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INTELLIGENT CLASSIFICATION OF AMMONIA CONCENTRATION
BASED ON ODOR PROFILE

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Thesis submitted in fulfillment of the requirements
for the award of the degree of
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ABSTRACT

This thesis presents the 'intelligent classification of ammonia concentration based on the standard of oil and gas industries wastewater discharge. The intelligent classification using signal processing is a well-known technique in many applications and as well in the oil and gas industry. The intelligent classification technique for ammonia concentration classification is a demanding technique especially in the environmental sector. Ammonia solution properties and ammonia solution preparations were studied in this thesis which commonly used in industry. The objectives of this thesis are to develop an intelligence classification of ammonia concentration based on the oil and gas industry wastewater discharge schedule and to analyze performance of the intelligent classification of ammonia concentration based on the oil and gas industry wastewater discharge schedule. In this thesis the ammonia odor profile has been pre- identified by chemist using four sensor array. The ammonia concentration was validated using a commercialized gas sensor and spectrophotometer to cross-validated e-nose instrument. The odor profile from two different samples; high (20 ppm and 25 ppm) and low (5 ppm, 10 ppm and 15ppm) concentration that have been normalized and visualized in a 2D plot to extract the unique patterns. The variance of the low and high concentration of ammonia odor profile has been identified as different group samples. This group samples have been analyzed statistically using Boxplot, calibration curve and proximity matrix. The thesis describes the statistical techniques to visualize the pattern and using mean features to classify between the low and high concentration. Two intelligent classification techniques have been used which are Artificial Neural Network (ANN) using the back-propagation approaches and then, the result of ANN model was cross-validated using CBR. Both ANN model and CBR classifier have been measured using several performance measures. From the results, it is observed that ANN model and CBR classifier are capable of classifying 100% of ammonia concentration odor profile from the water. The results can also significantly reduce the cost and time, and improve product reliability and customer confidence.

ABSTRAK

Tesis ini membentangkan klasifikasi pintar ammonia dari sisa air buangan berdasarkan standard industri minyak dan gas. Pengelasan pintar menggunakan pemprosesan isyarat adalah satu teknik yang terkenal dalam pelbagai aplikasi dan juga dalam industri minyak dan gas. Teknik pengelasan pintar untuk pengelasan ammonia adalah teknik yang ditunggu-tunggu terutama dalam sektor alam sekitar. Ciri-ciri Ammonia solution dan persediaan ammonia solution telah dikaji dalam tesis ini adalah yang biasa digunakan dalam industri. Objektif projek ini adalah untuk membangunkan klasifikasi pintar ammonia daripada sisa air buangan berdasarkan pada industri minyak dan gas mengikut pelepasan sisa air buangan yang telah ditetapkan oleh Jabatan Alam Sekitar (JAS) dan untuk menganalisis prestasi klasifikasi pintar ammonia daripada sisa air buangan berasaskan minyak dan gas industri. Dalam tesis ini profil bau ammonia telah pradikenalpasti oleh ahli kimia menggunakan empat sensor pelbagai. Kepekatan ammonia telah disahkan menggunakan sensor gas yang telah dikomersialkan untuk silang pengesahan menggunakan E-nose. Profil bau dari dua sampel berbeza; Kepekatan yang tinggi dan rendah telah dibiasakan dan digambarkan dalam plot 2D untuk mendapatkan corak yang unik. Kepelbagaiannya kepekatan rendah dan tinggi bagi profil bau ammonia telah dikenal pasti sebagai kumpulan sampel yang berbeza. Ini kumpulan sampel telah dianalisis menggunakan Boxplot, Calibration Curve dan Proximity matrix sebagai teknik statistik. Tesis menerangkan teknik statistik untuk menggambarkan corak dan menggunakan ciri-ciri min untuk mengklasifikasikan antara kepekatan rendah dan tinggi. Dua teknik klasifikasi pintar telah digunakan iaitu Rangkaian Neural Buatan (ANN) menggunakan pendekatan back-propagation. Kemudian, hasil daripada model ANN adalah untuk silang pengesahan menggunakan ‘Case- Based Reasoning’ (CBR). Kedua-dua pengelasan pintar bagi model ANN dan pengelas CBR telah diukur menggunakan beberapa prestasi. Daripada keputusan, didapati bahawa model ANN dan CBR pengelas mampu untuk mengklasifikasikan 100% daripada profil bau kepekatan ammonia. Keputusan menunjukkan bahawa E-nose mampu mengurangkan kos dan masa ke pasaran, dan meningkatkan kebolehpercayaan produk dan keyakinan pelanggan.

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APPENDIX A

List of publication as author

Zahari, M. F., Julius, T. A., Halim, F. A., Najib, M. S., Ghazali, K. H., & Mohd Azoddein, A. A. (2015). Intelligent Classification Hazardous Gas Using Sensors Array. *Journal of Advances in Information Technology*, Volume 6 (No. 4), 233–237.

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