

Optimization of Hybrid Electric Bus Control Strategy with Hybrid Optimization Algorithm

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Abstract: control strategy is one of the most decisive techniques in Hybrid Electric Bus (HEB) and directly influences the dynamic performance and fuel economy. For achieving the best fuel economy and keeping the battery for a long time, First, power analytic control strategy was built; then, the hybrid optimization algorithm (HOA) based on Multi-island genetic Algorithm (MIGA) and NLPQL was built by ISIGHT software. HOA is adopted in control strategy parameters of HEB optimization. The results show that the best result can be obtained in few iterative times by HOA, the calculation time was reduce by 12 hours, the fuel economy was improved by 12% and find the rules between control strategy parameters and fuel economy 、 the balance of the battery state of charge(SOC).

Introduction

The core and base of Hybrid electric Bus (HEB) is multi-energy powertrain control system. The objective of the HEB's control strategy reduce fuel consumption 、 exhaust emission and keep the battery for a long time to save when the automobile remains its original dynamics^[1-3]. As core of power distribution directly, the control strategy affects the dynamic, economic of the vehicle. So it is important to research the optimization of control strategy in theoretical study and application.

It is transferred the multi objective problem to single objective problem using the method of weighting, its pareto optimal solution was obtained with genetic algorithm, but Premature convergence has been a main problem in genetic algorithms ,which converges on local optimum not global optimal solution^[4-6]. A software which combines genetic algorithm with traditional gradient algorithm SQP is put forward and it makes the best use of the ability of the global search ability of genetic algorithm and gradient algorithm fast convergence. But it use ADVISOR, which is largely different from the actual situation to get the imprecise result^[7-8]. According to combine MATLAB and AVL Cruise, an genetic algorithm to optimize control strategy was adopted, but it not only needs write genetic algorithm by hand, but also takes long time to optimize^[9].

In order to solve the problems occurred during the control strategy optimization, the hybrid optimization algorithm with composing of MIGA and NLPQL is built by SIGHT software platform .After global exploration by MIGA get the optimum areas, and then NLPQL come into optimum areas to get the local optimization. This optimization algorithm is used to optimal control strategy of HEB to get the best fuel economy and keep the balance of battery SOC.

HEB simulation analysis model and control strategy

Simulation analysis model is a powerful tool to study HEB control strategy. As a forward simulation software, compared with reverse simulation software, AVL Cruise not only realize dynamic calculation ,but also have advantages of the higher calculation accuracy and realistic .So it is fit to the research of control strategy.

1.1. HEB simulation analysis model based on AVL Cruise

HEB uses a typical dual-clutch and Single-axle Parallel configuration show as fig.1. It includes an engine, battery, a motor and AMT. Engine can work independently to supply the power. The driving motor is not only driving unit but also a system being used starting engine and generating electricity (ISG). The battery connects the motor, which can keep the balance of the charging and discharging based on control strategy. Some key parameters of HEB as show in Table 1.

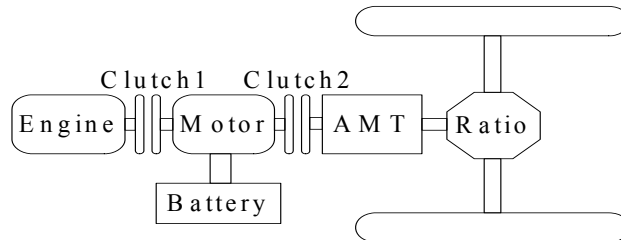


Figure 1 . Structure of HEB

Table 1 . Several key parameters of HEB

	Parameter	Type and Value
Engine	Type	4-cylinder diesel engine
	Max Power[kw]	132
	Max Torque(Speed) [kw/r·min ⁻¹]	655(250)
Battery	Type	lithium battery
	Capacity [Ah]	70
	Rated voltage[V]	358
Motor	Type	Ac induction motor
	Rated power [kw]	50
	Rated speed[r·min ⁻¹]	1450

1.2 Based on the analytic power control strategy of HEB

For the feature of HEB, the power analytic control strategy was build, which based on the power required of driver. In order to get the required torque of the HEB, driver give signals by electronic accelerator pedal or brake peda to distribute the torque from engine and motor and give consideration to the balance of the battery SOC by control strategy. Torque analysis for driver as show in fig 2

According to the layout and dynamic performance of HEB, when required torque is positive, the power management mode was established as show in fig 3. Engine optimal torque shows as $T_{emax} \cdot Q_{emax}$, engine cutoff torque shows as $T_{emin} \cdot Q_{emin}$, the engine launch speed shows as n_{low} .

The number describe as the work modes. When required torque is upper than $T_{emax} \cdot Q_{emax}$, the work mode is the hybrid mode shows as number 1; When required torque between $T_{emax} \cdot Q_{emax}$ and $T_{emin} \cdot Q_{emin}$. the work mode is engine work lonely mode shows as number 2; When the battery SOC is lower than its low limit and required torque is upper than $T_{emin} \cdot Q_{emin}$, engine works in $T_{emax} \cdot Q_{emax}$, an additional torque to charge the battery shows as number 3. When required torque is lower than $T_{emin} \cdot Q_{emin}$ and the battery SOC is lower than its low limit, engine supply drive torque and additional torque to charge the battery show as number 4. When required torque is lower than $T_{emin} \cdot Q_{emin}$ and the battery SOC is lower than its low limit, motor drive HEB show as number 5.

When required torque is negative or zero, according to battery SOC, whether recover the breaking energy.

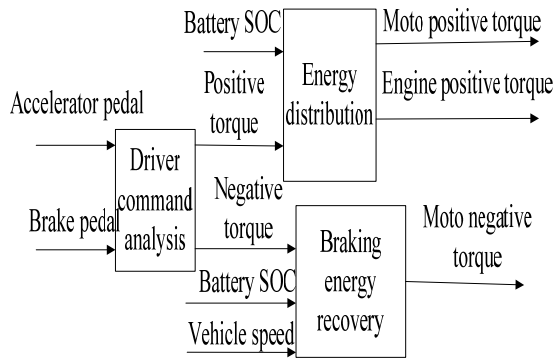


Fig. 2 . Energy management mode

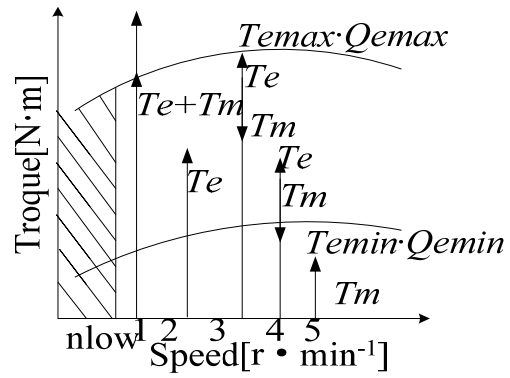


Fig. 3 . Torque analysis for driver

Optimization of HEB control strategy

The aim of optimization for HEB control strategy is to get to the best economy, lowest emission and keep a long life for battery without sacrificing the dynamic performance, which is the goal Optimization of HEB control strategy. In essence, this leads to the solution of a constrained nonlinear programming problem whose expressions are as follow:

$$\begin{cases} \min F(x) \\ s.t. \quad g_j(x) \geq 0 \quad j=1,2,\dots,m \\ \quad \quad x_j^{max} \leq x_j \leq x_j^{min} \quad i=1,2,\dots,n \end{cases} \quad (1)$$

$F(x)$ is the objective function, $\{A(x)B(x)\dots W(x)\}$ are the different function, x_j the control strategy parameter. x_j^{max} and x_j^{min} are the control strategy parameter's upper value and lower value; $g_i(x)$ is the constraint condition.

2.1 Objective function of HEB

In this paper, optimization of HEB control strategy aims at the reduction of fuel consumption and preventing the battery from over charging or over discharging to keep the long life of the SOC while maintaining the driving performance .However, minimum fuel consumption and minimum fluctuation of battery SOC are conflicting in nature, so the objective function can be described as follows:

$$\min \begin{cases} F(x) = Q_{fuel}(x) \\ F(x) = |W_{(SOCend - SOCini)}(x)| \end{cases} \quad (2)$$

Where $Q_{fuel}(x)$ is the function of minimum fuel consumption; $|W_{(SOCend - SOCini)}|$ is minimum fluctuation of battery SOC. $SOCend$ and $SOCini$ is the final state value and initial state value.

2.2 Control strategy parameters for optimization

HEB control strategy is a complex, nonlinear question, which has much more different control parameters ,but it is almost impossible to optimize all of the parameters .In this paper, we choose these independent parameters, which have the most notable effect on the error probability and can be controlled directly are used as the design variables of the objective function as table 2 shows. C_SOC_hi means the up limit coefficient, C_SOC_low means the lower limit coefficient, C_Trq_hi means engine optimal torque coefficient, C_Trq_low means engine cutoff torque coefficient.

Table 2. Pre-optimization control strategy parameters

Variable name	C SOC hi	C SOC low	C Trq hi	C Trq low
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Lower bound	0.5	0.05	0.5	0.05
Upper bound	1	0.5	1	0.5

2.3 Constraint conditions for optimization

The optimization of control strategy meets driving performance, but there are more and more driving performance parameters to be chosen for HEB. In this paper, the driving parameters of HEB are in accordance with prototype bus as described in table 3.

Table 3. Constraint conditions

Constraint	Description
Maximum speed[km/h]	≥ 80
Maximum inclination	$\leq 12\%$
Acceleration time of 0~50 km/h[s]	≤ 20
Chinese typical city bus absolute error[km/h]	≤ 3

HOA and optimization for HEB

As an excellent global optimum algorithm, Genetic Algorithm has been applied widely in the optimization of HEV, but Genetic Algorithms get local optimization too easily and converge slowly. In order to get over the disadvantages of the slow convergence speed and stagnation behavior, a new hybrid optimization algorithm (HOA) that combine MIGA and NLPQL is proposed.

3.1 Design of HOA

MIGA^[10] can divide the whole swarm into several sub-swarms, executes the operation of selection, cross over and mutation in a subpopulation, and periodically do immigration operation among the different islands. The pseudo-and some individual regular transfer from the different sub-swarms, this transfer make MIGA higher effective, avoid premature convergence and apt to combine with other algorithms. But It also needs to spend much time to get optimization. As a local numerical optimum on method, NLPQL^[11] can solve non-linear, constrained optimize problems well, the algorithm combines NLPQL and genetic algorithm together, which makes genetic algorithm in the early stage play a powerful global search. It is easy to converge to the global optimum solution; in the later stage, NLPQL are used to deal with the overall situation of pre-optimum solution, And it makes full use of NLPQL's latter part of the power of local search and eventually converges to the local optimal solution.

3.2 Optimization for control strategy with HOA

The HOA been integrated in ISIGHT software platform, which combined AVL-Cruise and MATLAB to optimize the control strategy of HEB. The sketch map of integration as show in fig 4. The optimized flow as followed:

- (1) Initialization of the control strategy and parameter design
- (2) The sub-swarm were randomly assigned, and make the generation evolution.
- (3) Individual fitness should be calculated under the dynamic constraint. According to the order of probabilities P_c , the evolution was operated order crossover operation, and according to the order of probabilities P_m , the mutation operated.
- (4) During the process of exploring the objective value in every iteration step, some individuals transfer from different islands to get the optimization.
- (5) When MIGA finish, NLPQL locate the optimal areas to search the optimization result automatically.

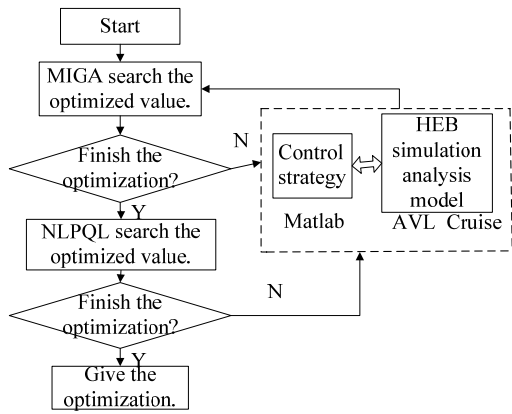


Fig. 4. The sketch map of integration

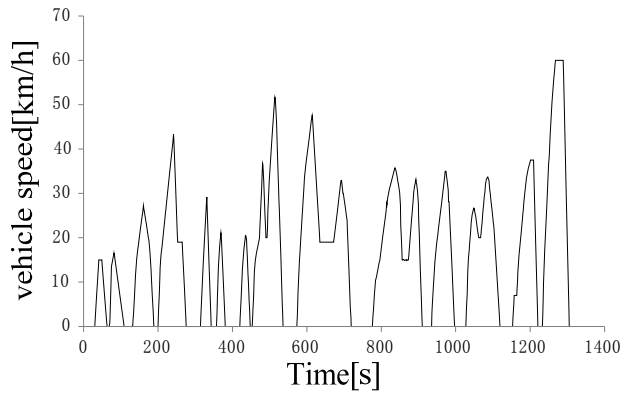


Fig. 5. CBC

Simulation Results And Analysis

In order to validate the validity of HOA, the vehicle simulation model is constructed in AVL Cruise, control strategy is constructed by MATLAB/Simulink. HOA is constructed by ISIGHT. Simulation works with typical city-bus driving-cycle in China (CBC) shows as fig 5. After 2000 generations evolved by MIGA alone , 1000 generations evolved by HOA, included 40 gradient optimization by NLPQL, and MIGA and HOA have the same numbers of island and the crossover and mutation rates The simulation optimal result used MIGA 、 HOA and non-optimal result of control strategy are listed in Table 4.

Table 4. Result of simulation

Optimization variables and Objective	Non-optimal	MIGA	HOA
C_Trq_hi	0.8	0.92	0.99
C_Trq_low	0.3	0.28	0.26
C_SOC_hi	0.8	0.82	0.79
C_SOC_low	0.3	0.43	0.44
Fuel consumption[L/100km)	31	28.23	27.98
Power consumption[Kwh/100km]	2.23	1.23	-0.62
Integrated fuel consumption[L/100km]	31.69	28.59	27.78

Optimal procedures for fuel consumption and power consumption with HOA as show in fig 6.in the early evolution stages, it searches more quickly and efficiently in the whole global optimization. After the evolution of four hundred generations, it converges to the global optimization .but some evolved individuals escape from local optimal area to explore the global areas for overcoming the premature. At the last time of optimal procedures, NLPQL locate the optimum areas automatically, which eventually converges to the optimal solution by means of the gradient technique quickly and can't explore global optimum areas.

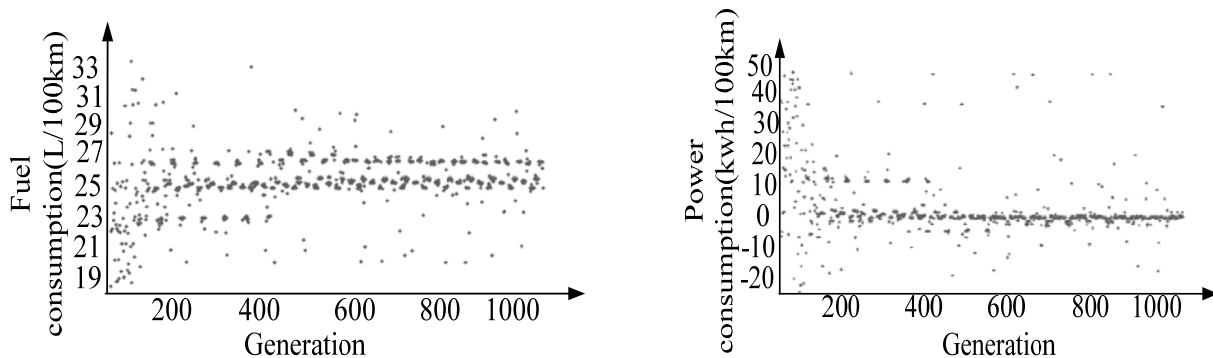


Fig. 6. Optimal procedures for fuel consumption and power consumption

Compared with the case of non-optimization, the Hundred-kilometer power consumption is reduced from 2.23(kwh/100km) to -0.62(kwh/100km), battery SOC maintains balance basically.

The hundred-kilometer fuel consumption reduces by 10%, integrated fuel consumption reduced by 12%. Compared with MIGA and HOA, Simulation results show that it's no advantage for the fuel consumption and power consumption basically. But it take 24 hours to get the optimum solution by MIGA and 12 hours to get the global optimization by HOA, it greatly enhanced search efficiency compared with MIGA.

Relationship between control strategy parameter and objective function shows as fig 7. In order to keep the balance of the battery SOC, fuel consumption should keep from 27.5(L/100km) to 28.4(L/100km);the engine optimal torque coefficient keep from 0.88 to 0.99;the engine cutoff torque coefficient keep from 0.2 to 0.27.;the battery SOC low limit coefficient keep from 0.2 to 0.27and the battery SOC up limit coefficient should keep from 0.75 to 0.82.the result shows that fuel consumption and power consumption is a pair of a mutually conflicting objective functions, the balance of the battery SOC and reducing the fuel consumption depend on multiple control strategy parameters. These results can provide important design guidance of control strategy for hardware-in-loop test.

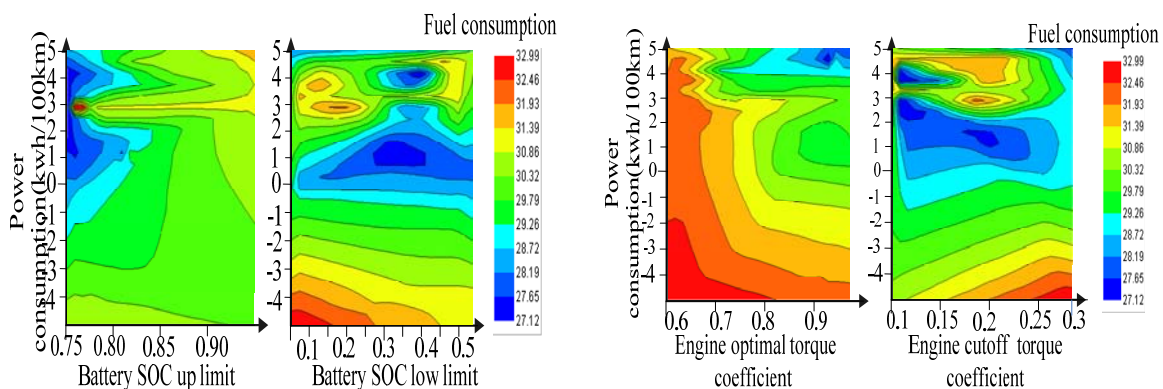


Fig.7. Relationship between control strategy parameter and objective function

Conclusion

The HOA been integrated in ISIGHT software platform, which combined AVL-Cruise and MATLAB to optimal the control strategy of HEB. Compared with using manual programming, it takes little time to design HOA .The simulation result shows that:

(1)The calculation results show that HOA improve fuel economy and keep the balance of the battery SOC when the automobile remains its original dynamics. Hundred-kilometer integrated fuel consumption is reduced by 12% and kept the balance of the battery SOC by -0.67.

(2)Compared with MIGA, HOA can get global optimal solution and reduce search time, optimization time was shortened by 12 hours.

(3)Rule between control strategy parameters and objective function is obtained, which give valuable guidance for hardware-in-loop test.

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