## Participatory Democracy with Imperfect Information

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#### 2 The Benchmark Model





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#### The Setting

- The Aam Aadmi Party government on Saturday conducted public meetings in six constituencies of Delhi in the second phase of its participatory budgeting exercise. ...
- Going to the public for budget has helped the government understand local needs, which cannot be addressed from the Secretariat... said Dwarka MLA Adarsh Shastri.

The Hindu, NEW DELHI, April 26, 2015

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- Participatory democracy is a process of collective decision making
- Real life experiences: Brazilian municipalities, state level in Rio Grande del Sul (Brazil), in West Bengal and Kerala (India) (For details see Fung and Wright (2001))
- Formal model of participatory democracy inspired by Aragone's and Sanchez-Pages (2008):

citizens decide whether to attend a meeting relevant for the final policy choice

decide whether or not to reelect the incumbent politician

 Issues: uncertainty due to delegated participatory discussions inefficiency (lack of experience) in understanding and / or transmitting the policy choice of the electorate to the government

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- Electorate of unit mass. k local groups mass  $n_i$ , i = 1, 2, ..., k such that  $\sum_{i=1}^{k} n_i = 1$ .
- Ideal policy choice for each group and governing party  $\theta_1, \theta_2, ..., \theta_k$  and  $\theta_0$
- Delegates: Government officials or local level party workers
- Output: Unbiased but noisy signal,  $a_i$ , of  $\theta_i$ .
- Messengers operate independently

 $a_i \sim F(\theta_i, \sigma^2)$ 

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## The Benchmark Model

#### Support function

$$S(\theta_i, \hat{\theta}) = S((\theta_i - \hat{\theta})^2) = 1 - s(\theta_i - \hat{\theta})^2$$

#### Government's satisfaction function

$$D( heta_0,\hat{ heta})=D(( heta_0-\hat{ heta})^2)=1-d( heta_0-\hat{ heta})^2$$

#### Government's objective

$$\Pi(S_1, S_2, \dots S_k, D) = \Pi(\sum_{i=1}^k n_i S(\theta_i, \hat{\theta}), D(\theta_0, \hat{\theta}))$$
$$= \alpha \sum^k n_i S(\theta_i, \hat{\theta}) + (1 - \alpha) D(\theta_0, \hat{\theta})$$

i=1

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#### Final choice of policy

$$egin{aligned} \hat{ heta} : & \sum_{i=1}^k w_i a_i + w_0 heta_0, \ \textit{where} \ w_i \geq 0 \ \textit{and} \ \sum_{i=0}^k w_i = 1. \ & ( heta_i - \hat{ heta}) \ \sim \mathcal{F}(( heta_i - \sum w_j heta_j), \sigma^2 \sum w_j^2). \end{aligned}$$

Similarly

$$(\theta_0 - \hat{\theta}) \sim F((\theta_0 - \sum w_j \theta_j), \sigma^2 \sum w_j^2).$$

Denote  $E(\hat{\theta}) = \sum w_j \theta_j$  by  $\theta$ .

$$E(\Pi) = 1 - \left[\alpha s \sum_{i=1}^{k} n_i (\theta_i, \theta)^2 + (1 - \alpha) d(\theta_0, \theta)^2\right] - \left[\alpha s + (1 - \alpha) d\right] \sigma^2 \sum w_j^2$$
(1)

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### Optimisation with no uncertainty: $\sigma^2 = 0$

Maximisation with respect to  $\hat{\theta}$  yields

$$\hat{\theta} = \frac{\alpha s \sum_{i=1}^{k} n_i \theta_i + (1-\alpha) d\theta_0}{\alpha s + (1-\alpha) d} = \theta^{\mathsf{C}}, \text{ say.}$$

#### Optimisation with uncertainty:

$$\hat{\theta} = \frac{\alpha s \sum_{i=1}^{k} n_i \theta_i + (1-\alpha) d\theta_0}{\alpha s + (1-\alpha) d} \times \frac{\sum \theta_j^2}{(\sigma^2 + \sum \theta_j^2)} = \theta^C \frac{\sum \theta_j^2}{(\sigma^2 + \sum \theta_j^2)} = \theta^U, \text{ say.}$$

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#### Certain

$$\Pi^{C} = 1 - \alpha s \left\{ \sum n_{i} \theta_{i}^{2} - 2\theta^{C} \sum n_{i} \theta_{i} + (\theta^{C})^{2} \right\} - (1 - \alpha) d(\theta_{0} - \theta^{C})^{2}$$
(2)

#### Uncertain

$$E(\Pi)^{U} = 1 - \alpha s \left\{ \sum n_{i} \theta_{i}^{2} - 2\theta^{U} \sum n_{i} \theta_{i} + (\theta^{U})^{2} \right\} - (1 - \alpha) d(\theta_{0} - \theta^{U})^{2}$$
$$-(\alpha s + (1 - \alpha) d)\sigma^{2} \sum w_{j}^{2}$$
(3)

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Prefernces on a circle

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Figure: Preference geometry

#### Remark 1:

(a) Payoff is decreasing in volatility.

**(b)** Both  $E(\Pi)^U$  and  $\Pi^C$  are decreasing in  $\sum n_i \theta_i^2$ , inter-group heterogeneity in the electorate.

Both are also decreasing in  $(\theta_0 - \theta^C)^2$ .

(c) Figure 1: The impact of distortion different for linear preference (with explicit presence of extreme groups). It will certainly hurt some groups (and/or the government).

For a circular preference pattern, this may be mitigated automatically.

#### Uncertainty $\Rightarrow$ distortion

So interest of the incumbent: noise is reduced. Better delegation or more interaction.

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## Extensions

#### Meeting cost c > 0 per capita.

Notation  $\theta^{A-i} = \frac{1}{1-n_i} \sum_{j \neq i} n_j \theta_j$ . So,  $(\theta^A - \theta^{A-i}) = n_i(\theta_i - \theta^{A-i}) =$  the aggregate signal from all other groups (except *i*<sup>th</sup>).

$$\theta^{C} = \frac{\alpha sn_{i}(\theta_{i} - \theta^{A-i}) + \alpha s\theta^{A-i} + (1-\alpha)d\theta_{0}}{\alpha s + (1-\alpha)d} = \frac{\alpha sn_{i}(\theta_{i} - \theta^{A-i})}{\alpha s + (1-\alpha)d} + \theta^{C-i}$$

Therefore

$$(\theta^{C} - \theta_{i})^{2} = \left(\frac{\alpha sn_{i}(\theta_{i} - \theta^{A-i})}{\alpha s + (1 - \alpha)d} + (\theta^{C-i} - \theta_{i})\right)^{2}$$

Participation will be optimal if  $|\theta_i - \theta^{A-i}|$  greater than  $\eta_i$ 

#### Remark 2:

- (a) bigger groups participate more.
- (b) extreme groups always participate

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## Extensions

#### Intra-group heterogeneity

Assumption:  $a_i \sim F(\theta_i, \sigma_i^2)$ 

- For moderate groups incentive for participation will be actually smaller.
- Extreme groups will all participate and with more vigour.

#### Strategic behaviour

Electorate indicating their choice as different from the true  $\theta_i$ .

For preferences on a line, reporting will be upward biased for groups with above average  $\theta_i$ . A strategic filtering of the signals needed by the government.

For preferences on a circle,

- If diametrically opposite: truth will become a dominant strategy.
- When closer: incentives for misrepresentation reappears!

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# Thank you

## Comments are welcome

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