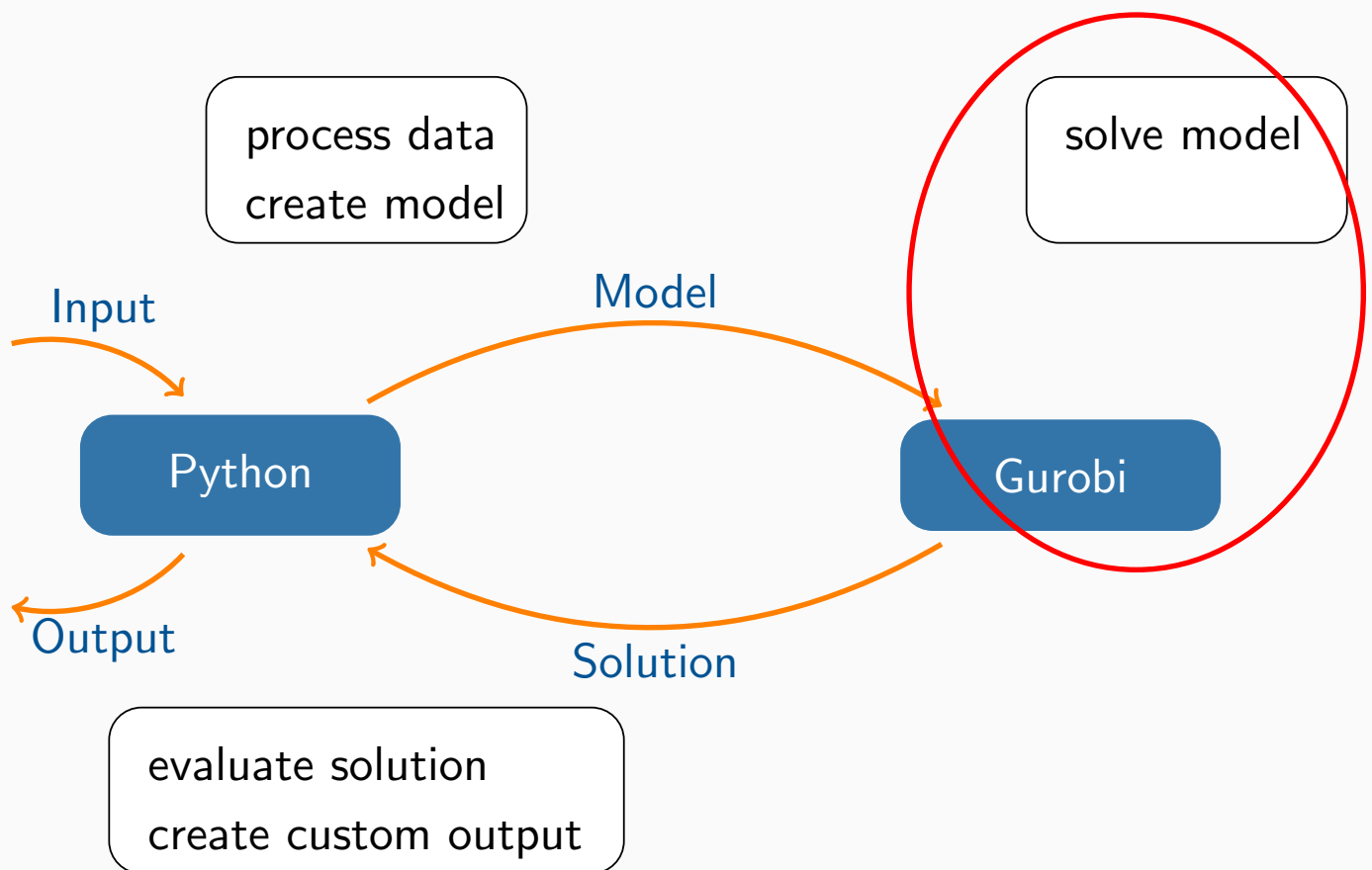


Output Interpretation

Gurobi - LP solver



Presolve

```
Optimize a model with 1500 rows,  
  2250000 columns and 4497000 nonzeros  
Coefficient statistics:  
  Matrix range      [1e+00, 1e+00]  
  Objective range   [1e+00, 1e+02]  
  Bounds range      [0e+00, 0e+00]  
  RHS range         [1e+00, 2e+01]  
Presolve removed 0 rows and 1611 columns  
Presolve time: 4.37s  
Presolved: 1500 rows, 2248389 columns, 4496778 nonzeros
```

Presolve - Example

$$x_1 + x_2 + x_3 \geq 15$$

$$x_1 \leq 7$$

$$x_2 \leq 3$$

$$x_3 \leq 5$$

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delete constraint and variables in order to reduce the problem

LP methods

- ▶ 3 different methods
 - ▶ primal/dual simplex
 - ▶ robust
 - ▶ easy to restart after model modification
 - ▶ barrier
 - ▶ can be run on multiple cores
- ▶ concurrent optimization

Concurrent Optimization

- ▶ run simplex *and* barrier at the same time
- ▶ first one to finish reports solution
- ▶ use multiple cores
- ▶ fastest choice for general model

Primal Simplex

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-1.2573140e+02	2.968000e+03	5.593346e+11	7s
3297	1.2166285e+05	0.000000e+00	6.532640e+07	10s
4619	8.3232592e+04	0.000000e+00	1.162234e+08	15s
7000	5.2154173e+04	0.000000e+00	4.234078e+06	25s
9189	3.7601369e+04	0.000000e+00	1.177763e+07	35s
10706	3.0986489e+04	0.000000e+00	2.645246e+06	45s
12019	2.7488411e+04	0.000000e+00	4.758955e+06	56s
13844	2.3646778e+04	0.000000e+00	3.840044e+05	65s
15403	2.1713789e+04	0.000000e+00	3.055039e+04	75s
17347	2.1105977e+04	0.000000e+00	1.777696e+01	85s
17419	2.1108270e+04	0.000000e+00	0.000000e+00	91s

Solved in 17419 iterations and 90.82 seconds
Optimal objective 2.110826998e+04

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Primal Dual Relations

Primal:

$$\begin{aligned} \min c^\top x \quad & \text{s.t.} \\ Ax &\leq b \end{aligned}$$

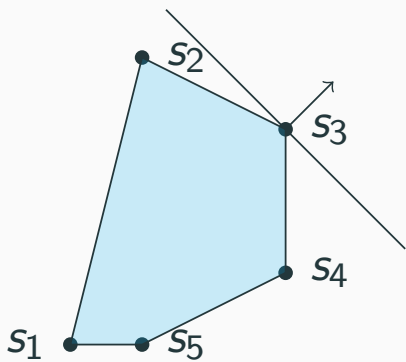
Dual:

$$\begin{aligned} \max b^\top y \quad & \text{s.t.} \\ A^\top y &= c \\ y &\geq 0 \end{aligned}$$

Primal Dual Relations

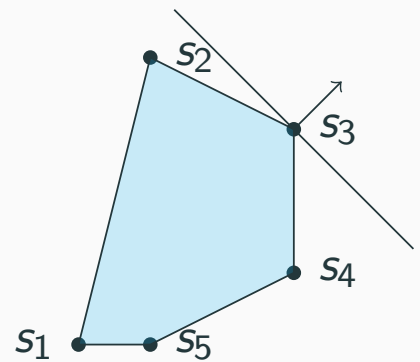
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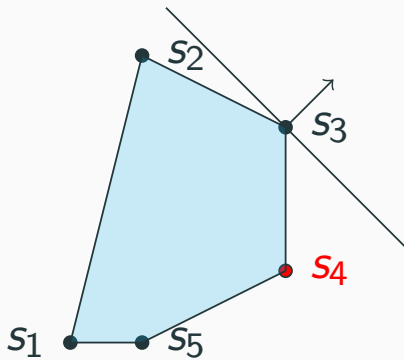
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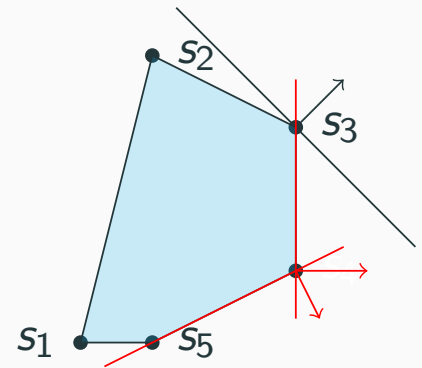
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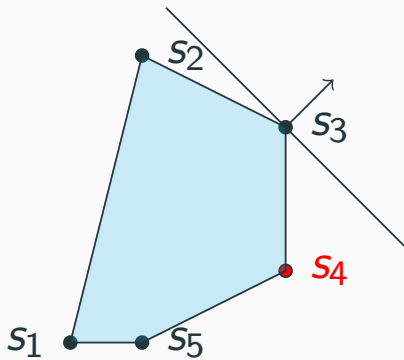
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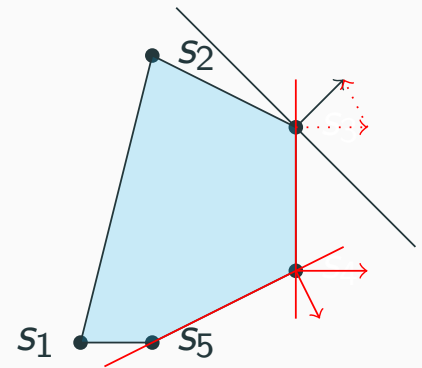
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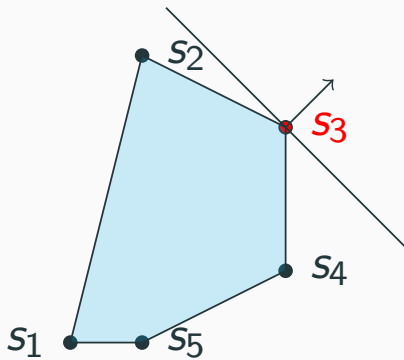
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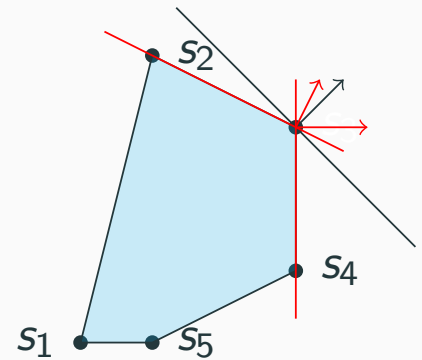
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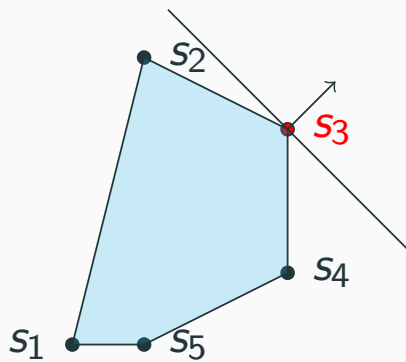
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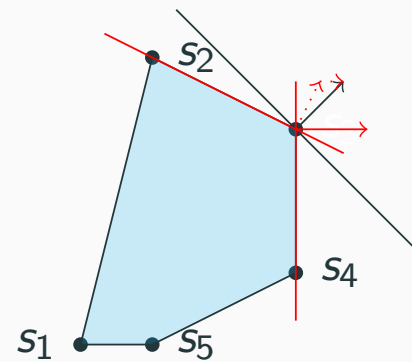
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Dual:

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$$c^\top \bar{x} = \bar{y}^\top A \bar{x} \stackrel{(*)}{=} \bar{y}^\top b$$

Dual Simplex

- ▶ simplex on dual problem
- ▶ obtain primal solution as discussed before
- ▶ dominant algorithm used in MIP solving

Barrier Method

Idea:

$$\begin{aligned} \min c^T x \quad & \text{s.t.} \\ Ax & \leq b \end{aligned}$$

x

Barrier Method

Idea:

$$\begin{array}{ll} \min f(x) & \text{s.t.} \\ c_i(x) \leq 0 \end{array}$$

with f and c_i linear

Barrier Method

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$$\min f(x) - \mu \sum_i \log(c_i(x))$$

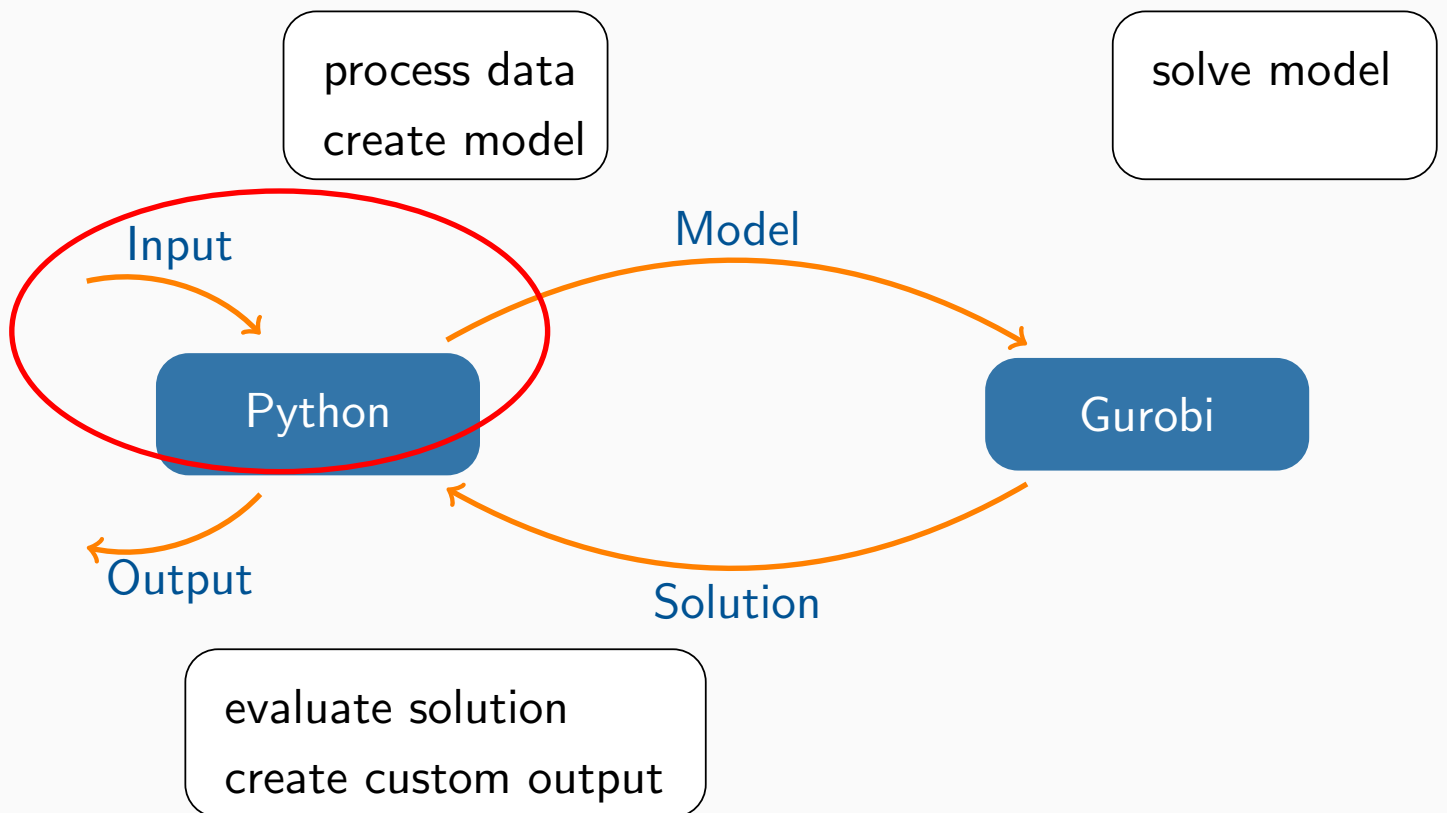
Barrier Method

$$\min f(x) - \mu \sum_i \log(c_i(x))$$

- ▶ $\mu \rightarrow 0$ and solve as nonlinear optimization problem
- ▶ also known as Interior Point Method
- ▶ few, but expensive calculations
- ▶ not clear whether *warm start* is doable

Advanced Input Methods

Advanced Input



Reading Data from Files

Advantages:

- ▶ separation between data and code
- ▶ faster replacement and sharing of data
- ▶ more flexible code
- ▶ clearer code

File Formats

- ▶ excel (pandas, xlrd)
- ▶ csv (csv)
- ▶ json (json)
- ▶ basically any text-file of some custom format (write parser)

The JSON File Format

```
{  
  "oil_types": [  
    "heavy", "medium", "light"  
  ],  
  "processes": [0, 1],  
  "production": {  
    "heavy": [2, 1], "medium": [2, 2], "light": [1, 4]  
  },  
  "demand": {  
    "heavy": 3, "medium": 5, "light": 4  
  },  
  "process_cost": [3, 5]  
}
```

Data Types

- ▶ Boolean
- ▶ Number (integer or floating point)
- ▶ String
- ▶ Array [] (ordered list of elements of arbitrary type)
- ▶ Object {} (unordered collection of name-value pairs)

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Translates straight-forward to Python!

Reading JSON files in Python

```
import json

with open("data.json") as json_file:
    data = json.load(json_file)
```

Read text files

- ▶ generally, input files not given as *json* or *csv* or *xsl/x*
- ▶ any custom format
- ▶ solution: read file line by line according to its syntax and create list/dictionary/object

Reading general textfiles in Python

```
filename = 'data.txt'
with open(filename, "r") as file:
    line = file.readline()
    while line:
        print(line.split("separator"))
```

Write text files

- ▶ generate different datasets to have stable collection of example data
- ▶ straightforward in python (very similar to *print()*)

Writing textfiles in Python

```
filename = 'data.txt'
with open(filename, "w") as file:
    file.write("sometext\n")
    file.close()
```
