## Smoothed Analysis of Multiobjective Optimization

### Heiko Röglin Department of Quantitative Economics



### July 2010 DIMAP Summer School

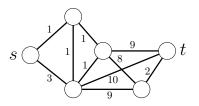
based on joint work with Rene Beier, Shang-Hua Teng, and Berthold Vöcking

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## **Optimization Problems**

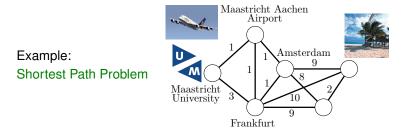
**Single-criterion Optimization Problem:** min f(x) subject to  $x \in S$ .

Example: Shortest Path Problem



## **Optimization Problems**

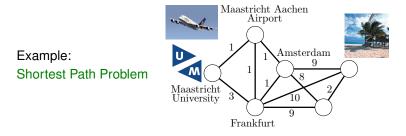
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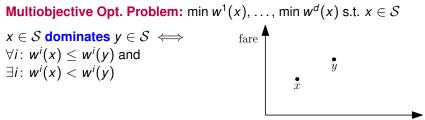
**Multiobjective Opt. Problem:** min  $f_1(x), \ldots, \min f_d(x)$  s.t.  $x \in S$ . Usually, there is no solution that is simultaneously optimal for all  $f_i$ .

#### Question

What can we do algorithmically to support the decision maker?

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## Pareto-optimal Solutions



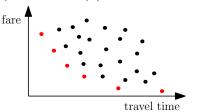
travel time

## Pareto-optimal Solutions

**Multiobjective Opt. Problem:** min  $w^1(x), \ldots, \min w^d(x)$  s.t.  $x \in S$ 

 $x \in S$  dominates  $y \in S \iff$  $\forall i: w^i(x) \le w^i(y)$  and  $\exists i: w^i(x) < w^i(y)$ 

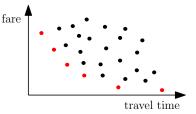
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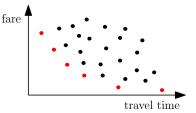
Often the Pareto curve is generated:

- Pareto curve limits options for decision maker.
- Monotone functions are optimized by Pareto-optimal solutions, e.g.,  $\lambda_1 w^1(x) + \ldots + \lambda_d w^d(x)$  or  $w^1(x) \cdots w^d(x)$ .

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#### **Central Question**

How large is the Pareto curve?

### Model

#### Linear Binary Optimization Problem

- set of feasible solutions S ⊆ {0, 1}<sup>n</sup>
   solution x = (x<sub>1</sub>,..., x<sub>n</sub>) ∈ S consists of *n* binary variables
- *d* linear objective functions:

$$\forall i \in \{1,\ldots,d\}: \min w^i(x) = w_1^i x_1 + \cdots + w_n^i x_n$$

S can encode arbitrary combinatorial structure, e.g., for a given graph, all paths from *s* to *t*, all Hamiltonian cycles, all spanning trees, ...

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#### How large is the Pareto curve?

- Exponential in the worst case for almost all problems.
- In practice, often few Pareto optimal solutions.



Example: Train Connections w.r.t. travel time, fare, number of train changes [Müller-Hannemann, Weihe 2001]

#### **Smoothed Analysis**

- $S \subseteq \{0,1\}^n$ , *d* objectives: min  $w^i(x) = w_1^i x_1 + \cdots + w_n^i x_n$
- Every coefficient  $w_j^i$  is an independent random variable following a probability density  $f_j^i : [-1, 1] \rightarrow [0, \phi]$ .

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- each  $w_i^i$  uniformly at random from interval of length  $1/\phi$
- $w^i_i$  are Gaussians, adversary specifies means,  $\phi \sim 1/\sigma$
- $\phi$  large  $\approx$  worst case  $\phi$  small  $\approx$  average case

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 $Q_d(n, \phi) = \max_{\mathcal{S}, f_j^i} \mathbf{E} \left[ \text{number of Pareto-optimal sol. for } \mathcal{S} \text{ and } f_j^i \right]$ 

### Results

**Bicriteria Optimization** (d = 2):

Beier, Vöcking (STOC 2003)

For any S and any  $f_j^i$ ,  $Q_2(n, \phi) = O(n^4 \phi)$ . There are S and  $f_j^i$  such that  $Q_2(n, \phi) = \Omega(n^2)$ .

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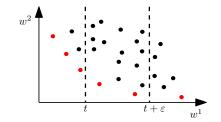
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### **Multiobjective Optimization (***d* **arbitrary constant):**

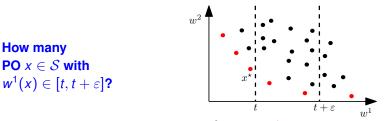
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For any S and any  $f_j^i$ ,  $Q_d(n, \phi) = O((n\phi)^{h(d)})$  for some function h. For any  $c \in \left[1, \sqrt{\log(n)}\right], Q_d(n, \phi)^c = O((n\phi)^{c \cdot h(d)}).$ 



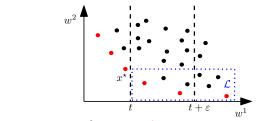
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Lemma [Beier, Vöcking (STOC 2004)]

For every  $\varepsilon \geq 0$  and  $t \in \mathbb{R}$ ,  $\Pr[\Lambda^1(t) \leq \varepsilon] \leq n\phi\varepsilon$ .

 $w^2$   $x^*$   $\lambda_2(t)$   $x^*$  L  $\lambda_1(t)$   $\Lambda_3(t)$   $t + \varepsilon$  $w^1$ 

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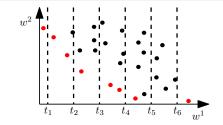
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#### Lemma [R., Teng (FOCS 2009)]

For every  $\varepsilon \geq 0, z \in \mathbb{N}$ , and  $t \in \mathbb{R}$ ,

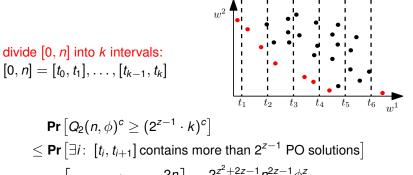
$$\Pr\left[\Lambda^{2^{z-1}}(t) \le \varepsilon\right] \le 2^{z^2+z} n^z \phi^z \varepsilon^{z-1}.$$

## **Higher Moments**



## divide [0, n] into *k* intervals: $[0, n] = [t_0, t_1], \dots, [t_{k-1}, t_k]$

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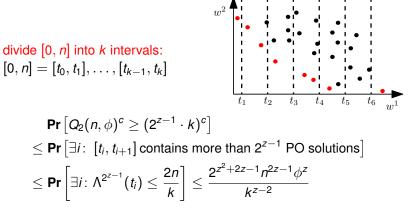


$$\Pr\left[Q_2(n,\phi)^c \ge (2^{z-1} \cdot k)^c\right]$$
  

$$\leq \Pr\left[\exists i: [t_i, t_{i+1}] \text{ contains more than } 2^{z-1} \text{ PO solutions}\right]$$
  

$$\leq \Pr\left[\exists i: \Lambda^{2^{z-1}}(t_i) \le \frac{2n}{k}\right] \le \frac{2^{z^2+2z-1}n^{2z-1}\phi^z}{k^{z-2}}$$

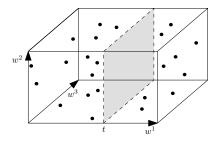
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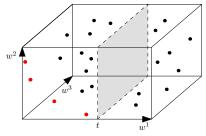
#### R., Teng (FOCS 2009)

For any S and any  $f_j^i$  and every  $c \in [1, \sqrt{\log(n)}]$ ,  $\mathbf{Q}_{\mathbf{d}}(\mathbf{n}, \phi)^{\mathbf{c}} = (\mathbf{n}^2 \phi)^{\mathbf{c}(1+\mathbf{o}(1))}$ .

How many PO  $x \in S$  with  $w^1(x) \in [t, t + \varepsilon]$ ? min  $w^2(x), \dots, \min w^d(x)$ s.t.  $w^1(x) \leq t$  and  $x \in S$ 



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#### *d* = 2

optimal solution  $x^*$ 

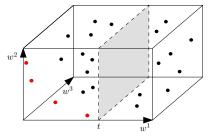
 $d \ge 3$  $\mathcal{P}^{\star} = \text{set of PO solutions}$ 

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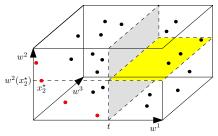


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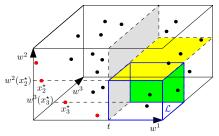


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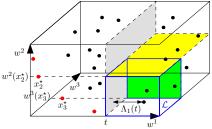
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 $\begin{aligned} \boldsymbol{d} &\geq \boldsymbol{3} \\ \mathcal{P}^{\star} &= \boldsymbol{set of PO solutions} \\ (\boldsymbol{x}_{2}^{\star}, \dots, \boldsymbol{x}_{d}^{\star}) \in (\mathcal{P}^{\star})^{d-1} \\ \mathcal{L} &= \{\boldsymbol{x} \in \mathcal{S} \mid \forall i \colon \boldsymbol{w}^{i}(\boldsymbol{x}) < \boldsymbol{w}^{i}(\boldsymbol{x}_{i}^{\star}) \} \end{aligned}$ 

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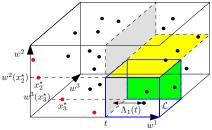
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Every (d - 1)-tuple from  $\mathcal{P}$  defines a loser gap. Induction: Use bound for  $Q_{d-1}(n, \phi)^{d-1}$  to bound  $Q_d(n, \phi)$ .

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#### Theorem

$$\forall d \ \forall c \in [1, \sqrt{\log(n)}]: Q_d(n, \phi)^c = O((n\phi)^{c \cdot h(d)}) \text{ for } h(d) = 2^{d-3}d!.$$

#### **Further Results**

- polynomial bound for  $Q_d(n, \phi)^c$  for any constants *c* and *d*  $\Rightarrow$  first concentration bounds for  $|\mathcal{P}|$
- extension to integer case  $\mathcal{S} \subseteq \{-m, -m+1, \dots, m-1, m\}^n$
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#### **Open Questions**

- $h(d) = 2^{d-3}d!: (n\phi)^{O(d)}$  possible?
- Iower bounds?
- higher moments, stronger concentration?

# Thank you for your attention!



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