Abstract:

We investigate systems of indirect voting based on the law of Penrose, in which each representative in the voting body receives the number of votes (voting weight) proportional to the square root of the population he or she represents. For a generic population distribution the quota required for the qualified majority can be set in such a way that the voting power of any state is proportional to its weight. For a specific distribution of population the optimal quota has to be computed numerically. We analyse a toy voting model for which the optimal quota can be estimated analytically as a function of the number of members of the voting body. This result, combined with the normal approximation technique, allows us to design a simple, efficient, and flexible voting system which can be easily adopted for varying weights and number of players.

Keywords: power indices; weighted voting games; optimal quota; Penrose square root law; normal approximation

JEL classification: C71; D71
1 Introduction

A game theory approach proved to be useful to analyse voting rules implemented by various political or economical bodies. Since the pioneering contributions of Lionel Penrose who originated the mathematical theory of voting power just after the World War II [1], this subject has been studied by a number of researchers, see, e.g. [2, 3] and references therein.

Although the current scientific literature contains several competing definitions of voting indices, which quantitatively measure the voting power of each member of the voting body, one often uses the original concept of Penrose. The a priori voting power in his approach is proportional to the probability that a vote cast by a given player in a hypothetical ballot will be decisive: should this country decide to change its vote, the winning coalition would fail to satisfy the qualified majority condition. Without any further information about the voting body it is natural to assume that all potential coalitions are equally likely. This very assumption leads to the concept of Penrose-Banzhaf index (PBI) called so after John Banzhaf who introduced this index independently in 1965 [4].

Recent research on voting power was partially stimulated by the political debate on the voting system used in the Council of Ministers of the European Union (EU). The double majority system endorsed in 2004 by The Treaty Establishing a Constitution for Europe, based on ‘per capita’ and ‘per state’ criteria, was criticized by several authors [5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15] who pointed out that it is favorable to the most and to the least populated EU countries at the expense of all medium size states. Ironically, a similar conclusion follows from a book written fifty years earlier by Penrose [16, p.73], who also discovered this drawback of a ‘double majority’ system.

In search for an optimal two-tier voting system (where a set of constituencies of various size elect one delegate each to a decision-making body) Penrose considered first a direct election in a state consisting of $N$ voters and proved that the voting power of a single citizen decays as \(1/\sqrt{N}\), provided that the votes are uncorrelated [1]. To compensate this effect he suggested that the a priori voting power of each representative in the voting body should behave proportionally to $\sqrt{N}$ making the citizens’ voting power in all states equal (the Penrose square root law).

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1 Penrose wrote: ‘[...] if two votings were required for every decision, one on a *per capita* basis and the other upon the basis of a single vote for each country. This system [...] would be inaccurate in that it would tend to favour large countries.’
To achieve such a balance, one may attribute the voting weight of each state proportional to the square root of its population, and such voting systems were discussed by several experts also in the EU context [17, 18, 19, 20, 21, 22, 23, 25, 24, 26, 27, 28, 11, 29, 12]. The quota \( q \) for the qualified majority is still a free parameter of the system and can be optimized in such a way that the mean discrepancy \( \Delta \) between the voting power of each state and the rescaled root of its population is minimal.

For a concrete distribution of population in the EU consisting of 25 (resp. 27) member states it was found [12, 30], see also [31], that the discrepancy exhibits a sharp minimum around a critical quota \( q_* \sim 62\% \) (resp. 61.5\%) falling down to a negligible value. Therefore, the Penrose square root system with this quota is optimal, in the sense that every citizen in each member state of the Union has the same voting power (measured by the Penrose-Banzhaf index), i.e. the same influence on the decisions taken by the European Council. Such a voting system occurs to give a larger voting power to the largest EU states than the Treaty of Nice but smaller than the European Constitution, and thus has been christened by the media as the ‘Jagiellonian Compromise’.

The existence of such a critical quota \( q_* \) for which the rescaled PBI indices of all states are approximately equal to their voting weights, is not restricted to this particular distribution of population in the EU. On the contrary, it seems to be a rather generic behaviour which was found by means of numerical simulations for typical random distributions of weights in the voting body generated with respect to various probability measures [12, 32, 33]. The value of \( q_* \) depends to some extent on a given realization of the random population distribution, but more importantly, it varies considerably with the number \( M \) of the member states. In the limit \( M \to \infty \) the optimal quota seems to tend to 50\%, in consistence with the so called Penrose limit theorem [34, 35].

Working with random probability distributions it becomes difficult to get any analytical prediction concerning the functional dependence of \( q_* \) on the number \( M \) of voting states. Therefore in this work we propose a toy model in which an analytical approach is feasible. We compute the PBIs for this model distribution of population consisting of \( M \) states and evaluate the discrepancy \( \Delta \) as a function of the quota \( q \). The optimal quota \( q_* \) is defined as the value at which the quantity \( \Delta(q) \) achieves its minimum. This reasoning performed for an arbitrary number of states \( M \) allows us to derive an explicit dependence \( q_*(M) \). Results obtained analytically for this particular model occur to be close to those received earlier in numerical experiments for random samples.
Thus we are tempted to design a simple scheme of indirect voting based on the square root law of Penrose supplemented by a rule setting the approximate value of the optimal quota \( q^* \) as a function of the number of players \( M \). The normal approximation of the number of votes achieved by all possible coalitions provides another estimate of the optimal quota as a function of the quadratic mean of all the weights.

This work is organized as follows. In Sect. 2 we recall the definition of Penrose-Banzhaf index and define the optimal quota. Sect. 3 provides a description of the toy model of voting in which one player is \( c \) times stronger than all other players. We describe the dependence of the optimal quota in this model on the number of voters for \( c = 2 \) and \( c = 3 \). In Sect. 4 we discuss the optimal quota applying an alternative technique of normal approximation. The paper is concluded in Sect. 5, where we design a complete voting system. The heuristic proof of the validity of the normal approximation method is given in Appendix.

2 A priori voting power and critical quota

Consider a set of \( M \) members of the voting body, each representing a state with population \( N_k, k = 1, \ldots, M \). Let us denote by \( w_k \) the voting weight attributed to \( k \)-th representative. We work with renormalized quantities, so that \( \sum_{i=1}^{M} w_i = 1 \), and we assume that the decision of the voting body is taken if the sum of the weights of all members of the coalition exceeds the given quota \( q \in [0.5, 1] \), i.e. we consider so called (canonical) weighted majority voting game \([q, w_1, \ldots, w_M] \), see [2].

To analyse the voting power of each member one has to consider all \( 2^M \) possible coalitions and find out the number \( \omega \) of winning coalitions which satisfy the qualified majority rule adopted. The quantity \( A := \omega/2^M \) measures the decision-making efficiency of the voting body, i.e. the probability that it would approve a randomly selected issue. Coleman called this quantity the power of a collectivity to act [36]. For a thorough discussion of this concept, see [34, Ch. 6].

The absolute Penrose–Banzhaf index (PBI) \( \psi_k \) of the \( k \)-th state is defined as the probability that a vote cast by \( k \)-th representative is decisive. This happens if \( k \) is a critical voter in a coalition, i.e. the winning coalition with \( k \) ceases to fulfil the majority requirements without \( k \). Assuming that all \( 2^M \) coalitions are equally likely, we see that the PBI of the \( k \)-th state depends
only on the number $\omega_k$ of winning coalitions that include this state. Namely, the number $\eta_k$ of these coalitions, where a vote of $k$ is decisive, is given by:

$$\eta_k = \omega_k - (\omega - \omega_k) = 2\omega_k - \omega,$$

and so the absolute Penrose-Banzhaf index of the $k$–th state is equal to $\psi_k = \eta_k/2^M - 1$. To compare these indices for decision bodies consisting of different number of players, it is convenient to define the normalised PBIs:

$$\beta_k := \frac{\psi_k}{\sum_{i=1}^M \psi_i} = \frac{\eta_k}{\sum_{i=1}^M \eta_i}$$

($k = 1, \ldots, M$) fulfilling $\sum_{i=1}^M \beta_i = 1$.

In the Penrose voting system one sets the voting weights proportional to the square root of the population of each state, i.e. $w_k = \sqrt{N_k}/\sum_{i=1}^M \sqrt{N_i}$ for $k = 1, \ldots, M$. For any level of the quota $q$ one may compute numerically the power indices $\beta_k$. The Penrose rule would hold perfectly if the voting power of each state would be proportional to the square root of its population. Hence, to quantify the overall representativeness of the voting system one can use the mean discrepancy $\Delta$, defined as the root mean square deviation:

$$\Delta := \sqrt{\frac{1}{M} \sum_{i=1}^M (\beta_i - w_i)^2}.$$  

The optimal quota $q_*$ is defined as the quota for which the mean discrepancy $\Delta$ is minimal. Note that this quota is not unique and usually there is a whole interval of optimal points. However, the length of this interval decreases with increasing number of voters.

Studying the problem for a concrete distribution of population in the European Union, as well as using a statistical approach and analyzing several random distributions of population we found [12, 30] that in these cases all $M$ ratios $\beta_k/w_k$ ($k = 1, \ldots, M$), plotted as a function of the quota $q$, cross approximately near a single point, i.e.

$$\beta_k \approx w_k$$

for $k = 1, \ldots, M$. In other words, the discrepancy $\Delta$ at this critical quota $q_*$ is negligible. The existence of the critical quota was confirmed numerically in a recent study by Chang, Chua, and Machover [32]. (This does not contradict the fact that there is a wide range of quotas, where the mean discrepancy is small [29, 37].) In the next section we propose a toy model, for which a rigorous analysis of this numerical observation is possible.
Consider a voting body of \( M \) members and denote by \( w_k \), \( k = 1, \ldots, M \) their normalized voting weights. Assume now that a single large player with weight \( w_L := w_1 \) is the strongest one, while remaining \( m := M - 1 \) players have equal weights \( w_S := w_2 = \cdots = w_M = (1 - w_L)/m \). We may assume that \( w_L \leq 1/2 \), since in the opposite case, for some values of \( q \), the strongest player would become a ‘dictator’ and his relative voting power would be equal to unity. Furthermore, we assume that the number of small players \( m \) is larger than two, and we introduce a parameter \( c := w_L/w_S \) which quantifies the difference between the large player and the other players. Thus we consider the weighted voting game \([q; c+m, 1/c+m, \ldots, 1/c+m]\), where the population distribution is characterized by only two independent parameters, say, the number of players \( M \) and the ratio \( c \). Sometimes it is convenient to use as a parameter of the model the weight \( w_L \), which is related with the ratio \( c \) by the formula \( c = mw_L/(1 - w_L) \). On the other hand, the qualified majority quota \( q \), which determines the voting system, is treated as a free parameter and will be optimized to minimize the discrepancy (3). Note that a similar model has been analysed in [38].

To avoid odd-even oscillations in the discrepancy \( \Delta (q) \) we assume that \( c \geq 2 \). To compute the PBI\( s \) of all the players we need to analyse three kinds of possible winning coalitions. The vote of the large player is decisive if he forms a coalition with \( k \) of his colleagues, where \( k < mq/(1 - w_L) \) and \( k \geq m(q - w_L)/(1 - w_L) \). Using the notion of the roof, i.e. the smallest natural number larger than or equal to \( x \), written as \( \lfloor x \rfloor := \min \{ n \in \mathbb{N} : n \geq x \} \), we may put

\[
j_1 := \left\lfloor \frac{m(q - w_L)}{1 - w_L} \right\rfloor - 1 \tag{5}
\]

and

\[
j_2 := \left\lfloor \frac{mq}{1 - w_L} \right\rfloor - 1 \tag{6}
\]

and recast the above conditions into the form

\[
j_1 + 1 \leq k \leq j_2 \tag{7}
\]

On the other hand, there exist two cases where the vote of a small player is decisive. He may form a coalition with \( j_2 \) other small players, or, alternatively, he may form a coalition with the large player and \( j_1 \) small players.
With these numbers at hand, we may write down the absolute Penrose–Banzhaf indices for both players. The a priori voting power of the larger player can be expressed in terms of binomial symbols:

$$\psi_L := \psi_1 = 2^{-m} \sum_{k=j_1+1}^{j_2} \binom{m}{k},$$

(8)

while the voting power for all the small players is equal and reads:

$$\psi_S := \psi_2 = \cdots = \psi_M = 2^{-m} \left[ \binom{m-1}{j_1} + \binom{m-1}{j_2} \right].$$

(9)

It is now straightforward to renormalize the above results according to (2) and use the normalized indices $\beta_L$ and $\beta_S$ to write an explicit expression for the discrepancy (3), which depends on the quota $q$. Searching for an ideal system we want to minimize the discrepancy

$$\Delta(q) = \frac{1}{\sqrt{M}} \sqrt{(\beta_L - w_L)^2 + m (\beta_S - w_S)^2}$$

$$= \frac{1}{\sqrt{m}} \left| \beta_L - \frac{c}{c + m} \right|$$

$$= \frac{1}{\sqrt{m}} \left| \frac{\sum_{k=j_1+1}^{j_2} \binom{m}{k}}{m \binom{m-1}{j_1} + \binom{m-1}{j_2}} + \sum_{k=j_1+1}^{j_2} \binom{m}{k} - \frac{c}{c + m} \right|$$

$$= \frac{1}{\sqrt{m}} \left| \frac{\sum_{k=[d-c-1]}^{d-1} \binom{m}{k}}{m \binom{m-1}{[d-c-1]} + \binom{m-1}{[d-1]} + \sum_{k=[d-c]}^{d-1} \binom{m}{k}} - \frac{c}{c + m} \right|,$$

(10)

where $d := mq / (1 - w_L) = (m + c) q$.

In principle, one may try to solve this problem looking first for the optimal $d$, and then computing the optimal quota $q^*$, but due to the roof in the bounds of the sum the general case is not easy to work with.

The problem simplifies significantly if we set $c = 2$, considering the $M$–point weight vector $(w_L, w_L/2, \ldots, w_L/2)$, where $w_L = 2 / (M + 1)$. 

7
In such a case (10) becomes

\[ \Delta(q) = \frac{1}{\sqrt{m}} \left| \frac{m}{r-2} + \frac{m}{r-1} + m \left( \frac{m-1}{r-3} + \frac{m-1}{r-1} \right) - \frac{2}{m+2} \right| \]

\[ = \frac{1}{(m+2)\sqrt{m}} \left| \frac{m^2 - 4mr + 5m + 4r^2 - 12r + 8}{m^2 - 2mr + 4m + 2r^2 - 6r + 5} \right|, \]  

(11)

where \( r := \lceil d \rceil = \lceil q(M+1) \rceil \). To analyse this dependence we introduce a new variable

\[ t := r - M/2 - 1 = \lceil q(M+1) \rceil - M/2 - 1 , \]

(12)

obtaining

\[ \Delta(t) = \frac{2}{(M+1)\sqrt{M-1}} \left| \frac{M - 4t^2}{M^2 + 4t^2} \right| \]

\[ = \frac{4}{(M+1)\sqrt{M-1}} \left| \frac{\sqrt{M} + 2t}{M^2 + 4t^2} \right| \left| \sqrt{M}/2 - t \right| . \]  

(13)

In principle, one can minimize this expression finding \( \min \Delta(t) = 0 \) for \( t_* = \sqrt{M}/2 \), see Fig.1a. However, due to the presence of the roof function in (12), \( \Delta(q) \) is not a continuous function of the quota, and, consequently, the optimization problem \( \min \Delta(q) \) does not have a unique solution and the minimal value may be greater than 0, see Fig.1b. Nevertheless, applying (12) and (13), one can show that there exists an optimal quota \( q_*(M) \) in the interval

\[ \frac{M + \sqrt{M}}{2(M+1)} \leq q_*(M) \leq \frac{2 + M + \sqrt{M}}{2(M+1)} . \]  

(14)

This means that for a large number \( M \) of players the optimal quota behaves exactly as

\[ q_*(M) \simeq q_*(M) := \frac{1}{2} \left( 1 + \frac{1}{\sqrt{M}} \right) . \]  

(15)

Although this is an asymptotic formula, it works also for a moderate number of states. Moreover, it follows from (11) and (14) that the minimal mean discrepancy \( \Delta(q_*(M)) \leq 8/M^3 \).
Figure 1: a) The ‘mean discrepancy’ $\Delta(t)$ as a function of the parameter $t$; b) The mean discrepancy $\Delta(R)$ as a function of the parameter $R$ (in both cases $c = 2, M = 27$)

Surprisingly, the efficiency of the system given by

$$A(q_s(M)) = \frac{\sum_{k=r(M)-2}^{M-1} \binom{M-1}{k} + \sum_{k=r(M)-1}^{M-2} \binom{M-2}{k}}{2^M}, \quad (16)$$

where $r(M) := \lceil (M + 1) q_s(M) \rceil$, does not decrease with the number of players to 0. On the contrary, it is always larger than $15/128 \approx 0.117$ and tends to $\frac{3}{4}(1 - \Phi(1)) \approx 0.119$ for $M \to \infty$.

Analogous considerations for $c = 3$ give similar result:

$$\frac{1 + M + \sqrt{M}}{2(M+1)} \leq q_s(M) \leq \frac{3 + M + \sqrt{M}}{2(M+1)}, \quad (17)$$

and so, also in this case, $q_s(M) \approx \frac{1}{2} (1 + 1/\sqrt{M})$.

4 Normal approximation

Let us have a closer look at the approximate formula (15) for the optimal quota. In the limit $M \to \infty$ the optimal quota tends to $1/2$ in agreement with the Penrose limit theorem [34, 35]. Numerical values of the approximate optimal quota $q_s$ obtained in our toy model for $c = 2$ and $c = 3$ are consistent, with an accuracy up to two per cent, with the data obtained numerically by
averaging quotas over a sample of random weights distributions \cite{12,33}.

Furthermore, the above results belong to the range of values of the quota for qualified majority, which have been used in practice or recommended by experts on designing the voting systems.

Consider now a voting body of \( M \) members and denote by \( w_k, \ k = 1, \ldots, M \), their normalized voting weights fulfilling \( \sum_{i=1}^{M} w_i = 1 \). Feix et al. proposed (also in the EU context) yet another method of estimating the optimal quota for the weighted voting game \([q; w_1, \ldots, w_M]\), where \( q \in [0.5, 1] \) \cite{34}. They considered the histogram \( n \) of the sum of weights (number of votes) achieved by all possible coalitions

\[
N(q) := \sum_{z \leq q} n(z) \approx \int_{-\infty}^{q} \frac{1}{\sigma \sqrt{2\pi}} \exp\left(-\frac{(x-m)^2}{2\sigma^2}\right) dx = \Phi\left(\frac{q-m}{\sigma}\right),
\]

and assumed that it allows the normal approximation with the mean value \( m = \frac{1}{2} \sum_{i=1}^{M} w_i = \frac{1}{2} \) and the variance \( \sigma^2 = \frac{1}{4} \sum_{i=1}^{M} w_i^2 \), i.e.

\[
\psi_k \approx \sqrt{\frac{2}{\pi e}} \frac{w_k}{\sqrt{\sum_{i=1}^{M} w_i^2}} \quad \text{(20)}
\]

and, in consequence,

\[
\beta_k \approx w_k \quad \text{(21)}
\]

\(^2\)Nevertheless, one can construct an artificial model with very different values of optimal quota. In this aim, it is enough to consider one ‘small’ state and an even number of ‘large’ states with equal population (i.e. \( c < 1 \) in our toy model), see \cite{34,35}. As Lindner stressed: ‘experience suggests that such counter-examples are atypical, contrived exceptions’.
The validity of this method depends on the accuracy of the normal approx-
imation for the absolute Banzhaf indices (see Appendix). The necessary
condition for the latter is
\[
\max_{j=1,\ldots,M} w_j \ll \sqrt{\sum_{i=1}^{M} w_i^2}.
\] (22)

For the thorough discussion of the problem see [39, 40, 34, 31]. For the
Penrose voting system, where \( w_k \sim \sqrt{N_k} \) (\( k = 1, \ldots, M \)), (22) is equivalent
to
\[
\max_{j=1,\ldots,M} N_j \ll \sum_{i=1}^{M} N_i,
\] (23)

which means that the population of each country is relatively small when
compared with the total population of all countries. One can easily check
that it is more likely that (22) holds in this case than when the weights are
proportional to the population.

By making use of (21) we arrive at the following weights-dependent ap-
proximation formula for the optimal quota:
\[
q_s \simeq q_n (w_1, \ldots, w_M) := \frac{1}{2} \left( 1 + \sqrt{\sum_{i=1}^{M} w_i^2} \right).
\] (24)

This approximation of the optimal quota can be directly compared with the
approximation (15) obtained for the toy model. Since \( \sum_{i=1}^{M} w_i = 1 \) implies
\( \sum_{i=1}^{M} w_i^2 \geq 1/M \), it follows that
\[
q_s (M) = \frac{1}{2} \left( 1 + \frac{1}{\sqrt{M}} \right) \leq \frac{1}{2} \left( 1 + \frac{1}{\sqrt{M_{\text{eff}}}} \right) = q_n,
\] (25)

where \( M_{\text{eff}} := 1/\sum_{i=1}^{M} w_i^2 \) is equal to the effective number of players. (This
quantity was introduced by Laakso and Taagepera [41] and is the inverse of
the more widely used Herfindahl–Hirschman index of concentration [42, 43],
see also [44].) The equality in (25) holds if and only if all the weights are
equal. For the Penrose voting system we have
\[
q_n = \frac{1}{2} \left( 1 + \sqrt{\frac{\sum_{i=1}^{M} N_i}{\sum_{i=1}^{M} \sqrt{N_i}}} \right),
\] (26)
where \( N_k \) stands for the population of the \( k \)-th country. For the toy model we get \( q_n = \frac{1}{2} \left( 1 + \sqrt{\frac{M+c-1}{M+c-1}} \right) \approx q_s (M) \) for large \( M \).

Both approximations \( q_s \) and \( q_n \) are consistent with an accuracy up to two per cent, with the optimal quotas \( q_s \) obtained for the Penrose voting system applied retrospectively to the European Union (see Tab. 1 below). Observe that in this case the approximation of the optimal quota \( q_s \) by \( q_n \) is better for larger number of states, where the normal approximation of the histogram is more efficient.

<table>
<thead>
<tr>
<th>( M )</th>
<th>15</th>
<th>25</th>
<th>27</th>
</tr>
</thead>
<tbody>
<tr>
<td>year</td>
<td>1995</td>
<td>2004</td>
<td>2007</td>
</tr>
<tr>
<td>( q_s ) [%]</td>
<td>62.9</td>
<td>60.0</td>
<td>59.6</td>
</tr>
<tr>
<td>( q_s ) [%]</td>
<td>64.4</td>
<td>62.0</td>
<td>61.5</td>
</tr>
<tr>
<td>( q_n ) [%]</td>
<td>64.9</td>
<td>62.2</td>
<td>61.6</td>
</tr>
</tbody>
</table>

Tab. 1. Comparison of optimal quotas for the Penrose voting system applied to the EU (\( q_s \)) and for two approximations (\( q_s, q_n \)).

Applying the normal approximation one can easily explain why the efficiency \( A \) of our system does not decrease when the number of players \( M \) increases. We have

\[
A(q_s) \geq A(q_n) \approx 1 - \mathcal{N}(q_n) \approx \int_{m+\sigma}^{\infty} \frac{1}{\sigma \sqrt{2\pi}} \exp \left( -\frac{(x - m)^2}{2\sigma^2} \right) \, dx . \tag{27}
\]

The right-hand side of this inequality does not depend neither on \( m \) nor on \( \sigma \), and it equals \( 1 - \Phi(1) \approx 0.159 \), where \( \Phi \) is the standard normal cumulative distribution function.

### 5 Complete voting system

We shall conclude this paper proposing a complete voting system based on the Penrose square root law and the toy model. The system consists of a single criterion only and is determined by the following two rules:

A. The voting weight attributed to each member of the voting body of size \(M\) is proportional to the square root of the population he or she represents;

B. The decision of the voting body is taken if the sum of the weights of members of a coalition exceeds the quota \(q_s = (1 + 1/\sqrt{M})/2\).

Alternatively, one can set the quota to \(q_n = (1 + (\sum_{i=1}^{M} w_i^2)^{1/2})/2\), if the weights are known, or just take the optimal quota \(q^*\) which, however, requires more computational effort.

Such a voting system is extremely simple, since it is based on a single criterion. It is objective and so cannot a priori handicap a given member of the voting body. The quota for qualified majority is considerably larger than 50\% for any size of the voting body of a practical interest. Thus the voting system is also moderately conservative. Furthermore, the system is representative and transparent: the voting power of each member of the voting body is (approximately) proportional to his voting weight. However, as a crucial advantage of the proposed voting system we would like to emphasize its extendibility: if the size \(M\) of the voting body changes, all one needs to do is to set the voting weights according to the square root law and adjust the quota accordingly. As the number \(M\) grows, the efficiency of the system does not decrease.

The formulae for the quotas \(q_s(M)\) and \(q_n\) can be also applied in other weighted voting games. Note that for a fixed number of players the quota \(q_s(M)\) does not depend on the particular distribution of weights in the voting body. This feature may be relevant, e.g. for voting bodies in stock companies where the voting weights of stockholders depend on the proportion of stock that investors hold and may vary frequently.

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References


Appendix

Consider a weighted voting game \( [q; w_1, \ldots, w_M] \), where \( q \in [0.5, 1] \) and \( \sum_{i=1}^{M} w_i = 1 \). Set \( m := \frac{1}{2} \sum_{i=1}^{M} w_i = \frac{1}{2} \) and \( \sigma^2 := \frac{1}{4} \sum_{i=1}^{M} w_i^2 \). Let \( j = 1, \ldots, M \). We put \( m_j := m - w_j/2 \) and \( \sigma_j^2 := \sigma^2 - w_j^2/4 \).

Assume that for \( j = 1, \ldots, M \) and \( q \in [0.5, 1] \) the absolute Banzhaf index

\[
\psi_j = \Pr \left( \left\{ I \subset \{1, \ldots, M\} : q - w_j \leq \sum_{i \in I, i \neq j} w_i < q \right\} \right) \tag{A1}
\]

admits the following normal approximation:

\[
\psi_j \approx \Phi \left( \frac{q - m_j}{\sigma_j} \right) - \Phi \left( \frac{q - w_j - m_j}{\sigma_j} \right), \tag{A2}
\]

where \( \Phi (\cdot; \mu, d) \) stands for the normal cumulative distribution function with mean \( \mu \) and standard deviation \( d \). Hence

\[
\psi_j \approx \Phi \left( \frac{q - m_j}{\sigma_j} \right) - \Phi \left( \frac{q - w_j - m_j}{\sigma_j} \right). \tag{A3}
\]

Put \( q = q_n := m + \sigma \). Then

\[
\psi_j \approx \Phi \left( \frac{m + \sigma - m_j}{\sigma_j} \right) - \Phi \left( \frac{m + \sigma - w_j - m_j}{\sigma_j} \right) = \Phi \left( \frac{\sigma + \frac{1}{2} w_j}{\sigma_j} \right) - \Phi \left( \frac{\sigma - \frac{1}{2} w_j}{\sigma_j} \right) = \Phi \left( \sqrt{\frac{1 + v_j}{1 - v_j}} \right) - \Phi \left( \sqrt{\frac{1 - v_j}{1 + v_j}} \right), \tag{A4}
\]

where \( v_j := w_j / 2\sigma = w_j / \sqrt{\sum_{i=1}^{M} w_i^2} \). If \( w_j \ll \sqrt{\sum_{i=1}^{M} w_i^2} \), then \( v_j \ll 1 \), and both \( \sqrt{(1 + v_j) / (1 - v_j)} \) and \( \sqrt{(1 - v_j) / (1 + v_j)} \) are close to 1. Near this point the standard normal density function \( \Phi' \) is almost linear and so

\[
\Phi \left( \frac{\sigma + \frac{1}{2} w_j}{\sigma_j} \right) - \Phi \left( \frac{\sigma - \frac{1}{2} w_j}{\sigma_j} \right) \approx \Phi' \left( \frac{\sigma}{\sigma_j} \right) \frac{w_j}{\sigma_j}. \tag{A5}
\]
From \( (A4) \) and \( (A5) \) we deduce that

\[
\psi_j \approx \Phi' \left( \frac{\sigma}{\sigma_j} \frac{w_j}{\sigma_j} \right) = 1 \sqrt{\frac{2}{\pi}} \exp \left( -\frac{\sigma^2}{2\sigma_j^2} \right) = \sqrt{\frac{2}{\pi}} \left( \sum_{i=1}^{M} w_i^2 \right) \left( \sum_{i=1}^{M} w_i^2 \right) - w_j^2 \exp \left( -\frac{1}{2 (1 - v_j^2)} \right) = \sqrt{\frac{2}{\pi} e v_j + o (v_j^2)} . \tag{A6}
\]

Consequently,

\[
\psi_j \approx \sqrt{\frac{2}{\pi e}} \left( \sum_{i=1}^{M} w_i^2 \right) + o (v_j^2) , \tag{A7}
\]

and so

\[
\frac{\beta_j}{w_j} \approx 1 . \tag{A9}
\]

The accuracy of this approximation depends highly on the accuracy of the normal approximation \( (A2) \).

Note that for the quota \( q = 1/2 \) we get (see [34, 35] for the formal proof)

\[
\psi_j \approx \Phi' (0) \frac{w_j}{\sigma_j} = \sqrt{\frac{2}{\pi}} \sqrt{\left( \sum_{i=1}^{M} w_i^2 \right) - w_j^2} + a (v_j^2) . \tag{A10}
\]

Hence in this case \( \beta_j \) need not be as close to \( w_j \) as for \( q = q_n \).