

**Abstract tube associated with a perturbed  
polyhedron and multidimensional normal  
probability calculation**

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## 1. Introduction

- ▶ Multidimensional normal probability calculation
- ▶ An integration technique for simplicial cones

## 2. Abstract tubes

- ▶ Inclusion-exclusion
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- ▶ Lexicographic perturbation in an outer direction
- ▶ Abstract tube
- ▶ Construction of  $\mathcal{F}(\varepsilon)$
- ▶ Studentized range statistic — Numerical example

Summary

# 1. Introduction

## Multidimensional normal probability calculation

- ▶ A closed convex polyhedron in  $\mathbb{R}^n$ :

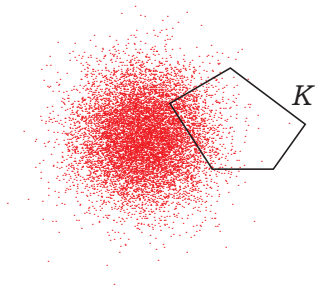
$$K = \{x \in \mathbb{R}^n \mid A^\top x \leq b\}$$

where

$$A = (a_1, \dots, a_m)_{n \times m}, \quad b = (b_1, \dots, b_m)_{m \times 1}^\top$$

- ▶ Let  $x \sim N_n(0, I_n)$ , i.e.,  $n$ -dim standard Gaussian vector. Our primary purpose is to calculate

$$P(K) := \Pr(x \in K)$$



## Multidimensional normal probability calculation (cont)

- ▶ Application — Multiple comparisons, e.g., distribution of

$$\Pr \left( \max_{1 \leq i < j \leq k} \frac{|x_i - x_j|}{\sqrt{\sigma_i^2 + \sigma_j^2}} \leq c \right)$$

where  $x_i \sim N(\mu_i, \sigma_i^2)$  independently.  
(Tukey's studentized range statistic)

## An integration technique for simplicial cones

- ▶ When the column vectors  $a_i$  of  $A = (a_1, \dots, a_m)$  are linearly independent,  $K$  is a cone.  
Precisely,  $K$  is a **simplicial cone**, or the **direct sum of a simplicial cone  $K_1$  and a linear subspace  $L$** , i.e.,  $K = K_1 \oplus L$ .



simplicial cone



Non-simplicial cone

- ▶ For such  $K$ , Miwa, Hayter and Kuriki (2003, JRSS, B) proposed an “**successive numerical integration technique**” based on Markov property for computing  $\Pr(x \in K)$ .
- ▶ Implemented as an R library **mvtnorm**
- ▶ Then, how to deal with the general case?

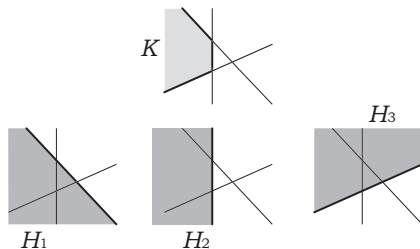
## 2. Abstract tubes

# Inclusion-exclusion

- ▶ Abstract tubes = “Polyhedral inclusion-exclusion identity”
- ▶ Rewrite the polyhedron  $K$  as the intersection of half spaces:

$$K = \bigcap H_i, \quad H_i = \{x \mid a_i^\top x \leq b\}$$

- ▶ Example:  $K = H_1 \cap H_2 \cap H_3$



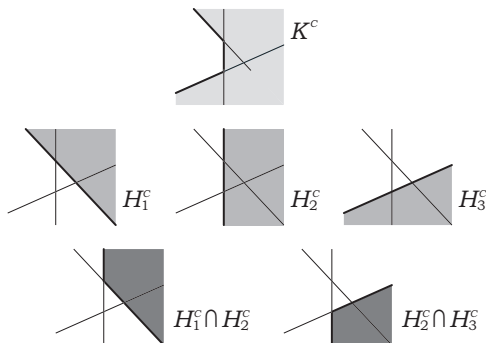


## Inclusion-exclusion (cont)

- ▶ Since  $K^c = H_1^c \cup H_2^c \cup H_3^c$ ,

$$1 - P(K) = P(H_1^c) + P(H_2^c) + P(H_3^c) \\ - P(H_1^c \cap H_2^c) - P(H_2^c \cap H_3^c)$$

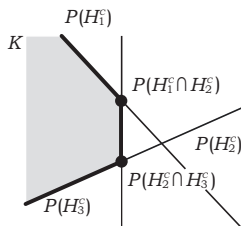
Note:  $-P(H_1^c \cap H_3^c) + P(H_1^c \cap H_2 \cap H_3^c) = 0$



- ▶  $H_i^c$ ,  $H_i^c \cap H_j^c$  are simplicial (or simplicial  $\oplus$  linear subspace).  
“Successive numerical integration technique” can be used.

## Inclusion-exclusion (cont)

- ▶  $P(H_1^c)$ ,  $P(H_2^c)$ ,  $P(H_3^c)$  correspond to the 3 edges of  $K$ .  
 $P(H_1^c \cap H_2^c)$  and  $P(H_2^c \cap H_3^c)$  correspond to the 2 vertices of  $K$ .



- ▶ Note: The identity holds for any probability measure  $P(\cdot)$ .  
Not only for Gaussian probability.  
(e.g., discrete distribution)

## Inclusion-exclusion (cont)

- ▶  $H_i^c, H_i^c \cap H_j^c$  are not necessarily simplicial (or simplicial  $\oplus$  linear subspace).
- ▶ Counter example (Pyramid in  $\mathbb{R}^3$ )

$$H_1 : -x_1 - x_2 + x_3 \leq 1$$

$$H_2 : -x_1 + x_2 + x_3 \leq 1$$

$$H_3 : +x_1 + x_2 + x_3 \leq 1$$

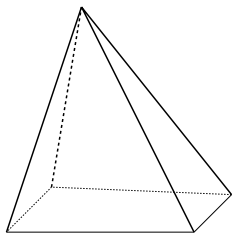
$$H_4 : +x_1 - x_2 + x_3 \leq 1$$

$$\begin{aligned} 1 - P(K) &= P(H_1^c \cup H_2^c \cup H_3^c \cup H_4^c) \\ &= P(H_1^c) + P(H_2^c) + \dots - P(H_1^c \cap H_2^c) + \dots \\ &\quad - P(H_1^c \cap H_2^c \cap H_3^c \cap H_4^c) \end{aligned}$$

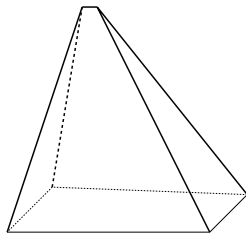
The last term is not simplicial.

## Inclusion-exclusion (cont)

- ▶ This difficulty comes from the fact that:  
4 facets in  $\mathbb{R}^3$  are not in general position.



Pyramid  
(not in general position)



Perturbed pyramid  
(in general position)

## Weak Abstract tube (Naiman & Wynn)

- ▶ Perturbed polyhedron

$$K(\epsilon\delta) = \{x \in \mathbb{R}^n \mid A^\top x \leq b + \epsilon\delta\}$$

where  $\epsilon \in \mathbb{R}$  and  $\delta = (\delta_1, \dots, \delta_n)^\top$  is a direction vector.

- ▶ Let  $\mathcal{F}(\epsilon\delta)$  be the set of all faces of  $K(\epsilon\delta)$ .

- ▶ **Proposition** (Naiman & Wynn, 1997)

(i) For a suitable  $\delta$  and for all  $\epsilon$  such that  $|\epsilon| \ll 1$ ,

$$1 - P(K) = \sum_{J \in \mathcal{F}(\epsilon\delta)} (-1)^{|J|-1} P\left(\bigcap_{i \in J} H_i^c\right)$$

for any continuous probability  $P(\cdot)$ . “Weak abstract tube”

(ii)  $|J| \leq n$

### 3. Our proposal and results

## Lexicographic perturbation in an outer direction

- ▶ We propose the use of the following lexicographic perturbation vector

$$\varepsilon = (\varepsilon, \varepsilon^2, \dots, \varepsilon^n)^\top$$

where  $\varepsilon > 0$  is an **infinitesimal positive number**.

- ▶ Let  $\mathcal{F}(\varepsilon)$  be the set of all faces of the infinitesimally perturbed polyhedron  $K(\varepsilon)$ .

► **Proposition**

(i)

$$1 - P(K) = \sum_{J \in \mathcal{F}(\varepsilon)} (-1)^{|J|-1} P\left(\bigcap_{i \in J} H_i^c\right)$$

for all probability measure  $P(\cdot)$ . “Abstract tube” in the strict sense.

(ii)  $|J| \leq \text{rank}(A)$

(iii)  $\bigcap_{i \in J} H_i^c = K_1 \oplus L$ , where

$K_1$  :  $|J|$ -dim simplicial cone

$L$  :  $(n - |J|)$ -dim linear subspace



## Construction of $\mathcal{F}(\varepsilon)$

- ▶ To construction of  $\mathcal{F}(\varepsilon)$ , for each subset  $J \subset \{1, \dots, m\}$ , determine the feasible (existence of a solution) of the system:

$$a_i^\top x - (b_i + \varepsilon^i) = 0 \quad (i \in J)$$

$$a_i^\top x - (b_i + \varepsilon^i) \leq 0 \quad (i \notin J)$$

If a solution exists, then  $J \in \mathcal{F}(\varepsilon)$ . Conducted by the linear programming (LP).

- ▶ The term  $b_i + \varepsilon^i$  is treated as a polynomial in  $\varepsilon$ .

For

$$f(\varepsilon) = \sum c_i \varepsilon^i \quad \text{and} \quad g(\varepsilon) = \sum d_i \varepsilon^i$$

let

$$f(\varepsilon) \geq g(\varepsilon) \quad \Leftrightarrow \quad (c_0, c_1, \dots, c_n) \geq (d_0, d_1, \dots, d_n)$$

(lexicographically)

- ▶ This LP is called **lexicographic method**.

## Studentized range statistic — Numerical example

- ▶ The polyhedron defined by the studentized range statistic is

$$K = \left\{ x \in \mathbb{R}^k \mid \frac{|\sigma_i x_i - \sigma_j x_j|}{\sqrt{\sigma_i^2 + \sigma_j^2}} \leq c, \forall i < j \right\}$$

- ▶ In the balanced case  $\sigma_1^2 = \dots = \sigma_k^2$ , because

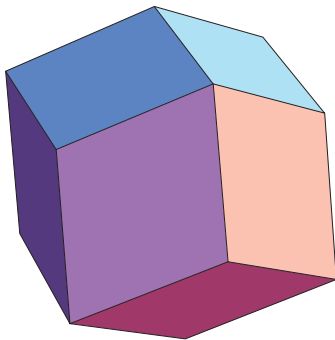
$$\frac{x_1 - x_2}{\sqrt{2}} = c \quad \text{and} \quad \frac{x_2 - x_3}{\sqrt{2}} = c \quad \text{and} \quad \frac{x_3 - x_4}{\sqrt{2}} = -c$$

$$\text{imply} \quad \frac{x_1 - x_4}{\sqrt{2}} = c$$

The facets of  $K$  are not in general position.

## Studentized range statistic (cont)

►  $K_1 = K \cap \{x \mid x_1 = \dots = x_k\}^\perp$



Studentized range polytope  $K_1$   
in the balanced case ( $k = 4$ )

## Studentized range statistic (cont)

- ▶ Number of terms in the abstract tube  $|\mathcal{F}(\varepsilon)|$

$k$	2	3	4	5	6
$m = k(k - 1)$	2	6	12	20	30
$ \mathcal{F}(\varepsilon) $	2	12	62	320	1682

$m$  is the number of facets (inequalities)

- ▶ Although the lexicographic perturbation depends on the order,  $|\mathcal{F}(\varepsilon)|$  looks unchanged even if the order of inequalities is changed.

## Studentized range statistic (cont)

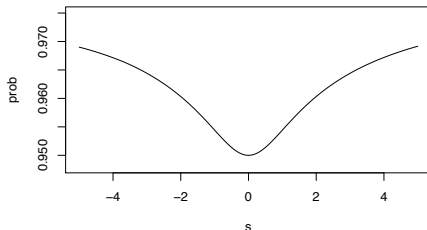
- ▶ Hayter (1984) proved the Tukey-Kramer conjecture that

$$\min_{\sigma_1, \dots, \sigma_k} \Pr(x \in K)$$

is attained iff  $\sigma_1 = \dots = \sigma_k$ .

- ▶  $k = 5$ ,

$$(\sigma_i^2) = \left(1, 10^{\frac{1}{4}s}, 10^{\frac{2}{4}s}, 10^{\frac{3}{4}s}, 10^s\right) \quad (s = 0 \Leftrightarrow \sigma_i^2 \equiv 1)$$



## Summary

- ▶ We discussed a method for computing  $\Pr(x \in K)$ , where  $x \sim N_n(0, I_n)$  and  $K$  is any convex polyhedron.
- ▶ We proposed the use of the abstract tube with a lexicographic perturbation in an outer direction.
  - (i) The lexicographic method of LP is useful in the construction.
  - (ii) Each term is simplicial and Miwa, et al. (2003)'s “successive numerical integration technique” can be used for its calculation.
  - (iii) The proposed abstract tube is applicable to any probability measures such as discrete distribution.
- ▶ Reference:  
S. Kuriki, T. Miwa, and A. J. Hayter (2011), arXiv:1110.2824