

Data-Driven Compatible Schemes for Structure-Preserving Surrogate Models

Jonas A. Actor¹, Anthony Gruber¹, Andy Huang¹, Eric Cyr¹, Xiaozhe Hu², Nat Trask³

¹Sandia National Laboratories ²Tufts University ³University of Pennsylvania



Given:
high-fidelity data that obeys physics laws

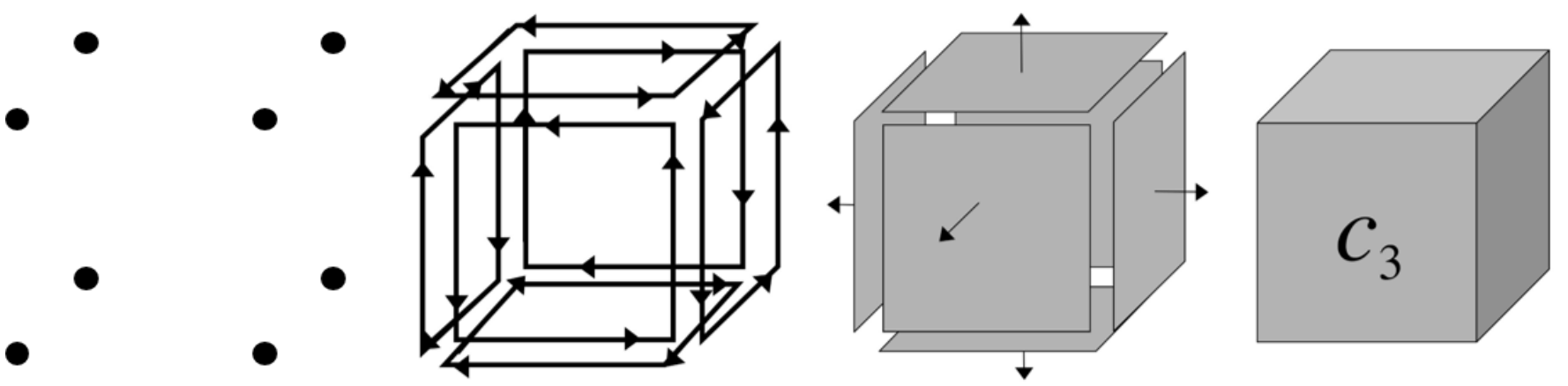
Goal:
recover surrogate model preserving physics structure

Challenges

- 1 How to **enforce structure preservation** from sampled point data?
- 2 How to extract a generalizable **parameterized model** from small amounts of data?
- 3 How to guarantee **stability** and **convergence** properties?

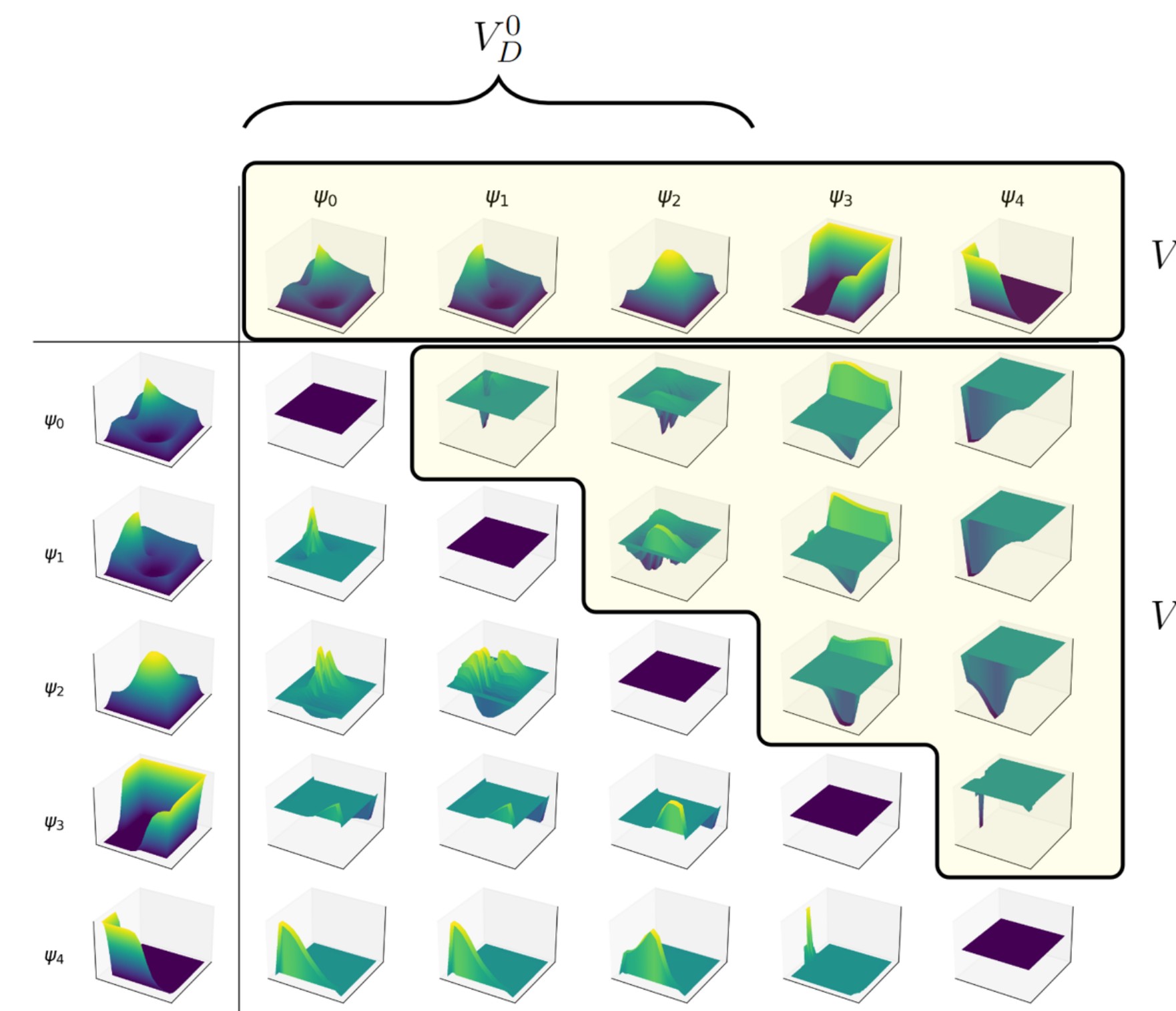
Chain Complexes as Surrogate Models?

- Guarantee Exact Sequence property
- Enforce Stokes' Theorem
- Ensure compatibility between spaces



Chain Complex Model Parameterization

4 Construct V^1 variational space



5 Introduce learnable (co)boundary metrics

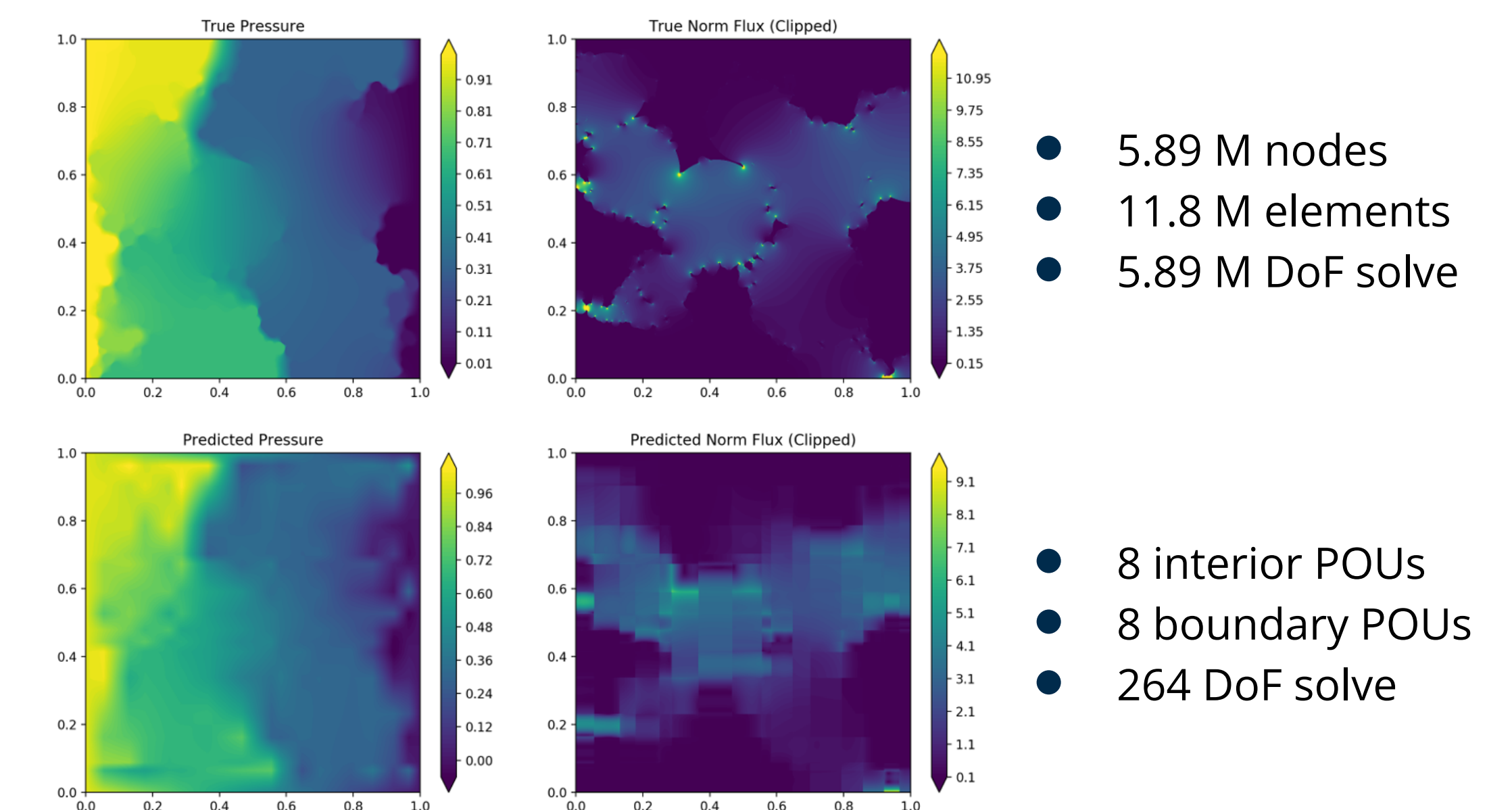
$$\begin{array}{ccccccc}
 C^0 & \xleftarrow{d_0^*} & C^1 & \xleftarrow{d_1^*} & C^2 & \xleftarrow{d_2^*} & C^3 & \xleftarrow{d_3^*} & \dots & \xleftarrow{d_{d-1}^*} & C^d \\
 \downarrow D_0^{-1} & & \downarrow D_1^{-1} & & \downarrow D_2^{-1} & & \downarrow D_3^{-1} & & & & \downarrow D_d^{-1} \\
 C^0 & \xleftarrow{\delta_0^*} & C^1 & \xleftarrow{\delta_1^*} & C^2 & \xleftarrow{\delta_2^*} & C^3 & \xleftarrow{\delta_3^*} & \dots & \xleftarrow{\delta_{d-1}^*} & C^d \\
 \downarrow B_0 & & \downarrow B_1 & & \downarrow B_2 & & \downarrow B_3 & & & & \downarrow B_d \\
 C^0 & \xrightarrow{d_0} & C^1 & \xrightarrow{d_1} & C^2 & \xrightarrow{d_2} & C^3 & \xrightarrow{d_3} & \dots & \xrightarrow{d_{d-1}} & C^d
 \end{array}$$

6 Minimize model residual against data

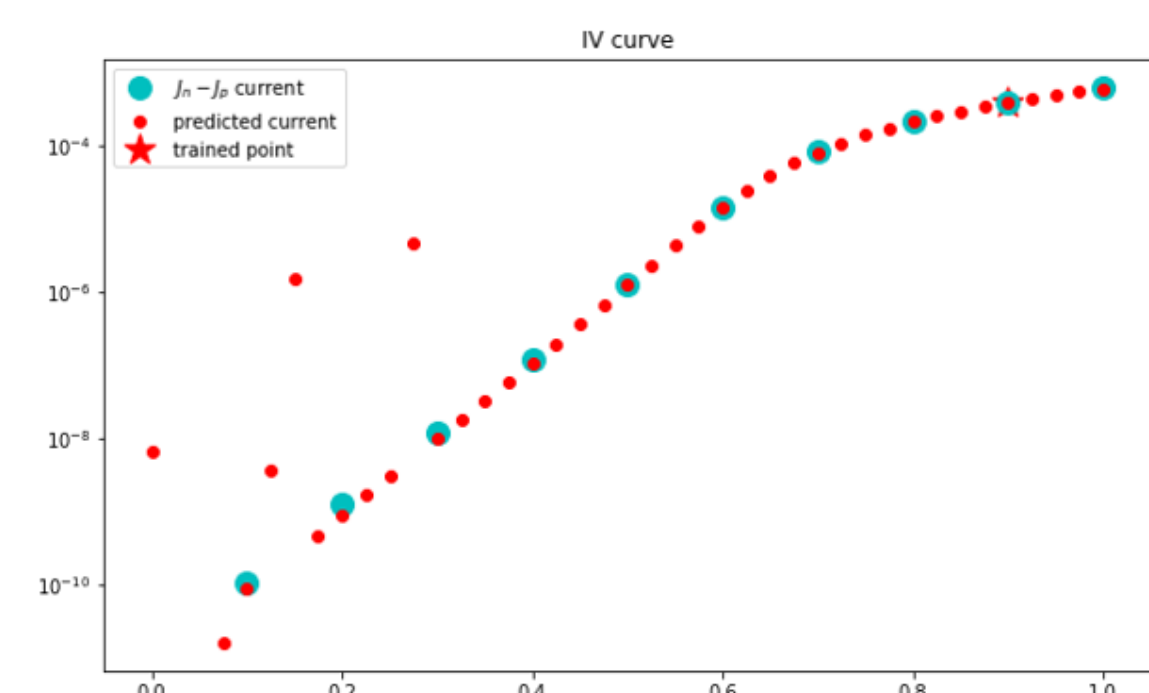
$$\min_{\substack{\text{POU parameters } \theta \\ \text{metrics } \mathbf{D}_0, \mathbf{D}_1, \mathbf{B}_0, \mathbf{B}_1}} \left\| p_{\text{data}} - \sum_i \hat{p}_i \psi_i \right\|_2^2 + \left\| \mathbf{F}_{\text{data}} - \sum_{ij} \hat{F}_{ij} \psi_{ij} \right\|_2^2$$

such that
$$\begin{bmatrix} M_1 & M_1 D_1^{-1} \delta_0 D_0 \\ B_0^{-1} \delta_0^* B_1 M_1 & 0 \end{bmatrix} \begin{bmatrix} \hat{F} \\ \hat{p} \end{bmatrix} = \begin{bmatrix} g \\ f \end{bmatrix}$$

Darcy Flow



Drift-Diffusion



- Train to fit an applied voltage of $f_\theta = 0.9$
- Solution changes nonlinear from linear to exponential regime
- Measure current $I = (\mathbf{J}_n - \mathbf{J}_p) \cdot \vec{n}$ at boundary