```
In [18]:
                                                                                           H
using JuMP
In [139]:
                                                                                           H
using NLPModelsJuMP, NLPModels
In [19]:
                                                                                           H
using GLPK
In [28]:
                                                                                           H
using Distributions
In [30]:
                                                                                           H
using Random
In [153]:
using DataFrames
                                                                                           H
In [154]:
using CSV
In [20]:
H = 4
J = 3
T^d = 1
T^s = 1
\Xi = 1
Out[20]:
1
Ψ
In [24]:
\psi = [100;010;001;100];
In [26]:
\chi = zeros(H,J);
```

```
H
In [38]:
Random.seed! (1234)
\mu = [1.2, 0.8, 0.6]
\Sigma = [0.5 \ 0.0 \ 0.0];
     0.0 0.5 0.0;
     0.0 0.0 0.5
d\xi = reshape(rand(MvNormal(\mu, \Sigma), \Xi), (J,T^s,\Xi))
d\xi = round.(d\xi, digits = 2)
Out[38]:
3×1×1 Array{Float64, 3}:
[:, :, 1] =
 1.81
 0.16
 0.25
In [264]:
                                                                                                     H
z p
Out[264]:
4-element Vector{Int64}:
  7000
  9000
```

In [275]:

```
z^p = [7; 9; 11;7] * 1000
z^c = z^p *2
z^m = fill(0, H, J)
for h=1
         z^{m}[h, 2] = (1/2) * z^{p}[h]
         z^{m}[h, 3] = (3/4) * z^{p}[h]
end
for h=2
         z^{m}[h, 3] = (1/2) * z^{p}[h]
         z^{m}[h, 1] = (3/4) * z^{p}[h]
end
for h=3
         z^{m}[h, 1] = (1/2) * z^{p}[h]
         z^{m}[h, 2] = (3/4) * z^{p}[h]
end
for h=4
         z^{m}[h, 2] = (1/2) * z^{p}[h]
         z^{m}[h, 3] = (3/4) * z^{p}[h]
end
z^{\, m}
```

Out[275]:

```
4×3 Matrix{Int64}:
    0    3500   5250
6750    0   4500
5500   8250    0
    0   3500   5250
```

main model

```
In [259]: ▶
```

```
main = Model(GLPK.Optimizer)
@variable(main, ψ[1:H , 1:J], Bin)
@variable(main, 0 ≤ χ[1:H , 1:J])
@variable(main, α[1:H , 1:J , 1:T<sup>d</sup>], Bin)
@variable(main, -10000000 ≤ θ)

x = all_variables(main)[1 : (H * J * 2) ];
```

```
In [261]:
                                                                                              M
@objective(main, Min, sum(z^p[h] * \psi[h,j] * (T^d + T^s) + z^m[h,j] * \chi[h,j]
                            for h in 1:H for j in 1:J))
UndefVarError: z<sup>™</sup> not defined
Stacktrace:
 [1] macro expansion
   @ C:\Users\e29115\.julia\packages\MutableArithmetics\8xkW3\src\rewrite.j
1:279 [inlined]
 [2] macro expansion
   @ C:\Users\e29115\.julia\packages\JuMP\klrjG\src\macros.jl:1260 [inlined]
 [3] top-level scope
   @ .\In[261]:1
 [4] eval
   @ .\boot.jl:360 [inlined]
 [5] include_string(mapexpr::typeof(REPL.softscope), mod::Module, code::Stri
ng, filename::String)
   @ Base .\loading.jl:1116
```

 $\sum_{h\in\Psi}\sum_{j\in\Phi}z_h^p\psi_{hj}\left\{T^d+T^s\right\}+\sum_{h\in\Psi}\sum_{j\in\Phi}z_{hj}^m\chi_{hj}+\mathbb{E}_{\xi}\left[Q\left(\psi_{hj},\chi_{hj},d_{jt}(\xi)\right)\right]$

sub model

```
γ
```

Stage 1

In []:

```
function sub(\psi, \chi) sub = Model(GLPK.Optimizer) @variable(sub, \alpha[1:H \ , 1:J \ , 1:T^s \ , 1:\Xi], Bin ) @variable(sub, \gamma\xi[1:J \ , 1:T^s \ , 1:\Xi]); @objective(sub, Min, (1/\Xi) * sum(z^c[j] * \gamma\xi[j,t,\xi] for j in 1:J for t in 1:T^s for \xi in con1_S2 = @constraint(sub, demand_met[j in 1:J, t in 1:T^s, \xi in 1:\Xi], sum(\alpha[h,j,t,\xi] * (\psi[h,j] + \chi[h,j]) for h in 1:H ) + \gamma[j,t,\xi] \geq d\xi[j,t,\xi] ); con2_s2 = @constraint(sub, permanent_allocability[h in 1:H, f in 1:J, f in 1:T^s, f in f in 1:f in 1:f
```

```
\begin{aligned} & \underset{\gamma_{jt}(\xi)}{\textit{minimise}} & \quad \frac{1}{n} \sum_{t \in \backslash \mathbf{mT}^s} \sum_{j \in \Phi} \sum_{\xi \in \Xi} z_j^c \gamma_{jt}(\xi) \\ & \sum_{j \in \Phi} \alpha_{hjt}(\xi) \leq 1 & \forall h \in \Psi, \quad \forall t \in \backslash \mathbf{mT}^s \quad \forall \xi \in \Xi = \{1, \dots, n\} \\ & \alpha_{hjt}(\xi) \leq \psi_{hj} + M \chi_{hj} & \forall h \in \Psi \quad \forall j \in \Phi \quad \forall t \in \backslash \mathbf{mT}^s \quad \forall \xi \in \Xi = \{1, \dots, n\} \end{aligned}
```

In [96]:

Out[96]:

```
4×3×1×1 Array{Float64, 4}:

[:,:,1,1] =

1.0 0.0 0.0

0.0 1.0 0.0

0.0 0.0 1.0

1.0 0.0 0.0
```

defining a function for second stage integer dual

when we want to get the dual the x (first stage variables are considered as fixed)

when we want to get the coefficents of x, x should be variable

therefore, we need to define two models one for dual and one for coefficient

In [234]: ▶

```
\# when we want to get the dual the x (first stage variables are considered as fixed)
\# when we want to get the coefficents of x, x should be variable
# therefore, we need to define two models one for dual and one for
function sub_dual(\psi , \chi)
    sub_for_dual = Model(GLPK.Optimizer)
    @variable(sub_for_dual, \alpha[1:H,1:J,1:T^s, 1:\Xi])
    @variable(sub_for_dual, 0 \le \gamma[1:J, 1:T^s, 1:\Xi])
    @objective(sub_for_dual, Min, sum(z^c[j] * \gamma[j,t,\xi] for j in 1:J for t in 1:T<sup>s</sup> for \xi in
    for h in 1:H
            for j in 1:J
                 for t in 1:Ts
                     for \xi in 1:\Xi
                         if 1[h, j, t, \xi] == 0
                             con = @constraint(sub_for_dual, \alpha[h,j,t,\xi] == 0)
                         else
                             con = @constraint(sub_for_dual, \alpha[h,j,t,\xi] == 1)
                         end
                     end
                 end
            end
        end
    con1_S2 = @constraint(sub_for_dual, demand_met[ j in 1:J , t in 1:T<sup>s</sup> , ξ in 1:Ξ],
         d\xi[j,t,\xi] \leq sum(\alpha[h,j,t,\xi] * (\psi[h,j] + \chi[h,j]) for h in 1:H ) + \gamma[j,t,\xi]
    con2_s2 = @constraint(sub_for_dual, permanent_allocability[h in 1:H , j in 1:J , t in 1
                 \alpha[h,j,t,\xi] \leq \psi[h,j] + 10 * \chi[h,j] );
    con3_s2 = @constraint(sub_for_dual, no_more_than_one_station[h in 1:H , t in 1:T<sup>s</sup> , ξ i
                sum( \alpha[h,j,t,\xi] for j in 1:J) \leq 1 );
    optimize!(sub for dual)
   #print(sub_for_dual)
    con_equal = all_constraints(sub_for_dual, AffExpr, MOI.EqualTo{Float64})
    con_less = all_constraints(sub_for_dual, AffExpr, MOI.LessThan{Float64})
    \lambda 1 = dual.(con_equal)
    \lambda 2 = dual.(con less)
    \lambda = append!(\lambda 1, \lambda 2)
    no_con_equal = length(con_equal)
    no_con_less = length(con_less)
    no_all_con = no_con_equal + no_con_less;
    @show no con equal
    @show no_con_less
    @show no all con;
    return λ
end
\psi = [100;010;001;100];
\chi = zeros(H,J)
sub\_dual(\psi, \chi)
```

```
no\_con\_equal = 12
no_con_less = 19
no all con = 31
Out[234]:
31-element Vector{Float64}:
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
```

function for coefficients of ψ and χ

In [224]:

```
function sub coeff()
    sub_for_coeff = Model(GLPK.Optimizer)
                                     \psi[1:H , 1:J], Bin )
    @variable(sub_for_coeff,
    @variable(sub_for_coeff, 0 \le \chi[1:H, 1:J])
    @variable(sub_for_coeff, \alpha[1:H, 1:J, 1:T^s, 1:\Xi])
    @variable(sub_for_coeff, 0 \le \gamma[1:J, 1:T^s, 1:\Xi])
    @objective(sub_for_coeff, Min, sum(z^c[j] * \gamma[j,t,\xi] for j in 1:J for t in 1:T's for \xi i
    for h in 1:H
             for j in 1:J
                  for t in 1:Ts
                      for \xi in 1:\Xi
                           if 1[h,j,t,\xi] == 0
                               con = @constraint(sub_for_coeff, \alpha[h,j,t,\xi] == 0)
                           else
                               con = @constraint(sub_for_coeff, \alpha[h,j,t,\xi] == 1)
                           end
                      end
                  end
             end
         end
    con1_S2 = @NLconstraint(sub_for_coeff, demand_met[ j in 1:J , t in 1:T<sup>s</sup> , ξ in 1:Ξ],
          d\xi[j,t,\xi] \leq sum(\alpha[h,j,t,\xi] * (\psi[h,j] + \chi[h,j]) \text{ for } h \text{ in } 1:H ) + \gamma[j,t,\xi]
    con2_s2 = @constraint(sub_for_coeff, permanent_allocability[h in 1:H , j in 1:J , t in
                  \alpha[h,j,t,\xi] \leq \psi[h,j] + 10 * \chi[h,j]);
    con3_s2 = @constraint(sub_for_coeff, no_more_than_one_station[h in 1:H , t in 1:T<sup>s</sup> , ξ
                sum( \alpha[h,j,t,\xi] for j in 1:J) \leq 1 );
    vr = all_variables(sub_for_coeff)
    vr_index = [vr[i].index.value for i in 1:length(vr)]
    df = DataFrame(variable = vr , index = vr_index);
    #@show df
    nlp = MathOptNLPModel(sub for coeff)
    q = zeros(nlp.meta.nvar)
    jac(nlp, q)
    A1 = jac(nlp, q)[:, 1:24]
    return A1
end
sub coeff()
```

Out[224]:

31×24 SparseArrays.SparseMatrixCSC{Float64, Int64} with 48 stored entries:



```
H
In [246]:
\psi = [ 100; 010; 001; 100];
\chi = zeros(H,J)
sub_dual(\psi, \chi)
sub_coeff()
cut = @constraint(main, \theta \ge \text{sub\_dual}(\psi, \chi)' * sub_coeff()
no\_con\_equal = 12
no\_con\_less = 19
no_all_con = 31
no\_con\_equal = 12
no\_con\_less = 19
no_all_con = 31
Out[246]:
1×24 adjoint(::Vector{Float64}) with eltype Float64:
0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 ... 0.0 0.0 0.0 0.0 0.0 0.0 0.
In [ ]:
                                                                                             H
```