

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

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AUTONOMOUS SYSTEMS
AND SOFTWARE PROGRAM

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:

network architecture

activation

network structure

loss

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network architecture		loss
activation	network structure	
identity		
ReLU		
polynomial		

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network architecture		loss
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identity	fully-connected	
ReLU	convolutional	
polynomial	group equivariant	

Algebra & Geometry \Rightarrow Neural Network Theory

Algebraic settings:

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identity	fully-connected	squared-error loss	= Euclidean dist
ReLU	convolutional	Wasserstein distance	= polyhedral dist.
polynomial	group equivariant	cross-entropy	\approx KL divergence

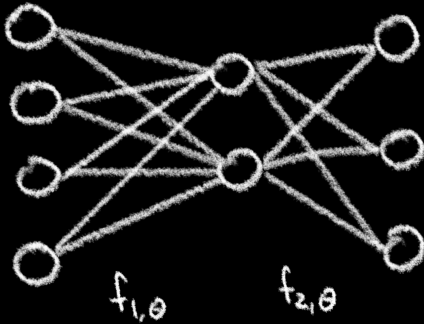
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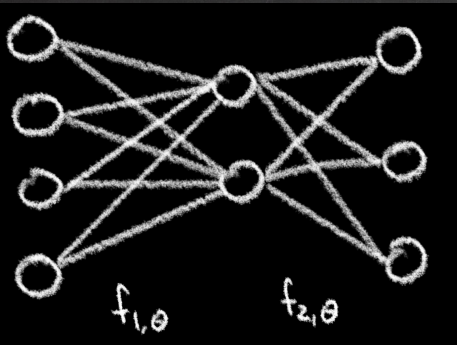
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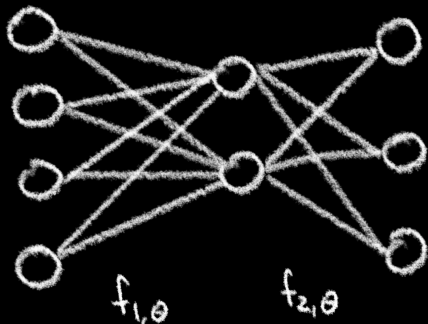


This network parametrizes linear maps:

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

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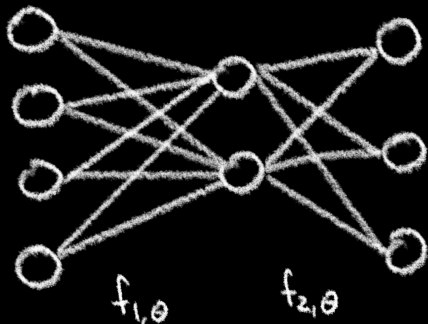
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Its **function space** is

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

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In general: $\mu : \mathbb{R}^{k_1 \times k_0} \times \mathbb{R}^{k_2 \times k_1} \times \dots \times \mathbb{R}^{k_L \times k_{L-1}} \longrightarrow \mathbb{R}^{k_L \times k_0},$

$$(W_1, W_2, \dots, W_L) \longmapsto W_L \dots W_2 W_1.$$

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

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Its inputs and outputs are 3×3 images:

a_{11}	a_{12}	a_{13}
a_{21}	a_{22}	a_{23}
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 $\in \mathbb{R}^9$.

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Which are invariant?

example cont'd

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is represented by the permutation matrix

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

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$$\Leftrightarrow$$

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

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$W \in \mathbb{R}^{9 \times 9}$
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$$\Leftrightarrow W \cdot P_\sigma = P_\sigma \cdot W.$$

$W \in \mathbb{R}^{9 \times 9}$
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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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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What are **all** ways to parametrize $\mathcal{I}^\sigma \cap \mathcal{M}_r$ with autoencoders?

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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.

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 $G \subset \mathcal{S}_n$.

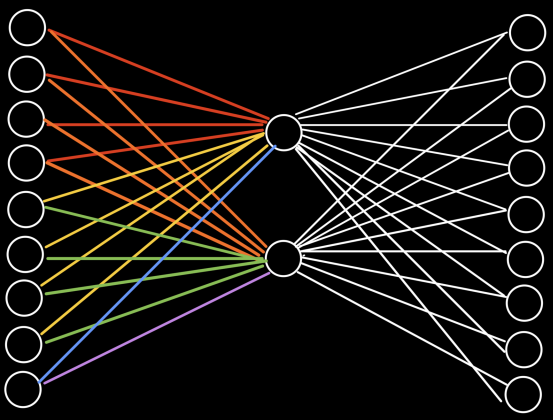
The set of G -invariant $W \in \mathbb{R}^{m \times n}$ is \mathcal{I}^σ for some $\sigma \in \mathcal{S}_n$.

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

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

running example

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



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 $\sigma \in \mathcal{S}_n$ 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

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

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

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

$$P = P_\sigma$$

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

$$P\text{-equivariant matrices}$$

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

running example

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

$P =$ diagonalization of P_σ

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

P -equivariant matrices

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

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

P -equivariant matrices

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

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

running example

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There are 10 ways how W can have rank 2:

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other blocks are 0

running example

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

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

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

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

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

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

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

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 $d_m := \#\{j \text{ such that } m \mid \ell_j\}$.

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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

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

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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

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

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

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

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

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

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 $\ell_j := \text{length}(\pi_j)$.

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$$\cong \prod_{m \in \mathbb{Z}_{>0}} \prod_{u \in (\mathbb{Z}/m\mathbb{Z})^\times} \{A \in \mathbb{C}^{d_m \times d_m} \mid \text{rank}(A) \leq r_{m,u}\}.$$

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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

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

P -equivariant matrices

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & & & & & & \\ a_{21} & a_{22} & a_{23} & & & & & & \\ a_{31} & a_{32} & a_{33} & & & & & & \\ & & & b_{11} & b_{12} & & & & \\ & & & b_{21} & b_{22} & & & & \\ & & & & & c_1 & -c_2 & d_1 & -d_2 \\ & & & & & c_2 & c_1 & d_2 & d_1 \\ & & & & & e_1 & -e_2 & f_1 & -f_2 \\ & & & & & e_2 & e_1 & f_2 & f_1 \end{bmatrix}$$

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There are 4 ways how W can have rank 2:

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Equivariance over \mathbb{R}

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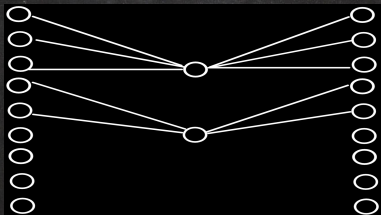
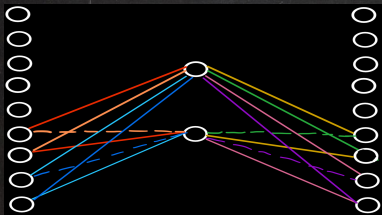
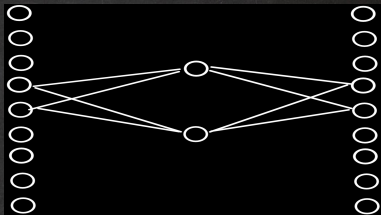
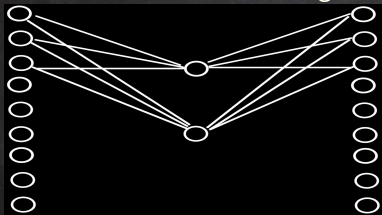
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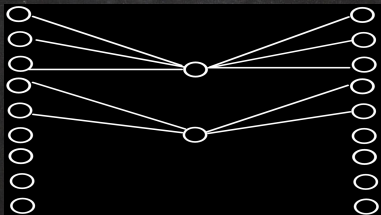
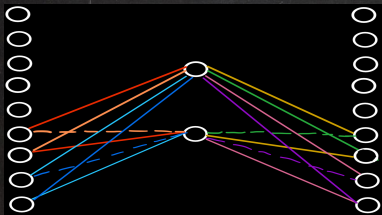
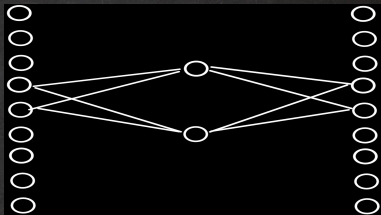
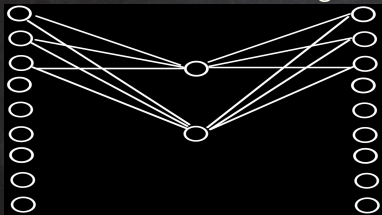
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This works in general for $\mathcal{E}^\sigma \cap \mathcal{M}_r \subset \mathbb{R}^{n \times n}$!

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

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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 \Rightarrow

algebraic geometry \Rightarrow

data science \Rightarrow



Bernd Sturmfels

Kathlén Kohn

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Metric Algebraic Geometry



Historical Snapshot

Polars
Foci
Envelopes

Critical Equations

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

Computations

Gröbner Bases
Parameter Continuation Theorem
Polynomial Homotopy Continuation

Polar Degrees

Polar Varieties
Projective Duality
Chern Classes

Wasserstein Distance

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

Curvature

Plane Curves
Algebraic Varieties
Volumes of Tubular Neighborhoods

Reach and Offset

Medial Axis and Bottlenecks
Offset Hypersurfaces
Offset Discriminant

Voronoi Cells

Voronoi Basics
Algebraic Boundaries
Degree Formulas
Voronoi meets Eckart–Young

Condition Numbers

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

Machine Learning

Neural Networks
Convolutional Networks
Learning Varieties

Maximum Likelihood

Kullback–Leibler Divergence
Maximum Likelihood Degree
Scattering Equations
Gaussian Models

Tensors

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

Computer Vision

Multiview Varieties
Grassmann Tensors
3D Reconstruction from
Unknown Cameras

Volumes of Semialgebraic Sets

Calculus and Beyond
D-Modules
SDP Hierarchies

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