

X[4]=1.0X[5]=1.0X[6]=1.0X[7]=1.0X[8]=1.0X[9]=1.0X[10]=1.0Y[1,1]=-0.0Y[1,2]=-0.0Y[1,3]=-0.0Y[1,4]=-0.0Y[1,5]=1.0Y[1, 6] = -0.0Y[1,7]=-0.0Y[1,8]=-0.0Y[1,9]=1.0Y[1,10]=-0.0Y[1,11]=1.0Y[1,12]=-0.0Y[1,13]=0.0Y[1,14]=-0.0Y[1, 15] = -0.0Y[1,16] = -0.0Y[1,17]=-0.0Y[1, 18] = -0.0Y[1,19]=-0.0Y[1,20]=0.0Y[2,1]=-0.0Y[2,2]=1.0Y[2,3]=-0.0Y[2,4]=-0.0Y[2,5]=-0.0Y[2,6]=-0.0Y[2,7]=-0.0Y[2,8]=-0.0Y[2,9]=-0.0Y[2,10] = -0.0Y[2,11]=-0.0Y[2,12] = -0.0Y[2,13]=-0.0Y[2,14]=-0.0Y[2, 15] = -0.0Y[2,16]=-0.0Y[2,17]=-0.0Y[2,18]=-0.0Y[2,19]=-0.0Y[2,20] = -0.0Y[3,1]=-0.0Y[3,2]=-0.0Y[3,3]=-0.0Y[3,4]=-0.0Y[3,5]=-0.0Y[3, 6] = -0.0Y[3,7]=-0.0Y[3,8]=-0.0Y[3,9]=-0.0Y[3,10]=-0.0Y[3,11]=-0.0Y[3,12]=-0.0Y[3,13]=-0.0Y[3,14] = -0.0Y[3, 15] = -0.0Y[3,16] = -0.0Y[3, 17] = 1.0Y[3,18]=1.0Y[3,19]=-0.0Y[3,20]=-0.0Y[4,1]=0.0Y[4,2]=-0.0Y[4,3]=-0.0Y[4,4] = -0.0Y[4,5] = -0.0Y[4,6] = -0.0Y[4,7]=-0.0Y[4,8] = -0.0Y[4,9]=-0.0Y[4,10] = -0.0Y[4,11]=0.0Y[4,12]=-0.0Y[4,13]=1.0Y[4,14] = -0.0Y[4,15]=-0.0Y[4,16]=-0.0Y[4,17] = -0.0Y[4,18] = -0.0Y[4,19]=-0.0Y[4,20] = -0.0Y[5,1]=-0.0Y[5,2]=-0.0Y[5,3]=-0.0Y[5,4]=1.0Y[5,5]=-0.0Y[5,6]=-0.0Y[5,7]=1.0Y[5,8]=-0.0Y[5, 9] = -0.0Y[5,10]=-0.0Y[5,11]=-0.0Y[5, 12] = -0.0Y[5,13]=-0.0Y[5,14]=-0.0Y[5, 15] = -0.0Y[5,16] = -0.0Y[5,17]=-0.0Y[5, 18] = -0.0Y[5,19]=-0.0Y[5,20]=-0.0Y[6,1]=1.0Y[6,2]=-0.0Y[6,3]=-0.0Y[6,4]=-0.0Y[6,5]=-0.0Y[6,6] = -0.0Y[6,7]=-0.0Y[6,8]=-0.0Y[6,9]=-0.0Y[6,10]=0.0Y[6,11]=-0.0Y[6,12] = -0.0Y[6,13]=0.0Y[6,14]=1.0Y[6, 15] = -0.0Y[6,16]=-0.0Y[6,17] = -0.0Y[6,18] = -0.0Y[6,19]=-0.0Y[6,20]=0.0Y[7,1]=-0.0Y[7,2]=-0.0Y[7,3]=1.0Y[7,4]=-0.0Y[7,5]=-0.0Y[7,6] = -0.0Y[7,7] = -0.0Y[7,8]=1.0Y[7,9]=-0.0Y[7,10]=-0.0Y[7,11] = -0.0Y[7,12]=-0.0Y[7,13] = -0.0Y[7,14] = -0.0Y[7, 15] = -0.0Y[7,16] = -0.0Y[7, 17] = -0.0Y[7,18] = -0.0Y[7,19]=-0.0Y[7,20]=-0.0Y[8,1]=-0.0Y[8,2]=-0.0Y[8,3]=-0.0Y[8,4]=-0.0Y[8,5] = -0.0Y[8,6]=1.0Y[8,7]=-0.0Y[8,8]=-0.0Y[8,9]=-0.0Y[8,10] = -0.0Y[8,11]=-0.0Y[8,12]=-0.0Y[8,13]=-0.0Y[8,14]=-0.0Y[8, 15] = 1.0Y[8,16]=-0.0Y[8,17]=-0.0Y[8,18]=-0.0Y[8,19]=-0.0Y[8,20]=1.0Y[9,1]=0.0Y[9,2]=-0.0Y[9,3]=-0.0Y[9,4]=-0.0Y[9,5]=-0.0Y[9,6]=-0.0Y[9,7]=-0.0Y[9,8]=-0.0Y[9,9]=-0.0Y[9,10]=1.0Y[9,11]=-0.0Y[9,12]=-0.0Y[9,13]=-0.0Y[9,14]=-0.0Y[9, 15] = -0.0Y[9,16]=-0.0Y[9,17]=-0.0Y[9,18]=-0.0Y[9,19]=1.0Y[9,20]=0.0Y[10,1]=-0.0Y[10,2]=-0.0Y[10,3]=-0.0Y[10,4]=-0.0Y[10,5]=-0.0Y[10,6]=-0.0Y[10,7]=-0.0Y[10,8]=-0.0Y[10,9]=-0.0Y[10,10]=-0.0Y[10,11]=-0.0Y[10,12]=1.0Y[10,13] = -0.0Y[10,14]=-0.0Y[10, 15] = -0.0Y[10,16]=1.0Y[10,17] = -0.0Y[10, 18] = -0.0Y[10,19]=-0.0Y[10,20] = -0.0arcos_activos=[k for k in arcos if Y[k].x>0.99] model FLP b.ObjVal Out[23]: 1137338.1398864486 In [24]: model FLP b.Runtime Out[24]: 0.051860809326171875 model FLP b relaxed = model FLP b.relax() model FLP b relaxed.optimize() Gurobi Optimizer version 9.1.2 build v9.1.2rc0 (win64) Thread count: 4 physical cores, 8 logical processors, using up to 8 threads Optimize a model with 41 rows, 210 columns and 630 nonzeros Model fingerprint: 0x9405ff5c Coefficient statistics: Matrix range [1e+00, 8e+02] Objective range [2e+03, 1e+05] [1e+00, 1e+00] Bounds range RHS range [1e+00, 1e+00] Presolve time: Presolved: 41 rows, 210 columns, 630 nonzeros Time Iteration Objective Primal Inf. Dual Inf. 7.9767608e+05 2.273438e+02 0.000000e+00 1.1076967e+06 0.000000e+00 0.000000e+00 0 Solved in 39 iterations and 0.01 seconds Optimal objective 1.107696670e+06 for v in model FLP b relaxed.getVars(): print(str(v.VarName)+'='+str(round(v.x,2))) X[1]=1.0X[2]=1.0X[3]=1.0X[4] = 0.0X[5]=1.0X[6] = 0.82X[7]=1.0X[8]=1.0X[9]=1.0X[10] = 0.79Y[1,1]=0.0Y[1,2]=0.0Y[1,3]=0.0Y[1,4]=0.0Y[1,5]=0.59Y[1,6]=0.0Y[1,7]=0.0Y[1,8]=0.0Y[1,9]=1.0Y[1,10]=0.0Y[1,11]=1.0Y[1,12]=0.0Y[1,13]=0.0Y[1,14]=0.0Y[1, 15] = 0.0Y[1,16]=0.0Y[1,17]=0.0Y[1,18]=0.0Y[1,19]=0.0Y[1,20]=0.43Y[2,1]=0.0Y[2,2]=1.0Y[2,3]=0.0Y[2,4]=0.0Y[2,5]=0.41Y[2,6]=0.0Y[2,7]=0.0Y[2,8]=0.0Y[2,9]=0.0Y[2,10]=0.0Y[2,11]=0.0Y[2,12]=0.0Y[2,13]=0.0Y[2,14]=0.0Y[2,15]=0.0Y[2,16]=0.0Y[2,17]=0.0Y[2,18]=0.0Y[2,19]=0.0Y[2,20]=0.0Y[3,1]=0.0Y[3,2]=0.0Y[3,3]=0.0Y[3,4]=0.0Y[3,5]=0.0Y[3, 6] = 0.0Y[3,7]=0.0Y[3,8]=0.0Y[3,9]=0.0Y[3,10]=0.0Y[3,11]=0.0Y[3,12]=0.0Y[3,13]=0.0Y[3,14]=0.0Y[3, 15] = 0.0Y[3,16]=0.12Y[3.17]=1 0 Y[3, 18] = 1.0Y[3,19]=0.0Y[3,20]=0.0Y[4,1]=0.0Y[4,2]=0.0Y[4,3]=0.0Y[4,4]=0.0Y[4,5]=0.0Y[4,6]=0.0Y[4,7]=0.0Y[4,8]=0.0Y[4,9]=0.0Y[4,10]=0.0Y[4,11]=0.0Y[4,12]=0.0Y[4,13]=0.0Y[4,14]=0.0Y[4,15]=0.0Y[4,16]=0.0Y[4,17]=0.0Y[4,18]=0.0Y[4,19]=0.0Y[4,20]=0.0Y[5,1]=0.0Y[5,2]=0.0Y[5,3]=0.0Y[5,4]=0.58Y[5,5]=0.0Y[5,6]=0.0Y[5,7]=1.0Y[5,8]=0.0Y[5, 9] = 0.0Y[5,10]=0.0Y[5,11]=0.0Y[5, 12] = 0.0Y[5,13]=0.0Y[5,14]=0.0Y[5, 15] = 0.0Y[5,16]=0.88Y[5,17]=0.0Y[5, 18] = 0.0Y[5,19]=0.0Y[5,20]=0.0Y[6,1]=0.0Y[6,2]=0.0Y[6,3]=0.51Y[6,4]=0.0Y[6,5]=0.0Y[6,6]=0.0Y[6,7]=0.0Y[6,8]=0.0Y[6,9]=0.0Y[6,10]=0.0Y[6,11]=0.0Y[6,12]=0.0Y[6,13]=0.0Y[6,14]=1.0Y[6, 15] = 0.0Y[6,16]=0.0Y[6,17]=0.0Y[6,18]=0.0Y[6,19]=0.0Y[6,20]=0.36Y[7,1]=0.0Y[7,2]=0.0Y[7,3]=0.0Y[7,4]=0.0Y[7,5]=0.0Y[7,6]=0.0Y[7,7]=0.0Y[7,8]=1.0Y[7,9]=0.0Y[7,10]=0.0Y[7,11]=0.0Y[7,12]=0.0Y[7,13]=0.0Y[7,14]=0.0Y[7, 15] = 0.0Y[7,16]=0.0Y[7.17]=0.0Y[7,18]=0.0Y[7,19]=0.72Y[7,20]=0.0Y[8,1]=0.0Y[8,2]=0.0Y[8,3]=0.0Y[8,4]=0.0Y[8,5]=0.0Y[8,6]=1.0Y[8,7]=0.0Y[8,8]=0.0Y[8,9]=0.0Y[8,10]=0.0Y[8,11]=0.0Y[8,12]=0.0Y[8,13]=1.0Y[8,14]=0.0Y[8,15]=1.0Y[8,16]=0.0Y[8,17]=0.0Y[8,18]=0.0Y[8,19]=0.0Y[8,20]=0.21Y[9,1]=1.0Y[9,2]=0.0Y[9,3]=0.49Y[9,4]=0.0Y[9,5]=0.0Y[9,6]=0.0Y[9,7]=0.0Y[9,8]=0.0Y[9, 9] = 0.0Y[9,10]=1.0Y[9,11]=0.0Y[9,12]=0.0Y[9,13]=0.0Y[9,14]=0.0Y[9,15]=0.0Y[9,16]=0.0Y[9,17]=0.0Y[9,18]=0.0Y[9,19]=0.28Y[9,20]=0.0Y[10,1]=0.0Y[10,2]=0.0Y[10,3]=0.0Y[10,4]=0.42Y[10,5]=0.0Y[10, 6] = 0.0Y[10,7]=0.0Y[10,8]=0.0Y[10, 9] = 0.0Y[10,10]=0.0Y[10,11]=0.0Y[10, 12] = 1.0Y[10,13]=0.0Y[10,14]=0.0Y[10, 15] = 0.0Y[10,16]=0.0Y[10,17]=0.0Y[10, 18] = 0.0Y[10,19]=0.0Y[10,20]=0.0model_FLP_b_relaxed.ObjVal Out[27]: 1107696.6702895337 model_FLP_b_relaxed.Runtime Out[28]: 0.013963699340820312 In [29]: GAP = (model_FLP_b.ObjVal-model_FLP_b_relaxed.ObjVal)/model_FLP_b.ObjVal round(GAP*100, 2) Out[29]: 2.61 **Grafico Solucion** plt.figure(figsize=(14,11)); plt.scatter(x_bodegas, y_bodegas, color="blue", marker="s"); plt.scatter(x_clientes, y_clientes, color="red"); for i in range(len(I)): $plt.annotate("\$Bodega_{\$d}\$"\$(i+1),(x_bodegas[i]+1,y_bodegas[i]-1),size=13, color="blue");$ for j in range(len(J)): plt.annotate("\$Cliente_{%d}\$"%(j+1),(x_clientes[j]+1,y_clientes[j]-1.8),size=13, color="red",weight="bold") for n in arcos_activos: i=n[0] j=n[1] plt.plot([x_bodegas[i-1],x_clientes[j-1]],[y_bodegas[i-1],y_clientes[j-1]],color="green") plt.xlabel("Eje x"); plt.ylabel("Eje y"); plt.title("Solución \$Y_{ij}\$"); plt.show(); Solución Yij 100 Cliente₁₉ Cliente₂₀ Cliente₁ Cliente₁₀ Cliente₃ Bodega₇ Bodegag Cliente₁₄ 80 **■**Bodega₆ Cliente₈ Bodega₁ 60 Cliente₅ Cliente₁₁ Cliente₉ ■ Bodega₂ ■ Bodega₄ Cliente₂ Bodega₈ Cliente₇ Cliente₆ Cliente₁₅ 40 Cliente₁₃ Cliente₄ Cliente₁₇ 20 ■ Bodega₅ Bodega₃ Cliente Bodega₁₀ Cliente₁₈ Cliente₁₆ 80 40 Eje x Parte II: Lot-Sizing Problem Capacitado Mono-Producto a) Caso variable: (8)**Conjuntos:** • Periodos (T) = 10 Parámetros: P_t = Costo unitario de producción en el periodo t F_t = Costo fijo de ordenar en el periodo t H_t = Costo unitario de almacenamiento en el periodo t D_t = Demanda de unidades para el periodo t M_t = Valor mas grande posible para el periodo t $M_t = \sum_{\scriptscriptstyle a=t}^T D_s \qquad orall t = 1, \ldots, T$ (9)np.random.seed(2) T = [t+1 for t in range(10)] $P = \{t:np.random.randint(100,200) \text{ for } t \text{ in } T\}$ $K = \{t:np.random.randint(1000,2000) \text{ for } t \text{ in } T\}$ $H = \{t:np.random.randint(50,200) \text{ for } t \text{ in } T\}$ $D = \{t:np.random.randint(1000,5000) for t in T\}$ sumador=0 for t in range(len(T)): sumador+=D[t+1] Q =sumador/len(D) $M = \{t:sum(list(D.values())[t-1:])$ for t in $T\}$ Variables de decisión: X_t = Unidades a producir en el periodo t Y_t = 1, si se produce en el periodo t I_t = Unidades almacenadas en el periodo t model LZP a = Model('model LZP a') #model_LZP_a.setParam('LogToConsole', 0) X = model LZP a.addVars(T, vtype=GRB.CONTINUOUS, name='X') Y = model LZP a.addVars(T, vtype=GRB.BINARY, name='Y') I = model LZP a.addVars([0]+T, vtype=GRB.CONTINUOUS, name='I') Modelo: $Min\sum_{t=1}^{T}P_{t}*X_{t}+\sum_{t=1}^{T}K_{t}*Y_{t}+\sum_{t=1}^{T}H_{t}*I_{t}$ (10)(11) $I_t - X_t - I_{t-1} = -D_t \qquad \forall t = 1, \dots, T$ (12) $X_t \leq M_t * Y_t \qquad orall t = 1, \ldots, T$ (13) $X_t \leq Q \qquad orall t = 1, \ldots, T$ (14) $I_0,I_T=0$ (15) $X_t, I_t \geq 0 \qquad orall t = 1, \dots, T$ (16) $Y_t \in \{0,1\}$ (17)In [34]: objective = None objective = quicksum(P[t]*X[t] for t in T)+quicksum(K[t]*Y[t] for t in T)+quicksum(H[t]*I[t] for t in T) model LZP a.setObjective(expr=objective, sense=GRB.MINIMIZE) $model LZP \ a.addConstrs(I[t]-X[t]-I[t-1]==-1*D[t] \ for \ t \ in \ T)$ model LZP a.addConstrs(X[t] <= M[t] *Y[t] for t in T)</pre> model LZP a.addConstrs(X[t] <= Q for t in T)</pre> model LZP a.addConstr(I[0]==0) model LZP a.addConstr(I[T[-1]]==0) model LZP a.update() model LZP a.optimize() Gurobi Optimizer version 9.1.2 build v9.1.2rc0 (win64) Thread count: 4 physical cores, 8 logical processors, using up to 8 threads Optimize a model with 32 rows, 31 columns and 62 nonzeros Model fingerprint: 0x22427338 Variable types: 21 continuous, 10 integer (10 binary) Coefficient statistics: [1e+00, 3e+04] Matrix range Objective range [5e+01, 2e+03] Bounds range [1e+00, 1e+00] RHS range [1e+03, 5e+03] RHS range Presolve removed 32 rows and 31 columns Presolve time: 0.00s Presolve: All rows and columns removed Explored 0 nodes (0 simplex iterations) in 0.01 seconds Thread count was 1 (of 8 available processors) Solution count 1: 5.78025e+06 Optimal solution found (tolerance 1.00e-04) Best objective 5.780249000000e+06, best bound 5.780249000000e+06, gap 0.0000% for v in model LZP a.getVars(): print(str(v.VarName)+'='+str(round(v.x,2))) X[1] = 3157.0X[2] = 3157.0X[3] = 3157.0X[4] = 3157.0X[5] = 3157.0X[6] = 3157.0X[7] = 3157.0X[8] = 3157.0X[9] = 3157.0X[10] = 3157.0Y[1]=1.0Y[2]=1.0Y[3]=1.0Y[4]=1.0Y[5]=1.0Y[6]=1.0Y[7]=1.0Y[8]=1.0Y[9]=1.0Y[10]=1.0I[0]=0.0I[1]=1902.0I[2] = 664.0I[3]=1472.0I[4] = 2517.0I[5] = 894.0I[6] = 1725.0I[7] = 356.0I[8] = 1138.0I[9]=12.0I[10]=0.0produccion=[X[k].x for k in T] inventario=[I[k].x for k in T] produccion, inventario Out[36]: ([3157.0, 3157.0, 3157.0, 3157.0, 3157.0, 3157.0, 3157.0, 3157.0, 3157.0, 3157.0], [1902.0, 664.0, 1472.0, 2517.0, 894.0, 1725.0, 356.0, 1138.0, 12.0, 0.0]) model LZP a.ObjVal Out[37]: 5780249.0 model LZP a.Runtime Out[38]: 0.01296234130859375 model_LZP_a_relaxed = model_LZP_a.relax() model_LZP_a_relaxed.optimize() Gurobi Optimizer version 9.1.2 build v9.1.2rc0 (win64) Thread count: 4 physical cores, 8 logical processors, using up to 8 threads Optimize a model with 32 rows, 31 columns and 62 nonzeros Model fingerprint: 0x5e60b4de Coefficient statistics: Matrix range [1e+00, 3e+04] Objective range [5e+01, 2e+03] [1e+00, 1e+00] [1e+uu, _. [1e+03, 5e+03] Bounds range RHS range Presolve removed 32 rows and 31 columns Presolve time: 0.01s Presolve: All rows and columns removed Iteration Objective Primal Inf. Dual Inf. Time 5.7693979e+06 0.000000e+00 0.000000e+00 Solved in 0 iterations and 0.01 seconds Optimal objective 5.769397856e+06 In [40]: for v in model LZP a relaxed.getVars(): print(str(v.VarName)+'='+str(round(v.x,2))) X[1] = 3157.0X[2] = 3157.0X[3] = 3157.0X[4] = 3157.0X[5] = 3157.0X[6] = 3157.0X[7] = 3157.0X[8] = 3157.0X[9] = 3157.0X[10] = 3157.0Y[1]=0.1Y[2]=0.1Y[3] = 0.12Y[4] = 0.13Y[5] = 0.15Y[6] = 0.19Y[7] = 0.22Y[8] = 0.32Y[9] = 0.42Y[10]=1.0I[0] = 0.0I[1]=1902.0I[2] = 664.0I[3]=1472.0I[4] = 2517.0I[5] = 894.0I[6]=1725.0I[7] = 356.0I[8]=1138.0I[9]=12.0I[10]=0.0In [41]: model LZP a relaxed.ObjVal Out[41]: 5769397.856098097 In [42]: model LZP a relaxed.Runtime Out[42]: 0.008975982666015625 In [43]: GAP = (model LZP a.ObjVal-model LZP a relaxed.ObjVal)/model LZP a.ObjVal round (GAP*100,2) Out[43]: 0.19 **Grafico Solucion** In [44]: fig = plt.figure(figsize=(15,5)) ax = fig.add subplot(121) ax.plot(T,produccion,label="Produccion (unidades)") ax.set title("Produccion \$x_{t}\$") $ax.set_ylim(0.0)$ ax.set xlim(1) ax.legend() ax = fig.add subplot(122)ax.plot(T,inventario,label="Inventario (unidades)") ax.set_title("Inventario \$I {t}\$") ax.set xlim(1) ax.legend() Out[44]: <matplotlib.legend.Legend at 0x1e5db041a90> Inventario It Produccion x_t 2500 Inventario (unidades) 3000 2000 2500 2000 1500 1500 1000 1000 500 500 Produccion (unidades) 0 10 10 b) Caso variable: W_{ts} (18)In [45]: X, Y, I = None, None, None In [46]: model LZP b = Model('Model LZP b') #model_LZP_b.setParam('LogToConsole',0) In [47]: TS = [(t,s) for t in T for s in T[t-1:]]In [48]: W = model_LZP_b.addVars(TS, vtype=GRB.CONTINUOUS, name='W') Y = model_LZP_b.addVars(T, vtype=GRB.BINARY, name='Y') I = model_LZP_b.addVars([0]+T, vtype=GRB.CONTINUOUS, name='I') objective = None $objective = quicksum(P[t]*W[(t,s)] \ \textbf{for} \ t,s \ \textbf{in} \ TS) + quicksum(K[t]*Y[t] \ \textbf{for} \ t \ \textbf{in} \ T) + quicksum(H[t]*I[t] \ \textbf{for} \ T) + quicksum(H[t]*I[t] \$ model LZP b.setObjective(expr=objective, sense=GRB.MINIMIZE) $\bmod LZP \ b.addConstrs((I[t]-quicksum(W[(t,s)] \ \textbf{for} \ s \ \textbf{in} \ T[t-1:]) - I[t-1] == -1*D[t] \ \textbf{for} \ t \ \textbf{in} \ T), \ name = '(1)')$ model LZP b.addConstrs((quicksum(W[(i,t)] for i in T[:t]) == D[t] for t in T), name='(2)') model LZP b.addConstrs(($W[(t,s)] \le D[s] \times Y[t]$ for t,s in TS), name='(3)') model LZP b.addConstrs((quicksum(W[(t,s)] for s in T[t-1:]) <= Q for t in T), name='(4)') model LZP b.addConstr(I[0]==0, name='(5)') model LZP b.addConstr(I[T[-1]]==0, name='(6)') model LZP b.update() model LZP b.optimize() Gurobi Optimizer version 9.1.2 build v9.1.2rc0 (win64) Thread count: 4 physical cores, 8 logical processors, using up to 8 threads Optimize a model with 87 rows, 76 columns and 297 nonzeros Model fingerprint: 0xd71a53b1 Variable types: 66 continuous, 10 integer (10 binary) Coefficient statistics: [1e+00, 5e+03] Matrix range Objective range [5e+01, 2e+03] Bounds range [1e+00, 1e+00] RHS range [1e+03, 5e+03] Presolve removed 79 rows and 54 columns Presolve time: 0.00s Presolved: 8 rows, 22 columns, 44 nonzeros Variable types: 22 continuous, 0 integer (0 binary) Root relaxation: objective 5.780249e+06, 5 iterations, 0.00 seconds Current Node Objective Bounds Expl Unexpl | Obj Depth IntInf | Incumbent BestBd Gap | It/Node Time 5780249.0000 5780249.00 0.00% - Os * 0 0 Explored 0 nodes (5 simplex iterations) in 0.01 seconds Thread count was 8 (of 8 available processors) Solution count 1: 5.78025e+06 Optimal solution found (tolerance 1.00e-04) Best objective 5.780249000000e+06, best bound 5.780249000000e+06, gap 0.0000% for v in model_LZP_b.getVars(): print(str(v.VarName)+'='+str(round(v.x,2)))

X[1]=1.0 X[2]=1.0X[3]=1.0





