

# Generalized Decomposition

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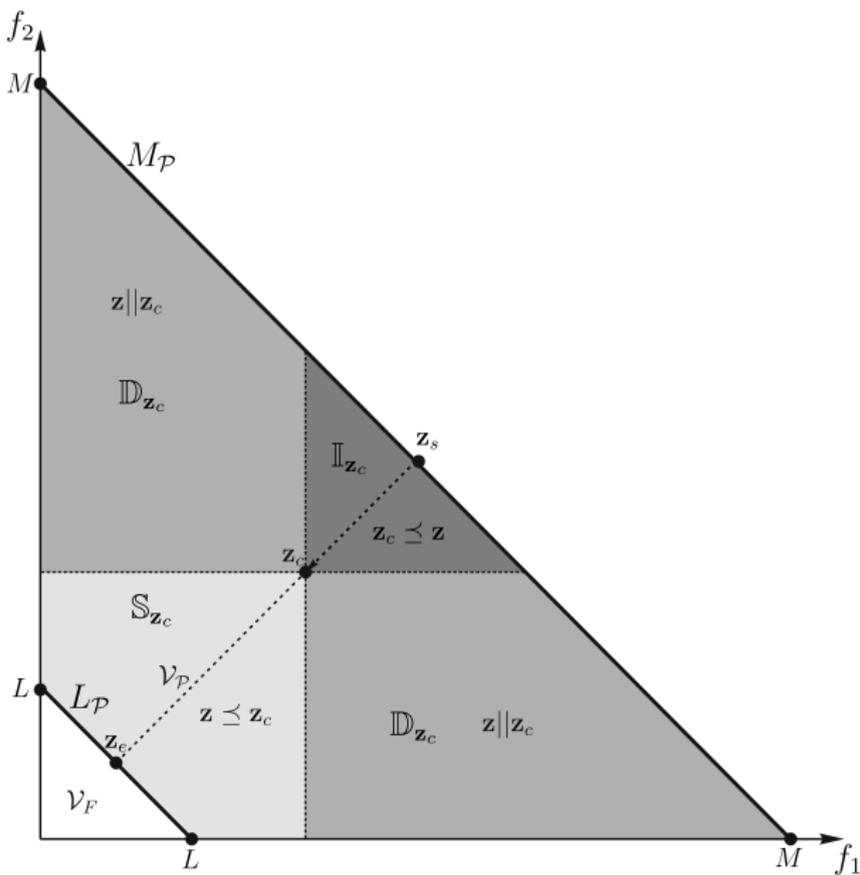
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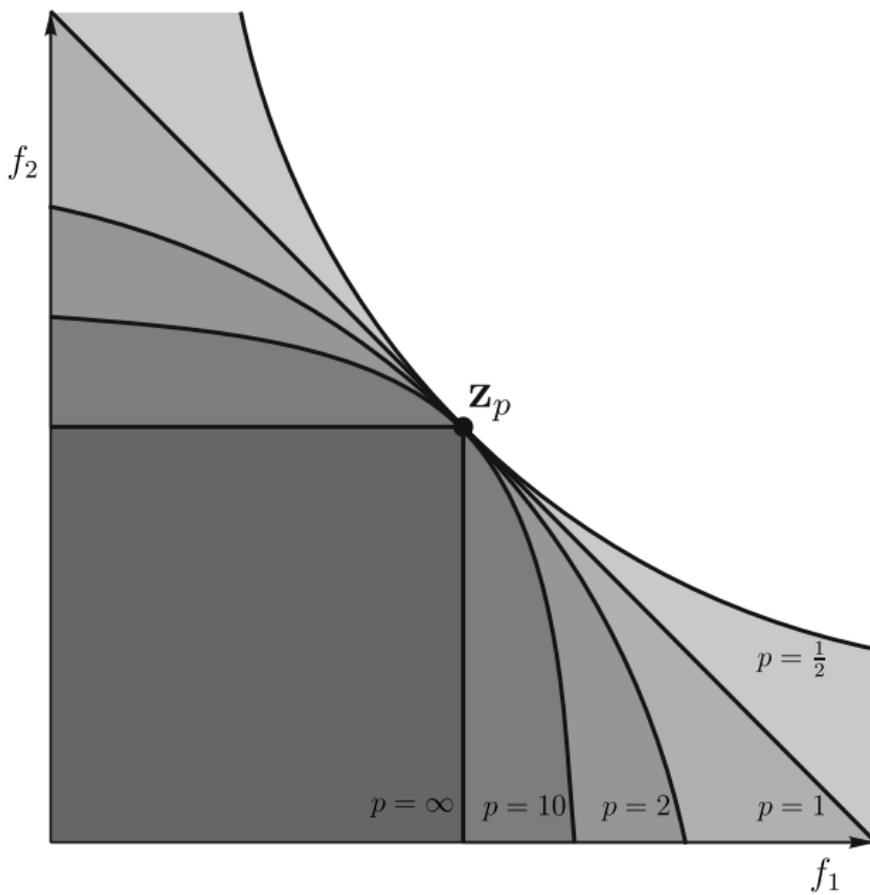
# Contents

- 1 Motivation
- 2 Methods for Selecting Weighting Vectors
- 3 Generalised Decomposition
- 4 A Glimpse Of Preference Articulation

## Why decomposition for many-objective problems?

- Search can be directed at will.
- Variety of norms and scalarizing functions (we'll see why this is beneficial).
- No incomparable solutions with respect to the  $\ell_p$ -norm used.
- It is unclear how to distribute solutions along the Pareto front
- Not every scalarizing function can guarantee that all Pareto optimal solutions are reachable.





## Evenly Distributed

### Evenly distributed weighting vectors

$$\left\{ \frac{0}{H}, \frac{1}{H}, \dots, \frac{H}{H} \right\}, \quad (1)$$

## Uniformly Distributed

### Uniformly distributed weighting vectors

$$\mathbf{w} = \{w_1, \dots, w_k\},$$
$$w_i = 1 - \sum_{m=1}^{i-1} w_m - (\mathcal{U}(0, 1))^{i-k}, \text{ for all } i = 1, \dots, k. \quad (2)$$

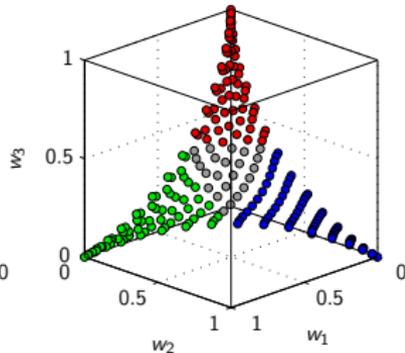
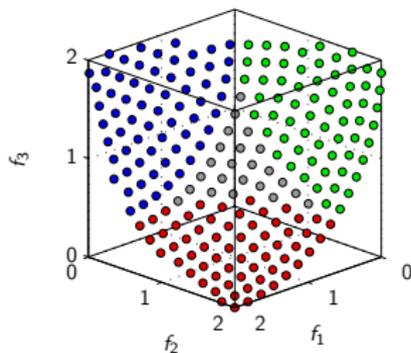
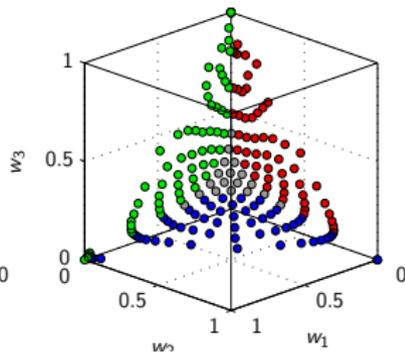
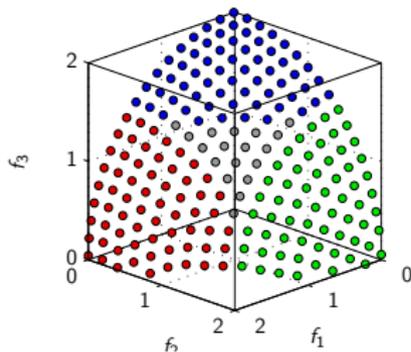
## Optimal Weighting Vector Choice

### Definition

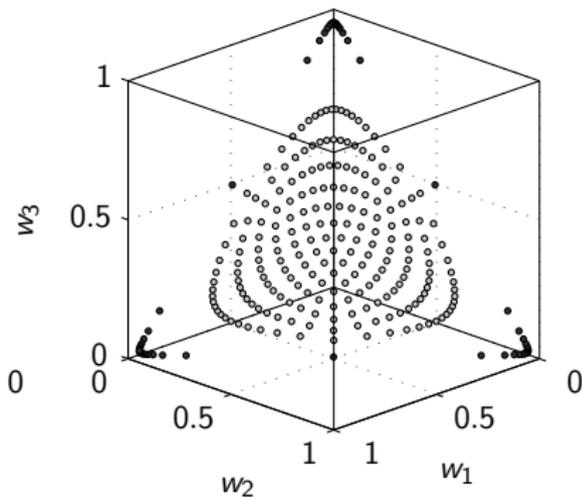
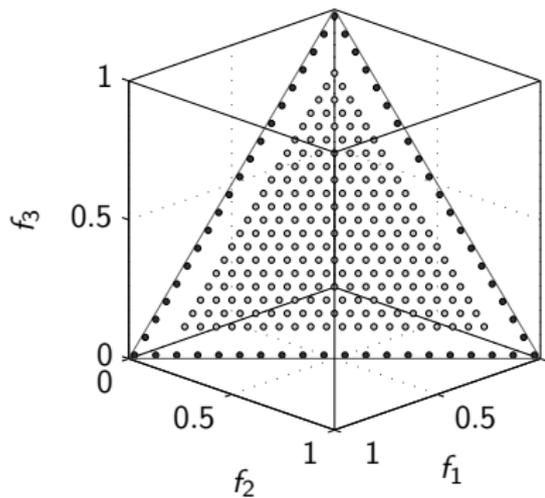
Generalised Decomposition

$$\begin{aligned} \min_{\mathbf{w}} & \|\mathbf{w} \circ \mathbf{F}(\mathbf{x})\|_p \\ \text{s.t.} & \mathbf{1}^T \mathbf{w} = 1 \\ & \mathbf{w} \succeq 0, \mathbf{F}(\mathbf{x}) \succeq 0. \end{aligned} \tag{3}$$

# Pareto Front Geometries



## Practical Considerations



## Practical Considerations (cont'd)

### Definition

$$\begin{aligned} \min_{\mathbf{x}} g_{\rho}(\mathbf{x}, \mathbf{w}^s, \mathbf{z}^*) &= \|\mathbf{w}^s \circ (\|\mathbf{F}(\mathbf{x}) - (\mathbf{z}^* - \epsilon)\| \\ &+ \rho \sum_{i=1}^k |f_i(\mathbf{x}) - (z_i^* - \epsilon)|)\|_{\rho} \end{aligned} \quad (4)$$

$\forall s = \{1, \dots, N\},$   
subject to  $\mathbf{x} \in \mathbf{S},$

## Practical Considerations (cont'd)

### Definition

$$\min_{\mathbf{w}} \|\mathbf{w} \circ (\mathbf{F}(\mathbf{x}) + \rho \cdot C(k))\|_p,$$

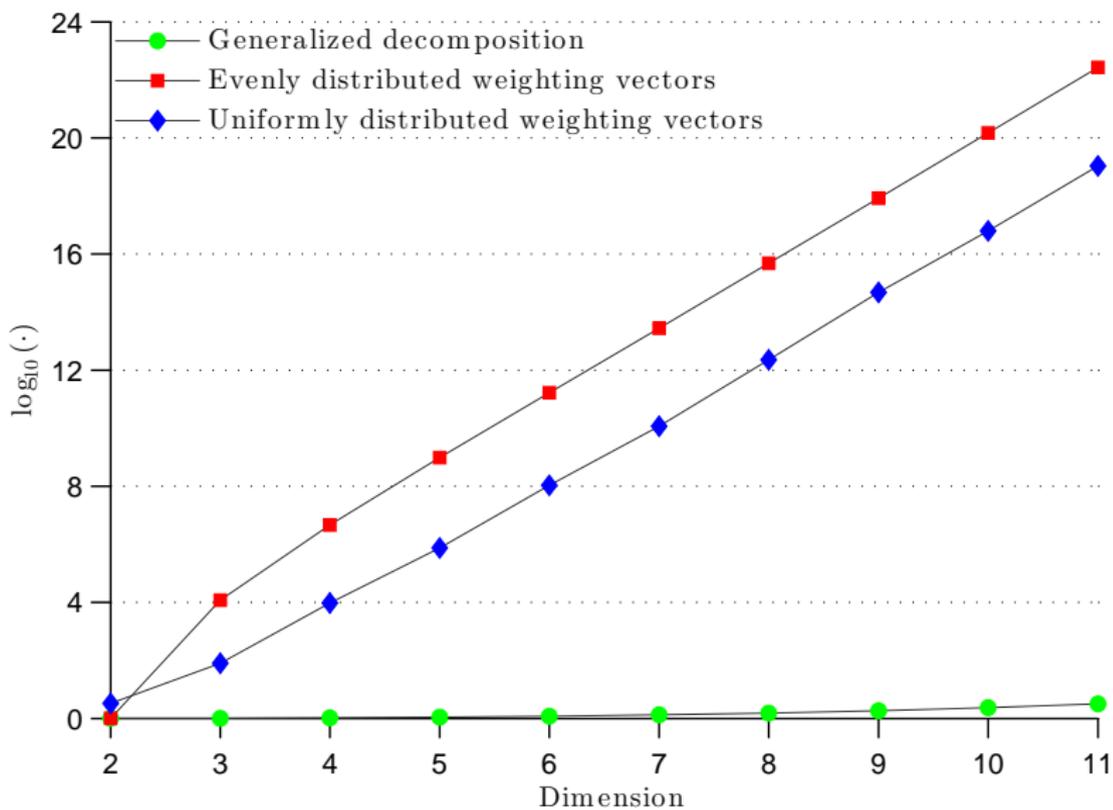
$$\text{subject to } \sum_{i=1}^k w_i = 1, \quad (5)$$

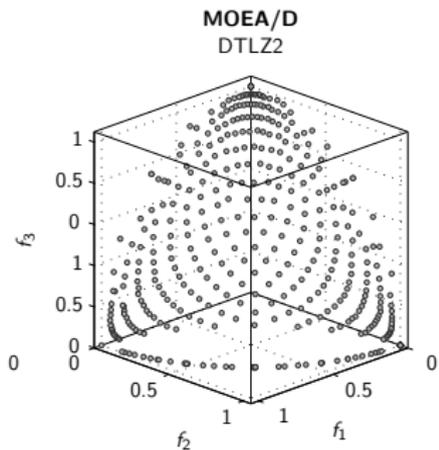
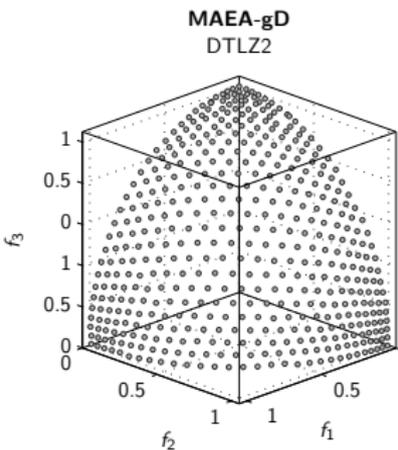
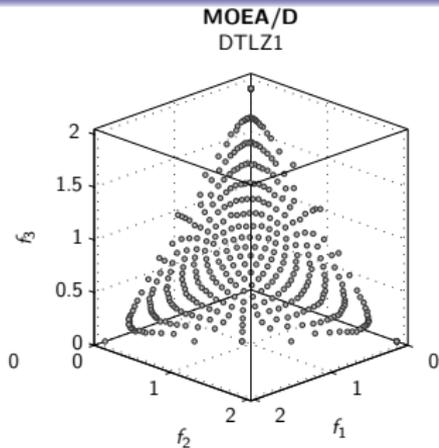
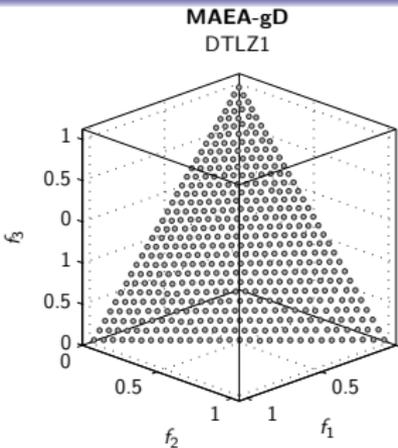
$$\text{and } w_i \geq 0, \forall i \in \{1, \dots, k\}, \mathbf{F}(\mathbf{x}) \geq 0.$$

## Definition

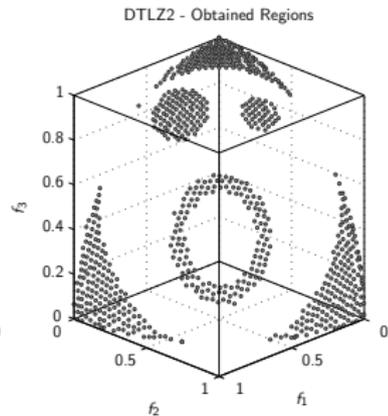
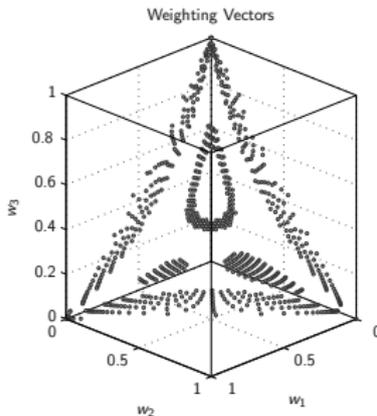
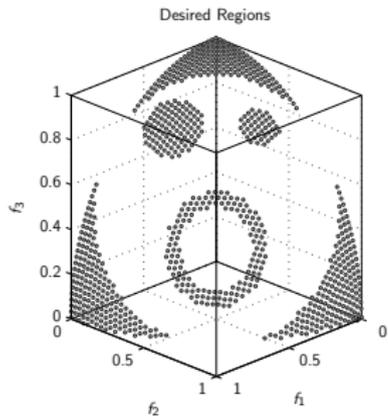
Coulomb Potential Energy

$$E(\mathbf{x}; s) = \sum_{1 \leq i < j \leq N} \|\mathbf{x}_i - \mathbf{x}_j\|^{-s}, \quad s > 0 \quad (6)$$





# Generalized Decomposition - Preference Articulation



## Conclusions

Given a measure and a scalarizing function and information about the Pareto front geometry, generalized decomposition can:

- Identify the optimal weighting vectors that minimize the given measure.
- Direct solutions in specific regions of the Pareto front.

## Why a Different Scalarization?

