

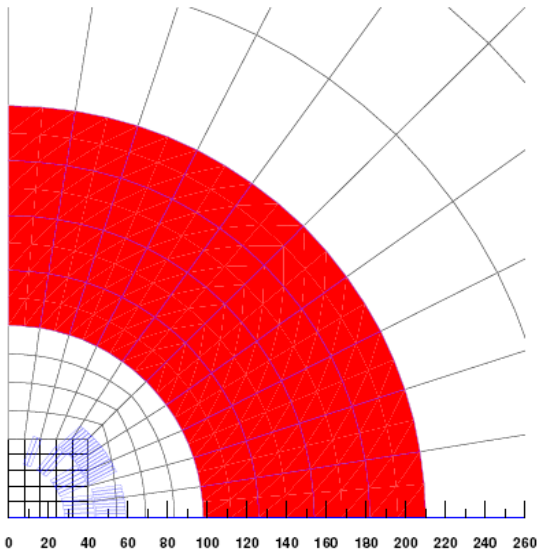
Foundations of Analytical and Numerical Field Computation

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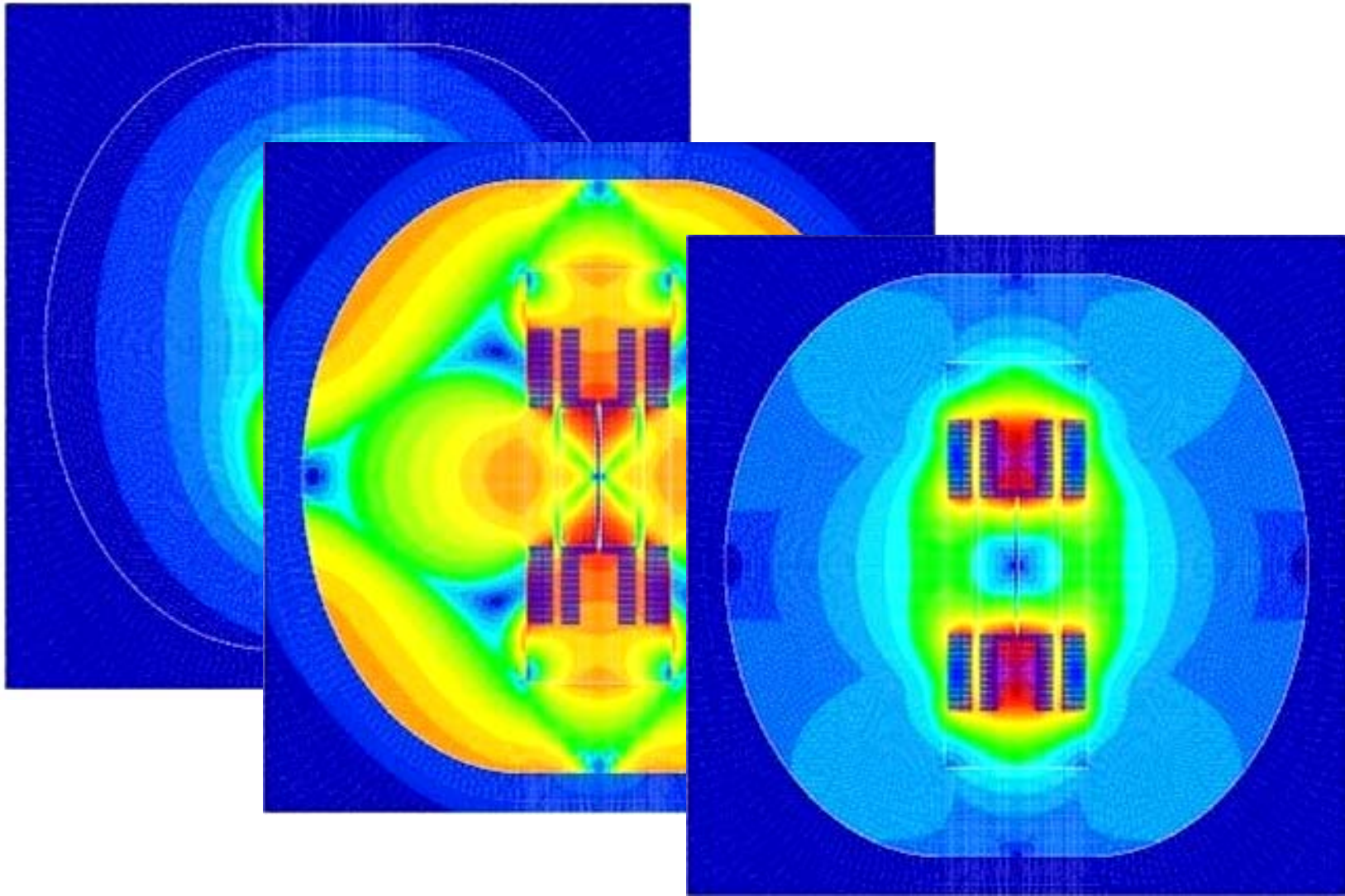
$$\mathbf{A} = \mathbf{A}_s + \mathbf{A}_r \qquad \mathbf{B} = \mu_0 \mathbf{H}_s + \text{curl } \mathbf{A}_r$$

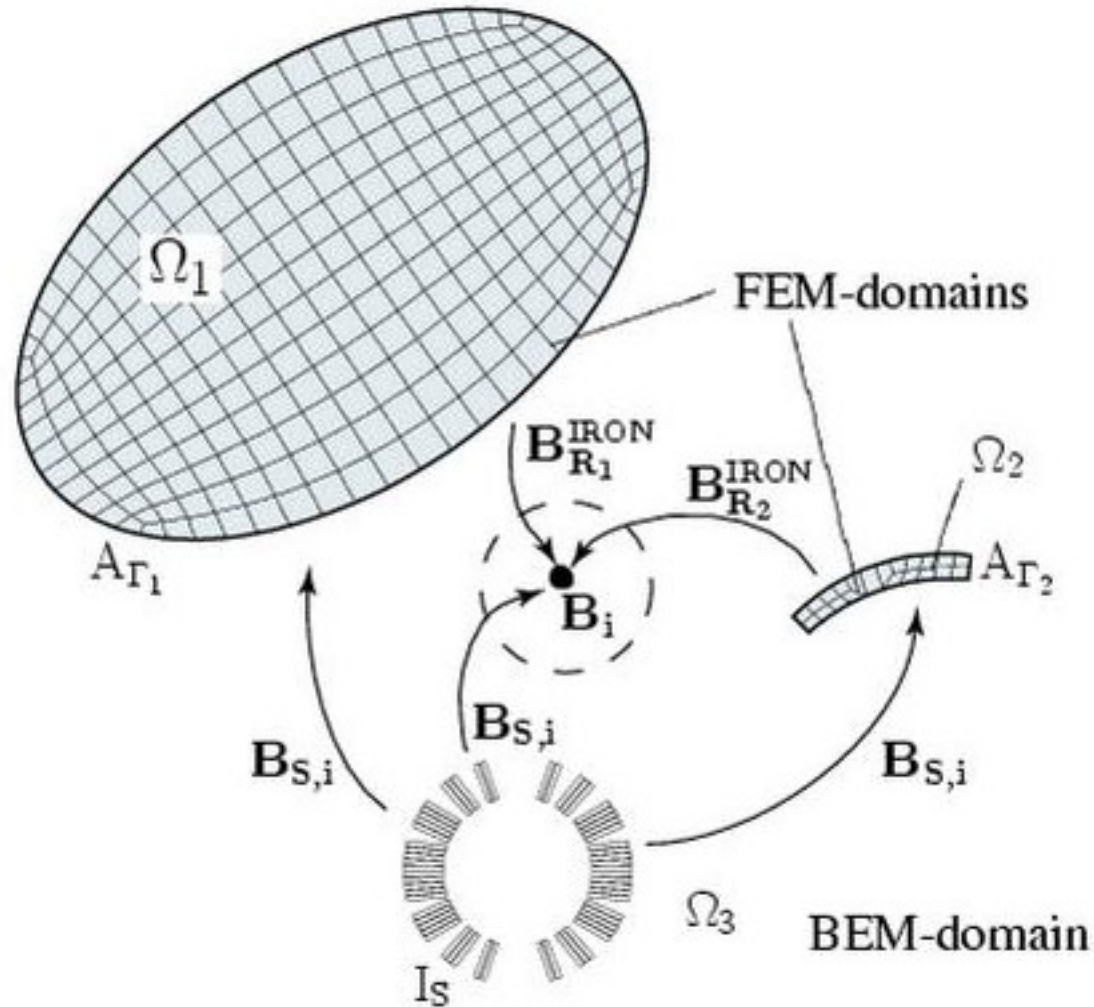
$$\text{curl } \frac{1}{\mu} \text{curl } (\mathbf{A}_r + \mathbf{A}_s) - \text{grad } \frac{1}{\mu} \text{div } (\mathbf{A}_r + \mathbf{A}_s) = \mathbf{J}$$

$$\begin{aligned} \text{curl } \frac{1}{\mu} \text{curl } \mathbf{A}_r - \text{grad } \frac{1}{\mu} \text{div } \mathbf{A}_r &= \mathbf{J} - \text{curl } \frac{1}{\mu} \text{curl } \mathbf{A}_s \\ &= \text{curl } \mathbf{H}_s - \text{curl } \frac{\mu_0}{\mu} \mathbf{H}_s \\ &= \text{curl } \left(\mathbf{H}_s - \frac{\mu_0}{\mu} \mathbf{H}_s \right) \end{aligned}$$



Advantages: No meshing of the coil, no cancellation errors, distinction between source field and iron magnetization





$$-\frac{1}{\mu_0} \nabla^2 \mathbf{A} = \mathbf{J} + \text{curl } \mathbf{M}$$

in Ω_i ,

$$\mathbf{A} \cdot \mathbf{n} = 0$$

on Γ_H ,

$$\frac{1}{\mu_0} \text{div } \mathbf{A} = 0$$

on Γ_B ,

$$\mathbf{n} \times (\mathbf{A} \times \mathbf{n}) = \mathbf{0}$$

on Γ_B ,

$$\frac{1}{\mu} (\text{curl } \mathbf{A}) \times \mathbf{n} = \mathbf{0}$$

on Γ_H ,

$$\left[\frac{1}{\mu_0} \text{div } \mathbf{A}_a \right]_{\text{ai}} = 0$$

on Γ_{ai} ,

$$\frac{1}{\mu_0} (\text{curl } \mathbf{A}_i - \mu_0 \mathbf{M}) \times \mathbf{n}_i + \frac{1}{\mu_0} (\text{curl } \mathbf{A}_a) \times \mathbf{n}_a = \mathbf{0}$$

on Γ_{ai} ,

$$[\mathbf{A}]_{\text{ai}} = \mathbf{0}$$

on Γ_{ai} .

$$\frac{1}{\mu_0} \int_{\Omega_i} \text{grad}(\mathbf{A} \cdot \mathbf{e}_a) \cdot \text{grad} w_a d\Omega_i - \frac{1}{\mu_0} \oint_{\Gamma_{ai}} \left(\frac{\partial \mathbf{A}}{\partial n_i} - (\mu_0 \mathbf{M} \times \mathbf{n}_i) \right) \cdot \mathbf{w}_a d\Gamma_{ai} = \int_{\Omega_i} \mathbf{M} \cdot \text{curl} \mathbf{w}_a d\Omega_i$$

$$\mathbf{Q}_{\Gamma_{ai}} := -\frac{\partial \mathbf{A}_{\Gamma_{ai}}^{\text{BEM}}}{\partial n_a} - \frac{\partial \mathbf{A}_i^{\text{FEM}}}{\partial n_i} - (\mu_0 \mathbf{M} \times \mathbf{n}_i) + \frac{\partial \mathbf{A}_a^{\text{BEM}}}{\partial n_a} = \mathbf{0} \quad \text{on } \Gamma_{ai}$$

$$\frac{1}{\mu_0} (\text{curl} \mathbf{A}_i^{\text{FEM}} - \mu_0 \mathbf{M}) \times \mathbf{n}_i + \frac{1}{\mu_0} (\text{curl} \mathbf{A}_a^{\text{BEM}}) \times \mathbf{n}_a = \mathbf{0} \quad \text{on } \Gamma_{ai}$$

$$\frac{1}{\mu_0} \int_{\Omega_i} \text{grad}(\mathbf{A} \cdot \mathbf{e}_a) \cdot \text{grad} w_a d\Omega_i - \frac{1}{\mu_0} \oint_{\Gamma_{ai}} \mathbf{Q}_{\Gamma_{ai}} \cdot \mathbf{w}_a d\Gamma_{ai} = \int_{\Omega_i} \mathbf{M} \cdot \text{curl} \mathbf{w}_a d\Omega_i$$

$$[K] \{A_x\} - [T] \{Q_x\} = \{F_x(\mathbf{M})\}$$

Vector Laplace

$$\nabla^2 \mathbf{A} = -\mu_0 \mathbf{J}, \quad \text{in } \Omega_a$$

Weighted Residual

$$\int_{\Omega_a} \nabla^2 A w \, d\Omega_a = - \int_{\Omega_a} \mu_0 J w \, d\Omega_a$$

From Green's second theorem:

$$\int_{\Omega_a} A \nabla^2 w \, d\Omega_a = - \int_{\Omega_a} \mu_0 J w \, d\Omega_a + \int_{\Gamma_{ai}} A \frac{\partial w}{\partial n_a} \, d\Gamma_{ai} - \int_{\Gamma_{ai}} \frac{\partial A}{\partial n_a} w \, d\Gamma_{ai}$$

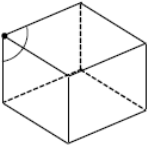


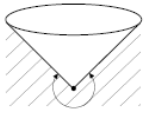
$$u^*(\mathbf{r}, \mathbf{r}') := w = \frac{1}{4\pi |\mathbf{r} - \mathbf{r}'|}$$

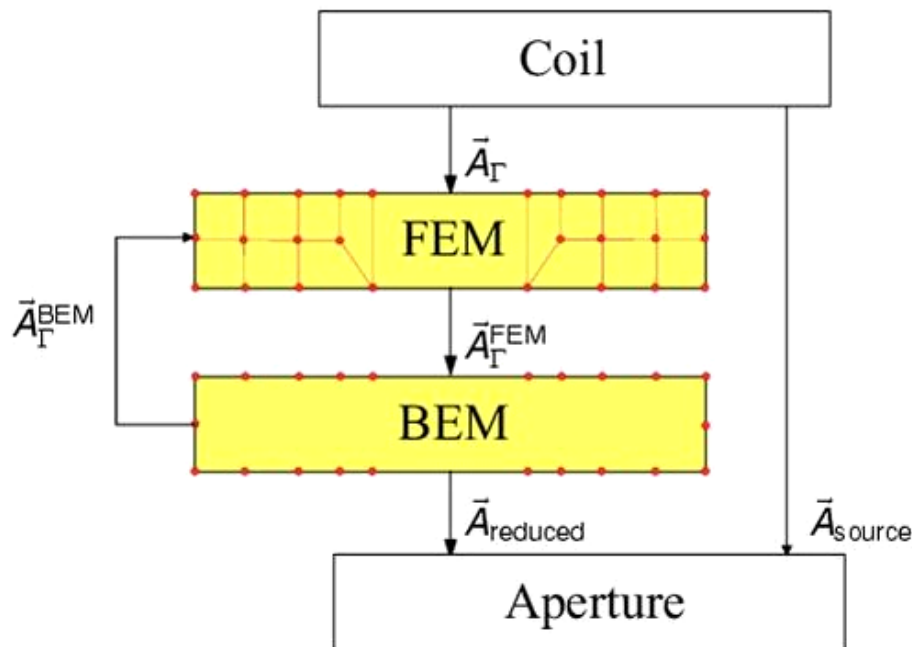
$$\nabla^2 w = -\delta(|\mathbf{r} - \mathbf{r}'|)$$

$$q^*(\mathbf{r}, \mathbf{r}') := \frac{\partial u^*}{\partial n_a} = \frac{\partial w}{\partial n_a} = -\frac{(\mathbf{r} - \mathbf{r}') \cdot \mathbf{n}_a}{4\pi |\mathbf{r} - \mathbf{r}'|^3}$$

$$\int_{\Omega} A(\mathbf{r}) \nabla^2 w \, d\Omega = \int_{\Omega} A(\mathbf{r}) \delta(|\mathbf{r} - \mathbf{r}'|) \, d\Omega = A(\mathbf{r}')$$

$$\frac{\Theta}{4\pi} A(\mathbf{r}) + \int_{\Gamma_{ai}} Q_{\Gamma_{ai}} u^*(\mathbf{r}, \mathbf{r}') d\Gamma_{ai} + \int_{\Gamma_{ai}} A_{\Gamma_{ai}} q^*(\mathbf{r}, \mathbf{r}') d\Gamma_{ai} = \int_{\Omega_a} \mu_0 J u^*(\mathbf{r}, \mathbf{r}') d\Omega_a.$$

| | | | | |
|-----------------------|---|---|---|---|
| Ω_a |  |  |  |  |
| | 90° corner | 90° cone inner | half space | 90° cone outer |
| Θ | $\frac{1}{2} \pi$ | $(2 - \sqrt{2})\pi$ | 2π | $(2 + \sqrt{2})\pi$ |
| $\frac{\Theta}{4\pi}$ | $\frac{1}{8}$ | $\frac{2-\sqrt{2}}{4}$ | $\frac{1}{2}$ | $\frac{2+\sqrt{2}}{4}$ |



BEM

$$[G]\{Q_z\} + [H]\{A_z\} = \{A_{z,s}\}$$

FEM

$$[K]\{A_z\} - [T]\{Q_z\} = \{F_z(\mathbf{M})\}$$

$$\{Q\} = -[G]^{-1}[H]\{A\} + [G]^{-1}\{A\}$$

$$([K] + [T][G]^{-1}[H])\{A\} = \{F(\mathbf{M})\} + [T][G]^{-1}\{A_s\}$$

$$[K]\{A\} = \{F(A_s, M)\}$$

yields an iteration scheme

$$\{A_{k+1}\} = [K]^{-1}\{F(A_s, M_k)\}$$

Subtracting $\{A_k\}$ from both sides yields:

$$\{\Delta A_k\} := \{A_{k+1}\} - \{A_k\} = [K]^{-1}\{R_k\}$$

where the residual is defined by

$$\{R_k\} = \{F(A_s, M_k)\} - [K]\{A_k\}$$

Introduce a *relaxation parameter* ω .

$$\{A_{k+1}\} = \{A_k\} + \omega\{\Delta A_k\}$$

with $\omega_0 = 1$ and

$$\omega_k = \frac{\omega_{k-1}}{1 - \frac{\{\Delta A_k\} \cdot \{\Delta A_{k-1}\}}{\|\{\Delta A_{k-1}\}\|^2}}$$

1. Set iteration index $k = 0$, and initialize vector potentials $\{A_0\} = \{0\}$.
2. For $k = 0, 1, 2, \dots$, unit convergence Do:
3. Compute the force vector $\{F(A_s, M_k)\}$ and the residual $\{R_k\}$.
4. If $\frac{\|\{R_k\}\|}{\|\{F(A_{sk}, M_k)\}\|} < \varepsilon$ Goto 8.
5. Calculate the step sizes $\{\Delta A_k\} = [K]^{-1}\{R_k\}$.
6. Choose the relaxation parameter.
7. $\{A_{k+1}\} = \{A_k\} + \omega\{\Delta A_k\}$
8. End Do.

Advantages

- If direct solvers are used, the stiffness matrix needs to be inverted only once.
- The method is globally convergent
- No derivative of the M(B) curve is required.

and disadvantages

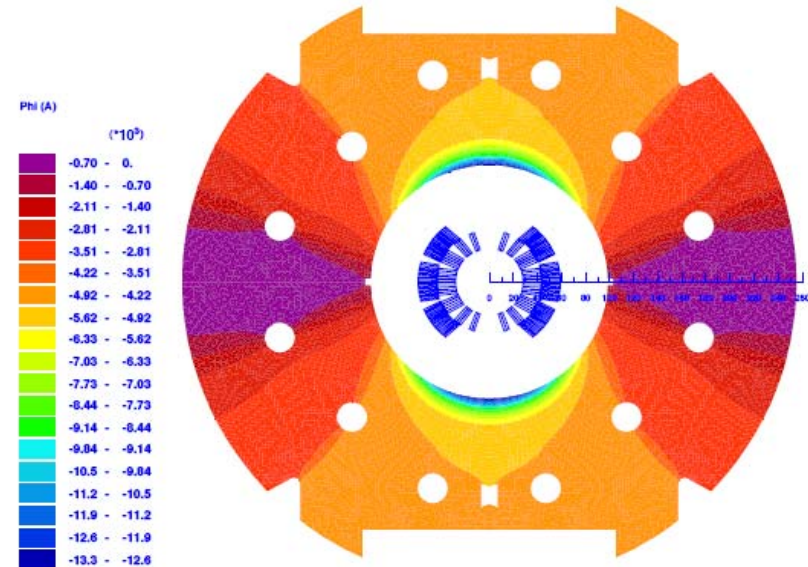
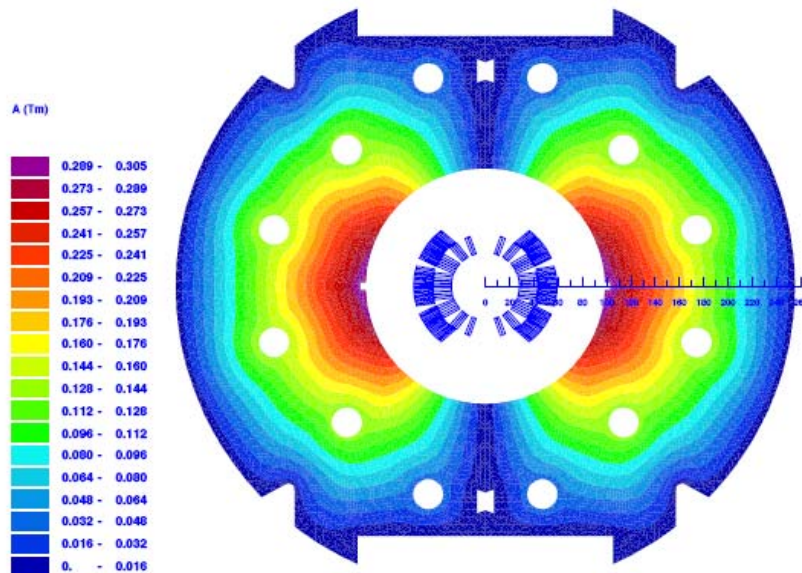
- The number of iteration steps is high.
- Even in the case of linear media the M(B)-iteration is necessary.

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          GMRES STEP 23, RESIDUAL= 0.1014E-05 ( -59.52 DB)
          GMRES STEP 24, RESIDUAL= 0.2402E-06 ( -65.78 DB)
INFO 240: GMRES STEP 24, RESIDUAL= 0.2402E-06 ( -65.78 DB).
          >>>>>>>>>> NEWTON STEP 4, RESIDUAL= 0.3917E-03 ( -38.38 DB)
COMPUTING THE GLOBAL MATRIX...
          GMRES STEP 25, RESIDUAL= 0.9083E+00 (  0.00 DB)
          GMRES STEP 26, RESIDUAL= 0.1915E-04 ( -46.76 DB)
          GMRES STEP 27, RESIDUAL= 0.6527E-05 ( -51.44 DB)
          GMRES STEP 28, RESIDUAL= 0.2510E-05 ( -55.59 DB)
          GMRES STEP 29, RESIDUAL= 0.4689E-06 ( -62.87 DB)
INFO 240: GMRES STEP 29, RESIDUAL= 0.4689E-06 ( -62.87 DB).
          >>>>>>>>>> NEWTON STEP 5, RESIDUAL= 0.1299E-03 ( -43.22 DB)
COMPUTING THE GLOBAL MATRIX...
          GMRES STEP 30, RESIDUAL= 0.9083E+00 (  0.00 DB)
          GMRES STEP 31, RESIDUAL= 0.3846E-05 ( -53.73 DB)
          GMRES STEP 32, RESIDUAL= 0.1670E-05 ( -57.36 DB)
          GMRES STEP 33, RESIDUAL= 0.7543E-06 ( -60.81 DB)
INFO 240: GMRES STEP 33, RESIDUAL= 0.7543E-06 ( -60.81 DB).
          >>>>>>>>>> NEWTON STEP 6, RESIDUAL= 0.3413E-04 ( -49.05 DB)
COMPUTING THE GLOBAL MATRIX...
          GMRES STEP 34, RESIDUAL= 0.9083E+00 (  0.00 DB)
          GMRES STEP 35, RESIDUAL= 0.2110E-05 ( -56.34 DB)
          GMRES STEP 36, RESIDUAL= 0.1095E-05 ( -59.19 DB)
          GMRES STEP 37, RESIDUAL= 0.3836E-06 ( -63.74 DB)
INFO 240: GMRES STEP 37, RESIDUAL= 0.3836E-06 ( -63.74 DB).
          >>>>>>>>>> NEWTON STEP 7, RESIDUAL= 0.7897E-05 ( -55.43 DB)
INFO 240: >>>>>>>>>> NEWTON STEP 7, RESIDUAL= 0.7897E-05 ( -55.43 DB).
INFO 296: SOLVER INFORMATION
          NUMBER OF TIME STEPS           = 1
          NUMBER OF NON-LINEAR STEPS     = 7
          AVERAGE PER TIME STEP         = 7.0
          NUMBER OF LINEAR STEPS        = 37

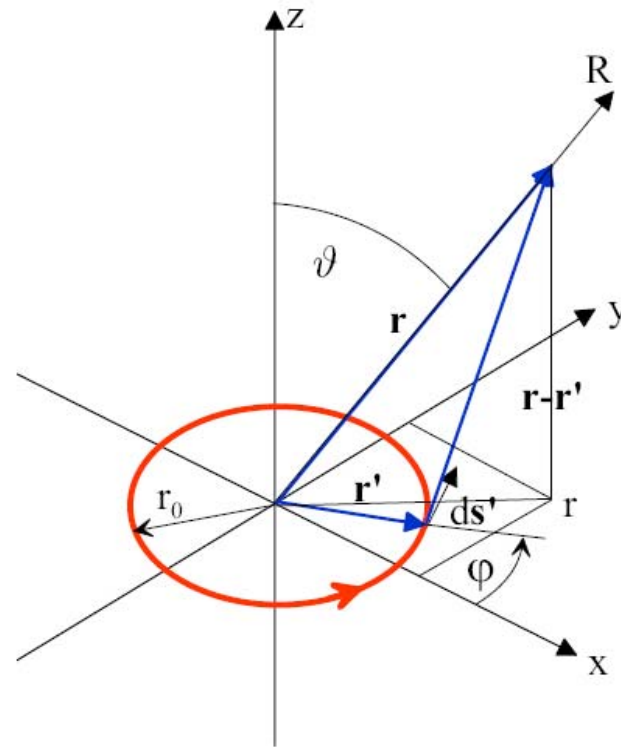
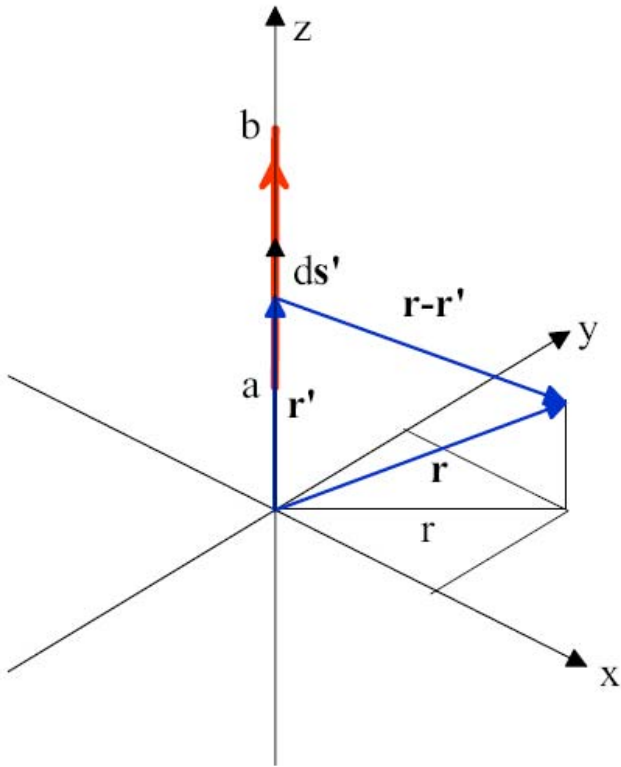
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Always check convergence of your computation



| | | | | | | |
|---------------------------|-------|-------|------|------|------|------|
| Number of finite elements | 60 | 178 | 449 | 787 | 2799 | 6233 |
| Total scalar potential | 65.8 | 72.1 | 13.0 | 5.0 | 3.8 | 15.7 |
| Vector potential | -40.5 | -27.4 | -7.4 | -4.8 | -3.8 | 25.0 |

Pre-processing (Current sources)



$$G(\mathbf{r}, \mathbf{r}') = \frac{1}{4\pi|\mathbf{r} - \mathbf{r}'|} \quad \nabla^2 \mathbf{A} = -\mu_0 \mathbf{J}$$

$$\begin{aligned} \mathbf{B}(\mathbf{r}) &= \text{curl } \mathbf{A}(\mathbf{r}) & A_{x,y,z}(\mathbf{r}) &= \frac{\mu_0}{4\pi} \int_V \frac{J_{x,y,z}(\mathbf{r}')}{|\mathbf{r} - \mathbf{r}'|} dV'. \\ &= \frac{\mu_0}{4\pi} \int_V \text{curl}_{\mathbf{r}} \left(\frac{\mathbf{J}(\mathbf{r}')}{|\mathbf{r} - \mathbf{r}'|} \right) dV' \\ &= \frac{\mu_0}{4\pi} \int_V \left(\frac{1}{|\mathbf{r} - \mathbf{r}'|} \text{curl}_{\mathbf{r}} \mathbf{J}(\mathbf{r}') - \mathbf{J}(\mathbf{r}') \times \text{grad}_{\mathbf{r}} \left(\frac{1}{|\mathbf{r} - \mathbf{r}'|} \right) \right) dV' \\ &= \frac{\mu_0}{4\pi} \int_V \frac{\mathbf{J}(\mathbf{r}') \times (\mathbf{r} - \mathbf{r}')}{|\mathbf{r} - \mathbf{r}'|^3} dV'. \end{aligned}$$

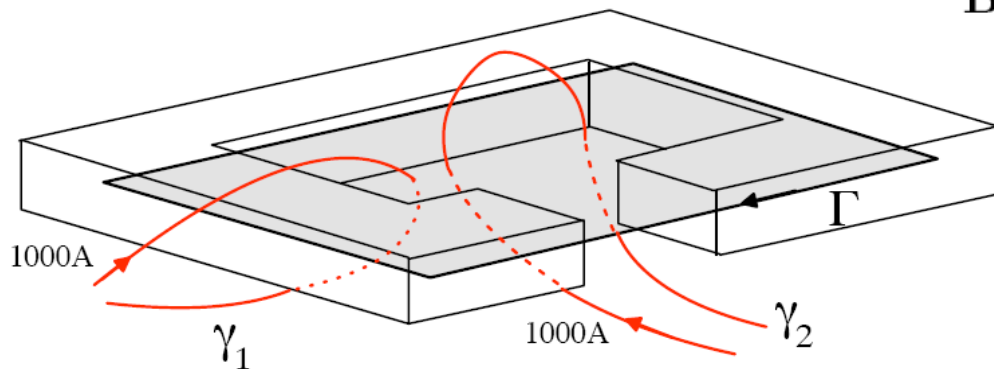
$$\begin{aligned} \text{div } \mathbf{A}(\mathbf{r}) &= \frac{\mu_0}{4\pi} \int_V \text{div} \left(\frac{\mathbf{J}(\mathbf{r}')}{|\mathbf{r} - \mathbf{r}'|} \right) dV' \\ &= \frac{\mu_0}{4\pi} \int_V \left(\mathbf{J}(\mathbf{r}') \cdot \text{grad}_{\mathbf{r}} \left(\frac{1}{|\mathbf{r} - \mathbf{r}'|} \right) + \frac{1}{|\mathbf{r} - \mathbf{r}'|} \text{div}_{\mathbf{r}} \mathbf{J}(\mathbf{r}') \right) dV' \\ &= \frac{\mu_0}{4\pi} \int_V \mathbf{J}(\mathbf{r}') \cdot \text{grad}_{\mathbf{r}} \left(\frac{1}{|\mathbf{r} - \mathbf{r}'|} \right) dV' = -\frac{\mu_0}{4\pi} \int_V \mathbf{J}(\mathbf{r}') \cdot \text{grad}_{\mathbf{r}'} \left(\frac{1}{|\mathbf{r} - \mathbf{r}'|} \right) dV' \\ &= -\frac{\mu_0}{4\pi} \int_V \left(\text{div}_{\mathbf{r}'} \left(\frac{\mathbf{J}(\mathbf{r}')}{|\mathbf{r} - \mathbf{r}'|} \right) - \frac{1}{|\mathbf{r} - \mathbf{r}'|} \text{div}_{\mathbf{r}'} \mathbf{J}(\mathbf{r}') \right) dV' \\ &= -\frac{\mu_0}{4\pi} \int_V \text{div}_{\mathbf{r}'} \left(\frac{\mathbf{J}(\mathbf{r}')}{|\mathbf{r} - \mathbf{r}'|} \right) dV' = -\frac{\mu_0}{4\pi} \int_{\partial V} \frac{\mathbf{J}(\mathbf{r}')}{|\mathbf{r} - \mathbf{r}'|} \cdot d\mathbf{a}' = 0. \end{aligned} \quad (13.2)$$

Current loops have always be closed, must not leave the problem domain

$$\mathbf{J}(\mathbf{r}')dV' = \mathbf{J}(\mathbf{r}')da'ds' = I ds'$$

$$\mathbf{A}(\mathbf{r}) = \frac{\mu_0 I}{4\pi} \oint_{\gamma} \frac{ds'}{|\mathbf{r} - \mathbf{r}'|}$$

$$\mathbf{B}(\mathbf{r}) = \frac{\mu_0 I}{4\pi} \oint_{\gamma} \frac{ds' \times (\mathbf{r} - \mathbf{r}')}{|\mathbf{r} - \mathbf{r}'|^3}$$

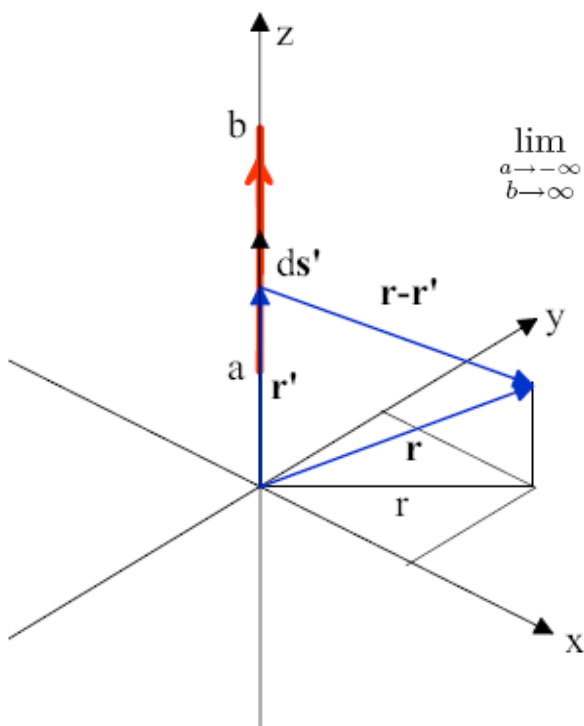


$$\int_{\partial a} \mathbf{H} \cdot ds = \int_a \mathbf{J} \cdot da = NI \text{ where } \Gamma := \partial a$$

$$NI = I \sum_{k=1}^K \text{link}(\Gamma, \gamma_k) = I \sum_{k=1}^K \text{int}(a, \gamma_k)$$

$$\text{link}(\Gamma, \gamma_k) = \frac{1}{4\pi} \oint_{\Gamma} \oint_{\gamma_k} \frac{ds' \times (\mathbf{r} - \mathbf{r}') \cdot ds}{|\mathbf{r} - \mathbf{r}'|^3}$$

$$\begin{aligned}
 A_z &= \frac{\mu_0 I}{4\pi} \int_a^b \frac{dz'}{|\mathbf{r} - \mathbf{r}'|} = \frac{\mu_0 I}{4\pi} \int_a^b \frac{dz'}{\sqrt{x^2 + y^2 + (z - z')^2}} \\
 &= \frac{-\mu_0 I}{4\pi} \left[\ln \left((z - z') + \sqrt{x^2 + y^2 + (z - z')^2} \right) \right]_a^b \\
 &= \frac{\mu_0 I}{4\pi} \ln \frac{z - a + \sqrt{x^2 + y^2 + (z - a)^2}}{z - b + \sqrt{x^2 + y^2 + (z - b)^2}}.
 \end{aligned}$$



$$\begin{aligned}
 \lim_{\substack{a \rightarrow -\infty \\ b \rightarrow \infty}} \ln \frac{z - a + \sqrt{x^2 + y^2 + (z - a)^2}}{z - b + \sqrt{x^2 + y^2 + (z - b)^2}} &= \lim_{\substack{a \rightarrow -\infty \\ b \rightarrow \infty}} \ln \frac{-a + |a| \sqrt{1 + \frac{x^2 + y^2}{a^2}}}{-b + |b| \sqrt{1 + \frac{x^2 + y^2}{b^2}}} \\
 &= \lim_{\substack{a \rightarrow -\infty \\ b \rightarrow \infty}} \ln \frac{-a - a \left(1 + \frac{x^2 + y^2}{2a^2} + \dots\right)}{-b + b \left(1 + \frac{x^2 + y^2}{2b^2} + \dots\right)} \\
 &= \lim_{\substack{a \rightarrow -\infty \\ b \rightarrow \infty}} \ln \frac{-2a}{-b + b + \frac{x^2 + y^2}{2b}} \\
 &= \lim_{\substack{a \rightarrow -\infty \\ b \rightarrow \infty}} \ln \frac{-4ab}{x^2 + y^2},
 \end{aligned}$$

Caution: Infinitely long line currents have infinite energy

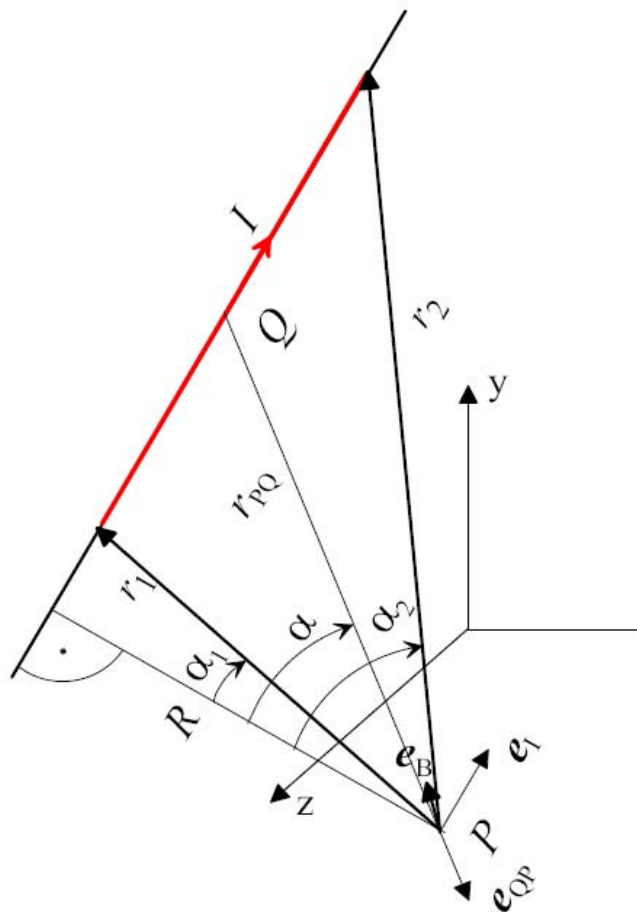
$$A_z = \lim_{\substack{a \rightarrow -\infty \\ b \rightarrow \infty}} \frac{\mu_0 I}{4\pi} \ln \frac{-4ab}{x^2 + y^2}$$

$$= \lim_{\substack{a \rightarrow -\infty \\ b \rightarrow \infty}} \frac{\mu_0 I}{4\pi} \ln \left(\frac{-4ab}{x_0^2 + y_0^2} \right) - \frac{\mu_0 I}{4\pi} \ln \left(\frac{x^2 + y^2}{x_0^2 + y_0^2} \right)$$

$$\mathbf{A} = -\frac{\mu_0 I}{4\pi} \ln \left(\frac{x^2 + y^2}{x_0^2 + y_0^2} \right) \mathbf{e}_z = -\frac{\mu_0 I}{2\pi} \ln \left(\frac{r}{R_{\text{ref}}} \right) \mathbf{e}_z$$

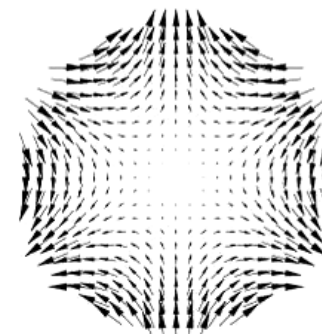
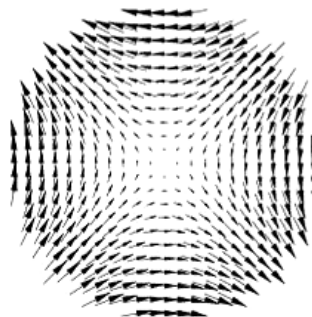
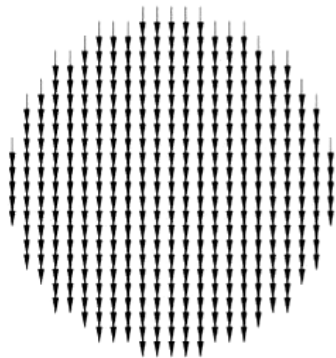
**Problem solved, but reference radius has physical significance:
Return path for sum-currents**

$$r_{PQ} = \frac{\vec{R}}{\cos \alpha}, ds = \frac{R}{\cos^2 \alpha} d\alpha, \mathbf{e}_I \times \mathbf{e}_{QP} = \cos \alpha \mathbf{e}_B$$



$$\begin{aligned} \mathbf{B} &= \frac{\mu_0 I}{4\pi R} [\sin \alpha_2 - \sin \alpha_1] \mathbf{e}_B \\ &= \frac{\mu_0 I}{4\pi} \frac{\cos \alpha_2 + \cos \alpha_1}{R} \frac{\sin \alpha_2 - \sin \alpha_1}{\cos \alpha_2 + \cos \alpha_1} \mathbf{e}_B \\ &= \frac{\mu_0 I}{4\pi} \left(\frac{1}{|\mathbf{r}_1|} + \frac{1}{|\mathbf{r}_2|} \right) \frac{\sin(\alpha_2 - \alpha_1)}{1 + \cos(\alpha_2 - \alpha_1)} \mathbf{e}_B \\ &= \frac{\mu_0 I}{4\pi} \left(\frac{1}{|\mathbf{r}_1|} + \frac{1}{|\mathbf{r}_2|} \right) \frac{\sin(\alpha_2 - \alpha_1)}{1 + \frac{\mathbf{r}_1 \cdot \mathbf{r}_2}{|\mathbf{r}_1| |\mathbf{r}_2|}} \frac{\mathbf{r}_1 \times \mathbf{r}_2}{|\mathbf{r}_1| |\mathbf{r}_2| \sin(\alpha_2 - \alpha_1)} \\ &= \frac{\mu_0 I}{4\pi} \frac{|\mathbf{r}_1| + |\mathbf{r}_2|}{|\mathbf{r}_1| |\mathbf{r}_2| + \mathbf{r}_1 \cdot \mathbf{r}_2} \frac{\mathbf{r}_1 \times \mathbf{r}_2}{|\mathbf{r}_1| |\mathbf{r}_2|} \end{aligned}$$

Easy to program equation. Still remember: Closed loops



$$\varphi_k = \frac{2\pi k}{N}, \quad k = 0, 1, 2, \dots, N - 1$$

$$A_n(r_0) \approx \frac{2}{N} \sum_{k=0}^{N-1} B_r(r_0, \varphi_k) \cos n\varphi_k$$

$$B_n(r_0) \approx \frac{2}{N} \sum_{k=0}^{N-1} B_r(r_0, \varphi_k) \sin n\varphi_k.$$

Discrete Fourier Transform

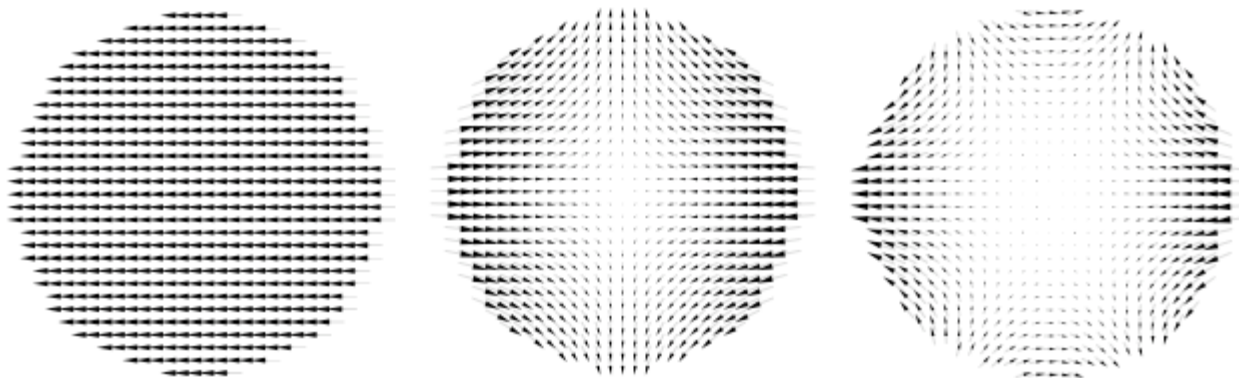
$$B_r(r_0, \varphi) = \sum_{n=1}^{\infty} (B_n(r_0) \sin n\varphi + A_n(r_0) \cos n\varphi)$$

Don't bother with the Fast Fourier Transform, but consider to use the vector-potential directly

$$A_z(r_0, \varphi) = \sum_{n=1}^{\infty} (\mathcal{F}_n(r_0) \cos n\varphi + \mathcal{E}_n(r_0) \sin n\varphi)$$

$$B_n(r_0) = \frac{-n \mathcal{F}_n}{r_0}, \quad A_n(r_0) = \frac{n \mathcal{E}_n}{r_0}$$

And please: Never use holomorphic continuation, it's just to inaccurate



$$B_R(R_0, \vartheta) = \sum_{n=1}^{\infty} A_n P_n(\cos \vartheta)$$

$$A_n(R_0) = \frac{2n+1}{2} \int_0^{\pi} B_R(R_0, \vartheta) P_n(\cos \vartheta) \sin \vartheta \, d\vartheta$$

$$P_n(x) = \sum_{m=0}^M (-1)^m \frac{(2n-2m)!}{2^n r! (n-m)! (n-2r)!} x^{n-2m}$$

To derive this result would take 5 pages. So take it from the “horse’s mouth” (and don’t do Fourier series analysis).

$$X_p(x) = \mathcal{A}_p \cosh px + \mathcal{B}_p \sinh px \qquad \phi_m(x, y) = \sum_{p=0}^{\infty} X_p(x) Y_p(y)$$

$$Y_p(y) = \mathcal{C}_p \cos py + \mathcal{D}_p \sin py,$$

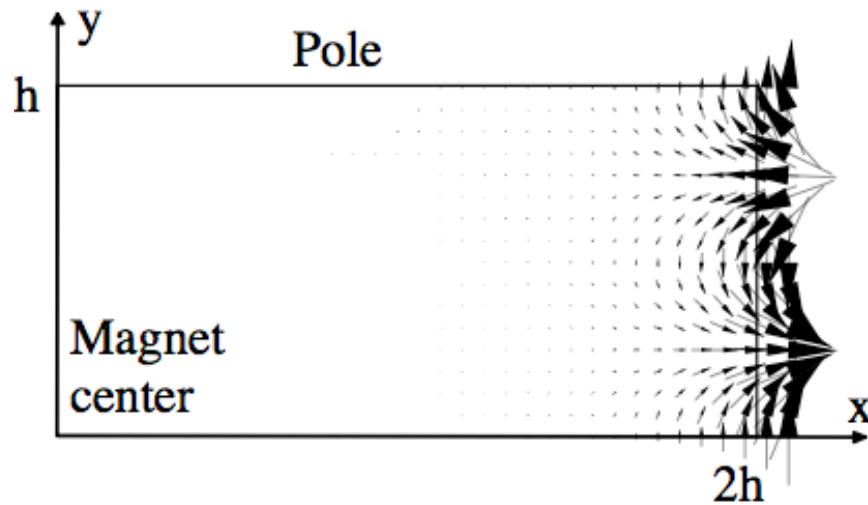
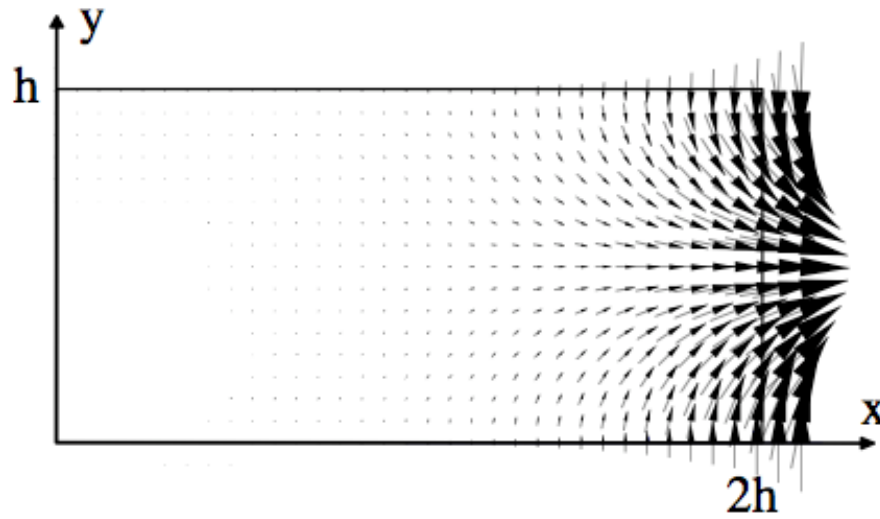
$$B_y(-x, y) = B_y(x, y) \quad \rightarrow \quad X_p(x) = X_p(-x)$$

$$B_y(x, -y) = B_y(x, y) \quad \rightarrow \quad \left. \frac{dY_p(y)}{dy} \right|_{y=-y_0} = \left. \frac{dY_p(y)}{dy} \right|_{y=y_0}$$

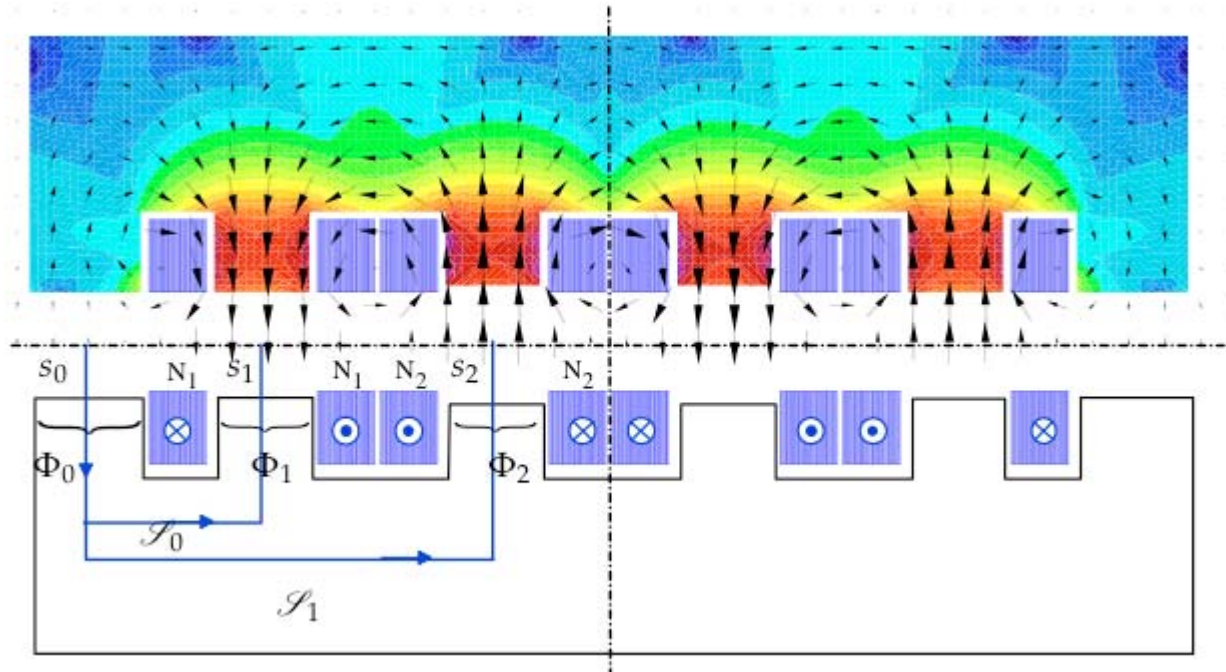
$$B_x = \sum_{p=1}^{\infty} \mathcal{U}_p \sinh px \sin py \qquad B_x(x, \pm h) = 0, \qquad p = \frac{n\pi}{h}$$

$$B_x(x, y) = \sum_{n=1}^{\infty} \mathcal{U}_n \sinh \left(\frac{n\pi}{h} x \right) \sin \left(\frac{n\pi}{h} y \right),$$

$$B_y(x, y) = \mathcal{U}_0 + \sum_{n=1}^{\infty} \mathcal{U}_n \cosh \left(\frac{n\pi}{h} x \right) \cos \left(\frac{n\pi}{h} y \right)$$



$$\Phi_i = \sum_{j=1}^n L_{ij} I_j$$



$$U = L^d \frac{dI}{dt}$$

$$U(t) = \frac{d\Phi}{dt} = \frac{d(LI)}{dt} = L \frac{dI}{dt} + I \frac{dL}{dt}$$

$$dL = \frac{\partial L}{\partial I} dI + \frac{\partial L}{\partial t} dt$$

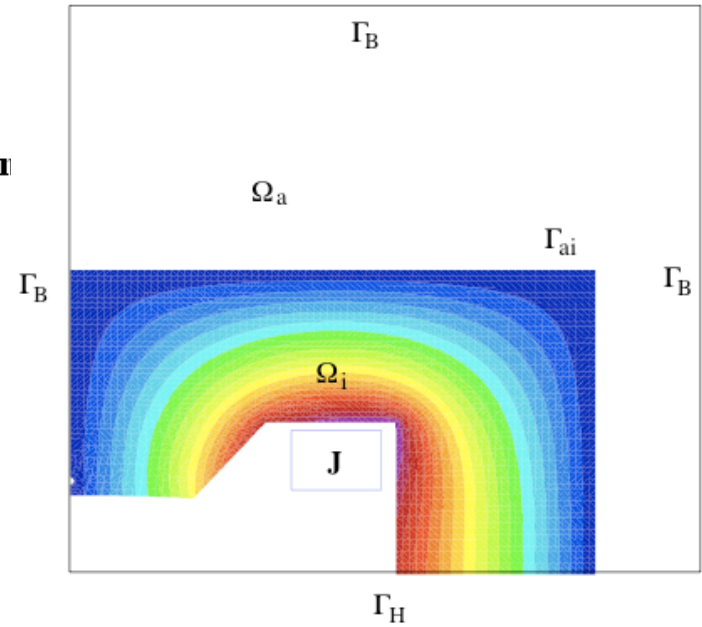
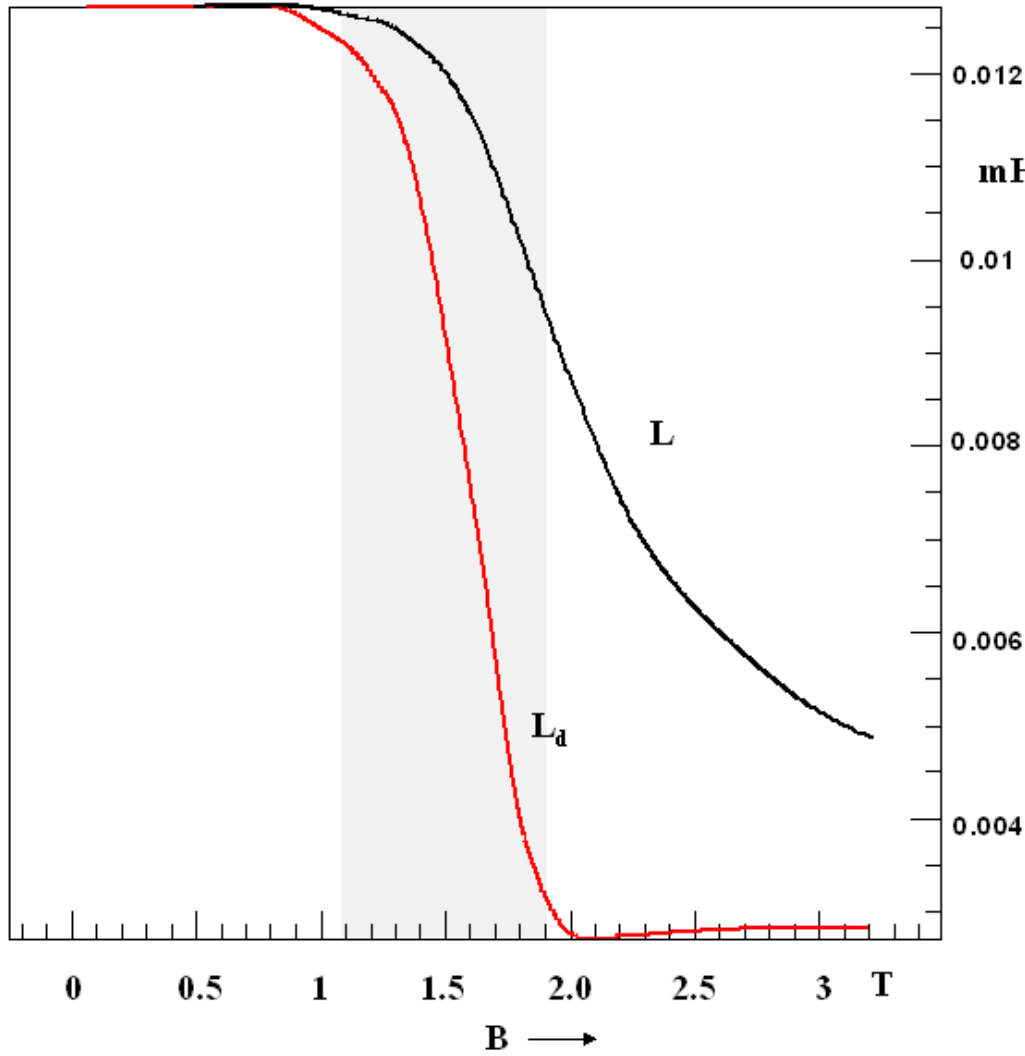
$$U(t) = \left(\frac{\partial L}{\partial I} I + L \right) \frac{dI}{dt} + I \frac{\partial L}{\partial t}$$

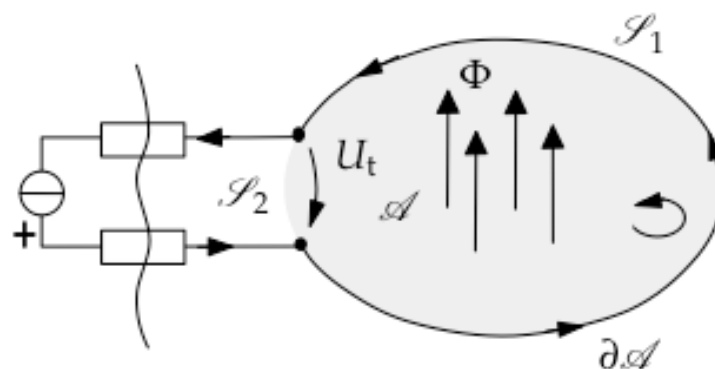
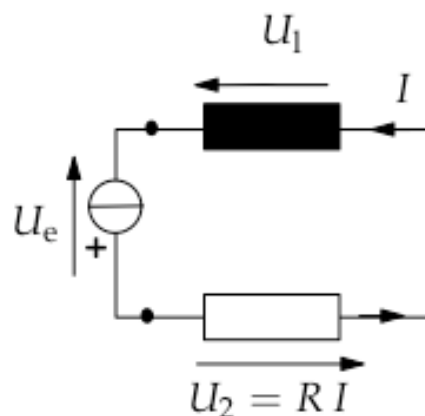
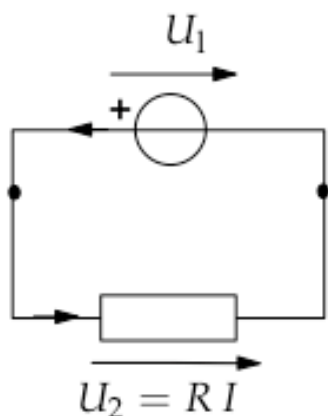
$$L^d = L + I \frac{\partial L}{\partial I} = \frac{d\Phi}{dI}$$

$$W = \frac{1}{2} L^W I^2$$

$$\begin{aligned} W &= \int_0^t U I d\tau = \int_0^t \frac{d\Phi(I(\tau))}{d\tau} I(\tau) d\tau = \int_0^t \frac{d\Phi(I(\tau))}{dI} \frac{dI(\tau)}{d\tau} I(\tau) d\tau \\ &= \int_0^{I(t)} \frac{d\Phi}{dI} I dI = \int_0^{I(t)} L^d I dI, \end{aligned}$$

$$L^W = \frac{2}{I(t)^2} \int_0^{I(t)} L^d I dI = \int_0^1 2\lambda L^d(\lambda I) d\lambda$$





$$U(\partial \mathcal{A}) = - \frac{d}{dt} \Phi(\mathcal{A})$$

Faradays law

Generator case

$$U_1 = U_2 = U_t = - \frac{d}{dt} \Phi(\mathcal{A})$$

Load case

$$U_1 = U_e - U_2 = -U_t = \frac{d}{dt} \Phi(\mathcal{A})$$

$$\mathbf{F}_m = \int_{\mathcal{V}} \frac{1}{\mu_0} (\text{curl } \mathbf{B}) \times \mathbf{B} dV$$

$$(\text{curl } \mathbf{B}) \times \mathbf{B} = \text{div } \mathbf{S}_m - \mathbf{B} \text{ div } \mathbf{B}$$

$$\mathbf{S}_m := (\sigma_{ij}) = \begin{pmatrix} B_x^2 - \frac{1}{2}|\mathbf{B}|^2 & B_x B_y & B_x B_z \\ B_y B_x & B_y^2 - \frac{1}{2}|\mathbf{B}|^2 & B_y B_z \\ B_z B_x & B_z B_y & B_z^2 - \frac{1}{2}|\mathbf{B}|^2 \end{pmatrix}$$

$$\mathbf{F}_m = \int_{\mathcal{V}} \frac{1}{\mu_0} \text{div } \mathbf{S}_m dV = \int_{\partial\mathcal{V}} \frac{1}{\mu_0} \mathbf{S}_m \cdot \mathbf{n} da$$

$$\mathbf{S}_m \cdot \mathbf{n} = (\mathbf{B} \cdot \mathbf{n})\mathbf{B} - \frac{1}{2}|\mathbf{B}|^2 \mathbf{n}$$

$$\mathbf{F}_m = \int_{\partial\mathcal{V}} \left(\frac{1}{\mu_0} (\mathbf{B} \cdot \mathbf{n})\mathbf{B} - \frac{1}{2\mu_0} |\mathbf{B}|^2 \mathbf{n} \right) da$$

$$X \subseteq \mathbb{R}^n \quad (x_1, x_2, \dots, x_n)^T \in X$$

$$\min\{f(\mathbf{x})\} \quad f : X \rightarrow \mathbb{R} \quad \text{Subject to}$$

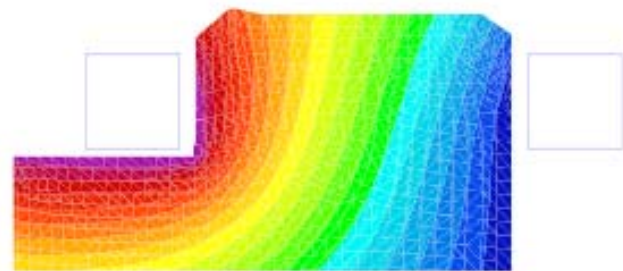
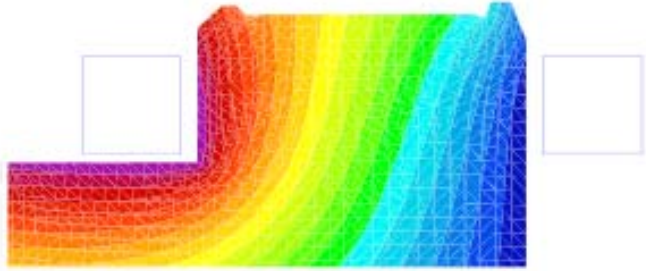
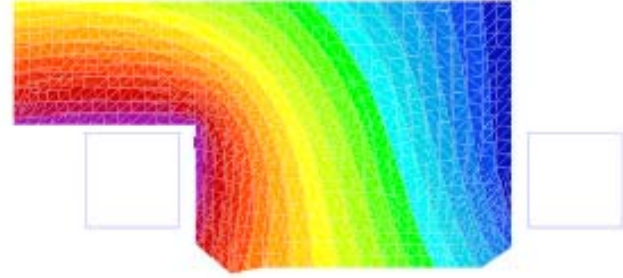
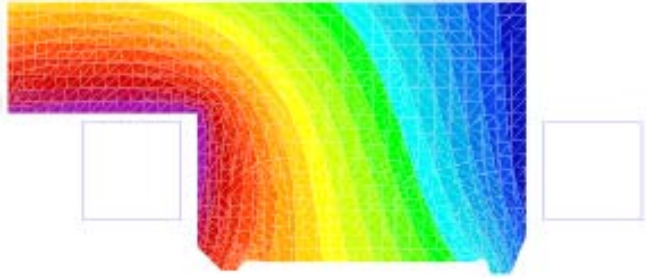
$$g_i(\mathbf{x}) \leq 0, \quad i = 1, 2, \dots, m,$$

$$h_j(\mathbf{x}) = 0, \quad j = 1, 2, \dots, p,$$

$$x_{l,\text{lower}} < x_l < x_{l,\text{upper}}, \quad l = 1, 2, \dots, n$$

$$M = \{\mathbf{x} \in X \mid g_i(\mathbf{x}) \leq 0; h_j(\mathbf{x}) = 0; x_{l,\text{lower}} < x_l < x_{l,\text{upper}}; \\ \forall i = 1, \dots, m; j = 1, \dots, p; l = 1, \dots, n\}.$$

$$\min\{f(\mathbf{x}) \mid \mathbf{x} \in M\}$$

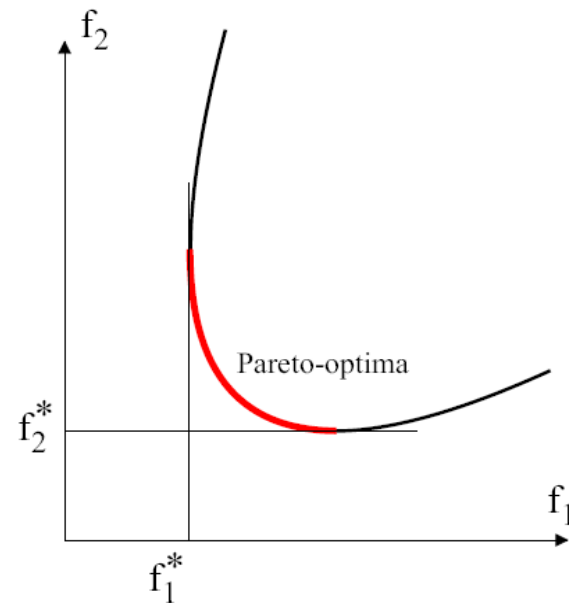
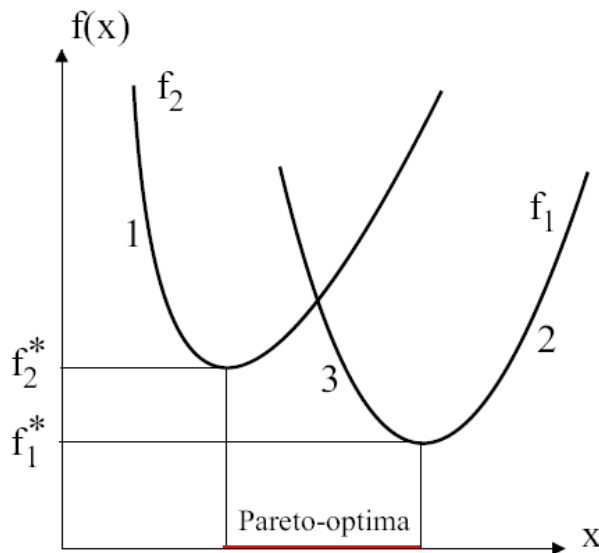


$$\text{MIN } \{\mathbf{f}(\mathbf{x})\} = \text{MIN } \{f_1(\mathbf{x}), f_2(\mathbf{x}), \dots, f_K(\mathbf{x})\}$$

A Pareto optimal solution \mathbf{x}^* is given if there exists no solution with

$$f_k(\mathbf{x}) \leq f_k(\mathbf{x}^*) \quad \forall k \in [1, K],$$

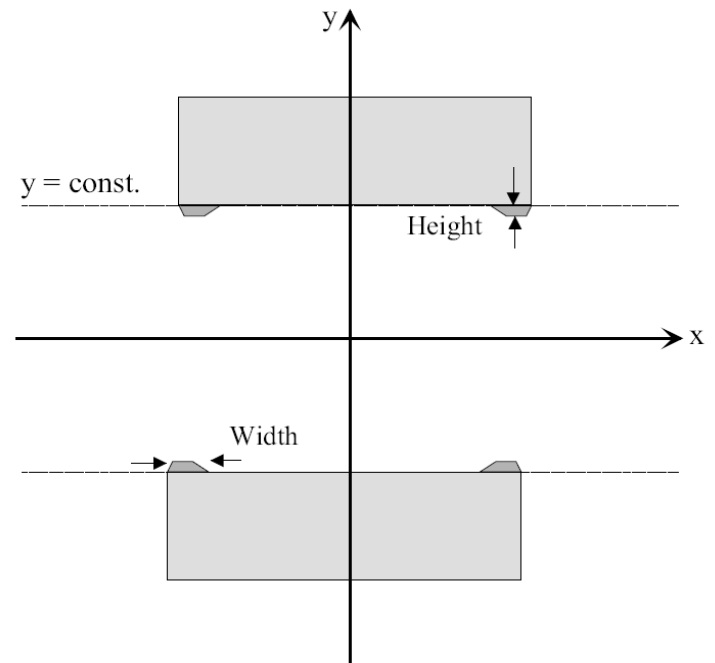
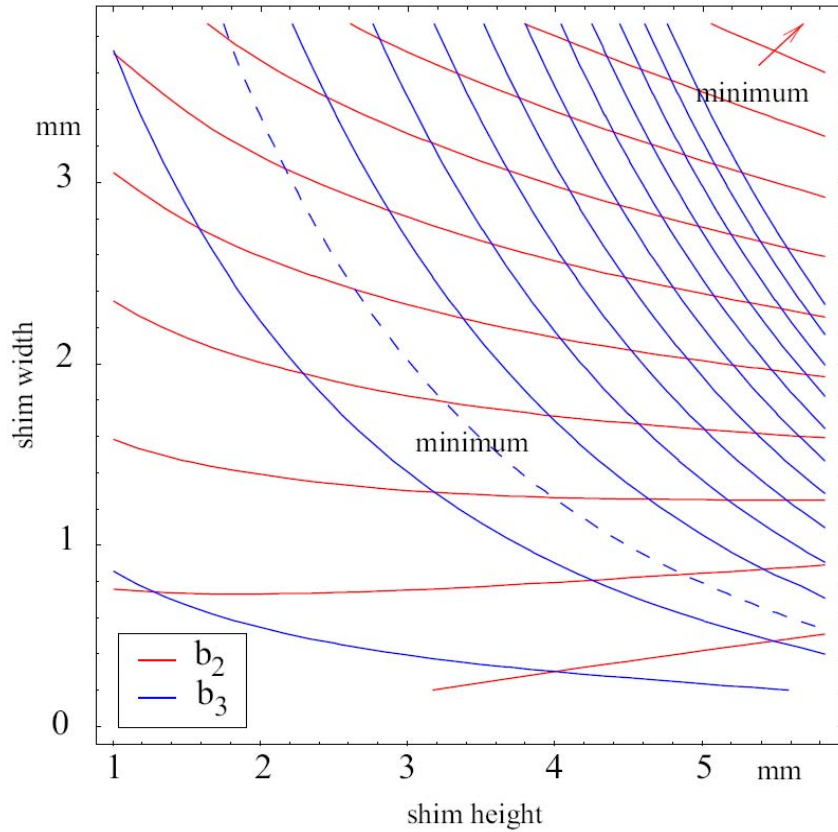
$$f_k(\mathbf{x}) < f_k(\mathbf{x}^*) \quad \text{for at least one } k \in [1, K]$$



- There are only Pareto-optimal solutions
 - Decision making
 - Treatment of nonlinear constraints
 - Optimization algorithms

- The objective conflict is the characteristic of real world optimization problems

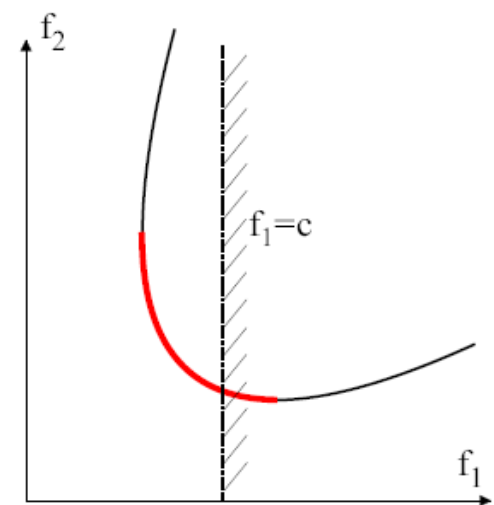
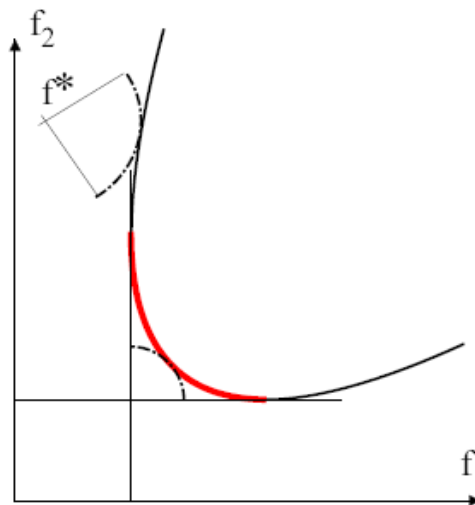
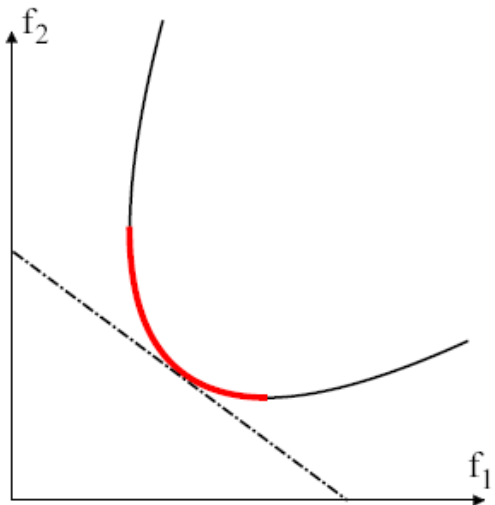
- Fuzzy objectives in the concept phase



$$\min \left\{ u(\mathbf{f}(\mathbf{x})) := \sum_{k=1}^K t_k f_k(\mathbf{x}) \mid \mathbf{x} \in M \right\}$$

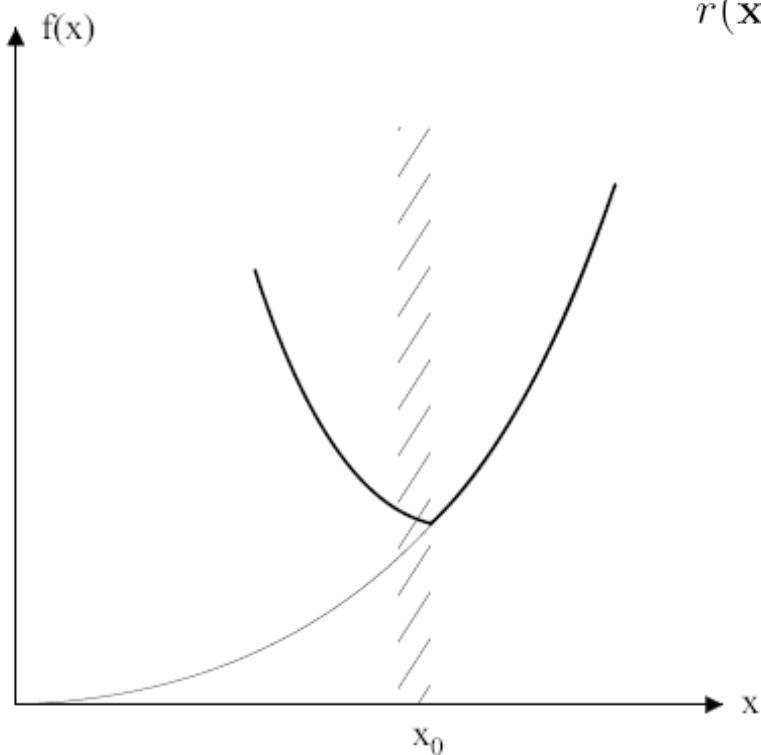
$$\min \left\{ \|\mathbf{z}(\mathbf{x})\|_2^2 := \sum_{k=1}^K (t_k (f_k^*(\mathbf{x}) - f_k(\mathbf{x})))^2 \mid \mathbf{x} \in M \right\}$$

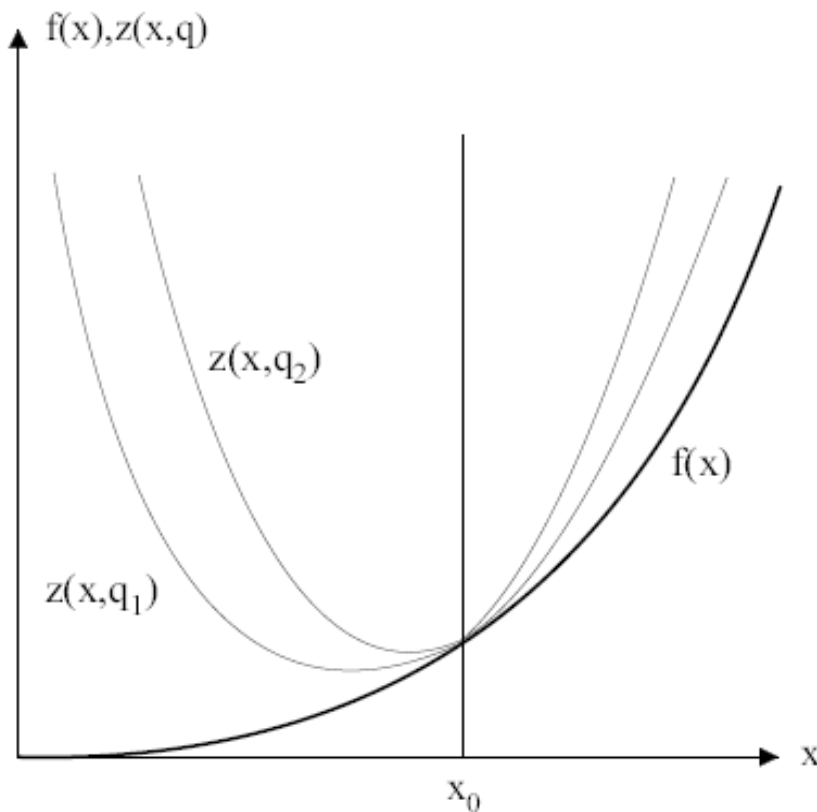
$$\min \{ f_i(\mathbf{x}) \} \quad \text{s.t.} \quad f_k(\mathbf{x}) - r_k \leq 0$$



$$p(\mathbf{x}) = \begin{cases} f(\mathbf{x}) & \text{no bound violated} \\ f(\mathbf{x}^*) + r(\mathbf{x}) & \text{bound violated} \end{cases}$$

$$r(\mathbf{x}) = \sum_l r_l \begin{cases} (x_l - x_{l,\text{upper}})^2 & \text{if } x_l > x_{l,\text{upper}} \\ (x_{l,\text{lower}} - x_l)^2 & \text{if } x_l < x_{l,\text{lower}} \\ 0 & \text{otherwise} \end{cases}$$





$$z(\mathbf{x}, \mathbf{p}, \mathbf{q}) = f_i(\mathbf{x}) + \sum_{k=1}^{m+K-1} p_k \cdot \max^2\{(0, g_k(\mathbf{x}) - d_k)\} + \sum_{j=1}^p q_j (h_j(\mathbf{x}) - c_j)^2$$

| Search methods | | |
|---------------------------------|-------------------------|------|
| Direct search | Gauss-Seidel | |
| EXTREM | Jacob | 1982 |
| Rosenbrock | Rosenbrock | 1960 |
| Powell | Powell | 1965 |
| Flexible Polyhedron search | Nelder-Mead | 1964 |
| Hooke-Jeeves | Hooke-Jeeves | 1962 |
| Gradient methods | | |
| Steepest descend | Cauchy | 1847 |
| Newton's method | Newton | 1700 |
| Levenberg-Marquard | Levenberg Marquard | 1963 |
| Conjugate gradient (CG) | Fletcher-Reeves | 1964 |
| Quasi-Newton | Davidon-Fletcher-Powell | 1959 |
| Stochastic and neural computing | | |
| Evolutionary | Rechenberg | 1964 |
| Genetic algorithms | Fogel-Holland | 1987 |
| Neural computing (ANN) | Aarts-Korst | 1989 |

