

# Identifying Subspaces from Canonical Projections

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Applied Algebra Seminar  
November 13<sup>th</sup>, 2014

Robert Nowak and Nigel Boston

# Outline

- ▶ Introduction
- ▶ Problem Description
- ▶ Setup
- ▶ The Answer
- ▶ Sketch of the proof
- ▶ Application
- ▶ Conclusions



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# Introduction

We have lots of data



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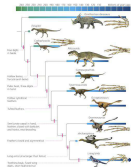
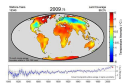
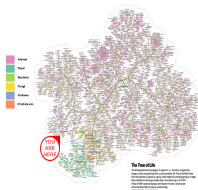
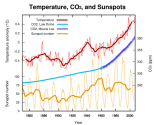
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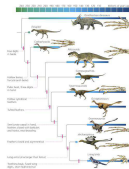
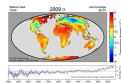
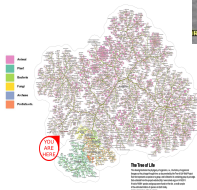
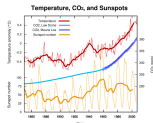
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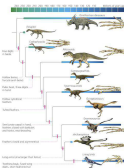
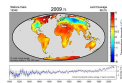
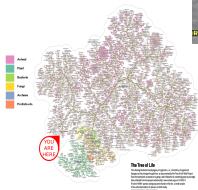
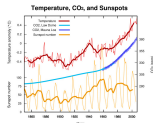
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$$\begin{array}{c} \text{movies} \end{array} \left\{ \begin{array}{c} \left[ \begin{array}{cccccccccc} 4 & \cdot & 1 & 1 & 5 & 4 & 2 & 5 & 3 & \cdot \\ \cdot & 5 & \cdot & 1 & \cdot & \cdot & \cdot & 4 & \cdot & 5 \\ 1 & \cdot & 2 & \cdot & 2 & 5 & 1 & \cdot & \cdot & 4 \\ 2 & \cdot & 5 & 4 & \cdot & 1 & \cdot & 2 & \cdot & 4 \\ \cdot & 5 & 1 & \cdot & 5 & \cdot & \cdot & \cdot & 1 & \cdot \\ 1 & 5 & \cdot & 2 & \cdot & 2 & 1 & \cdot & \cdot & 5 \end{array} \right] \end{array} \right.$$

$\underbrace{\hspace{15em}}_{\text{people}}$

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$$\text{movies} \left\{ \begin{bmatrix} 4 & \cdot & 1 & 1 & 5 & 4 & 2 & 5 & 3 & \cdot \\ \cdot & 5 & \cdot & 1 & \cdot & \cdot & \cdot & 4 & \cdot & 5 \\ 1 & \cdot & 2 & \cdot & 2 & 5 & 1 & \cdot & \cdot & 4 \\ 2 & \cdot & 5 & 4 & \cdot & 1 & \cdot & 2 & \cdot & 4 \\ \cdot & 5 & 1 & \cdot & 5 & \cdot & \cdot & \cdot & 1 & \cdot \\ 1 & 5 & \cdot & 2 & \cdot & 2 & 1 & \cdot & \cdot & 5 \end{bmatrix} \right.$$

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Nobody has seen every movie.

# Introduction

But if we knew *who* would like *what*...

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$$\text{movies} \left\{ \begin{bmatrix} 4 & 5 & 1 & 1 & 5 & 4 & 2 & 5 & 3 & 1 \\ 2 & 5 & 3 & 1 & 4 & 1 & 4 & 4 & 3 & 5 \\ 1 & 3 & 2 & 5 & 2 & 5 & 1 & 3 & 2 & 4 \\ 2 & 1 & 5 & 4 & 3 & 1 & 5 & 2 & 5 & 4 \\ 3 & 5 & 1 & 1 & 5 & 1 & 4 & 3 & 1 & 3 \\ 1 & 5 & 3 & 2 & 2 & 2 & 1 & 4 & 5 & 5 \end{bmatrix} \right.$$

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We would be able to make good recommendations!



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$$\begin{matrix} \text{clothes} & \left\{ \begin{array}{l} \left[ \begin{array}{cccccccccc} 4 & \cdot & 1 & 1 & 5 & 4 & 2 & 5 & 3 & \cdot \\ \cdot & 5 & \cdot & 1 & \cdot & \cdot & \cdot & 4 & \cdot & 5 \\ 1 & \cdot & 2 & \cdot & 2 & 5 & 1 & \cdot & \cdot & 4 \\ 2 & \cdot & 5 & 4 & \cdot & 1 & \cdot & 2 & \cdot & 4 \\ \cdot & 5 & 1 & \cdot & 5 & \cdot & \cdot & \cdot & 1 & \cdot \\ 1 & 5 & \cdot & 2 & \cdot & 2 & 1 & \cdot & \cdot & 5 \end{array} \right] \end{array} \right. \end{matrix}$$

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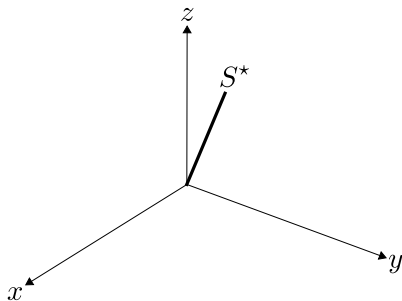
- ▶ Linear algebra is one of our favorite tools...
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- ▶ But what if data are missing?
  - ▶ ?????
- ▶ There is great interest on extending usage of linear algebra to incomplete datasets.
- ▶ That is what we are studying.

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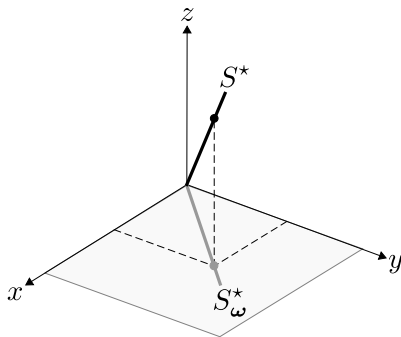
## Problem description

$S^* := r$ -dimensional subspace of  $\mathbb{R}^d$ ,  $r < d$ .



## Problem description

$S_\omega^* :=$  Projection of  $S^*$  onto a canonical subspace.

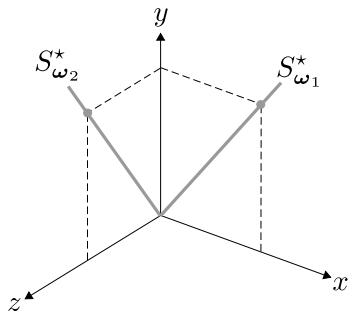


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Suppose I don't tell you  $S^*$ ...

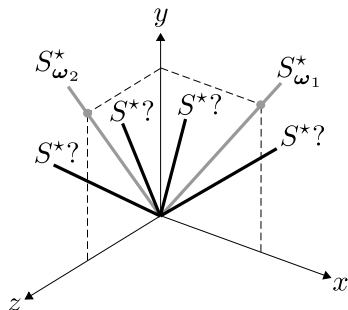
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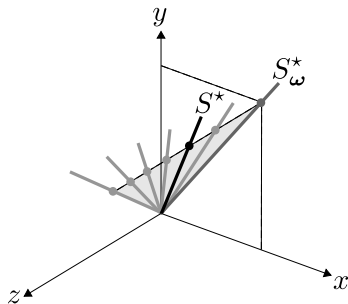
Can you uniquely determine  $S^*$  from this set of projections?



# Problem description

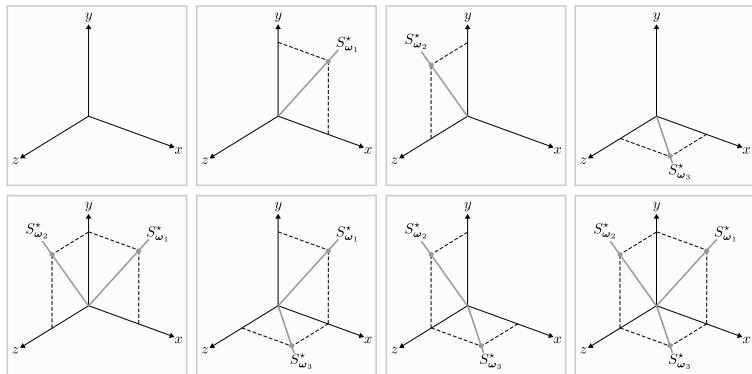
Is this even possible?

There might be many subspaces that agree with the projections.



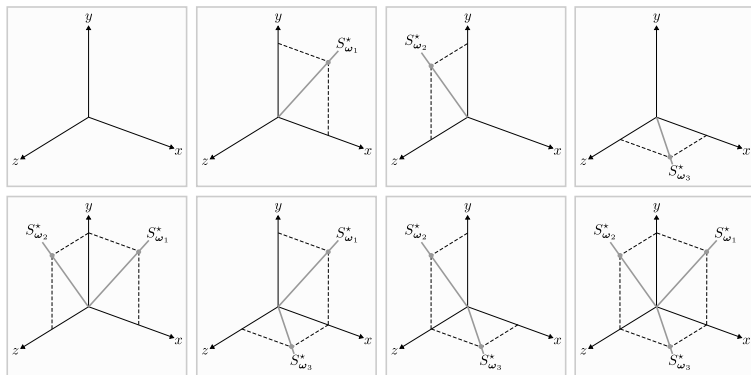
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Well... it depends on which set of projections I give you.



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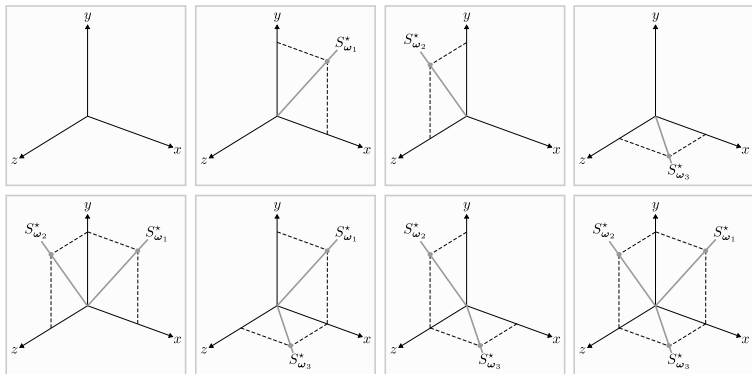
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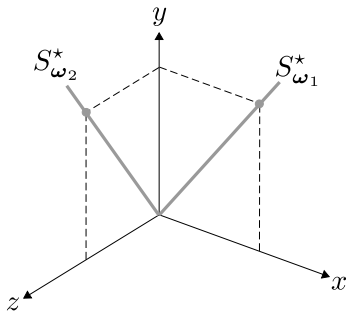
This is what we answer here: which are *the good sets*.

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- ▶ Problem Description ✓
- ▶ **Setup**
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# Setup

The columns of  $\Omega$  will index the given projections.



$$\Omega = \begin{bmatrix} \omega_1 & \omega_2 \\ 1 & 0 \\ 1 & 1 \\ 0 & 1 \end{bmatrix}$$

# Setup

- ▶  $\text{Gr}(r, \mathbb{R}^d) :=$  Grassmannian manifold of  $r$ -dimensional subspaces in  $\mathbb{R}^d$ .

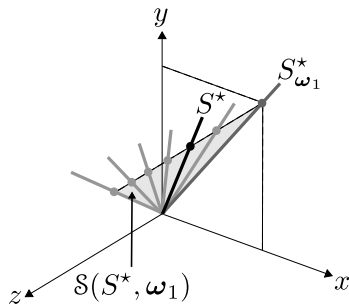
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- ▶  $\text{Gr}(r, \mathbb{R}^d) :=$  Grassmannian manifold of  $r$ -dimensional subspaces in  $\mathbb{R}^d$ .
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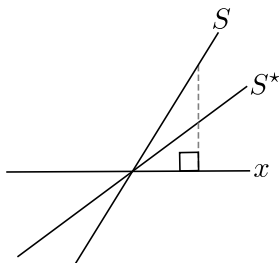


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- ▶  $S^*$  is  $r$ -dimensional.

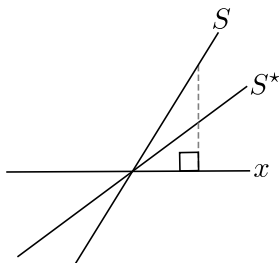
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- ▶  $\Rightarrow$  Assume w.l.o.g. that all projections are onto  $r + 1$  canonical coordinates.

# Setup

- For any matrix  $\Omega'$  formed with a subset of the columns in  $\Omega$ :

$$\Omega' = \underbrace{\begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}}_{n(\Omega') := \# \text{columns}} \left. \vphantom{\begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}} \right\} m(\Omega') := \# \text{nonzero rows}$$

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- $d - r$  projections are *necessary*, so we will assume w.l.o.g.

$$n(\Omega) = d - r.$$

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# The Answer

## Theorem (Pimentel-Alarcón, Nowak, Boston, '14)

*For almost every  $S^\star$ , with respect to the uniform measure over  $\text{Gr}(r, \mathbb{R}^d)$ ,  $S^\star$  is the only subspace in  $\mathcal{S}(S^\star, \Omega)$  if and only if for every matrix  $\Omega'$  formed with a subset of the columns in  $\Omega$ ,*

$$m(\Omega') \geq n(\Omega') + r.$$



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There is a set of measure zero of *bad* subspaces that we wouldn't identify.

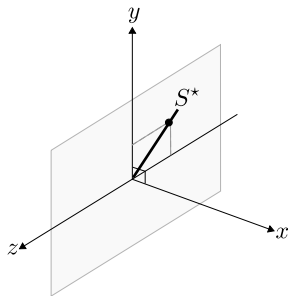
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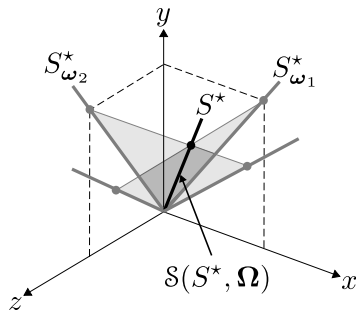
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This is what we want!



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Every subset of  $n$  columns of  $\Omega$  has at least  $n + r$  nonzero rows.



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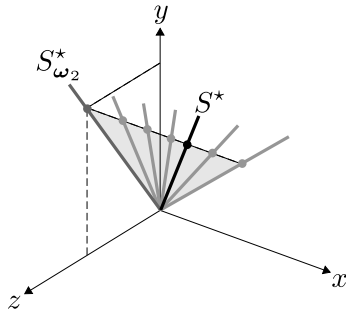
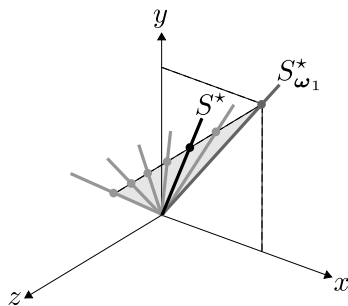
$$\Omega = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \Rightarrow \text{Check: } \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 & 1 \\ 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 1 & 1 \\ 1 & 0 \\ 0 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 1 \\ 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 1 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

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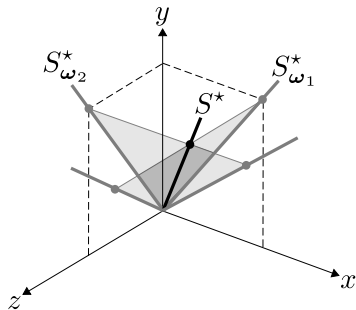
## Sketch of the proof

We will find the subspaces that agree with *each* projection.



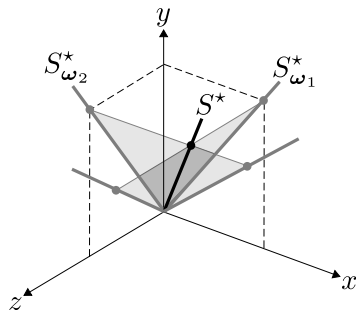
## Sketch of the proof

Then find the intersection.



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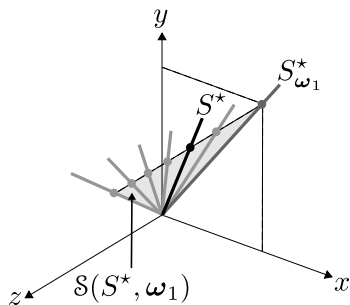
Then find the intersection.



If the intersection only contains one subspace, then ;)

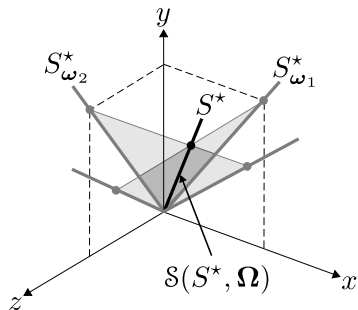
## Sketch of the proof

$\mathcal{S}(S^*, \omega_i) :=$  Set of  $r$ -dimensional subspaces matching  $S^*$  on  $\omega_i$ .



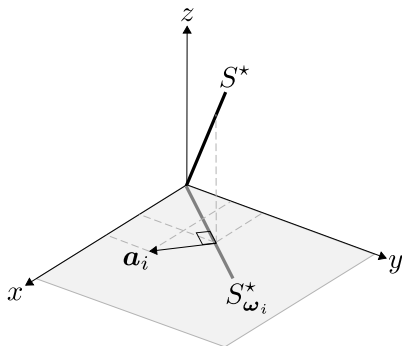
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$$\mathcal{S}(S^*, \Omega) = \bigcap_i \mathcal{S}(S^*, \omega_i).$$



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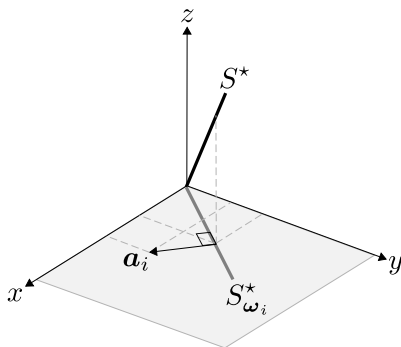
$\mathbf{a}_i :=$  Vector orthogonal to the  $i^{th}$  projection.





## Sketch of the proof

$\mathbf{a}_i :=$  Vector orthogonal to the  $i^{th}$  projection.



An entry in  $\mathbf{a}_i$  is zero iff the corresponding entry in  $\boldsymbol{\omega}_i$  is zero.

# Sketch of the proof

One great thing:

- ▶ Every subspace in  $\mathcal{S}(S^*, \omega_i)$  is orthogonal to  $\mathbf{a}_i$ .

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Cool!  $\Rightarrow$

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$$\mathbf{A} = [ \mathbf{a}_1 \mid \cdots \mid \mathbf{a}_N ].$$

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Cool!  $\Rightarrow$

- ▶ Construct

$$\mathbf{A} = [ \mathbf{a}_1 \mid \cdots \mid \mathbf{a}_N ].$$

- ▶ Every  $S \in \mathcal{S}(S^*, \Omega)$  must be contained in

$$\ker \mathbf{A}^\top.$$

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# Sketch of the proof

What we really want is to determine  $\dim \ker \mathbf{A}^T$ .

- ▶ If  $\dim \ker \mathbf{A}^T > r$   
 $\Rightarrow$  There are many subspaces that agree with the projections
- ▶ If  $\dim \ker \mathbf{A}^T = r$   
 $\Rightarrow$  Only  $S^*$  will agree with the projections.

# Sketch of the proof

What we really want is to determine  $\dim \ker \mathbf{A}^\top$ .

- ▶ If  $\dim \ker \mathbf{A}^\top > r$   
 $\Rightarrow$  There are many subspaces that agree with the projections
- ▶ If  $\dim \ker \mathbf{A}^\top = r$   
 $\Rightarrow$  Only  $S^\star$  will agree with the projections. Moreover,

$$S^\star = \ker \mathbf{A}^\top$$

.



## Sketch of the proof

- For any matrix  $\mathbf{A}'$  formed with a subset of the columns in  $\mathbf{A}$ :

$$\mathbf{A}' = \underbrace{\begin{bmatrix} a_{11} & 0 \\ a_{21} & a_{22} \\ 0 & a_{32} \\ 0 & 0 \end{bmatrix}}_{n(\mathbf{A}') := \# \text{columns}} \left. \vphantom{\begin{bmatrix} a_{11} & 0 \\ a_{21} & a_{22} \\ 0 & a_{32} \\ 0 & 0 \end{bmatrix}} \right\} m(\mathbf{A}') := \# \text{nonzero rows}$$

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- ▶ We want  $\dim \ker \mathbf{A}^T = r$ , so  $\mathbf{A}$  better have  $d - r$  linearly independent columns.

# Sketch of the proof

We know how to deal with  $\mathbf{A}$  using linear algebra!

- ▶ Through some technical details:

**Lemma (Pimentel-Alarcón, Nowak, Boston, '14)**

*For almost every  $S^*$ , the columns of  $\mathbf{A}$  are linearly dependent if and only if  $m(\mathbf{A}') < n(\mathbf{A}') + r$  for some matrix  $\mathbf{A}'$  formed with a subset of the columns in  $\mathbf{A}$ .*

## Sketch of the proof

The zero entries of  $\mathbf{\Omega}$  and  $\mathbf{A}$  are in the same positions.

$$\mathbf{\Omega} = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \iff \mathbf{A}' = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & 0 & 0 \\ 0 & a_{32} & 0 \\ 0 & 0 & a_{43} \end{bmatrix}$$

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Then

$$m(\Omega') \geq n(\Omega') + r \iff m(A') \geq n(A') + r$$



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- ▶ Lff  $S^*$  is the only subspace in  $\mathcal{S}(S^*, \Omega)$ .



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- ▶ Introduction
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# Application

## Low-Rank Matrix Completion (LRMC)

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- ▶ Given a subset of entries in a rank  $r$  matrix, exactly recover *all* of the missing entries.

$$\mathbf{x}_{\Omega} = \begin{bmatrix} 1 & \cdot & 3 & \cdot \\ 1 & 2 & \cdot & \cdot \\ \cdot & 2 & 3 & \cdot \\ \cdot & \cdot & \cdot & 4 \\ \cdot & \cdot & \cdot & 4 \end{bmatrix} \Rightarrow \hat{\mathbf{x}} = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 1 & 2 & 3 & 4 \\ 1 & 2 & 3 & 4 \\ 1 & 2 & 3 & 4 \\ 1 & 2 & 3 & 4 \end{bmatrix}$$

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- ▶  $\sim$  Identifying the subspace spanned by the columns,  $S^*$ . Here

$$\hat{S} = \text{span} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

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And the real subspace is

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What if these assumptions are not met? How can we validate a completion?

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## Corollary (Pimentel-Alarcón, Nowak, Boston, '14)

*Let the columns of  $\mathcal{X}$  be drawn independently according to  $\mu$ , an absolutely continuous distribution with respect to the Lebesgue measure on  $S^*$ . Suppose  $\mathcal{X}_{\Omega}$  can be partitioned into two sets of columns,  $\mathcal{X}_{\Omega_1}$  and  $\mathcal{X}_{\Omega_2}$ , such that  $\Omega_2$  satisfies the conditions of the subspace identifiability theorem.*

*Let  $\hat{S}$  be the output of running an LRMC algorithm on  $\mathcal{X}_{\Omega_1}$ . Then for almost every  $S^*$ , and almost surely with respect to  $\mu$ ,  $\mathcal{X}_{\Omega_2}$  fits in  $\hat{S}$  if and only if  $\hat{S} = S^*$ .*

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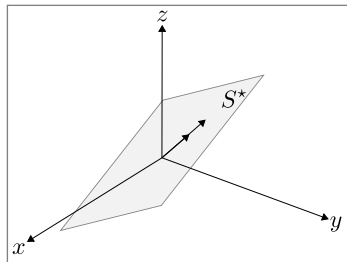
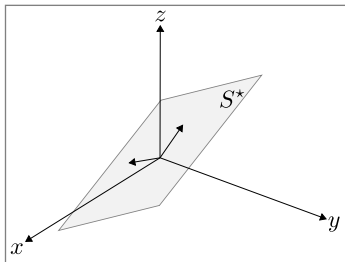
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- ▶ It is possible to uniquely identify an  $r$ -dimensional subspace  $S^*$  from its projections onto  $\Omega$ .
- ▶ If and only if every subset of  $n$  columns of  $\Omega$  has at least  $n + r$  nonzero rows.
- ▶ Whence  $S^* = \ker \mathbf{A}^T$ .

Thanks.