

Mathematical Programming: Modelling and Applications

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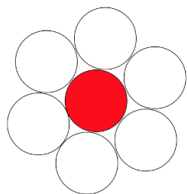
Outline

1 Kissing number problem

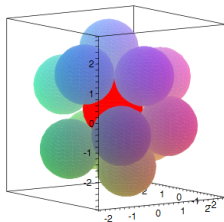
Kissing number problem

Some examples

$$n = 6, K = 2$$



$$n = 12, K = 3$$



more dimensions

n	τ (lattice)	τ (nonlattice)
0	0	
1	2	
2	6	
3	12	
4	24	
5	40	
6	72	
7	126	
8	240	
9	272	(306)*
10	336	(500)*
11	438	(582)*
12	756	(840)*
13	918	(1130)*
14	1422	(1582)*
15	2340	
16	4320	
17	5346	
18	7398	
19	10668	
20	17400	
21	27720	
22	49896	

Kissing number problem: exercise 1

Use AMPL to implement a MINLP formulation of the Kissing Number Problem (as an **optimization problem**).

Hint 1: [.dat](#) files are given for small/medium instances.

Hint 2: look at Maculan, Michelon, Smith' model (1995).

Hint 3: you can use additional variables to reformulate the binary products.

Hint 4: you can use the [knpcheck.run](#) script (which is given) to check your solution (in other words, you can include it at the bottom of your `.run` file).

Hint 5: use [Baron](#).

Set a threshold (for example, CPU Time limit 100 s., i.e. `maxtime=100` in the options of your `.run` script) and test these instances:

(a) $n=12$ and $k=3$

(b) $n=7$ and $k=2$

In theory, your program should be able to accomplish (a) but not (b).

Then, download [knp.mod](#) and [knp.run](#). Your implementation should be quite similar.

Kissing number problem: exercise 2

Use AMPL to implement the Kissing Number Problem (as a **decision problem**).

Hint: look at Kucherenko, Belotti, Liberti, Maculan' model (2007).

Set again a threshold (for example, CPU Time limit 100 s., i.e. `maxtime=100` in the options of your `.run` script) and test again these instances:

- (a) $n=12$ and $k=3$
- (b) $n=7$ and $k=2$

In theory, the program should not be able to accomplish neither (a) nor (b). Then, download [knpfeas18.mod](#) and [knpfeas18.run](#). Your implementation should be quite similar.

Change the constraint on α , to see if something changes.

Reminder: SDP

SEMIDEFINITE PROGRAMMING (SDP) refers to optimization problems where decision variables are a symmetric matrix that is required to be semidefinite positive. The standard form is

$$\left. \begin{array}{l} \min \quad C \bullet X \\ \forall i \leq m \quad A^i \bullet X = b^i \\ X \succeq 0, \end{array} \right\} \quad (1)$$

where:

- C and A^i (for $i \leq m$) are $n \times n$ symmetric matrices
- X is an $n \times n$ matrix of decision variables,
- $b^i \in \mathbb{R}$ for all $i \leq m$
- for two $n \times n$ matrices $L = (\lambda_{ij})$, $M = (\mu_{ij})$ we have $L \bullet M = \sum_{i,j \leq n} \lambda_{ij} \mu_{ij}$
- $X \succeq 0$, i.e. X is positive semidefinite

Reminder: SDP

To make Eq. (1) clearer, write out the componentwise product \bullet of the matrices:

$C = (c_{jh})$, $A^i = (a_{jh}^i)$ and $X = (x_{jh})$:

$$\begin{aligned} \min \quad & \sum_{j,h \leq n} c_{jh} x_{jh} \\ \forall i \leq m \quad & \sum_{j,h \leq n} a_{jh}^i x_{jh} = b^i. \end{aligned}$$

This is just an LP subject to a semidefinite constraint $X \succeq 0$. What does $X \succeq 0$ mean? X , square and symmetric matrix, with real values is PSD if $\forall v \in \mathbb{R}^n$, we have that the scalar $v^T X v \geq 0$.

This is the same as requiring the decision variables x_{jh} to take values such that, when they are arranged in a $n \times n$ array, the resulting matrix A has non-negative eigenvalues.

Kissing number problem: exercise 3

Use Octave to implement an SDP relaxation of the KNP (decision version, [KBLM07])

Then, download [SDP_KBLM07.m](#) (your code should not be very different).

Test some instances:

- (b) $n=7$ and $k=2$
- (c) $n=8$ and $k=2$
- (d) $n=9$ and $k=2$
- (e) ...

Can you explain this behaviour, with growing n ?

Kissing number problem: exercise 4

Apparently, this first version of a SDP relaxation of KNP (i.e. [SDP_KBLM07.m](#)) is feasible independently from n . This makes it not really useful (cf. *Uselessness Theorem* in Liberti's slides)

Look for an improved SDP relaxation that deals with this drawback.

Initially, focus on **2D case** only ($k = 2$).

Hint: add a constraint on the dimension of the sdp matrix (i.e. decision variable).

Hint 2: for example, estimate the perimeter of the central sphere "consumed" by the placements.

Kissing number problem: exercise 5

Extend your code to deal with 3D case ($k = 3$).

Hint: this would be an approximation.

Kissing number problem: Upper bounds

Download [delsartebnd.m](#), [delsartelp.m](#) and [gegenbauer.m](#) and test it in Octave in the following way:

```
[b,r] = delartebnd(3, 60, 10);
```

Similarly, Download [pfenderbnd.m](#) and [pfenderlp.m](#) and test it in Octave in the following way:

```
[p,s] = pfenderbnd(3, 60, 10);
```

Your outcome should be, respectively:

delsartebnd: bound positivity in [0,0.000100], accepting

pfenderbnd: bound positivity in [0,0.000100], accepting

Can you explain what this means?

Kissing number problem: scripting

In the end, look in these scripts:

- how to optimize unconstrained functions within Octave using `sqp()` (examples in `dgpnlp.m`, `delsartebnd.m` and `pfenderbnd.m`).
- how to call `glpk()` within octave (examples in `delsartelp.m` and `pfenderlp.m`).
- how to compute Gegenbauer polynomial coefficients.