$

$$p \cdot f = p \sum_{i=1}^{s} q_j f_j = \sum_{i=1}^{s} p q_j f_j, \quad p q_j \in k[x_1 \dots x_n] \text{ so}$$

 $p \cdot f \in \langle f_1 \dots f_s \rangle$

Done.

The 2 properties in Ex. 2 are defining properties of ideals in the ring $k[x_1 \dots x_n]$

Definition 3 (1.5). Let $I \subset k[x_1 \dots x_n], I \neq \emptyset$ I ideal if

- (a) $f + g \in I$, $\forall f, g \in I$
- (b) $pf \in I$, $\forall f \in I$, arbitrary $p \in k[x_1 \dots x_n]$

Thus $\langle f_1 \dots f_s \rangle$ is an ideal by Ex. 2.

we call it the ideal generated by $f_1 \dots f_s$.

Exercise 3. Suppose
$$\exists$$
 ideal J , $f_1 \dots f_s \in J$ s.t. $J \subset \langle f_1 \dots f_s \rangle$ if $f \in \langle f_1 \dots f_s \rangle$, $f = \sum_{i=1}^s p_i f_i$, $p_i \in k[x_1 \dots x_n]$

 $\forall i = 1 \dots s, p_i f_i \in J$ and so $\sum_{i=1}^s p_i f_i \in J$, by def. of J as an ideal.

$$\langle f_1 \dots f_s \rangle \subseteq J \qquad \Longrightarrow J = \langle f_1 \dots f_s \rangle$$

 $\Longrightarrow \langle f_1 \dots f_s \rangle$ is smallest ideal in $k[x_1 \dots x_n]$ containing $f_1 \dots f_s$

Exercise 4. For
$$I = \langle f_1 \dots f_s \rangle$$

 $J = \langle g_1 \dots g_t \rangle$

 $I = J \text{ iff } s = t \text{ and } \forall f \in I, \ f = \sum_{i=1}^t q_i g_i \text{ and if } 0 = \sum_{i=1}^t q_i g_i, \ q_i = 0, \quad \forall i = 1 \dots t, \text{ and if } 0 = \sum_{i=1}^s p_i f_i, \quad p_i = 0, \quad I \text{ ideal if } f + g \in I \quad \forall f, g \in I$ $\forall i = 1 \dots s$

Definition 4 (1.6).

$$\sqrt{I} = \{g \in k[x_1 \dots x_n] | g^m \in I \text{ for some } m \ge 1\}$$

e.g.
$$x + y \in \sqrt{\langle x^2 + 3xy, 3xy + y^2 \rangle}$$

in $\mathbb{Q}[x, y]$ since

$$(x+y)^3 = x(x^2 + 3xy) + y(3xy + y^2) \in \langle x^2 + 3xy, 3xy + y^2 \rangle$$

- (Radical Ideal Property) \forall ideal $I \subset k[x_1 \dots x_n], \sqrt{I}$ ideal, $\sqrt{I} \supset I$
- (Hilbert basis Thm.) \forall ideal $I \subset k[x_1 \dots x_n]$ \exists finite generating set,

i.e. $\exists \{f_1 \dots f_2\} \subset k[x_1 \dots x_n] \text{ s.t. } I = \langle f_1 \dots f_s \rangle$

• (Division Algorithm in k[x]) $\forall f, g \in k[x]$ (EY: in 1 variable) $\forall f, g \in k[x] \text{ (in 1 variable)}$ $f = qq + r, \exists !$ quotient q, \exists remainder r

7.2.

7.3. Gröbner Bases.

Definition 5 (3.1). Gröbner basis for $I \equiv G = \{q_1 \dots q_k\} \subset I$ s.t. $\forall f \in I$, LT(f) divisible by $LT(q_i)$ for some i

• (Uniqueness of Remainders) let ideal $I \subset k[x_1 \dots x_n]$ division of $f \in k[x_1 \dots x_n]$ by Grö bner basis for I, produces f = q + r, $q \in I$, and no term in r divisible by any element of LT(I)

7.4. Affine Varieties. affine n-dim. space over k $k^n = \{(a_1 \dots a_n) | a_1 \dots a_n \in k\}$ \forall polynomial $f \in k[x_1 \dots x_n], (a_1 \dots a_n) \in k^n$ $f:k^n\to k$ $f(a_1 \dots a_n)$ s.t. $x_i = a_i$ i.e. if $f = \sum_{\alpha} c_{\alpha} x^{\alpha}$ for $c_{\alpha} \in k$, then $f(a_1 \dots a_n) = \sum_{\alpha} c_{\alpha} a^{\alpha} \in k$, where $a^{\alpha} = a_1^{\alpha_1} \dots a_n^{\alpha_n}$ **Definition 6** (4.1). affine variety $V(f_1 ... f_s) = \{(a_1 ... a_n) | (a_1 ... a_n) \in k^n, f_1(x_1 ... x_n) = ... = f_s(x_1 ... x_n) = 0\}$ subset $V \subset k^n$ is affine variety if $V = V(f_1 \dots f_s)$ for some $\{f_i\}$, polynomial $f_i \in k[x_1 \dots x_n]$ • (Equal Ideals Have Equal Varieties) If $\langle f_1 \dots f_s \rangle = \langle g_1 \dots g_t \rangle$ in $k[x_1 \dots x_n]$, then $\mathbf{V}(f_1 \dots f_s) = \mathbf{V}(g_1 \dots g_t)$ so, recap if $\langle f_1 \dots f_s \rangle = \langle g_1 \dots g_t \rangle$ in $k[x_1 \dots x_n]$, then $V(f_1 \dots f_s) = V(g_1 \dots g_t)$ Recall Hilbert basis Thm. \forall ideal $I \subset k[x_1 \dots x_n]$ $I = \langle f_1 \dots f_s \rangle$ \implies if I = J, then V(I) = V(J)think of V defined by I, rather than $f_1 = \cdots = f_s = 0$ Exercise 3. Recall Def. 1.5 Let $I \subset k[x_1 \dots x_n]$ $pf \in I$, $\forall f \in I$ arbitrary $p \in k[x_1 \dots x_n]$ Let $f, g \in I(V)$ $(f+q)(a_1 \dots a_n) = f(a_1 \dots a_n) + g(a_1 \dots a_n) = 0 + 0 = 0$ $f+q \in I(V)$ $pf(a_1 \dots a_n) = p(a_1 \dots a_n) f(a_1 \dots a_n) = 0$ $pf \in I(V)$ Then I(V) an ideal. $V = V(x^2)$ in \mathbb{R}^2 $I = \langle x^2 \rangle$ in $\mathbb{R}[x, y], I = \{px^2 | p \in k[x, y]\}$ $I \subset I(V)$, since $px^2 = 0$ for $x^2 = 0$, (0, b), $b \in \mathbb{R}$ But $p(x,y) = x \in I(V)$, as $I(V) = \{ f \in k[x_1 \dots x_n] | f(a_1 \dots a_n) = 0, \forall (a_1 \dots a_n) \in V \}$ p(0,b) = x = 0But $x \notin I$ Exercise 4. $I \subset \sqrt{I}$ Recall Def. 1.6 $\sqrt{I} = \{g \in k[x_1 \dots x_n] | g^m \in I \text{ for some } m \ge 1\}$ $\forall f \in I, f = f^1, m = 1, \text{ so } f \in \sqrt{I}, \quad I \subset \sqrt{I}$ Hilbert basis thm., \forall ideal $I \subset k[x_1 \dots x_n]$ s.t. $I = \langle f_1 \dots f_s \rangle$ $\{V(I) = \{(a_1...a_n) | (a_1...a_n) \in k^n, f_1(a_1...a_n) = \dots = f_s(a_1...a_n) = 0\}$ $\mathbf{I}(\mathbf{V}(I)) = \{ f \in k[x_1 \dots x_n] | f(a_1 \dots a_n) = 0 \quad \forall (a_1 \dots a_n) \in V(I) \}$ Let $g \in \sqrt{I}$, $g^m \in I$, $g^m = g^{m-1}g$

$$g^{m}(a_{1} \dots a_{n}) = 0 = g^{m-1}(a_{1} \dots a_{n})g(a_{1} \dots a_{n}) = 0$$
. Then $g(a_{1} \dots a_{n}) = 0$ or $g^{m-1}(a_{1} \dots a_{m}) = 0$

as $g^m \in I$, and V(I) is s.t. $f_1(a_1 \dots a_n) = \dots = f_s(a_1 \dots a_n) = 0$ for $I = \langle f_1 \dots f_s \rangle$

• (Strong Nullstellensatz) if k algebraically closed (e.g. \mathbb{C}), I ideal in $k[x_1 \dots x_n]$, then

$$\mathbf{I}(\mathbf{V}(I) = \sqrt{I}$$

• (Ideal-variety correspondence) Let k arbitrary field

$$I \subset I(V(I))$$
$$V(I(V)) = V \quad \forall V$$

Additional Exercises for Sec.4. Exercise 6.

8. Solving Polynomial Equations

8.1.

8.2. Finite-Dimensional Algebras. Gröbner basis $G = \{g_1 \dots g_t\}$ of ideal $I \subset k[x_1 \dots x_n]$, recall def.: Gröbner basis $G = \{g_1 \dots g_t\} \subset I$ of ideal $I, \forall f \in I, \mathrm{LT}(f)$ divisible by $\mathrm{LT}(g_i)$ for some i $f \in k[x_1 \dots x_n]$ divide by G produces $f = g + r, g \in I, r$ not divisible by any $\mathrm{LT}(I)$ uniqueness of r $f \in k[x_1 \dots x_n]$ divide by G,

Recall from Ch. 1, divide $f \in k[x_1 \dots x_n]$ by G, the division algorithm yields

$$(2.1) f = h_1 g_1 + \dots + h_t g_t + \overline{f}^G$$

where remainder \overline{f}^G is a linear combination of monomials $x^\alpha \notin \langle \operatorname{LT}(I) \rangle$ since Gröbner basis, $f \in I$ iff $\overline{f}^G = 0$ $\forall f \in k[x_1 \dots x_n]$, we have coset $[f] = f + I = \{f + h | h \in I\}$ s.t. [f] = [g] iff $f - g \in I$ We have a 1-to-1 correspondence

remainders
$$\leftrightarrow$$
 cosets

$$\overline{f}^G \leftrightarrow [f]$$

algebraic

$$\overline{f}^G + \overline{g}^G \leftrightarrow [f] + [g]$$
$$\overline{f}^G \cdot \overline{g}^G \leftrightarrow [f] \cdot [g]$$

 $B = \{x^{\alpha} | x^{\alpha} \notin \langle \operatorname{LT}(I) \rangle \} \text{ is a basis of } A, \text{ basis monomials, standard monomials} \\ 20141023 \text{ EY's take} \\ \forall [f] \in A = k[x_1 \dots x_n]/I, \quad [f] = p_i b_i; \quad b_i \in B = \{x^{\alpha} | x^{\alpha} \notin \langle \operatorname{LT}(I) \rangle \} \\ \text{For } I = \langle G \rangle \\ \text{e.g. } G = \{x^2 + \frac{3}{2}xy + \frac{1}{2}y^2 - \frac{3}{2}x - \frac{3}{2}y, xy^2 - x, y^3 - y \} \\ \langle \operatorname{LT}(I) \rangle = \langle x^2, xy^2, y^3 \rangle \\ \text{e.g. } B = \{1, x, y, xy, y^2 \} \\ [f] \cdot [g] = [fg] \\ \text{e.g. } f = x, g = xy, [fg] = [x^2y] \\ \text{now } f = h_1 q_1 + \dots + h_t q_t + \overline{f}^G$

8.3.

8.4. Solving Equations via Eigenvalues and Eigenvectors.

9. Resultants

10. Computation in Local Rings

10.1. Local Rings.

ERNEST YEUNG ERNESTYALUMNI@GMAIL.COM

Definition 7 (1.1).

$$k[x_1 \dots x_n]_{\langle x_1 \dots x_n \rangle} \equiv \{\frac{f}{g} | \text{ rational functions } \frac{f}{g} \text{ of } x_1 \dots x_n \text{ with } g(p) \neq 0 \text{ at } p\}$$

main properties of $k[x_1 \dots x_n]_{\langle x_1 \dots x_n \rangle}$

Proposition 1 (1.2). Let $R = k[x_1 \dots x_n]_{\langle x_1 \dots x_n \rangle}$. Then

- (a) R subring of field of rational functions $k(x_1...x_n) \supset k[x_1...x_n]$
- (b) Let $M = \langle x_1 \dots x_n \rangle \subset R$ (ideal generated by $x_1 \dots X_n$ in R) Then $\forall \frac{f}{g} \in R \backslash M$, $\frac{f}{g}$ unit in R (i.e. multiplicative inverse in R)
- (c) M maximal ideal in R

Exercise 1. if $p = (a_1 \dots a_n) \in k^n$, $R = \{\frac{f}{g} | f, g \in k[x_1 \dots x_n], g(p) \neq 0\}$

- (a) R subring of field of rational functions $k(x_1 ... x_n)$
- (b) Let M ideal generated by $x_1 a_1 \dots x_n a_n$ in RThen $\forall \frac{f}{g} \in R \backslash M$, $\frac{f}{g}$ unit in R (i.e. multiplicative inverse in R)
- (c) M maximal ideal in R

Proof. let $p = (a_1 \dots a_n) \in k^n$ let $g_1(p) \neq 0$, $g_2(p) \neq 0$

$$\frac{f_1}{g_1} + \frac{f_2}{g_2} = \frac{f_1 g_2 + f_2 g_1}{g_1 g_2} \qquad g_1(p) g_2(p) \neq 0 \text{ so } \frac{f_1}{g_1} + \frac{f_2}{g_2} \in R$$

$$\frac{f_1}{g_1} \cdot \frac{f_2}{g_2} = \frac{f_1 f_2}{g_1 g_2} \qquad g_1(p) g_2(p) \neq 0 \text{ so } \frac{f_1}{g_1} \frac{f_2}{g_2} \in R$$

 $f = \frac{f}{I} \in R$, $\forall f \in k[x_1 \dots x_n]$, so $k[x_1 \dots x_n] \subset R$

EY: 20141027, to recap,

Let $V = k^n$

Let $p = (a_1 \dots a_n)$

single pt. $\{p\}$ is (an example of) a variety

$$I(\{p\}) = \{x_1 - a_1 \dots x_n - a_n\} \subset k[x_1 \dots x_n]$$

 $R \equiv k[x_1 \dots x_n]_{\langle x_1 - a_1 \dots x_n - a_n \rangle}$

$$R = \{\frac{f}{g} | \text{ rational function } \frac{f}{g} \text{ of } x_1 \dots x_n, g(p) \neq 0, p = (a_1 \dots a_n) \}$$

Prop. 1.2. properties

- (a) R subring of field of rational functions $k(x_1 ... x_n) = k(x_1 ... x_n) \subset R$
- (b) $M = \langle x_1 \dots a_1 \dots x_n a_n \rangle \subset R$. ideal generated by $x_1 a_1 \dots x_n a_n$ Then $\forall \frac{f}{g} \in R \setminus M$, $\frac{f}{g}$ unit in R (\exists multiplicative inverse in R)
- (c) M maximal ideal in R.

in R we allow denominators that are not elements of this ideal $I(\{p\})$

Definition 8 (1.3). local ring is a ring that has exactly 1 maximal ideal

Proposition 2 (1.4). ring R with proper ideal $M \subset R$ is local ring if $\forall \frac{f}{g} \in R \backslash M$ is unit in R

localization Ex. 8, Ex. 9 parametrization

Exercise 2.

$$x = x(t) = \frac{-2t^2}{1+t^2}$$
$$y = y(t) = \frac{2t}{1+t^2}$$

$$\begin{array}{ll} k[t]_{\langle t \rangle} & \frac{-2t^2}{1+t^2} \text{ rational function of } t. \ 1+t^2 \neq 0 \\ \text{if } k = \mathbb{C} \text{ or } \mathbb{R} \end{array}$$

Consider set of convergent power series in n variables

(3)
$$k\{x_1 \dots x_n\} = \{ \sum_{\alpha \in \mathbb{Z}_{>0}^n} c_\alpha x^\alpha | c_\alpha \in k, \text{ series converges in some open } U \ni 0 \in k^n \}$$

Consider set $k[[x_1 \dots x_n]]$ of formal power series

(4)
$$k[[x_1 \dots x_n]] = \{ \sum_{\alpha \in \mathbb{Z}_{\geq 0}^n} c_{\alpha} x^{\alpha} | c_{\alpha} \in k \} \text{ series need not converge}$$

variety V

$$k[x_1 \dots x_n]/\mathbf{I}(V)$$
 variety V

10.2. Multiplicities and Milnor Numbers. if I ideal in $k[x_1 ... x_n]$, then denote $Ik[x_1 ... x_n]_{\langle x_1 ... x_n \rangle}$ ideal generated by I in larger ring $k[x_1 ... x_n]_{\langle x_1 ... x_n \rangle}$

Definition 9 (2.1). Let I 0-dim. ideal in $k[x_1...x_n]$, so V(I) consists of finitely many pts. in k^n . Assume $(0...0) \in V(I)$ multiplicity of $(0...0) \in V(I)$ is

$$dim_k k[x_1 \dots x_n]_{\langle x_1 \dots x_n \rangle} / Ik[x_1 \dots x_n]_{\langle x_1 \dots x_n \rangle}$$

generally, if $p = (a_1 \dots a_n) \in V(I)$ multiplicity of $p, m(p) = \dim k[x_1 \dots x_n]_M / Ik[x_1 \dots x_n]_M$

$$\dim k[x_1 \dots x_n]_M / Ik[x_1 \dots x_n]_M$$

localizing $k[x_1 \dots x_n]$ at maximal ideal $M = I(\{p\}) = \langle x_1 - a_1 \dots x_n - a_n \rangle$

11.

12.

- 13. Polytopes, Resultants, and Equations
- 14. POLYHEDRAL REGIONS AND POLYNOMIALS
- 14.1. Integer Programming. Prop. 1.12.

Suppose 2 customers A, B ship to same location

A: ship 400 kg pallet taking up $2 m^3$ volume

B: ship 500 kg pallet taking up $3 m^3$ volume

shipping firm trucks carry up to 3700 kg, up to $20 m^3$

B's product more perishable, paying \$ 15 per pallet

A pays \$ 11 per pallet

How many pallets from A, B each in truck to maximize revenues?

(5)
$$4A + 5B \le 37$$

$$2A + 3B \le 20$$

$$A, B \in \mathbb{Z}_{>0}^*$$

maximize 11A + 15B

integer programming.

max. or min. value of some linear function

$$l(A_1 \dots A_n) = \sum_{i=1}^n c_i A_i$$

on set $(A_1 \dots A_n) \in \mathbb{Z}_{>0}^n$ s.t.

3. Finally, by introducing additional variables; rewrite linear constraint inequalities as equalities. The new variables are called "slack variables"

(6)
$$(1.4) a_{ij}A_j = b_i, A_j \in \mathbb{Z}_{\geq 0}$$

introduce indeterminate z_i , \forall equation in (1.4)

$$z_i^{a_{ij}A_j} = z_i^{b_i}$$

m constraints

$$\prod_{i=1}^{m} z_i^{a_{ij}A_j} = \prod_{i=1}^{m} z_i^{b_i} = \left(\prod_{i=1}^{m} z_i^{a_{ij}}\right)^{A_j}$$

Proposition 3 (1.6). Let k field, define $\varphi : k[w_1 \dots w_n] \to k[z_1 \dots z_m]$ by

$$\varphi(w_j) = \prod_{i=1}^m z_i^{a_{ij}} \qquad \forall j = 1 \dots n$$

and

$$\varphi(g(w_1 \dots w_n)) = g(\varphi(w_1) \dots \varphi(w_n))$$

 \forall general polynomial $g \in k[w_1 \dots w_n]$

Then $(A_1 \ldots A_n)$ integer pt. in feasible region iff $\varphi : w_1^{A_1} \ldots w_n^{A_n} \mapsto z_1^{b_1} \ldots z_m^{b_m}$

Exercise 3.

Now

$$\varphi(w_j) = \prod_{i=1}^m z_i^{a_{ij}}$$
$$z_i^{a_{ij}A_j} = z_i^{b_i}$$

If $(A_1
ldots A_n)$ an integer pt. in feasible region, $a_{ij}A_j = b_i$

6 ERNEST YEUNG ERNESTYALUMNI@GMAIL.COM

$$z_i^{a_{ij}A_j} = z_i^{b_i} = \prod_{j=1}^n z_i^{a_{ij}A_j} \Longrightarrow \prod_{j=1}^n \prod_{i=1}^m (z_i^{a_{ij}})^{A_j} = \prod_{i=1}^m z_i^{b_i} = \prod_{j=1}^n \varphi(w_j)^{A_j} = \prod_{j=1}^n \varphi(w_j)^{A_j} = \varphi\left(\prod_{j=1}^n w_j^{A_j}\right) = \prod_{i=1}^m z_i^{b_i}$$

since $\varphi(g(w_1 \dots w_n)) = g(\varphi(w_1) \dots \varphi(w_n))$

If
$$\varphi: \prod_{j=1}^n w_j^{A_j} \mapsto \prod_{i=1}^m z_i^{b_i}$$

$$\varphi\left(\prod_{j=1}^{n} w_{j}^{A_{j}}\right) = \prod_{j=1}^{n} (\varphi(w_{j}))^{A_{j}} = \prod_{i=1}^{m} z_{i}^{b_{i}} = \prod_{j=1}^{n} \left(\prod_{i=1}^{m} z_{i}^{a_{ij}}\right)^{A_{j}} \Longrightarrow \prod_{j=1}^{n} z_{i}^{a_{ij}A_{j}} = z_{i}^{b_{i}}$$

or $a_{ij}A_j = b_i$. So $(A_1 \dots A_n)$ integer pt.

Exercise 4

$$\prod_{i=1}^{m} z_i^{b_i} = \prod_{i=1}^{m} \prod_{j=1}^{n} z_i^{a_{ij} A_j} = \prod_{j=1}^{n} \left(\prod_{i=1}^{m} z_i^{a_{ij}} \right)^{A_j} = \prod_{j=1}^{n} \varphi(w_j)^{A_j} = \varphi\left(\prod_{j=1}^{n} w_j^{A_j} \right)^{A_j}$$

So if given $(b_1 \dots b_m) \in \mathbb{Z}^m$, and for a given a_{ij} , $a_{ij}A_j = b_i$

For $m \leq n$, then a_{ij} is surjective, so $\exists A_j$ s.t. $\prod_{i=1}^m z_i^{b_i} = \varphi\left(\prod_{j=1}^n w_j^{A_j}\right)$

Proposition 4 (1.8). Suppose $f_1 \dots f_n \in k[z_1 \dots z_m]$ given

Fix monomial order in $k[z_1 \dots z_n, w_1 \dots w_n]$ with elimination property:

 \forall monomial containing 1 of z_i greater than any monomial containing only w_i

Let G Gröbner basis for ideal

$$I = \langle f_1 - w_1 \dots f_n - w_n \rangle \subset k[z_1 \dots z_m, w_1 \dots w_n]$$

 $\forall f \in k[z_1 \dots z_m], let \overline{f}^{\mathcal{G}}$ be remainder on division of f by \mathcal{G} Then

- (a) polynomial f s.t. $f \in k[f_1 \dots f_n]$ iff $g = \overline{f}^{\mathcal{G}} \in k[w_1 \dots w_n]$
- (b) if $f \in k[f_1 \dots f_n]$ as in part (a), $g = \overline{f}^{\mathcal{G}} \in k[w_1 \dots w_n]$

then $f = g(f_1 \dots f_n)$, giving an expression for f as polynomial in f_i

(c) if $\forall f_i, f \text{ monomials}, f \in k[f_1 \dots f_n],$ then q also a monomial.

14.2. Integer Programming and Combinatorics.

15. Algebraic Coding Theory

16. The Berlekamp-Massey-Sakata Decoding Algorithm

Gröbner Bases, Martin R. Albrecht of the DTU Crypto Group

Part 3. Algebraic Topology

cf. Bredon (1997) [3]

17. SIMPLICIAL COMPLEXES

cf. pp. 245, from Sec. 21 Simplicial Complexes of Ch. 4 Homology Theory in Bredon (1997) [3] $\mathbf{v}_0, \dots \mathbf{v}_n \in \mathbb{R}^{\infty}$, "affinely independent" if they span an affine *n*-plane, i.e.

if
$$\left(\sum_{i=0}^{n} \lambda_i \mathbf{v}_i = 0, \sum_{i=0}^{n} \lambda_i = 0\right)$$
, then $\Longrightarrow \forall \lambda_i = 0$

If not, then, e.g. $\lambda_0 \neq 0$, assume $\lambda_0 = -1$, and solve the equations to get

$$\mathbf{v}_0 = \sum_{i=1}^n \lambda_i \mathbf{v}_i$$
$$\sum_{i=1}^n \lambda_i = 1$$

i.e. \mathbf{v}_0 is in affine space spanned by $\mathbf{v}_1 \dots \mathbf{v}_n$.

If $\mathbf{v}_0, \dots \mathbf{v}_n$ affinely independent, then

(7)
$$\sigma = (\mathbf{v}_0, \dots \mathbf{v}_n) = \{ \sum_{i=0}^n \lambda_i \mathbf{v}_i | \sum_{i=0}^n \lambda_i = 1, \lambda_i \ge 0 \}$$

is "affine simplex" spanned by \mathbf{v}_i ; also convex hull of \mathbf{v}_i .

 $\forall k \leq n, k$ -face of σ is any affine simplex of form $(\mathbf{v}_{i_1}, \dots \mathbf{v}_{i_k})$, where vertices all distinct, so are affinely independent.

Definition 10. (geometric) simplicial complex K := collection of affine simplices s.t.

- (1) $\sigma \in K \Longrightarrow any face of \sigma \in K$: and
- (2) $\sigma, \tau \in K \Longrightarrow \sigma \cap \tau$ is a face of both σ and τ , or $\sigma \cap \tau = \emptyset$

If K simplicial complex, $|K| = \bigcup \{\sigma | \sigma \in K\} \equiv \text{"polyhedron" of } K$

Definition 11 (Def. 21.2 of Bredon (1997) [3]). polyhedron := space X if \exists homeomorphism $h: |K| \xrightarrow{\approx} X$ for some simplicial complex K. h, K is triangulation of X; (map h, complex K)

Let K finite simplicial complex.

Choose ordering of vertices $\mathbf{v}_0, \mathbf{v}_1 \dots$ of K.

If $\sigma = (\mathbf{v}_{\sigma_0}, \dots \mathbf{v}_{\sigma_n})$ is simplex of K, where $\sigma_0 < \dots < \sigma_n$, then

let
$$f_{\sigma}: \Delta_n \to |K|$$
 be

$$f_{\sigma} = [\mathbf{v}_{\sigma_b}, \dots \mathbf{v}_{\sigma_n}]$$

in notation of Def. 1.2. Bredon (1997) [3].

Then this gives CW-complex structure on |K| with f_{σ} as characteristic maps.

Part 4. Graphs, Finite Graphs

18. Graphs, Finite Graphs, Trees

Serre (1980) [4]

cf. Chapter I. Trees and Amalgams, Section 1 Amalgams, Subsection 1.1 Direct limits of Serre (1980) [4]

Let $(G_i)_{i \in I}$, family of groups.

 \forall pair (i,j), let F_{ij} = set of homomorphisms of G_i into G_j

Want: group $G = \lim_{i \to \infty} G_i$ and

$$\{f_i|f_i:G_i\to G\}$$
s.t. $f_j\circ f=f_i\quad \forall\, f\in F_{ij}$

group G and family $\{f_i\}$ universal in that

(*) if H group, if $\{h_i|h_i:G_i\to H;h_j\circ f=h_i \quad \forall f\in F_{ij}\},$

then $\exists !h: G \to H \text{ s.t. } h_i = h \circ f_i$

i.e. $\operatorname{Hom}(G, H) \simeq \varprojlim \operatorname{Hom}(G_i, H)$, the inverse limit being taken relative to F_{ij} . i.e. G direct limit of G_i relative to the F_{ij} .

Proposition 5. \exists ! pair G, family $(f_i)_{i \in I}$, i.e. (pair consisting of G, $(f_i)_{i \in I}$, unique up to unique isomorphism.

Proof. Define G by generators and relations.

Take generating family to be disjoint union of those for G_i .

relations - xyz^{-1} where $x, y, z \in G_i$, $z = xy \in G_i$

$$xy^{-1}$$
 where $x \in G_i$, $y \in G_i$, $y = f(x)$ for at least $f \in F_{ij}$.

Thus, existence of G, $\{f_i\}$.

G represents functor $H \mapsto \lim \operatorname{Hom}(G_i, H)$.

Thus, uniqueness (also from universal property).

e.g. groups A, G_1, G_2 , homomorphisms $f_1: A \to G_1$.

$$f_2:A\to G_2$$

G obtained by amalgamating A in G_1, G_2 by $f_1, f_2 \equiv G_1 *_A G_2$.

1 can have $G = \{1\}$, even though f_1, f_2 non-trivial.

Application: (Van Kampen Thm.)

Let topological space X be covered by open U_1, U_2 .

Suppose $U_1, U_2, U_{12} = U_1 \cap U_2$ arcwise connected.

Let basept. $x \in U_{12}$.

Then $\pi_1(X;x)$ obtained by taking 3 groups

$$\pi_1(U_1; x), \pi_1(U_2; x), \pi_1(U_{12}; x)$$

and amalagamating them according to homomorphism

$$\pi_1(U_{12};x) \to \pi_1(U_1;x)$$

$$\pi_1(U_{12};x) \to \pi_1(U_2;x)$$

Exercise 1. Let homomorphisms $f_1: A \to G_1$ amalgam $G = G_1 *_A G_2$.

$$f_2:A\to G_2$$

Define subgroups A^n, G_1^n, G_2^n , of A, G_1, G_2 recursively by

$$A^1 = \{1\}$$

$$G_1^1 = \{1\}$$

$$G_2^1 = \{1\}$$

 A^n = subgroup of A generated by $f_1^{-1}(G_1^{n-1})$ and $f_2^{-1}(G_2^{n-1})$

$$G_1^n = \text{subgroup of } G_i \text{ generated by } f_i(A^n)$$

Let A^{∞}, G_i^{∞} be unions of A^n, G_i^n resp.

Show that f_i defines injection $A/A^{\infty} \to G_i/G_i^{\infty}$.

So the amalgamation is $G \simeq G_1/G_1^{\infty} *_{A/A^{\infty}} G_2/G_2^{\infty}$.

Take the first induction case (for intuition about the solution).

$$A^{2} = \langle f_{1}^{-1}(G_{1}^{1}), f_{2}^{-1}(G_{2}^{1}) \rangle = \langle f_{1}^{-1}(\{1\}), f_{2}^{-1}(\{1\}) \rangle$$

$$G_{i}^{2} = f_{i}(A^{2})$$

Let
$$f_i(a) = f_i(b) \in G_i/G_i^{\infty}$$
; $a, b \in A/A^{\infty}$.

Then since $f_i(a), f_i(b) \in G_i/G_i^{\infty}, f_i(a), f_i(b) \in \{gG_i^{\infty}|g \in G_i\}$ (quotient is defined to be the set of all left cosets of G_i^{∞} , which has to be a normal subgroup for G_i/G_i^{∞} to be a quotient group).

Since $a, b \in A/A^{\infty}$, suppose we take $a, b \in A$.

And suppose we take

$$f_i(a) = f_i(a)G_i^{\infty} = f_i(a)f_i(A^{n_a}) = f_i(aA^{n_a})$$

 $f_i(b) = f_i(b)G_i^{\infty} = f_i(b)f_i(A^{n_b}) = f_i(bA^{n_b})$

Taking f_i^{-1} (recall for group homomorphisms, they map inverse of element of 1st. group to inverse of image of this element).

 $aA^{n_a} = bA^{n_b} \in A/A^{\infty}$ (This is okay as we've "quotiented out A^{∞} ; so indeed, they're equal)

cf. Subsection 1.2 Structure of amalgams of Serre (1980) [4]

Suppose given group A, family of groups $(G_i)_{i \in I}$, and, $\forall i \in I$, injective homomorphism $A \to G_i$.

 $*_AG_i \equiv$ direct limit (cf. no. 1.1) of family (A, G_i) with respect to these homomorphisms, call it *sum* (in category theory sense, i.e. product) of G_i with A amalgamated.

e.g.
$$A = \{1\},\$$

 $*G_i \equiv \text{free product of } G_i.$

18.0.1. reduced word. $\forall i \in I$, choose set S_i of right coset representations of G_i modulo A, assume $1 \in S_i$.

 $(a, s) \mapsto as$ is bijection of $A \times S_i$ onto G_i ,

$$A \times (S_i - \{1\}) \to G_i - A \text{ (onto)}$$

Let
$$\mathbf{i} = (i_1 \dots i_n), n \ge 0, i_j \in I, \text{ s.t.}$$

$$(8) i_m \neq i_{m+1} \text{ for } 1 \leq m \leq n-1$$

cf. (T) of Serre (1980) [4].

So reduced word m is defined as

$$m = (a; s_1 \dots s_n)$$

where $a \in A, s_1 \in S_{i_1} \dots s_n \in S_{i_n}$, and $s - j \neq 1 \,\forall j$.

 $f \equiv \text{canonical homomorphism of } A \text{ into group } G = *_A G_i$

 $f_i \equiv \text{canonical homomorphism of } G_i \text{ into group } G = *_A G_i$

EY: 20170611 (Further explanations, basic examples, from me):

Given A, $\{G_i\}_{i\in I}$, injective (group) homomorphisms $\{f_i: A \to G_i\}_i$.

 $G_i \backslash f_i(A) = \{ f_i(A)g | g \in G_i \}.$

Right coset representation of $f_i(A)g \mapsto g$.

e.g.
$$A, G_1, G_2, f_1: A \to G_1$$
.

$$f_2:A\to G_2$$

$$G_1 \backslash f_1(A) = \{ f_1(A)g | g \in G_1 \}$$

$$G_2 \backslash f_2(A) = \{ f_2(A)g | g \in G_2 \}$$

 $\mathbf{i} = (i_1 \dots i_n), i_j \in I, i_m \neq i_{m+1} \text{ for } 1 \leq m \leq n-1.$

Consider (1212...12)

 $m = (a; f_1g_2f_3g_4 \dots f_{2n-1}, g_{2n})$ where f's $\in S_1 \subset G_1$, g's $\in S_2 \subset G_2$. and so

Definition 12 (reduced word). reduced word of type i, m,

$$(9) m = (a; s_1 \dots s_n)$$

where
$$a \in A, s_1 \in S_{i_1}, \dots s_n \in S_{i_n}, s_j \neq 1 \quad \forall j,$$

 $\mathbf{i} = (i_1 \dots i_n), i_j \in I, s.t. \ i_m \neq i_{m+1} \ for \ 1 \leq m \leq n-1,$
with $S_i = \{q | q \in f_i(A)q \in f_i(A)G_i\}$

Theorem 1 (1 of Serre (1980) [4]). $\forall g \in G, \exists sequence i s.t. i_m \neq i_{m+1} for 1 \leq m \leq n-1 and reduced word$

$$m = (a; s_1 \dots s_n)$$

of type i s.t.

$$g = f(a)f_{i_1}(s_1)\dots f_{i_n}(s_n)$$

Furthermore, \mathbf{i} and m unique.

Remark. Thm. 1 implies f; f_i injective.

Then identify A and G_i with images $f(A), f_i(G_i)$ in G, and reduced decomposition (*) of $g \in G$

$$g = as_1 \dots s_n, \quad a \in A, s_1 \in S_{i_1} - \{1\} \dots s_n \in S_{i_n} - \{1\}$$

Likewise, $G_i \cap G_i = A$ if $i \neq j$.

In particular, $S_i - \{1\}$ pairwise disjoint in G.

Proof. Let $X_i \equiv \text{set of reduced words of type } \mathbf{i}, X = [X_i]$

Make G act on X.

In view of universal property of G, sufficient to make $\forall i, G_i$ act,

check action induced on A doesn't depend on i

Suppose then that $i \in I$, and let $Y_i = \text{set of reduced words of form } (1; s_1 \dots s_n)$, with $i_1 \neq i$.

EY: 20170611 Recall that

$$S_i = \{g | g \in f_i(A)g \in f_i(A)G_i\}$$

 $A \times S_i \to G_i \text{ onto}$
 $A \times (S_i - \{1\}) \to G_i - A \text{ onto}$
 $(a, s) \mapsto as \text{ bijection}$

Let Y_i = set of reduced words of form $(1; s_1 \dots s_n) = \{(1; s_1 \dots s_n) | 1 \in A; s_1 \in S_{i_1} \dots s_n \in S_{i_n}; \mathbf{i} = (i_1 \dots i_n), i_j \in I \text{ s.t. } i_m \neq i_{m+1} \text{ for } 1 \leq m \leq n-1\}.$

$$A \times Y_i \to X = \coprod_i X_i$$

$$(a, (1; s_1 \dots s_n)) \mapsto (a; s_1 \dots s_n)$$

$$A \times \{S_i - \{1\}\} \times Y_i \to X$$

$$((a, s), (1; s_1 \dots s_n)) \mapsto (a; s, s_1 \dots s_n)$$

and remember that $X_i = \text{set of reduced words of type } \mathbf{i}$.

It's clear that this yields a bijection $A \times Y_i \bigcup A \times (S_i - \{1\}) \times Y_i \to X$.

Let $x \in X$. Then $x \in X_i$ for some **i**. So x is a reduced word of type **i**: $x = (a; s_1 \dots s_n)$. Then clearly $x = (a; s_1 \dots s_n) \mapsto (a, (1; s_1 \dots s_n)) \in A \times Y_i$.

cf. pp. 13, Sec. 2. Trees, 2.1 Graphs of Serre (1980) [4]

Definition 13 (1. of Serre (1980) [4]).
$$\operatorname{graph} \Gamma = (X,Y,Y \to X \times X,Y \to Y), \text{ where } \operatorname{set} X = \operatorname{vert} \Gamma \operatorname{set} Y = \operatorname{edge} \Gamma \operatorname{Serre} Y \to X \times X$$

$$Y \to X \times X$$

$$Y \mapsto (o(y),t(y))$$

$$Y \to Y$$

$$Y \mapsto \overline{y}$$

s.t. $\forall y \in Y, \ \overline{y} = y, \ \overline{y} \neq y, \ o(y) = t(\overline{y}).$ vertex $P \in X$ of Γ . (oriented) edge $y \in Y, \ \overline{y} \equiv inverse$ edge. origin of $y := vertex \ o(y) = t(\overline{y}).$ terminus of $y := vertex \ t(y) = o(\overline{y})$ extremities of $y := \{o(y), t(y)\}$ If 2 vertices adjacent, they're extremities of some edge. orientation of graph $\Gamma = Y_+ \subset Y = edge \ \Gamma$ s.t. $Y = Y_+ \coprod \overline{Y}_+$. It always exists. oriented graph defined, up to isomorphism, by giving 2 sets X, Y_+ and $Y_+ \to X \times X$. corresponding set of edges is $Y = Y_+ \coprod \overline{Y}_+$ where $\overline{Y}_+ \equiv copy$ of Y_+

18.0.2. Realization of a Graph. cf. Realization of a Graph in Serre (1980) [4]. Let graph Γ , $X = \text{vert}\Gamma$, $Y = \text{edge}\Gamma$. topological space $T = X \coprod Y \times [0,1]$, where X,Y provided with discrete topology. Let R be finest equivalence relation on T for which

$$(y,t) \equiv (\overline{y}, 1-t)$$

$$(y,0) \equiv o(y) \qquad \forall y \in Y, \forall t \in [0,1]$$

$$(y,1) \equiv t(y)$$

quotient space real(Γ) = T/R is realization of graph Γ . (realization is a functor which commutes with direct limits). Let $n \in \mathbb{Z}^+$. Consider oriented graph of n+1 vertices $0,1,\ldots n$,

Definition 14. path (of length n) in graph Γ is morphism c of Path_n into Γ

orientation given by n edges $[i,i+1],\, 0 \leq i < n, \ o([i,i+1]) = i$ t([i,i+1]) = i+1

For $n \geq 1$,

 $(y_1 \dots y_n)$ sequence of edges $y_i = c([i-1,i])$ s.t.

$$t(y_i) = o(y_{i+1}), \qquad 1 \le i < n \text{ determine } c$$

□ If $P_i = c(i)$, c is a path from P_0 to P_n , and P_0 and P_n are extremities of the path c. pair of form $(y_i, y_{i+1}) = (y_i, \overline{y}_i)$ in path is **backtracking**. path (of length n-2), from P_0 to P_n given (for n>2) by $(y_1 \dots y_{i-1}, y_{i+2} \dots y_n)$ If ∃ path from P to Q in Γ , ∃ one without backtracking (by induction) direct limit Path_∞ = \varinjlim Path_n provides notion of infinite path. Path_∞ ∋ infinite sequence (y_1, y_2, \dots) of edges s.t. $t(y_i) = o(y_{i+1}) \quad \forall i \geq 1$.

Definition 15 (connected graph; Def. 3 of Serre (1980) [4]). graph connected if \forall 2 vertices, 2 vertices are extremities of at least 1 path.

maximal connected subgraphs (under relation of inclusion) are connected components of graph.

18.0.3. Circuits. Let $n \in \mathbb{Z}^+$, $n \ge 1$. Consider

set of vertices $\mathbb{Z}/n\mathbb{Z}$, orientation given by n edges [i, i+1], $(i \in \mathbb{Z}/n\mathbb{Z})$ with o([i, i+1]) = it([i, i+1]) = i+1

Definition 16 (circuit; Def. 4 of Serre (1980) [4]). circuit (length n) in graph is subgraph isormorphic to Circ_n.

i.e. subgraph = path $(y_1 \dots y_n)$, without backtracking, s.t. $P_i = t(y_i)$, $(1 \le i \le n)$ distinct, s.t. $P_n = o(y_1)$

$$n = 1$$
 case: Circ₁, $\mathbb{Z}/\mathbb{Z} = \{0\}$, 1 edge, $[0, 1]$, $0 \in \mathbb{Z}/1\mathbb{Z}$, $o([0, 1]) = 0$
 $t([0, 1]) = 1$

Note Circ₁ has automorphism of order 2, which changes its orientation, i.e.

 \exists automorphism $\sigma \in \text{Aut}(\text{Circ}_1) \text{ s.t. } |\sigma| = 2, \text{ i.e. } \sigma^2 = 1.$

loop := circuit of length 1; so loop $\in \overline{\text{Circ}}_1$.

path (y_1) , $P_1 = t(y_1) = o(y_1)$.

n = 2 case: Circ₂, $\mathbb{Z}/2\mathbb{Z} = \{0, 1\}$, 2 edges [0, 1], [1, 2],

path
$$(y_1, y_2)$$
, $(1 \le i \le 2)$, $P_1 = t(y_1)$
 $P_2 = t(y_2) = o(y_1)$

18.1. Combinatorial graphs. Let $(X,S) \equiv \text{simplicial complex of dim.} \leq 1$, with

 $X \equiv \text{set}$

 $S \equiv \text{set of subsets of } X \text{ with 1 or 2 elements, containing all the 1-element subsets.}$ associates with it a graph $\Gamma = (X, \{(P, Q)\})$.

X is its set of vertices.

edges =
$$\{(P,Q) \in X \times X\}$$
 s.t. $P \neq Q$, $\{P,Q\} \in S$, with $\overline{(P,Q)} = (Q,P)$

$$o(P,Q) = P$$

$$t(P,Q) = Q$$

In this graph, 2 edges with same origin and same terminus are equal. This is equivalent to (see following Def.)

Definition 17 (combinatorial; Def. 5 of Serre (1980) [4]). graph is combinatorial if it has no circuit of length ≤ 2

Conversely, it's easy to see that

every combinatorial graph Γ derived (up to isomorphism) by construction above from simplicial complex (X, S), where $X = \text{vert}\Gamma$

 $S = \text{set of subset } \{P, Q\} \text{ of } X \text{ s.t. } P \text{ and } Q \text{ either adjacent or equal.}$

Part 5. Tensors, Tensor networks; Singular Value Decomposition, QR decomposition, Density Matrix Renormalization Group (DMRG), Matrix Product states (MPS)

19. Introductions to Tensor Networks

José Barbon (IFT-CSIC, Univ. Autonoma de Madrid) gave the https://youtu.be/nsxgAOAEgbg for the workshop "Black Holes, Quantum Information, Entanglement, and all that," (29 May-1 June, 2017, with the organizing committee of Thibault Damour (IHES), Vasily Pestun (IHES), Eliezer Rabinovici (IHES & Hebrew Univ. of Jerusalem).

In the talk,

cf. 43:13

The church of the doubled Hilbert space. Any thermal box can be obtained by tracing over a second identical copy, if appropriately entangled into a global pure state.

$$\rho_R = \operatorname{Tr}_L \sum_n C_n \Psi_n^L \otimes \Psi_n^R$$

$$(C_n)_{\text{thermal}} = \left[\frac{e^{-\beta E_n}}{\sum_m e^{-\beta E_M}}\right]^{1/2}$$

But!!

If the entanglement basis is taken to be the high-energy band of two "entangled" CFTs ...

$$|TFD\rangle \sim \sum_{E} e^{-\beta E_n/2} |E_n\rangle_L \otimes |E_n\rangle_R$$

neglecting the tiny e^{-S} spacings. we can approximate by continuous spectrum of fields in the background of an AdS black hole, to get ...

$$\int_{E} e^{-\beta E/2} |E\rangle_{L} \otimes |E\rangle_{R}$$

The HH state of the bulk fields!

cf. 46:16

SLOGAN: EPR = ER Maldacena-Susskind

Accumulating a density of entanglement of $S \gg 1$ well-separated Bell pairs within a transversal size of order $(GS)^{1/2}$ seems to generate a geometrical bridge of area GS.

cf. 49:26

Parametrizing complexity of entanglement. Pick a tensor decomposition of Hilbert space of dimension $\exp(S)$ into S factors of O(1) dimension.

$$\mathcal{H} = \mathcal{H}_1 \otimes \mathcal{H}_2 \otimes \cdots \otimes \mathcal{H}_S$$

A tensor of S indices gives a generic state.

cf. 50:27

The decomposition of the big tensor in small building blocks gives a notion of "complexity of entanglement" rather simple entanglement pattern

somewhat more complex entanglement pattern

picture from M von Raamsdonk

cf. 55:10

A list of open questions & problems.

- Need exactly calculable toy models of AdS/CFT along the lines of SYK model
- Give a "renormalized" definition of quantum complexity for continuum CFTs
- Can tensor networks describe bulk gravitons?
- What is the space-time meaning of quantum complexity saturation?
- Can we define approximate local observables for black hole inferiors?
- Are there obstructions related to firewalls and/or fuzzballs?

Workshop introductory overview by José Barbon for the Institut des Hautes Études Scientifiques (IHÉS) gave me the first impetus to understand tensor networks as I sought to also understand the condensates of entanglement pairs within the black hole.

A Google search for introductions to tensor networks that are on arxiv ("Introduction Tensor Network arxiv") yielded Bridgeman and Chubb's course notes (bf. Bridgeman and Chubb (2017) [7]).

ERNEST YEUNG ERNESTYALUMNI@GMAIL.COM

19.1. List of stuff I want to look at/do/study. I would like to compare/contrast the following:

- Rotman (2010) [5], Ch. 8, but starting from 8.4 Tensor Products, pp. 574
- Jeffrey Lee (2009) [6], Ch. 7 Tensors

Maldacena and Susskind (2013) [11]

Lectures on Gravity and Entanglement. Mark Van Raamsdonk [14]

- Consider as physical system AdS-Schwarzchild black hole
- CFT
- PFL Lectures on Conformal Field Theory in $D \geq 3$ Dimensions, Rychkov (2016) [12].

Evenbly and Vidal (2011) [13], Tensor network states and geometry

Loose ends (might not be useful links)

- https://arxiv.org/pdf/1506.06958.pdf
- https://arxiv.org/pdf/1512.02532.pdf One-point Functions in AdS/dCFT from Matrix Product States

19.2. Tensor operations; Tensor properties.

19.2.1. rank. $r = \text{rank tensor of dim. } d_1 \times \cdots \times d_r \text{ is element of } \mathbb{C}^{d_1 \times \cdots \times d_r}$

Tensor product

$$[A \otimes B]_{i_1 \dots i_r, j_1 \dots j_s} := A_{i_1 \dots i_r} \cdot B_{j_1 \dots j_s}$$

19.2.2. Trace. Given tensor A, xth, yth indices have identical dims. $(d_x = d_y)$, partial trace over these 2 dims. is simply joint summation over that index

(12)
$$[\operatorname{Tr}_{x,y} A]_{i_1 \dots i_{x-1} i_{x+1} \dots i_{y-1} i_{y+1} \dots i_r} = \sum_{\alpha=1}^{d_x} A_{i_1 \dots i_{x-1} \alpha i_{x+1} \dots i_{y-1} \alpha i_{y+1} \dots i_r}$$

19.2.3. Contraction.

19.2.4. Group and splitting, Bridgeman and Chubb (2017) [7]. "Rank is a rather fluid concept in the study of tensor networks." Bridgeman and Chubb (2017) [7].

 $\mathbb{C}^{a_1 \times \cdots \times a_n} \simeq \mathbb{C}^{b_1 \times \cdots \times b_m}$ isomorphic as vector spaces if $\prod_i a_i = \prod_i b_i$.

We can "group" or "split" indices to lower or raise rank of given tensor, resp.

Consider contracting 2 arbitrary tensors.

If we group together indices which are and are not involved in contraction,

"It should be noted that not only is this reduction to matrix multiplication pedagogically handy, but this is precisely the manner in which numerical tensor packages perform contraction, allowing them to leverage highly optimised matrix multiplication code." (cf. Bridgeman and Chubb (2017) [7]; check this)

"Owing to freedom in choice of basis, precise details of grouping and splitting aren't unique." (cf. Bridgeman and Chubb (2017) [7]).

1 specific choice of convention:

tensor product basis, defining basis on product space by product of respective bases.

"The canonical use of tensor product bases in quantum information allows for grouping and splitting described above to be - dealt with implicitly."

$$|0\rangle \otimes |1\rangle \equiv |0\rangle$$

and precisely this grouping,

$$|0\rangle \otimes |1\rangle \in \operatorname{Mat}_{\mathbb{C}}(2,2), \text{ whilst}$$

$$|01\rangle \in \mathbb{C}^4$$

Suppose rank n+m tensor T, group its first n indices, last m indices together.

$$T_{I,J} := T_{i_1...i_n,j_1...j_m}$$

where

$$I := i_1 + d_1^{(i)} i_2 + d_1^{(i)} d_2^{(i)} i_3 + \dots + d_1^{(i)} \dots d_{n-1}^{(i)} i_n$$

$$J := j_1 + d_1^{(j)} j_2 + d_1^{(j)} d_2^{(j)} j_3 + \dots + d_1^{(j)} \dots d_{m-1}^{(j)} j_m$$

EY: 20170627 to elaborate, consider a functor flatten that does what's described above, in the context of category theory (and so this is the generalization):

$$\mathbb{K}^{d_{1}^{(i)}} \times \mathbb{K}^{d_{2}^{(i)}} \times \cdots \times \mathbb{K}^{d_{n}^{(i)}} \times \mathbb{K}^{d_{1}^{(j)}} \times \mathbb{K}^{d_{2}^{(j)}} \times \cdots \times \mathbb{K}^{d_{m}^{(j)}} \xrightarrow{\text{flatten}} \mathbb{K}^{\prod_{p=1}^{n} d_{p}^{(i)}} \times \mathbb{K}^{\prod_{q=1}^{m} d_{q}^{(j)}}$$

$$T_{i_{1}...i_{n},j_{1}...j_{m}} \xrightarrow{\text{flatten}} T_{I,J}$$

$$\{0,1,...d_{1}^{(i)}\} \times \{0,1,...d_{2}^{(i)}\} \times \cdots \times \{0,1,...d_{n}^{(i)}\} \times \{0,1,...d_{1}^{(j)}\} \times \{0,1,...d_{2}^{(j)}\} \times \cdots \times \{0,1,...d_{m}^{(j)}\} \xrightarrow{\text{flatten}}$$

$$\xrightarrow{\text{flatten}} \{0,1,\cdots \prod_{p=1}^{n} d_{p}^{(i)} - 1\} \times \{0,1,\cdots \prod_{q=1}^{m} d_{q}^{(j)} - 1\}$$

$$(i_{1},i_{2},...i_{n},j_{1},j_{2}...j_{m}) \xrightarrow{\text{flatten}} (I,J) := (i_{1}+d_{1}^{(i)}i_{2}+\cdots+d_{1}^{(i)}...d_{n-1}^{(i)}i_{n},j_{1}+d_{1}^{(j)}j_{2}+\cdots+d_{1}^{(j)}...d_{m-1}^{(j)}j_{m})$$

It doesn't make sense to call this "row-major" or "column-major" ordering generalization, because we are not dealing with only 2 indices where we can definitely say the first index indexes the "row" and the second index indexes the "column." At most, possibly, you can alternatively have this:

$$(i_1 \dots i_n, j_1 \dots j_m) \xrightarrow{\text{flatten}} (I, J) := (d_2^{(i)} \dots d_n^{(i)} i_1 + d_3^{(i)} \dots d_n^{(i)} i_2 + \dots + i_n, d_2^{(j)} \dots d_m^{(j)} j_1 + \dots + j_m)$$

Note that this is all 0-based counting (i.e. we start counting from 0 just like in C,C++,Python, etc.). If you really wanted 1-based counting, you'd have to complicate the above formulas as such:

$$(I,J) := (i_1 + d_1^{(i)}(i_2 - 1) + \dots + d_1^{(i)} \dots d_{n-1}^{(i)}(i_n - 1), j_1 + d_1^{(j)}(j_2 - 1) + \dots + d_1^{(j)} \dots d_{m-1}^{(j)}(j_m - 1))$$

Note that formulas are easily checked by pluggin in the minimum and maximum values for the indices and seeing if they make sense (e.g. plug in $(0,0,\ldots,0)$ for all indices for 0-based counting and make sure you get back I=0 or J=0).

19.3. Singular Value Decomposition.

$$T_{I,J} = \sum_{\alpha} U_{I,\alpha} S_{\alpha,\alpha} \overline{V}_{J,\alpha}$$

$$\operatorname{Mat}_{\mathbb{K}}(N,M) \xrightarrow{\operatorname{SVD}} \operatorname{Mat}_{\mathbb{K}}(N,P) \times \operatorname{Mat}_{\mathbb{K}}(P,P) \times \operatorname{Mat}_{\mathbb{K}}(M,P)$$

$$T_{I,J} \xrightarrow{\operatorname{SVD}} U_{I,\alpha}, S_{\alpha,\alpha}, \overline{V}_{I,\alpha} \text{ s.t.}$$

$$T_{I,J} = \sum_{\alpha} U_{I,\alpha} S_{\alpha,\alpha} \overline{V}_{J,\alpha}$$

$$T = USV^{\dagger}$$

For the higher-dimensional version of SVD

$$\mathbb{K}^{d_1^{(i)}} \otimes \cdots \otimes \mathbb{K}^{d_N^{(i)}} \otimes \mathbb{K}^{d_1^{(j)}} \otimes \cdots \otimes \mathbb{K}^{d_M^{(j)}} \xrightarrow{\text{flatten}} \operatorname{Mat}_{\mathbb{K}}(N, M) \xrightarrow{\operatorname{SVD}} \operatorname{Mat}_{\mathbb{K}}(N, P) \times \operatorname{Mat}_{\mathbb{K}}(P, P) \times \operatorname{Mat}_{\mathbb{K}}(M, P) \xrightarrow{\text{splitting}}$$

$$\xrightarrow{\text{splitting}} \mathbb{K}^{d_1^{(i)}} \otimes \cdots \otimes \mathbb{K}^{d_N^{(i)}} \otimes \mathbb{K}^P \times \operatorname{Mat}_{\mathbb{K}}(P, P) \times \mathbb{K}^{d_1^{(j)}} \otimes \cdots \otimes \mathbb{K}^{d_M^{(j)}} \otimes \mathbb{K}^P$$

$$T_{i_1 \dots i_N, j_1 \dots j_M} = \sum_{\alpha} U_{i_1 \dots i_N, \alpha} S_{\alpha, \alpha} \overline{V}_{j_1 \dots j_M, \alpha}$$

20. Density Matrix Renormalization Group; Matrix Product States (MPS)

cf. Sec. 4, Matrix Product States (MPS) of Schollwöck [9].

Necessarily, given matrix $M \in \operatorname{Mat}_{\mathbb{K}}(M, N)$ (notation in Bridgeman and Chubb (2017) [7] and CUDA Toolkit Documentation; I will follow the notation in Schollwöck [9] since his A, B denote specific physical meaning).

For

$$U \in \operatorname{Mat}_{\mathbb{K}}(N_A, \min(N_A, N_B)) \text{ s.t. } UU^{\dagger} = 1$$

 $S \in \operatorname{Mat}_{\mathbb{K}}(\min(N_A, N_B), \min(N_A, N_B))$

s.t. S diagonal with nonnegative $S_{aa}=s_a$, i.e. $S_{ij}=\delta_{ij}s_i$ s.t. $s_i\geq 0 \quad \forall i=1,2,\ldots \min{(N_A,N_B)}$.

 $r \equiv (Schmidt)$ rank of M := number of nonzero singular values.

Assume $s_1 \geq s_2 \geq \cdots \geq s_r \geq 0$.

 $V^{\dagger} \in \operatorname{Mat}_{\mathbb{K}}(\min(N_A, N_B), N_B) \text{ s.t. } V^{\dagger}V = 1.$

$$\operatorname{Mat}_{\mathbb{K}}(N_{A}, N_{B}) \xrightarrow{\operatorname{SVD}} U_{\mathbb{K}}(N_{A}, \min{(N_{A}, N_{B})}) \times \operatorname{diag}_{\mathbb{K}}(\min{(N_{A}, N_{B})}) \times U_{\mathbb{K}}(\min{(N_{A}, N_{B})}, N_{B})$$

$$M \vdash \longrightarrow USV^{\dagger}$$

Optimal approximation of M (rank r by matrix M' (rank r' < r) property.

In Frobenius norm $||M||_F^2 := \sum_{i,j} |M_{ij}|^2$, induced by inner product $\langle M|N\rangle = \operatorname{tr} M^{\dagger} N$. Indeed,

$$\operatorname{tr} M^{\dagger} N = (M^{\dagger})_{ik} N_{ki} = \overline{M}_{ki} N_{ki}$$

and so for

(18)
$$M' = US'V^{\dagger}, \qquad S' = \operatorname{diag}(s_1, s_2 \dots s_{r'}, 0 \dots)$$

cf. Eq. (19) of Schollwöck [9], i.e. 1 sets all but 1st r' singual values to 0.

Use singular value decomposition (SVD) to derive Schmidt decomposition of general quantum state. \forall pure state $|\psi\rangle$ on AB,

$$|\psi\rangle = \sum_{i,j} \Psi_{ij} |i\rangle_A |j\rangle_B$$

where $\{|i\rangle_A\}, \{|j\rangle_B\}$ orthonormal bases of A, B ((complex) Hilbert spaces), with dim. N_A, N_B , respectively. Let $\Psi_{i,j} \in \operatorname{Mat}_{\mathbb{K}}(N_A, N_B)$.

Then reduced density operators $\hat{\rho}_A, \hat{\rho}_B$ are such that

$$\widehat{\rho}_A = \operatorname{tr}_B |\psi\rangle\langle\psi|$$
 $\widehat{\rho}_B = \operatorname{tr}_A |\psi\rangle\langle\psi|$

In matrix form,

$$\rho_A = \Psi \Psi^{\dagger}$$

$$\rho_B = \Psi^{\dagger} \Psi$$

Indeed,

$$(\rho_A)_{ij} = \Psi_{ik}\overline{\Psi}_{jk}$$

$$(\rho_B)_{ij} = \overline{\Psi}_{ki}\Psi_{kj}$$

$$|\psi\rangle\langle\psi| = \sum_{i,j} \Psi_{ij}|i\rangle_A|j\rangle_B \sum_{l,m} \overline{\Psi}_{lm}\langle l|_A\langle m|_B$$

$$\operatorname{tr}_B|\psi\rangle\langle\psi| = \sum_{i,j} \Psi_{ik}\overline{\Psi}_{jk}|i\rangle_A\rangle j|_A$$

In matrix form.

$$\rho_A = \Psi \Psi^{\dagger}$$

$$\rho_B = \Psi^{\dagger} \Psi$$

Carry out SVD on Ψ in Eq. (20) of Schollwöck [9].

$$|\psi\rangle = \sum_{i,j} \Psi_{ij} |i\rangle_A |j\rangle_B$$

$$|\psi\rangle = \sum_{ij} \Psi_{ij} |i\rangle_A |j\rangle_B = \sum_{ij} \sum_{a=1}^{\min{(N_A,N_B)}} U_{ia} S_{aa} \overline{V}_{ja} |i\rangle_A |j\rangle_B = \sum_{a=1}^{\min{(N_A,N_B)}} \sum_i U_{ia} |i\rangle_A s_a \sum_j \overline{V}_{ja} |j\rangle_B = \sum_{a=1}^{\min{(N_A,N_B)}} s_a |a\rangle_A |a\rangle_B$$

Due to orthogonality of $U, V^{\dagger}, \{|a\rangle_A\}, \{|a\rangle_B\}$ orthonormal, and can be extended to be orthonormal bases of A, B.

If we restrict the sum to run only over the $r \leq \min(N_A, N_B)$ positive nonzero singular values (i.e., for $\sum_{a=1}^{\min(N_A, N_B)}$, a > 0 $\forall a \leq r$, and so

$$|\psi\rangle = \sum_{a=1}^{r} s_a |a\rangle_A |a\rangle_B$$

r=1 (classical) product states. $|\psi\rangle = s_1|1\rangle_A|1\rangle_B$.

r > 1 entangled (quantum) states.

Schmidt decomposition on reduced density operators for A and B:

$$\widehat{\rho}_A = \sum_{a=1}^r s_a^2 |a\rangle_A \langle a|_A$$

$$\widehat{\rho}_B = \sum_a^r s_a^2 |a\rangle_B \langle a|_B$$

Respective eigenvectors are left and right singular vectors.

Von Neumann entropy can be read off:

$$S_{A|B}(|\psi\rangle) = -\operatorname{tr}\widehat{\rho}_A \log_2 \widehat{\rho}_A = -\sum_{a=1}^r s_a^2 \log_2 s_a^2$$

In view of large size of Hilbert spaces, approximate $|\psi\rangle$ by some $|\widetilde{\psi}\rangle$ spanned over state spaces A, B that have dims. r' only. Since 2-norm of $|\psi\rangle$,

$$\||\psi\rangle\|_2^2 = \sum_{ij} |\Psi_{ij}|^2 = \|\Psi\|_F^2$$

since

$$\|\psi\|_{2}^{2} = \sum_{a=1}^{r} s_{a}^{2} = \sum_{ij} |\Psi_{ij}|^{2}$$

iff $\{|i\rangle\}, \{|j\rangle\}$ orthonormal. Optimal approx. of 2-norm given by optimal approx. of Ψ by $\overline{\Psi}$ in Frobenius norm, where $\overline{\Psi}$ is matrix of rank r'.

 $\overline{\Psi} = US'V^{\dagger}, S' = \operatorname{diag}(s_1, \dots s_{r'}, 0 \dots)$ from above.

 \Longrightarrow Schmidt decomposition of approximate state

(19)
$$|\overline{\Psi}\rangle = \sum_{a=1}^{r'} s_a |a\rangle_A |a\rangle_B$$

cf. Eq. (27) of Schollwöck [9], where s_a must be rescaled if normalization desired.

12

20.1. **QR decomposition.** cf. 4.1.2. of Schollwöck [9]. If actual value of singular values not used explicitly, then use *QR decomposition*.

QR decomposition: $\forall M \in \operatorname{Mat}_{\mathbb{K}}(N_A, N_B)$,

(20) M = QR, $Q \in U_{\mathbb{K}}(N_A)$, i.e. $Q^{\dagger}Q = 1 = QQ^{\dagger}$, $R \in \operatorname{Mat}_{\mathbb{K}}(N_A, N_B)$ s.t. upper triangular, i.e. $R_{ij} = 0$ if i > j thin QR decomposition: assume $N_A > N_B$. Then bottom $N_A - N_B$ rows of R are 0, so

$$M = Q \begin{bmatrix} R_1 \\ 0 \end{bmatrix} = \begin{bmatrix} Q_1 & Q_2 \end{bmatrix} \begin{bmatrix} R_1 \\ 0 \end{bmatrix} = Q_1 R_1$$
$$Q_1 \in \operatorname{Mat}_{\mathbb{K}}(N_A, N_B)$$
$$R_1 \in \operatorname{Mat}_{\mathbb{K}}(N_B, N_B)$$

While $Q_1^{\dagger}Q_1 = 1$ in general $Q_1Q_1^{\dagger} \neq 1$

21. Matrix Product States (MPS)

cf. Section 4.13 Decomposition of arbitrary quantum states into MPS of Schollwöck [9]. Consider lattice of L sites, d-dim. local state spaces $\{\sigma_i\}_{i=1,...L}$. Most general pure quantum state on lattice (assume normalized)

(21)
$$|\psi\rangle = \sum_{\sigma_1 \dots \sigma_L} c_{\sigma_1 \dots \sigma_L} |\sigma_1 \dots \sigma_L\rangle$$

cf. Eq. (30) of Schollwöck [9]

21.1. Left-canonical matrix product state. cf. Schollwöck [9],

Consider the process of refactoring or "flattening", which I claim to be a functor flatten:

$$|\psi\rangle \in \mathcal{H} \text{ s.t. } \dim \mathcal{H} = d^L \mapsto \Psi \in \operatorname{Mat}_{\mathbb{K}}(d, d^{L-1})$$

$$\Psi_{\sigma_1,(\sigma_2...\sigma_L)} = c_{\sigma_1...\sigma_L}$$

(22)
$$\xrightarrow{\text{SVD}} c_{\sigma_1...\sigma_L} = \Psi_{\sigma_1,(\sigma_2...\sigma_L)} = \sum_{a}^{r_1} U_{\sigma_1,a_1} S_{a_1,a_1} (V^{\dagger})_{a_1,(\sigma_2...\sigma_L)} \equiv \sum_{a_1}^{r_1} U_{\sigma_1,a_1} c_{a_1,\sigma_2...\sigma_L}$$

i.e.

$$(\mathbb{K}^d)^L \to \operatorname{Mat}_{\mathbb{K}}(1, r) \times \operatorname{Mat}_{\mathbb{K}}(r_1 d, d^{L-2})$$
$$c_{\sigma_1 \dots \sigma_L} \mapsto A_{\sigma_1}^{\sigma_1}, \Psi_{(a_1 \sigma_2), (\sigma_3 \dots \sigma_L)}$$

s.t.

$$c_{\sigma_1...\sigma_L} = \sum_{a_1}^{r_1} A_{a_1}^{\sigma_1} \Psi_{(a_1\sigma_2),(\sigma_3...\sigma_L)}$$

where rank $r_1 < d$.

$$U \in \operatorname{Mat}_{\mathbb{K}}(d, \min(d, r)) = \operatorname{Mat}_{\mathbb{K}}(d, r)$$

Consider d row vectors A^{σ_1} , $A_{a_1}^{\sigma_1} = U_{\sigma_1,a_1}$.

$$c_{a_1\sigma_2...\sigma_L} = \sum_{a_1}^{r_1} A_{a_1}^{\sigma_1} \Psi_{(a_1,\sigma_2),(\sigma_3...\sigma_L)}$$
 with

$$\Psi_{(a_1\sigma_2),(\sigma_3...\sigma_L)} \in \operatorname{Mat}_{\mathbb{K}}(r_1d,d^{L-2})$$

So from Eq. (34) of Schollwöck [9],

(23)
$$c_{\sigma_1...\sigma_L} = \sum_{\sigma_1}^{r_1} \sum_{\sigma_2}^{r_2} A_{a_1}^{\sigma_1} U_{(a_1\sigma_2),a_2} S_{a_2,a_2}(V^{\dagger})_{a_2,(\sigma_3...\sigma_L)} = \sum_{\sigma_1}^{r_1} \sum_{\sigma_2}^{r_2} A_{a_1}^{\sigma_1} A_{a_1,a_2}^{\sigma_2} \Psi_{(a_2\sigma_3),(\sigma_4...\sigma_L)}$$

ERNEST YEUNG ERNESTYALUMNI@GMAIL.COM

So for

$$U \in \operatorname{Mat}_{\mathbb{K}}(d, r_1 \times r_2) \mapsto \{A^{\sigma_2}\}_{\sigma_2}, \qquad |\{A^{\sigma_2}\}_{\sigma_2}| = d, \qquad A^{\sigma_2} \in \operatorname{Mat}_{\mathbb{K}}(r_1, r_2)$$

 $A_{a_1,a_2}^{\sigma_2} = U_{(a_1,\sigma_2),a_2}$ and multiplied S and V^{\dagger} ,

$$SV^{\dagger} \mapsto \Psi \in \operatorname{Mat}_{\mathbb{K}}(r_2d, d^{L-3}); \qquad r_2 \le r_1d \le d^2$$

and so continuing the application of SVD and refactoring (what I call applying the flatten functor)

$$\xrightarrow{\text{SVD}} c_{\sigma_1 \dots \sigma_L} = \sum_{a_1 \dots a_{L-1}} A_{a_1}^{\sigma_1} A_{a_1 a_2}^{\sigma_2} \dots A_{a_{L-2}, a_{L-1}}^{\sigma_{L-1}} A_{a_L - 1}^{\sigma_L} \equiv A^{\sigma_1} A^{\sigma_2} \dots A^{\sigma_{L-1}} A^{\sigma_L}$$

21.1.1. Matrix Product State (definition)

Definition 18 (Matrix Product State).

$$|\psi\rangle = \sum_{\sigma_1...\sigma_L} A^{\sigma_1} A^{\sigma_2} \dots A^{\sigma_{L-1}} A^{\sigma_L} |\sigma_1 \dots \sigma_L\rangle$$

Maximally, the dims. are

$$(1 \times d), (d \times d^2) \dots (d^{L/2-1} \times d^{L/2}), (d^{L/2} \times d^{L/2-1}) \dots (d^2 \times d), (d \times 1)$$

Since \forall SVD, $U^{\dagger}U = 1$,

$$\delta_{a_{l},a'_{l}} = \sum_{a_{l-1}a_{l}} (U^{\dagger})_{a_{l},(a_{l-1}\sigma_{l})} U_{(a_{l-1}\sigma_{l}),a'_{l}} = \sum_{a_{l-1}\sigma_{l}} (A^{\sigma_{l}})^{\dagger}_{a_{l},a_{l-1}} A^{\sigma_{l}}_{a_{l-1},a'_{l}} = \sum_{\sigma_{l}} ((A^{\sigma_{2}})^{\dagger} A^{\sigma_{l}})_{a_{l},a'_{l}}$$

or

(25)
$$\sum_{\sigma_l} (A^{\sigma_l})^{\dagger} A^{\sigma_l} = 1$$

cf. Eq. (38) of Schollwöck [9],

If for $\{A^{\sigma_l}\}_{\sigma_l}$, $\sum_{\sigma_l} (A^{\sigma_l})^{\dagger} A = 1$, $\{A^{\sigma_l}\}_{\sigma_l}$ are **left-normalized**; matrix product states that consist of only left-normalized matrices are **left-canonical**.

View Density Matrix Renormalization Group (DMRG) decomposition of universe into blocks A and B, split lattice into parts A, B, where A comprises sites 1 through l and B sites l + 1 through L.

$$|a_l\rangle_A = \sum_{\sigma_1...\sigma_l} (A^{\sigma_1} A^{\sigma_2} ... A^{\sigma_l})_{a_l,1} |\sigma_1...\sigma_l\rangle$$

$$|a_l\rangle_B = \sum_{\sigma_{l+1}...\sigma_L} (A^{\sigma_{l+1}} A^{\sigma_{l+2}} ... A^{\sigma_L})_{a_l,1} |\sigma_{l+1}...\sigma_L\rangle$$

s.t. matrix product state (MPS) is

$$|\psi\rangle = \sum_{a_l} |a_l\rangle_A |a_l\rangle_B$$

21.1.2. Summarize this procedure of constructing, from a pure state, the matrix product state (version) by successive application Singular Value Decomposition (SVD) from the Category Theory point of view. Consider all applications of SVD to get to a matrix

$$(\mathbb{K}^d)^L \xrightarrow{\text{SVD}} (\text{Mat}_{\mathbb{K}}(1, r_1))^d \times (\text{Mat}_{\mathbb{K}}(r_1, r_2))^d \times \cdots \times (\text{Mat}_{\mathbb{K}}(r_{L-2}, r_{L-1}))^d \times (\text{Mat}_{\mathbb{K}}(r_{L-1}, 1))^d$$

$$c_{\sigma_1...\sigma_L} \longmapsto c_{\sigma_1...\sigma_L} = \sum_{a_1...a_{L-1}} A_{a_1}^{\sigma_1} A_{a_1 a_2}^{\sigma_2} \dots A_{a_{L-2},a_{L-1}}^{\sigma_{L-1}} A_{a_{L-1}}^{\sigma_L}$$

product state (MPS):

and remember the maximal values that the r_i 's can take:

$$r_1 \le d$$
 $r_{L/2} \le d^{L/2}$ $r_{L-2} \le d^2$ $r_{L-2} \le d^2$ $r_{L/2+1} \le d^{L/2-1}$ $r_{L-1} \le d$

Let us explicitly note the functors (that were applied) flatten (and its inverse), and the application of SVD, explicitly:

$$(\mathbb{K}^d)^L \xrightarrow{\text{flatten}^{-1}} \operatorname{Mat}_{\mathbb{K}}(d, d^{L-1}) \xrightarrow{\text{SVD}} U_{\mathbb{K}}(d, r_1) \times \operatorname{diag}_{\mathbb{K}}(r_1) \times U_{\mathbb{K}}(r_1, d^{L-1}) \xrightarrow{\cong} (\operatorname{Mat}_{\mathbb{K}}(1, r_1))^d \times \operatorname{Mat}_{\mathbb{K}}(r_1, d^{L-2}) \xrightarrow{\text{flatten}} (\operatorname{Mat}_{\mathbb{K}}(1, r_1))^d \times (\mathbb{K}^{r_1}) \times (\mathbb{K}^d)^{L-1}$$

$$c_{\sigma_1...\sigma_L} \xrightarrow{\text{flatten}^{-1}} c_{\sigma_1...\sigma_L} = \Psi_{\sigma_1,(\sigma_2...\sigma_L)} \xrightarrow{\text{SVD}} \Psi_{\sigma_1,(\sigma_2...\sigma_L)} = \sum_{a_1}^{r_1} U_{\sigma_1 a_1} S_{a_1,a_1}(V^{\dagger})_{a_1,(\sigma_2...\sigma_L)} \xrightarrow{\cong} c_{a_1 \sigma_2...\sigma_L} = \sum_{a_1}^{r_1} A_{a_1}^{\sigma_1} \Psi_{(a_1,a_2),(\sigma_3...\sigma_L)} \xrightarrow{\text{flatten}} c_{a_1 \sigma_2...\sigma_L} = \sum_{a_1}^{r_1} A_{a_1}^{\sigma_1} C_{a_1 \sigma_2...\sigma_L}$$

with \cong in this case denoting an isomorphism (clearly).

In considering some kind of recursive algorithm, so to repeat some series of steps until a matrix product state is obtained, consider this:

$$(\mathbb{K}^d)^L \longrightarrow (\mathrm{Mat}_{\mathbb{K}}(1, r_1))^d \times \mathbb{K}^{r_1} \times (\mathbb{K}^d)^{L-1}$$

$$c_{\sigma_1...\sigma_L} \longmapsto c_{\sigma_1...\sigma_L} = \sum_{a_1}^{r_1} A_{a_1}^{\sigma_1} c_{a_1\sigma_2...\sigma_L}$$

So in summary, to obtain matrix product states, starting from a matrix,

$$\operatorname{Mat}_{\mathbb{K}}(d,d^{L-1}) \longrightarrow (\operatorname{Mat}_{\mathbb{K}}(1,r_{1}))^{d} \times \operatorname{Mat}_{\mathbb{K}}(r_{1}d,d^{L-2}) \longrightarrow \cdots \longrightarrow (\operatorname{Mat}_{\mathbb{K}}(1,r_{1}))^{d} \times (\operatorname{Mat}_{\mathbb{K}}(r_{1},r_{2}))^{d} \times \cdots \times (\operatorname{Mat}_{\mathbb{K}}(r_{n-1},r_{n}))^{d} \times (\operatorname{Mat}_{\mathbb{K}}(r_{n}d,d^{L-(n+1)}))^{d}$$

$$\Psi_{\sigma_{1},(\sigma_{2}...\sigma_{L})} \longmapsto \sum_{a_{1}}^{r_{1}} A_{a_{1}}^{\sigma_{1}} \Psi_{(a_{1},\sigma_{2}),(\sigma_{3}...\sigma_{L})} \longmapsto \cdots \longmapsto \sum_{a_{1},a_{2},...a_{n}}^{r_{1}} A_{a_{1}}^{\sigma_{2}} A_{a_{1}a_{2}}^{\sigma_{2}} \dots A_{a_{n-1}a_{n}}^{\sigma_{n}} \Psi_{(a_{n}\sigma_{n+1}),(\sigma_{n+2}...\sigma_{L})}$$

(26)

21.2. Right-canonical matrix product state. cf. Schollwöck [9],

We can start from right in order to obtain

$$c_{\sigma_{1}...\sigma_{L}} = \Psi_{(\sigma_{1}...\sigma_{L-1}),\sigma_{L}} = \sum_{a_{L-1}} U_{(\sigma_{1}...\sigma_{L-1}),a_{L-1}} S_{a_{L-1},a_{L-1}} (V^{\dagger})_{a_{L-1},\sigma_{L}} = \sum_{a_{L-1}} \Psi_{(\sigma_{1}...\sigma_{L-2}),(\sigma_{L-1}a_{L-1})} B_{a_{L-1}}^{\sigma_{L}} = \sum_{a_{L-1},a_{L-2}} U_{(\sigma_{1}...\sigma_{L-2}),a_{L-2}} S_{a_{L-2},a_{L-2}} (V^{\dagger})_{a_{L-2},(\sigma_{L-1}a_{L-1})} B_{a_{L-1}}^{\sigma_{L}} = \sum_{a_{L-2},a_{L-1}} \Psi_{(\sigma_{1}...\sigma_{L-3}),(\sigma_{L-2}a_{L-2})} B_{a_{L-2},a_{L-1}}^{\sigma_{L-1}} B_{a_{L-1}}^{\sigma_{L}} = \dots$$

or consider

$$(\mathbb{K}^d)^L \xrightarrow{\text{flatten}^{-1}} \operatorname{Mat}_{\mathbb{K}}(d^{L-1}, d) \xrightarrow{\text{SVD}} U_{\mathbb{K}}(d^{L-1}, r_{L-1}) \times \operatorname{diag}_{\mathbb{K}}(r_{L-1}, d) \xrightarrow{\cong} \operatorname{Mat}_{\mathbb{K}}(d^{L-2}, dr_{L-1}) \times (\operatorname{Mat}_{\mathbb{K}}(r_{L-1}, 1))^d \xrightarrow{\text{SVD}} \operatorname{Mat}_{\mathbb{K}}(d^{L-2}, dr_{L-1}) \times \operatorname{Mat}_{\mathbb{K}}(d^{L-2},$$

$$c_{\sigma_{1}...\sigma_{L}} \models \underbrace{\text{flatten}^{-1}} \\ c_{\sigma_{1}...\sigma_{L}} \models \underbrace{\text{SVD}} \\ c_{\sigma_{1}...\sigma_{L}} = \underbrace{\sum_{a_{L-1}}^{r_{L-1}} U_{(\sigma_{1}...\sigma_{L-1}),a_{L-1}} S_{a_{L-1},a_{L-1}}} \\ c_{\sigma_{1}...\sigma_{L}} = \underbrace{\sum_{a_{L-1$$

$$\underline{\underline{SVD}} \longrightarrow U_{\mathbb{K}}(d^{L-2}, r_{L-2}) \times \operatorname{diag}_{\mathbb{K}}(r_{L-2}) \times U_{\mathbb{K}}(r_{L-2}, dr_{L-1}) \times (\operatorname{Mat}_{\mathbb{K}}(r_{L-1}, 1))^{d} \xrightarrow{\cong} \operatorname{Mat}_{\mathbb{K}}(d^{L-3}, dr_{L-2}) \times (\operatorname{Mat}_{\mathbb{K}}(r_{L-2}, r_{L-1}))^{d} \times (\operatorname{Mat}_{\mathbb{K}}(r_{L-1}, 1))^{d}$$

$$\begin{array}{c} VD \\ \longrightarrow \\ C\sigma_1...\sigma_L = \sum_{a_{L-1},a_{L-2}} U_{(\sigma_1...\sigma_{L-2}),a_{L-2}} S_{a_{L-2},a_{L-2}} & V^{\dagger})_{a_{L-2},(\sigma_{L-1}a_{L-1})} B^{\sigma_L}_{a_{L-1}} \\ \longrightarrow \\ C\sigma_1...\sigma_L = \sum_{a_{L-1},a_{L-2}} V_{(\sigma_1...\sigma_{L-2}),(\sigma_{L-2}a_{L-2})} B^{\sigma_{L-1}}_{a_{L-2},(\sigma_{L-1}a_{L-1})} B^{\sigma_L}_{a_{L-1}} \\ \longrightarrow \\ C\sigma_1...\sigma_L = \sum_{a_{L-1},a_{L-2}} \Psi_{(\sigma_1...\sigma_{L-3}),(\sigma_{L-2},a_{L-2})} B^{\sigma_{L-1}}_{a_{L-2},a_{L-1}} B^{\sigma_L}_{a_{L-1}} \\ \longrightarrow \\ C\sigma_1...\sigma_L = \sum_{a_{L-1},a_{L-2}} V_{(\sigma_1...\sigma_{L-3}),(\sigma_{L-2},a_{L-2})} B^{\sigma_{L-1}}_{a_{L-2},a_{L-1}} B^{\sigma_L}_{a_{L-1}} \\ \longrightarrow \\ C\sigma_1...\sigma_L = \sum_{a_{L-1},a_{L-2}} V_{(\sigma_1...\sigma_{L-3}),(\sigma_{L-2},a_{L-2})} B^{\sigma_L}_{a_{L-2},a_{L-1}} B^{\sigma_L}_{a_{L-1}} \\ \longrightarrow \\ C\sigma_1...\sigma_L = \sum_{a_{L-1},a_{L-2}} V_{(\sigma_1...\sigma_{L-3}),(\sigma_{L-2},a_{L-2})} B^{\sigma_L}_{a_{L-2},a_{L-1}} B^{\sigma_L}_{a_{L-1}} \\ \longrightarrow \\ C\sigma_1...\sigma_L = \sum_{a_{L-1},a_{L-2}} V_{(\sigma_1...\sigma_{L-3}),(\sigma_{L-2},a_{L-2})} B^{\sigma_L}_{a_{L-2},a_{L-1}} B^{\sigma_L}_{a_{L-1}} \\ \longrightarrow \\ C\sigma_1...\sigma_L = \sum_{a_{L-1},a_{L-2}} V_{(\sigma_1...\sigma_{L-3}),(\sigma_{L-2},a_{L-2})} B^{\sigma_L}_{a_{L-2},a_{L-1}} B^{\sigma_L}_{a_{L-2},a_{L-2}} \\ \longrightarrow \\ C\sigma_1...\sigma_L = \sum_{a_{L-1},a_{L-2}} V_{(\sigma_1...\sigma_{L-3}),(\sigma_{L-2},a_{L-2})} B^{\sigma_L}_{a_{L-2},a_{L-1}} B^{\sigma_L}_{a_{L-2},a_{L-2}} \\ \longrightarrow \\ C\sigma_1...\sigma_L = \sum_{a_{L-1},a_{L-2}} V_{(\sigma_1...\sigma_{L-3}),(\sigma_L,a_{L-2})} B^{\sigma_L}_{a_{L-2},a_{L-2}} B^{\sigma_L}_{a_{L-2},a_{L-2}} \\ \longrightarrow \\ C\sigma_1...\sigma_L = \sum_{a_{L-1},a_{L-2}} V_{(\sigma_1...\sigma_{L-3}),(\sigma_L,a_{L-2})} B^{\sigma_L}_{a_{L-2},a_{L-2}} \\ \longrightarrow \\ C\sigma_1...\sigma_L = \sum_{a_{L-1},a_{L-2}} V_{(\sigma_1...\sigma_{L-2}),(\sigma_L,a_{L-2})} B^{\sigma_L}_{a_{L-2},a_{L-2}} \\ \longrightarrow \\ C\sigma_1...\sigma_L = \sum_{a_{L-1},a_{L-2}} V_{(\sigma_1...\sigma_{L-2}),(\sigma_L,a_{L-2})}$$

with \cong in this case denoting an isomorphism (clearly).

And so we can explicitly state the recursion step, for the purpose of writing numerical implementations/algorithms: $\forall l = 1, 2 \dots L$,

$$\operatorname{Mat}_{\mathbb{K}}(d^{L-l}, dr_{L-(l-1)}) \longrightarrow \operatorname{Mat}_{\mathbb{K}}(d^{L-(l+1)}, dr_{L-l}) \times (\operatorname{Mat}_{\mathbb{K}}(r_{L-l}, r_{L-(l-1)}))^{d}$$

$$\Psi_{(\sigma_1...\sigma_{L-l}),(\sigma_{L-(l-1)}a_{L-(l-1)})} \longmapsto \Psi_{(\sigma_1...\sigma_{L-l}),(\sigma_{L-(l-1)}a_{L-(l-1)})} = \sum_{a_{L-l}} \Psi_{(\sigma_1...\sigma_{L-(l+1)}),(\sigma_{L-l}a_{L-l})} B_{a_{L-l},a_{L-(l-1)}}^{\sigma_{L-(l-1)}a_{L-(l-1)}}$$

and we finally obtained, after successive applications SVD, the matrix product state:

$$(\mathbb{K}^d)^L \longrightarrow \operatorname{Mat}_{\mathbb{K}}(d^{L-1}, d) \longrightarrow (\operatorname{Mat}_{\mathbb{K}}(1, r_1))^d \times (\operatorname{Mat}_{\mathbb{K}}(r_1, r_2))^d \times \cdots \times (\operatorname{Mat}_{\mathbb{K}}(r_{L-2}, r_{L-1}))^d \times (\operatorname{Mat}_{\mathbb{K}}(r_{L-1}, 1))^d$$

$$c_{\sigma_1...\sigma_L} \longmapsto \Psi_{(\sigma_1...\sigma_{L-l}),\sigma_L} \longmapsto c_{\sigma_1...\sigma_L} = \sum_{a_1...a_{L-1}} B_{a_1}^{\sigma_1} B_{a_1 a_2}^{\sigma_2} \dots B_{a_{L-2} a_{L-1}}^{\sigma_{L-1}} B_{a_{L-1}}^{\sigma_L}$$

Since

$$(27) V^{\dagger}V = 1$$

, then

(28)
$$\delta_{a_l a'_l} = \sum_{\sigma_m a_m} (V^{\dagger})_{a_l (\sigma_m a_m)} V_{(\sigma_m a_m) a'_l} = \sum_{\sigma_m a_m} B^{\sigma_m}_{a_l a_m} \overline{B}^{\sigma_m}_{a'_l a_m} \Longrightarrow \sum_{\sigma_m} \left[B^{\sigma_m} (B^{\sigma_m})^{\dagger} = 1 \right]$$

The *B*-matrices that obey this condition are referred to as **right-normalized** matrices. A matrix product state (MPS) entirely consisting of a product of these right-normalized matrices is called **right-canonical**.

21.2.1. Numerical implementation; both in BLAS and cuBLAS. As stated in the CUDA Toolkit Documentation v8.0 for cu-SOLVER, under section 5.3.6. cusolverDn<t>gesvd() and Remark 1, gesvd "only supports" m>=n, for matrix you want to decompose $A \in \operatorname{Mat}_{\mathbb{K}}(m,n)$. So number of rows must be greater than or equal to number of columns. And so we can only consider right-normalized matrices in a practical implementation.

I suspect it's the same in BLAS.

Consider the very first step, l=1, in a procedure to calculate the matrix product state.

$$\operatorname{Mat}_{\mathbb{K}}(d^{L-1}, d) \xrightarrow{\operatorname{SVD}} U_{\mathbb{K}}(d^{L-1}, r_{L-1}) \times \operatorname{diag}_{\mathbb{K}}(r_{L-1}) \times U_{\mathbb{K}}(r_{L-1}, d) \xrightarrow{\cong} \operatorname{Mat}_{\mathbb{K}}(d^{L-2}, dr_{L-1}) \times (\operatorname{Mat}_{\mathbb{K}}(r_{L-1}, 1))^{d}$$

$$\Psi_{(\sigma_{1}...\sigma_{L-1}),\sigma_{L}} \stackrel{\text{SVD}}{\longmapsto} = \sum_{a_{L-1}}^{r_{L-1}} U_{(\sigma_{1}...\sigma_{L-1}),a_{L-1}} S_{a_{L-1},a_{L-1}} (V^{\dagger})_{a_{L-1},\sigma_{L}} \stackrel{\cong}{\longmapsto} \frac{U_{(\sigma_{1}...\sigma_{L-1}),a_{L-1}} S_{a_{L-1},a_{L-1}} = \Psi_{(\sigma_{1}...\sigma_{L-2}),(\sigma_{L-1}a_{L-1})}}{(V^{\dagger})_{a_{L-1},\sigma_{L}}} = B_{a_{L-1}}^{\sigma_{L}}$$

with \cong in this case denoting an isomorphism, the *reshaping* of a matrix into different matrix size dimensions, which should be the inverse of a "flatten" functor, which I'll denote as flatten⁻¹ as well (and this is this same isomorphism we're talking about).

Let's deal with the specific procedure of flatten⁻¹, how it reshapes indices in accordance with different matrix size dimensions, and with the so-called "stride" when going from, say, 2-dimensional indices to a "flattened" 1-dimensional index.

Note also as a practical numerical implementation design point, LAPACK's linear algebra BLAS library package and CUBLAS assumes *column*-major ordering.

Consider i = 1, 2, ... L - 1 (for site i) (or for 0-based counting, starting to count from 0, i = 0, 1, ... L - 2; be aware of this difference as in practical numerical implementation, in C, C++, Python, it assumes 0-based counting).

For a state space of dimension d, we can consider the specific example of d=2, representing say a spin-1/2 system. Then index σ_i can be 0 or 1: $\sigma_i \in \{0,1\}$. In general, $\sigma_i \in \{0,1,\ldots d-1\}$. I may use d or 2 in the context of the number of states (basis vectors) of the spin system (state vector space).

Consider site i. Suppose the spin system there interacts most with sites i-1, i+1, and then next sites i-2, i+2, etc. So the values at $\sigma_{i-1}, \sigma_{i+1}$, etc. are most important in calculating interactions with spin system at site i.

Then we seek this reshaping of the matrix index - assuming 0-based counting/ordering, for l=1:

$$\{0,1\}^{L-1} \longrightarrow \{0,1,\dots 2^{L-1}-1\}$$

$$(\sigma_0, \sigma_1, \dots \sigma_{L-2}) \overset{\text{(flatten)}^{-1}}{\longmapsto} I_{L-1} := \sigma_0 + 2\sigma_1 + \dots + 2^i \sigma_i + \dots + 2^{L-2} \sigma_{L-2} = \sum_{i=0}^{L-2} 2^i \sigma_i$$

In this way, states of a site i are closest in memory addresses in the allocation of a 1-dim. array, on CPU or GPU memory, so that memory access operations should be efficient.

Assuming SVD doesn't change the striding, and defining the result of matrix multiplication:

$$U_{(\sigma_0,\sigma_1...\sigma_{L-2}),a_{L-1}}S_{a_{L-1},a_{L-1}} =: (US)_{(\sigma_0...\sigma_{L-2}),a_{L-1}} \in \operatorname{Mat}_{\mathbb{K}}(d^{L-1},r_{L-1})$$

We can reshape (i.e. $(flatten)^{-1}$) in such a manner:

$$\operatorname{Mat}_{\mathbb{K}}(d^{L-1}, r_{L-1}) \xrightarrow{\qquad \qquad } \operatorname{Mat}_{\mathbb{K}}(d^{L-2}, dr_{L-1})$$

$$(US)_{(\sigma_{0} \dots \sigma_{L-2}), a_{L-1}} \xrightarrow{\qquad \qquad } \Psi_{(\sigma_{0}, \sigma_{1}, \dots \sigma_{L-3}), (\sigma_{L-2} a_{L-1})} \xrightarrow{\qquad \qquad } \{0, 1, \dots 2^{L-1} - 1\} \times \{0, 1, \dots r_{L-1} - 1\} \xrightarrow{\qquad \qquad } \{0, 1, \dots 2^{L-2} - 1\} \times \{0, 1, \dots dr_{L-1} - 1\}$$

$$I_{L-1}, a_{L-1} \xrightarrow{\qquad \qquad } I_{L-1} \mod 2^{L-2}, \frac{I_{L-1}}{2^{L-2}} + da_{L-1}$$

Reshaping V^{\dagger} at iteration l=1 can be done as follows:

$$U_{\mathbb{K}}(r_{L-1}, d) \xrightarrow{\qquad \qquad } (\operatorname{flatten})^{-1} \longrightarrow (\operatorname{Mat}_{\mathbb{K}}(r_{L-1}, 1))^{d}$$

$$(V^{\dagger})_{a_{L-1}, \sigma_{L-1}} \longmapsto (\operatorname{flatten})^{-1} \longrightarrow (V^{\dagger})_{a_{L-1}, \sigma_{L-1}} = B_{a_{L-1}}^{\sigma_{L-1}}$$

$$\{0, 1, \dots, r_{L-1} - 1\} \times \{0, 1, \dots, d-1\} \xrightarrow{\text{(flatten})^{-1}} (\{0, 1, \dots, r_{L-1} - 1\})^{d}$$

$$a_{L-1}, \sigma_{L-1} \longmapsto (\operatorname{flatten})^{-1} \longrightarrow a_{L-1}$$

Let's do this same procedure, reshaping or (flatten) $^{-1}$, for a general l iteration.

$$\begin{split} \operatorname{Mat}_{\mathbb{K}}(d^{L-l}, r_{L-l}) & \xrightarrow{\qquad \qquad } \operatorname{Mat}_{\mathbb{K}}(d^{L-(l+1)}, dr_{L-l}) \\ & (US)_{(\sigma_0 \dots \sigma_{L-(l+1)}), a_{L-l}} & \xrightarrow{\qquad \qquad } \operatorname{Mat}_{\mathbb{K}}(d^{L-(l+1)}, dr_{L-l}) \\ & \{0, 1, \dots d^{L-l} - 1\} \times \{0, 1, \dots r_{L-l} - 1\} & \xrightarrow{\qquad \qquad } \{0, 1, \dots d^{L-(l+1)} - 1\} \times \{0, 1, \dots dr_{L-l} - 1\} \\ & I_{L-l}, a_{L-l} & \xrightarrow{\qquad \qquad } \operatorname{Ind}_{\mathbb{K}}(r_{l-l}, dr_{l-l}) & \xrightarrow{\qquad \qquad } \operatorname{Mat}_{\mathbb{K}}(r_{l-l}, r_{L-(l+1)}) + da_{L-l} \\ & U_{\mathbb{K}}(r_{L-l}, dr_{L-(l-1)}) & \xrightarrow{\qquad \qquad } \operatorname{Mat}_{\mathbb{K}}(r_{L-l}, r_{L-(l-1)}))^d \\ & (V^{\dagger})_{a_{L-l}, (\sigma_{L-l}a_{L-(l-1)})} & \xrightarrow{\qquad \qquad } (\operatorname{flatten})^{-1} & (V^{\dagger})_{a_{L-l}, (\sigma_{L-l}a_{L-(l-1)})} & = B_{a_{L-l}, a_{L-(l-1)}}^{\sigma_{L-l}} \\ & \{0, 1, \dots r_{L-l} - 1\} \times \{0, 1, \dots dr_{L-(l-1)} - 1\} & \xrightarrow{\qquad \qquad } (\operatorname{flatten})^{-1} & (\{0, 1, \dots r_{L-1} - 1\} \times \{0, 1, \dots r_{L-(l-1)} - 1\})^d \\ & a_{L-l}, (\sigma_{L-l}a_{L-(l-1)}) := a_{L-l}, \sigma_{L-1} + da_{L-(l-1)} & \xrightarrow{\qquad \qquad } (\operatorname{flatten})^{-1} \\ & a_{L-l}, (\sigma_{L-l}a_{L-(l-1)}) := a_{L-l}, \sigma_{L-1} + da_{L-(l-1)} & \xrightarrow{\qquad } (\operatorname{flatten})^{-1} \\ & a_{L-l}, (\sigma_{L-1}a_{L-(l-1)}) := a_{L-l}, \sigma_{L-1} + da_{L-(l-1)} & \xrightarrow{\qquad } (\operatorname{flatten})^{-1} \\ & a_{L-l}, (\sigma_{L-1}a_{L-(l-1)}) := a_{L-l}, \sigma_{L-1} + da_{L-(l-1)} & \xrightarrow{\qquad } (\operatorname{flatten})^{-1} \\ & a_{L-l}, (\sigma_{L-1}a_{L-(l-1)}) := a_{L-l}, \sigma_{L-1} + da_{L-(l-1)} & \xrightarrow{\qquad } (\operatorname{flatten})^{-1} \\ & a_{L-l}, (\sigma_{L-1}a_{L-(l-1)}) := a_{L-l}, \sigma_{L-1} + da_{L-(l-1)} & \xrightarrow{\qquad } (\operatorname{flatten})^{-1} \\ & a_{L-l}, (\sigma_{L-1}a_{L-(l-1)}) := a_{L-l}, \sigma_{L-1} + da_{L-(l-1)} & \xrightarrow{\qquad } (\operatorname{flatten})^{-1} \\ & a_{L-l}, (\sigma_{L-1}a_{L-(l-1)}) := a_{L-l}, \sigma_{L-1} + da_{L-(l-1)} & \xrightarrow{\qquad } (\operatorname{flatten})^{-1} \\ & a_{L-l}, (\sigma_{L-1}a_{L-(l-1)}) := a_{L-l}, \sigma_{L-1} + da_{L-(l-1)} & \xrightarrow{\qquad } (\operatorname{flatten})^{-1} \\ & a_{L-l}, (\sigma_{L-1}a_{L-(l-1)}) := a_{L-l}, \sigma_{L-1} + da_{L-(l-1)} & \xrightarrow{\qquad } (\operatorname{flatten})^{-1} \\ & a_{L-l}, (\sigma_{L-1}a_{L-(l-1)}) := a_{L-l}, \sigma_{L-1} + da_{L-(l-1)} & \xrightarrow{\qquad } (\operatorname{flatten})^{-1} \\ & a_{L-l}, (\sigma_{L-1}a_{L-(l-1)}) := a_{L-l}, \sigma_{L-1} + da_{L-(l-1)} & \xrightarrow{\qquad } (\operatorname{flatten})^{-1} \\ & a_{L-l}, (\sigma_{L-1}a_{L-(l-1)}) := a_{L-l}, \sigma_{L-1} + da_{L-(l-1)} & \xrightarrow{\qquad } (\operatorname{flatten})^{-1} \\ & a_{L-1}, (\sigma_{L-1}a_{L-(l-$$

References

- [1] David A. Cox. John Little. Donal O'Shea. Using Algebraic Geometry. Second Edition. Springer. 2005. ISBN 0-387-20706-6 QA564.C6883 2004
- [2] David Cox, John Little, Donal O'Shea. Ideals, Varieties, and Algorithms: An Introduction to Computational Algebraic Geometry and Commutative Algebra, Fourth Edition, Springer
- [3] Glen E. Bredon. Topology and Geometry. Graduate Texts in Mathematics (Book 139). Springer; Corrected edition (October 17, 1997). ISBN-13: 978-0387979267
- [4] Jean-Pierre Serre (Author), J. Stilwell (Translator). Trees (Springer Monographs in Mathematics) 1st ed. 1980. Corr. 2nd printing 2002 Edition. ISBN-13: 978-3540442370
- [5] Joseph J. Rotman, Advanced Modern Algebra (Graduate Studies in Mathematics) 2nd Edition, American Mathematical Society; 2 edition (August 10, 2010), ISBN-13: 978-0821847411
- [6] Jeffrey M. Lee. Manifolds and Differential Geometry, Graduate Studies in Mathematics Volume: 107, American Mathematical Society, 2009. ISBN-13: 978-0-8218-4815-9
- [7] Jacob C. Bridgeman and Christopher T. Chubb. Hand-waving and Interpretive Dance: An Introductory Course on Tensor Networks: Lecture Notes. arXiv:1603.03039 [quant-ph]
- [8] Ulrich Schollwoeck. The density-matrix renormalization group. Rev. Mod. Phys. 77, 259 (2005) arXiv:cond-mat/0409292 [cond-mat.str-el]
- [9] Ulrich Schollwoeck. The density-matrix renormalization group in the age of matrix product states. Annals of Physics 326, 96 (2011). arXiv:1008.3477 [cond-mat.str-el]
- [10] José L.F. Barbón and Eliezer Rabinovici. "Holographic Complexity And Spacetime Singularities." arXiv:1509.09291 [hep-th]
- [11] Juan Maldacena, Leonard Susskind. "Cool horizons for entangled black holes." arXiv:1306.0533 [hep-th]
- [12] Slava Rychkov. "EPFL Lectures on Conformal Field Theory in $D \ge 3$ Dimensions." arXiv:1601.05000 [hep-th]
- [13] G. Evenbly, G. Vidal. "Tensor network states and geometry." arXiv:1106.1082 [quant-ph]
- [14] Mark Van Raamsdonk. "Lectures on Gravity and Entanglement." arXiv:1609.00026 [hep-th]