### A Theory of Crowdfunding Dynamics

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#### Crowdfunding:

Open call to many small contributors (usually online).

#### Reward-based crowdfunding:

- Funders buy a product that the entrepreneur will produce *after* the campaign.
- Crowdfunding is a form of pre-selling to cover production costs.
- Funders bid the price of the crowdfunding product. (we call them bidders)

### Canonical format

- Aggregate Fund Threshold: funding goal
- AoN: succeeds if funds reach goal by deadline, else fails

S (success)  $\rightarrow$  Bids paid, production occurs, rewards delivered;  $\mathcal{F}$  (failure)  $\rightarrow$  Bids reimbursed, no production.

#### • Reward-based AoN crowdfunding is a popular choice because:

It is easy to initiate in practice;

Gives to entrepreneurs a market-test that protects from demand uncertainty.

# Popular reward-based crowdfunding platforms

#### • Kickstarter (US)

235 ML financed projects, 7 BN USD raised from 2009 as of 2023.

• Indiegogo (US)

800 K financed projects, 2 BN USD raised since 2009 as of 2023.

#### • Kisskissbankbank (FR)

33 K financed project, 150 ML EUR raised from 2009 as of 2023.

• Verkami (SP) 12 K financed projects, 50 ML EUR raised from 2010 as of 2023.

# Produzioni Dal Basso (IT) 8 K financed projects, 23 ML EUR raised from 2005 as of 2023.

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There is also crowdfunding taking place outside big platforms. (Radiohead's album "In Rainbows" was financed with crowdfunding in 2007 directly on the band's website)

# Anatomy of a crowdfunding campaign

#### RICKSTARTER



Objective investigation and the two range of the same sequences and an experimental discourse and provide Oblective and the same discourse strong sectors and same and the same provide and the same sequence.

# Statistics from Kickstarter

• Bids:

Heavily right-skewed funding distribution: 60 bidders in median, most bidding once per campaign.



Bidders per campaign

- Duration: often 30 or 60 days.
- Goal: 5 K USD in median (median pledge: 25 USD), but heterogeneous across categories: (bigger for technology, games, design; smaller for music, food, crafts).
- Success rate: 30-40%.

# Empirical puzzle

• Average time profile of bids is a "U-shape" for successful campaigns and an "L-shape" for failed campaigns.



Bidding (as threshold percentage) over time. Bids averaged over all 03/01/2014-02/05/2014, Kickstarter campaigns, conditioning on success (in green), failure (in red).

- What incentives drive these patterns?
- Can we use them to assess the welfare implications of current crowdfunding practices?
- Can we use them to optimize platform and campaign design?

# Determinants of bidding dynamics

- Bid dynamics originate when production uncertainty faced by bidders combines with sunk bidding costs:
- Aggregate Fund Threshold: funding goal  $\implies$  strategic complementarity
- AON: succeeds if funds reach goal by deadline, else fails

S (success)  $\rightarrow$  Bids paid, production occurs, rewards delivered;  $\mathcal{F}$  (failure) $\rightarrow$  Bids reimbursed, no production

 $\implies$  no sunk costs in prior works

- But inspection involves sunk costs.
- So real-time disclosure of bidding aggregates creates dynamic interactions:
- Bidders sunk the bidding cost only if their estimate of the campaign's success probability is sufficiently high.

- Characterize co-evolution of bidding and project success rates.
  - Pivotality often substantial given sequential moves (30% of campaigns attract < 20 bids)</li>
- Welfare implications of funding disclosure (transparency).
- Optimal price and threshold.
- Price dynamics via price menus with limited quantities.

### **Closest Papers**

- Alaei et al (2019)
- Deb et al (2021)
- Ellman & Hurkens (2019)

### Campaign

- $\circ \ {\rm Price} \ p$
- Bidder gap  $g_0 \in \mathbb{N}_+$ .

(everyone pays the same price: funding gap  $\rightarrow$  bidder gap)

- deadline  $\tau \in \mathbb{R}_+$  (outcome determined at  $\tau_+$ ).
- Time  $t \to \text{time left } \tau t$

### Bidders

• Poisson arrival rate  $\lambda$  of bidders over  $[0, \tau]$ 

• Valuations 
$$v_t = \begin{cases} v & q \\ 0 & 1-q \end{cases}$$

• Inspection costs  $c_t$  drawn from CDF  $F(\cdot)$  on [0,q]

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- Bidder t observes the funding state (t, g) on arrival
  - He also knows  $(\mathbf{c_t}, g_0, \tau, p, \lambda, q)$
- Success probability (funding) state (t, g):

$$S_{(t,g)} \triangleq \mathbb{P}(\mathcal{S}|(t,g)) \equiv \mathbb{E}_{(t,g)} \left( \mathbbm{1}_{g_{\tau_+} \leq 0} \right)$$

- Strategies  $a : (t, g, c) \rightarrow \{A, B, C\}$  and their expected payoffs
  - A no inspection, no bid  $\rightarrow U^A_{(t,q)} = 0$
  - B no inspection, bid
  - C valuation-contingent bid  $\rightarrow$

$$U_{(t,g)}^{C} = q S_{(t,g)}^{\text{bid}} d - c_{t} \qquad d = v - p \qquad S_{(t,g)}^{\text{bid}} = S_{(t,g-1)}$$

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# Equilibrium

#### Unique PBE; threshold strategies:

$$\begin{split} \mathbf{C} \text{ if } c_t &\leq q S_{(t,g)}^{\text{bid}} \triangleq q S_{(t,g-1)}, \text{ else } \mathbf{A}.\\ \text{Bidding rate: } \beta_{(t,g)} &= \lambda q F\left(q S_{(t,g)}^{\text{bid}}\right). \end{split}$$

#### Success rate

Using terminal conditions,  $S_{(\tau,g)} = 0$  for g > 0,  $S_{(\tau,g)} = 1$  for  $g \le 0$ ,

$$S_{(t,g)} = \int_t^\tau \left[ \exp\left(-\int_t^x \beta_{(u,g)} \,\mathrm{d}u\right) \beta_{(x,g)} \right] S_{(x,g-1)} \,\mathrm{d}x \tag{REC-}S)$$

• (REC-S) is an opaque expression, so we use stochastic calculus and a martingale insight to disentangle it.

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• (REC-S) is an opaque expression, so we use stochastic calculus and a martingale insight to disentangle it.

• Two effects describe the evolution of bidding rates from an initial date t, at a known gap g, to a future date x with a stochastic future gap  $g_x$ :

$$\mathbb{E}_{(t,g)}\left(\beta_{(x,g_x)}\right) - \beta_{(t,g)} = \lambda q \left[\mathcal{E}_x^{(t,g)} + \mathcal{N}_x^{(t,g)}\right]$$

• By LIE,  $S_{(t,g_t)}$  is a martingale w.r.t. to the natural filtration of  $g_t$ . That is,

$$\mathbb{E}_{(t,g)}\left(S_{(x,g_x)}\right) = S_{(t,g)}$$

• On the other hand, *Pivotality*,

$$\Delta S_{(t,g_t)} \triangleq S_{(t,g_t-1)} - S_{(t,g_t)}$$

is a supermartingale (decreases in expectation).

- Strategic complementarity falls (on average) since a later bidder expects to have fewer bidders arriving after him.
- Combining the two insights, the bid-conditional success rate

$$S_{(t,g_t)}^{\text{bid}} = S_{(t,g_t)} + \Delta S_{(t,g_t)}$$

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- The first is the **pivotality effect** (**PE**):  $\mathcal{E}_x^{(t,g)}$ .
- PE is the effect of the *expected* decreasing pivotality trend.
- It is *always* negative or null (in trivial cases).

$$\mathcal{E}_x^{(t,g)} \triangleq F\left(\mathbb{E}_{(t,g)}\left(qS_{(x,g_x)}^{\mathrm{bid}}\right)\right) - F\left(qS_{(t,g)}^{\mathrm{bid}}\right)$$

- The second effect is the News effect (NE):  $\mathcal{N}_x^{(t,g)}$ .
- NE is the effect of the variance in the funding gap: Bidding is stochastic simply because of preference shocks and random arrivals. Bidders learn good or bad news depending on the gap they observe upon arrival.
- NE is determined by the sensitivity to news of the bidder population:
  - Null NE with linear cumulative inspection cost distribution F
  - Negative NE with a concave  $F (\downarrow \text{ returns to good news})$
  - Positive NE with a convex F ( $\uparrow$  returns to good news)
- Technically, the sign of the NE results from Jensen's inequality:

$$\mathcal{N}_{x}^{(t,g)} \triangleq \mathbb{E}_{(t,g)} \left( F\left(qS_{(x,g_{x})}^{\mathrm{bid}}\right) \right) - F\left(\mathbb{E}_{(t,g)}\left(qS_{(x,g_{x})}^{\mathrm{bid}}\right) \right)$$

- Combining the two effects:
  - $\implies$  Bid rate falls with concave or linear F
  - $\implies$  Bid rate rises with convex *F* if  $|\mathcal{N}| > |\mathcal{E}|$

# Empirically salient funding patterns



Density





- We assume E's utility from success = R & neglect revenues.
- Realized arrival sequence  $\{t_n\}, n = 1, 2, ..., N$ .
- Bidder surplus:

$$V = \mathbb{E}\left(\sum_{n=1}^{N} u^{a(t_n, g_n, c_n)}\right) \underset{\text{Poisson arrivals}}{=} \lambda \tau \mathbb{E}\left(U_{(t, g_t)}^{a(t, g_t, c_t)}\right)$$

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- In the baseline model the gap is disclosed in real time.
- Is transparency good for welfare?
- To answer this question, we compare our baseline (SEQ) to a no-information benchmark (SIM).
- In SIM, bidders arrive sequentially but do not observe  $g_t$ .
- Equilibrium strategies use a constant threshold type  $\hat{c}_{(t,g)} = \hat{c}$

### Welfare comparison of SIM and SEQ

- While in SIM surplus contributions are constant over time, in SEQ they decrease.
- DP and NE manifest themselves also in the welfare analysis:
  - 1 (infra-marginals): Surplus of types  $c_t \leq \hat{c}$  decreases due to DP.
  - 2 (semi-marginals): good-news effects from a bid activates types  $c_t \in (\hat{c}_{(t,g)}, \hat{c}_{(t,g-1)}].$
- SEQ yields higher welfare than SIM for polarized inspection costs.
- SIM yields higher welfare than SEQ when inspection costs are sufficiently concentrated (homogenous case).
- Intermediate information structures can improve on both SIM and SEQ.

### Main takeaways

- Characterization of bidding profiles with 2 new forces: pivotality and news effects
  - These effects combined with heterogeneous costs explain observed bidding profiles.
- Cost diversity and adverse conditions can explain real-time disclosure.
  - Partial disclosure can raise welfare
- Framework reveals price-threshold tradeoff, identifies optimal price dynamics, and allows for study of welfare effect of gap-dependent pricing.

#### Next:

- Endogenous timing via option to delay (E and F 2021).
- Common value.

# Thank you for your attention!

For the paper and the blog post, scan the QR codes below:



Paper



**BSE** Focus