



# Optimization with Gurobi and Python

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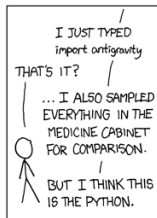
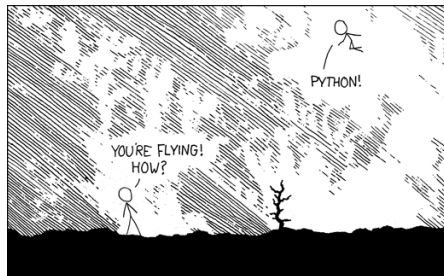
# Gurobi – a one-page explanation

- Optimization system by Z. **Gu**, E. **Rothberg**, and R. **Bixby**



- Very high performance, cutting-edge solvers:
  - linear programming
  - quadratic programming
  - mixed-integer programming
- Advanced presolve methods
- MILP and MIQP models:
  - cutting planes
  - powerful solution heuristics
- Free academic license

# Why Python?



- Everything can be done after loading a module!
- Optimization allowed:  

```
import gurobipy
```
- Use algorithms on graphs:  

```
import networkX  
import matplotlib
```
- Allows levitation:  

```
import antigravity (?)
```

# Python — a one-page explanation

- Simple types: bools, integers, floats, strings (*immutable*)
- Complex types:
  - lists: sequences of elements (of any type; *mutable*)
    - indexed by an integer, from 0 to size-1
    - `A=[1,5,3,7], A.append(5), a=A.pop(), a=A[3], A[4]="abc", A.remove(6), A.sort()`
  - tuples: as lists, but *immutable* → may be used as indices
    - `T=(1,5,3,7), t=T[3]`
  - dictionaries: mappings composed of pairs *key, value* (*mutable*)
    - indexed by an integer, from 0 to size-1
    - `D = {}, D[872]=6, D["pi"]=3.14159, D[(1,7)]=3`
- Iteration:

*lists:*

```
for i in A:  
    print i
```

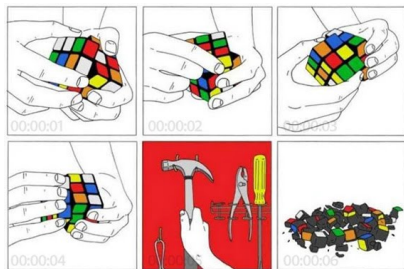
*dictionaries:*

```
for i in D:  
    print i, D[i]
```

*cycles:*

```
i = 0  
while i < 10:  
    print i  
    i += 1
```

# Putting things together



- import the `gurobipy` module
- create a `model` object
  - add variables
  - add constraints
- *[debug?]*
- solve
- report solution

# Hello world example

minimize  $3000x + 4000y$   
subject to:  $5x + 6y \geq 10$   
 $7x + 5y \geq 5$   
 $x, y \geq 0$

```
000000010100011011000000100101100011
01011101000100011111111110100000100
001011000011010111011010110110010001
100000101011001000100001110001001111
110010110100110110100111101111011110
01000100111010100110011000011010
#include <stdio.h>
0011010001010001110
001int main()
001{
011000 printf("Hello World");
0001111 return 42;
010001000111010010110001101000011010
001101111010111011110000001010001110
00100010101100100111011101000101111
00111001101010111000101010100011000
0110000011011111101010011111000110
```

```
from gurobipy import *
model = Model("hello")
```

```
x = model.addVar(obj=3000, vtype="C", name="x")
y = model.addVar(obj=4000, vtype="C", name="y")
model.update()
```

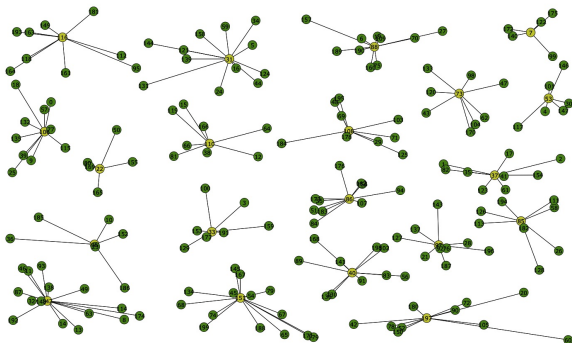
```
L1 = LinExpr([5,6],[x,y])
model.addConstr(L1,">",10)
L2 = LinExpr([7,5],[x,y])
model.addConstr(L2,">",5)
```

```
model.ModelSense = 1 # minimize
model.optimize()
```

```
if model.Status == GRB.OPTIMAL:
    print "Opt. Value=",model.ObjVal
    print "x* =", x.X
    print "y* =", y.X
```

# The $k$ -median problem

- facility location problem of min-sum type
- $n$  customers
- $m$  positions for facilities (at some customer's coordinates)
- $k$  maximum open facilities
- minimize service time summed for all the customers
- (Euclidean distance, random uniform  $(x, y)$  coordinates)



# The $k$ -median problem — formulation



- $n$  customers,  $m$  facilities
- variables:
  - $x_{ij} = 1$  if customer  $i$  is served by facility  $j$
  - $y_j = 1$  if facility  $j$  is open
- ① all customers must be served
- ② maximum of  $k$  open facilities
- ③ customer  $i$  can be served by  $j$  only if  $j$  is open
- ④ minimize total, accumulated service time

$$\begin{array}{ll}\text{minimize} & \sum_i \sum_j c_{ij} x_{ij} \\ \text{subject to} & \sum_j x_{ij} = 1 \quad \forall i \\ & \sum_j y_j = k \\ & x_{ij} \leq y_j \quad \forall i, j \\ & x_{ij} \in \{0, 1\} \quad \forall i \\ & y_j \in \{0, 1\} \quad \forall j\end{array}$$



# The $k$ -median problem — Python/Gurobi model

```
def kmedian(m, n, c, k):  
    model = Model("k-median")  
    y,x = {}, {}  
    for j in range(m):  
        y[j] = model.addVar(obj=0, vtype="B", name="y[%s]"%j)  
        for i in range(n):  
            x[i,j] = model.addVar(obj=c[i,j], vtype="B", name="x[%s,%s]"%(i,j))  
    model.update()  
  
    for i in range(n):  
        coef = [1 for j in range(m)]  
        var = [x[i,j] for j in range(m)]  
        model.addConstr(LinExpr(coef,var), "=", 1, name="Assign[%s]"%i)  
    for j in range(m):  
        for i in range(n):  
            model.addConstr(x[i,j], "<", y[j], name="Strong[%s,%s]"%(i,j))  
    coef = [1 for j in range(m)]  
    var = [y[j] for j in range(m)]  
    model.addConstr(LinExpr(coef,var), "=", rhs=k, name="k _ median")  
  
    model.update()  
    model.__data = x,y  
    return model
```



# The $k$ -median problem — preparing data

```
import math
import random
def distance(x1, y1, x2, y2):
    return math.sqrt((x2-x1)**2 + (y2-y1)**2)

def make_data(n):
    x = [random.random() for i in range(n)]
    y = [random.random() for i in range(n)]
    c = {}
    for i in range(n):
        for j in range(n):
            c[i,j] = distance(x[i],y[i],x[j],y[j])
    return c, x, y
```

# The $k$ -median problem — calling and solving

```
n = 200
c, x_pos, y_pos = make_data(n)
m = n
k = 20
model = kmedian(m, n, c, k)

model.optimize()
x,y = model.__data
edges = [(i,j) for (i,j) in x if x[i,j].X == 1]
nodes = [j for j in y if y[j].X == 1]
print "Optimal value=", model.ObjVal
print "Selected nodes:", nodes
print "Edges:", edges
```

# The $k$ -median problem — plotting

```
import networkx as NX
import matplotlib.pyplot as P
P.ion() # interactive mode on
G = NX.Graph()

other = [j for j in y if j not in nodes]
G.add_nodes_from(nodes)
G.add_nodes_from(other)
for (i,j) in edges:
    G.add_edge(i,j)

position = {}
for i in range(n):
    position[i]=(x_pos[i],y_pos[i])

NX.draw(G, position, node_color='y', nodelist=nodes)
NX.draw(G, position, node_color='g', nodelist=other)
```



# The $k$ -median problem — solver output

Optimize a model with 40201 rows, 40200 columns and 120200 nonzeros

Presolve time: 1.67s

Presolved: 40201 rows, 40200 columns, 120200 nonzeros

Variable types: 0 continuous, 40200 integer (40200 binary)

Found heuristic solution: objective 22.1688378

Root relaxation: objective 1.445152e+01, 2771 iterations, 0.55 seconds

Nodes		Current Node		Objective Bounds				Work	
Expl	Unexpl	Obj	Depth	IntInf	Incumbent	BestBd	Gap	It/Node	Time
0	0	14.45152	0	92	22.16884	14.45152	34.8%	-	2s
H	0	0			14.4528610	14.45152	0.01%	-	2s

Cutting planes:

Gomory: 1

Zero half: 1

Explored 0 nodes (2771 simplex iterations) in 2.67 seconds

Thread count was 1 (of 8 available processors)

Optimal solution found (tolerance 1.00e-04)

Best objective 1.445286097717e+01, best bound 1.445151681275e+01, gap 0.0093%

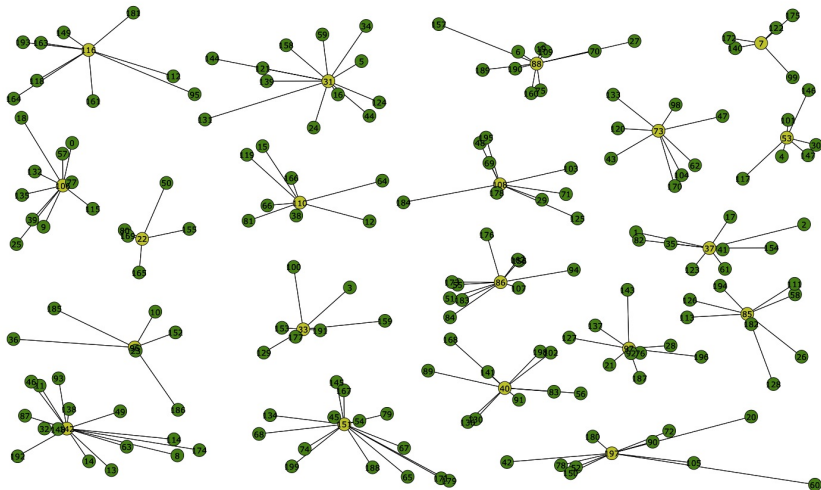
Optimal value= 14.4528609772

Selected nodes: [7, 22, 31, 33, 37, 40, 53, 73, 85, 86, 88, 96, 97, 106, 108, 110, 116, 142, 151, 197]

Edges: [(57, 106), (85, 85), (67, 151), (174, 142), (139, 31), (136, 40), (35, 37), (105, 106), (105, 108)]

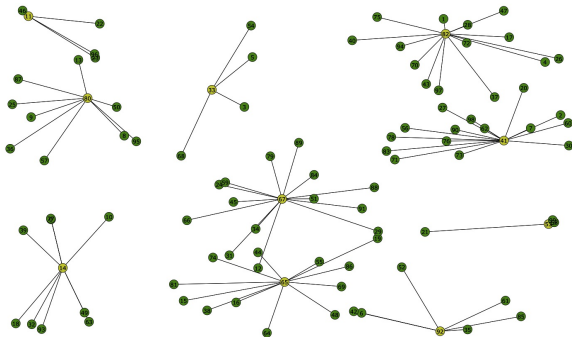
max c: 0.257672494705

## The $k$ -median problem: solution



# The $k$ -center problem

- facility location problem of min-max type
- $n$  customers
- $m$  positions for facilities (at some customer's coordinates)
- $k$  maximum open facilities
- minimize service time for the **latest-served** customer
- (Euclidean distance, random uniform  $(x, y)$  coordinates)



# The $k$ -center problem — formulation (min-max type)

- $x_{ij} = 1$  if customer  $i$  is served by facility  $j$
- $y_j = 1$  if a facility  $j$  is open
- 1 all customers must be served
- 2 maximum of  $k$  open facilities
- 3 customer  $i$  can be served by  $j$  only if  $j$  is open
- 4 update service time for the latest-served customer

minimize  
subject to

$$\begin{aligned} z \\ \sum_j x_{ij} &= 1 & \forall i \\ \sum_j y_j &= k \\ x_{ij} &\leq y_j & \forall i, j \\ c_{ij} x_{ij} &\leq z & \forall i, j \\ x_{ij} &\in \{0, 1\} & \forall i, j \\ y_j &\in \{0, 1\} & \forall j \end{aligned}$$



# The $k$ -center problem — Python/Gurobi model

```
def kcenter(m, n, c, k):
    model = Model("k-center")
    z = model.addVar(obj=1, vtype="C", name="z")
    y, x = {}, {}
    for j in range(m):
        y[j] = model.addVar(obj=0, vtype="B", name="y[%s]"%j)
        for i in range(n):
            x[i,j] = model.addVar(obj=0, vtype="B", name="x[%s,%s]"%(i,j))
    model.update()
    for i in range(n):
        coef = [1 for j in range(m)]
        var = [x[i,j] for j in range(m)]
        model.addConstr(LinExpr(coef,var), "=", 1, name="Assign[%s]"%i)
    for j in range(m):
        for i in range(n):
            model.addConstr(x[i,j], "<", y[j], name="Strong[%s,%s]"%(i,j))
    for i in range(n):
        for j in range(n):
            model.addConstr(LinExpr(c[i,j],x[i,j]), "<", z, name="Max_x[%s,%s]"%(i,j))
    coef = [1 for j in range(m)]
    var = [y[j] for j in range(m)]
    model.addConstr(LinExpr(coef,var), "=", rhs=k, name="k_center")
    model.update()
    model.__data = x,y
    return model
```

# The $k$ -center problem — solver output

Optimize a model with 20101 rows, 10101 columns and 50000 nonzeros

Presolve removed 100 rows and 0 columns

Presolve time: 0.35s

Presolved: 20001 rows, 10101 columns, 49900 nonzeros

Variable types: 1 continuous, 10100 integer (10100 binary)

Found heuristic solution: objective 0.9392708

Found heuristic solution: objective 0.9388764

Found heuristic solution: objective 0.9335182

Root relaxation: objective 3.637572e-03, 13156 iterations, 1.88 seconds

Nodes		Current Node		Objective Bounds			Work	
Expl	Unexpl	Obj	Depth	IntInf	Incumbent	BestBd	Gap	It/Node Time
0	0	0.00364	0	9255	0.93352	0.00364	100%	- 3s

[...]

H	7	0			0.2187034	0.21870	0.0%	603	454s
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Cutting planes:

Gomory: 1

Zero half: 2

Explored 7 nodes (83542 simplex iterations) in 454.11 seconds

Thread count was 1 (of 8 available processors)

Optimal solution found (tolerance 1.00e-04)

Best objective 2.187034280810e-01, best bound 2.187034280810e-01, gap 0.0%

Optimal value= 0.218703428081

Selected nodes: [12, 14, 23, 33, 41, 51, 53, 72, 80, 92]

Edges: [(53, 53), (36, 80), (54, 33), (69, 12), (39, 14), (86, 51), (99, 53), (37, 41), (49, 14), (26, 72), (2

- $k$ -median instance:  $n = m = 200$ ,  $k = 20$ , CPU = 5s
- $k$ -center instance:  $n = m = 100$ ,  $k = 10$ , CPU = 454s
- $k$ -center: for an instance that is **half size** the one solved for  $k$ -median, used **almost ten times** more CPU
- can we do better?

# The $k$ -center problem — formulation (min type)

- $a_{ij} = 1$  if customer  $i$  **can be** served by facility  $j$
- $y_j = 1$  if a facility  $j$  is open
- $\xi_i = 1$  if customer  $i$  cannot be served
- parameter: distance  $\theta$  for which a client can be served
  - if  $c_{ij} < \theta$  then set  $a_{ij} = 1$
  - else, set  $a_{ij} = 1$
- ① either customer  $i$  is served or  $\xi_i = 1$
- ② maximum of  $k$  open facilities

$$\begin{array}{ll}\text{minimize} & \sum_i \xi_i \\ \text{subject to} & \sum_j a_{ij} y_j + \xi_i \geq 1 \quad \forall i \\ & \sum_j y_j = k \\ & \xi_i \in \{0, 1\} \quad \forall i \\ & y_j \in \{0, 1\} \quad \forall j\end{array}$$

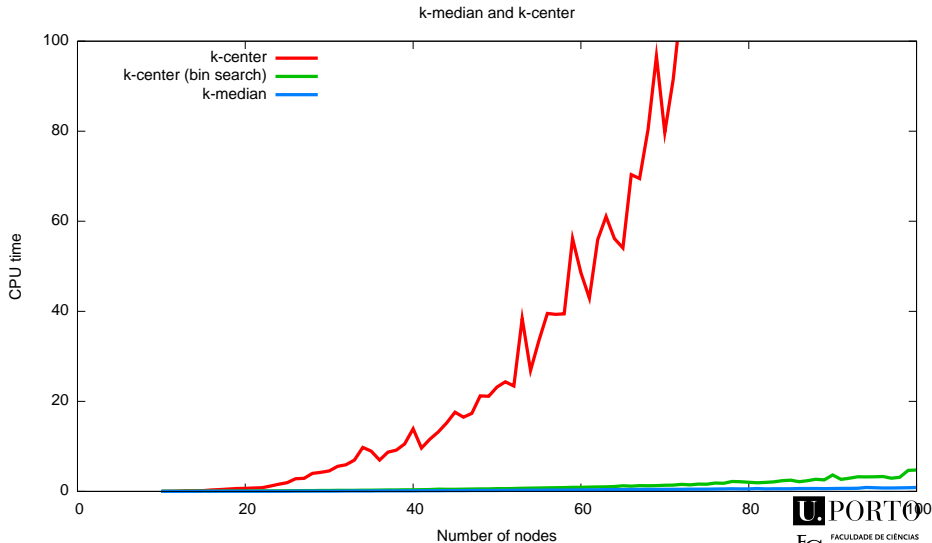
# The $k$ -center problem — model for binary search

```
def kcenter(m, n, c, k, max_c):
    model = Model("k-center")
    z, y, x = {}, {}, {}
    for i in range(n):
        z[i] = model.addVar(obj=1, vtype="B", name="z[%s]"%i)
    for j in range(m):
        y[j] = model.addVar(obj=0, vtype="B", name="y[%s]"%j)
        for i in range(n):
            x[i,j] = model.addVar(obj=0, vtype="B", name="x[%s,%s]"%(i,j))
    model.update()
    for i in range(n):
        coef = [1 for j in range(m)]
        var = [x[i,j] for j in range(m)]
        var.append(z[i])
        model.addConstr(LinExpr(coef,var), "=", 1, name="Assign[%s]"%i)
    for j in range(m):
        for i in range(n):
            model.addConstr(x[i,j], "<", y[j], name="Strong[%s,%s]"%(i,j))
    coef = [1 for j in range(m)]
    var = [y[j] for j in range(m)]
    model.addConstr(LinExpr(coef,var), "=", rhs=k, name="k_center")
    model.update()
    model.__data = x,y,z
    return model
```

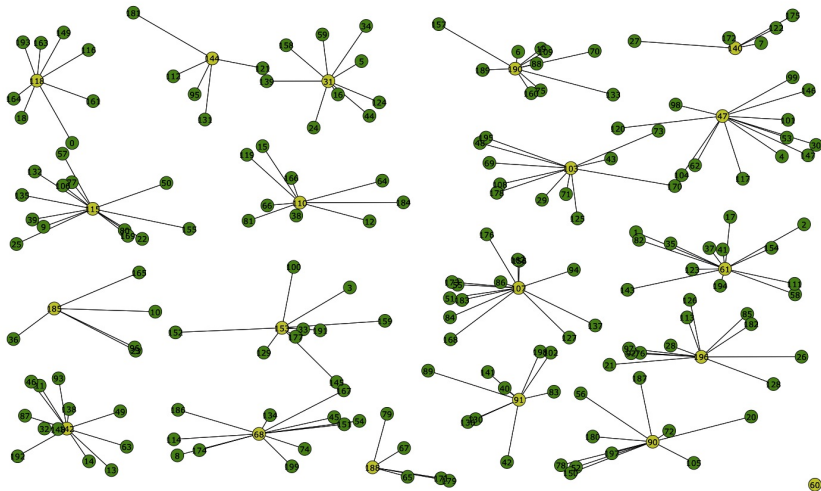
# The $k$ -center problem — binary search

```
def solve_kcenter(m, n, c, k, max_c, delta):
    model = kcenter(m, n, c, k, max_c)
    x,y,z = model.__data
    LB = 0
    UB = max_c
    while UB-LB > delta:
        theta = (UB+LB) / 2.
        for j in range(m):
            for i in range(n):
                if c[i,j]>theta:
                    x[i,j].UB = 0
                else:
                    x[i,j].UB = 1.0
        model.update()
        model.optimize()
        infeasibility = sum([z[i].X for i in range(m)])
        if infeasibility > 0:
            LB = theta
        else:
            UB = theta
            nodes = [j for j in y if y[j].X == 1]
            edges = [(i,j) for (i,j) in x if x[i,j].X == 1]
    return nodes, edges
```

# The $k$ -center problem: CPU usage



# The $k$ -center problem: solution





# The $k$ -median (left) and $k$ -center (right) solutions

