



**UNIVERSITI PUTRA MALAYSIA**

**APPLICATION OF DATA MINING TECHNIQUES FOR ECONOMIC  
EVALUATION OF AIR POLLUTION IMPACT AND CONTROL**

**IING LUKMAN**

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EVALUATION OF AIR POLLUTION IMPACT AND CONTROL**

**IING LUKMAN**

**Thesis Submitted to the School of Graduate Studies, Universiti Putra Malaysia,  
in Fulfilment of the Requirements for the Degree of Doctor of Philosophy**

**August 2007**





*To all my family,*

*My teachers, IPB, UPM,*

*and Tanah Air.....Indonesia*



Abstract of thesis presented to the Senate of Universiti Putra Malaysia in fulfilment of requirement for the degree of Doctor of Philosophy

**APPLICATION OF DATA MINING TECHNIQUES FOR ECONOMIC  
EVALUATION OF AIR POLLUTION IMPACT AND CONTROL**

By

**IING LUKMAN**

**August 2007**

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**Faculty : Environmental Studies**

In this research we examine aspects of the interdependence between economic development and the use of environmental and natural resources assets from global data published by United Nations. For that purpose, we use data mining techniques.

Data mining techniques applied in this thesis were: 1) Group method of data handling (GMDH), originally from engineering, introducing principles of evolution - inheritance, mutation and selection - for generating a network structure systematically to develop the automatic model, synthesis, and its validation; 2) The weighted least square (WLS) and step wise regression were also applied for some cases; 3) The classification-based association rules were applied.

Data sets for this research consist of two sets integration data of air quality data and macroeconomic data of the cross-country data of World Development Indicator 2003 (WDI 2003), and from [www.nationmaster.com](http://www.nationmaster.com). The results from [www.nationmaster.com](http://www.nationmaster.com) were as follows: the corruption index was strongly related to the urban SO<sub>2</sub> concentration. The corruption index along with NO<sub>x</sub> emission has big contribution to the debt. Debt is the debt of the home country to the foreign country or external debt or foreign debt.

The result from WDI 2003 shows that the mortality rate of children under five years old depended on sanitation and water facilities obtained from GMDH results. However, the results from stepwise regression shows that mortality rate was dependent on annual deforestation, particulate matter, nationally protected area, but the big contribution was from annual deforestation.

Based on GMDH, new Gross National Income (GNI) formula was found. Previously GNI was known as Gross National Product (GNP). It was different from the common formula of GNP. The formula or equation model of urban SO<sub>2</sub> concentration was also found through the GMDH algorithms. The results were then compared to WLS and Stepwise regression.

The debt was found by GMDH to be dependent on the corruption index as well as urban SO<sub>2</sub> concentration. Corruption index along with NO<sub>x</sub> emission were related to debt.

Results from weighted least square using SAS software showed that the corruption index was significant to the concentration of urban SO<sub>2</sub>.

Results from classification rules of the WDI 2003 data showed that the more energy imports net from foreign country was associated with the smaller in adjusted net saving in home country. Energy imports net were calculated as energy use in oil equivalents. This indicated that if the energy imports net was higher, then the adjusted net saving was small, and then CO<sub>2</sub> emissions was small also. Thus, to reduce global warming in home country, a country can import energy from foreign country. According to the result from association rules on nationamaster.com data there were indication that corruption index was related with higher urban SO<sub>2</sub> concentration, and inflation.

Results from association rules of item sets shows that the urban SO<sub>2</sub> always follows the direction of corruption index. In addition, if any country wants to reduce the urban SO<sub>2</sub> concentration, more works can be conducted on controlling corruption index than controlling SO<sub>2</sub> emission per populated area.

Abstrak tesis yang dikemukakan kepada Senat Universiti Putra Malaysia  
sebagai memenuhi keperluan untuk ijazah Doktor Falsafah

**APLIKASI TEKNIK PERLOMBONGAN DATA UNTUK PENILAIAN  
EKONOMI BAGI IMPAK DAN KAWALAN PENCEMARAN UDARA**

Oleh

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**Ogos 2007**

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Dalam kajian ini, aspek saling bergantung antara pembangunan ekonomi dengan penggunaan harta alam sekitar dan sumber semulajadi daripada data global yang diterbitkan oleh Bangsa-bangsa Bersatu dikaji. Bagi tujuan itu, kami gunakan kaedah-kaedah perlombongan data.

Kaedah perlombongan data yang digunakan dalam tesis ini adalah: 1) Kaedah kumpulan dalam pengendalian data (GMDH) berasal dari bidang kejuruteraan yang memperkenalkan prinsip evolusi –warisan, mutasi dan pilihan – untuk menjana satu struktur rangkaian yang teratur dalam pembinaan model automatik, sintesis and pengesahannya; 2) Regresi kuasa-dua berpemberat (WLS) dan regresi langkah cerdik juga diaplikasikan untuk sesetengah kes; 3) Peraturan asosiasi berasaskan klasifikasi digunakan.

Set data untuk kajian ini terdiri daripada dua set integrasi data dari data kualiti udara dan data makroekonomi dari kumpulan data serata dunia dari penunjuk pembinaan dunia 2003 (WDI 2003), dan juga dari [www.nationmaster.com](http://www.nationmaster.com). Keputusan dari [www.nationmaster.com](http://www.nationmaster.com) adalah seperti berikut: indeks rasuah berkait rapat dengan kandungan urban  $SO_2$ . Indeks rasuah berserta kepulan  $NO_x$  memiliki sumbangan yang besar kepada hutang. Hutang bermakna hutang sesebuah negara sendiri ke negara asing atau hutang ke dunia luar atau hutang ke negara luar.

Keputusan dari WDI 2003 menunjukkan bahawa kadar kematian kanak-kanak dibawah umur lima tahun bergantung kepada kebersihan dan fasiliti air yang diperolehi dari keputusan GMDH. Walau bagaimanapun, keputusan dari regresi langkah cerdik menunjukkan kadar kematian bergantung kepada kepupusan setiap tahun kawasan perhutanan, jirim debu halus, kawasan lindungan negara, tetapi sumbangan besar yalah dari kepupusan setiap tahun kawasan perhutanan.

Berasaskan pada GMDH, kemudiannya formula pendapatan kotor kebangsaan (GNI) yang baru telah ditemui. Dahulunya GNI dikenali sebagai hasil kotor kebangsaan (GNP). Ianya berbeza dengan formula biasa hasil kotor kebangsaan. Formula atau persamaan model kandungan urban  $SO_2$  juga ditemui menggunakan algoritma GMDH. Keputusannya kemudian diperbandingkan dengan keputusan dari WLS dan regresi langkah cerdik.



Hutang ditemui oleh GMDH sebagai bergantung pada indeks rasuah sepertimana kandungan urban  $\text{SO}_2$ . Index rasuah bersama dengan keputulan  $\text{NO}_x$  berkait kepada hutang. Keputusan dari WLS menggunakan perisian SAS menunjukkan bahawa indeks rasuah penting kepada kandungan urban  $\text{SO}_2$ .

Keputusan dari peraturan klasifikasi daripada data WDI 2003 menunjukkan bahawa lebih banyaknya impot bersih tenaga dari negara asing berkaitan dengan simpanan bersih terubahsuai yang kecil di negara sendiri. Impot tenaga bersih dicongak sebagai penggunaan tenaga yang disamakan dengan penggunaan minyak. Ini bermakna bahawa jika impot bersih tenaga adalah tinggi, maka simpanan bersih terubahsuai adalah kecil, dan keputulan  $\text{CO}_2$  pun kecil. Maka, untuk mengurangkan kepanasan global di negara sendiri, sesebuah negara boleh mengimpot tenaga dari negara asing. Berasaskan pada keputusan dari peraturan asosiasi pada data nationmaster.com ada tanda bahawa indeks rasuah berkait dengan kandungan urban  $\text{SO}_2$  dan juga inflasi.

Keputusan dari peraturan asosiasi daripada set-set item menunjukkan bahawa urban  $\text{SO}_2$  selalu mengikut hala indeks rasuah. Dengan demikian, jika sesebuah negara ingin mengurangkan kadar kandungan urban  $\text{SO}_2$ , lebih banyak kerja yg boleh dilaksanakan dengan mengawal indeks rasuah daripada dengan mengawal keputulan  $\text{SO}_2$  pada setiap kawasan penduduk.

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I certify that an Examination Committee met on 20 April 2007 to conduct the final examination of Iing Lukman on his Doctor of Philosophy thesis entitled “Data Mining for Economic Evaluation of Air Pollution Impacts and Controls” in accordance with Universiti Pertanian Malaysia (Higher Degree) Act 1980 and Universiti Pertanian Malaysia (Higher Degree) Regulations 1981. The Committee recommends that the candidate be awarded the relevant degree. Members of the Examination Committee are as follows:

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