

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

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WASP | WALLENBERG AL.
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

Algebra & Geometry \Rightarrow Neural Network Theory

Algebraic settings:

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:

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

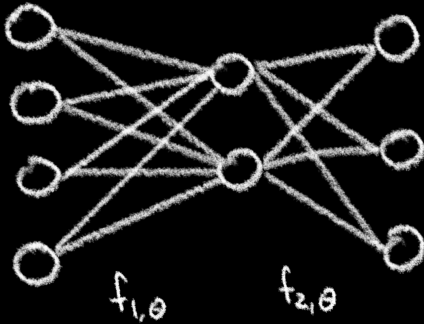
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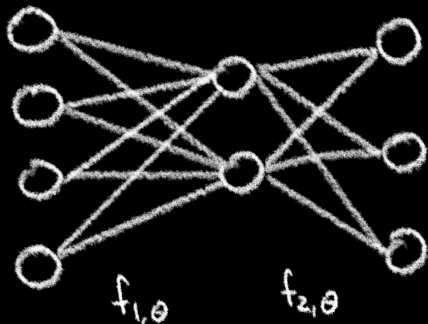
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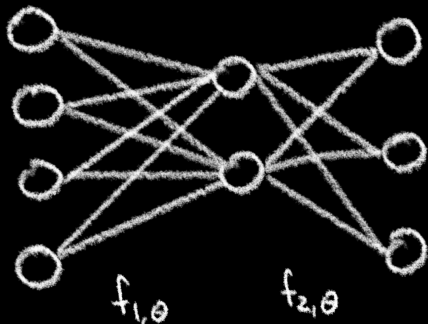


This network parametrizes linear maps:

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

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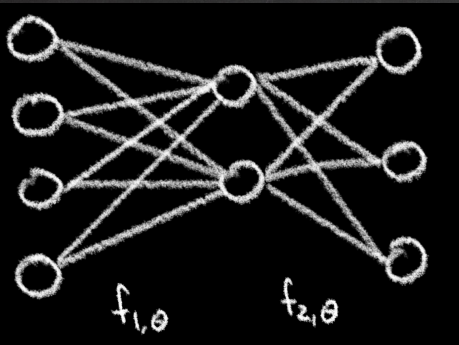
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Its **function space** is

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

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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}
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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

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

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

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

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

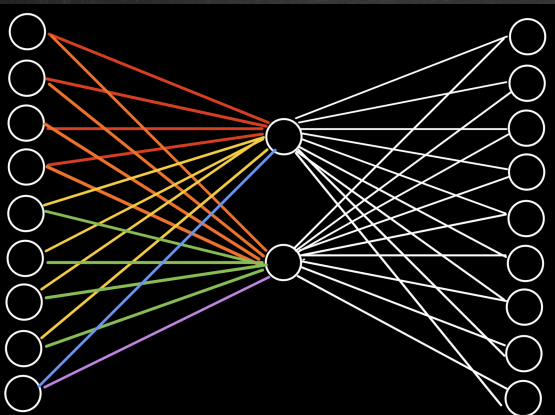
What are all ways to parametrize $\mathcal{I}^\sigma \cap \mathcal{M}_r$ with autoencoders?

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.

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

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

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$$P = P_\sigma$$

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

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

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

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

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

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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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

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

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

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

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

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

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

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

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

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

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

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

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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$$\sum_{m \in \mathbb{Z}_{>0}} \sum_{u \in (\mathbb{Z}/m\mathbb{Z})^\times} r_{m,u} = r, \quad \text{where } 0 \leq r_{m,u} \leq d_m.$$

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The component indexed by $(r_{m,u})$ is

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

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

There are 4 ways how W can have rank 2:

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There are 4 ways how W can have rank 2:

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

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\rightsquigarrow 3 components of $\mathcal{E}^\sigma \cap \mathcal{M}_2$

\rightsquigarrow 1 component of $\mathcal{E}^\sigma \cap \mathcal{M}_2$

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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\Rightarrow we can list all irreducible components of $\mathcal{E}^\sigma \cap \mathcal{M}_r$,
parametrize them via autoencoders,
understand their algebraic properties such as dimension, degree, ...

Which of these 4 components is best ??

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Euclidean distance degree

Lemma: Given a sufficient amount of training data that is sufficiently generic, training a network with function space \mathcal{M} using the **squared-error loss** means to solve an optimization problem of the form

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There are 4 ways how W can have rank 2:

- ◆ One of the diagonal blocks has rank 2;
with EDdeg 3, 1, and 2, respectively.
- ◆ Two first 2 blocks have rank 1;
with EDdeg $3 \cdot 2 = 6$.

\rightsquigarrow 3 components of $\mathcal{E}^\sigma \cap \mathcal{M}_2$

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

differential geometry \Rightarrow

algebraic geometry \Rightarrow

data science \Rightarrow

Paul Breiding, Kathlén Kohn and Bernd Sturmfels

Metric Algebraic Geometry

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