Risk allocation games are cooperative games that are used to attribute the risk of a financial entity to its divisions. In this paper, we extend the literature on risk allocation games by incorporating liquidity considerations. A liquidity policy specifies state-dependent liquidity requirements that a portfolio should obey. To comply with the liquidity policy, a financial entity may have to liquidate part of its assets, which is costly.

The definition of a risk allocation game under liquidity constraints is not straightforward, since the presence of a liquidity policy leads to externalities. We argue that the standard worst case approach should not be used here and present an alternative definition. We show that the resulting class of transferable utility games coincides with the class of totally balanced games. It follows from our results that also when taking liquidity considerations into account there is always a stable way to allocate risk.

1. Introduction

If a financial enterprise (bank, insurance company, investment fund, etc.) consists of several divisions (individuals, products, subportfolios, risk factors, etc.), not only is it important to measure the risk of the entire financial enterprise properly, but also to allocate this risk to the divisions using a proper risk allocation method.

Risk allocation has many applications: the division of capital reserves among business units by financial institutions; strategic decision making regarding new business lines; product pricing; performance measurement; the formation of risk limits; see Denault (2001), Kalkbrener (2005), Buch and Dorfleitner (2008), Homburg and Scherpereel (2008), Kim and Hardy (2009), Csóka et al. (2009) and Menchero and Davis (2011).

The various applications have the following question in common: how to allocate the risk of the financial enterprise over its constituents? A natural approach to answer this question comes from cooperative game theory. First, one defines a risk allocation game, the cooperative game where a player is a portfolio and the payoffs of a coalition are negatively related to the risk of the coalition's portfolio.

Next, the risk of a portfolio is measured by an appropriate method. In particular, we advocate to use coherent measures of risk (Artzner et al., 1999), which are defined by four axioms: monotonicity, subadditivity, positive homogeneity, and translation invariance. Csóka et al. (2007) show that these axioms are supported by a natural general equilibrium approach to measure risk.

Of these axioms, subadditivity is the most essential one, and captures the notion that the risk of the financial enterprise is at most as large as the sum of the risks of its divisions.

Finally, one can use one of the point-valued solution concepts in cooperative game theory like the Shapley value (Shapley, 1953) or the nucleolus (Schmeidler, 1969) to attribute risk to the players, or one of the set-valued solution concepts like the core (Gillies, 1959) to determine stable allocations of risk.

Liquidity is a major concern in financial markets. This is most certainly the case in times of liquidity crises (Brunnermeier and Pedersen (2008) list the Black Monday in 1987, the U.S.-Iraq war in 1990, the fall of LTCM in 1998, and the subprime crisis in 2007), but also in normal times the liquidity of an asset is of tremendous importance. In recognition of this fact, MSCI has recently introduced LiquidityMetrics, whereby the liquidity is estimated for a large range of assets (Acerbi and Szekeres, 2013).
Acerbi and Scandolo (2008) extend the axioms of coherent risk measures to incorporate two types of liquidity considerations. First, requirements on the composition of the portfolio, captured by the so-called liquidity policy and, second, the liquidity of asset markets as expressed by marginal demand curves, corresponding to the order books of the assets at a given future point in time.

In this paper, we incorporate liquidity considerations in risk allocation. Doing so is not straightforward, since the imposition of a liquidity constraint generates strong externalities among the divisions of a financial enterprise. Indeed, any division of the enterprise can liquidate some of its assets in order to satisfy the liquidity policy, and the more assets one division liquidates, the less assets have to be liquidated by other divisions. The standard approaches to define a function in characteristic function form in the presence of externalities are $\alpha$-effectiveness and $\beta$-effectiveness as suggested by Aumann (1961), thereby generalizing the two-player case treated by von Neumann and Morgenstern (1944) to the case with an arbitrary number of players.

The notion of $\alpha$-effectiveness defines the payoff of a coalition as the payoff it can achieve irrespective of the actions taken by its complement. In a sense, the coalition acts first in anticipation of the worst actions its complement can take. The notion of $\beta$-effectiveness is less stringent and defines the payoff of a coalition as what it can achieve for sure given the worst actions of its complement, so now the complementary coalition acts first.

We will argue that as soon as the liquidity considerations are non-trivial, the two standard approaches should not be applied. We also argue that the approach where the portfolio of the coalition’s complement is ignored, or equivalently, put equal to zero, is not satisfactory. Rather than taking a worst-case approach for the behavior of the complementary coalition, we will fix the portfolio holdings of the players outside the coalition to their initial values, thereby placing the burden of satisfying the liquidity policy entirely on the coalition itself, but not putting an extra burden because of adversary behavior of the coalition’s complement. The resulting cooperative games with transferable utility are called risk allocation games with liquidity constraints where, as we demonstrate in an example, adding liquidity considerations can lead to a strikingly different risk allocation.

Our main theorem claims that the class of risk allocation games with liquidity constraints is totally balanced, thereby generalizing the result for risk allocation games without liquidity constraints in Csóka et al. (2009). A direct consequence of this result is that risk allocation games with liquidity constraints have a non-empty core. Thus it follows that even when taking liquidity considerations into account, there is always a stable way to allocate risk.

It has been shown in Csóka et al. (2009) that any totally balanced game is generated by some risk allocation game without liquidity constraints. Since a risk allocation game without liquidity considerations results as a special case when the liquidity policy is trivial and assets are perfectly liquid, it holds that any totally balanced game is generated by some risk allocation game with liquidity constraints. We therefore obtain an equivalence between the class of risk allocation games with liquidity constraints and the class of totally balanced games.

The structure of the paper is as follows. In Section 2 we set up risk environments with liquidity considerations and in Section 3 we define risk allocation games with liquidity constraints. Section 4 contains our main theorem and Section 5 concludes.

2. Risk environments with liquidity considerations

Acerbi and Scandolo (2008) study coherent measures of risk in a framework where portfolios are subject to liquidity considerations. Csóka et al. (2009) study risk allocation games that are generated by coherent measures of risk in a set-up where liquidity considerations are absent. In this section we extend the analysis of Csóka et al. (2009) and we define risk environments that take liquidity considerations into account. We denote such risk environments by $(N, J, S, \pi, \theta, m, L, p)$.

The group of players in a risk environment is denoted by $N$, it consists of the $n$ divisions of a financial enterprise, and it is referred to as the firm. Each division holds cash as well as assets belonging to a set $J$. The initial portfolio $\theta^0 = (\theta^0_i, \theta^0_j) \in \mathbb{R} \times \mathbb{R}^J$ of division $i \in N$ shows the amounts of cash and assets held initially by division $i$.

The initial portfolio of the firm is given by the aggregate portfolio $\theta(N) = \sum_{i \in N} \theta^0_i$. We denote the space of portfolios by $P = \mathbb{R} \times \mathbb{R}^J$.

Cash has a number of special properties which we explain in the sequel. The addition of an amount of cash $a \in \mathbb{R}$ to a portfolio $p \in P$ is denoted by $p \oplus a$ and results in the portfolio $q \in P$ defined by $q_0 = p_0 + a$ and $q_j = p_j, j \in J$. The tuple of the initial portfolios of the various divisions is denoted by $\theta = (\theta^0_i)_{i \in N}$.

The future value of the initial portfolio is subject to uncertainty. One out of a set $S$ of possible states of nature materializes in the future, where state of nature $s \in S$ occurs with probability $\pi_s > 0$. Clearly, it holds that $\sum_{s \in S} \pi_s = 1$. The value of the initial portfolio in state $s$ depends on the order books for the various assets and the liquidity policy of the firm, both of which are allowed to be state dependent. We follow Çetin et al. (2004) and Jarrow and Protter (2005) in modeling the order book for asset $j$ in state $s$ by a marginal demand curve $m_j$. A function is càdlàg if it is right continuous with left limits and càdlàg if it is left continuous with right limits.

Definition 2.1. The marginal demand curve (MDC) for asset $j \in J$ in state $s \in S$ is given by the map $m_j^*: \mathbb{R} \setminus \{0\} \rightarrow \mathbb{R}$ satisfying

(i) $m_j^*(x) \geq m_j^*(x')$ if $x < x'$;

(ii) $m_j^*$ is càdlàg at $x < 0$ and càdlàg at $x > 0$.

The amount $m_j^*(x)$ for $x > 0$ expresses the marginal bids that have been made to buy an amount of asset $j$. Similarly, $m_j^*(x)$ for $x < 0$ represents the marginal asks that have been made for an amount of asset $j$ to be sold. We call $m_j^*(0^-)$ the best bid and $m_j^*(0^+)$ the best ask price of asset $j$.

The part of the MDC corresponding to negative (positive) values of $x$ corresponds to the demand (supply) of the asset by others. MDCs can therefore be used to represent both demand and supply side constraints in liquidity. Note that the MDC is not defined at zero, and continuity with left/right limits allows for jumps to arbitrarily low numbers for positive values of $x$ and to arbitrarily high numbers for negative values of $x$.

So far the issue of liquidity has not been considered at all in risk allocation games. Implicitly, it has been assumed that the MDCs are all flat, corresponding to perfectly liquid asset markets.

Definition 2.2. Asset $j \in J$ is perfectly liquid if for every $s \in S$, there is $c \in \mathbb{R}$ such that for all $x \in \mathbb{R} \setminus \{0\}$, $m_j^*(x) = c$.

Since the constant $c$ is allowed to depend on $s$, the price against which an asset can be bought or sold is allowed to be stochastic, even when the asset market is perfectly liquid.

The MDC can be used to calculate the liquidation value of a portfolio.

Definition 2.3. The liquidation mark-to-market value of a portfolio $p \in P$ in state $s \in S$ is defined by

$$\ell^s(p) = p_0 + \sum_{j \in J} \int_0^{p_j} m_j^*(x) dx. \tag{1}$$
The liquidation mark-to-market value of a portfolio equals the portfolio’s amount of cash plus the proceeds of selling all long positions minus the payments needed to close short positions. Acerbi and Scandolo (2008) prove the following result.

**Proposition 2.4.** For every \( s \in S \), the function \( \ell^s : P \mapsto \mathbb{R} \) is continuous and concave.

The set of portfolios attainable from some given portfolio \( p \in P \) in state \( s \in S \) is given by
\[
A^s(p) = \{ q \in P | q_0 = \ell^s(p_0, p_1 - q_1) \}.
\]
The portfolio \( q \in A^s(p) \) is obtained by liquidating the amounts \( p_1 - q_1 \) of assets in \( J \) and adding the proceeds to \( p_0 \).

**Proposition 2.5.** For every \( s \in S \), for every \( p \in P \), the set \( A^s(p) \) is closed.

**Proof.** Let \( (q^r)_{r \in \mathbb{N}} \) be a sequence in \( A^s(p) \) with limit \( q \in P \). We have to show that \( q \in A^s(p) \). For every \( r \in \mathbb{N} \), it holds that
\[
q^r_0 = \ell^s(p_0, p_1 - q^r_1).
\]
By Proposition 2.4 it holds that \( \ell^s \) is continuous, so
\[
q_0 = \ell^s(p_0, p_1 - q_1).
\]
It now follows from the definition that \( q \in A^s(p) \). \( \square \)

The liquidity policy (Acerbi and Scandolo, 2008) incorporates the requirements imposed by a regulator or the contractual obligations that have to be met, and specifies that the portfolio of the firm should belong to the set \( L^s \subseteq P \) in state of nature \( s \in S \). We denote \( L = \bigcup_{s \in S} L^s \). The state dependence of the liquidity policy enables us to model regulatory risk as well as short sale constraints which depend on market conditions as expressed in the prevailing MDC.

The literature on risk allocation games has so far ignored liquidity policies.

**Definition 2.6.** The liquidity policy is trivial if for every \( s \in S \) it holds that \( L^s = P \). Throughout the paper we make the following assumption on \( L \).

**Assumption 2.7.** For every \( s \in S \), it holds that

(i) \( L^s \) is closed and convex;
(ii) for every \( p \in L^s \), for every \( a > 0 \), \( p \oplus a \in L^s \);
(iii) \( A^s(\partial(N)) \cap L^s \neq \emptyset \).

The first two items in Assumption 2.7 are inherited from Acerbi and Scandolo (2008). Closedness as required in Assumption 2.7. (i) is a standard technical assumption. Convexity means that if two portfolios are acceptable, then so is their weighted average. Assumption 2.7. (ii) implies that it is always acceptable to have more cash. Assumption 2.7. (iii) guarantees that in every state there is a feasible choice to meet the requirements of the liquidity policy. Assumption 2.7 is clearly satisfied if the liquidity policy is trivial.

In case all assets are perfectly liquid, the liquidity policy is irrelevant. This is the case which has been studied in the existing literature on risk allocation games. Even if the liquidity policy is trivial, the liquidity of assets can matter if there is a bid-ask spread.

For a portfolio \( p \in P \), we denote the assets hold long by \( J^+(p) = \{ j \in J | p_j > 0 \} \) and the assets held short by \( J^-(p) = \{ j \in J | p_j < 0 \} \).

**Definition 2.8.** The uppermost mark-to-market value of a portfolio \( p \in P \) in state \( s \in S \) is defined by
\[
u^s(p) = p_0 + \sum_{j \in J^+(p)} m^s_j(0^+) p_j + \sum_{j \in J^-(p)} m^s_j(0^-) p_j.
\]

The uppermost mark-to-market value of a portfolio can be interpreted as the value of a portfolio in the long run. Long positions are valued using the best bid prices and short positions using the best ask prices. In case all asset markets are perfectly liquid, the uppermost mark-to-market value of a portfolio is equal to its liquidation mark-to-market value. Acerbi and Scandolo (2008) prove the following result.

**Proposition 2.9.** For every \( s \in S \), the function \( u^s : P \mapsto \mathbb{R} \) is continuous, concave, and positive homogeneous of degree one.

Two portfolios \( p, q \in P \) are said to be concordant if \( p_j q_j \geq 0 \) for all \( j \in J \). Two portfolios are concordant if there is no asset which is held long in one portfolio and short in the other. It is easily verified that \( u^s \) is additive for concordant portfolios. The concavity and positive homogeneity of \( u^s \) imply that \( u^s \) is superadditive on \( P \), i.e.,
\[
u^s(p) + u^s(q) \leq u^s(p + q) \text{ for all } p, q \in P.
\]

**Corollary 2.10.** For every \( s \in S \), the function \( u^s \) is superadditive on \( P \) and additive for concordant portfolios.

Given some state \( s \in S \), the firm might have to liquidate part of its assets to obtain a portfolio in \( L^s \). The initial portfolio’s value in state \( s \) is given by the highest attainable uppermost mark-to-market value satisfying the liquidity policy. These considerations lead us to the following definition.

**Definition 2.11.** The realization vector \( X(N) \in \mathbb{R}^L \) for the firm is equal to
\[
X^s(N) = \sup\{u^s(q)|q \in A^s(\partial(N)) \cap L^s, s \in S\}.
\]

Since \( A^s(\partial(N)) \cap L^s \neq \emptyset \) by Assumption 2.7, it follows that \( X^s(N) \) is not equal to \(-\infty\). We will show in Proposition 3.5 that \( X^s(N) \) is bounded from above by \( u^s(\partial(N)) \), so \( X^s(N) \) is finite.

The following example shows why in Eq. (3) we need a supremum rather than a maximum.

**Example 2.12.** Consider the case where \( \partial(N) = (\partial_0(N), \partial_1(N)) \in \mathbb{R}^2 \), so we have two assets, called 1 and 2, and cash. There is no uncertainty, so the cardinality of \( S \) is one. The liquidity policy \( L \) specifies that the asset portfolio \( q_0 \) satisfy \( q_0 \geq (\varepsilon_1, -1, -1) \) and \( (q_1 + 1)(q_2 + 1) \geq 1 \). Moreover, for simplicity assume there are no constraints on cash holdings. We assume the initial asset portfolio to be \( \partial(N) = (0, -1, -1) \) and we assume asset 2 to be perfectly liquid.

Trading in asset 1 involves liquidity costs. For instance, consider the case where there is a simple bid-ask spread, so for some \( \gamma > 0 \), \( m_1(x) = 1 + \gamma \text{ if } x < 0 \) and \( m_1(x) = 1 \text{ if } x > 0 \). Since we have to go to a portfolio of assets where holdings of both assets strictly exceed \(-1 \), we would like to buy \( \varepsilon > 0 \) of asset 1, buy \( 1/\varepsilon \) of asset 2, and go short in cash, resulting in a portfolio \( q(e) = (-1 + \gamma)e - m_2(0^-)/e, -1 + \varepsilon, -1 + 1/\varepsilon \). It is straightforward to verify that \( u(q(e)) = u(\partial(N) - xe) \). The uppermost mark-to-market value would be maximized, and liquidity costs would be minimized, by taking the smallest possible \( e \), something which clearly does not exist, so we need a supremum rather than a maximum in Eq. (3).

We will show in Proposition 3.6 that optimal portfolios exist under the mild and reasonable additional assumption that going infinitely short or long involves liquidity costs. Under such an additional assumption, one can use a maximum in Eq. (3).

Artzner et al. (1999) have introduced coherent measures of risk. A measure of risk is a function \( \rho : \mathbb{R}^+ \mapsto \mathbb{R} \), measuring the risk of a realization vector from the perspective of the present. It
corresponds to the minimal amount of cash the regulated agent has to add to his portfolio, and to invest in a reference instrument today, such that the risk involved in the portfolio is acceptable to the regulator. We assume that the reference instrument has payoff 1 in each state of nature at \( t = 1 \), thus its realization vector is 1^t = (1, \ldots, 1). It is most natural to think of it as a zero coupon bond. The price of the reference instrument can be thought of as the discount factor and is denoted by \( \delta \).

**Definition 2.13.** A function \( \rho : \mathbb{R}^3 \to \mathbb{R} \) is a coherent measure of risk if it satisfies the following axioms.

1. **Monotonicity:** for all \( X, Y \in \mathbb{R}^3 \) such that \( Y \geq X \), we have \( \rho(Y) \leq \rho(X) \).
2. **Subadditivity:** for all \( X, Y \in \mathbb{R}^3 \), we have \( \rho(X + Y) \leq \rho(X) + \rho(Y) \).
3. **Positive homogeneity:** for all \( X \in \mathbb{R}^3 \) and \( h \in \mathbb{R}_+ \), we have \( \rho(hX) = h\rho(X) \).
4. **Translation invariance:** for all \( X \in \mathbb{R}^3 \) and \( a \in \mathbb{R} \), we have \( \rho(X + aY) = \rho(X) - \alpha a \).

This completes the definition of a risk environment with liquidity considerations \((N, J, S, \pi, \theta, m, L, \rho)\).

### 3. Risk allocation games with liquidity constraints

A cooperative game with transferable utility consists of a set of players \( N \) and a value function \( v : \mathcal{P}^N \to \mathbb{R} \), which assigns to every coalition \( C \subseteq 2^N \) of players a worth \( v(C) \in \mathbb{R} \). By assumption it holds that \( v(\emptyset) = 0 \). An allocation is a vector \( y \in \mathbb{R}^N \) where \( y_i \) is the payoff of player \( i \in N \). An allocation \( y \) yields payoff \( y(C) = \sum_{i \in C} y_i \) to the members of coalition \( C \). An allocation \( y \in \mathbb{R}^N \) is called efficient if \( v(N) = v(N, \text{individually rational}) \) if \( y_i \geq v(i) \) for all \( i \in N \), and coherently rational if \( y(C) \geq v(C) \) for all \( C \subseteq 2^N \).

**Example 3.1.** Consider the case where we have one asset, \( J = \{1\} \), and no uncertainty. The liquidity policy \( L \) specifies that the firm’s portfolio \( q \) should satisfy \( q_1 \geq -1 \). The firm has two divisions, \( N = \{1, 2\} \), with identical initial portfolios given by \( \theta^1 = (1, -1) \) and \( \theta^2 = (0, -1) \). Consider the case where there is a simple bid-ask spread, so for some \( \gamma > 0 \), we have \( m_1(x) = 1 + \gamma x \) if \( x < 0 \) and \( m_1(x) = 1 \) if \( x > 0 \). Under \( \alpha \)-effectiveness, division 1’s realization is equal to

\[
X(\{1\}) = \sup \{ u \in \mathbb{R} \mid \exists q^1 \in A(\theta^1), \exists q^2 \in A(\theta^2), q^1 + q^2 \\
\in L \text{ and } u(q^1) \geq u \}.
\]

For \( i = 1, 2 \), we have that \( q^i \in A(\theta^i) \) if and only if

\[
[q^i_1 \geq -1 \text{ and } q^i_0 = -(q^i_1 + 1)(1 + \gamma)] \text{ or } [q^i_1 \leq -1 \text{ and } q^i_0 = -q^i_1].
\]

Since for any choice of \( q^1 \in A(\theta^1) \) there is \( q^2 \in A(\theta^2) \) such that \( q^1 + q^2 \not\in L \), in fact any \( q^2 \) such that \( q^2_1 < -q^1_1 - 1 \) would do, we find that \( X(\{1\}) = -\infty \). Notice that the same conclusion would follow even in the absence of a bid-ask spread.

The notion of \( \beta \)-effectiveness is less stringent and defines the payoff of a coalition as what it can achieve for sure given the worst actions of its complement, so now the complementary coalition acts first. We continue Example 3.1 by demonstrating that also \( \beta \)-effectiveness leads to undesirable consequences.

**Example 3.2.** Consider the primitives of Example 3.1. Under \( \beta \)-effectiveness, division 1’s realization is equal to

\[
X(\{1\}) = \sup \{ u \in \mathbb{R} \mid \exists q^1 \in A(\theta^1), \exists q^2 \in A(\theta^2), q^1 + q^2 \\
\in L \text{ and } u(q^1) \geq u \}.
\]

Consider some \( q^2 \in A(\theta^2) \) with \( q^2_1 \leq -1 \). To satisfy \( q^1 + q^2 \in L \), it should hold that \( q^1_1 \geq -q^2_1 - 1 \) and therefore that \( q^1_1 \geq 0 \). It can easily be computed that

\[
u(q^1) = -\gamma q^1_1 \leq -\gamma(q^2_1 + 1) = -\gamma q^1_1.
\]

Since \( q^2_1 \) can be chosen arbitrarily negative, we find that \( X(\{1\}) = -\infty \).

Both \( \alpha \)-effectiveness and \( \beta \)-effectiveness lead to undesirable properties of the value function, even in the simplest of examples. Both approaches share the feature that the complement of a coalition \( C \) is supposed to take the worst possible action. Rather than making such an extreme assumption, we will instead assume that the complement of coalition \( C \) remains inactive, so the burden of satisfying the liquidity policy will be put entirely on coalition \( C \). The portfolios which are attainable for coalition \( C \) in state \( s \in S \) are given by \( A^s(\theta(C)) \), where \( \theta(C) = \sum_{i \in C} \theta^i \). Inactivity of the complementary coalition means that those divisions stick to their initial portfolio, which equals \( \theta(N \setminus C) \) in the aggregate.

**Definition 3.3.** Given a risk environment with liquidity considerations \((N, J, S, \pi, \theta, m, L, \rho)\) and a coalition of divisions \( C \subseteq 2^N \), the realization vector \( X(C) \) of coalition \( C \) is defined by

\[
X(C) = \sup \{ u(q) \mid q \in A^s(\theta(C)) \text{ and } q + \theta(N \setminus C) \in L^1 \} \quad s \in S.
\]

When calculating \( X(C) \), we take the portfolios of the divisions outside the coalition as fixed, and liquidate the portfolios of the divisions in \( C \) in such a way that the resulting portfolio of the firm is attainable and satisfies the liquidity policy. Applying Definition 3.3 for the grand coalition \( N \) results in

\[
X(N) = \sup \{ u(q) \mid q \in A^s(\theta(N)) \cap L^1 \}.
\]

which is in accordance with Eq. (3).

The next result is useful in simplifying our expression for the realization vector of a coalition.
Proposition 3.4. For every $s \in S$, for every $C \in 2^N$, it holds that $q \in A^i(\theta_i(C))$ if and only if $q + \theta_i(N \setminus C) \in A^i(\theta_i(N))$.

Proof. The result follows from the observation that $q \in A^i(\theta_i(C))$
if and only if
$$q_0 = \ell^i(\theta_i(C), \theta_i(C) - q) = \theta_i(C) + \ell^i(0, \theta_i(C) - q)$$
if and only if
$$q_0 + \theta_i(N \setminus C) = \theta_i(C) + \ell^i(0, \theta_i(C) - q) + \theta_i(N \setminus C)$$
$$= \ell^i(\theta_i(N), \theta_i(N) - q_j - \theta_j(N \setminus C))$$
if and only if
$$q + \theta_i(N \setminus C) \in A^i(\theta_i(N)).$$

Proposition 3.4 shows that we can compute the realization vector of coalition $C$ as

$$X^i(C) = \sup\{u^i(q) | q + \theta_i(N \setminus C) \in A^i(\theta_i(N)) \cap L^i\}, \quad s \in S. \quad (4)$$

The next result provides bounds on the value of $X^i(C)$, implying that this value is not equal to $-\infty$ or $+\infty$.

Proposition 3.5. For every $s \in S$, for every $q \in A^i(\theta_i(N)) \cap L^i$, it holds that

$$u^i(q - \theta_i(N \setminus C)) \leq X^i(C) \leq u^i(q) \cap C \in 2^N.$$  

Proof. Consider some $q \in A^i(\theta_i(N)) \cap L^i$, where the latter set is non-empty by Assumption 2.7. Using (4), it follows that $u^i(q - \theta_i(N \setminus C)) \leq X^i(C)$. Next, consider some $q \in A^i(\theta_i(C))$. It holds that

$$q_0 = \ell^i(\theta_i(C), \theta_i(C) - q_j) = \theta_i(C) + \sum_{j \in J} \int_0^{\theta_j(C) - q_j} m_j^i(x)dx.$$  

It follows that

$$u^i(q) = \theta_i(C) + \sum_{j \in J^i(q)} \int_0^{\theta_j(C) - q_j} m_j^i(x)dx$$
$$+ \sum_{j \in J^i(q)} \int_0^{\theta_j(C) - q_j} m_j^i(x)dx + m_j^i(0^+)q_j$$
$$\leq \theta_i(C) + \sum_{j \in J^i(q)} \int_0^{\theta_j(C) - q_j} m_j^i(x)dx + m_j^i(0^+)q_j.$$  

Notice that irrespective of the sign of $\theta_j(C) - q_j$ it holds that

$$\int_0^{\theta_j(C) - q_j} m_j^i(x)dx \leq m_j^i(0^+)(\theta_j(C) - q_j)$$
and
$$\int_0^{\theta_j(C) - q_j} m_j^i(x)dx \leq m_j^i(0^+)(\theta_j(C) - q_j).$$

Consider some $j \in J^i(q)$. If $\theta_j(C) > 0$, then

$$\int_0^{\theta_j(C) - q_j} m_j^i(x)dx + m_j^i(0^+)q_j \leq m_j^i(0^+)(\theta_j(C) - q_j) + m_j^i(0^+)q_j$$
$$= m_j^i(0^+)(\theta_j(C)).$$  

If $\theta_j(C) < 0$, then

$$\int_0^{\theta_j(C) - q_j} m_j^i(x)dx + m_j^i(0^+)q_j \leq m_j^i(0^+)(\theta_j(C) - q_j) + m_j^i(0^+)q_j$$
$$= m_j^i(0^+)(\theta_j(C)).$$  

Consider some $j \in J^i(q)$. If $\theta_j(C) > 0$, then

$$\int_0^{\theta_j(C) - q_j} m_j^i(x)dx + m_j^i(0^+)q_j \leq m_j^i(0^+)(\theta_j(C) - q_j) + m_j^i(0^+)q_j$$
$$= m_j^i(0^+)(\theta_j(C)).$$  

We therefore find that

$$u^i(q) \leq \theta_i(C) + \sum_{j \in J^i(q)} m_j^i(0^+)\theta_j(C) + \sum_{j \in J^i(q)} m_j^i(0^+)\theta_j(C)$$
$$= \theta_i(C) + m_j^i(0^+)\theta_j(C).$$  

which completes the proof. □

If the liquidity policy is trivial, then it holds that $\theta_i(N) \in A^i(\theta_i(N)) \cap L^i$, so Proposition 3.5 gives $X^i(C) = u^i(\theta_i(N))$. If, moreover, there are no bid-ask spreads, it holds that $X^i(C) = \sum_n X^i(\{i\})$, the case which has been studied in the literature so far. In case the liquidity policy is trivial, but assets are not perfectly liquid and there is a bid-ask spread, it is still the case that the realization of division $i \in N$ in state $s$ is given by $X^i(\{i\}) = u^i(\theta_i(N))$, but it is no longer necessarily the case that the realization vector of a coalition of divisions $C \in 2^N$ is given by $X^i(C) = \sum_{i \in C} X^i(\{i\})$ as now coalitions can save on costs related to bid-ask spreads by combining their portfolios.

We show next that the supremum in Eq. (4) can be replaced by a maximum under the mild condition that going infinitely short or long in an asset involves liquidity costs.

Proposition 3.6. Assume that for every $s \in S$, for every $j \in J$, there is $x^*$ such that $m_j^i(x^*) > m_j^i(0^+)$ and there is $x^*$ such that $m_j^i(x^*) < m_j^i(0^+)$. Then for every $C \in 2^N$ it holds that

$$X^i(C) = \max\{u^i(q) | q + \theta_i(N \setminus C) \in A^i(\theta_i(N)) \cap L^i\}, \quad s \in S.$$  

Proof. Consider some $s \in S$ and let $(q^n_{j \in C})_{n \in \mathbb{N}}$ be a sequence such that $(q^n_{j \in C} + \theta_i(N \setminus C) \in A^i(\theta_i(N)) \cap L^i$ and $u^i(q^n_{j \in C})$ converges to $X^i(C)$. Suppose there is an asset $j \in J$ such that for some appropriately chosen subsequence, $\lim_{n \to \infty} (q^n_{j \in C}) = -\infty$. We will derive a contradiction. By repeating the steps in the proof of Proposition 3.5, we derive that

$$u^i(q^n_{j \in C}) \leq \theta_i(C) + \sum_{j \in J \setminus C} m_j^i(0^+)\theta_j(C) + \sum_{j \in J \setminus C} m_j^i(0^+)\theta_j(C)$$
$$+ \sum_{j \in J \setminus C} \int_0^{\theta_j(C) - q_j} m_j^i(x)dx + m_j^i(0^+)q_j$$
$$\leq \theta_i(C) + \sum_{j \in J \setminus C} m_j^i(0^+)\theta_j(C) + \sum_{j \in J \setminus C} m_j^i(0^+)\theta_j(C)$$
$$+ \sum_{j \in J \setminus C} \int_0^{\theta_j(C) - q_j} m_j^i(x)dx + m_j^i(0^+)q_j,$$

where the right-hand side tends to $-\infty$ if $r \to \infty$ since $m_j^i(0^+) > m_j^i(x^*)$. This contradicts the conclusion of Proposition 3.5, which establishes that $X^i(C)$ is bounded from below.

We have shown that for every $j \in J$ there is no subsequence of $(q^n_{j \in C})_{n \in \mathbb{N}}$ such that $\lim_{n \to \infty} (q^n_{j \in C}) = -\infty$. By the straightforward analogous argument, we can show that for every $j \in J$ there is no subsequence of $(q^n_{j \in C})_{n \in \mathbb{N}}$ such that $\lim_{n \to \infty} (q^n_{j \in C}) = \infty$. It follows that the sequence $(q^n_{j \in C})_{n \in \mathbb{N}}$ is bounded. By continuity of $e^r$ on $P$ as asserted in Proposition 2.4, it follows that the sequence $(q^n_{j \in C})_{n \in \mathbb{N}}$ is bounded. Without loss of generality, the sequence $(q^n_{j \in C})_{n \in \mathbb{N}}$ can be assumed to converge to some $q \in P$. The set $A^i(\theta_i(N)) \cap L^i$ is closed as the intersection of two sets which are closed by Proposition 2.5 and
Assumption 2.7. It follows that \( q + \theta(N \setminus C) \in A^i(\theta(N)) \cap L^1 \). By Proposition 2.9 it holds that \( u^i \) is continuous, so \( X^i(C) = u^i(q) \).

Since the existing literature on risk allocation assumes perfect liquidity, and we would like to incorporate perfect liquidity as a special case, we refrain from assumptions additional to Assumption 2.7, and continue with the formulation of Eq. (4) involving a supremum.

Definition 3.7. Given a risk environment with liquidity considerations \((N, J, S, \pi, \theta, m, L, \rho)\), the risk allocation game with liquidity constraints is the game \((N, v)\), where the value function \( v : 2^N \to \mathbb{R} \) is defined by

\[
v(C) = -\rho(X(C)), \quad C \in 2^N. \tag{5}
\]

Let \( \Gamma_n \) denote the family of risk allocation games with liquidity constraints with set of players \( N \). In such a game, according to Eq. (5), the larger the risk of any subset of portfolios, the lower its worth.

Example 3.8. Consider a firm with \( n = 2 \) divisions, where each division has invested into one asset and cash, \( J = \{1\} \). The portfolios of the divisions are \( \theta^1 = (20, -6) \) and \( \theta^2 = (26, -7) \), so both divisions have short positions in the risky asset. We assume \( S = (1, 2) \) with both states having equal probability of occurrence.

In state 1 the MDC of the risky asset is given by

\[
m_1^1(x) = \begin{cases} 
5 & \text{if } x < 0, \\
4 & \text{if } x > 0.
\end{cases}
\]

In state 2 the MDC of the risky asset is given by

\[
m_2^1(x) = \begin{cases} 
10 & \text{if } x < -2, \\
9 & \text{if } -2 \leq x < -1, \\
8 & \text{if } -1 \leq x < 0, \\
7 & \text{if } 0 \leq x < 1, \\
6 & \text{if } 1 \leq x.
\end{cases}
\]

We consider the deterministic liquidity policy specified by

\[
L^1 = L^2 = \{ p \in P | p_1 \geq -10 \}.
\]

In both states the liquidity policy of the firm does not allow to short the risky asset by more than 10 units. According to the initial portfolios, the two divisions together are \( 6 + 7 = 13 \) units short in the risky asset. We set \( \delta = 1 \) and take the maximum loss as the coherent measure of risk, so \( \rho(X) = \max_{x \in S} -X \).

First, let us consider state 1. This element of the realization vector of division 1 is calculated as follows. Looking at the MDC of the risky asset in state 1, we see that the firm can buy the risky asset for a price of 5, and it can sell more risky assets for a price of 4. Since currently the two divisions together are shorting 13 units from the risky asset and the liquidity policy allows to short at most 10 units, at least 3 units should be bought. If 3 units are bought for a price of \( 5 = 3 \times 5 = 15 \), division 1 ends up with the portfolio \( q^1(1) = (5, -3) \) in state 1 by trading \( t^1(1) = -15, 3 \). The uppermark mark-to-market value of \( q^1(1) \) is \( u^1(q^1(1)) = 5 - 3 \times 5 = -10 \). It is easy to check that we get the same value if division 1 buys anywhere between 3 to 6 units from the risky asset. Due to the bid-ask spread, buying back more than 6 units would result in a loss of 1 for each additional unit, since long positions would be valued at 4 per unit. It follows that the realization vector of division 1 in state 1 is \( X^1(1) = -10 \). Similarly, the realization vector of division 2 in state 1 is \( X^2(1) = -9 \). For the firm itself, we have \( \theta^1 + \theta^2 = (46, -13) \), and an optimal portfolio \( q^1(1, 2) \) satisfying the liquidity constraint is any convex combination of the portfolios \( (31, -10) \) and \( (-19, 0) \), resulting in \( X^1(1, 2) = -19 \).

Next, let us analyze state 2. Again, division 1 should buy at least 3 units of the risky asset to satisfy the liquidity policy. Buying 3 units and trading \( t^1(1) = -27, 3 \) will result in \( q^1(1) = (7, -3) \) and \( u^1(q^1(1)) = -7 - 3 \times 8 = -31 \). Buying more than 3 units would be costly. For instance, buying 4 units by trading \( t^1(1) = -37, 4 \) would result in \( q^1(1) = (17, -2) \) and \( u^1(q^1(1)) = -17 - 2 \times 8 = -33 \), hence division 1 will only buy 3 units. The remaining short position will be valued at the best ask price, and \( X^1(2) = -31 \). Similarly, the realization vector of division 2 in state 2 is \( X^2(2) = -33 \). For the firm itself, since \( \theta^1 + \theta^2 = (46, -13) \), the optimal trade is \( t^1(2) = -27, 3 \), leading to \( q^1(1, 2) = (19, -10) \) and \( X^1(1, 2) = u^1(q^1(1, 2)) = 19 - 10 \times 8 = -61 \). The calculations and the resulting cooperative game are summarized in Table 1.

Note that in state 1 the realization vector of coalition \( \{1, 2\} \) is additive over its members’ realization vectors, which is due to the flat MDCs and the concordant portfolios. In state 2 coalition \( \{1, 2\} \) gains since it only has to buy back once 3 units of the risky asset at a higher price. Also note that if the original portfolio of the grand coalition satisfies the liquidity policy and the portfolios of the divisions are concordant, then we get additivity for the realization vector. On the other hand, if for the same asset with a positive bid-ask spread some divisions have had long positions and some other divisions have had short positions, then we would get superadditivity.

For two-player games, most single-valued solution concepts, and in particular the Shapley value and the nucleolus, would share the surplus of the grand coalition over the individual worths equally over the players. In this example, this would result in the allocation \( (-29.5, -31.5) \). The firms should add 61 units of cash to its portfolio to make it acceptable to the regulator, and the risk allocation over the divisions corresponds to 29.5 units of cash for division 1 and 31.5 units of cash for division 2.

Rather than assuming the worst-case for the behavior of the opponents, we have utilized the fact that in a risk environment with liquidity considerations, divisions have initial portfolios, which means that it is meaningful to speak about inaction of a division. We have calculated the realization vector of a coalition by assuming inactivity of the complementary coalition.

Another alternative which comes to mind when defining the realization vector of a coalition, is to assume that the complementary coalition is going to hold the 0 portfolio. Alternatively, one could think of this assumption as ignoring the complementary coalition. It is not hard to see that this is not an attractive alternative. Consider for instance the risk environment with liquidity considerations of Example 3.8. When the complementary coalition holds the 0 portfolio, it is easy to calculate that \( v(1) = -28 \), \( v(2) = -30 \), and \( v(1, 2) = -61 \). The resulting risk allocation game with liquidity constraints is not superadditive. The reason is that in this example, it is easier for a single division to satisfy the short-sales constraint of 10 units than it is for the entire firm. Failure of superadditivity is not natural for the problem under consideration, as the entire firm has more actions at its disposal to satisfy the liquidity policy than a single division.

The next example demonstrates how important it is to take liquidity considerations into account in risk allocation.

<table>
<thead>
<tr>
<th>Table 1</th>
<th>A risk allocation game with liquidity constraints.</th>
</tr>
</thead>
<tbody>
<tr>
<td>State/C</td>
<td>X(1)</td>
</tr>
<tr>
<td>s = 1</td>
<td>10</td>
</tr>
<tr>
<td>s = 2</td>
<td>31</td>
</tr>
</tbody>
</table>

\(\rho(X(C))\) = 31, \(d(C)\) = -31
Consider a firm with two risky assets, $J = \{1, 2\}$. The portfolios of the divisions are $\theta^1 = (-16, 5, 0)$ and $\theta^2 = (-14, 0, 5)$, so the divisions have long positions in different assets. We assume $S = \{1, 2\}$ with both states having equal probability of occurrence.

In state 1 the MDCs of the risky assets are given by

$$m_1^1(x) = \begin{cases} 4.4 & \text{if } x < 0, \\ 3.4 & \text{if } x > 0, \\ 2.6 & \text{if } x = 0 \end{cases}$$

$$m_2^1(x) = \begin{cases} 3.6 & \text{if } x < 0, \\ 2.6 & \text{if } x > 0. \end{cases}$$

Except for the bid-ask spread, both assets are liquid. In state 2 the MDCs of the risky assets coincide and they are less liquid,

$$m_1^2(x) = m_2^2(x) = \begin{cases} 4 & \text{if } x < 0, \\ 3 & \text{if } 0 < x < 1, \\ 2 & \text{if } x > 1. \end{cases}$$

We consider the deterministic liquidity policy specified by

$$L_1 = L_2 = \{p \in [p_0, p_0] \geq -23\}.$$
It holds that

\[ \sum_{c \in C} \lambda_c \rho^p(c) = - \sum_{c \in C} \lambda_c \rho(X(c)) \]

\[ = - \sum_{c \in C} \rho(\lambda_c X(c)) \]

\[ \leq - \rho \left( \sum_{c \in C} (\lambda_c X(c)) \right) \]

\[ \leq - \rho \left( \left( \sum_{c \in C} (\lambda_c u^q(c) + e) \right) \right) \]

\[ \leq - \rho \left( \left( \sum_{c \in C} \sum_{i \in S} (\lambda_c u^q(i)) \right) + \sum_{c \in C} \lambda_c e \right) \]

\[ = - \rho \left( \sum_{c \in C} \lambda_c (u^q(c)) \right) \]

\[ \leq - \rho(X(D)) + \delta \sum_{c \in C} \lambda_c \]

\[ = \tau^p(D) + \delta \sum_{c \in C} \lambda_c. \]

Since \( \sum_{c \in C} \lambda_c \leq n \), we have that \( \delta \sum_{c \in C} \lambda_c \) tends to zero as \( \varepsilon \) tends to zero. Therefore, when taking the limit as \( \varepsilon \to 0 \), the chain of equalities and inequalities (8)–(15) imply that

\[ \sum_{c \in C} \lambda_c \rho^p(c) \leq \tau^p(D), \]

so the game \( (D, \rho^p) \) is balanced.

Eq. (8) follows from Eqs. (5) and (9) follows from the positive homogeneity of \( \rho \). Inequality (10) is due to the subadditivity of \( \rho \). Eq. (11) plugs in Eq. (6). Inequality (12) is true because of the positive homogeneity and superadditivity of \( u \) (following from Proposition 2.9), and the monotonicity of \( \rho \). Eq. (13) makes use of the fact that \( \rho \) satisfies translation invariance. Due to the monotonicity of \( \rho \), Inequality (14) is true if we can prove that for each \( s \in S \) it holds that

\[ u^q \left( \sum_{c \in C} \lambda_c q_i(c) \right) \leq X(D). \]

Consider any \( s \in S\). Take any \( i \in D \), define \( c^i = \{ c \in 2^D | i \in c \} \) and \( D^i = D \setminus c^i \). We have that

\[ \sum_{c \in C} \lambda_c u^q(c) = \sum_{c \in C} \lambda_c u^q_i(c) + \sum_{c \in C} \lambda_c u^q_i(c) \]

\[ \leq \sum_{c \in C} \lambda_c u^q_i(c) + \sum_{c \in C} \lambda_c u^q_i(\theta(c)) \]

\[ = u^q \left( \sum_{c \in C} \lambda_c (\theta(c)) \right) \]

\[ = u^q \left( \sum_{c \in C} \lambda_c (\theta(D) + \theta(c)) \right) \]

\[ = u^q \left( \sum_{c \in C} \lambda_c (\theta(D) + \theta(c)) \right). \]

where the first inequality follows from Proposition 3.5 and the second inequality from Corollary 2.10.

We define \( \theta(D) = \sum_{c \in C} \lambda_c (\theta(D) + \theta(c)) \) and show that \( \theta(D) + \theta(N \setminus D) \in L^1 \). For every \( c \in C \) it holds that \( q^i(c) + \theta(N \setminus c) \in L^1 \). Therefore, it holds that

\[ \theta(D) + \theta(c) + \theta(N \setminus D) = q^i(c) + \theta(D \setminus c) + \theta(N \setminus D) \]

\[ = q^i(c) + \theta(N \setminus c) \in L^1. \]

Since \( \sum_{c \in C} \lambda_c \theta^c = 1 \) and \( L^1 \) is convex, we have that \( p^\theta(D) + \theta(N \setminus D) \in L^1 \). Finally, we define \( q^\theta(D) \) by \( q^\theta(D) = p^\theta(D) \) and

\[ q^\theta(D) = \theta(D) = \sum_{c \in C} \lambda_c t^r(c). \]

Since \( \ell \) is concave, we have

\[ q^\theta(D) \geq \sum_{c \in C} \lambda_c \ell \left( \theta(D) + t^r(c) \right) = \theta(D) - \sum_{c \in C} \lambda_c t^r(c) = \rho^\theta(D), \]

so \( u^q(q^\theta(D)) \geq u^p(p^\theta(D)) \). By definition, it holds that \( q^\theta(D) \in A^*(\theta(D)) \). Assumption 2.7 (ii) implies that \( q^\theta(D) + \theta(N \setminus D) \in L^1 \). It follows that

\[ \sum_{c \in C} \lambda_c u^q(q^\theta(c)) \leq u^p(p^\theta(D)) \leq u^q(q^\theta(D)) \leq X(D). \]

Note that all the axioms of coherent measures of risk were used in the proof of Proposition 4.3.

Not only is it true that all risk allocation games with liquidity constraints are totally balanced. It also holds that any totally balanced game can be generated by a risk allocation game with liquidity constraints. It suffices to take one perfectly liquid asset with a trivial liquidity policy and apply the construction of Csőka et al. (2009) [Proposition 3.2]. Thus we have the following theorem.

**Theorem 4.4.** The class of risk allocation games with liquidity constraints coincides with the class of totally balanced games, \( \Gamma_{\mathrm{tr}} = \Gamma_{\mathrm{tb}} \).

5. Conclusion

Liquidity is of crucial importance when assessing the risk involved in an asset portfolio. Nevertheless, the literature on risk allocation games has so far ignored this important aspect. When we add a liquidity policy to a risk environment, we obtain an environment that is characterized by pervasive externalities. Indeed, when attributing the risk of a firm to its divisions, the question whether a single division satisfies the liquidity policy cannot be answered without making assumptions about the behavior of the complementary divisions.

We argue that the standard ways to deal with externalities, \( x \)-effectiveness and \( y \)-effectiveness are not appropriate here, and that it is also not possible to simply ignore the complementary divisions. Since in our framework the property rights of divisions are well-defined, it is meaningful to speak about inactivity of a coalition. We then say that a coalition of divisions satisfies the liquidity policy if it does so when the complementary coalition is inactive.

Our main result demonstrates that this approach leads to risk allocation games that are totally balanced. It is therefore possible to attribute the risk of a firm to its divisions in a stable way, since the core of a totally balanced game is evidently non-empty. In fact, the core of any subgame of a totally balanced game is non-empty as well, so the risk of any division can also be attributed in a stable way to its subdivisions too.

**References**


