

# Good-Citizen Lottery

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## A broader research question

How does random, as opposed to meritocratic, allocation of the awards and penalties via a lottery affect efficiency of resource allocations and social welfare?

## Good-Citizen Lottery: Question

- Public bads incur social costs. We aim to collectively minimize the production of public bads.
- “Evil pays more”: Good citizens are typically unpaid for their good deeds.

**Question:** Would a lottery paying one of the **good citizens** whose prize is funded by bad citizens be effective in reducing social costs?

# Motivation

- Check this [partially fake news](#).



There's a speed camera lottery in Stockholm, Sweden where drivers who drive at or under the speed limit are entered to win money. The prize fund comes from the fines paid by people who were speeding.

Greenhouse gas emissions, littering, illegal parking, and speeding are examples of situations where good citizen behaviors are costlier.

(Example that I want to avoid: Voter turnout lottery funded by penalties collected from no turnouts)

## Previous Studies

It is not rare to use lotteries for nonstandard situations.

- Kim (2021): vaccination lottery
- Kim (2023): penalty lottery
- Gerardi et al. (2016), Duffy and Matros (2014): turnout lottery
- Kearny et al. (2010), Filiz-Ozbay et al. (2015): savings lottery
- Morgan (2000), Morgan and Sefton (2002): lottery to fund public goods
- Björkman Nyqvist et al. (2018): lottery incentivizing safer sexual behavior
- Volpp et al. (2008), Levitt et al. (2016): lottery for habit formation

# Setup

(Can be more general)

- $n$  citizens
- Each player chooses either  $S$  (safely abide by law) or  $V$  (violate it).
- Citizen  $i$  accrues a benefit of acting  $V$ ,  $B_i > 0$ .
- Incomplete monitoring capacity: With probability  $p \in (0, 1)$ , action  $V$  is monitored and fined  $F$ .  $p$  is fixed.
- Assume  $B_i - pF > 0$  for all  $i$ , so bad behavior is beneficial. (Otherwise the question becomes trivial.)
- The payoff of choosing  $S$  is 0 for 'most' cases, but when selected as a winner of the lottery, it is  $kF$ , where  $k$  is the number of players who chose  $V$  and got monitored.

## Illustration

- Consider three homogeneous citizens,  $B_i = B$  for all  $i$ .
- sort of a coordination game: neither (S,S,S) nor (V,V,V) is an equilibrium (under some parametric restrictions).
- Given two others playing S: I get 0 if I play S, and I get  $B - pF > 0$  if I play V, so I should play V.
- Given two others playing V: I get  $2pF$  if I play S, and I get  $B - pF$  if I play V, so as long as  $p > \frac{B}{3F}$ , I should play S.
- A symmetric MSNE: playing S with probability  $\delta$ , where

$$\delta = \frac{3 - \sqrt{4B/pF - 3}}{2}$$

(Note that we assume  $B - pF > 0$  so  $4B/pF > 4$ .)

# Analysis

For  $n + 1$  citizens,

- the expected payoff of playing S:

$$\begin{aligned} & \binom{n}{0} \delta^n (1 - \delta)^0 0 + \binom{n}{1} \delta^{n-1} (1 - \delta) \frac{pF}{n} + \binom{n}{2} \delta^{n-2} (1 - \delta)^2 \frac{pF}{n-1} + \cdots + \binom{n}{n} \delta^0 (1 - \delta)^n pF \\ &= \sum_{i=1}^n \binom{n}{i} \delta^{n-i} (1 - \delta)^i \frac{pF}{n+1-i} \end{aligned}$$

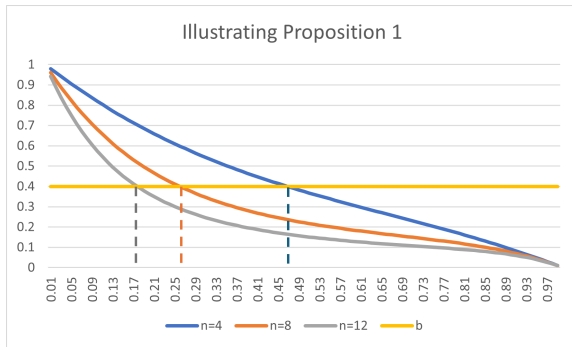
the expected payoff of playing V:  $B - pF$

- $\delta^* \in (0, 1)$  such that  $B - pF = \sum_{i=1}^n \binom{n}{i} \delta^{n-i} (1 - \delta)^i \frac{pF}{n+1-i}$
- Let  $b := \frac{B}{pF} - 1 \in (0, 1)$ , the normalized excess benefit to the expected cost.  $\delta^*$  such that  $b = \sum_{i=1}^n \binom{n}{i} \delta^{n-i} (1 - \delta)^i \frac{1}{n+1-i}$



# Analysis

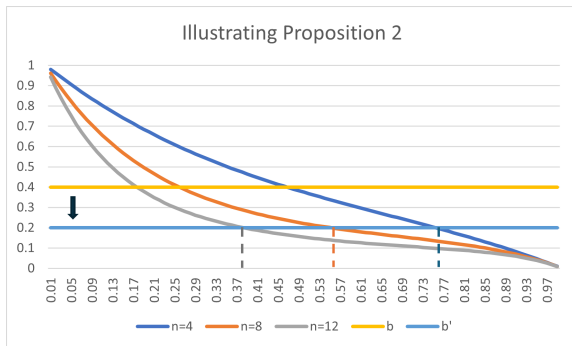
Proposition 1:  $\frac{\Delta \delta^*}{\Delta n} < 0$  if  $b \in (0, 1)$ .



- Intuition: The benefit of  $V$  is fixed. Given the same  $\delta$ , the benefit of  $S$  monotone decreases in  $n$ .
- Voluntary contributions for public good provision tend to decrease in  $n$ . This good-citizen lottery predicts it similarly.
- If  $b > 1$ , then  $\delta^* = 0$  is a corner solution.  $\frac{\Delta \delta^*}{\Delta n} = 0$  if  $b > 1$ .

# Analysis

Proposition 2:  $\frac{d\delta^*}{dp} > 0$  if  $b \in (0, 1)$ .



- Larger  $p \Rightarrow$  more likely to play  $S$ , as long as  $b \in (0, 1)$ .
- Intuition: The excess benefit decreases. Very straightforward.

# Analysis

There are some factors not yet analyzed but worth mentioning.

## Risk preferences

- ⇒ It is not straightforward to tell which action is more “risky”.  
A relatively straightforward claim is, **for sufficiently large  $n$ , risk-averse subjects’  $\delta^*$  is larger**. The winning prob. of the lottery is negligible, while the payoffs of  $V$  are volatile. For small  $n$ ,  $V$  may be more attractive.

## Subjective probability weighting

- ⇒ Some people subjectively overestimate small probabilities.  
Those who **overestimate the small probability of winning the lottery would be more willing to play  $S$** .

# Testable Hypotheses

## Hypothesis

*The overall compliance (the fraction of  $S$  actions) decreases in the population size if  $b \in (0, 1)$ .*

## Hypothesis

*The overall compliance (the fraction of  $S$  actions) increases in monitoring capacity  $p$ .*

## Hypothesis

*For a sufficiently large  $n$ , the more risk averse, the citizens would prefer to choose  $S$  more.*

## Hypothesis

*The more over-weighting the small probabilities, the more likely the citizen chooses  $S$ .*

## Remarks on the null hypotheses

- The null hypotheses are not “normative”: I am not claiming that the findings must be consistent with what theory says.
- Those are benchmarks. We can learn more from what is different from theory, not from what is “as predicted”.

# Experimental Design

Case of 18 subjects,  $p = 0.3$

Varying  $n$  within subjects, varying  $p$  between subjects. In each session of 18 subjects, they play 12 similar games, wherein:

- A subject is randomly assigned to a group whose size is  $n \in \{3, 6, 9, 18\}$ . Their task is to choose one of the two items: a white ball and a box.
- Unwrapping the box, a subject gets a red ball with probability  $p = 0.3$  and a blue ball with probability  $1 - p$ .
- By getting a blue ball, a subject earns a payoff of 240. A red ball is associated with a payoff of 40.
- Choosing the white ball earns 100. Also, one of the group members who chose the white ball is randomly selected to get an additional payoff of  $200k$ , where  $k$  is the number of the members who got the red ball.
- The experiment currency unit is tokens. (1 token = 100 KRW)

## Experimental Design, cont'd

The subjects repeat the game in the mixed order in terms of  $n$ .

| Round | 1         | 2 | 3  | 4 | 5 | 6 | 7  | 8 | 9 | 10 | 11 | 12 |
|-------|-----------|---|----|---|---|---|----|---|---|----|----|----|
| $n$   | 6         | 9 | 18 | 3 | 9 | 6 | 18 | 3 | 9 | 6  | 3  | 18 |
| P03   | $p = 0.3$ |   |    |   |   |   |    |   |   |    |    |    |
| P05   | $p = 0.5$ |   |    |   |   |   |    |   |   |    |    |    |

- To control for the potential order effect, the same mixed order is used for all sessions. (Having randomly mixed order for sufficiently many sessions would be ideal. I struggled for recruiting subjects.)
- Given those parameters,  $b > 1$  when  $p = 0.3$ . Thus, treatment P03 works as a baseline because there must be no treatment effect in  $n$ , that is, no S for all  $n$ . With  $p = 0.5$ , proposition 1 works.

## Experimental Design, cont'd

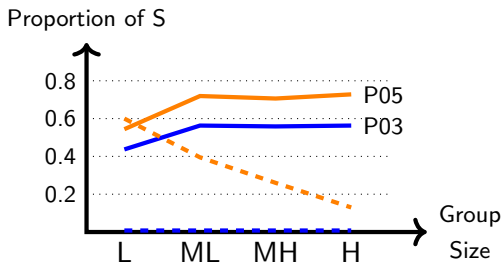
- Then, risk preference and subjective probability weighting are elicited using a simple survey. Post-experimental survey include typical things as well.
- One of the 12 rounds is randomly selected to be paid.



# Experiment Procedure

- Zoom-administered real-time online experiment
- LIONESS (Live Interactive ONline Experimental Server Software)
- at SKKU, in Korean
- 4 sessions each for P03 and P05.  $74 + 76 = 150$  participants
- Random regrouping
- On average 18,800KRW; min 5,000KRW, max 75,000KRW.
- Starbucks e-gift cards corresponding to the cash value

## Result 1



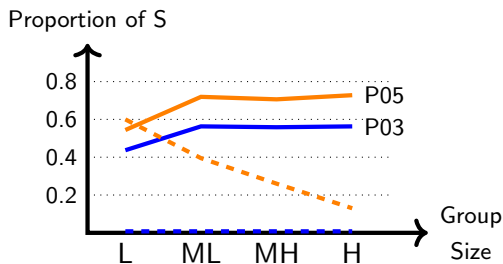
(Dashed lines are theoretical predictions when  $n = 18$ .)

In P03, the proportion of ball choices is larger than 0 ( $p < 0.001$ ).

In P05, the proportion increases in the group size ( $p = 0.031$ ).

This result rejects Hypothesis 1.

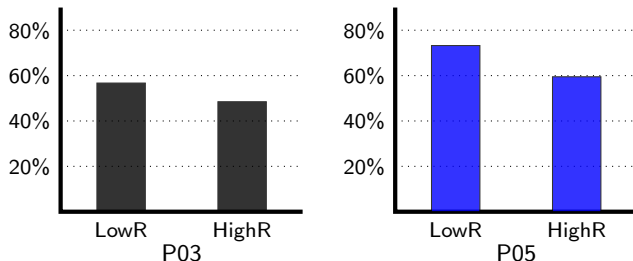
## Result 2



The proportion of ball choices in P05 is significantly larger than that in P03 ( $p < 0.001$ ), supporting Hypothesis 2.

What/who drives S then?

## Result 3



(HighR: risk-taking tendency is high.)

Figure 1: Proportion of Ball Choices By Risk Preference

- More risk averse  $\Rightarrow$  more likely to choose S, supporting Hyp 3.
- Aversion to strategic uncertainty seems to matter less. If it matters, V could have been more frequent for LowR subjects.

## Result 4

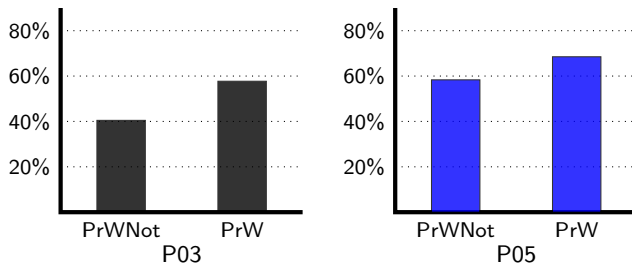


Figure 2: Proportion of Ball Choices By Probability Weighting Tendency

- Those who tend to subjectively weight small probabilities more are more likely to choose S, supporting Hypothesis 4.
- ⇒ Citizen lottery may work with a large population as well.

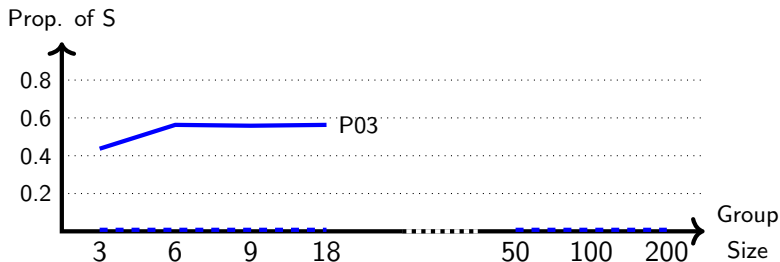
## Take-Away Messages (so far)

- Theory predicts (and experimental findings show) that the proportion of good citizens
  - decreases (increases) in the population size;
  - increases (increases) in monitoring capacity;
  - increases (increases) in risk aversion;
  - increases (increases) in probability-weighting tendency.
- Experimental findings support the last three. The first one turns out to be a complete opposite.
- It is a good sign. It suggest that the citizen lottery can work even for a large population.
- A bit of speculation: if tendencies of overweighting small probs  $\propto$  reckless production of public bads, citizen lottery can be more effective.

## New supplementary evidence

A common criticism: A group of (at most) 20 doesn't seem like a "large" group. Do you see the similar results when a group size becomes really large?

- I conducted an online experiment at Prolific. The data from 400 subjects were collected on Feb, 2025.
- The same setup with P03, but the group sizes are 50, 100, and 200. (Orders were randomized.)



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