

# Electromagnetic Gaussian beams and Riemannian geometry

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## Introduction

- A *Gaussian beam* is a solution to Maxwell's equations that look like this:
- A Gaussian beam propagate along a curve:
- **Motivation:** Gaussian beams could (maybe) be used to solve the traveltime problem in EM.

## Initial assumptions

- Everything is smooth
- $M$  3-manifold, bounded/unbounded
- Media is anisotropic, non-homogeneous, no time or frequency dependence
- $\varepsilon, \mu$  real, positive definite, simultaneously diagonalizable. For some orthogonal matrix  $R$ ,

$$\varepsilon = R^{-1} \cdot \begin{pmatrix} \varepsilon_1 & & \\ & \varepsilon_2 & \\ & & \varepsilon_3 \end{pmatrix} \cdot R \quad \mu = R^{-1} \cdot \begin{pmatrix} \mu_1 & & \\ & \mu_2 & \\ & & \mu_3 \end{pmatrix} \cdot R$$

## Gaussian beam definition

$$\text{Trial: } E(x, t) = \text{Re}\{e^{iP\theta(x,t)} E_0(x, t)\}$$

$P > 0$  large constant,  $E_0$  complex 1-form,  $\theta$  complex phase function

- Trial is a “separation of variables”
- Let  $c: I \rightarrow M$  be a curve, and let

$$\begin{aligned}\phi(t) &= \theta(c(t), t), \\ p_j(t) &= \frac{\partial \theta}{\partial x^j}(c(t), t), \\ H_{ij}(t) &= \frac{\partial^2 \theta}{\partial x^i \partial x^j}(c(t), t),\end{aligned}$$

Then  $E$  is a G.B. on  $c$  if  $\phi(t), p_i(t)$  are real, and  $\text{Im } H_{ij}$  is positive definite.

## Basic property of Gaussian beams

Let us fix  $t$ , and let

$$z = z(x) = x - c(t)$$

then by Taylor's theorem,

$$\theta(x(z), t) = \phi(t) + p_i(t) \cdot z^i + \frac{1}{2} H_{ij}(t) z^i z^j + \dots$$

Then

$$\begin{aligned} |\exp(iP\theta)| &= \exp\left(-\frac{P}{2} z^T \cdot \text{Im } H \cdot z\right) \\ &= \text{Gaussian bell curve} \end{aligned}$$

## Propagating Gaussian beams

- *Hamilton-Jacobi* equation for  $\theta$ :  $\frac{\partial \theta}{\partial t} = h(d\theta)$  where Hamiltonian  $h: T^*M \rightarrow \mathbb{R}$  ( $h = h_{\pm}$  depend only on media)
- Expanding both sides as a Taylor series at  $z = 0$  gives sufficient condition on  $c, \phi, p, H$ :
  - $\phi$  is constant
  - $(c, p)$  is a solution to Hamilton's equations

$$\begin{aligned}\frac{dc^i}{dt} &= \frac{\partial h}{\partial \xi_i} \circ (c, p), \\ \frac{dp_i}{dt} &= -\frac{\partial h}{\partial x^i} \circ (c, p).\end{aligned}$$

- $H$  is a solution to a matrix Riccati equation (depending on  $h$ )

## Geometry of $TM$ vs. geometry of $T^*M$

### $TM$

- On  $TM$ , a curve  $c: I \rightarrow M$  has a canonical lift,  $\hat{c}: I \rightarrow TM$ .
- “Traditional” geometry (metric tensor, curvature tensors, etc.) exist on  $TM$ .
- Finsler norm  $F: TM \rightarrow \mathbb{R}$ , (in Riemannian case:  
$$F(y) = \sqrt{g_{ij}(x)y^i y^j}, y \in T_x M$$
  
→ Geodesic equation → Geodesics

### $T^*M$

- $T^*M$  has a canonical symplectic structure.
- Hamiltonian function:  $h: T^*M \rightarrow \mathbb{R}$   
→ Hamilton equations → bicharacteristics

## Legendre transformation

- In mechanics: Hamilton eqs.  $\leftrightarrow$  Euler-Lagrange eqs.
- Mathematical interpretation: Legendre transformation:  
Symplectic geometry  $\leftrightarrow$  Riemann-Finsler geometry
- **Legendre transformation:** bijection  $L: T^*M \rightarrow TM$  that preserves structure. Examples:
  - $h$  strongly convex, 1-homogeneous  $\longrightarrow$  Finsler geometry
  - $h = \sqrt{\text{pos. def. quadratic form}}$   $\longrightarrow$  Riemann geometry



## Legendre transformation for quadratic Hamiltonians

- Suppose  $h: T^*M \rightarrow \mathbb{R}$  has the form

$$h(x, \xi) = \sqrt{h^{ij}(x)\xi_i\xi_j}.$$

where  $h^{ij}$  is positive definite for each  $x$ .

$$\begin{aligned} \Rightarrow \quad L: T^*M &\rightarrow TM, \\ \xi_i dx^i &\mapsto h^{ij}\xi_j \frac{\partial}{\partial x^i} \end{aligned}$$

is a bijection, and

$$\|y\| = h \circ L^{-1}(y) = \sqrt{h_{ij}(x)y^i y^j}, \quad h_{ij} = (h^{ij})_{ij}^{-1}$$

is a norm induced by a Riemannian inner product  $g_{ij} = h_{ij}$ .

- $(c, p)$  is a bicharacteristic w.r.t.  $h \Rightarrow c$  is geodesic w.r.t.  $g$ .
- $c$  is geodesic w.r.t.  $g \Rightarrow L^{-1} \circ \hat{c}$  is a bicharacteristic w.r.t.  $h$

## Hamiltonians $h_{\pm}$

- For  $(x, \xi) \in T^*M$ , let

$$M(x, \xi) = \begin{pmatrix} \varepsilon_1 & & \\ & \ddots & \\ & & \mu_3 \end{pmatrix} \cdot \begin{pmatrix} & \xi \times I \\ -\xi \times I & \end{pmatrix}$$

- $h_{\pm} : T^*M \rightarrow \mathbb{R}$  are continuous functions such that

$$\sigma(M(x, \xi)) = \pm\{0, h_+(x, \xi), h_-(x, \xi)\},$$

and  $h_{\pm}(x, \xi) \geq 0$ .

- $\sigma(M(x, \xi))$  is coordinate invariant, but  $h_{\pm}$  are not well defined.

**Example:** In isotropic media,  $\varepsilon = \varepsilon(x)I$ ,  $\mu = \mu(x)I$ ,

$$h_{\pm}(x, z) = \frac{1}{\sqrt{\varepsilon(x)\mu(x)}} \|z\|.$$

**Computer algebra:** If  $(x, \xi) \in T^*M$  and  $\eta = \det R R \cdot \xi$ , then  $\lambda \in \sigma(M(x, \xi))$  if and only if

$$\lambda^2(\lambda^4 - \|S \cdot \eta\|^2 \lambda^2 + \|F \cdot \eta\|^2 \|N \cdot \eta\|^2) = 0,$$

where

$$e_i = 1/\sqrt{\varepsilon_i}, \quad m_i = 1/\sqrt{\mu_i},$$

$$S = \text{diag} \left( \sqrt{e_2^2 m_3^2 + e_3^2 m_2^2}, \sqrt{e_1^2 m_3^2 + e_3^2 m_1^2}, \sqrt{e_1^2 m_2^2 + e_2^2 m_1^2} \right),$$

$$F = \text{diag} (e_2 e_3, e_1 e_3, e_1 e_2),$$

$$N = \text{diag} (m_2 m_3, m_1 m_3, m_1 m_2),$$

$\sqrt{\cdot}$  is the pos. square root,

$\|\cdot\|$  Euclidean norm.

- Quadratic solution formula:

$$\sigma(M(x, \xi)) = \pm \left\{ 0, \frac{1}{\sqrt{2}} \sqrt{\|S \cdot \eta\|^2 \pm \sqrt{D(x, \xi)}} \right\}, \quad (1)$$

where  $\pm$ -signs are independent and  $D$  is the 4-homogeneous polynomial

$$D(x, \xi) = \|S \cdot \eta\|^4 - 4\|F \cdot \eta\|^2 \|N \cdot \eta\|^2.$$

- This gives

$$h_{\pm}(x, \xi) = \frac{1}{\sqrt{2}} \sqrt{\|S \cdot \eta\|^2 \pm \sqrt{D(x, \xi)}}.$$

## The $\Delta_{ij}$ -symbols

- For  $i, j = 1, 2, 3$  let

$$\Delta_{ij} = e_i^2 m_j^2 - e_j^2 m_i^2.$$

- If two  $\Delta_{ij}$ -symbols vanish, then all symbols vanish.
- 3 possibilities: all  $\Delta_{ij} = 0$ , one  $\Delta_{ij} = 0$ , none  $\Delta_{ij} = 0$ .
- $h_{\pm}$ -functions behave qualitatively differently depending on how many of the  $\Delta_{ij}$ -symbols vanish.
- Computer algebra:

$$\begin{aligned} D &= (\Delta_{23}\xi_1^2 + \Delta_{13}\xi_2^2 + \Delta_{12}\xi_3^2)^2 - 4\Delta_{12}\Delta_{23}\xi_1^2\xi_3^2 \\ &= (\Delta_{23}\xi_1^2 + \Delta_{13}\xi_2^2 + \Delta_{21}\xi_3^2)^2 - 4\Delta_{21}\Delta_{13}\xi_2^2\xi_3^2 \\ &= (\Delta_{32}\xi_1^2 + \Delta_{13}\xi_2^2 + \Delta_{12}\xi_3^2)^2 - 4\Delta_{13}\Delta_{32}\xi_1^2\xi_2^2. \end{aligned}$$

## The three media classes

**Proposition 0.1** (Characterization of media I,  $\Delta_{ij}$  all zero)

*The following are equivalent:*

1. *All  $\Delta_{ij}$ -symbols vanish.*
2. *At least two  $\Delta_{ij}$ -symbols vanish.*
3.  $h_+ = h_-$
4. *The medium matrices satisfy*

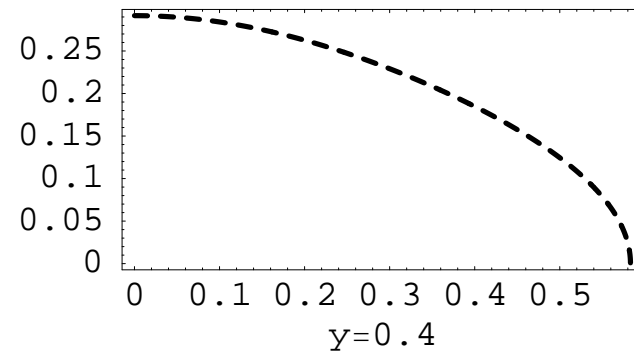
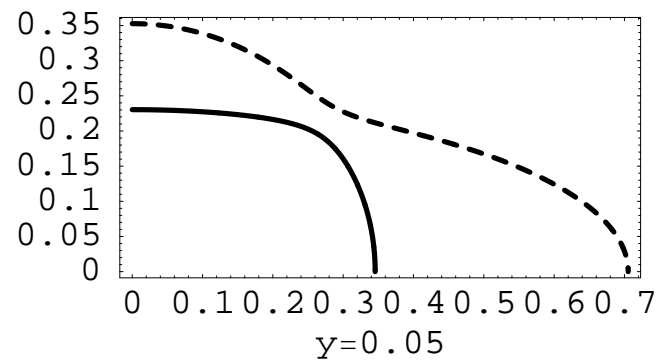
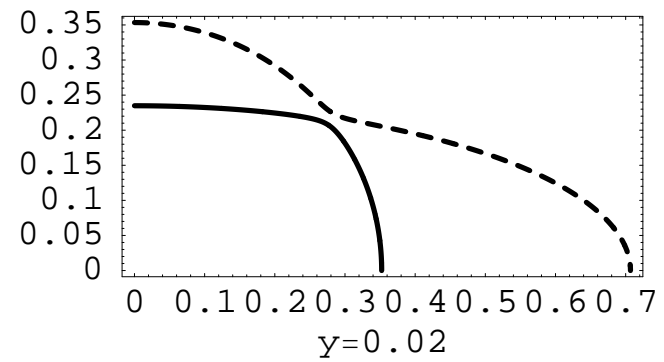
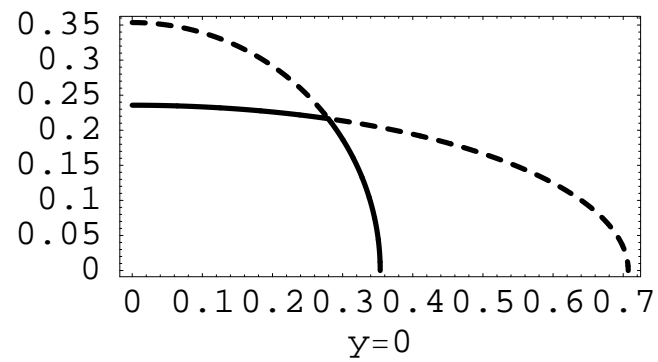
$$\varepsilon = \rho\mu$$

*for some  $\rho > 0$ .*

**Proposition 0.2 (Characterization of media II)** *The following are equivalent:*

1. *At least one of the  $\Delta_{ij}$ -symbols vanishes.*
2.  *$h_{\pm} = \sqrt{\text{pos. def. quadratic form}}$*
3.  *$h_{\pm}$  are smooth on  $\mathbb{R}^3 \setminus \{0\}$ .*
4.  *$h_{-}$  is convex on  $\mathbb{R}^3$ .*

# Characterization of media III: All $\Delta_{ij} \neq 0$





## Examples

$\Delta_{23} = 0$ , that is,  $\varepsilon_3\mu_2 = \varepsilon_2\mu_3$

$$g_{+,ij}(x) = (R^{-1} \cdot \text{diag}(\varepsilon_3\mu_2, \varepsilon_1\mu_3, \varepsilon_1\mu_2) \cdot R)_{ij}$$

$$g_{-,ij}(x) = (R^{-1} \cdot \text{diag}(\varepsilon_3\mu_2, \varepsilon_3\mu_1, \varepsilon_2\mu_1) \cdot R)_{ij}$$

Media:  $(\varepsilon_1, \varepsilon_2, \varepsilon_2), (\mu_1, \mu_2, \mu_2),$

$$g_{+,ij}(x) = \mu_2(R^{-1} \cdot \text{diag}(\varepsilon_2, \varepsilon_1, \varepsilon_1) \cdot R)_{ij}$$

$$g_{-,ij}(x) = \varepsilon_2(R^{-1} \cdot \text{diag}(\mu_2, \mu_1, \mu_1) \cdot R)_{ij}$$

Media:  $(\varepsilon_1, \varepsilon_2, \varepsilon_2), (\mu_1, \mu_1, \mu_1),$

$$g_{+,ij}(x) = \mu_2(R^{-1} \cdot \text{diag}(\varepsilon_2, \varepsilon_1, \varepsilon_1) \cdot R)_{ij}$$

$$g_{-,ij}(x) = \varepsilon_2\mu_2\delta_{ij}$$

Media:  $(\varepsilon_1, \varepsilon_1, \varepsilon_1), (\mu_1, \mu_1, \mu_1),$

$$g_{\pm,ij} = \varepsilon_1\mu_1\delta_{ij}$$