

# 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



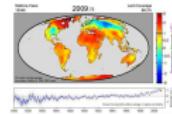
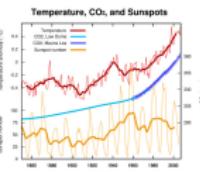
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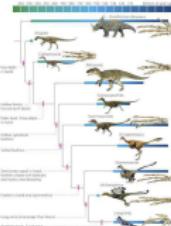
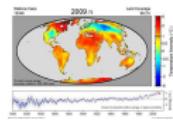
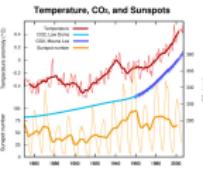
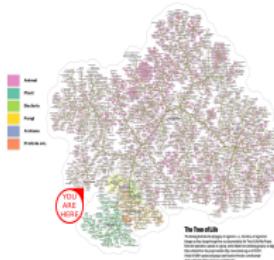
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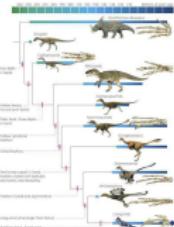
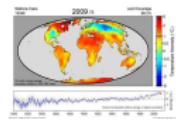
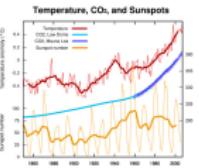
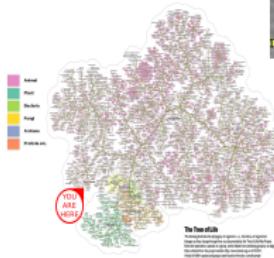
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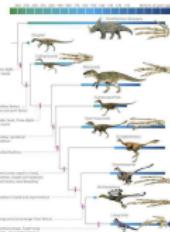
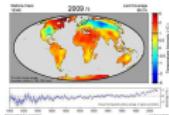
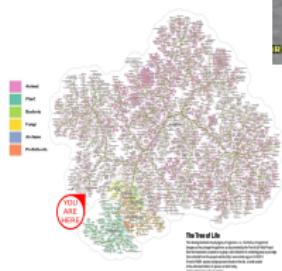
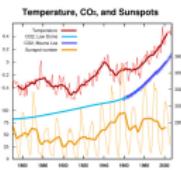
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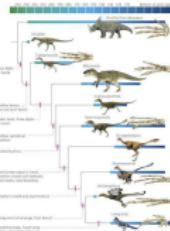
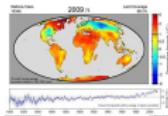
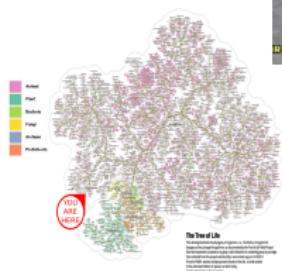
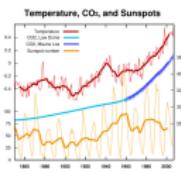
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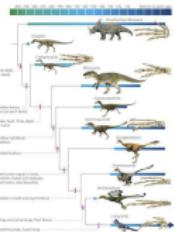
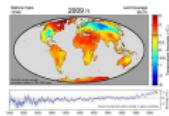
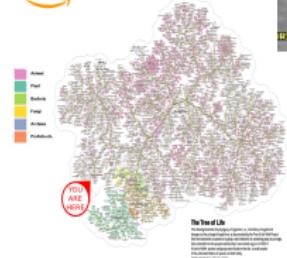
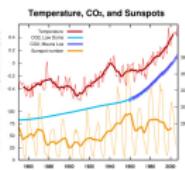
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Nobody has seen every movie.

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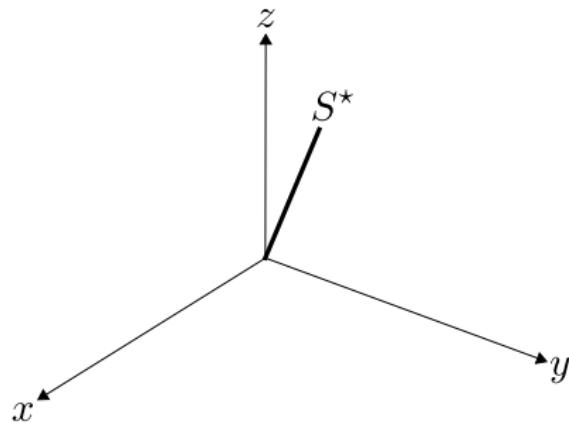
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- ▶ There is great interest on extending usage of linear algebra to incomplete datasets.
- ▶ **That is what we are studying.**

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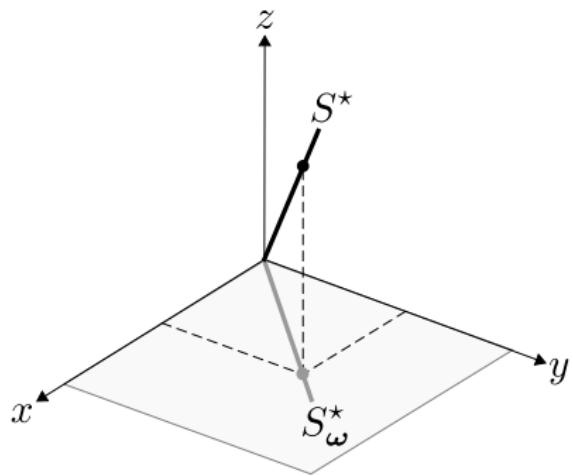
## Problem description

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



## Problem description

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

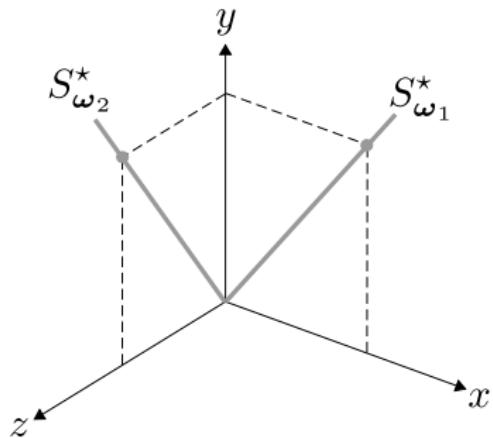


## Problem description

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

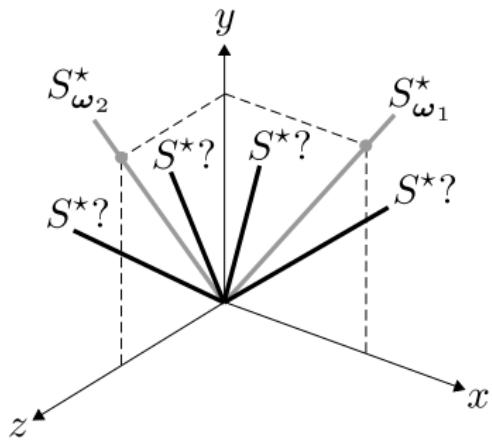
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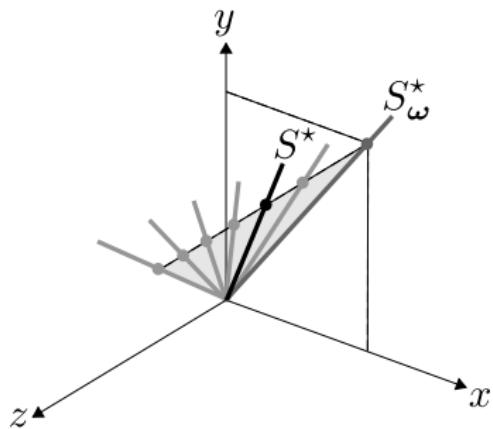


Can you uniquely determine  $S^*$  from this set of projections?

## Problem description

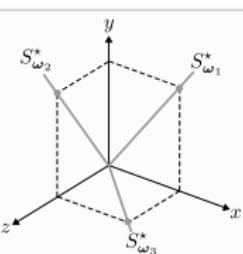
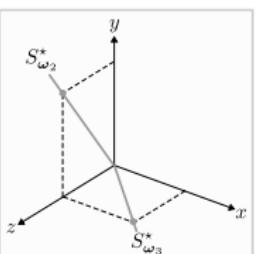
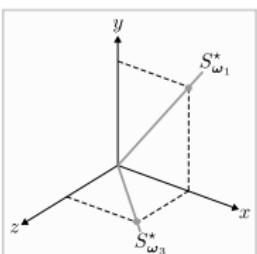
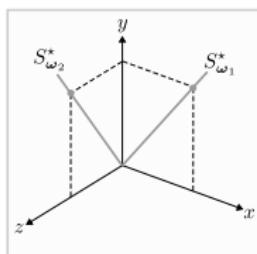
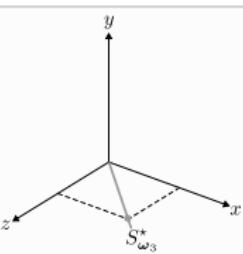
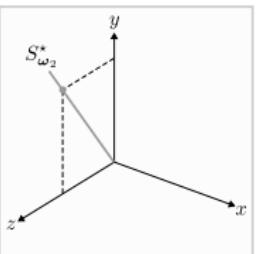
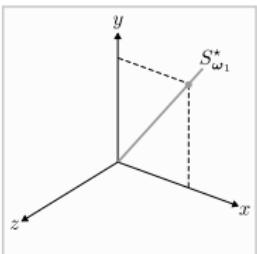
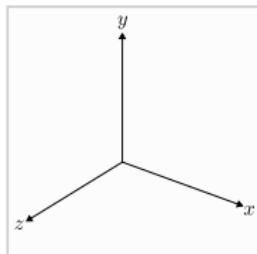
Is this even possible?

There might be many subspaces that agree with the projections.



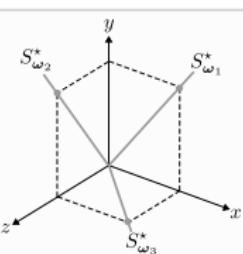
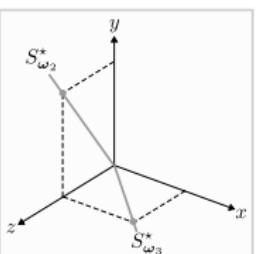
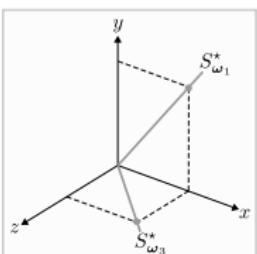
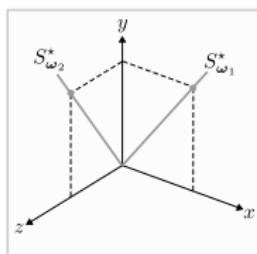
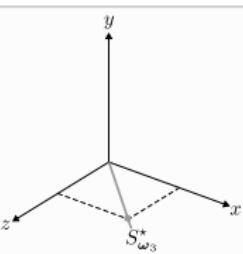
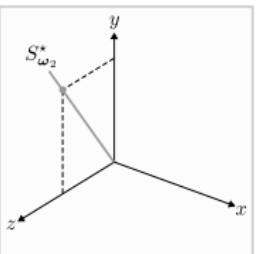
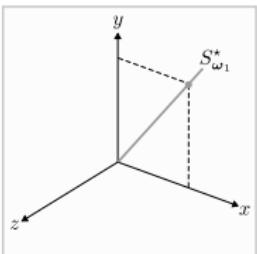
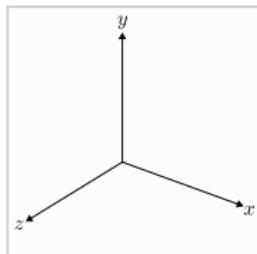
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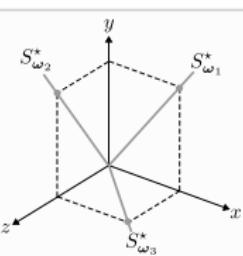
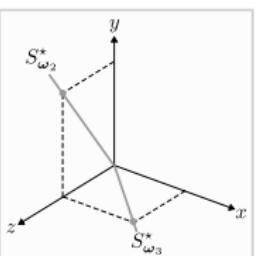
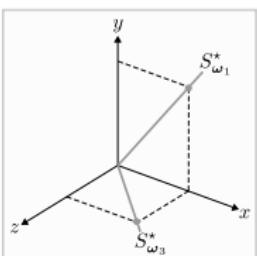
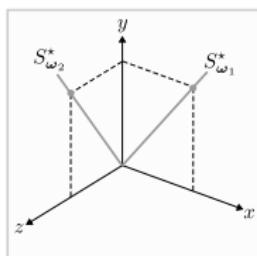
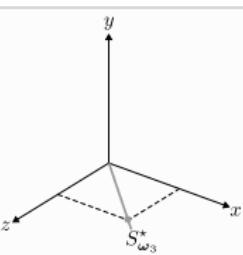
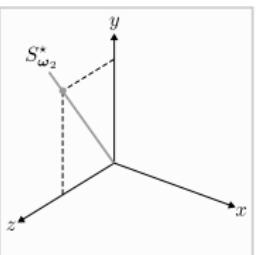
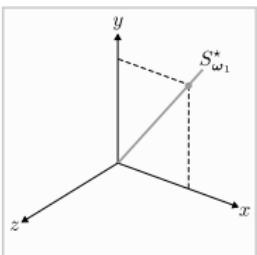
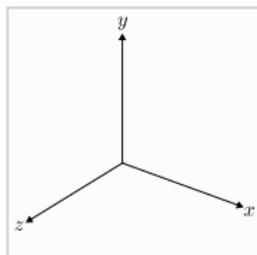
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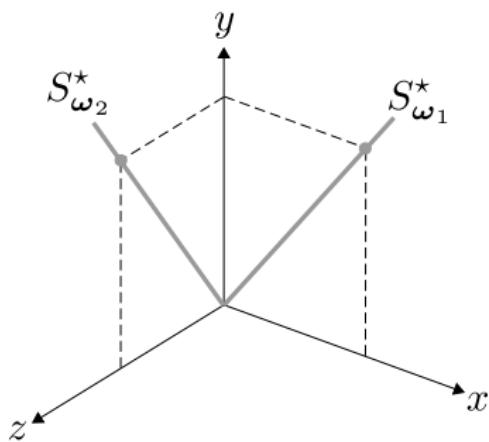
This is what we answer here: which are *the good sets*.

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## Setup

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



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

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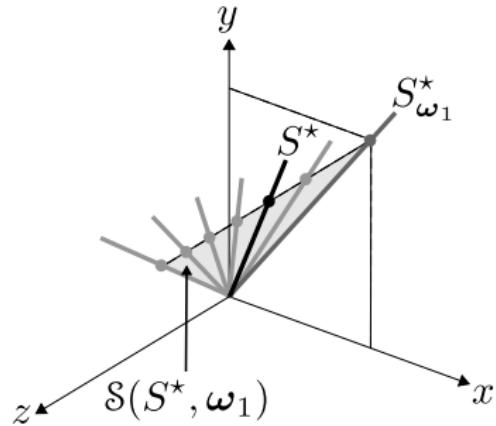
- ▶  $\text{Gr}(r, \mathbb{R}^d) :=$  Grassmannian manifold of  $r$ -dimensional subspaces in  $\mathbb{R}^d$ .

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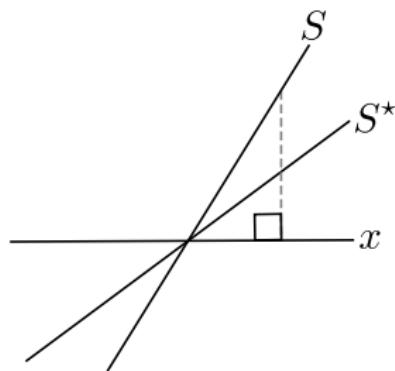


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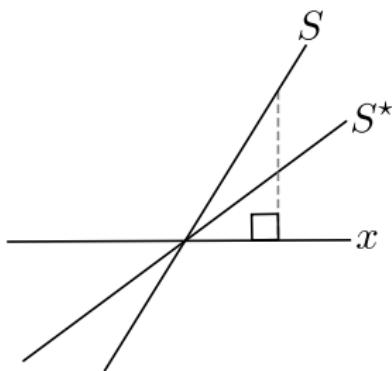
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- ▶ ⇒ Assume w.l.o.g. that all projections are onto  $r + 1$  canonical coordinates.

## Setup

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

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

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

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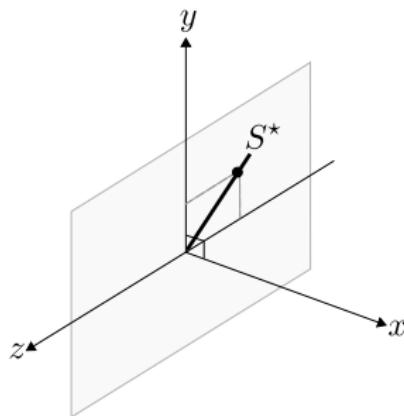
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For almost every  $S^*$ , with respect to the uniform measure over  $\text{Gr}(r, \mathbb{R}^d)$ ,  $S^*$  is the only subspace in  $\mathcal{S}(S^*, \Omega)$  if and only if for every matrix  $\Omega'$  formed with a subset of the columns in  $\Omega$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

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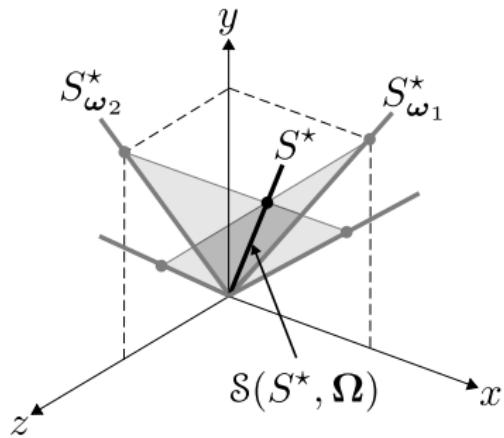
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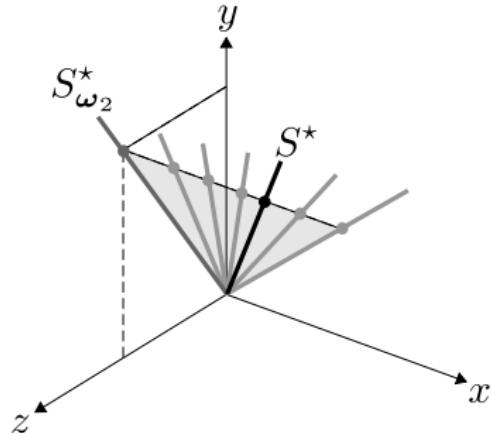
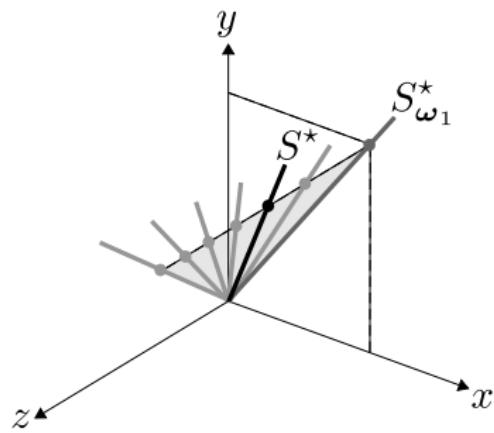
$$\Omega = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \Rightarrow \text{Check: } \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 & 1 \\ 1 & 0 \\ 0 & 1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

# Outline

- ▶ Introduction
- ▶ Problem Description ✓
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- ▶ Sketch of the proof
- ▶ Application
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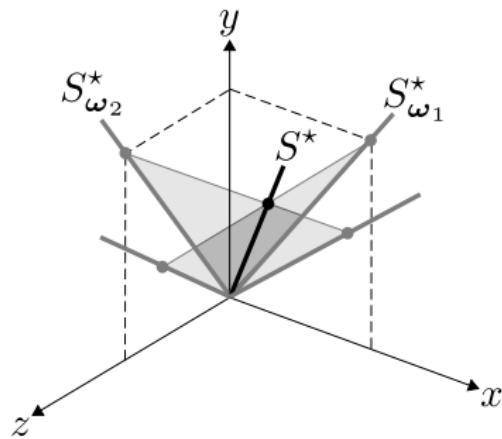
## Sketch of the proof

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



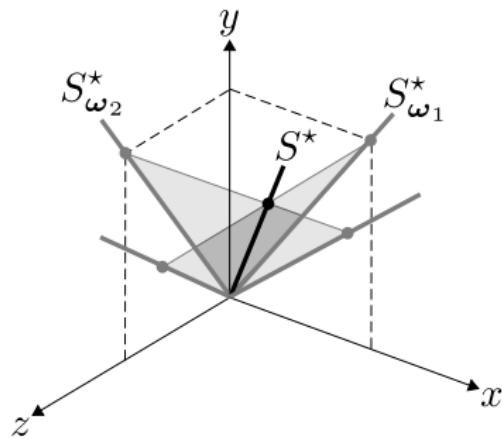
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Then find the intersection.



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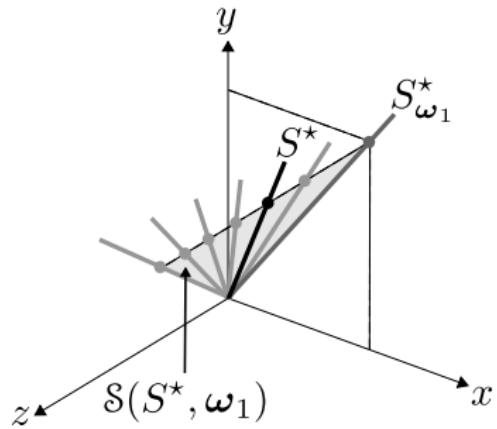
Then find the intersection.



If the intersection only contains one subspace, then ;)

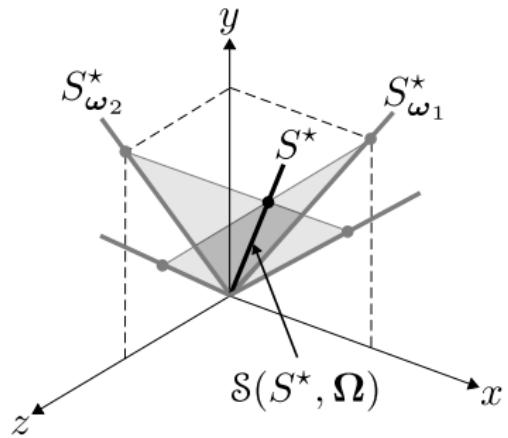
## Sketch of the proof

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



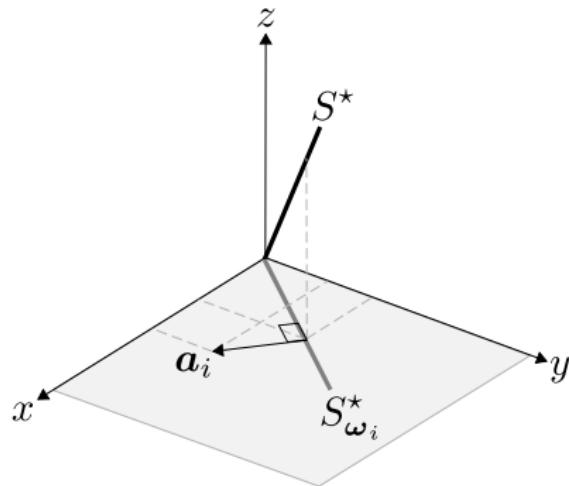
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$$\mathcal{S}(S^*, \Omega) = \bigcap_i \mathcal{S}(S^*, \omega_i).$$



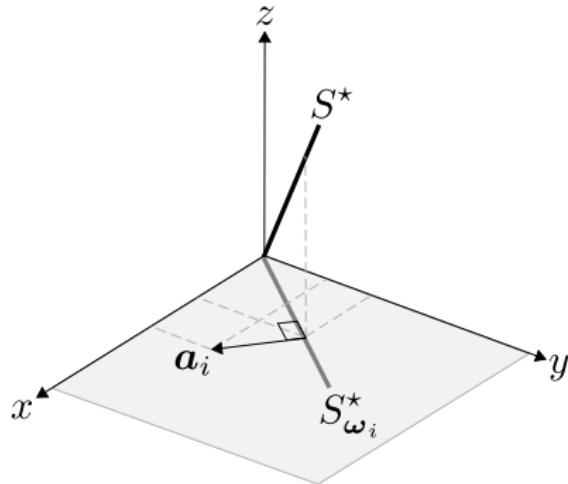
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$a_i :=$  Vector orthogonal to the  $i^{th}$  projection.



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An entry in  $a_i$  is zero iff the corresponding entry in  $\omega_i$  is zero.

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One great thing:

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

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## Sketch of the proof

One great thing:

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Cool!  $\Rightarrow$

- ▶ Construct

$$\mathbf{A} = [ \ a_1 \ | \ \cdots \ | \ a_N \ ].$$

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

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

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- ▶ If  $\dim \ker A^T = r$   
⇒ Only  $S^*$  will agree with the projections. Moreover,

$$S^* = \ker A^T$$

## Sketch of the proof

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

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

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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  $\mathbf{A}$ .*

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

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

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Then

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

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## 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.

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

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- ~ Identifying the subspace spanned by the columns,  $S^*$ . Here

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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.*

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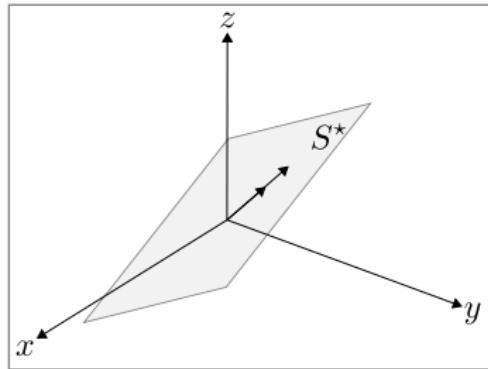
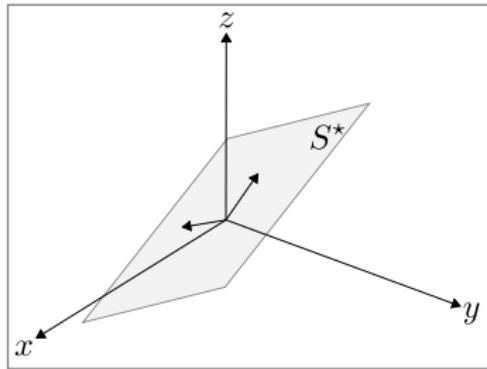
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Now we know that:

- ▶ 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  $S^* = \ker A^T$ .

Thanks.