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## The Study For Convergence Enhanced Genetic Algorithm Development

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## THE STATUS OF CURRENT STUDY

Two major problems of conventional GA

- A low precision, especially for highly multi-modal functions, due to premature convergence
- A lack of stability with CGA when treating with functions presenting narrow valleys

The current study

- Hybrid method
  - CGA : global search algorithm + DFP : local search algorithm
  - CGA : global search algorithm + LZGA : local search algorithm ← New Method
- Successive zooming genetic algorithm ← SZGA ← New Method

Introduce the stability of the optimized solution based on a theory of probability

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## HYBRID ALGORITHM I (MGA+DFP)

The SGA (simple GA) is improved to MGA (micro GA) by using some techniques like tournament selection along with the elitist strategy.

The DFP method as a local search algorithm can converge to a local optimum value without preliminary knowledge.

But the proposed hybrid algorithm converges to a global optimum value.

**Flow chart combining MGA with DFP**

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## HYBRID ALGORITHM II (MGA+LZGA)

**Flow chart combining MGA with local GA**

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## CONCAVITY CONDITION

A MGA is performed generation by generation until the objective function changes no more, then we have approximate optimum solution at  $Z_{local}$ .

The gradients of the objective function along the design variables are checked, if the concavity condition (Hummelblau 1977) is satisfied at the boundary of small zoomed area.

If the condition is not satisfied, the small zoomed area is increased by .

After many iterations we finally have the concavity conditions at the boundary of final zoomed area centered at  $Z_{local}$ .

With the elitist solution from the global GA (approximate optimum solution,  $Z_{global}$ ) and the concavity condition, we have the optimum point within the final zoomed area.

**Signs of gradients at concavity condition**

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## SUCCESSIVE ZOOMING GENETIC ALGORITHM

**Parameter Definition**

- $n$  : maximum number of variables
- $G$  : number of total generations
- $g$  : number of sub-generations in a zooming
- $z$  : number of zooming operations
- $p$  : number of populations in a sub-generation
- $e$  : number of fitness evaluations at sub-generations in a zooming ( $n \times p \times e$ )
- $\delta$  : zooming factor

**Flow chart of SZGA**

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## SUCCESSIVE ZOOMING METHOD

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### Zooming Technique

Main idea : smart reduction of search space around the candidate optimum point, continuous zooming factor

Zooming procedure : search domain is reduced to  $(x_{opt} - (\alpha^k/2), x_{opt} + (\alpha^k/2))$ , then a new initial population is generated within the new boundaries and an MGA applied again.

Convergence criteria :  $\frac{x_{N_{zoom}} - x_{N_{zoom-1}}}{x_{N_{zoom}}} < \delta$

(error ratio =  $\delta = 10^{-6}$ ; will be different according to each problem)

If we use a  $\alpha$  greater than the value required for the specified value of reliability, that is the backtracking. This kind of backtracking is continuous while the backtracking in DFE is discrete (digital)

$X_1$  : Global variable(dimensionless)  
 $Z_1$  : Local variable(dimensionless in zoomed range)  
 $k : 1 \sim N_{zoom}$

Length of search range = 1

Length of search range =  $\alpha^{k-1}$

Length of search range =  $\alpha^k$

$X_1 : 0-1$   
 $Z_1 : 0-1$

$X_2 : X_{(k-1)opt} - \frac{\alpha^{k-1}}{2} - X_{(k-1)opt} + \frac{\alpha^{k-1}}{2}$   
 $Z_2 : 0-1$

$X_{k+1} : X_{kopt} - \frac{\alpha^k}{2} - X_{kopt} + \frac{\alpha^k}{2}$   
 $Z_{k+1} : 0-1$

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## RELIABILITY OF SOLUTION

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### P: Optimum point within zoom interval

$X$  : local coordinate,  $Z$  : global coordinate

The range of  $X$  and  $Z$  are 0-1.0, plus the fitness  $F(Z)$  is also between 0 and 1.0

The relation of  $Z_{zoom} = \alpha^{N_{zoom}-1}$  is accomplished, where  $Z_{zoom}$  as related to the resolution of the solutions, means the dimensionless search domain length after  $N$  zooming.

For example, when  $N_{zoom} = 6$  and  $\alpha = 0.1$ ,  $Z_{zoom}$  becomes  $1.0E-05$

$N_{zoom}$	1	2	3	4	5	6
$Z_{zoom}$	$\alpha^1$	$\alpha^2$	$\alpha^3$	$\alpha^4$	$\alpha^5$	$\alpha^6$

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## RELIABILITY OF RANDOM ALGORITHM

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In the case of a random search, the probability that at least one search point falls within the zoom interval :

$$P_{rand}^{at} = 1 - (1 - (\alpha/2)^{N_{zoom}})^{N_{zoom}}$$

$\alpha$  : the probability that one search point falls within  $\alpha$  for the 2-D case (red zone) range during one function evaluation.

$1 - \alpha$  : the probability that one search point falls out of  $\alpha$  for the 2-D case range during one function evaluation.

$(1 - \alpha)^{N_{zoom}}$  : the probability that every search point falls out of  $\alpha$  for the 2-D case range during a function evaluation.

$1 - (1 - \alpha)^{N_{zoom}}$  : the probability that at least one search point falls within  $\alpha$  for the 2-D case range during an function evaluation.

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## RELIABILITY OF GENETIC ALGORITHM

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In the case of a GA, the probability becomes larger than or equal to the probability of using a random search.

$$P_{GA}^{at} \geq P_{rand}^{at} = 1 - (1 - (\alpha/2)^{N_{zoom}})^{N_{zoom}}$$

After progressing 100 generation, the probability that at least one search point falls within the zoom interval

$$P_{GA}^{at} = 1 - (1 - (\alpha/2)^{N_{zoom}})^{N_{zoom}}$$

To express the degree of improvement, the improvement factor  $D_{zoom}$  is introduced, the exact definition for which can be defined as the ratio of the number of individuals within the zoom interval between a random search and a GA

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## FINAL RELIABILITY

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After  $N$  zooming, the final existence probability of a search point within the zoom interval, which is the probability the optimum point will not be missed.

The GA case is

$$P_{GA}^{at} = 1 - (1 - (\alpha/2)^{N_{zoom}})^{N_{zoom}}$$

For example, consider two-dimensional cases with the following parameters :

$N_{zoom}$	1	2	3	4	5	6
$P_{GA}^{at}$	0.9182	0.9933	0.9999	0.9999	0.9999	0.9999
$P_{rand}^{at}$	0.6526	0.9672	0.9999	0.9999	0.9999	0.9999

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## APPLICATION OF STRUCTURE OPTIMUM DESIGN

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Selecting the location of a radial arm is very important because the size of the member supporting the working stress is affected by the location of the radial arm.

Weight minimization is obtained by minimizing the volumes because the density is constant. (Structure Engineering and Mechanics)

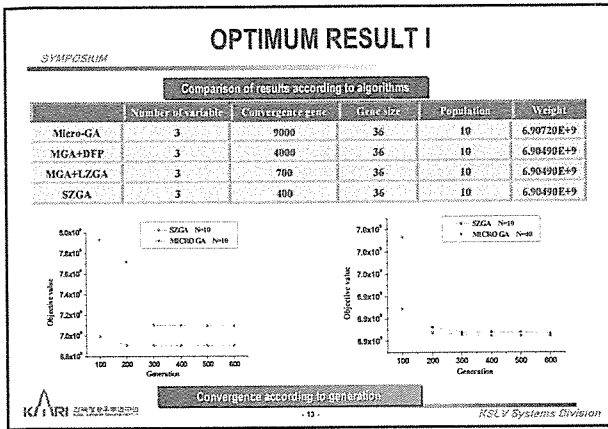
### Formulation

Minimize  $V_{gate} = \max. \text{ volume of gate,}$

subject to  $\sigma \leq \sigma_{allow}$

$n$  : dimensionless lengths

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### OPTIMUM RESULT II

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Optimum results of a 3-arm type

Three-arm	Value of optimization	Optimum value	
		Girder only	Girder and Arm
Ration of length	$R_1$	1.66670E-01	9.41670E-02
	$R_2$	3.33330E-01	4.062740E-01
	$R_3$	3.33330E-01	3.67890E-01
	$R_4$	1.66667E-01	7.52030E-02
Reaction (kg/mm)	$R_5$	3.01600E+01	3.04620E+01
	$R_6$	1.09030E+01	4.91990E+01
	$R_7$	9.10070E+01	6.24090E+01
Mument (kg mm/mm)	$M_8$	4.65420E+04	2.92310E+04
	$M_9$	6.97690E+03	1.23310E+04
	$M_{10}$	8.10750E+04	2.52310E+04
Section modulus (mm)	$Z_1$	4.07610E+06	1.26850E+06
	$Z_2$	3.37810E+06	1.05130E+06

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### CONCLUSION

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- ✓ Two hybrid algorithm and SZGA were proposed to improve the local-tuning ability
  - GA+DFP incorporates the advantages of both a genetic algorithm and local search algorithm
  - LZGA used GA for global solution and LGA for local solution
  - SZGA presents a GA with successive zooming that involves a smart reduction of the search space : The reliability of the identified solution is deduced from a probability theory and an improvement factor  $p_{im}$  is introduced to express the degree of probability amelioration.
- ✓ The weight of the radial gate was minimized using an optimization technique involving a genetic algorithm.

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