



**Lecture 10:
Multi-Objective
Optimization Exercises**



Multi-Objective Optimization in GOSET

- GOSET employ an **elitist GA** for the multi-objective optimization problem
- **Diversity control algorithms** are also employed to prevent over-crowding of the individuals in a specific region of the solution space
- The non-dominated solutions are identified using the recursive algorithm proposed by **Kung** et al.



Schaffer's problem

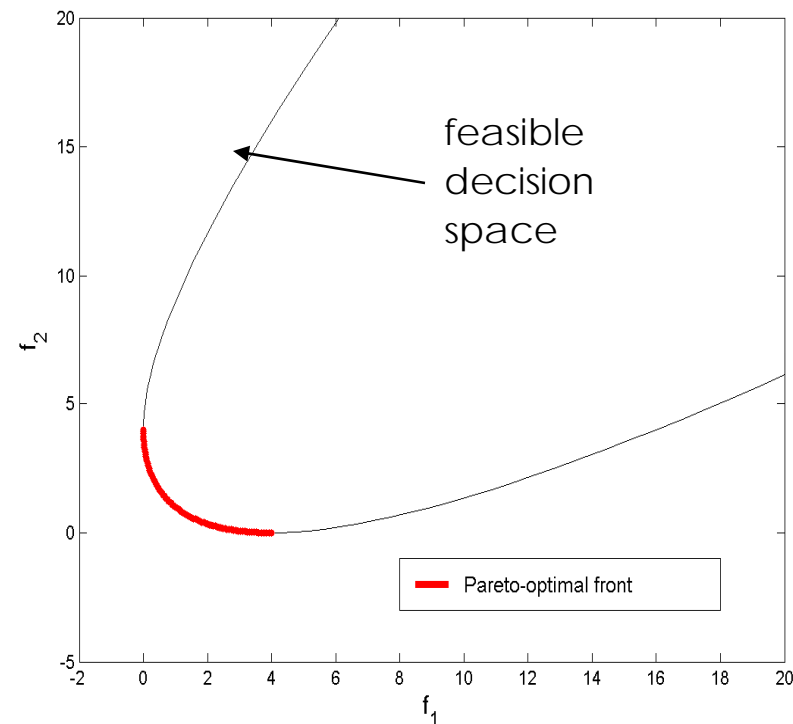
Problem Statement

- Schaffer's problem is a single variable problem with two objectives to be minimized

$$\min f_1(x) = x^2$$

$$\min f_2(x) = (x - 2)^2$$

$$-10 < x < 10$$



- The Schaffer's problem has Pareto optimal solutions $x \in [0, 2]$



Fitness Function

- Two objectives have positive values and they are to be minimized
- The fitness function values are defined to be the negative of the objective function values



schaffer.m

```
% Schaffer's problem
```

```
function [f] = schaffer (x)
```

```
f(1,1) = -x(1)^2;
```

```
f(2,1) = -(x(1)-2)^2;
```

schaffer_problem.m

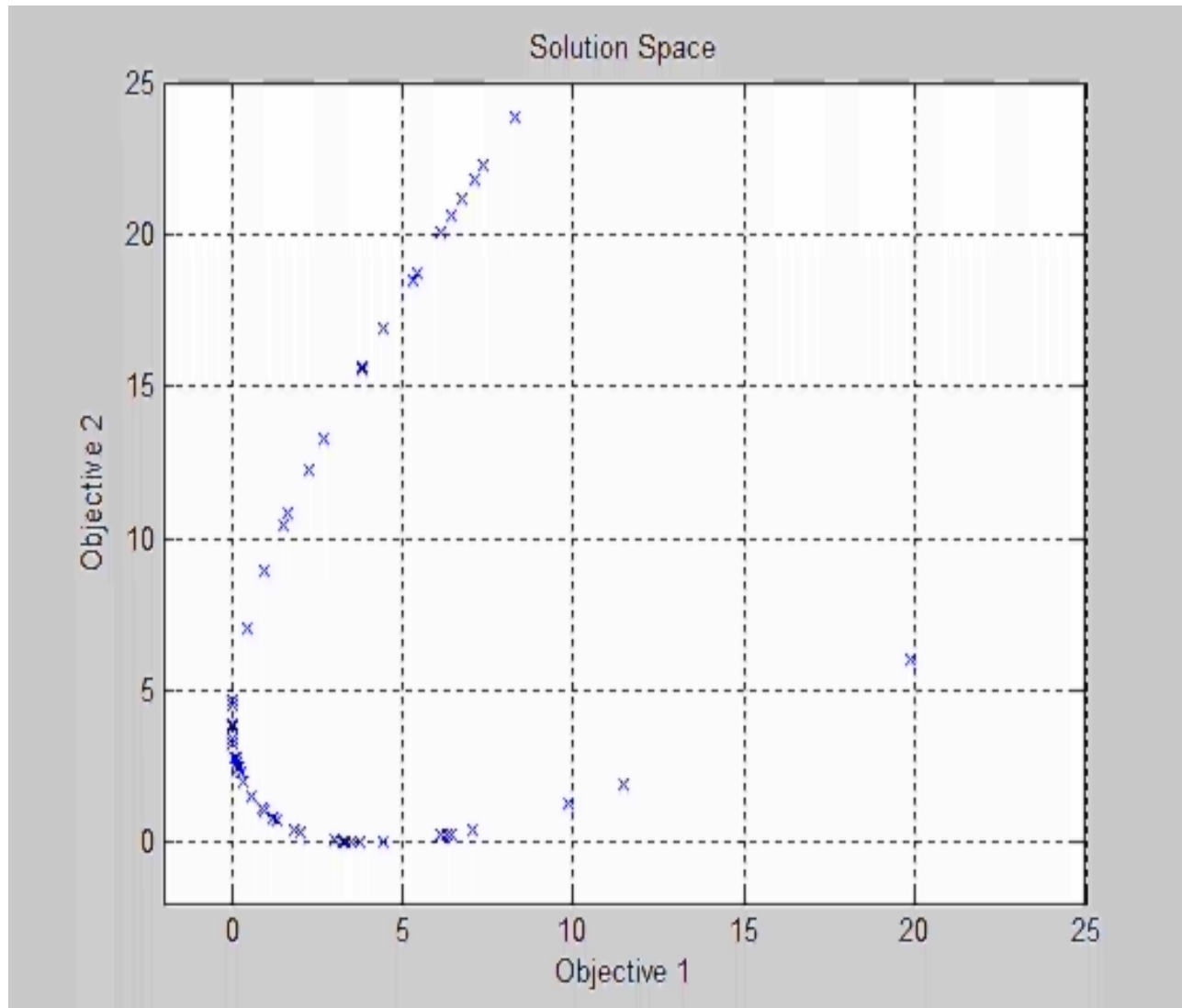
```
% Optimization of Schaffer's Problem

% Initialize the parameters
GAP = gapdefault(2);
GAP.fp_ngen = 20;
GAP.fp_obj = 0;
GAP.sc_alg = 1;
GAP.op_list = [];
GAP.pp_list = [1 2];
GAP.pp_style = [1 1];
GAP.pp_sign = [-1 -1];
GAP.pp_axis = [-2 25 -2 25];

%           x1
% gene      1
GAP.gd_min  = [ -10  ];
GAP.gd_max  = [  10  ];
GAP.gd_type = [   2  ];
GAP.gd_cid  = [   1  ];

[P,GAS]=gaoptimize(@schaffer_fit,GAP,[],[],[],[]);
```

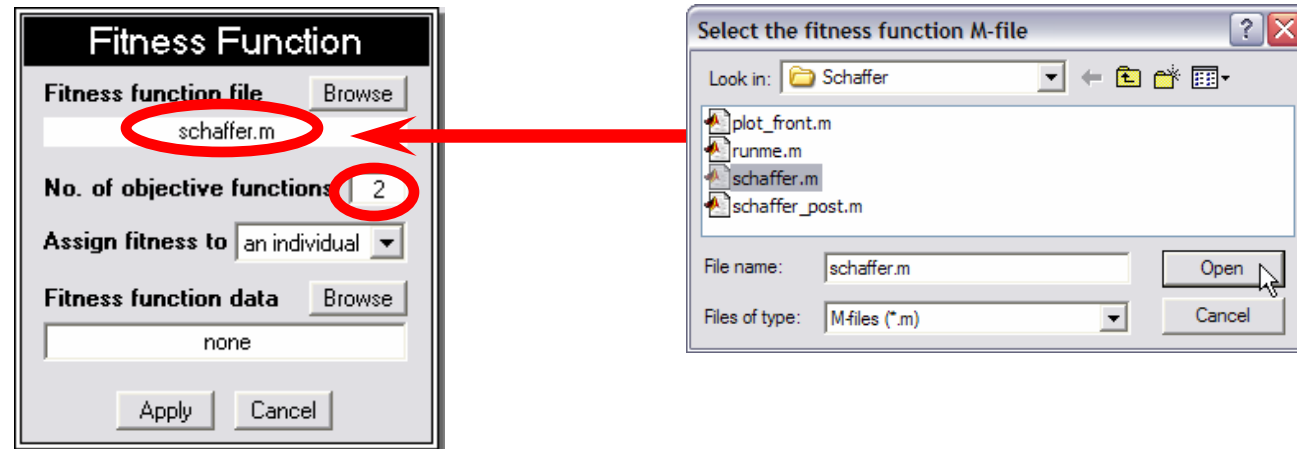
Pareto Plot (100 Generations)



GUI approach

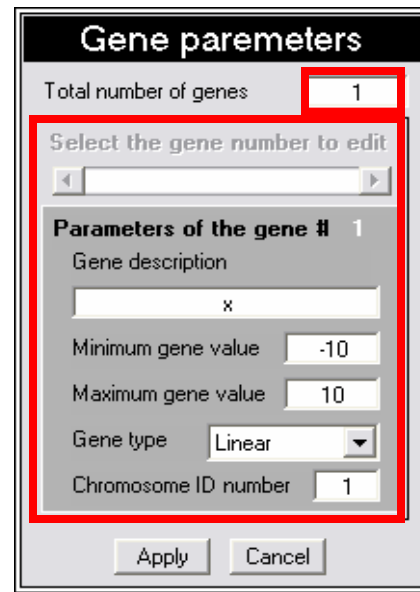
■ Define the fitness function

- In the fitness function window, browse and select the fitness function 'schaffer.m' defined in the command line approach
- Schaffer's problem has two objective functions and thus the number of objective functions is set to 2



■ Define the gene parameters

- There are only 1 parameter values in this problem, so set the total number of genes to **1**



Gene parameters

Total number of genes: 1

Select the gene number to edit

Parameters of the gene # 1

Gene description: x

Minimum gene value: -10

Maximum gene value: 10

Gene type: Linear

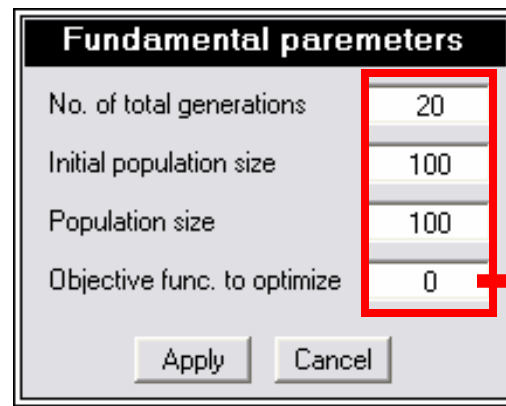
Chromosome ID number: 1

Apply Cancel

- The minimum and maximum gene value for x are to **-10** and **10**, **'Linear'** gene type is used

■ Define the fundamental parameters

- In the fundamental parameter input field, set the number of total generations to **20**, the initial population size to **100**, the population size to **100**
- As there are two objective functions, ‘objective function to optimize’ is set to **0** then click ‘apply’



Fundamental parameters	
No. of total generations	20
Initial population size	100
Population size	100
Objective func. to optimize	0

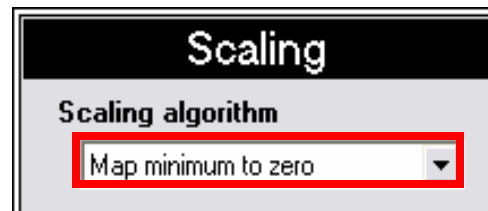
Apply Cancel

Caution!!!

Set ‘Objective function to optimize to 0 for multi-objective optimization problem

■ Define the scaling parameters

- The scaling algorithm is set to ‘Map minimum to zero’



Scaling	
Scaling algorithm	Map minimum to zero

■ Define the plotting parameters

- To display the Pareto plot, ‘Objectives for plot’ for the objective plot is set to ‘[]’ and the ‘Objectives for plot’ for the Pareto plot is set to ‘[1 2]’
- ‘Plot scale’ for the Pareto plot parameters are linear for all objectives
- Sign of fitness is set to -1 for both objectives
- Axis limits are defined as [-2 25 -2 25]

The screenshot shows a dialog box with three main sections: 'Objective plot', 'Pareto plot parameters', and 'Distribution parameter'. The 'Objective plot' section has 'Objectives for plot' set to an empty array '[]', 'Plot scale' set to '[1 1]', and 'Sign of fitness' set to '[1 1]'. The 'Pareto plot parameters' section has 'Objectives for plot' set to '[1 2]', 'Plot scale' set to '[1 1]', and 'Sign of fitness' set to '[-1 -1]'. The x-axis label is 'Objective 1', the y-axis label is 'Objective 2', and the z-axis label is 'Objective 3'. The 'Pareto plot title' is 'Solution Space'. The 'Axis limits for Pareto plot' are set to '[-2 25 -2 25]'. The 'Distribution parameter' section has 'Distribution plot type' set to 'plot histograms', 'Number of individuals to plot' set to 100, and 'Number of bins' set to 20. There are 'Apply' and 'Cancel' buttons at the bottom.

Objective plot	
Objectives for plot	[]
Plot scale	[1 1]
Sign of fitness	[1 1]

Pareto plot parameters	
Objectives for plot	[1 2]
Plot scale	[1 1]
Sign of fitness	[-1 -1]
x-axis label	Objective 1
y-axis label	Objective 2
z-axis label	Objective 3
Pareto plot title	Solution Space
Axis limits for Pareto plot	[-2 25 -2 25]

Distribution parameter	
Distribution plot type	plot histograms
Number of individuals to plot	100
Number of bins	20

- Define the output reporting level
 - For the text and graphical report, check the reporting option is set to **‘Text and plot’**



- Start GOSET



Tanaka Problem



Problem Statement

- Tanaka problem is a constrained optimization problem with two objectives to be minimized

$$\min f_1(x_1, x_2) = x_1$$

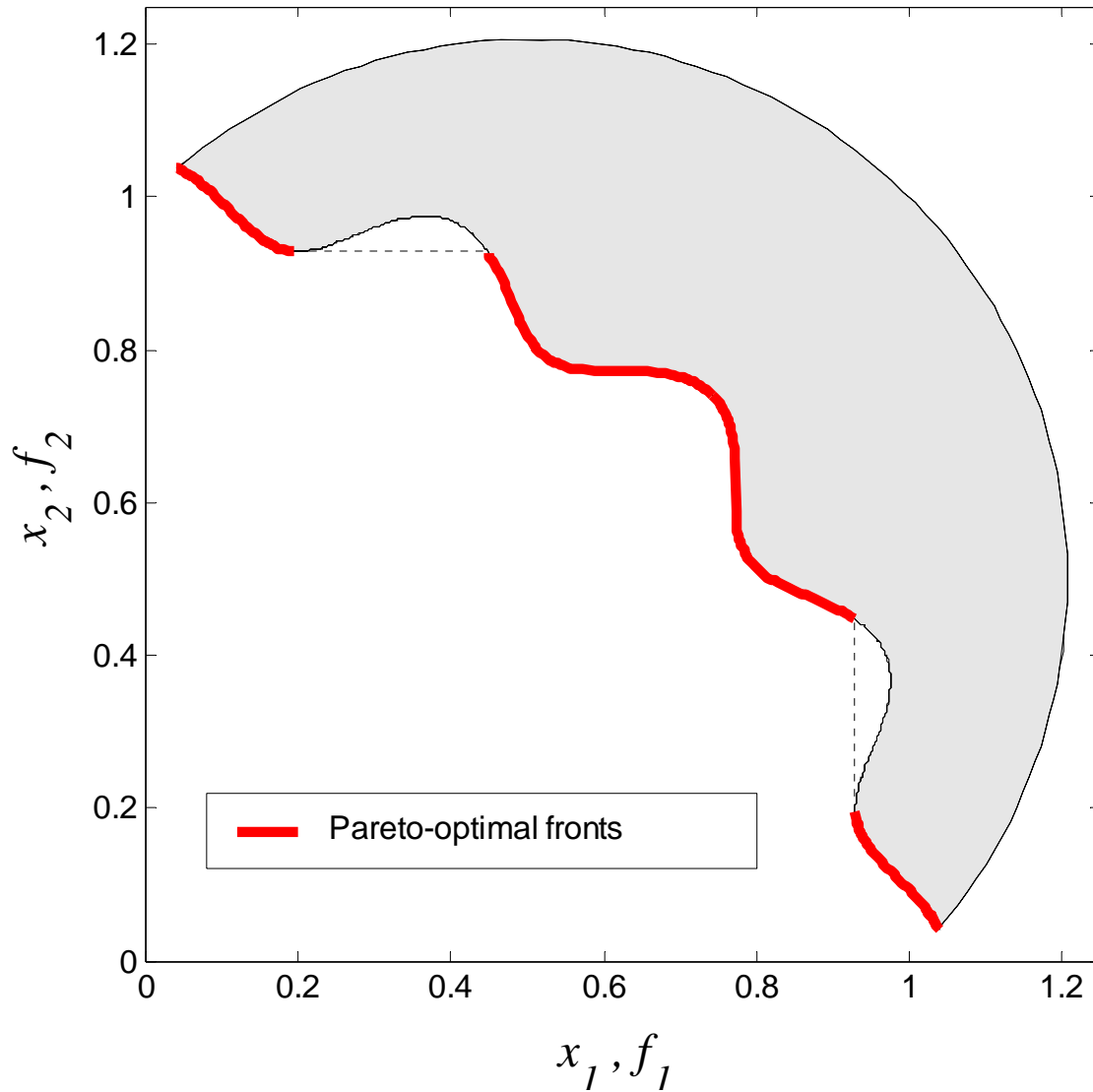
$$\min f_2(x_1, x_2) = x_2$$

$$\text{subject to } C_1(x_1, x_2) = x_1^2 + x_2^2 - 1 - 0.1 \cos\left(16 \arctan \frac{x_1}{x_2}\right) \geq 0, \quad 0 \leq x_1 \leq \pi,$$

$$C_2(x_1, x_2) = (x_1 - 0.5)^2 + (x_2 - 0.5)^2 \leq 0.5, \quad 0 \leq x_2 \leq \pi.$$

- Note the variable space is also the objective space

Tanaka Problem





Fitness Function

- Two objectives have positive values and they are to be minimized
- The fitness function values are defined to be the negative of the objective function values
- Infeasible solutions are assigned with -10 to reduce the chance of surviving



tanaka.m

```
% Tanaka problem (1995)
```

```
function [f] = tanaka(x)
```

```
C1 = x(1)^2+x(2)^2-1-0.1*cos(16*atan(x(1)/x(2))) >= 0;
```

```
C2 = (x(1)-0.5)^2+(x(2)-0.5)^2 <= 0.5;
```

```
if C1 & C2
```

```
    f(1,1) = -x(1);
```

```
    f(2,1) = -x(2);
```

```
else
```

```
    f(1,1) = -10;
```

```
    f(2,1) = -10;
```

```
end
```

tanaka_problem.m

```
% Initialize the parameters
GAP = gapdefault(2);

GAP.fp_ipop = 200;
GAP.fp_npop = 200;
GAP.fp_ngen = 200;
GAP.fp_obj = 0;
GAP.sc_alg = 6;           % sigma-truncation
GAP.op_list = [];       % list of objectives for objective plots
GAP.pp_list = [1,2];    % list parameters for Pareto plot
GAP.pp_sign = [-1,-1];  % sign of fitness for each objective
GAP.pp_axis=[0 1.25 0 1.25]; % axis limits for Pareto plot

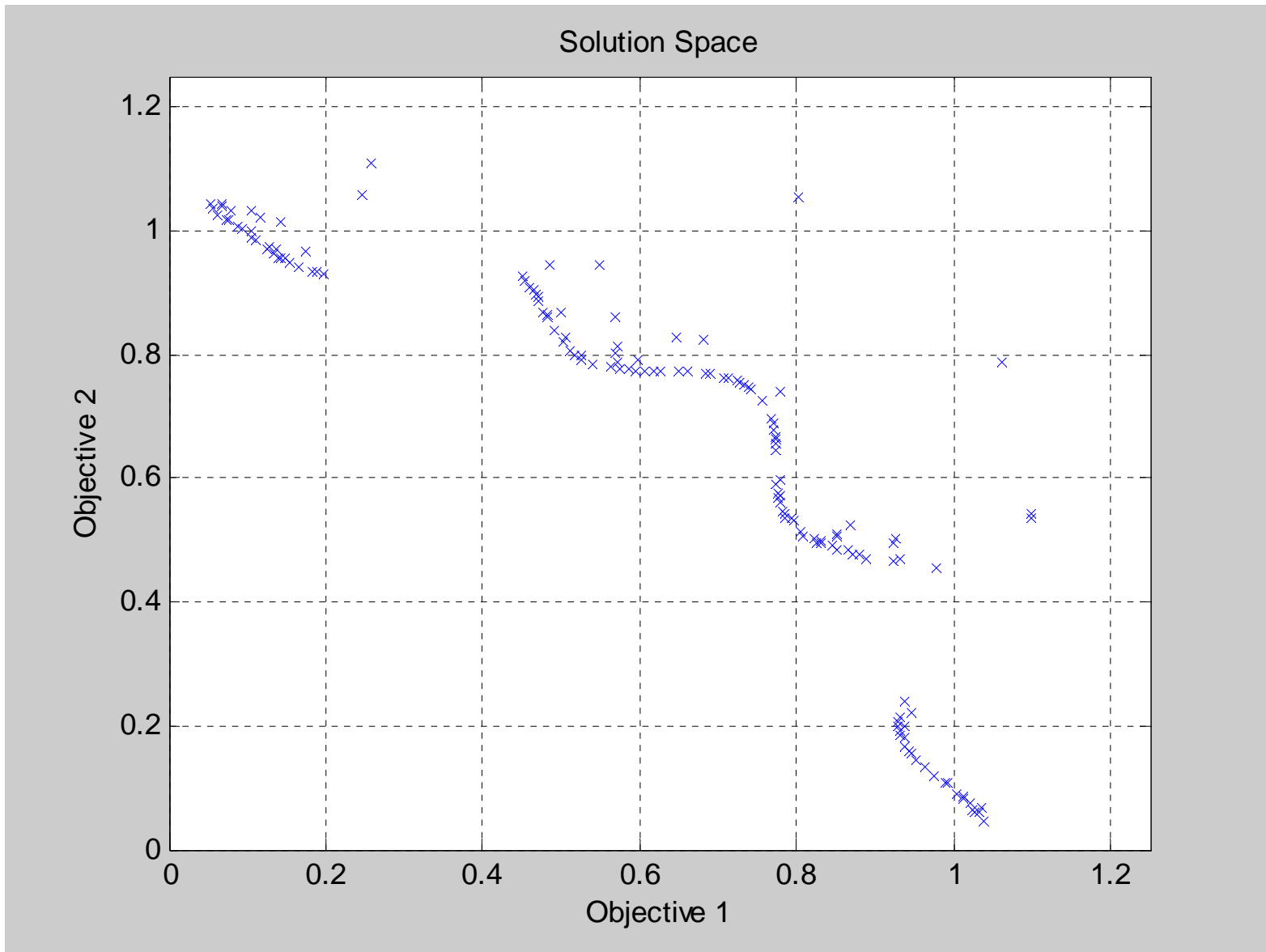
%           x1      x2
% gene      1      2
GAP.gd_min = [ 0      0  ];
GAP.gd_max = [ pi     pi  ];
GAP.gd_type = [ 2      2  ];
GAP.gd_cid  = [ 1      1  ];

% perform optimization
[P,GAS]=gaoptimize(@tanaka,GAP,[],[],[],[]);

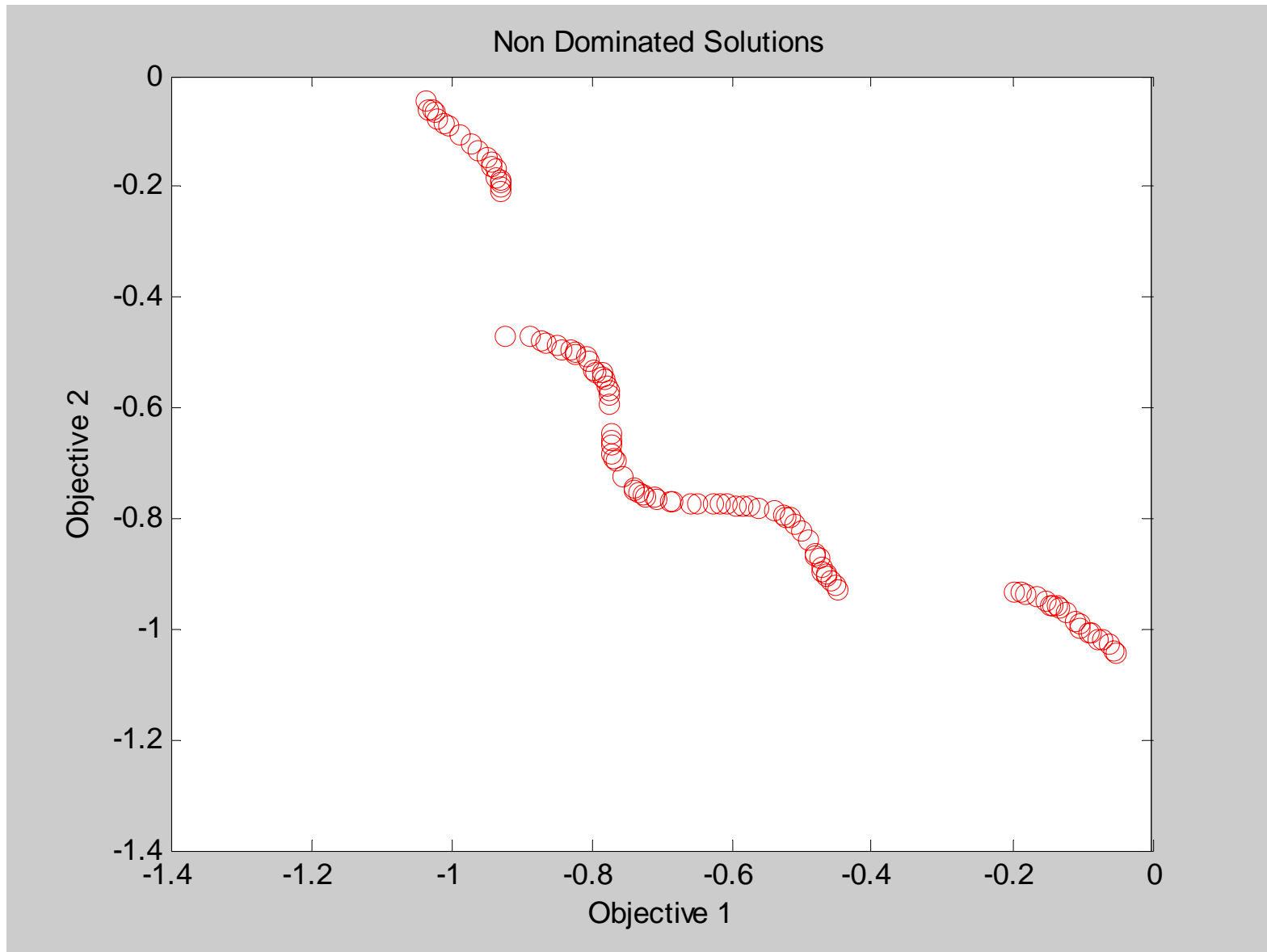
% find the nondominated set
nd=nondom(P.mfit,1);
ndi=find(nd==1);
figure(2);
plot(P.mfit(1,ndi),P.mfit(2,ndi),'ro');
title('Non Dominated Solutions');
xlabel('Objective 1');
ylabel('Objective 2');
```



Pareto Plot



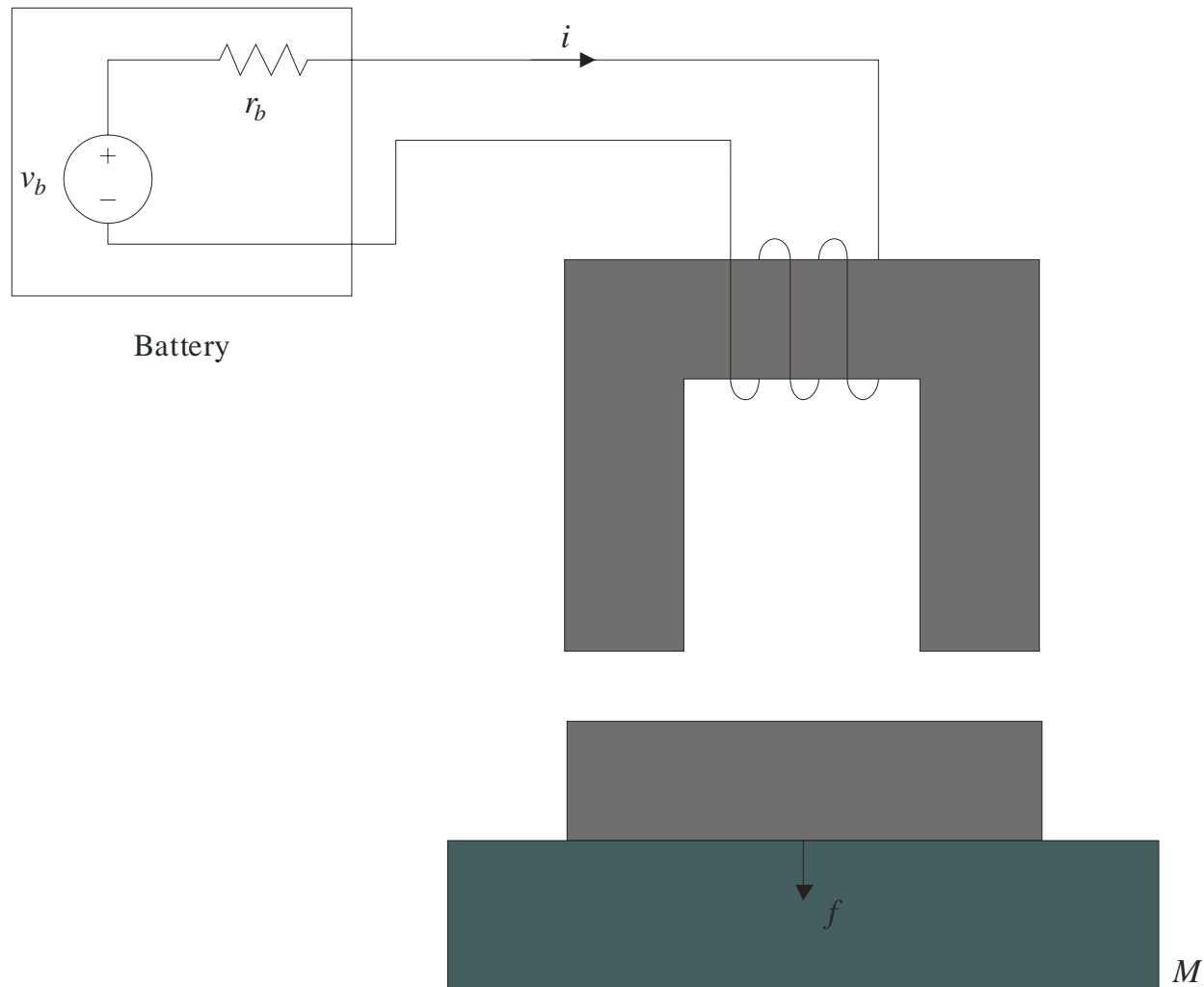
Non Dominated Solutions





Return of the ... Electromagnet

Recall Our Electromagnet Design



Multi Objective Fitness Function

$$\mathbf{f} = \begin{cases} \begin{bmatrix} \frac{1}{m_1 + \varepsilon} \\ \frac{1}{i + \varepsilon} \end{bmatrix} & c = 1 \\ \left(\sum_{c=1}^{10} c_i - 10 \right) \begin{bmatrix} 1 \\ 1 \end{bmatrix} & c < 1 \end{cases}$$



New Matlab Files

- `Electromagnet_MultiObjective_Fitness.m`
- `Electromagnet_MultiObjective_Design.m`
- `Electromagnet_MultiObjective_Design_Review.m`

Fitness Function Changes

■ Electromagnet_Fitness

```
% compute the fitness
if (cmin == 1)
    fitness = 1/(m1+1e-6);
else
    fitness = sum(c)-length(c);
end
```

■ Electromagnet_MultiObjective_Fitness

```
% compute the fitness
if (cmin == 1)
    fitness = [1/(m1+1e-6); 1/(i+1e-6)];
else
    fitness = (sum(c)-length(c))*[1; 1];
end
```



Electromagnet_MultiObjective_Design.m

```
% Electromagnet Design
%
% Written by:
% Ricky Chan for S.D. Sudhoff
% School of Electrical and Computer Engineering
% 1285 Electrical Engineering Building
% West Lafayette, IN 47907-1285
% E-mail: sudhoff@ecn.purdue.edu

% Set Up Population
GAP          = gapdefault(2);
GAP.fp_ngen = 2500;
GAP.fp_ipop = 500;
GAP.fp_npop = 500;
GAP.mc_alg  = 4.0;
GAP.dt_alg  = 3;

% Set Up Reporting
GAP.op_list=[];           % No objective plot
GAP.pp_list=[1,2];       % Pareto plot
GAP.pp_xl='Inverse Mass'; % Axis labels
GAP.pp_yl='Inverse Current';
GAP.pp_axis=[0.0 2 0.0 1]; % Axis limits

% Set Up Migrations
GAP.mg_nreg=5;           % number of regions
GAP.mg_tmig=100;        % mean time between migrations
GAP.mg_pmig=0.05;       % probability of a individual migrating

% Units
mm=1.0e-3;
cm=1.0e-2;
```

Electromagnet_MultiObjective_Design.m

```
% Problem Requirement Data
Psi.M      = 10; % Mass of object (Kg)
Psi.G      = 9.8; % Gravity (m/s2)
Psi.g      = 0.5*cm; % Gap Width (m)
Psi.rhow   = [ 8900; 2701]; % Wire - Kg/m3
Psi.sigmax = [ 58.0; 35.4]*1e6; % Wire - A/Ohm
Psi.jmaxw   = [ 7.62; 6.14]*1e6; % Wire - A/m2
Psi.descw   = ['Copper '; 'Aluminum']; % Wire - Description
Psi.bmaxm   = [ 1.4; 0.7; 1.2]; % Steel - Bsat, T
Psi.rhom    = [7064.1; 8069.4; 7892.1]; % Steel - Density Kg/m3
Psi.myum    = [ 15000; 6000; 3500]; % Steel - Perm., (relative)
Psi.descm   = ['Microsil '; ... % Steel - Description
              'Superperm 80'; ...
              'Superperm 49'];
Psi.imax    = 6; % Maximum current, A
Psi.lmax    = 10e-2; % Maximum length
Psi.pfmax   = 0.7; % Maximum packing factor
Psi.vb      = 12; % Battery Voltage
Psi.rb      = 0.5; % Battery Resistance

% Genetic Mapping
%
GAP.gd_min  = [ 1 1e-8 10 1 1*cm 0.1 1*mm 0.1 1*mm 0.2 0.2 1*mm];
GAP.gd_max  = [ 2 1e-4 1e3 3 10*cm 1.0 20*cm 1.0 5*cm 2.0 2.0 10*cm];
GAP.gd_type = [ 1 3 3 1 3 3 3 3 3 3 3 3 ];
GAP.gd_cid  = [ 1 1 1 1 1 1 1 1 1 1 1 1 ];

% Solve Problem
[fp,GAS]= gaoptimize(@Electromagnet_MultiObjective_Fitness,GAP,Psi,[],[],[]);

% Save run
s=input('Type 1 to Save Run ');
if (s==1)
    save sample_multi_objective_run
end
```



Electromagnet_MultiObjective_Design_Review.m

```
% Electromagnet Design Review
% Illustrates the evolution of the electromagnet design
%
% Written by:
% S.D. Sudhoff
% School of Electrical and Computer Engineering
% 1285 Electrical Engineering Building
% West Lafayette, IN 47907-1285
% E-mail: sudhoff@ecn.purdue.edu

clear all

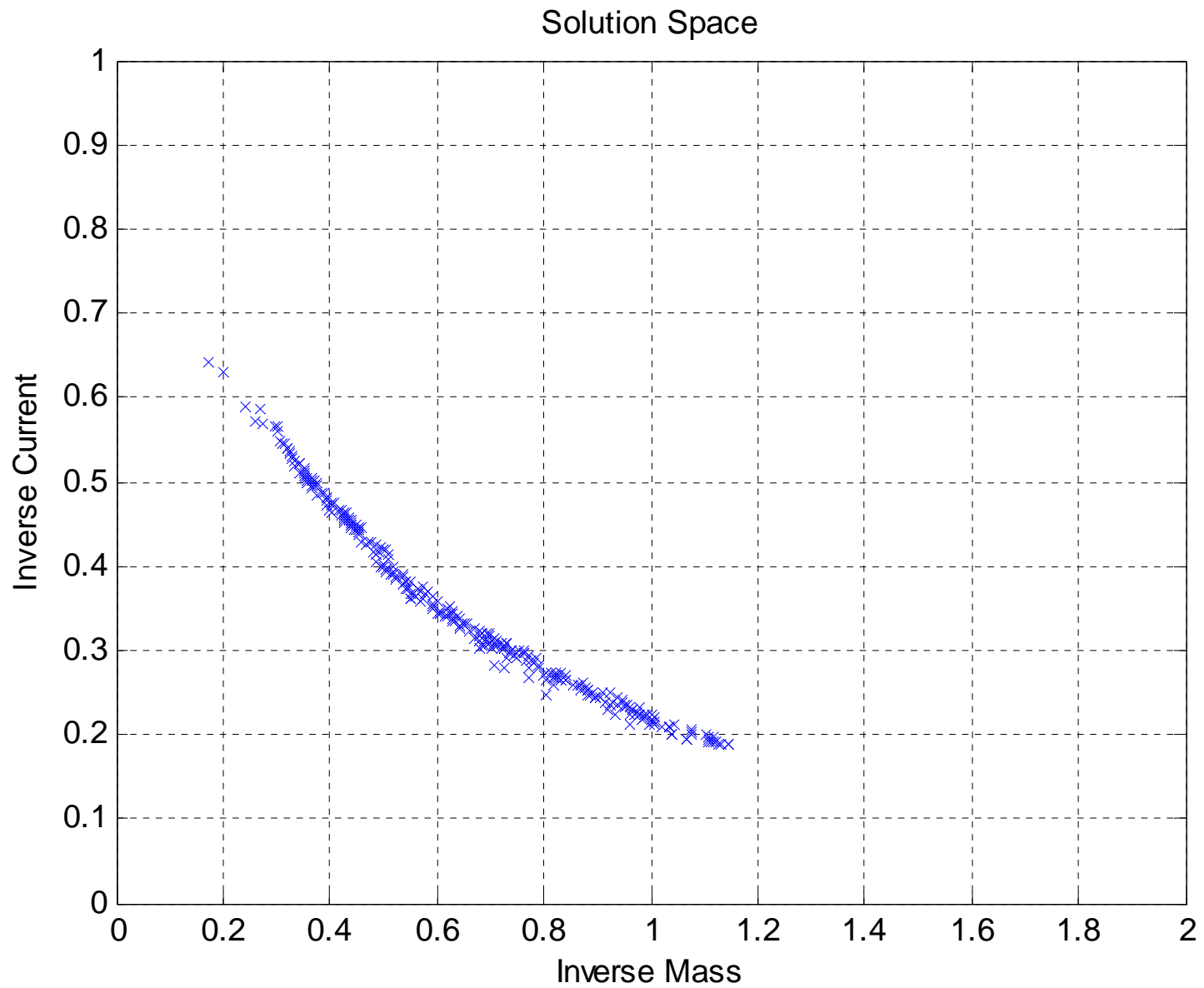
load sample_multi_objective_run

% Plot Pareto Optimal Front in Objective Space
non_dominated_status=nondom(fP.mfit,1);
non_dominated_indices=find(non_dominated_status==1);
Mass=1./fP.mfit(1,non_dominated_indices)-1e-6;
Current=1./fP.mfit(2,non_dominated_indices)-1e-6;
figure(2)
plot(Mass,Current,'ro');
xlabel('Mass, kg');
ylabel('Current, A');
title('Trade Off Between Mass and Current Draw');

% Minimum Mass Design
mmd=GAS.bestgenes(:,GAS.cg,1);
Electromagnet_MultiObjective_Fitness(mmd,Psi,3);

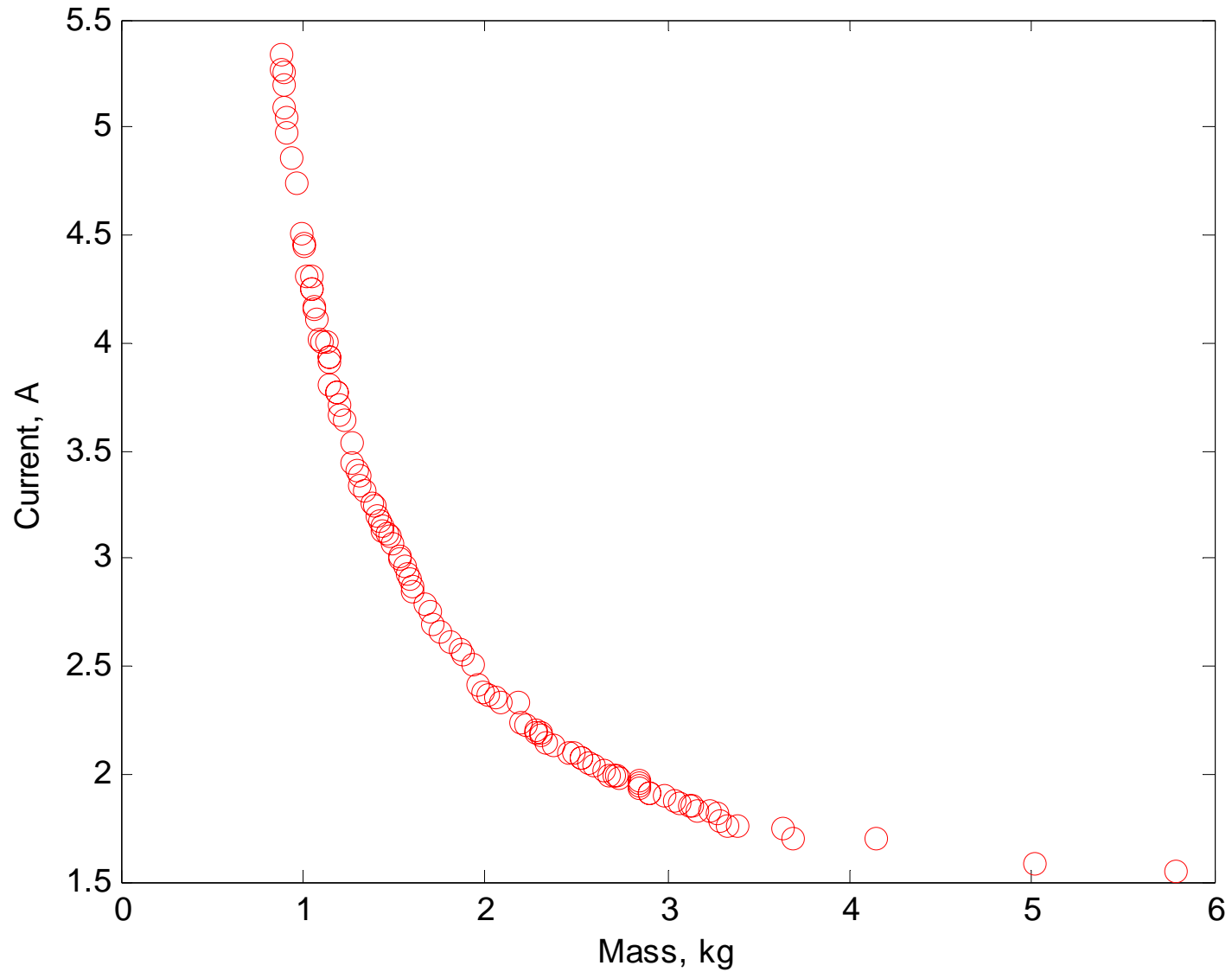
% Minimum Current Design
mcd=GAS.bestgenes(:,GAS.cg,2);
Electromagnet_MultiObjective_Fitness(mcd,Psi,5);
```

Pareto Plot



Design Space

Trade Off Between Mass and Current Draw





Design Extremes

- Design 1

- Mass: 0.874 kg

- Current: 5.33 A

- Design 2

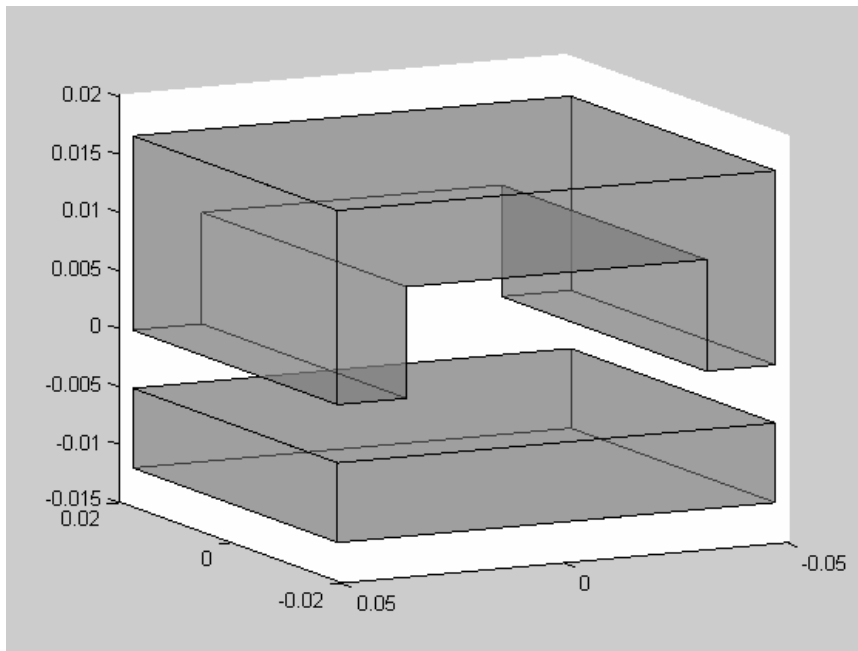
- Mass: 5.79 kg

- Current: 1.56 A

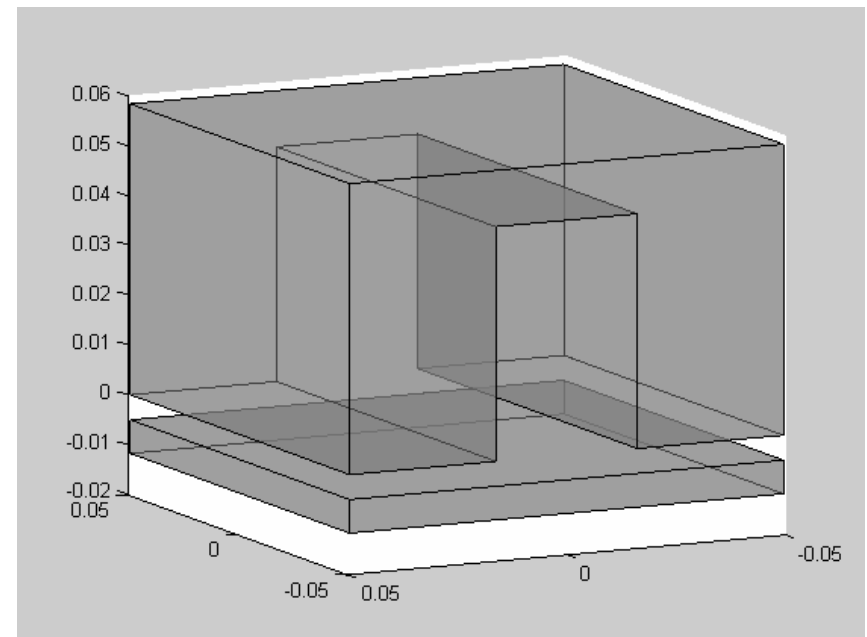


Design Extremes

Design 1

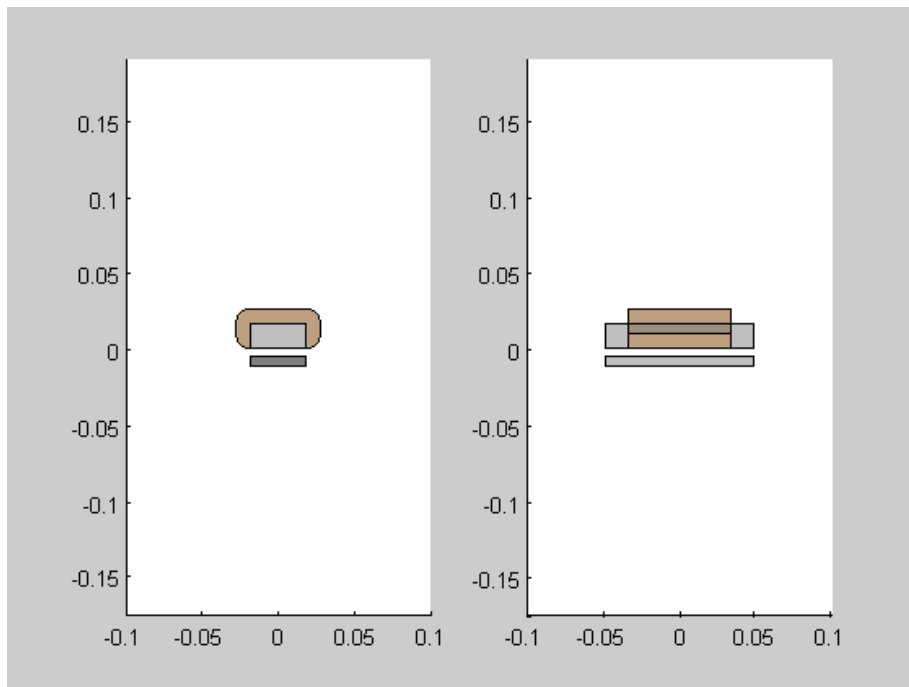


Design 2



Design Extremes

Design 1



Design 2

