

PREDICTING ENGINE PERFORMANCE
AND EXHAUST EMISSIONS OF A SPARK
IGNITION ENGINE FUELLED WITH 2-
BUTANOL-GASOLINE BLENDS USING RSM
AND ANN MODELS

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BLENDS USING RSM AND ANN MODELS

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ABSTRAK

Penyelidikan eksperimen dalam pengujian enjin menggunakan bahan bakar alternatif selalunya tertakluk kepada operasi enjin, yang memerlukan masa dan pembiayaan kos bahan yang mahal. Atas sebab-sebab ini, kajian ini bertujuan untuk meramalkan prestasi enjin dan pelepasan ekzos menggunakan 2-butanol-gasoline bahan api yang dicampur dengan nisbah peratusan 5:95 (GBu5), 10:90 (GBu10) dan gasoline 15:85 (GBU-15) kepada 2-butanol, masing-masing, yang dikendalikan dalam empat silinder, empat lejang bahan api 4G93 Mitsubishi pada kedudukan pendikit lebar 30%, 50% dan 70% menggunakan metodologi permukaan tindak balas (RSM) dan rangkaian saraf tiruan (ANN). Berdasarkan eksperimen penyiasatan tersebut, pada 30%, 50% dan 70% daripada kedudukan pendikit lebar, 2-butanol-gasoline bahan api dicampur menunjukkan peningkatan kuasa brek enjin, brek tork dan kecekapan haba brek dengan meningkatkan kandungan 2-butanol dalam bahan api petrol. Prestasi enjin menunjukkan peningkatan dalam kuasa brek, brek tork dan brek thermal kecekapan dalam purata 2% hingga 15% dan 0.2% kepada 1.5%, masing-masing, untuk semua kedudukan pendikit diuji berkenaan dengan meningkatkan kandungan 2-butanol dalam bahan api petrol. Untuk pelepasan ekzos, ianya telah dicatatkan, penurunan yang sekata bagi nitrogen oksida (NO_x), carbon monoksida (CO), carbon dioksida (CO_2) dan hydrocarbon yang tidak terbakar (HC) untuk GBu5, GBu10 dan GBU-15, secara purata sebanyak 7.1%, 13.7%, dan 19.8% daripada G100, masing-masing, lebih jarak kelajuan 1000 hingga 4000 RPM. Kandungan pelepasan lain menunjukkan CO dan HC lebih rendah tetapi CO_2 lebih tinggi dari 2500 hingga 4000 RPM untuk bahan api campuran. Seterusnya, kelajuan enjin, bahan api campuran 2-butanol dan kedudukan pendikit enjin dan hasil dari prestasi enjin dan ekzos ciri-ciri pelepasan telah digunakan sebagai input dan output untuk metodologi RSM dan ANN. Berdasarkan model RSM, ciri-ciri prestasi mendedahkan bahawa kenaikan 2-butanol dalam bahan api yang dicampurkan membawa kepada peningkatan aliran kuasa brek, brek tork dan kecekapan terma brek. Bagaimanapun, bahan api brek didapati sedikit lebih tinggi diperhatikan. Tambahan pula, RSM model ini mencadangkan bahawa kehadiran 2-butanol mempamerkan trend penurunan kepada NO_x , CO dan HC, bagaimanapun trend yang lebih tinggi dapat diperhatikan untuk pelepasan CO_2 dimana keputusan ini adalah selari dengan keputusan eksperimen. Sementara itu, bagi ANN pula, kedua-dua lapisan tersembunyi model ANN dilatih tansig-logsig gabungan fungsi pengaktifan menghasilkan koefisien korelasi terbaik, R pada nilai 0.9995 terhadap gabungan fungsi pengaktifan lain yang dinilai. Walau bagaimanapun, untuk mencapai model ramalan yang lebih tepat, semua konfigurasi dinilai lebih lanjut oleh tambahan analisis kesilapan statistik dan korelasi metrik, iaitu Mean Absolute Percentage Error (MAPE), Mean Squared Error (MSE), Root Mean Squared Error (RMSE), Theil U2, Nash-Sutcliffe Efficiency (NSE) and Kling-Gupta Efficiency (KGE). Penilaian, gabungan fungsi pengaktifan terbaik untuk kuasa brek, BSFC, BTE, NO_x , CO, dan model ANN bagi CO_2 adalah konfigurasi tansig-logsig. Bagi tork Brek dan HC, kombinasi tansig memberikan ramalan yang lebih baik. Ia boleh ditunjukkan dengan tepat daripada kajian bahawa model ANN yang maju mempunyai ketepatan ramalan yang lebih tinggi berbanding dengan model RSM.

ABSTRACT

Experimental investigation in engine testing using alternative fuels always subjected to more engine operation, time-consuming and require expensive cost of materials. For these reasons, this study is aimed to predict the engine performance and exhaust emissions using 2-butanol-gasoline blended fuels with percentage volume ratios of 5:95 (GBu5), 10:90 (GBu10) and 15:85 (GBu15) of gasoline to 2-butanol, respectively, operated in a four-cylinder, four-stroke port fuel 4G93 Mitsubishi spark ignition engine at 30%, 50% and 70% of throttle position using artificial neural network and response surface methodology techniques. Based on the experimental investigation, at 30%, 50% and 70% of throttle position, 2-butanol-gasoline blended fuels indicated an improvement in engine brake power, brake torque and brake thermal efficiency with increasing 2-butanol content in the gasoline fuels. The engine performance indicated improvement in brake power, brake torque and brake thermal efficiency in the average of 2 to 15% and 0.2% to 1.5%, respectively, for all of the tested throttle position with respect to increasing the 2-butanol content in the gasoline fuel. For exhaust emissions, it was recorded that, a significant decreased of NO_x , CO, CO_2 and HC for GBu5, GBu10 and GBu15, by an average of 7.1%, 13.7%, and 19.8% than G100, respectively, over a speed range of 1000 to 4000 RPM. Other emission contents indicate lower CO and HC but higher CO_2 from 2500 to 4000 RPM for the blended fuels. The engine speeds, 2-butanol blended fuels and engine throttle position and results from the engine performance and exhaust emissions characteristics was then used as the input and output for the for the artificial neural network and response surface methodology. Based on the RSM model, performance characteristics revealed that the increment of 2-butanol in the blended fuels lead to the increasing trends of brake power, brake torque and brake thermal efficiency. Nonetheless, a marginally higher brake specific fuel consumption was observed. Furthermore, the RSM model suggests that the presence of 2-butanol exhibits a decreasing trend of NO_x , CO, and HC, however, a higher trend was observed for CO_2 exhaust emissions, which are in accordance with the experimental results. Meanwhile, for ANN it was shown that the two hidden layer ANN model trained with the tansig-logsig activation function combination yields the best correlation coefficient, R at a value of 0.9995 against other activation function combinations evaluated. However, to attain a higher fidelity prediction model, all the configurations are further assessed by additional statistical error and correlation metrics, namely Mean Absolute Percentage Error (MAPE), Mean Squared Error (MSE), Root Mean Squared Error (RMSE), Theil U2, Nash-Sutcliffe Efficiency (NSE) and Kling-Gupta Efficiency (KGE). Following the evaluation, the best activation function combination for the brake power, BSFC, BTE, NO_x , CO, and CO_2 ANN predictive models is the tansig-logsig configuration. As for Brake torque and HC, the tansig combination provides a better prediction. It can be conclusively shown from the study that the developed ANN models have a higher predictive accuracy as compared to the RSM model.

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LIST OF SYMBOLS

Abbreviations

| | |
|---|------------------------------|
| $^{\circ}\text{CA}$ | degree of crank angle |
| AP_e | average effective pressure |
| $\text{CH}_3\text{CH}_2\text{CH}_2\text{CH}_2\text{OH}$ | n-butanol |
| $\text{CH}_3\text{CH}_2\text{CHOHCH}_3$ | 2-butanol |
| $\text{C}_n\text{H}_{2n+1}\text{OH}$ | alcohol structure formula |
| CH_3OH | methanol structure formula |
| $\text{C}_2\text{H}_5\text{OH}$ | ethanol |
| $\text{C}_3\text{H}_7\text{OH}$ | propanol |
| $\text{C}_4\text{H}_9\text{OH}$ | butanol |
| $(\text{CH}_3)_2\text{CH}_2\text{CHOH}$ | iso-butanol |
| $(\text{CH}_3)_3\text{COH}$ | tert-butanol |
| P_e | effective power |
| P_{max} | maximum pressure |
| Nm | newton metre |
| R | correlation coefficient |
| R^2 | coefficient of determination |
| T_{e_x} | exhaust gas temperature |
| rpm | revolutions per minute |
| -OH | hydroxyl group |

LIST OF ABBREVIATIONS

Abbreviations

| | |
|--------------------|--|
| AFR | air fuel-ratio |
| ANN | artificial neural network |
| AI | artificial intelligence |
| bTDC | before top dead centre |
| BDC | bottom dead centre |
| BMEP | brake mean effective pressure |
| BSFC | brake specific fuel consumption |
| BTE | brake thermal efficiency |
| BVP | butanol Volume Percentage |
| CI | compression ignition |
| CO | carbon monoxide |
| CO ₂ | carbon dioxide |
| COP21 | Conference of the Parties |
| COV | coefficient of variation |
| DA | direction accuracy |
| DLS | damped least-squares |
| ECI-multi | electronically controlled multi-point fuel injection |
| EU | european union |
| FF | feedforward |
| FIT | fuel injection timing |
| FIP | fuel injection pressure |
| G100 | Gasoline |
| GBu5 | 5% 2-butanol + 95 gasoline |
| GBu10 | 10% 2-butanol + 90% gasoline |
| GBu15 | 15% 2-butanol + 85% gasoline |
| GNA | gauss–newton algorithm |
| GHG | greenhouse gasses |
| H _n OME | honne oil methyl ester |
| IMEP | indicated mean effective pressure |
| HC | unburned hydrocarbon |
| HCCI | homogenous charge compression ignition engine |

| | |
|------------------------|--------------------------------|
| KGE | klings-gupta efficiency |
| LR | linear regression |
| LHV | lower heating value |
| Logsig | hyperbolic log-sigmoid |
| MAPE | mean absolute percentage error |
| CAD HRR _{max} | location of heat release rate |
| CAD P _{max} | location of maximum pressure |
| CuHRR | cumulative heat release rate |
| CI | compression ignition |
| MLP | multilayer perceptron |
| MSE | mean square error |
| MSRE | mean square root error |
| NO _x | nitrogen oxides |
| PME | peanut methyl ester |
| ppm | parts per million |
| purelin | linear function |
| RBF | radial basis function |
| RSM | response surface method |
| RSE | relative standard error |
| RMSE | root mean square error |
| rpm | revolutions per minute |
| SI | spark ignition |
| SHL | single hidden layer |
| SOHC | single overhead camshaft |
| Tansig | hyperbolic tangent sigmoid |
| TDC | top dead centre |
| THL | two hidden layer |
| trainbfg | quasi-Newton backpropagation |
| trainrp | resilient backpropagation |
| trainscg | scaled conjugate gradient |
| traingdx | variable learning rate |
| UN | united nation |
| WCO | waste cooking oil |
| WTO | wide throttle open |

| | |
|-------------------|--|
| NSE | nash–sutcliffe coefficient of efficiency |
| <i>n</i> -butanol | primary butyl alcohol |
| 2-butanol | secondary butyl alcohol |

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