

Introduction: Python language

- Python is a widely used general-purpose, high level programming language
- Free, open-source
- Python provides a clean, intuitive syntax
- Many built-in libraries and third-party extensions
- Simple, but extremely powerful

<https://www.learnpython.org/>

You can launch the Spyder IDE, which is included in Anaconda3
Type statements or expressions in console:

```
In [1]: print("Hello world")
Hello world
In [2]: x = 12**2
In [3]: x/2
72
In [4]: # this is a comment
```

To write a program, put commands in a file
hello.py
`print("Hello world")`
`x = 12**2`
`print(x)`



Run file (F5) *hello.py*
Hello world
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Introduction: Python language

Python Variables:

- You do not need to declare variables before using them, or declare their type
- Need to assign (initialize)
- Numbers: integers/floating point numbers; strings; lists; dictionaries,...
- INDEXING STARTS FROM 0!

Lists: declared by []

- Flexible arrays, *not* linked lists
 - `a = [99, "bottles of beer", ["on", "the", "wall"]]`
- Same operators as for strings
 - `a+b, a*3, a[0], a[-1], a[1:], len(a)`
- Item and slice assignment
 - `a[0] = 98`
 - `a[1:2] = ["bottles", "of", "beer"]`
`# -> [98, "bottles", "of", "beer", ["on", "the", "wall"]]`
 - `del a[-1]`
`# -> [98, "bottles", "of", "beer"]`

Introduction: Python language

Dictionaries, declared by {}

- Hash tables, "associative arrays"
 - `d = {"duck": "eend", "water": "water"}`
- Lookup:
 - `d["duck"]` # -> "eend"
 - `d["back"]` # raises *KeyError* exception
- Delete, insert, overwrite:
 - `del d["water"]` # {"duck": "eend", "back": "rug"}
 - `d["back"] = "rug"` # {"duck": "eend", "back": "rug"}
 - `d["duck"] = "duik"` # {"duck": "duik", "back": "rug"}
- Keys, values, items:
 - `d.keys()` -> ["duck", "back"]
 - `d.values()` -> ["duik", "rug"]
 - `d.items()` -> [("duck", "duik"), ("back", "rug")]
- Presence check:
 - `d.has_key("duck")` # -> 1; `d.has_key("spam")` -> 0
- Values of any type; keys almost any
 - { "name": "Guido",
 "age": 43,
 ("hello", "world"): 1,
 42: "yes",
 "flag": ["red", "white", "blue"] }

Introduction: Python language

Control structures: (grouping indentation)

<code>if condition:</code> <code>statements</code>	<code>while condition:</code> <code>statements</code>
<code>[elif condition:</code> <code>statements] ...</code>	<code>for var in sequence:</code> <code>statements</code>
<code>else:</code> <code>statements</code>	 <code>break</code> <code>continue</code>

Note: *Spacing matters!*

Control structure scope dictated by indentation

Introduction: Python language

Functions, procedures

```
def name(arg1, arg2, ...):  
    """documentation"""    # optional doc string  
    statements  
  
    return expression        # from function  
    return                    # from procedure (returns None)
```

Example:

```
def gcd(a, b):  
    """greatest common divisor"""  
    while a != 0:  
        a, b = b%a, a    # parallel assignment  
    return b  
  
>>> gcd.__doc__  
'greatest common divisor'  
  
>>> gcd(12, 20)  
4
```

Introduction: Python language

Modules

- Collection of stuff in *foo.py* file
 - functions, classes, variables
- Importing modules:

```
import re
print( re.match("[a-z]+", s) )
from re import match
print( match("[a-z]+", s) )
```
- Import with rename:

```
import re as regex
from re import match as m
```

Major Python Packages

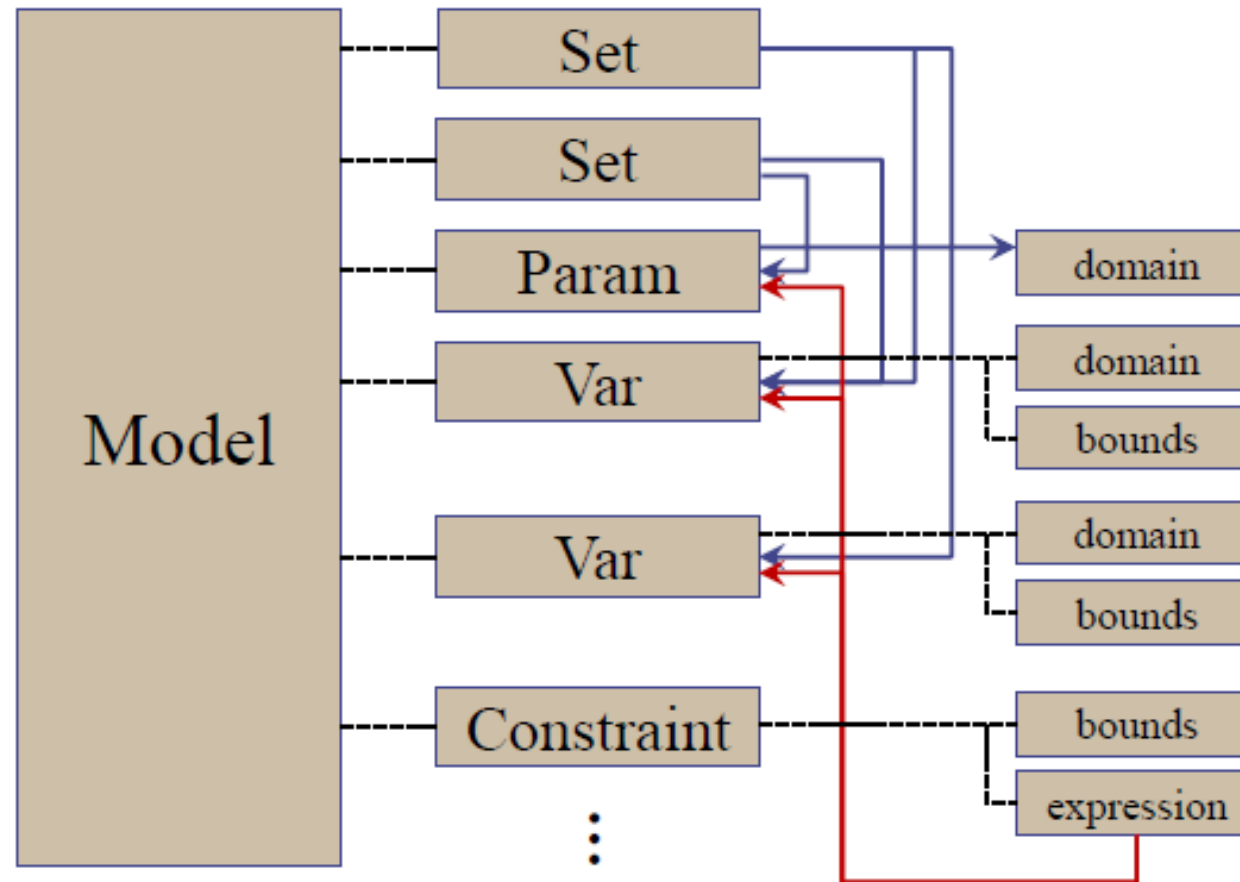
- SciPy
 - Scientific Python for mathematics and engineering
 - <http://www.scipy.org>
- Numpy
 - Numeric array package
 - <http://www.numpy.org/>
- Matplotlib
 - 2D plotting library
 - <http://matplotlib.org/>
- Pandas
 - Data structures and analysis
 - <http://pandas.pydata.org/>
- IPython
 - Interactive Python shell
 - <http://ipython.org/>

Introduction: Pyomo fundamentals

<https://jckantor.github.io/ND-Pyomo-Cookbook/>

<http://www.pyomo.org/documentation>

- Pyomo is an *object model* for describing optimization problems
 - The fundamental objects used to build models are *Components*



Introduction: Pyomo fundamentals – the MODEL

```
from pyomo.environ import *
```



Every Pyomo model starts with this; it tells Python to load the Pyomo Modeling Environment

```
model = ConcreteModel()
```



Create an instance of a *Concrete* model

- Concrete models are immediately constructed
- Data must be present *at the time* components are defined



Local variable to hold the model we are about to construct

- While not required, by convention we use “**model**”
- If you choose to name your model something else, you will need to tell the Pyomo script the object name through the command line

Introduction: Pyomo fundamentals – the VARIABLES

```
model.a_variable = Var(within = NonNegativeReals)
```

↑
The name you assign the object to becomes the object's name, and must be unique in any given model.

↑
“within” is optional and sets the variable domain (“domain” is an alias for “within”)

↑
Several pre-defined domains, e.g., “Binary”

```
model.a_variable = Var(bounds = (0, None))
```

↑
Same as above: “domain” is assumed to be Reals if missing

Introduction: Pyomo fundamentals – the OBJECTIVE FUNCTION

```
model.x = Var( initialize=-1.2, bounds=(-2, 2) )  
model.y = Var( initialize= 1.0, bounds=(-2, 2) )
```

```
model.obj = Objective(  
    expr= (1-model.x)**2 + 100*(model.y-model.x**2)**2,  
    sense= minimize )
```

If “sense” is omitted, Pyomo
assumes minimization

“expr” can be an expression,
or any function-like object
that returns an expression

Introduction: Pyomo fundamentals – the SETS to generate and manage indexes

- Any iterable object can be an index, e.g., lists:

- `IDX_a = [1,2,5]`
 - `DATA = {1: 10, 2: 21, 5:42};`
`IDX_b = DATA.keys()`

- Sets: objects for managing multidimensional indices

- `model.IDX = Set(initialize = [1,2,5])`

Note: capitalization matters:
Set = Pyomo class
set = native Python set

Like indices, Sets can be
initialized from any iterable

- Sets of sequential integers are common

- `model.IDX = Set(initialize=range(5))`
 - `model.IDX = RangeSet(5)`

Note: **RangeSet** is 1-based.
This gives [1, 2, 3, 4, 5]

Note: Python **range** is 0-based.
This gives [0, 1, 2, 3, 4]

- You can provide lower and upper bounds to RangeSet

- `model.IDX = RangeSet(0, 4)`

This gives [0, 1, 2, 3, 4]

Introduction: Pyomo fundamentals – the CONSTRAINTS

$$x_{w,c} \leq y_w \quad \forall w \in W, c \in C$$

```
W = ['Harlingen', 'Memphis', 'Ashland']  
C = ['NYC', 'LA', 'Chicago', 'Houston']  
model.x = Var(W, C, bounds=(0,1))  
model.y = Var(W, within=Binary)
```

For indexed constraints, you provide a “rule” (function) that returns an expression (or tuple) for each index.

Each dimension of the indices is a separate argument to the rule

```
def warehouse_active_rule(m, w, c):  
    return m.x[w,c] <= m.y[w]  
model.warehouse_active = Constraint(W, C, rule=warehouse_active_rule)
```

Rule is called once for every entry w in W
crossed with every entry c in C!

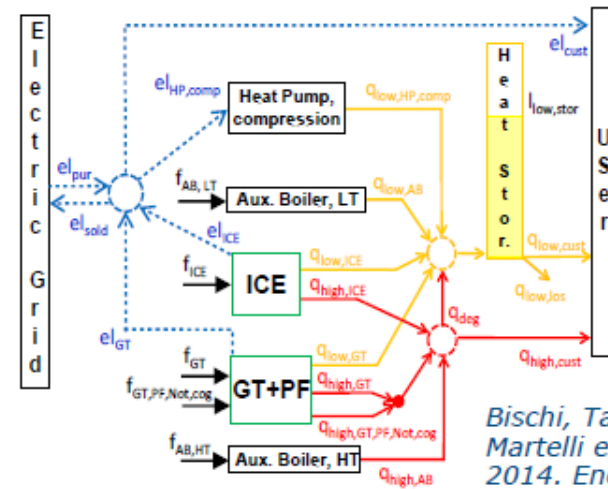
The MES optimal operation MILP model:

The basic steps of a simple modeling process are as follows:

1. create an instance of a model using Pyomo modeling components
2. pass this instance to a solver to find a solution
3. report and analyze results from the solver

MILP PWL FOR DAILY AND WEEKLY PROBLEMS

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*Bischi, Taccari,
Martelli et al.
2014. Energy, Vol.
74*

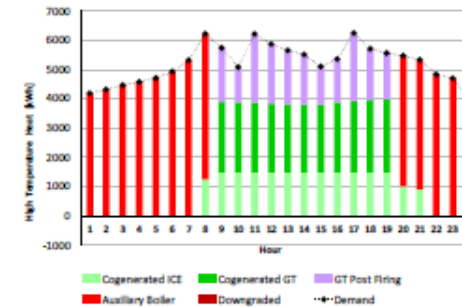
Computational time:

1 day operation: < 1 sec

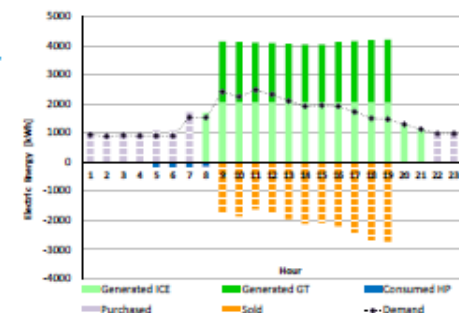
1 week operation: < 2 min

Up to 18% primary energy saving compared to usual operation strategies!

High temperature heat



Electricity



The MES optimal operation MILP model

Sets

I = set of all units

T = set of all time instants (24 hours in a day)

Subsets:

- Machines_fuel
- Machines_el
- Machines_th
- Machines_el_cons

Variables

- 'MIMO' units with 1 degree of freedom (ICE, ORC, boiler, heat pump,...)

$f_{i,t}$ = input fuel of unit i at time t

$q_{i,t}$ = output heat of unit i at time t

$p_{i,t}$ = output electricity of unit i at time t

$el_in_{i,t}$ = input electricity of unit i at time t



$z_{i,t} \in \{0,1\}$ = 1 if unit i is on at time t , 0 if unit i is off at time t

$d_{SU_{i,t}} \in \{0,1\}$ = 1 if unit i starts-up at time t , 0 otherwise

The MES optimal operation MILP model

Sets

I = set of all units

T = set of all time instants (24 hours in a day)

S = set of all storages

Subsets:

- Storages_el
- Storages_th

Variables

- Storages

$u_{s,t}$ = level of energy (heat or electricity) of storage s at time t

$Charge_{s,t}$ = energy charged to storage s at time t

$Discharge_{s,t}$ = energy discharged from storage s at time t

- Electricity sold/bought from the grid

El_{buy_t} = electricity bought from the grid at time t

El_{sell_t} = electricity sold to the grid at time t

ALL CONTINUOUS VARIABLES
ARE POSITIVE REALS

The MES optimal operation MILP model

Parameters

d_{el_t} = demand of electricity at time t

d_{th_t} = demand of heat at time t

$El_{price_{sell,t}}$ = electricity selling price to grid (constant) at time t

$El_{price_{buy,t}}$ = electricity buying price from grid (constant) at time t

NG_{price} = natural gas price (constant)

$Biomass_{price}$ = biomass price (constant)

The MES optimal operation MILP model

Parameters

$Fuel_cost_i$ = fuel cost of unit i , for machines $i \in I_f$

a_{th_i} = coefficient of heat production map of unit i

b_{th_i} = constant term of heat production map of unit i

Linear performance maps:

$$q_{i,t} = a_{th_i} f_{i,t} + b_{th_i} z_{i,t}$$

$$q_{i,t} = a_{th_i} el_in_{i,t} + b_{th_i} z_{i,t}$$

a_{el_i} = coefficient of power production map of unit i

b_{el_i} = constant term of power production map of unit i

$$p_{i,t} = a_{el_i} f_{i,t} + b_{el_i} z_{i,t}$$

The MES optimal operation MILP model

Parameters

For machines $i \in I$:

$\min In_i$ = minimum input of unit i (if on)

$\max In_i$ = maximum input of unit i (if on)

$RUlim_i$ = ramp-up limit of unit i (only for ORC)

$\max_n_SU_i$ = maximum number of daily start-ups of unit i (only for ORC)

OM_i = fixed O&M costs of unit i

SU_i = fixed start-up cost of unit i

The MES optimal operation MILP model

Parameters

For storages

$maxC_s$ = maximum capacity (heat or electricity) of storage s

$eta_{ch_s}, eta_{disch_s}$ = charge/discharge efficiency of electric storage s

eta_{diss_s} = thermal loss efficiency of thermal storage s

The MES optimal operation MILP model

Objective function (to be minimized):

$$OBJ = machines_fuel_cost + machines_OM_cost + machines_SU_cost + grid_SellBuy$$

Where

$$machines_fuel_cost = \sum_{i \in I_f} \sum_{t \in T} Fuel_cost_i \cdot f_{i,t}$$

$$machines_OM_cost = \sum_{i \in I} \sum_{t \in T} OM_i \cdot z_{i,t}$$

$$machines_SU_cost = \sum_{i \in I} \sum_{t \in T} SU_i \cdot dSU_{i,t}$$

$$grid_SellBuy = \sum_{t \in T} El_{price_buy} \cdot El_{buy_t} - \sum_{t \in T} El_{price_sell} \cdot El_{sell_t}$$

The MES optimal operation MILP model

Constraints:

Electricity balance

$$\sum_{i \in I_{el}} p_{i,t} - \sum_{i \in I_{elcons}} el_in_{i,t} + El_{buy_t} - El_{sell_t} + \sum_{s \in S_{el}} Discharge_{s,t} - \sum_{s \in S_{el}} Charge_{s,t} = d_{el_t}$$

$$\forall t \in T$$

The MES optimal operation MILP model

Constraints:

Heat balance

$$\sum_{i \in I_{th}} q_{i,t} + \sum_{s \in S_{th}} Discharge_{s,t} - \sum_{s \in S_{th}} Charge_{s,t} = d_{th_t}$$

$$\forall t \in T$$

The MES optimal operation MILP model

Constraints:

El. power production of units

$$p_{i,t} = a_{el_i} f_{i,t} + b_{el_i} z_{i,t} \quad \forall i \in I_{el}, \quad \forall t \in T$$

Heat production ($i \in I_{th}$) of units with fuel input or electricity input (heat pump)

$$q_{i,t} = a_{th_i} f_{i,t} + b_{th_i} z_{i,t} \quad \forall i \in I_f, \quad \forall t \in T$$

$$q_{i,t} = a_{th_i} el_in_{i,t} + b_{th_i} z_{i,t} \quad \forall i \in I_{el_cons}, \quad \forall t \in T$$

The MES optimal operation MILP model

Constraints:

Operating range of units – minimum input

$$f_{i,t} \geq \min In_i z_{i,t} \quad \forall i \in I_f, \quad \forall t \in T$$

$$el_in_{i,t} \geq \min In_i z_{i,t} \quad \forall i \in I_{el_cons}, \quad \forall t \in T$$

Operating range of units – maximum input

$$f_{i,t} \leq \max In_i z_{i,t} \quad \forall i \in I_f, \quad \forall t \in T$$

$$el_in_{i,t} \leq \max In_i z_{i,t} \quad \forall i \in I_{el_cons}, \quad \forall t \in T$$

The MES optimal operation MILP model

Constraints:

Logical constraints for start-up of units

$$z_{i,t} - z_{i,t-1} \leq dSU_{i,t}$$

$$\forall i \in I, \quad \forall t \in T$$

NOTE: case
when $t == 0$

Ramp-up constraints

$$f_{i,t} - f_{i,t-1} \leq RUlim_i z_{i,t} \quad \forall i \in I_f, \quad \forall t \in T$$

NOTE: case
when $t == 0$

$$el_in_{i,t} - el_in_{i,t-1} \leq RUlim_i z_{i,t} \quad \forall i \in I_{el_cons}, \forall t \in T$$

The MES optimal operation MILP model

Constraints:

Maximum number of start-ups per day (only for ORC)

$$\sum_{t \in T} dSU_{i,t} \leq 1$$

The MES optimal operation MILP model

Constraints:

NOTE: case
when $t == 0$

Thermal storage energy balance:

$$u_{s,t} - u_{s,t-1} \cdot \eta_{diss,s} = Charge_{s,t} - Discharge_{s,t}$$

$$\forall s \in S_{th}, \quad \forall t \in T$$

Electric storage energy balance:

$$u_{s,t} - u_{s,t-1} = Charge_{s,t} \cdot \eta_{ch,s} - Discharge_{s,t} / \eta_{disch,s}$$

$$\forall s \in S_{el}, \quad \forall t \in T$$

Storage capacity

$$u_{s,t} \leq MaxC_s \quad \forall s \in S, \quad \forall t \in T$$

The MES optimal operation MILP model

Solve model

```
pyo.SolverFactory('CBC', mipgap = 0.005).solve(m).write()
```

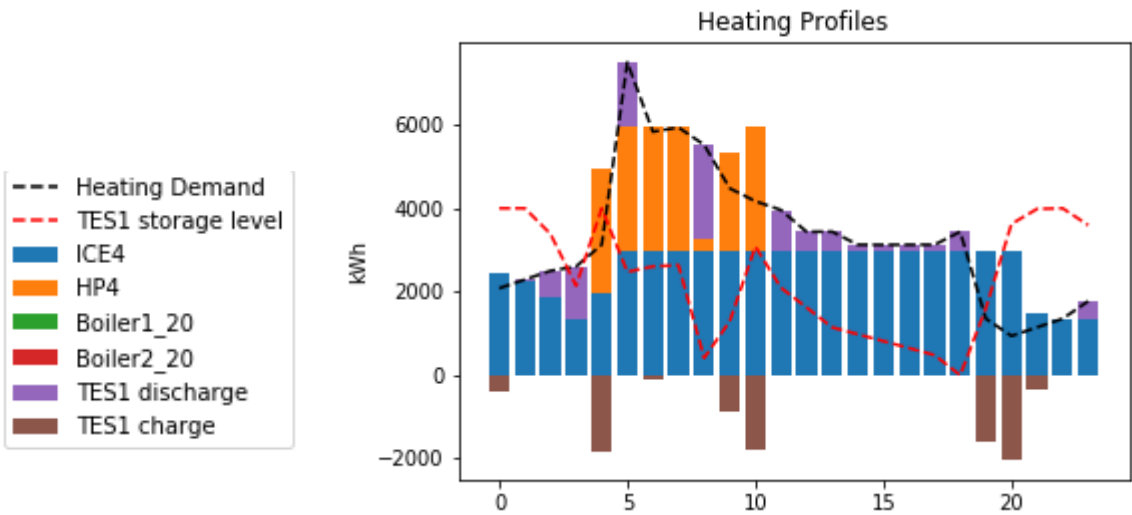
Save results

```
.value
```

The MES optimal operation MILP model

Barplots

Day 1



Day 2

