

Engineering Applications of Nonlinear Optimization

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ABSTRACT

- Brief description of interior-point methods for NLP.
- Engineering Applications
 - Finite Impulse Response (FIR) filter design
 - Antenna Array design
 - Telescope design
 - Stable orbits for the n -body problem

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LOQO: An Interior-Point Code for NLP



LOQO solves problems in the following form:

$$\begin{array}{ll} \text{minimize} & f(x) \\ \text{subject to} & b \leq h(x) \leq b + r, \\ & l \leq x \leq u \end{array}$$

The functions $f(x)$ and $h(x)$ must be twice differentiable (at least at points of evaluation).

The standard *interior-point paradigm* is used:

- Add slacks.
- Replace nonnegativities with barrier terms in objective.
- Write first-order optimality conditions.
- Rewrite optimality conditions in primal-dual symmetric form.
- Use Newton's method to get search directions...

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Interior-Point Paradigm Continued

- Use Newton's method to get search directions:

$$\begin{bmatrix} -H(x, y) - D & A^T(x) \\ A(x) & E \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix} = \begin{bmatrix} \nabla f(x) - A^T(x)y \\ -h(x) + \mu Y^{-1}e \end{bmatrix}.$$

Here, D and E are diagonal matrices involving slack variables,

$$H(x, y) = \nabla^2 f(x) - \sum_{i=1}^m y_i \nabla^2 h_i(x) + \lambda I, \text{ and } A(x) = \nabla h(x),$$

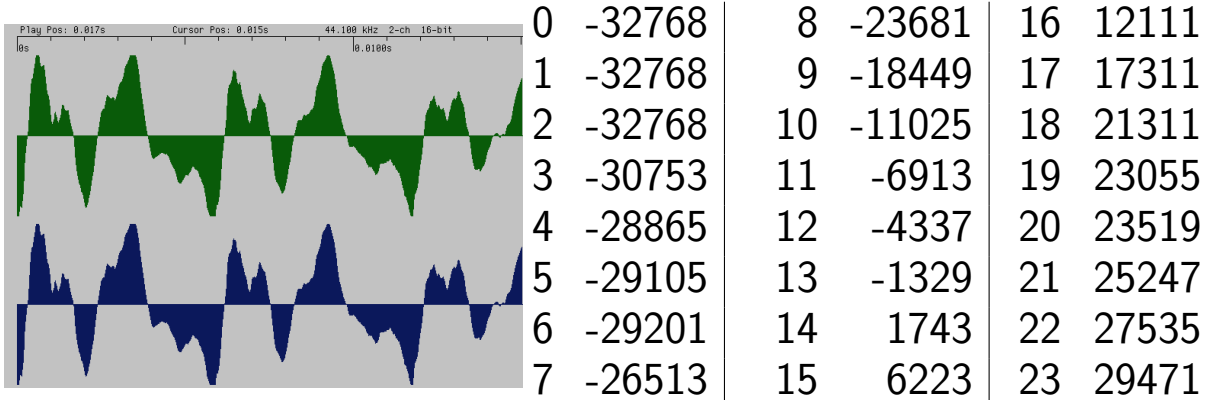
where λ is chosen to ensure appropriate descent properties.

- Compute step lengths to ensure positivity of slack variables.
- Shorten steps further to ensure a reduction in either infeasibility or in the barrier function—a myopic, or Markov, *filter*. (N.B.: We no longer use a merit function.)
- Step to new point and repeat.

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Finite Impulse Response (FIR) Filter Design

- Audio is stored digitally in a computer as a stream of short integers: $u_k, k \in \mathbb{Z}$.
- When the music is played, these integers are used to drive the displacement of the speaker from its resting position.
- For CD quality sound, 44100 short integers get played per second per channel.



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FIR Filter Design—Continued

- A *finite impulse response (FIR) filter* takes as input a digital signal and convolves this signal with a finite set of fixed numbers h_0, \dots, h_n to produce a filtered output signal:

$$y_k = \sum_{i=-n}^n h_{|i|} u_{k-i}.$$

- Sparring the details, the output power at frequency ν is given by

$$|H(\nu)|^2$$

where

$$H(\nu) = \sum_{k=-n}^n h_{|k|} e^{2\pi i k \nu} = h(0) + 2 \sum_{k=1}^n h_k \cos(2\pi k \nu),$$

- Similarly, the mean absolute deviation from a flat frequency response over a frequency range, say $\mathcal{L} \subset [0, 1]$, is given by

$$\frac{1}{|\mathcal{L}|} \int_{\mathcal{L}} |H(\nu) - 1| d\nu$$



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Filter Design: Woofer, Midrange, Tweeter

$$\text{minimize} \quad \int_0^1 |H_w(\nu) + H_m(\nu) + H_t(\nu) - 1| d\nu$$

$$\text{subject to} \quad -\epsilon \leq H_w(\nu) \leq \epsilon, \quad \nu \in W = [.2, .8]$$

$$-\epsilon \leq H_m(\nu) \leq \epsilon, \quad \nu \in M = [.4, .6] \cup [.9, .1]$$

$$-\epsilon \leq H_t(\nu) \leq \epsilon, \quad \nu \in T = [.7, .3]$$

where

$$H_i(\nu) = h_0^i + 2 \sum_{k=1}^n h_k^i \cos(2\pi k\nu), \quad i = W, M, T$$

h_k^i = filter coefficients, i.e., **decision variables**



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Conversion to a Linear Programming Problem

$$\text{minimize } \int_0^1 t(\nu) d\nu$$

$$\text{subject to } t(\nu) \leq H_w(\nu) + H_m(\nu) + H_t(\nu) - 1 \leq t(\nu) \quad \nu \in [0, 1]$$

$$-\epsilon \leq H_w(\nu) \leq \epsilon, \quad \nu \in W$$

$$-\epsilon \leq H_m(\nu) \leq \epsilon, \quad \nu \in M$$

$$-\epsilon \leq H_t(\nu) \leq \epsilon, \quad \nu \in T$$



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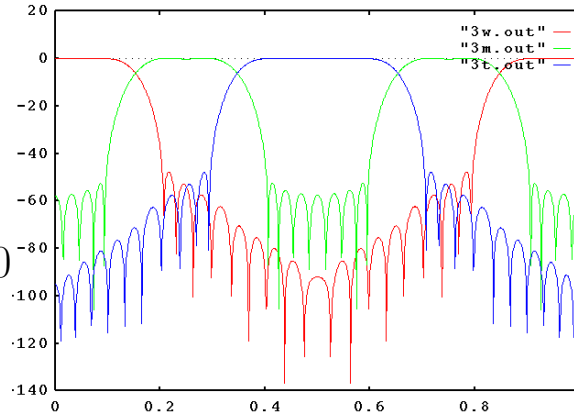
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Specific Example

filter length: $n = 14$

integral discretization: $N = 1000$



Demo: [orig-clip](#) [woofer](#) [midrange](#) [tweeter](#) [reassembled](#)

Ref: J.O. Coleman, U.S. Naval Research Laboratory,

CISS98 paper available: enr.umbc.edu/~jeffc/pubs/abstracts/ciss98.html



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Antenna Array Design

- Given: an array (linear or 2-D) of radar antennae.
- An incoming signal generates an output signal at each antenna.
- A linear combination of the signals is made to produce one total output signal.
- Coefficients of the linear combination can be chosen to accentuate and/or attenuate the output signal's strength as a function of the input signal's source direction.



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2-D Antenna-Array Design Problem

$$\begin{aligned} & \text{minimize} && \int_S |A(p)|^2 ds \\ & \text{subject to} && A(p_0) = 1, \end{aligned}$$

where

$$A(p) = \sum_{l \in \{\text{array elements}\}} w_l e^{-2\pi i p \cdot x_l}, \quad p \in S$$

w_l = complex-valued **design weight** for array element l

S = subset of unit hemisphere: sidelobe directions

x_l = spatial coord vector for array element l

p_0 = “look” direction



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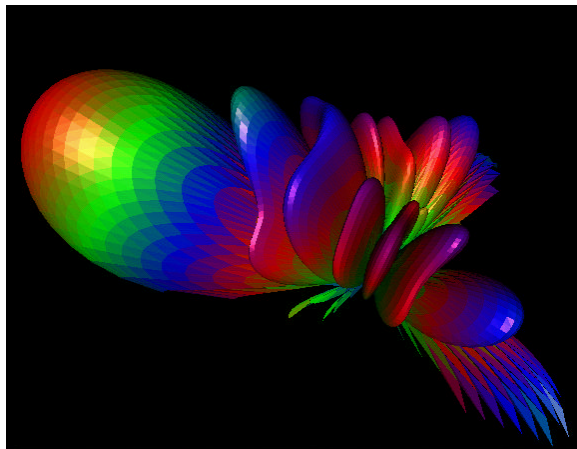
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Specific Example: Hexagonal Lattice of 61 Elements

$$\rho = -20 \text{ dB} = 0.01$$

$S = 889$ points outside 20° from look direction

$p_0 = 40^\circ$ from zenith



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Shape Optimization (Telescope Design)

The problem is to design and build a space telescope that will be able to “see” planets around nearby stars (other than the Sun).

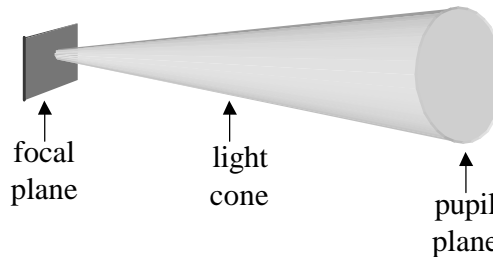
Consider a telescope. Light enters the front of the telescope. This is called the *pupil plane*.

The telescope focuses all the light passing through the pupil plane from a given direction at a certain point on the *focal plane*, say $(0, 0)$.

However, the wave nature of light makes it impossible to concentrate all of the light at a point. Instead, a small disk, called the *Airy disk*, with diffraction rings around it appears.

These diffraction rings are bright relative to any planet that might be orbiting a nearby star and so would completely hide the planet. The Sun, for example, would appear 10^{10} times brighter than the Earth to a distant observer.

By placing a mask over the pupil, one can design the shape and strength of the diffraction rings. The problem is to find an optimal shape so as to put a very deep *null* very close to the Airy disk.



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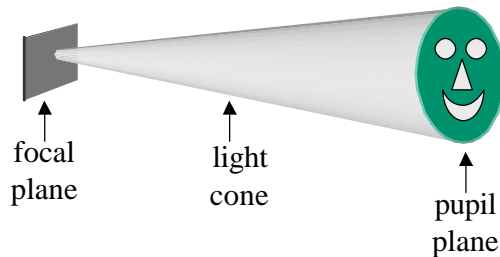
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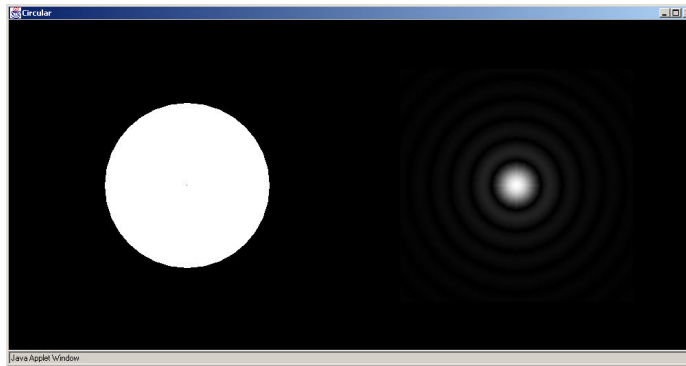
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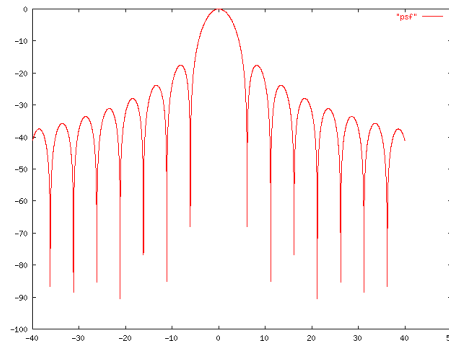
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Airy Disk and Diffraction Rings

A conventional telescope has a circular opening as depicted by the left side of the figure. Visually, a star then looks like a small disk with rings around it, as depicted on the right.



The rings grow progressively dimmer as this log-plot shows:



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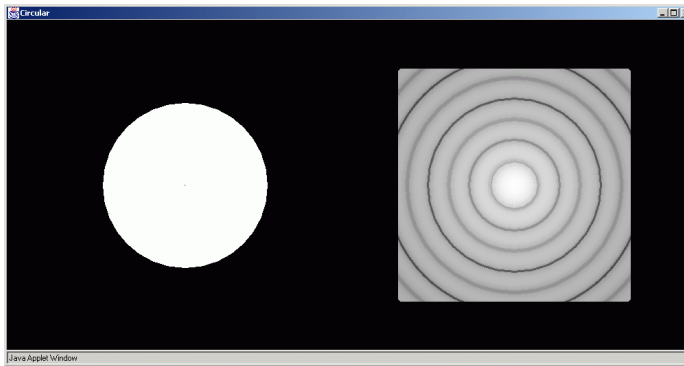
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Airy Disk and Diffraction Rings—Log Scaling

Here's the same Airy disk from the previous slide plotted using a logarithmic brightness scale with $10^{-11} = -110\text{dB}$ set to black:



The problem is to find an aperture mask, i.e. a pupil plane mask, that yields a -100 dB *null* somewhere near the first diffraction ring. A *hard problem!* Such a null would appear almost black in this log-scaled image.



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Electric Field

Consider an aperture mask consisting of an opening given by

$$\left\{ (x, y) : -\frac{1}{2} \leq x \leq \frac{1}{2}, -A(x) \leq y \leq A(x) \right\}.$$

We only consider masks that are symmetric with respect to both the x and y axes. Hence, the function $A()$ is a nonnegative even function.

In such a situation, the electric field $E(\xi, \zeta)$ is real and also symmetric about both the x and y axes. It is given by

$$\begin{aligned} E(\xi, \zeta) &= \int_{-\frac{1}{2}}^{\frac{1}{2}} \int_{-A(x)}^{A(x)} e^{i(x\xi+y\zeta)} dy dx \\ &= 4 \sum_j \int_0^{\frac{1}{2}} \cos(x\xi) \frac{\sin(A(x)\zeta)}{\zeta} dx \end{aligned}$$

The intensity of the light at (ξ, ζ) is given by the square of the electric field.



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Maximizing Throughput

Because of the symmetry, we only need to optimize in the first quadrant:

$$\text{maximize } 4 \int_0^{\frac{1}{2}} A(x) dx$$

$$\begin{aligned} \text{subject to } & -10^{-5} E(0,0) \leq E(\xi, \zeta) \leq 10^{-5} E(0,0), \quad \text{for } (\xi, \zeta) \in \mathcal{O} \\ & 0 \leq A(x) \leq 1/2, \quad \text{for } 0 \leq x \leq 1/2 \end{aligned}$$

The objective function is the total open area of the mask. The first constraint guarantees 10^{-10} light intensity throughout a desired region of the focal plane, and the remaining constraint ensures that the mask is really a mask.

If the set \mathcal{O} is a subset of the x -axis, then the problem is entirely linear (a linear programming problem).



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One Pupil w/ On-Axis Constraints

$$\mathcal{O} = \{(\xi, 0) : \xi_0 \leq \xi \leq \xi_1\}$$

$$\xi_0 = 4$$

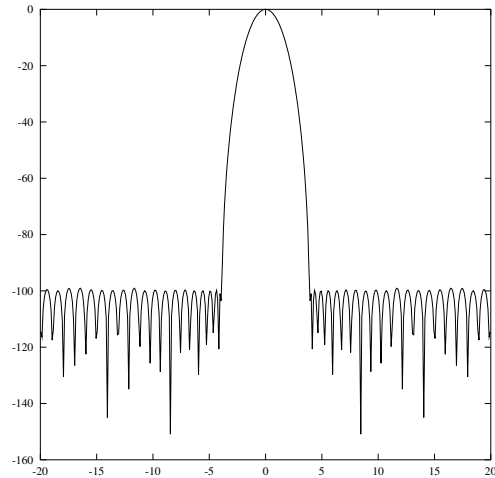
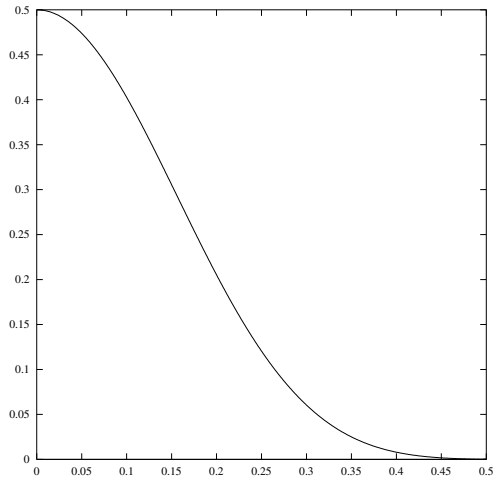
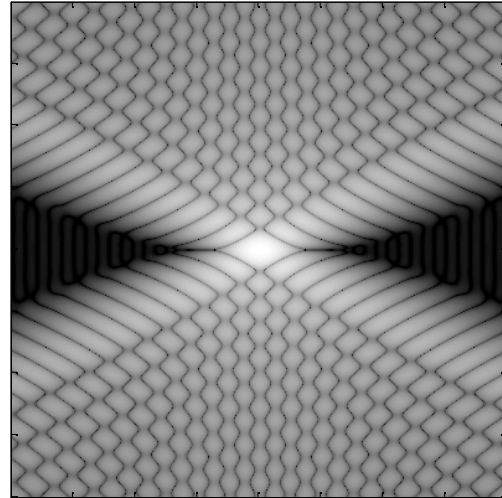
$$\xi_1 = 40$$

Log scale:

white = 0dB,

black = -110dB

Thruput = 37%



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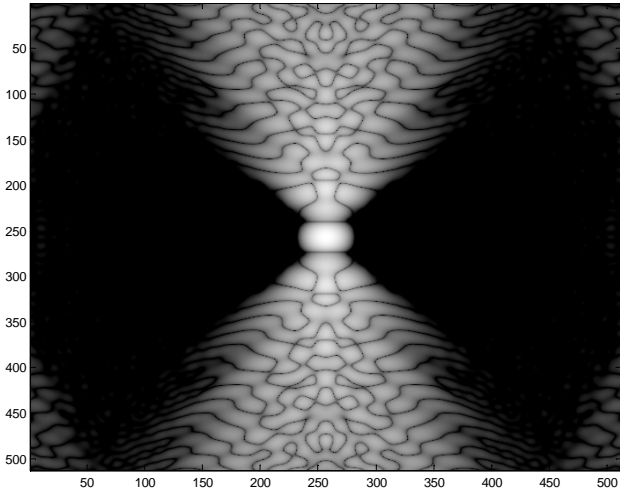
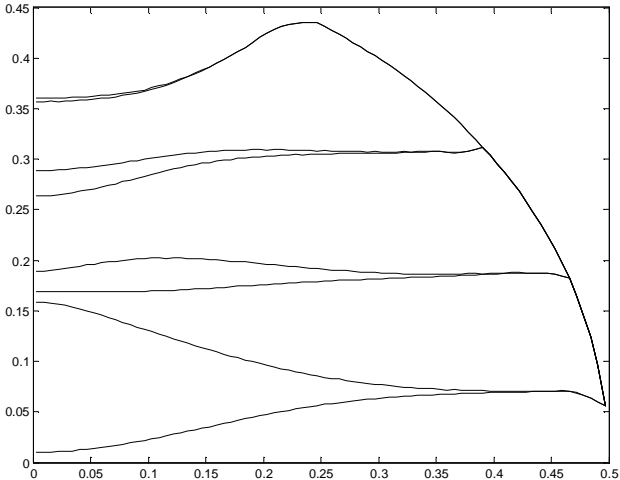
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Best Mask



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Circularly Symmetric Masks

- My original question was “Why not work with circularly symmetric optics?” In this case, one could think of making a variable filter. That is, at point (x, y) have the filter transmit a fraction $A(x, y)$ of the light.
- Such a filter is called an *apodization*.
- The answer is that apodizations are hard to make *accurately*.
- For small working bands, the square-aperture masks are essentially bang-bang all-or-nothing masks.
- It suggests looking for similar circularly symmetric masks.
- They can be thought of as apodizations in which the apodizing function $A(r)$ is zero-one valued.
- On the next few slides we derive the formulas for circularly symmetric apodization and then restrict attention to the zero-one valued case.



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Circularly Symmetric Apodization

Instead of a square mask, we consider now a circularly symmetric apodized aperture:

$$E(\xi, \zeta) = \int_0^{a/2} \int_{-\pi}^{\pi} A(r) e^{-2\pi i k(x\xi + y\zeta)/f} r d\theta dr$$

where, of course, $x = r \cos \theta$ and $y = r \sin \theta$.

WLOGWMAT, $\zeta = 0$ and hence we look at

$$\begin{aligned} E(\xi) &= \int_0^{a/2} r A(r) \left(\int_{-\pi}^{\pi} e^{-i \frac{2\pi k \xi}{f} r \cos \theta} \right) d\theta dr \\ &= \int_0^{a/2} 2\pi r A(r) J_0 \left(\frac{2\pi k r \xi}{f} \right) dr \end{aligned}$$



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Circularly Symmetric Masks



Let

$$A(r) = \begin{cases} 1 & r_{2j} \leq r \leq r_{2j+1}, \\ 0 & \text{otherwise,} \end{cases} \quad j = 0, 1, \dots, m - 1$$

where

$$0 \leq r_0 \leq r_1 \leq \dots \leq r_{2m-1} \leq a/2.$$

The integral on the previous slide can now be written as a sum of integrals and each of these integrals can be explicitly integrated to get:

$$E(\xi) = \sum_{j=0}^{m-1} \frac{f}{k\xi} \left(r_{2j+1} J_1 \left(\frac{2\pi k\xi r_{2j+1}}{f} \right) - r_{2j} J_1 \left(\frac{2\pi k\xi r_{2j}}{f} \right) \right).$$

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Circularly Symmetric Masks Optimization Problem

$$\text{maximize } \sum_{j=0}^{m-1} \pi(r_{2j+1}^2 - r_{2j}^2)$$

$$\text{subject to: } -10^{-5}E(0) \leq E(\xi) \leq 10^{-5}E(0), \quad \text{for } \xi_0 \leq \xi \leq \xi_1$$

where $E(\xi)$ is the function of the r_j 's given on the previous slide.



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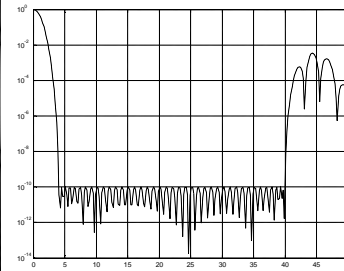
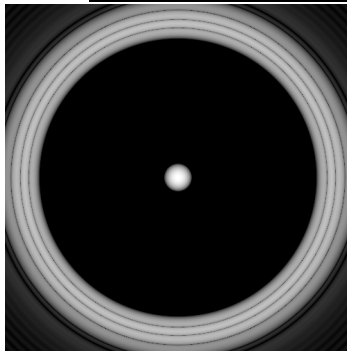
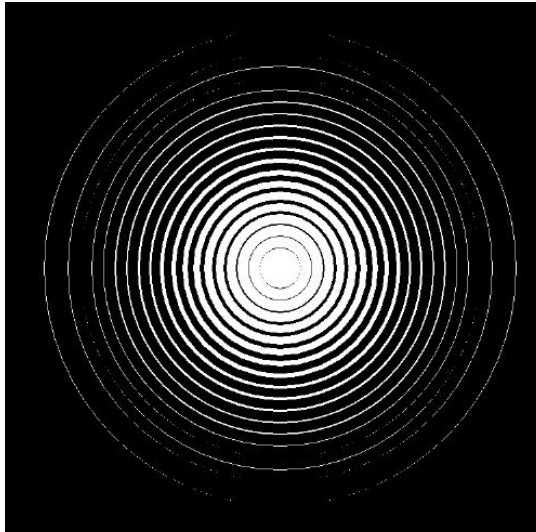
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$\xi_0 = 4\lambda f/a$ and $\xi_1 = 48\lambda f/a$ and $m = 26$



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Celestial Mechanics—Periodic Orbits

- Find periodic orbits for the planar gravitational n -body problem.

- Minimize action:

$$\int_0^{2\pi} (K(t) - P(t))dt,$$

- where $K(t)$ is kinetic energy,

$$K(t) = \frac{1}{2} \sum_i m_i (\dot{x}_i^2(t) + \dot{y}_i^2(t)),$$

- and $P(t)$ is potential energy,

$$P(t) = - \sum_{i < j} \frac{m_i m_j}{\sqrt{(x_i(t) - x_j(t))^2 + (y_i(t) - y_j(t))^2}}.$$

- Subject to periodicity constraints:

$$x_i(2\pi) = x_i(0), \quad y_i(2\pi) = y_i(0).$$

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Periodic Solutions



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We assume solutions can be expressed in the form

$$x(t) = a_0 + \sum_{k=1}^{\infty} (a_k^c \cos(kt) + a_k^s \sin(kt))$$

$$y(t) = b_0 + \sum_{k=1}^{\infty} (b_k^c \cos(kt) + b_k^s \sin(kt))$$

The variables a_0 , a_k^c , a_k^s , b_0 , b_k^c , and b_k^s are the *decision variables* in the optimization model.

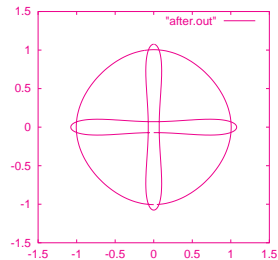
Choreographies and the Ducati Orbit



Recently, Montgomery and Chencinier (2001) and then others discovered a host of new solutions to the equimass n -body problem. They call these solutions *choreographies* because all of the bodies follow the same path — they are simply spread out uniformly along this path. They found these orbits by *minimizing* the action functional.

I reproduced these choreographies using AMPL and LOQO and noticed that all except one are *unstable*.

This inspired me to look for stable solutions. I found a few, including the one that M. Todd called the *Ducati* solution.



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Limitations of the Model

- The infinite sum gets truncated to a finite sum. This amounts to adding constraints. Hence, the solution might be suboptimal. That is, the trajectory obtained might not satisfy the equations of motion.
- Masses must be positive.
- Model can't solve 2-body problem w/ eccentricity (see next section).



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Elliptic Solutions to the 2-Body Problem



An ellipse with semimajor axis a , semiminor axis b , and having its left focus at the origin of the coordinate system is given parametrically by:

$$x(t) = f + a \cos t, \quad y(t) = b \sin t,$$

where $f = \sqrt{a^2 - b^2}$ is the distance from the focus to the center of the ellipse.

However, this is *not* the trajectory of a mass in the 2-body problem. Such a mass will travel faster around one focus than around the other. We need to introduce a time-change function $\theta(t)$:

$$x(t) = f + a \cos \theta(t), \quad y(t) = b \sin \theta(t).$$

This function θ must be increasing and must satisfy $\theta(0) = 0$ and $\theta(2\pi) = 2\pi$.

The optimization model can be used to find (a discretization of) $\theta(t)$ automatically by letting it be a vector of variables and adding appropriate monotonicity and boundary constraints.

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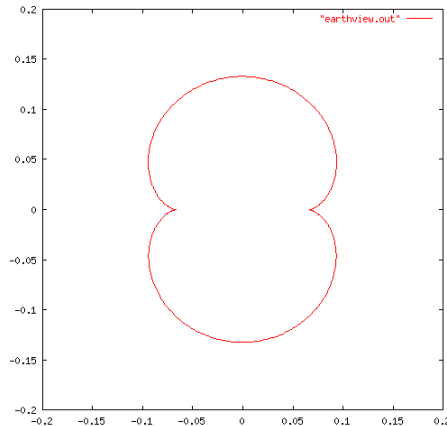
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A Hill-Type Solution to the Eccentric Sun-Earth System



Using an eccentricity $e = f/a = 0.0167$ and appropriate Sun and Earth masses, we can find a periodic Hill-Type satellite trajectory in which the satellite orbits the Earth once per year.



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