

# Multivariate Higher Order Monotonicity – and its Preservation

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# 1. Basic notions

$I_1, \dots, I_d \subseteq \mathbb{R}$  intervals (always non-degenerate)

$I := I_1 \times \dots \times I_d, f : I \rightarrow \mathbb{R}$

For  $s \in I, h \in \mathbb{R}_+^d$  such that also  $s + h \in I$

$$(E_h f)(s) := f(s + h)$$

$$\Delta_h := E_h - E_0, \text{ i.e. } (\Delta_h f)(s) := f(s + h) - f(s)$$

Since  $\{E_h \mid h \in \mathbb{R}_+^d\}$  is commutative, so is  $\{\Delta_h \mid h \in \mathbb{R}_+^d\}$ .

For any  $h, \Delta_h^0 f := f$  (also for  $h = 0$ ), but  $\Delta_0 f = 0 \forall f$ .

For  $\mathbf{n} = (n_1, \dots, n_d) \in \mathbb{N}_0^d$  and  $h = (h_1, \dots, h_d) \in \mathbb{R}_+^d$

$$\Delta_h^{\mathbf{n}} := \Delta_{h_1 e_1}^{n_1} \Delta_{h_2 e_2}^{n_2} \cdots \Delta_{h_d e_d}^{n_d}$$

(where  $e_1, \dots, e_d$  are standard unit vectors), so that  $(\Delta_h^{\mathbf{n}} f)(s)$  is defined for  $s, s + \sum_{i=1}^d n_i h_i e_i \in I$ .

We first consider  $d = 1$ ,  $\mathbf{n} = n \in \mathbb{N}$ ,  $I \subseteq \mathbb{R}$  and put

$$\sigma_n(x_1, \dots, x_n) := x_1 + \cdots + x_n.$$

Assume  $I = [0, 1]$ , hence  $f : [0, 1] \rightarrow \mathbb{R}$ . For  $t \in [0, 1[$  and  $h > 0$ ,  $t + nh \leq 1$ ,

$$\begin{aligned} (\Delta_h^n f)(t) &= f(t + nh) - \binom{n}{1} f(t + (n-1)h) \pm \dots + (-1)^n f(t) \\ &= \left[ \Delta_{(h, \dots, h)}^{(1, \dots, 1)} (f \circ \sigma_n) \right] \left( \frac{t}{n}, \dots, \frac{t}{n} \right). \end{aligned}$$

If a univariate  $f$  is  $C^\infty$ , then

$$\Delta_h^n(f) \geq 0 \quad \forall h > 0 \quad \Leftrightarrow \quad f^{(n)} \geq 0.$$

If a multivariate  $f$  is  $C^\infty$ , then

$$\Delta_h^n(f) \geq 0 \quad \forall h \in \mathbb{R}_+^d \quad \Leftrightarrow \quad f_{\mathbf{n}} := \frac{\partial^{|\mathbf{n}|} f}{\partial x_1^{n_1} \dots \partial x_d^{n_d}} \geq 0$$

**Definition 1.**

$I \subseteq \mathbb{R}^d$ ,  $\mathbf{n} \in \mathbb{N}_0^d \setminus \{0\}$ ,  $f : I \rightarrow \mathbb{R}$  is  $\mathbf{n}$ - $\uparrow$  (read “ $\mathbf{n}$ -increasing”) iff

$$(\Delta_h^{\mathbf{p}} f)(s) \geq 0 \quad \forall s \in I, h \in \mathbb{R}_+^d, \mathbf{p} \in \mathbb{N}_0^d, 0 \not\leq \mathbf{p} \leq \mathbf{n}$$

such that  $s_j + p_j h_j \in I_j \quad \forall j \in [d] := \{1, \dots, d\}$ .

For a  $C^\infty$  function  $f$  then

$$f \text{ is } \mathbf{n}\text{-}\uparrow \quad \Leftrightarrow \quad f_{\mathbf{p}} \geq 0 \text{ for } 0 \not\leq \mathbf{p} \leq \mathbf{n}.$$

- For  $d = 1$ ,  $\mathbf{n} = n \in \mathbb{N}$ ,  $f$  is  $1$ - $\uparrow$  iff  $f$  is (weakly) increasing, and  $f$  is  $2$ - $\uparrow$  iff  $f$  is increasing and convex.
- $d = 2$ ,  $f(s_1, s_2) := (s_1 s_2 - a)_+$ ,  $a > 0$   
 $f$  is  $(1, 0)$ - $\uparrow$ ,  $(0, 1)$ - $\uparrow$ ,  $(1, 1)$ - $\uparrow$  (will be shown later), but not  $(2, 2)$ - $\uparrow$ :

$$\left( \Delta_{(\sqrt{a}, \sqrt{a})}^{(2,2)} f \right) (0) = -a.$$

## 2. The univariate case

### Theorem 1.

$f : [0, 1[ \rightarrow \mathbb{R}_+$  is  $n$ - $\uparrow$  ( $n \geq 2$ ) iff  $\exists! a_0, \dots, a_{n-2} \geq 0$  and a measure  $\mu$  on  $[0, 1[$  such that

$$f(t) = a_0 + a_1 t + \dots + a_{n-2} t^{n-2} + \int (t - a)_+^{n-1} d\mu(a).$$

$f$  is continuous and for  $n > 2$  ( $n - 2$ ) times continuously differentiable.

## 2. The univariate case

Let  $K_n := \{f : [0, 1] \rightarrow \mathbb{R}_+ \mid f \text{ is } n\text{-}\uparrow, f(1) = 1\}$ , and  $f_a(t) := (t - a)_+ / (1 - a)$  for  $0 \leq a < 1$ ,  $f_1 := \mathbf{1}_{\{1\}}$ ,  $E_n := \{1, t, \dots, t^{n-2}\} \cup \{f_a^{n-1} \mid a \in [0, 1]\}$ ,  $n \geq 2$

### Corollary 1.

$K_n$  is a Bauer simplex,  $\text{ex}(K_n) = E_n$  for  $n \geq 2$ .

### Corollary 2.

If  $f : [0, 1[ \rightarrow \mathbb{R}_+$  is “absolutely monotone”, i.e.  $n\text{-}\uparrow \forall n \in \mathbb{N}$ , then

$$f(t) = \sum_{j \geq 0} a_j t^j \quad \text{where } a_j \geq 0 \forall j.$$

### Corollary 3.

$K_\infty := \bigcap_{n \geq 1} K_n$  is a Bauer simplex with  $\text{ex}(K_\infty) = \{1, t, t^2, \dots\} \cup \{\mathbf{1}_{\{1\}}\}$ .



### 3. The multivariate case

First some special situations and examples.

①  $f_j : I_j \rightarrow \mathbb{R}$ ,  $1 \leq j \leq d$ ,  $f := f_1 \otimes \cdots \otimes f_d$ , i.e.

$$f(s) = \prod_{i=1}^d f_i(s_i)$$

Then

$$(\Delta_h^{\mathbf{p}} f)(s) = \prod_{j=1}^d (\Delta_{h_j}^{p_j} f_j)(s_j)$$

for  $\mathbf{p} \in \mathbb{N}_0^d$ ,  $h \in \mathbb{R}_+^d$ , implying for  $f_j \geq 0$

$$f \text{ is } \mathbf{n}\text{-}\uparrow \Leftrightarrow f_j \text{ is } n_j\text{-}\uparrow \forall j.$$

- 2  $f(s) := \sum_{j=1}^d f_j(s_j)$  (a “tensor sum”)

Then

$$\Delta_h^{\mathbf{p}} f(s) = \begin{cases} \Delta_{h_i}^{p_i} f_i(s_i) & \text{if } \mathbf{p} = p_i \mathbf{e}_i, p_i \geq 1 \\ = 0 & \text{if } \mathbf{p} \text{ has more than one positive entry.} \end{cases}$$

$$f \text{ is } \mathbf{n}\text{-}\uparrow \Leftrightarrow f_j \text{ is } n_j\text{-}\uparrow \forall j$$

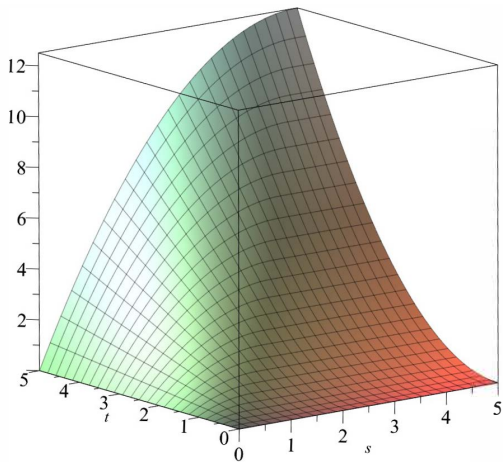
( $f_j \geq 0$  **not** necessary)

3  $d = 2,$

$$\begin{aligned}
 f(s, t) &:= \int_0^\infty \underbrace{\mathbf{1}_{[a, \infty[}(s)}_{1\text{-}\uparrow} \underbrace{(t - a)_+}_{2\text{-}\uparrow} da \text{ on } \mathbb{R}_+^2 \\
 &\quad \underbrace{\hspace{10em}}_{(1,2)\text{-}\uparrow} \\
 &= \int_0^{s \wedge t} (t - a)_+ da = (s \wedge t) \cdot \left( t - \frac{1}{2}(s \wedge t) \right)
 \end{aligned}$$

$f$  is  $(1, 2)$ - $\uparrow$  (and not more!)

### 3. The multivariate case



$$f(s, t) := (s \wedge t) \cdot \left( t - \frac{1}{2}(s \wedge t) \right)$$

### 3. The multivariate case

For  $\mathbf{n} \in \mathbb{N}^d$ ,  $\mathbf{n} \geq \mathbf{2}_d$  we consider

$$K_{\mathbf{n}} := \{f : [0, 1]^d \rightarrow \mathbb{R}_+ \mid f \text{ is } \mathbf{n}\text{-}\uparrow, f(\mathbf{1}_d) = 1\},$$

obviously convex and compact, and

$$E_{\mathbf{n}} := E_{n_1} \otimes \cdots \otimes E_{n_d} = \{f_1 \otimes \cdots \otimes f_d \mid f_i \in E_{n_i} \forall i\}.$$

#### Theorem 2.

For  $\mathbf{n} \geq \mathbf{2}_d$   $K_{\mathbf{n}}$  is a Bauer simplex, and  $\text{ex}(K_{\mathbf{n}}) = E_{\mathbf{n}}$ .

For example, if  $d = 2$ , any  $f : [0, 1]^2 \rightarrow \mathbb{R}_+$  which is  $(2, 2)$ - $\uparrow$ , has the form

$$f(s, t) = \int_{E_2 \times E_2} \rho_1(s)\rho_2(t) d\mu(\rho_1, \rho_2),$$

hence in case  $f(s, 0) = f(0, t) = 0 \forall s, t$ ,

$$f(s, t) = \int_{[0,1]^2} f_a(s)f_b(t) d\mu(a, b)$$

for some measure  $\mu$  on  $[0, 1]^2$ .

Let us now consider the special important case

$$\mathbf{n} = \mathbf{1}_d = (1, \dots, 1).$$

For  $d = 2$ ,  $h = (h_1, h_2) \in \mathbb{R}_+^2$

$$\begin{aligned} \left( \Delta_h^{(1,1)} f \right) (s, t) &= f(s + h_1, t + h_2) - f(s, t + h_2) \\ &\quad - f(s + h_1, t) + f(s, t) \end{aligned}$$

$$\left( \Delta_h^{(1,0)} f \right) (s, t) = f(s + h_1, t) - f(s, t)$$

$$\left( \Delta_h^{(0,1)} f \right) (s, t) = f(s, t + h_2) - f(s, t).$$

### 3. The multivariate case

For general  $d \in \mathbb{N}$ ,  $h \in \mathbb{R}_+^d$

$$\begin{aligned}(\Delta_h^{1^d} f)(s) &= f(s+h) - f(s_1, s_2+h_2, \dots, s_d+h_d) - \dots \\ &\quad - f(s_1+h_1, \dots, s_{d-1}+h_{d-1}, s_d) \\ &\quad + f(s_1, s_2, s_3+h_3, \dots, s_d+h_d) + \dots \\ &\quad \dots + (-1)^d f(s).\end{aligned}$$

If  $f$  is the distribution function (“d.f.”) of some measure  $\mu$ , say on  $\mathbb{R}_+^d$ , i.e.

$$f(s) = \mu([0, s]),$$

then

$$(\Delta_h^{1^d} f)(s) = \mu(]s, s+h]) \geq 0.$$



**Theorem 3.**

Let  $\emptyset \neq I_j \subseteq \overline{\mathbb{R}}$  be arbitrary (not necessarily intervals),  $j \leq d$ ,  
 $I = I_1 \times \cdots \times I_d$ ,  $f : I \rightarrow \mathbb{R}_+$ . Then

*$f$  is the d.f. of some measure on  $\bar{I}$*

*$\Leftrightarrow f$  is  $\mathbf{1}_d$ - $\uparrow$  and right-continuous.*

The proof relies on the fact, that  $f(\geq 0)$  is  $\mathbf{1}_d$ - $\uparrow$  iff  $f$  is completely monotone on the semigroup  $(I, \wedge)$  iff  $f$  is positive definite on  $(I, \wedge)$ .

In dimension one any  $2$ - $\uparrow$  function is right-continuous; if a multivariate  $f$  is (only)  $\mathbf{1}_d$ - $\uparrow$ , this need **not** be the case:

$$f = \mathbf{1}_{]0,1]^2} \text{ on } [0, 1]^2.$$

However, for  $\mathbf{n} \in \mathbb{N}^d$  each  $\mathbf{n}$ - $\uparrow$   $f$  is the pointwise limit of some net of  $\mathbf{n}$ - $\uparrow$  right-continuous functions.

(If  $\mathbf{n} \geq \mathbf{2}_d$ , then  $f$  itself is right-continuous.)

**Application.** Mean values as distribution functions

For  $x \in ]0, \infty[^d$  and  $t \in \mathbb{R}$  consider

$$M_t(x) := \left( \frac{1}{d} \sum_{i=1}^d x_i^t \right)^{\frac{1}{t}} \quad \text{for } t \neq 0$$

$$M_0(x) := \left( \prod_{i=1}^d x_i \right)^{\frac{1}{d}} \quad \left( = \lim_{t \rightarrow 0} M_t(x) \right)$$

$$M_\infty(x) := \max_{i \leq d} x_i \quad \left( = \lim_{x \rightarrow \infty} M_t(x) \right)$$

$$M_{-\infty}(x) := \min_{i \leq d} x_i \quad \left( = \lim_{t \rightarrow -\infty} M_t(x) \right).$$

(If  $t < 0$  and  $x_i = 0$  for some  $i$ , then  $M_t(x) = 0$ .)

The function  $\mathbb{R} \ni t \mapsto M_t(x)$  (for non-constant  $x$ ) is continuous and strictly increasing from  $\min x_i$  to  $\max x_i$ .

Since  $M_t(1, \dots, 1) = 1$ , these mean values are candidates for d.f.s of probability measures on  $[0, 1]^d$ .

**Theorem 4.**

$M_t|_{[0,1]^d}$  is a d.f.  $\Leftrightarrow t \in [-\infty, \frac{1}{d-1}] \cup \{\frac{1}{d-2}, \dots, \frac{1}{2}, 1\}$   
 ( $t \in [-\infty, 1]$  for  $d = 2$ )

How to prove this?

$$M_t = f_t \circ \left( \frac{1}{d} \sum_{i=1}^d x_i^t \right), \quad f_t(s) := s^{1/t} \text{ on } ]0, \infty[$$

For  $t > 0$ ,  $\sum_{i=1}^d x_i^t$  is a tensor sum of increasing functions, hence  $\mathbf{1}_d$ - $\uparrow$ .

We need to know which functions on  $\mathbb{R}_+$  preserve this property!

### 3. The multivariate case

The following result was essentially shown by P. M. Morillas (2005):

#### Theorem 5.

*Let  $I \subseteq \mathbb{R}^d$  and  $J \subseteq \mathbb{R}$  be intervals,  $g : I \rightarrow J$ ,  $f : J \rightarrow \mathbb{R}$ . Then, if  $g$  is  $\mathbf{1}_d$ - $\uparrow$  and  $f$  is  $d$ - $\uparrow$ ,  $f \circ g$  is again  $\mathbf{1}_d$ - $\uparrow$ .*

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Let's apply this to the mean values  $M_t$  (for  $t > 0$ ):

- for  $t \in \{1, \frac{1}{2}, \frac{1}{3}, \dots\}$  we have  $f_t(s) = s^k$  for some  $k \in \mathbb{N}$ , hence  $f_t$  is absolutely monotone, and  $M_t$  a d.f.

**Theorem 5.**

Let  $I \subseteq \mathbb{R}^d$  and  $J \subseteq \mathbb{R}$  be intervals,  $g : I \rightarrow J$ ,  $f : J \rightarrow \mathbb{R}$ . Then, if  $g$  is  $\mathbf{1}_d$ - $\uparrow$  and  $f$  is  $d$ - $\uparrow$ ,  $f \circ g$  is again  $\mathbf{1}_d$ - $\uparrow$ .

Let's apply this to the mean values  $M_t$  (for  $t > 0$ ):

- for  $t \in \{1, \frac{1}{2}, \frac{1}{3}, \dots\}$  we have  $f_t(s) = s^k$  for some  $k \in \mathbb{N}$ , hence  $f_t$  is absolutely monotone, and  $M_t$  a d.f.
- for  $t \in ]0, 1] \setminus \{1, \frac{1}{2}, \frac{1}{3}, \dots\}$ :

$$\begin{aligned} \left(x^{\frac{1}{t}}\right)^{(k)} &= \frac{1}{t} \left(\frac{1}{t} - 1\right) \cdots \left(\frac{1}{t} - (k-1)\right) x^{\frac{1}{t}-k} \\ \frac{1}{t} > k-1 &\Rightarrow \dots > 0 \\ \Leftrightarrow t < \frac{1}{k-1} \end{aligned}$$

hence  $t < \frac{1}{d-1}$  is sufficient for  $f_t$  to be  $d$ - $\uparrow$ . (in fact also necessary;  $t > 1$  will be dealt with later)



Before considering  $M_t$  for  $t < 0$ , this remark: Let  $\alpha, \beta > 0$

then  $x_i \mapsto -x_i^{-\alpha}$  is increasing on  $]0, \infty[$

$\Rightarrow x \mapsto -\sum x_i^{-\alpha}$  is  $\mathbf{1}_d$ - $\uparrow$  on  $]0, \infty[^d$

$s \mapsto s^{-\beta}$  is completely monotone on  $]0, \infty[$

$\Leftrightarrow s \mapsto (-s)^{-\beta}$  is absolutely monotone on  $] -\infty, 0[$

Now let  $t < 0$ ,  $\alpha := -t$ ,  $\beta := -\frac{1}{t}$

$\Rightarrow M_t(x) = d^{-1/t} \left[ -\left( -\sum x_i^{-\alpha} \right) \right]^{-\beta}$  is  $\mathbf{1}_d$ - $\uparrow$ ,

again by Theorem 5.

An important supplement to Theorem 5:

#### Theorem 6.

Let  $I \subseteq \mathbb{R}^d$  be an interval (always non-degenerate),

$\sigma_d(x) := \sum_{i=1}^d x_i$ ,  $J := \sigma_d(I)$ ,  $f : J \rightarrow \mathbb{R}$ . Then

$$f \text{ is } d\text{-}\uparrow \Leftrightarrow f \circ \sigma_d \text{ is } \mathbf{1}_d\text{-}\uparrow.$$

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$$f \text{ is } d\text{-}\uparrow \Leftrightarrow f \circ \sigma_d \text{ is } \mathbf{1}_d\text{-}\uparrow.$$

Here “ $\Rightarrow$ ” follows from Theorem 5. For “ $\Leftarrow$ ” it is sufficient to consider  $I = [0, \frac{1}{d}]^d$  and  $J = [0, 1]$ . Then for  $t \in [0, 1[$ ,  $h > 0$ ,  $t + k \cdot h \leq 1$  ( $k \leq d$ )

$$\begin{aligned} (\Delta_h^k f)(t) &= f(t + kh) - \binom{k}{1} f(t + (k-1)h) + \cdots + (-1)^k f(t) \\ &= \left( \Delta_{(h, \dots, h)}^{\mathbf{1}_k, \mathbf{0}_{d-k}} (f \circ \sigma_d) \right) \left( \frac{t}{d}, \dots, \frac{t}{d} \right) \end{aligned}$$

However: attention!

We'd need  $\frac{t}{d} + h \leq \frac{1}{d}$ , or  $t + dh \leq 1$ , but only know  $t + kh \leq 1$ .

#### Lemma 1.

*$J \subseteq \mathbb{R}$  interval,  $f : J \rightarrow \mathbb{R}$ ,  $k \in \mathbb{N}$ . If  $\exists h_0 > 0$  such that  $(\Delta_h^k f)(t) \geq 0 \forall t \in J, h \in ]0, h_0]$  with  $t + kh \in J$ , then the same holds  $\forall h > 0$  such that  $t + kh \in J$ .*

To finish the proof of Theorem 6, choose  $h_0 := \frac{1-t}{d}$ .

### 3. The multivariate case

However: attention!

We'd need  $\frac{t}{d} + h \leq \frac{1}{d}$ , or  $t + dh \leq 1$ , but only know  $t + kh \leq 1$ .

#### Lemma 1.

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To finish the proof of Theorem 6, choose  $h_0 := \frac{1-t}{d}$ .

#### Corollary 4.

*$f : [0, 1] \rightarrow \mathbb{R}$  is  $d$ - $\uparrow$  iff  $f \circ M_1$  is  $\mathbf{1}_d$ - $\uparrow$ .*

A natural question:

Suppose  $F$  is a two-dimensional d.f.,  $G$  a three-dimensional d.f., for which functions  $f$  on  $[0, 1]^2$  is then always  $f \circ (F \times G)$  a five-dimensional d.f.?

Apart from normalisation, when is  $f(F(x), G(y))$   $\mathbf{1}_5$ - $\uparrow$ ?

#### Theorem 7.

Let  $I_1 \subseteq \mathbb{R}^{n_1}, \dots, I_d \subseteq \mathbb{R}^{n_d}$  be intervals,  $g_1 : I_1 \rightarrow [0, 1]$   $\mathbf{1}_{n_1}$ - $\uparrow$ ,  $\dots$ ,  $g_d : I_d \rightarrow [0, 1]$   $\mathbf{1}_{n_d}$ - $\uparrow$ ,  $\mathbf{n} := (n_1, \dots, n_d)$ . If  $f : [0, 1]^d \rightarrow \mathbb{R}_+$  is  $\mathbf{n}$ - $\uparrow$ , then  $f \circ (g_1 \times \dots \times g_d)$  is  $\mathbf{1}_{|\mathbf{n}|}$ - $\uparrow$ .

### 3. The multivariate case

#### Theorem 7.

Let  $I_1 \subseteq \mathbb{R}^{n_1}, \dots, I_d \subseteq \mathbb{R}^{n_d}$  be intervals,  $g_1 : I_1 \rightarrow [0, 1]$   $\mathbf{1}_{n_1}$ - $\uparrow$ ,  $\dots$ ,  $g_d : I_d \rightarrow [0, 1]$   $\mathbf{1}_{n_d}$ - $\uparrow$ ,  $\mathbf{n} := (n_1, \dots, n_d)$ . If  $f : [0, 1]^d \rightarrow \mathbb{R}_+$  is  $\mathbf{n}$ - $\uparrow$ , then  $f \circ (g_1 \times \dots \times g_d)$  is  $\mathbf{1}_{|\mathbf{n}|}$ - $\uparrow$ .

#### Proof.

We may assume  $f(\mathbf{1}_d) = 1$ , and also  $\mathbf{n} \geq \mathbf{2}_d$ . Then (Theorem 2)

$$f(s) = \int_{E_{\mathbf{n}}} \prod_{i=1}^d \rho_i(s_i) d\mu(\rho_1, \dots, \rho_d)$$

for some probability measure  $\mu$  on  $E_{\mathbf{n}}$ . So

$$f \circ (g_1 \times \dots \times g_d) = \int \bigotimes_{i=1}^d (\rho_i \circ g_i) d\mu(\rho_1, \dots, \rho_d)$$

where each  $\rho_i \circ g_i$  is  $\mathbf{1}_{n_i}$ - $\uparrow$  (Theorem 5), therefore  $\bigotimes_{i=1}^d (\rho_i \circ g_i)$  is  $\mathbf{1}_{|\mathbf{n}|}$ - $\uparrow$ , and so is then  $f \circ (g_1 \times \dots \times g_d)$  as a mixture of those.  $\square$



A special case would be  $g_i = \sigma_{n_i}$  ( $\forall i$ ) on suitable  $n_i$ -dimensional intervals. As a generalisation of Theorem 6 we have

$$(\sigma_{\mathbf{n}} := \sigma_{n_1} \times \cdots \times \sigma_{n_d})$$

#### Theorem 8.

*Let  $I_1 \subseteq \mathbb{R}^{n_1}, \dots, I_d \subseteq \mathbb{R}^{n_d}$  be non-degenerate intervals,  $J_i := \sigma_{n_i}(I_i)$ ,  $J := J_1 \times \cdots \times J_d$  and  $f : J \rightarrow \mathbb{R}$ . Then*

$$f \text{ is } \mathbf{n} \text{-}\uparrow \iff f \circ \sigma_{\mathbf{n}} \text{ is } \mathbf{1}_{|\mathbf{n}|} \text{-}\uparrow$$

(similar proof!)

Finally, we get a natural generalisation of Theorem 5, our **first main result**:

#### Theorem 9.

Let  $g_i : I_i \rightarrow [0, 1]$  be  $\mathbf{m}_i$ - $\uparrow$ , where  $\mathbf{m}_i \in \mathbb{N}^{n_i}$ . Put  $g := g_1 \times \cdots \times g_d : I_1 \times \cdots \times I_d \rightarrow [0, 1]^d$ . If  $f : [0, 1]^d \rightarrow \mathbb{R}$  is  $(|\mathbf{m}_1|, \dots, |\mathbf{m}_d|)$ - $\uparrow$  then  $f \circ g$  is  $(\mathbf{m}_1, \dots, \mathbf{m}_d)$ - $\uparrow$ .

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Let  $g_i : I_i \rightarrow [0, 1]$  be  $\mathbf{m}_i$ - $\uparrow$ , where  $\mathbf{m}_i \in \mathbb{N}^{n_i}$ . Put  $g := g_1 \times \cdots \times g_d : I_1 \times \cdots \times I_d \rightarrow [0, 1]^d$ . If  $f : [0, 1]^d \rightarrow \mathbb{R}$  is  $(|\mathbf{m}_1|, \dots, |\mathbf{m}_d|)$ - $\uparrow$  then  $f \circ g$  is  $(\mathbf{m}_1, \dots, \mathbf{m}_d)$ - $\uparrow$ .

#### Example.

$d = 2$ ,  $n_1 = 2$ ,  $n_2 = 3$ ,  $\mathbf{m}_1 = (2, 4)$ ,  $\mathbf{m}_2 = (3, 3, 2)$ . If  $g_1$  (bivariate) is  $(2, 4)$ - $\uparrow$ ,  $g_2$  (trivariate) is  $(3, 3, 2)$ - $\uparrow$ , and  $f$  is  $(6, 8)$ - $\uparrow$ , then  $f \circ (g_1 \times g_2)$  is  $(2, 4, 3, 3, 2)$ - $\uparrow$  (as a function of 5 variables).

### 3. The multivariate case

Proof.

$$f(s) = \int \bigotimes_{i=1}^d \rho_i(s_i) d\mu(\rho)$$

$\mu$  on  $E_{(|\mathbf{m}_1|, \dots, |\mathbf{m}_d|)}$

$$\Rightarrow f \circ g = \int \bigotimes_{i=1}^d \rho_i \circ g_i d\mu(\rho)$$

$g_i$  is  $\mathbf{m}_i$ - $\uparrow$ , equiv.  $g_i \circ \sigma_{\mathbf{m}_i}$  is  $\mathbf{1}_{|\mathbf{m}_i|}$ - $\uparrow$

$$f \circ g \circ \sigma_{(\mathbf{m}_1, \dots, \mathbf{m}_d)} = \int \underbrace{\bigotimes_{i=1}^d \rho_i \circ g_i \circ \sigma_{\mathbf{m}_i}}_{\text{is } \mathbf{1}_{|\mathbf{m}_1| + \dots + |\mathbf{m}_d|} \text{-}\uparrow} d\mu(\rho)$$

$\Rightarrow f \circ g$  is  $(\mathbf{m}_1, \dots, \mathbf{m}_d)$ - $\uparrow$ . □

**Special cases:**

$$g \mathbf{n}\text{-}\uparrow, f |\mathbf{n}|\text{-}\uparrow \Rightarrow f \circ g \mathbf{n}\text{-}\uparrow$$

$$g n\text{-}\uparrow, f n\text{-}\uparrow \Rightarrow f \circ g n\text{-}\uparrow$$

(if defined...)

## 4. $k$ -increasing functions ( $k \in \mathbb{N}!$ ) in any dimension

Let  $I \subseteq \mathbb{R}$ ,  $f : I \rightarrow \mathbb{R}$  continuous,  $k \in \mathbb{N}$ . Then

$$\Delta_h^k(f)(t) \geq 0 \quad \forall t \in I, h > 0, t + kh \in I$$

is equivalent with

$$(\Delta_{h_1} \Delta_{h_2} \dots \Delta_{h_k} f)(t) \geq 0 \quad \forall t \in I, h_i > 0, t + \sum_{i=1}^k h_i \in I.$$

(Boas-Widder (1940), easy to see)

The following notion now seems natural:

### Definition 2.

$I \subseteq \mathbb{R}^d$  interval,  $f : I \rightarrow \mathbb{R}$ ,  $k \in \mathbb{N}$ . Then  $f$  is called  $k$ -increasing (“ $k$ - $\uparrow$ ”)

iff  $\forall j \in [k], \forall h^{(1)}, \dots, h^{(j)} \in \mathbb{R}_+^d, \forall s \in I$  such that  
 $s + h^{(1)} + \dots + h^{(j)} \in I$

$$(\Delta_{h^{(1)}} \dots \Delta_{h^{(j)}} f)(s) \geq 0.$$

#### 4. $k$ -increasing functions ( $k \in \mathbb{N}!$ ) in any dimension

- For  $k = 2$  these functions are known as *ultramodular*.
- For  $d = 1$  this definition is the known one.
- Already for  $d = 2$  increasing convexity and being  $2$ - $\uparrow$  are incomparable properties: on  $\mathbb{R}_+^2$  the product is  $2$ - $\uparrow$ , but not convex; and the Euclidean norm is convex, however not  $2$ - $\uparrow$ :

$$\left( \Delta_{e_1} \Delta_{e_2} \sqrt{x^2 + y^2} \right) (0) = \sqrt{2} - 2$$

There is a surprisingly close connection to  $\mathbf{n}$ - $\uparrow$  functions:

### Theorem 10.

Let  $I \subseteq \mathbb{R}^d$  be an interval,  $d, k \in \mathbb{N}$ ,  $f : I \rightarrow \mathbb{R}$ . Then

$$f \text{ is } k\text{-}\uparrow \Leftrightarrow f \text{ is } \mathbf{n}\text{-}\uparrow \quad \forall \mathbf{n} \in \mathbb{N}_0^d \text{ with } 0 < |\mathbf{n}| \leq k.$$

Furthermore:

$\forall m \in \mathbb{N}$ ,  $\forall$  interval  $J \subseteq \mathbb{R}^m$ ,  $\forall$  positive affine  $\varphi : \mathbb{R}^m \rightarrow \mathbb{R}^d$   
such that  $\varphi(J) \subseteq I$ , also  $f \circ \varphi$  is  $k$ - $\uparrow$ .



**Theorem 10.**

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Furthermore:

$$\forall m \in \mathbb{N}, \forall \text{ interval } J \subseteq \mathbb{R}^m, \forall \text{ positive affine } \varphi : \mathbb{R}^m \rightarrow \mathbb{R}^d \\ \text{such that } \varphi(J) \subseteq I, \text{ also } f \circ \varphi \text{ is } k\text{-}\uparrow.$$

**Corollary 5.**

$I \subseteq \mathbb{R}^d$ ,  $B \subseteq \mathbb{R}$  intervals,  $g : I \rightarrow B$  and  $f : B \rightarrow \mathbb{R}$  both  $k\text{-}\uparrow$ , then so is  $f \circ g$ .

Because:  $0 < |\mathbf{n}| \leq k \Rightarrow f \text{ } |\mathbf{n}|\text{-}\uparrow, g \text{ } \mathbf{n}\text{-}\uparrow \Rightarrow f \circ g \text{ } \mathbf{n}\text{-}\uparrow.$

**Lemma 2.**

$I \subseteq \mathbb{R}^{d_1}$ ,  $J \subseteq \mathbb{R}^{d_2}$  intervals,  $f : I \rightarrow \mathbb{R}_+$ ,  $g : J \rightarrow \mathbb{R}_+$  both  $k$ - $\uparrow$   
 $\Rightarrow f \otimes g$   $k$ - $\uparrow$  on  $I \times J$ . In case  $I = J$  the product  $f \cdot g$  is also  $k$ - $\uparrow$ .

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**Proof.**

$[\Delta_{(h^{(1)}, h^{(2)})}^{\mathbf{m}, \mathbf{n}}(f \otimes g)](x, y) = (\Delta_{h^{(1)}}^{\mathbf{m}} f)(x) \cdot (\Delta_{h^{(2)}}^{\mathbf{n}} g)(y)$ . For  $|\mathbf{m}, \mathbf{n}| = |\mathbf{m}| + |\mathbf{n}| \leq k$  both factors are  $\geq 0$  ( $\mathbf{m} = 0$  or  $\mathbf{n} = 0$  is possible, therefore  $f \geq 0$ ,  $g \geq 0$ ).

For  $I = J$ , let  $\varphi : \mathbb{R}^d \rightarrow \mathbb{R}^{2d}$  be given by  $\varphi(x) := (x, x)$ , a linear positive map, with  $\varphi(I) \subseteq I \times I$ . Therefore

$$(f \otimes g) \circ \varphi = f \cdot g \text{ is also } k\text{-}\uparrow. \quad \square$$

### Examples.

- Each monomial  $f(x) = \prod_{i=1}^d x_i^{n_i}$  ( $n_i \in \mathbb{N}$ ) is  $k$ - $\uparrow$  on  $\mathbb{R}_+^d \forall k \in \mathbb{N}$ .  
 $\prod_{i=1}^d x^{c_i}$  ( $c_i > 0$ ) is  $k$ - $\uparrow$  on  $\mathbb{R}_+^d$  at least for  $k \leq c_i + 1 \forall i$ .

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- For  $a > 0$  the function  $f(x, y) := (xy - a)_+$  is 2- $\uparrow$ , since  $(t - a)_+$  is 2- $\uparrow$  on  $\mathbb{R}_+$ . So, by Theorem 10,  $f$  is  $(1, 1)$ - $\uparrow$ , but not  $(2, 2)$ - $\uparrow$  as we saw earlier. It is even not  $(1, 2)$ - $\uparrow$ :  $\left(\Delta_{\frac{1}{2}, 1}^{(1, 2)} f\right) \left(\frac{1}{2}, 1\right) = -\frac{1}{2}$ , for  $a = 1$ .

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- $f(x, y, z) := xy + xz + yz - xyz$  on  $[0, 1]^3$   
 Then  $f_1 = y + z - yz \geq 0$ ,  $f_{(1, 1)} = 0$ ,  $f_{(1, 2)} = 1 - z \geq 0$ ,  
 $f_{(1, 2, 3)} = -1$  i.e.  $f$  is  $2$ - $\uparrow$ , but not  $3$ - $\uparrow$ .



Intermezzo: **Bernstein polynomials**

$$b_{i,r}(t) := \binom{r}{i} t^i (1-t)^{r-i}, \quad r \in \mathbb{N}, \quad i \in \{0, 1, \dots, r\}, \quad t \in \mathbb{R}$$

For  $\mathbf{i} = (i_1, \dots, i_d) \in \{0, 1, \dots, r\}^d$

$$B_{\mathbf{i},r} := b_{i_1,r} \otimes \dots \otimes b_{i_d,r}.$$

For any  $f : [0, 1]^d \rightarrow \mathbb{R}$  the associated Bernstein polynomials  $f^{(1)}, f^{(2)}, \dots$  are defined by

$$f^{(r)} := \sum_{\mathbf{0} \leq \mathbf{i} \leq \mathbf{r}_d} f\left(\frac{\mathbf{i}}{r}\right) B_{\mathbf{i},r}$$

For each continuity point  $x$  of  $f$  we have

$$f^{(r)}(x) \rightarrow f(x), \quad r \rightarrow \infty.$$

#### 4. $k$ -increasing functions ( $k \in \mathbb{N}!$ ) in any dimension

For each continuity point  $x$  of  $f$  we have

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In the following, the “upper right boundary” of  $[0, 1]^d$  will play a role.

For  $\alpha \subseteq [d]$  let  $T_\alpha := \{x \in [0, 1]^d \mid x_i < 1 \Leftrightarrow i \in \alpha\}$ . Then

$$[0, 1]^d = \bigcup_{\alpha \subseteq [d]} T_\alpha \text{ is a disjoint union}$$

$$T_\emptyset = \{\mathbf{1}_d\}, \quad T_{[d]} = [0, 1[^d$$

and  $\bigcup_{\alpha \subsetneq [d]} T_\alpha$  is called the *upper right boundary* of  $[0, 1]^d$ .

#### 4. $k$ -increasing functions ( $k \in \mathbb{N}!$ ) in any dimension

It is easy to show, that on each part  $T_\alpha$  ( $\alpha \subsetneq [d]$ ) of this boundary the restriction  $f|_{T_\alpha}$  has as its Bernstein polynomials the restrictions  $f^{(r)}|_{T_\alpha}$ . Thus we have the

#### Lemma 3.

*Let  $f : [0, 1]^d \rightarrow \mathbb{R}$  have the property that each restriction  $f|_{T_\alpha}$  for  $\emptyset \neq \alpha \subseteq [d]$  is continuous. Then  $\lim_{r \rightarrow \infty} f^{(r)}(x) = f(x) \forall x \in [0, 1]^d$ , i.e.  $f^{(r)}$  converges pointwise to  $f$ .*

(Note that  $f^{(r)}(\mathbf{1}_d) = f(\mathbf{1}_d) \forall r$ .)

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(Note that  $f^{(r)}(\mathbf{1}_d) = f(\mathbf{1}_d) \forall r$ .)

##### Lemma 4.

Let  $f : [0, 1]^d \rightarrow \mathbb{R}$  be  $2$ - $\uparrow$ . Then

- (i)  $f$  is continuous iff  $f$  is continuous in  $\mathbf{1}_d$ .
- (ii)  $f$  is right-continuous and on  $[0, 1]^d$  continuous.

#### 4. $k$ -increasing functions ( $k \in \mathbb{N}!$ ) in any dimension

##### Theorem 11.

$f : [0, 1]^d \rightarrow \mathbb{R}$ ,  $\mathbf{2}_d \leq \mathbf{n} \in \mathbb{N}_0^d$ ,  $2 \leq k \in \mathbb{N}$ .

④  $f \mathbf{n}\text{-}\uparrow \Rightarrow$  each  $f^{(r)}$  is  $\mathbf{n}\text{-}\uparrow$  and  $f^{(r)} \rightarrow f$  pointwise

④  $f k\text{-}\uparrow \Rightarrow$  each  $f^{(r)}$  is  $k\text{-}\uparrow$  and  $f^{(r)} \rightarrow f$  pointwise

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We can now tackle another natural question on the preservation of monotonicity, related but different to the previous one.

If  $g_1, \dots, g_m : I \rightarrow [0, 1]$  are d.f.s on some  $d$ -dimensional interval,

$$g = (g_1, \dots, g_m) : I \rightarrow [0, 1]^m, \text{ i.e. } g(s) = (g_1(s), g_2(s), \dots),$$

for which functions  $f$  on  $[0, 1]^m$  is  $f \circ g$  again a d.f.?

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for which functions  $f$  on  $[0, 1]^m$  is  $f \circ g$  again a d.f.?

For  $d = 1$   $f$  has just to be increasing (and right-cont.), for  $d = 2$  this is not sufficient:

$$\begin{aligned} g_1(s, t) &:= \frac{s+t}{2}, \quad g_2(s, t) := st, \quad f = 1_{[(\frac{1}{2}, \frac{1}{2}), (1, 1)]} \\ &\Rightarrow \left[ \Delta_{(\frac{1}{2}, \frac{1}{2})}^{(1, 1)} f \circ (g_1, g_2) \right] \left( \frac{1}{2}, \frac{1}{2} \right) = -1. \end{aligned}$$



Here is our **second main result**:

### Theorem 12.

Let  $f : [0, 1]^m \rightarrow \mathbb{R}_+$  be  $d$ - $\uparrow$  ( $d \geq 2$ ), and let  $g_1, \dots, g_m : \mathbb{R}^d \rightarrow [0, 1]$  be d.f.s of (subprobability) measures. Then, also  $f \circ (g_1, \dots, g_m)$  is a d.f. on  $\mathbb{R}^d$ .

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Idea of proof:

$$g := (g_1, \dots, g_m), \quad h := f \circ g$$

$h$  is right-continuous (since  $f$  is by Lemma 1).

To show:  $h$  is  $\mathbf{1}_d$ - $\uparrow$ !

Because of Theorem 11 we may assume  $f$  to be  $C^\infty$ .

##### Idea of proof:

- 1 Also  $g_1, \dots, g_d$  are  $C^\infty$ .

An explicit and rather complicated generalization of the usual multivariate chain rule and of Faà di Bruno's formula leads to the result (Constantine & Savits, TAMS 1996).



## Idea of proof:

- 1 Also  $g_1, \dots, g_d$  are  $C^\infty$ .

An explicit and rather complicated generalization of the usual multivariate chain rule and of Faà di Bruno's formula leads to the result (Constantine & Savits, TAMS 1996).

- 2 To show: for  $x \in \mathbb{R}^d$  and  $\xi \in \mathbb{R}_+^d$

$$\left(\Delta_\xi^{1^d} h\right)(x) = h(x + \xi) \mp \dots + (-1)^d h(x) \geq 0$$

$\exists C^\infty$  d.f.s  $\tilde{g}_1, \dots, \tilde{g}_m$  such that  $\tilde{g}_i|_B = g_i|_B \forall i \leq d$ , where  $B := \{x + \sum_{i \in \alpha} \xi_i e_i \mid \alpha \subseteq [d]\}$ .

$$\Rightarrow 0 \leq \Delta_\xi^{1^d}(f \circ \tilde{g})(x) = \left(\Delta_\xi^{1^d} h\right)(x). \quad \square$$

For  $d = 2$  this result was proved in 2011 (Klement et al., Inf. Sc.).

### Corollary 6.

Let  $m, d, k \in \mathbb{N}$ ,  $J \subseteq \mathbb{R}^m$  and  $I \subseteq \mathbb{R}^d$  intervals,

$g = (g_1, \dots, g_m) : I \rightarrow J$ ,  $f : J \rightarrow \mathbb{R}$ ,  $\mathbf{n} \in \mathbb{N}^d$ .

- (i) If each  $g_i$  is  $\mathbf{n}$ - $\uparrow$ , and  $f$  is  $|\mathbf{n}|$ - $\uparrow$ , then  $f \circ g$  is  $\mathbf{n}$ - $\uparrow$
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**Proof.**

(i) By Theorem 8 each  $g_i \circ \sigma_{\mathbf{n}}$  is  $\mathbf{1}_{|\mathbf{n}|}$ - $\uparrow$ , hence so is by Theorem 12

$$f \circ (g_1 \circ \sigma_{\mathbf{n}}, \dots, g_m \circ \sigma_{\mathbf{n}}) = (f \circ g) \circ \sigma_{\mathbf{n}},$$

and again Theorem 8 shows  $f \circ g$  to be  $\mathbf{n}$ - $\uparrow$ .

(ii) For any  $\mathbf{n} \in \mathbb{N}^d$    with  $|\mathbf{n}| \leq k$  each  $g_i$  is  $\mathbf{n}$ - $\uparrow$ ,

hence  $f \circ g$  is  $\mathbf{n}$ - $\uparrow$ . By Theorem 10  $f \circ g$  is  $k$ - $\uparrow$ . □

## 5. Two related notions of monotonicity

$$\nabla f := -\Delta f, \text{ i.e. } (\nabla_h f)(s) := f(s) - f(s + h)$$

### Definition 3.

$f : I \rightarrow \mathbb{R}$  is  $\mathbf{n}$ - $\downarrow$  (“ $\mathbf{n}$ -decreasing”) iff

$$(\nabla_h^{\mathbf{p}} f)(s) \geq 0 \quad \forall s \in I, \quad h \in \mathbb{R}_+^d, \quad 0 \preceq \mathbf{p} \leq \mathbf{n}.$$

And  $f$  is  $\mathbf{n}$ - $\updownarrow$  (“ $\mathbf{n}$ -alternating”) iff

$$(\nabla_h^{\mathbf{p}} f)(s) \leq 0 \quad \forall s \in I, \quad h \in \mathbb{R}_+^d, \quad 0 \preceq \mathbf{p} \leq \mathbf{n}.$$

## Remarks.

- (a)  $f$  is  $\mathbf{n}$ - $\downarrow$  on  $I \Leftrightarrow f(-\cdot)$  is  $\mathbf{n}$ - $\uparrow$  on  $-I$
- (b)  $f$  is  $\mathbf{n}$ - $\updownarrow$  on  $I \Leftrightarrow -f(-\cdot)$  is  $\mathbf{n}$ - $\uparrow$  on  $-I$
- (c) A  $C^\infty$  function  $f$  is  $\mathbf{n}$ - $\downarrow$  iff  $(-1)^{|\mathbf{p}|} f_{\mathbf{p}} \geq 0 \forall 0 \preceq \mathbf{p} \leq \mathbf{n}$ , and  $f$  is  $\mathbf{n}$ - $\updownarrow$  iff  $(-1)^{|\mathbf{p}|} f_{\mathbf{p}} \leq 0$  instead.



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For  $d = 1$ :

$f$  is 2- $\uparrow \Leftrightarrow f$  is increasing and convex

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$f \geq 0$  is  $n\text{-}\updownarrow \forall n \in \mathbb{N}$  (" $\infty\text{-}\updownarrow$ ")  $\Leftrightarrow f$  is a Bernstein function

An easy consequence of Theorem 1 is

### Williamson's theorem.

If  $f : ]0, \infty[ \rightarrow \mathbb{R}_+$  is  $n$ - $\downarrow$ ,  $n \geq 2$ , then

$$f(s) = \int (1 - cs)_+^{n-1} d\mu(c)$$

where  $\mu$  is a measure on  $\mathbb{R}_+$ .

( $n$ - $\downarrow$  functions are often called “ $n$ -monotone”)

A (more recent) generalization reads:

If  $f : ]0, \infty[^d \rightarrow \mathbb{R}_+$  is  $\mathbf{n}$ - $\downarrow$ ,  $\mathbf{n} \geq \mathbf{2}_d$ , then

$$f(s) = \int \prod_{i=1}^d (1 - c_i s_i)_+^{n_i-1} d\mu(c)$$

with  $\mu$  a measure on  $\mathbb{R}_+^d$ .

An interesting appearance of 3- $\uparrow$  functions:

For  $x, y, z \in \mathbb{R}$  we have always

$$|x + y| + |y + z| + |z + x| \leq |x| + |y| + |z| + |x + y + z|$$

the so-called *Hornich-Hlawka inequality*. This can be generalized as follows:

Let  $f : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  be 3- $\uparrow$ , then  $\forall x, y, z \in \mathbb{R}$

$$f(|x+y|) + f(|y+z|) + f(|z+x|) \leq f(|x|) + f(|y|) + f(|z|) + f(|x+y+z|).$$

$f = \text{id}$  gives the original inequality, which also holds for vectors  $x, y, z$ . The above generalization for  $x, y, z \in \mathbb{R}^d$  can be shown for  $f(t) = \sqrt{t}$ ,  $f(t) = \sqrt[4]{t}$ ,  $f(t) = \sqrt[8]{t}$ , ..., but is open for other (Bernstein) functions.

## 5. Two related notions of monotonicity

The remarks given before are useful even in dimension one, as shown in the following

### Example.

$$\varphi(t) := -\log(1 - e^{-t}), \quad t \in ]0, \infty[$$

Then  $\varphi$  is *completely montone*, i.e.  $n$ - $\downarrow$  for each  $n \in \mathbb{N}$ , and this was shown in an article from 2018 by using so-called Eulerian numbers (of permutations). It follows also from

$$\varphi(-\cdot) = \underbrace{[-\log(1 - \cdot)]}_{n\uparrow} \circ \underbrace{\exp}_{n\uparrow} \quad (\text{on } ]-\infty, 0[)$$

using the Bernstein function  $\log(1 + t)$ :

$\varphi(-\cdot)$  is  $n$ - $\uparrow$  as composition of two such functions, hence  $\varphi$  is  $n$ - $\downarrow \forall n \in \mathbb{N}$ .

Of special importance is again  $\mathbf{n} = \mathbf{1}_d$ . Non-negative  $\mathbf{1}_d$ - $\downarrow$  functions are (essentially) *survival-functions*, i.e. of the form  $\mu([s, \infty])$  for some measure  $\mu$ .

Non-negative  $\mathbf{1}_d$ - $\uparrow$  are (essentially) *co-survival functions*, i.e. of the form  $\mu([s, \infty]^c)$  for some  $\mu$ .

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A particular subclass of the latter is of special interest:

A d.f.  $F$  on  $\mathbb{R}_+^d$  is called a *simple multivariate extreme value distribution* iff

$$(F(tx))^t = F(x) \quad \forall x \in \mathbb{R}_+^d, \forall t > 0$$

and if  $F$  has *standard Fréchet margins*, defined by the (one-dimensional) d.f.  $\exp\left(-\frac{1}{u}\right)$  for  $u > 0$ . Then  $F(x) = 0$  if  $x_i = 0$  for some  $i$ , and  $0 < F(x) < 1$  else.

$$f(x) := -\log F\left(\frac{1}{x}\right) \quad \text{where } \frac{1}{x} := \left(\frac{1}{x_1}, \frac{1}{x_2}, \dots\right)$$

is called a *stable tail dependence function* (STDF). It is a function  $f : \mathbb{R}_+^d \rightarrow \mathbb{R}_+$  with the properties

- (i)  $f$  is homogeneous, i.e.  $f(tx) = tf(x) \quad \forall t > 0, \forall x$
- (ii)  $f(e_i) = 1 \quad \forall i = 1, \dots, d$
- (iii)  $\max_{i \leq d} x_i \leq f(x) \leq \sum_{i=1}^d x_i$
- (iv)  $f$  is convex

but this is a full characterization of STDFs only for  $d = 2$ . In higher dimensions, this had been an open problem for some time.



The answer I could give is as follows:

### Theorem 13.

A function  $f : \mathbb{R}_+^d \rightarrow \mathbb{R}$  is a STDF if and only if  $f$  is homogeneous,  $\mathbf{1}_d$ - $\updownarrow$ , and  $f(e_1) = \dots = f(e_d) = 1$ .

In this case  $f$  is the co-survival function of a homogeneous Radon measure  $\mu$  on  $[0, \infty]^d \setminus \{\infty\}$ , i.e.

$$f(x) = \mu \left( [x, \infty]^c \right) \quad , \quad x \in \mathbb{R}_+^d.$$

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$$f(x) = \mu\left([x, \infty]^d\right), \quad x \in \mathbb{R}_+^d.$$

In this case  $f$  has the unique integral representation

$$f(x) = f(\mathbf{1}_d) \cdot \int \max_{i \leq d}(c_i x_i) d\nu(c),$$

$\nu$  being a probability measure on  $\{c \in \mathbb{R}_+^d \mid \max_{i \leq d} c_i = 1\}$ .

### Examples.

- The classical norms  $f_p(x) := \left(\sum_{i=1}^d x_i^p\right)^{\frac{1}{p}}$  for  $p \geq 1$ , up to  $f_\infty(x) := \max_{i \leq d} x_i$  (GUMBEL)
- $f(x, y) = \frac{x^2 + xy + y^2}{x + y}$  on  $\mathbb{R}_+^2$
- $\sum_i x_i - \sum_{i < j} (x_i^p + x_j^p)^{\frac{1}{p}} \pm \dots + (-1)^{d-1} (\sum_i x_i^p)^{\frac{1}{p}}$  for  $p < 0$  (GALAMBOS)

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Which (univariate) functions preserve  $\mathbf{n}\text{-}\downarrow$  (multivariate) functions?

Answer: exactly those preserving  $\mathbf{n}\text{-}\uparrow$  ones, because of Remark (a).

And which ones preserve  $\mathbf{n}\text{-}\updownarrow$  functions:

- $g \mathbf{n}\text{-}\updownarrow, f \mathbf{|n|}\text{-}\updownarrow \Rightarrow f \circ g$  is  $\mathbf{n}\text{-}\updownarrow$

since  $-(f \circ g)(-\cdot) = [-f(-\cdot)] \circ [-g(-\cdot)]$ ,

- $g \mathbf{n}\text{-}\updownarrow, f \mathbf{|n|}\text{-}\downarrow \Rightarrow f \circ g$  is  $\mathbf{n}\text{-}\downarrow$

since  $f \circ g(-\cdot) = [f(-\cdot)] \circ [-g(-\cdot)]$ .

**An open problem:**

$$K := \{f : [0, 1]^2 \rightarrow \mathbb{R}_+ \mid f \text{ is } 2\text{-}\uparrow, f(1, 1) = 1\}$$

Then  $K$  is compact and convex, multiplicatively stable.

- Is  $K$  a Bauer simplex?
- Determine  $\text{ex}(K)$ !

I could prove that

$$f_c \circ (f_a \otimes f_b) \in \text{ex}(K)$$

$\forall a, b, c \in [0, 1] (f_a(t) = (t - a)_+ / (1 - a), f_1 := 1_{\{1\}}).$

Are there other extreme points?