

# A Modified Formulation of Optimal Source–Sink Matching in Carbon Capture and Storage Systems with Time, Injection Rate and Capacity Constraints

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Carbon capture and storage (CCS) is considered as one of the leading options in mitigating carbon dioxide emissions into the atmosphere. It involves collecting and compressing of relatively pure carbon dioxide from a given source, then storing it into a various locations or also called sinks. Multiple variables limits the implementation of CCS such as injection rate limit of sink, capacity of sink, start and end of operating life of source and sink, and other socio-economical aspects that the CCS retrofit will do to the existing carbon source facility. The proposed approach focuses on temporal issues where operating lives of sources and sink may not completely coincide. In 2012, Tan et al. developed an optimization model for optimal source sink matching in carbon capture and storage systems with time, injection rate, and capacity constraints. This paper improves the existing optimization model by eliminating inconsistencies in the optimization results. Previous case studies presented by Tan, et al. were re-evaluated and a new case study was added to further show improvement made by the modified model. The new model provides more accurate and robust results compared to the original model and eliminates inconsistencies that required human intervention in order to extract the correct conclusion. It also simplified the model, requiring lesser variables and solver iterations, and making it more suited for large scale set up optimization.

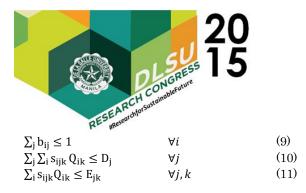
Keywords: Carbon capture and storage, source–sink matching, mixed integer linear programming

# 1. INTRODUCTION

In 2012, Tan et al.<sup>(1)</sup> developed an optimization model for optimal source sink matching in carbon capture and storage systems with time, injection rate, and capacity constraints. This paper aims to improve the existing optimization model and eliminate inconsistencies in the optimization results.

The summarized optimization model by Tan et al. is as follows (eq.  $1 \sim 11$ ):

$\max \sum_k \sum_j \sum_i Q_{ik} s_{ijk}$		(1)
$b_{ijk}S_i^L \leq s_{ijk} \leq b_{ijk}S_i^U$	$\forall i, j, k$	(2)
$\mathbf{b}_{ijk} \in \{0,1\}$	$\forall i, j, k$	(3)
$b_{ijk} \le b_{ij(k+1)}$	$\forall i,j \;\; \forall k,k+1 \; \in K$	(4)
$s_{ijk} \le s_{ij(k+1)}$	$\forall i,j \ \forall k,k+1 \in K$	(5)
$s_{ij(k+1)} - s_{ijk} \le (b_{ij(k+1)} - b_{ijk})$	o <sub>ijk</sub> )S <sup>U</sup>	
	$\forall i,j \ \forall k,k+1 \in K$	(6)
$b_{ij}T^{L} \leq \sum_{k} Q_{ik}b_{ijk} \leq b_{ij}T^{U}$	∀i,j	(7)
$b_{ij} \in \{0,1\}$	$\forall i, j$	(8)



In the model by Tan et al., three decision variables were utilized namely  $b_{ij}$ ,  $b_{ijk}$  and  $s_{ijk}$ . These variables dictate where the connection, b, will be from source ito sink j, on what period k they will connect, and what will be the allowed maximum CO<sub>2</sub> rate, s, will be obtained from the source. These variables are correlated by the following constraints:

$$b_{ijk}S_i^{L} \le s_{ijk} \le b_{ijk}S_i^{U} \qquad \forall i, j, k \qquad (2)$$
  
Sinc  $\le S_{ii}(r, t) \qquad \forall i, i \ \forall k, k+1 \in K \qquad (5)$ 

$$\begin{aligned} \delta_{ijk} &\leq s_{ij(k+1)} & \forall i, j \quad \forall k, k+1 \in K \quad (5) \\ S_{ijk} &= S_{ijk} \leq (b_{ijk}, c_{ij}) - b_{ijk} S_{ijk}^U \end{aligned}$$

$$\begin{array}{ccc} s_{ij(k+1)} & s_{ijk} \geq (s_{ij(k+1)} & s_{ijk})s_i \\ & \forall i, j & \forall k, k+1 \in K \quad (6) \\ s_{ij}T^L \leq \sum_k Q_{ik} s_{ijk} \leq s_{ij}T^U & \forall i, j \quad (7) \end{array}$$

Equation 2 limits  $s_{ijk}$  so that it will only have a value between  $S_{i^L}$  and  $S_{i^U}$  while the multiplier  $b_{ijk}$  ensures that there will only be a flow rate when a connection is present between source and sink. Equations 6 and 5 combined limit the value of  $s_{ijk}$  such that it will retain a common value for the whole connection period between source and sink. Equation 7 ensures that sum of the periods where an active connection between source *i* to sink *j* for periods k ( $\Sigma_k Q_{ik} b_{ijk}$ ) is within the minimum  $T^L$  and maximum  $T^U$  viable period considered for connection between source *i* to sink *j* to be established ( $b_{ij}$ =1).

Decision variables  $b_{ij}$  and  $b_{ijk}$  both represent a connection between source *i* to sink *j* with the latter having an additional dimension which is the specific period, *k* (every 5 years). It is therefore expected that

the conclusions obtained from these variables must be consistent.

### 2. METHODOLOGY

### CASE STUDY 1

Using Case Study 1 presented by Tan et al., the results were reproduced using Lingo v.14 platform<sup>(2)</sup>. The parameters and results are as follows:

Table 1: Case Study 1 Parameters (From Tan et al.)

Sources	Flow Rate	Time of	Maximum
	(Mt/y)	Flow (y)	Capture (Mt)
1	10	0-20	200
2	2.5	0-30	75
3	4	0-30	120
4	4	0-25	100
5	6	10-30	120
Total	26.5	n/a	615
Sinks	Injection	Start	Maximum
	Limit (Mt/y)	Time (y)	Storage (Mt)
А	10	0	400
В	10	5	500
Total	20	n/a	900

Table 2:	Case	Study	1	Results	(From	Tan et a	1.)

	testites (110im 1	an ee an,
Sink A	Sink B	Source Summary
10 Mt/y	0	200 Mt (100%)
(t=0 to 20)		
0	0	0 Mt (0%)
0	4 Mt/y	100 Mt (83%)
	(t=5 to 30)	
0	0	0 Mt (0%)
0	6 Mt/y	120 Mt (100%)
	(t=10 to 30)	
mary		
200 Mt (50%)	220 Mt (44%)	420 Mt of CO2
2	00 Mt (50%)	00 Mt (50%) 220 Mt (44%)

Table 3: Case Study 1 bij Variable Results	(Original Optimization Model Raw Results)

	j(1)	j(2)
i (1)	1	0
i (2)	0	0
i (3)	0	1
i (4)	0	0
i (5)	0	1

#### Table 4: Case Study 1 bijk Variable Results (Original Optimization Model Raw Results)

		Sink 1 (j=1)							Sink 2	2 ( j=2)		
	k(1)	k(2)	k(3)	k(4)	k(5)	k(6)	k(1)	k(2)	k(3)	k(4)	k(5)	k(6)
i (1)	1	1	1	1	1	1	0	0	0	0	0	1
i (2)	0	0	0	0	0	0	0	0	0	0	0	0
i (3)	0	0	0	0	0	0	0	1	1	1	1	1
i (4)	0	0	0	0	0	1	0	0	0	0	0	1
i (5)	0	0	0	0	0	0	0	0	1	1	1	1



 Table 5: Case Study 1 s<sub>ijk</sub> Variable Results (Original Optimization Model Raw Results)

			Sink 1	L (j=1)					Sink 2	2 ( j=2)		
j=1	k(1)	k(2)	k(3)	k(4)	k(5)	k(6)	k(1)	k(2)	k(3)	k(4)	k(5)	k(6)
i (1)	10	10	10	10	10	10	0	0	0	0	0	10
i (2)	0	0	0	0	0	0	0	0	0	0	0	0
i (3)	0	0	0	0	0	0	0	4	4	4	4	4
i (4)	0	0	0	0	0	4	0	0	0	0	0	4
i (5)	0	0	0	0	0	0	0	0	6	6	6	6
$\Sigma_{\mathrm{i}} s_{ijk}$	10	10	10	10	10	14	0	4	10	10	10	24
$\Sigma_{\mathrm{i}} \mathcal{S}_{ijk} Q_{ik}$	10	10	10	10	0	0	0	4	10	10	10	10

We can observe that only the variable  $b_{ij}$  was consistent with the optimization results presented by Tan et al. Although manual selection of values from the results can be done and the same optimized results can be obtained, such task is prone to human error and shows weakness in the optimization model since it cannot, by itself, provide the optimized solution to the problem.

### MODIFICATIONS OF THE OPTIMIZATION MODEL

The optimization model results show multiple inconsistencies between  $b_{ijk}$  and  $b_{ij}$  for Sources 1 and 4. It also violated restrictions set upon by equations 7 and 9 where a connection should only exist from one source to a sink with a minimum viable period of operation. This inconsistency was due to the lack of restriction on variable  $b_{ijk}$ . This can be traced back to equation 7 and 9.

$$b_{ij}T^{L} \leq \sum_{k} Q_{ik}b_{ijk} \leq b_{ij}T^{U} \quad \forall i, j$$
(7)

$$\sum_{j} b_{ij} \le 1 \qquad \qquad \forall i \qquad (9)$$

The restriction in equation 7 has a loophole where when the value of  $b_{ij}$  and  $Q_{ik}$  happens to be both zero (0),  $b_{ijk}$  variables after last period of  $Q_{ik}=1$  can obtain a value of 1 or 0 without violating any constraints. Equation 9 is also rendered useless in restraining the values of  $b_{ijk}$  due to the loophole in equation 7. This is what happened in Case study 1 where the last period of Source 1 and Source 4 obtained a value of 1. This loophole can be neutralized by modifying equation 7 and 9 to 7' and 9'.

$$\begin{split} b_{ijk_{(Last)}} T^{L} &\leq \sum_{k} Q_{ik} b_{ijk} \leq b_{ijk_{(Last)}} T^{U} \qquad \forall i,j \qquad (7') \\ \sum_{j} b_{ijk_{(Last)}} \leq 1 \qquad \qquad \forall i \qquad (9') \end{split}$$

The variable  $b_{ij}$  is eliminated from the model and replaced by  $b_{ijk(\text{Last})}$ . The variable  $b_{ij}$  in equation 7 and 9 was used as a representative for the connection between source *i* to sink *j*. The variable  $b_{ijk(\text{Last})}$  can also be used as a representative since it is assumed that the connection between source *i* to sink *j* will be retained up to the end of the planning horizon (last period *k*). This is reinforced by equation 4. By this modification, even if  $Q_{ik}$  and  $b_{ijk}$  are zero (0),  $b_{ijk}$  will be forced to take a zero value since equation 4 will force all  $b_{ijk}$  before (Last) to also be zero (0).

$$b_{ijk} \le b_{ij(k+1)} \qquad \forall i, j \ \forall k, k+1 \in K \quad (4)$$

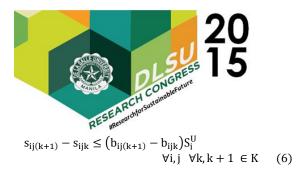
The current constraints for the variable  $s_{ijk}$  allow it to have a positive value even if the source no longer provides CO<sub>2</sub> to the sink. This was allowed since  $s_{ijk}$ variable is multiplied by  $Q_{ik}$  which zeroes out the value when the source stops operating in equations 1, 10, and 11. This result to  $s_{ijk}$  values that are inconsistent with the final conclusion obtained from the optimization model. This is reflected in source 1 where  $s_{ijk}$  retains a value of 10 for periods k=5 and k=6 even if it is no longer operational. This inconsistency is eliminated by reformulating how  $s_{ijk}$ is constrained in the model. Refer to the following:

$$\begin{aligned} & b_{ijk}S_i^L \leq s_{ijk} \leq b_{ijk}S_i^U & \forall i, j, k \\ & b_{iik}Q_{ik}S_i^L \leq s_{iik} \leq b_{iik}Q_{ik}S_i^U & \forall i, j, k \end{aligned}$$

Equation 2 restricts  $s_{ijk}$  to have a value between  $S_i^{l}$  and  $S_i^{l}$  while the factor  $b_{ijk}$  dictates where it will be a positive or a zero value. The proposed equation 2' reinforces the restriction by including the factor  $Q_{ik}$ . This will force the variable  $s_{ijk}$  to have a positive value only during the time source i is operational. This will contradict with the existing equation 5 where it forces  $s_{ijk}$  to keep its positive value up to the end of the planning horizon. It is therefore necessary to modify this equation to equation 5' in order to allow  $s_{ijk}$  to have a zero value once source i stops operating before the end of the planning horizon.

$$\begin{aligned} s_{ijk} &\leq s_{ij(k+1)} & \forall i, j \ \forall k, k+1 \in K \quad (5) \\ s_{iik} &* Q_{i(k+1)} \leq s_{ii(k+1)} & \forall i, j \ \forall k, k+1 \in K \quad (5') \end{aligned}$$

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$\sum_{j} \sum_{i} s_{ijk} \le D_j$	∀j	(10')
$\sum_{i} s_{ijk} \le E_{jk}$	∀j, k	(11')

Another effect of equation 2' is the simplification of equations 1, 10, and 11 since the factor  $Q_{ik}$  will no longer be necessary. This results to equations 1', 10', and 11'.

$\max \sum_{\mathbf{k}} \sum_{j} \sum_{i} s_{ijk}$	(1')
Table 6: Ontimization Model Comparison	Summary

A comparison table of the original optimizat	tion
model and the modified optimization model can	be
seen in table 6. Both the original model and	the
modified model are calculated as a Mixed Inte	ger
Linear Program (MILP).	-

Original Model from Tan et al.			Modified Model		
MILP			MILP		
may S S S O g		(1)	may F. F. F. c		(1')
$\max \sum_{k} \sum_{j} \sum_{i} Q_{ik} s_{ijk}$			$\max \sum_{k} \sum_{j} \sum_{i} s_{ijk}$		
$b_{ijk}S_i^L \le s_{ijk} \le b_{ijk}S_i^U$	∀i,j,k	(2)	$b_{ijk}Q_{ik}S_i^L \le s_{ijk} \le b_{ijk}Q_{ik}S_i^U$	∀i,j,k	(2')
$b_{ijk} \in \{0,1\}$	∀i,j,k	(3)	$b_{ijk} \in \{0,1\}$	$\forall i, j, k$	(3)
$b_{ijk} \le b_{ij(k+1)}$	$\forall i,j \ \forall k,k+1 \in K$	(4)	$b_{ijk} \le b_{ij(k+1)}$	$\forall i,j \ \forall k,k+1 \in K$	(4)
$s_{ijk} \le s_{ij(k+1)}$	$\forall i,j \ \forall k,k+1 \in K$	(5)	$s_{ijk}Q_{i(k+1)} \le s_{ij(k+1)}$	$\forall i,j \ \forall k,k+1 \in K$	(5')
$s_{ij(k+1)} - s_{ijk} \le (b_{ij(k+1)} - b_{ijk})S_i^U$	$\forall i,j \;\; \forall k,k+1 \; \in K$	(6)	$s_{ij(k+1)} - s_{ijk} \le (b_{ij(k+1)} - b_{ijk})S_i^U$	$\forall i,j \ \forall k,k+1 \in K$	(6)
$b_{ij}T^L \le \sum_k Q_{ik}b_{ijk} \le b_{ij}T^U$	$\forall i, j$	(7)	$b_{ijk_{LAST}}T^L \le \sum_k Q_{ik}b_{ijk} \le b_{ijk_{LAST}}T^U$	∀i,j	(7')
$b_{ij} \in \{0,1\}$	$\forall i, j$	(8)	(deleted)	(deleted)	(8)
$\sum_{j} b_{ij} \leq 1$	$\forall i$	(9)	$\sum_{j} b_{ijk_{LAST}} \leq 1$	$\forall i$	(9')
$\sum_{j} \sum_{i} s_{ijk} Q_{ik} \le D_j$	$\forall j$	(10)	$\sum_j \sum_i s_{ijk} \le D_j$	$\forall j$	(10')
$\sum_{i} s_{ijk} Q_{ik} \le E_{jk}$	∀j,k	(11)	$\sum_{i} s_{ijk} \le E_{jk}$	$\forall j, k$	(11')

Presented in Tables 7 and 8 are the results of Case Study 1 using the Modified Optimization Model.

			Sink 1	l (j=1)			Sink 2 (j=2)								
	k(1)	k(2)	k(3)	k(4)	k(5)	k(6)	k(1)	k(2)	k(3)	k(4)	k(5)	k(6)			
i (1)	1	1	1	1	1	1	0	0	0	0	0	0			
i (2)	0	0	0	0	0	0	0	0	0	0	0	0			
i (3)	0	0	0	0	0	0	0	1	1	1	1	1			
i (4)	0	0	0	0	0	0	0	0	0	0	0	0			
i (5)	0	0	0	0	0	0	0	0	1	1	1	1			

Table 7: C	ase Study	$1 b_{ijk}$ V	Variable	Results	(Modifi	ed O	ptimization	Model Rav	v Results)

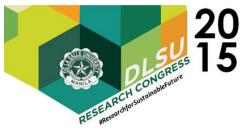
Table 8: Case Study 1 sijk Variable Results (Modified Optimization Mo	del Raw Results)
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			Sink 1	( j=1)					Sink 2	2 ( j=2)		
j=1	k(1)	k(2)	k(3)	k(4)	k(5)	k(6)	k(1)	k(2)	k(3)	k(4)	k(5)	k(6)
i (1)	10	10	10	10	0	0	0	0	0	0	0	0
i (2)	0	0	0	0	0	0	0	0	0	0	0	0
i (3)	0	0	0	0	0	0	0	4	4	4	4	4
i (4)	0	0	0	0	0	0	0	0	0	0	0	0
i (5)	0	0	0	0	0	0	0	0	6	6	6	6
$\Sigma_{\mathrm{i}} s_{ijk}$	10	10	10	10	0	0	0	4	10	10	10	10

The same objective value of 420 Mt of CO2 was obtained using the modified optimization model. We can also observe that all the inconsistencies between the results of Tan et al. and the raw data from the model have been eliminated. This eliminates the need for human intervention to properly pick out the correct results from the model.

### CASE STUDY 2

Case 2 from Tan, et al. was also evaluated to test consistency of results of the modified optimization model. Both original and modified model obtained an objective value of 530 MT of CO2. The modified



model has also successfully eliminated the inconsistencies in the results. CASE STUDY 3

To further magnify the problem of the original optimization model, Case Study 3 is presented below. Case study 3 considers a longer planning horizon of 40 years.

Sources	Flow Rate	Time of	Maximum
	(Mt/y)	Flow (y)	Capture (Mt)
1	10	0-20	200
2	2.5	0-30	75
3	4	0-35	140
4	4	0-25	100
<b>5</b>	6	10-40	180
Total	26.5	n/a	695
Sinks	Injection	Start	Maximum
	Limit (Mt/y)	Time (y)	Storage (Mt)
А	10	0	400
В	10	5	500
Total	20	n/a	900

Sources	Sink A	Sink B	Source
			Summary
1	10 Mt/y (t=0	0	200 Mt (100%)
	to 20)		
2	0	0	0 Mt (0%)
3	0	4 Mt/y (t=5	120 Mt (86%)
		to 35)	
4	0	4 Mt/y (t=5	80 Mt (80%)
		to 25)	
<b>5</b>	6 Mt/y	0	120 Mt (67%)
	(t=20 to 40)		
Sink Sum	mary		
	320 Mt (80%)	200Mt(40%)	520 Mt of CO2

Table 11: Case Study 3  $b_{ij}$  Variable Results (Original Optimization Model Raw Results)

1		
	j(1)	j(2)
i (1)	1	0
i (2)	0	0
i (3)	0	1
i (4)	0	1
i (5)	1	0

Table 12: Case Study 3 *b*<sub>ijk</sub> Variable Results (Original Optimization Model Raw Results)

				Sink 1	l (j=1)			Sink 2 ( j=2)								
	k(1)	k(2)	k(3)	k(4)	k(5)	k(6)	k(7)	k(8)	k(1)	k(2)	k(3)	k(4)	k(5)	k(6)	k(7)	k(8)
i (1)	1	1	1	1	1	1	1	1	0	0	0	0	0	1	1	1
i (2)	0	0	0	0	0	0	1	1	0	0	0	0	0	0	1	1
i (3)	0	0	0	0	0	0	0	1	0	1	1	1	1	1	1	1
i (4)	0	0	0	0	0	1	1	1	0	1	1	1	1	1	1	1
i (5)	0	0	0	0	1	1	1	1	0	0	0	0	0	0	0	0

Table 13: Case Study 3 sijk Variable Results (Original Optimization Model Raw Results)

				Sink	1 ( j=1)				Sink 2 ( j=2)								
j=1	k(1)	k(2)	k(3)	k(4)	k(5)	k(6)	k(7)	k(8)	k(1)	k(2)	k(3)	k(4)	k(5)	k(6)	k(7)	k(8)	
i (1)	10	10	10	10	10	10	10	10	0	0	0	0	0	10	10	10	
i (2)	0	0	0	0	0	0	2.5	2.5	0	0	0	0	0	0	2.5	2.5	
i (3)	0	0	0	0	0	0		4	0	4	4	4	4	4	4	4	
i (4)	0	0	0	0	0	4	4	4	0	4	4	4	4	4	4	4	
i (5)	0	0	0	0	6	6	6	6	0	0	0	0	0	0	0	0	
$\Sigma_{\mathrm{i}} s_{ijk}$	10	10	10	10	10	<b>20</b>	22.5	<b>26.5</b>	0	8	8	8	8	18	20.5	20.5	
$\Sigma_{\mathrm{i}} S_{ijk} Q_{ik}$	10	10	10	10	6	6	6	6	0	8	8	8	8	4	4	0	

#### Table 14: Case Study 3 bijk Variable Results (Modified Optimization Model Raw Results)

				Sink 1	L (j=1)			Sink 2 (j=2)								
	k(1)	k(2)	k(3)	k(4)	k(5)	k(6)	k(7)	k(8)	k(1)	k(2)	k(3)	k(4)	k(5)	k(6)	k(7)	k(8)
i (1)	1	1	1	1	1	1	1	1	0	0	0	0	0	0	0	0
i (2)	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
i (3)	0	0	0	0	0	0	0	0	0	1	1	1	1	1	1	1
i (4)	0	0	0	0	0	0	0	0	0	1	1	1	1	1	1	1
i (5)	0	0	0	0	1	1	1	1	0	0	0	0	0	0	0	0



Table 15: Case Study 3 sijk Variable Results (Modified Optimization Model Raw Results)

				Sink 1	L (j=1)				Sink 2 (j=2)								
j=1	k(1)	k(2)	k(3)	k(4)	k(5)	k(6)	k(7)	k(8)	k(1)	k(2)	k(3)	k(4)	k(5)	k(6)	k(7)	k(8)	
i (1)	10	10	10	10	0	0	0	0	0	0	0	0	0	0	0	0	
i (2)	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
i (3)	0	0	0	0	0	0	0	0	0	4	4	4	4	4	4	0	
i (4)	0	0	0	0	0	0	0	0	0	4	4	4	4	0	0	0	
i (5)	0	0	0	0	6	6	6	6	0	0	0	0	0	0	0	0	
$\Sigma_{\mathrm{i}} S_{ijk}$	10	10	10	10	6	6	6	6	0	8	8	8	8	4	4	0	

As we can see in Case Study 3, the problem with the inconsistent results increases with the increase of inactive period of the source after the connection. Both original model and modified model obtained an objective value of 520 Mt of CO2.

# 3. RESULTS AND DISCUSSION

In terms of calculation speed and complexity, please refer to Table 16 for Model Statistics. The optimization is done using B-and-B solver in Lingo version 14 on a Core i3, 2.3 GHz processor. The modified model involves lesser variables due to removal of  $b_{ij}$  variable. It can be noticed that there are lesser iterations for Case 2 which is the largest system compared to Case 1 and 3. This indicates easier convergence of model for larger systems. Although computational time for original and modified model is negligible, it can be assumed that the modified model can handle larger systems better.

Table 16: Comparison of Original and Modified Model Statistics

	Original			Modified		
Model Statistics	CS1	CS2	CS3	CS1	CS2	CS3
Objective Value	420	530	520	420	530	520
Solver Iteration	0	344	20	19	246	19
Total Variables	126	186	166	116	171	156
Integer Variables	65	95	85	55	80	75
Total Constraints	311	463	415	311	463	415
Total Nonzeros	894	1326	1164	886	1315	1164
Computational	<1	<1	<1	<1	<1	<1
Time, (s)						

# 4. CONCLUSIONS

A modified optimization model for optimal source sink matching in carbon capture and storage systems with time, injection rate, and capacity constraints has been formulated. It provides more accurate and robust results compared to the original model and eliminates inconsistencies that required human intervention in order to extract the correct conclusion.

# 5. NOMENCLATURE (1)

### Indices

I Set of CO<sub>2</sub> sources (i = 1, 2, ..., m)J Set of CO<sub>2</sub> sinks (j = 1, 2, ..., n)

K Set of Planning Periods (k = 1, 2, ... o)

### Parameters

- $D_j$  CO<sub>2</sub> storage limit of sink j
- $E_{jk}$  CO<sub>2</sub> injection rate limit of sink *j* in period *k*
- $Q_{ik}$  Binary parameter denoting the existence of source i in period k
- $S_{i^L}$  Lower limit for CO<sub>2</sub> flowrate from source *i*
- $S_i^{U}$  Upper limit for CO<sub>2</sub> flowrate from source *i*
- $T^L$  Minimum viable duration of connectivity
- $T^U$  Duration of planning horizon

### Variables

- $b_{ij}$  Binary variable denoting the existence CO2 stream from source *i* to sink *j* at any time within the planning horizon
- $b_{ijk}$  Binary variable denoting the existence CO2 stream from source *i* to sink *j* in period *k*
- $s_{ijk}$  CO2 flowrate from source *i* to sink *j* in period *k*

## 6. REFERENCES

- Tan, R., Aviso, K., Bandyopadhyay, S., & Ng, D. (2103). Optimal source-sink matching in carbon capture and storage systems with time, injection rate, and capacity constraints. *Environmental Progress & Sustainable Energy*, 32(2), 411-416.
- (2) Lindo Systems, Inc. Lingo: the Modelling Language and Optimizer; Chicago: 2013.