On the dual of the solvency cone

Andreas Löhne

Friedrich-Schiller-Universität Jena

Joint work with:

Birgit Rudloff (WU Wien)

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Simplest solvency cone example



$$\begin{pmatrix} \mathsf{Rupee} \\ \mathsf{Euro} \end{pmatrix} \text{-portfolios:} \qquad \begin{pmatrix} 130 \\ -1 \end{pmatrix} \rightarrow \begin{pmatrix} 0 \\ 0 \end{pmatrix} \\ \begin{pmatrix} -110 \\ 1 \end{pmatrix} \rightarrow \begin{pmatrix} 0 \\ 0 \end{pmatrix} \qquad K = \operatorname{cone} \left\{ \begin{pmatrix} 130 \\ -1 \end{pmatrix}, \begin{pmatrix} -110 \\ 1 \end{pmatrix} \right\}$$

$$price \text{ systems:} \qquad \begin{pmatrix} 1 \\ 130 \end{pmatrix} \begin{pmatrix} 1 \\ 110 \end{pmatrix} \\ \begin{pmatrix} 1/130 \\ 1 \end{pmatrix} \begin{pmatrix} 1/110 \\ 1 \end{pmatrix} \qquad K^+ = \operatorname{cone} \left\{ \begin{pmatrix} 1 \\ 130 \end{pmatrix}, \begin{pmatrix} 1 \\ 110 \end{pmatrix} \right\}$$

Solve a problem stated in

Bouchard, B., Touzi, N. (2000): Explicit solution to the multivariate super-replication problem under transaction costs, Ann. Appl. Probab.

> "'provide explicitly a generating family for the polar [or dual] cone [of K_d for d > 2]"

Basic facts about transportation problem

$$s_{1} > 0$$

$$s_{2} > 0$$

$$s_{2} > 0$$

$$s_{2} > 0$$

$$s_{3} > 0$$

$$A = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ -1 & -1 & 0 & -1 & 0 \\ 0 & 0 & -1 & 0 & -1 \end{pmatrix}$$

$$s_{3} > 0$$

$$Variables: x = \begin{pmatrix} x_{14} \\ x_{24} \\ x_{25} \\ x_{34} \\ x_{35} \end{pmatrix} \quad Neg. \ cost: \ c = \begin{pmatrix} c_{14} \\ c_{24} \\ c_{25} \\ c_{34} \\ c_{35} \end{pmatrix} \quad Supply \ s = \begin{pmatrix} s_{1} \\ s_{2} \\ s_{3} \\ s_{4} \\ s_{5} \end{pmatrix}$$

 $\max c^T x$ s.t. Ax = s, $x \ge 0$

Dual transportation problem

 $y_{1} \ge y_{4} + c_{14}$ $y_{2} \ge y_{4} + c_{24}$ $y_{2} \ge y_{5} + c_{25}$ $y_{3} \ge y_{4} + c_{34}$ $y_{3} \ge y_{5} + c_{35}$

(primal problem = feasibility problem)

c = 0

 $y_1 \ge y_4$

 $s^T y = 0$

Modified transportation problem

$$s_{1} > 0$$

$$s_{2} > 0$$

$$s_{2} > 0$$

$$s_{2} > 0$$

$$s_{3} > 0$$

$$A = \begin{pmatrix} \pi_{14} & 0 & 0 & 0 & 0 \\ 0 & \pi_{24} & \pi_{25} & 0 & 0 \\ 0 & 0 & 0 & \pi_{34} & \pi_{35} \\ -1 & -1 & 0 & -1 & 0 \\ 0 & 0 & -1 & 0 & -1 \end{pmatrix}$$

$$s_{3} > 0$$

$$Variables: x = \begin{pmatrix} x_{14} \\ x_{25} \\ x_{34} \\ x_{35} \end{pmatrix} \quad Neg. \ cost: c = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix} \quad Supply \ s = \begin{pmatrix} s_{1} \\ s_{2} \\ s_{3} \\ s_{4} \\ s_{5} \end{pmatrix}$$

 $\max c^T x$ s.t. Ax = s, $x \ge 0$

Modified transportation problem (dual)

 $\min s^T y \quad \text{ s.t. } \quad A^T y \ge c$



 $\pi_{14} \cdot y_1 \ge y_4$ $\pi_{24} \cdot y_2 \ge y_4$ $\pi_{25} \cdot y_2 \ge y_5$ $\pi_{34} \cdot y_3 \ge y_4$ $\pi_{35} \cdot y_3 \ge y_5$ $s^T y = 0$

Definition (solvency cone)

Let $d \in \{2, 3, \ldots\}$, $V = \{1, \ldots, d\}$ and let $\Pi = (\pi_{ij})$ be a $(d \times d)$ -matrix such that

 $\forall i \in V : \quad \pi_{ii} = 1, \tag{1}$

$$\forall i, j \in V : \quad 0 < \pi_{ij}, \tag{2}$$

$$\forall i, j, k \in V : \quad \pi_{ij} \le \pi_{ik} \pi_{kj}, \tag{3}$$

$$\exists i, j, k \in V : \quad \pi_{ij} < \pi_{ik} \pi_{kj}. \tag{4}$$

Sometimes, (3) and (4) is replaced by

$$\forall i, j \in V, \ \forall k \in V \setminus \{i, j\} : \quad \pi_{ij} < \pi_{ik} \pi_{kj}.$$
(5)

The polyhedral convex cone

$$K_d := \operatorname{cone} \left\{ \pi_{ij} e^i - e^j | ij \in V \times V \right\}$$

is called solvency cone induced by Π .

The dual cone

$$K_d^+ := \left\{ y \in \mathbb{R}^d | \ \forall x \in K_d : \ x^T y \ge 0 \right\} \ \dots \ \text{(positive) dual cone of } K_d$$

Proposition 1. One has $K_d^+ = \{y \in \mathbb{R}^d | \forall i, j \in V : \pi_{ij}y_i \ge y_j\}$.

Proof: obvious Recall: $K_d := \operatorname{cone} \left\{ \pi_{ij} e^i - e^j | ij \in V \times V \right\}$

Proposition 2. One has $\mathbb{R}^d_+ \setminus \{0\} \subseteq \operatorname{int} K_d$ and $K^d_d \setminus \{0\} \subseteq \operatorname{int} \mathbb{R}^d_+$.

Proof: Follows from (1) to (4), a separation argument is used.

Proposition 3. One has $K_d \cap -\mathbb{R}^d_+ = \{0\}$.

Proof: Elementary.

Feasible tree solution

 $V = \{1, \ldots, d\}$ (P,N) ... bi-partition of V, i.e., $\emptyset \neq P \subsetneq V$, $N = V \setminus P$ G(P,N) ... bi-partite digraph with arc set $E = P \times N$

 $y \in \mathbb{R}^d$ is called generated by a tree T if T is a spanning tree of G(P, N) such that

$$\forall ij \in E(T) \subseteq P \times N : \ \pi_{ij}y_i = y_j > 0.$$
(6)

 $y \in \mathbb{R}^d$ is called feasible with respect to (P, N) if

$$\forall ij \in P \times N : \pi_{ij} y_i \ge y_j > 0.$$
(7)

y is called feasible tree solution w.r.t (P, N) if both properties hold.

Feasible tree solution

 $V = \{1, 2, 3, 4, 5, 6, 7\}, P = \{1, 2, 3, 4\}, N = \{5, 6, 7\}$



Tree solution: $\pi_{ij}y_i = y_j$ for $ij \in E(T)$

Feasible tree solution

 $V = \{1, 2, 3, 4, 5, 6, 7\}, P = \{1, 2, 3, 4\}, N = \{5, 6, 7\}$



Feasibility: e.g. $\pi_{37}y_3 \ge y_7$

Characterization of K_d^+

Theorem 1. For $y \in \mathbb{R}^d$, the following statements are equivalent. (i) y is an extremal direction of K_d^+ ; (ii) y is a feasible tree solution w.r.t. some bipartition (P, N) of V.

Questions:

Existence of extremal directions/feasible tree solutions

Construction of extremal directions/feasible tree solutions

Structure of extremal directions/feasible tree solutions

Degree vectors

$$\deg_T(P) = \begin{pmatrix} 1\\3\\1\\1 \end{pmatrix}$$

$$\frac{P}{1}$$

$$\frac{1}{3}$$



 $c \in \mathbb{N}^P$ is called *P*-configuration if $\sum_{i \in P} c_i = d - 1$ $b \in \mathbb{N}^N$ is called *N*-configuration if $\sum_{i \in N} b_i = d - 1$



 $c \in \mathbb{N}^P$ is called *P*-configuration if $\sum_{i \in P} c_i = d - 1$ $b \in \mathbb{N}^N$ is called *N*-configuration if $\sum_{i \in N} b_i = d - 1$



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 $c \in \mathbb{N}^P$ is called *P*-configuration if $\sum_{i \in P} c_i = d - 1$ $b \in \mathbb{N}^N$ is called *N*-configuration if $\sum_{i \in N} b_i = d - 1$

Existence of feasible tree solutions

Theorem 2. For every bi-partition (P, N) of V and every P-configuration $c \in \mathbb{N}^P$ there exists a feasible tree solution $y \in \mathbb{R}^d$ generated by a spanning tree T of the bi-partite graph G(P, N) with $\deg_T(P) = c$.

An analogous statement holds if an *N*-configuration is given.







 $k \in \arg \max\{y_j/\pi_{1j} \mid j \in N\}$

1 4



Is there an *N*-configuration $b \in \mathbb{N}^N$ and a feasible tree solution *y* generated by *T* such that $b = \deg_T(N)$ and $c = \deg_T(P)$?

1 (4)









 $k \in \arg\min\{y_i \cdot \pi_{ij} \mid i \in P\}$

Remaining question:



Given a *P*-configuration $c \in \mathbb{N}^P$.

Is there an *N*-configuration $b \in \mathbb{N}^N$ and a feasible tree solution y generated by T such that $b = \deg_T(N)$ and $c = \deg_T(P)$?

 $\mathcal{T}(H)$... set of all spanning trees of a graph H

Lemma 1. Let H = H(P, N) be a bi-partite graph. Then

 $|\{\deg_T(P)| T \in \mathcal{T}(H)\}| = |\{\deg_T(N)| T \in \mathcal{T}(H)\}|.$



Sang-Il Oum \rightarrow Postnikov 2009 (about generalized permutohedra)

For a feasible tree solution y, define subgraph H(y) of G = G(P, N)

$$V(H(y)) := V(G), E(H(y)) := \{ ij \in P \times N | \pi_{ij}y_i = y_j \}$$

 $\mathcal{P}(y) := \{ \deg_T(P) | T \in \mathcal{T}(H(y)) \}$ $\mathcal{N}(y) := \{ \deg_T(N) | T \in \mathcal{T}(H(y)) \}$

Lemma 2. Let x, y be two feasible tree solutions such that $x \neq \alpha y$ for all $\alpha > 0$. Then

$$\mathcal{P}(x) \cap \mathcal{P}(y) = \emptyset$$
 and $\mathcal{N}(x) \cap \mathcal{N}(y) = \emptyset$.

Illustration of Lemma 1 and Lemma 2





Consequences of Theorem 1 and 2

Corollary 1. Assume that also (5) holds. Let x, y be two feasible tree solutions with respect to bi-partitions (P_x, N_x) and (P_y, N_y) of V, respectively. Then $(P_x, N_x) \neq (P_y, N_y)$ implies $x \neq \alpha y$ for all $\alpha > 0$. Moreover, K_d^+ has at least $2^d - 2$ extremal directions.

Corollary 2.
$$K_d^+$$
 has at most $\sum_{p=1}^{d-1} {d-2 \choose p-1} {d \choose p}$ extremal directions.

Example. The upper bound in Corollary 2 cannot be improved.

Let the non-diagonal entries be pairwise different prime numbers such that

$$\left(\min\left\{\pi_{ij} \mid ij \in V \times V, i \neq j\right\}\right)^2 > \max\left\{\pi_{ij} \mid ij \in V \times V, i \neq j\right\}$$

Example. d = 20, $\pi_i i = 1$, $\pi_{12} = 59$, $\pi_{12} = 61 \dots \pi_{20,19} = 2713$

$$59^2 > 2713 \implies (5)$$

 K_{20}^+ has exactly $\sum_{p=1}^{19} {18 \choose p-1} {20 \choose p} = 35.345.263.800$ extremal directions.

 $P = \{5, 6, 7, 8, 9, 10, 11\}, N = \{1, \dots, 4, 12, \dots, 20\}.$

 $\binom{d-2}{p-1} = \binom{18}{6} = 18564 P$ -configurations for this bi-partition (p := |P|).

$$c = (3, 2, 4, 2, 2, 2, 4)^T \in \mathbb{N}^P$$

Algorithm (Matlab, about 15 minutes):

 $y = \left(\frac{487 \cdot 757}{503 \cdot 859}, \frac{491 \cdot 757}{503 \cdot 859}, \frac{619 \cdot 947 \cdot 1367}{677 \cdot 953 \cdot 1427}, \frac{757}{859}, \frac{757}{503 \cdot 859}, \frac{947 \cdot 1367}{677 \cdot 953 \cdot 1427}, \frac{1}{859}, \frac{1367}{953 \cdot 1427}, \frac{1}{859}, \frac{1367}{953 \cdot 1427}, \frac{1}{109}, \frac{1}{953 \cdot 1427}, \frac{1}{1427}, \frac{1327}{1427}, \frac{947 \cdot 1367}{1427}, \frac{1367}{1427}, \frac{1373}{1427}, \frac{829}{859}, \frac{839}{859}, \frac{839 \cdot 1249}{859 \cdot 1237}, \frac{1109}{1117}, 1\right)^{T}$ $b = (1, 1, 1, 2, 1, 2, 2, 1, 1, 2, 1, 1, 3)^{T} \in \mathbb{N}^{N}$ Special case 1

$$\pi_{ii} := 1 \text{ and } \pi_{ij} := a_j/b_i \ (i \neq j), \\ 0 < b_i \le a_i \text{ for all } i \in V, \\ 0 < b_k < a_k \text{ for at least one } k \in V$$

$$\Rightarrow (1) \text{ to } (4)$$

Recursion formula

$$Y_{2} = \begin{pmatrix} a_{1} & b_{1} \\ b_{2} & a_{2} \end{pmatrix} \qquad Y_{d} = \begin{pmatrix} b_{1} & & a_{1} \\ Y_{d-1} & \vdots & Y_{d-1} & \vdots \\ & & b_{d-1} & & a_{d-1} \\ a_{d} & \dots & a_{d} & a_{d} & b_{d} & \dots & b_{d} & b_{d} \end{pmatrix}$$

Direct description

$$K_d^+ = \operatorname{cone} \left\{ y \in \mathbb{R}^d | (P, N) \text{ bi-part. of } V, \forall i \in P : y_i = b_i, \forall j \in N : y_j = a_j \right\}$$

•

Consequence

$$K_d^+$$
 has at most $2^d - 2$ extremal directions.

Special case 2

$$\pi_{ii} := 1 \text{ and } \pi_{ij} := a_j/b_i \ (i \neq j), \\ 0 < b_i < a_i \text{ for all } i \in V, \end{cases} \Rightarrow (1) \text{ to } (5)$$

The same as in special case 1, but now

 K_d^+ has exactly $2^d - 2$ extremal directions.

Special case 3

$$\pi_{ii} := 1 \text{ and } \pi_{ij} := a_j/b_i \ (i \neq j), \\ 0 < b_i \le a_i \text{ for all } i \in V, \\ 0 < b_k < a_k \text{ for at least one } k \in V \end{cases} \Rightarrow (1) \text{ to } (4)$$

 $b_k = a_k$ for some $k \in V$

Recursion formula (w.l.o.g. $a_1 = b_1 = 1$)

$$Y_2 = \begin{pmatrix} 1 & 1 \\ a_2 & b_2 \end{pmatrix} \qquad Y_d = \begin{pmatrix} Y_{d-1} & Y_{d-1} \\ a_d & \dots & a_d & b_d & \dots & b_d \end{pmatrix}$$

Direct description

$$K_d^+ = \operatorname{cone} \left\{ y \in \mathbb{R}^d | Q \subseteq V \setminus \{k\}, \forall i \in Q : y_i = b_i, \forall j \in V \setminus Q : y_j = a_j \right\}.$$

Consequence

 K_d^+ has at most 2^{d-1} extremal directions.

References

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