Tensor Clusters: The hard case

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Applications

Data

Results of measurement & computation, e.g.

Mercury is 2.5cm high in tube.

Information

data used to make a decision, e.g.

Thermometer reads 38° F; I'll wear a coat.

The Data Problem

Turn data into information.

Nutrition Matrix

Nutrients×Produce data collected and used on diet problems.

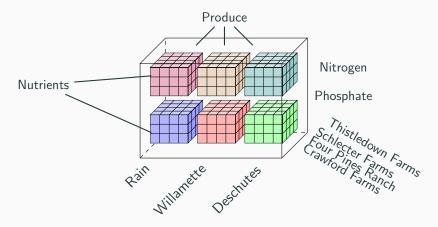
	Apple	Beef	Egg	Filbert	 Strawberry
Carbs	13.8	1	2	4.8	 7.7
Protein	0.3	20.7	17.1	3.9	0.7
Fat	0.2	7.4	14.4	1.3	0.3
:	:				

Substitute other examples $minerals \times mines$, pollutants $\times water$ -source, keywords $\times authors$

Nutrition Tensor

Reality is more complex...

Nutrients \times Produce \times Farms \times Water Source \times Fertilizer



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First Approximation of Tensor

Fox coefficients K and set $[n] = \{1, \dots, n\}$.

A hypermatrix (some call it a "tensor") is a function $\Gamma: [d_1] \times \cdots \times [d_\ell] \to K$.

$$K^{d_1 \times \cdots \times d_\ell} = \{\Gamma : [d_1] \times \cdots \times [d_\ell] \to K\}.$$

Technical matter: For general sets X_1, \ldots, X_ℓ $\Gamma: X_1 \times \cdots \times X_\ell \to K$ has *finite support* (only finitely many nonzero values).

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Data collection chooses bases for convenience/practicality/instrumentation/safety/laws/...

- Measure each fruit, farm, fertilizer separately
- Choose nutrients a lab can measure (Carbs, Sugar, Vit. A, Vit. B...)

Data collection is basis dependent; Likely some qualities of nutrition are basis independent.

The Tensor-Data ProblemFind basis invariant properties of tensor data.
(This is now a math problem!)

Second Approximation of Tensor

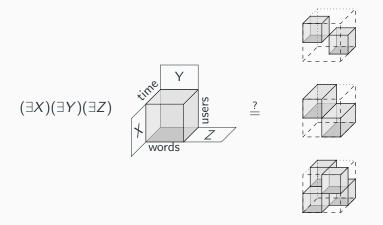
A function $\langle \Gamma | : K^{X_1} \times \cdots \times K^{X_\ell} \rightarrow K$ where

$$\langle \Gamma | u_1, \ldots, u_\ell \rangle = \sum_{x_1 \in X_1} \cdots \sum_{x_\ell \in X_\ell} \Gamma_{x_1 \cdots x_\ell} u_{1x_1} \cdots u_{\ell x_\ell}.$$

→? Just reminds me to tell you this is same "multi-linear"

$$\langle t|u_a + \lambda \tilde{u}_a, u_{\bar{a}} \rangle = \langle t|u_a, u_{\bar{a}} \rangle + \lambda \langle t|\tilde{u}_a, u_{\bar{a}} \rangle$$

Example: Find change of basis matrices to get clusters



Then ask: What type? Are they unique? How to find them?

The Tensor Product

Main Idea

 $\mathsf{Data}\ \mathsf{table} \equiv \mathsf{Multiplication}\ \mathsf{Table}$

So study multiplication tables.

For finite sets

$$\boxtimes : R^m \times R^n \rightarrowtail R^{m \times n} \quad u \boxtimes v = uv^{\dagger} = \begin{bmatrix} u_1v_1 & \cdots & u_1v_n \\ \vdots & & \vdots \\ u_mv_1 & \cdots & u_mv_n \end{bmatrix}.$$

In general possibly infinite matrices, but finite support

$$\boxtimes : R^X \times R^Y \rightarrow R^{X \times Y} \qquad (u \boxtimes v)_{xy} = u_x v_y.$$

The Tensor Product

A right *R*-modules has a presentation

$$U_R = \operatorname{Pres}_R \langle X \mid S \rangle = R^X / \operatorname{Span} S$$

A left R-module has a presentation

$$_RV = \operatorname{Pres}_R\langle Y \mid T \rangle = R^Y / \operatorname{Span} T$$

Their **Tensor Product** is

$$U \otimes_R V = R^{X \times Y} / (R^X \boxtimes T + S \boxtimes R^Y)$$

along with the induced function

$$\otimes: U \times V \rightarrowtail U \otimes_R V$$
$$x \otimes y = x \boxtimes y + (R^X \boxtimes T + S \boxtimes R^Y).$$

$$U_R = \operatorname{Pres}_R \langle X \mid S \rangle = R^X / \operatorname{Span} S;$$

 $_R V = \operatorname{Pres}_R \langle Y \mid T \rangle = R^Y / \operatorname{Span} T$

Example

$$(\mathbb{Z}/2\mathbb{Z} \oplus \mathbb{Z}/6\mathbb{Z} \oplus \mathbb{Z}) \otimes (\mathbb{Z}/4\mathbb{Z} \oplus \mathbb{Z}/8\mathbb{Z})$$

$$\begin{bmatrix} \mathbb{Z} & \mathbb{Z} \\ \mathbb{Z} & \mathbb{Z} \\ \mathbb{Z} & \mathbb{Z} \end{bmatrix} / \begin{pmatrix} \begin{bmatrix} \mathbb{Z} \\ \mathbb{Z} \\ \mathbb{Z} \end{bmatrix} \begin{bmatrix} 4\mathbb{Z} & 8\mathbb{Z} \end{bmatrix} + \begin{bmatrix} 2\mathbb{Z} \\ 6\mathbb{Z} \\ 0\mathbb{Z} \end{bmatrix} \begin{bmatrix} \mathbb{Z} & \mathbb{Z} \end{bmatrix} \\ = \begin{bmatrix} \mathbb{Z} & \mathbb{Z} \\ \mathbb{Z} & \mathbb{Z} \end{bmatrix} / \begin{bmatrix} 4\mathbb{Z} + 2\mathbb{Z} & 8\mathbb{Z} + 2\mathbb{Z} \\ 4\mathbb{Z} + 6\mathbb{Z} & 8\mathbb{Z} + 6\mathbb{Z} \\ 4\mathbb{Z} + 0\mathbb{Z} & 8\mathbb{Z} + 0\mathbb{Z} \end{bmatrix} \\ = \begin{bmatrix} \mathbb{Z}/2\mathbb{Z} & \mathbb{Z}/2\mathbb{Z} \\ \mathbb{Z}/2\mathbb{Z} & \mathbb{Z}/2\mathbb{Z} \\ \mathbb{Z}/4\mathbb{Z} & \mathbb{Z}/8\mathbb{Z} \end{bmatrix}$$

$$\mathbb{Q} = \left\langle \frac{1}{1!}, \frac{1}{2!}, \frac{1}{3!}, \ldots \right\rangle = \left\langle e_1, e_2, \ldots \mid e_n = (n+1)e_{n+1} \right\rangle$$

$$\mathbb{Z}^{[1]\times\mathbb{N}}\Bigg/\Bigg(2\mathbb{Z}\boxtimes\mathbb{Z}^\mathbb{N}+\mathbb{Z}\boxtimes\begin{bmatrix}1 & -2 & & \\ & 1 & -3 & \\ & & \ddots & \ddots\end{bmatrix}\Bigg)$$

$$\begin{bmatrix} 1 & -2 \\ & 1 & -3 \\ & & \ddots & \ddots \\ 2 & 0 & & & \\ & 2 & 0 & & & \\ & & \ddots & \ddots & & \\ \end{bmatrix}$$

$$\mathbb{Q} = \left\langle \frac{1}{1!}, \frac{1}{2!}, \frac{1}{3!}, \dots \right\rangle = \left\langle e_1, e_2, \dots \mid e_n = (n+1)e_{n+1} \right\rangle$$

$$\mathbb{Z}^{[1]\times\mathbb{N}}\Bigg/\Bigg(2\mathbb{Z}\boxtimes\mathbb{Z}^\mathbb{N}+\mathbb{Z}\boxtimes\begin{bmatrix}1&-2&&\\&1&-3&\\&&\ddots&\ddots&\end{bmatrix}\Bigg)$$

$$\begin{bmatrix} 1 & -2 & & & \\ & 1 & -3 & & \\ & & \ddots & \ddots \\ 2 & 0 & & & \\ & 2 & 0 & & \\ & & \ddots & \ddots \end{bmatrix} \sim \begin{bmatrix} 1 & -2 & & \\ & 1 & -3 & & \\ & & \ddots & \ddots \\ 0 & 4 & & \\ & 2 & 0 & & \\ & & \ddots & \ddots \end{bmatrix}$$

$$\mathbb{Q} = \left\langle \frac{1}{1!}, \frac{1}{2!}, \frac{1}{3!}, \ldots \right\rangle = \left\langle e_1, e_2, \ldots \mid e_n = (n+1)e_{n+1} \right\rangle$$

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$$\begin{bmatrix} 1 & -2 & & & \\ & 1 & -3 & & \\ & & \ddots & \ddots \\ 2 & 0 & & & \\ & 2 & 0 & & \\ & & \ddots & \ddots \end{bmatrix} \sim \begin{bmatrix} 1 & -2 & & \\ & 1 & -3 & & \\ & & \ddots & \ddots \\ 0 & 0 & & \\ & & 0 & 6 & \\ & & & \ddots & \ddots \end{bmatrix}$$

$$\mathbb{Q} = \left\langle \frac{1}{1!}, \frac{1}{2!}, \frac{1}{3!}, \dots \right\rangle = \left\langle e_1, e_2, \dots \mid e_n = (n+1)e_{n+1} \right\rangle$$

$$\mathbb{Z}^{[1]\times\mathbb{N}}\Bigg/\Bigg(2\mathbb{Z}\boxtimes\mathbb{Z}^\mathbb{N}+\mathbb{Z}\boxtimes\begin{bmatrix}1&-2&&\\&1&-3&\\&&\ddots&\ddots&\end{bmatrix}\Bigg)$$

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$$\mathbb{Z}/2\mathbb{Z}\otimes\mathbb{Q}$$

Solution
$$\mathbb{Q} = \left\langle \frac{1}{1!}, \frac{1}{2!}, \frac{1}{3!}, \ldots \right\rangle = \left\langle e_1, e_2, \ldots \mid e_n = (n+1)e_{n+1} \right\rangle$$

$$\mathbb{Z}^{[1] \times \mathbb{N}} / \left(2\mathbb{Z} \boxtimes \mathbb{Z}^{\mathbb{N}} + \mathbb{Z} \boxtimes \begin{bmatrix} 1 & -2 & \\ & 1 & -3 & \\ & & \ddots & \ddots \end{bmatrix} \right)$$

$$\mathbb{Z}^{1 imes\mathbb{N}}igg/\mathbb{Z}oxtimesegin{bmatrix} 1 & -2 & & & \ & 1 & -3 & & \ & & \ddots & \ddots \end{bmatrix}\cong\mathbb{Q}$$

Theory

Tensor products are distributive

$$(u + \tilde{u}) \otimes v = u \otimes v + \tilde{u} \otimes v$$

$$u \otimes (v + \tilde{v}) = u \otimes v + u \otimes \tilde{v}$$

$$(U \oplus \tilde{U}) \otimes_{R} V = U \otimes_{R} V \oplus \tilde{U} \otimes_{R} V$$

$$U \otimes_{R} (V \oplus \tilde{V}) = U \otimes_{R} V \oplus U \otimes \tilde{V}$$

Theory

Tensor products have a 1.

$$U \otimes_R R \cong U$$

$$R \otimes_R V \cong V$$

Theory

Sometimes it is said tensor products are associative and commutative, that is nonsense

$$\mathbb{R}^{\otimes}$$

But in special cases, e.g. if R is commutative

$$U \otimes_R (V \otimes_R W) \cong (U \otimes_R V) \otimes_R W$$

 $U \otimes_R V \cong V \otimes_R U.$

But be warned $U_1 \otimes \cdots \otimes U_n$ is generally not defined!

Universal Mapping Property

Further fact $ur \boxtimes v = (ur)v^{\dagger} = u(rv)^{\dagger} = u \boxtimes rv$; so $ur \otimes v = u \otimes rv$.

Theorem

If $*: U \times V \rightarrowtail W$ is distributive and $\forall r \in R$, ur * v = u * rv then $\exists ! \widehat{*} : U \otimes_R V \to W$ where

$$u*v=\widehat{*}(u\otimes v).$$

In particular $U \otimes_R V$ does not depend on choice of presentations.

Whitney Tensor Product

Every module has its "regular" presentation

$$U_R = \operatorname{Pres}_R \langle e_u, u \in U \mid e_{u+\tilde{u}} = e_u + e_{\tilde{u}}, e_{ur} = e_u r \rangle$$
 $R_R V = \operatorname{Pres}_R \langle e_v, v \in V \mid e_{v+\tilde{v}} = e_v + e_{\tilde{v}}, e_{rv} = r e_v \rangle$

With these presentations we recover Whitney's definition (i.e. your textbook definition)

$$U \otimes_{R} V = R^{U \times V} / \left\langle \begin{array}{c} e_{u+\tilde{u}} \otimes e_{v} = e_{u} \otimes e_{v} + e_{\tilde{u}} \otimes e_{v}, \\ e_{u} \otimes e_{v+\tilde{v}} = e_{u} \otimes e_{v} + e_{u} \otimes e_{\tilde{v}}, \\ e_{ur} \otimes e_{v} = e_{u} \otimes e_{rv} \end{array} \right\rangle$$

So established methods are a special case.

Similar ideas picking up on matrix-vector product $R^{m \times n} \times R^n \longrightarrow R^m$ give rise to

$$U \oslash V \times V \rightarrow U$$

Which behave like fractions, e.g.

$$A \oslash K \cong A$$
 $A \to K \oslash (K \oslash A)$

$$A \oslash (B \otimes C) = A \oslash B \oslash C$$

I.e.

$$\mathsf{hom}(C \otimes B, A) \cong \mathsf{hom}(C, \mathsf{hom}(B, A))$$

Solving for Coordinates

 $\begin{array}{l} \textbf{Applications} \\ \text{We } \textit{have} \text{ a tensor.} & \text{Why would we want a tensor product?} \end{array}$

What if
$$R = R_1 \oplus R_2$$
?
$$(R_1 \oplus R_2)^X \longrightarrow U_R = \operatorname{Span} X$$

$$\times \qquad \qquad \times$$

$$(R_1 \oplus R_2)^Y \longrightarrow {}_RV = \operatorname{Span} Y$$

$$\downarrow \boxtimes \qquad \qquad \downarrow \otimes$$

$$(R_1 \oplus R_2)^{X \times Y} \longrightarrow U \otimes_{R_1 \oplus R_2} V$$

What if $R = R_1 \oplus R_2$?

$$R_1^X \oplus R_2^X \longrightarrow U_1 \oplus U_2$$

$$\times \qquad \qquad \times$$

$$R_1^Y \oplus R_2^Y \longrightarrow V_1 \oplus V_2$$

$$\downarrow \boxtimes \qquad \qquad \qquad \downarrow \otimes$$

$$R_1^{X \times Y} \oplus R_2^{X \times Y} \longrightarrow (U_1 \otimes_{R_1} V_1) \oplus (U_2 \otimes_{R_2} V_2)$$

hat's one of these:

That's one of these: \(\sum_{--1}\sum_{-1}\) ... for which people are hunting!

Solve for **universal** *R*!

$$(?)^{X} \longrightarrow U_{?} = \operatorname{Span} X = U$$

$$\times \times \times \times$$

$$(?)^{Y} \longrightarrow P \cdot V = \operatorname{Span} Y = V$$

$$\downarrow \boxtimes \qquad \qquad \downarrow \otimes \qquad \qquad \downarrow *$$

$$(?)^{X \times Y} \longrightarrow U \otimes_{?} V \longrightarrow W$$

Solution is the Centroid

$$C(*) = \{(X, Y, Z) \mid Xu * v = Z(u * v) = u * (Yv)\}$$

Cluster Algorithm (W. 2008)

$$C(*) = \{(X,Y,Z) \mid Xu*v = Z(u*v) = u*(Yv)\}$$
 Random $\alpha \in C(*)$, if $\min_{\alpha}(x) = a(x)b(x)$ with $\deg a(x), \deg b(x) > 0$ and
$$1 = \operatorname{GCD}(a(x),b(x)) = s(x)a(x) + t(x)b(x)$$

Then
$$e = s(\alpha)a(\alpha)$$
; $f = t(\alpha)b(\alpha)$
Return $U = eU \oplus fU$, $V = eV \oplus fV$, $W = eW \oplus fW$.
Else try another α .

Theorem W. The resulting clusters are unique.

Adjoints

Universal Property

Theorem Brooksbank-Wilson.

The **adjoint algebra** is the largest set of operators over which we can for tensor product.

Proof of Universality of Adjoints

Proof. Suppose
$$\Omega \to \operatorname{End}(U) \times \operatorname{End}(V)^{op}$$