# Power measures and the sequential query process 

Mark C. Wilson<br>UMass Amherst

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## TU games and simple games

- A monotone TU-game is a cooperative game $G=(X, v)$ given in coalitional form: for the player set $X$ there is a map $v: 2^{X} \rightarrow \mathbb{R}$ with $v(\emptyset)=0$, and $A \subseteq B$ implies $v(A) \leq v(B)$.


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\partial_{i} v(A)=v(A)-v(A \backslash\{i\}) .
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## Motivating example

- Devise a less crude measure of the manipulability of a voting rule.
$\rightarrow$ The most commonly used measure simply gives 0 or 1 for a given profile, depending on whether it is manipulable by some coalition or not.
$\rightarrow$ A more sophisticated measure is the size of the minimum manipulating coalition.
- W/e want to see how easy it would be to assemble a manipulating coalition - the query model is one idea.
- Concrete example: Borda rule, 2 voters $a b c, 1$ voter $b a c, 1$ voter cba.


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## Collective and individual measures

- The process is equivalently described via first choosing a probability measure $\mu_{n}$ on $\{0, \ldots, n\}$, sampling from it and then choosing a subset of that size uniformly at random.

The function $Q_{F}^{*}(G)$ is a measure of decisiveness that generalizes Coleman's index.

- The marginal $q_{F}^{*}(G)$ of $Q_{F}^{*}(G)$ is a power index.
$\rightarrow$ The simplest $F$ (corresponding to uniformly choosing the size of the subset) gives a new measure, that we call $Q_{0}^{*}(G)$
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## Generalization to TU-games

- The obvious definition is

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Q_{F}^{*}(G)=\sum_{S \subseteq X}\binom{n}{|S|} \mu_{n}(|S|) v(S)
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| Name | $F(n, k)$ | $\mu_{n}(k)$ | Value on voting game |
| :--- | :--- | :--- | :--- |
| Coleman/Banzhaf | $2^{-n} \sum_{j \geq k}\binom{n}{j}$ | $\operatorname{Bin}(1 / 2)$ | $3 / 4=0.75$ |
| Shapley | $1-H_{k-1} / H_{n}$ | $\operatorname{Unif}(1 . . n)$ | $36 / 50=0.72$ |
| $Q_{0}^{*} / q_{0}^{*}$ | $1-k /(n+1)$ | $\operatorname{Unif}(0 . . n)$ | $2 / 3 \approx 0.67$ |

## Another connection to the Shapley value

- In applications of Shapley the grand coalition is always winning and we seek to divide up the surplus.
$\Rightarrow$ In some applications the grand coalition is not winning (e.g. the voting game example above).
- If we allow this, our $q_{F}^{*}$ are precisely analogous to semivalues and the characterization theorem of Dubey, Neyman \& Weber (1981) extends naturally.
- In this new model, $a_{0}^{*}$ is the exact analog of the Shapley value.


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## Future work

- An entire class of values/power indices and potentials/decisiveness indices remains to be explored.
$\rightarrow$ Axiomatic characterization of these new values.
- Finding applications of new values such as $q_{0}^{*}$. We give applications to manipulability of voting rules. Perhaps machine learning applications?
- $Q_{0}^{*}$ appears better than Coleman in discriminating between simple games, and should be studied more.


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