

Appendix of “Integration of Preferences in Decomposition Multi-Objective Optimization”

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1 Choosing Pivot Point

In NUMS, the pivot point \mathbf{w}^p is used to represent the ROI on the simplex Ψ^m . In this paper, we use the projection of the DM supplied aspiration level vector \mathbf{z}^r onto Ψ^m to serve this purpose. In theory, finding such projection is equivalent to solve the following minimization problem:

$$\mathbf{w}^p = \underset{\mathbf{w} \in \Psi^m}{\operatorname{argmin}} \|\mathbf{w} - \mathbf{z}^r\| \quad (1)$$

As discussed in [1], the solution of (1) is nontrivial and there does not have an explicit form. Here, we use the theorem developed in [1] to find such projection. Algorithm 1 gives the corresponding pseudo-code and interested readers are recommended to [1] for detailed mathematical deductions.

Algorithm 1: Projection of $\mathbf{z}^r \in \mathbb{R}^m$ onto the simplex Ψ^m

Input: DM supplied aspiration level vector \mathbf{z}^r

Output: Pivot point \mathbf{w}^p on Ψ^m

1 Sort \mathbf{z}^r in ascending order as $z_1^r \leq \dots \leq z_m^r$;

2 **for** $i \leftarrow m - 1$ **to** 1 **do**

3 $t_i \leftarrow \frac{\sum_{j=i+1}^m z_j^r - 1}{m - i}$;

4 **if** $t_i \geq z_i^r$ **then**

5 $\hat{t} \leftarrow t_i$;

6 **return** $\mathbf{w}^p \leftarrow [\mathbf{z}^r - \hat{t}]_+$

7 $\hat{t} \leftarrow \frac{\sum_{j=1}^m z_j^r - 1}{m}$;

8 **return** $\mathbf{w}^p \leftarrow [\mathbf{z}^r - \hat{t}]_+$

2 Proof of Theorem 1

Proof. Let us use a specific example shown in Fig. 1 to prove this theorem. Suppose the reference points are originally generated by the Das and Dennis’s method [2]. Therefore, reference points are distributed on an unit m -simplex. Let the centroid of this simplex, i.e., $\mathbf{w}^p = (\frac{1}{m}, \dots, \frac{1}{m})^T$, be the

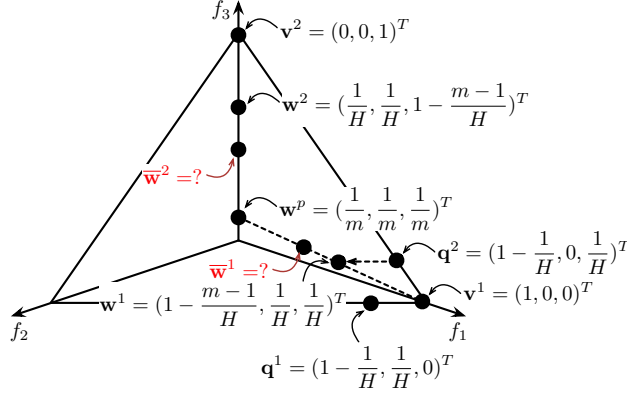


Figure 1: Illustrative example for η computation.

pivot point. Let us consider $\mathbf{v}^1 = (1, 0, 0)^T$ and $\mathbf{v}^2 = (0, 0, 1)^T$ as two vertices of an edge of this simplex. Obviously, the length of each edge of this simplex is the same, i.e., $\|\mathbf{v}^1 - \mathbf{v}^2\| = \sqrt{2}$. Suppose \mathbf{w}^1 and \mathbf{w}^2 are two reference points inside this simplex and closest to \mathbf{v}^1 and \mathbf{v}^2 . Since reference points are generated in a structured manner, we can use a geometric method to find the coordinates of \mathbf{w}^1 and \mathbf{w}^2 . As shown in Fig. 1, \mathbf{q}^1 and \mathbf{q}^2 are two reference points lying on the two edges and closest to \mathbf{v}^1 . Obviously, \mathbf{w}^1 is a linear combination of \mathbf{v}^1 , \mathbf{q}^1 and \mathbf{q}^2 as:

$$w_i^1 = (q_i^2 - v_i^1) + q_i^1 \quad (2)$$

where $i \in \{1, \dots, m\}$. In summary, we can have $\mathbf{w}^1 = (1 - \frac{m-1}{H}, \dots, \frac{1}{H})^T$ and $\mathbf{w}^2 = (\frac{1}{H}, \dots, 1 - \frac{m-1}{H})^T$. Based on our non-uniform mapping scheme, we have the new locations of \mathbf{w}^1 and \mathbf{w}^2 can be calculated as:

$$\bar{\mathbf{w}}^i = \mathbf{w}^p + t \times \mathbf{u}^i \quad (3)$$

where $i \in \{1, 2\}$, $\mathbf{u}^i = \frac{\mathbf{v}^i - \mathbf{w}^p}{\|\mathbf{v}^i - \mathbf{w}^p\|}$ and

$$t = d - d\left(\frac{d-D}{d}\right)^{\frac{1}{\eta+1}} \quad (4)$$

where $d = \|\mathbf{v}^i - \mathbf{w}^p\|$ and $D = \|\mathbf{w}^i - \mathbf{w}^p\|$. Let $q = \frac{d-D}{d}$, we have:

$$t = d(1 - q^{\frac{1}{\eta+1}}) \quad (5)$$

In order to have the extent of ROI become size of τ of the EF, we have the following equation:

$$\frac{\|\bar{\mathbf{w}}^1 - \bar{\mathbf{w}}^2\|}{\|\mathbf{v}^1 - \mathbf{v}^2\|} = \tau \quad (6)$$

Since $\|\mathbf{v}^1 - \mathbf{v}^2\| = \sqrt{2}$, we have:

$$\|\bar{\mathbf{w}}^1 - \bar{\mathbf{w}}^2\| = \sqrt{2}\tau \quad (7)$$

Using equation (3) to substitute $\bar{\mathbf{w}}^1$ and $\bar{\mathbf{w}}^2$ in equation (7), we have:

$$\|t \times (\mathbf{u}^1 - \mathbf{u}^2)\| = \sqrt{2}\tau \quad (8)$$

Using equation (5) to substitute t in equation (8), we have:

$$d(1 - q^{\frac{1}{\eta+1}}) \times \|\mathbf{u}^1 - \mathbf{u}^2\| = \sqrt{2}\tau \quad (9)$$

By substitution, we have:

$$\begin{aligned}
(1 - q^{\frac{1}{\eta+1}}) \times \|\mathbf{v}^1 - \mathbf{v}^2\| &= \sqrt{2}\tau \\
\implies q^{\frac{1}{\eta+1}} &= 1 - \tau \\
\implies \eta &= \frac{\log q}{\log(1 - \tau)} - 1
\end{aligned} \tag{10}$$

Since the coordinates of \mathbf{v}^1 , \mathbf{w}^1 and \mathbf{w}^p are known, we have:

$$\begin{aligned}
d &= \sqrt{(1 - \frac{1}{m})^2 + (m-1)\frac{1}{m^2}} \\
&= \sqrt{1 - \frac{1}{m}}
\end{aligned} \tag{11}$$

and

$$\begin{aligned}
D &= \sqrt{(1 - \frac{m-1}{H} - \frac{1}{m})^2 + (m-1)(\frac{1}{m} - \frac{1}{H})^2} \\
&= \sqrt{1 - \frac{1}{m}(1 - \frac{m}{H})} \\
&= (1 - \frac{m}{H})d
\end{aligned} \tag{12}$$

Based on equation (11) and equation (12), we have:

$$\begin{aligned}
q &= \frac{d - D}{d} \\
&= \frac{d - (1 - \frac{m}{H})d}{d} \\
&= \frac{m}{H}
\end{aligned} \tag{13}$$

■

3 Proof of Corollary 1

Proof. As discussed in Section III-D, we should set $\eta > 0$ in the NUMS. Thus, based on Theorem 1, we have:

$$\frac{\log \frac{m}{H}}{\log(1 - \tau)} > 1 \tag{14}$$

Since $\frac{m}{H} < 1$ and $1 - \tau < 1$, we have:

$$\log \frac{m}{H} \leq \log(1 - \tau) \implies 0 < \tau < 1 - \frac{m}{H} \tag{15}$$

■

4 Proof of Corollary 2

Proof. The proof of this corollary is similar to the Theorem 1. Let us use Fig. 1 for illustration again. As for the reference points \mathbf{w}^1 and \mathbf{w}^2 , we should have the following relationship after the

non-uniform mapping:

$$\frac{\|\bar{\mathbf{w}}^1 - \bar{\mathbf{w}}^2\|}{\|\mathbf{w}^1 - \mathbf{w}^2\|} = \tau \quad (16)$$

Since $\|\mathbf{w}^1 - \mathbf{w}^2\| = \sqrt{2}(1 - \frac{m}{H})$, we have:

$$\|\bar{\mathbf{w}}^1 - \bar{\mathbf{w}}^2\| = \sqrt{2}(1 - \frac{m}{H})\tau \quad (17)$$

Using equation (3) to substitute $\bar{\mathbf{w}}^1$ and $\bar{\mathbf{w}}^2$ in equation (17), we have:

$$\|t \times (\mathbf{u}^1 - \mathbf{u}^2)\| = \sqrt{2}(1 - \frac{m}{H})\tau \quad (18)$$

Using equation (5) to substitute t in equation (18), we have:

$$d(1 - q^{\frac{1}{\eta+1}}) \times \|\mathbf{u}^1 - \mathbf{u}^2\| = \sqrt{2}(1 - \frac{m}{H})\tau \quad (19)$$

Since $\|\mathbf{v}^1 - \mathbf{v}^2\| = \sqrt{2}$, by substitution, we have:

$$\begin{aligned} 1 - q^{\frac{1}{\eta+1}} &= (1 - \frac{m}{H})\tau \\ \implies \eta &= \frac{\log q}{\log[1 - (1 - \frac{m}{H})\tau]} - 1 \end{aligned} \quad (20)$$

where $q = \frac{m}{H}$ according to equation (13). ■

5 Proof of Corollary 3

Proof. Since $\eta > 0$, according to equation (20), we have:

$$\frac{\log q}{\log[1 - (1 - \frac{m}{H})\tau]} > 1 \quad (21)$$

Since $\frac{m}{H} < 1$ and $1 - (1 - \frac{m}{H})\tau < 1$, we have:

$$\log \frac{m}{H} < \log [1 - (1 - \frac{m}{H})\tau] \implies 0 < \tau < 1 \quad (22)$$
■

6 Plots of Final Populations

This section provides the visual comparisons of different preference-based EMO algorithms. In particular, we plot the final solutions obtained by different algorithms that achieve the best R-IGD value.

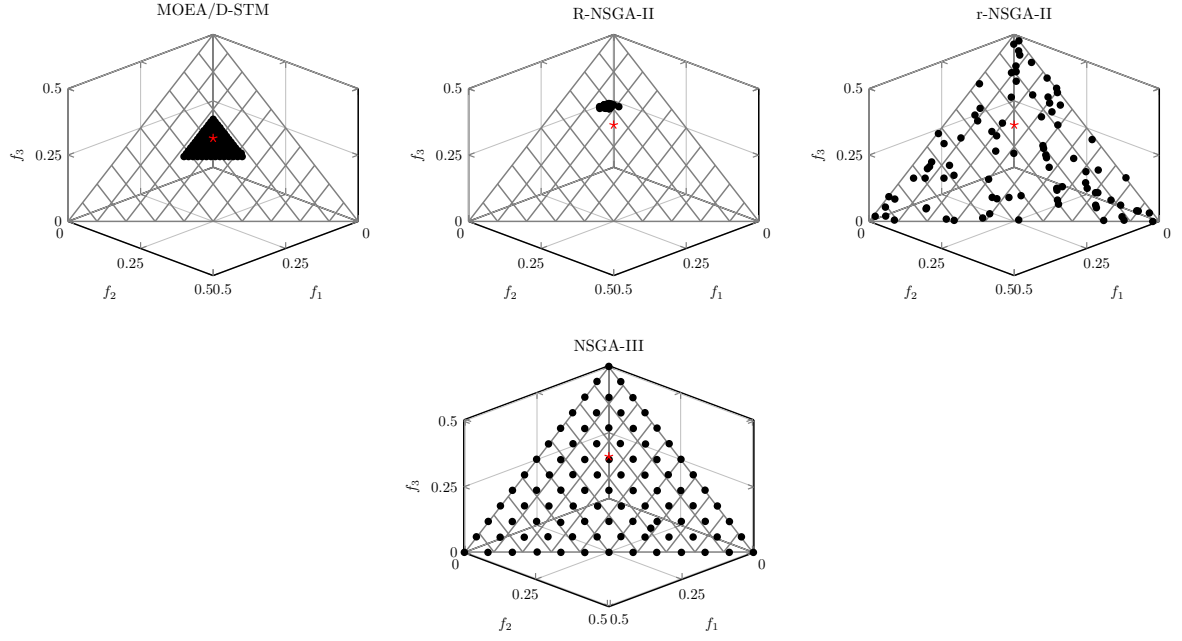


Figure 2: Comparisons on 3-objective DTLZ1 where $\mathbf{z}^r = (0.05, 0.05, 0.2)$.

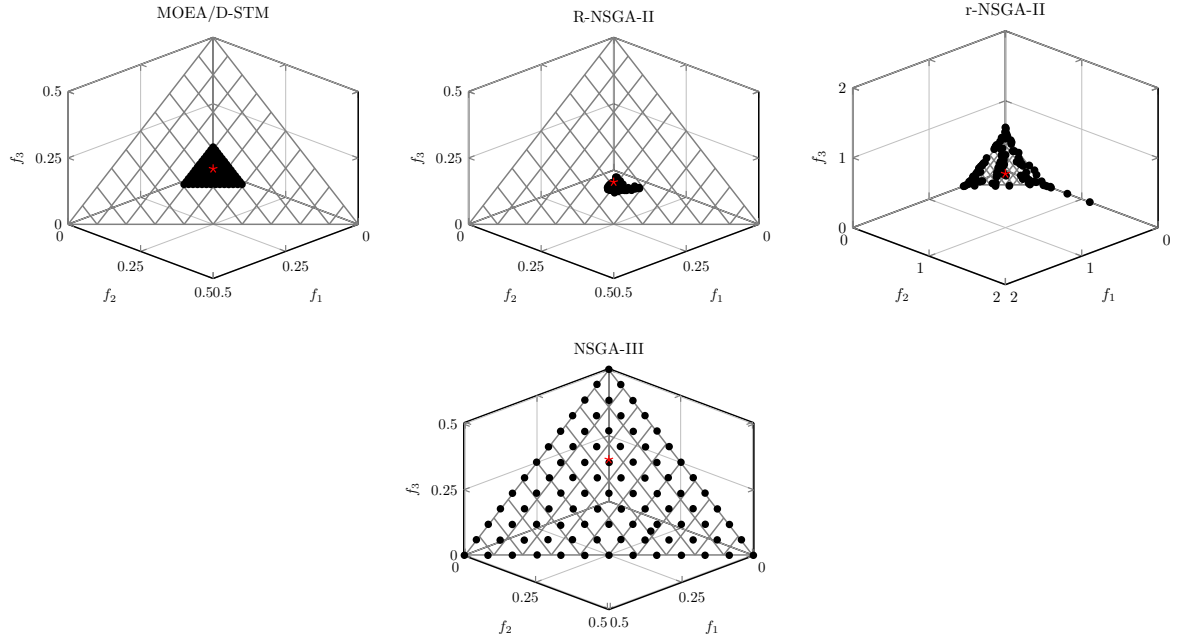


Figure 3: Comparisons on 3-objective DTLZ1 where $\mathbf{z}^r = (0.3, 0.3, 0.2)$.

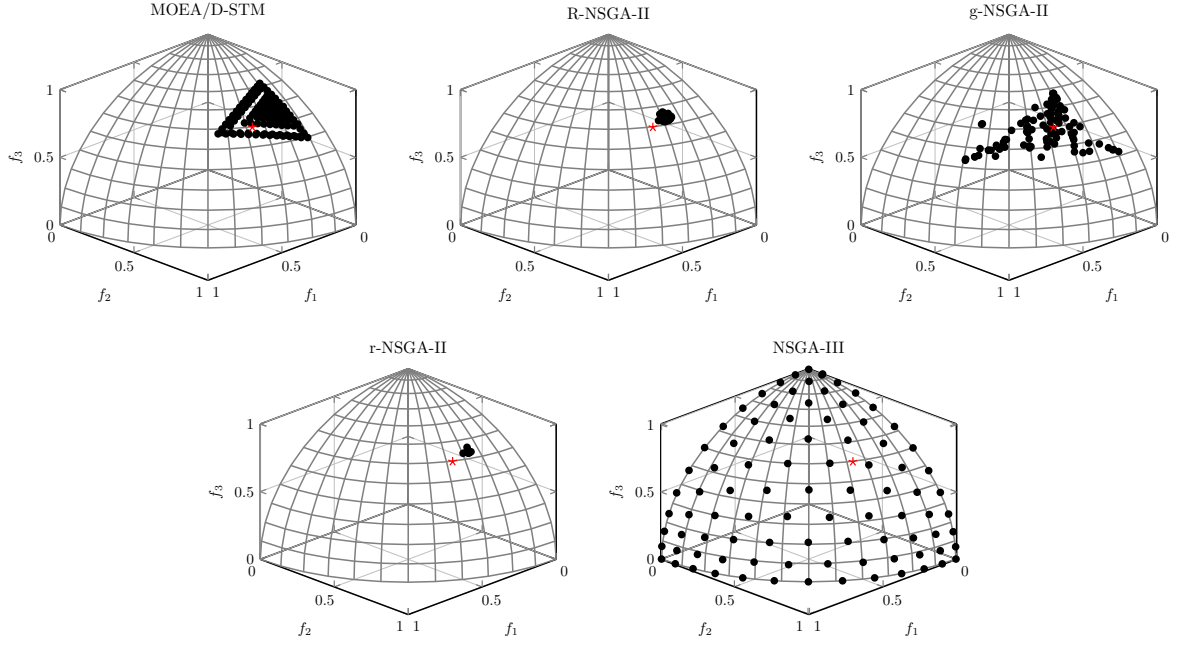


Figure 4: Comparisons on 3-objective DTLZ2 where $\mathbf{z}^r = (0.2, 0.5, 0.6)$.

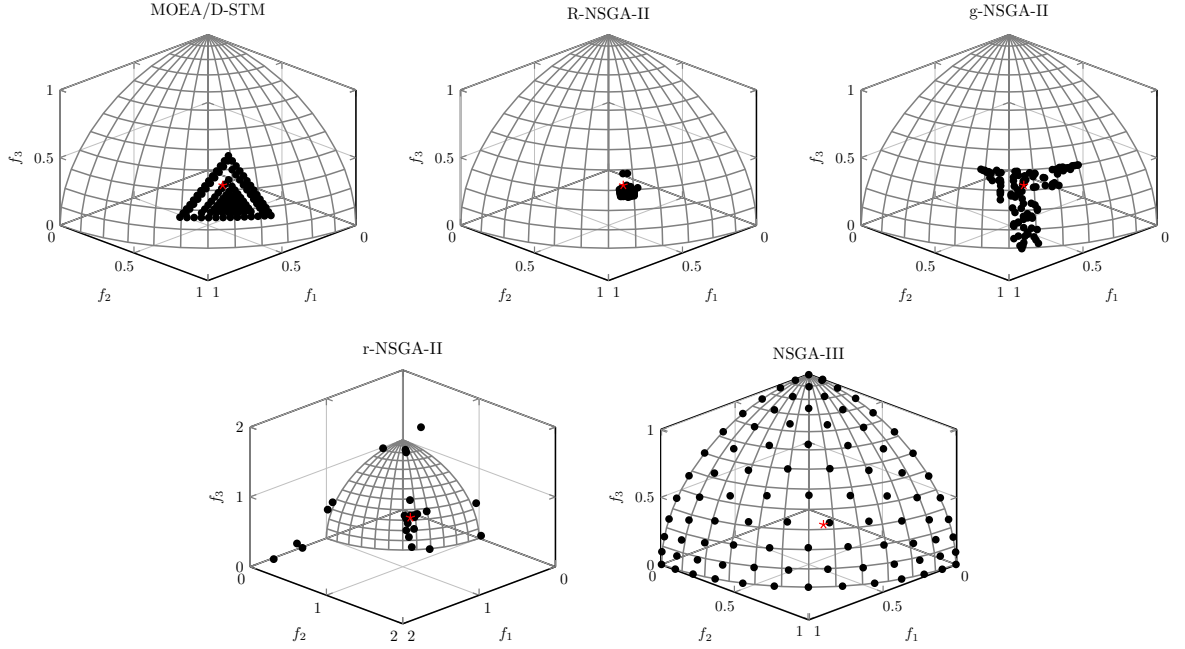


Figure 5: Comparisons on 3-objective DTLZ2 where $\mathbf{z}^r = (0.7, 0.8, 0.5)$.

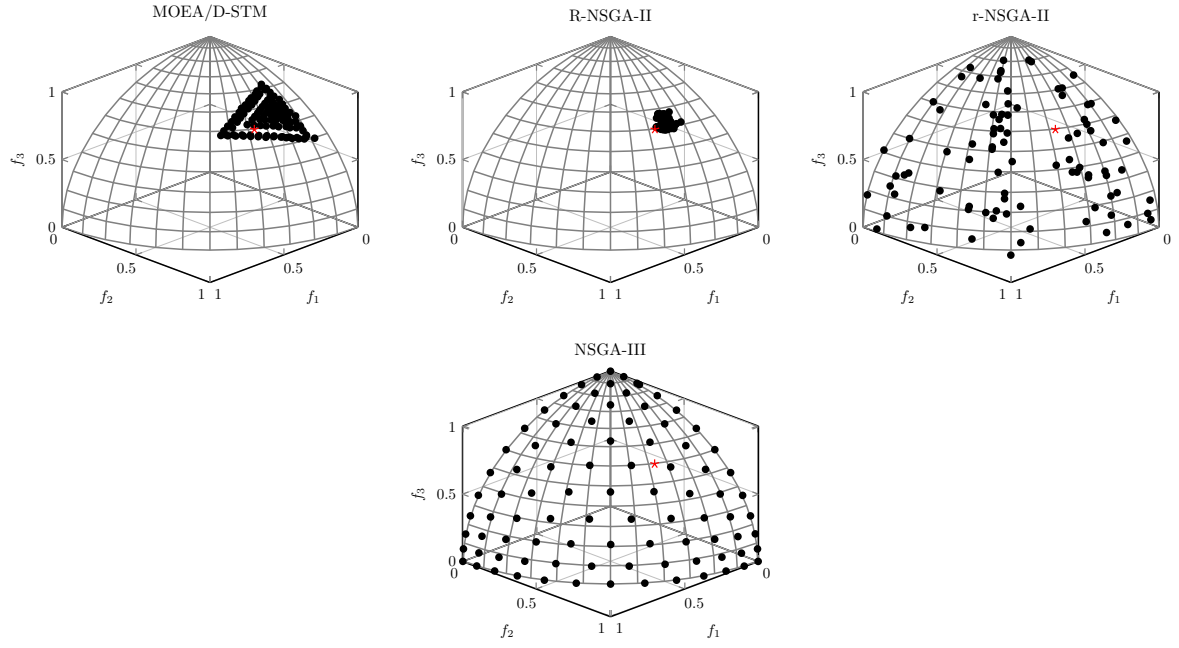


Figure 6: Comparisons on 3-objective DTLZ3 where $\mathbf{z}^r = (0.2, 0.5, 0.6)$.

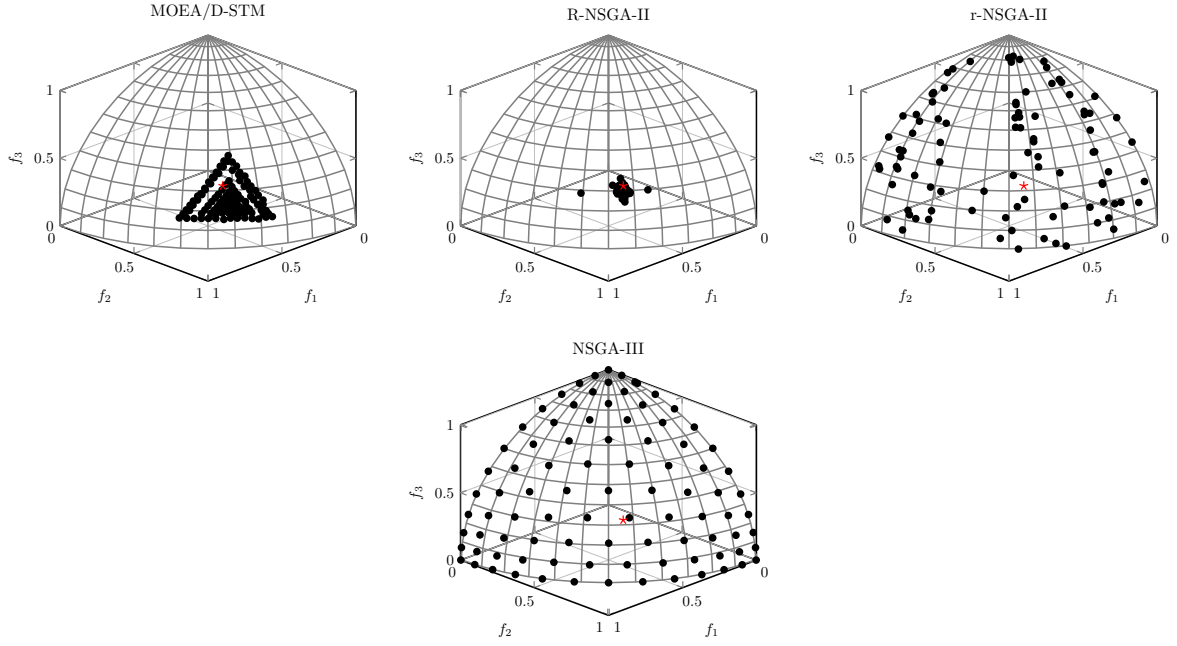


Figure 7: Comparisons on 3-objective DTLZ3 where $\mathbf{z}^r = (0.7, 0.8, 0.5)$.

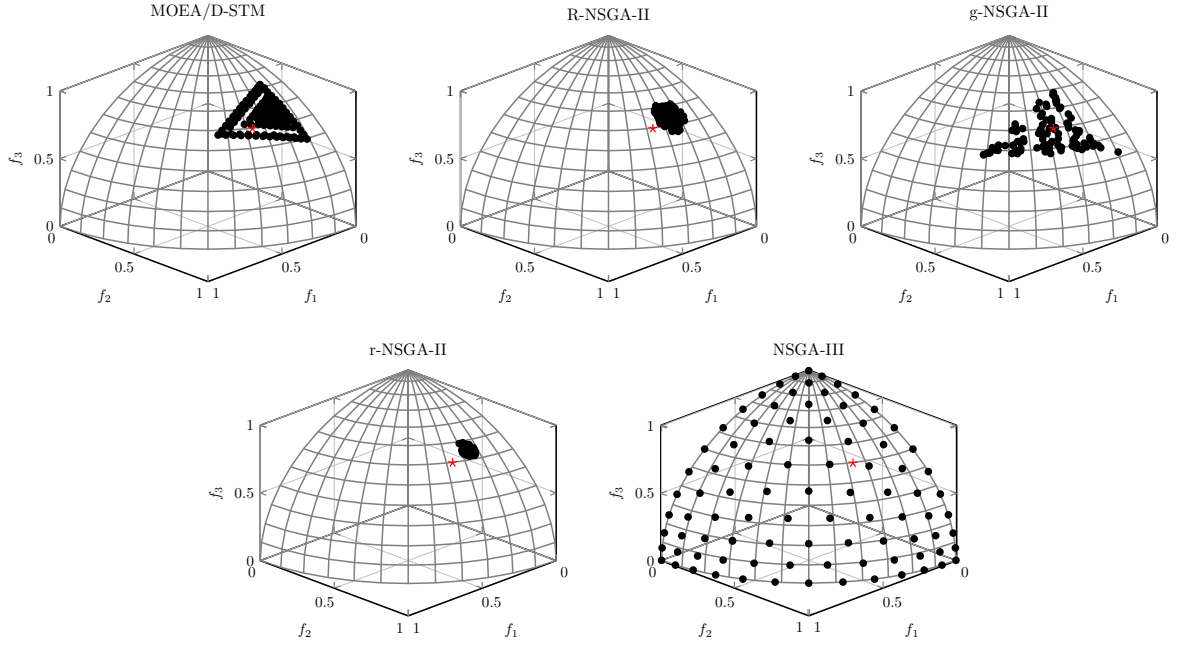


Figure 8: Comparisons on 3-objective DTLZ4 where $\mathbf{z}^r = (0.2, 0.5, 0.6)$.

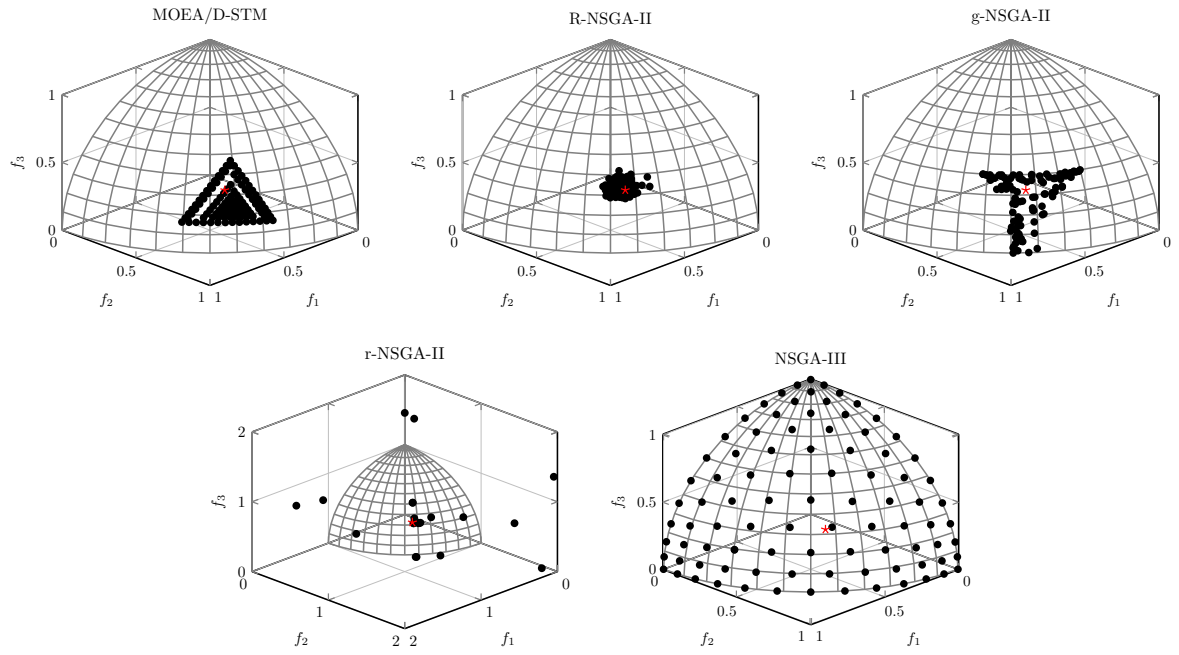


Figure 9: Comparisons on 3-objective DTLZ4 where $\mathbf{z}^* = (0.7, 0.8, 0.5)$.

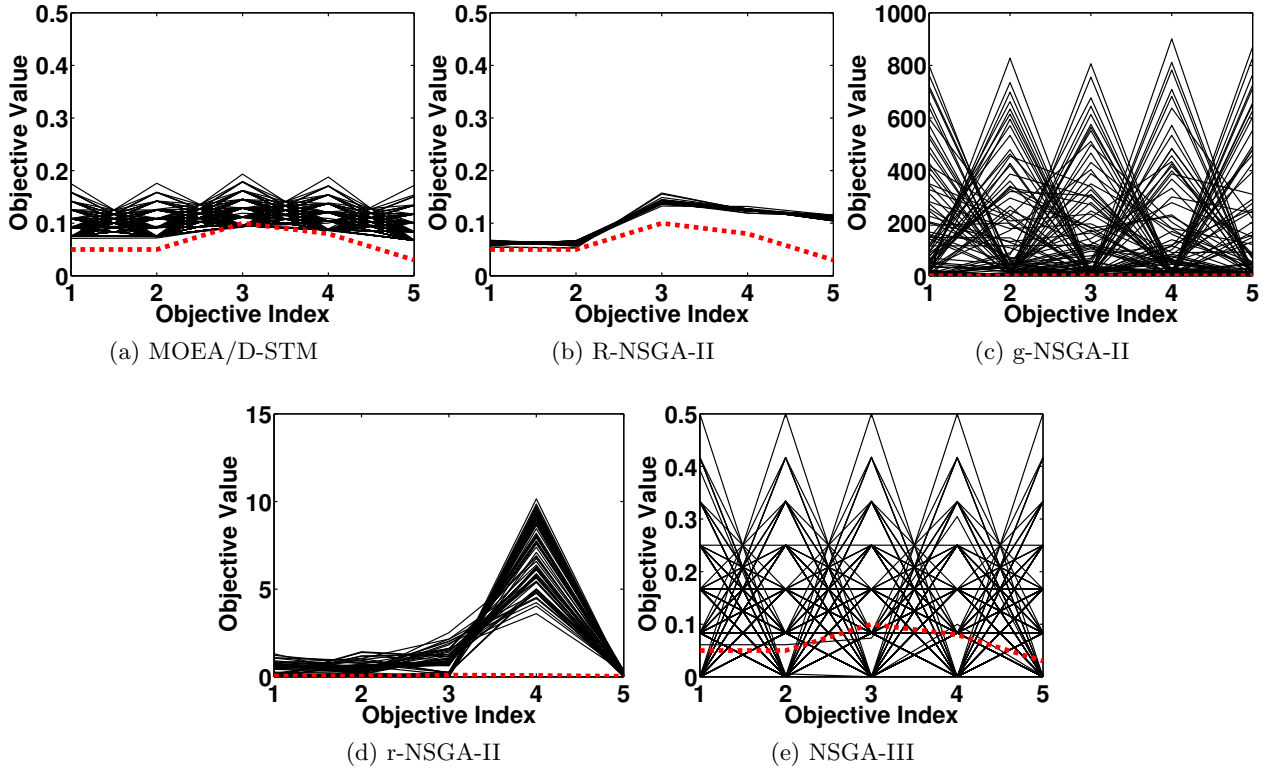


Figure 10: Comparisons on 5-objective DTLZ1 where $\mathbf{z}^r = (0.05, 0.05, 0.1, 0.08, 0.03)^T$.

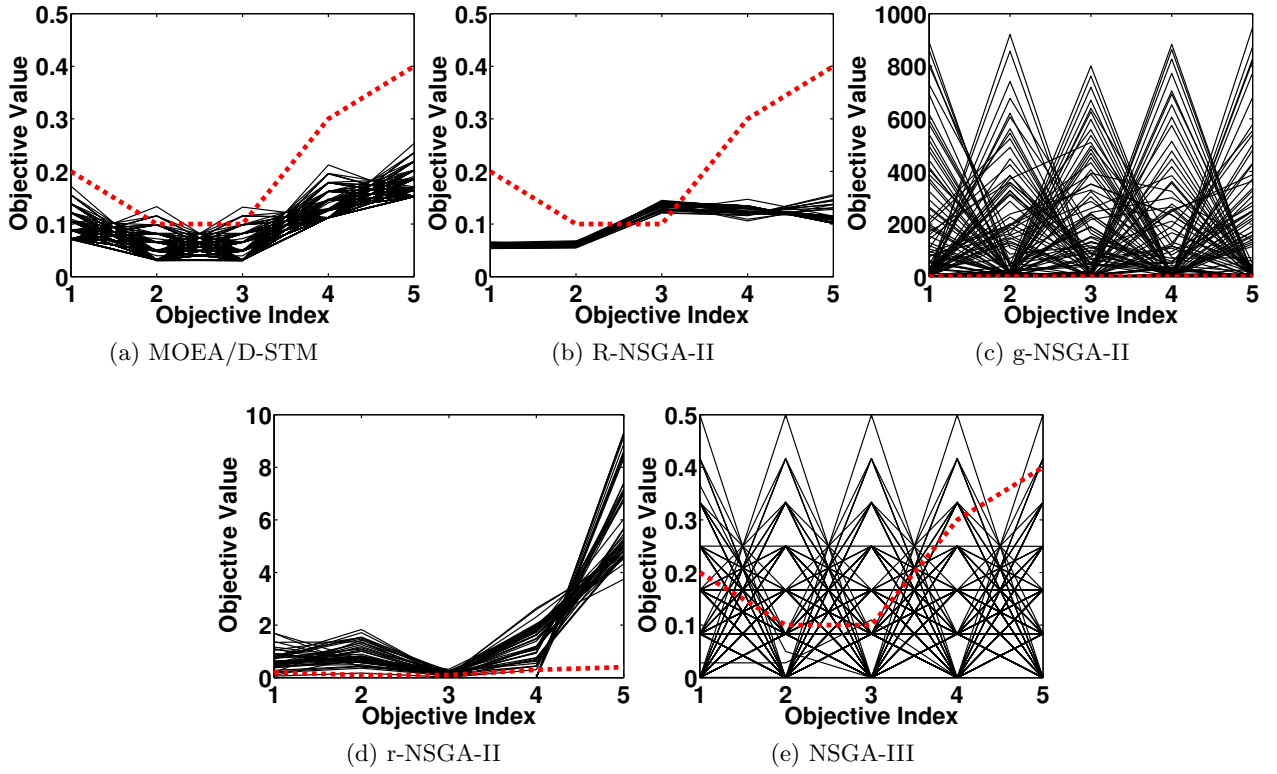


Figure 11: Comparisons on 5-objective DTLZ1 where $\mathbf{z}^r = (0.2, 0.1, 0.1, 0.3, 0.4)^T$.

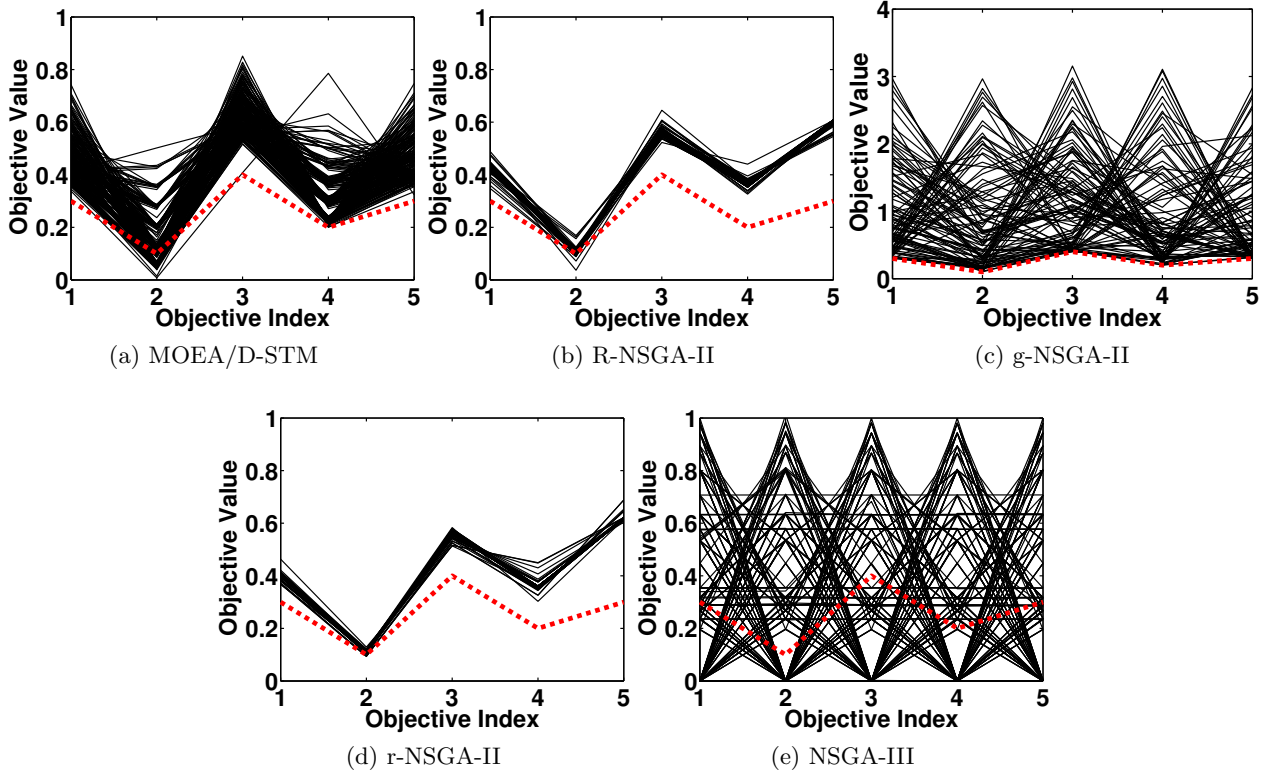


Figure 12: Comparisons on 5-objective DTLZ2 where $\mathbf{z}^r = (0.3, 0.1, 0.4, 0.2, 0.3)^T$.

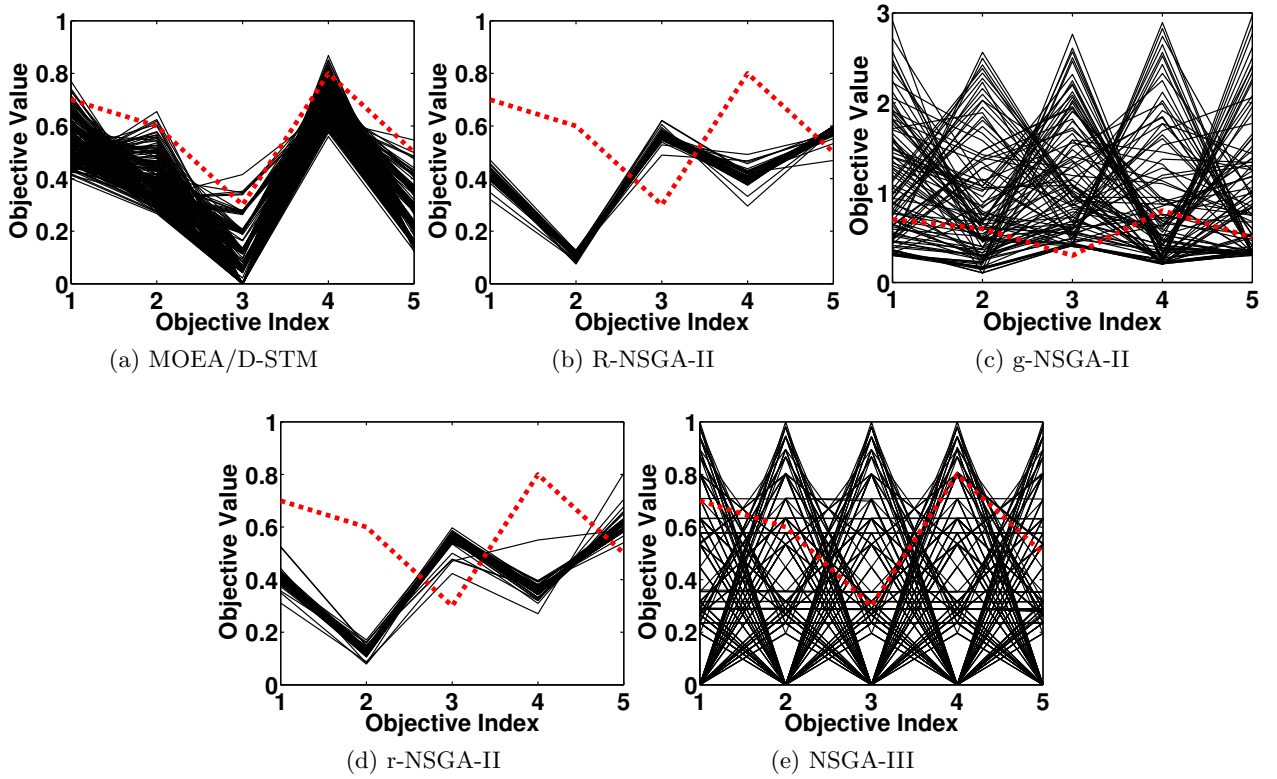


Figure 13: Comparisons on 5-objective DTLZ2 where $\mathbf{z}^r = (0.7, 0.6, 0.3, 0.8, 0.5)^T$.

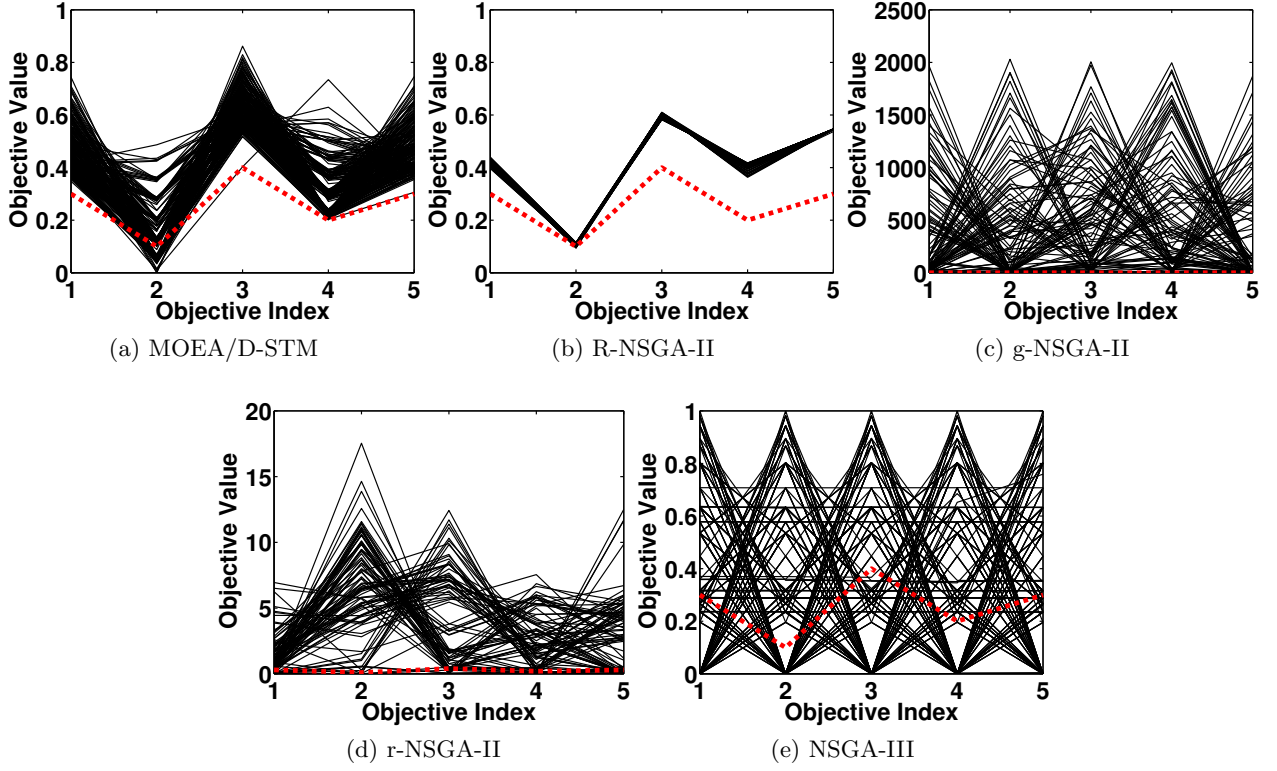


Figure 14: Comparisons on 5-objective DTLZ3 where $\mathbf{z}^r = (0.3, 0.1, 0.4, 0.2, 0.3)^T$.

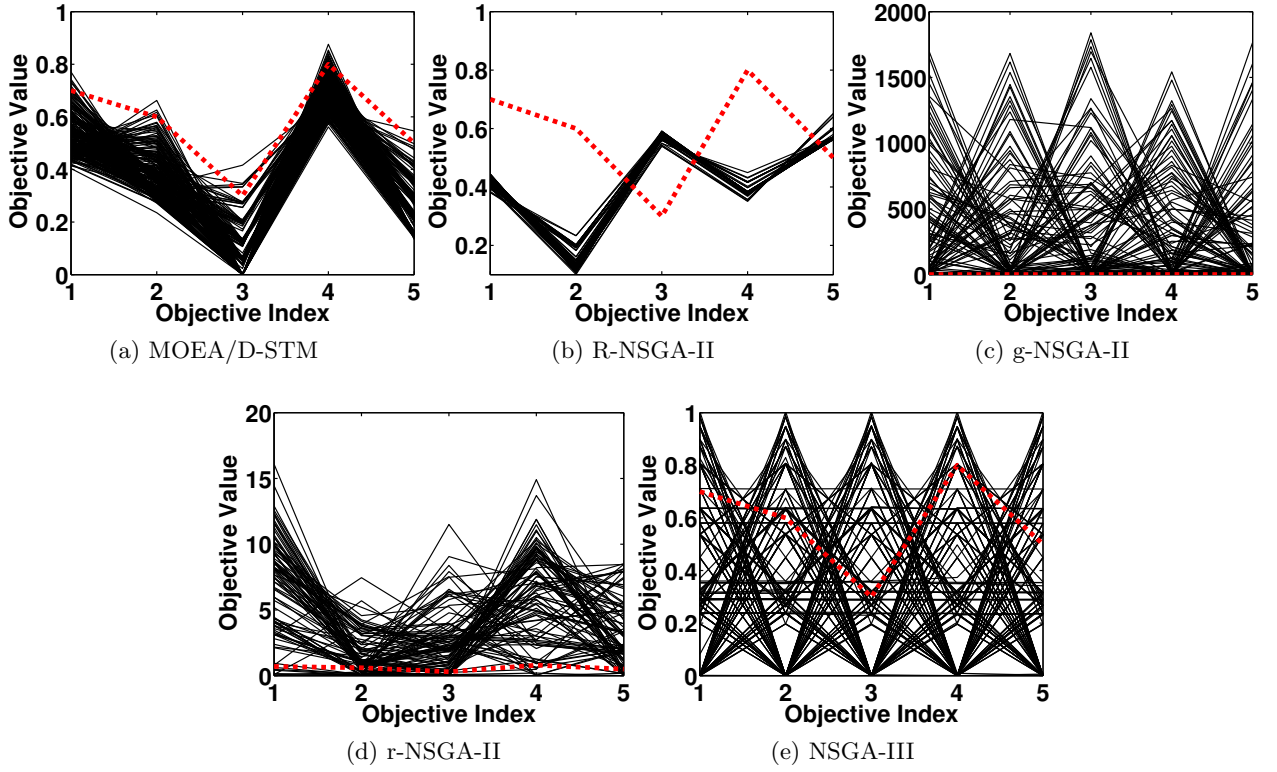


Figure 15: Comparisons on 5-objective DTLZ3 where $\mathbf{z}^r = (0.7, 0.6, 0.3, 0.8, 0.5)^T$.

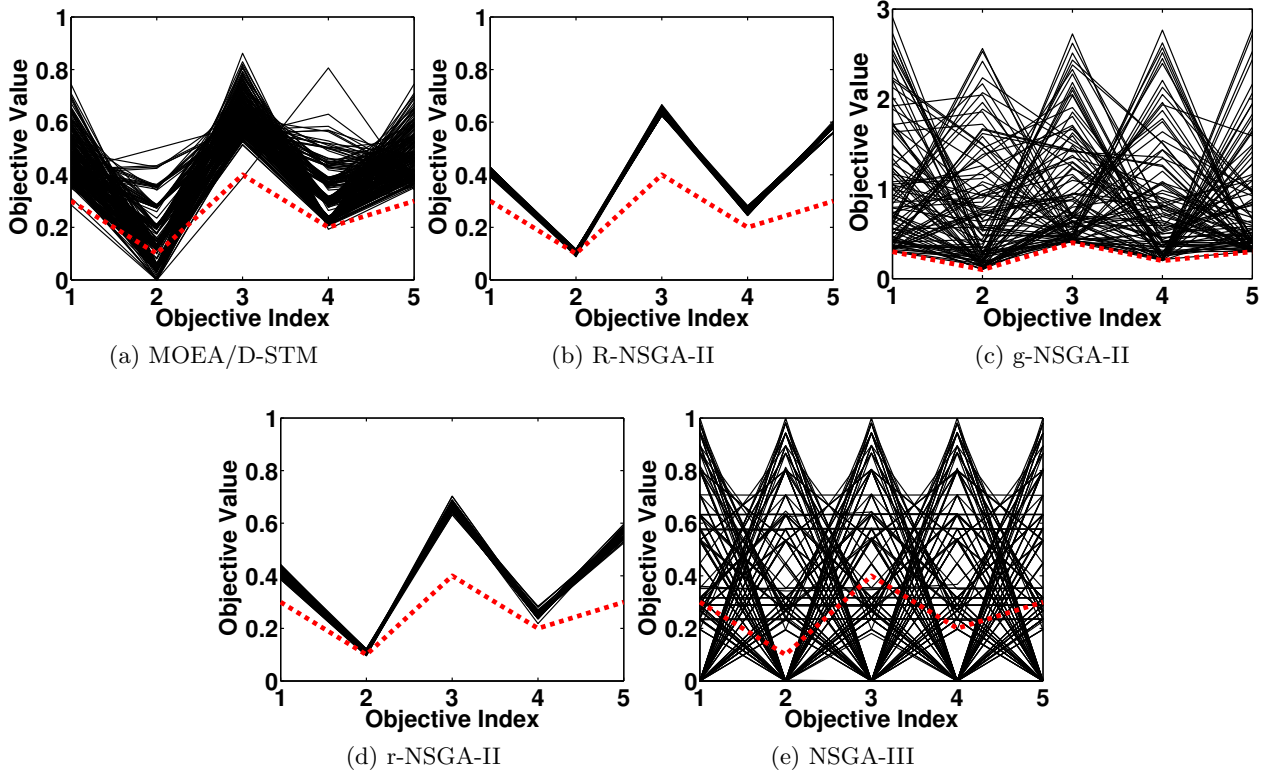


Figure 16: Comparisons on 5-objective DTLZ4 where $\mathbf{z}^r = (0.3, 0.1, 0.4, 0.2, 0.3)^T$.

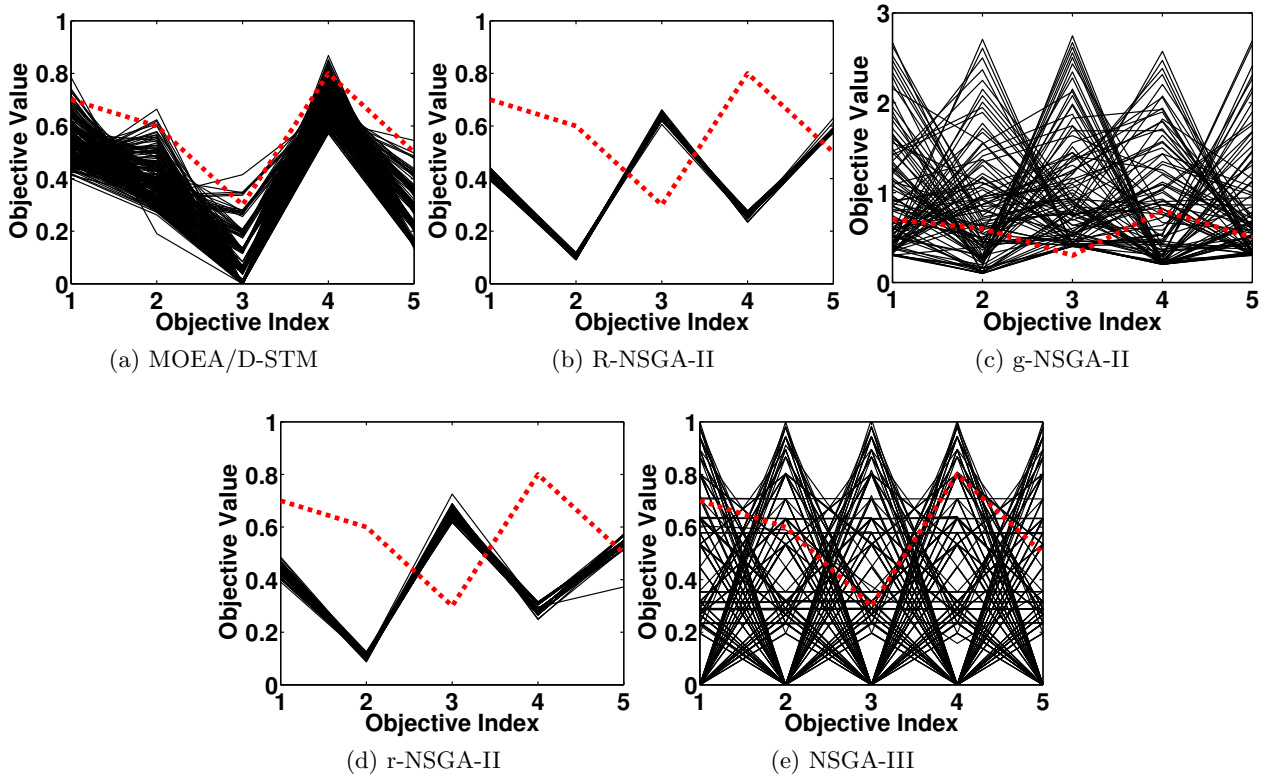


Figure 17: Comparisons on 5-objective DTLZ4 where $\mathbf{z}^r = (0.7, 0.6, 0.3, 0.8, 0.5)^T$.

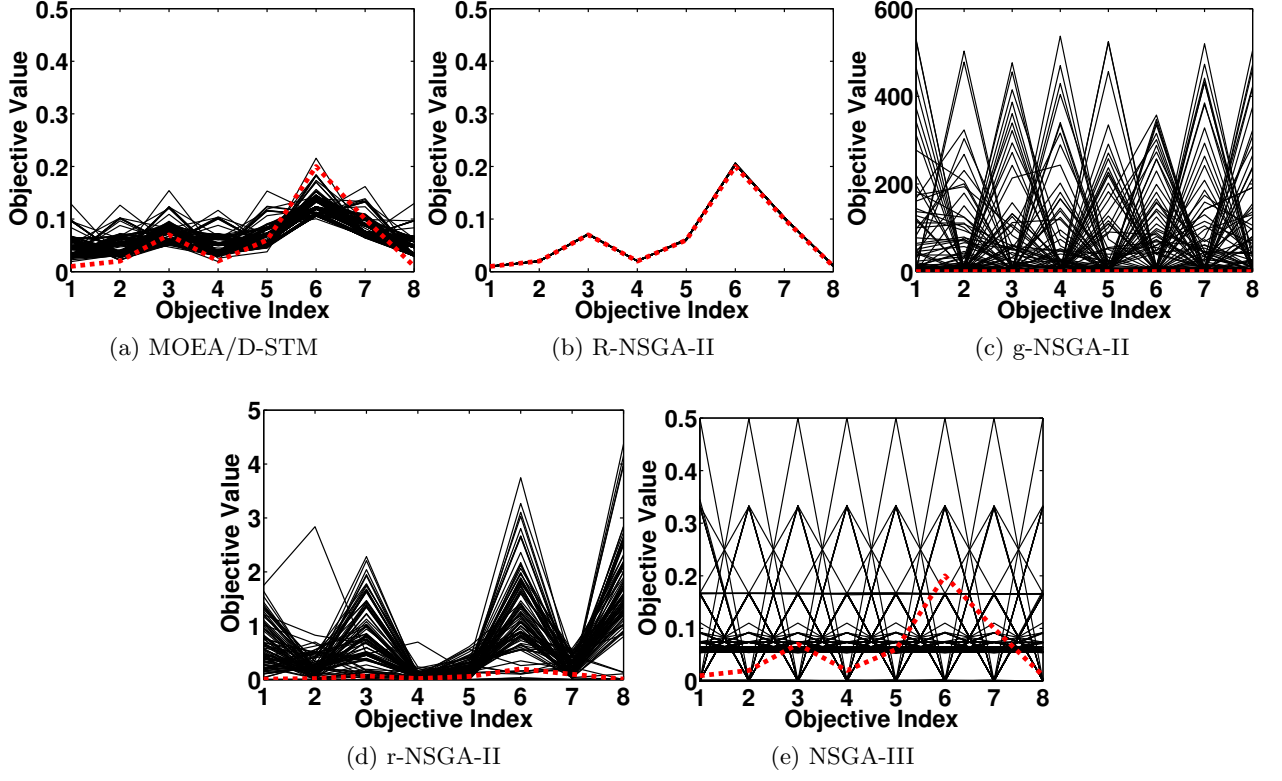


Figure 18: Comparisons on 8-objective DTLZ1 where $\mathbf{z}^r = (0.01, 0.02, 0.07, 0.02, 0.06, 0.2, 0.1, 0.01)^T$.

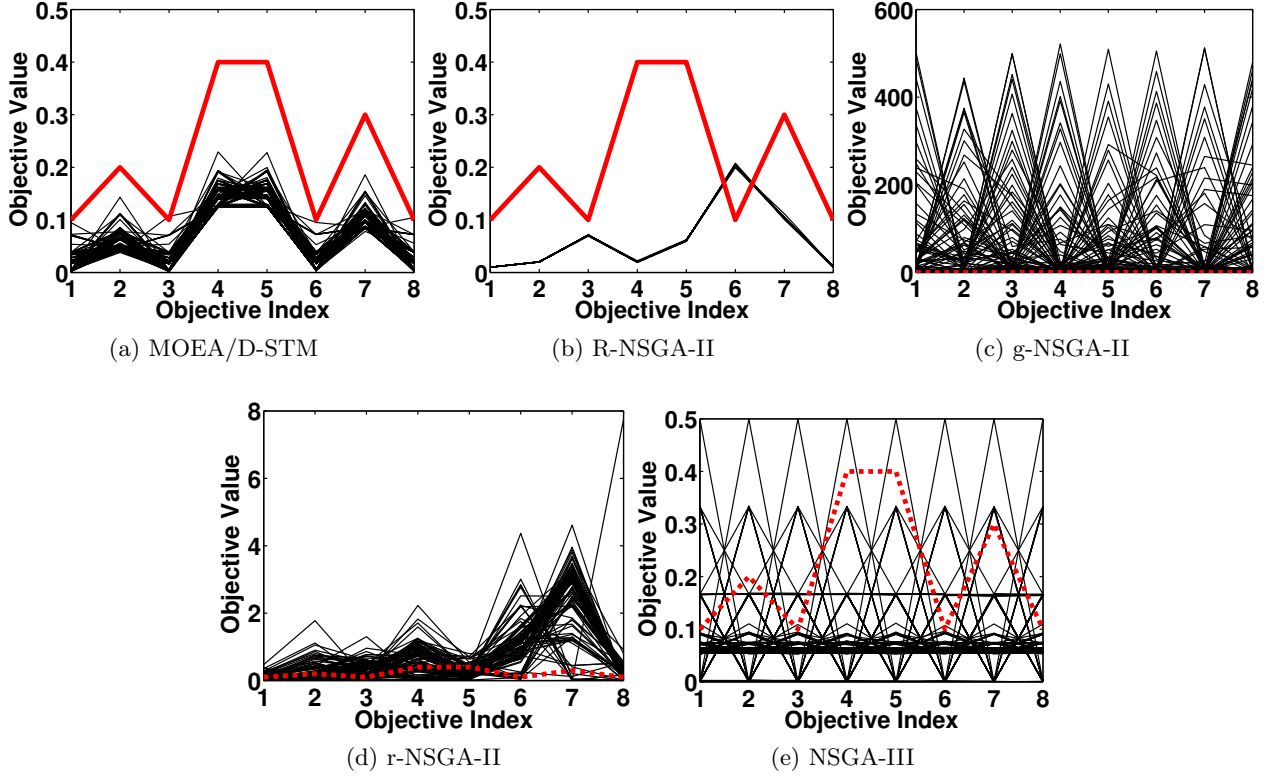


Figure 19: Comparisons on 8-objective DTLZ1 where $\mathbf{z}^r = (0.1, 0.2, 0.1, 0.4, 0.4, 0.1, 0.3, 0.1)^T$.

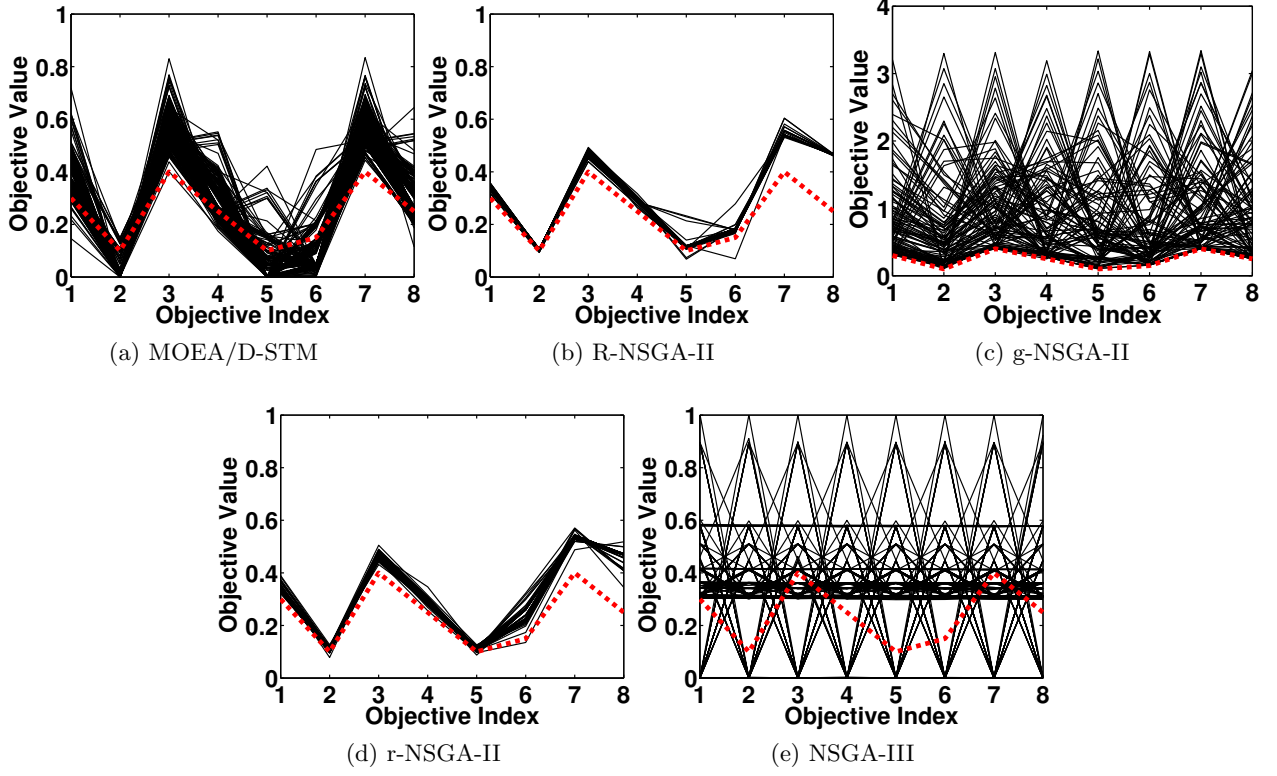


Figure 20: Comparisons on 8-objective DTLZ2 where $\mathbf{z}^r = (0.3, 0.1, 0.4, 0.25, 0.1, 0.15, 0.4, 0.25)^T$.

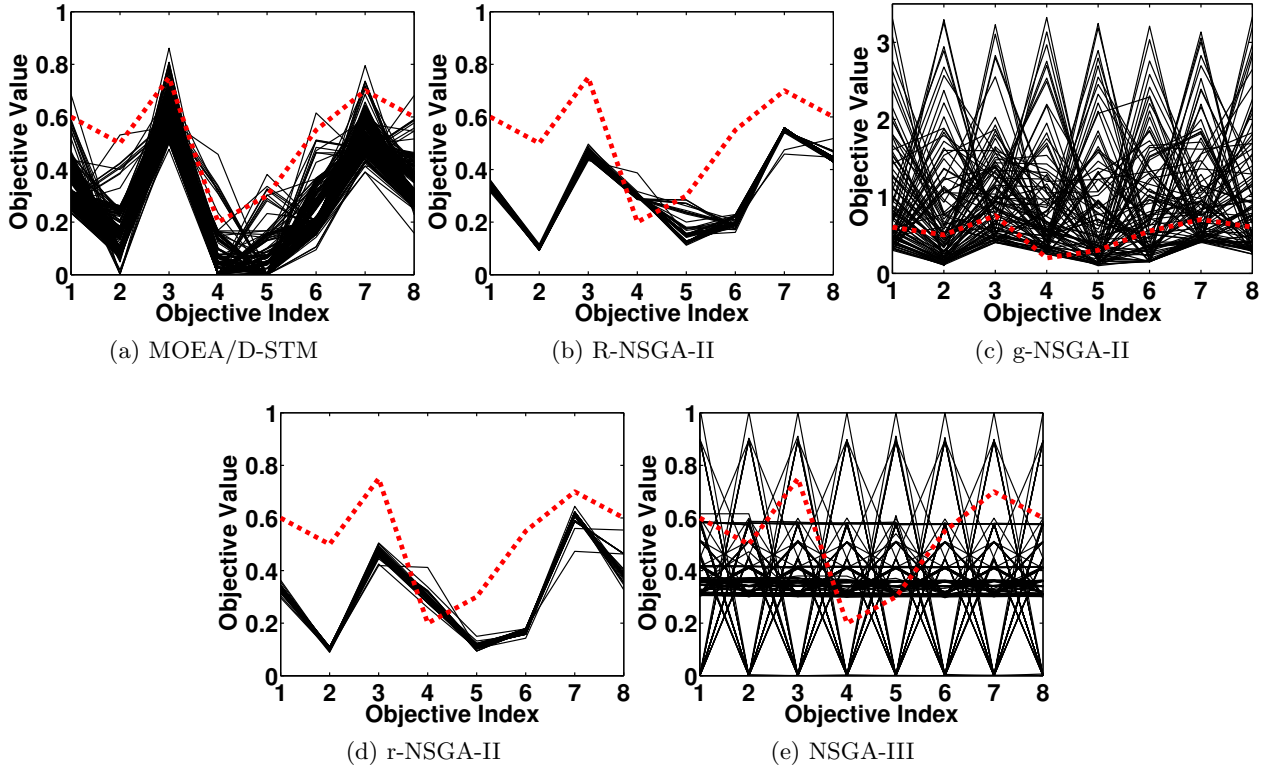


Figure 21: Comparisons on 8-objective DTLZ2 where $\mathbf{z}^r = (0.6, 0.5, 0.75, 0.2, 0.3, 0.55, 0.7, 0.6)^T$.

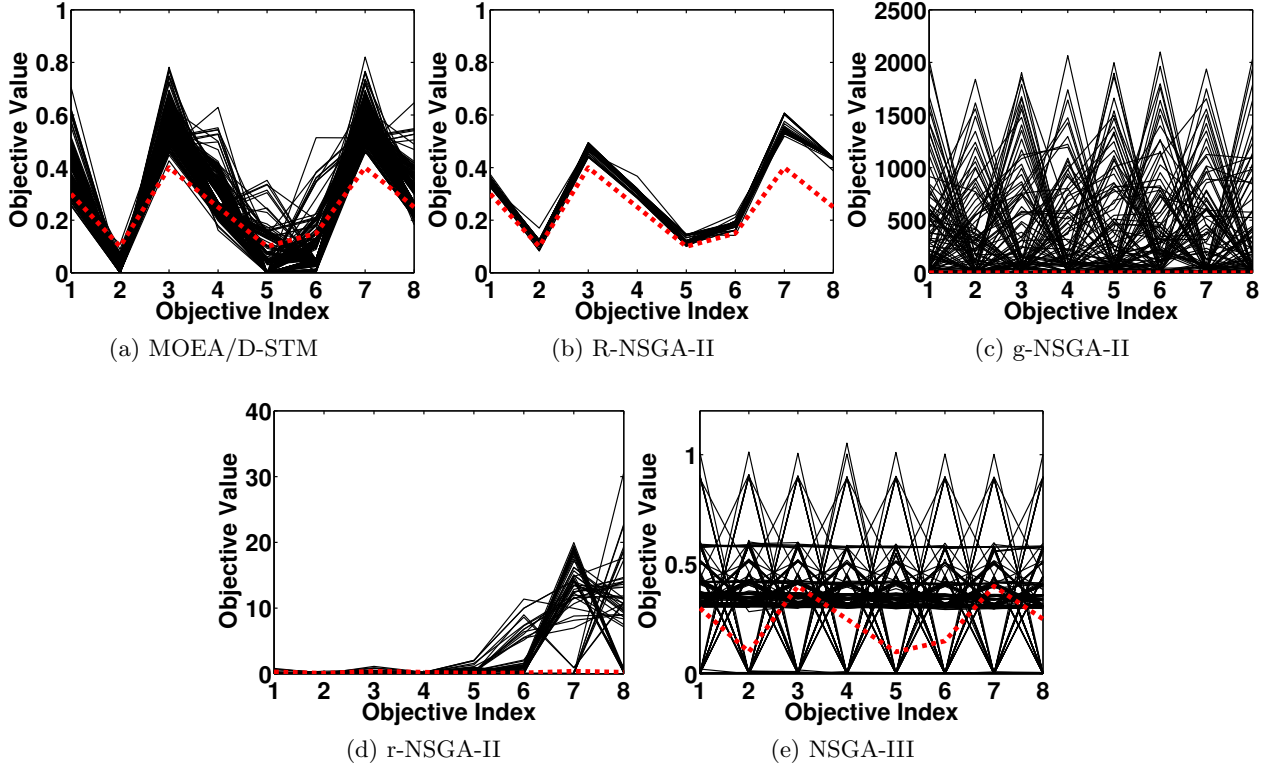


Figure 22: Comparisons on 8-objective DTLZ3 where $\mathbf{z}^r = (0.3, 0.1, 0.4, 0.25, 0.1, 0.15, 0.4, 0.25)^T$.

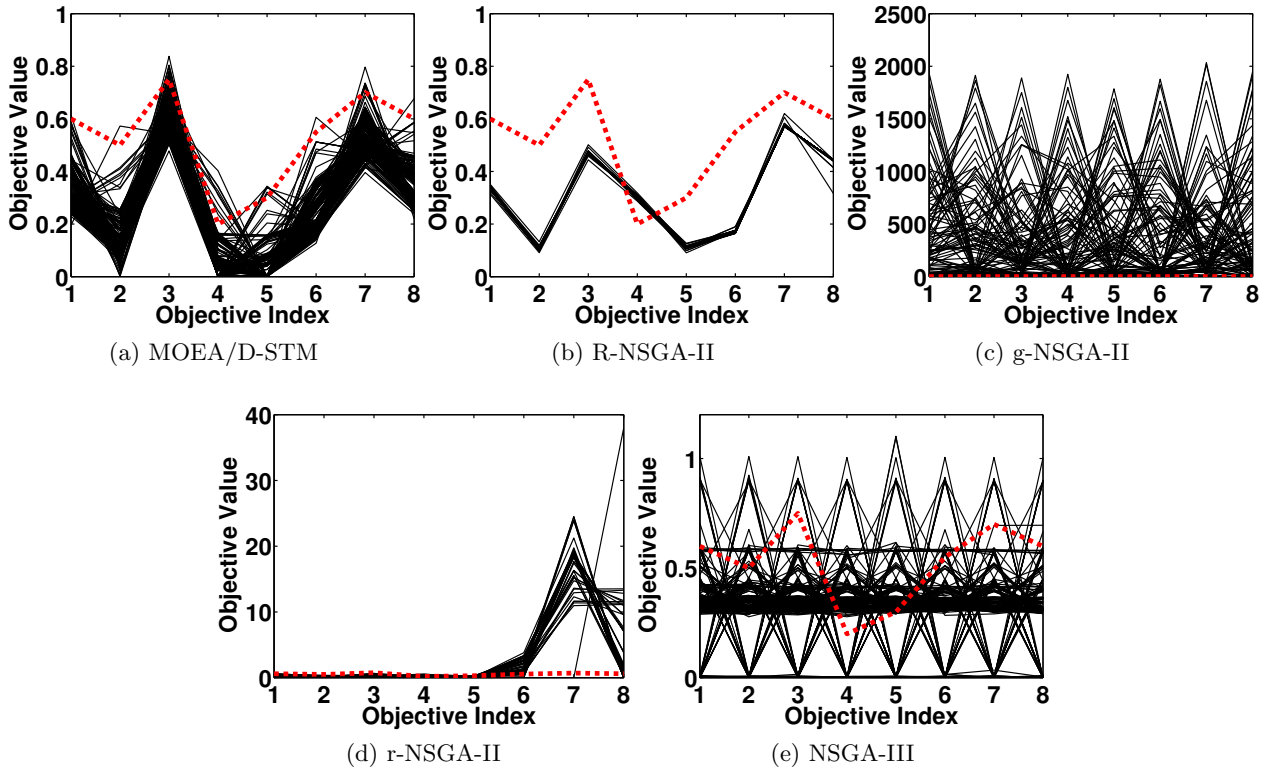


Figure 23: Comparisons on 8-objective DTLZ3 where $\mathbf{z}^r = (0.6, 0.5, 0.75, 0.2, 0.3, 0.55, 0.7, 0.6)^T$.

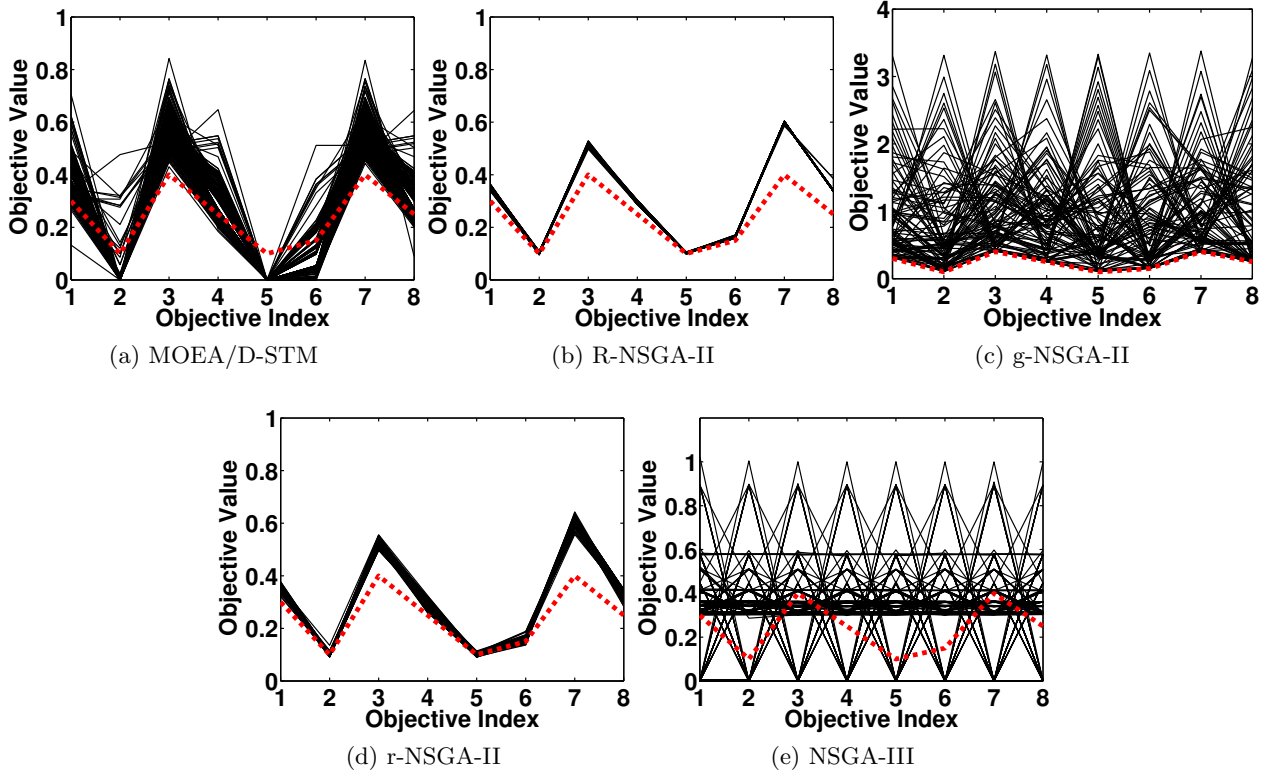


Figure 24: Comparisons on 8-objective DTLZ4 where $\mathbf{z}^r = (0.3, 0.1, 0.4, 0.25, 0.1, 0.15, 0.4, 0.25)^T$.

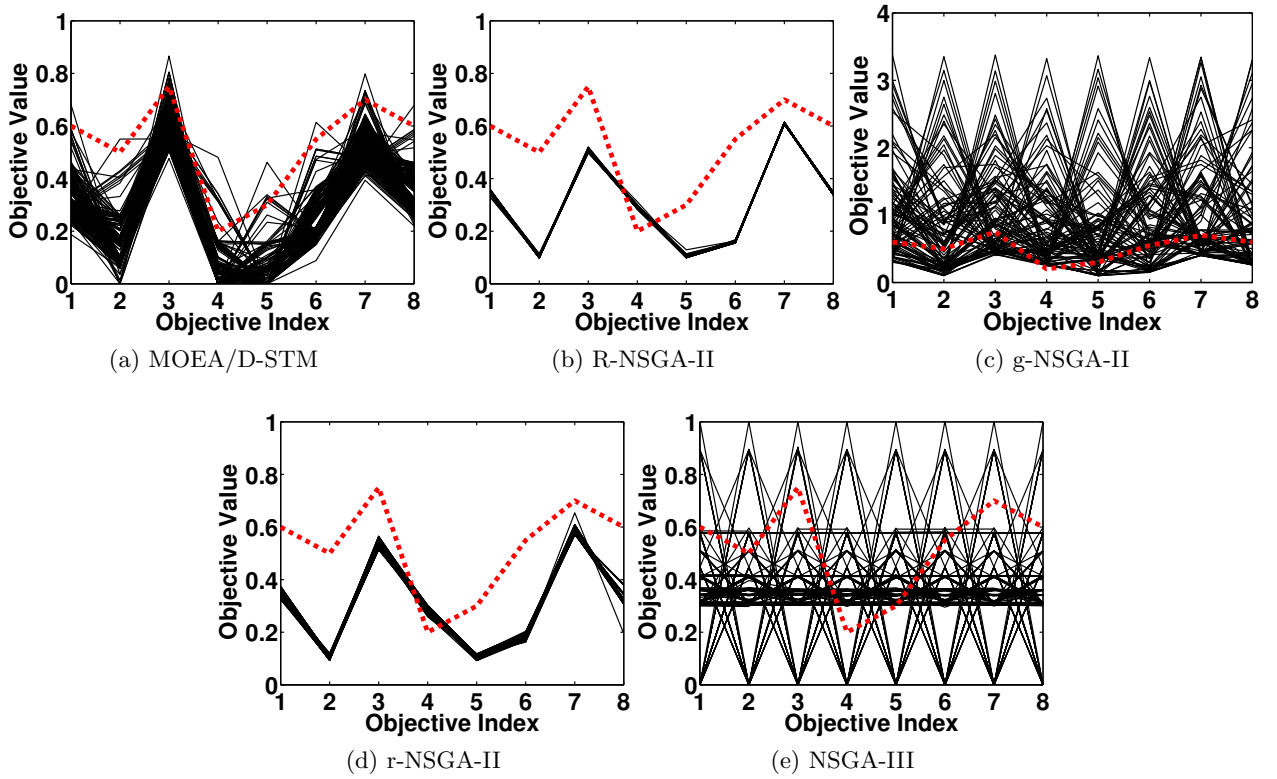


Figure 25: Comparisons on 8-objective DTLZ4 where $\mathbf{z}^r = (0.6, 0.5, 0.75, 0.2, 0.3, 0.55, 0.7, 0.6)^T$.

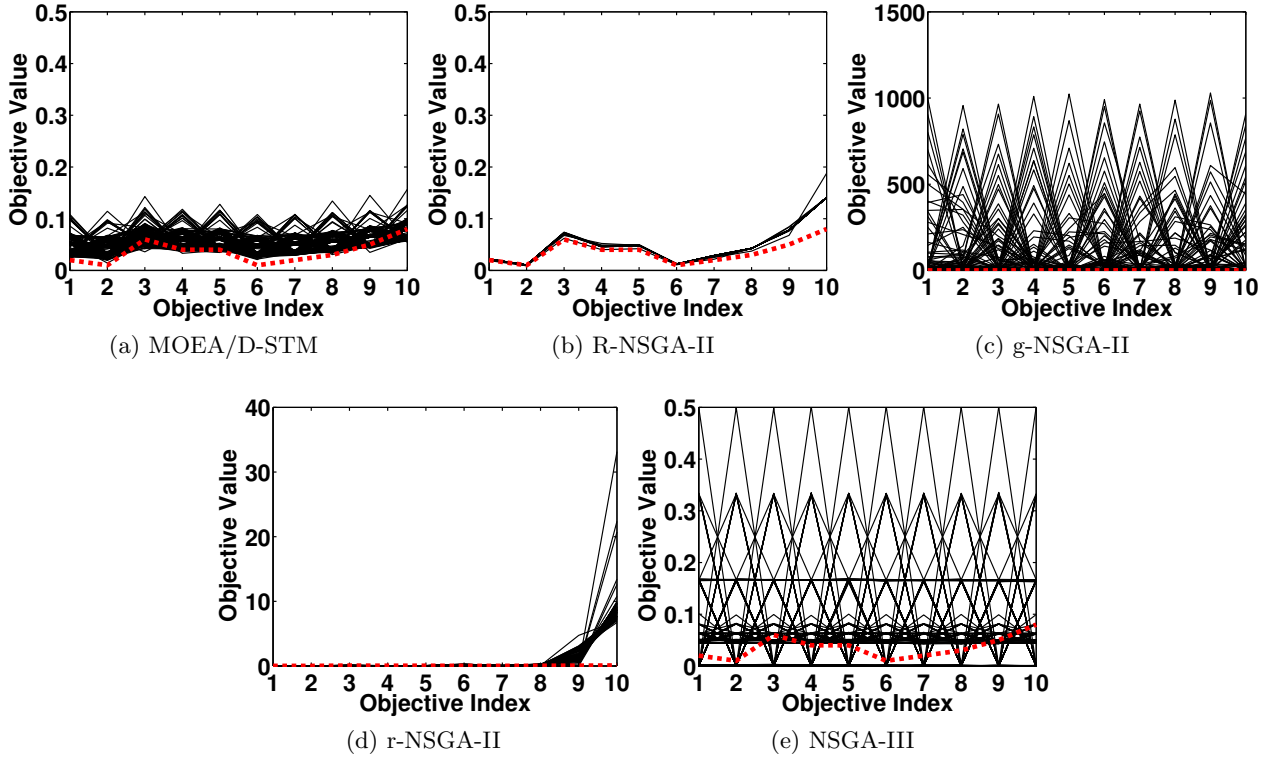


Figure 26: Comparisons on 10-objective DTLZ1 where $\mathbf{z}^r = (0.02, 0.01, 0.06, 0.04, 0.04, 0.01, 0.02, 0.03, 0.05, 0.08)^T$.

References

- [1] Y. Chen and X. Ye, “Projection onto a simplex,” *ArXiv e-prints*, Jan 2011.
- [2] I. Das and J. E. Dennis, “Normal-Boundary Intersection: A new method for generating the pareto surface in nonlinear multicriteria optimization problems,” *SIAM Journal on Optimization*, vol. 8, pp. 631–657, 1998.

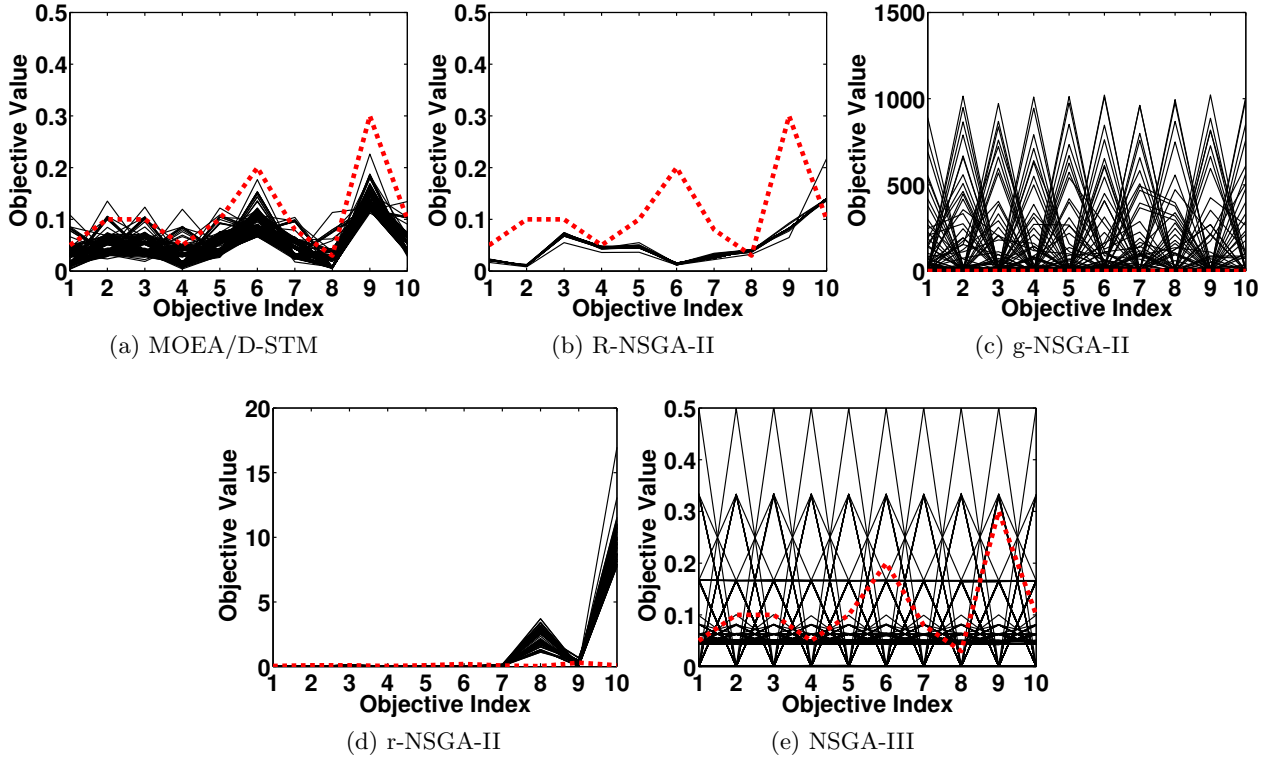


Figure 27: Comparisons on 10-objective DTLZ1 where $\mathbf{z}^r = (0.05, 0.1, 0.1, 0.05, 0.1, 0.2, 0.08, 0.03, 0.3, 0.1)^T$.

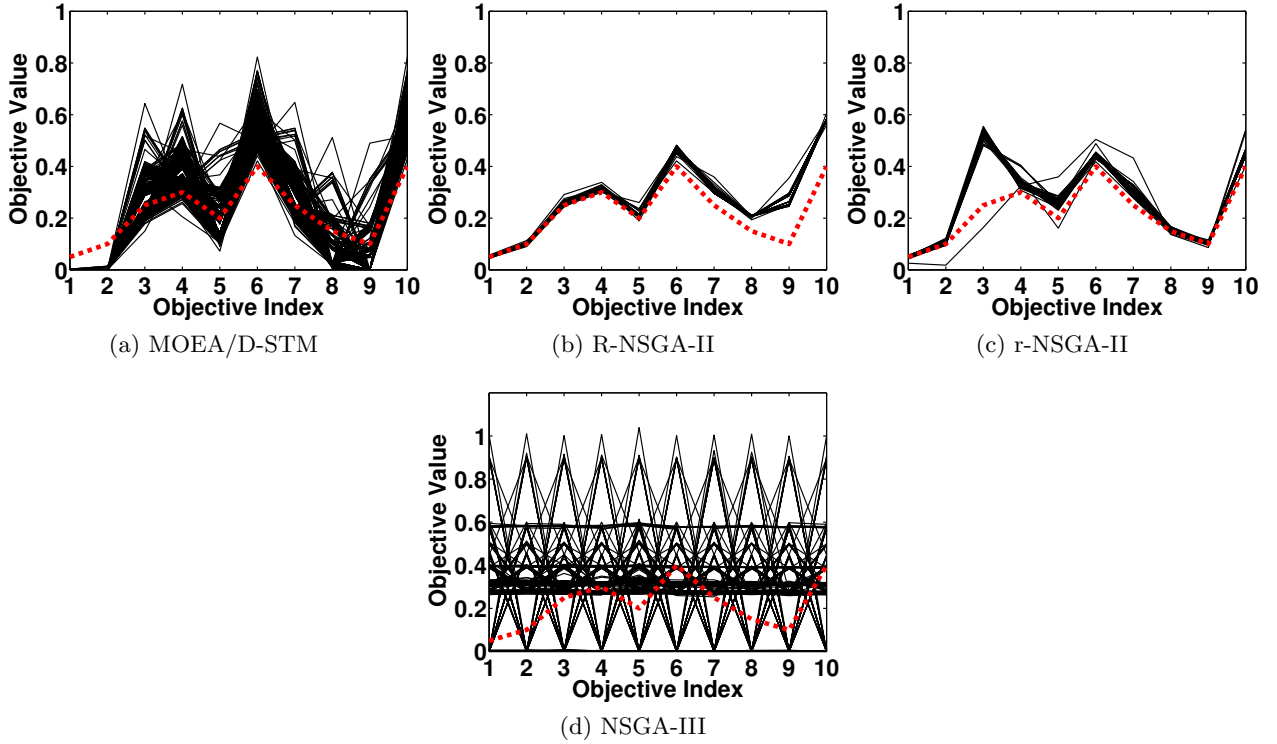


Figure 28: Comparisons on 10-objective DTLZ2 where $\mathbf{z}^r = (0.05, 0.1, 0.25, 0.3, 0.2, 0.4, 0.25, 0.15, 0.1, 0.4)^T$.

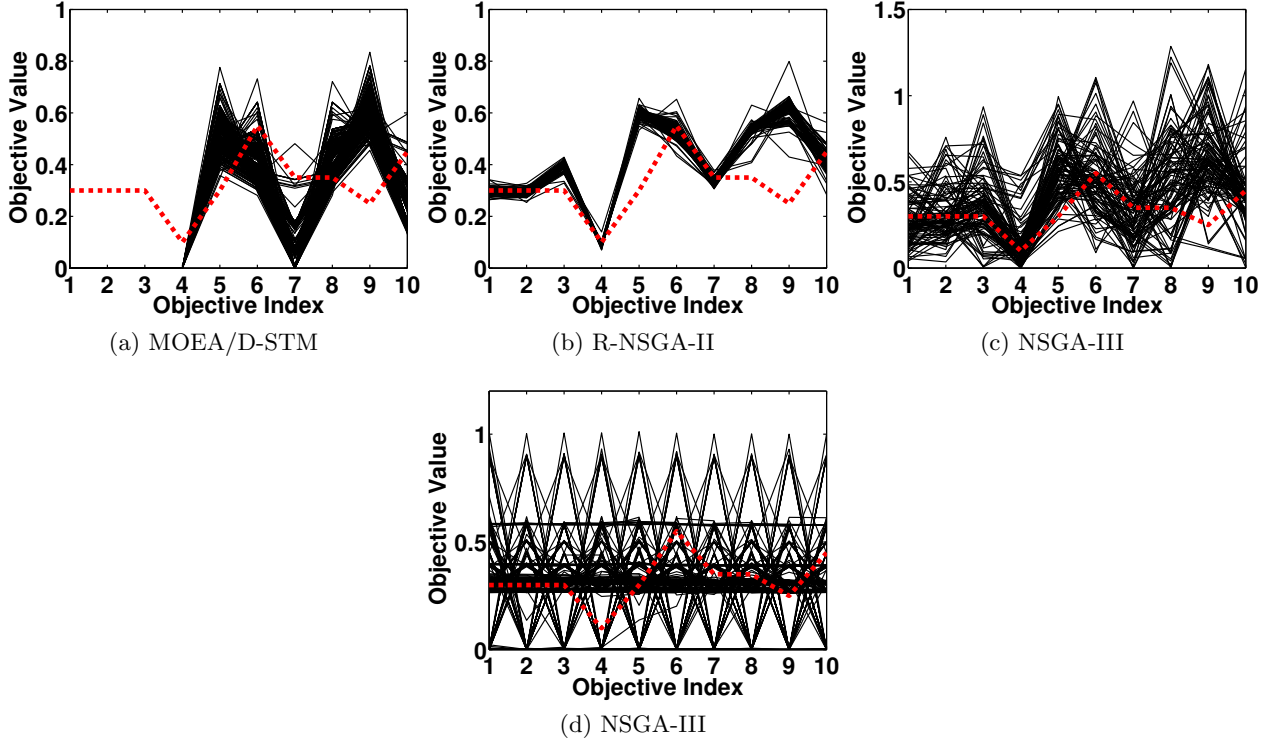


Figure 29: Comparisons on 10-objective DTLZ2 where $\mathbf{z}^r = (0.3, 0.3, 0.3, 0.1, 0.3, 0.55, 0.35, 0.35, 0.25, 0.45)^T$.

where $\mathbf{z}^r =$

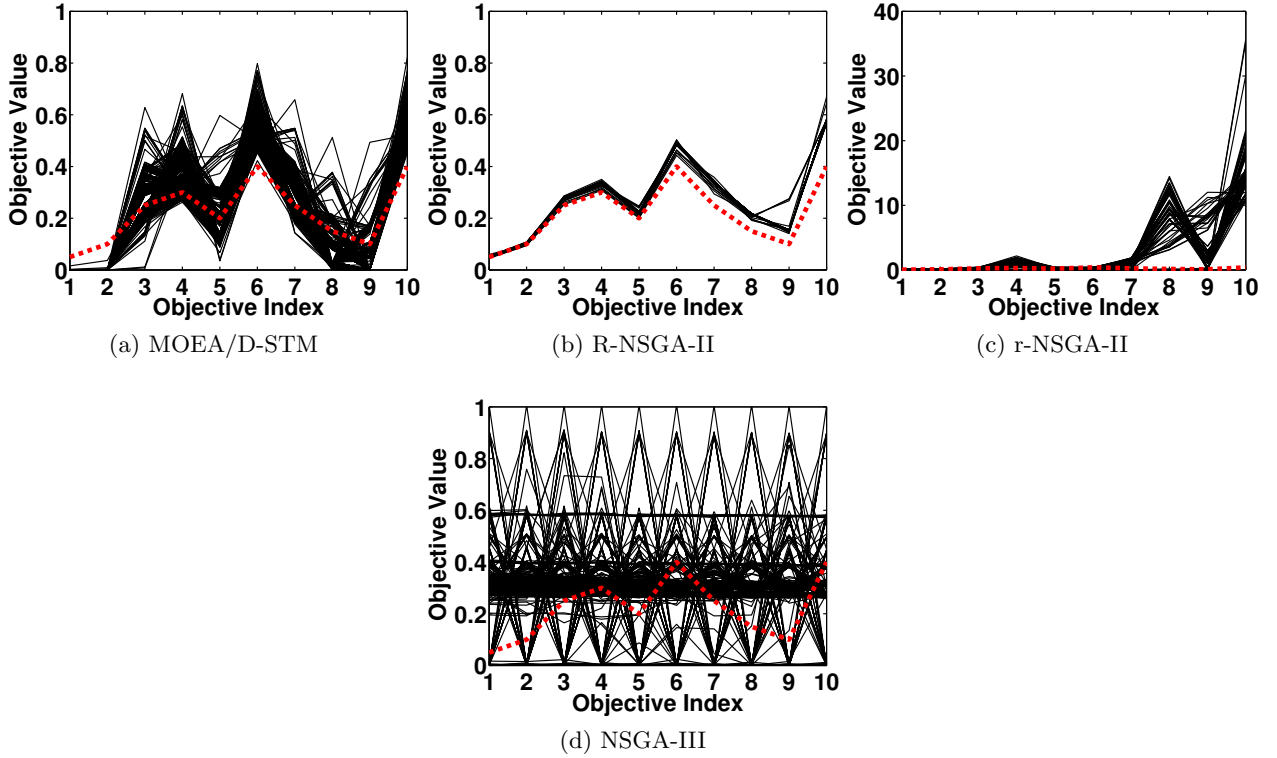


Figure 30: Comparisons on 10-objective DTLZ3 where $\mathbf{z}^r = (0.05, 0.1, 0.25, 0.3, 0.2, 0.4, 0.25, 0.15, 0.1, 0.4)^T$.

where $\mathbf{z}^r =$

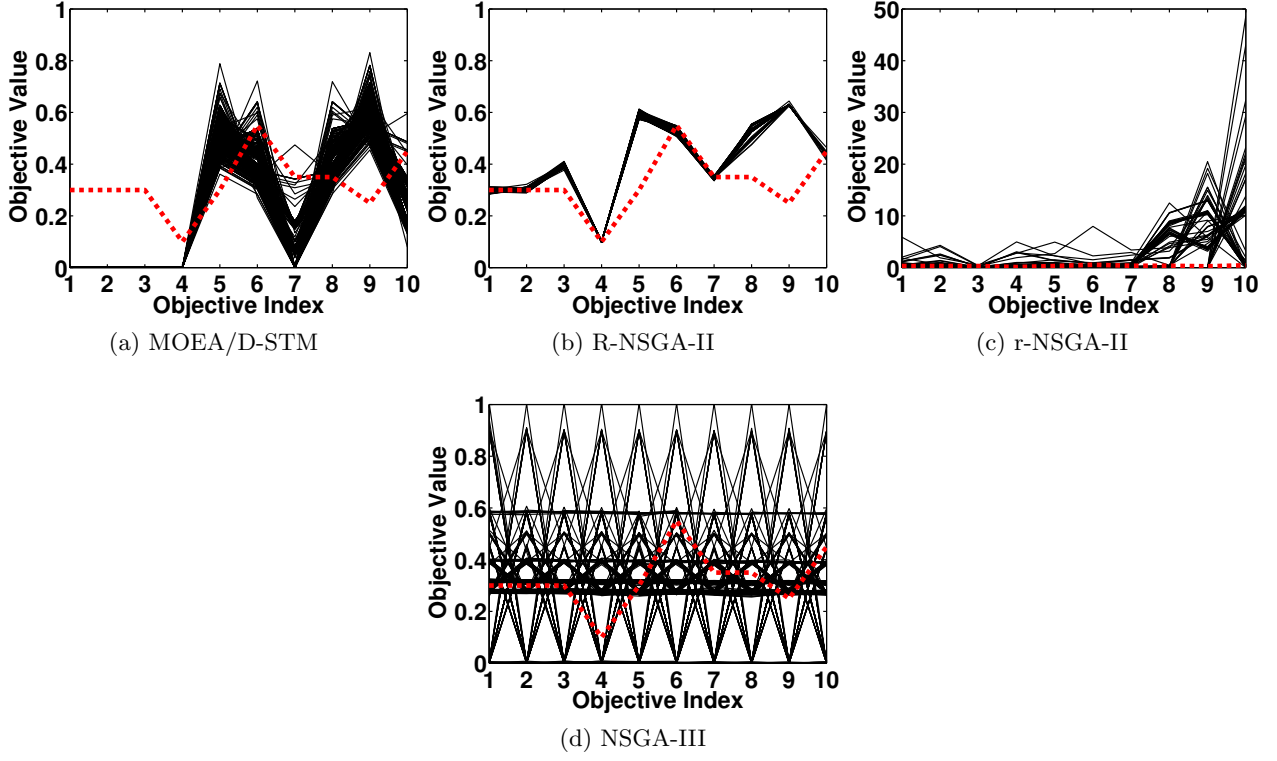


Figure 31: Comparisons on 10-objective DTLZ3 where $\mathbf{z}^r = (0.3, 0.3, 0.3, 0.1, 0.3, 0.55, 0.35, 0.35, 0.25, 0.45)^T$.

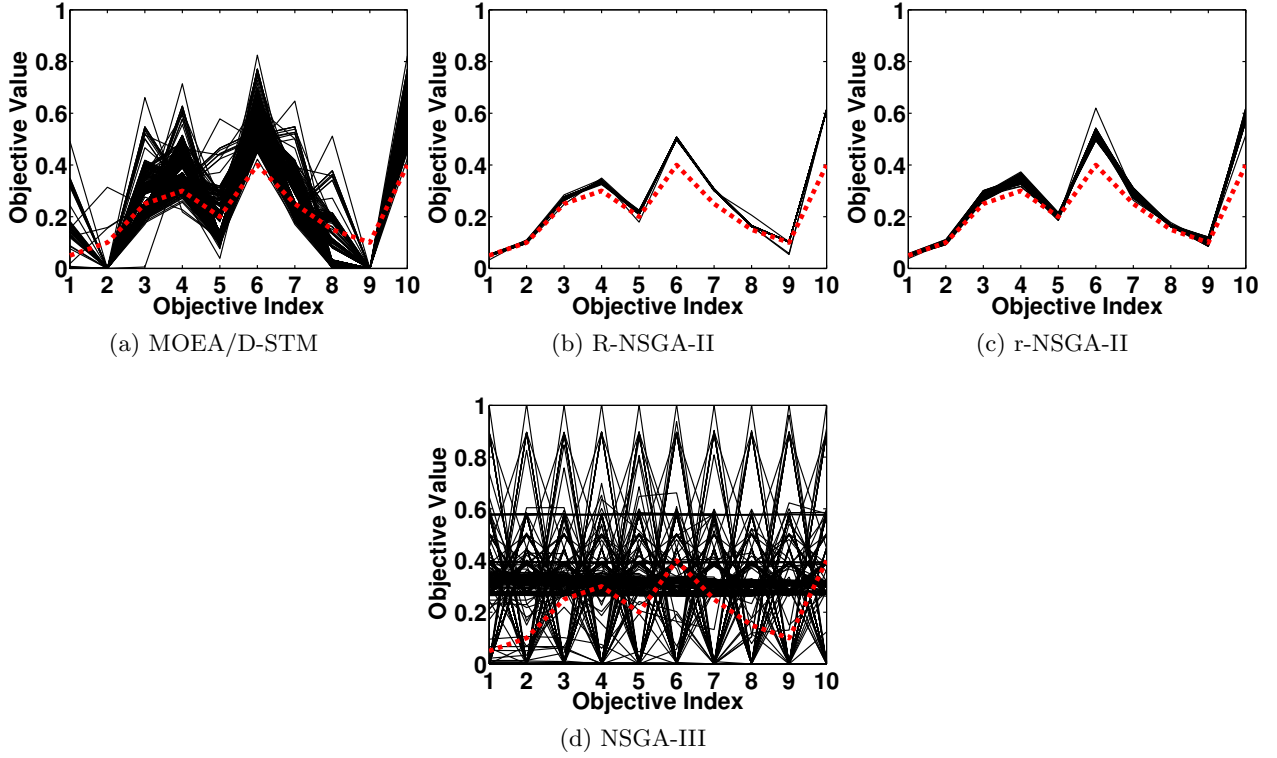


Figure 32: Comparisons on 10-objective DTLZ4 where $\mathbf{z}^r = (0.05, 0.1, 0.25, 0.3, 0.2, 0.4, 0.25, 0.15, 0.1, 0.4)^T$.

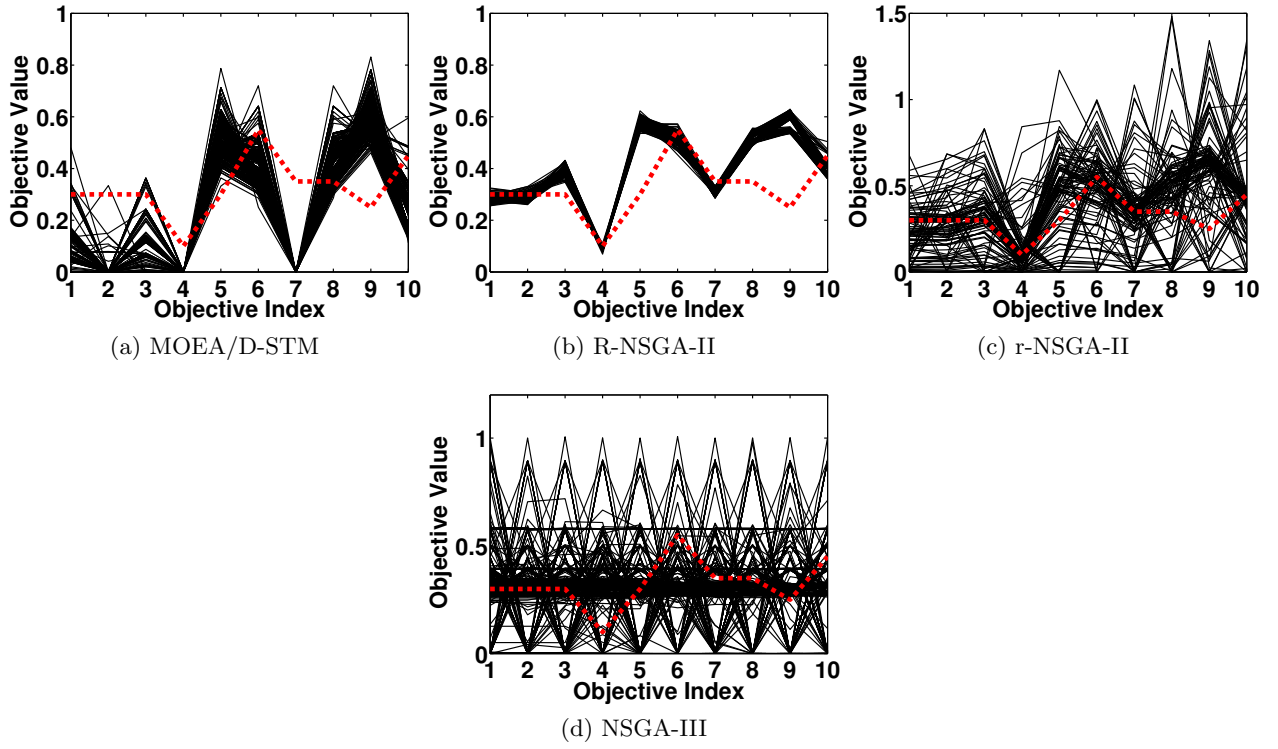


Figure 33: Comparisons on 10-objective DTLZ4 $(0.3, 0.3, 0.3, 0.1, 0.3, 0.55, 0.35, 0.35, 0.25, 0.45)^T$.

where $\mathbf{z}^r =$