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1 Introduction

1.1 Problem Background



180.6 km^3 $9500 \text{ m}^3/\text{s}$

fi

fi

fi

fi

fi

fi

fi

fi

1.2 Our Work

i	
X	
$Series(X)$	
$Supply(X)$	
$Risk(X)$	
P	
$Q(X; t)$	
$k(X)$	
N	
n	
M	
L	
d	
α	
$h(X)$	

4.1 Behavior of Water Flow

$$v = \frac{k}{n} R^{2/3} S^{1/2} \quad (1)$$

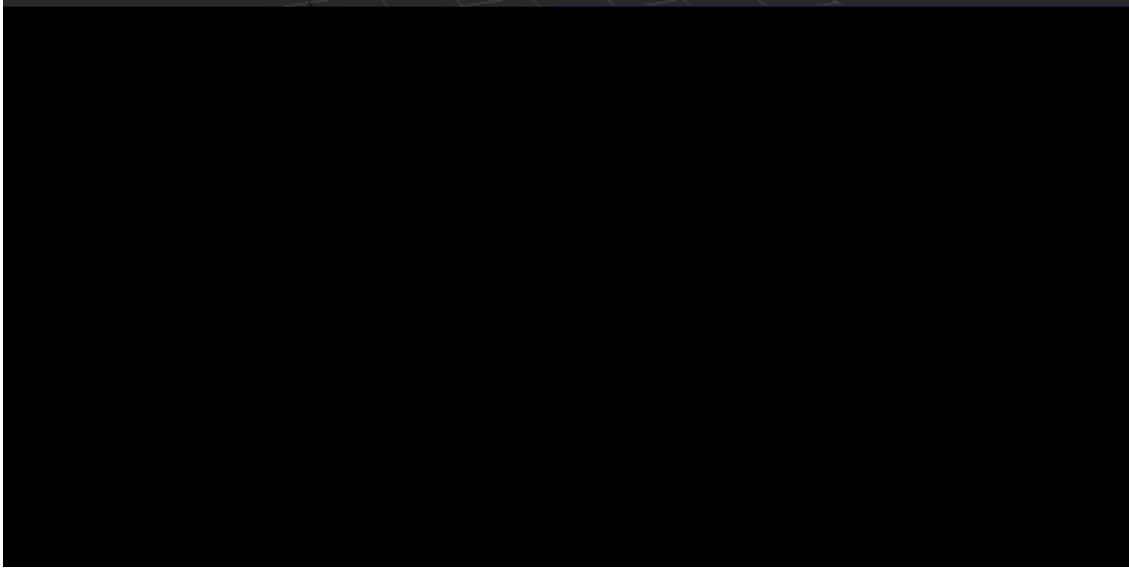
- v : velocity
- n : Manning's roughness coefficient
- R : hydraulic radius (L, f, m)
- S : slope (L/L)
- k : coefficient

$$CA(X) = \sum_{i=0}^n \frac{1}{3} (A_i + A_{i+1} + \sqrt{A_i \times A_{i+1}}) \times \Delta L$$

4.2 Analysis of A Single Dam

capacity storage

$$CA(X) = \sum_{i=0}^n \frac{1}{3} (A_i + A_{i+1} + \sqrt{A_i \times A_{i+1}}) \times \Delta L$$



- V
- A
- ΔL

$$\begin{aligned}
 CA(X) &= \frac{1}{3} \left(\frac{1}{2} d \times h + d \times h \right) \times \frac{h(X)}{k(X)} \\
 &= \frac{h(X)^2}{3k(X)} \left(\frac{3}{2} d \right) \\
 &= \frac{h(X)^2 d}{2k(X)} \quad ()
 \end{aligned}$$

building cost.

$$P = \rho g h \Delta$$

voir,

dam,

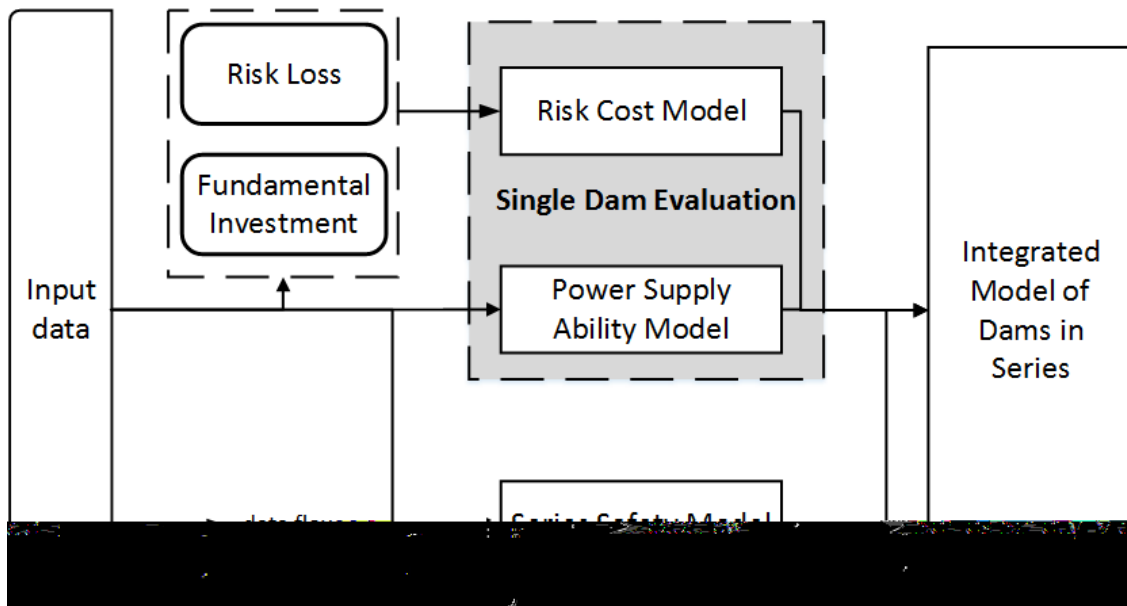
reser-

$\beta,$

$$\begin{aligned}
 Cost(X) &= \lambda \frac{\rho g h(X) \times d \times h(X)}{2\sigma} + \beta \\
 &= \lambda \frac{\rho g d h(X)^2}{2\sigma} + \beta
 \end{aligned}
 \tag{ }$$

4.3 iMoDS: Integrated Model of Dams in Series

4.3.1 Prerequisites



$$k(X) = \frac{dE}{dX} \Big|_{X=i}$$

$k(X)$

$k(X)$

4.3.2 Risk Cost Model

cost safety fi

the extent that a person focus more on safety than merely cost.

Cost - Fundamental Investment

P

$$= \frac{Cost(X)}{P^{1.1}} \quad ()$$

P

Safety - Risk Loss

Loss Risk

$$[1 - (1 - P)]$$

fi

collapse probability * (1+ indirect damage index)*Volumetric flow rate

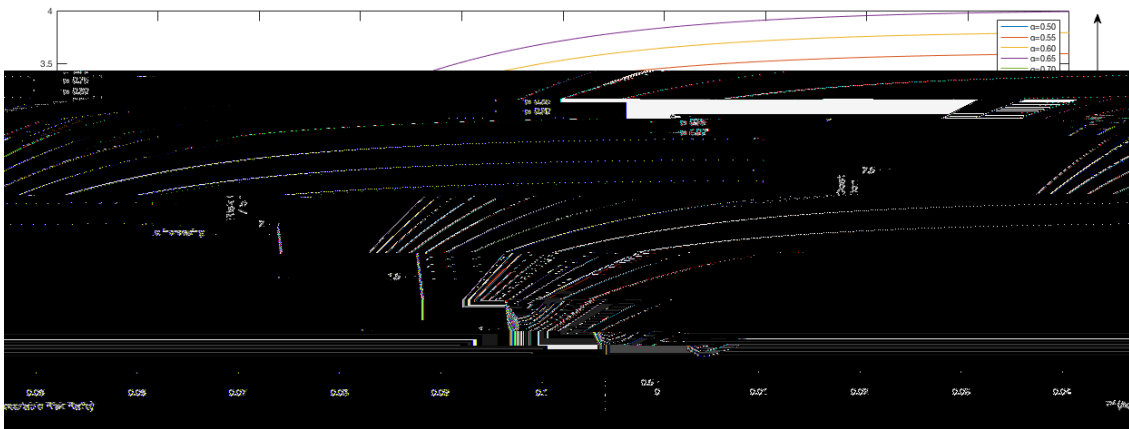
$$= [1 - (1 - P)] \alpha (1 + \lambda) Q(X; \bar{t}) \quad ()$$

α fi

$$Risk(X) =$$

P Risk Value.

α



X , P Risk Value, α

Observation 1

Illustration Risk Value P Fundamental Investment Risk Loss

Observation 2

Illustration f_i , α , P , Risk Cost

4.3.3 Series Safety Model

Series Safety Model

$$Series(X) = \frac{\bar{d}}{e^{\sum_{i=1}^{n-1} \frac{(d_i - \bar{d})^2}{\bar{d}^2}}} \quad ()$$

\bar{d}

4.3.4 Power Supply Ability Model

$$Supply(X) = \eta \rho g h(X) Q(X; \bar{t}) \quad ()$$

- η
- ρ
- g

4.3.5 Integration

- *Series(X)* **Positively** *Series(X)'*
- *Risk(X)* *Risk(X)'*
- *Supply(X)* **Positively** *Supply(X)'*

[0, 1].

$$x' = \frac{x - \min(x)}{\max(x) - \min(x)}$$

- x
- x'

$Risk(X)'$, $Supply(X)'$ **positively**, $Risk(X)$ **negatively**, $Series(X)'$, $Supply(X)$

$$Series(X)' = Series(X) \times \frac{n-1}{L}$$

$$Risk(X)' = \min(Risk(X))/Risk(X)$$

$$Supply(X)' = Supply(X) \times \frac{n-1}{L}$$

$Series(X) \in [-1, 0]$, **Final Value**, $Supply(X)$
Integrated Model of Dams in Series

$$\max \quad \text{Objective} = Series(X)' + \sum_{i=1}^n Risk(X)' + \sum_{i=1}^n Supply(X)'$$

$$\text{s.t. } X \leq L, \quad \forall X \in [1, n]$$

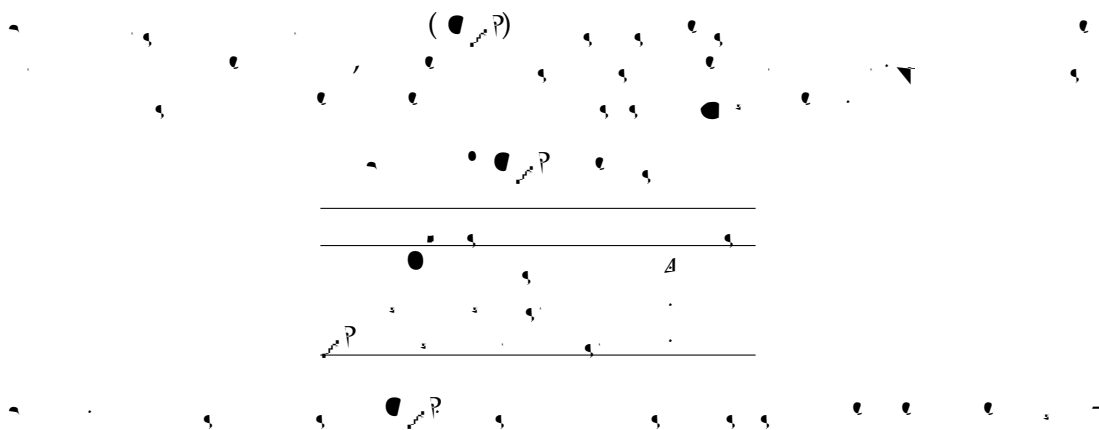
$$\sum_{i=1}^n CA(X) \geq M$$

$$10 \leq n \leq 20, \quad n \in \mathbb{Z}$$

$$X \geq 0, \quad \forall X \in [1, n]$$

$$\begin{aligned} \text{Objective} &= Series(X)' + \sum_{i=1}^n Risk(X)' + \sum_{i=1}^n Supply(X)' \\ &= \frac{\bar{d}}{e^{\sum_{i=1}^{n-1} \frac{(d_i - \bar{d})^2}{\bar{d}^2}}} \times \frac{n-1}{L} + \sum_{i=1}^n \min(Risk(X)) / \left\{ \frac{Cost(X)}{P^{1.1}} + [1 - (1 - P)^n] \alpha (1 + \lambda) Q(X; \bar{t}) \right\} + \sum_{i=1}^n \rho g \eta L k(X) Q(X; \bar{t}) \end{aligned}$$

4.3.6 Ranking Submodels with AHP



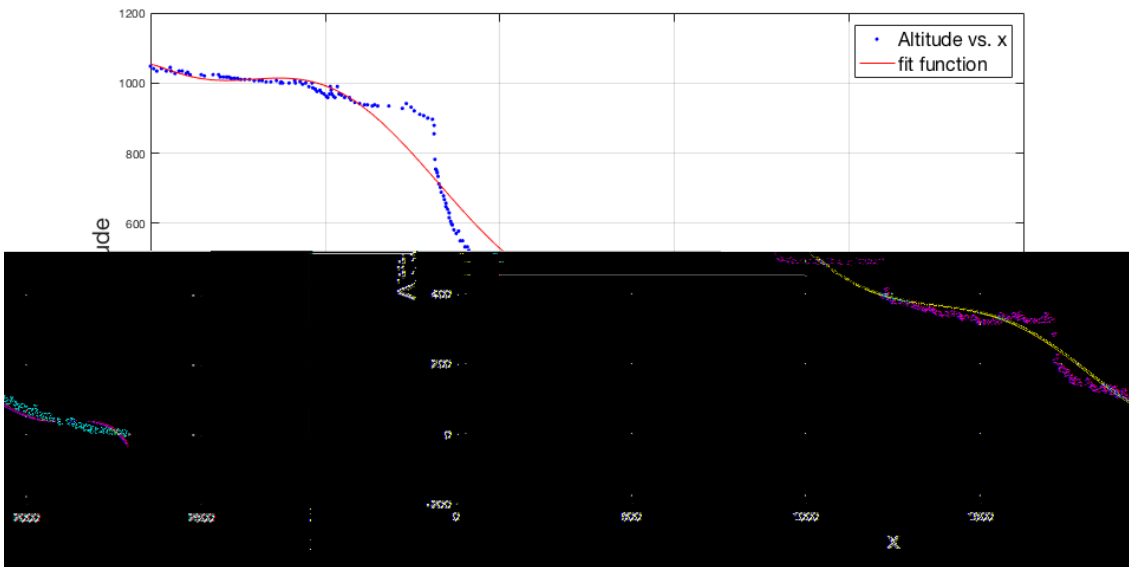
$$\begin{aligned}
 \max \quad & 0.35 \text{Series}(X)' + 0.54 \sum_{=1}^n \text{Risk}(X)' + 0.11 \sum_{=1}^n \text{Supply}(X)' \\
 \text{s.t} \quad & X \leq L, \quad \forall X \in [1, n] \\
 & \sum_{=1}^n CA(X) \geq M \\
 & 10 \leq n \leq 20, \quad n \in \mathbb{Z} \\
 & X \geq 0, \quad \forall X \in [1, n]
 \end{aligned}$$

5 Implementation

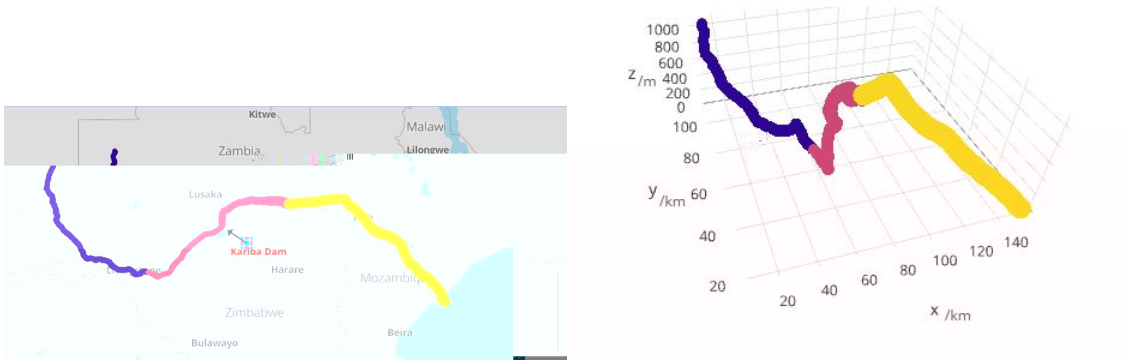
5.1 Data

eighth-power polynomials.

$$f(X)$$



$$Q(X; \bar{t})$$



()

fl

fl

5.2 Number and placement of the new dams

5.2.1 Whole Procedures

{X}

{Xi}

5.2.2 Determining the number and placement by genetic algorithm

fi

()

fi

fi, h(X)

{Xi}, P n. fi

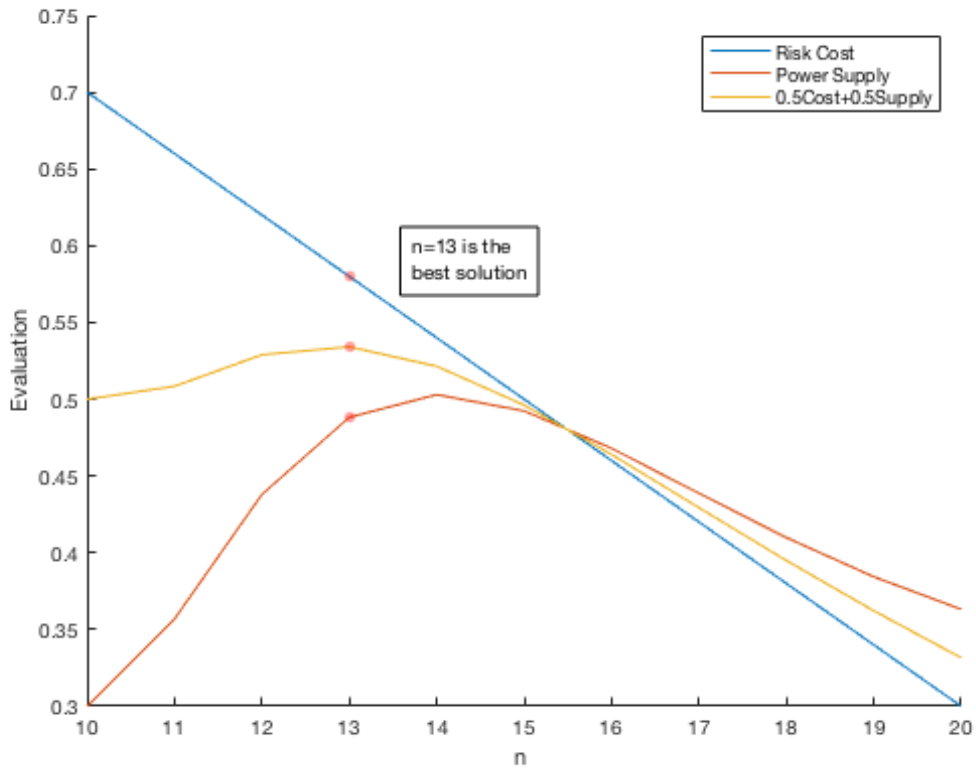
genetic algorithm

Risk Cost

min (i)

P (i)

M,



Final Value , n

Series(X) , , , , n.

fi , , , n = 13 , Pf = 0.0024.

1, 13, X

5.2.3 Determining the height of each dam

risk cost power supply fi fl
 P , h(X) , Series(X) , final value (iMoDS model)

6.2 Protection for Lake Kariba

fi

$$Protection = \sum_{i=1}^n CA(X)N(X_L, \sigma^2)$$

X_L

Protection

6.3 Guidance for Emergency Water Flow Situations

fi

fi

fi

fi

Method

fi

fi

fi

fi

fi

fi

fi

Method

fi

fi

fi

fi

6.4 Guidance for Extreme Water Flows

$Q(X; \bar{t})$ is the mean annual flood, $Q(X; t)$ is the flood with return period t .

The flood m is defined by the equation:

$$m = p + km$$

where p is the mean annual flood and k is the frequency factor.

$$\frac{dm}{dt} = -p - km$$

$$m = \frac{ce^{-kt} - p}{k}$$

$$m(0) = M, m(T) = 0$$

$$p = \frac{Mke^{-kT}}{1 - e^{-kT}}$$

The flood $Q(X; t)$ is defined by the equation:

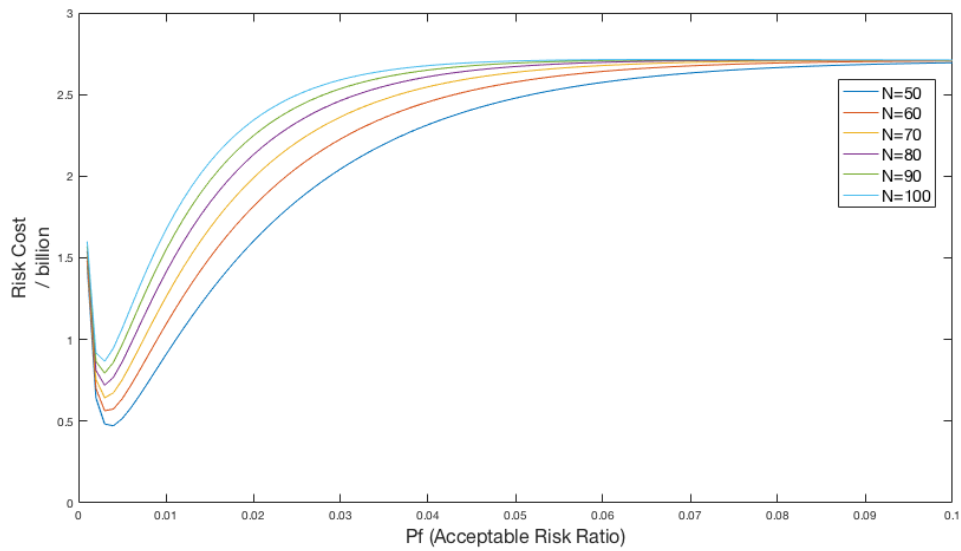
$$Q(X; t) = p + kQ(X; t)$$

where p is the mean annual flood and k is the frequency factor.

7 Model Analysis

7.1 Sensitivity Analysis

7.1.1 Impact of Planned Working Years N for a Dam



X , P Risk Value, N
 N Risk Loss, N

7.1.2 Impact of Extreme Condition ratio α

α

7.2 Strengths and Weaknesses

7.2.1 Strengths

The strengths of the project are its focus on the most vulnerable populations, its comprehensive approach to addressing the needs of these populations, and its strong partnerships with community organizations. The project's focus on the most vulnerable populations, including the elderly, the disabled, and the homeless, is a key strength. The project's comprehensive approach to addressing the needs of these populations, including providing food, clothing, and shelter, is another key strength. The project's strong partnerships with community organizations, such as the local food bank and the homeless shelter, are also a key strength.

7.2.2 Weaknesses

The weaknesses of the project are its limited budget, its lack of funding for transportation, and its limited reach. The project's limited budget is a key weakness, as it restricts the number of people that can be served. The project's lack of funding for transportation is another key weakness, as it prevents people from accessing the services. The project's limited reach is also a key weakness, as it only serves a small portion of the population in need.

Brief Assessment Report

$$FR(T) = \begin{cases} 3 \times 10^{-4} & 0 \leq T < 50 \\ 1 - e^{-\left(\frac{t+143.67}{100}\right)^3} & T \geq 50 \end{cases} \quad ()$$



Option 1: Repairing the existing Kariba Dam

Potential Cost

$$FR(t) = 1 - e^{-\left(\frac{t+143.67}{100}\right)^3}, \quad ()$$

t =

Y C,

$$294 + C \times \int_0^{-2017} FR(t) dt \quad ()$$

Benefits

$$B = \int_0^T (FR(t) - C) dt$$

Shortcomings

$$S = \int_0^T C dt$$

Option 2: Rebuilding the existing Kariba Dam

Potential Cost

$$C = 241.4 + \frac{241.4}{27.2} \times 1.8 = \$15.975$$

$$7440 + C \times \int_0^{2017} FR(t) dt$$

Benefits

$$B = \int_0^T (FR(t) - C) dt$$

Shortcomings

$$S = \int_0^T C dt$$

Option 3: Removing the Kariba Dam and replacing it with a series of ten to twenty smaller dams along the Zambezi River

Potential Costs

$$C = (1 + 10\% * n + 15\%) * \$15.975$$

$n = 13, \$39.139$

Benefits

$$B = \int_0^T (FR(t) - C) dt$$

Shortcomings

$$S = \int_0^T C dt$$

Appendices

Appendix A Implemented Genetic Algorithm

```

n=13;% need adjust
B_sub1=[zeros(n,1);0.00001];
%B_sub1 means the VLB of the Xi
B_sub2=[ones(n,1)*2100;0.1];
%B_sub2 means the VUB of the Xi
B=[B_sub1,B_sub2];
% constraint of Xi
initPop=initializega(n,B,'fitness');
[x endPop,bPop,trace]=ga(B,'fitness',[],initPop,[1e-6 1 1],
    'maxGenTerm',10000,'normGeomSelect',...
    [0.08],['arithXover'],[2],'nonUnifMutation',[2 10000 3]);
%3000 generations
xx=sort(x(1:14));
% sites selection sorts as ascending order
x=[xx,x(15)];
% the last one is the evaluation value

```

Appendix B Fitness Function

```

function[sol,eval]=fitness(sol,~)
n=13;
x=sol(1,1:n+1);
xx=sort(x(1:n));
x=[xx,x(n+1)];
p1=0;
p2=0;
cc=0;
N=50;
Ca=300;
BB=6000;
aa1=400;
rr=0.8;
pp=1005;
yy=0.5;
l=200;
Cmin=2e6;
Pm=1e8;
% above are the constant parameter setting
wei=[0.12,0.54,0.34];
% wei means the weight of three key function
for i=1:n
    k=my(x(i));
    Q=my2(x(i));
    R=(1-(1-x(n+1))^N)*Q*aa1*(1+rr);
    C=((Ca/n)*k+BB)*(1/(x(n+1))^1.1);
    Power=Q*pp*yy*k*1*9.8/1000;
    p1=p1+Power;
    p2=p2+C+R;
end
p1=p1/Pm
p2=Cmin/p2

```

```
for j=1:n-1
    cc=cc+((x(n)-x(1))/(n-1)-(x(j+1)-x(j)))^2;
end
Sa=((x(n)-x(1))/exp(cc/((x(n)-x(1))/(n-1))^2))/2100
eval=p1*wei(1)+Sa*wei(3)+p2*wei(2)
end
```
