Rebate rules in reward-based crowdfunding: Introducing the bid-cap rule

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Introduction	Contribution	Design	Hypotheses	Results
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Definition of crowdfunding

• Raising money from (many) people (via an internet platform)

Types of crowdfunding

- Donation-based
- Equity-based
- Lending-based
- Reward-based
 - Backers get non-monetary reward if their pledge exceeds pre-set entry fee

Introduction	Contribution	Design	Hypotheses	Results	С
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- Reward-based
 - Backers get non-monetary reward if their pledge exceeds pre-set entry fee
 - Allows project creator to contract with future customers **before** investment costs are sunk

Introduction	Contribution	Design	Hypotheses	Results	Conclusion
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All-or-nothing rule superior to *keep-it-all rule* (Coats et al., 2009; Cumming et al., 2020; Strausz, 2017; Wash and Solomon, 2014)

 \rightarrow Still, many projects are unsuccessful ${\scriptstyle \scriptsize \mbox{\tiny Wickstarter}}$

Introduction	Contribution	Design	Hypotheses	Results
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All-or-nothing rule superior to keep-it-all rule (Coats et al., 2009; Cumming et al., 2020; Strausz, 2017; Wash and Solomon, 2014)

 \rightarrow Still, many projects are unsuccessful \blacktriangleright Kickstarter

Solutions to this include:

- Encourage early contributions (Ansink et al., 2017; Solomon et al., 2015)
- Dissemination of positive opinions (Comeig et al., 2020)
- Highlighting of specific projects (Corazzini et al., 2015)
- Timing of promotions (Li and Wang, 2019)

 \rightarrow All these options aimed at increasing the backer base and helping backers to coordinate

Introduction	Contribution	Design	Hypotheses	Results	Conclusion
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What to do if the number of backers is exhausted, but pledges are short of the provision point? \rightarrow A residual public good game arises

A recently proposed solution to this: refund bonuses (Zubrickas, 2014; Cason and Zubrickas, 2017, 2019; Cason et al., 2021)

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 \rightarrow How practical is this really?

Introduction	Contribution	Design	Hypotheses	Results	Conclusion
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 \rightarrow How practical is this really?

Different solution: rebates of excess pledges

 \rightarrow all excess pledges above funding goal are returned to backers according to some rule

 \rightarrow works for threshold public goods (see: Marks and Croson, 1998; Rondeau et al., 1999; Spencer et al., 2009; Donazzan et al., 2016)

Introduction	Contribution	Design	Hypotheses	Results	Conclusion
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In this paper:

- Introduce rebate rules to the reward-based crowdfunding setting
- Establish the novel **bid-cap** rule
- Adaption of proportional rebate rule to reward-based crowdfunding
- Experimentally test these rebate rules

Introduction	Contribution	Design	Hypotheses	Results	Conclusion
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- Experimentally test these rebate rules

Preview of findings:

- Under both rebate rules increased bids and project successes compared to the all-or-nothing rule
- Under the **bid-cap** rule there is <u>less</u> variance in payments, <u>less</u> overbidding and <u>less</u> free riding compared to the proportional rebate rule

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Model

- *N* active individuals $i \in \{1, ..., N\}$ with endowment E_i
- One **passive** individual ("project creator")
- Active individuals decide on pledge $b_i \in [0, E_i]$ towards project
- If $\sum b_i \ge PP$ (exogenous Provision Point) \Rightarrow project realized
- Upon project realization **active** individuals are considered **investors** and receive valuation v_i^a iff $b_i \ge r$ (reservation price)Not Found
- **Passive** individual receives v^P iff $\sum b_i \ge PP$ else a default d

 $\implies \sum v_i \ge PP > N \cdot r$ (socially desirable, residual public good game)

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Introduction	Contribution	Design	Hypotheses	Results	Conclusion
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Experiment

3 Treatments

All-or-nothing: Excess pledges are payed to the project creator

Proportional rebate: Excess pledges are payed back to investors proportional to their excess pledge

Bid-Cap: Algorithm determines the smallest maximal pledge (cap) which high bidders have to pay

- 40 active people per treatment (44 total)
- One shot game followed by a surprise 10 time repetition with random individual valuation $v_i \sim \text{unif}\{30, 60\}$

Introduction	Contribution	Design	Hypotheses	Results	Conclusion
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All-or-nothing

Active's payoff π_i^a

$$\pi_i^a = \begin{cases} E_i - b_i + v_i & \text{if } \sum b_i \ge PP \text{ and } b_i \ge r\\ E_i - b_i & \text{if } \sum b_i \ge PP \text{ and } b_i < r\\ E_i & \text{if } \sum b_i < PP \end{cases}$$

Passive's payoff

$$\pi^{p} = \begin{cases} v^{p} + (\sum b_{i} - PP) & \text{if} \quad \sum b_{i} \ge PP \\ d & \text{if} \quad \sum b_{i} \ge PP \end{cases}$$

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Introduction	Contribution	Design	Hypotheses	Results	Conclusion
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Proportional rebate

Rebate proportional to excess pledge e_i := max{0, b_i - r}
Active's payoff π_i

$$\pi_i^a = \begin{cases} E_i - b_i + v_i + \frac{e_i}{\sum e_i} \left(\sum b_i - PP \right) & \text{if} \quad \sum b_i \ge PP \text{ and } b_i \ge r\\ E_i - b_i & \text{if} \quad \sum b_i \ge PP \text{ and } b_i < r\\ E_i & \text{if} \quad \sum b_i < PP \end{cases}$$

Introduction	Contribution	Design	Hypotheses	Results	Conclusion
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Passive's payoff

$$\pi^{p} = \begin{cases} v^{p} & \text{if } \sum b_{i} \ge PP \\ d & \text{if } \sum b_{i} \ge PP \end{cases}$$

Introduction 000	Contribution	Design 0000●0	Hypotheses	Results 0000000	Conclusion
Bid-cap	0		0	000000	00

• Cut-off pledge $\bar{b} > r$ such that *PP* is exactly met determined by recursive algorithm: • Explanation

Introduction	Contribution	Design	Hypotheses	Results	Conclusion
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Bid-cap					

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Active's payoff π_i

$$\pi_i^a = \begin{cases} E_i - b_i + v_i + (b_i - \overline{b}) & \text{if} \quad \sum b_i \ge PP \text{ and } b_i \ge \overline{b} \\ E_i - b_i + v_i & \text{if} \quad \sum b_i \ge PP \text{ and } b_i \in [r, \overline{b}) \\ E_i - b_i & \text{if} \quad \sum b_i \ge PP \text{ and } b_i < r \\ E_i & \text{if} \quad \sum b_i < PP \end{cases}$$

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Introduction	Contribution	Design	Hypotheses	Results	Conclusion
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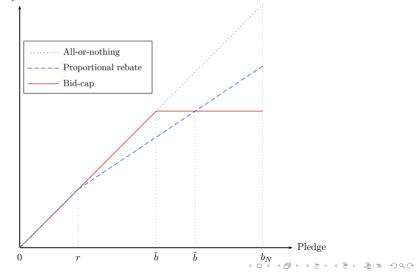
Passive's payoff

$$\pi^{P} = egin{cases} v^{p} & ext{if} & \sum b_{i} \geq PP \ d & ext{if} & \sum b_{i} \geq PP \end{cases}$$



Pledge to paid by treatment

Payment



Introduction	Contribution	Design	Hypotheses	Results	Conclusion
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Hypothesis 1: The pledges will be higher under the rebate rules compared with the all-or-nothing model.

Hypothesis 2: The project realization rates will be higher under the rebate rules compared to the all-or-nothing model.

Hypothesis 3: The variance of payments will be smaller under the bid-cap rule compared to the proportional rebate rule.

All main hypothesis preregistered under: https://aspredicted.org/blind.php?x=X97_FHC



Introduction 000	Contribution O	Design 000000	Hypotheses O	Results ●000000	Conclusior 00
_		All-or-nothing	Proportional	Bid-cap	_
_	Part 1:				_
	$\overline{\text{Mean Pledges } b_i}$	28	33.75 ^a	33.08	
	-	(13.91)	(14.00)	(13.48)	
	Demand revelation b_i/v_i	0.62 ^b	0.75 ⁶	0.74 ^{<i>b</i>}	
		(0.27)	(0.31)	(0.30)	
	Proportion of projects funded	0.25	0.75	0.75	
	Payment when project funded	31.2	30	30	
		(10.69)	(10.07)	(6.54)	
	<u>Part 2:</u>				
, i	Mean pledges b_i	27.84	35.77 ^a	33.63 ^a)	
		<u>(</u> 14.07)	(17.12)	(14.41)	
	Demand revelation b_i/v_i	0.63^{b}	0.82^{b}	0.77 ^b	
		(0.30)	(0.40)	(0.35)	
	Proportion of projects funded	0.35	0.88	0.85	
	Payment when project funded	32.56	30	30	
		(13.28)	(12.23)	(9.28)	
_	^a Significantly different from sy	mmetric equilibri	um prediction of 30		_

 a Significantly different from symmetric equilibrium prediction of 30. b Bids are significantly different from valuation

Table: Descriptive statistics of Part 1 and Part 2 by experimental condition with standard deviations in brackets.

Introduction 000	n Contribution O	Design 000000	Hypotheses O	Results o●ooooo	Conclusior 00
-		All-or-nothin	g Proportional	Bid-cap	_
-	Part 1:				
	$\overline{\text{Mean pledge } b_i}$	28	33.75 ^a	33.08	
		(13.91)	(14.00)	(13.48)	
	Demand revelation b_i/v_i	0.62 ^b	0.75 ⁶	0.74 ^b	
		(0.27)	(0.31)	(0.30)	
	Proportion of projects funded	0.25	0.75	0.75	
	Payment when project funded	31.2	30	30	
		(10.69)	(10.07)	(6.54)	
	Part 2:	. ,	. ,	. ,	
	Mean pledge <i>b_i</i>	27.84	35.77 ^a	33.63 ^a	
		(14.07)	(17.12)	(14.41)	
	Demand revelation b_i/v_i	0.63^{b}	0.82^{b}	0.77^{b}	
		<u>(0.30)</u>	(0.40)	<u>(0.35)</u>	
	Proportion of projects funded	0.35	0.88	0.85	
	Payment when project funded	32.56	30	30	
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^aSignificantly different from symmetric equilibrium prediction of 30.

^b Pledges are significantly different from valuation

Table: Descriptive statistics of Part 1 and Part 2 by experimental condition with standard deviations in brackets.

Introduction	Contribution	Design	Hypotheses	Results	Conclusion
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Regression results

	Part 1 (One round)	Part 2 ((Ten rounds)
	$b_i \in [0, 65]$	$b_i \in [0, 65]$	$Funded \in \{0,1\}$
Proportional	5.75*	× 7.93***`\	<pre>/ 0.53**``,</pre>
	(2.950)	(2.690)	(0.207)
Bid-cap	5.07*	`、5.79**、	`、_0.50** _/
	(2.888)	(2.507)	(0.217)
Constant	28.00***	27.84***	0.35*
	(1.950)	(1.693)	(0.188)
Level of observations	Subject	Subject	Group
Number of observations	120	1200	120

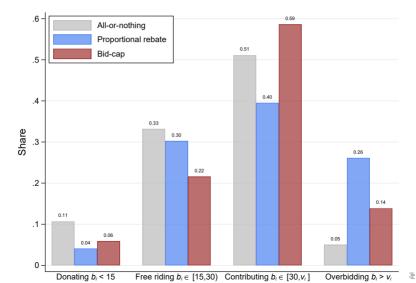
Postestimation Wald tests to compare proportional rebate and bid-cap treatments:

 H_0 : Proportional rebate = bid-cap p = 0.83 p = 0.44 p = 0.86

Standard errors in parentheses. Estimation by OLS regression with robust standard errors for Part 1 and estimation by random-effects regression with clustering on level of observations for Part 2. The baseline category is All-or-nothing in all specifications. *, ** and **** denote significance at the 10%, 5% and 1% level, respectively.

Introduction	Contribution	Design	Hypotheses	Results	Conclusion
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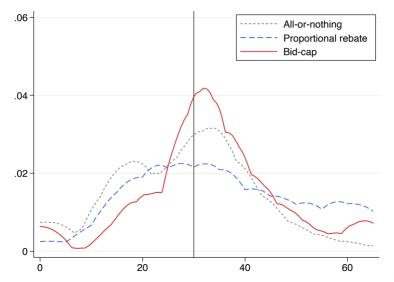
Biding behavior pooled



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Introduction	Contribution	Design	Hypotheses	Results	Conclusion
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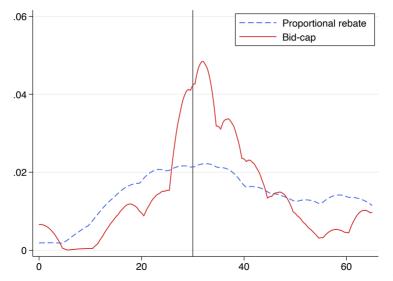
Kernel density of pledges



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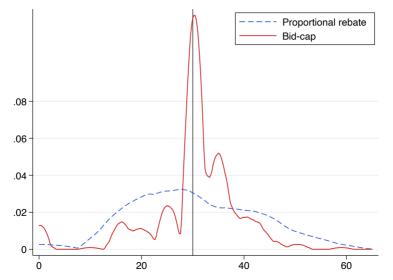
Kernel density of pledges (only funded projects)



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Kernel density of payments (only funded projects)



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Introduction	Contribution	Design	Hypotheses	Results	Conclusion
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Summary of findings:

- Rebate rules improve project success rates in reward-based crowdfunding by enticing backers to place higher pledges.
- Pledging is similar between proportional rebate and bid-cap rules, Although bid-cap induces less variance in payments, less free riding and less overbiding

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Thank you for your attention!

Any Questions?

Link to the working paper:



Explanatio 00 Kickstarter Data

References

Marginal penalty of over-pledging All-or-nothing

$$rac{\partial \pi^a_i}{\partial b_i} = -1$$

Marginal penalty of over-pledging Proportional rebate

$$\frac{\partial \pi_i^a}{\partial b_i} = \begin{cases} -1 + \frac{\left(\sum b_i - PP\right)\left(\sum e_i - e_i\right) + \left(e_i \sum e_i\right)}{\left(\sum e_i\right)^2} & \text{if } b_i \ge r\\ -1 & \text{if } b_i < r \end{cases}$$

Marginal penalty of over-pledging Bid-cap

$$\frac{\partial \pi_i^a}{\partial b_i} = \begin{cases} 0 & \text{if } b_i \ge \bar{b} \\ -1 & \text{if } b_i < \bar{b} \end{cases}$$

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Kickstarter Data 000 References

Regression on bids - equilibrium prediction

	Pa	art 1 (One roun	d)	Part 2 (Ten rounds)			
	$b_{i} - 30$	$b_{i} - 30$	$b_i - 30$	$b_i - 30$	$b_{i} - 30$	$b_i - 30$	
Constant	-2.00 (1.950)	3.75* (2.213)	3.08 (2.131)	-2.16 (1.706)	5.77*** (2.106)	3.63* (1.862)	
Treatment	All-or- nothing	Proportional rebate	Bid-cap	All-or- nothing	Proportional rebate	Bid-cap	
Observations	40	40	40	400 400		400	

Standard errors in parentheses. Estimation by OLS regression with robust standard errors for Part 1 and estimation by random-effects regression with clustering on subject level for Part 2. *, ** and *** denote significance at the 10%, 5% and 1% level, respectively.

Appendix 00•

Explanati

Kickstarter Data

References

Regression on *bids* – *valuation*

	Pa	art 1 (One roun	d)	Pa	Part 2 (Ten rounds)			
	$b_i - v_i$	$b_i - v_i$	$b_i - v_i$	$b_i - v_i$	$b_i - v_i$	$b_i - v_i$		
Constant	-17.00*** (1.950)	-11.25*** (2.213)	-11.93*** (2.131)	-16.18*** (1.644)	-8.25*** (2.134)	-10.83*** (1.962)		
Treatment Observations	All-or- nothing 40	Proportional rebate 40	Bid-cap 40	All-or- nothing 400	Proportional rebate 400	Bid-cap 400		

Standard errors in parentheses. Estimation by OLS regression with robust standard errors for Part 1 and estimation by random-effects regression with clustering on subject level for Part 2. *, ** and *** denote significance at the 10%, 5% and 1% level, respectively.

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Explanatory example:

- Check if *PP donations* is reached if all that want the good pay lowest pledge
- If **yes** all pay lowest pledge and excess contributions equally rebated among investores
- If not lowest bidder(s) pay the lowest pledge and it is checked if all others pay second highest pledge the PP reached
- If **yes** lowest bidders pay lowest all others pay second lowest pledge and potential excess is equally split among the people paying the most
- If **not** continue process

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Explanation

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- Consider an ordered sequence of unique pledges $(b_1, b_2, ..., b_N)$
- Suppose that $\sum b_i > PP$ and each of the first j bids is smaller than r
- Algorithm checks if $(N j) \cdot b_{j+1} \ge PP \sum_{i=1}^{j} b_i$
 - If yes all N-j Individuals pay $b_{j+1} \frac{1}{N-j} \cdot excess$
 - If no check if $(N-j-1) \cdot b_{j+2} > PP \sum_{1}^{j} b_i b_{j+1}$
 - If yes Indiv. (j + 1) pays b_{j+1} all other N j 1 pay $b_{j+2} - \frac{1}{N-j-1} \cdot excess$
 - If no check if $(N j 2) \cdot b_{j+3} > PP \sum_{1}^{j} -b_{j+1} b_{j+2}$ • ...

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Kickstarter

All projects on kickstarter.com

Projects and Dollars

Category	Launched	Total	Successful	Unsuccessful	Live	Live	Success
	Projects	Dollars	Dollars	Dollars	Dollars	Projects	Rate
All	601,692	\$7.51 B	^{\$} 6.88 B	^{\$} 587 M	\$41 M	2,914	40.84%

Successful projects on kickstarter.com

Successfully Funded Projects

Most successfully funded projects raise less than \$10,000, but a growing number have reached six, seven, and even eight figures. Currently funding projects that have reached their goals are not included in this chart — only projects whose funding is complete.

Category	Successfully Funded Projects	Less than \$1,000 Raised	\$1,000 to \$9,999 Raised	\$10,000 to \$19,999 Raised	\$20,000 to \$99,999 Raised	\$100 K to \$999,999 Raised	\$1 M Raised
All	244,514	31,415	129,534	35,379	37,210	10,218	758

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Kickstarter

Unsuccessful projects on kickstarter.com

Unsuccessfully Funded Projects

Funding on Kickstarter is all-or-nothing in more ways than one. While 9% of projects finished having never received a single pledge 79% of projects that raised more than 20% of their goal were successfully funded.

Category	Unsuccessfully Funded Projects	0% Funded	1% to 20% Funded	21% to 40% Funded	41% to 60% Funded	61% to 80% Funded	81% to 99% Funded
All	354,264	56,271	231,971	36,820	16,200	7,036	5,964

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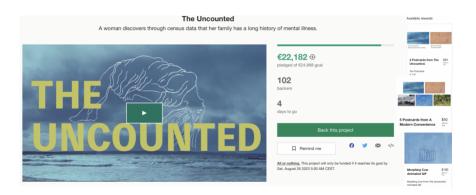
https://www.kickstarter.com/help/stats?ref=global-footer

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Explanation 00 Kickstarter Data 00●

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Example project



/web/20230823090623/https://www.kickstarter.com/projects/thecounted/the-uncounted?ref=section-film-projectcollection-6-staff-picks-category-ending-soon

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Explanatio 00 Kickstarter Data 000 References

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