

# **CSE/ECE 848**

## **Introduction to**

# **Evolutionary Computation**

### **Module 5, Lecture 22, Part 1**

## **Engineering Component**

## **Design**

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# Overview

- EC methods are routinely applied to engineering design and process optimization tasks
  - Flexibility in modifying EC methods
    - Plug-and-play (!), easily customizable, modular, unified single and multi-objective, multi-level, robust/reliable, dynamic, etc.
  - No hard requirement for gradients
  - Often, an approximate feasible solution is enough
- Hybrid EC-local search approaches are more efficient

# Six-Objective Blackbox Problem from Practice

- **Black-box** functions computed from responses surfaces from expensive simulation models
- **145 input variables:**  $\mathbf{x} = [x_1, x_2, \dots, x_{145}]$
- **147 responses:**  $\mathbf{y} = [y_1, y_2, \dots, y_{147}]$
- A single call of **eval()**, for any input  $\mathbf{x} = [x_1, x_2, \dots, x_{145}]$  returns the corresponding responses  $\mathbf{y} = [y_1, y_2, \dots, y_{147}]$  and Jacobian  $\mathbf{J}$ :  $(\mathbf{y}, \mathbf{J}) = \text{eval}(\mathbf{x})$ , where  $\dim(\mathbf{J}) = 147 \times 145$  and  $\mathbf{J}_{i,j} = \partial y_i / \partial x_j$
- Gradients are available!

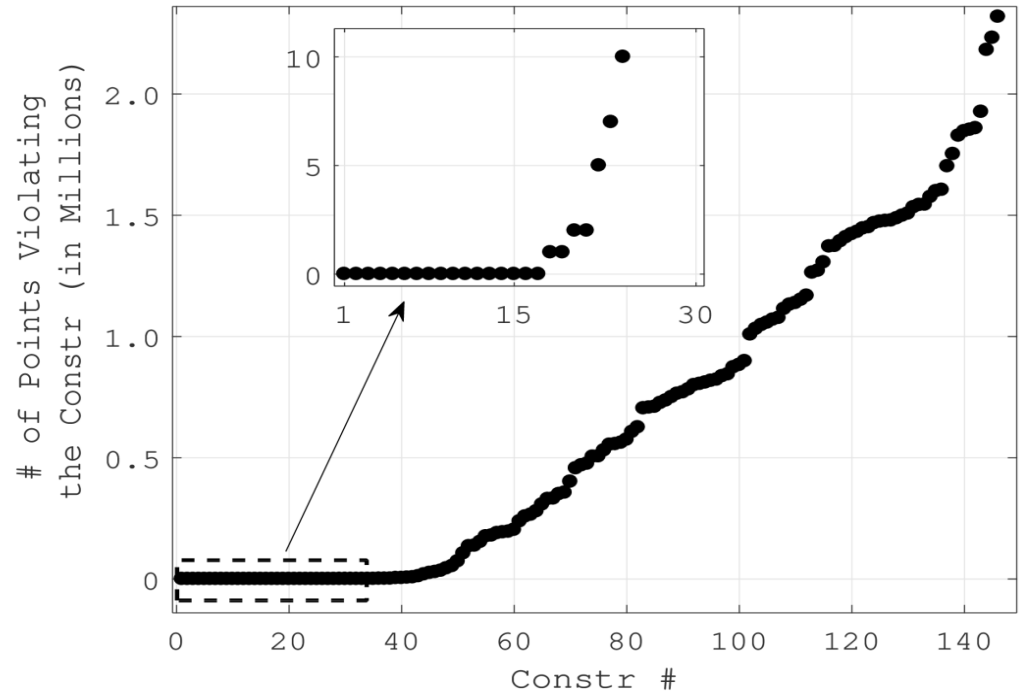
Reference: Gaur, A., Talukder, K., Deb, K., Tiwari, S., Xu, S., and Jones, D. (2020). Unconventional Optimization for Achieving Well-Informed Design Solutions for an Automobile Industry. *Engineering Optimization*, 52 (9). 1542–1560.

# The Problem (Contd...)

- Each design variable is bounded and **discretized** at 0.05 interval
- **146 Constraints:** Responses  $[y_2, y_3, \dots, y_{147}]$  have either upper or lower bound, but not both
- **6 Objectives:**  $y_1$  is the **primary objective** and  $[y_6, y_{14}, y_{29}, y_{108}, y_{146}]$  are **secondary objectives**. The rest of the  $y_i$ 's are constraints
  - One supplied solution  $y_1 = 184$  (improvement by 10 is target!)
- The problem has a very narrow and disconnected feasible region, because of 146 inequality constraints (mostly active) and discrete

# Feasibility of Search Space

- 2.5 million Latin Hypercube samples. Here, x-axis is the response ID and y-axis shows how many solutions are violating them.
- No feasible solution out of 2.5M random solutions
- Matlab's `fmincon()` failed to produce a feasible solution beyond 184

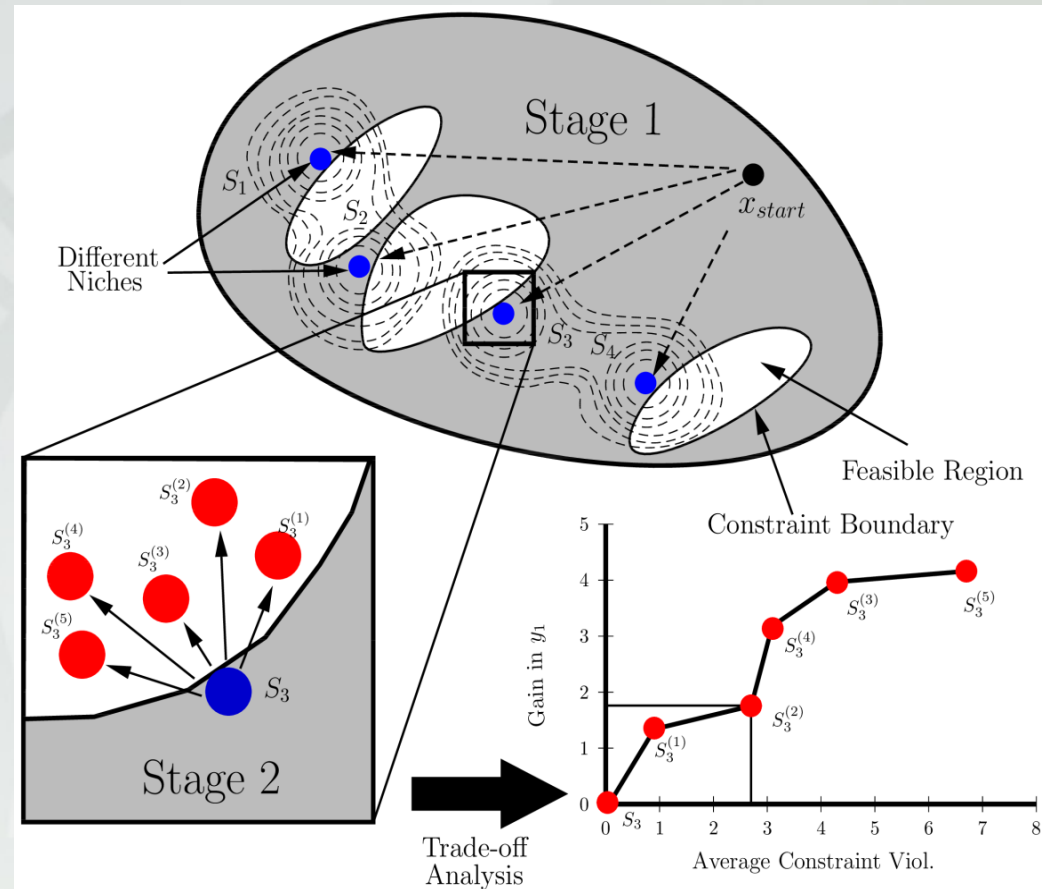


# Goal of the Study

- **Convergence:** Solutions in an **acceptable range** of  $y_1$ , given a start solution with  $y_1 = 184$
- **Trade-offs:** Need to know trade-offs in secondary objectives for good/acceptable designs
- **Niche/Diversity:** Need diverse designs (in  $x$ -space)
- **Computational Budget:** Maximum 50,000 calls to `eval()`
- **Post Optimality DSS:** How to choose a single preferred solution from multiple acceptable solutions?
- **Sensitivity analysis:** Choose a preferred solution based on trade-off between  $y_1$  and constraint violation

# Multi-objective Trade-off Analyzer (MOTRAN)

- **Stage1:** Find a diverse set of solutions
- **Stage2:** Perform sensitivity analysis to choose a preferred solution



# MOTRAN Stage-1

## (Many-Objective Optimization)

- 6 Objective handling through **NSGA-III** (Deb and Jain, IEEE-TEVC, 2014)
- **Custom NSGA-III** to protect best feasible solutions of Obj-1
  - Add archive of feasible solutions in **Merged Population**
  - Apply Derivative-based **Local-Search** to every new feasible solution and save all intermediate feasible solutions in archive
  - Update reference directions to **protect best ten  $y_1$  and x-niche** solutions with  $d_{\text{niche}}$  distance
  - Need diverse designs having similar Obj-1 values
- Provide **few promising choices** to DM
  - **Post-optimization DS program** : Choose most diverse three solutions with acceptable values of Obj-1 at the end



# MOTRAN Stage-1: Local-Search

A point-based approach

0)  $\mathbf{x} \leftarrow \mathbf{x}_{start}$

1) Evaluate  $\mathbf{x}$

$$(y_{1x}, \mathbf{g}_x, \mathbf{y}_{1x}', \mathbf{G}_x) = \Psi(\mathbf{x})$$

$$\mathbf{x} = [x_1, x_2, \dots, x_{145}]$$

$y_{1x}$ : Obj-1 value at  $\mathbf{x}$

$\mathbf{g}_x = [g_1, g_2, \dots, g_{146}]$ : Constraint values at  $\mathbf{x}$

$\mathbf{y}_{1x}' = \partial y_1 / \partial x_i$ : Gradient values of  $y_1$  at  $\mathbf{x}$

$\mathbf{G}_{x_i, j} = \frac{\partial g_i}{\partial x_j}$ : Gradient of Constraint  $g_i$  w.r.t Variable  $x_j$

2) Find  $x_k$  having Max-Feasible-Descent (expected) as per gradient values and a step size of  $\delta_k = \pm 0.05$  (based on  $\partial y_1 / \partial x_k = \mp$ )

$$CV_k = \sum_i \nabla g_i(x_k) \delta_k$$

Choose the  $x_k$  having min  $CV_k$

3) Evaluate  $\mathbf{z} = [x_1, x_2, \dots, x_k + \delta_k, \dots, x_{145}]$

$$(y_{1z}, \mathbf{g}_z, \mathbf{y}_{1z}', \mathbf{G}_z) = \Psi(\mathbf{z})$$

4) If (z is Feasible) AND ( $y_{1z} < y_{1x}$ )

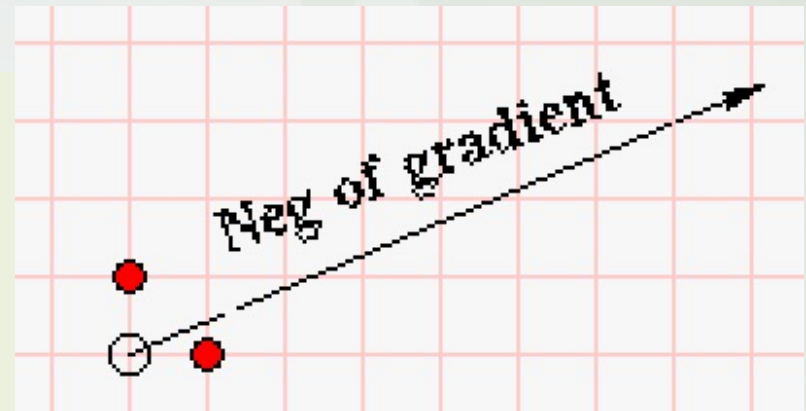
$$\mathbf{x} \leftarrow \mathbf{z}$$

Go to Step-2

Else

Return  $\mathbf{x}$

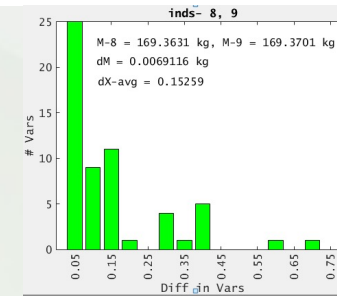
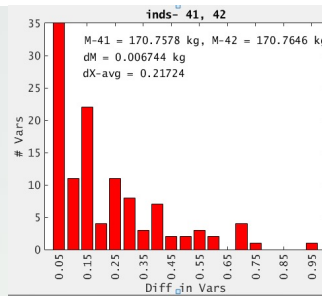
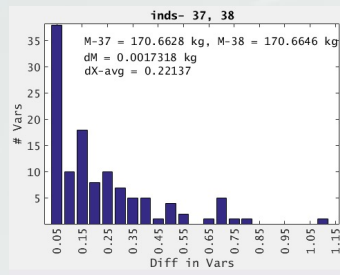
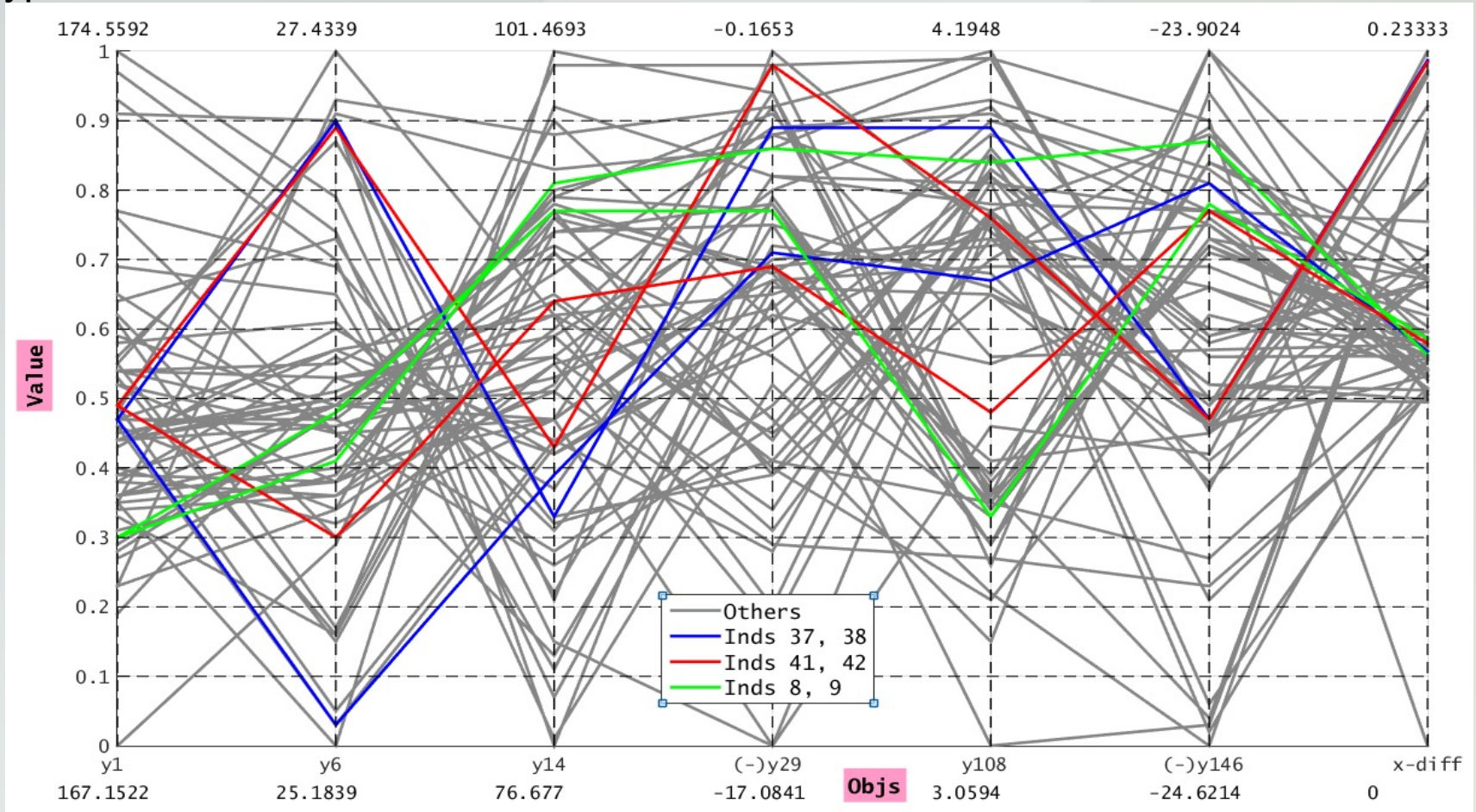
Stop!



# MOTRAN Stage-1 (Results)

A Typical NSGA-III Run:

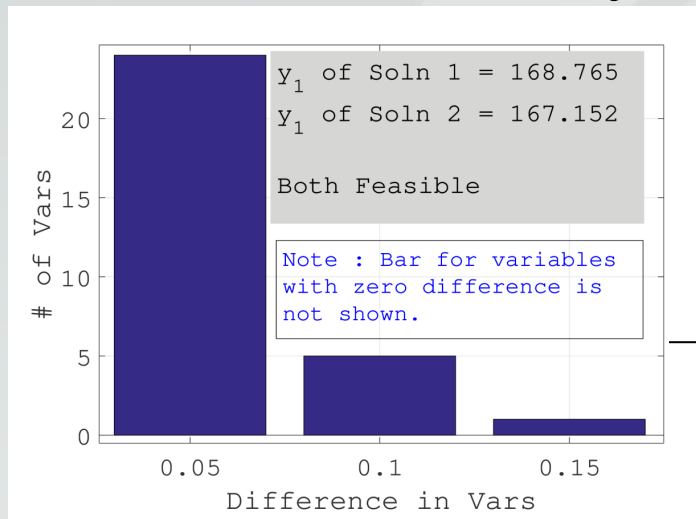
Modified NSGA-III Results



# MOTRAN Stage-1

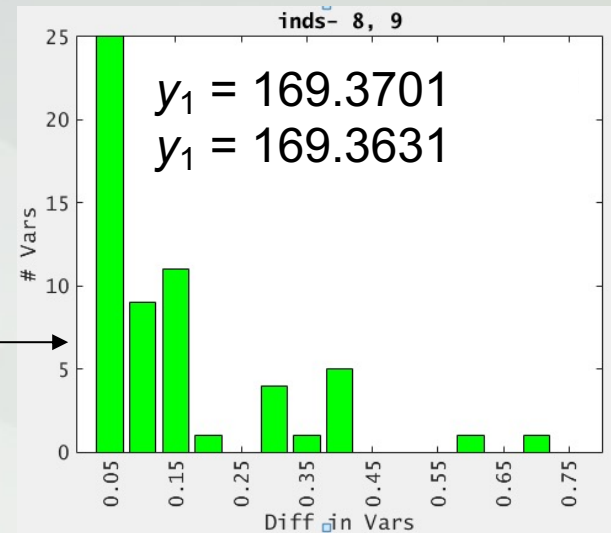
## (Diversity Comparison to LS)

- 1.613 difference in  $y_1$



Improvement in Diversity

- 0.007 difference in  $y_1$

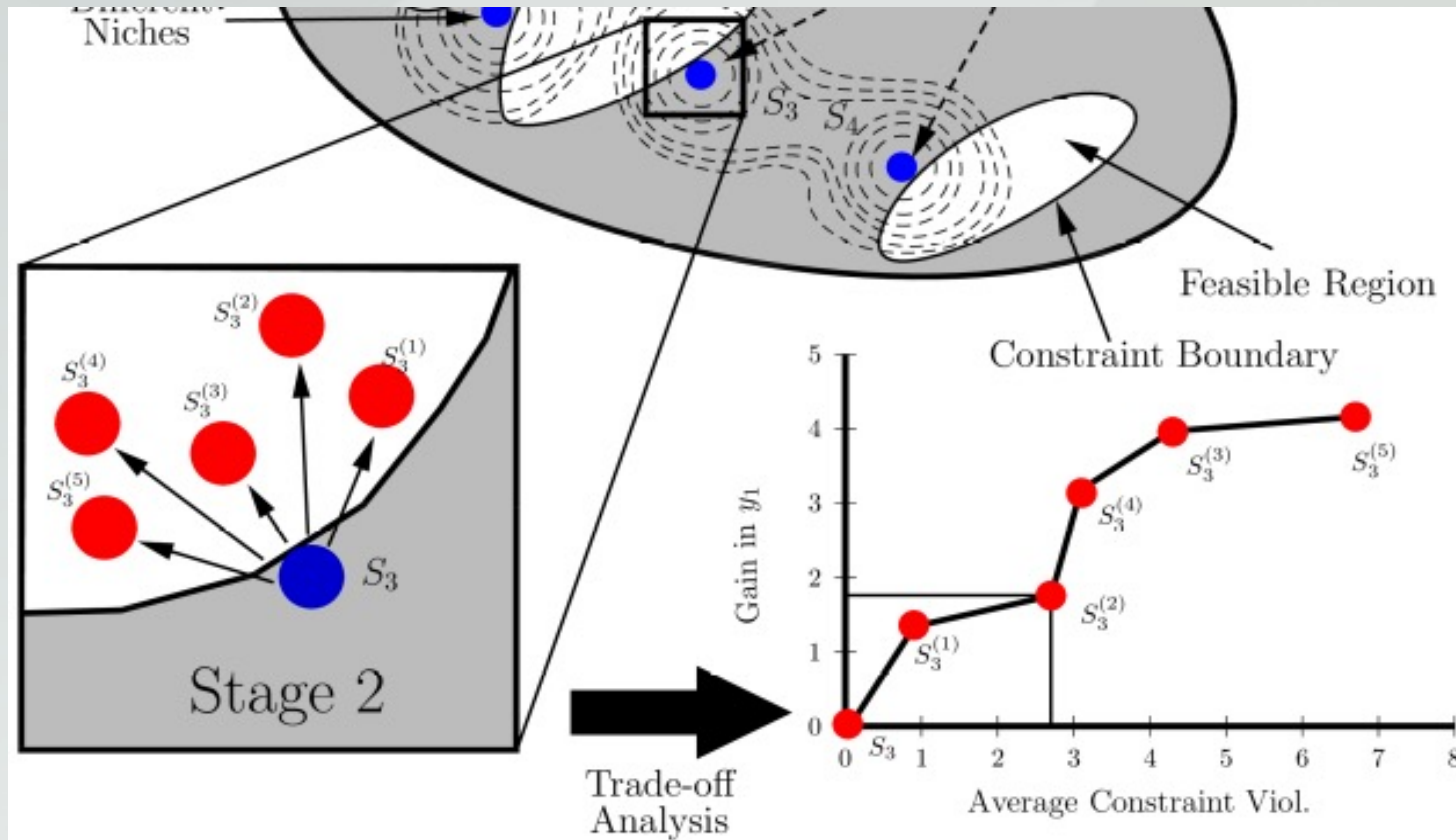


- 30 variables different
- As large as 3 steps different

- 63 variables different
- As large as 14 steps different

EMO capable of finding more diverse solns.

# MOTRAN Stage-2



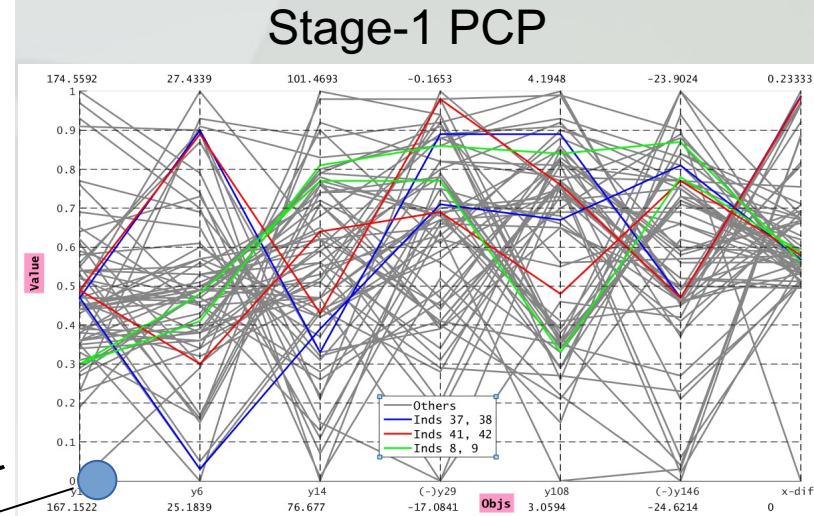
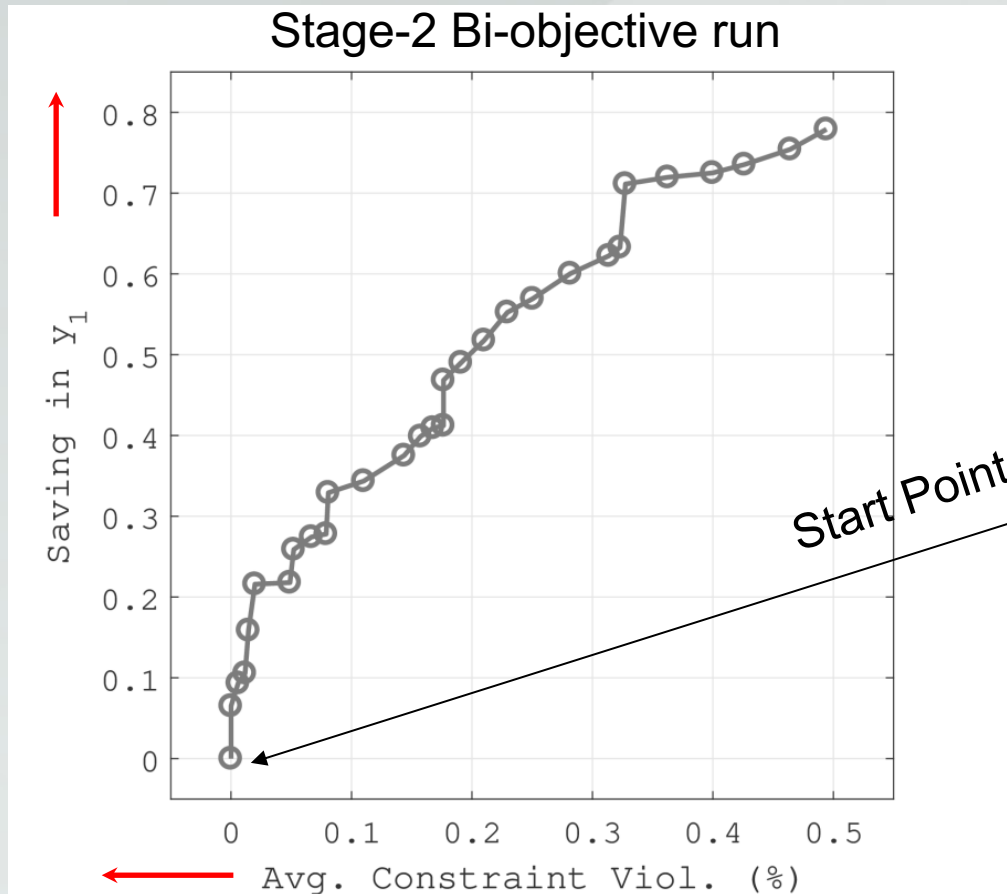
- Stage2: Sensitivity of best  $y_1$  solutions

# MOTRAN Stage-2

- Two Objectives:
  - Maximize Improvement in  $y_1$
  - Minimize Average Constraint Violation
- Apply **NSGA-II**
- Search restricted within neighborhood
  - No feasible 1-jump is effective
  - Max jump is limited to **3-5 variables** and go one step in each variable, then follow the same CV check as in Phase 1
- Low Computational budget of **5,000 FEs**
- New points aided with **Local-Search**



# MOTRAN Stage-2



$y_1 = 167.16$

- Improvement of  $y_1$  of 0.8 in Stage 2 due to focused search

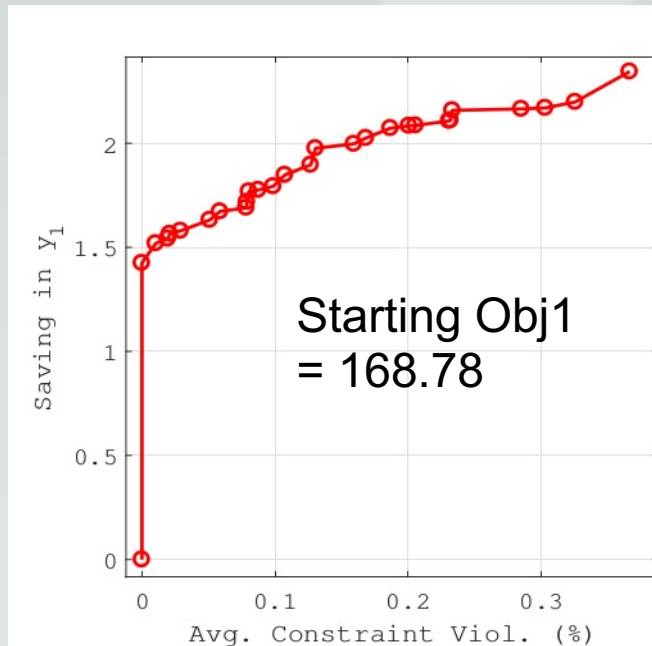
$y_1 = 166.36$

- Start Point : Lowest Obj-1 from Stage-1 or a chosen point

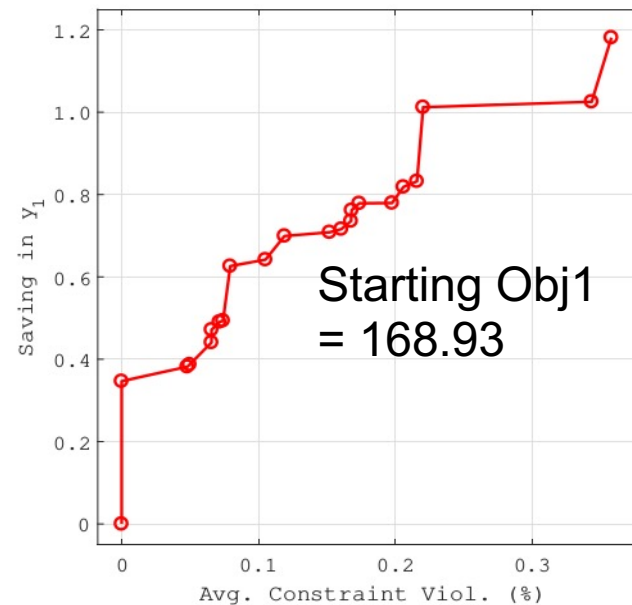
# MOTRAN Stage-2

## (Comparison of Diff. Stage1 Solutions)

Green solutions



(a)  $y_1 = 167.3$



(b)  $y_1 = 168.55$

- Neighborhood comparison of two solns. with similar Obj-1 but different Niches

# End of Module 5, Lecture 22, Part 1

- Real world problem solving is different from the academic benchmark studies
- Many problems require multiple optimization techniques to come up with desired solutions
  - Many conflicting objectives, niching, uncertainty handling etc.
  - Local Search for faster convergence
  - $y_1$  reduced from 184 to 166.36 (17.64 reduction, 10 was desired)
- Humans are in charge!