Universidad de los Andes

Laboratorio 4
Manejo básico de Pyomo

Modelado, Optimización y Simulación

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Agenda

- Conjuntos
- Parámetros
- Variables
- Ejemplos
  - Función objetivo
  - Restricciones
    - Sumatorias y “para todo”.

Conjuntos

- Definición básica:

```
23 Model.N={1,2,3,4,5}
```

- Definición usando la función RangeSet:

```
19 numNodes=5
20
21 Model.N=RangeSet(1, numNodes)
```

- Definición usando cadenas de caracteres:

```
25 Model.N = {"Nodo1", "Nodo2", "Nodo3", "Nodo4", "Nodo5"}
```

- Operaciones entre conjuntos:

```
>>> model.I = model.A | model.D # union
>>> model.J = model.A & model.D # intersection
>>> model.K = model.A - model.D # difference
>>> model.L = model.A ^ model.D # exclusive-or
```

Parámetros

- Vectores:

```
14 numProyectos=8
15
16 M.p=RangeSet(1, numProyectos)
17
18 M.valor=Param(M.p, mutable=True)
19
20 M.valor[1]=2
21 M.valor[2]=5
22 M.valor[3]=4
23 M.valor[4]=2
24 M.valor[5]=6
25 M.valor[6]=3
26 M.valor[7]=1
27 M.valor[8]=4
```



```
14 numProyectos=8
15
16 M.p=RangeSet(1, numProyectos)
17
18 M.valor=Param(M.p, mutable=True)
19
20 for i in M.p:
21     M.valor[i]=2
```

Parámetros

- Matrices:

Forma 1:

```
numNodes=5

N=RangeSet(1, numNodes)

cost={(1,1):999, (1,2):5,   (1,3):2,   (1,4):999, (1,5):999,\
      (2,1):999, (2,2):999, (2,3):999, (2,4):999, (2,5):8,\
      (3,1):999, (3,2):999, (3,3):999, (3,4):3,   (3,5):999,\
      (4,1):999, (4,2):999, (4,3):999, (4,4):999, (4,5):2,\
      (5,1):999, (5,2):999, (5,3):999, (5,4):999, (5,5):999}
```

Forma 2:

```
numNodes=5

Model.N=RangeSet(1, numNodes)

Model.cost=Param(Model.N, Model.N, mutable=True)

for i in Model.N:
    for j in Model.N:
        Model.cost[i,j]=999

Model.cost[1,2]=5
Model.cost[1,3]=2
Model.cost[2,5]=8
Model.cost[3,4]=3
Model.cost[4,5]=5
```

Variables

- Variable simple (sin dimensiones):
  - Model.x=Var()
- Dominio de las variables:
  - Forma 1: Model.x=Var(N, domain=NonNegativeReals)
  - Forma 2: Model.x=Var(N, within=NonNegativeReals)
- Limite superior e inferior de las variables:
  - Ej1: Model.x=Var(domain=NonNegativeReals, bounds=(0,6))
  - Ej2: Model.x=Var(N, domain=NonNegativeReals, bounds=(0,6))

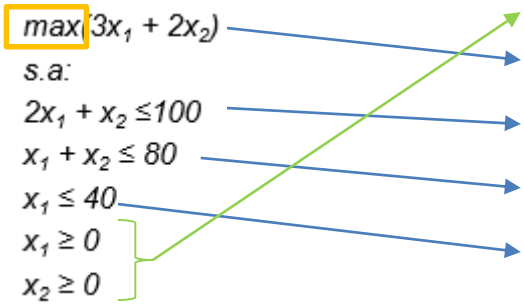
Variables

- Dominios posibles para las variables:

`Reals`	The set of floating point values
`PositiveReals`	The set of strictly positive floating point values
`NonPositiveReals`	The set of non-positive floating point values
`NegativeReals`	The set of strictly negative floating point values
`NonNegativeReals`	The set of non-negative floating point values
`PercentFraction`	The set of floating point values in the interval [0,1]
`Integers`	The set of integer values
`PositiveIntegers`	The set of positive integer values
`NonPositiveIntegers`	The set of non-positive integer values
`NegativeIntegers`	The set of negative integer values
`NonNegativeIntegers`	The set of non-negative integer values
`Boolean`	The set of boolean values, which can be represented as False/True, 0/1, 'False'/'True' and 'F'/'T'
`Binary`	The same as 'Boolean'

Ejemplos

- Caso Woodcarving:




```
7  Model = ConcreteModel()
8
9  Model.x = Var([1,2], domain=NonNegativeReals)
10
11 Model.obj = Objective(expr = 3*Model.x[1] + 2*Model.x[2], sense=maximize)
12
13 Model.res1 = Constraint(expr = 2*Model.x[1] + Model.x[2] <= 100)
14
15 Model.res2 = Constraint(expr = Model.x[1] + Model.x[2] <= 80)
16
17 Model.res3 = Constraint(expr = Model.x[1] <= 40)
18
19 SolverFactory('glpk').solve(Model)
20
21 Model.display()
```

  - **Nota**: si en el campo donde aparece *"sense=maximize"* no se especifica nada, por defecto Pyomo assume que estamos minimizando.

Ejemplos

- Caso Proyectos:

$$x_i = \begin{cases} 0 \\ 1 \end{cases}$$

$$max \sum_i g_i * x_i$$

$$\sum_i x_i = 2$$

```
1
2  from __future__ import division
3  from pyomo.environ import *
4
5  from pyomo.opt import SolverFactory
6
7  Model = ConcreteModel()
8
9  # Sets and Parameters
10 numProyectos=8
11
12 #p=[1, 2, 3, 4, 5, 6, 7, 8]
13 p=RangeSet(1, numProyectos)
14
15 valor={1:2, 2:5, 3:4, 4:2, 5:6, 6:3, 7:1, 8:4}
16
17 # Variables
18 Model.x = Var(p, domain=Binary)
19
20 # Objective Function
21 Model.obj = Objective(expr = sum(Model.x[i]*valor[i] for i in p), sense=maximize)
22
23 # Constraints
24 Model.res1 = Constraint(expr = sum(Model.x[i] for i in p) == 2)
25
26 # Applying the solver
27 SolverFactory('glpk').solve(Model)
28
29 Model.display()
30
```

Ejemplos

- Caso Mínimo Costo:

$$X_{ij} = \begin{cases} 1 \rightarrow \text{Escojo el enlace } (i,j) \\ 0 \rightarrow \text{NO Escojo el enlace } (i,j) \end{cases}$$

$$\min \sum_{i \in N} \sum_{j \in N} C_{ij} X_{ij}$$

$$\sum_{j \in N} X_{ij} = 1 \quad \forall i / i = 1$$

$$\sum_{i \in N} X_{ij} = 1 \quad \forall j / j = 5$$

$$\sum_{j \in N} X_{ij} - \sum_{j \in N} X_{ji} = 0 \quad \forall i / i \notin \{1,5\}$$

```
29 # VARIABLES*******************************************************************************
30 Model.x = Var(N,N, domain=Binary)
31
32 # OBJECTIVE FUNCTION*************************************************************************
33 Model.obj = Objective(expr = sum(Model.x[i,j]*cost[i,j] for i in N for j in N))
34
35 # CONSTRAINTS*******************************************************************************
36 def source_rule(Model,i):
37     if i==1:
38         return sum(Model.x[i,j] for j in N)==1
39     else:
40         return Constraint.Skip
41
42 Model.source=Constraint(N, rule=source_rule)
43
44 def destination_rule(Model,j):
45     if j==5:
46         return sum(Model.x[i,j] for i in N)==1
47     else:
48         return Constraint.Skip
49
50 Model.destination=Constraint(N, rule=destination_rule)
51
52 def intermediate_rule(Model,i):
53     if i!=1 and i!=5:
54         return sum(Model.x[i,j] for j in N) - sum(Model.x[j,i] for j in N)==0
55     else:
56         return Constraint.Skip
57
58 Model.intermediate=Constraint(N, rule=intermediate_rule)
59
60 # APPLYING THE SOLVER************************************************************************
61 SolverFactory('glpk').solve(Model)
62
63 Model.display()
```

Tips

- Cómo introducimos un *"tal que"* en una sumatoria?
  - Forma 1:

$$\sum_{i/i\neq 1}^{p} x_i = 2$$

```
40 Model.res1 = Constraint(expr = sum(Model.x[i] for i in Model.p if i!=1) == 2)
```

  - Forma 2:

$$\sum_{j\in N/j\neq 2} x_{ij} = 1 \quad \forall i \in N|i = 1$$

```
36 def source_rule(Model,i):
37     if i==1:
38         return sum(Model.x[i,j] for j in N if j!=2)==1
39     else:
40         return Constraint.Skip
41
42 Model.source=Constraint(N, rule=source_rule)
```

Tips

- Otro método para crear una restricción que tenga un ‘para todo’:
  - Ejemplo 1:

$$\sum_{\mathrm{i}\in N} x_{ij} = 1 \quad \forall \mathrm{j} \in N | \mathrm{j} = 1$$

```
166 Model.res1=ConstraintList()
167 for j in N:
168     if j == 1:
169         Model.res1.add( sum(Model.x[i,j] for i in N ) == 1 )
```

  - Ejemplo 2:

$$\sum_{\mathrm{k}\in M} x_{ijk} = 1 \quad \forall \mathrm{i,j} \in N | i = 1$$

```
171 Model.res2=ConstraintList()
172 for i in N:
173     for j in N:
174         if i == 1:
175             Model.res2.add( sum(Model.x[i,j,k] for k in M ) == 1 )
```

Solvers

- Problemas LP y MIP:

```
61 SolverFactory('glpk').solve(Model)
62
63 Model.display()
```

- Problemas NLP:

```
28 SolverFactory('ipopt').solve(model)
29
30 model.display()
```

- Problemas MINLP:

```
26 SolverFactory('mindtpy').solve(model, mip_solver='glpk', nlp_solver='ipopt')
27
28 model.display()
```