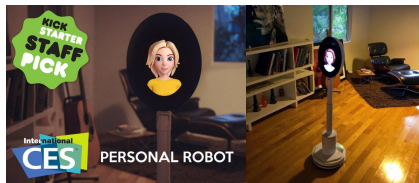


Learning through Crowdfunding

Gilles Chemla and Katrin Tinn

2017

Example



PERSONAL ROBOT

Target: \$50 000

Raised \$161 537 (323%)

274 backers

Rewards:

\$4 – 52 backers, “thank you”

\$19 – 49 backers, “thank you” and T-shirt

\$995 – 86 backers (200 available), Pre-order

\$995 – 23 backers (50 available), Pre-order special version (development kit)

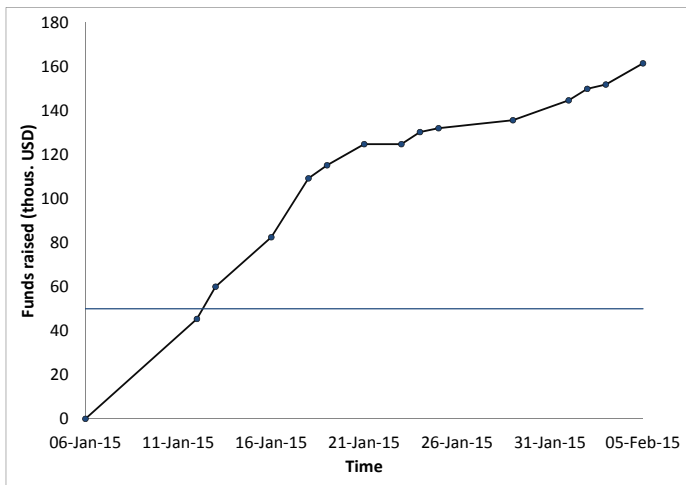
91-99%

\$995 – 4 backers (10 available) Pre-order special version (research kit)

\$1195 – 26 backers (31 available), Pre-order

\$4975 – 1 (10 available), Pre-order 10 with company logo

Timing of backer contributions (Personal Robot)



Questions

- ▶ What is the main source of value creation for firms and backers?
- ▶ Why restrict the pool of backers to future consumers?
- ▶ What are the characteristics of firms that benefit most?
- ▶ How can firms commit to deliver their products to future consumers?
- ▶ Why are third-party platforms needed?

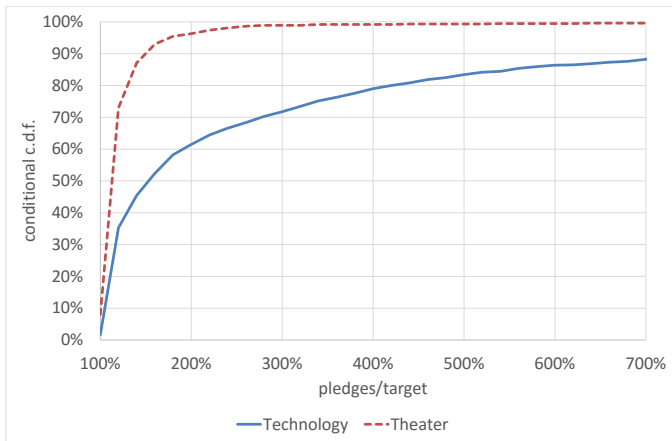
Sector and projects

- ▶ \$5.5bn (2015) up from \$4bn (2014), \$19bn expected in 2021. (Massolution and Statista)
- ▶ Innovative consumer products (Technology, Design and Gaming) raise most funds - 61% of all funds collected; \$55K-\$90K per project compared to \$21K overall average. (Kickstarter)
- ▶ Many projects raise funds comparable to VC/Angel investments - around 4000 projects raised over \$100K; 240 projects raised over \$1M. Pebble Technology \$20.3M (2015), \$10.3M (2012), \$12M (2016). (Kickstarter)
- ▶ Average contribution per backer is noticeable - e.g., \$200K on average of technology project. (Kickstarter)

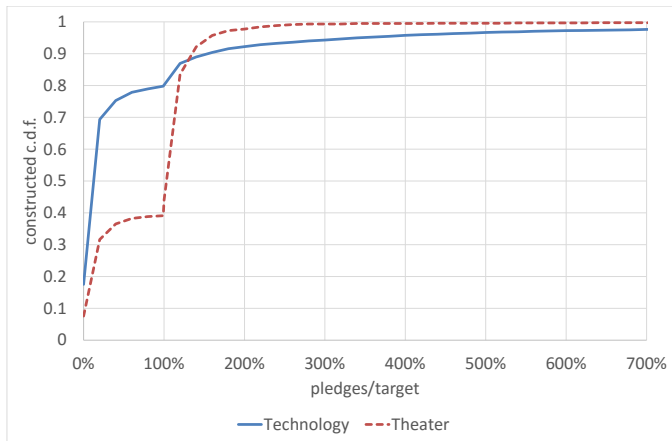
Moral hazard

- ▶ Backers contribute during a fixed length campaign.
- ▶ Funds are passed on before the firm invests and delivers rewards.
- ▶ Nevertheless most projects deliver the rewards (Mollick 2014).

Successful theatre vs. technology projects.



Overall theatre vs. technology projects.



Our paper

- ▶ Reward-based crowdfunding enables firms to credibly **learn about demand**.
- ▶ Real option value of learning: better investment decisions.
- ▶ We derive the optimal scheme, analyze existing schemes.
- ▶ Value of learning mitigates moral hazard.
- ▶ We derive empirical predictions.

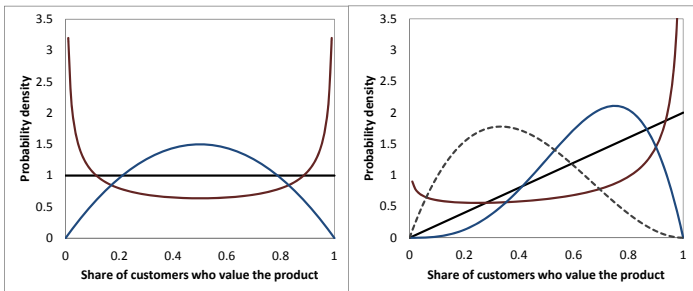
Alternative explanations

- ▶ Belleflamme, Lambert, Schwienbacher 2012 and Varian 2013 focus on backer preferences:
 - ▶ price discrimination - but products are often pre-sold at a discount.
 - ▶ backers are pivotal - but systematic oversubscription.
- ▶ Strausz 2016 and Ellman and Hurkens 2017 consider pre-selling, and contribute to debate about the importance of moral hazard.
 - ▶ preselling without "consumer survey" feature of crowdfunding - but innovative consumer products seem to benefit the most and credit constraints are not the main reason for participation.

Setting

- ▶ The firm has N potential consumers; fraction $\theta \in [0, 1]$ has valuation 1 and $1 - \theta$ has valuation 0.
- ▶ θ is unknown to the firm, prior distribution $\theta \sim Be(\alpha, \beta)$, where $\alpha = \lambda\theta_0$ and $\beta = \lambda(1 - \theta_0)$.
- ▶ Many possible prior beliefs including uniform prior ($Be(1, 1)$)

Examples of possible prior beliefs



Setting

- ▶ All agents are rational and risk neutral, discount factor is $\delta < 1$.
- ▶ Crowdfunding at date 0.
- ▶ The firm decides whether to invest I at date 1.
- ▶ If the firm invests, it produces and sells at date 2.
- ▶ No credit constraints.

Benchmark

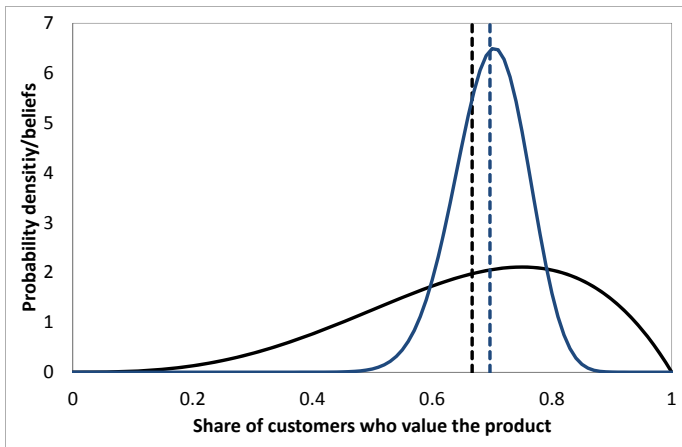
- ▶ $M \leq N$ consumers frictionlessly reveal their valuation at date 0.
- ▶ The firm has incentives to invest iff

$$-I + \delta m + \delta (N - M) \mathbb{E} [\theta|m] \geq 0,$$

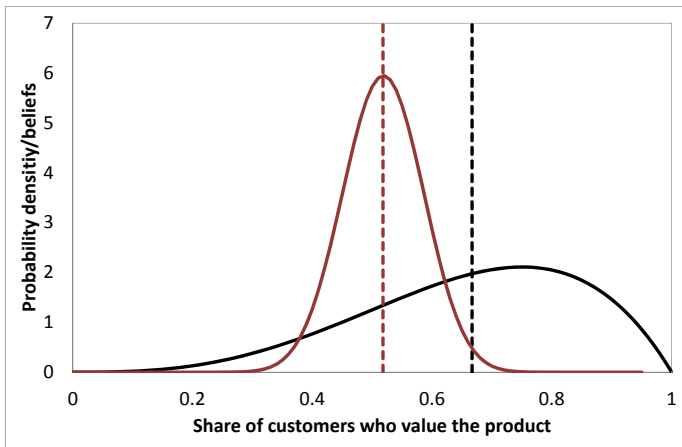
where m is the number of consumers with valuation 1 in sample M .

- ▶ Threshold: the firm invests if $m \geq \bar{m}$

Learning example with $Be(4, 2)$: 35 out of 50 customers pre-order the product



Learning example with $Be(4, 2)$: 25 out of 50 customers pre-order the product



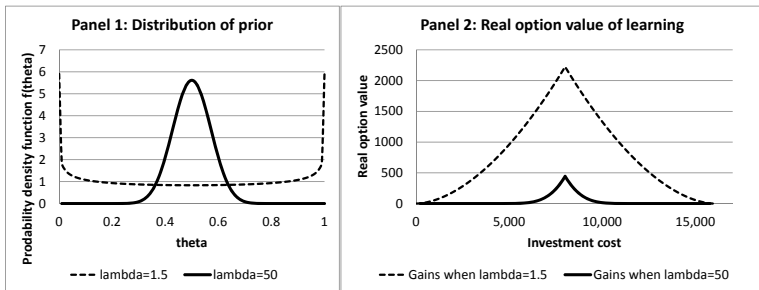
Insights from the benchmark model

Low investment cost I . No gain from pre-selling.

Wide range of intermediate investment costs. Value of learning is positive and maximized at the ex-ante breakeven point.

Higher uncertainty about demand increases the value of learning.

Example on the value of learning



Crowdfunding

- ▶ Pre-selling is an efficient way to learn about preferences.
- ▶ But firms cannot commit to money back guarantees.
- ▶ Further, firms cannot either commit to limited campaign length and transparency.

Crowdfunding

- ▶ Third party platform indirectly mitigate moral hazard:
 - ▶ Transparency during the campaign.
 - ▶ Limited length of campaigns.

All-or-nothing crowdfunding - setting

- ▶ Timing during date 0:
 - ▶ **Morning of date 0:** the firm decides whether to launch a campaign.
 - ▶ **Mid-day of date 0:** the firm sets a target \bar{m}' and pre-ordering price p_0 .
 - ▶ **Afternoon of date 0:** M potential backers observe each other's decisions and decide whether to participate.
 - ▶ **Evening of date 0:** The firm gets $p_0 m$ iff $m \geq \bar{m}'$.
- ▶ We allow for reputation costs χ that may depend on whether the firm meets its target.
- ▶ Platforms are competitive and the intermediation cost is Z .

Decisions of the firm

- ▶ The firm always extracts all the high valuation consumer surplus.
- ▶ The firm invests iff

$$\delta m + \delta (N - M) E [\theta | m] - I \geq \delta m - \chi_Y \text{ if } m \geq \bar{m}'$$

$$\delta m + \delta (N - M) E [\theta | m] - I - \chi_N \geq 0 \text{ if } m < \bar{m}'$$

- ▶ Two thresholds: the lowest target the firm can commit to ($\bar{m}^{*'}),$ and the investment threshold (\bar{m}^*).

All-or-Nothing - results

1. If reputation costs are small, the firm sets a target higher than optimal, and may want to invest after failure. Crowdfunding is possible as long as M is low enough.
2. If the reputation cost of no-delivery is intermediate and the cost of a failed campaign is high, the firm sets target higher than optimal and invests only if it meets the target.
3. If the reputation cost of no-delivery is high, the all-or-nothing scheme achieves the first best
 - ▶ **As long as the reputation cost of failure is small, the firm's expected profit is nearly as high as the first best!**

Keep-it-all - results

- ▶ Both schemes can lead to the same outcome.
- ▶ In general, the firm profit is lower under keep-it-all, and cannot achieve the first best.
- ▶ With high reputation costs there is an additional inefficiency: despite low demand, the firm may have to invest in order to avoid the reputation cost of failure.

Empirical implications

- ▶ Successful projects are oversubscribed (especially when uncertainty is high).
- ▶ If moral hazard is severe, firms must set the target "too high":
 - ▶ Some firms continue after failure.
 - ▶ High target implies that completion ratio= $\text{pledges}/\text{target}$ should be below 1 on average. Cumming et. al. (2015) finds an average completion ratio of $0.403 < 1$ (based on Indiegogo data).

Empirical implications

- ▶ Shorter campaigns are associated with a higher success rate (see e.g., Mollick 2014).
- ▶ Platforms should (and do) take active steps to hide information about failed projects.
- ▶ Pre-orders are sold at par or at a discount.
- ▶ Complementarity with other sources of funding.
- ▶ Statistical structure to assess prior beliefs (e.g., effect of uncertainty).

Conclusion

- ▶ Crowdfunding is beneficial due to learning about demand, even without credit constraints.
- ▶ The value of crowdfunding comes from the option to avoid suboptimal investments. Firms with high uncertainty and intermediate investment costs gain most.
- ▶ Moral hazard is mitigated by third party platforms which can implement transparency, short campaigns (and reputation costs).
- ▶ "All-or-nothing" schemes dominate "keep-it-all" schemes.