# Masterclass

May 13, 2021

Combinatorial Optimization course, FEE CTU in Prague. Created by Industrial Informatics Department.

```
[1]: %pip install -i https://pypi.gurobi.com gurobipy import gurobipy as g import matplotlib.pyplot as plt import numpy as np import networkx as nx
```

```
Looking in indexes: https://pypi.gurobi.com
Requirement already satisfied: gurobipy in
/home/benedond/anaconda3/lib/python3.7/site-packages (9.1.0)
Note: you may need to restart the kernel to use updated packages.
```

# 1 Automatized model tuning

As an example, we will revisit the model for Game of Fiver which for larger board sizes took considerable time to solve:

```
m.params.PreDual = 1
m.params.outputflag = 0

m.optimize()

X = [[int(round(x[i,j].X)) for j in range(1,n+1)] for i in range(1,n+1)]
return m.runtime
```

```
[3]: # run with default parameters
def_time = game_of_fivers(21)
print('Time with default settings: {}s'.format(def_time))
```

Using license file /home/benedond/Apps/gurobi/gurobi910/gurobi.lic Academic license - for non-commercial use only - expires 2021-06-18 Gurobi Optimizer version 9.1.0 build v9.1.0rc0 (linux64)

Thread count: 2 physical cores, 4 logical processors, using up to 4 threads Optimize a model with 442 rows, 970 columns and 2734 nonzeros

Model fingerprint: 0xaa07108c

Variable types: 0 continuous, 970 integer (529 binary)

Coefficient statistics:

Matrix range [1e+00, 2e+00]
Objective range [1e+00, 1e+00]
Bounds range [1e+00, 1e+00]
RHS range [1e+00, 1e+00]

Presolve removed 1 rows and 96 columns

Presolve time: 0.01s

Presolved: 441 rows, 874 columns, 2533 nonzeros

Variable types: 0 continuous, 874 integer (518 binary)

Root relaxation: objective 9.585736e+01, 1221 iterations, 0.06 seconds

	Node	s	Cu	rrent	Node	Э	Object	ive Bo	ınds		Worl	X.
Ex	cpl Un	expl	Obj	Depth	Int	tInf	Incumbent	Bes <sup>-</sup>	tBd	Gap	It/Node	Time
	0	0	95.85	736	0	440	_	95.85	736	_	_	0s
	0	0	96.94	696	0	468	_	96.94	696	_	-	0s
	0	0	96.98	3572	0	464	_	96.98	572	_	-	0s
	0	0	96.98	726	0	465	_	96.98	726	_	-	0s
	0	0	96.98	919	0	470	_	96.98	919	_	-	0s
	0	0	96.98	928	0	471	_	96.98	928	_	-	0s
	0	0	97.67	'888	0	479	_	97.678	388	_	-	0s
	0	0	97.71	186	0	478	_	97.71	186	_	_	0s
	0	0	97.71	.283	0	479	_	97.71	283	_	-	0s
	0	0	98.44	142	0	485	_	98.44	142	_	-	0s
	0	0	98.51	.498	0	491	_	98.51	498	_	-	0s
	0	0	98.51	.785	0	492	_	98.51	785	_	-	0s
	0	0	99.37	022	0	490	_	99.37	022	_	_	0s

```
0
           0 99.45836
                              495
                                               99.45836
                                                                        0s
    0
                             498
           0 99.48538
                           0
                                               99.48538
                                                                        0s
    0
           0 99.49044
                           0
                             500
                                               99.49044
                                                                        0s
    0
           0 99.49082
                              503
                                               99.49082
                                                                        0s
                           0
    0
           0 100.38463
                              506
                                           - 100.38463
                                                                        0s
     0
           0 100.69590
                             511
                                           - 100.69590
                                                                        0s
    0
           0 101.33193
                              520
                                           - 101.33193
                                                                        0s
    0
           0 102.31607
                              529
                                           - 102.31607
                                                                        0s
    0
           0 102.60194
                             538
                                           - 102.60194
                           0
                                                                        0s
    0
           0 103.02475
                           0
                             529
                                           - 103.02475
                                                                        0s
    0
                                                                        1s
           0 103.51409
                              536
                                           - 103.51409
                           0
    0
           0 103.65124
                              543
                                           - 103.65124
                                                                        1s
    0
           0 103.98002
                             541
                                           - 103.98002
                           0
                                                                        1s
     0
           0 104.68522
                              551
                                           - 104.68522
                                                                        1s
    0
           0 105.01278
                              553
                                           - 105.01278
                                                                        1s
    0
           0 105.20101
                              565
                                           - 105.20101
                           0
                                                                        1s
    0
           0 105.20101
                           0
                              565
                                           - 105.20101
                                                                        1s
    0
           2 105.21072
                          0
                             565
                                           - 105.21072
                                                                        1s
        413 108.80917
                                           - 105.44028
                                                                 142
                                                                        5s
   493
                          11
                              525
  793
        622 114.07780
                          26
                              655
                                           - 114.07780
                                                                 131
                                                                       10s
  833
        649 117.40460
                          30
                              663
                                           - 117.40460
                                                                 125
                                                                       15s
   906
        698
             118.68468
                          46
                             747
                                           - 118.68468
                                                                 115
                                                                       20s
   982
        748
             122.43603
                         31 749
                                              119.20375
                                                                 106
                                                                       25s
                                 245.0000000 119.31481 51.3%
H 1007
                                                                 131
                                                                       26s
        723
```

#### Cutting planes:

MIR: 259

Flow cover: 50 Zero half: 132

Explored 1007 nodes (137125 simplex iterations) in 26.90 seconds Thread count was 4 (of 4 available processors)

Solution count 1: 245

Optimal solution found (tolerance 1.00e-04)
Best objective 2.450000000000e+02, best bound 2.45000000000e+02, gap 0.0000%
Time with default settings: 26.91558289527893s

```
[4]: # lets try to tune some of the params
model = g.read('fivers_n21.lp')

model.params.tuneResults = 1
model.params.TuneTimeLimit = 30  # how much time to invest into the tuning

model.tune()
if model.tuneResultCount > 0:
```

```
model.getTuneResult(0)
model.write('fivers_tuned_params.prm')
```

Read LP format model from file fivers\_n21.lp
Reading time = 0.00 seconds
: 442 rows, 970 columns, 2734 nonzeros
Changed value of parameter tuneResults to 1
 Prev: -1 Min: -1 Max: 200000000 Default: -1
Changed value of parameter TuneTimeLimit to 30.0
 Prev: -1.0 Min: -1.0 Max: inf Default: -1.0

Solving model using baseline parameter set with TimeLimit=3s

Testing candidate parameter set 1...

Default parameters

Solving with random seed #1 ...

Optimize a model with 442 rows, 970 columns and 2734 nonzeros

Model fingerprint: 0xd83b2255

Variable types: 0 continuous, 970 integer (529 binary)

Coefficient statistics:

Matrix range [1e+00, 2e+00]
Objective range [1e+00, 1e+00]
Bounds range [1e+00, 1e+00]
RHS range [1e+00, 1e+00]

Presolve removed 1 rows and 96 columns

Presolve time: 0.01s

Presolved: 441 rows, 874 columns, 2533 nonzeros

Variable types: 0 continuous, 874 integer (518 binary)

Root relaxation: objective 9.585736e+01, 1221 iterations, 0.06 seconds

Node	s	1		Curren	t N	lod	е	Ι	Object	ive	Bounds		-	Wo	rk
Expl Un	expl		Obj	j Dep	th	In	tInf		Incumbent		BestBd	Gaj	) I	It/Nod	e Time
0	0		95.8	35736		0	440		_	95	.85736		-	-	0s
0	0		96.9	94696		0	468		_	96	.94696		-	-	0s
0	0		96.9	98572		0	464		_	96	.98572		-	-	0s
0	0		96.9	98726		0	465		_	96	.98726		_	-	0s
0	0		96.9	98919		0	470		_	96	.98919		-	-	0s
0	0		96.9	98928		0	471		_	96	.98928		_	-	0s
0	0		97.6	37888		0	479		_	97	.67888		_	-	0s
0	0		97.7	1186		0	478		_	97	.71186		-	-	0s
0	0		97.7	1283		0	479		_	97	.71283		_	-	0s
0	0		98.4	14142		0	485		_	98	.44142		_	-	0s
0	0		98.5	51498		0	491		_	98	.51498		_	_	0s

0	0	98.51785	0	492	-	98.51785	_	-	0s
0	0	99.37022	0	490	-	99.37022	_	_	0s
0	0	99.45836	0	495	-	99.45836	_	-	0s
0	0	99.48538	0	498	-	99.48538	_	-	0s
0	0	99.49044	0	500	-	99.49044	_	_	0s
0	0	99.49082	0	503	-	99.49082	_	-	0s
0	0	100.38463	0	506	-	100.38463	-	-	0s
0	0	100.69590	0	511	-	100.69590	-	-	0s
0	0	101.33193	0	520	-	101.33193	-	-	0s
0	0	102.31607	0	529	-	102.31607	-	-	0s
0	0	102.60194	0	538	-	102.60194	-	-	0s
0	0	103.02475	0	529	-	103.02475	-	-	0s
0	0	103.51409	0	536	-	103.51409	-	-	1s
0	0	103.65124	0	543	-	103.65124	-	-	1s
0	0	103.98002	0	541	-	103.98002	-	-	1s
0	0	104.68522	0	551	-	104.68522	-	-	1s
0	0	105.01278	0	553	-	105.01278	-	-	1s
0	0	105.20101	0	565	-	105.20101	-	-	1s
0	0	105.20101	0	565	-	105.20101	-	-	1s
0	2	105.21072	0	565	-	105.21072	-	-	1s

#### Cutting planes:

Cover: 2

Implied bound: 1

MIR: 110 Zero half: 67

RLT: 2

Explored 197 nodes (32326 simplex iterations) in 3.01 seconds Thread count was 4 (of 4 available processors)

Solution count 0

Time limit reached

Best objective -, best bound 1.06000000000e+02, gap -

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Solving with random seed #2 ...

Optimize a model with 442 rows, 970 columns and 2734 nonzeros

Model fingerprint: 0xd83b2255

Variable types: 0 continuous, 970 integer (529 binary)

Coefficient statistics:

Matrix range [1e+00, 2e+00]
Objective range [1e+00, 1e+00]
Bounds range [1e+00, 1e+00]
RHS range [1e+00, 1e+00]
Presolve removed 1 rows and 96 columns

Presolve time: 0.01s

Presolved: 441 rows, 874 columns, 2533 nonzeros

Variable types: 0 continuous, 874 integer (518 binary)

Root relaxation: objective 9.585736e+01, 1219 iterations, 0.06 seconds

Nodes		Current	Nod	e	Objec	ctive Bounds	1	Wor	K
Expl Une	xpl	Obj Depth	ı In	tInf	_		Gap	It/Node	Time
0	0	95.85736	0	440	_	95.85736	_	_	0s
0	0	96.93040	0	461	-	96.93040	_	-	0s
0	0	97.04389	0	467	-	97.04389	_	_	0s
0	0	97.05700	0	466	-	97.05700	_	_	0s
0	0	97.05720	0	467	-	97.05720	_	_	0s
0	0	98.36331	0	475	_	98.36331	_	-	0s
0	0	98.42652	0	478	-	98.42652	_	_	0s
0	0	98.44018	0	481	-	98.44018	_	_	0s
0	0	98.44254	0	482	_	98.44254	_	-	0s
0	0	98.44382	0	483	_	98.44382	_	-	0s
0	0	98.44385	0	484	_	98.44385	_	-	0s
0	0	99.19140	0	489	_	99.19140	_	-	0s
0	0	99.25976	0	489	_	99.25976	_	-	0s
0	0	99.29452	0	492	_	99.29452	_	-	0s
0	0	99.29710	0	494	_	99.29710	_	-	0s
0	0	99.29844	0	497	_	99.29844	_	-	0s
0	0	99.61365	0	499	-	99.61365	_	_	0s
0	0	99.74041	0	500	-	99.74041	_	_	0s
0	0	99.75620	0	506	-	99.75620	_	_	0s
0	0	99.75901	0	506	-	99.75901	_	_	0s
0	0	100.37690	0	502	-	100.37690	_	_	0s
0	0	100.51491	0	509	-	100.51491	_	_	0s
0	0	100.52406	0	510	-	100.52406	_	_	0s
0	0	100.52712	0	513	-	100.52712	_	-	0s
0	0	100.83827	0	515	-	100.83827	_	_	0s
0	0	100.83909	0	518	-	100.83909	_	_	0s
0	0	101.90654	0	509	-	101.90654	_	-	0s
0	0	102.66904	0	519	-	102.66904	_	-	0s
0	0	103.01378	0	527	-	103.01378	_	-	0s
0	0	103.65683	0	532	-	103.65683	_	-	1s
0	0	104.33427	0	536	-	104.33427	_	-	1s
0	0	105.34836	0	534	-	105.34836	_	_	1s
0	0	105.93595	0	558	_	105.93595	_	_	1s
0	0	106.55636	0	561	-	106.55636	_	-	1s
0	0	106.85965	0	567	-	106.85965	_	-	1s
0	0	107.17770	0	566	-	107.17770	_	-	1s
0	0	107.51395	0	567	-	107.51395	_	-	1s
0	0	107.94968	0	580	-	107.94968	_	-	1s
0	0	108.36590	0	578	_	108.36590	_	-	1s

0	0	108.88869	0	585	-	108.88869	-	-	-	1s
0	0	109.36345	0	590	-	109.36345	_	-	•	1s
0	0	109.81196	0	587	-	109.81196	_	-	•	2s
0	0	110.10839	0	593	_	110.10839	_	_		2s
0	0	110.10839	0	592	_	110.10839	_	_		2s
0	2	110.11907	0	592	_	110.11907	_	_		2s

#### Cutting planes:

Cover: 2

Implied bound: 2

MIR: 93

Zero half: 101

RLT: 1

Explored 76 nodes (18148 simplex iterations) in 3.01 seconds Thread count was 4 (of 4 available processors)

Solution count 0

Time limit reached

Best objective -, best bound 1.11000000000e+02, gap -

\_\_\_\_\_\_

Solving with random seed #3 ...

Optimize a model with 442 rows, 970 columns and 2734 nonzeros

Model fingerprint: 0xd83b2255

Variable types: 0 continuous, 970 integer (529 binary)

Coefficient statistics:

Matrix range [1e+00, 2e+00]
Objective range [1e+00, 1e+00]
Bounds range [1e+00, 1e+00]
RHS range [1e+00, 1e+00]
Presolve removed 1 rows and 96 columns

Presolve time: 0.01s

Presolved: 441 rows, 874 columns, 2533 nonzeros

Variable types: 0 continuous, 874 integer (518 binary)

Root relaxation: objective 9.585736e+01, 1228 iterations, 0.05 seconds

Nodes		Cu	rrent N	lode	)	Object	cive Bounds	- 1	Worl	Σ.
Expl Unexp	1	Obj	Depth	Int	Inf	Incumbent	BestBd	Gap	It/Node	Time
0	0	95.85	736	0	440	_	95.85736	_	-	0s
0	0	96.97	025	0	463	_	96.97025	_	_	0s
0	0	97.08	247	0	464	-	97.08247	-	-	0s
0	0	97.09	562	0	467	_	97.09562	_	-	0s
0	0	97.09	796	0	467	_	97.09796	_	_	0s

0	0	97.09853	0	466	-	97.09853	_	-	0s
0	0	97.09853	0	467	_	97.09853	_	-	0s
0	0	98.07699	0	472	_	98.07699	_	-	0s
0	0	98.19686	0	480	_	98.19686	_	-	0s
0	0	98.20215	0	482	_	98.20215	_	-	0s
0	0	98.20253	0	482	_	98.20253	_	-	0s
0	0	98.66571	0	489	_	98.66571	_	-	0s
0	0	98.72609	0	491	-	98.72609	_	-	0s
0	0	98.73460	0	495	_	98.73460	_	_	0s
0	0	98.74101	0	499	-	98.74101	_	-	0s
0	0	98.74154	0	501	-	98.74154	_	-	0s
0	0	100.07621	0	497	-	100.07621	_	-	0s
0	0	100.09689	0	498	-	100.09689	_	-	0s
0	0	100.09785	0	497	_	100.09785	_	_	0s
0	0	100.88550	0	512	-	100.88550	-	-	0s
0	0	101.78031	0	512	-	101.78031	-	-	0s
0	0	102.31611	0	520	-	102.31611	-	-	0s
0	0	103.06033	0	531	-	103.06033	-	-	0s
0	0	104.07059	0	523	-	104.07059	-	-	0s
0	0	104.67974	0	549	-	104.67974	-	-	0s
0	0	105.52274	0	550	-	105.52274	-	-	1s
0	0	106.03969	0	562	-	106.03969	-	-	1s
0	0	106.38329	0	572	-	106.38329	-	-	1s
0	0	106.58346	0	578	-	106.58346	-	-	1s
0	0	106.58346	0	577	-	106.58346	-	-	1s
0	2	106.59237	0	575	_	106.59237	_	_	1s

## Cutting planes:

Cover: 2 MIR: 105 StrongCG: 2 Zero half: 79

RLT: 1

Explored 221 nodes (33883 simplex iterations) in 3.01 seconds Thread count was 4 (of 4 available processors)

Solution count 0

Time limit reached

Best objective -, best bound 1.07000000000e+02, gap -

Begin tuning (baseline 3 no\_solution)...

Testing candidate parameter set 2...

```
Heuristics 0.5
```

Solving with random seed #1  $\dots$  MIP gap 56.7% Solving with random seed #2  $\dots$  MIP gap 54.7% Solving with random seed #3  $\dots$  MIP gap 56.3%

Improvement found:

baseline: 3 no\_solution
improved: mean MIP gap 55.9%

Total elapsed tuning time 18s (12s remaining)

------

Testing candidate parameter set 3...

MIPFocus 2

Solving with random seed #1 ... MIP gap -

Progress so far:

 ${\tt baseline: 3 no\_solution}$ 

best: mean MIP gap 55.9%

Total elapsed tuning time 21s (9s remaining)

\_\_\_\_\_\_

Testing candidate parameter set 4...

Heuristics 0.5 ZeroHalfCuts 1

Solving with random seed #1  $\dots$  MIP gap 56.7% Solving with random seed #2  $\dots$  MIP gap 54.7% Solving with random seed #3  $\dots$ 

Tune time limit reached.

\_\_\_\_\_\_

Tested 3 parameter sets in 30.00s

Baseline parameter set: 3 no\_solution

Default parameters

# Name 0 1 2 Avg Std Dev 0 Model - - - - -

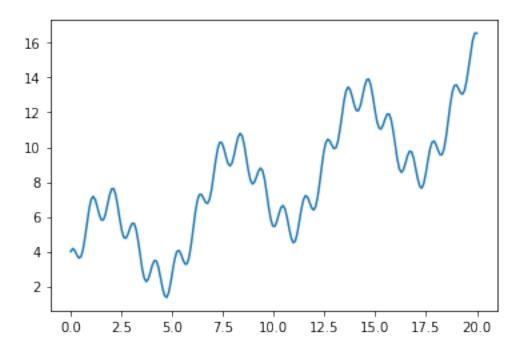
```
Improved parameter set 1 (mean MIP gap 55.9%):
            Heuristics 0.5
     # Name
                                                  Avg Std Dev
     0 Model
                                                          0.88
                     56.7%
                              54.7%
                                       56.3%
                                                55.9%
[5]: print('Time before tuning: {}s'.format(def_time))
    print('Time after tuning: {}s'.format(game_of_fivers(21, {'CutPasses': 10, __
     →'PreDual':1})))
    Time before tuning: 26.91558289527893s
    Changed value of parameter CutPasses to 10
       Prev: -1 Min: -1 Max: 200000000 Default: -1
    Changed value of parameter PreDual to 1
       Prev: -1 Min: -1 Max: 2 Default: -1
    Time after tuning: 8.29596495628357s
```

## 2 Nonlinear constrains

```
[6]: n = 200
t = np.linspace(0, 20, n)
y = 3*np.sin(t)+np.cos(6*t)+0.5*t+3

plt.plot(t, y)
```

[6]: [<matplotlib.lines.Line2D at 0x7f2d77bb2f90>]



```
[7]: m = g.Model()
     u = m.addVar(vtype=g.GRB.CONTINUOUS)
     v = m.addVar(vtype=g.GRB.CONTINUOUS)
     m.addGenConstrPWL(u, v, t, y)
     m.setObjective(v)
     m.optimize()
     plt.plot(t, y)
     {\tt plt.plot(u.x,\ v.x,\ marker='o',\ markersize=8,\ color="red")}
    Gurobi Optimizer version 9.1.0 build v9.1.0rc0 (linux64)
    Thread count: 2 physical cores, 4 logical processors, using up to 4 threads
```

Optimize a model with 0 rows, 2 columns and 0 nonzeros

Model fingerprint: 0x8d885f17 Model has 1 general constraint

Variable types: 2 continuous, 0 integer (0 binary)

Coefficient statistics:

Matrix range [0e+00, 0e+00] Objective range [1e+00, 1e+00] Bounds range [0e+00, 0e+00] [0e+00, 0e+00] RHS range Presolve added 1 rows and 197 columns Presolve time: 0.00s

Presolved: 1 rows, 199 columns, 199 nonzeros

Presolved model has 1 SOS constraint(s)

Variable types: 199 continuous, 0 integer (0 binary)

Root relaxation: objective 1.364267e+00, 0 iterations, 0.00 seconds

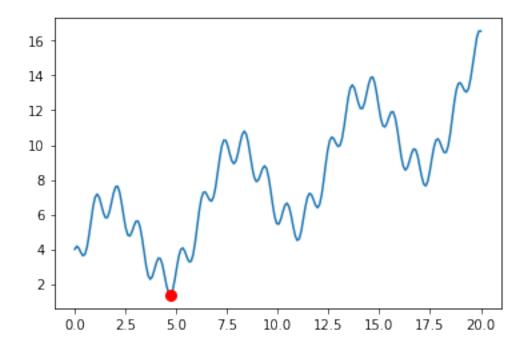
	Node	es		Cu	rrent 1	Vode		Objecti	ve Bounds		Worl	K
E	xpl Ur	nexpl	I	Obj	Depth	IntInf	I	Incumbent	BestBd	Gap	It/Node	Time
*	0	0				0	1	3642670	1 36427	0.00%	_	Ωs

Explored 0 nodes (0 simplex iterations) in 0.01 seconds Thread count was 4 (of 4 available processors)

Solution count 1: 1.36427

Optimal solution found (tolerance 1.00e-04)
Best objective 1.364266996602e+00, best bound 1.364266996602e+00, gap 0.0000%

[7]: [<matplotlib.lines.Line2D at 0x7f2d7609a610>]



[]:

# 3 Solution pool

/home/benedond/anaconda3/lib/python3.7/site-packages/networkx/drawing/nx\_pylab.py:579: MatplotlibDeprecationWarning: The iterable function was deprecated in Matplotlib 3.1 and will be removed in 3.3. Use np.iterable instead.

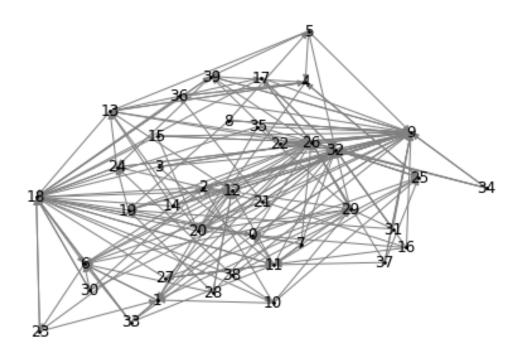
if not cb.iterable(width):

/home/benedond/anaconda3/lib/python3.7/site-

packages/networkx/drawing/nx\_pylab.py:676: MatplotlibDeprecationWarning:

The iterable function was deprecated in Matplotlib 3.1 and will be removed in 3.3. Use np.iterable instead.

if cb.iterable(node\_size): # many node sizes



```
[9]: s = 28
t = 4
E = dg.edges()
E = dict.fromkeys(E)
```

```
V = dg.nodes()
np.random.seed(69)
w = np.random.randint(0, 50, len(E))
m = g.Model()
x = m.addVars(E.keys(), vtype=g.GRB.BINARY, ub=1, obj=w)
m.addConstr(g.quicksum([x[s, j] for k, j in E if k == s]) == 1)
m.addConstr(g.quicksum([x[i, t] for i, k in E if k == t]) == 1)
for i in V:
  if i not in [s, t]:
    m.addConstr(g.quicksum([x[i, j] for k, j in E if k == i]) == g.
 \rightarrowquicksum([x[j, i] for j, k in E if k == i]))
m.setParam(g.GRB.Param.PoolSolutions, 3)
m.setParam(g.GRB.Param.PoolSearchMode, 2) # k-best solutions
m.optimize()
sols = [0]*m.solcount
colorlist = ['r', 'g', 'b']
print('Found {} solutions.'.format(m.solcount))
for sol_idx in range(m.solcount):
  print('Sol no. {}'.format(sol_idx+1))
  sols[sol_idx] = []
  m.setParam(g.GRB.Param.SolutionNumber, sol_idx)
  for i, j in E:
    if x[i, j].xn > 0.5:
      print(i, j)
      sols[sol_idx] += [(i, j)]
nx.draw(dg, pos, node_color='k', node_size=3, edge_color='grey')
for k in range(m.solcount):
  nx.draw_networkx_edges(dg, pos, edgelist=sols[k], edge_color=colorlist[k],_
 \rightarrowwidth=3)
Changed value of parameter PoolSolutions to 3
  Prev: 10 Min: 1 Max: 200000000 Default: 10
Changed value of parameter PoolSearchMode to 2
  Prev: 0 Min: 0 Max: 2 Default: 0
Gurobi Optimizer version 9.1.0 build v9.1.0rc0 (linux64)
Thread count: 2 physical cores, 4 logical processors, using up to 4 threads
Optimize a model with 40 rows, 174 columns and 331 nonzeros
Model fingerprint: 0x0bb97d12
Variable types: 0 continuous, 174 integer (174 binary)
```

#### Coefficient statistics:

Matrix range [1e+00, 1e+00] Objective range [1e+00, 5e+01] Bounds range [1e+00, 1e+00] RHS range [1e+00, 1e+00]

Found heuristic solution: objective 314.0000000

Presolve removed 17 rows and 73 columns

Presolve time: 0.00s

Presolved: 23 rows, 101 columns, 185 nonzeros

Variable types: 0 continuous, 101 integer (101 binary)

Found heuristic solution: objective 43.0000000

Root relaxation: objective 4.100000e+01, 13 iterations, 0.00 seconds

	Nodes	- 1	Cu	rrent l	Vode		Object	ive Boun	ds		Work	Σ
Ex	pl Unex	cpl	Obj	Depth	${\tt IntInf}$	-	${\tt Incumbent}$	BestB	d	Gap	It/Node	Time
	•	^			0			44 0000	•	0 00%		•
*	0	0			0	4.	1.0000000	41.0000	0 (	0.00%	_	0s

Optimal solution found at node 0 - now completing solution pool...

Nodes	Cu	rrent N	lode	Pool	s	Worl	X.		
	- 1				Worst		-		
Expl Unexpl	L	Obj	Depth	${\tt IntInf}$	Incumbent	${\tt BestBd}$	Gap	It/Node	Time
0 (	)		-	0	43.00000	41.00000	4.65%	-	0s
0 (	)		-	0	43.00000	41.00000	4.65%	_	0s
0 2	2		-	0	43.00000	41.00000	4.65%	_	0s

Explored 76 nodes (80 simplex iterations) in 0.05 seconds Thread count was 4 (of 4 available processors)

Solution count 3: 41 42 42

No other solutions better than 42

Optimal solution found (tolerance 1.00e-04)

Best objective 4.100000000000e+01, best bound 4.10000000000e+01, gap 0.0000% Found 3 solutions.

Sol no. 1

19 26

25 4

26 32

28 19

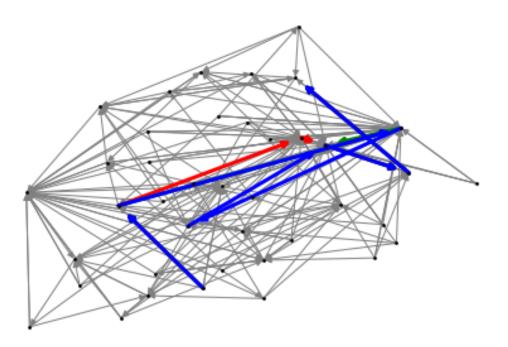
32 25

Sol no. 2

9 32

19 9

```
28 19
32 25
Sol no. 3
9 20
19 9
20 32
25 4
28 19
32 25
```



# 4 Bratley revisited: CP model

```
[10]: # Visualization
import matplotlib.pyplot as plt

def plot_solution(s, p):
    """
    s: solution vector
    p: processing times
    """
    fig = plt.figure(figsize=(10,2))
```

```
ax = plt.gca()
          ax.set_xlabel('time')
          ax.grid(True)
          ax.set_yticks([2.5])
          ax.set_yticklabels(["oven"])
          eps = 0.25 # just to show spaces between the dishes
          ax.broken_barh([(s[i], p[i]-eps) for i in range(len(s))], (0, 5),
                         facecolors=('tab:orange', 'tab:green', 'tab:red', 'tab:
       →blue', 'tab:gray'))
[27]: # Load data
      path = "./bratley_data/instances/public_3.txt"
      with open(path, "r") as f_in:
          lines = f_in.readlines()
          n = int(lines[0].strip())
          r,d,p = [],[],[]
          for i in range(n):
              (pi, ri, di) = list(map(int, lines[1+i].split()))
              r.append(ri)
              p.append(pi)
              d.append(di)
      print("r", r, "d", d, "p", p, sep="\n")
     [0, 4, 4, 4, 10, 9, 2, 1, 3, 1, 0, 4]
     [10, 39, 39, 39, 30, 30, 12, 49, 7, 20, 39, 39]
     [3, 2, 2, 2, 1, 4, 5, 5, 3, 5, 3, 2]
[11]: # Generate data
      from numpy import random as rnd
      import numpy as np
      rnd.seed(15)
      n = 200
      p = [rnd.randint(1,100) for i in range(n)]
      r = [0 \text{ for i in } range(n)]
```

r[i] = int(round(r[i-1] + rnd.exponential(0.5\* sum(p)/len(p))))

for i in range(1,n):

```
d = [int(round(r[i] + p[i] + rnd.exponential(100*sum(p)/len(p))))) for i in
 \rightarrowrange(n)]
print("r", r, "d", d, "p", p, sep="\n")
[0, 17, 47, 59, 64, 67, 68, 88, 114, 191, 191, 279, 287, 288, 294, 318, 350,
460, 476, 509, 573, 584, 589, 620, 641, 792, 793, 872, 929, 949, 972, 997, 1006,
1031, 1069, 1143, 1145, 1145, 1151, 1167, 1170, 1189, 1191, 1193, 1227, 1244,
1327, 1363, 1371, 1446, 1549, 1557, 1576, 1586, 1646, 1665, 1677, 1711, 1711,
1722, 1742, 1759, 1761, 1778, 1803, 1832, 1840, 1844, 1899, 1938, 1947, 1967,
2018, 2109, 2123, 2288, 2300, 2306, 2312, 2333, 2337, 2370, 2461, 2530, 2535,
2560, 2580, 2613, 2645, 2668, 2696, 2710, 2761, 2801, 2825, 2864, 2871, 2875,
2880, 2884, 2890, 2891, 2955, 2978, 2990, 3072, 3113, 3138, 3169, 3216, 3269,
3277, 3312, 3343, 3431, 3445, 3483, 3483, 3517, 3531, 3581, 3623, 3649, 3656,
3662, 3671, 3681, 3683, 3698, 3708, 3759, 3768, 3770, 3830, 3838, 3913, 3921,
3978, 3982, 4015, 4018, 4025, 4046, 4073, 4099, 4149, 4163, 4213, 4220, 4244,
4244, 4250, 4254, 4267, 4289, 4291, 4315, 4319, 4368, 4370, 4373, 4379, 4405,
4420, 4444, 4475, 4509, 4514, 4537, 4562, 4568, 4594, 4611, 4642, 4643, 4662,
4683, 4726, 4779, 4789, 4876, 4877, 4878, 4903, 4910, 4948, 4956, 4990, 4998,
5003, 5012, 5086, 5102, 5109, 5141, 5166, 5185, 5212, 5246, 5274]
[1653, 1274, 683, 4237, 7291, 7022, 499, 6566, 956, 12361, 2273, 4740, 7231,
5693, 17535, 5851, 3754, 5734, 5748, 1871, 3036, 8860, 6743, 10257, 4376, 6886,
2355, 920, 3985, 7539, 3220, 1430, 9802, 8269, 17657, 9754, 2953, 18576, 5789,
6649, 13608, 20906, 1256, 5419, 9168, 2149, 9160, 3764, 2085, 7151, 8404, 4097,
8132, 9647, 5384, 3132, 7837, 2736, 3852, 20142, 3259, 5922, 3564, 8043, 6688,
6967, 6871, 4696, 14848, 2245, 6409, 7160, 6515, 4009, 5834, 2989, 4071, 3748,
3628, 2634, 8579, 7688, 14355, 10790, 8446, 9242, 17217, 2769, 5092, 14926,
12880, 9399, 4836, 5952, 10726, 5712, 3161, 24580, 6862, 6771, 5007, 5309,
25457, 3713, 14273, 11156, 15824, 20413, 10546, 6995, 5869, 6110, 5992, 5980,
10857, 14021, 18216, 5134, 9709, 3772, 6917, 4419, 5122, 13725, 6241, 10165,
9315, 8992, 6508, 4788, 21040, 8278, 17517, 14468, 18058, 14332, 5059, 10607,
24110, 15952, 29200, 4824, 7139, 9630, 6140, 12606, 4968, 7599, 10169, 5389,
6909, 5582, 12165, 5693, 12741, 10125, 4938, 9310, 12376, 5695, 10949, 8007,
9125, 6158, 9109, 11212, 4913, 11588, 4675, 8023, 6372, 13904, 6884, 4960, 6218,
6002, 11671, 5589, 5503, 9818, 5247, 19407, 5599, 5839, 5414, 6941, 11527, 6971,
14055, 5395, 6641, 5544, 9417, 6893, 9865, 5285, 9039, 6887, 14091, 5398]
[73, 13, 6, 1, 29, 28, 72, 76, 86, 48, 94, 18, 32, 24, 33, 63, 11, 16, 69, 40,
38, 20, 45, 78, 61, 30, 80, 16, 57, 50, 2, 32, 97, 86, 27, 35, 76, 51, 66, 54,
71, 42, 35, 41, 23, 64, 80, 57, 29, 5, 8, 67, 43, 97, 8, 25, 61, 46, 84, 50, 54,
30, 77, 89, 77, 34, 3, 89, 43, 82, 52, 63, 24, 94, 99, 88, 19, 91, 91, 17, 78,
91, 33, 71, 5, 29, 85, 36, 29, 70, 55, 65, 74, 85, 57, 47, 39, 36, 15, 83, 93,
91, 53, 83, 42, 36, 68, 20, 80, 81, 33, 55, 34, 35, 34, 29, 88, 61, 9, 11, 76,
44, 78, 62, 99, 80, 86, 77, 64, 68, 45, 61, 28, 42, 85, 85, 63, 92, 1, 69, 14,
6, 60, 4, 53, 49, 12, 66, 75, 69, 38, 28, 85, 28, 73, 20, 96, 58, 32, 18, 18,
```

71, 72, 49, 26, 32, 40, 89, 36, 43, 69, 26, 58, 11, 48, 35, 44, 35, 69, 67, 81,

78, 18, 24, 61, 54, 41, 11, 55, 71, 56, 96, 90, 5, 86, 97, 47, 12, 10, 27]

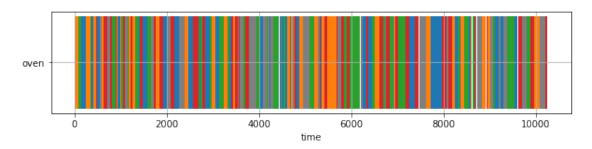
#### 4.1 CP model

```
[12]: from docplex.cp.model import CpoModel
      from docplex.cp.config import context
      import sys
      # Create model
      m = CpoModel()
      # - add variables
      tasks = [m.interval_var(name="task{:d}".format(i), optional=False, size=p[i])__
      \rightarrow for i in range(n)]
      seq = m.sequence_var(tasks, name='seq')
      # - set objective
      m.add(m.minimize(m.max([m.end of(tasks[i]) for i in range(n)])))  # minimize_
      \hookrightarrow C_max
      # - add constraints
      for i in range(n):
          m.add(m.start_of(tasks[i]) >= r[i]) # release time
          m.add(m.end_of(tasks[i]) <= d[i]) # deadline</pre>
      m.add(m.no_overlap(seq)) # one task executed at one time
      # Solve the model
      msol = m.solve(TimeLimit=10, LogVerbosity="Normal", LogPeriod=1, Workers=1)
      # Print the solution
      print()
      if msol.is_solution():
          starts = [msol.get_value(tasks[i])[0] for i in range(n)]
          print(*starts, sep="\n")
      else:
          print("No solution found.")
      print("Done")
```

----

```
5467
8354
5284
Done
```

# [13]: plot\_solution(starts, p)



#### 4.2 ILP model

```
[14]: import gurobipy as g
      m = g.Model()
      # - add variables
      s = m.addVars(n, vtype=g.GRB.CONTINUOUS, lb=0)
      x = \{\}
      for i in range(n):
          for j in range(i + 1, n):
              x[i, j] = m.addVar(vtype=g.GRB.BINARY)
      Cmax = m.addVar(vtype=g.GRB.CONTINUOUS, obj=1)
      # - add constraints
      for i in range(n):
          m.addConstr(s[i] + p[i] <= Cmax)</pre>
          m.addConstr(s[i] >= r[i])
          m.addConstr(s[i] + p[i] <= d[i])</pre>
      M = max(d)
      for i in range(n):
          for j in range(i + 1, n):
              m.addConstr(s[i] + p[i] \le s[j] + M*(1-x[i, j]))
              m.addConstr(s[j] + p[j] \le s[i] + M*x[i, j])
      # call the solver --
      m.optimize()
```

```
print()
if m.SolCount > 0:
    starts = [s[i].X for i in range(n)]
else:
    print("No solution was found.")
print("Done")
```

Gurobi Optimizer version 9.1.0 build v9.1.0rc0 (linux64)

Thread count: 2 physical cores, 4 logical processors, using up to 4 threads

Optimize a model with 40400 rows, 20101 columns and 120200 nonzeros

Model fingerprint: 0x74609c7a

Variable types: 201 continuous, 19900 integer (19900 binary)

Coefficient statistics:

Matrix range [1e+00, 3e+04] Objective range [1e+00, 1e+00] Bounds range [1e+00, 1e+00] RHS range [1e+00, 3e+04]

Presolve removed 9123 rows and 4458 columns

Presolve time: 0.19s

Presolved: 31277 rows, 15643 columns, 93438 nonzeros

Variable types: 201 continuous, 15442 integer (15442 binary)

Root simplex log...

Iteration	Objective	Primal Inf.	Dual Inf.	Time
20011	5.3531148e+03	1.228794e+03	0.000000e+00	5s
25781	5.3530000e+03	0.000000e+00	0.000000e+00	8s

Root relaxation: objective 5.353000e+03, 25781 iterations, 7.27 seconds Total elapsed time = 10.40s

Nodes			Cu	rrent 1	lode	Э	Objec	tive	Bounds		-	Wo	rk
Expl Unex	pl	Ob	j	Depth	Int	Inf	Incumbent	. ]	BestBd	Gap		It/Nod	e Time
0	0	5439.	000	000	0	516	_	5439	.00000		_	_	12s
0	0	5439.	000	000	0	756	_	5439	.00000		_	_	14s
0	0	5439.	000	000	0	683	-	5439	.00000		-	-	15s
0	0	5439.	000	000	0	365	-	5439	.00000		_	_	16s
0	0	5439.	000	000	0	365	-	5439	.00000		_	_	17s
0	0	5439.	000	000	0	347	_	5439	.00000	-	_	_	19s
0	0	5439.	000	000	0	399	_	5439	.00000	-	_	_	19s
0	0	5439.	000	000	0	323	_	5439	.00000		_	_	22s
0	0	5439.	000	000	0	346	_	5439	.00000		_	_	22s
0	0	5439.	000	000	0	327	_	5439	.00000		_	_	24s
0	0	5439.	000	000	0	332	_	5439	.00000		_	_	24s

0	0	5439.00000	0	329	- 5439.00000	_	_	26s
0	0	5439.00000	0	352	- 5439.00000	_	_	26s
0	0	5439.00000	0	307	- 5439.00000	_	_	29s
0	0	5439.00000	0	307	- 5439.00000	_	_	29s
0	2	5439.00000	0	307	- 5439.00000	_	_	33s
203	161	infeasible	65		- 5442.86736	_	58.8	35s
732	538	5480.00000	14	456	- 5445.29630	_	62.3	42s
735	540	5568.00000	45	476	- 5445.29630	_	62.1	45s
740	544	5475.00000	6	434	- 5445.29630	_	61.7	52s
741	544	5529.22963	21	434	- 5445.29630	_	61.6	55s
744	549	5445.29630	17	464	- 5445.29630	_	43.3	60s
777	577	5627.00000	25	371	- 5449.00000	_	47.3	65s
877	644	5836.00000	54	351	- 5449.00000	_	51.1	70s
999	731	5938.00000	80	444	- 5449.00000	_	55.6	75s
1150	877	5897.00000	110	439	- 5449.00000	_	54.2	80s

## Cutting planes:

Cover: 7

Implied bound: 5

MIR: 5

Relax-and-lift: 8

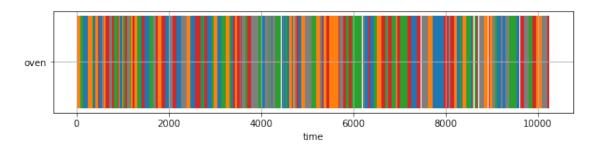
Explored 1233 nodes (218830 simplex iterations) in 80.55 seconds Thread count was 4 (of 4 available processors)

Solution count 0

Solve interrupted
Best objective -, best bound 5.44900000000e+03, gap -

No solution was found. Done

## [15]: plot\_solution(starts, p)



# 5 TSP

```
[22]: import math
      from collections import namedtuple
      import gurobipy as g
      Point = namedtuple("Point", ['x', 'y'])
      def length(point1, point2):
          return int(round(math.sqrt((point1.x - point2.x)**2 + (point1.y - point2.
       \rightarrowy)**2))) # CP works with int only
      class TSP:
          def __init__(self):
              D = None # distance matrix
              points = None # vertices
          def load_instance(self, path):
              input data file = open(path, 'r')
              input_data = ''.join(input_data_file.readlines())
              # parse the input
              lines = input_data.split('\n')
              nodeCount = int(lines[0])
              points = []
              for i in range(1, nodeCount+1):
                  parts = lines[i].split()
                  points.append(Point(float(parts[0]), float(parts[1])))
              # distance matrix
              D = [[0 for _ in range(nodeCount+1)] for _ in range(nodeCount+1)] #__
       \rightarrow Add dummy vertex last
              for i in range(nodeCount):
                  for j in range(nodeCount):
                      D[i][j] = length(points[i], points[j])
              # the last vertex is the same as the first one
              for i in range(nodeCount):
                  D[i][-1] = D[i][0]
                  D[-1][i] = D[0][i]
              self.D = D
              self.points = points
              return self
```

```
[23]: instances = [
              {"inst": TSP().load_instance("./tsp_data/tsp_5_1"),
               "init": [0, 1, 2, 4, 3]},
              {"inst": TSP().load_instance("./tsp_data/tsp_51_1"),
               \rightarrow11, 42, 37, 20, 25, 1, 31, 22, 48, 32, 17, 49, 39, 50, 38, 15, 44, 14, 16, \square
       \rightarrow29, 43, 21, 30, 12, 23, 34, 24, 41, 27, 36, 6, 26, 47, 33]},
              {"inst": TSP().load_instance("./tsp_data/tsp_70_1"),
               "init": [0, 35, 50, 11, 57, 2, 56, 27, 21, 49, 58, 53, 41, 36, 38, 52, __
       →6, 5, 7, 51, 55, 68, 46, 67, 24, 16, 44, 39, 22, 1, 14, 15, 20, 29, 28, 45, ⊔
       →12, 31, 18, 26, 3, 59, 9, 25, 4, 10, 61, 43, 32, 8, 64, 54, 48, 62, 13, 19, U
       \rightarrow60, 42, 37, 66, 40, 17, 30, 23, 69, 33, 65, 34, 47, 63]},
              {"inst": TSP().load instance("./tsp data/tsp 100 1"),
               "init": [0, 6, 69, 61, 76, 35, 84, 11, 9, 26, 72, 47, 40, 94, 81, 60, 
       \rightarrow64, 66, 8, 23, 70, 59, 33, 67, 43, 37, 65, 71, 19, 15, 75, 14, 53, 46, 5, \Box
       \rightarrow29, 80, 38, 91, 57, 41, 50, 12, 55, 98, 39, 24, 68, 2, 28, 73, 87, 48, 85, \square
       \rightarrow21, 96, 42, 77, 16, 7, 10, 74, 30, 18, 17, 34, 22, 99, 93, 51, 3, 89, 13, \Box
       \rightarrow31, 44, 62, 25, 82, 86, 54, 1, 27, 45, 88, 79, 97, 49, 90, 20, 63, 52, 92, \Box
       \rightarrow95, 78, 83, 32, 4, 56, 58, 36]},
              {"inst": TSP().load_instance("./tsp_data/tsp_200_1"),
               "init": [0, 103, 62, 192, 5, 48, 89, 148, 117, 9, 128, 83, 136, 23, 
       →37, 108, 177, 181, 98, 106, 35, 160, 125, 131, 123, 58, 73, 20, 145, 71, ⊔
       →111, 46, 97, 22, 114, 112, 178, 59, 61, 163, 119, 154, 141, 34, 85, 26, 11, ⊔
       →19, 146, 130, 166, 76, 164, 179, 60, 24, 80, 101, 134, 68, 167, 129, 188, ⊔
       →158, 102, 172, 88, 168, 41, 30, 79, 55, 199, 132, 144, 96, 180, 196, 3, 64, ⊔
       →65, 195, 25, 186, 151, 110, 183, 147, 69, 21, 15, 87, 143, 162, 93, 150, U
       →115, 17, 78, 52, 165, 18, 191, 198, 118, 109, 74, 135, 156, 173, 7, 113, 91, 11
       →159, 57, 176, 50, 86, 56, 6, 8, 105, 153, 174, 82, 54, 107, 121, 33, 28, 45, ⊔
       →116, 124, 133, 189, 42, 2, 13, 197, 157, 40, 70, 99, 187, 47, 127, 138, 137, ⊔
       →170, 29, 171, 182, 161, 84, 67, 72, 122, 49, 43, 169, 175, 190, 193, 194, ⊔
       \rightarrow149, 38, 185, 95, 155, 51, 77, 104, 4, 142, 36, 32, 75, 12, 94, 81, 1, 63,
       \rightarrow39, 120, 53, 140, 66, 27, 92, 126, 90, 44, 184, 31, 100, 152, 14, 16, 10, \Box
       →139]}
      ]
```

```
[24]: inst_id = 4
inst = instances[inst_id]["inst"]
init_order = instances[inst_id]["init"]
```

#### [25]: INITIALIZE = False

#### 5.1 ILP

```
[26]: nodeCount = len(inst.points)
      points = inst.points
      # Create model
      m = g.Model("tsp")
      # - add variables
      x = m.addVars(nodeCount, nodeCount, vtype=g.GRB.BINARY, name="x")
      u = m.addVars(nodeCount, vtype=g.GRB.INTEGER, lb=0, name="u")
      # - set objective
      obj = g.quicksum(g.quicksum(inst.D[i][j]*x[i,j] for j in range(nodeCount)) for⊔
      →i in range(nodeCount))
      m.setObjective(obj, g.GRB.MINIMIZE)
      # - add constraints
      m.addConstrs((1 == g.quicksum(x[i,j] for j in range(nodeCount)) for i in_
      →range(nodeCount)))
      m.addConstrs((1 == g.quicksum(x[j,i] for j in range(nodeCount)) for i in_
      →range(nodeCount)))
      for i in range(1, nodeCount):
          for j in range(1, nodeCount):
              m.addConstr(u[i]-u[j]+1 <= nodeCount*(1-x[i,j]))</pre>
      # Initialization
      if INITIALIZE:
          for i, order in enumerate(init_order):
              u[order].start = i
      m.Params.TimeLimit = 10
      m.optimize()
      # Print the solution
      print()
      if m.SolCount > 0:
          obj = m.objVal
          print("Objective {}".format(obj))
          order = [u[i].X for i in range(nodeCount)]
          indices = range(nodeCount)
          s = sorted(zip(order,indices), key=lambda x: x[0])
          print([x[1] for x in s])
      else:
```

```
print("No solution was found.")
print("Done")
Changed value of parameter TimeLimit to 10.0
  Prev: inf Min: 0.0 Max: inf Default: inf
Gurobi Optimizer version 9.1.0 build v9.1.0rc0 (linux64)
Thread count: 2 physical cores, 4 logical processors, using up to 4 threads
Optimize a model with 40001 rows, 40200 columns and 198405 nonzeros
Model fingerprint: Oxceefc9da
Variable types: 0 continuous, 40200 integer (40000 binary)
Coefficient statistics:
 Matrix range
                   [1e+00, 2e+02]
 Objective range [1e+01, 4e+03]
 Bounds range
                   [1e+00, 1e+00]
 RHS range
                   [1e+00, 2e+02]
Presolve removed 199 rows and 200 columns
Presolve time: 0.32s
Presolved: 39802 rows, 40000 columns, 197808 nonzeros
Variable types: 0 continuous, 40000 integer (39801 binary)
Deterministic concurrent LP optimizer: primal and dual simplex
Showing first log only...
Concurrent spin time: 0.01s
Solved with dual simplex
Root relaxation: objective 2.312896e+04, 677 iterations, 0.22 seconds
                  Current Node
                                        Objective Bounds
                                                                    Work
Expl Unexpl | Obj Depth IntInf | Incumbent
                                                 BestBd
                                                          Gap | It/Node Time
     0
           0 23128.9600
                           0 411
                                           - 23128.9600
                                                                        3s
Explored 1 nodes (772 simplex iterations) in 10.30 seconds
Thread count was 4 (of 4 available processors)
Solution count 0
Time limit reached
Best objective -, best bound 2.31290000000e+04, gap -
No solution was found.
```

Done

#### 5.2 CP

```
[27]: from docplex.cp.model import CpoModel
      from docplex.cp.config import context
      import sys
      node_count = len(inst.points)
      points = inst.points
      # Create model
      m = CpoModel()
      # - add variables
      cities = [m.interval_var(name="city{:d}".format(i), optional=False, size=1) for_u
      →i in range(node_count+1)]
      seq = m.sequence_var(cities, name='seq', types=([i for i in range(node_count)]__
      →+ [0]))
      # - set objective
      m.add(m.minimize(m.max([m.end_of(cities[i]) for i in range(len(cities))]) -_u
      →len(cities)))
      # - add constraints
      m.add(m.first(seq, cities[0])) # start from city 0
      m.add(m.last(seq, cities[-1])) # repeat the same city last
      m.add(m.no_overlap(seq, inst.D, True))
      # Solve the model
      msol = m.solve(TimeLimit=10, LogVerbosity="Verbose", LogPeriod=1, Workers=1)
      # Print the solution
      print()
      if msol.is_solution():
          ovals = msol.get_objective_values()
          print("Objective {}".format(ovals[0]))
          starts = [msol.get_value(cities[i])[0] for i in range(len(cities)-1)]
          indices = range(len(cities)-1)
          s = sorted(zip(starts,indices), key=lambda x: x[0])
          print([x[1] for x in s])
      else:
          print("No solution found.")
      print("Done")
```

Objective 31962

[0, 103, 62, 192, 5, 48, 89, 148, 117, 9, 128, 83, 136, 23, 37, 108, 177, 181, 98, 106, 125, 35, 160, 131, 123, 58, 73, 20, 145, 71, 111, 97, 22, 114, 112, 178, 59, 61, 163, 119, 154, 141, 34, 85, 26, 11, 19, 146, 130, 166, 76, 164, 179, 60, 24, 80, 101, 134, 68, 167, 129, 188, 158, 102, 172, 88, 168, 41, 30, 79, 55, 199, 132, 144, 96, 180, 196, 3, 64, 65, 195, 25, 186, 151, 110, 183, 147, 69, 21, 15, 87, 143, 162, 93, 150, 115, 17, 78, 52, 165, 18, 191, 198, 118, 109, 74, 135, 156, 173, 7, 113, 91, 159, 57, 176, 50, 86, 56, 6, 8, 105, 153, 174, 82, 54, 107, 121, 33, 28, 45, 116, 124, 133, 189, 42, 2, 13, 197, 157, 40, 70, 99, 187, 47, 127, 138, 137, 170, 29, 171, 182, 161, 84, 67, 72, 122, 49, 43, 169, 175, 190, 193, 194, 149, 38, 185, 95, 155, 51, 77, 104, 4, 142, 36, 32, 75, 12, 94, 81, 1, 63, 39, 120, 53, 140, 66, 27, 92, 126, 90, 44, 184, 31, 100, 152, 14, 16, 10, 139, 46]
Done

#### 5.3 Comparison

Best objective found by ILP and CP model under 10s timelimit. (without initialization)

instance	ILP	СР
5	3	3
50	496	441
70	994	710
100	-	22247
200	-	31962

[]: