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Recalibrating the Reported Rates of Return to Food and Agricultural R&D

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Introduction

More than half a century has passed since Zvi Griliches published the first formal estimate of the returns to food and agricultural R&D in the Journal of Political Economy. Since then economists have published a large number of similar estimates. Alston et al. (2000) reported 292 studies with 1,886 evaluations of the payoffs to agricultural R&D in terms of the internal rate of return (IRR) or benefit-cost ratio (BCR). The average IRR was 73.1% per year, which is indicative of persistent underinvestment. But rather than ramping up these investments, growth in agricultural R&D spending for each of the past four decades has slowed worldwide and particularly in rich countries. One explanation for this apparent underinvestment is that rate of return estimates have simply been dismissed as unbelievably high, implying economists got it

Alston et al. (2011) argued that economists did get it wrong and have systematically overstated the returns to R&D due to their reliance on the IRR and its assumption that the financing rate of an investment equals the reinvestment rate of return earned by its beneficiaries. To correct this bias, they proposed relaxing this assumption and employing the modified internal rate of return (MIRR) (Hirshleifer 1958). For U.S. Department of Agriculture and state agricultural experiment station R&D investments from 1949 to 2002, they found the IRR (22.7%) was more than twice the MIRR (9.9%).

Objectives

The purpose of this research was to reexamine past estimates of the rates of return to food and agricultural R&D in light of the MIRR to see if a similar pattern emerges across the literature. Specific objectives included:

- Assemble a comprehensive database of estimates of the rate of return to food and agricultural R&D.
- Use the information available from past studies to estimate the MIRR when feasible.
- Compare the IRR and MIRR estimates under alternative assumptions regarding the reinvestment rate.

Data

The database assembled for this project includes 2,186 evaluations of returns to R&D published in 359 studies from 1958 to 2011. Of these, 95 percent reported the IRR, 26 percent reported a BCR, and 21 percent reported both. Most reported the time over which the benefits and costs of investments were evaluated. Most also reported the discount rate for computing BCRs when they were reported. Investments covered in the dataset include those sponsored by governments, non-governmental organizations, and private companies. These investments covered a wide range of commodities from many regions of the world. The sources include studies published in books, journals, and a large amount of grey literature (e.g., evaluation reports and studies published by various international and national agencies).

Calculation of the MIRR requires information on the cost and benefit profiles, which are seldom reported. However, these profiles can be estimated given the time frame and estimates of the IRR and BCR — information that was reported by a quarter of the studies and 302 (13.8%) of the evaluations. Table 1 compares the distributions of the IRR and BCR for this subsample to the overall sample.

Table 1 Internal Rate of Return (IRR, in percentage) versus

Benefit-Cost Ratio (BCR)											
		Obs.	Mean	S.D.	Min.	1st Quantile	Median	3rd Quantile	Max.		
IRR	Overall	2077	74.3	196.3	-47.5	24.0	43.0	74.0	5645.0		
	Subsample	302	51.8	59.8	0.1	18.9	32.9	75.6	677.0		
BCR	Overall	568	23.3	30.7	0.0	3.2	11.0	31.9	199.0		
	Subsample	302	29.0	30.6	0.1	6.9	20.4	40.1	176.0		

Methods

Setting the date of the initial and final R&D investments to 0 and T_{c} , and the date of the initial and final benefits from these investments to T_b and T_b , the MIRR can be defined implicitly by

(1)
$$\frac{B}{C} \sum_{t=T_b}^{T} w_t^b (1+\rho)^{T-t} = \sum_{t=0}^{T_c} w_t^c (1+MIRR)^{T-t}$$

where C and B are the aggregate nominal costs and benefits accruing from the investment; $\mathbf{w^c} = (w_0^c, ..., w_{T_c}^c)$ and $\mathbf{w^b} =$ $(w_{T_b}^b, ..., w_T^b)$ are the distributions of these aggregate costs and benefits over time; and ρ is the reinvestment rate. While T_c , T_b and T are readily available in most studies, and ρ is typically taken as exogenous in MIRR calculations, the specific cost and benefit profiles (i.e., $\mathbf{w}^{c}C$ and $\mathbf{w}^{b}B$) are not typically reported.

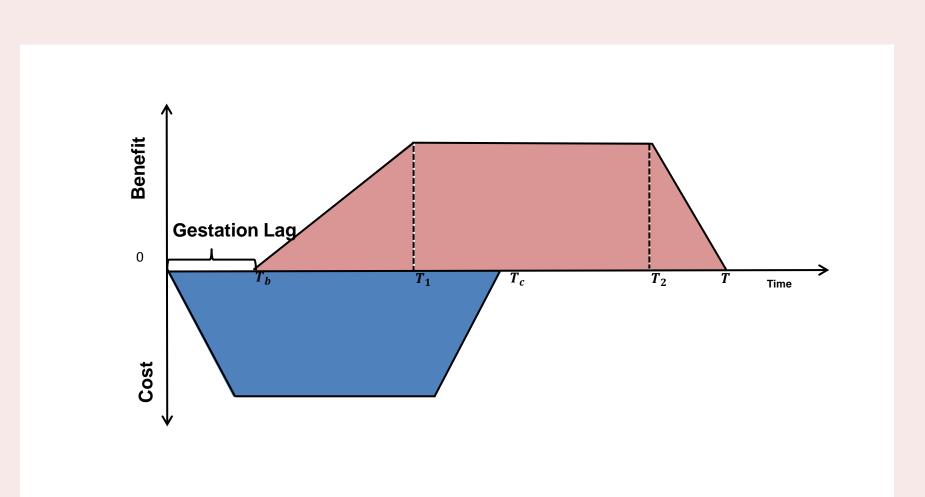
Calculating the MIRR is not feasible if there is no information on the cost and benefit profiles. Fortunately, knowing IRR, BCR, T_c , T_b , T_c , and the discount rate (δ) used to calculate the BCR does provide information that can be used to estimate these profiles. To see how, note that the definitions of the IRR and BCR imply

(2)
$$\frac{B}{C} = \frac{\sum_{t=0}^{T_C} w_t^C (1 + IRR)^{-t}}{\sum_{t=T_b}^{T_c} w_t^b (1 + IRR)^{-t}}$$

(3)
$$BCR - \frac{\sum_{t=0}^{T_C} w_t^C (1 + IRR)^{-t} \sum_{t=T_b}^{T_c} w_t^b (1 + \delta)^{-t}}{\sum_{t=T_b}^{T_c} w_t^b (1 + IRR)^{-t} \sum_{t=0}^{T_C} w_t^C (1 + \delta)^{-t}} = 0.$$

This reduces the problem to finding the distributions for **w**^c and **w**^b that satisfy equation (3). While conceptually straightforward, this problem is computationally impractical. This computational impracticality can be overcome by approximating w^c and w^b with





a flexible family of distributions that can be characterized by a parsimonious parameter space.

Let $\mathbf{w^c}(\alpha^c)$ and $\mathbf{w^b}(\alpha^b)$ be parameterized distributions for the costs and benefits where α^c and α^b are parameter vectors. The problem can then be framed as

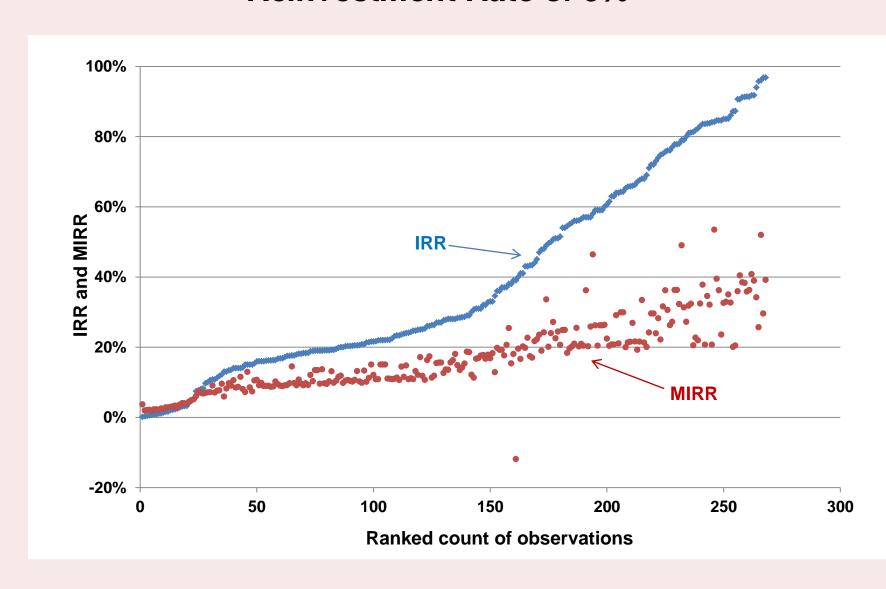
(4)
$$\min_{\boldsymbol{\alpha^{c}}, \boldsymbol{\alpha^{b}}} \left| BCR - \frac{\sum_{t=0}^{T_{c}} w_{t}^{c}(\boldsymbol{\alpha^{c}})(1+IRR)^{-t} \sum_{t=T_{b}}^{T} w_{t}^{b}(\boldsymbol{\alpha^{b}})(1+\delta)^{-t}}{\sum_{t=T_{b}}^{T} w_{t}^{b}(\boldsymbol{\alpha^{b}})(1+IRR)^{-t} \sum_{t=0}^{T_{c}} w_{t}^{c}(\boldsymbol{\alpha^{c}})(1+\delta)^{-t}} \right|.$$

The problem in equation (4) can be solved for evaluations that include both the IRR and BCR by assuming the distributions of costs and benefits are trapezoidal as illustrated in Figure 1. With these trapezoidal distributions, the optimization problem can be solved robustly and efficiently using a grid search. Once estimates for α^c and α^b are obtained, estimates for B/C and MIRR follow immediately from equations (1) and (2).

Results

Figure 2 shows the reported IRR and the estimated MIRR assuming a reinvestment rate of 5%. Table 2 shows the sensitivity of these results for alternative assumptions regarding the is 2.5 times larger; though the IRR is not universally larger than the IRR. For low IRRs (those less than the reinvestment rate), the MIRR tends to be higher. But this is a relatively small percentage (<10% for all three reinvestment rates) of observations. Alternatively, for relatively large IRRs, the MIRR is

Figure 2 Comparison of IRR and MIRR using a Reinvestment Rate of 5%



smaller, with the difference tending to increase with the IRR.

While the MIRRs are generally smaller than the IRRs, they still suggest a competitive return to investments in agricultural R&D. reinvestment rate. Compared to the mean MIRR, the mean IRR With a 5% reinvestment rate, 90% of the MIRRs exceed 7.1%, while half exceed 17.4%. For 3% and 10% reinvestment rates, 90% exceed 6.0% and 10.3% respectively, while half exceed 16.6% and 19.9%. It is also worth noting that for the IRRs, 90% exceed 10.1% and half exceed 33%.

Table 2 Sensitivity Test on Reinvestment Rate (ρ)*

	IRR	M	<i>=)</i>	
	(percentage)	$\rho = 3\%$	ρ = 5%	$\rho = 10\%$
Mean	52	18	20	22
Minimum	0	-11	-12	3
1 st Quantile	19	9	10	13
Median	33	16	17	20
3 rd Quantile	76	25	26	29
Maximum	677	268	269	273
	vith <i>IRR</i> <= <i>MIRR</i> of total sample)	17 (5.7%)	22 (7.3%)	29 (9.7%)
	with <i>IRR</i> > <i>MIRR</i> of total sample)	283 (94.3%)	278 (96.7%)	271 (90.3%)

*Our grid search for the best-fit distributions included 302 observations. Unique distributions were identified for all but two observations, which had BCRs equal to 1 and a discount rate equal to the IRR.

Discussion & Conclusion

Over the past half century economists have published many estimates of the returns to food and agricultural R&D. The result of this effort suggests the investments have paid handsomely. Despite estimates of high returns, agricultural R&D spending growth has slowed or stalled. Alston et al. (2011) challenged previous estimates of the IRR due to the implausible assumption that the cost of financing these investments equals the rate of return to beneficiaries. They further showed how the MIRR, which does not rely on this assumption, yields estimates that are about half the size of the IRR for agricultural R&D in the United States.

This research reexamined estimates of returns to R&D using the MIRR to see if a similar pattern emerges throughout the literature. The results are even more striking, with the IRR 2.5 times larger than the MIRR on average. While reexamining all previous studies was not possible, the results are conservative because the difference in IRR and MIRR tends to increase as the IRR increases, and the studies that were reexamined had lower IRRs than others on average. While the MIRRs are typically lower than the IRRs, they still suggest that the vast majority of investments yielded a competitive rate of

The evident failure of economic evidence to sway R&D decisions has profound consequences. Alston, Babcock and Pardey (2010) concluded that there has been a widespread slowdown in agricultural productivity growth, consistent with a prior and persistent ratcheting down in agricultural R&D spending growth. If R&D-induced shifts in global food supplies fall short of demand for agricultural output, affordable access to food will be further curtailed, with inevitable adverse consequences for the chronically hungry worldwide.

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