

DM - Decision Maker

PO - Pareto Optimal

Method  
Assignment  
(A)

Assignment  
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Assignment  
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Apriori <u>Prerequisite</u>	Interactive	Posteriori
Requires well-known knowledge about the value function prior to problem solution.	Does not require any prior knowledge about any value function.	Requires Mathematical programming methods & EA knowledge.
<u>Role of DM</u> : DM is actively involved & must indicate the assumption before the optimization process.	DM is involved in the search process & decides upon which parts of PO front are considered with the optimization program.	DM will be provided a set of PO solutions & the most suitable one will be selected which provides good diversity & good convergence.
<u>Methods</u> - Weighting method - E constraint method - Lexicographic ordering	- Pairwise Comparison - Reference point method - Classification of objectives	- EMO - NSGA-II

## Weighting Method. (Working Principle) (5.1)

- Uses scalarization function:

- Converts a MOP to SOP.

i.e; Minimize  $(f_1(\vec{x}), f_2(\vec{x}), \dots, f_m(\vec{x}))$

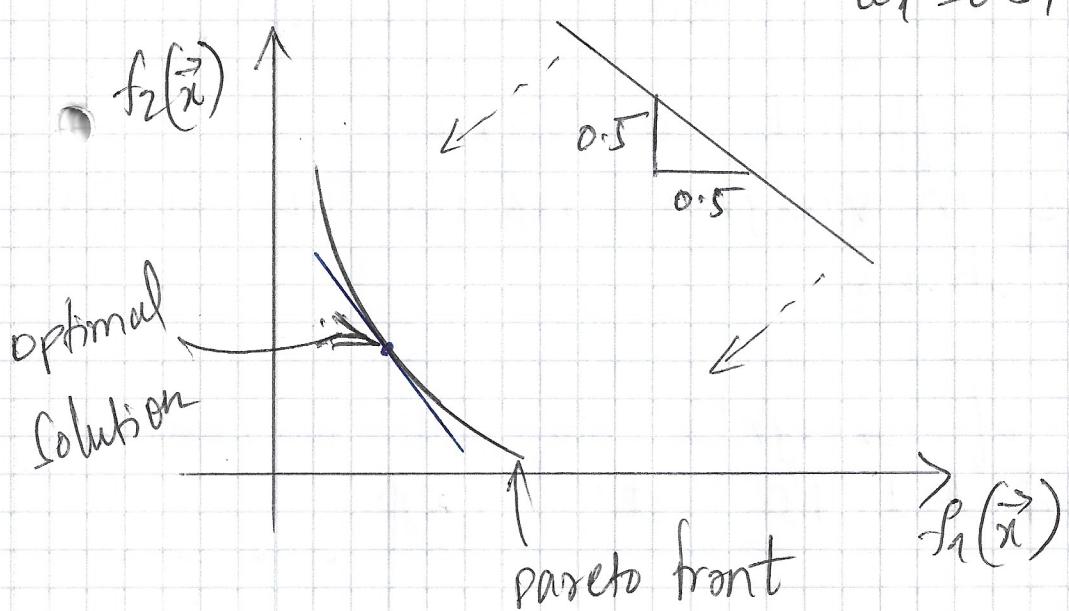
to

$$\text{Minimize } (w_1 f_1(\vec{x}) + w_2 f_2(\vec{x}) + \dots + w_m f_m(\vec{x}))$$

where  $\sum_{i=1}^m w_i = 1$

- It finds one optimal solution out of many.
- uses weights indicating the slope (position of solution).

$$w_1 = 0.5, w_2 = 0.5$$



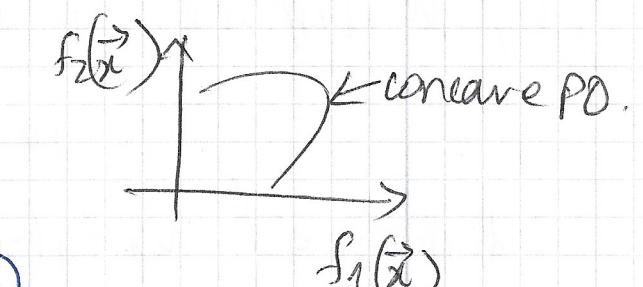
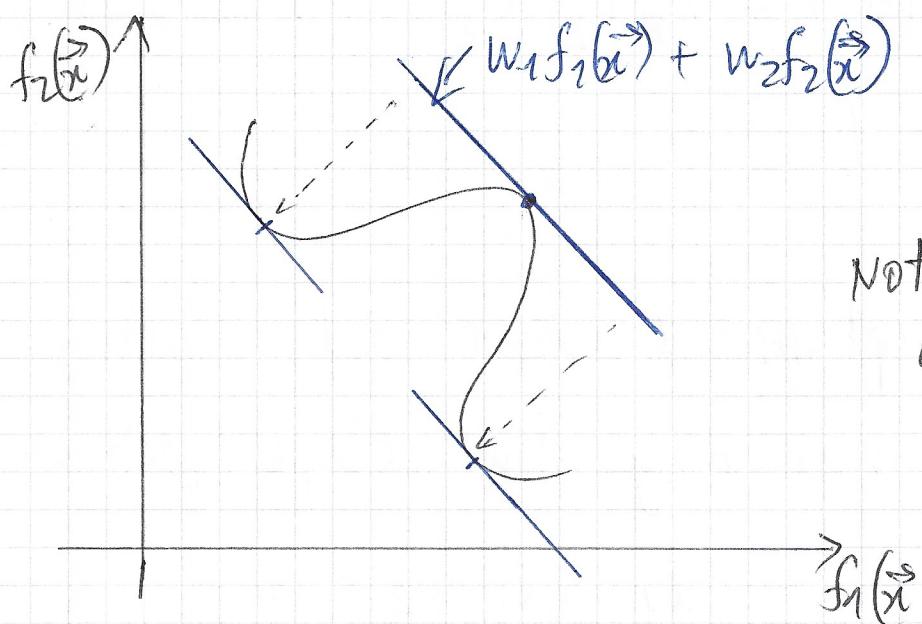
Advantages: For convex problems,  $f_2$

one optimal solution is guaranteed.

$f_1$

Drawbacks:

① Non convex problems-

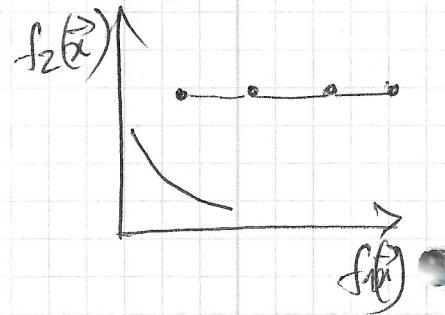


not able to provide  
optimal solution.

② Weights are difficult to learn.

Small change in weights results in choosing  
different solutions.

③ Distribution of weights do not guarantee  
evenly distributed PO set.



④ Preference of weights

(Some preferences won't be  
considered as strongly as DM preferences)

ex: Selecting a location for long term stay.

Location	Rent (e)	Uni <sub>dist</sub> (km)	Sup <sub>dist</sub> (km)	Result
A	230	5	5	2.8
B	185	6	7	2.2
C	190	7	7	2.7
wts.	0.8	0.1	0.1	

Solution set: { a (2,6), b (2,4), c (3.5,3.5), d (4,2), e (8,1) }.

Case #1.  $(w_1, w_2) := (0.5, 0.5)$   $\sum_{i=1}^2 w_i = 1$

$w_1 f_1(a) + w_2 f_2(a)$	$0.5 \times 2 + 0.5 \times 6$	4
$w_1 f_1(b) + w_2 f_2(b)$	$0.5 \times 2 + 0.5 \times 4$	3
$w_1 f_1(c) + w_2 f_2(c)$	$0.5 \times 3.5 + 0.5 \times 3.5$	3.5
$w_1 f_1(d) + w_2 f_2(d)$	$0.5 \times 4 + 0.5 \times 2$	3
$w_1 f_1(e) + w_2 f_2(e)$	$0.5 \times 8 + 0.5 \times 1$	4.5

$\min(4, 3, 3.5, 3, 4.5)$  two optimal solutions  
b & d.

Case #2.  $(w_1, w_2) := (0.1, 0.9)$   $\sum_{i=1}^2 w_i = 1$

$w_1 f_1(a) + w_2 f_2(a)$	$0.1 \times 2 + 0.9 \times 6$	5.6
$w_1 f_1(b) + w_2 f_2(b)$	$0.1 \times 2 + 0.9 \times 4$	3.8
$w_1 f_1(c) + w_2 f_2(c)$	$0.1 \times 3.5 + 0.9 \times 3.5$	3.5
$w_1 f_1(d) + w_2 f_2(d)$	$0.1 \times 4 + 0.9 \times 2$	2.2
$w_1 f_1(e) + w_2 f_2(e)$	$0.1 \times 8 + 0.9 \times 1$	1.7

$\min(5.6, 3.8, 3.5, 2.2, 1.7) \Rightarrow$  one optimal solution 'e'.

## 5.2. $\epsilon$ -Constraint Method (working principle)

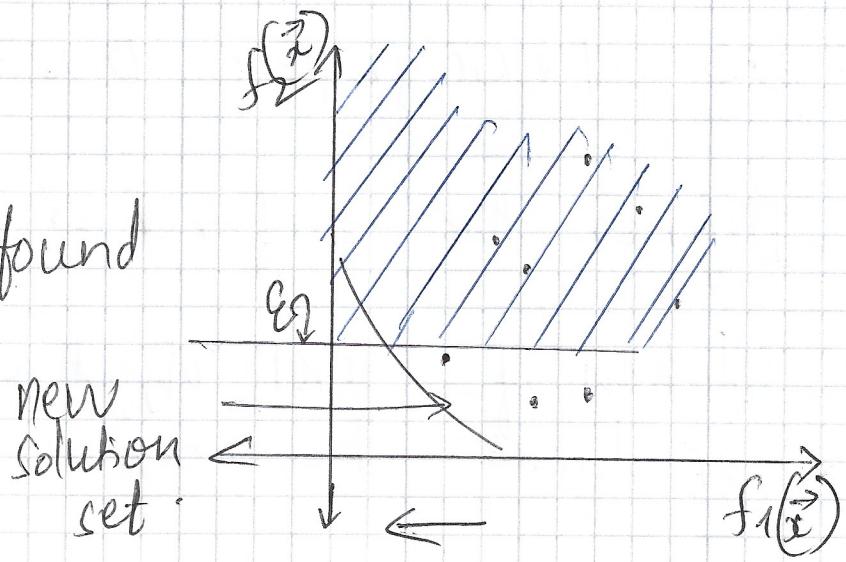
Convert a MOP to SOP by keeping one objective as optimization function & rest as constraints.

MOP:  $\min(f_1(\vec{x}), f_2(\vec{x}), \dots, f_m(\vec{x}))$  s.t.  $\vec{x} \in S$

$$\min(f_1(\vec{x})) \quad | \quad f_2(\vec{x}) \leq \epsilon_2$$

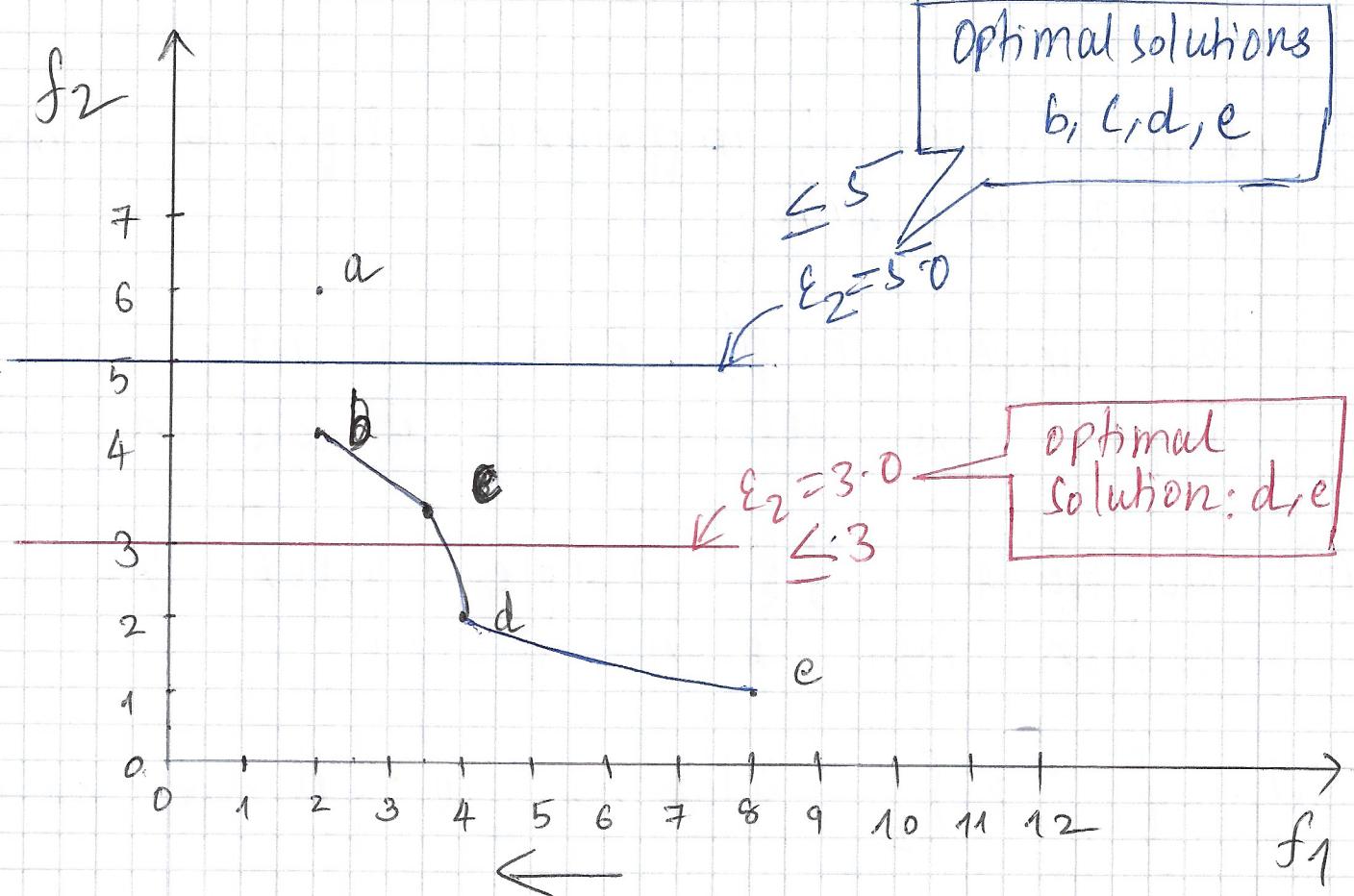
## Advantage:

- PO Solution can be found easily.



## Drawback:

- Pareto front should be determined in advance without which putting the  $\epsilon$  (hyperparameter) may be difficult.



### 5.3 Lexicographic Ordering (Working Principle)

- Defines the preferences among objectives.
- Solution of first objective is influenced by or drives the remaining objectives.

#### Advantages -

- ① Solution is PO
- ② Successive decision making is possible.

#### Drawbacks -

- ① Absolute order of importance depends from person-to-person.
- ② Less important objectives have very less impact on deciding final solution.
- ③ No Trade off possible after the solution has been filtered

If  $f_1 \gg f_2$  Solutions a. & b are chosen

If  $f_2 \gg f_1$  Solution e is chosen.