

Geometry of Linear Neural Networks that are Equivariant / Invariant under Permutation Groups

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joint work with

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Algebra & Geometry \Rightarrow Neural Network Theory

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Geometric questions:

1. How does the network architecture affect the geometry of the function space?
2. How does the geometry of the function space impact the training of the network?

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Algebra & Geometry \Rightarrow Neural Network Theory

Algebraic settings:

| activation | network architecture | network structure | loss |
|------------|----------------------|-------------------|------|
| | | | |

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| identity | network structure | |
| ReLU | | |
| polynomial | | |

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| identity | fully-connected | |
| ReLU | convolutional | |
| polynomial | group equivariant | |

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| ReLU | convolutional | Wasserstein distance | = polyhedral dist. |
| polynomial | group equivariant | cross-entropy | \cong KL divergence |

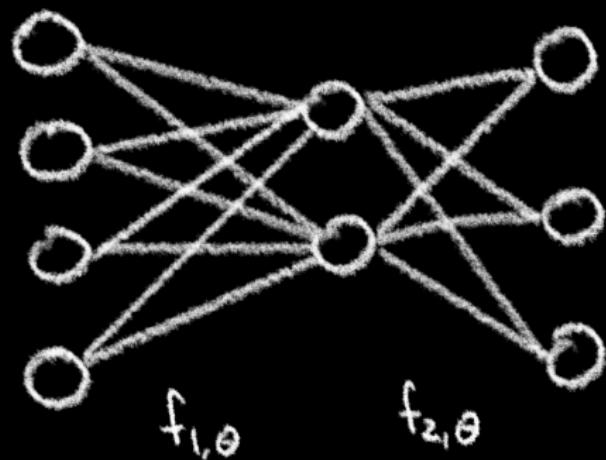
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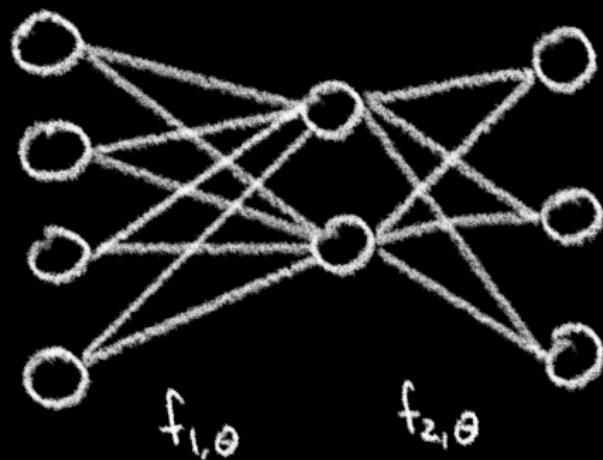
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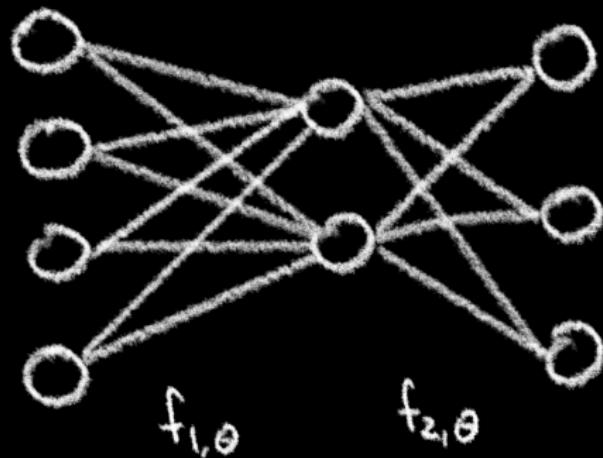


This network parametrizes linear maps:

$$\begin{aligned}\mu : \mathbb{R}^{2 \times 4} \times \mathbb{R}^{3 \times 2} &\longrightarrow \mathbb{R}^{3 \times 4}, \\ (W_1, W_2) &\longmapsto W_2 W_1.\end{aligned}$$

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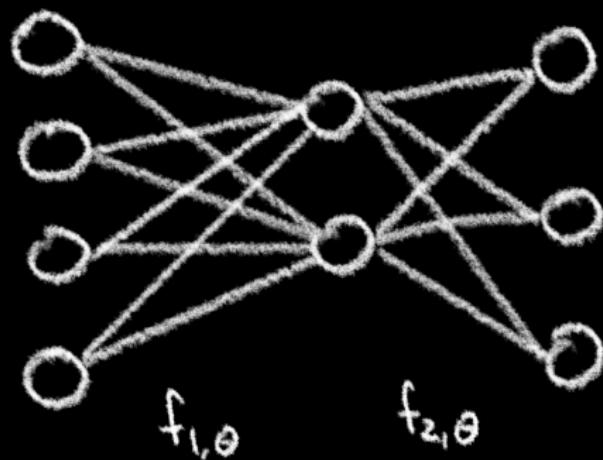
$$\mu : \mathbb{R}^{2 \times 4} \times \mathbb{R}^{3 \times 2} \longrightarrow \mathbb{R}^{3 \times 4},$$
$$(W_1, W_2) \longmapsto W_2 W_1.$$

Its **function space** is

$$\mathcal{M}_2 = \{W \in \mathbb{R}^{3 \times 4} \mid \text{rank}(W) \leq 2\}.$$

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In general: $\mu : \mathbb{R}^{k_1 \times k_0} \times \mathbb{R}^{k_2 \times k_1} \times \dots \times \mathbb{R}^{k_L \times k_{L-1}} \longrightarrow \mathbb{R}^{k_L \times k_0},$
 $(W_1, W_2, \dots, W_L) \longmapsto W_L \cdots W_2 W_1.$

Its function space $\mathcal{M}_r = \text{im}(\mu) = \{W \in \mathbb{R}^{k_L \times k_0} \mid \text{rank}(W) \leq r\}$, where $r := \min(k_0, \dots, k_L)$, is an algebraic variety.

Linear Group-Equivariant Networks

Running Example

Consider an autoencoder $\mu : \mathbb{R}^{2 \times 9} \times \mathbb{R}^{9 \times 2} \longrightarrow \mathbb{R}^{9 \times 9}$, $(W_1, W_2) \longmapsto W_2 W_1$

Linear Group-Equivariant Networks

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| | | |
|----------|----------|----------|
| a_{11} | a_{12} | a_{13} |
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Its inputs and outputs are 3×3 images:

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$$\in \mathbb{R}^9.$$

Consider the clockwise rotation by 90° :

$$\sigma : \mathbb{R}^9 \longrightarrow \mathbb{R}^9,$$

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Which are invariant?

example cont'd

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is represented by the permutation matrix

$$P_\sigma = \left[\begin{array}{cccc|ccc|c} 0 & 0 & 0 & 1 & & & & \\ 1 & 0 & 0 & 0 & & 0 & & 0 \\ 0 & 1 & 0 & 0 & & & & \\ 0 & 0 & 1 & 0 & & & & \\ \hline & & & & 0 & 0 & 0 & 1 \\ & & & & 1 & 0 & 0 & 0 & 0 \\ 0 & & & & 0 & 1 & 0 & 0 & \\ 0 & & & & 0 & 0 & 1 & 0 & 1 \end{array} \right]$$

example cont'd

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$$\begin{array}{|ccc|} \hline a_{11} & a_{12} & a_{13} \\ \hline a_{21} & a_{22} & a_{23} \\ \hline a_{31} & a_{32} & a_{33} \\ \hline \end{array} \longmapsto \begin{array}{|ccc|} \hline a_{31} & a_{21} & a_{11} \\ \hline a_{32} & a_{22} & a_{12} \\ \hline a_{33} & a_{23} & a_{13} \\ \hline \end{array}$$

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$$W \in \mathbb{R}^{9 \times 9}$$

is equivariant under σ

$$\Leftrightarrow$$

$$W \cdot P_\sigma = P_\sigma \cdot W.$$

example cont'd

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$$W \in \mathbb{R}^{9 \times 9}$$

is invariant under σ

$$\Leftrightarrow$$

$$W \cdot P_\sigma = W.$$

example cont'd

$W \in \mathbb{R}^{9 \times 9}$ is equivariant under σ iff

$$W = \left[\begin{array}{cccc|cccc|c} \alpha_1 & \alpha_2 & \alpha_3 & \alpha_4 & \beta_1 & \beta_2 & \beta_3 & \beta_4 & \varepsilon_3 \\ \alpha_4 & \alpha_1 & \alpha_2 & \alpha_3 & \beta_4 & \beta_1 & \beta_2 & \beta_3 & \varepsilon_3 \\ \alpha_3 & \alpha_4 & \alpha_1 & \alpha_2 & \beta_3 & \beta_4 & \beta_1 & \beta_2 & \varepsilon_3 \\ \alpha_2 & \alpha_3 & \alpha_4 & \alpha_1 & \beta_2 & \beta_3 & \beta_4 & \beta_1 & \varepsilon_3 \\ \hline \gamma_1 & \gamma_2 & \gamma_3 & \gamma_4 & \delta_1 & \delta_2 & \delta_3 & \delta_4 & \varepsilon_4 \\ \gamma_4 & \gamma_1 & \gamma_2 & \gamma_3 & \delta_4 & \delta_1 & \delta_2 & \delta_3 & \varepsilon_4 \\ \gamma_3 & \gamma_4 & \gamma_1 & \gamma_2 & \delta_3 & \delta_4 & \delta_1 & \delta_2 & \varepsilon_4 \\ \gamma_2 & \gamma_3 & \gamma_4 & \gamma_1 & \delta_2 & \delta_3 & \delta_4 & \delta_1 & \varepsilon_4 \\ \hline \varepsilon_1 & \varepsilon_1 & \varepsilon_1 & \varepsilon_1 & \varepsilon_2 & \varepsilon_2 & \varepsilon_2 & \varepsilon_2 & \varepsilon_5 \end{array} \right].$$

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The linear space \mathcal{E}^σ of σ -equivariant $W \in \mathbb{R}^{9 \times 9}$

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The linear space \mathcal{E}^σ of σ -equivariant $W \in \mathbb{R}^{9 \times 9}$ intersected with the function space $\mathcal{M}_2 = \{W \in \mathbb{R}^{9 \times 9} \mid \text{rank}(W) \leq 2\}$ of our autoencoder is an algebraic variety with

- ◆ 10 irreducible components over \mathbb{C}
- ◆ 4 irreducible components over \mathbb{R}

takeaway message

There is **no** neural network whose function space is $\mathcal{E}^\sigma \cap \mathcal{M}_2$!

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There is **no** neural network whose function space is $\mathcal{E}^\sigma \cap \mathcal{M}_2$!

Any neural network can parametrize at most one of the real irreducible components of $\mathcal{E}^\sigma \cap \mathcal{M}_2$.

example cont'd

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The linear space \mathcal{I}^σ of σ -invariant $W \in \mathbb{R}^{9 \times 9}$ intersected with the function space $\mathcal{M}_2 = \{W \in \mathbb{R}^{9 \times 9} \mid \text{rank}(W) \leq 2\}$ is an irreducible algebraic variety $\cong \{A \in \mathbb{R}^{9 \times 3} \mid \text{rank}(A) \leq 2\}$.

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The set of G -invariant $W \in \mathbb{R}^{m \times n}$ is \mathcal{I}^σ for some $\sigma \in \mathcal{S}_n$.

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The set of G -invariant $W \in \mathbb{R}^{m \times n}$ is \mathcal{I}^σ for some $\sigma \in \mathcal{S}_n$.

What are **all ways to parametrize** $\mathcal{I}^\sigma \cap \mathcal{M}_r$ **with autoencoders?**

Invariance

Consider $\mathcal{M}_r = \{W \in \mathbb{R}^{m \times n} \mid \text{rank}(W) \leq r\}$ and $\sigma \in \mathcal{S}_n$.

Decompose $\sigma = \pi_1 \circ \pi_2 \circ \dots \circ \pi_k$ into disjoint cycles.

Lemma: The linear space \mathcal{I}^σ of σ -invariant $W \in \mathbb{R}^{m \times n}$ consists of all matrices W whose columns indexed by π_i are equal, for all $i = 1, 2, \dots, k$. Hence, $\mathcal{I}^\sigma \cap \mathcal{M}_r \cong \{W \in \mathbb{R}^{m \times k} \mid \text{rank}(W) \leq r\}$ is an irreducible variety.

Lemma: Let $G \subset \mathcal{S}_n$.

The set of G -invariant $W \in \mathbb{R}^{m \times n}$ is \mathcal{I}^σ for some $\sigma \in \mathcal{S}_n$.

What are **all ways to parametrize** $\mathcal{I}^\sigma \cap \mathcal{M}_r$ **with autoencoders?**

Lemma: $\{(A, B) \in \mathbb{R}^{m \times k} \times \mathbb{R}^{k \times n} \mid \text{rank}(AB) = k, AB \in \mathcal{I}^\sigma\} =$

Invariance

Consider $\mathcal{M}_r = \{W \in \mathbb{R}^{m \times n} \mid \text{rank}(W) \leq r\}$ and $\sigma \in \mathcal{S}_n$.

Decompose $\sigma = \pi_1 \circ \pi_2 \circ \dots \circ \pi_k$ into disjoint cycles.

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The set of G -invariant $W \in \mathbb{R}^{m \times n}$ is \mathcal{I}^σ for some $\sigma \in \mathcal{S}_n$.

What are **all ways to parametrize** $\mathcal{I}^\sigma \cap \mathcal{M}_r$ **with autoencoders?**

Lemma: $\{(A, B) \in \mathbb{R}^{m \times k} \times \mathbb{R}^{k \times n} \mid \text{rank}(AB) = k, AB \in \mathcal{I}^\sigma\} = \{A \in \mathbb{R}^{m \times k} \mid \text{rank}(A) = k\} \times \{B \in \mathbb{R}^{k \times n} \mid \text{columns indexed by } \pi_i \text{ are equal}\}$
 $\Rightarrow \sigma$ induces weight sharing on the encoder!

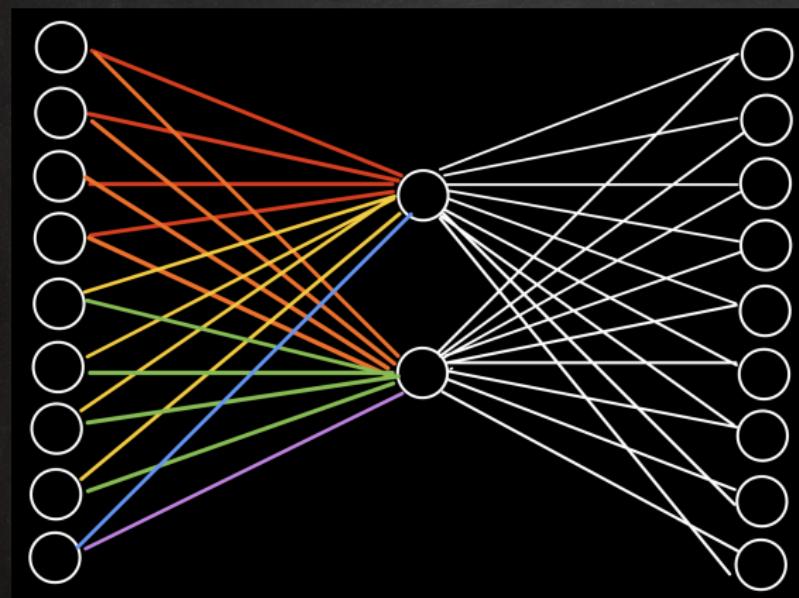
running example

$$\sigma : \mathbb{R}^9 \longrightarrow \mathbb{R}^9,$$

| | | |
|----------|----------|----------|
| a_{11} | a_{12} | a_{13} |
| a_{21} | a_{22} | a_{23} |
| a_{31} | a_{32} | a_{33} |

\longmapsto

| | | |
|----------|----------|----------|
| a_{31} | a_{21} | a_{11} |
| a_{32} | a_{22} | a_{12} |
| a_{33} | a_{23} | a_{13} |



has function space $\mathcal{I}^\sigma \cap \mathcal{M}_2$

Equivariance

Consider $\mathcal{M}_r = \{W \in \mathbb{R}^{n \times n} \mid \text{rank}(W) \leq r\}$ and $\sigma \in \mathcal{S}_n$ represented by
 $P_\sigma \in \mathbb{R}^{n \times n}$.

Equivariance

Consider $\mathcal{M}_r = \{W \in \mathbb{R}^{n \times n} \mid \text{rank}(W) \leq r\}$ and $\sigma \in S_n$ represented by

$$P_\sigma \in \mathbb{R}^{n \times n}.$$

Idea: Let $T \in \text{GL}_n$.

W is P_σ -equivariant iff $T^{-1}WT$ is $T^{-1}P_\sigma T$ -equivariant.

Equivariance

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$$P_\sigma \in \mathbb{R}^{n \times n}.$$

Idea: Let $T \in \text{GL}_n$.

W is P_σ -equivariant iff $T^{-1}WT$ is $T^{-1}P_\sigma T$ -equivariant.

This base change also preserves rank!

running example

$$\sigma : \mathbb{R}^9 \longrightarrow \mathbb{R}^9, \quad \begin{array}{|c|c|c|} \hline a_{11} & a_{12} & a_{13} \\ \hline a_{21} & a_{22} & a_{23} \\ \hline a_{31} & a_{32} & a_{33} \\ \hline \end{array} \longmapsto \begin{array}{|c|c|c|} \hline a_{31} & a_{21} & a_{11} \\ \hline a_{32} & a_{22} & a_{12} \\ \hline a_{33} & a_{23} & a_{13} \\ \hline \end{array}$$

$$P = P_\sigma$$

P -equivariant matrices

$$\left[\begin{array}{cccc|cc} 0 & 0 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ \hline & & & & 0 & 0 \\ 0 & & & & 0 & 0 \\ 0 & & & & 0 & 0 \\ 0 & & & & 0 & 0 \\ \hline 0 & & & & 0 & 1 \end{array} \right]$$

$$\left[\begin{array}{cccc|cccc|c} \alpha_1 & \alpha_2 & \alpha_3 & \alpha_4 & \beta_1 & \beta_2 & \beta_3 & \beta_4 & \varepsilon_3 \\ \alpha_4 & \alpha_1 & \alpha_2 & \alpha_3 & \beta_4 & \beta_1 & \beta_2 & \beta_3 & \varepsilon_3 \\ \alpha_3 & \alpha_4 & \alpha_1 & \alpha_2 & \beta_3 & \beta_4 & \beta_1 & \beta_2 & \varepsilon_3 \\ \alpha_2 & \alpha_3 & \alpha_4 & \alpha_1 & \beta_2 & \beta_3 & \beta_4 & \beta_1 & \varepsilon_3 \\ \hline \gamma_1 & \gamma_2 & \gamma_3 & \gamma_4 & \delta_1 & \delta_2 & \delta_3 & \delta_4 & \varepsilon_4 \\ \gamma_4 & \gamma_1 & \gamma_2 & \gamma_3 & \delta_4 & \delta_1 & \delta_2 & \delta_3 & \varepsilon_4 \\ \gamma_3 & \gamma_4 & \gamma_1 & \gamma_2 & \delta_3 & \delta_4 & \delta_1 & \delta_2 & \varepsilon_4 \\ \gamma_2 & \gamma_3 & \gamma_4 & \gamma_1 & \delta_2 & \delta_3 & \delta_4 & \delta_1 & \varepsilon_4 \\ \hline \varepsilon_1 & \varepsilon_1 & \varepsilon_1 & \varepsilon_1 & \varepsilon_2 & \varepsilon_2 & \varepsilon_2 & \varepsilon_2 & \varepsilon_5 \end{array} \right]$$

running example

$$\sigma : \mathbb{R}^9 \longrightarrow \mathbb{R}^9, \quad \begin{array}{|c|c|c|} \hline a_{11} & a_{12} & a_{13} \\ \hline a_{21} & a_{22} & a_{23} \\ \hline a_{31} & a_{32} & a_{33} \\ \hline \end{array} \longmapsto \begin{array}{|c|c|c|} \hline a_{31} & a_{21} & a_{11} \\ \hline a_{32} & a_{22} & a_{12} \\ \hline a_{33} & a_{23} & a_{13} \\ \hline \end{array}$$

$P = \text{diagonalization of } P_\sigma$

$$\left[\begin{array}{cccc|cc} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & i & 0 & 0 & 0 & 0 \\ 0 & 0 & -1 & 0 & 0 & 0 \\ 0 & 0 & 0 & -i & 0 & 0 \end{array} \right]$$

P -equivariant matrices

$$\left[\begin{array}{cccc|cccc|c} a_{11} & 0 & 0 & 0 & a_{12} & 0 & 0 & 0 & 0 & a_{13} \\ 0 & c_{11} & 0 & 0 & 0 & c_{12} & 0 & 0 & 0 & 0 \\ 0 & 0 & b_{11} & 0 & 0 & 0 & b_{12} & 0 & 0 & 0 \\ 0 & 0 & 0 & d_{11} & 0 & 0 & 0 & d_{12} & 0 & 0 \\ \hline a_{21} & 0 & 0 & 0 & a_{22} & 0 & 0 & 0 & 0 & a_{23} \\ 0 & c_{21} & 0 & 0 & 0 & c_{22} & 0 & 0 & 0 & 0 \\ 0 & 0 & b_{21} & 0 & 0 & 0 & b_{22} & 0 & 0 & 0 \\ 0 & 0 & 0 & d_{21} & 0 & 0 & 0 & d_{22} & 0 & 0 \\ \hline a_{31} & 0 & 0 & 0 & a_{32} & 0 & 0 & 0 & 0 & a_{33} \end{array} \right]$$

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P -equivariant matrices

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running example

$$W = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \\ & b_{11} & b_{12} \\ & b_{21} & b_{22} \\ & c_{11} & c_{12} \\ & c_{21} & c_{22} \\ & d_{11} & d_{12} \\ & d_{21} & d_{22} \end{bmatrix}$$

There are 10 ways how W can have rank 2:

running example

$$W = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \\ & b_{11} & b_{12} \\ & b_{21} & b_{22} \\ & c_{11} & c_{12} \\ & c_{21} & c_{22} \\ & d_{11} & d_{12} \\ & d_{21} & d_{22} \end{bmatrix}$$

There are 10 ways how W can have rank 2:

- ◆ One of the diagonal blocks has rank 2;
other blocks are 0

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There are 10 ways how W can have rank 2:

- ◆ One of the diagonal blocks has rank 2; \rightsquigarrow 4 components of $\mathcal{E}^\sigma \cap \mathcal{M}_2$
other blocks are 0

running example

$$W = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \\ & b_{11} & b_{12} \\ & b_{21} & b_{22} \\ & c_{11} & c_{12} \\ & c_{21} & c_{22} \\ & d_{11} & d_{12} \\ & d_{21} & d_{22} \end{bmatrix}$$

There are 10 ways how W can have rank 2:

- ◆ One of the diagonal blocks has rank 2; \rightsquigarrow 4 components of $\mathcal{E}^\sigma \cap \mathcal{M}_2$
other blocks are 0
- ◆ Two distinct blocks have rank 1;
other blocks are 0

running example

$$W = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \\ & b_{11} & b_{12} \\ & b_{21} & b_{22} \\ & c_{11} & c_{12} \\ & c_{21} & c_{22} \\ & d_{11} & d_{12} \\ & d_{21} & d_{22} \end{bmatrix}$$

There are 10 ways how W can have rank 2:

- ◆ One of the diagonal blocks has rank 2;
other blocks are 0 \rightsquigarrow 4 components of $\mathcal{E}^\sigma \cap \mathcal{M}_2$
- ◆ Two distinct blocks have rank 1;
other blocks are 0 \rightsquigarrow 6 components of $\mathcal{E}^\sigma \cap \mathcal{M}_2$

Equivariance

Consider $\mathcal{M}_r = \{W \in \mathbb{R}^{n \times n} \mid \text{rank}(W) \leq r\}$ and $\sigma \in \mathcal{S}_n$ represented by $P_\sigma \in \mathbb{R}^{n \times n}$.

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 Chop W into blocks following the same pattern!

$$P_\sigma = \left[\begin{array}{cccc|cc} 0 & 0 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ \hline & & & & 0 & 0 \\ & & & & 0 & 0 \\ 0 & & & & 0 & 0 \\ & & & & 0 & 0 \\ \hline 0 & & & & 0 & 1 \end{array} \right]$$

$$W = \left[\begin{array}{cccc|cc} \alpha_1 & \alpha_2 & \alpha_3 & \alpha_4 & \beta_1 & \beta_2 & \beta_3 & \beta_4 & \varepsilon_3 \\ \alpha_4 & \alpha_1 & \alpha_2 & \alpha_3 & \beta_4 & \beta_1 & \beta_2 & \beta_3 & \varepsilon_3 \\ \alpha_3 & \alpha_4 & \alpha_1 & \alpha_2 & \beta_3 & \beta_4 & \beta_1 & \beta_2 & \varepsilon_3 \\ \alpha_2 & \alpha_3 & \alpha_4 & \alpha_1 & \beta_2 & \beta_3 & \beta_4 & \beta_1 & \varepsilon_3 \\ \hline \gamma_1 & \gamma_2 & \gamma_3 & \gamma_4 & \delta_1 & \delta_2 & \delta_3 & \delta_4 & \varepsilon_4 \\ \gamma_4 & \gamma_1 & \gamma_2 & \gamma_3 & \delta_4 & \delta_1 & \delta_2 & \delta_3 & \varepsilon_4 \\ \gamma_3 & \gamma_4 & \gamma_1 & \gamma_2 & \delta_3 & \delta_4 & \delta_1 & \delta_2 & \varepsilon_4 \\ \gamma_2 & \gamma_3 & \gamma_4 & \gamma_1 & \delta_2 & \delta_3 & \delta_4 & \delta_1 & \varepsilon_4 \\ \hline \varepsilon_1 & \varepsilon_1 & \varepsilon_1 & \varepsilon_1 & \varepsilon_2 & \varepsilon_2 & \varepsilon_2 & \varepsilon_2 & \varepsilon_5 \end{array} \right]$$

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Lemma: A matrix $W \in \mathbb{R}^{n \times n}$ is σ -equivariant iff
 each block is a (possibly non-square) circulant matrix.

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 Chop W into blocks following the same pattern!

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Lemma: A matrix $W \in \mathbb{R}^{n \times n}$ is σ -equivariant iff
 each block is a (possibly non-square) circulant matrix.

$$\begin{bmatrix} a & a & a \\ a & a & a \\ a & a & a \\ a & a & a \end{bmatrix}, \quad \begin{bmatrix} a & a & a \\ a & a & a \\ a & a & a \end{bmatrix}, \quad \begin{bmatrix} a & b & a & b \\ b & a & b & a \end{bmatrix}, \quad \dots$$

Equivariance

Consider $\mathcal{M}_r = \{W \in \mathbb{R}^{n \times n} \mid \text{rank}(W) \leq r\}$ and $\sigma \in \mathcal{S}_n$.

Decompose $\sigma = \pi_1 \circ \dots \circ \pi_k$ into disjoint cycles and let $\ell_j := \text{length}(\pi_j)$.

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Diagonalize P_σ and sort the eigenvalues. This yields the diagonal matrix P .

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Lemma: A matrix $W \in \mathbb{R}^{n \times n}$ is P -equivariant iff its block diagonal with $\#(\mathbb{Z}/m\mathbb{Z})^\times$ many blocks of size $d_m \times d_m$, where $d_m := \#\{j \text{ such that } m|\ell_j\}$.

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$$P = \begin{bmatrix} 1 & & & & \\ & 1 & & & \\ & & 1 & & \\ & & & -1 & \\ & & & & -1 \\ & & & & i \\ & & & & & i \\ & & & & & & -i \\ & & & & & & & -i \end{bmatrix} \quad W = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \\ & b_{11} & b_{12} \\ & b_{21} & b_{22} \\ & c_{11} & c_{12} \\ & c_{21} & c_{22} \\ d_{11} & d_{12} \\ d_{21} & d_{22} \end{bmatrix}$$

$$\ell_1 = 4, \ell_2 = 4, \ell_3 = 1$$

Equivariance

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$$P = \begin{bmatrix} 1 & & & & \\ & 1 & & & \\ & & 1 & & \\ & & & -1 & \\ & & & & -1 \\ & & & & & i \\ & & & & & & i \\ & & & & & & & -i \\ & & & & & & & & -i \end{bmatrix} \quad W = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \\ & & b_{11} & b_{12} \\ & & b_{21} & b_{22} \\ & & c_{11} & c_{12} \\ & & c_{21} & c_{22} \\ & & d_{11} & d_{12} \\ & & d_{21} & d_{22} \end{bmatrix}$$

$$\ell_1 = 4, \ell_2 = 4, \ell_3 = 1$$

$$d_1 = 3, d_2 = 2, d_3 = 0, d_4 = 2, d_5 = 0, \dots$$

Equivariance

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$$P = \begin{bmatrix} 1 & & & & \\ & 1 & & & \\ & & 1 & & \\ & & & -1 & \\ & & & & -1 \\ & & & & i \\ & & & & & i \\ & & & & & & -i \\ & & & & & & & -i \end{bmatrix} \quad W = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \\ & & b_{11} & b_{12} \\ & & b_{21} & b_{22} \\ & & c_{11} & c_{12} \\ & & c_{21} & c_{22} \\ & & d_{11} & d_{12} \\ & & d_{21} & d_{22} \end{bmatrix}$$

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$$\#(\mathbb{Z}/1\mathbb{Z})^\times = 1, \#(\mathbb{Z}/2\mathbb{Z})^\times = 1, \#(\mathbb{Z}/4\mathbb{Z})^\times = 2$$

Equivariance

Consider $\mathcal{M}_r = \{W \in \mathbb{R}^{n \times n} \mid \text{rank}(W) \leq r\}$ and $\sigma \in \mathcal{S}_n$.

Decompose $\sigma = \pi_1 \circ \dots \circ \pi_k$ into disjoint cycles and let $\ell_j := \text{length}(\pi_j)$.

Diagonalize P_σ and sort the eigenvalues. This yields the diagonal matrix P .

Lemma: A matrix $W \in \mathbb{R}^{n \times n}$ is P -equivariant iff its block diagonal with $\#(\mathbb{Z}/m\mathbb{Z})^\times$ many blocks of size $d_m \times d_m$, where $d_m := \#\{j \text{ such that } m|\ell_j\}$.

Theorem: The irreducible components of $\mathcal{E}^\sigma \cap \mathcal{M}_r$ over \mathbb{C} are in 1-to-1 correspondence with the integer solutions $(r_{m,u})$ of

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Example: to diagonalize

$$\begin{bmatrix} 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

, use base change

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & i & -1 & -i \\ 1 & -1 & 1 & -1 \\ 1 & -i & -1 & i \end{bmatrix}$$

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$$\rightsquigarrow \frac{1}{2} \begin{bmatrix} 1 & \sqrt{2} & 1 & 0 \\ 1 & 0 & -1 & -\sqrt{2} \\ 1 & -\sqrt{2} & 1 & 0 \\ 1 & 0 & -1 & \sqrt{2} \end{bmatrix} \in O_4(\mathbb{R})$$

running example

$$\sigma : \mathbb{R}^9 \longrightarrow \mathbb{R}^9, \quad \begin{array}{|c|c|c|} \hline a_{11} & a_{12} & a_{13} \\ \hline a_{21} & a_{22} & a_{23} \\ \hline a_{31} & a_{32} & a_{33} \\ \hline \end{array} \longmapsto \begin{array}{|c|c|c|} \hline a_{31} & a_{21} & a_{11} \\ \hline a_{32} & a_{22} & a_{12} \\ \hline a_{33} & a_{23} & a_{13} \\ \hline \end{array}$$

$P = P_\sigma$ after $O_9(\mathbb{R})$ -base change

P -equivariant matrices

$$\left[\begin{array}{ccc|cc} 1 & & & & \\ & 1 & & & \\ & & 1 & & \\ & & & -1 & \\ & & & & -1 \\ \hline & & & 0 & 1 \\ & & & -1 & 0 \\ & & & & 0 & 1 \\ & & & & & -1 & 0 \end{array} \right] \quad \left[\begin{array}{cccc} a_{11} & a_{12} & a_{13} & \\ a_{21} & a_{22} & a_{23} & \\ a_{31} & a_{32} & a_{33} & \\ \hline b_{11} & b_{12} & & \\ b_{21} & b_{22} & & \\ \hline c_1 & -c_2 & d_1 & -d_2 \\ c_2 & c_1 & d_2 & d_1 \\ e_1 & -e_2 & f_1 & -f_2 \\ e_2 & e_1 & f_2 & f_1 \end{array} \right]$$

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There are **4** ways how W can have rank 2:

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In general: After the $O_n(\mathbb{R})$ -base change, the σ -equivariant matrices become block diagonal:

- ◆ at most 2 blocks are arbitrary (corresponding to eigenvalues ± 1 of P_σ);
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For $z = a + ib \in \mathbb{C}$, define $\mathcal{R}(z) := \begin{bmatrix} a & -b \\ b & a \end{bmatrix}$.

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$$\text{rank}(\mathcal{R}(M)) = 2 \cdot \text{rank}(M)$$

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$$\begin{aligned} & \cong \{A \in \mathbb{R}^{d_1 \times d_1} \mid \text{rank}(A) \leq r_{1,1}\} \times \{A \in \mathbb{R}^{d_2 \times d_2} \mid \text{rank}(A) \leq r_{2,1}\} \\ & \times \prod_{m>2} \prod_{\substack{u \in (\mathbb{Z}/m\mathbb{Z})^\times, \\ \frac{1}{2} < \frac{u}{m} < 1}} \mathcal{R}(\{A \in \mathbb{C}^{d_m \times d_m} \mid \text{rank}(A) \leq r_{m,u}\}). \end{aligned}$$

Which of these 4 components is best ??

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Parametrizing equivariant functions with autoencoders

There is **no** neural network whose function space is $\mathcal{E}^\sigma \cap \mathcal{M}_2$!

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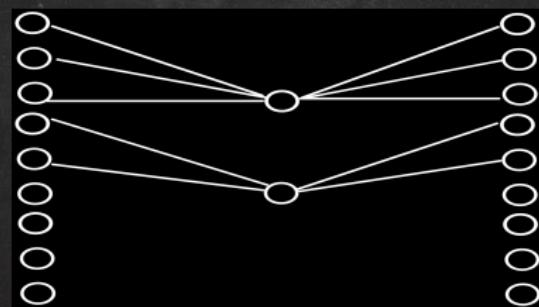
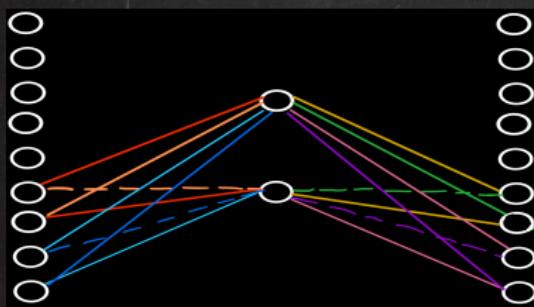
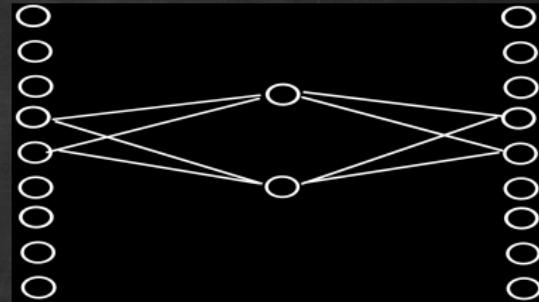
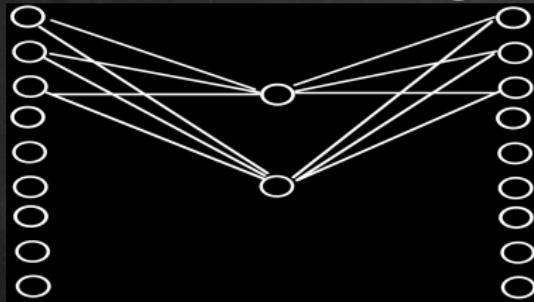
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But: Each irreducible component of $\mathcal{E}^\sigma \cap \mathcal{M}_2$ is the function space of an autoencoder.

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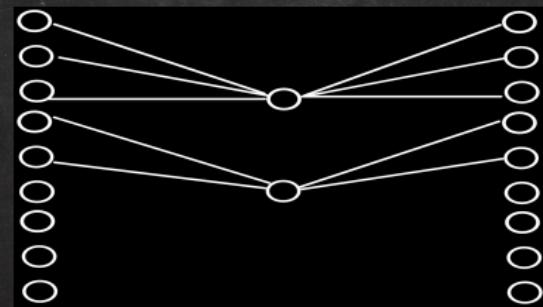
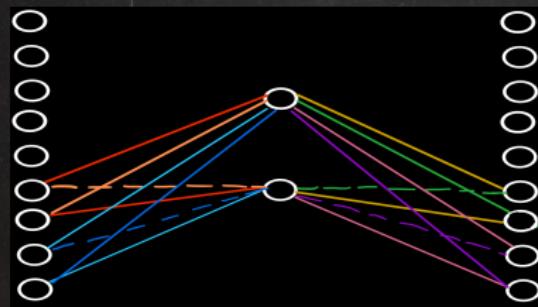
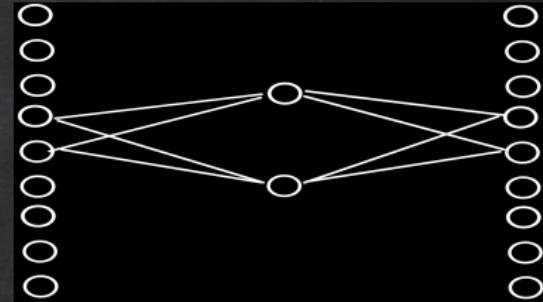
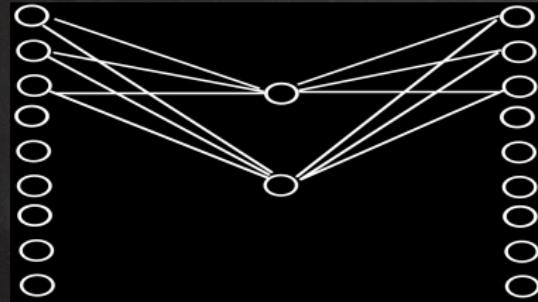
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This works in general for $\mathcal{E}^\sigma \cap \mathcal{M}_r \subset \mathbb{R}^{n \times n}$!

Euclidean distance optimization

Consider a function space $\mathcal{M} \subset \mathbb{R}^{m \times n}$. Given training data $X \in \mathbb{R}^{n \times d}$ and $Y \in \mathbb{R}^{m \times d}$, the **squared-error loss** is

$$\mathcal{M} \rightarrow \mathbb{R}, \quad W \mapsto \|WX - Y\|_F^2.$$

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Lemma: If $\text{rank}(XX^\top) = n$ (which holds for a sufficient amount of training data that is sufficiently generic), minimizing the squared-error loss is equivalent to minimizing the weighted Euclidean distance

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Now let's assume that XX^\top is (close to) a multiple of the identity, and \mathcal{M} is an irreducible component of $\mathcal{E}^\sigma \cap \mathcal{M}_r$.

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Orthogonal base changes do not affect the standard Euclidean distance!
Hence, our task is

$$\min_{\tilde{W} \in \tilde{\mathcal{M}}} \|\tilde{W} - \tilde{U}\|_F^2, \quad (1)$$

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Euclidean distance minimization on these blocks typically has a unique local minimum, easily found by SVD (Eckart-Young theorem)!

Data science requires us to rethink the schism between mathematical disciplines!

differential geometry ⇒
algebraic geometry ⇒
data science ⇒



Metric Algebraic Geometry



Historical Snapshot

Polars
Foci
Envelopes

Curvature

Plane Curves
Algebraic Varieties
Volumes of Tubular Neighborhoods

Maximum Likelihood

Kullback–Leibler Divergence
Maximum Likelihood Degree
Scattering Equations
Gaussian Models

Critical Equations

Euclidean Distance Degree
Low-Rank Matrix Approximation
Invitation to Polar Degrees

Reach and Offset

Medial Axis and Bottlenecks
Offset Hypersurfaces
Offset Discriminant

Tensors

Tensors and their Rank
Eigenvectors and Singular Vectors
Volumes of Rank-One Varieties

Computations

Gröbner Bases
Parameter Continuation Theorem
Polynomial Homotopy Continuation

Voronoi Cells

Voronoi Basics
Algebraic Boundaries
Degree Formulas
Voronoi meets Eckart–Young

Computer Vision

Multiview Varieties
Grassmann Tensors
3D Reconstruction from Unknown Cameras

Polar Degrees

Polar Varieties
Projective Duality
Chern Classes

Condition Numbers

Errors in Numerical Computations
Matrix Inversion and Eckart–Young
Condition Number Theorems
Distance to the Discriminant

Volumes of Semialgebraic Sets

Calculus and Beyond
D-Modules
SDP Hierarchies

Wasserstein Distance

Polyhedral Norms
Optimal Transport & Independence Models
Wasserstein meets Segre–Veronese

Machine Learning

Neural Networks
Convolutional Networks
Learning Varieties

Sampling

Homology from Finite Samples
Sampling with Density Guarantees
Markov Chains on Varieties
Chow goes to Monte Carlo

Open PhD and Postdoc Positions!

- ◆ **PhD position in Algebraic Geometry & Computer Vision**
<https://kathlenkohn.github.io/phd>
- ◆ **PhD position in Geometric Combinatorics**
with Katharina Jochemko
- ◆ **Postdoc position in Algebraic Geometry applied to Machine Learning & Computer Vision**
<https://kathlenkohn.github.io/postdoc>
- ◆ **Researcher position in Graphical Models and Algebraic Statistics**
with Liam Solus