

# Fixing Gauge and Rank Deficiency

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SIAM CONFERENCE ON  
APPLIED LINEAR ALGEBRA



## Equivalent Statements

Fix gauge to fix rank deficiency

Input  $\mathbf{A} \in \mathbb{C}_\rho^{m \times n}$ , construct  $\mathbf{A} \in \mathbb{C}_\rho^{m \times \rho}$

$$\text{Given } \mathbf{A} = \left[ \begin{array}{c|c} \mathbf{U}_{\mathcal{R}} & \mathbf{U}_{\mathcal{N}} \end{array} \right] \left[ \begin{array}{c|c} \mathbf{S} & \mathbf{0} \\ \hline \mathbf{0} & \mathbf{0} \end{array} \right] \left[ \begin{array}{c} \mathbf{V}_{\mathcal{R}}^* \\ \hline \mathbf{V}_{\mathcal{N}}^* \end{array} \right], \dim(\mathbf{V}_{\mathcal{N}}) \rightarrow 0$$

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## Central Questions

What does a gauge function measure? Displacement in superfluous degrees of freedom.

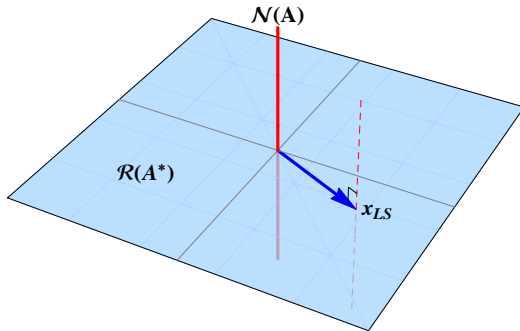
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$$\mathbf{A} \quad x \quad = \quad b$$
$$\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \quad \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad = \quad \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$



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$$x_{soln} \quad = \quad x_{particular} \quad + \quad x_{homogeneous}$$

$$x \quad = \quad \begin{bmatrix} b_1 \\ 0 \end{bmatrix} \quad + \quad \alpha \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

# Electrodynamics: Example

Maxwell's equations

divergence	curl
------------	------

$$\nabla \cdot \mathbf{D} = 4\pi\rho \quad \nabla \times \mathbf{E} = -\frac{1}{c} \frac{\partial \mathbf{B}}{\partial t}$$

$$\nabla \cdot \mathbf{B} = 0 \quad \nabla \times \mathbf{H} = \frac{4\pi}{c} \mathbf{J} + \frac{1}{c} \frac{\partial \mathbf{D}}{\partial t}$$

# Electrodynamics: Vector and Scalar Potentials

$$\mathbf{B} = \nabla \times \mathbf{A}$$

$$\mathbf{E} = -\nabla\phi - \frac{1}{c} \frac{\partial \mathbf{A}}{\partial t}$$

# Electrodynamics: Gauge Fixing Conditions

Coulomb:

$$\nabla \cdot \mathbf{A}(\mathbf{r}, t) = 0$$

Lorenz:

$$\partial^\mu A_\mu = 0$$

or...

$$\nabla \cdot \mathbf{A}(\mathbf{r}, t) + \frac{1}{c^2} \frac{\partial \phi}{\partial t} = 0$$

# Scalar Potentials $\phi$

Sobolev Space:

$$W^{1,2}(\Omega) = \{\phi \in L^2(\Omega) : \partial_x^1 \phi \in L^2(\Omega)\}$$

# Prototype Vector Field Equation

$$\mathbf{E} = -\nabla\phi$$

Inverse problem: Measure  $\mathbf{E}$  find  $\phi$

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## Least Squares: Problem

$$\mathbf{A}x = b$$

- system matrix  $\mathbf{A}: \mathbb{C}^n \mapsto \mathbb{C}^m$
- data vector  $b \in \mathbb{C}^m$

Least squares solution

$$x_{LS} = \left\{ x \in \mathbb{C}^n : \|\mathbf{A}x - b\|_2^2 \text{ is minimized} \right\}$$



# Least Squares: Solution

$$\mathbf{A}x = b$$

$$x_{LS} = \mathbf{A}^\dagger b + \left( \mathbf{I}_n - \mathbf{A}^\dagger \mathbf{A} \right) y, \quad y \in \mathbb{C}^n$$

or...

$$x_{LS} = \mathbf{A}^\dagger b + \mathbf{P}_{\mathcal{R}(\mathbf{A}^*)} y, \quad y \in \mathbb{C}^n$$

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# Least Squares: Invariance

$$x_* = \mathbf{A}^\dagger b$$

$$x_* \rightarrow x_* + \mathbf{P}_{\mathcal{R}(\mathbf{A}^*)} y$$

$$\mathbf{A}(x_*) = \mathbf{A}(x_* + \mathbf{P}_{\mathcal{R}(\mathbf{A}^*)} y)$$

# Approximating Measurement

- ① Select **modes**: basis functions  $\{g_\nu(x)\}_{\nu=1}^n$
- ② Find amplitudes  $a$  to describe measured function  $f(x)$

$$f(x) \approx a_1 g_1(x) + a_2 g_2(x) + a_3 g_3(x) + \cdots = \sum_{\nu=1}^n a_\nu g_\nu(x)$$

# Merit Function

$$M(a) = \sum_{k=1}^m r_k^2$$

$$r_k = f(x_k) - \sum_{\nu=1}^n a_{\nu} g_{\nu}(x_k)$$

$$M(a) = \sum_{k=1}^m \left( f(x_k) - \sum_{\nu=1}^n a_{\nu} g_{\nu}(x_k) \right)^2$$

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## Example: Vectors

$$g(x) = \{1, x, x^2, x^3\}$$

$$\mathbf{1} = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}, \quad \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{bmatrix}, \quad \mathbf{x}^2 = \begin{bmatrix} x_1^2 \\ x_2^2 \\ \vdots \\ x_m^2 \end{bmatrix}, \quad \mathbf{x}^3 = \begin{bmatrix} x_1^3 \\ x_2^3 \\ \vdots \\ x_m^3 \end{bmatrix}$$



# Gauge Fixing Conditions

$$\partial_1 M \quad \mathbf{r} \cdot \mathbf{1} = 0 \quad \sum_{k=1}^m r_k = 0$$

$$\partial_2 M \quad \mathbf{r} \cdot \mathbf{x} = 0 \quad \sum_{k=1}^m r_k x_k = 0$$

$$\partial_3 M \quad \mathbf{r} \cdot \mathbf{x}^2 = 0 \quad \sum_{k=1}^m r_k x_k^2 = 0$$

$$\partial_4 M \quad \mathbf{r} \cdot \mathbf{x}^3 = 0 \quad \sum_{k=1}^m r_k x_k^3 = 0$$

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## Summary

Forming the normal equations  
=  
imposing a gauge condition

# Measurement of Average Gradient

Partition a domain:

$$\Omega = \bigcup_k \omega_k$$

Interval

$$\omega = \{x \in \mathbb{R} : a < x < b\}$$

Average gradient

$$\langle \nabla \phi(x) \rangle_\omega = \phi(b) - \phi(a)$$

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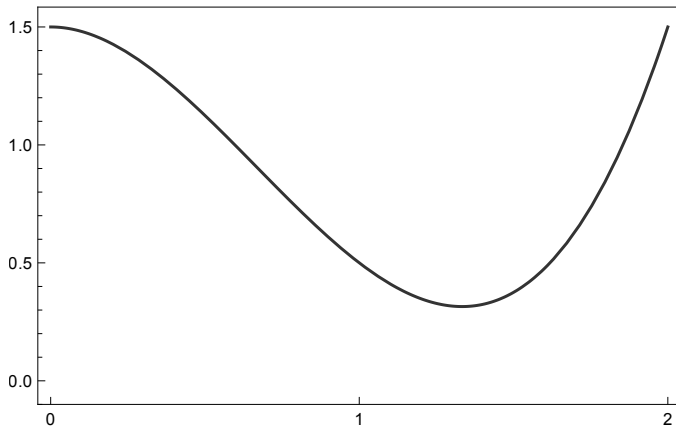
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# Source of Rank Defect

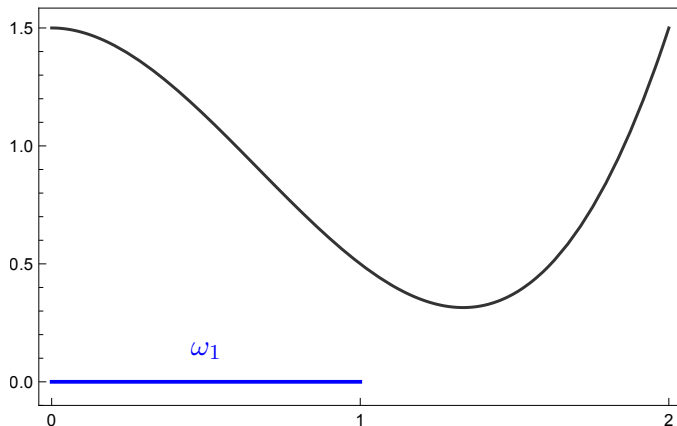
$$D_x \phi(x) = D_x (\phi(x) + \text{const})$$

# Scalar Potential $\phi$

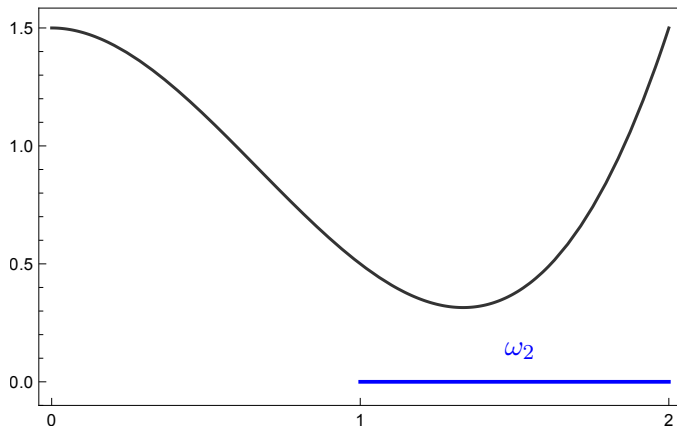




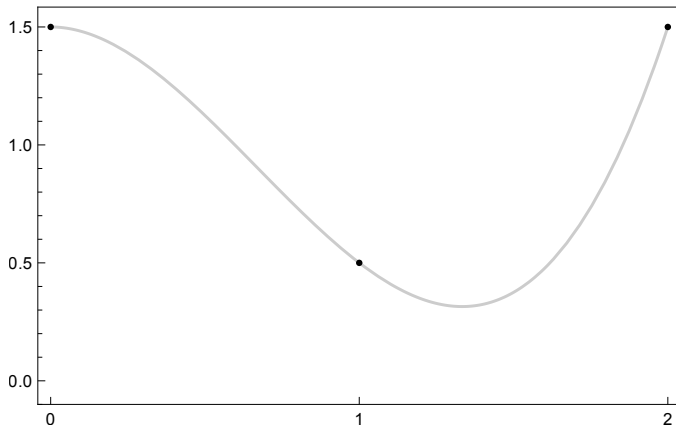
# Scalar Potential $\phi$ : Zone 1



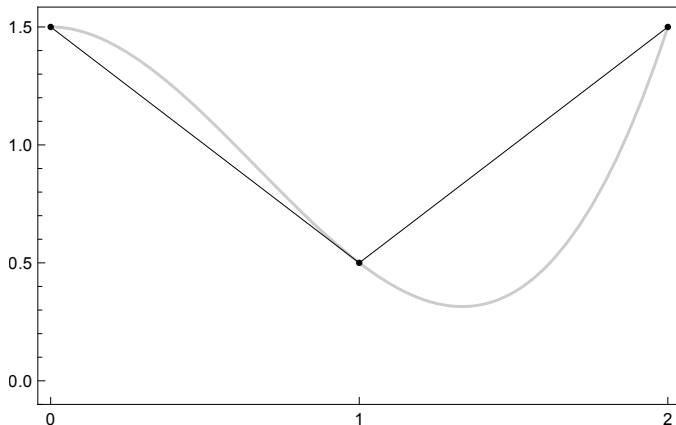
## Scalar Potential $\phi$ : Zone 2



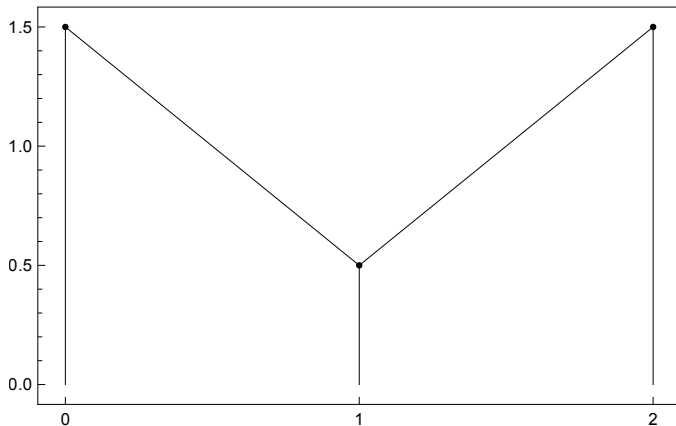
## Scalar Potential $\phi$ : Endpoints



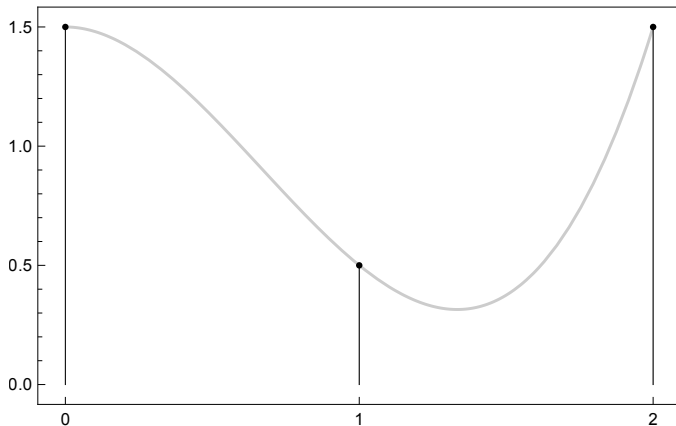
## Scalar Potential $\phi$ : Average Gradient



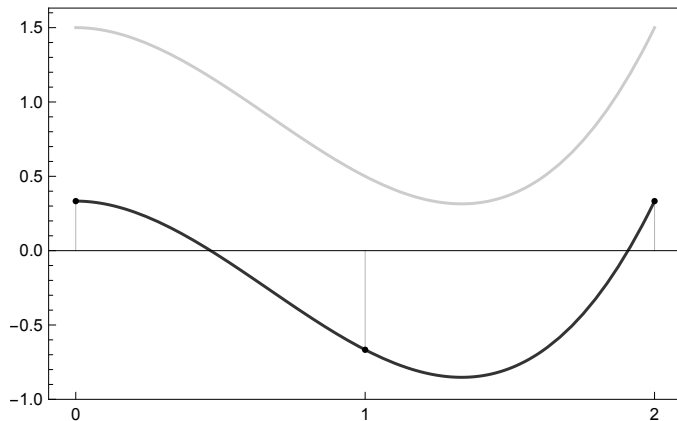
# Scalar Potential $\phi$ : Measurement and Solution



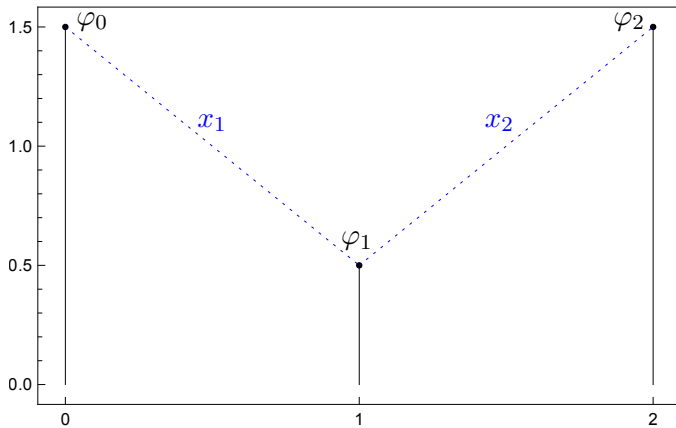
## Scalar Potential $\phi$ : Input and Output



# Scalar Potential $\phi$ : Least Squares Solution

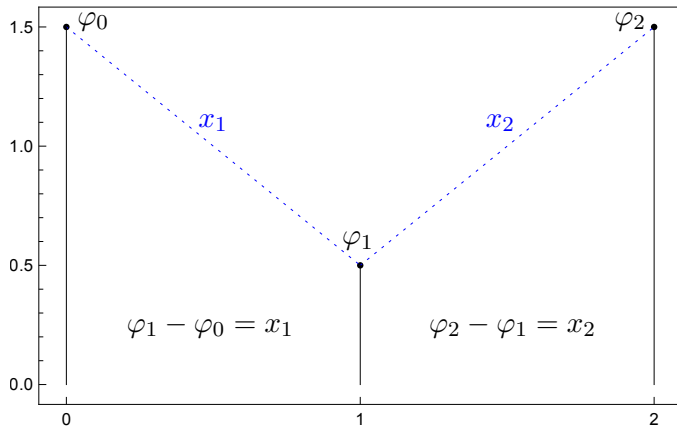


# Scalar Potential $\phi$ : Rosetta Stone





# Scalar Potential $\phi$ : Rosetta Stone



# Linear System

## Linear system

$$\begin{bmatrix} -1 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix} \begin{bmatrix} \varphi_0 \\ \varphi_1 \\ \varphi_2 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

## Least squares solution

$$\begin{bmatrix} \varphi_0 \\ \varphi_1 \\ \varphi_2 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} -2 & -1 \\ 1 & -1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \alpha \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

# Linear System

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# Gauge Condition

Least squares solution

$$\begin{bmatrix} \varphi_0 \\ \varphi_1 \\ \varphi_2 \end{bmatrix} = \frac{1}{3} \underbrace{\begin{bmatrix} -2 & -1 \\ 1 & -1 \\ 1 & 2 \end{bmatrix}}_{\text{columns sum to 0}} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \alpha \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

## Gauge Fixing Condition

$$\sum_k \varphi_k = 0$$

$\Downarrow$

$$\begin{bmatrix} -1 & 1 \\ -2 & -2 \end{bmatrix} \begin{bmatrix} \varphi_0 \\ \varphi_1 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

# Linear System: Fixed Gauge

$$\varphi_0 + \varphi_1 + \varphi_2 = 0 \quad \Rightarrow \quad \varphi_2 = -\varphi_0 - \varphi_1$$

$$\varphi_2 - \varphi_1 = x_2 \quad \Rightarrow \quad -\varphi_0 - 2\varphi_1 = x_2$$

$$\begin{bmatrix} -1 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix} \begin{bmatrix} \varphi_0 \\ \varphi_1 \\ \varphi_2 \end{bmatrix} \Rightarrow \begin{bmatrix} -1 & 1 \\ -1 & -2 \end{bmatrix} \begin{bmatrix} \varphi_0 \\ \varphi_1 \end{bmatrix}$$

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## Compare Solutions

Least squares solution

Fixed gauge solution

$$\varphi_0 = \frac{1}{3}(-2x_1 - x_2) + \alpha \quad \varphi_0 = \frac{1}{3}(-2x_1 - x_2)$$

$$\varphi_1 = \frac{1}{3}(x_1 - x_2) + \alpha \quad \varphi_1 = \frac{1}{3}(x_1 - x_2)$$

$$\varphi_2 = \frac{1}{3}(x_1 + 2x_2) + \alpha$$

From gauge condition:

$$\varphi_2 = -\varphi_1 - \varphi_0 = \frac{1}{3}(x_1 + 2x_2)$$

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Least squares solution

Fixed gauge solution

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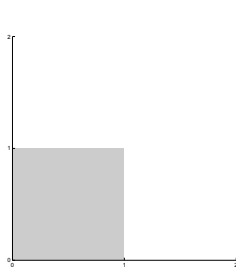
$$\varphi_1 = \frac{1}{3}(x_1 - x_2) + \alpha \quad \varphi_1 = \frac{1}{3}(x_1 - x_2)$$

$$\varphi_2 = \frac{1}{3}(x_1 + 2x_2) + \alpha$$

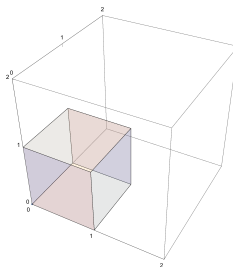
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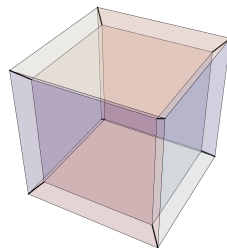
# Unit Cells



$d = 2$



$d = 3$



$d = 4$

# Source of Rank Defects

rank defect

invariance

1 
$$D_x \phi(x) = D_x (\phi(x) + c)$$

2 
$$\partial_x \phi(x, y) = \partial_x (\phi(x, y) + c_1)$$

$$\partial_y \phi(x, y) = \partial_y (\phi(x, y) + c_2)$$

3 
$$\partial_x \phi(x, y, z) = \partial_x (\phi(x, y, z) + c_1)$$

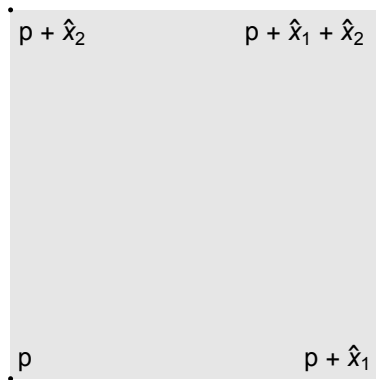
$$\partial_y \phi(x, y, z) = \partial_y (\phi(x, y, z) + c_2)$$

$$\partial_z \phi(x, y, z) = \partial_z (\phi(x, y, z) + c_3)$$

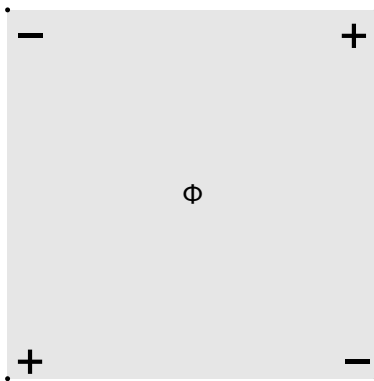
# Average Gradient

Motivates geometric interpretation of null space vectors

## Average Gradient: Unit Cell



## Average Gradient: Unit Cell



# Antiderivative

$$\Phi_{\mu}(x_1, x_2) = \int \phi(x_1, x_2) dx_{\mu}$$



# Average Gradient

$$\begin{aligned}\langle \partial_\mu \phi(x_1, x_2) \rangle_p &= \int \int \partial_\mu \phi(x_1, x_2) dx_\mu dx_\nu \\ &= \Phi_\nu(p) + \Phi_\nu(p + \hat{x}_1 + \hat{x}_2) - \Phi_\nu(p + \hat{x}_1) - \Phi_\nu(p + \hat{x}_2)\end{aligned}$$

# Bestiary

$\phi$ : ideal scalar field (smooth curve)

$\Phi$ : antiderivative of  $\phi$  (never shown)

$\varphi_k$ : approximation of  $\phi$  at point  $k$  (sticks)

# Average Gradient

we have

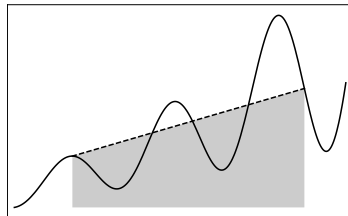
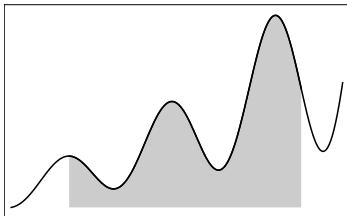
$$\Phi_1, \Phi_2$$

we want

$$\phi$$

# Trapezoidal Approximation

$$\int_a^b f(x)dx \approx \frac{b-a}{2} (f(a) + f(b))$$



# Trapezoidal Approximation

$$\int_a^b f(x)dx \approx \frac{b-a}{2} (f(a) + f(b))$$

$$|f''(x)| \leq M, \quad a < x < b$$

$$\left| \int_a^b f(x)dx - \frac{b-a}{2} (f(a) + f(b)) \right| \leq \frac{(b-a)^3}{12} M$$

# Trapezoidal Approximation

$$\int_a^b f(x)dx \approx \frac{b-a}{2} (f(a) + f(b))$$

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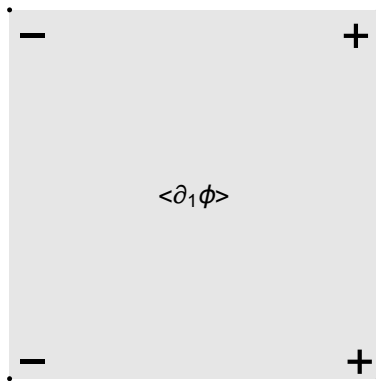
## Average Gradient: Approximation

$$\Phi_\nu(p) + \Phi_\nu(p + \hat{x}_1 + \hat{x}_2) - \Phi_\nu(p + \hat{x}_1) - \Phi_\nu(p + \hat{x}_2)$$

$$\langle \partial_1 \phi \rangle_p = \frac{1}{2} (\phi(p + \hat{x}_1 + \hat{x}_2) - \phi(p) + \phi(p + \hat{x}_1) - \phi(\mathbf{p} + \hat{\mathbf{x}}_2))$$

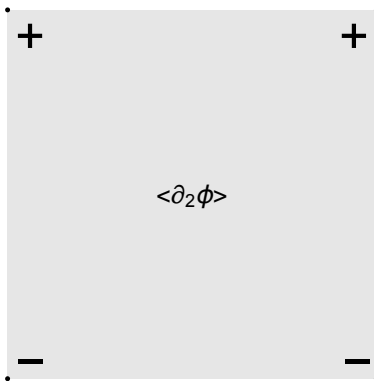
$$\langle \partial_2 \phi \rangle_p = \frac{1}{2} (\phi(p + \hat{x}_1 + \hat{x}_2) - \phi(p) - \phi(p + \hat{x}_1) + \phi(p + \hat{x}_2))$$

## Average Gradient: Unit Cell

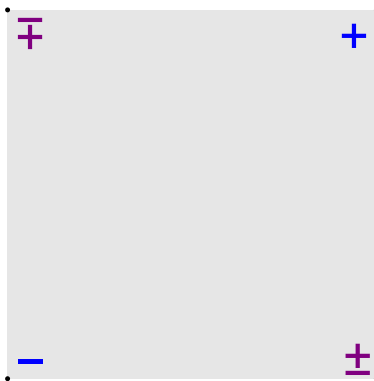




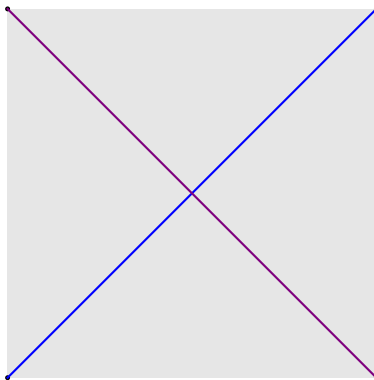
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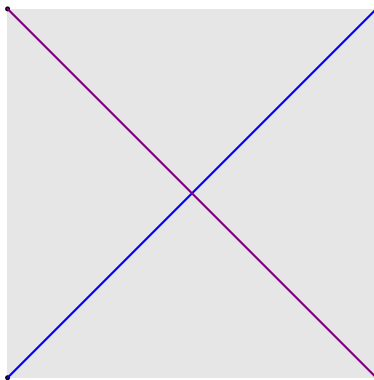
# Average Gradient: Unit Cell



# Average Gradient: Unit Cell



# Average Gradient: Next-Nearest Neighbor Interaction



## 2 Dimensions

Linear System:

$$\frac{1}{2} \begin{bmatrix} -1 & 1 & -1 & 1 \\ -1 & -1 & 1 & 1 \end{bmatrix} \begin{bmatrix} \varphi_0 \\ \varphi_1 \\ \varphi_2 \\ \varphi_3 \end{bmatrix} = \begin{bmatrix} x \\ y \end{bmatrix}$$

Least Squares Solution:

$$\begin{bmatrix} \varphi_0 \\ \varphi_1 \\ \varphi_2 \\ \varphi_3 \end{bmatrix} = \underbrace{\frac{1}{2} \begin{bmatrix} -1 & -1 \\ 1 & -1 \\ -1 & 1 \\ 1 & 1 \end{bmatrix}}_{\mathbf{A}^\dagger x} \begin{bmatrix} x \\ y \end{bmatrix} + \underbrace{\begin{bmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 \\ 0 & 1 & 1 & 0 \\ 1 & 0 & 0 & 1 \end{bmatrix}}_{\mathbf{P}_{\mathcal{R}(\mathbf{A}^*)}} \begin{bmatrix} z_1 \\ z_2 \\ z_3 \\ z_4 \end{bmatrix}$$

## 2 Dimensions

Linear System:

$$\frac{1}{2} \begin{bmatrix} -1 & 1 & -1 & 1 \\ -1 & -1 & 1 & 1 \end{bmatrix} \begin{bmatrix} \varphi_0 \\ \varphi_1 \\ \varphi_2 \\ \varphi_3 \end{bmatrix} = \begin{bmatrix} x \\ y \end{bmatrix}$$

Least Squares Solution:

$$\begin{bmatrix} \varphi_0 \\ \varphi_1 \\ \varphi_2 \\ \varphi_3 \end{bmatrix} = \underbrace{\frac{1}{2} \begin{bmatrix} -1 & -1 \\ 1 & -1 \\ -1 & 1 \\ 1 & 1 \end{bmatrix}}_{\mathbf{A}^\dagger x} \begin{bmatrix} x \\ y \end{bmatrix} + \underbrace{\begin{bmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 \\ 0 & 1 & 1 & 0 \\ 1 & 0 & 0 & 1 \end{bmatrix}}_{\mathbf{P}_{\mathcal{R}(\mathbf{A}^*)}} \begin{bmatrix} z_1 \\ z_2 \\ z_3 \\ z_4 \end{bmatrix}$$

## 2 Dimensions: Change of Coordinates

$$\begin{bmatrix} \xi \\ \eta \end{bmatrix} = \begin{bmatrix} x + y \\ x - y \end{bmatrix}$$

## 2 Dimensions

Linear System:

$$\frac{1}{2} \begin{bmatrix} -1 & 1 & -1 & 1 \\ -1 & -1 & 1 & 1 \end{bmatrix} \begin{bmatrix} \varphi_0 \\ \varphi_1 \\ \varphi_2 \\ \varphi_3 \end{bmatrix} = \begin{bmatrix} x \\ y \end{bmatrix}$$

Least Squares Solution:

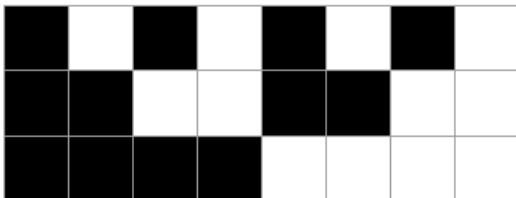
$$\begin{bmatrix} \varphi_0 \\ \varphi_1 \\ \varphi_2 \\ \varphi_3 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} -x - y \\ x - y \\ x + y \\ -x + y \end{bmatrix} = \frac{1}{2} \begin{bmatrix} -\xi \\ \eta \\ \xi \\ -\eta \end{bmatrix}$$



## Dimension 3: System Matrix **A** - Two Views

$$\mathbf{A} = \frac{1}{4} \begin{bmatrix} -1 & 1 & -1 & 1 & -1 & 1 & -1 & 1 \\ -1 & -1 & 1 & 1 & -1 & -1 & 1 & 1 \\ -1 & -1 & -1 & -1 & 1 & 1 & 1 & 1 \end{bmatrix}$$

$$\mathbf{A} = \frac{1}{4}$$



## Dimension 3: Linear System

$$\frac{1}{4} \begin{bmatrix} -1 & 1 & -1 & 1 & -1 & 1 & -1 & 1 \\ -1 & -1 & 1 & 1 & -1 & -1 & 1 & 1 \\ -1 & -1 & -1 & -1 & 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} \phi_0 \\ \phi_1 \\ \phi_2 \\ \phi_3 \\ \phi_4 \\ \phi_5 \\ \phi_6 \\ \phi_7 \end{bmatrix} = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

## Dimension 3

$$\mathbf{A} = \frac{1}{4} \begin{bmatrix} -1 & 1 & -1 & 1 & -1 & 1 & -1 & 1 \\ -1 & -1 & 1 & 1 & -1 & -1 & 1 & 1 \\ -1 & -1 & -1 & -1 & 1 & 1 & 1 & 1 \end{bmatrix}$$

$$\mathcal{N}(\mathbf{A}^*) = \begin{bmatrix} 1 & -1 & -1 & 1 & 0 & 0 & 0 & 0 \\ -1 & 1 & 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}^T$$

## Dimension 3: Column sums

Columns sum to 0

$$\mathbf{A}^\dagger = \frac{1}{2} \begin{bmatrix} -1 & -1 & -1 \\ 1 & -1 & -1 \\ -1 & 1 & -1 \\ 1 & 1 & -1 \\ -1 & -1 & 1 \\ 1 & -1 & 1 \\ -1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

## Dimension 3: Column sums, specific rows

Null space vector 1

$$\mathbf{A}^\dagger = \frac{1}{2} \begin{bmatrix} -1 & -1 & -1 \\ 1 & -1 & -1 \\ -1 & 1 & -1 \\ 1 & 1 & -1 \\ -1 & -1 & 1 \\ 1 & -1 & 1 \\ -1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

## Dimension 3: Column sums, specific rows

Null space vector 2

$$\mathbf{A}^\dagger = \frac{1}{2} \begin{bmatrix} -1 & -1 & -1 \\ 1 & -1 & -1 \\ -1 & 1 & -1 \\ 1 & 1 & -1 \\ -1 & -1 & 1 \\ 1 & -1 & 1 \\ -1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}$$

## Dimension 3: Column sums, specific rows

Null space vector 3

$$\mathbf{A}^\dagger = \frac{1}{2} \begin{bmatrix} -1 & -1 & -1 \\ 1 & -1 & -1 \\ -1 & 1 & -1 \\ 1 & 1 & -1 \\ -1 & -1 & 1 \\ 1 & -1 & 1 \\ -1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}$$

## Dimension 3: Column sums, specific rows

Null space vector 4

$$\mathbf{A}^\dagger = \frac{1}{2} \begin{bmatrix} -1 & -1 & -1 \\ 1 & -1 & -1 \\ -1 & 1 & -1 \\ 1 & 1 & -1 \\ -1 & -1 & 1 \\ 1 & -1 & 1 \\ -1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} -1 \\ 1 \\ 1 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$



## Dimension 3: Column sums, specific rows

Null space vector 5

$$\mathbf{A}^\dagger = \frac{1}{2} \begin{bmatrix} -1 & -1 & -1 \\ 1 & -1 & -1 \\ -1 & 1 & -1 \\ 1 & 1 & -1 \\ -1 & -1 & 1 \\ 1 & -1 & 1 \\ -1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

## 3 Dimensions: Least Squares Solution

$$\begin{bmatrix} \phi_0 \\ \phi_1 \\ \phi_2 \\ \phi_3 \\ \phi_4 \\ \phi_5 \\ \phi_6 \\ \phi_7 \end{bmatrix} = \frac{1}{2} \left[ \begin{array}{ccc|ccc} -1 & & & -1 & & \\ & 1 & & -1 & & \\ & & -1 & & 1 & \\ & & & 1 & & -1 \\ -1 & & & & 1 & \\ & -1 & & -1 & & 1 \\ & & 1 & & -1 & \\ & & & -1 & & 1 \\ & & & & 1 & \\ & & & & & 1 \end{array} \right] \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

$$+ \begin{bmatrix} 5 & -1 & -1 & 1 & -1 & 1 & 1 & 3 \\ -1 & 5 & 1 & -1 & 1 & -1 & 3 & 1 \\ -1 & 1 & 5 & -1 & 1 & 3 & -1 & 1 \\ 1 & -1 & -1 & 5 & 3 & 1 & 1 & -1 \\ -1 & 1 & 1 & 3 & 5 & -1 & -1 & 1 \\ 1 & -1 & 3 & 1 & -1 & 5 & 1 & -1 \\ 1 & 3 & -1 & 1 & -1 & 1 & 5 & -1 \\ 3 & 1 & 1 & -1 & 1 & -1 & -1 & 5 \end{bmatrix} \begin{bmatrix} \psi_0 \\ \psi_1 \\ \psi_2 \\ \psi_3 \\ \psi_4 \\ \psi_5 \\ \psi_6 \\ \psi_7 \end{bmatrix}$$

## 3 Dimensions: Gauge Fixing

Schematically:  $\left[ \begin{array}{c} \mathbf{I}_8 \\ -\mathcal{N}(\mathbf{A}^*) \end{array} \right] \Rightarrow$

$$\left[ \begin{array}{cccccccc} 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ \hline -1 & 1 & 1 & -1 & 0 & 0 & 0 & 0 \\ 1 & -1 & -1 & 0 & -1 & 0 & 0 & 0 \\ 0 & 0 & -1 & 0 & 0 & -1 & 0 & 0 \\ 0 & -1 & 0 & 0 & 0 & 0 & -1 & 0 \\ -1 & 0 & 0 & 0 & 0 & 0 & 0 & -1 \end{array} \right]$$

## 3 Dimensions: Reduced System Matrix

$$\mathbf{A}_r = \mathbf{A}\mathbf{Q} = \mathbf{A} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ -1 & 1 & 1 \\ 1 & -1 & -1 \\ 0 & 0 & -1 \\ 0 & -1 & 0 \\ -1 & 0 & 0 \end{bmatrix} = \begin{bmatrix} -1 & 1 & 0 \\ -1 & 0 & 1 \\ 0 & -1 & -1 \end{bmatrix}$$

## Equivalent Full Rank Condition

Linear System: 
$$\begin{bmatrix} -1 & 1 & 0 \\ -1 & 0 & 1 \\ 0 & -1 & -1 \end{bmatrix} \begin{bmatrix} \varphi_0 \\ \varphi_1 \\ \varphi_2 \end{bmatrix} = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

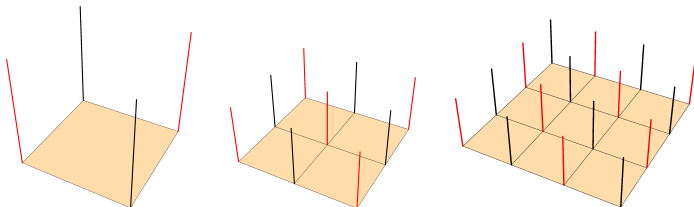
Direct Solution:

$$\begin{bmatrix} \varphi_0 \\ \varphi_1 \\ \varphi_2 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} -1 & -1 & -1 \\ 1 & -1 & -1 \\ -1 & 1 & -1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

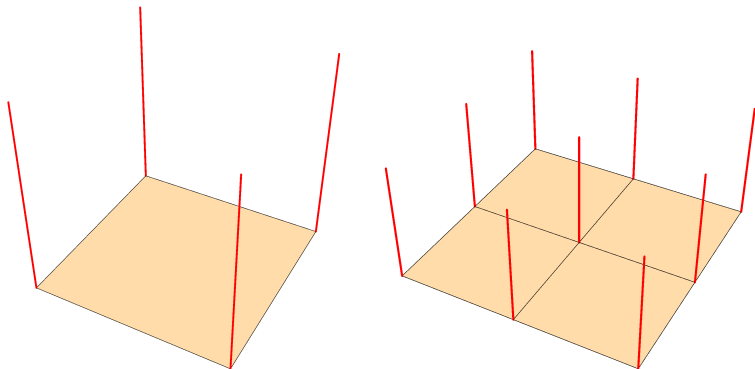
# Challenge

Construct geometries of arbitrary rank defect

## Rank Defect 2: Squares

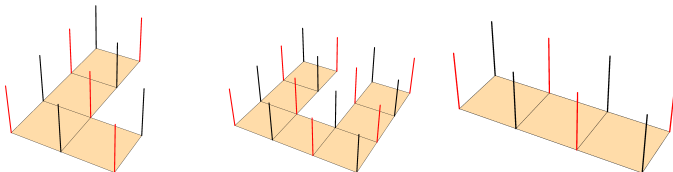


## Rank Defect 1: PBC

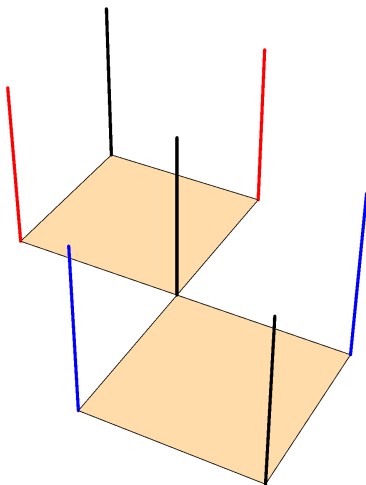




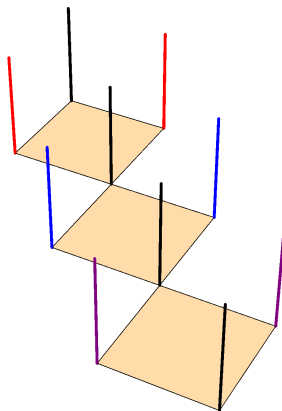
## Rank Defect 2: Irregulars



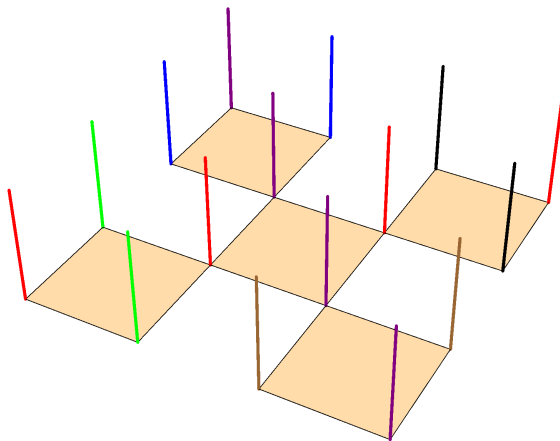
## Rank Defect 3



## Rank Defect 4



## Rank Defect 6



# Summary

① Measure vector field  $\mathbf{F}$

② Construct  $\mathbf{A} = \left[ \begin{array}{c|c} \mathbf{U}_{\mathcal{R}} & \mathbf{U}_{\mathcal{N}} \end{array} \right] \left[ \begin{array}{c|c} \mathbf{S} & \mathbf{0} \\ \hline \mathbf{0} & \mathbf{0} \end{array} \right] \left[ \begin{array}{c} \mathbf{V}_{\mathcal{R}}^* \\ \hline \mathbf{V}_{\mathcal{N}}^* \end{array} \right]$

③ Construct null space vectors

④ Create  $\mathbf{A} = \left[ \begin{array}{c|c} \mathbf{U}_{\mathcal{R}} & \mathbf{U}_{\mathcal{N}} \end{array} \right] \left[ \begin{array}{c} \mathbf{S} \\ \hline \mathbf{0} \end{array} \right] \left[ \begin{array}{c} \mathbf{V}_{\mathcal{R}}^* \end{array} \right]$

⑤ Solve for potential s.t.  $\mathbf{F} = \nabla \phi$



# Fixing Gauge and Rank Deficiency

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SIAM CONFERENCE ON  
APPLIED LINEAR ALGEBRA

