

Application of the Adjoint Method in gradient-based Optimization to the $\mathcal{M}1$ -Model in EPMA

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Forward Model: a k-ratio model based on the $\mathcal{M}1$ -Model

The spatial resolution in electron probe microanalysis (EPMA) is currently limited to a pixel size approximately as large as the interaction volume of the electrons.

Aim: increase the spatial resolution of EPMA to a pixel size smaller than the interaction volume

How: using **more sophisticated reconstruction methods** based on a model which precisely simulates electron transport and x-ray generation inside the material to resolve small concentration pixel

$\mathcal{M}1$ -Model to describe the electron fluence ψ^0

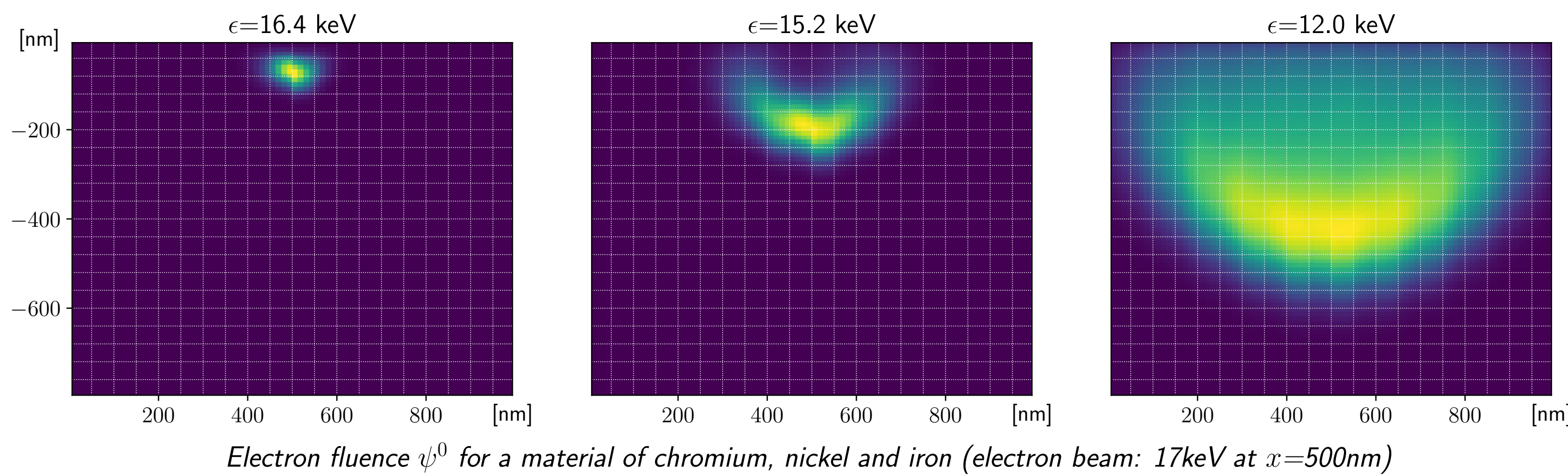
- deterministic transport/collision model of the electron fluence ψ^0 at energy ϵ and position \bar{x} in a material domain Q described by its mass concentration field $c(\bar{x})$
- first order moment approximation to the Boltzmann equation for electron transport in continuous slowing down approximation [1]

$$-\partial_\epsilon \left(S(c(\bar{x}), \epsilon) \begin{pmatrix} \psi^0(\bar{x}, \epsilon) \\ \psi^1(\bar{x}, \epsilon) \end{pmatrix} \right) + \nabla_x \left(\psi_{ME}^2(\psi^0, \psi^1) \right) + \begin{pmatrix} 0 & 0 \\ 0 & T(c(\bar{x}), \epsilon)I \end{pmatrix} \begin{pmatrix} \psi^0(\bar{x}, \epsilon) \\ \psi^1(\bar{x}, \epsilon) \end{pmatrix} = 0$$

$:= G(\psi(\bar{x}, \epsilon), c(\bar{x}))$

with: stopping power S , transport coefficient T , minimum entropy closure ψ_{ME}^2

Solution of the hyperbolic pde using the finite volume library CLAWPACK [2]

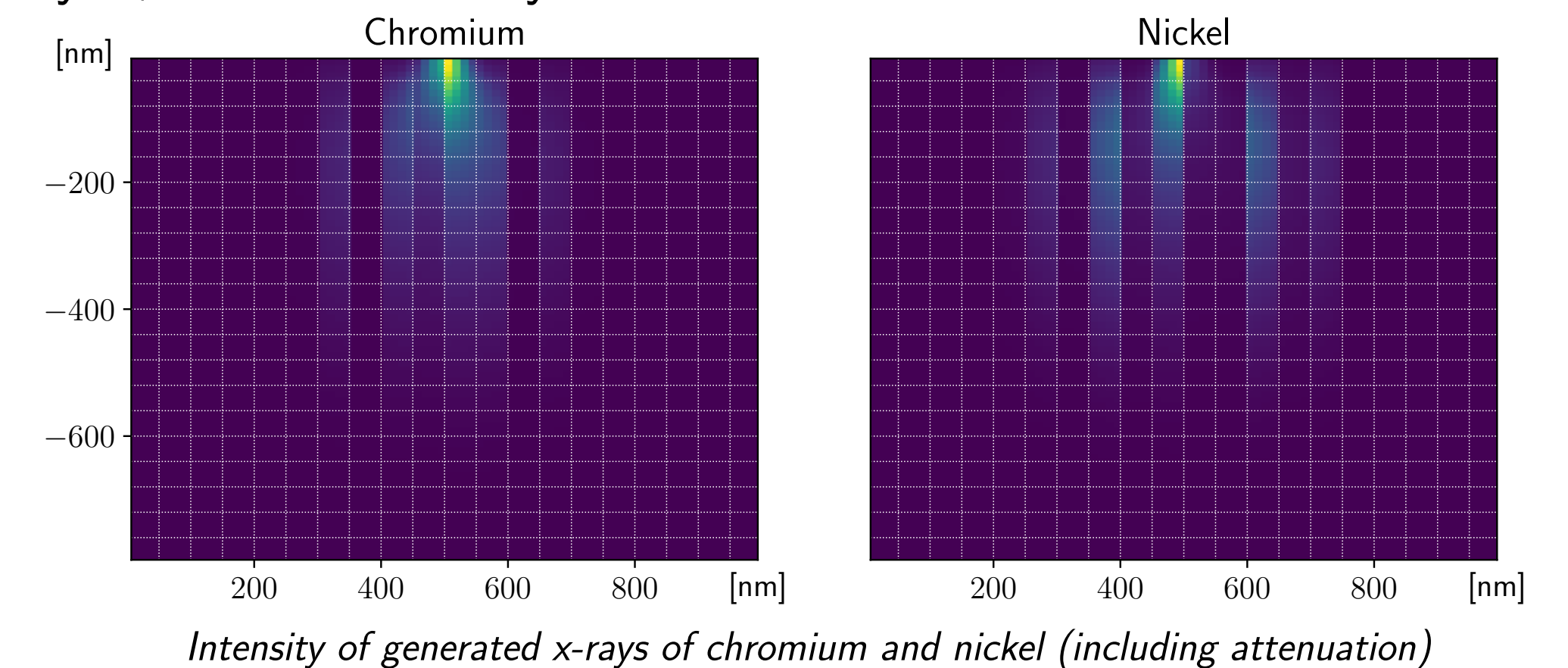


k-ratio model

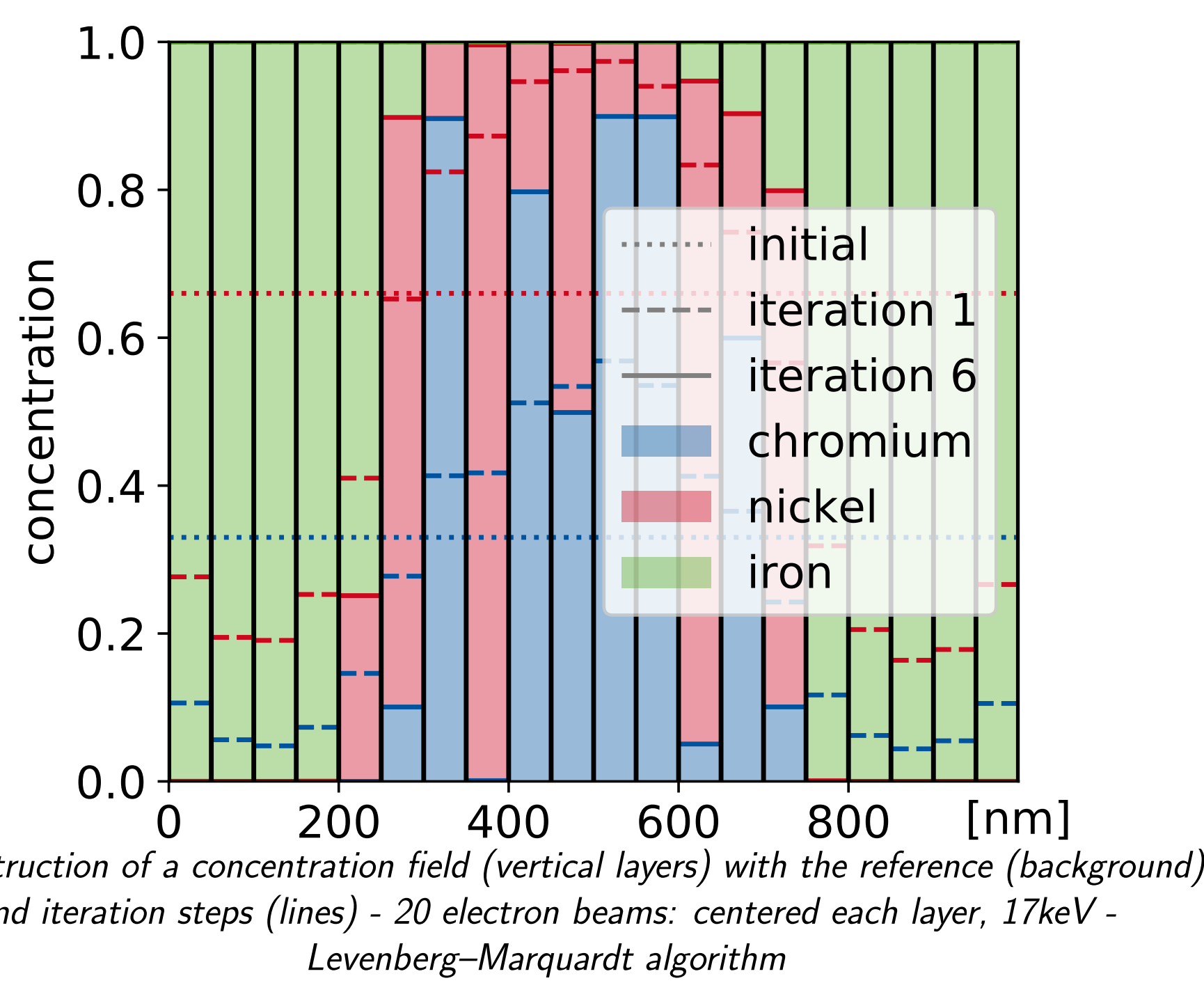
The k-ratio $k^{i,j}(c)$ can be computed as (i refers to an element, j to one of its characteristic k-ratios)

$$k^{i,j}(c) = \frac{1}{I_{std}^{i,j}} \int_Q e^{-\int_{d(\bar{x})}^{\mu^{i,j}(c(\bar{y}))} d\bar{y}} \int_{\epsilon_{cutoff}}^{\epsilon_{initial}} \omega_K^i \sigma_{ion}^{i,j}(\epsilon) N_V^i(c(\bar{x})) \int_{S^2} ||v(\epsilon)|| n(\bar{x}, v(\epsilon, \Omega)) d\Omega d\epsilon d\bar{x} = \psi^0(x, \epsilon)$$

with: standard intensity $I_{std}^{i,j}$, attenuation $e^{-\int \dots d\bar{y}}$, fluorescence yield ω_K^i , ionization cross section $\sigma_{ion}^{i,j}$, number density of atoms N_V^i , electron number density n , electron velocity v



Inverse Problem: reconstruct concentrations from k-ratios



Goal: given **experimental** k-ratios $\tilde{k}^{i,j}$ and a **model** $k^{i,j}(c)$ we search for the concentration c^* which **minimizes their squared error**

$$c^* = \arg \min_c \frac{1}{2} \sum_{i,j} (k^{i,j}(c) - \tilde{k}^{i,j})^2 =: J(c)$$

Gradient-based Optimization

$$c^{k+1} = c^k + f(c^k, \nabla_c J(c^k)) \quad \text{with initial } c^0$$

- iterative schemes based on the gradient of the objective function
- Levenberg-Marquardt algorithm (exploits least squares structure)

Gradient via Finite Differences

$$\frac{\partial J(c)}{\partial c_m} = \frac{J(c + \delta c_m e_m) - J(c)}{\delta c_m} \quad (\text{forward FD})$$

→ one pde solve per parameter c_m

Gradient via Adjoint State Method [4]

$$G(\psi, c) = 0 \quad (\text{forward model})$$

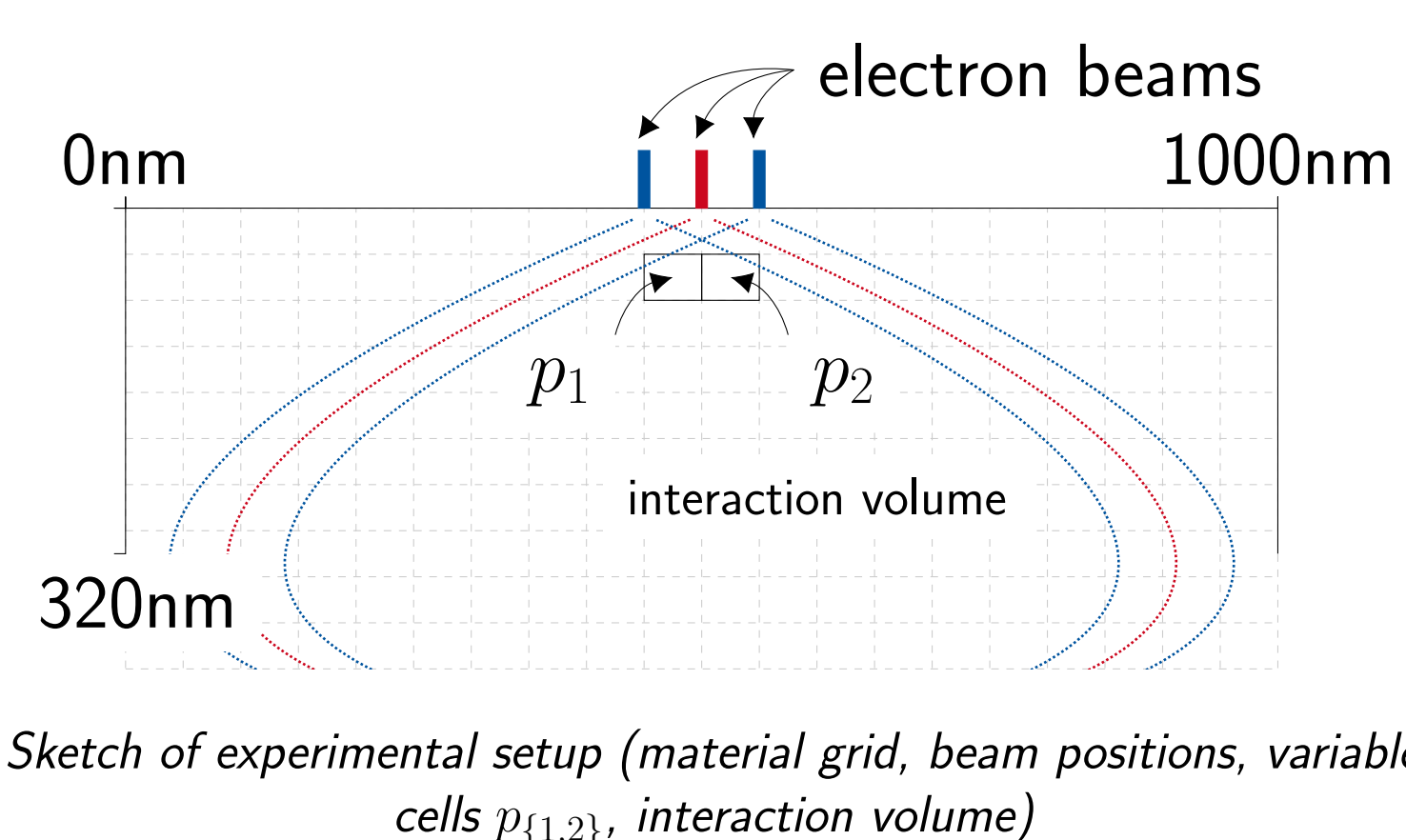
$$\frac{\partial J}{\partial \psi} - \left(\frac{\partial G}{\partial \psi} \right)^* \lambda = 0 \quad (\text{adjoint equation})$$

$$\nabla_c J(c) = \nabla_c h(\psi, c) - \langle \lambda, \nabla_c G(\psi, c) \rangle$$

with: the adjoint operator $(\cdot)^*$ and $J(c) = h(\psi, c)$

→ two pde solves per gradient

Investigation of the objective function (Example: variation of the concentration in two cells)



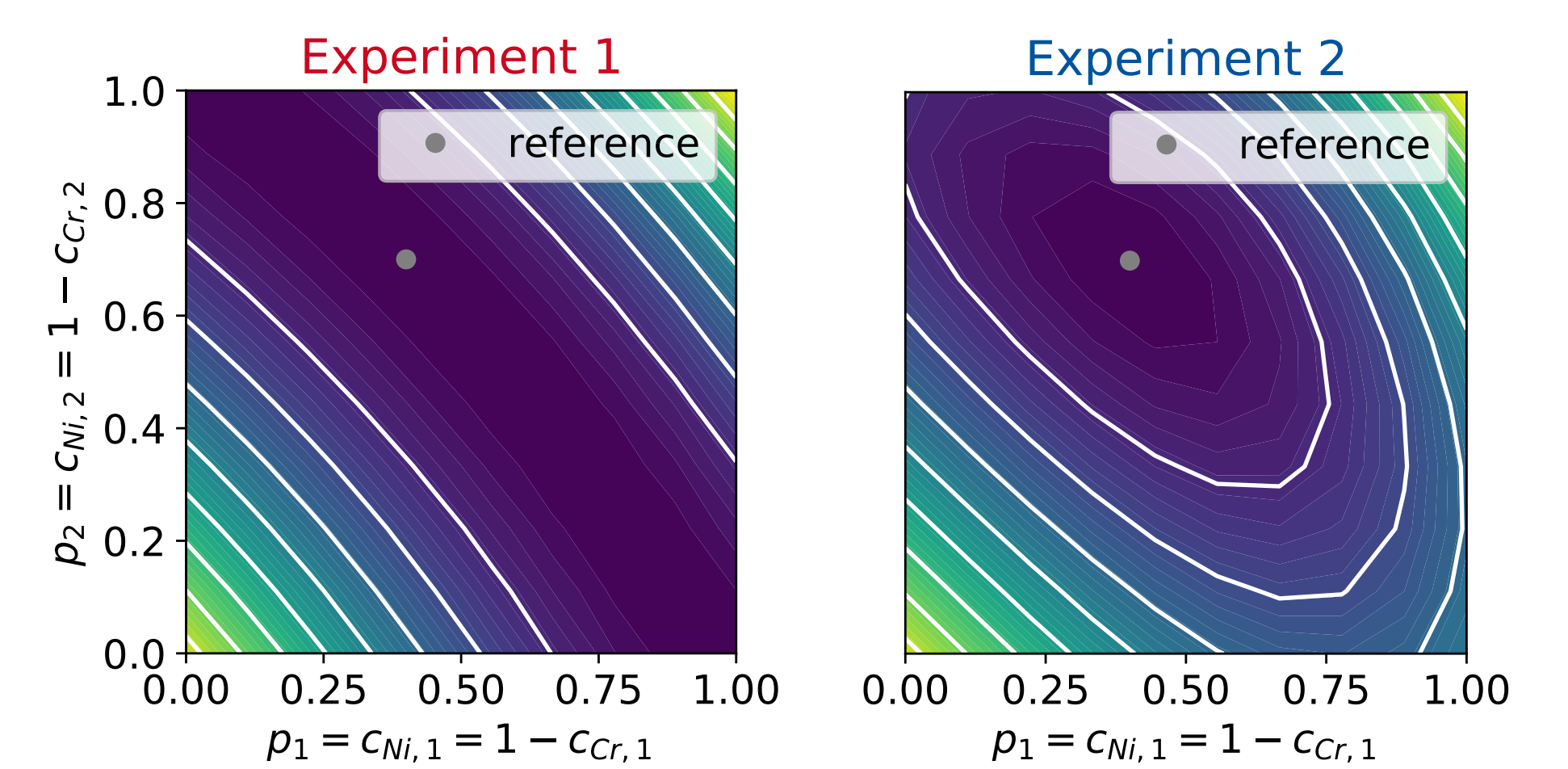
Assume: only $c_{Ni} = p_{\{1,2\}}$ and $c_{Cr} = 1 - p_{\{1,2\}}$ are variable

Result: Experiment 1 (one beam position): insufficient information to reconstruct $p_{\{1,2\}}$

Experiment 2 (two beam positions): unique minimum for $p_{\{1,2\}}$

Also: multiple beam energies for depth information [3]

Further investigation: Definiteness and eigenvalues/vectors of the hessian $\nabla_c^2 J(c)$ at the minimum



Next steps

- Investigation of the objective function and inverse problem with noisy measurements
- Uncertainty quantification for reconstructed parameters (e.g. confidence interval)

References

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