

# SPACE MAPPING OPTIMIZATION FOR ENGINEERING DESIGN: A TUTORIAL PRESENTATION

J.E. Rayas-Sánchez

Simulation Optimization Systems Research Laboratory and  
Department of Electrical and Computer Engineering  
McMaster University, Hamilton, Canada L8S 4K1

[bandler@mcmaster.ca](mailto:bandler@mcmaster.ca)  
[www.sos.mcmaster.ca](http://www.sos.mcmaster.ca)



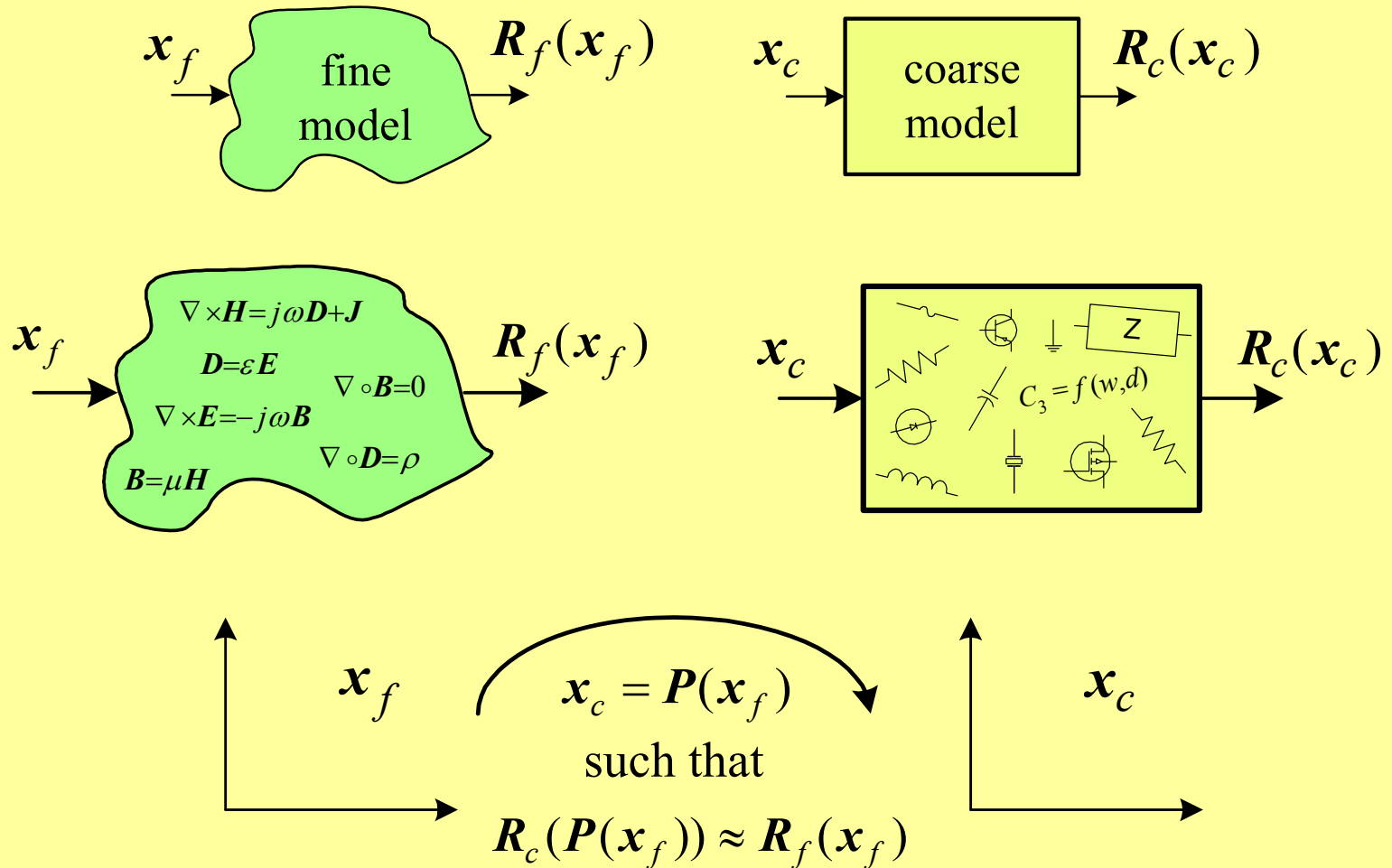
presented at

Workshop on Next Generation Optimization Methodologies for Wireless and Microwave Circuit Design, Hamilton, ON, June 28, 1999



# The Aim of Space Mapping

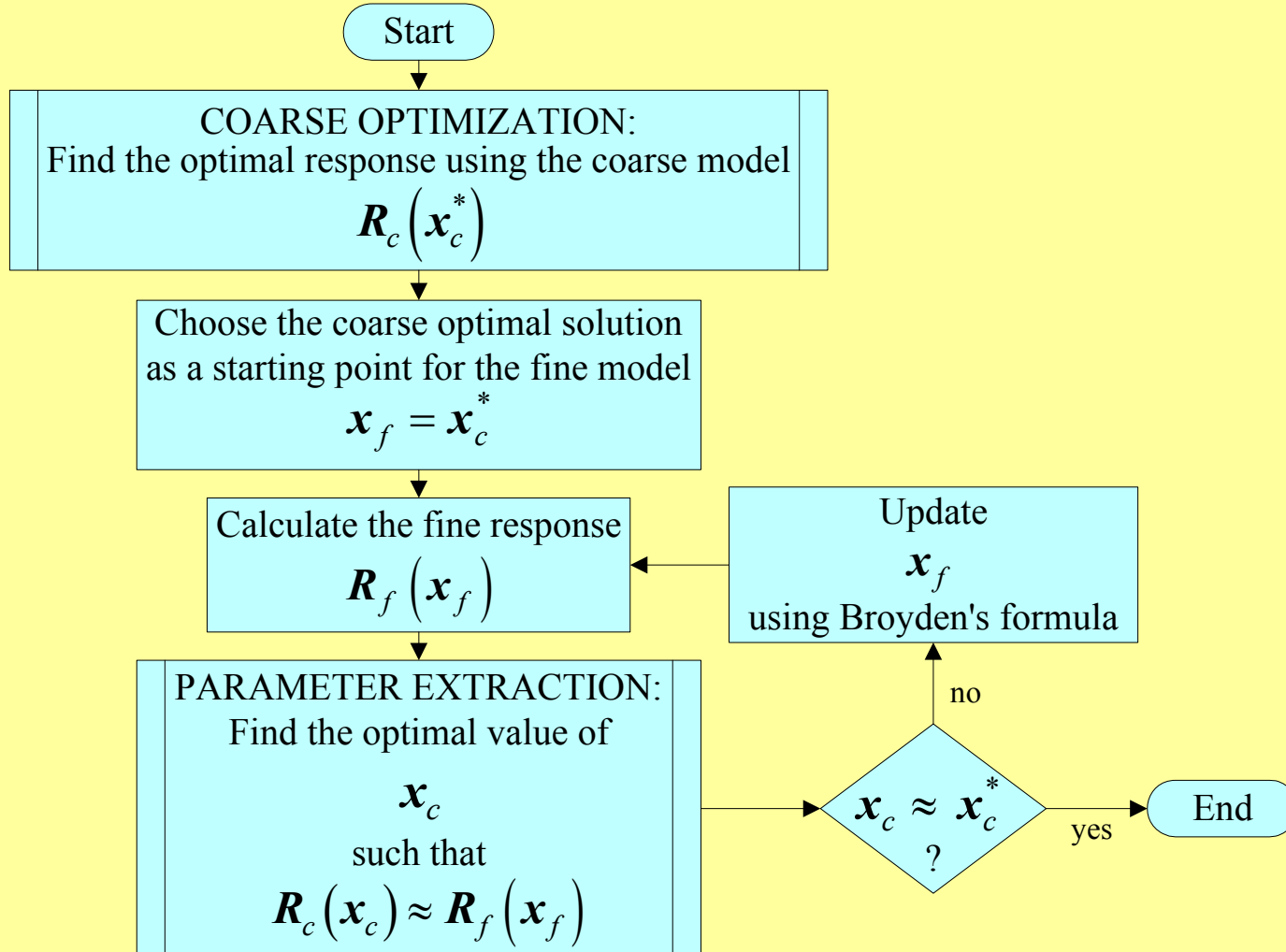
(Bandler et al., 1994-)





## Aggressive Space Mapping (ASM) Algorithm

(Bandler et al., 1995)





## **ASM Algorithm**

0. Initialize  $\mathbf{x}_f^{(1)} = \mathbf{x}_c^*$ ,  $\mathbf{B}^{(1)} = \mathbf{I}$ ,  $j = 1$

1. Evaluate  $\mathbf{R}_f(\mathbf{x}_f^{(1)})$



## ASM Algorithm

0. Initialize  $\mathbf{x}_f^{(1)} = \mathbf{x}_c^*$ ,  $\mathbf{B}^{(1)} = \mathbf{I}$ ,  $j = 1$

1. Evaluate  $\mathbf{R}_f(\mathbf{x}_f^{(1)})$

2. Extract  $\mathbf{x}_c^{(1)}$  such that

$$\mathbf{R}_c(\mathbf{x}_c^{(1)}) \approx \mathbf{R}_f(\mathbf{x}_f^{(1)})$$



## ASM Algorithm

0. Initialize  $\mathbf{x}_f^{(1)} = \mathbf{x}_c^*$ ,  $\mathbf{B}^{(1)} = \mathbf{I}$ ,  $j = 1$

1. Evaluate  $\mathbf{R}_f(\mathbf{x}_f^{(1)})$

2. Extract  $\mathbf{x}_c^{(1)}$  such that

$$\mathbf{R}_c(\mathbf{x}_c^{(1)}) \approx \mathbf{R}_f(\mathbf{x}_f^{(1)})$$

3. Evaluate  $\mathbf{f}^{(1)} = \mathbf{x}_c^{(1)} - \mathbf{x}_c^*$

Stop if  $\|\mathbf{f}^{(1)}\| \leq \eta$



## ASM Algorithm

0. Initialize  $\mathbf{x}_f^{(1)} = \mathbf{x}_c^*$ ,  $\mathbf{B}^{(1)} = \mathbf{I}$ ,  $j = 1$

1. Evaluate  $\mathbf{R}_f(\mathbf{x}_f^{(1)})$

2. Extract  $\mathbf{x}_c^{(1)}$  such that

$$\mathbf{R}_c(\mathbf{x}_c^{(1)}) \approx \mathbf{R}_f(\mathbf{x}_f^{(1)})$$

3. Evaluate  $\mathbf{f}^{(1)} = \mathbf{x}_c^{(1)} - \mathbf{x}_c^*$

Stop if  $\|\mathbf{f}^{(1)}\| \leq \eta$

4. Solve  $\mathbf{B}^{(j)} \mathbf{h}^{(j)} = -\mathbf{f}^{(j)}$  for  $\mathbf{h}^{(j)}$



## ASM Algorithm

0. Initialize  $\mathbf{x}_f^{(1)} = \mathbf{x}_c^*$ ,  $\mathbf{B}^{(1)} = \mathbf{I}$ ,  $j = 1$

1. Evaluate  $\mathbf{R}_f(\mathbf{x}_f^{(1)})$

2. Extract  $\mathbf{x}_c^{(1)}$  such that

$$\mathbf{R}_c(\mathbf{x}_c^{(1)}) \approx \mathbf{R}_f(\mathbf{x}_f^{(1)})$$

3. Evaluate  $\mathbf{f}^{(1)} = \mathbf{x}_c^{(1)} - \mathbf{x}_c^*$

Stop if  $\|\mathbf{f}^{(1)}\| \leq \eta$

4. Solve  $\mathbf{B}^{(j)} \mathbf{h}^{(j)} = -\mathbf{f}^{(j)}$  for  $\mathbf{h}^{(j)}$

5. Set  $\mathbf{x}_f^{(j+1)} = \mathbf{x}_f^{(j)} + \mathbf{h}^{(j)}$





## ASM Algorithm

0. Initialize  $\mathbf{x}_f^{(1)} = \mathbf{x}_c^*$ ,  $\mathbf{B}^{(1)} = \mathbf{I}$ ,  $j = 1$
1. Evaluate  $\mathbf{R}_f(\mathbf{x}_f^{(1)})$
2. Extract  $\mathbf{x}_c^{(1)}$  such that
$$\mathbf{R}_c(\mathbf{x}_c^{(1)}) \approx \mathbf{R}_f(\mathbf{x}_f^{(1)})$$
3. Evaluate  $\mathbf{f}^{(1)} = \mathbf{x}_c^{(1)} - \mathbf{x}_c^*$ 

Stop if  $\|\mathbf{f}^{(1)}\| \leq \eta$
4. Solve  $\mathbf{B}^{(j)} \mathbf{h}^{(j)} = -\mathbf{f}^{(j)}$  for  $\mathbf{h}^{(j)}$
5. Set  $\mathbf{x}_f^{(j+1)} = \mathbf{x}_f^{(j)} + \mathbf{h}^{(j)}$
6. Evaluate  $\mathbf{R}_f(\mathbf{x}_f^{(j+1)})$



## ASM Algorithm

0. Initialize  $\mathbf{x}_f^{(1)} = \mathbf{x}_c^*$ ,  $\mathbf{B}^{(1)} = \mathbf{I}$ ,  $j = 1$

1. Evaluate  $\mathbf{R}_f(\mathbf{x}_f^{(1)})$

2. Extract  $\mathbf{x}_c^{(1)}$  such that

$$\mathbf{R}_c(\mathbf{x}_c^{(1)}) \approx \mathbf{R}_f(\mathbf{x}_f^{(1)})$$

3. Evaluate  $\mathbf{f}^{(1)} = \mathbf{x}_c^{(1)} - \mathbf{x}_c^*$

Stop if  $\|\mathbf{f}^{(1)}\| \leq \eta$

4. Solve  $\mathbf{B}^{(j)} \mathbf{h}^{(j)} = -\mathbf{f}^{(j)}$  for  $\mathbf{h}^{(j)}$

5. Set  $\mathbf{x}_f^{(j+1)} = \mathbf{x}_f^{(j)} + \mathbf{h}^{(j)}$

6. Evaluate  $\mathbf{R}_f(\mathbf{x}_f^{(j+1)})$

7. Extract  $\mathbf{x}_c^{(j+1)}$  such that

$$\mathbf{R}_c(\mathbf{x}_c^{(j+1)}) \approx \mathbf{R}_f(\mathbf{x}_f^{(j+1)})$$



## ASM Algorithm

0. Initialize  $\mathbf{x}_f^{(1)} = \mathbf{x}_c^*$ ,  $\mathbf{B}^{(1)} = \mathbf{I}$ ,  $j = 1$

1. Evaluate  $\mathbf{R}_f(\mathbf{x}_f^{(1)})$

2. Extract  $\mathbf{x}_c^{(1)}$  such that

$$\mathbf{R}_c(\mathbf{x}_c^{(1)}) \approx \mathbf{R}_f(\mathbf{x}_f^{(1)})$$

3. Evaluate  $\mathbf{f}^{(1)} = \mathbf{x}_c^{(1)} - \mathbf{x}_c^*$

Stop if  $\|\mathbf{f}^{(1)}\| \leq \eta$

4. Solve  $\mathbf{B}^{(j)} \mathbf{h}^{(j)} = -\mathbf{f}^{(j)}$  for  $\mathbf{h}^{(j)}$

5. Set  $\mathbf{x}_f^{(j+1)} = \mathbf{x}_f^{(j)} + \mathbf{h}^{(j)}$

6. Evaluate  $\mathbf{R}_f(\mathbf{x}_f^{(j+1)})$

7. Extract  $\mathbf{x}_c^{(j+1)}$  such that

$$\mathbf{R}_c(\mathbf{x}_c^{(j+1)}) \approx \mathbf{R}_f(\mathbf{x}_f^{(j+1)})$$

8. Evaluate  $\mathbf{f}^{(j+1)} = \mathbf{x}_c^{(j+1)} - \mathbf{x}_c^*$

Stop if  $\|\mathbf{f}^{(j+1)}\| \leq \eta$



## ASM Algorithm

0. Initialize  $\mathbf{x}_f^{(1)} = \mathbf{x}_c^*$ ,  $\mathbf{B}^{(1)} = \mathbf{I}$ ,  $j = 1$

1. Evaluate  $\mathbf{R}_f(\mathbf{x}_f^{(1)})$

2. Extract  $\mathbf{x}_c^{(1)}$  such that

$$\mathbf{R}_c(\mathbf{x}_c^{(1)}) \approx \mathbf{R}_f(\mathbf{x}_f^{(1)})$$

3. Evaluate  $\mathbf{f}^{(1)} = \mathbf{x}_c^{(1)} - \mathbf{x}_c^*$

Stop if  $\|\mathbf{f}^{(1)}\| \leq \eta$

4. Solve  $\mathbf{B}^{(j)} \mathbf{h}^{(j)} = -\mathbf{f}^{(j)}$  for  $\mathbf{h}^{(j)}$

5. Set  $\mathbf{x}_f^{(j+1)} = \mathbf{x}_f^{(j)} + \mathbf{h}^{(j)}$

6. Evaluate  $\mathbf{R}_f(\mathbf{x}_f^{(j+1)})$

7. Extract  $\mathbf{x}_c^{(j+1)}$  such that

$$\mathbf{R}_c(\mathbf{x}_c^{(j+1)}) \approx \mathbf{R}_f(\mathbf{x}_f^{(j+1)})$$

8. Evaluate  $\mathbf{f}^{(j+1)} = \mathbf{x}_c^{(j+1)} - \mathbf{x}_c^*$

Stop if  $\|\mathbf{f}^{(j+1)}\| \leq \eta$

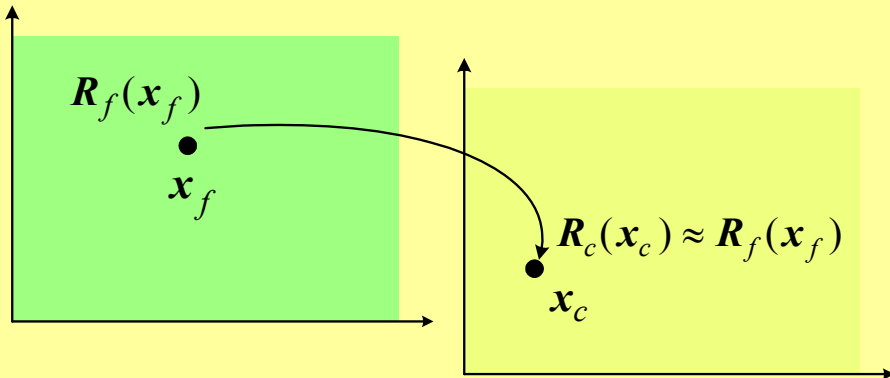
9. Update  $\mathbf{B}^{(j+1)} = \mathbf{B}^{(j)} + \frac{\mathbf{f}^{(j+1)} \mathbf{h}^{(j)T}}{\mathbf{h}^{(j)T} \mathbf{h}^{(j)}}$

10. Set  $j = j + 1$  ; go to Step 4



## Parameter Extraction

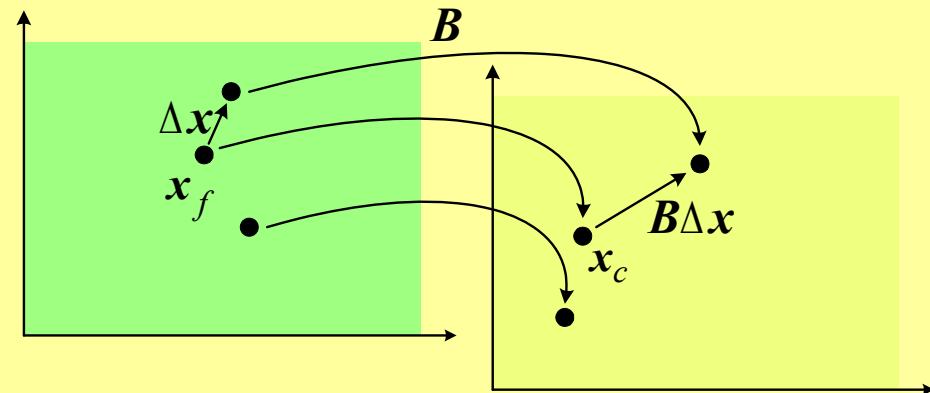
single point parameter extraction matches the responses of both models at a single point



it is formulated as

$$\min_{\mathbf{x}_c} \left\| \mathbf{R}_f(\mathbf{x}_f) - \mathbf{R}_c(\mathbf{x}_c) \right\|$$

multi-point parameter extraction simultaneously matches the responses at a number of corresponding points



it is formulated as

$$\min_{\mathbf{x}_c} \left\| [\mathbf{e}_1^T \quad \cdots \quad \mathbf{e}_i^T \quad \cdots]^T \right\|$$

where

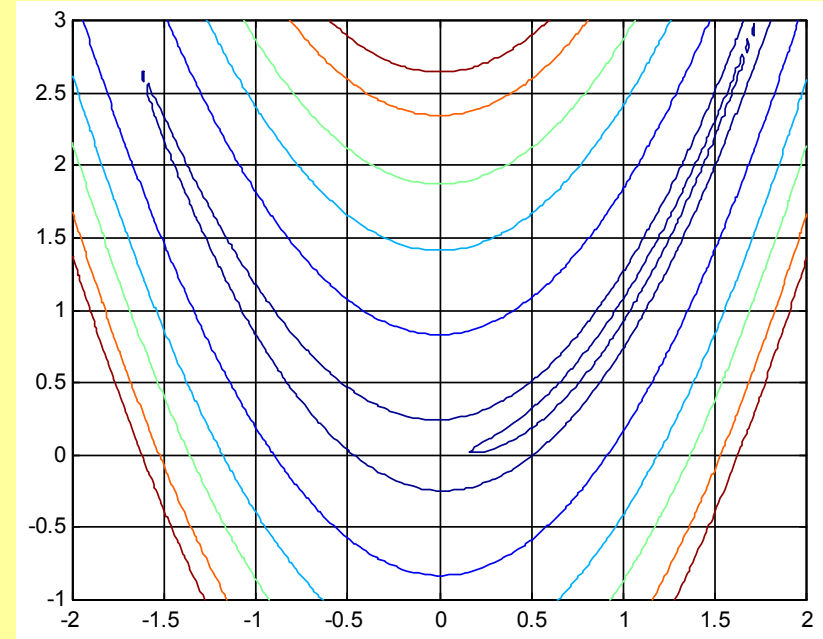
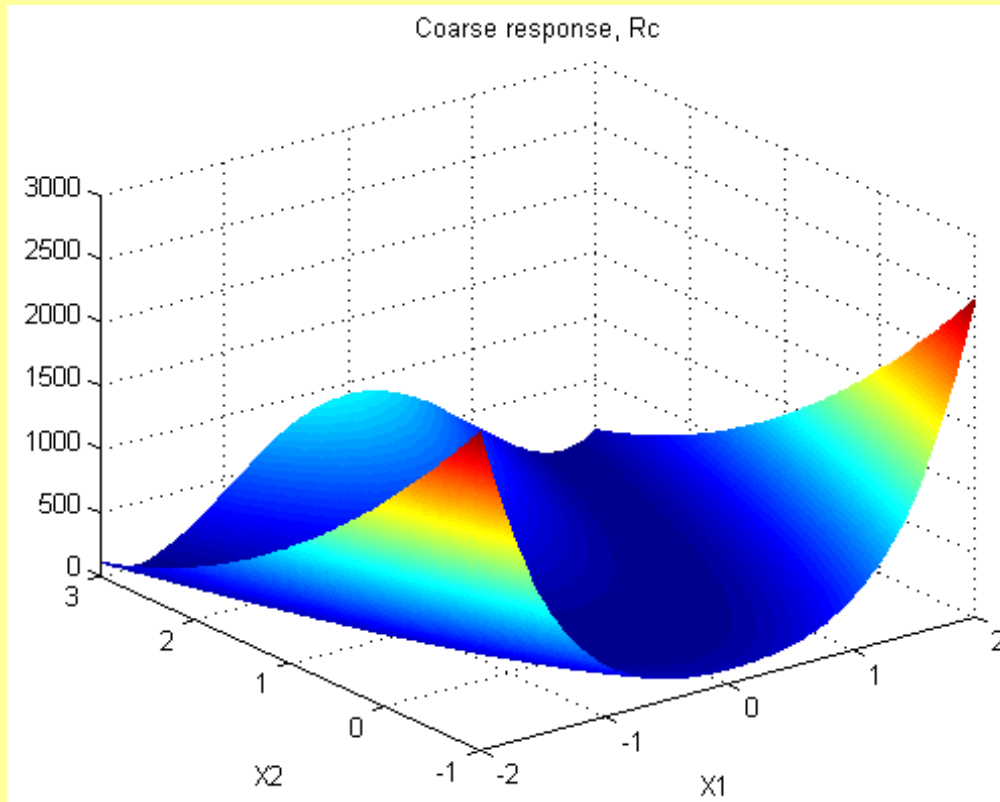
$$\mathbf{e}_i = \mathbf{R}_f(\mathbf{x}_f + \Delta \mathbf{x}_i) - \mathbf{R}_c(\mathbf{x}_c + \mathbf{B}\Delta \mathbf{x}_i)$$



## Illustration With Two Rosenbrock Functions

Coarse Model: Rosenbrock Function

$$R_c(\mathbf{x}) = 100(x_2 - x_1^2)^2 + (1 - x_1)^2, \text{ where } \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$



$$\mathbf{x}_c^* = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

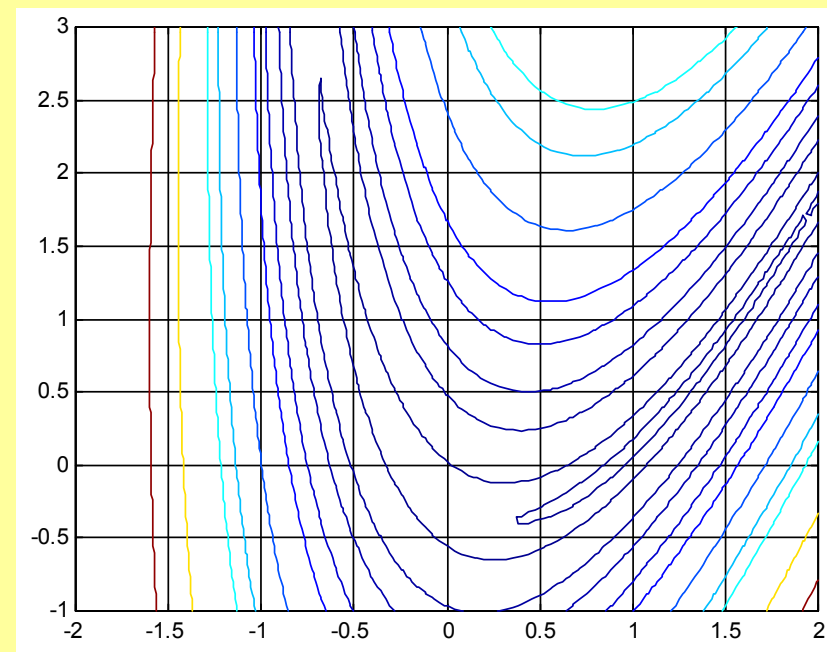
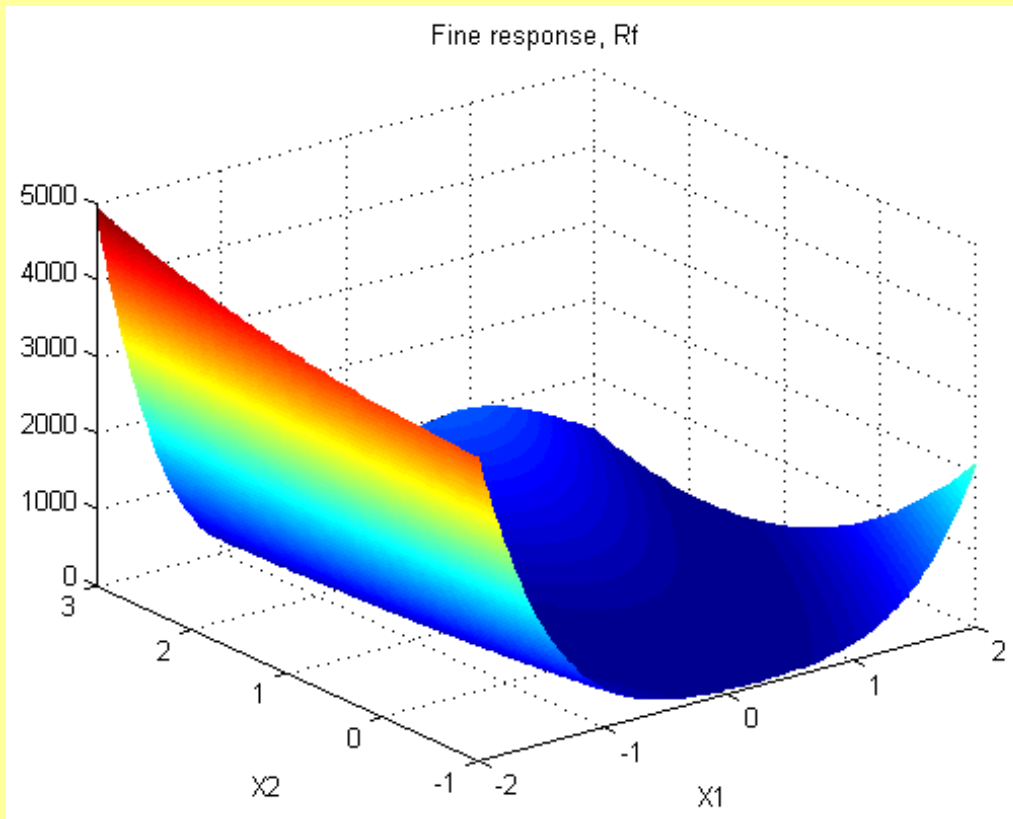
$$R_c^* = R_c(\mathbf{x}_c^*) = 0$$



## Illustration With Two Rosenbrock Functions (continued)

Fine Model Example: Transformed Rosenbrock Function

$$R_f(\mathbf{x}) = 100(u_2 - u_1^2)^2 + (1 - u_1)^2, \text{ where } \mathbf{u} = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 1.1 & -0.2 \\ 0.2 & 0.9 \end{bmatrix} \mathbf{x} + \begin{bmatrix} -0.3 \\ 0.3 \end{bmatrix}$$

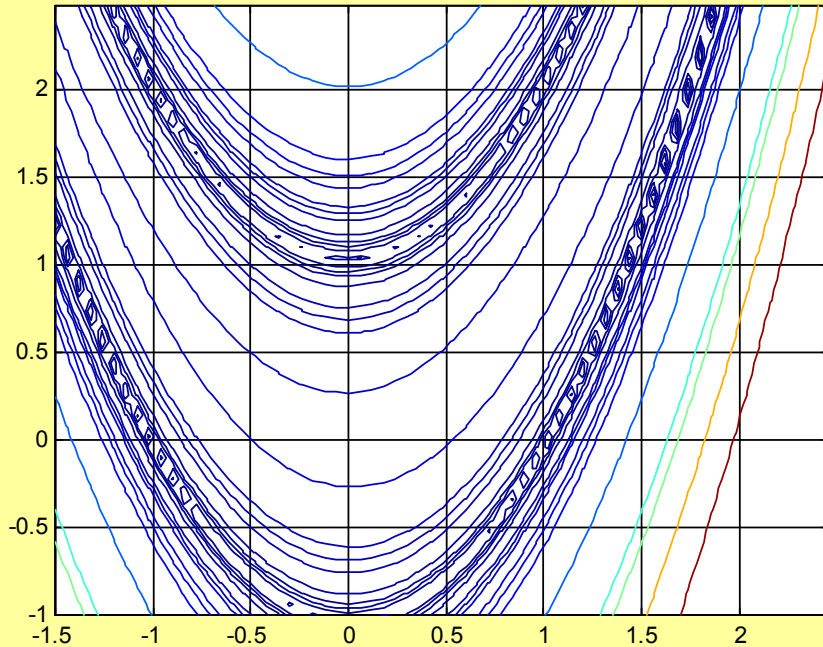


$$R_f(\mathbf{x}_c^*) = 108.32$$

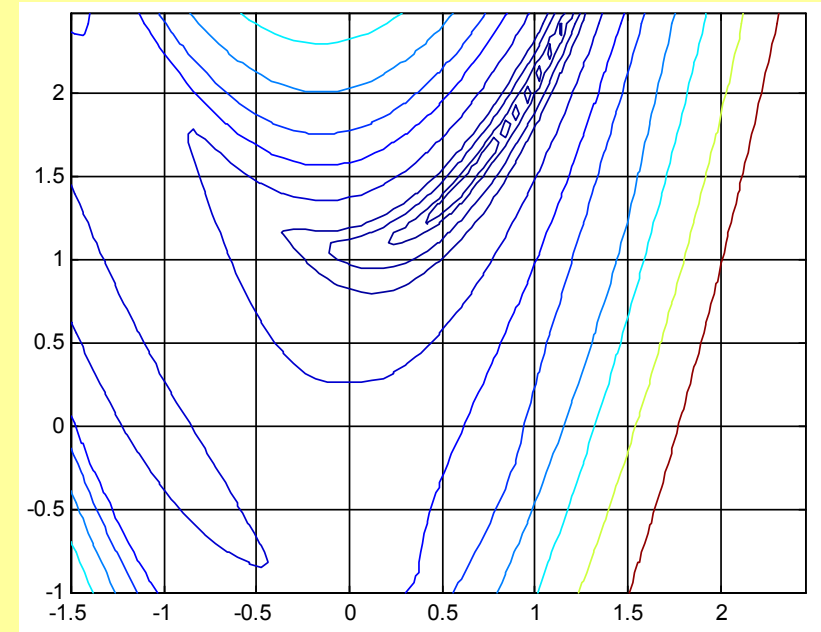


## First $l_1$ Parameter Extraction

single point parameter extraction



multi-point parameter extraction (with 2 additional points)



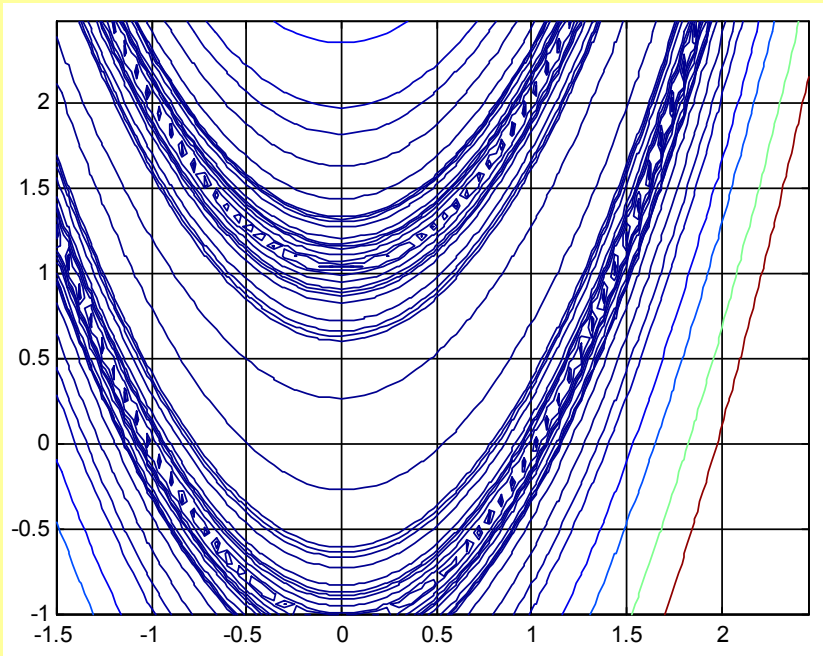
$$\mathbf{x}_c^{(1)} = \begin{bmatrix} 0.6 \\ 1.4 \end{bmatrix}$$



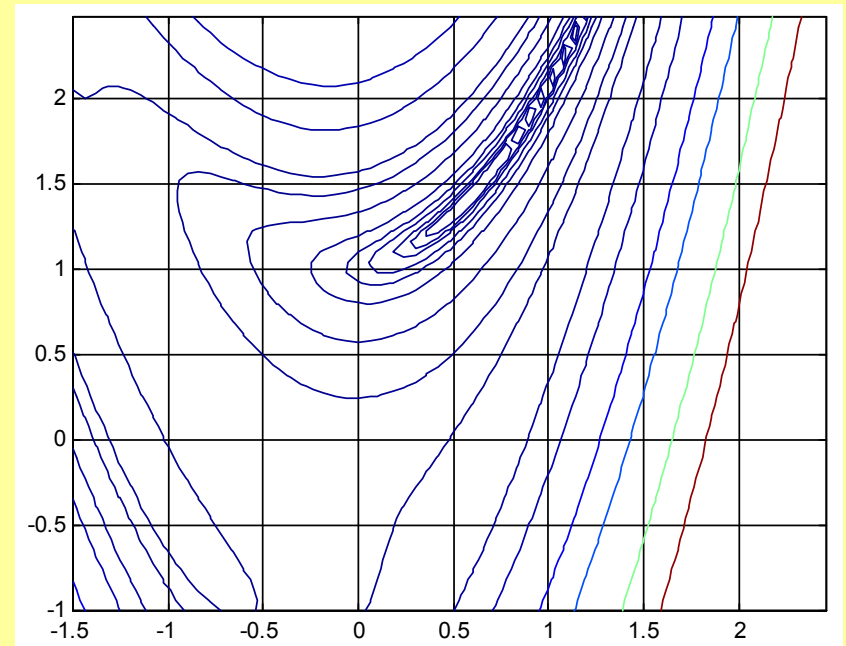


## First $l_2$ Parameter Extraction

single point parameter extraction



multi-point parameter extraction (with 2 additional points)



$$\mathbf{x}_c^{(1)} = \begin{bmatrix} 0.6 \\ 1.4 \end{bmatrix}$$



## Space Mapping Solution Process

$$0. \mathbf{x}_f^{(1)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \mathbf{B}^{(1)} = \mathbf{I}, j=1$$

$$1. R_f(\mathbf{x}_f^{(1)}) = 108.32$$

$$2. \text{When } \mathbf{x}_c^{(1)} = \begin{bmatrix} 0.6 \\ 1.4 \end{bmatrix}, R_c(\mathbf{x}_c^{(1)}) = R_f(\mathbf{x}_f^{(1)})$$

$$3. \mathbf{f}^{(1)} = \begin{bmatrix} 0.6 \\ 1.4 \end{bmatrix} - \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} -0.4 \\ 0.4 \end{bmatrix}$$

$$4. \text{Since } \mathbf{B}^{(1)} = \mathbf{I}, \mathbf{h}^{(1)} = \begin{bmatrix} 0.4 \\ -0.4 \end{bmatrix}$$

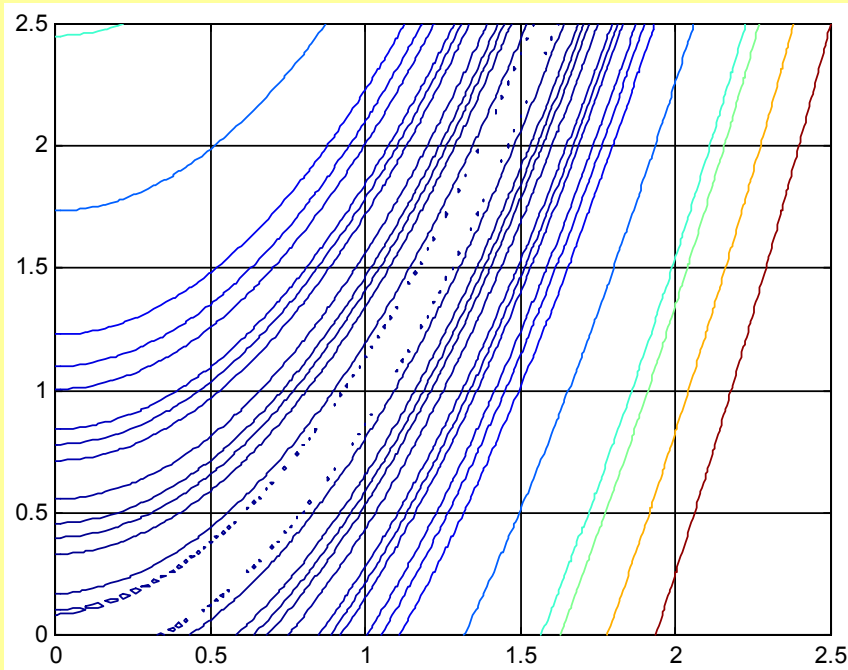
$$5. \text{Set } \mathbf{x}_f^{(2)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \begin{bmatrix} 0.4 \\ -0.4 \end{bmatrix} = \begin{bmatrix} 1.4 \\ 0.6 \end{bmatrix}$$

$$6. R_f(\mathbf{x}_f^{(2)}) = 1.8207$$

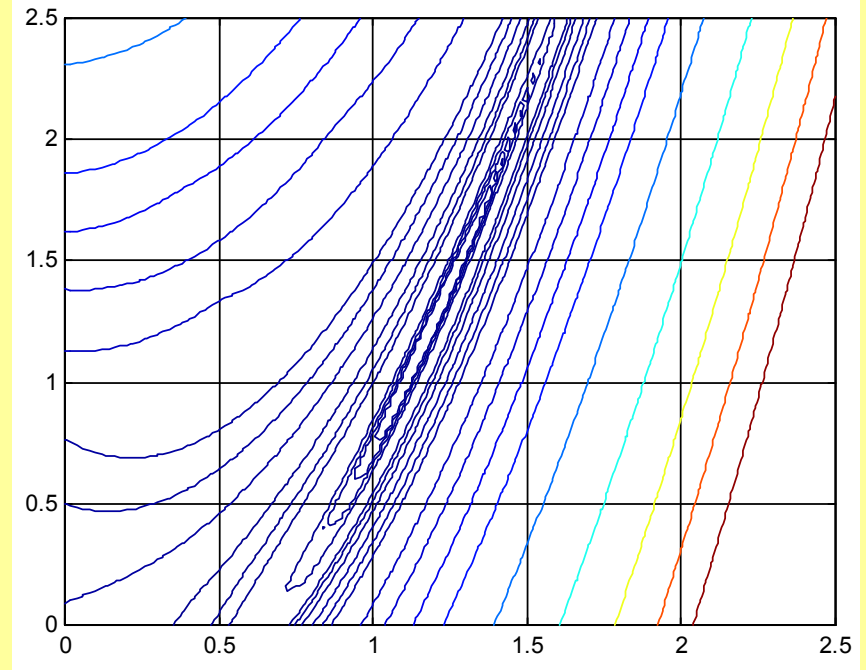


## Second $l_1$ Parameter Extraction

single point parameter extraction



multi-point parameter extraction (with 4 additional points)

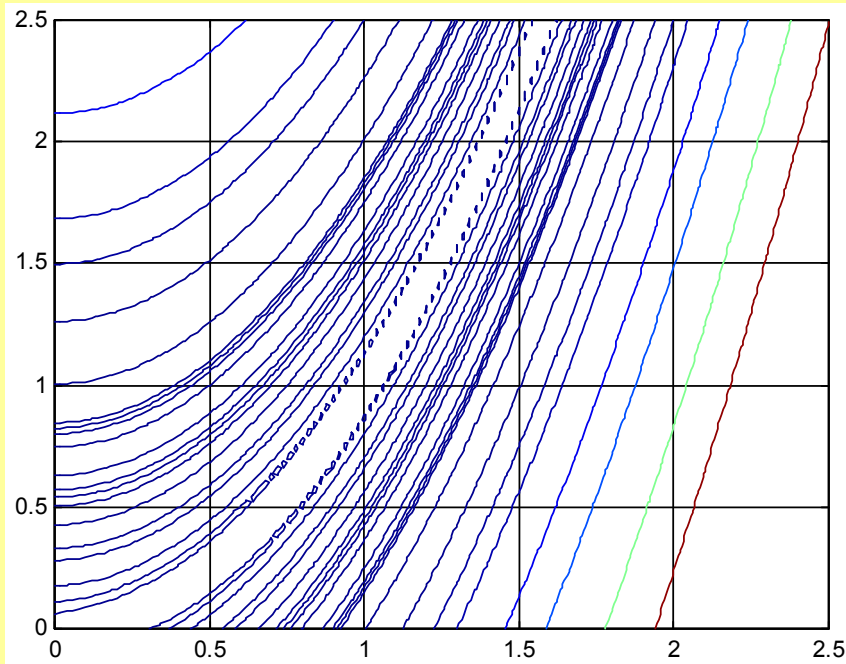


$$\mathbf{x}_c^{(2)} = \begin{bmatrix} 1.12 \\ 1.12 \end{bmatrix}$$

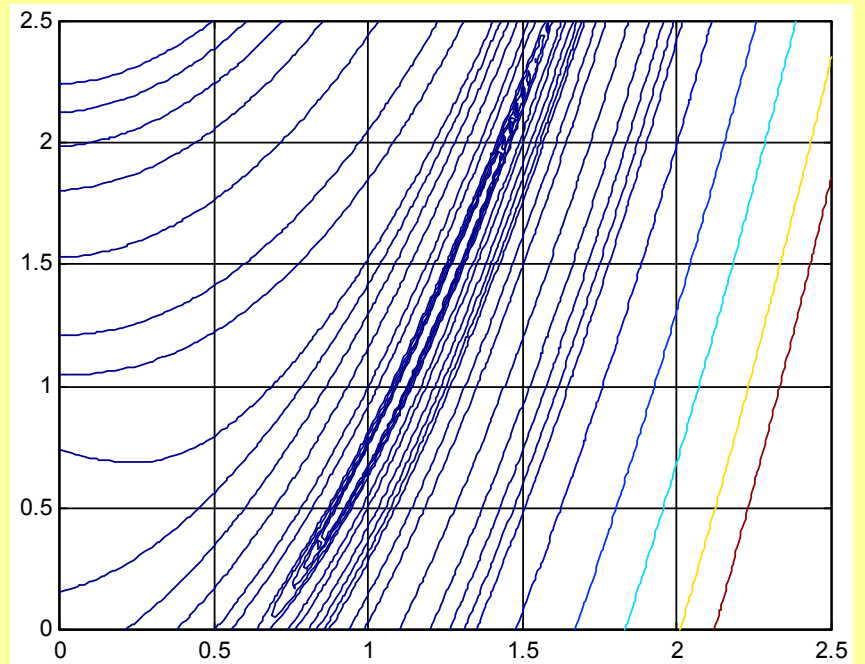


## Second $l_2$ Parameter Extraction

single point parameter extraction



multi-point parameter extraction (with 4 additional points)



$$\mathbf{x}_c^{(2)} = \begin{bmatrix} 1.12 \\ 1.12 \end{bmatrix}$$



## Space Mapping Solution Process (continued)

7. When  $\mathbf{x}_c^{(2)} = \begin{bmatrix} 1.12 \\ 1.12 \end{bmatrix}$ ,  $R_c(\mathbf{x}_c^{(2)}) = R_f(\mathbf{x}_f^{(2)})$

4b.  $\mathbf{h}^{(2)} = -\mathbf{B}^{(2)-1} \mathbf{f}^{(2)} = \begin{bmatrix} -0.12 \\ -0.12 \end{bmatrix}$

8.  $\mathbf{f}^{(2)} = \begin{bmatrix} 1.12 \\ 1.12 \end{bmatrix} - \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0.12 \\ 0.12 \end{bmatrix}$

5b. Set  $\mathbf{x}_f^{(3)} = \begin{bmatrix} 1.4 \\ 0.6 \end{bmatrix} + \begin{bmatrix} -0.12 \\ -0.12 \end{bmatrix} = \begin{bmatrix} 1.28 \\ 0.48 \end{bmatrix}$

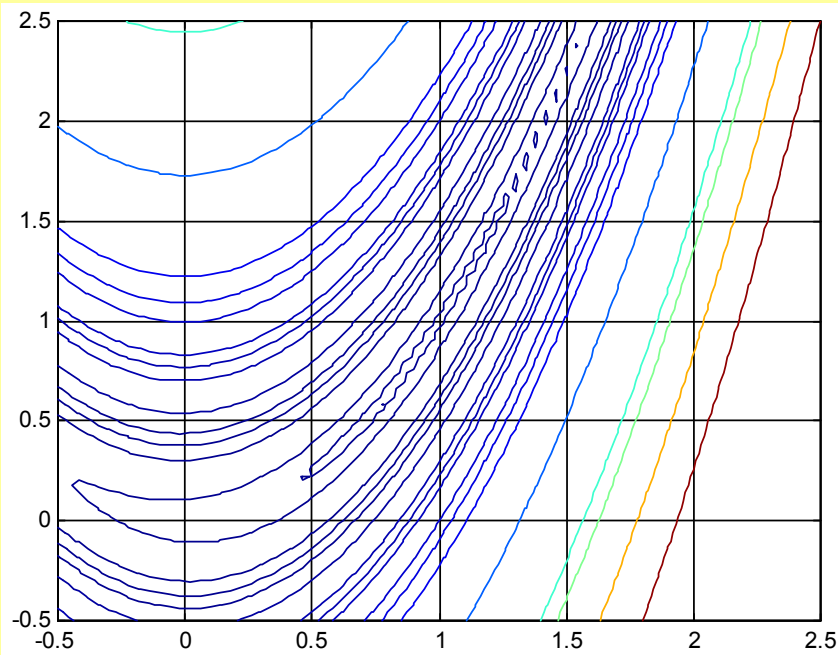
9.  $\mathbf{B}^{(2)} = \mathbf{B}^{(1)} + \frac{\mathbf{f}^{(2)} \mathbf{h}^{(1)T}}{\mathbf{h}^{(1)T} \mathbf{h}^{(1)}} = \begin{bmatrix} 1.15 & -0.15 \\ 0.15 & 0.85 \end{bmatrix}$

6b.  $R_f(\mathbf{x}_f^{(3)}) = 0.1308$

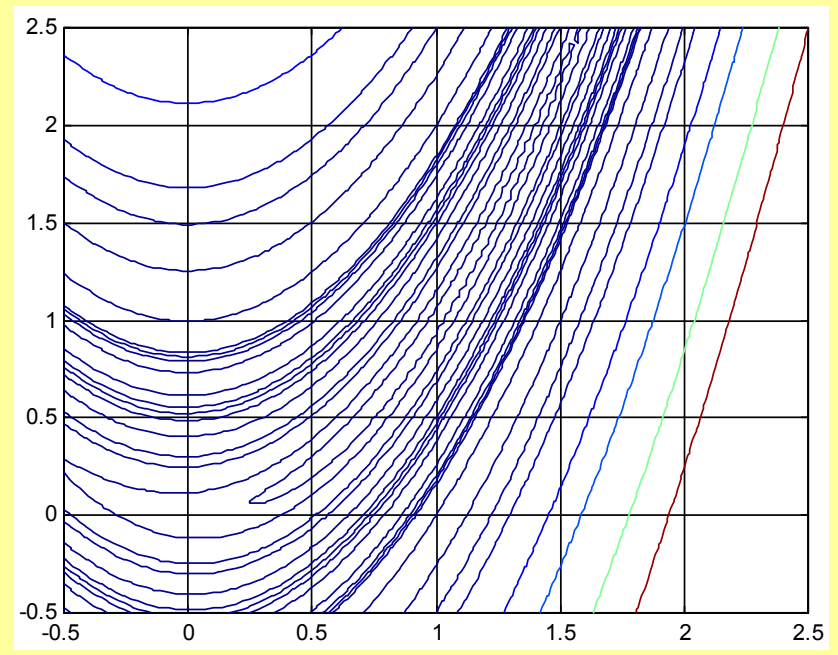


## Third Parameter Extraction

single  $l_1$  point parameter extraction



single  $l_2$  point parameter extraction



$$\mathbf{x}_c^{(3)} = \begin{bmatrix} 1.012 \\ 0.988 \end{bmatrix}$$



## Space Mapping Solution Process (continued)

7b. When  $\mathbf{x}_c^{(3)} = \begin{bmatrix} 1.012 \\ 0.988 \end{bmatrix}$ ,  $R_c(\mathbf{x}_c^{(3)}) = R_f(\mathbf{x}_f^{(3)})$

4c.  $\mathbf{h}^{(3)} = -\mathbf{B}^{(3)-1} \mathbf{f}^{(3)} = \begin{bmatrix} -0.0082 \\ 0.0151 \end{bmatrix}$

8b.  $\mathbf{f}^{(3)} = \begin{bmatrix} 1.012 \\ 0.988 \end{bmatrix} - \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0.012 \\ -0.012 \end{bmatrix}$

5c.  $\mathbf{x}_f^{(4)} = \begin{bmatrix} 1.28 \\ 0.48 \end{bmatrix} + \begin{bmatrix} -0.0082 \\ 0.0151 \end{bmatrix} = \begin{bmatrix} 1.2718 \\ 0.4951 \end{bmatrix}$

9b.  $\mathbf{B}^{(3)} = \mathbf{B}^{(2)} + \frac{\mathbf{f}^{(3)} \mathbf{h}^{(2)T}}{\mathbf{h}^{(2)T} \mathbf{h}^{(2)}} = \begin{bmatrix} 1.1 & -0.2 \\ 0.2 & 0.9 \end{bmatrix}$

6c.  $R_f(\mathbf{x}_f^{(4)}) = 9.2 \times 10^{-8}$ , then  $\bar{\mathbf{x}}_f = \mathbf{x}_f^{(4)}$   
and we can end the algorithm



## **Conclusions**

the Aggressive Space Mapping (ASM) algorithm with multiple point parameter extraction was illustrated making use of two analytical Rosenbrock functions

the improvement in the uniqueness of the parameter extraction due to multiple point matching was graphically illustrated

the ASM algorithm was executed in a step by step fashion, and the corresponding space mapped solution was accomplished after three SM iterations for the case of shifting, scaling and rotation of the input parameters