# Mathematical programming in computational social choice 

Gerdus Benade
benade@bu.edu


# Preference elicitation for participatory budgeting 



Gerdus Benade

# Preference elicitation for participatory budgeting 



Swaprava Nath


Ariel Procaccia


Nisarg Shah


Kobi Gal

## Let's warm up with a budgeting problem

- Coffee, transport and relaxation
- Each provides a different value per dollar spent $\left(v_{c}, v_{r}, v_{t}\right)$
- Allocate $x_{c}, x_{r}, x_{t}$ to each category
- Maximize utility gained from budget $B$

$$
\begin{array}{ll}
\max & v_{c} x_{c}+v_{r} x_{r}+v_{t} x_{t} \\
\text { s.t. } & x_{r}+x_{c}+x_{t} \leq B \\
& x_{r}, x_{c}, x_{t} \geq 0
\end{array}
$$



The Participatory Budgeting Project empowers people to decide together how to spend public money to deepen democracy, build stronger communities, and make public budgets more equitable and effective.

## PB BY THE NUMBERS


in public funding allocated through PB.


PB participants across the US \& Canada.

community-generated winning projects.

## CITY OF CAMBRIDGE PARTICIPATORY BUDGETING



How should voters express their preferences?

How should the votes be aggregated?

Can we limit the cognitive burden on voters?

## Model

- $n$ residents $N, m$ projects $A$
- Project $a$ has cost $c_{a}$, global budget $B=1$
- Voter $i$ has utility function $v_{i}$
- Utilities are additive

$$
v_{i}(S)=\sum_{a \in S} v_{i}(a) \text { for every } S \subseteq A
$$

## Model

- $n$ residents $N, m$ projects $A$
- Project $a$ has cost $c_{a}$, global budget $B=1$
- Voter $i$ has utility function $v_{i}$
- Utilities are additive and normalized

$$
v_{i}(A)=1 \text { for all } i \in N
$$

## Model

- $n$ residents $N, m$ projects $A$
- Project $a$ has cost $c_{a}$, global budget $B=1$
- Voter $i$ has utility function $v_{i}$, utility $v_{i a}$ for alternative $a$
- Utilities are additive and normalized

$$
\begin{gathered}
\max _{S} \quad \begin{array}{c}
\operatorname{sw}(S)=\sum_{a \in S} \sum_{i} v_{i}(a) \\
\text { subject to } c(S) \leq B
\end{array}, ~
\end{gathered}
$$

## Interlude: the knapsack problem

$$
\begin{array}{ll}
\max _{x} & \sum_{a}\left(\sum_{i} v_{i a}\right) x_{a} \\
\text { subject to } \quad \sum_{a} c_{a} x_{a} \leq B
\end{array}
$$

$$
\operatorname{KNAP}(v)
$$

## Interlude: the knapsack problem

$$
\begin{aligned}
& \max _{x} \quad \sum_{a}\left(\sum_{i} v_{i a}\right) x_{a} \\
& \text { subject to } \quad \sum_{a} c_{a} x_{a} \leq B \\
& x_{a} \in\{0,1\} \quad \forall a .
\end{aligned}
$$

$\operatorname{KNAP}(v)$

- Notice integer variables $x_{a}$
- LP solvable in polynomial time, IP takes exponential time in general
- Knapsack problems are 'easy': thousands of items takes seconds


## Model

- $n$ residents $N, m$ projects $A$
- Project $a$ has cost $c_{a}$, global budget $B=1$
- Voter $i$ has utility function $v_{i}$, utility $v_{i a}$ for alternative $a$
- Utilities are additive and normalized

$$
\begin{array}{cc}
\max _{S} & \operatorname{sw}(S)=\sum_{a \in S} \sum_{i} v_{i}(a) \\
& \text { subject to } c(S) \leq B
\end{array}
$$

- Assume voter $i$ submits vote $\rho_{i}$ consistent with $v_{i}: v_{i} \triangleright \rho_{i}$


## $\begin{aligned} & \text { Ranking } \\ & \text { by value }\end{aligned}>(0$ (ช)

## $\begin{aligned} & \begin{array}{l}\text { Ranking } \\ \text { by VFM }\end{array}>\text { ลix (0) }\end{aligned}>$



## Implicit utilitarian voting

- Randomized voting rule $f$ vote profile $\vec{\rho}$ as input, returns distribution over feasible sets of projects $f(\vec{\rho})$

$$
\operatorname{dist}(f, \vec{\rho})=\max _{\vec{v}: \vec{v} \triangleright \vec{\rho}} \frac{\max _{T: c(T) \leq B S W(T, \vec{v})}}{\mathbb{E}[S W(f(\vec{\rho}), \vec{v})]}(\text { Social welfare of } f \text { on } \vec{v})
$$

## Implicit utilitarian voting

- Randomized voting rule $f$ vote profile $\vec{\rho}$ as input, returns distribution over feasible sets of projects $f(\vec{\rho})$

$$
\begin{aligned}
& \operatorname{dist}(f, \vec{\rho})=\max _{\vec{v}: \vec{v} \triangleright \vec{\rho}} \frac{\max _{T: c(T) \leq B} S W(T, \vec{v})}{\mathbb{E}[\operatorname{sw}(f(\vec{\rho}), \vec{v})]} \\
& \operatorname{dist}(f)=\max _{\vec{p}} \max _{\vec{v}} \frac{\max _{T: c(T) \leq B} S w(T, \vec{v})}{\mathbb{E}[s w(f(\vec{\rho}), \vec{v})]} \\
& \begin{array}{l}
\text { Worst case } \\
\text { over inputs }
\end{array}
\end{aligned}
$$

## Decouple input format and aggregation

- Distortion of a voting rule:

$$
\operatorname{dist}(f)=\max _{\vec{\rho}} \max _{\vec{v}: \vec{v} \triangleright \vec{\rho}} \frac{\max _{T: c(T) \leq B} s w(T, \vec{v})}{\mathbb{E}[s w(f(\vec{\rho}), \vec{v})]}
$$

- Distortion of an input format: the distortion of its best voting rule

| Knapsack voting |  | $\widetilde{\Theta}(m)$ |
| :--- | :--- | :--- | :--- |
| Ranking by value | $\widetilde{\Theta}(\sqrt{m})$ |  |
| Ranking by VFM | $\widetilde{\Theta}(\sqrt{m})$ |  |
| Threshold approval | $0\left(\log ^{2} m\right)$ |  |

How should voters express their preferences?

How should the votes be aggregated?

Can we limit the cognitive burden on voters?

## How should the votes be aggregated?

Current practice: greedy aggregation based on number of approvals/appearances in a knapsack etc. (Goel et al.)

Use the input-specific distortion-minimizing voting rules:
Deterministic: $\quad f^{*}(\vec{\rho})=\underset{S}{\operatorname{argmin}} \max _{\vec{v}: \vec{v} \triangleright \vec{\rho}} \frac{\max _{T: c(T) \leq B} s w(T, \vec{v})}{s w(S, \vec{v})}$

## Distortion-minimizing voting rules via LP

Deterministic: $\quad f^{*}(\vec{\rho})=\operatorname{argmin} \max _{\vec{v}: \vec{v} \triangleright \vec{\rho}} \frac{\max _{T: c c}(T) \leq B s w(T, \vec{v})}{s w(S, \vec{v})}$

For every $S$ :
For every $\vec{v}: \vec{v} \triangleright \vec{\rho}$ :
Compute optimal solution to $\operatorname{KNAP}(\vec{v}), T$
Keep track of worst ratio $\operatorname{sw}(\mathrm{T}, \vec{v}) / \operatorname{sw}(S, \vec{v})$

Return $S$ with smallest worst-case ratio

## Distortion-minimizing voting rules via LP

Deterministic: $\quad f^{*}(\vec{\rho})=\operatorname{argmin} \max _{\vec{v}: \vec{v} \triangleright \vec{\rho}} \frac{\max _{T: c(T) \leq B} s w(T, \vec{v})}{s w(s, \vec{v})}$

For every $S$ :
For every $T$ :
Compute $\max _{\vec{v}: \vec{v} \triangleright \vec{\rho}} \operatorname{sw}(\mathrm{~T}, \vec{v}) / \operatorname{sw}(S, \vec{v})$

Return $S$ with smallest worst-case ratio

## Inner LP (S, $T$ fixed)

- $x_{S}, x_{T}$ are characteristic vectors of sets $S, T$ (fixed)
- $|A|$-dimensional
- $x_{S}(a)=1$ if alternative $a$ is in set $S$
- Variables $v_{i}(a)$ for every voter $i$, alternative $a$

$$
\begin{aligned}
& \max \frac{\sum_{a} \sum_{i} v_{i}(a) x_{T}(a)}{\sum_{a} \sum_{i} v_{i}(a) x_{S}(a)} \\
& \text { s.t. } \vec{v} \geq 0, \vec{v} \triangleright \vec{\rho}
\end{aligned}
$$

## Consistency constraints are linear

- For example, for a ranking $\rho_{i}$
- $v_{i}\left(\rho_{i}(1)\right) \geq v_{i}\left(\rho_{i}(2)\right) \geq \cdots \geq v_{-} i\left(\rho_{i}(m)\right)$
- Non-negativity + normalization constraints
- For threshold approval votes with threshold $t$
- $v_{i}(a) \geq t$ if $a$ approved
- $v_{i}(a) \leq t$ if not approved
- Non-negativity + normalization constraints


## Distortion-minimizing voting rules via LP

## For every $S$ :

For every $T$ :

$$
\begin{array}{r}
\text { Solve } \quad \max \frac{\sum_{a} \sum_{i} v_{i}(a) x_{T}(a)}{\sum_{a} \sum_{i} v_{i}(a) x_{S}(a)} \\
\text { s.t. } \vec{v} \geq 0, \vec{v} \triangleright \vec{\rho}
\end{array}
$$

LP with fractional objective:
Charnes-Cooper transformation

Return $S$ with smallest worst-case ratio

## Experiments

- Real-world participatory budgeting elections
- Held in Cambridge (MA) in 2015 and 2016
- Data provided by Ashish Goel and the Stanford Crowdsourced Democracy Team
- 10 projects, 4000 voters
- Real votes $\vec{\rho} \rightarrow$ consistent utility $\vec{v} \rightarrow$ votes in all formats $\rightarrow$ aggregate
- Measure social welfare ratio, compare 4 formats + greedy baselines

$$
\left(\frac{\max s w(T, \vec{v})}{\operatorname{sw}(f(\vec{\rho}), \vec{v})}\right)
$$




## How to Make Envy Vanish over Time



Alex Psomas


Aleks Kazachkov


Ariel Procaccia

How can we fairly allocate goods that arrive over time?

## How can we fairly]allocate goods that arrive over time?

Envy-freeness

No agent prefers another allocation to their own.

$$
E N V Y^{i j}=\max \left\{v_{i}\left(A_{j}\right)-v_{i}\left(A_{i}\right), 0\right\}
$$



## How can we (fairly)allocate goods that arrive over time?

Envy-freeness

No agent prefers another allocation to their own.

$$
E N V Y^{i j}=\max \left\{v_{i}\left(A_{j}\right)-v_{j}\left(A_{i}\right), 0\right\}
$$

Indivisible goods: EF1 (Envy-free up to 1 good)






$$
\left(E N V Y^{i j}=\max \left\{v_{i}\left(A_{j}\right)-v_{j}\left(A_{i}\right), 0\right\}\right)
$$

$$
E N V Y_{3}^{R B}=1.5-1=0.5
$$

$$
E N V Y_{3}^{B R}=1-0.5=0.5
$$



1

## Model

- Items arrive online in batches of size $k$
- $n$ agents
- Agent $i$ has value $v_{i t} \in[0,1]$ for item $t$
- Allocate to minimize $E\left[E N V Y_{T}\right]$
- Item values are chosen by adaptive adversary (maximizes $\mathrm{E}\left[E N V Y_{T}\right]$ )

Can we ensure vanishing envy after $T$ items?

$$
\left(\lim _{T \rightarrow \infty} \frac{E N V Y_{T}}{T}=0\right)
$$

## Can we ensure vanishing envy at time $T$ ?

$T$ batches of size 1
$T / k$ batches of size $k$


## Greedy deterministic allocation

Give to whoever has the highest envy

| Value agent 1 | $1 / 2$ | 1 | $\epsilon$ | 1 | $\epsilon$ | $\ldots$ |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| Value agent 2 | $1 / 2$ | $1 / 4$ | 1 | $\epsilon$ | 1 | $\ldots$ |

Envy agent 1

Envy agent 2

## Greedy deterministic allocation

Give to whoever has the highest envy

| Value agent 1 | $1 / 2$ | 1 | $\epsilon$ | 1 | $\epsilon$ | $\ldots$ |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| Value agent 2 | $1 / 2$ | $1 / 4$ | 1 | $\epsilon$ | 1 | $\ldots$ |

Envy agent 1

Envy agent 2

## Greedy deterministic allocation

Give to whoever has the highest envy

| Value agent 1 | $1 / 2$ | 1 | $\epsilon$ | 1 | $\epsilon$ | $\ldots$ |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| Value agent 2 | $1 / 2$ | $1 / 4$ | 1 | $\epsilon$ | 1 | $\ldots$ |
| Envy agent 1 | $-1 / 2$ |  |  |  |  |  |
| Envy agent 2 | $1 / 2$ |  |  |  |  |  |

## Greedy deterministic allocation

Give to whoever has the highest envy

| Value agent 1 | $1 / 2$ | 1 | $\epsilon$ | 1 | $\epsilon$ | $\ldots$ |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| Value agent 2 | $1 / 2$ | $1 / 4$ | 1 | $\epsilon$ | 1 | $\ldots$ |
| Envy agent 1 | $-1 / 2$ | $1 / 2$ |  |  |  |  |
| Envy agent 2 | $1 / 2$ | $1 / 4$ |  |  |  |  |

## Greedy deterministic allocation

Give to whoever has the highest envy

| Value agent 1 | $1 / 2$ | 1 | $\epsilon$ | 1 | $\epsilon$ | $\ldots$ |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| Value agent 2 | $1 / 2$ | $1 / 4$ | 1 | $\epsilon$ | 1 | $\ldots$ |
| Envy agent 1 | $-1 / 2$ | $1 / 2$ | $1 / 2-\epsilon$ |  |  |  |
| Envy agent 2 | $1 / 2$ | $1 / 4$ | $5 / 4$ |  |  |  |

## Greedy deterministic allocation

Give to whoever has the highest envy

| Value agent 1 | $1 / 2$ | 1 | $\epsilon$ | 1 | $\epsilon$ | $\ldots$ |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| Value agent 2 | $1 / 2$ | $(1 / 4$ | 1 | $\epsilon$ | 1 | $\ldots$ |
| Envy agent 1 | $-1 / 2$ | $1 / 2$ | $1 / 2-\epsilon$ | $3 / 2-\epsilon$ |  |  |
| Envy agent 2 | $1 / 2$ | $1 / 4$ | $5 / 4$ | $5 / 4-\epsilon$ |  |  |

## Greedy deterministic allocation

Give to whoever has the highest envy

| Value agent 1 | $1 / 2$ | 1 | $\epsilon$ | 1 | $\epsilon$ | $\cdots$ |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| Value agent 2 | $1 / 2$ | $(1 / 4$ | 1 | $\epsilon$ | 1 | $\cdots$ |
| Envy agent 1 | $-1 / 2$ | $1 / 2$ | $1 / 2-\epsilon$ | $3 / 2-\epsilon$ | $3 / 2-2 \epsilon$ | $\cdots$ |
| Envy agent 2 | $1 / 2$ | $1 / 4$ | $5 / 4$ | $5 / 4-\epsilon$ | $9 / 4-\epsilon$ | $\cdots$ |

$\left(E N V Y_{T} \approx T / 2\right)$

## Items arrive one at a time

- Pretend that every agent values every item at 1.)
- Allocate every item uniformly at random.
- Max change in $E N V Y_{i j}$ is 1 . T items arrive, $i$ gets it w.p. $1 / n$
- Random walk with step size $1, \mathrm{~T} / \mathrm{n}$ steps. Deviation $\sim \sqrt{T / n}$

Theorem. Allocating unit-valued items uniformly at random yields $E N V Y_{T} \in$ $O(\sqrt{T \log n / n})$ with high probability, and $E\left[E N V Y_{T}\right] \in O(\sqrt{T \log T / n})$.

## Items arrive in batches


(1, 1/2, 1/4)
$(1,1,1)$

(1/2, 0,1/2)
(0, 1/3,1/2)

(1/8, 3/4,1)



Next batch arrives
( $T / k$ batches)

## Intuition

- When $k=1, E N V Y_{T} \in \widetilde{\Theta}(\sqrt{T / n})$
- When $k=T, E N V Y_{T} \leq 1$

$$
k=1
$$

- Envy changed by $\leq 1$ per round
- $E N V Y^{i j}$ changed $\approx T / n$ times
- Bound: $E N V Y_{T} \in \widetilde{\Theta}(1 \cdot \sqrt{T / n})$

General $k$

- Envy changes by $\leq 1$ per round
- ENVY ${ }^{i j}$ may change $T / k$ times
- Bound: $E N V Y_{T} \in \widetilde{\Theta}(\sqrt{T / k})$ ?

Theorem. $E N V Y_{T, k} \in \Omega(\sqrt{T / k n})$ and $E N V Y_{T, k} \in \tilde{O}(\sqrt{T / k})$.

## Upper bound (batches)

## Problems

## General $k$

- Envy changes by $\leq 1$ per round
- ENVY ${ }^{i j}$ may change $T / k$ times
- Bound: $E N V Y_{T} \in \widetilde{\Theta}(\sqrt{T / k})$ ?
- Can we find 'balanced' allocations across batches?


We don't need EF1 (step size 1) per batch, any bounded constant change in envies will do.

Let's try rounding near-integral envy-free allocations.

## Finding near integral envy-free solutions

- $x_{i j}$ fraction of item $j$ given to agent $i$
- $v_{i j} \leq 1$ utility of agent $i$ for item $j$

How many fractional values?
Number of variables: $n m$
Number of const: $\mathrm{n}^{2}-\mathrm{n}+m$ So $\leq n^{2}-n+m$ pos.
variables
(Basically) tight, so $\sim n$
fractional items per agent
Rounding introduces $n$ envy

Feasible solution is envy-free. We have to round it to get allocation Rounding introduces envy: up to 1 per rounded item per agent

Theorem (Stromquist, 1980). Suppose $n$ agents have 'reasonable' valuations over $[0,1]$. Then there exists an envy-free division of the interval in which every agent receives a single contiguous segment.

Theorem (Stromquist, 1980). Suppose $n$ agents have 'reasonable' valuations over $[0,1]$. Then there exists an envy-free division of the interval in which every agent receives a single contiguous segment.


Theorem (Stromquist 1980). Suppose $n$ agents have `reasonable' valuations over $[0,1]$. Then there exists an envy-free division of the interval in which every agent receives a single contiguous segment.


Theorem (Stromquist 1980). Suppose $n$ agents have 'reasonable' valuations over $[0,1]$. Then there exists an envy-free division of the interval in which every agent receives a single contiguous segment.


- No player receives more than two fractional items.

Theorem (Stromquist 1980). Suppose $n$ agents have 'reasonable' valuations over $[0,1]$. Then there exists an envy-free division of the interval in which every agent receives a single contiguous segment.


- No player receives more than two fractional items.
- Rounding fractional items randomly guarantees envy changes by $\leq 4$

Theorem (Stromquist 1980). Suppose $n$ agents have `reasonable' valuations over $[0,1]$. Then there exists an envy-free division of the interval in which every agent receives a single contiguous segment.

- $x_{i j}$ fraction of item $j$ given to agent $i$
- $v_{i j} \leq 1$ utility of agent $i$ for item $j$
- Indicator variables $x_{i j}^{0}, x_{i j}^{1}$ : sum to 0 when $x_{i j}$ is fractional, sum to 1 o.w.

$$
\begin{aligned}
& \sum_{j} v_{i j} x_{i j} \geq \sum_{j} v_{i j} x_{k j} \forall i, k \\
& \quad \sum_{i} x_{i j}=1 \quad \forall j \\
& \quad x_{i j} \geq 0 \quad \forall i, j \\
& x_{i j}^{0} \leq x_{i j} \leq 1-x_{i j}^{1} \quad \forall i, j \\
& \sum_{j}\left(x_{i j}^{0}+x_{i j}^{1}\right) \geq m-2 \forall i \\
& x_{i j}^{0}, x_{i j}^{1} \in\{0,1\} \forall i, j
\end{aligned}
$$

## Envy-freeness <br> Envy-freeness

Every item assigned
Non-negativity
Indicator variable constraints
At most 2 fractional items per agent

## Questions

