# The geometry of rank-one tensor completion 

Thomas Kahle<br>Otto-von-Guericke Universität Magdeburg

joint work with Kaie Kubjas, Mario Kummer and Zvi Rosen

Given some of the entries of a tensor, does there exist a rank-one tensor that has the specified entries?

## Question

Given some of the entries of a tensor, does there exist a rank-one tensor that has the specified entries?

## Example: matrix completion*

Assume given are the diagonal entries of a matrix:

$$
X=\left(\begin{array}{cc}
x_{11} & ? \\
? & x_{22}
\end{array}\right)
$$

Do there exist values for the ? so that the matrix has rank-one?

## Question

Given some of the entries of a tensor, does there exist a rank-one tensor that has the specified entries?

## Example: matrix completion*

Assume given are the diagonal entries of a matrix:

$$
X=\left(\begin{array}{cc}
x_{11} & ? \\
? & x_{22}
\end{array}\right)
$$

Do there exist values for the ? so that the matrix has rank-one?

Yes!
Pick one value arbitrary, then $x_{11} x_{22}=x_{12} x_{21}$ fixes the fourth.
*see also Kubjas/Rosen
Matrix completion for the independence model.

## Answer

This can be answered with computational algebra:

- Rank-one tensors are a binomial algebraic set (Segre variety)
- Use elimination to understand its projections.


## Answer

This can be answered with computational algebra:

- Rank-one tensors are a binomial algebraic set (Segre variety)
- Use elimination to understand its projections.

Problem Solved! Except...
 of limineneral form of the main theorem of elimina we shall prove

### 14.1 Elimination Theory

The following classical result is enormously useful.
Theorem 14.1 (Main Theorem of Elimination Theory). If $X$ is any variety over an algebraically closed field $k$, and $Y$ is a Zariski closed subset of $X \times \mathbf{P}^{n}$, then the image of $Y$ under projection to $X$ is closed.

We want $k=\mathbb{R}$ and have inequalities too (probabilities are non-negative!).

## Example: joint probability distributions

Given the diagonal entries of a two-variate binary distribution

$$
X=\left(\begin{array}{cc}
x_{11} & ? \\
? & x_{22}
\end{array}\right) \quad \text { where } x_{i j}=\operatorname{Prob}\left(X_{1}=i, X_{2}=j\right)
$$

Do there exist values for the ? so that $X$ is the distribution of two independent binary random variables?

## Example: joint probability distributions

Given the diagonal entries of a two-variate binary distribution

$$
X=\left(\begin{array}{cc}
x_{11} & ? \\
? & x_{22}
\end{array}\right) \quad \text { where } x_{i j}=\operatorname{Prob}\left(X_{1}=i, X_{2}=j\right)
$$

Do there exist values for the ? so that $X$ is the distribution of two independent binary random variables?

Do there exist marginal distributions $p, q \in \Delta_{1}$ such that $x_{i j}=p_{i} q_{j}$ ? ( $\Delta_{m}=m$-diml. simplex)

## The geometric view

## Problem

Determine the image of the restricted parametrization map

$$
\begin{aligned}
\Delta_{1} \times \Delta_{1} & \rightarrow[0,1]^{2} \\
\left(p_{1}, p_{2}, q_{1}, q_{2}\right) & \mapsto\left(x_{11}, x_{22}\right)=\left(p_{1} q_{1}, p_{2} q_{2}\right) .
\end{aligned}
$$

Restrictions on the domain $\left(p, q \in \Delta_{1}\right)$ make this problem interesting.

$$
\left(p_{1}, p_{2}, q_{1}, q_{2}\right) \mapsto\left(x_{11}, x_{22}\right)=\left(p_{1} q_{1}, p_{2} q_{2}\right)
$$

subject to

$$
p_{2}=1-p_{1}, \quad q_{2}=1-q_{1}
$$



$$
\left(x_{11}-x_{22}\right)^{2}-2\left(x_{11}+x_{22}\right)+1 \geq 0
$$

$$
\begin{aligned}
\left(p_{1}, p_{2}, q_{1}, q_{2}\right) \mapsto & \left(x_{11}, x_{22}\right)=\left(p_{1} q_{1}, p_{2} q_{2}\right) \\
& \text { subject to }
\end{aligned}
$$

$$
p_{2}=1-p_{1}, \quad q_{2}=1-q_{1}, \quad p_{1} \geq 0, \quad q_{1} \geq 0
$$



$$
\begin{gathered}
\left(x_{11}-x_{22}\right)^{2}-2\left(x_{11}+x_{22}\right)+1 \geq 0 \\
x_{11} \geq 0, \quad 1+x_{11} \geq x_{22}
\end{gathered}
$$

$$
\begin{aligned}
&\left(p_{1}, p_{2}, q_{1}, q_{2}\right) \mapsto\left(x_{11}, x_{22}\right)=\left(p_{1} q_{1}, p_{2} q_{2}\right) . \\
& \text { subject to } \\
& p_{2}=1-p_{1} \geq 0, \quad q_{2}=1-q_{1} \geq 0, \quad p_{1} \geq 0, \quad q_{1} \geq 0
\end{aligned}
$$



$$
\begin{gathered}
\left(x_{11}-x_{22}\right)^{2}-2\left(x_{11}+x_{22}\right)+1 \geq 0 \\
1 \geq x_{11} \geq 0, \quad 1 \geq x_{22} \geq 0
\end{gathered}
$$

## Tensors

A tensor is an array of numbers from a field $\mathbb{F}$ indexed by

$$
D=\left[d_{1}\right] \times \cdots \times\left[d_{n}\right]
$$

where $d_{i} \geq 2$ are fixed integers and $[d]=\{1, \ldots, d\}$.

A partial tensor is an array of numbers from $\mathbb{F}$ indexed by a subset $E \subseteq D$. A completion of a partial tensor $S \in \mathbb{F}^{E}$ is a tensor $T \in \mathbb{F}^{D}$ such that the restriction $T_{\mid E}$ agrees with $S$.

## Parametrization of rank one tensors

Let $\mathbb{F}$ be one of the two fields $\mathbb{R}$ or $\mathbb{C}$. The set of rank-one tensors is the image of the parametrization

$$
\mathbb{F}^{d_{1}} \times \cdots \times \mathbb{F}^{d_{n}} \rightarrow \mathbb{F}^{D}, \quad\left(\theta_{1}, \ldots, \theta_{n}\right) \mapsto \theta_{1} \otimes \cdots \otimes \theta_{n}
$$

Convenient fact: A real point in the image has a real preimage.

The set of rank-one tensors is a toric variety (product of simplices), cut out by quadratic equations; the $(2 \times 2)$-minors of its flattenings.

## A remark on field dependence

A tensor with real entries and complex rank one has real rank one. This is false for partial tensors.

Consider the real $2 \times 2 \times 2$ partial tensor with third coordinate slices

$$
\left(\begin{array}{cc}
? & 1 \\
1 & ?
\end{array}\right), \quad\left(\begin{array}{cc}
1 & ? \\
? & -1
\end{array}\right) \quad \in \mathbb{R}^{2 \times 2 \times 2}
$$

$$
\left(\begin{array}{ll}
? & 1 \\
1 & ?
\end{array}\right), \quad\left(\begin{array}{cc}
1 & ? \\
? & -1
\end{array}\right) \quad \in \mathbb{R}^{2 \times 2 \times 2} .
$$

For a rank-one completion $T$, can assume $T=\binom{1}{a} \otimes\binom{1}{b} \otimes\binom{c}{d}$. Yields

$$
b c=1, \quad a c=1, \quad d=1, \quad a b d=-1 .
$$

Only two solutions:

$$
a= \pm i, \quad b= \pm i, \quad c=\mp i, \quad d=1
$$

## Proposition

The following are equivalent.

- Every real partial tensor $T_{E}$ with nonzero entries which is completable over the complex numbers is also completeable over the real numbers.
- The index of the lattice spanned by the columns of $A_{E}$ in its saturation is odd.
Moreover, given complex-completability, real-completability depends only on the signs of observed entries.


## Idea of the proof

Diagonalize binomial equations

$$
T_{e}=\theta_{1, e_{1}} \cdots \theta_{n, e_{n}}=\theta^{a_{e}}, \quad e \in E
$$

via Smith normal form of $A_{e}=\left(a_{e}\right)_{e \in E}$

## Next step

Impose semi-algebraic constraints on the domain of

$$
\mathbb{R}^{d_{1}} \times \cdots \times \mathbb{R}^{d_{n}} \rightarrow \mathbb{R}^{D}, \quad\left(\theta_{1}, \ldots, \theta_{n}\right) \mapsto \theta_{1} \otimes \cdots \otimes \theta_{n}
$$

For example, $\theta_{i} \in \Delta_{d_{i}-1}$ is a probability distribution:

- Non-negativity of entries of $\theta_{i}$.
- Linear constraints on the entries of $\theta_{i}$.


## Next step

Impose semi-algebraic constraints on the domain of

$$
\mathbb{R}^{d_{1}} \times \cdots \times \mathbb{R}^{d_{n}} \rightarrow \mathbb{R}^{D}, \quad\left(\theta_{1}, \ldots, \theta_{n}\right) \mapsto \theta_{1} \otimes \cdots \otimes \theta_{n}
$$

For example, $\theta_{i} \in \Delta_{d_{i}-1}$ is a probability distribution:

- Non-negativity of entries of $\theta_{i}$.
- Linear constraints on the entries of $\theta_{i}$.


## By the way, the Tarski-Seidenberg theorem yields

Semi-algebraic constraints on the parameters yield only semi-algebraic constraints on the image. In principle, they can be computed by eliminating quantifiers from the formula

$$
\exists \theta_{1,1}, \ldots, \exists \theta_{n, d_{n}} \in \mathbb{R} \text { such that } x_{1 \ldots 1}=\ldots
$$

We study first the algebraic boundary of the image of

$$
\Delta_{d_{1}-1} \times \cdots \times \Delta_{d_{n}-1} \rightarrow \mathbb{R}^{E}
$$

where $E \subseteq D$, and $\Delta_{m}$ is the probability simplex of dimension $m$.

The algebraic boundary of a semi-algebraic set $S \subseteq \mathbb{R}^{n}$ is the Zariskiclosure of the (Euclidean topology) boundary $\partial S=\mathrm{cl}(S) \backslash \operatorname{int}(S)$.

How to compute it
Compute the branch locus, the locus in the image where the rank of the Jacobian of the parametrization drops.


## Assume now

- the number of observations equals the number of parameters (i.e. Jacobian is square)
- every maximal-dimensional slice is observed


## Approach

- Sinn's Lemma: If a semi-algebraic set $S \subseteq \mathbb{R}^{k}$ is nonempty and contained in the closure of its interior and the same holds for $\mathbb{R}^{k} \backslash S$, then its algebraic boundary is of pure codimension one.
- Implicit function theorem: If an interior parameter point maps to a boundary tensor, the Jacobian determinant vanishes there.
- Argue that remaining components are all contained in coordinate hyperplanes.

Assume the observed entries of a $2 \times 2 \times 2$ tensor are $x_{211}, x_{121}, x_{112}$. Denote $l_{i}=1-\theta_{i}$ for $i=1,2,3$. The graph of the map is defined by

$$
I=\left\langle x_{211}-l_{1} \theta_{2} \theta_{3}, x_{121}-\theta_{1} l_{2} \theta_{3}, x_{112}-\theta_{1} \theta_{2} l_{3}\right\rangle .
$$

The Jacobian matrix of the parametrization map equals

$$
J=\left(\begin{array}{ccc}
-\theta_{2} \theta_{3} & l_{1} \theta_{3} & l_{1} \theta_{2} \\
l_{2} \theta_{3} & -\theta_{1} \theta_{3} & \theta_{1} l_{2} \\
\theta_{2} l_{3} & \theta_{1} l_{3} & -\theta_{1} \theta_{2}
\end{array}\right)
$$

and has determinant

$$
\theta_{1}^{2} \theta_{2} \theta_{3}+\theta_{1} \theta_{2}^{2} \theta_{3}+\theta_{1} \theta_{2} \theta_{3}^{2}-2 \theta_{1} \theta_{2} \theta_{3}=\theta_{1} \theta_{2} \theta_{3}\left(-\theta_{1}-\theta_{2}-\theta_{3}+2\right) .
$$

$\rightarrow$ Jacobian determinant is a monomial times a linear polynomial.

## Explanation of the linear polynomial

$\theta_{1}^{2} \theta_{2} \theta_{3}+\theta_{1} \theta_{2}^{2} \theta_{3}+\theta_{1} \theta_{2} \theta_{3}^{2}-2 \theta_{1} \theta_{2} \theta_{3}=\theta_{1} \theta_{2} \theta_{3}\left(-\theta_{1}-\theta_{2}-\theta_{3}+2\right)$.
Consider the matrix

$$
B_{E}=\left(\begin{array}{llll}
0 & 1 & 1 & 1 \\
1 & 0 & 1 & 1 \\
1 & 1 & 0 & 1
\end{array}\right)
$$

encoding in its columns which parameters $\theta_{i}$ contribute to a given observed entry, plus an extra column of ones.
The kernel of $B_{E}$ is spanned by $v=(-1,-1,-1,2)^{T}$, which yields the coefficients.

$$
I+\left\langle l_{p}\right\rangle=\left\langle x_{211}-l_{1} \theta_{2} \theta_{3}, x_{121}-\theta_{1} l_{2} \theta_{3}, x_{112}-\theta_{1} \theta_{2} l_{3},-\theta_{1}-\theta_{2}-\theta_{3}+2\right\rangle .
$$

Eliminating $\theta_{1}, \theta_{2}$, and $\theta_{3}$ yields a prime ideal generated by

$$
\begin{array}{r}
x_{211}^{4} x_{121}^{2}-2 x_{211}^{3} x_{121}^{3}+x_{211}^{2} x_{121}^{4}-2 x_{211}^{4} x_{121} x_{112}+2 x_{211}^{3} x_{121}^{2} x_{112}+2 x_{211}^{2} x_{121}^{3} x_{112} \\
-2 x_{211} x_{121}^{4} x_{112}+x_{211}^{4} x_{112}^{2}+2 x_{211}^{3} x_{121} x_{112}^{2}-6 x_{211}^{2} x_{121}^{2} x_{112}^{2}+2 x_{211} x_{121}^{3} x_{112}^{2}+x_{121}^{4} x_{112}^{2} \\
-2 x_{211}^{3} x_{112}^{3}+2 x_{211}^{2} x_{121} x_{112}^{3}+2 x_{211} x_{121}^{2} x_{112}^{3}-2 x_{121}^{3} x_{112}^{3}+x_{211}^{2} x_{112}^{4}-2 x_{211} x_{121} x_{112}^{4} \\
+x_{121}^{2} x_{112}^{4}-2 x_{211}^{3} x_{121}^{2}-2 x_{211}^{2} x_{121}^{3}+8 x_{211}^{3} x_{121} x_{112}-4 x_{211}^{2} x_{121}^{2} x_{112}+8 x_{211} x_{121}^{3} x_{112} \\
-2 x_{211}^{3} x_{112}^{2}-4 x_{211}^{2} x_{121}^{2} x_{112}^{2}-4 x_{211} x_{121}^{2} x_{112}^{2}-2 x_{121}^{3} x_{112}^{2}-2 x_{211}^{2} x_{112}^{3}+8 x_{211} x_{121} x_{112}^{3} \\
-2 x_{121}^{2} x_{112}^{3}+x_{211}^{2} x_{121}^{2}-10 x_{211}^{2} x_{121} x_{112}-10 x_{211} x_{121}^{2} x_{112}+x_{211}^{2} x_{112}^{2}-10 x_{211} x_{121} x_{112}^{2} \\
+x_{121}^{2} x_{112}^{2}+4 x_{211} x_{121} x_{112} .
\end{array}
$$

Observation: Eliminating from

$$
I+\left\langle\theta_{1} \theta_{2} \theta_{3}\left(-\theta_{1}-\theta_{2}-\theta_{3}+2\right)\right\rangle
$$

... and this is how it looks


- The surface is singular in dimension 1 (all along the boundary)
- The interior extends into negative coordinates


## Let

- $I=\left\langle x_{e}-\prod \theta, \ldots\right\rangle$ be the graph ideal of the parametrization.
- $l$ be the linear polynomial factor of the Jacobian determinant.
- $|E|$ be equal to the number of parameters.
- $E \subseteq D$ meet every maximal-dimensional slice.


## Algebraic boundary theorem

Eliminating the parameter variables from $I+\langle l\rangle$ yields a non-zero principal ideal generated by a non-constant irreducible polynomial $f$. The polynomial $q$ that defines the algebraic boundary of the completable region is the product of $f$ with some coordinates.

## Open Problems

- Determine the degree of $f$.
- How much do we know from the algebraic boundary?
- Non-square Jacobians ( $|E|$ small)

Have some answers for diagonal observations...

## Observing diagonal entries

- Consider $d \times d \times \cdots \times d$ tensor of order $n$.
- Let $E$ consists only of the $d$ diagonal entries (much fewer than number of parameters).
- Let $S_{n, d}$ be the set of diagonal entries that admit a completion to a rank-one "probability tensor" (independent multivariate discrete distribution).


## Theorem (Kubjas/Rosen)

$S_{n, d}$ is a semi-algebraic set and its algebraic boundary is known.
Furthermore

$$
S_{n, d}=\left\{x \in \mathbb{R}_{\geq 0}^{d}: \sum_{i=1}^{d} x_{i}^{\frac{1}{n}} \leq 1\right\}
$$

## Theorem

There are explicitly described polynomials $P_{n, d, i}(x), i=0, \ldots, n^{d-1}$ such that $x \in \mathbb{R}_{\geq 0}^{d}$ is an element of $S_{n, d}$ if and only if $P_{n, d, i}(x) \geq 0$ for all $0 \leq i<n^{d-1}$. If $n$ is odd, then $S_{n, d}=\left\{x \in \mathbb{R}_{\geq 0}^{d}: P_{n, d, 0}(x) \geq 0\right\}$.

Let $e_{i, d}$ denote the $i$ th elementary symmetric polynomial in $x_{1}, \ldots, x_{d}$.
$S_{3,2}(2 \times 2 \times 2$ tensors $)$ is defined by

$$
\begin{aligned}
x_{1}, x_{2} & \geq 0 \\
\left(1-e_{1,2}\right)^{3}-27 e_{2,2} & \geq 0 .
\end{aligned}
$$

$S_{2,3}(3 \times 3$ matrices $)$ is defined by

$$
\begin{aligned}
x_{1}, x_{2}, x_{3} & \geq 0 \\
1-e_{1,3} & \geq 0 \\
3\left(1-e_{1,3}\right)^{2}-4 e_{2,3} & \geq 0 \\
\left(1-e_{1,3}\right)\left(\left(1-e_{1,3}\right)^{2}-4 e_{2,3}\right)-16 e_{3,3} & \geq 0 \\
\left(\left(1-e_{1,3}\right)^{2}-4 e_{2,3}\right)^{2}-64 e_{3,3} & \geq 0
\end{aligned}
$$

For $d=n=2$ the two analyzed classes overlap. We get

$$
\begin{aligned}
x_{1}, x_{2} & \geq 0 \\
1-2\left(x_{1}+x_{2}\right)+\left(x_{1}-x_{2}\right)^{2} & \geq 0 \\
1-x_{1}-x_{2} & \geq 0
\end{aligned}
$$




The algebraic boundary misses $1-x_{1}-x_{2} \geq 0$ and is thus not the final answer, even in the easiest case.

## Conclusion

- The set of partial multi-variate independent probability distributions is semi-algebraic.
- In some cases we can find the algebraic boundary.
- We do want complete semi-algebraic descriptions.
- Semi-algebraic statistics is fun!


## Conclusion

- The set of partial multi-variate independent probability distributions is semi-algebraic.
- In some cases we can find the algebraic boundary.
- We do want complete semi-algebraic descriptions.
- Semi-algebraic statistics is fun!

