



# An overview of current optimization and physical bounds on antennas

Mats Gustafsson

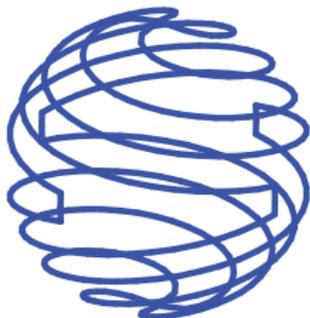
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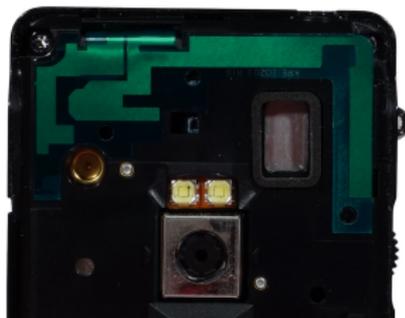
Lund University, Sweden

# Design of small antennas

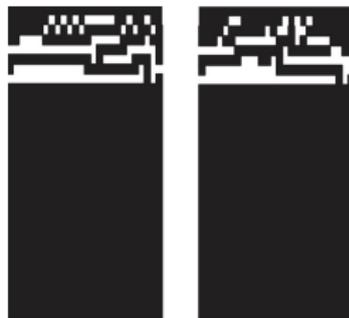
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Folded spherical helix



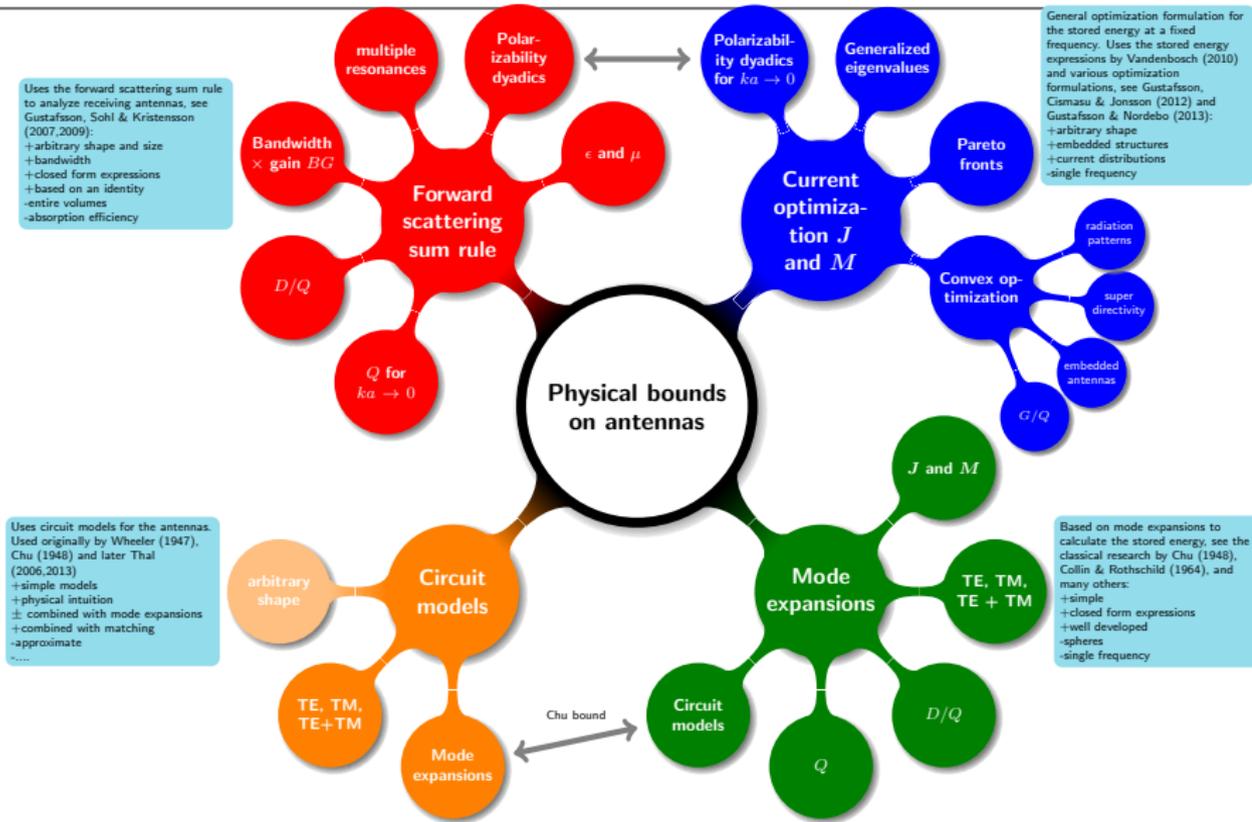
SonyEricsson P1i



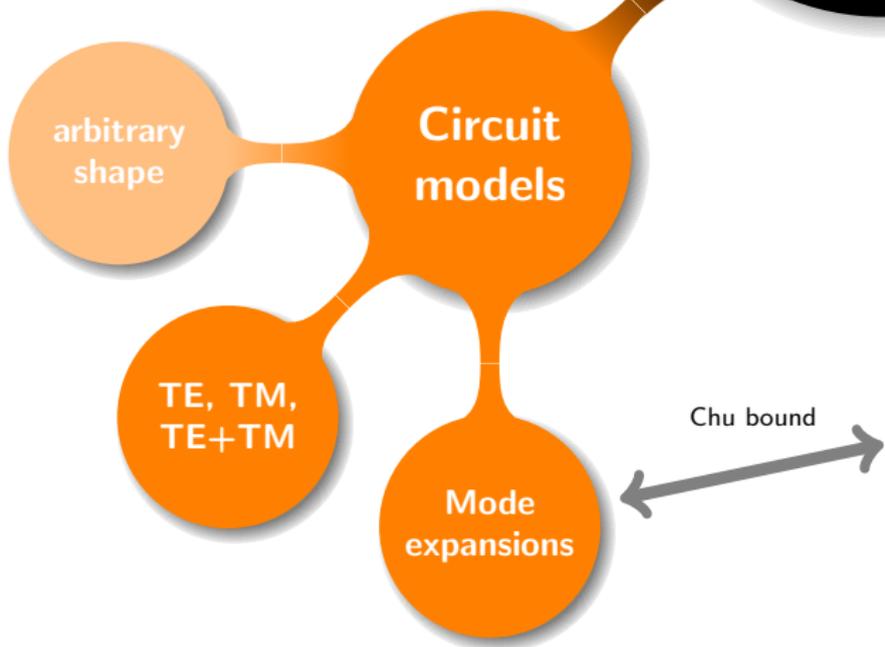
Fragmented patches

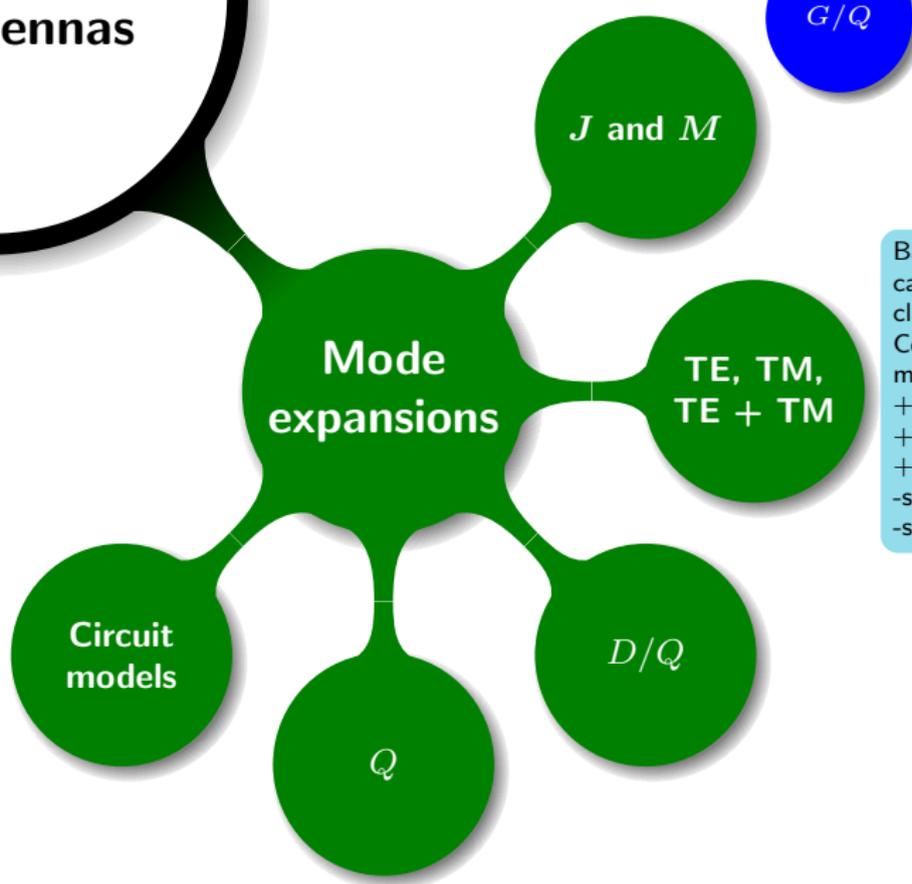
- ▶ There are many advanced methods to design small antennas.
- ▶ Often antennas embedded in structures.
- ▶ Performance in Q, bandwidth and efficiency.
- ▶ How does the performance depend on the design volume?
- ▶ What can we learn from performance bounds and optimal currents?
- ▶ Can we automate the design of optimal antennas?

# Physical bounds on antennas: methods



Uses circuit models for the antennas.  
Used originally by Wheeler (1947),  
Chu (1948) and later Thal  
(2006,2013)  
+simple models  
+physical intuition  
± combined with mode expansions  
+combined with matching  
-approximate  
-....



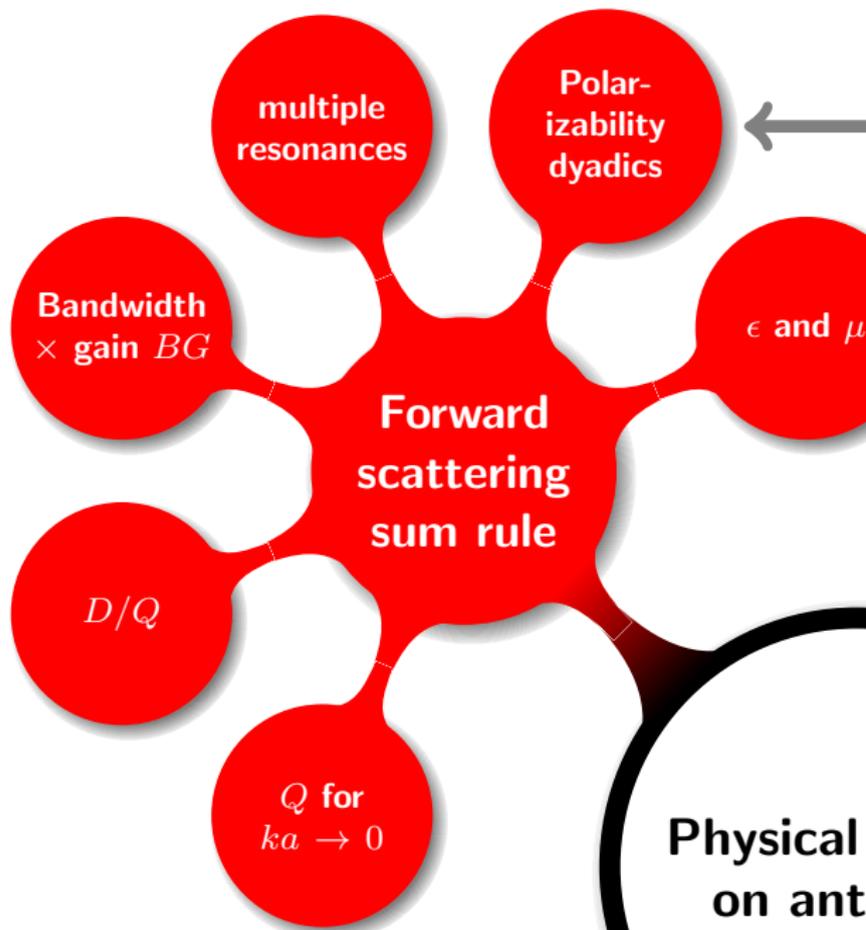


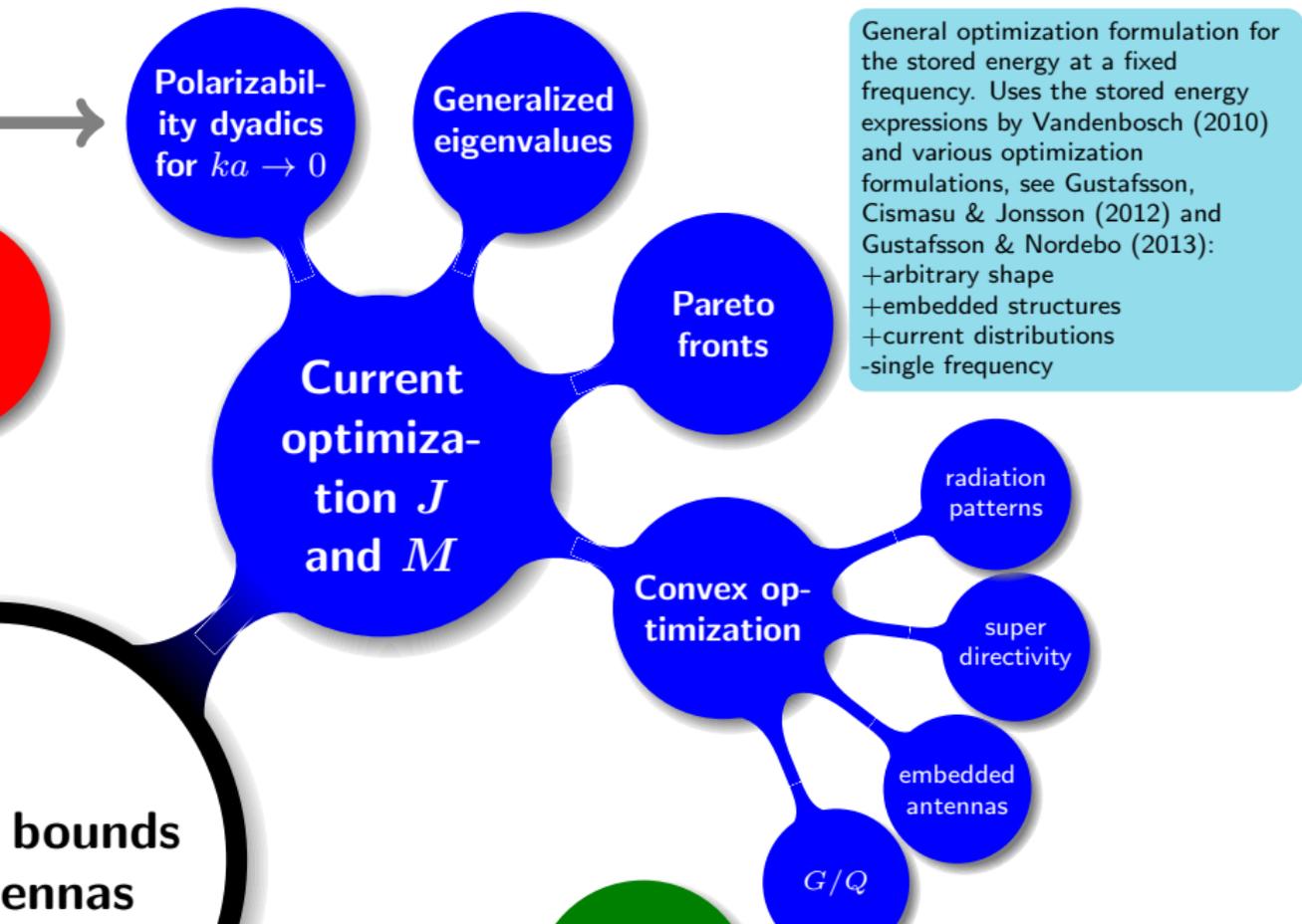
Based on mode expansions to calculate the stored energy, see the classical research by Chu (1948), Collin & Rothschild (1964), and many others:

- +simple
- +closed form expressions
- +well developed
- spheres
- single frequency

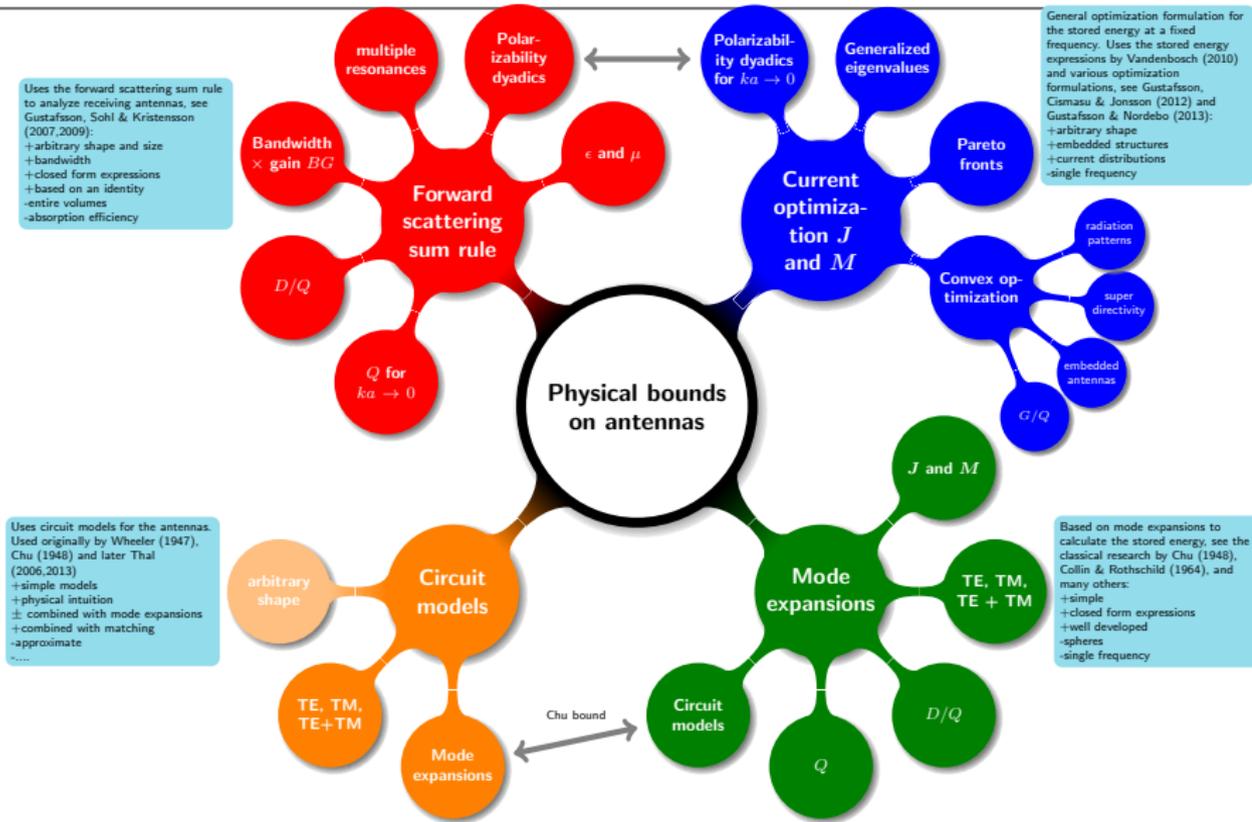
Uses the forward scattering sum rule to analyze receiving antennas, see Gustafsson, Sohl & Kristensson (2007,2009):

- +arbitrary shape and size
- +bandwidth
- +closed form expressions
- +based on an identity
- entire volumes
- absorption efficiency





# Physical bounds on antennas: methods



# Antenna and current optimization

Antenna design: produce the desired current distribution on the structure by shaping and choosing the materials.

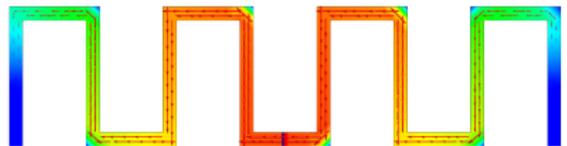
- ▶ Have a given maximal size of the antenna structure.
- ▶ Antenna optimization: determine the shape and material properties for optimal performance.
- ▶ Current optimization: determine an optimal current distribution from all possible currents in the available geometry.



Maximal size of the antenna.

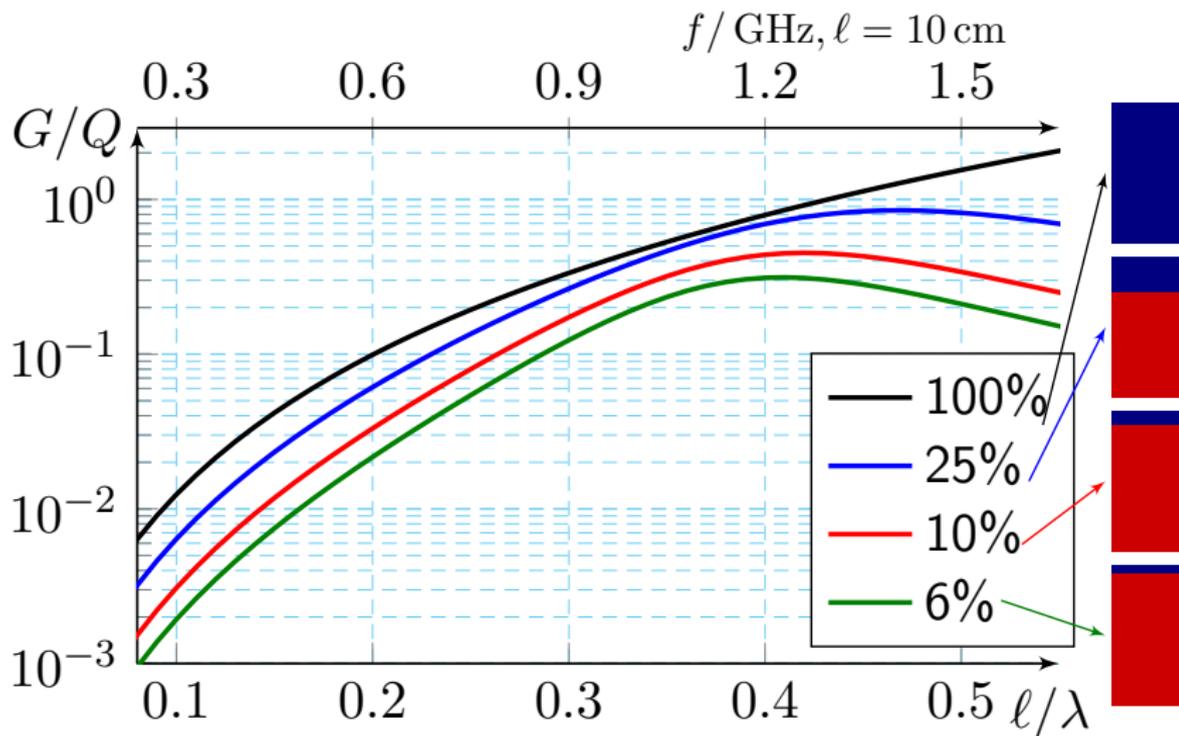


Antenna geometry with feed point.

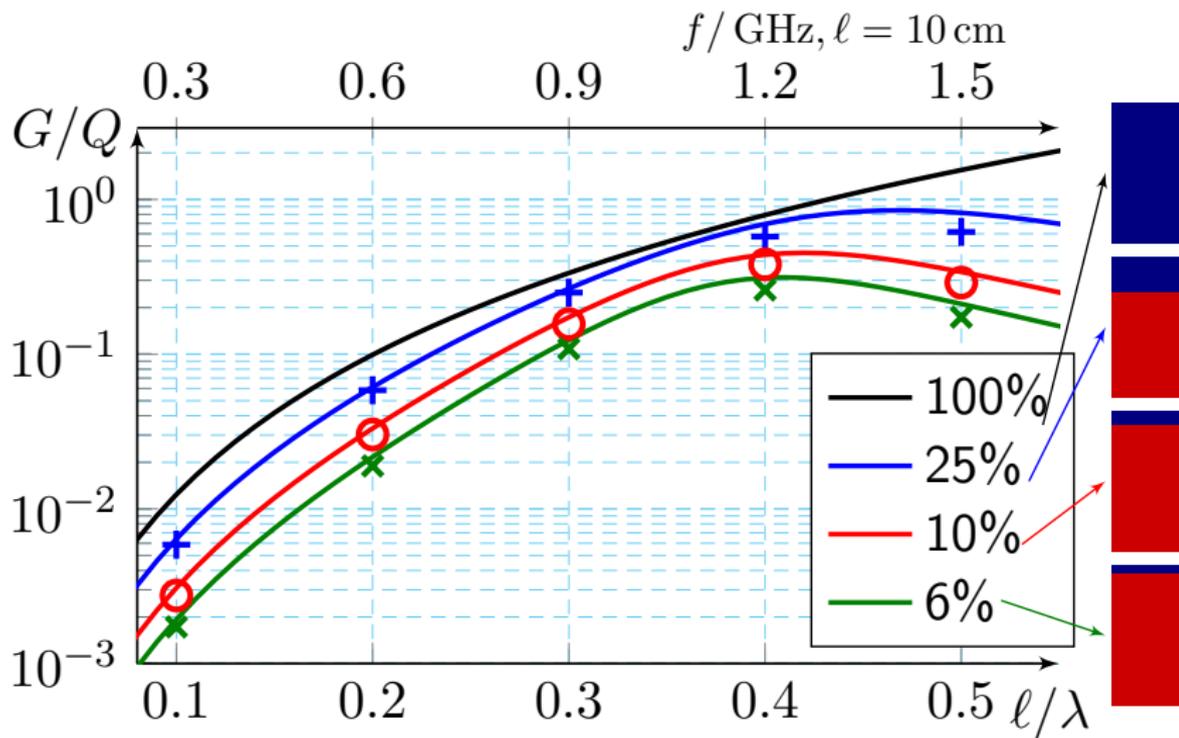


Current distribution on the antenna.

# Finite ground plane with {6, 10, 25, 100}% antenna region



# Finite ground plane with $\{6, 10, 25, 100\}\%$ antenna region



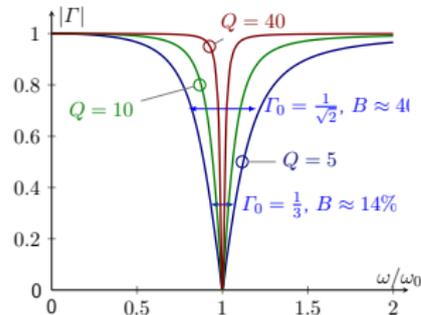
# Q-factor and single frequency evaluation

The Q-factor is defined as the ratio between the stored electric,  $W_e$ , and magnetic,  $W_m$ , energies and the dissipated power, *i.e.*,

$$Q = \frac{2\omega \max\{W_e, W_m\}}{P_{\text{rad}} + P_{\text{loss}}}.$$

Fractional bandwidth for single resonances

$$B \approx \frac{2}{Q} \frac{\Gamma_0}{\sqrt{1 - \Gamma_0^2}}$$



Reflection coefficient

$|\Gamma|$  for a RCL circuit  
with Q-factors

$Q = \{5, 10, 40\}$ .

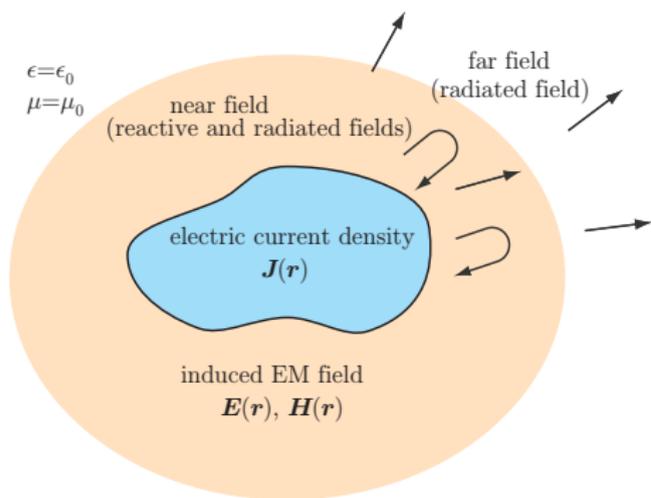
Fractional bandwidths  
for  $\Gamma_0 = \{1/\sqrt{2}, 1/3\}$ .

## Single frequency evaluation

use the Q-factor to estimate the bandwidth.  
Need to:

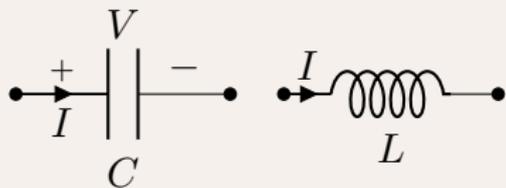
1. Compute the *stored energy*.
2. Solve the optimization problems.

# What is (stored) EM energy?



- ▶ Time average energy density  $\epsilon_0 |\mathbf{E}|^2/4$  and  $\mu_0 |\mathbf{H}|^2/4$ .
- ▶ What is stored and radiated?
- ▶ How can we express the (stored) energy in the current (density)?
- ▶ First, currents in free space.

## Lumped elements

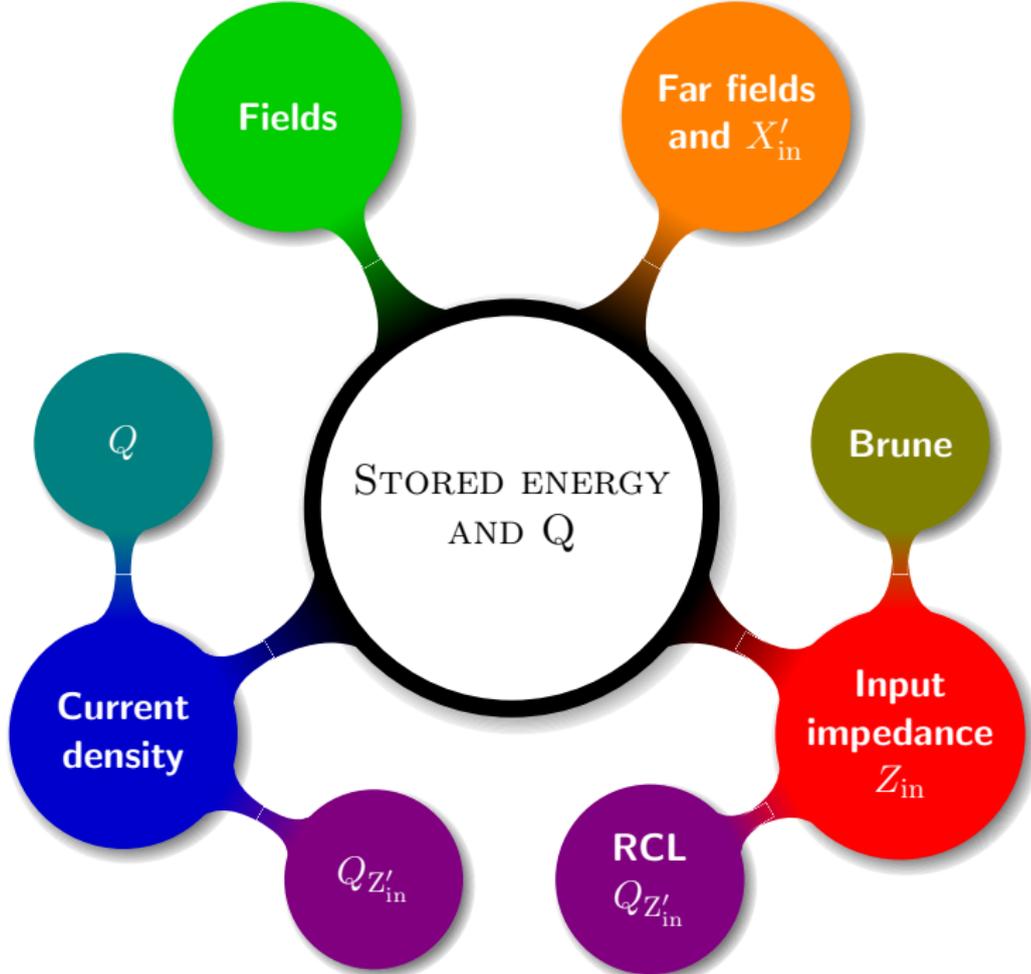


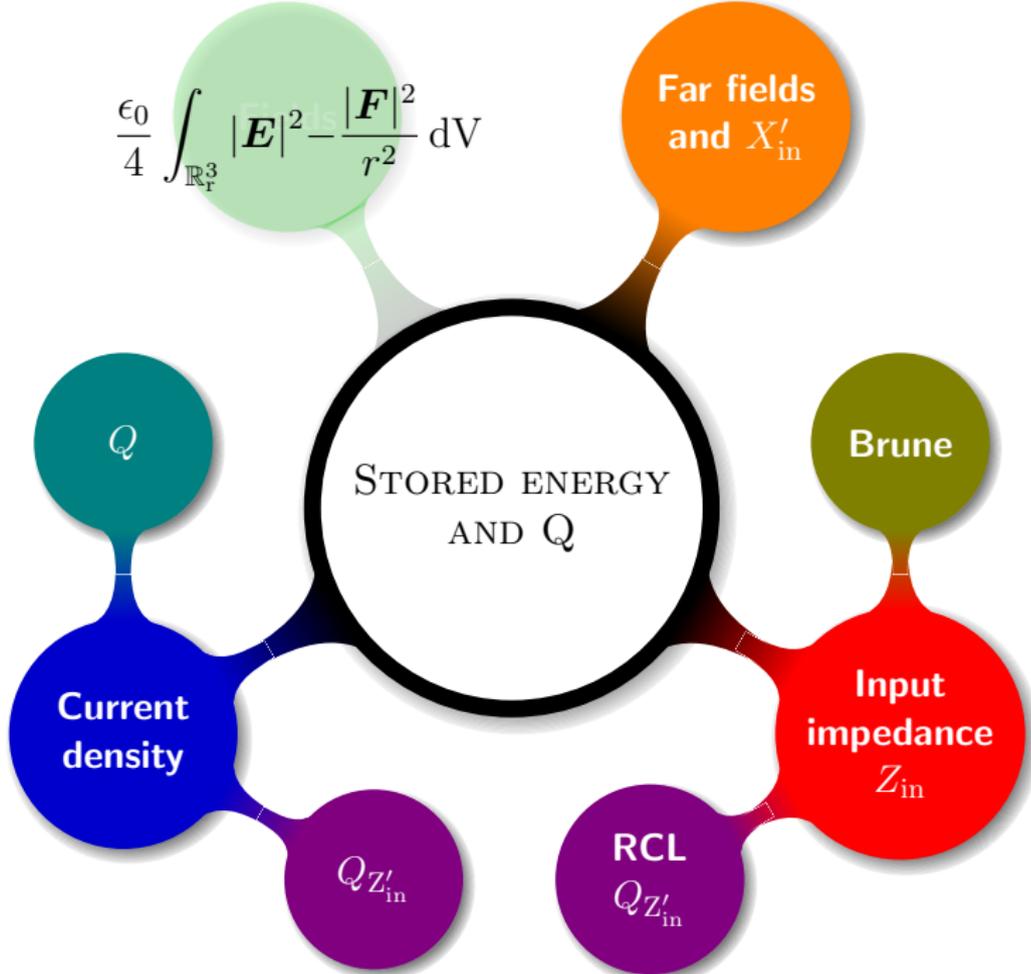
Time average stored energy in capacitors

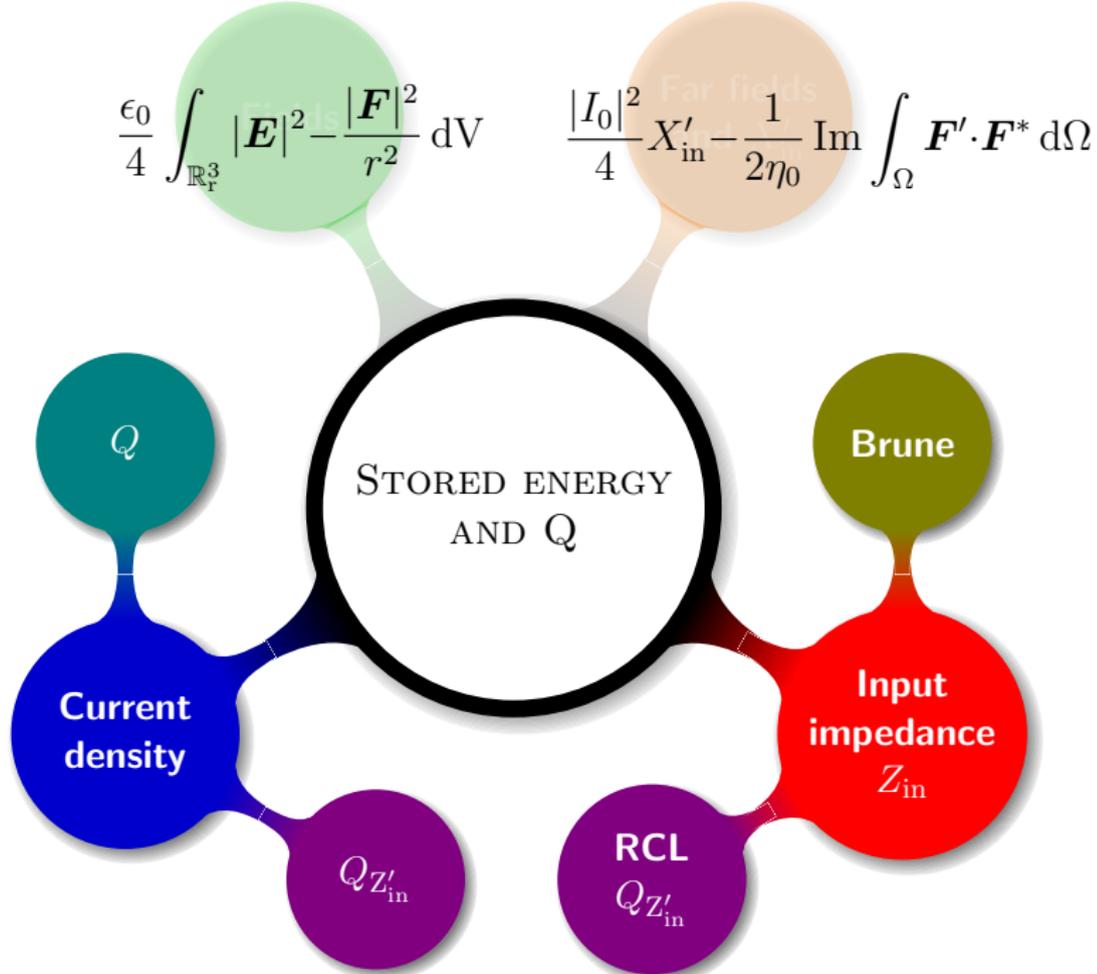
$$W_e = \frac{C|V|^2}{4} = \frac{|I|^2}{4\omega^2 C}$$

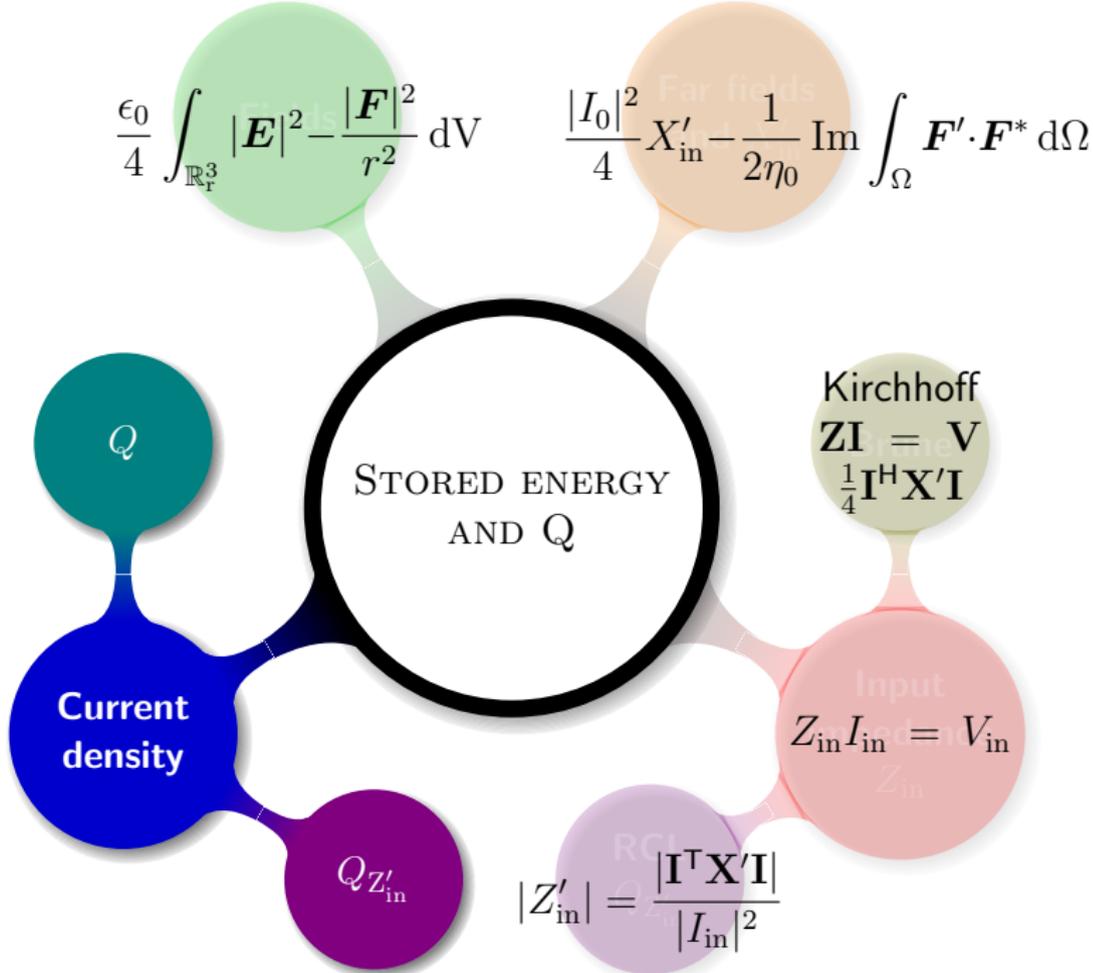
and in inductors

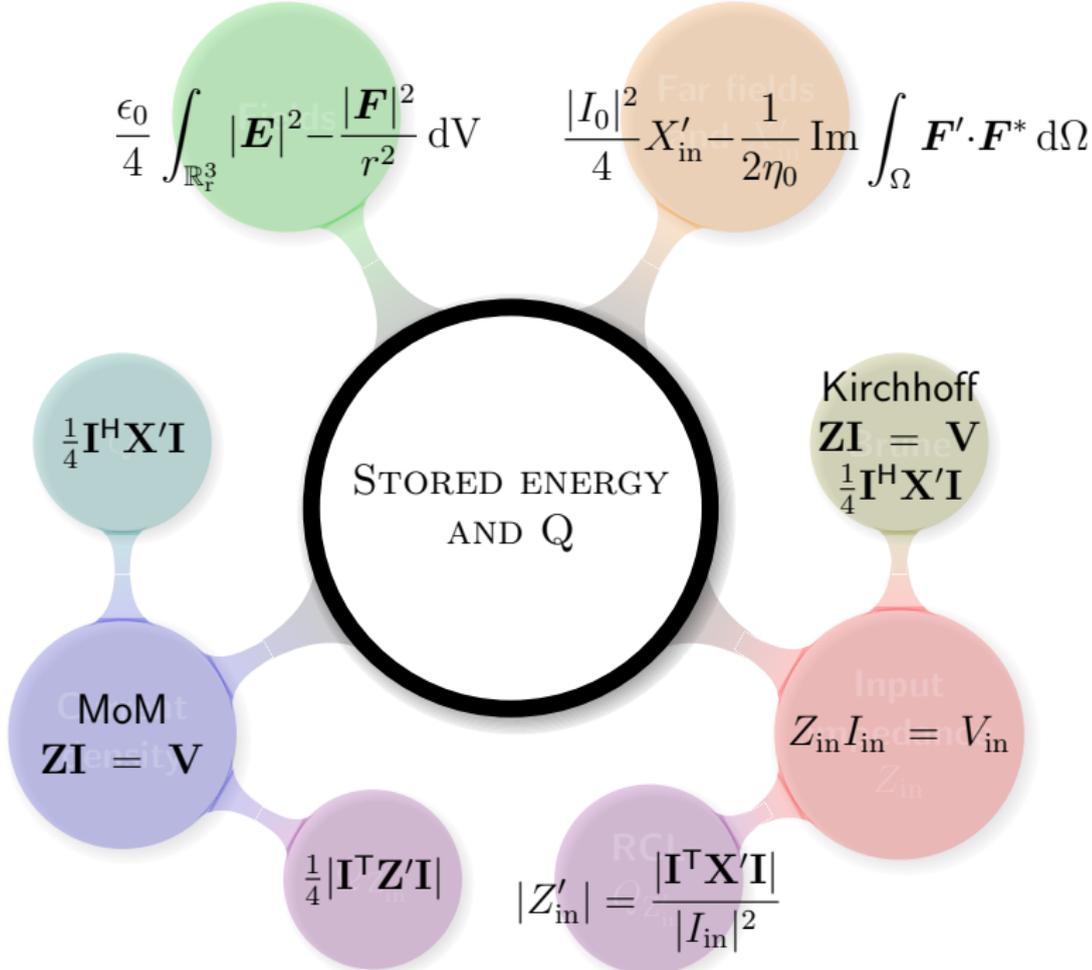
$$W_m = \frac{L|I|^2}{4}$$

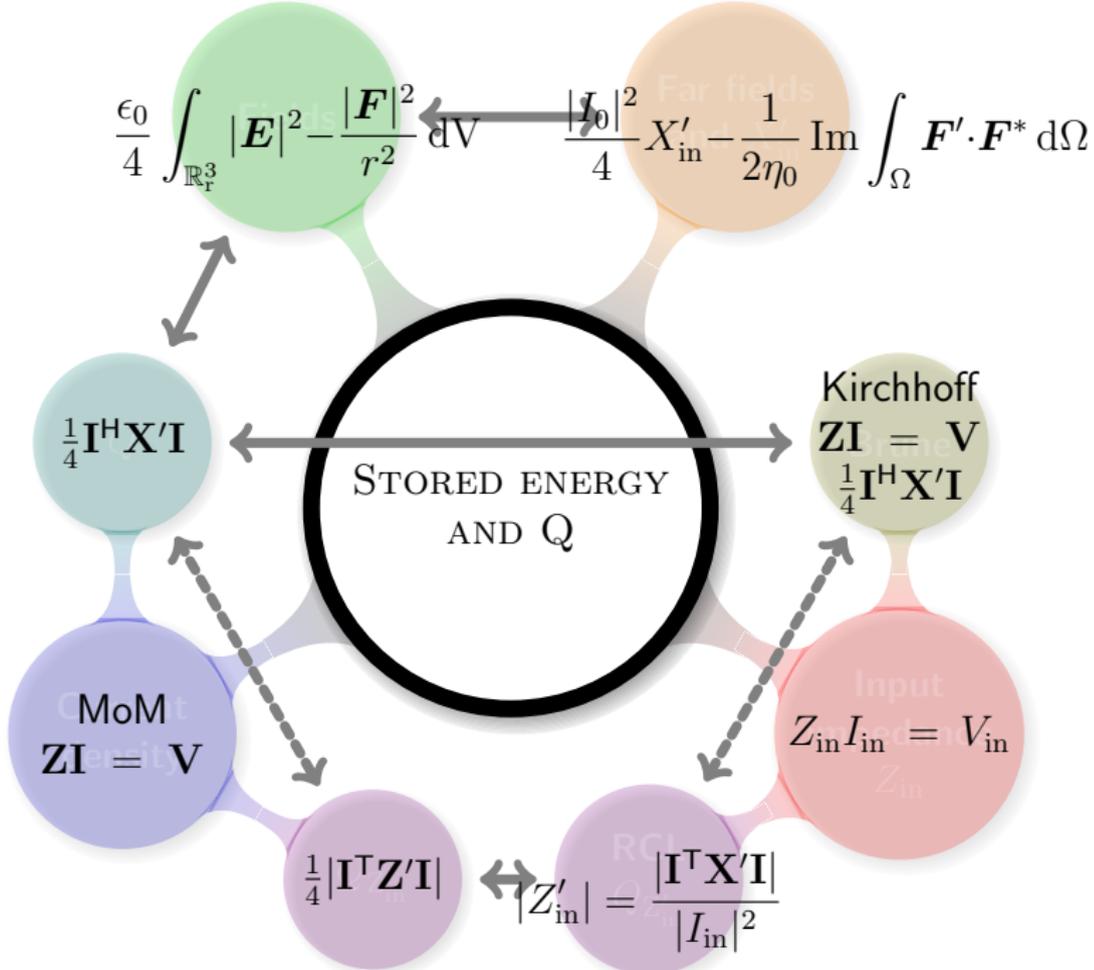


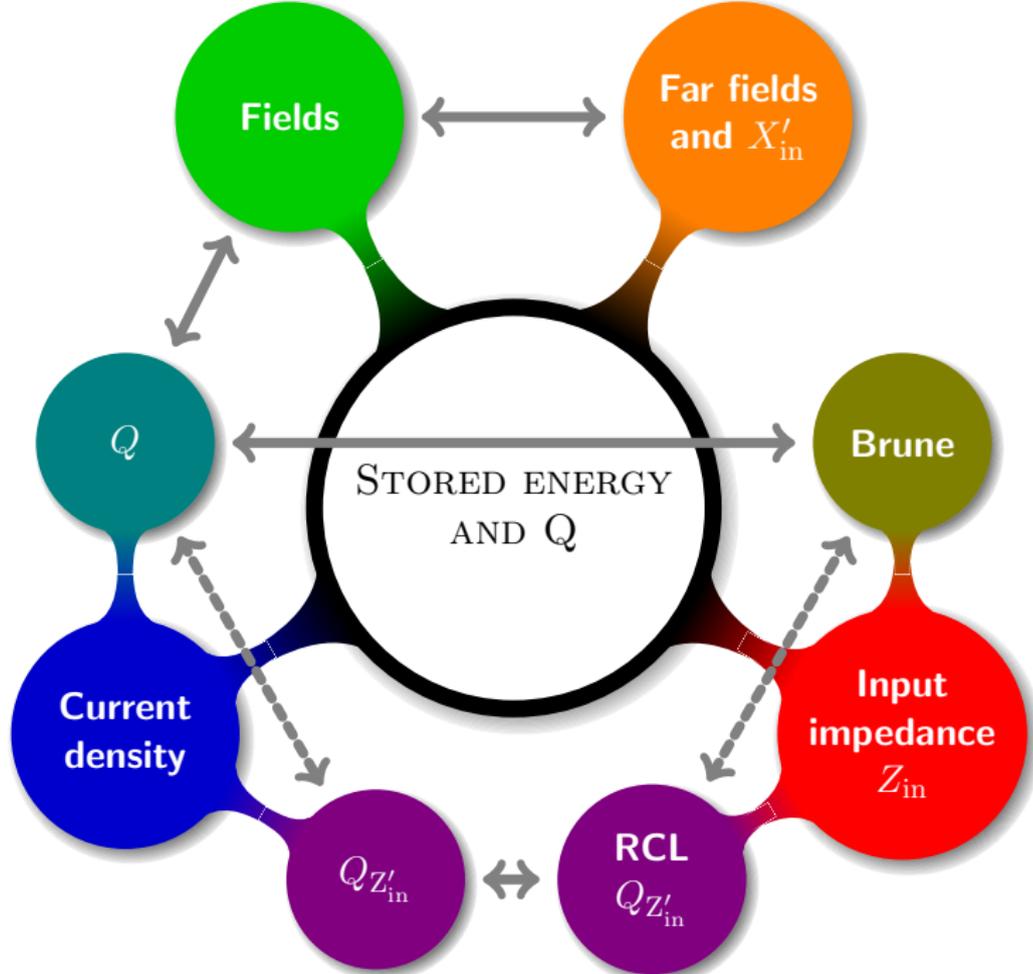












# Subtracted far field approach

$$W_F^{(E)} = \frac{\epsilon_0}{4} \int_{\mathbb{R}^3} |\mathbf{E}(\mathbf{r})|^2 - \frac{|\mathbf{F}(\hat{\mathbf{r}})|^2}{r^2} dV$$

Have shown that  $W_F^{(E)} = W_C^{(E)} + W_{c,0}$ :

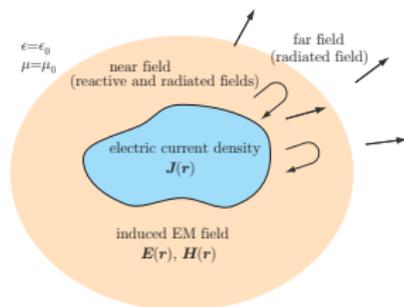
$$W_C^{(E)} = \frac{\eta_0}{4\omega} \int_V \int_V \nabla_1 \cdot \mathbf{J}_1 \nabla_2 \cdot \mathbf{J}_2^* \frac{\cos(kr_{12})}{4\pi kr_{12}} - (k^2 \mathbf{J}_1 \cdot \mathbf{J}_2^* - \nabla_1 \cdot \mathbf{J}_1 \nabla_2 \cdot \mathbf{J}_2^*) \frac{\sin(kr_{12})}{8\pi}$$

with  $\mathbf{J}_n = \mathbf{J}(\mathbf{r}_n)$ ,  $n = 1, 2$  and a **coordinate dependent** part

$$W_{c,0} = \frac{\eta_0}{4\omega} \int_V \int_V \text{Im} \{ k^2 \mathbf{J}_1 \cdot \mathbf{J}_2^* - \nabla_1 \cdot \mathbf{J}_1 \nabla_2 \cdot \mathbf{J}_2^* \} \frac{r_1^2 - r_2^2}{8\pi r_{12}} k_1(kr_{12})$$

where  ${}_1(z) = (\sin(z) - z \cos(z))/z^2$  is a spherical Bessel function.

Gustafsson & Jonsson: Stored electromagnetic energy and antenna Q, 2012

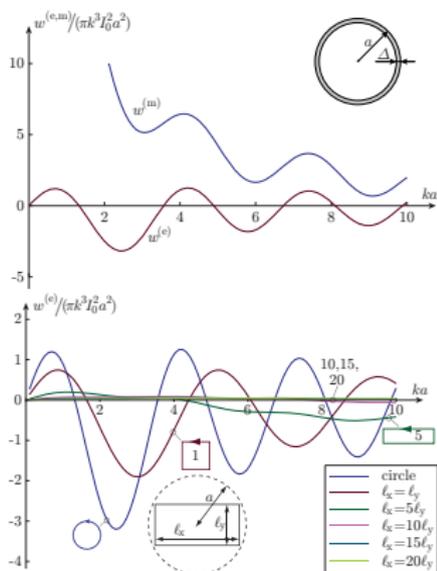


## Subtracted far field: comments

- Coordinate dependent for far-fields  $\mathbf{F}$  with

$$W_{c,0} - W_c = \frac{\epsilon_0}{4} \mathbf{d} \cdot \int_{\Omega} \hat{\mathbf{r}} |\mathbf{F}(\hat{\mathbf{r}})|^2 d\Omega \neq 0$$

- Identical coordinate independent part as for the stored energy introduced by Vandenbosch 2010 (Geyi 2003  $ka \ll 1$ ).
- Can produce negative values for larger structures.
- Difficult to generalize to antennas embedded in lossy media (no far field).



We now introduce an alternative approach to analyze antennas in lossy (dispersive) media.

# Frequency derivatives of impedance/admittance matrices

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Impedance and admittance matrices relate voltages and currents

$$\mathbf{Z}\mathbf{I} = \mathbf{V} \quad \text{or} \quad \mathbf{I} = \mathbf{Z}^{-1}\mathbf{V} = \mathbf{Y}\mathbf{V}$$

The (angular) frequency derivative of the admittance matrix is

$$\mathbf{Y}' = \frac{\partial \mathbf{Y}}{\partial \omega} = \frac{\partial \mathbf{Z}^{-1}}{\partial \omega} = -\mathbf{Z}^{-1}\mathbf{Z}'\mathbf{Z}^{-1} = -\mathbf{Y}\mathbf{Z}'\mathbf{Y}$$

*No complex conjugate.* Better to use quadratic forms with the transpose  $\mathbf{V}^T\mathbf{Y}'\mathbf{V}$  than Hermitian transpose  $\mathbf{V}^H\mathbf{Y}'\mathbf{V} = \mathbf{V}^{T*}\mathbf{Y}'\mathbf{V}$ .

For the case of a (frequency independent (MoM)) voltage source

$$Y_{\text{in}} = \frac{1}{Z_{\text{in}}} = \frac{\mathbf{V}^T\mathbf{Y}\mathbf{V}}{V_{\text{in}}^2} \quad \text{and} \quad V_{\text{in}}^2 Y'_{\text{in}} = \mathbf{V}^T\mathbf{Y}'\mathbf{V} = -\mathbf{I}^T\mathbf{Z}'\mathbf{I}$$

## $Z_{\text{in}}$ for antennas using MoM

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Use a method of moments (MoM) formulation of the electric field integral equation (EFIE). Impedance matrix  $\mathbf{Z} = \mathbf{R} + j\mathbf{X}$

$$\frac{Z_{mn}}{\eta} = j \iint_S \iint_S (k^2 \boldsymbol{\psi}_{m1} \cdot \boldsymbol{\psi}_{n2} - \nabla_1 \cdot \boldsymbol{\psi}_{m1} \nabla_2 \cdot \boldsymbol{\psi}_{n2}) \frac{e^{-jkR_{12}}}{4\pi k R_{12}} dS_1 dS_2$$

where  $\boldsymbol{\psi}_{n1} = \boldsymbol{\psi}_n(\mathbf{r}_1)$ ,  $\boldsymbol{\psi}_{n2} = \boldsymbol{\psi}_n(\mathbf{r}_2)$ ,  $n = 1, \dots, N$ , and  $R_{12} = |\mathbf{r}_1 - \mathbf{r}_2|$ .

The current density is  $\mathbf{J}(\mathbf{r}) = \sum_{n=1}^N I_n \boldsymbol{\psi}_n(\mathbf{r})$  with the expansion coefficients determined from

$$\mathbf{Z}\mathbf{I} = \mathbf{V} \quad \text{or} \quad \mathbf{I} = \mathbf{Z}^{-1}\mathbf{V} = \mathbf{Y}\mathbf{V}$$

where  $\mathbf{V}$  is a column matrix with the excitation coefficients. The input admittance is

$$Y_{\text{in}} = 1/Z_{\text{in}} = \mathbf{V}^T \mathbf{Y} \mathbf{V} / V_{\text{in}}^2$$

where  $Z_{\text{in}} = R_{\text{in}} + jX_{\text{in}}$  is the input impedance.

## $Q_{Z'_{in}}$ and $Q$ for antennas (fields)

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Differentiate the MoM impedance matrix

$$\begin{aligned} \frac{k}{\eta} \frac{\partial Z_{mn}}{\partial k} &= \int_V \int_V \mathbf{j} \left( k^2 \boldsymbol{\psi}_{m1} \cdot \boldsymbol{\psi}_{n2} + \nabla_1 \cdot \boldsymbol{\psi}_{m1} \nabla_2 \cdot \boldsymbol{\psi}_{n2} \right) \frac{e^{-jkR_{12}}}{4\pi k R_{12}} \\ &+ \left( k^2 \boldsymbol{\psi}_{m1} \cdot \boldsymbol{\psi}_{n2} - \nabla_1 \cdot \boldsymbol{\psi}_{m1} \nabla_2 \cdot \boldsymbol{\psi}_{n2} \right) \frac{e^{-jkR_{12}}}{4\pi} dS_1 dS_2 \end{aligned}$$

Differentiated input admittance

$$V_{in}^2 Y'_{in} = (\mathbf{V}^T \mathbf{Y} \mathbf{V})' = \mathbf{V}^T \mathbf{Y}' \mathbf{V} = -\mathbf{I}^T \mathbf{Z}' \mathbf{I}.$$

The stored energy determined from  $\mathbf{X}' = \text{Im } \mathbf{Z}'$

$$W_{e\mathbf{X}'} + W_{m\mathbf{X}'} = \frac{1}{4} \mathbf{I}^H \mathbf{X}' \mathbf{I}$$

is identical to the stored energy expressions introduced by Vandebosch (IEEE-TAP 2010).

## $Q$ and $Q_{Z'_{in}}$ for free-space self-resonant antennas

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Assume for simplicity a **self-resonant** antenna (circuit)

$$Q_{Z'_{in}} = \frac{\omega |Z'_{in}|}{2R_{in}} = \frac{\omega |\mathbf{I}^T \mathbf{Z}' \mathbf{I}|}{2\mathbf{I}^H \mathbf{R} \mathbf{I}}$$

and using MoM with the stores energy by Vandenbosch

$$Q = \frac{2\omega \max\{W_{e\mathbf{X}'}, W_{m\mathbf{X}'}\}}{P_d} = \frac{\omega \mathbf{I}^H \mathbf{X}' \mathbf{I}}{2\mathbf{I}^H \mathbf{R} \mathbf{I}}$$

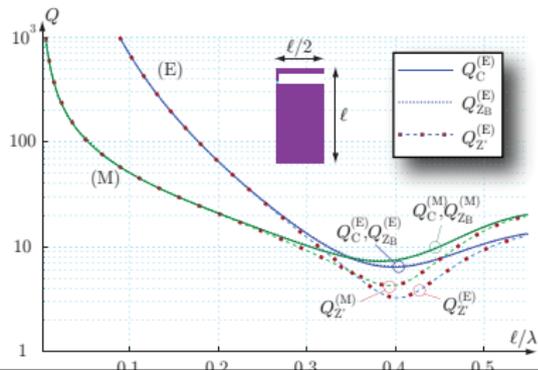
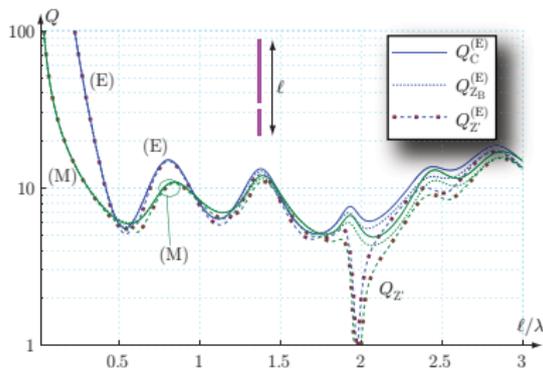
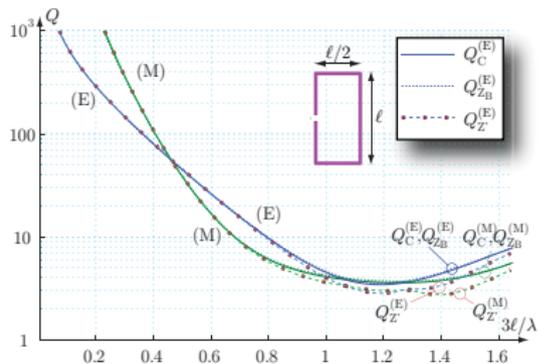
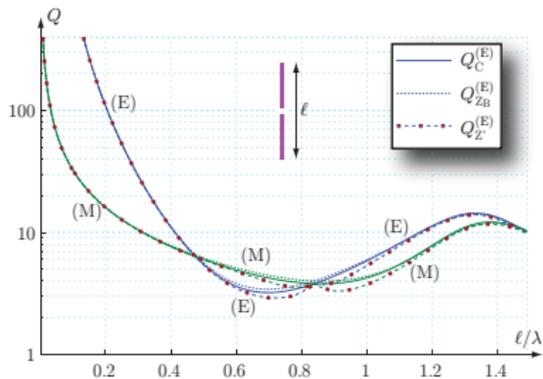
Transpose for  $Q_{Z'_{in}}$  and Hermitian transpose for  $Q$

- ▶  $\mathbf{I}^H \mathbf{X}' \mathbf{I} \geq 0$  for positive semidefinite matrices  $\mathbf{X}'$ .
- ▶  $|\mathbf{I}^T \mathbf{Z}' \mathbf{I}| = 0$  for some  $\mathbf{I}$  (rank  $> 1$ ).

See also Capek+etal. IEEE-TAP 2014 for  $Q_{Z'_{in}}$  using  $\mathbf{I}^H$  and  $\mathbf{I}'$ .

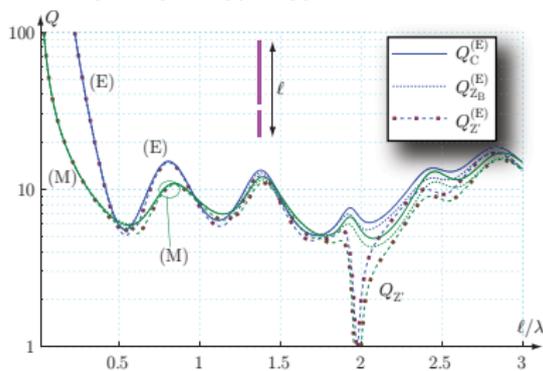
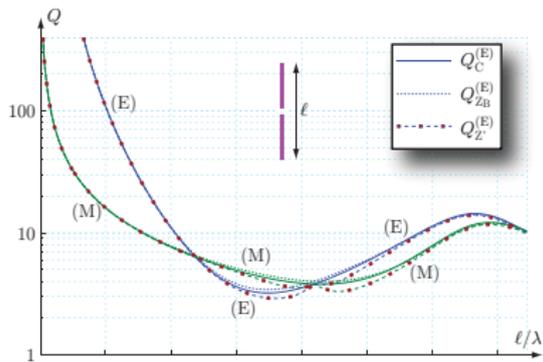
# Antenna examples (free space)

Q from stored energy expressed in the current density  $Q_C$ , Brune circuit  $Q_{Z_B}^{(E)}$ , and differentiated input impedance  $Q_{Z'}^{(E)}$



# Antenna examples (free space)

Q from stored energy expressed in the current density  $Q_C$ , Brune circuit  $Q_{Z_{in}^B}$ , and differentiated input impedance  $Q_{Z'_{in}}$



Q computed from

- ▶ the currents,  $Q_C$ .
- ▶ a circuit model synthesized from the input impedance using Brune synthesis (1931),  $Q_{Z_{in}^B}$ .
- ▶ differentiation of the (tuned) input impedance,

$$Q_{Z'_{in}} = \frac{\omega_0 |Z'_{in}|}{2R_{in}} = \omega_0 |I'|.$$

All agree for  $Q \gg 1$  but the Q from the differentiated impedance ( $Q_{Z'_{in}}$ ) is lower in some regions.

Which one is most accurate/best?

## Dispersive media

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The frequency derivative of the EFIE impedance matrix  $\mathbf{Z}$  is

$$\omega \frac{\partial \mathbf{Z}}{\partial \omega} = k \frac{\partial(\mathbf{Z}/\eta)}{\partial k} \frac{\eta \omega}{k} \frac{\partial k}{\partial \omega} + \omega \frac{\mathbf{Z}}{\eta} \frac{\partial \eta}{\partial \omega}$$

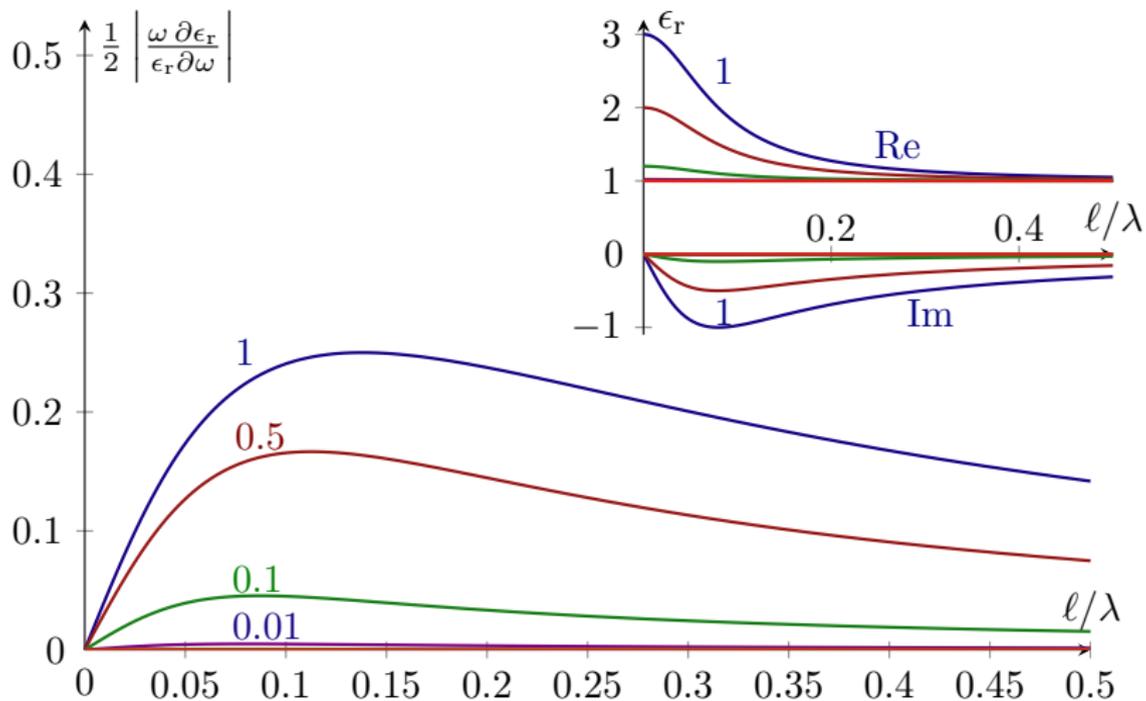
for a temporally dispersive background medium with  $k = \omega \sqrt{\epsilon \mu}$  and  $\eta = \sqrt{\mu/\epsilon}$ . The derivative simplifies to

$$\omega \frac{\partial \mathbf{Z}}{\partial \omega} = k \frac{\partial(\mathbf{Z}/\eta)}{\partial k} \eta \left( \frac{\omega \partial \epsilon}{2 \epsilon \partial \omega} + 1 \right) - \frac{\mathbf{Z}}{2} \frac{\omega \partial \epsilon}{\epsilon \partial \omega}$$

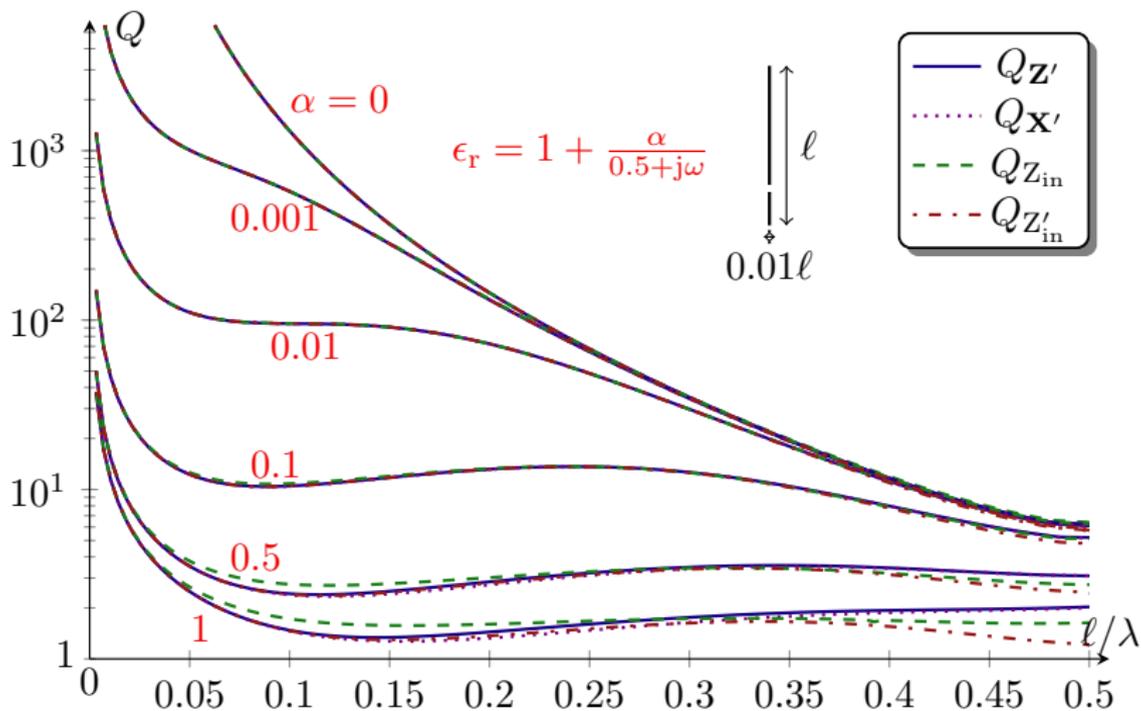
for the common case of a non-magnetic medium,  $\mu_r = 1$ .

Multiplication of the previously calculated derivative (with respect to the wavenumber  $k$  in the medium) with a factor that only depends on the medium. The factor  $\omega \epsilon' = (\omega \epsilon)' - \epsilon$  is similar to the classical approach used to define the energy density in dispersive media.

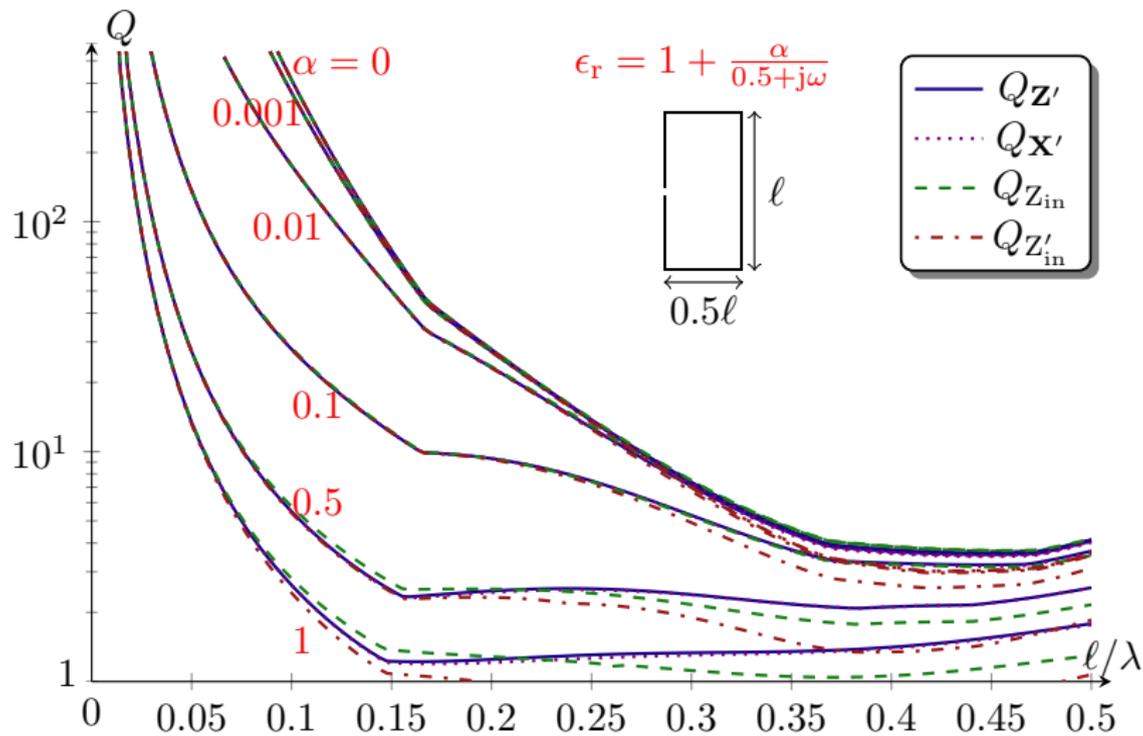
## Numerical examples: Debye media



# Numerical examples: Debye media



# Numerical examples: Debye media



## Stored EM energy matrices

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Method of Moments approximation (expand  $\mathbf{J}$  in basis functions)

$$W_e \approx \frac{1}{4\omega} \mathbf{I}^H \mathbf{X}_e \mathbf{I} \quad \text{stored E-energy, } \mathbf{X}_e \text{ electric reactance}$$

$$W_m \approx \frac{1}{4\omega} \mathbf{I}^H \mathbf{X}_m \mathbf{I} \quad \text{stored M-energy, } \mathbf{X}_m \text{ magnetic reactance}$$

$$P_{\text{rad}} \approx \frac{1}{2} \mathbf{I}^H \mathbf{R}_r \mathbf{I} \quad \text{radiated power}$$

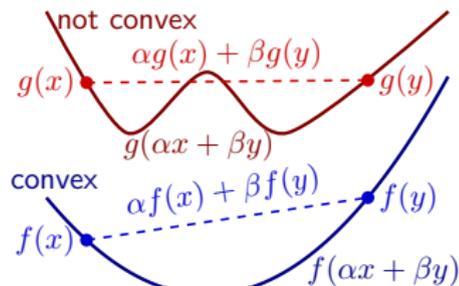
giving  $\mathbf{Z} = \mathbf{R}_r + j(\mathbf{X}_m - \mathbf{X}_e)$ . We also use

$$\mathbf{F} \approx \mathbf{F}^H \mathbf{I} (\text{far field}), \mathbf{E} \approx \mathbf{N}^H \mathbf{I} (\text{near field}), \mathbf{I}_2 \approx \mathbf{C}^H \mathbf{I}_1 (\text{induced current})$$

Pre-computed matrices used in the optimization.

# Convex optimization

$$\begin{aligned} &\text{minimize} && f_0(\mathbf{x}) \\ &\text{subject to} && f_i(\mathbf{x}) \leq 0, \quad i = 1, \dots, N_1 \\ &&& \mathbf{Ax} = \mathbf{b} \end{aligned}$$



where  $f_i(x)$  are convex, i.e.,  $f_i(\alpha\mathbf{x} + \beta\mathbf{y}) \leq \alpha f_i(\mathbf{x}) + \beta f_i(\mathbf{y})$  for  $\alpha, \beta \in \mathbb{R}$ ,  $\alpha + \beta = 1$ ,  $\alpha, \beta \geq 0$ .

Solved with efficient standard algorithms. No risk of getting trapped in a local minimum. A problem is 'solved' if formulated as a convex optimization problem.

Antenna performance expressed in the current density  $\mathbf{J}$ , e.g.,

- ▶ Radiated field  $\mathbf{F}(\hat{\mathbf{k}}) = -\hat{\mathbf{k}} \times \hat{\mathbf{k}} \times \int_V \mathbf{J}(\mathbf{r}) e^{j\mathbf{k} \cdot \mathbf{r}} dV$  is affine.
- ▶ Radiated power, stored electric and magnetic energies, and Ohmic losses are positive semi-definite quadratic forms in  $\mathbf{J}$ .

## Currents for maximal $G/Q$

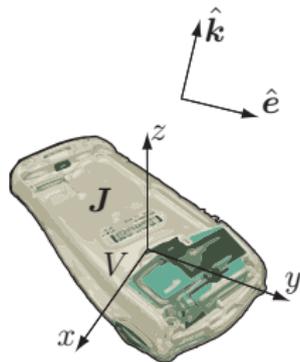
Determine a current density  $\mathbf{J}(\mathbf{r})$  in the volume  $V$  that maximizes the partial-gain Q-factor quotient  $G(\hat{\mathbf{k}}, \hat{\mathbf{e}})/Q$ .

- ▶ Partial radiation intensity  $P(\hat{\mathbf{k}}, \hat{\mathbf{e}})$

$$\frac{G(\hat{\mathbf{k}}, \hat{\mathbf{e}})}{Q} = \frac{2\pi P(\hat{\mathbf{k}}, \hat{\mathbf{e}})}{c_0 k \max\{W_e, W_m\}}$$

- ▶ Scale  $\mathbf{J}$  and reformulate  $P = 1$  as  $\hat{\mathbf{e}}^* \cdot \mathbf{F} = \mathbf{F}^H \mathbf{I} = 1$ .
- ▶ Convex optimization problem:

$$\begin{aligned} & \text{minimize} && W \\ & \text{subject to} && \mathbf{I}^H \mathbf{X}_e \mathbf{I} \leq W \\ & && \mathbf{I}^H \mathbf{X}_m \mathbf{I} \leq W \\ & && \mathbf{F}^H \mathbf{I} = 1 \end{aligned}$$



Determines a current density  $\mathbf{J}(\mathbf{r})$  in the volume  $V$  with minimal stored EM energy and unit partial radiation intensity.

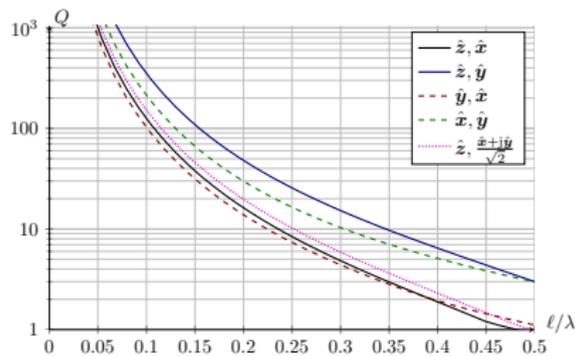
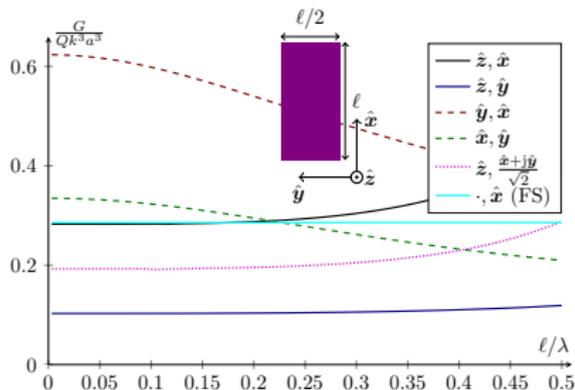
# Maximal $G(\hat{\mathbf{k}}, \hat{\mathbf{x}})/Q$ for planar rectangles

$$\begin{aligned} & \text{minimize} && W \\ & \text{subject to} && \mathbf{I}^H \mathbf{X}_e \mathbf{I} \leq W \\ & && \mathbf{I}^H \mathbf{X}_m \mathbf{I} \leq W \\ & && \mathbf{F}^H \mathbf{I} = 1 \end{aligned}$$

Solution for current densities confined to planar rectangles with side lengths  $\ell_x$  and  $\ell_y = 0.5\ell_x$ .

CVX code: <http://cvxr.com/cvx/>

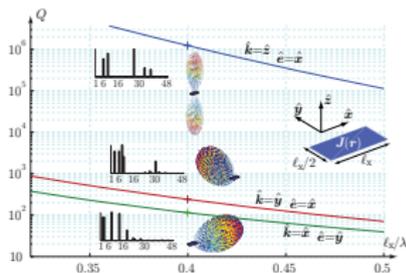
```
cvx_begin
    variable J(n) complex;
    minimize W
    subject to
        quad_form(J, Xe) <= W;
        quad_form(J, Xm) <= W;
        F' * J == 1;
cvx_end
```



# Example of current optimization formulations

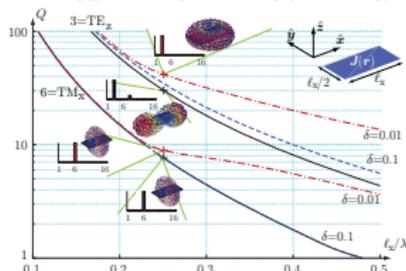
## Super directivity:

$$\begin{aligned} & \text{minimize}_{\mathbf{I}} \quad \max\{\mathbf{I}^H \mathbf{X}_e \mathbf{I}, \mathbf{I}^H \mathbf{X}_m \mathbf{I}\} \\ & \text{subject to} \quad \mathbf{F}^H \mathbf{I} = 1 \\ & \quad \quad \quad \mathbf{I}^H \mathbf{R}_r \mathbf{I} \leq 4\pi/(\eta_0 D_0) \end{aligned}$$



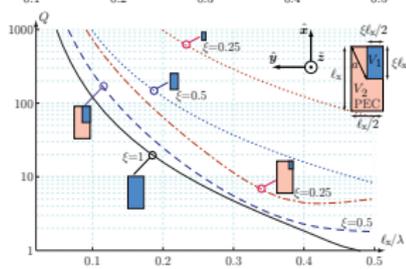
## Prescribed far field:

$$\begin{aligned} & \text{minimize} \quad \max\{\mathbf{I}^H \mathbf{X}_e \mathbf{I}, \mathbf{I}^H \mathbf{X}_m \mathbf{I}\} \\ & \text{subject to} \quad \int_{\Omega} |\mathbf{F}(\hat{\mathbf{k}}) - \mathbf{F}_0(\hat{\mathbf{k}})|^2 d\Omega_{\hat{\mathbf{k}}} < \delta \end{aligned}$$



## Embedded antennas:

$$\begin{aligned} & \text{minimize} \quad \max\{\mathbf{I}^H \mathbf{X}_e \mathbf{I}, \mathbf{I}^H \mathbf{X}_m \mathbf{I}\} \\ & \text{subject to} \quad \mathbf{F}^H \mathbf{I} = 1 \\ & \quad \quad \quad \mathbf{I}_2 = \mathbf{C}^H \mathbf{I}_1 \end{aligned}$$



# Why convex optimization?

**Solved** if formulated as a convex optimization problem.

Consider the  $G/Q$  problem

$$\begin{aligned} & \text{minimize} && \max\{\mathbf{I}^H \mathbf{X}_e \mathbf{I}, \mathbf{I}^H \mathbf{X}_m \mathbf{I}\} \\ & \text{subject to} && \mathbf{F}^H \mathbf{I} = 1 \end{aligned}$$

Many (optimization) algorithms can be used to solve this problem.

- ▶ Can *e.g.*, use any of the solvers included in CVX.
  - ▶ Very simple to use.
  - ▶ Good for small problems but less efficient for larger problems.
- ▶ A dedicated solver for quadratic programs.
  - ▶ More efficient for larger problems.
- ▶ Random search, eg genetic algorithms (GA), particle swarms,....
  - ▶ Very inefficient. Note you do not (should not) use (GA, ...) to solve *e.g.*,  $\mathbf{Ax} = \mathbf{b}$  (min.  $\|\mathbf{Ax} - \mathbf{b}\|$ ).
- ▶ We also use a dual formulation
  - ▶ Computational efficient for large problems.
  - ▶ Illustrates dual problems and posteriori error estimates.

## Relaxation and dual problem

---

An illustrative method is to use the inequality

$$W = \max\{W_e, W_m\} \geq \alpha W_e + (1 - \alpha)W_m = W_\alpha \quad \text{for } 0 \leq \alpha \leq 1$$

or with the matrices  $\mathbf{X}_e, \mathbf{X}_m$

$$W = \max\{\mathbf{I}^H \mathbf{X}_e \mathbf{I}, \mathbf{I}^H \mathbf{X}_m \mathbf{I}\} \geq W_\alpha = \mathbf{I}_\alpha^H (\alpha \mathbf{X}_e + (1 - \alpha) \mathbf{X}_m) \mathbf{I}_\alpha$$

or for the quotient  $G/Q$

$$\frac{G}{Q} = \frac{2\pi P}{\omega \max\{W_{e\alpha}, W_{m\alpha}\}} \leq \frac{2\pi P}{\omega W_\alpha} = \frac{2\pi P}{\omega(\alpha W_{e\alpha} + (1 - \alpha)W_{m\alpha})} = \frac{G_\alpha}{Q_\alpha}$$

Note  $P = 1$  is fixed in the optimization problem.

## Relaxation and dual problem

---

The inequality relaxes the  $G/Q$  optimization problem

$$\begin{aligned} &\text{minimize} && \max\{\mathbf{I}^H \mathbf{X}_e \mathbf{I}, \mathbf{I}^H \mathbf{X}_m \mathbf{I}\} \geq \mathbf{I}_\alpha^H (\alpha \mathbf{X}_e + (1 - \alpha) \mathbf{X}_m) \mathbf{I}_\alpha \\ &\text{subject to} && \mathbf{F}^H \mathbf{I} = 1 \end{aligned}$$

into

$$\begin{aligned} &\text{maximize}_\alpha \text{ minimize}_{\mathbf{I}_\alpha} && W_\alpha = \mathbf{I}_\alpha^H (\alpha \mathbf{X}_e + (1 - \alpha) \mathbf{X}_m) \mathbf{I}_\alpha \\ &\text{subject to} && \mathbf{F}^H \mathbf{I}_\alpha = 1 \\ &&& 0 \leq \alpha \leq 1 \end{aligned}$$

where for the quotient  $G/Q$  (note  $P = 1$ )

$$\frac{G}{Q} = \frac{2\pi P}{\omega \max\{W_{e\alpha}, W_{m\alpha}\}} \leq \frac{2\pi P}{\omega W_\alpha} = \frac{2\pi P}{\omega(\alpha W_{e\alpha} + (1 - \alpha) W_{m\alpha})} = \frac{G_\alpha}{Q_\alpha}$$

## Relaxation and dual problem

---

The dual problem

$$\begin{aligned} & \text{maximize}_{\alpha} \text{minimize}_{\mathbf{I}_{\alpha}} & W_{\alpha} &= \mathbf{I}_{\alpha}^{\text{H}} (\alpha \mathbf{X}_{\text{e}} + (1 - \alpha) \mathbf{X}_{\text{m}}) \mathbf{I}_{\alpha} \\ & \text{subject to} & & \mathbf{F}^{\text{H}} \mathbf{I}_{\alpha} = 1 \\ & & & 0 \leq \alpha \leq 1 \end{aligned}$$

is solved as a linear system (MoM equation) for fixed  $\alpha$  with

$$\mathbf{I}_{\alpha} = \frac{(\alpha \mathbf{X}_{\text{e}} + (1 - \alpha) \mathbf{X}_{\text{m}})^{-1} \mathbf{F}}{\mathbf{F}^{\text{H}} (\alpha \mathbf{X}_{\text{e}} + (1 - \alpha) \mathbf{X}_{\text{m}})^{-1} \mathbf{F}}$$

giving the optimization problem

$$\text{maximize}_{0 \leq \alpha \leq 1} W_{\alpha}$$

or

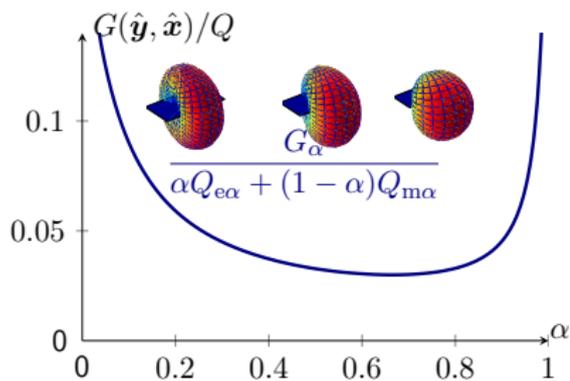
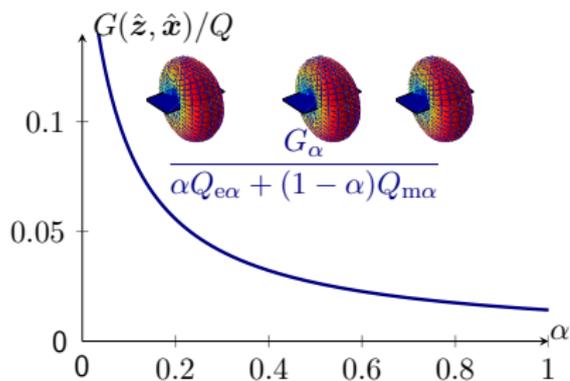
$$\text{minimize}_{0 \leq \alpha \leq 1} \frac{G_{\alpha}}{Q_{\alpha}} = \mathbf{F}^{\text{H}} (\alpha \mathbf{X}_{\text{e}} + (1 - \alpha) \mathbf{X}_{\text{m}})^{-1} \mathbf{F}$$

## Why convex optimization: illustration

The upper bound on  $G/Q|_{\text{ub}}$  is obtained by solving the dual (relaxed) problem, *i.e.*, finding the minimum of the (blue) curve

$$\frac{G}{Q}\Big|_{\text{ub}} \leq \frac{G_\alpha}{\alpha Q_{e\alpha} + (1-\alpha)Q_{m\alpha}}$$

This is efficiently solved by golden section search and parabolic interpolation.

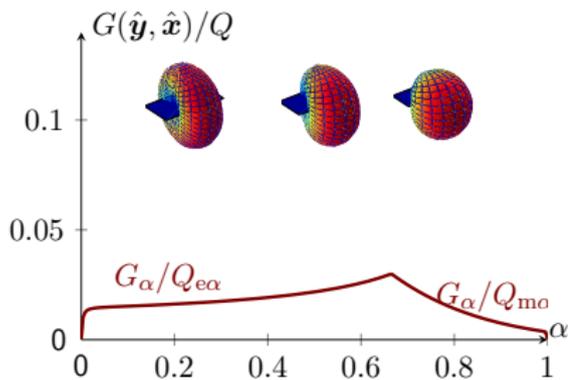
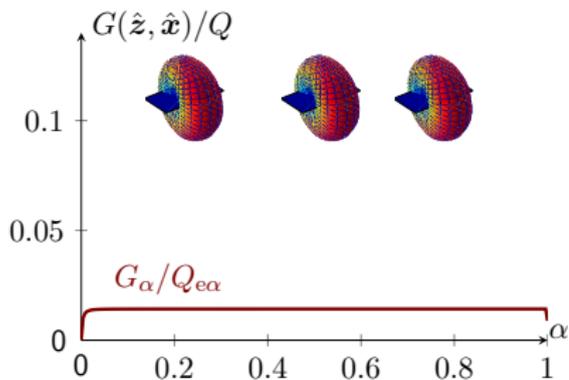


$$\ell/\lambda \approx 0.1 \text{ or } ka \approx 0.35$$

# Why convex optimization: illustration

We also compute the actual  $G/Q$  for the current  $\mathbf{I}_\alpha$  to get

$$\frac{G_\alpha}{\max\{Q_{e\alpha}, Q_{m\alpha}\}} \leq \left. \frac{G}{Q} \right|_{\text{ub}}$$



$\ell/\lambda \approx 0.1$  or  $ka \approx 0.35$

## Why convex optimization: illustration

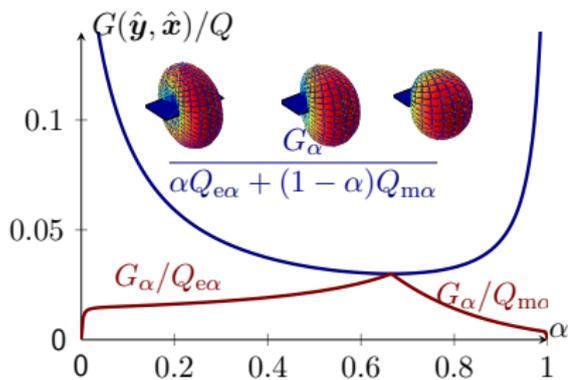
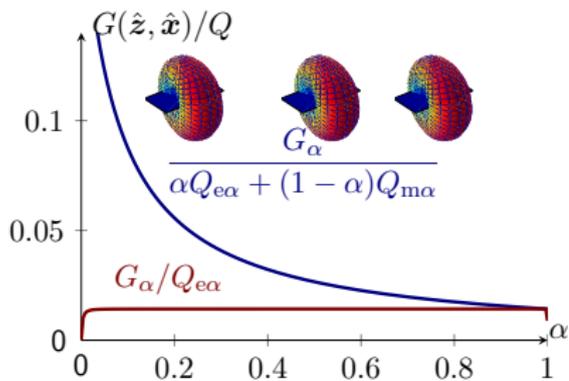
The upper bound on  $G/Q|_{\text{ub}}$  is obtained by solving the dual (relaxed) problem, *i.e.*, finding the minimum of the (blue) curve

$$\frac{G}{Q}\Big|_{\text{ub}} \leq \frac{G_\alpha}{\alpha Q_{e\alpha} + (1-\alpha)Q_{m\alpha}}$$

This is efficiently solved by golden section search and parabolic interpolation.

We also compute the actual  $G/Q$  for the current  $\mathbf{I}_\alpha$  to get

$$\frac{G_\alpha}{\max\{Q_{e\alpha}, Q_{m\alpha}\}} \leq \frac{G}{Q}\Big|_{\text{ub}}$$

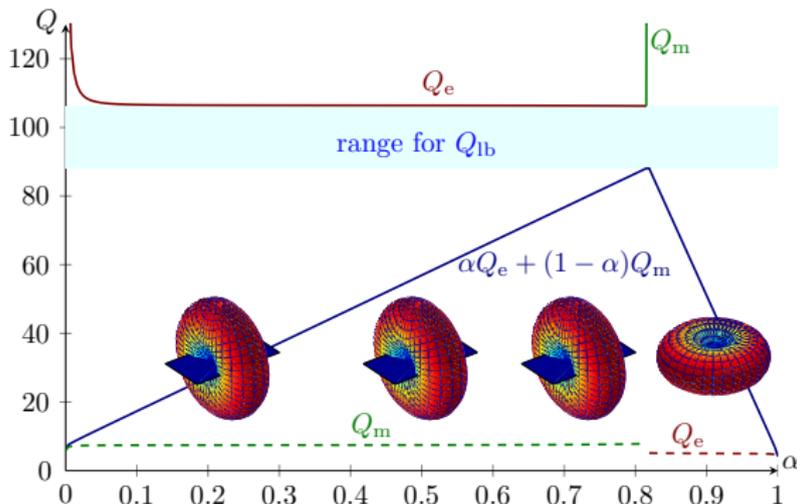


$\ell/\lambda \approx 0.1$  or  $ka \approx 0.35$

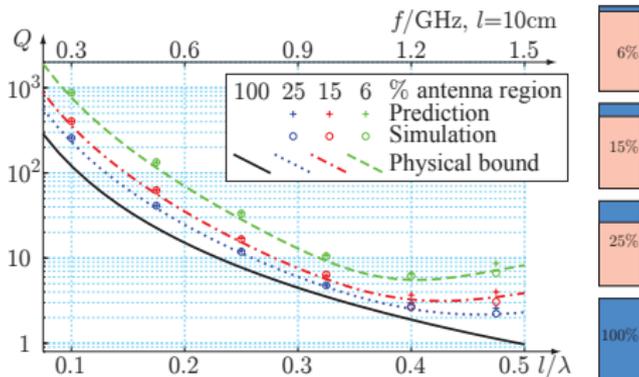
# Minimization of $Q$

The corresponding formulation for  $Q$  is not convex. For  $0 \leq \alpha \leq 1$

$$Q = \frac{\max\{\mathbf{I}^H \mathbf{X}_e \mathbf{I}, \mathbf{I}^H \mathbf{X}_m \mathbf{I}\}}{\mathbf{I}^H \mathbf{R}_r \mathbf{I}} = \max\{Q_e, Q_m\}$$
$$\geq \alpha Q_e + (1 - \alpha) Q_m = \frac{\mathbf{I}^H (\alpha \mathbf{X}_e + (1 - \alpha) \mathbf{X}_m) \mathbf{I}}{\mathbf{I}^H \mathbf{R}_r \mathbf{I}}$$



# Conclusions



- ▶ Current optimization for physical bounds.
- ▶ Stored energy from MoM reactance matrices (basically, already computed in most MoM codes for surface currents).
- ▶ Promising results for temporally dispersive media.
- ▶ Convex optimization (efficiently solved with a few  $\mathbf{Ax} = \mathbf{b}$ ).

## Minimization of $Q$

---

Compare maximization of  $G/Q$  with minimization of  $Q$ . Use the same inequality for  $0 \leq \alpha \leq 1$

$$Q = \frac{\max\{\mathbf{I}^H \mathbf{X}_e \mathbf{I}, \mathbf{I}^H \mathbf{X}_m \mathbf{I}\}}{\mathbf{I}^H \mathbf{R}_r \mathbf{I}} = \max\{Q_e, Q_m\}$$
$$\geq \alpha Q_e + (1 - \alpha) Q_m = \frac{\mathbf{I}^H (\alpha \mathbf{X}_e + (1 - \alpha) \mathbf{X}_m) \mathbf{I}}{\mathbf{I}^H \mathbf{R}_r \mathbf{I}}$$

The lower bound on  $Q$ ,  $Q_{lb}$ , is a minimization problem for a Rayleigh quotient solved as a generalized eigenvalue problem

$$\min(\alpha \mathbf{X}_e + (1 - \alpha) \mathbf{X}_m, \mathbf{R}_r)$$

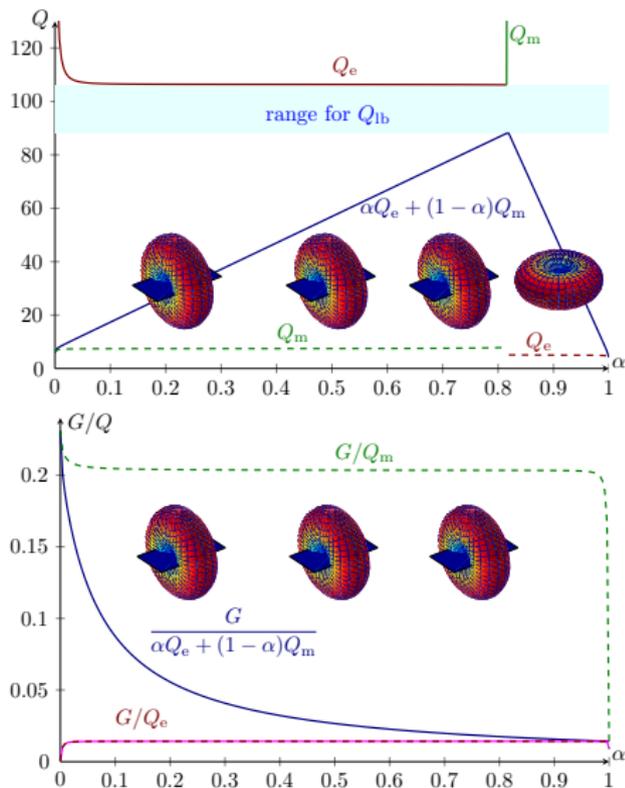
Let  $Q_{e\alpha}$  and  $Q_{m\alpha}$  denote the corresponding electric and magnetic Q-factors to get the estimate

$$\alpha Q_{e\alpha} + (1 - \alpha) Q_{m\alpha} \leq Q_{lb} \leq \max\{Q_{e\alpha}, Q_{m\alpha}\}$$

for the lower bound  $Q_{lb}$ .

# Numerical illustration of $\min. Q$ and $\max. G/Q$

- ▶ The formulation for  $\min. Q$  has a duality gap, *i.e.*, we have an interval for  $Q_{lb}$  here  $88 \leq Q_{lb} \leq 106$ .
- ▶ The optimization problem  $\min. Q$  is not convex.
- ▶ The formulation for  $\max. G/Q$  has no duality gap.
- ▶ This is common for many convex optimization problems.



# Why convex optimization? Simple algorithm

Consider the  $G/Q$  problem

$$\begin{aligned} & \text{minimize} && W = \max\{\mathbf{I}^H \mathbf{X}_e \mathbf{I}, \mathbf{I}^H \mathbf{X}_m \mathbf{I}\} \\ & \text{subject to} && \text{Im}\{\mathbf{F}^H \mathbf{I}\} = 1 \end{aligned}$$

There are many (optimization) algorithms that can be used to solve this problem. An illustrative method is to use ( $0 \leq \alpha \leq 1$ )

$$W = \max\{\mathbf{I}^H \mathbf{X}_e \mathbf{I}, \mathbf{I}^H \mathbf{X}_m \mathbf{I}\} \geq W_\alpha = \mathbf{I}_\alpha^H (\alpha \mathbf{X}_e + (1 - \alpha) \mathbf{X}_m) \mathbf{I}_\alpha$$

and (hence  $G/Q \leq G_\alpha/Q_\alpha$ ) to relax to the dual problem

$$\begin{aligned} & \text{maximize}_\alpha \text{ minimize}_{\mathbf{I}_\alpha} && W_\alpha = \mathbf{I}_\alpha^H (\alpha \mathbf{X}_e + (1 - \alpha) \mathbf{X}_m) \mathbf{I}_\alpha \\ & \text{subject to} && \text{Im}\{\mathbf{F}^H \mathbf{I}_\alpha\} = 1 \\ & && 0 \leq \alpha \leq 1 \end{aligned}$$

that is solved as a linear system (MoM equation) for fixed  $\alpha$  giving

$$\text{maximize}_{0 \leq \alpha \leq 1} W_\alpha \quad \text{with } \mathbf{I}_\alpha = \frac{(\alpha \mathbf{X}_e + (1 - \alpha) \mathbf{X}_m)^{-1} \mathbf{F}}{\mathbf{F}^H (\alpha \mathbf{X}_e + (1 - \alpha) \mathbf{X}_m)^{-1} \mathbf{F}} \quad (\text{relaxed problem})$$