#### A Quest for Knowledge

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#### Motivation

In his 1945 letter to Roosevelt—Science, the Endless Frontier—which paved the way for the creation of the NSF, Vannevar Bush emphasizes the importance of research and scientific freedom.

#### But...

- How do researchers act under scientific freedom?
- · What are implications for the evolution of knowledge?
- · How can funding institutions affect the researchers' actions?

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#### Framework

We propose a microfounded model of knowledge and research with three main features:

- 1. Existing knowledge determines benefit and cost of research.
- 2. Successful research improves conjectures about similar questions.
- 3. Researchers are free to choose which questions to study and to what extent.

#### We conceptualize research as

- $\cdot$  the selection of one out of many questions with correlated answers and
- the costly search for its answer building on existing knowledge.

Society values knowledge as it improves conjectures about optimal policies.

#### Contribution

#### Our framework endogenously links

- · the novelty of a research question and
- the probability of discovering its answer.

Expanding the knowledge frontier is more desirable than deepening the existing knowledge only if the area between the frontiers is sufficiently well-understood.

Apply model to classical topics in the economics of science:

- Evolution of knowledge: dynamic externality of knowledge creation.
  - → Short-run suboptimal novelty may improve the evolution of knowledge.
- Science funding: which choices can a budget-constrained funder implement?
  - → Derive implementable set of output and novelty.

#### Literature

#### · Science of Science:

Aghion, Dewatripont and Stein (2008), Bramoullé and Saint-Paul (2010), Foster, Rzhetsky and Evans (2015), Fortunato et al. (2018), Iaria, Schwarz and Waldinger (2018), Liang and Mu (2020), ...

#### · Discovering a Brownian path:

Rob and Jovanovic (1990), Callander (2011), Garfagnini and Strulovici (2016), Callander and Clark (2017), Prendergast (2019), Bardhi (2020), Bardhi and Bobkova (2021), ...

#### Agenda

- 1. Model of Knowledge
- 2. Benefit of Discovery
- 3. Cost of Research
- 4. Researcher's Choices
- 5. Application: Moonshots
- 6. Application: Science Funding

Model of Knowledge

#### Truth, Knowledge and Research Areas

**Questions:** Each  $x \in \mathbb{R}$  is a question.

**Answers:** The answer to x is the realization  $y(x) \in \mathbb{R}$  of a random variable Y(x).

**Truth:** The realization of a standard Brownian path determining all y(x).

Knowledge: Set of known question-answer pairs

$$\mathcal{F}_k = \{ (x_1, y(x_1)), \dots, (x_k, y(x_k)) \}$$
, with  $x_1 < x_2 < \dots < x_k$ .

Knowledge partitions questions into research areas

$$\{\underbrace{(-\infty, x_1)}_{\text{area } 0}, \underbrace{[x_1, x_2)}_{\text{area } 1}, \cdots, \underbrace{[x_{k-1}, x_k)}_{\text{area } k-1}, \underbrace{[x_k, \infty)}_{\text{area } k}\}.$$

Research area *i* has **length**  $X_i := x_{i+1} - x_i$ .

#### Conjectures

A **conjecture** is the distribution of the answer y(x) to a question x:  $G_X(Y|\mathcal{F}_k)$ .

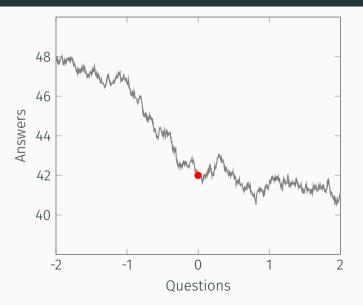
Brownian path determines answers:  $Y(x) \sim \mathcal{N}(\mu_X(Y|\mathcal{F}_k), \sigma_X^2(Y|\mathcal{F}_k))$  with

$$\mu_{X}(Y|\mathcal{F}_{k}) = \begin{cases} y(x_{1}) & \text{if } x < x_{1} \\ y(x_{i}) + (x - x_{i}) \frac{y(x_{i+1}) - y(x_{i})}{X_{i}} & \text{if } x \in [x_{i}, x_{i+1}) \\ y(x_{k}) & \text{if } x \geqslant x_{k} \end{cases}$$

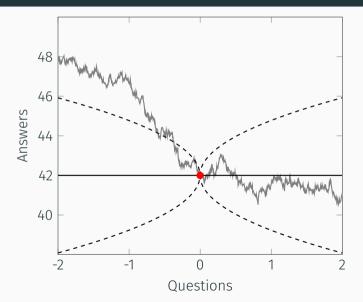
$$\sigma_{x}^{2}(Y|\mathcal{F}_{k}) = \begin{cases} x_{1} - x & \text{if } x < x_{1} \\ \frac{(x_{i+1} - x)(x - x_{i})}{X_{i}} & \text{if } x \in [x_{i}, x_{i+1}) \\ x - x_{k} & \text{if } x \ge x_{k}. \end{cases}$$

Model of Knowledge - Graphically

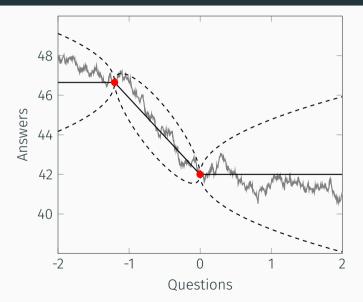
#### Truth and Knowledge



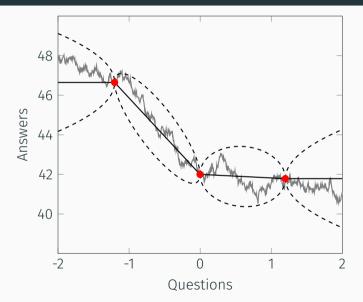
#### Conjectures



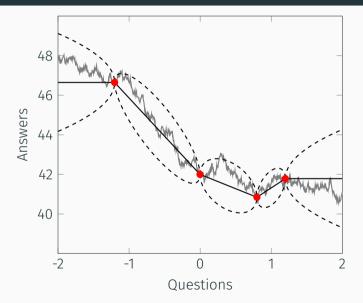
#### Expanding knowledge...



#### ...on both sides



#### Deepening Knowledge



# Society as Decision Maker

#### **Decision Making**

We represent society by a single decision maker that observes knowledge,  $\mathcal{F}_k$ . Society makes decisions on all questions,  $x \in \mathbb{R}$ , and can either

- make a proactive choice:  $a(x) \in \mathbb{R}$  or
- stick with status quo:  $a(x) = \emptyset$

with per-question payoffs

$$u(a(x),x) = \begin{cases} 1 - \frac{(a(x) - y(x))^2}{q} & \text{, if } a(x) \in \mathbb{R} \\ 0 & \text{, if } a(x) = \varnothing. \end{cases}$$

Status quo guarantees a finite payoff but proactive choices can be beneficial if sufficiently good—error tolerance of  $\sqrt{q}$ .

## Benefit of Discovery

#### What is the Value of Knowledge?

#### Jacob Marschak (1974):

Knowledge is useful if it helps to make the best decisions.

Hjort, Moreira, Rao and Santini (2021):

- · science fosters the adoption of effective policies and
- more precise information improves policies further.

#### The Value of Knowledge

Knowledge benefits society via more precise conjectures about optimal policies.

$$a^*(X) = \begin{cases} \mu_X(Y|\mathcal{F}_k) & \text{, if } \sigma_X^2(Y|\mathcal{F}_k) \leq q \\ \varnothing & \text{, if } \sigma_X^2(Y|\mathcal{F}_k) > q \end{cases}$$

Only if society's conjecture about the answer is sufficiently precise, a proactive choice is optimal.

Society's value of knowledge is

$$V(\mathcal{F}_k) := \int_{-\infty}^{\infty} \underbrace{\max \left\{ 1 - \frac{\sigma_X^2(Y|\mathcal{F}_k)}{q}, 0 \right\}}_{=u(a^*(X), X)} dX.$$

#### Benefit of a Discovery

The discovery of an answer y(x) to question x enhances knowledge

$$\mathcal{F}_{k+1} = \mathcal{F}_k \cup (x, y(x)).$$

The benefit of a discovery is the improvement in society's decision making

$$V(x; \mathcal{F}_k) := V(\mathcal{F}_k \cup (x, y(x))) - V(\mathcal{F}_k).$$

 $x_1$  and  $x_k$  are the frontiers of knowledge. A discovery

- expands knowledge if  $x \notin [x_1, x_k]$  and
- deepens knowledge if  $x \in [x_1, x_k]$ .

#### Change of Variables

The problem simplifies by focusing on

- the distance to knowledge,  $d(x; \mathcal{F}_k) := \min_{\xi \in \{x_1, \dots, x_k\}} |x \xi|$
- the length of the research area in which x lies, X.

Applying this rewriting to the variance,

$$\sigma^2(d;X) := \sigma_X^2(Y|\mathcal{F}_k) = \frac{d(X-d)}{X}.$$

Note that for expanding knowledge

$$\sigma^2(d; X = \infty) = d.$$

Benefit of discovery determined by the question's distance to existing knowledge d and the length of the research area X, V(d;X).

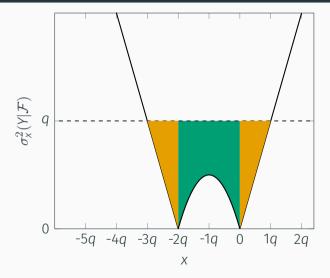
#### Benefit of Discovery - Characterization

#### Proposition

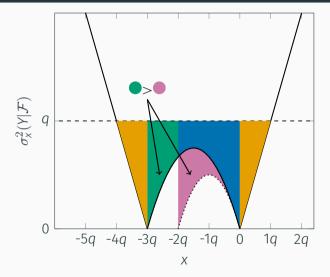
Consider a discovery (x, y(x)) in a research area of length X with distance to existing knowledge d. The benefit of the discovery is

$$V(d;X) = \frac{1}{6q} \left( 2X\sigma^2(d;X) + \mathbf{1}_{d>4q} \sqrt{d}(d-4q)^{3/2} + \mathbf{1}_{X-d>4q} \sqrt{X-d} (X-d-4q)^{3/2} - \mathbf{1}_{X>4q} \sqrt{X}(X-4q)^{3/2} \right).$$

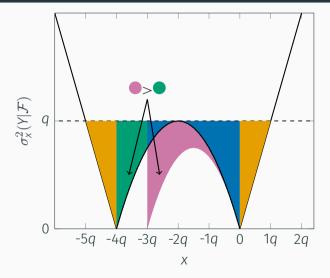
#### Benefit of Expanding Knowledge



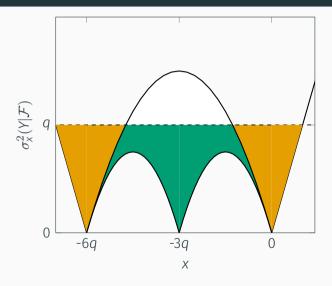
#### Benefit of Expanding Knowledge



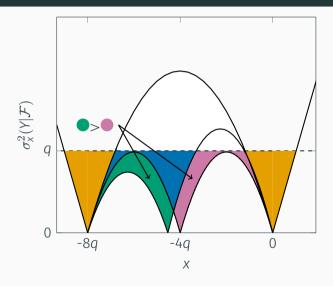
#### Benefit of Expanding Knowledge



#### Benefit of Deepening Knowledge



#### Deepening Knowledge



#### Benefit-Maximizing Distance

#### Corollary

The benefit-maximizing distance  $d^0(X)$  in a research area of length X has the following properties:

- If  $X = \infty$ ,  $d^0(\infty) = 3q$ .
- If  $X \le \widetilde{X}^0 \in (6q, 8q)$ ,  $d^0(X) = X/2$ .
- If  $X \in (\widetilde{X}^0, \infty)$ ,  $d^0(X) \in (3q, X/2)$ .
- $d^0(X)$  is increasing in X for  $X < \widetilde{X}^0$  and decreasing for  $X > \widetilde{X}^0$ .

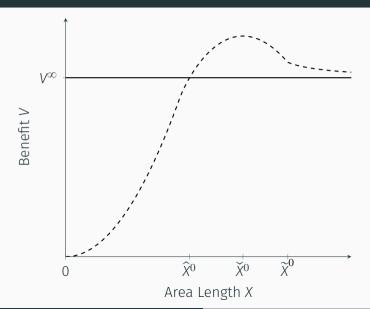
#### Properties of Benefit of Discovery

#### Corollary

There are two cutoff area lengths,  $4q < \hat{\chi}^0 < 6q < \check{\chi}^0 < 8q$ , such that:

- The maximum benefit of deepening knowledge in an area i is increasing in the area length if  $X_i < \check{X}^0$ ; it is decreasing if  $X_i > \check{X}^0$ .
- The benefit of optimally expanding knowledge by 3q dominates the benefit of deepening knowledge in area i if and only if  $X_i < \widehat{X}^0$ .

### Benefit of Discovery by Area Length



# Cost of Research

#### Research as Search for an Answer

The researcher searches for an answer y(x) by sampling an interval  $[a,b] \subseteq \mathbb{R}$ .

The researcher discovers the answer y(x) iff  $y(x) \in [a, b]$ .

Searching for an answer is costly:  $c([a,b]) = \eta(b-a)^2$ .

#### Proposition

Given a question x with distance d in a research area of length X, the lowest-cost search interval such that the answer is contained in the interval with probability  $\rho$  has length

$$2^{3/2}$$
erf $^{-1}(\rho)\sigma(d;X)$ 

and cost

$$c(\rho, d; X) = \eta 8 \left( erf^{-1}(\rho) \right)^2 \sigma^2(d; X).$$

Cost function is separable in probability  $\rho$  and precision of conjectures about y(x).

Researcher's Choice

#### How to Choose Research Questions?

#### Biologist and Nobel laureate Peter Medawar (1976):

Research is surely the art of the soluble. (...) Good scientists study the most important problems they think they can solve.

#### Researcher's Decision Problem

Researcher stands on shoulders of giants and observes  $\mathcal{F}_k$ .

Researcher's payoff consists of the benefit of discovery and the cost of search.

Researcher decides on a research question  $x \in \mathbb{R}$  and a search interval  $[a, b] \subseteq \mathbb{R}$ .

The choice of x and [a, b], can be reduced to a choice of

- · a research area denoted by its length, X,
- · a distance to existing knowledge, d,
- a success probability of search,  $\rho$ .

$$\max_{X \in \{X_0, \dots, X_R\}} \quad \max_{\substack{d \in [0, X/2], \\ \rho \in [0, 1]}} \rho V(d; X) - c(\rho, d; X)$$
 
$$=: U_R(X)$$

## Output & Novelty: Substitutes or Complements?

### Proposition

Suppose  $\eta > 0$ .

- 1. When the researcher expands knowledge, distance, d, and probability of discovery,  $\rho$ , are substitutes.
- 2. When the researcher deepens knowledge, d and  $\rho$  are
  - independent if  $X \leq 4q$ ,
  - complements if  $X \in (4q, \frac{5}{2-\sqrt{3/2}}q)$ ,
  - substitutes for  $d \in (0, \hat{d}(X))$  and complements for  $d \in (\hat{d}(X), \frac{X}{2})$  if  $X \in (\frac{5}{2-\sqrt{3/2}}q, 8q)$ ,
  - substitutes if X > 8q.

# Why Substitutes and Complements?

A ceteris paribus increase in novelty affects both

- the marginal benefit of  $\rho$ , V(d;X), and
- the marginal cost of  $\rho$ ,  $\frac{d}{d\rho} \left( erf^{-1}(\rho) \right)^2 \sigma^2(d;X)$ .

Success probability and novelty are complements if

$$\frac{\mathrm{d}}{\mathrm{d}d}\left(\frac{V(d;X)}{\sigma^2(d;X)}\right) > 0 \Longleftrightarrow \frac{V_d(d;X)}{V(d;X)} > \frac{\sigma_d^2(d;X)}{\sigma^2(d;X)}.$$

 $\frac{\sigma_d^2(d;X)}{\sigma^2(d;X)}$  is increasing and concave in X.

For X < 4q,  $V(d;X) \propto \sigma^2(d;X)$  implying that d and  $\rho$  are independent.

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When X just exceeds 4q, the increase in  $\frac{V_d(d;X)}{V(d;X)}$  accelerates as questions addressed proactively that were not before. d and  $\rho$  are complements. As X increases,  $\frac{\sigma_d^2(d;X)}{\sigma^2(d;X)}$  dominates for small d where  $\frac{\sigma_d^2(d;X)}{\sigma^2(d;X)}$  is highest implying that

d and  $\rho$  are substitutes.

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As  $d \to X/2$ , the marginal cost effect  $\sigma_d^2 \to 0$  implying that if  $V_d(d;x) > 0$  d and  $\rho$  are complements.

Whenever d is such that  $V_d(d;X) < 0$ , d and  $\rho$  are substitutes.

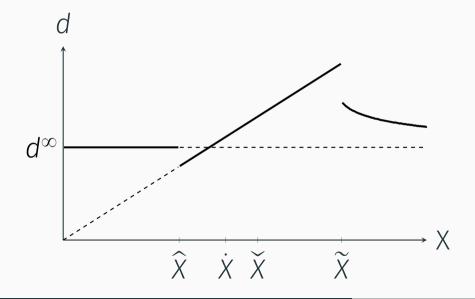
# Optimal Choice: Distance, Novelty and Research Area

## **Proposition**

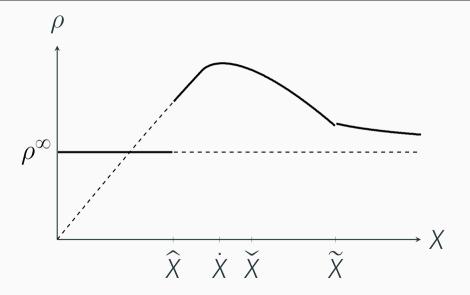
Suppose  $\eta > 0$ . There is a set of cutoff values  $\hat{X} \leq \dot{X} \leq \check{X} \leq X < 8q$  such that the following holds:

- The researcher expands knowledge if and only if all available research areas are shorter than  $\hat{X}$ .
- The researcher's payoffs,  $U_R(X)$  are single peaked with a maximum at  $\check{X}$ .
- The optimal choices of distance, d(X), and probability of discovery,  $\rho(X)$ , are non-monotone in X. The probability  $\rho(X)$  has a maximum at  $\dot{X}$ , the distance d(X) at  $\tilde{X}$ .

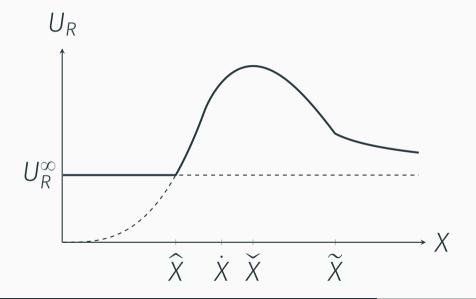
# Novelty by Area Length



# Output by Area Length



# Researcher's Value by Area Length



## Takeaways: Researcher

The microfounded model of research provides insights for classical topics in the economics of science.

- Output and novelty are endogenously linked via the cost of research and existing knowledge. They can be substitutes or complements.
  When expanding knowledge: more novelty → more risk.
- 2. Knowledge determines choice of research area, novelty and output. Research areas of intermediate length have high novelty and output.
- When knowledge is generated by a sequence of short-lived researchers, there is a dynamic externality:
  Research today affects the benefit of discoveries and the success probabilities in the future.

# Application: Moonshots

## **Incentivizing Moonshots**

Is it beneficial to a long-lived society to incentivize very novel research; i.e., more novel than myopically optimal, d > 3q?

Consider the following variant of our model.

- Time is discrete, t = 1, 2, ..., with initial knowledge  $\mathcal{F}_1 = \{0, y(0)\}$ .
- Society discounts time by  $\delta \in [0, 1)$ .
- Sequence of short-lived researchers that observe  $\mathcal{F}_t$  and decide on x and  $\rho$ . Assume symmetric strategies: Same  $\mathcal{F}_t \Rightarrow$  same choice of x and  $\rho$ .
- If research is successful, knowledge updates to  $\mathcal{F}_{t+1} = \mathcal{F}_t \cup (x, y(x))$ . Otherwise  $\mathcal{F}_{t+1} = \mathcal{F}_t$ .

In period 1, society can costlessly pick x and  $\rho$ .

# Optimality of Moonshots

Society maximizes

$$\max_{\mathsf{X},\rho} \ \mathbb{E}\left[\sum_{t=1}^{\infty} \delta^{t-1} \mathsf{V}(\mathcal{F}_{t+1})\right].$$

where choices in periods  $t \ge 2$  are made by researchers individually.

### Proposition

There is a non-empty interval  $(\underline{\eta},\overline{\eta})$  such that the decision maker strictly prefers a moonshot in t=1 for any  $\eta\in(\underline{\eta},\overline{\eta})$  provided  $\delta$  is larger than a critical discount factor  $\underline{\delta}(\eta)<1$ .

Low cost  $\rightarrow$  small distortion  $\rightarrow$  short-run losses dominate.

High cost  $\rightarrow$  small benefit of moonshot  $\rightarrow$  short-run losses dominate.

# Moonshots and Evolution of Knowledge ( $\eta = 1/8$ )



$$t = 1 -$$



$$t=1$$
  $-$ 

$$t=2-$$





$$t=3$$
  $\xrightarrow{\bullet}$   $\xrightarrow{\bullet}$   $\xrightarrow{\bullet}$   $\xrightarrow{\bullet}$   $\xrightarrow{\bullet}$ 



Application: Science Funding

## Science Funding

Under scientific freedom, researchers can freely choose their research questions.

Assume a funder with budget K has two instruments with relative price  $\kappa$ :

- 1. Cost reductions: lowering a researcher's cost by h,  $\eta = \eta_0 h$ .
- 2. Prizes: awarding a prize  $\zeta$  with probability  $\min\{\frac{\sigma^2(d;\mathcal{F}_k)}{s},1\}$  where s>3q.

Which novelty-output combinations can a budget-constrained funder incentivize?

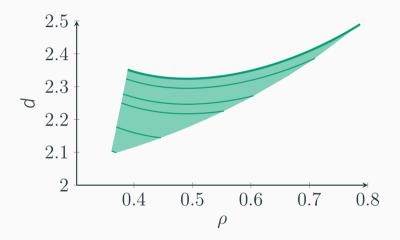
## Proposition

The research-possibility frontier  $d(\rho; \kappa, K)$  defined over  $[\underline{\rho}(\kappa, K), \overline{\rho}(\kappa, K)]$ 

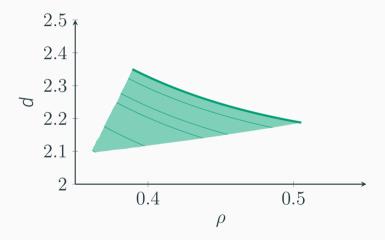
$$d(\rho; \kappa, K) = \min\{6q(K + S - \kappa\eta^{0}) \frac{\rho\tilde{c}_{\rho}(\rho) - \tilde{c}(\rho)}{2S\rho\tilde{c}_{\rho}(\rho) - S\tilde{c}(\rho) - \kappa\rho}, S\}.$$

To incentivize any d > 3q, the funder must award prizes for discoveries.

# Research-Possibility Frontier: Substitutes and Complements



# Research-Possibility Frontier: Substitutes and Complements



## **Maximizing Benefit of Discovery**

Consider a funding institution that maximizes the static benefit of expanding knowledge:

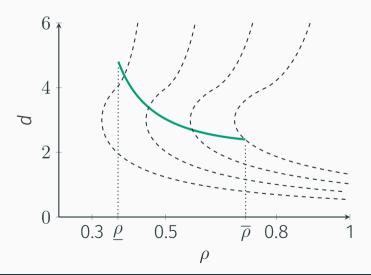
$$\max_{\zeta,\eta} \quad \rho V(d;\infty)$$
 s. t. 
$$\zeta + \kappa(\eta_0 - \eta) = K.$$

#### **Proposition**

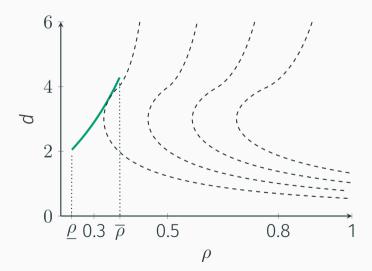
If output and novelty are substitutes from the funder's perspective, the optimal funding mix never incentivizes excessive novelty, d>3q.

If output and novelty are complements from the funder's perspective, the optimal funding mix might incentivize excessive novelty, d > 3q.

# Optimal Funding - Cost Reduction Cheap



# Optimal Funding - Cost Reduction Expensive





Conclusion

#### Conclusion

#### We propose a model of knowledge built on

- 1. a large pool of questions,
- 2. knowledge informing conjectures about related questions,
- 3. society applying knowledge to choose policies.

#### We conceptualize research as the

- 1. free choice of research questions and
- 2. and the costly search for their answers.

#### Our model

- · endogenously links novelty and research output and
- highlights the importance of existing knowledge for research and knowledge accumulation.

## Applications of Model

Model is tractable and widely applicable in the economics of science.

- 1. Evolution of science
  - Dynamic externality → suboptimal knowledge accumulation.
  - · Moonshots can improve evolution of knowledge.
- 2. Optimal research incentives
  - Novelty and output can be complements for funding institutions.
  - · Cost reductions alone cannot incentivize moonshots.
- 3. Consequences of null results (work in progress)
- 4. Innovation, competition, ...