Network-Node Stochastic Programming

Python Modeling Code for (c) and (d)

```
import itertools
from pyomo.environ import *
from pyomo.core import *
from pyomo.pysp.annotations import
(PySP ConstraintStageAnnotation,
StochasticConstraintBoundsAnnotation,
StochasticConstraintBodyAnnotation)
theta1_table = [1,1,1,1,1,1,1,0,0,0]
theta2_table = [1,1,1,1,1,1,1,0,0,0]
d1_rhs_table = [0,1,2,3,5]
d2_{rhs_table} = [0,1,2,3,5]
d3_rhs_table = [0,1,2,3]
model = ConcreteModel()
#-----C-----
# model.constraint_stage =
PySP_ConstraintStageAnnotation()
# model.stoch matrix =
StochasticConstraintBodyAnnotation()
# num_scenarios = len(theta1_table)*len(theta2_table)
# scenario data = dict(('Scenario'+str(i),
(theta1val, theta2val))
                       for i, (theta1val, theta2val)
in
                       enumerate
(itertools.product(theta1_table, theta2_table),1))
#-----D-----
model.constraint stage =
PySP_ConstraintStageAnnotation()
model.stoch rhs =
StochasticConstraintBoundsAnnotation()
model.stoch matrix =
StochasticConstraintBodyAnnotation()
num scenarios =
len(theta1_table)*len(theta2_table)*len(d1_rhs_table)
* len(d2_rhs_table) * len(d3_rhs_table)
scenario_data = dict(('Scenario'+str(i), (theta1val,
theta2val, d1_rhs_val, d2_rhs_val, d3_rhs_val))
                     for i, (theta1val, theta2val,
d1_rhs_val, d2_rhs_val, d3_rhs_val) in
                     enumerate
(itertools.product(theta1_table, theta2_table,
d1_rhs_table, d2_rhs_table, d3_rhs_table),1))
model.d1_rhs = Param(mutable=True, initialize=0.0)
model.d2 rhs = Param(mutable=True, initialize=0.0)
model.d3 rhs = Param(mutable=True, initialize=0.0)
model.x1 = Var(within=NonNegativeReals)
model.x2 = Var(within=NonNegativeReals)
model.x3 = Var(within=NonNegativeReals)
model.f11 = Var(within=NonNegativeReals)
model.f12 = Var(within=NonNegativeReals)
model.f21 = Var(within=NonNegativeReals)
model.f22 = Var(within=NonNegativeReals)
model.f31 = Var(within=NonNegativeReals)
model.f32 = Var(within=NonNegativeReals)
model.s1 = Var(within=NonNegativeReals)
model.s2 = Var(within=NonNegativeReals)
```

```
model.s3 = Var(within=NonNegativeReals)
model.theta1 = Param(mutable=True, initialize=0)
model.theta2 = Param(mutable=True, initialize=0)
model.FirstStageCost = Expression(initialize=0)
model.SecondStageCost =
Expression(initialize=model.s1+model.s2+model.s3)
model.budget =
Constraint(expr=model.x1+model.x2+4*model.x3 <= 10)</pre>
model.constraint stage.declare(model.budget,1)
#-----
# model.d1 =
Constraint(expr=model.f11+model.f12+model.s1 == 2.05)
# model.constraint stage.declare(model.d1,2)
# model.d2 =
Constraint(expr=model.f21+model.f22+model.s2 == 2.05)
# model.constraint_stage.declare(model.d2,2)
# model.d3 =
Constraint(expr=model.f31+model.f32+model.s3 == 1.5)
# model.constraint_stage.declare(model.d3,2)
#-----D-----
model.d1 =
Constraint(expr=model.f11+model.f12+model.s1 ==
model.d1 rhs)
model.constraint_stage.declare(model.d1,2)
model.stoch_rhs.declare(model.d1)
Constraint(expr=model.f21+model.f22+model.s2 ==
model.d2 rhs)
model.constraint_stage.declare(model.d2,2)
model.stoch_rhs.declare(model.d2)
model.d3 =
Constraint(expr=model.f31+model.f32+model.s3 ==
model.d3 rhs)
model.constraint stage.declare(model.d3,2)
model.stoch_rhs.declare(model.d3)
#-----
model.c1 =
Constraint(expr=model.f11+model.f22+model.f32-model.t
heta1*model.x1 <= 0)</pre>
model.constraint stage.declare(model.c1,2)
model.stoch_matrix.declare(model.c1,
variables=[model.x1])
model.c2 =
Constraint(expr=model.f12+model.f21+model.f32-model.t
heta2*model.x2 <= 0)</pre>
model.constraint stage.declare(model.c2,2)
model.stoch matrix.declare(model.c2,
variables=[model.x2])
Constraint(expr=model.f12+model.f22+model.f31-model.x
3 <= 0)
model.constraint stage.declare(model.c3,2)
model.obi =
Objective(expr=model.FirstStageCost+model.SecondStage
Cost)
```

```
def pysp_scenario_tree_model_callback():
    from pyomo.pysp.scenariotree.tree structure model
import \
        CreateConcreteTwoStageScenarioTreeModel
    st model =
CreateConcreteTwoStageScenarioTreeModel(num_scenarios
    first_stage = st_model.Stages.first()
    second_stage = st_model.Stages.last()
    # First Stage
    st_model.StageCost[first_stage] =
'FirstStageCost'
    st_model.StageVariables[first_stage].add('x1')
    st_model.StageVariables[first_stage].add('x2')
    st model.StageVariables[first stage].add('x3')
    # Second Stage
    st_model.StageCost[second_stage] =
'SecondStageCost'
    st_model.StageVariables[second_stage].add('f11')
    st_model.StageVariables[second_stage].add('f12')
    st model.StageVariables[second stage].add('f21')
    st_model.StageVariables[second_stage].add('f22')
    st_model.StageVariables[second_stage].add('f31')
    st model.StageVariables[second stage].add('f32')
    st_model.StageVariables[second_stage].add('s1')
    st_model.StageVariables[second_stage].add('s2')
    st model.StageVariables[second stage].add('s3')
    return st model
def pysp_instance_creation_callback(scenario_name,
node names):
    #
    # Clone a new instance and update the stochastic
    # parameters from the sampled scenario
    instance = model.clone()
    theta1_val, theta2_val, d1_rhs_val, d2_rhs_val,
d3_rhs_val = scenario_data[scenario_name]
   instance.theta1.value = theta1_val
    instance.theta2.value = theta2 val
 #-----D-----
    instance.d1_rhs.value = d1_rhs_val
    instance.d2 rhs.value = d2 rhs val
    instance.d3 rhs.value = d3 rhs val
    return instance
```

- (c) Total execution time=3.67 seconds
- (d) Total execution time=370.37 seconds

(C-1)SD input smps files

```
network_c.cor

* Source: Pyomo MPS Writer
* Format: Free MPS

*
NAME Scenario1
OBJSENSE
MIN
ROWS
N obj
L c_u_budget_
```

```
Eced1
E c_e_d2_
Eced3
L c_u_c1_
L c_u_c2_
L c_u_c3_
E c_e_ONE_VAR_CONSTANT
COLUMNS
  x1 c u budget 1
  x1 c u c1 -1
  x2 c_u_budget_ 1
  x2 c_u_c2_ -1
  x3 c u budget 4
  x3 c u c3 -1
  f11 c_e_d1_ 1
  f11 c u c1 1
  f12 c_e_d1_ 1
  f12 c_u_c2_ 1
  f12 c_u_c3_ 1
  f21 c_e_d2_ 1
  f21 c_u_c2_ 1
  f22 c e d2 1
  f22 c u c1 1
  f22 c_u_c3_ 1
  f31 c_e_d3_ 1
  f31 c u c3 1
  f32 c e d3 1
  f32 c_u_c1_ 1
  f32 c u c2 1
  s1 obj 1
  s1 c_e_d1_ 1
  s2 obj 1
  s2 c_e_d2_ 1
  s3 obj 1
  s3 c_e_d3 1
  ONE VAR CONSTANT obj 0
  ONE_VAR_CONSTANT c_e_ONE_VAR_CONSTANT 1
  RHS c u budget 10
  RHS c e d1 2.049999999999998
  RHS c e d2 2.049999999999998
  RHS c e d3 1.5
  RHS c_u_c1_ 0
  RHS c_u_c2_ 0
  RHS c u c3 0
  RHS c_e_ONE_VAR_CONSTANT 1
BOUNDS
LO BOUND x1 0
LO BOUND x2 0
LO BOUND x3 0
LO BOUND f11 0
LO BOUND f12 0
LO BOUND f21 0
LO BOUND f22 0
LO BOUND f31 0
LO BOUND f32 0
LO BOUND s1 0
LO BOUND s2 0
LO BOUND s3 0
ENDATA
```

```
network_c.tim

TIME network
PERIODS IMPLICIT
x1 obj TIME1
```

```
f11 c e d1 TIME2
ENDATA
```

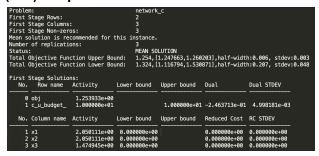
network_c_revised.sto						
STOCH r INDEP x1 x2 x1 x2 ENDATA	etail DISCRETE c_u_c11 c_u_c2_ 0 c_u_c1_ 0 c_u_c21	0.7 0.3 0.3 0.7				

(C-2) Output Console

Yului-MacBook:Debug yul\$./sd Beginning SD... Please enter the name of the problem files (eg. 'exags1p'): network c mkdir: ./sdoutput: File exists mkdir: /sdoutput/network c: File exists Reading problems from /sdinput/network c/network c.cor the filename is: ./sdinput/network c/network c.cor Specified objective sense: MINIMIZE Selected objective name: obj Selected RHS name: RHS Selected bound name: BOUND finished loading core file via external Solver. -0.700000 -0.700000 Vector x k from the mean problem :: 16.5999999913667068085 4.9999999949012696732 4.9999999935742867052 0.00000000028811110114 2.0499999981604874577 0.00000000003835065432 $2.0499999981604652532\ 0.0000000003835065432$ 0.0000000021140979251 1.44999999969580239956 0.0000000014588982868 0.0000000014588982868 $0.05000000009321703409\ 1.00000000000000000000$ Objective from the mean problem: 0.050000 num rv = 2; num_cipher = 1; ----- Replication No.0 ------6.000000 ---- Replication No.1 ------+++++Warning: Output names have been modified to conform to LP format. Warning: Output names have been modified to conform to LP format. +++Warning: Output names have been modified to conform to LP format. +Warning: Output names have been modified to conform to ++++++>***********Max pi***********: 6.000000 ----- Replication No.2 ------+++++++++++++++

```
Begin evaluation of average solution
.....
obs:10000 mean:1.250491 error: 0.041535
0.95 CI: [1.224521, 1.276461]
(ETC...)
Final Estimate
obs:172378 mean:1.253933
0.95 C.I.: [1.247663, 1.260203]
Ending SD...
Total running time: 7512.242000 ms
```

(C-3) Output Result



Hence, we get the optimal value of x1 = 2.0501, x2 = 2.0501, x3 = 1.4750

(D-1) smps files

network_d.cor Pyomo MPS Writer * Source: * Format: Free MPS NAME Scenario1 **OBJSENSE** MIN ROWS N obi L c_u_budget_ E c_e_d1_ E c_e_d2_ Eced3 L c u c1 L c_u_c2_ L c_u_c3_ E c e ONE VAR CONSTANT COLUMNS x1 c_u_budget_ 1 x1 c u c1 -1 x2 c_u_budget_ 1 x2 c_u_c2_ -1 x3 c_u_budget_ 4 x3 c_u_c3_ -1 f11 c_e_d1_ 1 f11 c_u_c1_ 1 f12 c e d1 1 f12 c_u_c2_ 1 f12 c u c3 1 f21 c e d2 1 f21 c_u_c2_ 1 f22 c_e_d2_ 1 f22 c u c1 1 f22 c u c3 1 f31 c_e_d3_ 1 f31 c u c3 1 f32 c_e_d3_ 1 f32 c_u_c1_ 1 f32 c u c2 1 s1 obj 1 s1 c_e_d1_ 1 s2 obj 1 s2 c e d2 1 s3 obj 1 s3 c_e_d3_ 1 ONE VAR CONSTANT obj 0 ONE_VAR_CONSTANT c_e_ONE_VAR_CONSTANT 1 RHS RHS c u budget 10 RHS c_e_d1_ 0 RHS c_e_d2_ 0 RHS c e d3 0 RHS c u c1 0 RHS c_u_c2_ 0 RHS c u c3 0 RHS c e ONE VAR CONSTANT 1 **BOUNDS** LO BOUND x1 0 LO BOUND x2 0 LO BOUND x3 0 LO BOUND f11 0 LO BOUND f12 0 LO BOUND f21 0 LO BOUND f22 0 LO BOUND f31 0

```
LO BOUND f32 0
LO BOUND s1 0
LO BOUND s2 0
LO BOUND s3 0
ENDATA
```

network_d.tim TIME network PERIODS IMPLICIT x1 obj TIME1 f11 c_e_d1_ TIME2 ENDATA

```
network d.sto
STOCH retail
INDEP
          DISCRETE
 RHS
          c e d1 0
                         0.05
 RHS
          c_e_d1_ 1
                         0.2
          c_e_d1_2
                         0.5
 RHS
 RHS
          c_e_d1_ 3
                         0.2
 RHS
          c_e_d1_ 5
                         0.05
                         0.05
 RHS
          c_e_d2_ 0
          c_e_d2_ 1
c_e_d2_ 2
 RHS
                         0.2
 RHS
                         0.5
          c_e_d2_3
 RHS
                         0.2
 RHS
          c_e_d2_5
                         0.05
 RHS
          c e d3 0
                         0.1
          c_e_d3_ 1
 RHS
                         0.4
 RHS
          c_e_d3_ 2
                         0.4
 RHS
          c e d3 3
                         0.1
 x1
         c_u_c1_ -1
                       0.7
                       0.3
 x1
         c_u_c1_ 0
         c_u_c2_ -1
 x2
                       0.7
 x2
         c_u_c2_0
                       0.3
ENDATA
```

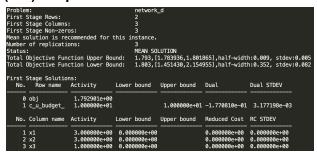
(E-1) Output Console

```
Yului-MacBook: Debug yul$ ./sd
Beginning SD...
Please enter the name of the problem files (eg. `exags1p'):
mkdir: ./sdoutput: File exists
Reading problems from ./sdinput/network_d/network_d.cor
the filename is : /sdinput/network_d/network_d.cor
Specified objective sense: MINIMIZE
Selected objective name: obj
Selected RHS
                 name: RHS
Selected bound
                 name: BOUND
finished loading core file via external Solver.
2.050000
2.050000
1.500000
-0.700000
-0.700000
Vector x_k from the mean problem ::
16.5999999913668489171 4.99999999949014206635
4.9999999935741357149 0.00000000028811109845
```

2.0499999981605007804 0.0000000003835065390 2.0499999981605185440 0.0000000003835065390 0.0000000021140979064 1.44999999969578707848 0.0000000014588982748 0.00000000145889827480.0500000009324059858 1.00000000000000000000 Objective from the mean problem: 0.050000 num_rv = 5; num cipher = 1; ------ Replication No.0 ------++++++++++Warning: Output names have been modified to conform to LP format. Warning: Output names have been modified to conform to LP format +>*******************************: 6.000000 ----- Replication No.1 ------+++++++Warning: Output names have been modified to conform to LP format. Warning: Output names have been modified to conform to LP format. +Warning: Output names have been modified to conform to LP format. >************* Max pi***********: 6.000000 ------ Replication No.2 ------+++++++++Warning: Output names have been modified to conform to LP format. Begin evaluation of average solution obs:10000 mean:1.810400 error: 0.039467 0.95 CI: [1.774675, 1.846125] obs:20000 mean:1.801300 error: 0.027890 0.95 CI: [1.776181, 1.826419] obs:30000 mean:1.803500 error: 0.022772 0.95 CI: [1.782965, 1.824035] obs:40000 mean:1.800275 error: 0.019738 0.95 CI: [1.782508, 1.818042] obs:50000 mean:1.796300 error: 0.017678 0.95 CI: [1.780422, 1.812178] obs:60000 mean:1.792733 error: 0.016168 0.95 CI: [1.778241, 1.807226] obs:70000 mean:1.794871 error: 0.014958 0.95 CI: [1.781448, 1.808295] obs:80000 mean:1.793413 error: 0.014011 0.95 CI: [1.780849, 1.805976]

```
.....
obs:90000 mean:1.795678 error: 0.013189
0.95 CI: [1.783836, 1.807519]
obs:100000 mean:1.797750 error: 0.012518
0.95 CI: [1.786498, 1.809002]
obs:110000 mean:1.794745 error: 0.011945
0.95 CI: [1.784026, 1.805465]
obs:120000 mean:1.794275 error: 0.011441
0.95 CI: [1.784010, 1.804540]
obs:130000 mean:1.792977 error: 0.011004
0.95 CI: [1.783112, 1.802842]
obs:140000 mean:1.791564 error: 0.010600
0.95 CI: [1.782069, 1.801060]
.....
obs:150000 mean:1.793207 error: 0.010239
0.95 CI: [1.784026, 1.802387]
.....
Final Estimate
obs:157253 mean:1.792901
0.95 C.I.: [1.783936, 1.801865]
Ending SD...
Total running time: 6902.213000 ms
```

(E-2) Output Result



Hence, we get the optimal value of x1 = 3.00, x2 = 3.00, x3 = 1.00