A Multi-Objective Unit Commitment Model for Setting Carbon Tax to Reduce CO₂ Emission: Thailand’s Electricity Generation Case

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Abstract

Carbon tax policy is a cost-effective instrument for emission reduction. However, setting the carbon tax is one of the challenging tasks for policy makers as it will lead to higher price of emission-intensive sources especially the utility price. In a large-scale power generation system, minimizing the operational cost and the environmental impact are conflicting objectives and it is difficult to find the compromise solution. This paper proposes a methodology of finding a feasible carbon tax rate on strategic level using the operational unit commitment model. We present a multi-objective mixed integer linear programming model to solve the unit commitment problem and consider the environmental impacts. The methodology of analyzing of the effect of carbon tax rates on the power generation, operating cost, and CO₂ emission is also provided. The trade-off relationship between total operating cost and total CO₂ emission is presented in the Pareto-optimal curve to analyze the feasible carbon tax rate that is influencing on electricity operating cost. The significant outcome of this paper is a modeling framework for the policy makers to determine the possible carbon tax that can be imposed on the electricity generation.

Keywords: CO₂ emission reduction; multi-objective unit commitment; carbon tax policy; pareto-optimal curve; energy policy

1. Introduction

Global warming and carbon dioxide (CO₂) emissions reduction have become important issues. The power sector is a major CO₂ polluter in many countries. Under the pressure of global warming, it is crucial for government to impose the effective policy to promote CO₂ emission reduction. Thailand’s CO₂ emissions levels are continuing to rise in accordance with the increased volume of national energy consumption.

There are several policies introduced to reduce CO₂ emissions. The most widely proposed is a carbon tax policy because it is a cost-effective instrument for emission reduction (Baranzini et al., 2000; Nagurney et al., 2006). Carbon tax is the tax paid by polluters who emit CO₂ from burning fossil fuels and releasing CO₂ into the atmosphere. In fact, CO₂ emissions are already implicitly taxed in every country even in those developing countries that the Kyoto protocol has not targeted (Baranzini et al., 2000). However, setting the carbon tax is a challenging task for policy makers as it will lead to higher price of emission-intensive sources, especially utility prices. The different carbon tax rate imposed in each country depends on the regulation and several factors.

Several researchers have formulated mathematical models and applied tools such as the HERMES macroeconomic model (Cosmo and Hyland, 2013) and the dynamic Computable General Equilibrium (CGE) model (Wachirarangsrikul et al., 2013; Thephkun et al., 2013) to explore the impact of the taxation of CO₂ on national economies that undertake carbon reduction in the future. Some of the studies estimate the optimal or appropriate carbon tax to impose on the electricity cost (Lu et al., 2010). Most of the studies divided the carbon tax into three scenarios which are a baseline (carbon tax is zero), the average carbon tax rate, and high tax rate. The results show that the CO₂ emission can be significantly reduced by imposing the fairly moderate tax rate (Wachirarangsrikul et al., 2013) and highest carbon tax (Cosmo and Hyland, 2013). The carbon tax policy shows the effectiveness and potential of CO₂ emission reduction and the positive impact on the economic and environmental improvement in long term.

Although, the proposed carbon tax rate shows the significant reduction of CO₂ emission, it cannot suggest a policy maker how much it costs to reduce the CO₂ emission to a certain amount based on the current carbon tax rate in Thailand and existing power generation capacity. It is important for the strategic policy maker to work in a cost effective way to reduce CO₂ emissions while maintaining minimum operating cost. In spite of this, the power sector intends to improve the operating efficiency and reduce overall operation
costs as much as possible while satisfying all demands.

Power generation planning is a challenging task because of the complications of the generation process, transmission, and generation of electric power which lead to a variety of issues in decision making. The power generator uses the unit commitment (UC) model to plan the operation of the generator unit whether to turn the unit on or off and the amount of power to be generated in each period. In general, the UC model is used to minimize the operating cost which mainly include the fuel cost. In this paper, the multi-objective unit commitment model is proposed with two conflicting objective functions: minimization of the operating cost and CO₂ emission. In the multi-objective problem, an optimal solution is generally more difficult to find. Therefore, the compromise solution can be the best for all conflicting objectives. Many solutions are presented in the Pareto-optimal solutions on a trade-off curve between the fuels cost and emission costs by using a multi-objective optimization (MO) model (Catalão et al., 2008) and another using the decommitment procedure of unit commitment (Yamashita et al., 2010). These solutions showed the trade-off points on the curve for a decision maker to select. However, the studies did not take a carbon tax into account. The models are tested on the simulated model which contains only 10 generating units (Yamashita et al., 2010).

From the strategic policy level view point, the impacts of various carbon tax rates on the operational level of unit scheduling are investigated. By considering different rates of the carbon tax, the impacts of CO₂ emission from strategic level planning that lead to the effect on the unit schedules and power generations of power system at the operational level are investigated. The model is coded and solved by using a commercial optimization package, IBM ILOG CPLEX, which is a computationally efficient solver that can handle large scale mixed integer programming problems.

2. Materials and Methods

2.1. Data of case study

A case study of a large scale power generation system in Thailand is used to demonstrate the methodology. The system comprises the combination of the thermal, gas turbine, hydro and combined cycle generators which are 51 power plants and 171 generators. The power plants are located in five main regions of Thailand: central, metro, north, northeast, and south regions. There are five main sources of fuel types: natural gas, coal/lignite, hydro, fuel oil, and diesel oil. The generating units generate electricity by using different types of fuel. Each fuel type has a different efficiency, price and CO₂ emission intensity. For daily planning, input of the models is the demand on Monday until 12:00 AM of day two. It is scheduled in a half-hour schedule with a time horizon of 24 hours (48 periods) with the time slot of 30 minutes as shown in Fig. 1.

There are two fuel modes in each generator which are the single and mixed fuel mode. The single fuel mode means the generating unit is allowed to use only one type of fuel to generate electricity. The mixed fuel mode means the generating unit is allowed to mix two types of fuel and the mixture is considered as a new fuel type.

The combination of fuel types produces different levels of CO₂ intensity. Therefore, we assign different

![Figure 1. Half an hour demand for one day for electricity in Thailand from EGAT data](image)
values of CO₂ intensity depending on the fuel type usage of each generating unit. The CO₂ intensity for each generator and fuel is shown in Table 1. For carbon tax imposed in Thailand, in order to estimate a carbon tax in Thailand, a baseline for setting a carbon tax is “Social cost of carbon (SCC)”. The social cost of carbon (SCC) is used by United States Environmental Protection Agency and other federal agencies to estimate the economic damages associated with a small increase in CO₂ emissions. However, the exact value of SCC is not defined. There are more than 200 estimated values of SCC starting from 15 baht/tCO₂ to 24,000 baht/tCO₂ (Policy Brief). Since, Thailand has not practically been imposed a carbon tax on electricity cost, therefore, this paper relies a carbon tax rate on IPCC Emission Trading value in 2009. Based on the Intergovernmental Panel on Climate Change (IPCC), an average carbon tax for Emission Trading Scheme (ET) in Europe Emission Trading Scheme (EU ETS) in 2009 is 349 baht/tCO₂ (Wattanakulcharat and Wongsa, 2011).

2.2. Problem Formulation

In this section, two models are introduced and formulated; the operating cost minimization (Model A) and the utility optimization model (Model B).

2.2.1. Operating Cost Minimization Model (Model A)

The model is formulated to solve the unit commitment (UC) problem to minimize total operating cost regardless of the CO₂ emission. The proportion of the fuel usage for each type is obtained by this model.

1) Objective Function: The objective function is the minimization of the total operating cost which includes the fuel cost and the startup cost of the generating units over the planning horizon, subjected to the load demand and other individual unit constraints such as the fuel usage. The objective function is formulated as follows:

\[
Min \sum_{i} \sum_{t} \left[ \left( \sum_{u \in U} \sum_{f \in F} c_{f}^{i} \times FU_{f}^{i} \times P_{f}^{i} \right) + \left( SC \times s_{i}^{i} \right) \right]
\]

(1)

Where \( I \) is set of power generating units (thermal, gas turbine, combined cycle and hydro), \( T \) is the length of the planning horizon, \( U \) is the total number of the generating units, \( F \) is the set of fuel types (lignite, natural gas and diesel oil).

The fuel cost function is calculated by the multiplication of cost of fuel type \( f \) used in unit \( i \) (\( c_{f}^{i} \)), the amount of fuel used in unit \( i \) at time \( t \) (\( FU_{f}^{i} \)), and the production of unit \( i \) at time \( t \) (\( P_{f}^{i} \)).

The startup cost \((SC)\) is the cost charged when any generating unit is started to generate the electricity. If the unit \( i \) is started at time \( t \), \( s_{i}^{i} \) has value of 1 and 0 otherwise. The startup cost is calculated by the startup cost of unit \( i \) multiplied by startup status.

2) General System and Unit Constraints:

System Load Balance Constraint: There are five zones of power generation in Thailand which are central, metro, northern, northeast, and southern zones. The total amount of production of unit \( i \) in zone \( z \) at time \( t \) (\( P_{i\ z}^{i} \)) plus the total of power transmitted through transmission line from demand zone \( i \) to demand zone \( j \) at time \( t \) (\( L_{i\ j}^{j} \)) minus the amount of water pumped by pump \( k \) in plant \( p \) at time \( t \) (\( pw_{kp}^{p} \)) must meet load demand \( d \) in zone \( z \) at time \( t \) (\( d_{z}^{z} \)). We also consider the transmission line capacity among zones and the pump power in this equation. The constraint is expressed as follows:

\[
\sum_{i} p_{i\ z}^{i} + \sum_{i \in Z, j \in J} L_{i\ j}^{j} - \sum_{j \in J, k \in K} p_{j\ k}^{p} = d_{z}^{z}, \quad \forall \ z \in Z, t \in T
\]

(2)

General Unit Constraints: This constraint is the turn on/off status for each generating unit. If the unit \( i \) is already operating when this scheduling horizon starts, then it is not turned on at the start as (3). If the unit \( i \) is already on when this scheduling horizon starts, then it will either be turned off at time 1 or remain operating at time 1 as (4). If the unit \( i \) is off when this scheduling horizon starts, then it is not turned off at the start as (5). If the unit \( i \) is off when this scheduling horizon starts, then the turn on variable must be the same as operating variable as (6). If the unit \( i \) is off at time \( t \) and on at

<table>
<thead>
<tr>
<th>Generator Type</th>
<th>Fuel Type</th>
<th>kgCO₂/MWh</th>
</tr>
</thead>
<tbody>
<tr>
<td>Thermal</td>
<td>The combination of natural gas, fuel oil, and diesel oil</td>
<td>622</td>
</tr>
<tr>
<td></td>
<td>The combination of natural gas and fuel oil</td>
<td>600</td>
</tr>
<tr>
<td></td>
<td>Lignite</td>
<td>914</td>
</tr>
<tr>
<td>Gas Turbine</td>
<td>The combination of natural gas and diesel oil</td>
<td>469</td>
</tr>
<tr>
<td>Combined Cycle</td>
<td>The combination of natural gas and diesel oil</td>
<td>469</td>
</tr>
<tr>
<td></td>
<td>Natural gas</td>
<td>426</td>
</tr>
</tbody>
</table>

Sources: Adopted from Sutham (2011) and EGAT (2010)
time \( t+1 \), then it was turned on at time \( t+1 \) as (7). The constraints are expressed as follows:

\[
s_i^0 = 0 \quad \forall \; i \in \{I \cup I_S \mid p_i^0 > 0\}
\]
\[
s_i^0 + o_i = 1 \quad \forall \; i \in \{I \cup I_S \mid p_i^0 > 0\}
\]
\[
s_i^0 = 0 \quad \forall \; i \in \{I \cup I_S \mid p_i^0 = 0\}
\]
\[
s_i^0 = o_i^0 \quad \forall \; i \in \{I \cup I_S \mid p_i^0 = 0\}
\]
\[
o_i^{t+1} - o_i^t \leq s_i^{t+1} \quad \forall \; i \in I \cap I_S, t \in \{1, \ldots, T-1\}
\]

Generation Limit Constraint: The power generation from the generating unit \( i \) at time \( t \) has upper and lower bound limits. The constraint is expressed as follows:

\[
o_i^t p_i^\text{min} \leq p_i^t \leq o_i p_i^\text{max} \quad \forall \; i \in I, t \in T
\]

Ramp-Up/Down Constraints: The generating unit’s ramp-up (\( \phi \)) and ramp-down (\( \phi \)) rate at time \( t \). For each ramping, we divided the equation into two stages which are the initial stage and the processing stage. Constraint (9) and (10) are the initial stages for ramp-up and ramp-down, respectively. Constraint (11) and (12) are the ramp-up and ramp-down limitation, respectively. The constraints are expressed as follows:

\[
p_i^t - p_i^0 \leq \phi_i \mu \quad \forall \; i \in I
\]
\[
p_i^0 - p_i^t \leq \phi_i \mu \quad \forall \; i \in I
\]
\[
p_i^{t+1} - p_i^t \leq \phi_i \mu \quad \forall \; i \in I
\]
\[
p_i^t - p_i^{t+1} \leq \phi_i \mu \quad \forall \; i \in I
\]

Minimum Up/Down Time Constraints: When the unit \( i \) is scheduled to start-up or shutdown, it has a duration of minimum up time (\( \varepsilon_i \)) or minimum down time (\( \bar{\varepsilon}_i \)) before the unit is up or shutdown because the generating unit cannot be immediately turned on or turned off. The constraints of minimum up/down time are expressed in (13) and (14) as follows:

\[
\sum_{t \leftarrow \varepsilon_i} s_i^t \leq o_i^t \quad \forall \; i \in I, t \in \{T \mid t > \varepsilon_i\}
\]
\[
\sum_{t \leftarrow \bar{\varepsilon}_i} s_i^t \leq 1 - o_i^t \quad \forall \; i \in I, t \in \{T \mid t > \bar{\varepsilon}_i\}
\]

Fuel Usage Constraints: There are two constraints for the fuel usage limitation. First, the total amount of fuel usages \( f \) of the unit \( i \) in plant \( p \) at time \( t \) (\( q_{fp}^i \)) must not exceed the maximum amount of fuel \( f \) available for plant \( p \) (\( q_{fp}^\text{max} \)). Second, the generating unit \( i \) cannot use more than one type of fuel \( f \) at time \( t \) (\( FU_i^f \)). The constraints are expressed as follows:

\[
\sum_{r \in I} q_{fp}^i \leq q_{fp}^\text{max} \quad \forall \; p \in P, f \in F
\]
\[
\sum_{f \in F} FU_i^f \leq 1 \quad \forall \; i \in I
\]

Spinning Reserve Constraint: The spinning reserve is provided to compensate when shortfalls occur. The total spinning reserve is greater than or equal to the spinning reserve required (\( \Delta \)). The constraint is expressed as follows:

\[
\sum_{i \in I} (p_i^\text{max} - o_i - p_i) \geq \Delta \quad \forall \; t \in T
\]

2.2.2. Utility Optimization Model (Model B)

The utility optimization model (Model B) is formulated using the multi-objective unit commitment technique to find the compromise solution that minimizes both operating cost and CO2 emission cost. The CO2 emission is converted to monetary cost by using the carbon tax rate. The objective consists of the fuel costs, start-up cost, operating cost, and CO2 emission costs. The formulation is presented as follows.

1) Objective Function: The objective function is the minimization of the total cost as formulated below:

\[
\text{Min} \sum_{i \in I} \sum_{t \in T} \left[ \sum_{f \in F} c_f^i \times FU_i^f \times P_i^t \right] + \left( SC_i + s_i \right) + \left( E_i \times P_i^t \times CT_i \right)
\]

\[
\text{CO2 emission for each generation: } E_i \cdot FC_f \times E_f
\]

The CO2 emission cost is the CO2 emission intensity from generating unit \( i \) (\( E_i \)) multiplied by the total production (\( P_i^t \)) and carbon tax rate from generating unit \( i \) (\( CT \)). The CO2 emission intensity (\( E_i \)) is calculated by fuel consumption (\( FC_f \)) multiplied by Emission Factor for each fuel type (\( E_f \)). The default value of emission factor is obtained from Revised 2006 IPCC Guideline for National Greenhouse Gas Inventories. Model B contains objective function with constraints (2)-(17).

3. Results and discussion

In this section, the analysis of the results obtained
3.1. Compromise Solution of Operating Cost, CO_2 emission Cost and CO_2 emission

In this section, we present the result of Model A and B using the mixed integer programming. The comparison is divided into three terms which are the total operating cost, the total CO2 emission cost and the total CO2 emission. For the total operating cost comparison, the Utility Optimization Model (Model B) has only 0.28 percent higher cost than the Operating Cost Minimization Model (Model A) which is considered acceptable. For the total CO2 emission cost, Model B is 1.23 percent lower than Model A. Moreover, the total CO2 emission from Model B is lower than Model A approximately 4.82 million kilograms of carbon dioxide (kgCO_2) or 4,816 tons of carbon dioxide (tCO2). Once, the multi-objective unit commitment approach is applied, Model B is not only minimize total operating cost, but also the total CO2 cost. Therefore, the main generators are in combined cycle power plants which lead to a lower amount of coal-fired consumption. Therefore, the CO2 emission is significantly dropped.

The results of CO2 emission comparison are shown in Table 3. The unit commitment optimization approach decreases the overall emission by 2.65 percent from Model A. The total proportion of emission of Model B from the thermal plants is decreased by 17.44 percent. The emission from the gas turbine and combined cycle plants are increased by 18.81 and 8.85, respectively. The system reduces the usage of lignite and shift to natural gas instead. The natural gas is typically expensive; therefore, it leads to higher total cost in Model B than Model A. Once the CO2 emission cost is involved in the objective function, the proportion of fuel type consumption is changed. For Model B, the fuel usage of lignite in thermal generator is decreased by 20.15 percent and the usage of natural gas is increased by 21.05 percent.

3.2. Sensitivity Analysis

In this section, the sensitivity analysis of Model B shows the comparison of the power generation for each generator by varying the carbon tax rates from 0.10 - 3.00 baht/kgCO2 (Fig. 2). The high carbon tax rate directly impacts the behavior of power generation, especially for the thermal and combined cycle plants. There is no significant change when the carbon tax rate is between 0.10 and 0.60 baht/kgCO2 which indicates that the prices are too small to trigger the change of system behavior compared with the operating cost. However, change occurs slowly when the prices are between 0.70 and 1.80 baht/kgCO2. The proportion of combined cycle plant is slowly increased while the thermal plant is slowly decreased. The obvious change is occurred when the price is greater than 1.90 baht/kgCO2.

The stacked chart of percent changed of power generation for thermal, gas turbine, and combined cycle plant is shown in Fig. 3. At the rate of 0.70 baht/kgCO2, the generation from thermal starts to decrease by 10 percent while the combined cycle is increased by 7.14 percent. This shows the carbon tax has higher effect than the operating cost in the thermal plant. On average, from the rate of 0.20 - 1.90 baht/kgCO2, the generation from thermal is decreased by 14.74 percent while the combined cycle is increased by 10.02 percent.

Table 2. Total cost comparison for each model

<table>
<thead>
<tr>
<th>Model</th>
<th>CO2 Emission Cost (THB)</th>
<th>Total Operating Cost (THB)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Operating Cost Minimization Model</td>
<td>136,723,966.31</td>
<td>633,856,023.21</td>
</tr>
<tr>
<td>Utility Optimization Model</td>
<td>135,043,207.04</td>
<td>635,609,670.08</td>
</tr>
</tbody>
</table>

Table 3. Emission comparison for each power plant type

<table>
<thead>
<tr>
<th>Plant Type</th>
<th>Emission (kgCO2)</th>
<th>% Change</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Model A</td>
<td>Model B</td>
</tr>
<tr>
<td>Thermal</td>
<td>201,056,448.00</td>
<td>171,202,150.53</td>
</tr>
<tr>
<td>Gas Turbine</td>
<td>29,027,026.93</td>
<td>35,750,944.71</td>
</tr>
<tr>
<td>Combined Cycle</td>
<td>139,715,145.47</td>
<td>153,286,247.06</td>
</tr>
<tr>
<td>Total</td>
<td>369,798,620.39</td>
<td>360,239,342.31</td>
</tr>
</tbody>
</table>
The significant change obvious occurs when the price is greater than 2.00 baht/kg CO\textsubscript{2}. On average, from the rate of 2.00 - 3.00 baht/kg CO\textsubscript{2}, the generation from thermal is decreased by 57.58 percent while the combined cycle is increased by 33.93 percent. This leads to higher CO\textsubscript{2} emission reduction.

Fig. 4 shows the percentage of change of the operating cost and the CO\textsubscript{2} emission when the carbon tax rate is increased by 0.10 baht/kg CO\textsubscript{2}. There are called the marginal operating cost and the marginal CO\textsubscript{2} emission, respectively. The obvious trade-off between the operating cost and the CO\textsubscript{2} emission occurs at the price of 0.70 baht/kg CO\textsubscript{2}. The CO\textsubscript{2} emission is decreased by 5.13 percent while the operating cost is increased only 1.66 percent. For the based case, the imposition of carbon tax at 0.349 baht/kg CO\textsubscript{2} reduces approximately 1.96 percent while the total operating cost increases by only 0.04 percent.
3.3. Pareto-Optimal Analysis of Utility Optimization Model

In this section, we present the Pareto-Optimal analysis by using the efficient frontier graph between the total operating cost and the CO2 emission of the utility optimization model. It is illustrated as a guideline for a strategic planner to determine the compromise solution for the power generation based on environmental considerations. Besides, we suggest the methodology to set an appropriate carbon tax that maximizes the reduction of the CO2 emission while the operating cost increases in the lowest proportion.

The trade-off graph of Model B is obtained by plotting the point representing the total operating cost and total CO2 emission for each carbon tax rate as shown in Fig. 5. This graph shows the different behavior of the increment of the total operating cost over the decrement of the total CO2 emission as the carbon tax rate increased. The marginal CO2 emission is the change in the CO2 emission when the carbon tax has an increment by unit. The marginal operating cost is the change in the operating cost when the carbon tax has an increment by unit.

When the carbon tax is changed from 0.60 to 0.70 baht/kgCO2, the marginal CO2 emission is decreased by 6.30 percent while the operating cost is increased only 1.72 percent. The rate of change of the operating cost and the CO2 emission is 1:4. From 1.90 to 2.00 baht/kgCO2, the marginal reduction of CO2 emission is equal to the marginal increment of operating cost which is 3 percent. From 2.40 to 2.50 baht/kgCO2 and the marginal reduction of CO2 emission is equal to the marginal increment of operating cost which is 5 percent. The suggestion from this relationship is the appropriate carbon tax range that can decrease a large proportion of CO2 emission is between 0.60 to 0.70 baht/kgCO2.

In the utility optimization model, the carbon tax at any rate can be set to test the effect of the operating cost and the CO2 emission instantly. Therefore, the computational time of the utility optimization model is fast and continent.

4. Conclusions

This paper presented a modeling framework to determine the possible carbon tax rates that can be imposed on the electricity operating cost for the policy makers. It was significant to know the best achievable level of the total operating cost and the CO2 emission based on the capacity of existing power generators. The Utility Optimization model was proposed in this paper to generate a good compromise solution for a multi-objective unit commitment problem in large-scale power generation planning. The operating cost was increase in small proportion compared with large amount of CO2 emission it reduced. The trade-off relationship between the operating cost and CO2 emission and the marginal cost of CO2 reduction were illustrated to use as a guideline for a planner to make an optimal decision for the unit commitment dispatch with the acceptable emission allowance.
The possible carbon tax rate that can be imposed to reduce CO₂ emissions in long term is between 0.60 to 0.70 baht/kgCO₂. In order to set the carbon tax on electricity cost, it is needed the involvement of government office and related parties. The appropriate carbon tax rate setting will lead to CO₂ emissions reduction in a long term. In the beginning phase, the lower carbon tax rate can be imposed on electricity cost to observe the impact on different perspective by using the carbon tax rate of 0.349 baht/kgCO₂ as a baseline. It might be applied for a few years. Once the CO₂ emissions can be decreased by this policy, the government sector can increase the carbon tax rate and reduce more emissions. The options that can be used to limit the CO₂ emissions from electricity generation include increasing the usage of renewable energy, using fuels with lower CO₂ emission per kWh produced and/or increasing the efficiency of the electricity production. In future research, the model should consider the share of renewable energy and formulate the equation that also involved renewable energy.

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References


IPCC Guidelines for National Greenhouse Gas Inventories. Chapter 2, Table 2.2 Default emission factors for stationary combustion in the energy industries, Energy; 2: 2.16. 2006.


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