

Introduction to Quantum Programming and Semantics

Lecture 6: Tensor networks

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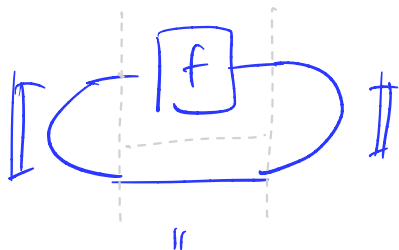
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Overview

- Cups and caps give ability to feedback
- Bases: linear maps vs matrices
- Sums of diagrams
- Tensor networks

Trace and dimension

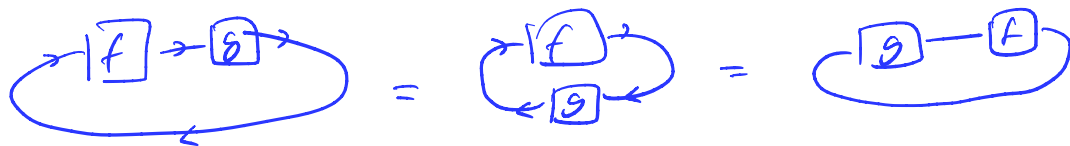
Feedback



is called the ^(total) trace of f

$$a+d = (1001) \begin{pmatrix} a & b & 0 & 0 \\ c & d & 0 & 0 \\ 0 & 0 & a & b \\ 0 & 0 & c & d \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \end{pmatrix} = \llbracket \text{ } \rrbracket \circ (f \otimes \text{id}) \circ \llbracket C \rrbracket$$

if $f: \mathbb{C}^2 \rightarrow \mathbb{C}^2$ then $\text{tr}(f) \in \mathbb{C}$
 is the linear-algebraic trace of f
 i.e. $f = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ then $\text{tr}(f) = a+d$



$$\begin{array}{c} A \\ \hline \boxed{f} \\ \hline B \\ \hline C \end{array} = \text{tr}_C(f): A \rightarrow B \quad \text{is the } \underline{\text{partial}} \text{ trace of } f$$

Dimensions

$$\text{Diagram: A rectangle labeled } A \text{ inside an oval labeled } A = \text{dimension of } A$$

$$\dim(A \otimes B) = \text{Diagram: An oval labeled } A \text{ containing a rectangle labeled } B = \bigcirc^A \bigcirc^B = \dim(A) \cdot \dim(B)$$

$$\dim(A \oplus B) = \dim(A) + \dim(B)$$

$$\dim(H) = \infty \Rightarrow \dim(H) + 1 = \dim(H) + \dim(\mathbb{C}) = \dim(H \oplus \mathbb{C}) = \dim(H) \quad \hookrightarrow$$

$$\Leftrightarrow H \simeq H \oplus \mathbb{C}$$

Bases

Orthonormal bases

linear map

$$f: \mathbb{C}^4 \rightarrow \mathbb{C}^3$$

need to
choose bases
for $\mathbb{C}^4$ and $\mathbb{C}^3$



3-by-4 matrix

$$\begin{pmatrix} \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots \end{pmatrix}$$

$$\langle e_i | e_j \rangle = 0 \quad \text{if } i \neq j$$

$$\langle e_i | e_i \rangle = 1$$

Sums of diagrams

Superposition

$$\sum_i \rightarrow \text{D} \leftarrow \text{D} \leftarrow = \text{---}$$

$$\sum_i \left(\begin{array}{ccc} \circ & \cdots & \circ \\ & \vdots & \\ \circ & \cdots & \circ \end{array} \right) \leftarrow i = \left(\begin{array}{ccc} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \end{array} \right)$$

↑
i

Tensor contraction

Tensor networks

$$\begin{aligned}
 & \text{Diagram 1: } h_{ij} \text{ (with } i, j \text{ as free indices)} \text{ is contracted with } f_{xz} \text{ (with } x \text{ as a summation index and } z \text{ as a free index). The result is } h_{ij} f_{xz} \text{ (with } i, j, z \text{ as free indices).} \\
 & = \sum_x h_{ij} f_{xz} \text{ (with } i, j, z \text{ as free indices).} \\
 & = \dots \\
 & = \sum_{k, y, z} h_{ij}^{x, l} g_y^z f_{zx}^{y, k}
 \end{aligned}$$

Special cases

tensor product: $(f \otimes g)_{ij}^{kl} = f_i^k g_j^l$

matrix multiplication: $(g \circ f)_{ij} = \sum_k f_i^k g_k^j$

trace $\text{tr}(f) = \sum_i f_i^i$

Summary:

- Graphical calculus automatically incorporates traces and dimensions
- It pays to be precise about choice of basis
- In computation it is handy to explicitly assume superposition
- Tensor networks are special case of graphical rewriting