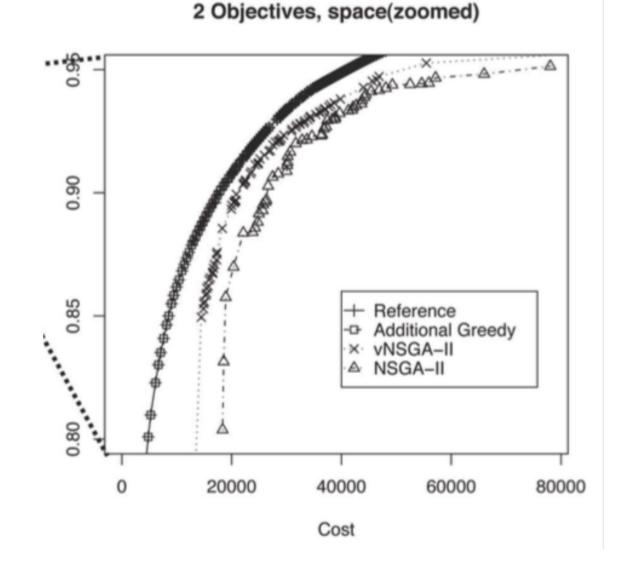
MOEA Evaluation Metrics

SEP592, Summer 2021 Shin Yoo

Comparing Pareto Fronts

- Comparison itself is not as straightforward as comparing two scalar values.
- There is no reference point, as the true Pareto front is usually not known.



Empirical Evaluation

- Empirical evaluation of different MOEAs becomes a meta-comparison: it is not only about the domain-specific quality, but it is also about the quality of the front itself.
- Properties that we want to evaluate:
 - Closeness to the true Pareto front
 - Diversity of the solutions on the Pareto front

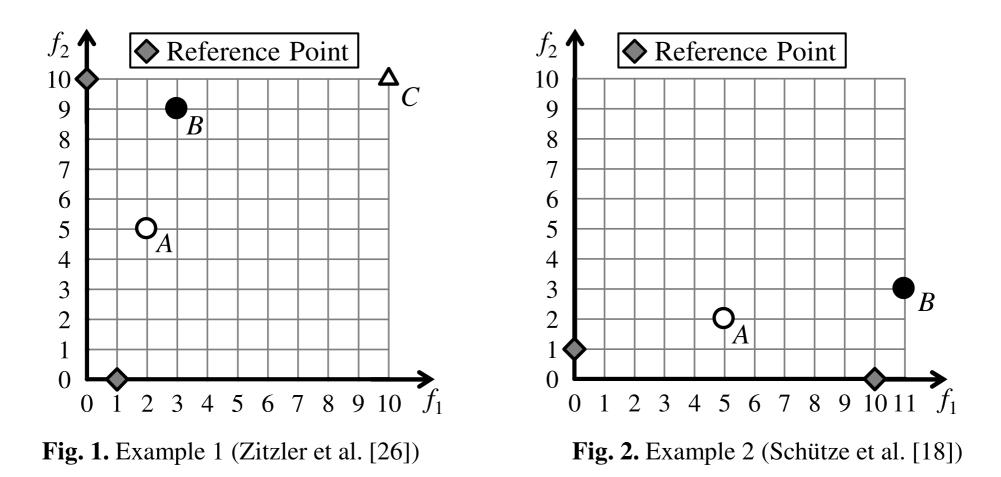
Closeness to true front

- There are cases where the true Front is known:
 - for example, benchmark optimisation problems.
- For cases where the true front is not known, we use what's called "reference front":
 - Collect all known solutions from all MOEAs involved.
 - Extract a single Pareto front from the collected solutions.
- Reference Pareto Front will include solutions contributed by different MOEAs.

Generational Distance (GD) and Inverted Generational Distance (IGD)

- GD: average of distance from each solution to its closest reference point.
- IGD: average of distance from each reference point to its closest solution
- The smaller, the better.

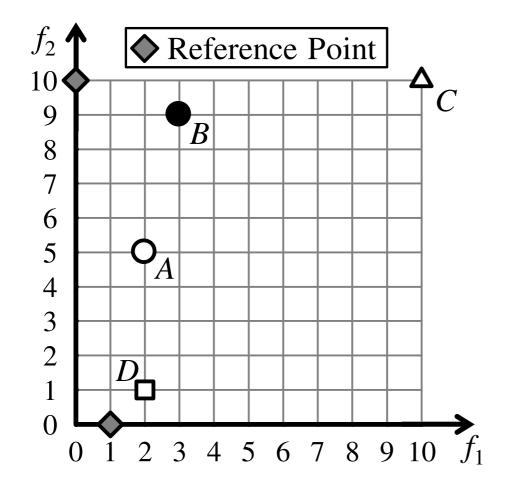
Weaknesses of GD



B has the shortest distance to its closest reference point, but arguably A is a better solution.

Modified Distance Calculation in Generational Distance and Inverted Generational Distance, Hisao Ishibuchi, Hiroyuki Masuda, Yuki Tanigaki, Yusuke Nojima, EMO 2015.

Weaknesses of GD



IGD is more sensitive to gaps.

$$IGD(A) = \frac{1}{2} \left(\sqrt{(2-0)^2 + (5-10)^2} + \sqrt{(2-1)^2 + (5-0)^2} \right) = 5.24,$$

$$IGD(D) = \frac{1}{2} \left(\sqrt{(2-0)^2 + (1-10)^2} + \sqrt{(2-1)^2 + (1-0)^2} \right) = 5.32.$$

Fig. 3. Example 3 with a new solution set D

Modified Distance Calculation in Generational Distance and Inverted Generational Distance, Hisao Ishibuchi, Hiroyuki Masuda, Yuki Tanigaki, Yusuke Nojima, EMO 2015.

Weaknesses of GD/IGD

- When the shape of the true Front is known as a continuous function: reference Pareto front, i.e. a set of points, is sampled from the function.
- How you sample can affect distances
 - For 11 reference points, IGD(Z, A) < IGD(Z, B)
 - For 21 reference points, IGD(Z, A) > IGD(Z, B)

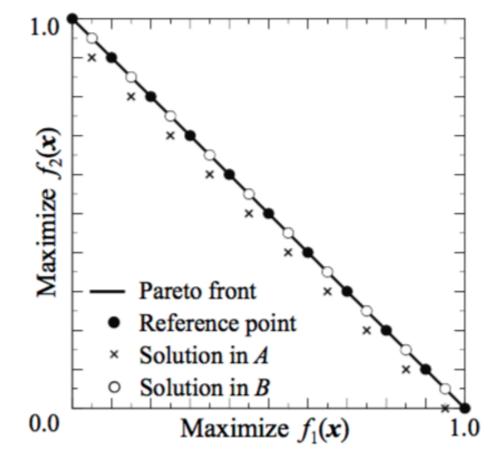


Fig. 3. Reference points (H = 10) and two solution sets A and B.

H. Ishibuchi, H. Masuda, Y. Tanigaki, and Y. Nojima. Difficulties in specifying reference points to calculate the inverted generational distance for many-objective optimization problems. In Computational Intelligence in Multi-Criteria Decision-Making (MCDM), 2014 IEEE Symposium on, pages 170–177, Dec 2014.

Weaknesses of IGD

- When using collected reference front, different MOEAs will contribute solutions with different characteristics:
 - some may show strong convergence
 - others may show greater diversity
- Again, this may results in sampling bias when evaluating an MOEA.

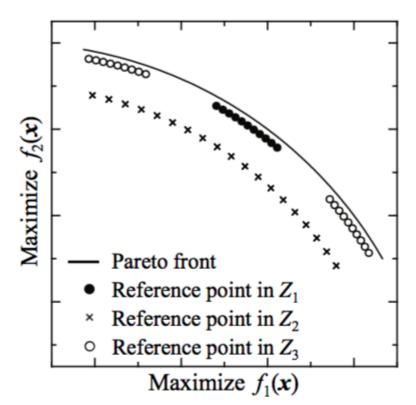


Fig. 7. Three typical situations of reference points: Concentration on the center of the Pareto front (Z_1) , uniform distribution over the entire Pareto front (Z_2) , and emphasis on the edges of the Pareto front (Z_3) .

H. Ishibuchi, H. Masuda, Y. Tanigaki, and Y. Nojima. Difficulties in specifying reference points to calculate the inverted generational distance for many-objective optimization problems. In Computational Intelligence in Multi-Criteria Decision-Making (MCDM), 2014 IEEE Symposium on, pages 170–177, Dec 2014.

Epsilon

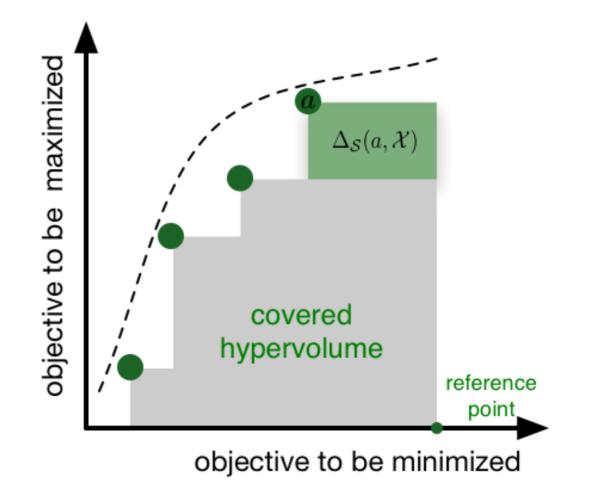
 Binary indicator, I_ε(A, B): intuitively, the amount of "shift" required to change B so that it is weakly dominated by A (i.e. A is not worse than B in all objectives).

$$I_{\epsilon}(A,B) = \inf_{\epsilon \in \mathbb{R}} \{ \forall z^2 \in B \exists z^1 \in A : z^1 \preceq_{\epsilon} z^2 \}$$

- $\boldsymbol{z}^1 \preceq_{\boldsymbol{\epsilon}} \boldsymbol{z}^2 \iff \forall i \in 1..n \, : \, \boldsymbol{z}^1_i \leq \boldsymbol{\epsilon} \cdot \boldsymbol{z}^2_i$
- $z^{1} \leq_{\epsilon+} z^{2} \iff \forall i \in 1..n : z^{1}_{i} \leq \epsilon + z^{2}_{i}.$ $I^{1}_{\epsilon}(A) = I_{\epsilon}(A, R).$

Hypervolume

- Intuitively, hypervolume measures the area (space) dominated by a given Pareto front.
- An unary indicator: does not need a reference front.
- Can be thought to measure both convergence and diversity.



Scaling and Normalisation

- The concept of Pareto optimality itself is independent from scale and normalisation: it is strictly based on partial order only.
- For quality indicators, normalisation may be necessary:
 - so that multiple objectives contribute equally to the indicators.

Simple Linear Scaling

- ... may be applicable, if the bounds are known.
- If not, there are other normalisation methods.
- ... which leads into the next topic :)

$$z'_i = \frac{z_i - z_i^{min}}{z_i^{max} - z_i^{min}}$$