Structural Tractability of Shapley and Banzhaf Values in Allocation Games

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Abstract
Allocation games are coalitional games defined in the literature as a way to analyze fair division problems of indivisible goods. The prototypical solution concepts for them are the Shapley value and the Banzhaf value. Unfortunately, their computation is intractable, formally #P-hard. Motivated by this bad news, structural requirements are investigated which can be used to identify islands of tractability. The main result is that, over the class of allocation games, the Shapley value and the Banzhaf value can be computed in polynomial time when interactions among agents can be formalized as graphs of bounded treewidth. This is shown by means of technical tools that are of interest in their own and that can be used for analyzing different kinds of coalitional games. Tractability is also shown for games where each good can be assigned to at most two agents, independently of their interactions.

1 Introduction
Coalitional game theory provides a solid mathematical framework to study scenarios where agents can obtain higher values by collaborating with each other rather than by acting in isolation (see, e.g., [Nisan et al., 2007; Osborne and Rubinstein, 1994]). In abstract terms, a coalitional game $G$ is a tuple $(N, v)$, where $N$ is a set of agents and $v$ is a function associating each coalition $C \subseteq N$ with the worth that agents in $C$ can guarantee to themselves. The worth can be freely distributed among the agents and, in fact, the crucial problem is to single out the most desirable distributions (of the worth associated with the grand-coalition $N$), usually called solution concepts, which can be perceived as fair and stable.

In this paper, we consider the class of allocation games, which provides a framework to analyze fair division problems where monetary compensations are allowed and utilities are quasi-linear [Moulin, 1992]: We are given an allocation scenario $A$ comprising a set of goods and a set of agents, and each agent is to be assigned at most one good she is interested in. Each good $g$ has a value $val(g) \in \mathbb{R}$ and the worth $v_A(C)$ associated with any coalition $C \subseteq N$ is the maximum overall value that can be obtained over the assignments to agents in $C$ only, also called allocations, hereinafter.

Example 1.1. Consider the allocation scenario $A_0$ that is reported in Figure 1, by using an intuitive graphical notation. We have a set $\{g_1, g_2, g_3, g_4\}$ of goods that have to be allocated to three agents. Each edge connects an agent to a good she is interested in. Edges in bold identify an optimal allocation.

For each $C \subseteq \{1, 2, 3\}$ with $C \neq \emptyset$, an optimal allocation restricted to the agents in $C$ is also reported. Then, the associated coalitional game is $G_{A_0} = (\{1, 2, 3\}, v_{A_0})$, where $v_{A_0}(\{1, 2, 3\})=6$, $v_{A_0}(\{1, 2\})=5$, $v_{A_0}(\{1, 3\}) = v_{A_0}(\{2, 3\})=4$, $v_{A_0}(\{1\}) = v_{A_0}(\{2\})=3$, and $v_{A_0}(\{3\})=1$.<ref>

Allocation games naturally arise in various application domains, ranging from house allocation to room assignment-rent division, to (cooperative) scheduling and task allocation, to protocols for wireless communication networks, and to queuing problems (see, e.g., [Moulin, 1992; Maniquet, 2003; Mishra and Rangarajan, 2007; Greco and Scarcello, 2014b] and the references therein). In these contexts (and when monetary transfers are possible), the prototypical solution concepts considered in the literature are the Shapley value [Shapley, 1953] and the Banzhaf value (or index) [Banzhaf, 1965]. However, it is well known that, in general, computing such values is #P-complete. This is a serious obstruction to their applicability in allocation scenarios involving many agents, and it motivates the design of approximation algorithms and the identification of subclasses of practical interest where exact computation can be carried out efficiently.

In the paper we focus on the latter approach. For a better understanding of the problem, we first strengthen the known hardness results to the case of goods with one possible value only. Then, we look for islands of tractability of allocations.
problems. To this end, we provide a characterization of the marginal contribution of an agent to any coalition in terms of certain properties of good allocations, which are not required to be optimal ones. Such a technical tool allows us to point out the tractability of allocation games where every good is shared (or claimed for) by two agents at most.

The main result of the paper, also based on the tool discussed above and on further ingredients exploiting constraint satisfaction techniques, is a polynomial-time algorithm for the computation of the Shapley value and the Banzhaf value in allocation games where agent interactions have a tree-like structure—formally, have bounded treewidth [Robertson and Seymour, 1984]. These games capture scenarios of practical interest. For instance, we analyzed an instantiation for the setting described in the Appendix A.1 of the work by [Greco and Scarcello, 2014b] and referring to an allocation problem arising in the Italian Research Assessment program. In particular, we analyzed the publications selected by the researchers at the University of Calabria for the period 2004-2010, by discovering that the treewidth of the underlying (co-authorship) graph, consisting of more than 500 nodes, is just 9.

Moreover, the main result and the technical tools used to get it have their own theoretical interest, since the analysis of the complexity of reasoning problems related to coalitional games has its own theoretical interest, since the analysis of graphs having bounded treewidth.

2 Formal Framework

Solution Concepts. Coalitional games can be formalized as tuples \( G = (N,v) \) where each coalition \( C \subseteq N \) is associated with a real value \( v(C) \) meant to encode the worth that agents in \( C \) obtain by collaborating with each other. A fundamental problem for a coalitional game \( G = (N,v) \) is to single out the most desirable outcomes, usually called solution concepts, in terms of appropriate notions of worth distributions, i.e., of payoff vectors of the form \((x_1, \ldots, x_N) \in \mathbb{R}^{|N|}\) where \( x_i + \cdots + x_i = v_i \) equals the worth associated with the whole set \( N \) of agents. In the paper, we focus on the Shapley value, which is a well-known solution concept such that the payoff associated with each agent \( i \in N \) is given by

\[
\phi_i(G) = \sum_{C \subseteq N \setminus \{i\}} \frac{|C|!(n - |C| - 1)!}{n!} \left( v(C \cup \{i\}) - v(C) \right),
\]

and on the Banzhaf value, for which the payoff of \( i \) is

\[
\beta_i(G) = \frac{1}{2^{n-1}} \sum_{C \subseteq N \setminus \{i\}} \left( v(C \cup \{i\}) - v(C) \right),
\]

where \( v(C \cup \{i\}) - v(C) \) is the marginal contribution of \( i \) to the coalition \( C \cup \{i\} \).

Allocation Scenario. Assume that a set \( G \) of goods have to be allocated to a set \( N = \{1, \ldots, n\} \) of agents. Each good \( g \in G \) is associated with a real value \( \text{val}(g) \in \mathbb{R} \), and each agent \( i \in N \) can receive at most one good taken from her set of interest \( \Omega(i) \subseteq G \). The tuple \( A = (N, \mathcal{G}, \Omega, \text{val}) \), with \( \Omega : N \rightarrow 2^G \) and \( \text{val} : G \rightarrow \mathbb{R} \), is an allocation scenario.

Goods are indivisible and unshareable. Hence, an allocation for \( A \) is a function \( \pi : N \rightarrow \mathcal{G} \cup \{\emptyset\} \) such that: (1) for each agent \( i \in N \), \( \pi(i) \neq \emptyset \) implies \( \pi(i) \in \Omega(i) \); and (2) for each pair \( i, i' \in N \) with \( i \neq i' \), \( \pi(i) \cap \pi(i') = \emptyset \) holds. We denote by \( \text{img}(\pi) \) the set of all goods in the image of \( \pi \), i.e., \( \text{img}(\pi) = \{\pi(i) \mid i \in N \land \pi(i) \neq \emptyset\} \).

By slightly abusing of notation, if \( S \subseteq G \) is a set of goods, then \( \text{val}(S) \) denotes the sum of their values. Moreover, if \( \pi \) is an allocation, then \( \text{val}(\pi) \) denotes the value of \( \text{img}(\pi) \). An allocation \( \pi \) is optimal (w.r.t. \( A \)) if \( \text{val}(\pi) \geq \text{val}(\pi') \) holds, for each allocation \( \pi' \). The value associated with any optimal allocation w.r.t. \( A \) is hereinafter denoted as \( \text{opt}(A) \).

Allocation Games. Let \( A = (N, \mathcal{G}, \Omega, \text{val}) \) be an allocation scenario and let \( C \subseteq N \) be a set of agents. The restriction of \( A \) to \( C \) is the sub-scenario \( A[C] = (C, \mathcal{G}, \Omega_C, \text{val}) \) where \( \Omega_C \) is the restriction of \( \Omega \) over \( C \). The allocation game induced by \( A \) is the tuple \( G_A = (N, \mathcal{v}_A) \), where \( \mathcal{v}_A : 2^N \rightarrow \mathbb{R} \) is such that \( \mathcal{v}_A(C) = \text{opt}(A[C]), \) for each \( C \subseteq N \).

The value of an empty set of goods is 0. Then, the definition trivializes for \( C = \emptyset \), with \( \mathcal{v}_A(\emptyset) = 0 \). Moreover, note that \( \mathcal{v}_A(C) \geq 0 \) holds, for each \( C \subseteq N \), since the allocation where no agent receives some good is a feasible one.

The following properties are known to hold on every pair \( C \subseteq N \) of sets of agents such that \( C \subseteq C \subseteq N \) [Greco and Scarcello, 2014b; Moulin, 1992]:

\begin{itemize}
  \item \( \text{allocation monotonicity}: \mathcal{v}_A(C) \geq \mathcal{v}_A(C') \).
  \item If \( \pi \) is an optimal allocation for \( A[C] \), then there is an optimal allocation \( \pi' \) for \( A[C'] \) such that \( \text{img}(\pi') \subseteq \text{img}(\pi) \).
  \item Submodularity: \( \mathcal{v}_A(C \cup \{i\}) - \mathcal{v}_A(C) \leq \mathcal{v}_A(C' \cup \{i\}) - \mathcal{v}_A(C') \), for each \( i \in N \setminus C \).
\end{itemize}

3 Intractability of Computation

Computing the Shapley value is a problem that has been shown to be \#P-complete on different classes of games (see, e.g., [Deng and Papadimitriou, 1994; Nagamochi et al., 1997; Bachrach and Rosenschein, 2009; Aziz and de Keijzer, 2014]), including the allocation games [Greco and Scarcello, 2014b]. In particular, hardness has been shown to hold even on instances whose goods have three possible values. Below, we observe the result by showing that there is no advantage in focusing on scenarios where all goods have the same value. To this end, we first focus on the Banzhaf value.

\begin{theorem}
Computing the Banzhaf value is \#P-hard on allocation games (under Turing reductions), even for scenarios \( A = (N, \mathcal{G}, \Omega, \text{val}) \) such that \( \{|\text{val}(g) \mid g \in G\}| = 1 \).
\end{theorem}

\begin{proof}
Let \((S \cup I, E)\) be a bipartite graph, hence with \( S \cap I = \emptyset \) and \( E \subseteq S \times I \). Computing the number of subsets \( C \subseteq S \) of vertices to which all vertices in \( I \) can be matched is \#P-hard [Colbourn et al., 1995].

Based on \((S \cup I, E)\), let us build the allocation scenario \( A = (S \cup \{S + 1\}, I, \Omega, \text{val}) \) where nodes in \( S \) (resp., \( I \)) are transparently viewed as the agents (resp., goods), where \( \text{val}(g) = 1 \) for each \( g \in I \), and where \( \Omega(i) = \{g \in I \mid g \}

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by that paper and the fact that, for each agent \(a\), 2009]. For its proof, we exploit some of the arguments of this latter concept turns out to be \(\#P\)-hard too. This proposition is established by showing that the

\[
\phi_i(G_A) = \sum_{h=0}^{n-1} \frac{n!}{h!(n-h-1)!} \beta_i(G_A, h),
\]

(1)

where, for each \(h \in \{0, \ldots, n-1\}\), it holds that \(\beta_i(G_A, h) = \sum_{C \subseteq N \setminus \{i\}, |C|=h} (\nu(C \cup \{i\}) - \nu(C))\).

From these results, it turns out that acting on the values of goods does not help very much in identifying tractable classes of instances. So, we next consider different kinds of restrictions based on the “interactions” that emerge among agents.

4 Characterizations of The Shapley Value

Throughout the section, assume that an allocation scenario \(A = (N, G, \Omega, \text{val})\) is given. Let \(\{w_1, \ldots, w_m\} = \{\text{val}(g) \mid g \in G\} \cup \{0\}\) be the set of all values associated with goods in \(G\) (plus the null value 0, if not present), and assume that \(w_1 > w_2 > \cdots > w_m\). W.l.o.g, assume also that \(w_m = 0\).

4.1 A Closer Look at Marginal Contributions

We start with a simple reformulation. Let \(i \in N\) be an agent, let \(h \in \{0, \ldots, n-1\}\), let \(\ell \in \{1, \ldots, m\}\), and let us denote by \#\(c_{i, \ell}(G_A, h)\) the number of coalitions \(C\) such that \(|C| = h\) and \(c_i(A^C \cup \{i\}) - c_i(A^C) \geq w_h\). Then, the coefficients \(\beta_i(G_A, h)\) in the expressions illustrated in Equation (1) can be rewritten as follows, by simple algebraic manipulations and by exploiting the monotonicity of allocation games.

**Theorem 4.1**. For each agent \(i \in N\) and \(h \in \{0, \ldots, n-1\}\),

\[
\beta_i(G_A, h) = w_1 \times \#\(c_{i, 1}(G_A, h)\) + \sum_{\ell=2}^{m} w_1 \times \left(\#\(c_{i, \ell}(G_A, h)\) - \#\(c_{i, \ell-1}(G_A, h)\)\right).
\]

Hence, counting the number of coalitions to which some given marginal contribution can be provided is deeply related to the computation of the Shapley and Banzhaf values of allocation games. We now further explore this specific task. For each “level” \(\ell \in \{1, \ldots, m\}\) over the possible values, an agent \(i \in N\) is said to be dependent at level \(\ell\) (short: \(\ell\)-dependent) if for each \(g \in \Omega(i)\) with \(\text{val}(g) \geq w_\ell\), there is an agent \(j \in N \setminus \{i\}\) such that \(g \in \Omega(j)\). In particular, note that any agent having goods with value at least \(w_\ell\) and which are not shared with any other agent is not dependent at level \(\ell\); in fact, all her marginal contributions are at least \(w_\ell\), independently of the coalition to be considered. Let \(G_i = (N_i, E_i)\) be the undirected graph where \(N_i\) is the set of all \(\ell\)-dependent agents and where \(\{i, j\} \in E_i\) if, and only if, there is a good \(g \in \Omega(i) \cap \Omega(j)\) with \(\text{val}(g) \geq w_\ell\). Then, a coalition \(R \subseteq N_i\) of agents is called a component at level \(\ell\) (short: \(\ell\)-component) if the subgraph of \(G_i\) induced by the nodes in \(R\) is connected. As a special case, if there is no good \(g \in \Omega(i)\) with \(\text{val}(g) \geq w_\ell\), then \(\{i\}\) is an \(\ell\)-component.

**Example 4.2**. Consider the scenario \(A_0\) reported in Figure 1. We have \(w_1 = \text{val}(g_1)\). Moreover, \(\{1, 2\}\) and \(\{3\}\) are the only subset-maximal components at level 1. Indeed, \(g_1 \in \Omega(1) \cap \Omega(2)\), and there is no good in \(\Omega(3)\) with value \(w_1\). If \(C \subseteq N\) is a coalition and \(i \notin C\) is an agent, then we denote by \(p_i^C(C)\) the \(i\)-part of \(C\) w.r.t. \(i\). This is the emptyset if \(i \notin N_i\); otherwise, \(p_i^C(C)\) is the subset-maximal (in fact, unique) \(\ell\)-component \(R \subseteq C \cup \{i\}\) with \(i \in R\). These concepts play a key role to characterize marginal contributions.

**Theorem 4.3**. Let \(C \subseteq N\) be a coalition and let \(i \in N \setminus C\) be an agent for which there is a good \(g \in \Omega(i)\) with \(\text{val}(g) \geq w_\ell\). Then, the following statements are equivalent:

(1) \(\nu_{\ell}(C \cup \{i\}) - \nu_{\ell}(C) \geq w_\ell\);

(2) there is an allocation \(\pi\) for \(A_{\ell}(C)\) such that \(\text{val}(\pi(j)) \geq w_\ell\) for each \(j \in p_i^C(C)\).

**Proof Idea.** If \(i \notin N_i\), then \(p_i^C(C) = \emptyset\) and (2) trivially holds. Moreover, its marginal contribution to any coalition is at least \(w_\ell\). So, (1) holds, too. In the remaining, consider the case where \(i \in N_i\), so that \(p_i^C(C)\) is non-empty. Let \(R = p_i^C(C) \setminus \{i\}\) and let \(S = C \setminus R\). Here, we show how to deal with the case \(S = \emptyset\). The result can be generalized by noticing that agents in \(S\) do not “interact” with agents in \(R\) (w.r.t. level \(w_\ell\)).

(1)\(\Rightarrow\)(2) Assume that (2) does not hold for optimal applications. That is, there is an optimal allocation \(\pi\) for \(A[R \cup \{i\}]\) and of an agent \(j' \in R \cup \{i\}\) such that \(\text{val}(\pi(j')) < w_\ell\). Consider the following two possible cases. First, assume that \(\text{val}(\pi(i)) < w_\ell\). Since the restriction of \(\pi\) over the agents in \(R\) is a feasible allocation for \(A[R]\), then we immediately get that \(\nu_{\ell}(R) \geq \text{val}(\pi) - \text{val}(\pi(i)) > \text{val}(\pi) - w_\ell\), and hence \(\nu_{\ell}(R \cup \{i\}) - \nu_{\ell}(R) < w_\ell\). Second, assume that \(\text{val}(\pi(i)) \geq w_\ell\). We start by observing that, due to the optimality of \(\pi\), for each agent \(j' \in R \cup \{i\}\) with \(\text{val}(\pi(j')) < w_\ell\), \(\{g \in \Omega(j') \wedge \text{val}(g) \geq w_\ell\} \subseteq \text{img}(\pi)\). That is, goods that might be in principle allocated to an agent \(j' \in R \cup \{i\}\) with \(\text{val}(\pi(j')) < w_\ell\) and having value at least \(w_\ell\) are actually allocated to some different agent in \(R \cup \{i\}\). Given that \(R \cup \{i\}\) is an \(\ell\)-component (and that, in particular, each agent is \(\ell\)-dependent), we are guaranteed about the existence of a
succession \( i = j_1, j_2, \ldots, j_k \) such that \( \pi(j'_1) \cap \Omega(j'_{k+1}) \neq \emptyset \), for each \( x \in \{1, \ldots, h-1\} \); and \( \text{val}(\pi(j'_k)) < \omega_x \). Consider then the function \( \overline{\pi}_*: R \rightarrow G \cup \{\emptyset\} \) with \( \overline{\pi}_*(j'_k) = \pi(j'_k) \), for each \( x \in \{1, \ldots, h-1\} \); and \( \overline{\pi}_*(j''_k) = \pi(j''_k) \), for each \( j'' \in \{j_2, \ldots, j_k\} \). Then, \( \overline{\pi}_* \) is an allocation for \( A[R] \) and we have that \( \text{val}(\overline{\pi}_*) = \text{val}(\pi) - \text{val}(\pi(j'_k)) \). Hence, \( v_A(R) \geq \text{val}(\pi) - \text{val}(\pi(j'_k)) > \text{val}(\pi) - \omega_x \).

That is, \( v_A(R \cup \{i\}) - v_A(R) < \omega_x \). In both cases, we have derived a contradiction with (1).

(2)\(\Rightarrow\)(1) Let \( \pi' \) be an allocation for \( A[R \cup \{i\}] \) such that \( \text{val}(\pi') \geq \omega_x \), for each \( j \in R \cup \{i\} \). We can show that there is an optimal allocation \( \pi \) for \( A[R \cup \{i\}] \) with the same property.

Because of the allocation monotonicity property, there is also an optimal allocation \( \pi_i \) for \( A[R] \) such that \( \text{img}(\pi_i) \subseteq \text{img}(\pi) \). Hence, for each \( j \in R \), \( \text{val}(\pi_i(j)) \geq \omega_x \). Now, observe that \( v_A(R \cup \{i\}) = \text{val}(\pi) \) and \( v_A(R) = \text{val}(\pi_i) \). So, \( v_A(R \cup \{i\}) - v_A(R) \) coincides with the value of one of the goods in \( \text{img}(\pi) \), and \( v_A(R \cup \{i\}) - v_A(R) \geq \omega_x \). \( \square \)

Example 4.4. By continuing with Example 4.2, note that \( \{1, 2\} = p_1^\ell(\{2, 3\}) \) holds, for \( \ell \in \{1, 2\} \). Therefore, the allocation for \( A_0[\{1, 2\}] \) depicted in Figure 1 witnesses, by Theorem 4.3, that \( v_{A_0}(\{1, 2, 3\}) - v_{A_0}(\{2, 3\}) \geq 2. \)

4.2 Bounded Sharing

Our analysis intensively uses Theorem 4.3. The first outcome is an island of tractability based on the notion of bounded sharing. Formally, for a given level \( \ell \), define the sharing degree of an allocation scenario \( \mathcal{S} \), denoted by \( sd_{\ell}(\mathcal{S}) \), as the maximum, over all goods \( g \) with \( \text{val}(g) \geq \omega_x \), of \( |\{j \in N \mid g \in \Omega(j)\}| \). Intuitively, it measures the maximum number of agents competing for the same good (with value at least \( \omega_x \)).

Theorem 4.5. The Shapley and Banzhaf values of allocation games \( G_A \) can be computed in polynomial time on scenarios \( \mathcal{S} = (N, G, \Omega, \text{val}) \) such that \( sd_{\ell}(\mathcal{S}) \leq 2 \), for each level \( \ell \).

Proof Idea. Let \( i \) be an agent in \( N \), and let \( h \in \{0, \ldots, n-1\} \). The line of the proof is to show that:

\[
#c^h_i(G_A, h) = \frac{(n-1)!}{(n-h-1)!h!} - X', \quad \text{where}
\]

- \( X' = 0 \), if \( h < |p_1^\ell(N \backslash \{i\})| - 1 \); or if \( i \) is not \( \ell \)-dependent, or the subgraph of \( G \) induced by the nodes in \( p_1^\ell(N \backslash \{i\}) \) contains a cycle, or there are two agents \( j \) and \( j' \) in \( p_1^\ell(N \backslash \{i\}) \) with \( |\Omega(j') \cap \Omega(j')| \cap \{g \mid \text{val}(g) \geq \omega_x\} | > 1 \).

- \( X' = \frac{(n-|p_1^\ell(N \backslash \{i\})|)!}{(n-h-1)!h!} |p_1^\ell(N \backslash \{i\})|! \), otherwise.

In particular, the value derives by analyzing the allocations of Theorem 4.3(2) on the scenario \( \mathcal{S} \) such that \( sd_{\ell}(\mathcal{S}) \leq 2 \). The result is then established because of this closed form, of Theorem 4.1, and of the expressions in Equation (1). \( \square \)

5 Bounded Treewidth

We now move to allocation games where the interactions among agents have a tree-like structure. We use the technical tools provided in Section 4, by combining them with CSP techniques that are of interest in their own.

For any scenario \( \mathcal{S} = (N, G, \Omega, \text{val}) \), let \( G(\mathcal{S}) = (N, E) \) be the undirected graph such that \( \{i, j\} \in E \) if, and only if, there is a good \( g \in \Omega(i) \cap \Omega(j) \). Moreover, recall that a tree decomposition of a graph \( G = (N, E) \) is a pair \( (T, \chi) \), where \( T = (V, F) \) is a tree, and \( \chi \) is a function assigning to each vertex \( p \in V \) of a set of nodes \( \chi(p) \subseteq N \), such that the following conditions are satisfied: (1) \( \forall b \in N, \exists p \in V \) such that \( b \in \chi(p) \); (2) \( \forall (b, d) \in E, \exists p \in V \) such that \( \{b, d\} \subseteq \chi(p) \); (3) \( \forall b \in N \), the set \( \{p \in V \mid b \in \chi(p)\} \) induces a connected subtree of \( T \). The width of \( (T, \chi) \) is \( \max_{p \in V} |\chi(p) - 1| \), and the treewidth of \( G \) (short: \( tw(G) \)) is the minimum width over all its tree decompositions (see, e.g., [Robertson and Seymour, 1984]).

5.1 Preliminaries on CSPs

A constraint satisfaction problem (short: CSP) instance is a triple \( \mathcal{I} = (Var, U, C) \), where \( Var \) is a finite set of variables, \( U \) is a finite domain of values, and \( C = \{C_1, C_2, \ldots, C_q\} \) is a finite set of constraints (see, e.g., [Dechter, 2005]).

Each constraint \( C_v \), for \( 1 \leq v \leq q \), is a pair \( (S_v, r_v) \), where \( S_v \subseteq Var \) is a set of variables called the constraint scope, and \( r_v \) is a constraint relation, i.e., a set of substitutions \( \theta : S_v \rightarrow U \) indicating the allowed combinations of simultaneous values for the variables in \( S_v \). A substitution from a set \( V \subseteq Var \) to \( U \) is often viewed as the set of pairs of the form \( X/u \), where \( \theta(X) = u \) is the value to which \( X \in V \) is mapped. For each variable \( X \in Var \), its domain is the set of all elements \( u \in U \) for which some constraint relation contains a substitution \( \theta \) with \( \theta(X) = u \). A substitution \( \theta \) satisfies \( C_v \) if its restriction to \( S_v \) occurs in \( r_v \). A solution to \( \mathcal{I} \) is a substitution \( \theta : Var \rightarrow U \) satisfying all constraints. The set of all solutions is denoted by \( \Theta(\mathcal{I}) \). If \( W \) is a set of variables, then \( \Theta(W, V) \) denotes the set of all solutions in \( \Theta(\mathcal{I}) \) restricted to the variables in \( W \). Variables outside \( W \) can be viewed as auxiliary ones—they are used for internal encoding activities, and they are not required in the output.

With each CSP instance \( \mathcal{I} \), we can naturally associate the graph \( G(\mathcal{I}) \) whose nodes are the variables and where there is an edge between any pair of variables appearing within the same scope. Deciding whether there is a solution (and compute one, if any) is generally NP-hard, but it is known to be feasible in polynomial time on classes of CSP instances \( \mathcal{I} \) whose associated graphs have treewidth bounded by some given constant [Gottlob et al., 2013]. Recently, these kinds of structural tractability results have been generalized to counting problems, as summarized below.

Theorem 5.1 (cf. [Pichler and Skritek, 2013; Greco and Scarcello, 2014a]). Counting the number of substitutions in \( \Theta(W, V) \) is feasible in polynomial time, on classes of CSP instances \( \mathcal{I} \) such that the treewidth of \( G(\mathcal{I}) \) is bounded by a constant, and the size of the domain of each variable not in \( W \) is bounded by some constant, too.

Note that, differently from the case of the standard decision and computation problems, the result is established under the additional condition that auxiliary variables have a bounded domain. If the condition is not met, then \#P-complete instances can be exhibited [Pichler and Skritek, 2013].

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5.2 CSP Encoding (for the Banzhaf Value)

In order to establish a tractability result, we shall encode the computation of the coefficients \( \#c^G_j(G_A, h) \) in terms of a counting problem over a suitably defined CSP instance and we shall then make use of Theorem 5.1. The challenge is to end up with an encoding using a constant number of values for the auxiliary variables. For instance, the natural encoding where some variable \( x_j \) (associated with an agent \( j \in N \)) can take as values the goods in \( \Omega(j) \) is not useful here. In fact, we propose an encoding that uses both the given allocation scenario \( A \) and a tree decomposition \( TD = (T, \chi) \) of \( G(A) \). The idea is that each good is associated with some distinguished vertex of \( T \), while suitable variables in the labels of the tree encode the roadmaps to reach such goods. In particular, their domain just contains the needed road signs (five values are enough). This is detailed below.

We start by building a tree decomposition with certain desirable properties. Let \( (T', \chi') \) be a tree decomposition of \( (G_A, h) \) whose width is \( k > 0 \). Note that, for each good \( g \in \Gamma \), we are guaranteed about the existence of a vertex \( v_g \) in \( T' \) such that \( \chi(v_g) \supseteq \{ j \mid g \in \Omega(j) \} \). Indeed, the agents in \( \{ j \mid g \in \Omega(j) \} \) form a clique in \( G(A) \).

In a pre-processing step, we modify \( (T', \chi') \) by adding a fresh vertex \( v''_g \) as a child of \( v_g \), whose label is \( \chi(v'') = \chi(v_g) \cup \{ j \mid g \in \Omega(j) \} \). By iterating over all goods, we get the desired tree where each good \( g \) is associated with a distinguished vertex (in fact, leaf) \( v''_g \) labeled by the agents to whom \( g \) can be allocated. Of course, the transformation is feasible in polynomial time. Eventually, we further transform the decomposition by making it binary: For each vertex \( v \) with children \( v_1, ..., v_n \), we can create a novel vertex \( \bar{v} \) as a child of \( v \) and with its label, by subsequently appending under it all these children but \( v_1 \). Let \( TD = (T', \chi') \) be the resulting tree decomposition, having the same width as \( (T', \chi') \).

**Example 5.2.** Figure 2 illustrates a width-2 tree decomposition \( TD_0 \) of \( G(A_0) \), by evidencing the vertices that are univocally associated with the goods in \( \{ g_1, g_2, g_3, g_4 \} \). Moreover, note that the decomposition is defined over a binary tree.

The input to our encoding is the allocation scenario \( A \), the agent \( i \in N \), the natural number \( \ell \), and the tree decomposition \( TD = (T, \chi) \). Note that, for the moment, we do not consider the size \( h \). Then, we define the encoding \( \xi \) such that \( \xi(A, i, \ell, TD) \) is the CSP instance \( \langle Var, U, C \rangle \), where

- \( Var = \bigcup_{j \in N} \{ X_j, Y_j \} \cup \{ X^v \mid v \text{ in } T \wedge v \in \chi(v) \} \)
- \( U = \{ 0, 1, \odot, \bigodot, \land \} \)

and where \( C \) is defined as follows, with constraint relations being reported in tabular form in Figure 3:

1. For each agent \( j \in N \) and vertex \( v \in T \) with \( v \in \chi(v) \), there is a constraint \( (S_{v,j}, r, \ell, \chi) \) with \( S_{v,j} = \{ X^v \} \).
2. For each good \( g \) with \( \text{val}(g) \geq \ell \) and each \( j \in \chi(v) \), there is a constraint \( (S_{g,j}, r, \ell, \chi) \) such that \( S_{g,j} = \{ X^g \} \).
3. For each agent \( j \in N \), there is a constraint \( (S_j, r, \ell, \chi) \) such that \( S_j = \{ Y_j, X_j \} \).
4. For each pair of agents \( j, j' \in N \) that are adjacent in \( G \), there is a constraint \( (S_{j,j'}, r, \ell, \chi) \) such that \( S_{j,j'} = \{ Y_j, X_j, Y_{j'}, X_{j'} \} \).
5. For each non-leaf vertex \( v \) whose right (resp., left) child is \( v_1 \) (resp., \( v_2 \)), and for each \( j \in \chi(v) \), there is a constraint \( (S_{v,j}, r, \ell, \chi) \) such that \( S_{v,j} = \{ X_j, X^{v_1}, X^{v_2} \} \).
6. For each good \( g \) with \( \text{val}(g) \geq \ell \) and for each pair \( j, j' \in \chi(v) \) with \( g \in \Omega(j) \) and \( \Omega(j') \), there is a constraint \( (S_{g,j,j'}, r, \ell, \chi) \) such that \( S_{g,j,j'} = \{ X_j^g, X_j'^g \} \).
7. No further constraint is in \( C \).

**Theorem 5.3.** The following properties hold:

(a) \( \xi(A, i, \ell, TD) \) can be built in polynomial time;
(b) the domain of each variable in \( \xi(A, i, \ell, TD) \) consists of at most 5 distinct elements;
(c) \( tw(G(\xi(A, i, \ell, TD))) \leq 5 \times (tw(G(A)) + 1) \);
(d) if \( \theta \) is a solution to \( \xi(A, i, \ell, TD) \), then \( R_0 = \{ j \mid \theta(Y_j) = 1 \wedge j \neq i \} \) is such that \( v_A(R_0 \cup \{ i \}) - v_A(R_0) \geq \ell \); if \( R \subseteq N \setminus \{ i \} \) is such that \( v_A(R \cup \{ i \}) - v_A(R) \geq \ell \), then there is a solution \( \theta \) to \( \xi(A, i, \ell, TD) \) with \( R = R_0 \); if \( \sum_{h=0}^{h-1} \#c^G_j(G_A, h) = |\Theta(\xi(A, i, \ell, TD), \{ Y_1, ..., Y_n \})|. \)
Proof Sketch. Property (a) and Property (b) are immediate. Concerning Property (c) note that, if TD = (T, χ), then the tuple (T, χ, ℓ) such that for each vertex v in T, χ(v) = \bigcup_{j \in \chi(v)} \{X_{v,j}, Y_{v,j}, X_{v,j}^\lor, X_{v,j}^\land \mid v \text{ is not a leaf}\} is a tree decomposition of G(ξ(A, i, ℓ, TD)). Note that the decomposition does not depend on i and ℓ. As an example, the modified label associated with the root node of the tree decomposition of the graph G(\mathcal{A}_0) in Example 5.2 is shown in Figure 2.

Concerning Property (d), assume that θ is a solution to ξ(A, i, ℓ, TD) and let R_0 be the set \{j \mid \theta(X_j) = 1 \land j \neq i\}. Let j be any agent in R_0 ∪ \{i\}. First, we claim that there is a vertex v^* in T such that \theta(X_{v^*}^\lor) = 0. By contradiction, assume there is no such vertex. Let v be the vertex in TD that is the closest to the root with j \in \chi(v). Because of the constraint (S_{v,j}, r_{v,j}) of type 1, we have that \theta(X_{v,j}^\lor) = 0. Consequently, the only clause that could evaluate to 1 is the constraint \theta(X_{v,j}^\lor) = 0. For each vertex v with \theta(X_{v,j}^\lor) = 0, there is a vertex v^* in T such that \theta(X_{v^*}^\lor) = 0. Moreover, for each vertex v with \theta(X_{v,j}^\lor) = 0, there is a vertex v^* such that \theta(X_{v^*}^\lor) = 0. Therefore, we can apply the argument again on w, and so top-down from w we can eventually reach a leaf v such that \theta(v) \in \{\chi(v)\}. But, this is impossible by the constraint (S_{v,j}, r_{v,j}) of type 1. So, we know that for each j \in R_0 ∪ \{i\}, there is a vertex v^* in T such that \theta(X_{v^*}^\lor) = 0. Concerning Property (e), we have that TD is a tree decomposition of G(ξ(A, i, ℓ, TD)), and hence, the Banzhaf value. In order to compute the contribution \#_{\mathcal{E}}(G_A, h) for each cardinality of the coalitions and, hence, the Shapley value by Theorem 4.1 and Equation (1), we need a way to filter, out of all possible solutions, those \theta such that [R_0] = h. This is not immediate (by preserving structural properties and the bound on the domains), so that a careful construction is in order.

5.3 From the Banzhaf Value to the Shapley Value

The encoding \xi discussed so far does not take h as a parameter. In fact, it just provides us a way to compute the value \sum_{n=0}^{n-1} \#_{\mathcal{E}}(G_A, h) and, hence, the Banzhaf value. In order to compute the contribution \#_{\mathcal{E}}(G_A, h) for each cardinality of the coalitions and, hence, the Shapley value by Theorem 4.1 and Equation (1), we need a way to filter, out of all possible solutions, those \theta such that [R_0] = h. This is not immediate (by preserving structural properties and the bound on the domains), so that a careful construction is in order.

Theorem 5.5. The Shapley value of allocation games \mathcal{G}_A can be computed in polynomial time on all allocation scenarios whose interaction graphs have bounded treewidth.

Proof Idea. Consider this class of allocation scenarios A with tw(G(A)) ≤ k, for some fixed natural number k. Then, a width-k tree decomposition TD of G(A) and the encoding I = ξ(A, i, ℓ, TD) can be computed in polynomial time. Let TD’ be a tree decomposition of G(\mathcal{I}) whose width is bounded by 5 \times (k + 1) (cf. Theorem 5.3). Consider the modified CSP instance \mathcal{I}’ = (\xi(\mathcal{I}, Y_1, ..., Y_n), h, TD’) = (Var’, U’, C’) such that: Var’ = Var ∪ \{W_v \mid v \text{ is in } T’\}; U’ = U ∪ \{0, ..., h\}; and C’ = C ∪ \{(S_{v,j}, r_{v,j}) \mid v \text{ is in } T’\}.

In particular, for each non-leaf vertex v in T’ with children v_1 and v_2, we have S_v = χ(v) ∪ \{W_v, W_{v_1}, W_{v_2}\}. Moreover, r_v contains all possible substitutions \theta over the variables in S_v such that \theta(W_v), \theta(W_{v_1}), \theta(W_{v_2}) ∈ \{0, ..., h\} and \theta(W_v) = \{y \in S_v \mid v = cri(j) ∧ \theta(Y_j) = 1\} = \theta(W_{v_1}) ∪ \theta(W_{v_2}), where cri(j) is the vertex v^* that is the closest to the root and such that Y_j ∈ χ(v^*). Additionally, if v is the root of T’, then we require that \theta(W_v) = h + 1 holds. Instead, if v is a leaf, then S_v = χ(v) ∪ \{W_v\}, and r_v contains all possible substitutions \theta over S_v such that \theta(W_v) = \{y \in S_v \mid v = cri(j) ∧ \theta(Y_j) = 1\}.

Note that tw(G(\mathcal{I}’)) ≤ tw(G(\mathcal{I})) + 3. Moreover, by Theorem 5.3 and the above encoding, it can be checked that |\Theta(\mathcal{I}’, Y_1, ..., Y_n)| coincides with \#_{\mathcal{E}}(G_A, h). We then get the Shapley value by using Theorem 4.1 and Equation (1). Unfortunately, we cannot apply Theorem 5.1 on I’ and \{Y_1, ..., Y_n\}, since the auxiliary variables W_v do not have a bounded domain. However, we can add such variables to the output variables without altering the number of solutions, because |\Theta(\mathcal{I}’, Y_1, ..., Y_n)| = |\Theta(\mathcal{I}’, Y_1, ..., Y_n) ∪ \{Var' \setminus Var\}| holds. Thus, Theorem 5.1 applied on I’ with output variables \{Y_1, ..., Y_n\} ∪ \{Var' \setminus Var\}, ensures that \#_{\mathcal{E}}(G_A, h) can be computed in polynomial-time. □
6 Conclusion

We have studied the problem of computing the Shapley value and the Banzhaf value of allocation games, which are coalition games implicitly (and succinctly) specified in terms of an underlying allocation scenario. We have shown that the problem is #P-complete, even in stringent settings. Motivated by this bad news, we identified islands of tractability by focusing either on scenarios with sharing degree at most 2 or such that the interactions among agents have a tree-like structure. This way, real world applications with useful structural properties can efficiently be dealt with. Moreover, the technical tools used to get the results may have a wider spectrum of applicability, beyond allocation problems.

A variant of the proposed framework considers scenarios where agents must necessarily get some good. In this case, it makes sense to have goods with negative values, too. In fact, we remark that our algorithms can be extended to manage these cases as well, by just considering suitable negative levels. Our work leaves open the technical question of whether tractability still holds over scenarios with sharing degree bounded by some constant greater than 2 (e.g., sd ≤ (A) ≥ 3). Moreover, it might stimulate further research to analyze the complexity of other solution concepts over allocation games, such as the nucleolus [Schmeidler, 1969]. Finally, we point out that since our technical elaborations are often rather involved, their immediate/naive implementation might be unpractical. Indeed, there is much room for practical improvements, for instance, by adopting implementation strategies used for decomposition methods in data-intensive applications (e.g., in the evaluation of SQL queries) and parallel solutions. This might constitute another interesting avenue of further research.

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References


