

Innovation, Growth and Structural Change in American Agriculture

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ABSTRACT. U.S. agriculture was transformed during the 20th century by waves of innovation with mechanical, biological, chemical, and information technologies. Compared with a few decades ago, today's agriculture is much less labor intensive and farms are much larger and more specialized, supplying a much-evolved market for farm products. Over recent decades, the global landscape for agricultural R&D has shifted away from farms, away from the public sector toward the private sector, and away from the United States toward agriculturally important middle-income countries, especially China, India and Brazil. U.S. investments in agricultural R&D are stalling even though meta-evidence shows that past U.S. investments in agricultural R&D have yielded very favorable returns: median reported benefit cost-ratios in the range of 12:1. Continuing U.S. investment and innovation will be required to preserve past productivity gains in the face of climate change, coevolving pests and diseases, and changing technological regulations—let alone increase productivity. Great potential exists for innovation in crop and livestock genetics and digital farming technologies to generate new products and production processes, but innovators have to overcome increasingly strong headwinds from social and political forces that seek to dictate technology choices.

Keywords: U.S. agriculture, R&D and productivity, biotechnology; digital technology; precision agriculture; technological regulation

1. Introduction

During the 20th century American agriculture was dramatically transformed and its role in the economy changed markedly. The progressive introduction and adoption of a host of technological innovations and other farming improvements enabled much more to be produced with less land and a lot less labor; farms became many fewer, much larger, and more specialized. However, while agriculture continued to grow, it shrank in relative importance. The U.S. farm population peaked at 32.5 million, 31.9% of the total U.S. population in 1916; since then it declined to 4.6 million in 2013, just 1.5% of the total. And, while agriculture's contribution to GDP increased (somewhat erratically) from 11.6% in 1889 to 17% in 1917, it declined steadily thereafter to just 0.97% in 2015. Innovation on and off farms played a central role.

This paper provides detailed documentation, evidence, and analysis of the past and ongoing sources of innovation and structural change in American agriculture. To begin, we describe the profound structural transformation of American agriculture over the course of the past century, and the implications for changing patterns of inputs, outputs, and productivity. Then we turn to a consideration of the sources of change. We pay particular attention to the impact of innovation as a driver of productivity and employment in the agricultural sector; to the role of private entrepreneurship in the process; and to the distinctive features of the agricultural sector—still largely atomistic and heavily dependent on a host of natural, often location-specific, inputs, changes in which can undercut past productivity gains—and its evolution (shrinking as a share of the economy). Digging deeper, we present evidence on inputs to and outputs from innovation, on the resulting gains in productivity, what those gains are worth, prospects for the coming decades, hurdles to be overcome, and roles played by government policies.

2. Special Features of Agriculture¹

Innovation in agriculture has many features in common with innovation more generally, but agriculture differs in terms of its industrial structure and the nature of market failures in innovation; the spatial dimensions of production and the site-specific nature of the technology; the biological nature of the production process; and the nature of food and farming as perceived by the broader public and groups that define technological regulations and requirements. It is helpful to have those differences in mind as we review the past and prospective changes in agriculture and the roles of entrepreneurship and innovation in shaping them.

Like other parts of the economy, agriculture is characterized by market failures associated with incomplete property rights over inventions. The atomistic structure of much of agriculture means that the attenuation of incentives to innovate is more pronounced than in other industries that are more concentrated in their industrial structure. On the other hand, unlike most innovations in manufacturing, food processing, or transportation, agricultural technology has a degree of site specificity because of the biological nature of agricultural production, in which appropriate technologies vary with changes in climate, soil types, topography, latitude, altitude, and distance from markets. The site-specific aspect circumscribes the potential for knowledge spillovers and the associated market failures that are exacerbated by the small-scale, competitive, atomistic industrial structure of much of production agriculture.

Agriculture is further distinguished by the biological and spatial nature of its production technology (Joglekar et al. 2016). Agricultural production takes up a lot of space—indeed, about 40% of the world’s land area is occupied by agriculture, of which 12% is used for crops (and

¹ Parts of this section draw from Pardey et al. (2010).

44% of U.S. land is in agriculture)—and the nature of the space varies in ways that are relevant for the choice of technology and the returns to innovations that are often very site-specific. The biological nature of agricultural production means that production processes take time, during which outcomes are susceptible to the influence of factors such as weather and pests that are difficult or costly to control.

The agricultural production consequences of pests and weather themselves vary in ways that are often uncontrolled and difficult to predict with present knowledge and technology, not only within a season but systematically over time and space. Climate change and the co-evolution and adaptation of pests and diseases means maintenance research is required to prevent yields from declining—the “Red Queen” effect as discussed by Olmstead and Rhode (2002) for instance.² These features of agriculture give rise to a demand for innovations that reduce the susceptibility of production to uncontrolled biotic and abiotic stresses and allow technology to adapt to changes in the farming environment or changes in technological regulations.

Agriculture is also subject to different kinds of public and policy scrutiny because, compared with most other industrial outputs, people care differently about food and the way it is produced, and increasingly so as they become richer. U.S. consumers are increasingly demanding foods that have “credence” attributes associated with the products and the processes used to produce them—such as organic, locally produced, and raised using humane livestock and poultry practices (see, e.g., Rausser et al. 2015, 2019). Alston (2019) discusses this demand and the related demand for technological regulation coming from what he terms the “woke farm and food policy reform movement,” which blames the agricultural and food industry for various

² For example, a 1986 survey of 744 U.S. agricultural scientists suggested that “maintenance research” accounted for around one-third of production-oriented agricultural research at that time (Adusei and Norton 1990).

societal ills (see, e.g., Willet et al. 2019). Some food processors, manufacturers, and retailers are requiring foods to be produced in ways that accommodate these demands (see, e.g., Saitone et al. 2015). All these forces have implications for the types of innovations that will be relevant for American farms in the coming decades—as they or forces like them were in the past.

3. Structure of American Agriculture

Land-saving and especially labor-saving innovations were central to the structural transformation of American agriculture in the 20th Century. The number of U.S. farms grew from 1.4 million in 1850 to a peak of 6.8 million in 1936 before falling to 2.9 million in 1970 and 2.1 million in 2017; land in farms peaked in 1954 (Figure 1; Pardey and Alston 2019). Farms became much larger and more-specialized in terms of their output mix, and the input mix shifted to use much less labor and a little less land, combined with more capital and purchased material inputs.³

[Figure 1: *U.S. Farm Area and Farm Numbers, 1850–2017*]

Farm Labor

These changes in the structure of agriculture entailed changes in the structure of the farm labor force, reflecting both the pull from growth in non-farm demand for labor, driving up the opportunity cost of farmers' time as well as the cost of hired farm labor, and the push from technological changes on farms that permitted more to be produced with much less labor and

³ These aggregate figures encompass highly diverse farm sizes and types. In 1920, the United States had 735 million acres on 6.4 million farms, at an average of 8.7 acres per farm. None of those farms had more than 1,000 acres. By 2012, the total number of farms had fallen by more than two-thirds compared with 1920. Now 9.2% of a total 2.1 million farms had more than 1,000 acres. Notably, more than 10% of today's "farms" have less than 10 acres, but many of today's small farms are "hobby farms," and many are part-time occupations for people for whom living on a farm is a life-style choice more than a way of making a living.

more land per farm. Farmers responded to these incentives and opportunities by consolidating farms and substituting other inputs for labor, in part by developing and adopting labor-saving innovations that favored higher land-labor ratios and larger optimal farm sizes.⁴ Figure 2 captures the main elements of the post-WWII changes in farm labor use. In panel a, total labor use—measured in hours per year, and taking account of the major shift to part-time farming—in agriculture fell by two-thirds, from 1949 to 2007. Within this total, operator labor fell by more than three-quarters and the hired labor share increased from 20% to 32%.⁵

[Figure 2: *Labor Use in U.S. Agriculture, 1949–2007*]

As farmers substituted other inputs for more-expensive labor the cost share of labor fell from more than 42% in 1949 to less than 30% in 2007. Mainly, farmers increased their use of materials inputs purchased off farm (such as seed, fuel, electricity, fertilizer and other agricultural chemicals, and hired machines); the total use of land and capital (measured with appropriate indexes) remained relatively constant—albeit with considerable variation in these details over time and among states.⁶ The mix of agricultural output has also changed over time, reflecting the effects of changes in U.S. and foreign demand for U.S. farm products, as well as

⁴ MacDonald et al. (2018) report that “By 2015, 51 percent of the value of U.S. farm production came from farms with at least \$1 million in sales, compared to 31 percent in 1991 (adjusted for price changes)... [Now] few farms specialize in a single crop, field crop operations increasingly grow just 2 or 3 crops, versus 4-6 crops previously. Livestock production continues to shift toward farms that produce no crops, and instead rely on purchased feed. ... Despite increased consolidation, most production continues to be carried out on family farms, which are owned and operated by people related to one another by blood or marriage. Family farms accounted for 90 percent of farms with at least \$1 million in sales in 2015.”

⁵ Growth of the rural non-farm economy has facilitated growth in off-farm employment for farm household members. Between 1930 and 2012, the share of full-time farm operators fell from 70% to 40% and their average number of days worked off-farm increased from 86.5 to 143.2.

⁶ Alston et al. (2010) provide detailed state-level and national data on inputs, outputs, and productivity in U.S. agriculture during the period 1949–2002 in a book-length treatment. Updated state-level (1949–2007) and backcast national data (1910–2007) are available on the InSTePP website (see Pardey et al. 2006 for data documentation), and discussed by Pardey and Alston (2019).

the effects of changing production possibilities enabled by new technologies.⁷ The category of nursery and greenhouse marketing in particular has grown relatively rapidly.

Agricultural Productivity

Reflecting these and other changes, over the past 100 years and more U.S. agricultural productivity grew rapidly—albeit unevenly over time and across states. Over the period 1910–2007, multifactor productivity (MFP) in U.S. agriculture grew at an average annual rate of 1.42%, reflecting average annual growth of 1.58% in the index of aggregate output and 0.16% in the index of inputs. These averages reflect shrinking total inputs (and even total output) in some states as they shifted out of agriculture, contrasted by comparatively rapid expansion in inputs and output in some other states. And they also reflect changes in the composition of inputs and outputs, as already discussed. Since WWII, MFP has grown generally rapidly in U.S. agriculture, but this reflects a surge (during the 1950s–1980s) followed by a slowdown such that MFP has been growing at about 1% per year since 1990, compared with about 2% per year during 1950–1990 (see Andersen et al. 2018, and Pardey and Alston 2019).⁸

[Table 1: *Growth Rates in U.S. Agricultural Input, Output, and Productivity 1910–2007*]

⁷ Among other changes, improved communications, electrification, transportation, and logistical infrastructure, meant that perishables and pre-prepared foods could be moved efficiently over much longer distances. This contributed to the changing spatial patterns of production.

⁸ This raises questions about the relative productivity performance of agriculture compared with other sectors of the U.S. economy. Following Jorgenson and Gollop (1992) it has become a stylized fact among agricultural and other economists to say productivity has grown comparatively quickly in agriculture. That may well have been true during the surge, but is less likely to be so since the slowdown. Consistent, comparable measures in suitably long time series are not readily available to permit definite answers to these questions, which are the subject of continuing study.

Off-farm Changes

The pre-and post-farm elements of the farm-based food and fiber supply chain have also been transformed, especially during the past half century. The pre-farm agribusiness industries that supply inputs used by farmers, many of which embody technological innovations—whether genetic, chemical, mechanical, or digital—have become more concentrated, more global, and more vertically integrated, while accounting for an increasing share of total value added by the sector (Figure 3). The post-farm sector, likewise has become more concentrated and more economically important, such that the farming sector represents an ever-shrinking share of the total food value chain. In 2017, the average farm share was down to 14.6 cents per dollar of food expenditure by consumers (USDA-ERS, 2019).

[Figure 3: *Concentration in the Food Retailing Sector*]

The income elasticity of demand for food per se is quite low for most Americans (see, e.g., Okrent and Alston 2011), but the demand for services associated with food and other “quality” attributes of food is much more income elastic. With rises in per capita incomes and the opportunity cost of time—especially for women as they have progressively entered the workforce—Americans have increasingly demanded more food away from home (accounting for more than half of food expenditure and more than one-third of calories since 2010) and more prepared foods and ready-to-eat foods for consumption at home (Okrent et al. 2018, Saksena et al. 2018) (Figure 4). In addition, they are demanding food differentiated in various ways according to both product attributes (related to nutrients, food safety, and so on) and process attributes (related to technologies used on farms such as genetically engineered varieties, organic practices, pesticide use, animal husbandry practices, and so on).

[Figure 4: *Food Expenditures at Home and Away from Home*]

The choices of technologies available to farmers and the types of food products available to consumers are increasingly being mediated by the food processing, manufacturing, and retailing sector (e.g., Saitone et al. 2015), cognizant of the evolving consumer marketplace and the influence of activist organizations (Rausser et al. 2015, 2019; Alston 2019). Indeed, private regulation of farm technologies imposed by food manufacturers and retailers might come to supplant government regulation in this domain. In turn, these shifts have implications for the demand for technologies expressed by farmers and the derived demand for investment in R&D and innovation, and ultimately these processes will shape the future path of demand for farm inputs, supply of farm outputs, productivity, and prices.

4. Investments in Innovation

Farmers are tinkerers. The 10,000 year history of agriculture and agricultural innovation includes only a century or two of organized science and other institutions that foster innovation (Ruttan 1982; Pardey et al. 1991). In the United States, since 1862—which marked both the establishment of the U.S. Department of Agriculture (USDA) and the passage of the Morrill Land Grant College Act—state and federal governments have become progressively more involved through public investments in and the performance of food and agricultural R&D. So too has the private sector, especially in more recent decades as the incentives for investing in food and agricultural innovation have strengthened.

Food and agricultural innovation investments also evolved in conjunction with, benefited from, and contributed to R&D spending directed to other sectors of the economy. During the past half century, the patterns of R&D spending overall (or gross domestic expenditures on R&D, GERD) and R&D spending directed to the food and agricultural sector (or agGERD) have continued to change in ways that we quantify and discuss in this section.

GERD versus agGERD

In 1953 the United States economy invested \$34.4 billion (2011 prices, and likewise below unless otherwise stated) in total (public and private) research and development (R&D). Over the subsequent six decades, GERD in deflated terms grew 13.5-fold (or 4.2% per year on average) to total \$465 billion in 2015 (Figure 5, Panel a). Over this same period, public and private investments in food and agricultural R&D carried out in the United States also grew markedly—albeit at a slightly slower average annual rate of 3.7% per year, again in constant prices—from \$1.27 billion in 1953 to \$12.6 billion in 2015 (Figure 5, Panel b).

[Figure 5: *GERD versus agGERD spending trends, 1950 - 2015*]

As a consequence of these differential rates of R&D spending growth, the share of food and agricultural R&D (agGERD) in total GERD gradually trended down from 3.7% in 1953 to 2.7% in 2015, in tandem with the secular decline in the agricultural share of overall economic activity. However, while the agGDP/GDP ratio shrank in a reasonably steady fashion, the reduction in the agricultural share of total R&D was less regular. Overall U.S. spending on R&D grew faster than spending on agricultural R&D during the 1950s. During the agricultural productivity surge of the 1960s and 1970s, spending on agGERD grew substantially faster than spending on GERD such that the food and agricultural share of total R&D spending peaked in 1977 at 4.2% (versus 2.5% in 1961). Thereafter, GERD grew faster than agGERD, such that by 2015 the food and agricultural share of GERD had reverted to the low point of the early 1960s.

Although the food and agricultural share of GERD has gradually declined over recent decades, the food and agricultural sector continues to invest more intensively in R&D than does the U.S. economy as a whole. Investments in food and agricultural R&D when expressed relative to agGDP grew steadily from just 0.7% in 1950 to a peak of 9.2% in 2002, shrinking

thereafter to 7.7% in 2015 (Figure 6). At 2.75% in 2015, the economy-wide intensity of R&D investments in the United States was half the corresponding food and agricultural research intensity, and roughly equivalent to the intensity of GERD investment that prevailed half a century earlier (i.e., 2.79% in 1964).

[Figure 6: *The Intensity of Investment in GERD versus agGERD, 1950 – 2015*]

Private versus Public Research

The private sector has long accounted for a significant share of U.S. GERD, averaging 70.2% of the total since 1953, albeit with a period during the 1960s and 1970s when the share of private R&D investments fell well below the long run average, bottoming out at 65.7% in 1975 (Figure 5, Panel a). While the private share of food and agricultural R&D (70.0% in 2015) is now roughly in line with the private share of research overall, this is a relatively recent development, with the private sector accounting for just over one-third of U.S. food and agricultural R&D in the early 1950s (Figure 5, Panel b).

Changes in the scope of U.S. intellectual property (IP) protection were associated with a rise in agricultural innovations coming from the corporate sector. Mechanical, chemical, storage, transport and processing inventions pertinent to food and agriculture have long been subject to patent protection (as well as copyright, trademark, trade secrecy and eventually other legal means) enabled by Article 1, Section 8 of the U.S. Constitution that became operational on March 4, 1789. However, little protection was afforded biological inventions such as new crop varieties (likewise for genetics-related innovations in the health sector). Trademarks and trade secrecy laws were applicable, but these did not protect against reverse engineering or self-

replication, and so the common practice of saving seeds for own reuse—or for sharing with other farmers or selling to them—did not constitute legal infringement of new seed varieties.

It was not until 1930 that legal forms of plant varietal rights were on offer in the United States, and the corporate share of such rights issued rose from 55% in the 1930s and 1940s to 82% by 2008 (Pardey et al. 2013). The 1930 Plant Patent Act covers asexually reproduced plants, a category that largely encompasses ornamental plants and fruits. Sexually reproduced crops, a category that includes grains, oilseed crops and grasses, gained intellectual property protection in 1970 by way of the Plant Variety Protection Act (PVP). A third form of protection became possible in 1980 with *Diamond v. Chakrabarty*, in which the U.S. Supreme Court narrowly found that “anything under the sun that is made by man” is patentable subject matter (Wright et al. 2007). In practice, this case and subsequent legal rulings clarified that plant varieties, parts of plants, genetically engineered organisms and gene products themselves were eligible for the same U.S. utility patents that cover most other inventions.

This IP landscape evolved hand-in-hand with important changes in the genetics and genomics sciences that support crop varietal development, notably mechanical (Taylor and Fauquet 2002) and bacterial (Gelvin 2003) means of manipulating genes to produce GMO crops, or more recent, more precise, and more cost-effective means to edit genes such as TALEN and Crisper-Cas9 technologies (Balten et al. 2017). The legislative and legal changes that gathered momentum in the 1970s and 1980s (Pardey et al. 2013) preceded a substantial rise in the amount of private research oriented towards biological innovations during the subsequent decades by firms such as Monsanto (acquired by Bayer in June 2018) and Pioneer-Dupont (now Corteva Agrisciences).

In addition, the share of food, beverage, and tobacco processing and manufacturing research in total food and agricultural R&D has continued to be a big part of the private R&D effort over the past 50 years (averaging 42.2% of the private U.S. total over the past two decades) as industry has sought to respond to changing consumer demands, including an increase in the share of food and beverages consumed away from home and the desire for prepared foods and those packaged in more convenient forms (Figure 4).

Notably, the more expansive IP protection afforded the food and agricultural sector in the past few decades was associated with a continuing increase (albeit at a slower rate) in the intensity of private food and agricultural R&D, while the intensity of public agGERD has trended down over the past 15 years (Figure 6, Panel b). In contrast, the intensity of public GERD has changed little since the mid-1960s, such that most of the increase in overall GERD intensity since the late 1970s is attributable to an upward trend in the intensity of private investments in R&D (Figure 6, Panel a).

These overall intensities mask considerable differences among sectors in the intensity of R&D investment by private firms. Table 2 reports aggregate firm sales and R&D spending data grouped into sectors, along with their corresponding intensity of investments (here measured relative to their net domestic sales). Pharmaceutical and medical firms, on average, spent \$12.9 on R&D for every \$100 of sales in 2015, while chemical companies averaged \$6.7 and information-related companies \$5.9, making them the three most R&D-intensive sectors in the NSF compilation. The InSTePP compilation of data on 132 firms undertaking research related to food and agriculture had an R&D intensity that averaged 1.2% in 2014 (Lee et al. 2020).⁹ But

⁹ InSTePP is the International Science and Technology Practice and Policy center at the University of Minnesota, details of which are at www.instepp.umn.edu.

there were marked differences in the intensity of research within that sector. Although total sales for food companies were more than double those of agricultural companies in 2014, food companies invested less in R&D such that their intensity ratios averaged just 0.7%, versus 2.6% for agriculture-related companies. Nonetheless, these data place food companies on par with “other non-manufacturing companies” in terms of their R&D intensities, while the intensity of research investments by agricultural companies exceeds that of the automobile, other manufacturing, other chemical, and finance sectors.

[Table 2: *Company sales and R&D intensity, 2014/15*]

While the increasing investments in innovation by private (food and agricultural) firms have obviated the need for some public research, much of the private research stands firmly on the shoulders of publicly performed research.¹⁰ The different roles played by public and private research are revealed to some extent by the substantial differences in the composition of the research performed by both sectors. Around 44% of the food and agricultural research performed by the public sector in the period 2007–2009 was considered “basic” research (Pardey and Beddow 2013), where the notional objective is the pursuit of new knowledge or ideas without specific applications in mind (OECD 2015, p. 29). Another 47% of public research was classified as “applied,” or research done to meet a specific need. Only 9% was deemed “developmental” and directed towards the production of specific products and processes with nearer-term commercial potential. This squares with the basic share (44%) of public GERD (NSB 2018, p. 4-3). In contrast, 78% of private GERD, was “developmental” in nature intended

¹⁰ For examples of public-private research complementarities in the food and agricultural sector see Fuglie and Schimmelpfennig (2000) and Pardey and Beddow (2013).

to develop prototypes, new processes, or products for commercialization, with only 16% of private research considered applied and 6% basic.

International Dimensions of U.S. Research

In 1960, the United States accounted for 18.6% of the entire world's expenditures on publicly performed food and agricultural R&D, but by 2015 that share had shrunk to 9.4% and the United States was eclipsed by China, which by 2013 had begun outspending the U.S. on *both* public and private food and agricultural research (Chai et al. 2019). Part of this shifting research geography reflects policy choices in the United States versus China and other countries—especially other large agricultural economies such as Brazil and India—regarding public spending and other forms of legislative support for agricultural R&D. Part of these changing international R&D relativities reflects more fundamental economic forces shaped by major changes in the economic geography of agriculture itself (Pardey et al. 2016).

In 1961, the United States accounted for 14.8% by value of the world's agricultural output, compared with China's 8.5%. Less than six decades later the tables had turned. The U.S. share of global agricultural production had declined to 10.4% in 2016, while China now accounted for almost one-quarter (23.7%) of the total, propelled by historically unprecedented and sustained rates of growth in Chinese agricultural production and productivity. These Chinese agricultural developments were enabled by several radical institutional reforms beginning in the late 1970s; notably the introduction of the household responsibility system for farming that incentivized farmers to increase output and spurred the off-farm migration of labor (Lin 1992), and the doubling down on investments in agricultural R&D (Chai et al. 2019). Meanwhile, the U.S. government first slowed and then of late scaled back public support for agricultural R&D.

These shifts in the global landscape for food and agricultural R&D also paralleled broader changes in the world's economic geography and the country composition of global GERD. In 1980, the United States accounted for 19.8% of the world's \$6.4 trillion (2009 purchasing power parity prices) GDP, compared with 2.1% for China. By 2014, while global GDP had grown to \$99.3 trillion the U.S. share had shrunk to 16.2% while China's had grown to 16.9%. Dehmer et al. (2019) estimated that over this same period, global GERD had grown from \$0.48 trillion in 1980 to \$1.67 trillion in 2014, the U.S. share of global GERD had declined a little from 31.2 to 27.0%, while China's share had increased dramatically from 1.2 to 20.0%.

Where in the rest-of-the-world agricultural R&D takes place matters as much as the amount and types of research conducted in the United States for the innovative future of U.S. agriculture. Just as genetic innovations conceived in the health sector have benefited agriculture (and vice versa), rest-of-world agricultural knowledge stocks have spatial spillover consequences for U.S. agriculture. However, space matters more for agriculture, and many agricultural innovations are site-specific. Consequently, taking wheat as an example, research targeted for agroecologies (or production systems) that are agroecologically distant from current or prospective wheat areas in the United States, are likely to be less consequential for wheat innovation in the United States than if they were targeted to U.S. agroecologies.¹¹ Thus, with an increasing share of the world's agricultural research taking place outside the United States, the global stock of scientific knowledge can be expected to have less relevance for innovations within U.S. production agriculture in the decades ahead relative to decades past.¹²

¹¹ At any given point in time hundreds of different wheat varieties are being grown or bred that are adapted to specific agroecologies, and the productive potential of any particular variety of wheat varies greatly depending on where in the world, precisely, that variety is to be grown.

¹² As discussed by Alston (2002) and emphasized by Alston et al. (2010), agricultural technology spillovers are

5. Payoffs to Investments in Agricultural Innovation

In the economic evidence on the payoffs to investment in R&D the internal rate of return (IRR) has been used as the statistic of choice to summarize the streams of costs and benefits associated with R&D activities that typically take (sometimes considerable) time for the research to be conducted, and years if not decades for the resulting innovations to diffuse and realize economic consequences. This is true, for example, for researchers summarizing the economic consequences of manufacturing R&D (see the tabulation in Hall et al. 2010, Table 2) and health related research (e.g., HERG et al. 2008; Deloitte 2014; Glover et al. 2014). Notwithstanding Griliches' (1958) objection to the use of IRRs in this context, it is also the preferred summary statistic for the now extensive literature on the returns to agricultural R&D, which he initiated.¹³

The InSTePP agricultural returns-to-research database (version 3) includes 3,426 rate-of-return estimates gleaned from 492 studies worldwide, of which 1,298 (37.9%) of the estimates evaluate research conducted in the United States (Rao et al. 2019a). Among the U.S estimates, 76% (986 estimates) report IRRs, while 312 report benefit-cost ratios (BCRs) (Table 3, upper half). Given the wide dispersion and positive skewness in the distribution of the reported rates of return, the median is a more informative measure of the central tendency than the mean. The median of the reported IRRs for U.S. agricultural R&D is 31.9% per year and the median of the

significant, and run in both directions, though not always symmetrically or spontaneously. Cognizant of this fact, private foundations based in the United States together with the U.S. government led the funding (as well as the founding) of the system of international agricultural research centers now known as the CGIAR. This institutional innovation was conceived for essentially humanitarian purposes to address the global food crisis of the 1960s, and was a primary source of the so-called "Green Revolution" technologies (see, e.g., Alston et al. 2006 and Wright 2012). It reduced but did not eliminate a global market failure in agricultural R&D that persists today. The donor countries also benefited by adopting the resulting technologies, as quantified, for example, by Pardey et al. (1996) in relation to the spillins of CGIAR crop varietal technologies into the United States.

¹³ With reference to using the IRR as a summary returns-to-research measure, Griliches (1958, p. 425) wrote "My objection to this procedure is that it values a dollar spent in 1910 at \$2,300 in 1933. This does not seem very sensible to me. I prefer to value a 1910 dollar at a reasonable rate of return on some alternative social investment."

reported BCRs is 12.0, roughly in line with the corresponding medians for the reported rest-of-world evidence.¹⁴

[Table 3: *Returns to Agricultural R&D*]

Although the IRR is merely a breakeven interest rate, equating the present values of costs and benefits, many policymakers (and even some economists!) treat IRRs as compounding rates of interest, analogous and comparable to the returns reported for financial products (e.g., mortgages, mutual funds and certificates of deposit). However, Hurley et al. (2014) showed that such an annualized percentage rate of return interpretation is generally incorrect and often leads to incredible implications.¹⁵ They also pointed out how the modified internal rate of return, MIRR (Lin 1976), offers an alternative to the IRR that can be reasonably interpreted as an annual percentage rate of return.

Hurley et al. (2014 and 2017) provide a detailed account of the properties of MIRR, and the implicit, often undesirable assumptions (not least in the returns-to-research context) made in the calculation of IRRs. One of the desirable properties of an MIRR is its one-to-one correspondence with a BCR if a common discount rate and research timeline are used to calculate the rates of return for different projects. Using the BCR-IRR relationship elucidated by Hurley et al. (2014), Rao et al. (2019b) developed and deployed a procedure to recalibrate the reported IRRs into a standardized set of imputed BCRs and MIRR, where the discount rates and

¹⁴ However, such direct comparisons of broad aggregates are of limited value given differences in the nature of the evidence across countries and over time in terms of what commodities and types of research are being evaluated, and the details of the evaluation methods themselves (Rao et al. 2019a).

¹⁵ As Alston et al. (2011, pp. 1271-2) showed, "... if the roughly \$4 billion invested in public agricultural R&D in 2005 earned a return of 50% per annum compounding over 35 years, by 2040 the accumulated benefits would be worth \$5,824,000 billion (2000 prices)—more than 100 times the projected U.S. GDP in 2040 and more than 10 times the projected global GDP in 2040."

research timelines are held constant, thus improving comparability among the estimates. The bottom half of Table 3 presents these imputed BCRs and MIRR for the United States and the rest-of-the world using a common discount rate of 5% and a research evaluation timeline (from the initiation of costs to the cessation of benefits) of 30 years, roughly the average timeline of the reported evidence. These results indicate a median BCR of 7.5 for investments in U.S. agricultural R&D, corresponding to an MIRR of 12.3% per year. The comparable median rest-of-world estimates are a BCR of 9.0 and an MIRR of 13.0% per year. These estimates are indicative of a sustained and substantial underinvestment in agricultural science both in the United States and globally. In spite of this government failure on top of market failure, innovation in American agriculture has accomplished a great deal and has contributed positively to global agricultural growth and change.

6. Clusters of Innovation in American Agriculture

Pardey and Alston (2019) discuss and document a century of transformative change in U.S. agriculture (1910–2007) in which they pay particular attention to the potential sources of a mid-century (1950s–1980s) surge and subsequent slowdown in farm productivity growth. They liken this farm productivity pattern to the earlier surge and slowdown in the broader economy identified by Gordon (2000, 2016), which he associated with great “clusters of inventions.”¹⁶ Borrowing those ideas and in a similar spirit, Pardey and Alston (2019) suggest that much of the

¹⁶ Rasmussen (1962) had a similar notion of technological clusters or sequential technological revolutions driving the arc of history regarding U.S. agriculture, noting that up to the time of his writing: “Two revolutions in American agriculture reflect the impact of technological change on farming during the past century. The first revolution saw the change from manpower to animal power, and centered about the Civil War. The second revolution saw the change from animal power to mechanical power and the adaptation of chemistry to agricultural production. It centered around the post-World War II period. The transition from animal power to mechanical power is virtually complete (Rasmussen, 1962, p. 578).”

past time path of U.S. agricultural input use, production, productivity, and prices can be understood in terms of clusters of agricultural inventions and the structural changes in the farm economy they enabled, including (1) “mechanical” (mostly labor-saving) technologies; (2) improved animal breeds and crop varieties and other “biological” innovations; (3) synthetic fertilizers, pesticides, and other “chemical” technologies; and (4) more recently, “information” technologies; all employed in conjunction with ever-evolving knowledge and improved understanding of and changes in agricultural production and practices. This section builds on the discussion of innovation clusters by Pardey and Alston (2019) paying greater attention to the more-recent period while placing the newer innovations in a longer-run historical context.

Mechanization played a large early role in the 20th Century transformation of U.S. agriculture. As well as much human labor, machines saved considerable amounts of land from having to be used to produce feed for horses and mules, and facilitated the consolidation of farms into many fewer and larger units. The tractor, in particular, saved millions of acres of land and the work of many men and women. In 1910, the United States had a total of 6.4 million farmers, farming 881 million acres using a total of 24 million horses and mules and just 1,000 tractors. After its peak in 1917 at 27 million animals, the stock of work horses and mules on U.S. farms dropped eventually to a low of 1.4 million in 1974 (Figure 7). Meanwhile, from just 50,000 tractors in 1917, the total grew to 4.5 million tractors in 1957.

[Figure 7: *American Agricultural Mechanization, 1867–2012*]

Along with tractors, farmers also adopted automobiles and motor trucks that were not developed specifically or exclusively for agriculture, as well as other machines that were useful only in agriculture, such as reapers, mowers, binders, and combines (Figure 8, Panel a). To a great extent this was a private-sector process, in which many of the innovations were patented

technologies embodied in tractors and related machines produced privately and sold to farmers. Public agricultural (and other) R&D was complementary but played a minor role here compared with its role in other types of innovations such as new crop varieties, and farming systems.

[Figure 8: *Adoption Paths for Selected Major U.S. Farming Innovations, 1920–2018*]

As can be seen in Panel a of Figure 8, biological innovations, in particular improved crop varieties that were responsive to chemical fertilizers, took center stage a little later—although they were clearly part of the story all along (Olmstead and Rhode 2008). For example, hybrid corn varieties were adopted rapidly in Iowa in the early 1930s, but it took until the 1960s for vastly improved hybrids to achieve 100 percent adoption throughout the United States (Griliches 1957, Dixon 1980, Hallauer and Miranda 1981). Varietal improvement has continued for corn and other crops, including food crops such as wheat and rice for which public investments have been more important. These innovations, with others, laid the foundation for genetically engineered (GE) hybrid corn varieties to be developed and adopted, beginning in 1996 (Fernandez-Cornejo et al. 2014). Similar, though typically less dramatic genetic innovations were common to many agricultural crop and livestock species, and contributed to the rapid rise of yields and aggregate productivity during the second half of the 20th century (Olmstead and Rhode 2008).

Chemical technologies for agriculture assumed importance beginning in the 1960s. The early 20th century invention of the Haber-Bosch process for the economical manufacturing of synthetic nitrogen fertilizer in particular was profoundly important for enhanced crop yields, especially when combined with complementary genetics and crop management practices. The U.S. on-farm adoption process for these fertilizers and associated varieties was notable in the 1960s and 1970s. Partly as a post-war dividend, synthetic pesticide technologies took off around

the same time, soon to become subject to environmental regulation following the publication of Rachel Carson's "Silent Spring" in 1962. A great many forms of agricultural chemical technologies have been developed, registered and approved for commercial use, and adopted by farmers (and in many cases subsequently deregistered or heavily regulated and disadopted)—including fertilizers, growth promotants, herbicides, fungicides, insecticides, antibiotics and other veterinary medicines.

More recently, genetically engineered (GE) crop varieties have been developed that could serve as better complements to or substitutes for these chemical technologies. Predominant among these are herbicide tolerant (HT) varieties of cotton, corn, soybeans, and canola, which permit herbicides to replace mechanical tillage for weed control, and pesticide inherent (PI) varieties (e.g., Bt cotton and corn) that reduce (or eliminate) the requirement for chemical pesticides for controlling specific pests. These technologies tend to predominate where they are available for adoption and confer significant benefits to farmers, consumers, and technology firms (see, e.g., Fernandez-Cornejo et al. 2014; Qaim 2016). Furthermore, herbicide tolerant technologies are complementary to conservation tillage practices (Perry et al. 2016), such that minimum-till and no-till systems are now used on a majority of acres of U.S. wheat (67% in 2017), corn (65%, 2016), and soybeans (70%, 2012) (Classen et al. 2018). Conservation tillage practices reap their own rewards in terms of reduced soil compaction, improved water infiltration (and so reduced runoff), promotion of soil fauna and biological processes, and increased soil organic matter, most of which have beneficial agricultural production and environmental outcomes.

To date, however, GE varieties have been developed and adopted widely for only a few crops (in particular soybeans, cotton, corn, canola, and papaya) and only in a few countries

(Qaim 2016). This reflects a combination of regulatory and market resistance, which has discouraged the development of technologies for these and other applications in the countries that are open to GE technologies; more so in those that are opposed. The United States is predominant both as a developer and adopter, but even in the United States the regulatory barriers are substantial, adding years of delay and hundreds of millions in costs (see, e.g., Kalaitzandonakes et al. 2006).¹⁷ Panel b of Figure 8 displays the U.S. adoption paths for GE varieties of corn and soybeans on an expanded shorter time scale, where they can be juxtaposed with adoption paths for digital (and related information) technologies used in U.S. crop production in the modern era. This Panel captures key elements of the current innovative landscape for agriculture, emphasizing genetic innovation and other data-intensive technologies.

Digital farming technologies (sometimes referred to as “precision agriculture technologies” or “variable rate technologies”), which help farmers gather information and adjust production practices according to changes in field conditions over space and time, are beginning to gain ground in U.S. farm production. Major examples include GPS (global positioning system)-based remote-sensing and guidance systems, soil and yield mapping based on GPS (perhaps in conjunction with satellite and aerial photography), and variable-rate technology (VRT) (Lowenberg-DeBoer and Erickson 2019; Jia et al. 2019). As discussed by Schimmelpfennig (2016), field operators can use GPS guidance systems to auto-steer tractors, combines, and other machines, which helps pinpoint precise field locations and reduce operator

¹⁷ Compared with the earlier path for hybrid corn, these ones are shorter reflecting the advantages of modern science and communications technologies, but once we allow for an additional decade of regulatory lags, the overall post-research development and adoption process is still on the order of 20 years for these technologies—like hybrid corn within Iowa.

fatigue. Guidance systems also save money by reducing over- and under-application of sprays and improving the precision of seeding.

Precision technologies can also help farm operators map their fields better. Yield monitors mounted on harvesters can be used to record data on yields with GPS coordinates that the operator can use to monitor changes in crop yield across the field and from year to year. These data can be combined with data from soil maps using related technology—based on core samples or soil sensors that use electrical conductivity to test soil—to better understand the sources of yield variation and act on the information. Lastly, variable-rate technology (VRT) allows farmers to customize the application of irrigation water, fertilizer, chemicals, and pesticides spatially and over time using data from remote sensors or GPS data—often from yield and soil maps or guidance systems. Farmers can even use VRT to plant different types of seeds or to apply different agricultural chemicals at different rates at different locations with a single pass of the tractor. And, looking forward, these technologies will be used to allow precise mechanical or chemical weeding and cultivation around individual plants, among other things.

Panel b of Figure 8 includes adoption curves for six types of technology in this context, as well as two types of GE crops (corn and soybeans). The plots for the GE crops start earlier, and rise more rapidly. By 2006, only 10 years after initial release, GE varieties had already been adopted on 60 percent of corn and 90 percent of soybean acres. In contrast, the adoption curves for most of the digital and related precision farming technologies are much flatter. Autosteer technology, which was first introduced around the same time as the GE crop varieties was used on 60 percent of acres in 2015, while the other digital technologies were less widely adopted. As discussed by Schimmelpfennig (2016) some of these technologies are simple to adopt and easy to use while others may come at a relatively high cost in terms of requiring specialized

equipment or specific skills. These factors, along with functionality, have contributed to the differences observed in the rate of development and adoption of these technologies. Many of these tools are knowledge- and skill-intensive, and their profitable use requires location-, application-, and site-specific adaptation, all of which takes time to work out. Their limited use to date reflects the fact that it is still early days for many of these technologies in terms of matching data to models and real-time processing of very large data sets, engineering the machinery to go with the data and knowledge, and software development. The technologies themselves have been changing rapidly and the prospects are very bright.

In addition to these types of technologies, which are most visibly apparent in the context of major field crops, other types of digital technologies are being developed and used in the context of specialty crops or livestock production, as well as field crops. For example, in crop production, automated irrigation systems based on sensors that measure precipitation, evapotranspiration, and soil moisture in the root zone are used to improve the efficiency of water use and save labor and are in their early stages of adoption. USDA, ERS (2020) report that “... fewer than 10 percent of irrigators make use of soil- or plant-moisture sensing devices or commercial irrigation scheduling services [and less than] 2 percent make use of computer-based simulation models to determine irrigation requirements based on consumptive-use needs by crop-growth stage under local weather conditions.”

Autonomous or articulated weeding machines that can selectively spray or physically cut out weeds based on GPS references or computer vision (or image identification) technologies, thereby saving labor and reducing the environmental burden of herbicide (see, e.g., Fennimore and Cutulle, 2019; Filmer 2019) are under development, and some are already being used in farmers’ fields (see, e.g., BlueRiver 2020), while drones and other self-driving machines fitted

with cameras, sniffers and other types of sensors are being developed and deployed to monitor the crop for drought and other stresses and check for pests and diseases. Analogous sensing technologies are being developed for monitoring soil nutrient (see, e.g., Teralytic 2020) and health status, including soil microbiome activity. Robotic and other devices fitted with sensors are also being developed and deployed to selectively harvest crops (e.g., apples, strawberries) with variable maturation dates. In livestock production, digital technologies are already widely used and their use is progressing: for example, companies such as Lely (2020) have already commercialized robotic milking machines and digital cow tag systems that monitor and help manage animal health and feeding status.

Growing from “just” \$2.6 billion in 2012 new agrifood startup companies attracted \$16.6 billion of investments worldwide in 2018 (Agfunder 2018, p. 15). The United States accounted for \$7.9 billion (48%) of the total, China \$3.5 billion (21%), and India \$2.4 billion (14%). Startups based in California accounted for almost one-third of global investment and two-thirds of all U.S. investment; 92% of the U.S. total if funding to firms located in Massachusetts, New York, North Carolina, Colorado and Minnesota are also included. Investments in startups closer to the farmer—spanning areas such as soil testing, pest detection, precision agriculture, digital agricultural management, agronomic data, and predictive analytics—totaled \$6.9 billion (41.5% of the global total), with firms focused on the off- (or vertical-) farm segments—including food processing and production companies (such as Impossible Foods, Zymergen and Bowery Farming Inc.) and, especially, food delivery companies (such as Instacart and DoorDash) attracting \$10 billion in risk capital.

7. The Next Wave of Agricultural Innovation: Ripple or Tsunami?

Looking forward, we can see great potential for new product and process innovations—in particular digital and other data- and knowledge-intensive technologies, including genetic innovations—that will enable more and better food, fiber, and industrial raw materials to be produced on farms at much lower cost and with a smaller environmental footprint, worldwide. Realizing this potential will matter for the future trajectory of global public goods including climate change, other natural resource stocks, the world food equation, poverty and related civil or military strife. The extent to which these opportunities will be captured, and when, will be determined to a great extent by forces outside agriculture and outside the R&D and technology sector. These forces will determine the availability and direction of resources available for public sector agricultural R&D, the regulations and rules governing the development and adoption of new farm and food technologies, and the demand for products depending on the technologies used to produce them.

Induced Innovation

As noted above, agriculture is unusual in that it faces knowledge depreciation arising from climate change and, in particular, the coevolution of pests and diseases. This gives rise to a demand for maintenance R&D—simply to preserve past productivity gains. Much of the past work on crop varietal innovations can be seen in this light. The demand for innovation on farms is also driven by changing factor supply conditions, evolving demand for farm products (now including feedstock for biofuels and other industrial raw materials as well as traditional feed, food and fiber), and the peculiar regulatory environment for agriculture related to issues including varietal technologies, animal welfare in livestock production, and landscape amenities (and dis-amenities) from agricultural production. Farmers also face a changing market

environment with demands for food products and food production processes mediated through private standards and mass media messages.

Over the long history, a major element of change was labor-saving innovation induced by farm labor scarcity. Past labor savings notwithstanding, reliable and timely availability of suitably skilled labor is a major concern of farmers today—especially in California’s labor demanding specialty crops—and they are actively seeking technological alternatives for harvesting, weeding, and a host of other farm operations.¹⁸ Farmers are also increasingly concerned over the reliability of natural rainfall and irrigation water, with variability and uncertainty in these dimensions exacerbated by climate change. Drought- and heat-tolerant varieties are being developed to mitigate these consequences (see, e.g., Cooper et al. 2014 and McFadden et al. 2019 in the case of drought tolerant corn). Information technologies combined with more precise and selective water delivery systems can reduce total water usage and vulnerability to drought.

Changing technological regulations generate a demand for replacement technologies. In recent years significant agricultural pesticides have been banned in some jurisdictions and are threatened in others owing to concern about their risks to the environment or human health. These include soil fumigants (e.g., methyl bromide), insecticides (e.g., neonicotinoids) and herbicides (e.g., glyphosate aka Roundup®). When significant pesticides are deregistered, farmers demand new solutions. In some cases, the alternative to a banned chemical is another chemical or new genetics, but sometimes it simply means technological regression. For

¹⁸ Today’s farm labor environment with its implications for the demand for labor-saving innovations is reminiscent of the period when the Bracero Program was terminated in 1964, stimulating the rapid deployment and adoption of the mechanical tomato harvester. Olmstead and Martin (1985) analyze the resulting controversy.

example, Roundup-resistant® varieties of corn, soybeans, and canola, combined with the herbicide glyphosate, permitted the widespread adoption of lo-till or no-till production systems that resulted in significant improvements in soil structure and reduced greenhouse gas emissions. If glyphosate were to be banned in the United States and Canada—as it has been (either totally or for selected uses) in some other countries recently—we could expect to see a reversion to older production systems using mechanical tillage for weed control and environmentally less benign herbicides. The pressure will be on to come up with an alternative to glyphosate that will be as effective for farmers and more acceptable to the regulators. This is a serious challenge.

Agriculture has generated various other environmental concerns related to air pollution (including greenhouse gases, particulate matter, and odors from livestock production) and water pollution (including nitrates in groundwater and surface water giving rise to human health and environmental issues). With increasing awareness of these issues, and the likelihood of government intervention in one form or another, demand is growing for alternative technologies that will enable more precise use of inputs and better control of unwanted outputs. Likewise, whether motivated by animal welfare concerns or other issues, new regulations over livestock production practices—such as castration; dehorning; pens for calves, sows, and egg-laying hens; use of antibiotics and other veterinary medicines—give rise to demand for new technologies.

In many instances, genetic innovations offer promising solutions to the problems created by the changing regulatory environment. However, genetic technologies also are subject to considerable regulatory weight. The science of genetic innovation has improved by leaps and bounds over recent decades, but society has placed arbitrary strictures (unsupported by scientific evidence) over some of the most powerful tools in the tool-kit available to the modern-day geneticist. In the United States genetically engineered crop varieties are subject to much greater

regulatory control than their conventionally bred counterparts, even though they pose no greater risk to human health or the environment (see, e.g., Qaim 2018). In many other countries GE crops are effectively banned. More recent innovations, such as gene-editing techniques, promise much greater possibilities for targeted genetic changes in commercial species, but they too might face serious regulatory barriers that could stifle that potential.¹⁹ Some countries have already opted to treat gene-edited varieties as GMOs, subject to severe restrictions (see, e.g., Wight 2018, regarding the European Court of Justice ruling regarding the use of gene editing in the EU).

It is not easy to get a good handle on the innovations in the pipeline or on the drawing board, especially since so much of what is going on is being undertaken privately, and in private—in particular when we talk about digital agriculture but also for some aspects of genetic innovations.²⁰ As we have discussed, genetic innovation in plants and animals includes the results from conventional breeding (albeit supported by the tools of modern biotechnology such as marker-assisted breeding), genetic engineering, and gene-editing. Much of the emphasis of this work tends to be focused on the main agricultural species and the main production systems, for sound economic reasons. Apart from yield potential, tolerance of abiotic stresses (drought, frost, and heat), and resistance to pests and diseases, crop geneticists are looking for various other agronomic advantages and product quality attributes. In the case of apples and table grapes, for example, fruit quality attributes are an important focus of private and public breeding efforts and varieties in use are changing rapidly.²¹

¹⁹ Van Eenennaam (2019) provides a review of the gene-editing targets for cattle, while Baltes et al. (2017) provide an overview of the gene editing opportunities and technical challenges for plant transformations.

²⁰ The public sector also is active in these areas and public-sector science is less secretive.

²¹ Alston and Sambucci (2019) discuss and document the rapid rate of innovation in table grape varieties in

As noted, digital farming innovations (including precision technologies and variable rate technologies) have the potential to save (and also reduce dependence on uncertain supplies of) labor and irrigation water; they also have the potential to save on materials and reduce environmental spillovers associated with fertilizers and pesticides. Some of these technologies also have the potential to reduce the requirement for farm labor to perform dangerous and arduous tasks that can be done better by machines. Some of these prospects will be enhanced by government policies and the political action of various interest groups, including the woke food policy movement, and others will be hampered.

Policy Perspectives

Government policy has been a central theme in our discussion of agricultural innovation because the government plays a central role both in contributing directly to the innovation process, as a major provider of agricultural R&D, and in setting the rules of the game that determine the supply and demand for agricultural innovations. In the current environment for agriculture, demands for private innovation investments are being influenced by government through the prospect of new regulations (or taxes) applied to agricultural production, including technological regulations and environmental regulations to reduce greenhouse gas emissions and other spillovers from agriculture; and through the influence of policy on the supply of inputs (especially labor and water) to agriculture, and on the markets for farm products. A more subtle influence of government is through changing support for public sector R&D (in terms of both the total investment and the balance of investments) influenced by a changing role of scientific

California, reflecting both public and increasingly private innovative activity. A total of 85 varieties are currently in production, and the mixture in vineyards is changing.

evidence in policy and shifting public preferences.²² These shifts create some opportunities for the private sector and foreclose others.

The National Academies of Science, Engineering and Medicine (2019) recently published a new agricultural research agenda for the United States entitled *Science Breakthroughs to Advance Food and Agricultural Research by 2030*. This report identifies innovative, emerging scientific advances for making the U.S. agricultural and food system more efficient, resilient, and sustainable. The report presents five priorities:

1. Increasing understanding of the animal, soil, and plant microbiomes and their broader applications across the food system.
2. Harnessing the potential of genomics and precision breeding to improve plant and animal traits.
3. Capitalizing on agri-food informatics to enable advanced analytics using data sciences, information technology, and artificial intelligence
4. Employing existing sensors and developing new sensing technologies to enable rapid detection and monitoring
5. Prioritizing transdisciplinary science and systems approaches.

A fundamental motivation for this effort was a concern over the shrinking total support for public agricultural R&D in the United States and the loss of direction in terms of the focus of the shrinking public funds. Among these five priorities, most of the topics entail significant opportunities for private entrepreneurial activity to generate proprietary research products.

²² For example, Pardey et al. (2013) show that the share of SAES (state agricultural experiment station) research oriented to farm productivity fell steadily from 69% in 1985 to 56% in 2009.

8. Conclusion

U.S. agriculture was transformed during the 20th century by waves of innovation with mechanical, biological, chemical, and information technologies. Compared with a few decades ago, today's agriculture is much less labor intensive and farms are much larger and more specialized, supplying a much-evolved market for farm products. Over recent decades, the global landscape for agricultural R&D has shifted away from farms, away from the public sector toward the private sector, and away from the United States toward agriculturally important middle-income countries, especially China, India and Brazil. Investments are stalling even though meta-evidence shows that past U.S. investments in R&D have yielded very favorable returns: median reported benefit cost-ratios in the range of 12:1. Continuing U.S. investment and innovation will be required to preserve past productivity gains in the face of climate change, coevolving pests and diseases, and changing technological regulations—let alone increase productivity. Great potential exists for innovation in crop and livestock genetics and digital farming technologies to generate new products and production processes, but innovators have to overcome increasingly strong headwinds from social and political forces that seek to dictate technology choices.

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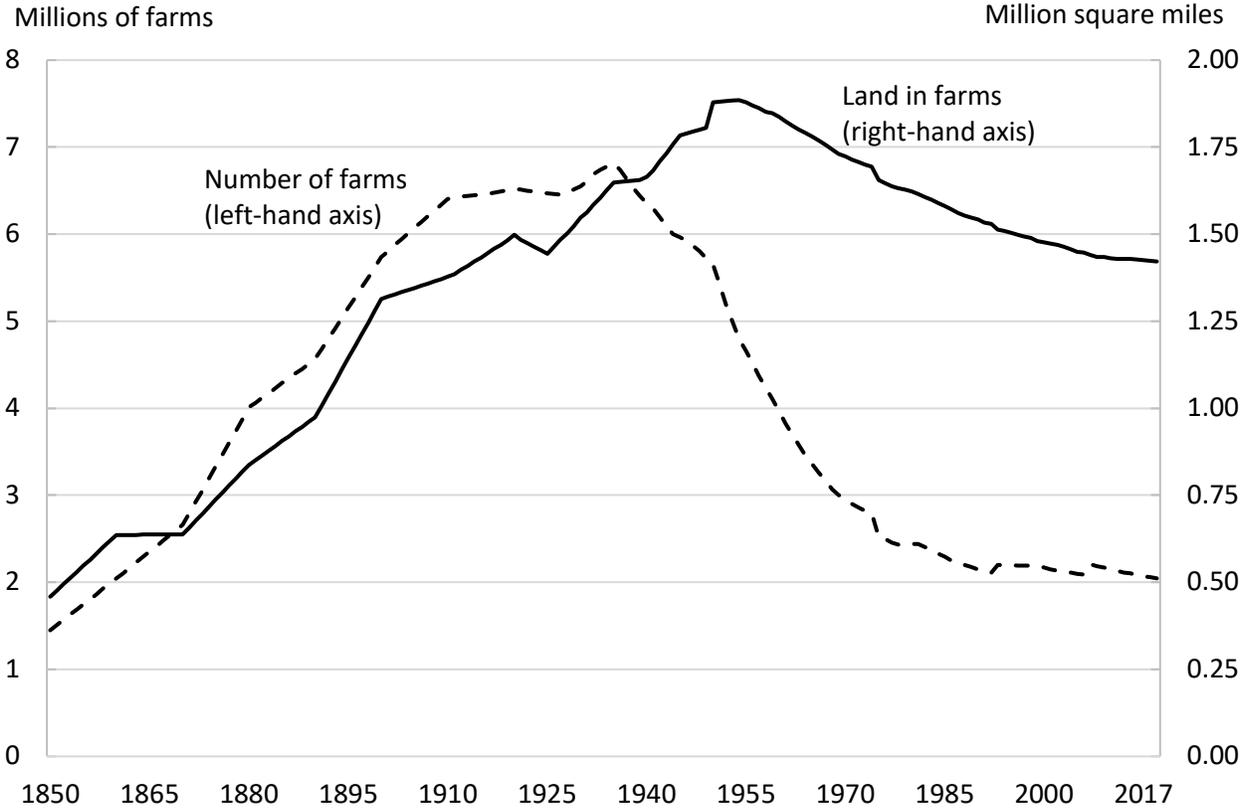
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Figure 1: U.S. farm area and farm numbers, 1850–2017

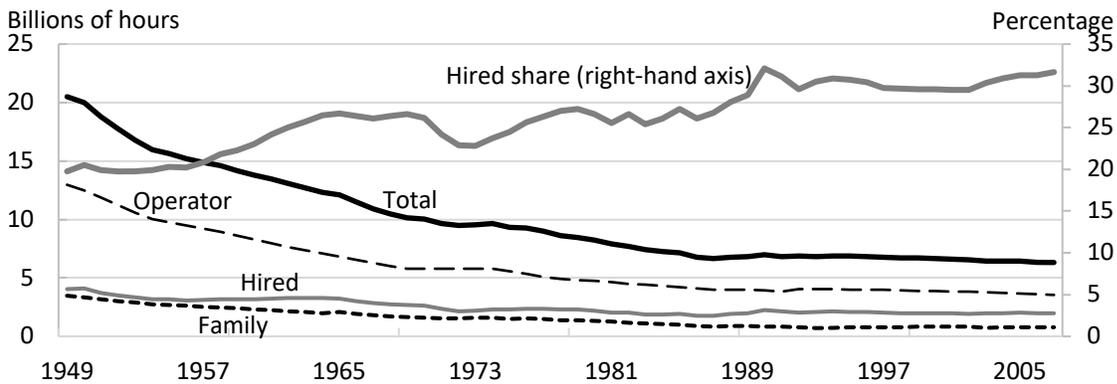


Sources: Pardey and Alston (2019, Figure 4).

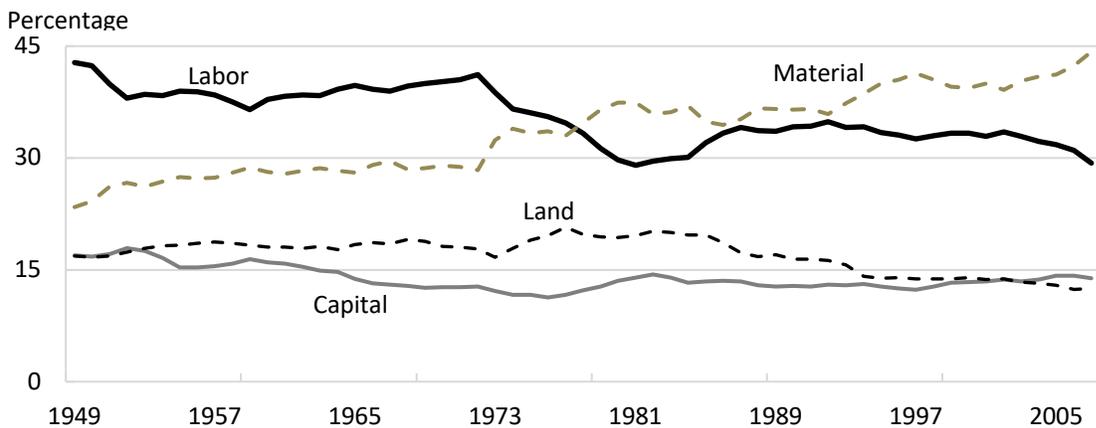
Notes: For number of farms, missing intercensal values were estimated by linear interpolation.

Figure 2: Labor use in U.S. agriculture, 1949–2012

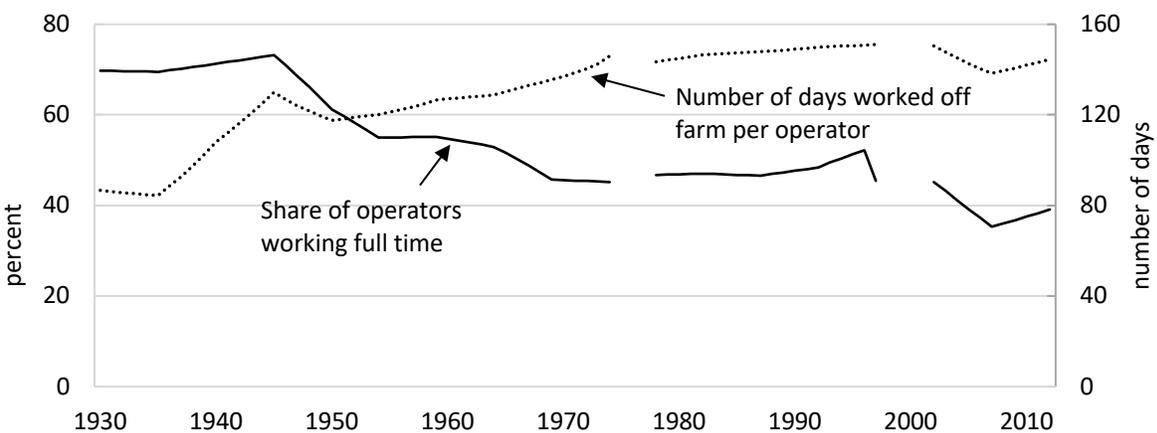
Panel a: Structure of labor



Panel b: Labor and other cost shares



Panel c: Part time farm operators



Sources: Pardey and Alston (2019, Figure 6).

Notes: In all census years up to 1997, the reported number of operators was set equal to the total number of farms. From 2002 onwards the census reported information on the total number of operators. If a farm had more than one operator, it was counted accordingly.

In all census years up to 1997, data are reported in terms of operators by days worked off-farm cohorts (e.g., 0 days, 1–49 days, 50–99 days etc.). For 2002, 2007, and 2012, data are reported in terms of cohorts of days worked off-farm by the principal operator. In 1974, data were collected only for individual or family operations (sole proprietorships) and partnerships. Thus, corporations and other types of organizations (e.g., cooperative, prison farms, grazing associations, and Indian reservations) were excluded (for more details, see U.S. Bureau of Census, 1977, Appendix A, page A4). In all other years, data on days worked off-farm were collected for all types of farms.

To calculate the average number of days worked off-farm per operator, we proceeded as follows. First, the total number of days worked off-farm in each cohort was estimated by multiplying the mid-point number of days worked off farm in each cohort (e.g., 25 days for 1-49 days, 75 days for 50-99 days etc. and 200 days for 200 and more) by the corresponding total number of operators. The total number of days worked off farms was obtained by summing the estimated number of days worked off-farm across cohorts. The number of days worked off-farm per operator is given by the total number of days worked off-farm divided by the total number of operators.

Data for intercensal years were estimated by linear interpolation.

The total number of operators working full time was estimated by subtracting the number of operators working off-farms from the total number of operators.

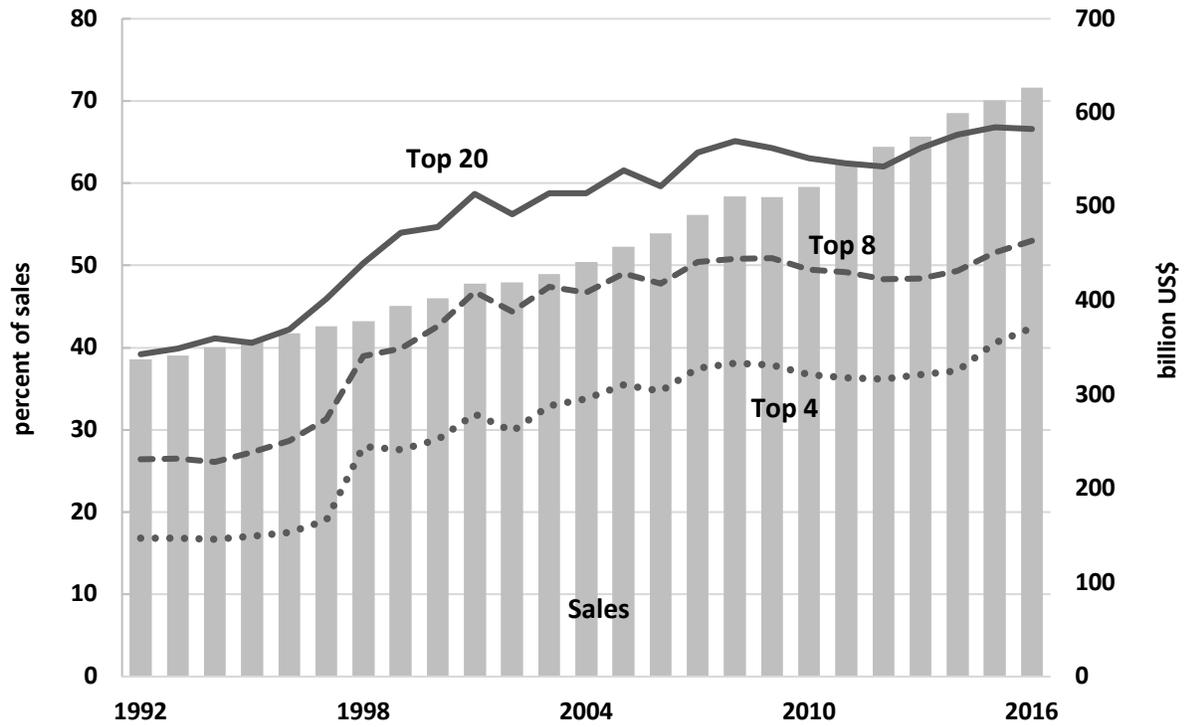
Table 1: Growth rates in U.S. agricultural input, output, and productivity 1910–2007

Period	Input	Output	Productivity Indexes		
			MFP	Labor	Land
			<i>percent per year</i>		
1910 – 1950	0.46	1.47	1.01	2.16	0.50
1950 – 1990	-0.21	1.76	1.97	4.07	1.92
1990 – 2007	0.31	1.39	1.08	1.90	2.04
1910 – 2007	0.16	1.58	1.42	2.90	1.35

Notes: All figures are annual averages of year-to-year changes.

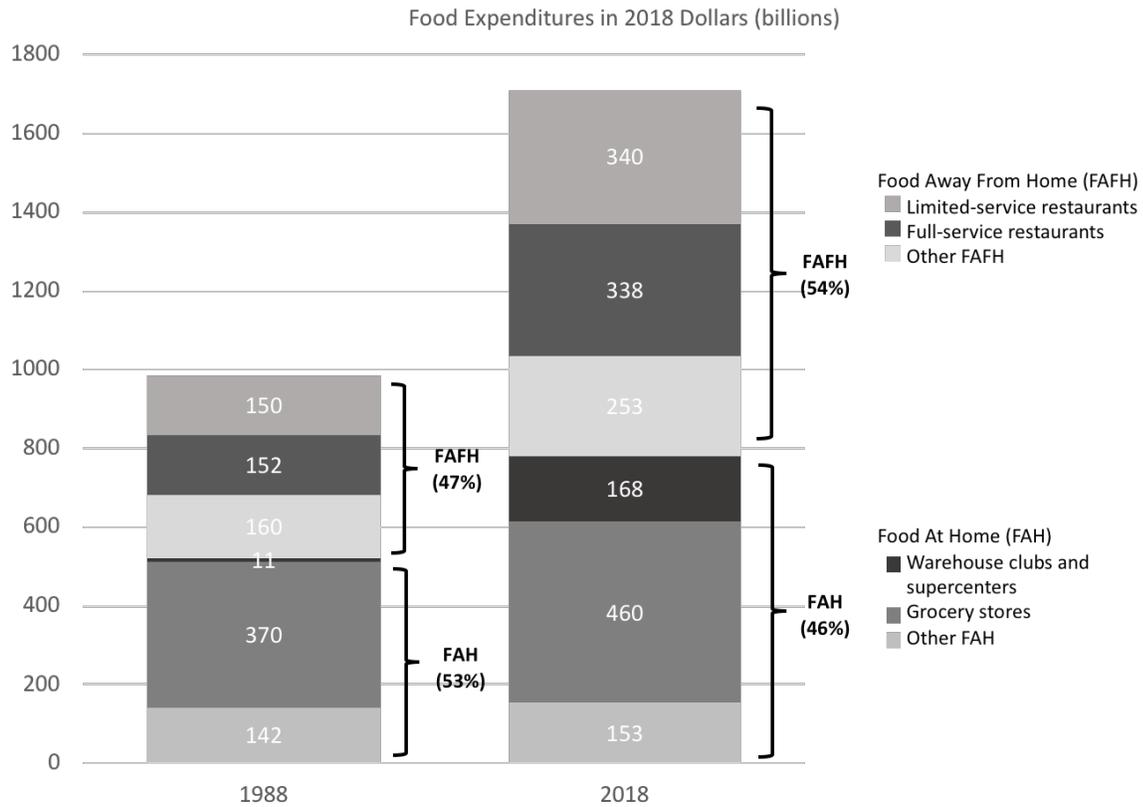
Source: Growth rates of productivity indexes were calculated by the authors using the InSTePP Production Accounts, version 5, augmented with data from USDA-ERS (1983). See Pardey and Alston (2019).

Figure 3: Top 4, 8, and 20 firms' shares of U.S. grocery store sales, 1992–2016



Sources: USDA, Economic Research Service, using data from U.S. Census Bureau, Monthly Retail Trade Survey, company annual reports, and industry sources. Sales based on North American Industry Classification System (NAICS).

Figure 4: Food expenditures at home and away from home, 1988 and 2018

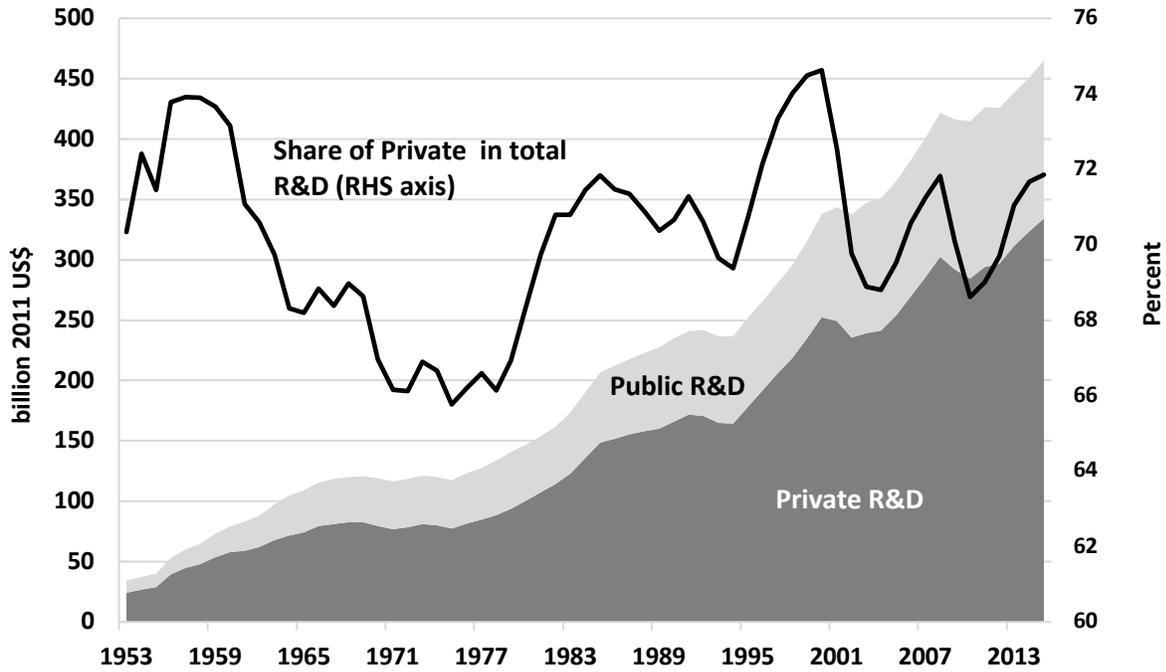


Sources: Calculated by USDA, Economic Research Service, from various sources. See details in Okrent et al. (2018).

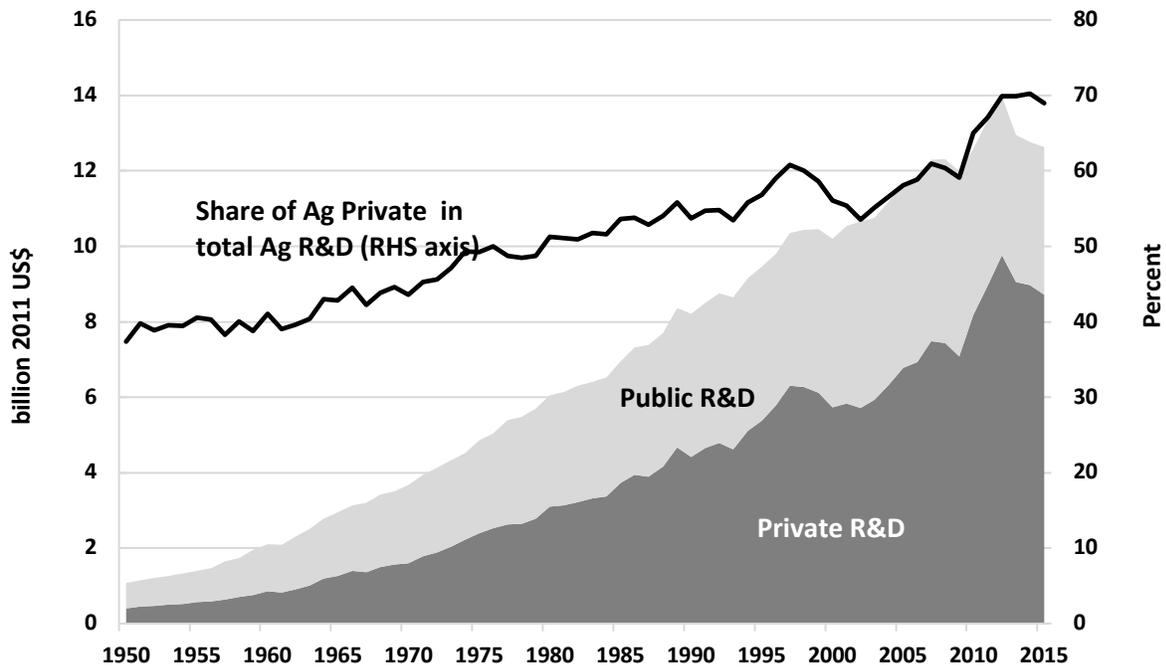
Notes: Nominal 1988 values inflated to 2018 values using CPI data from the U.S. Bureau of Labor Statistics: https://www.bls.gov/data/inflation_calculator.htm

Figure 5: GERD versus agGERD trends, 1950–2015

Panel a: GERD



Panel b: AgGERD

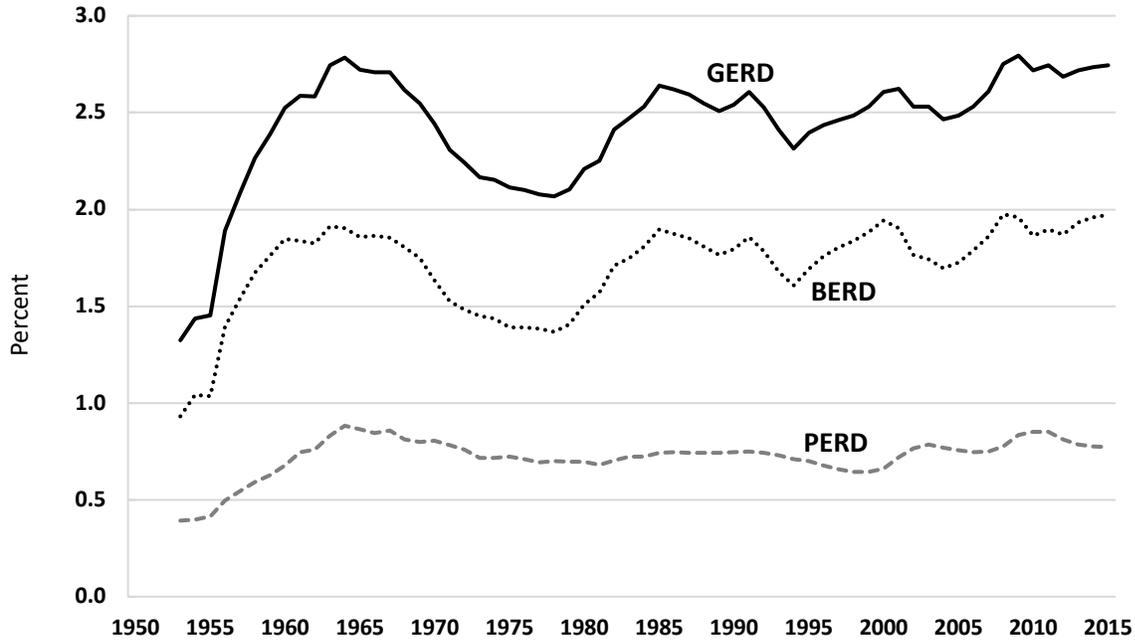


Sources: GERD data are from NSF (2019); AgGERD data are from InStePP International Innovation Accounts: Research and Development Spending (2019).

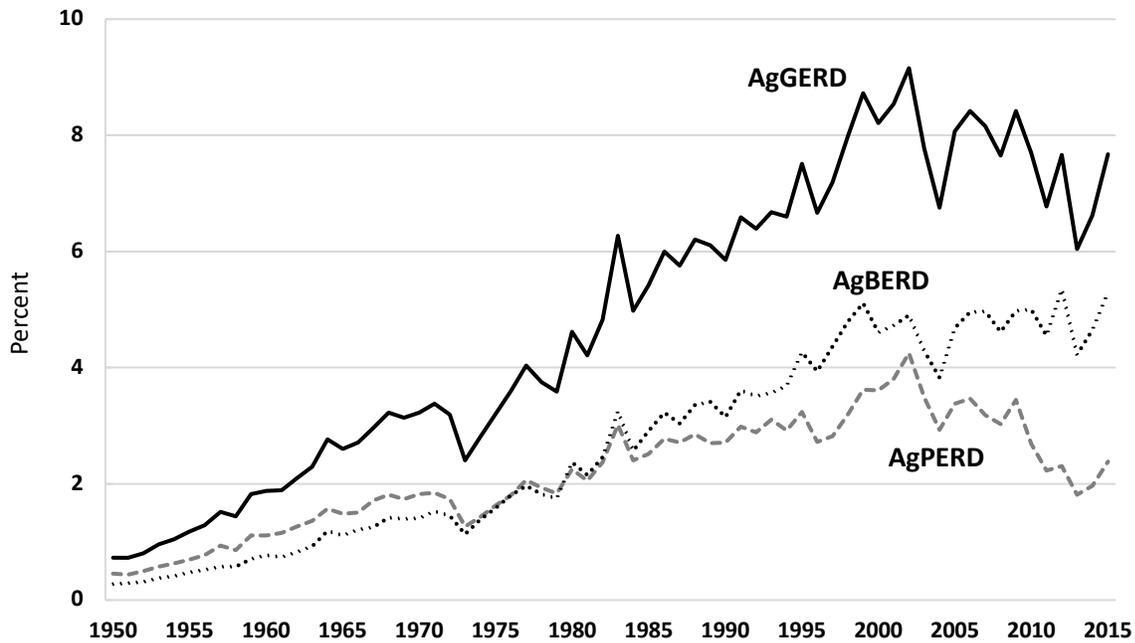
Notes: Public GERD include R&D expenditures from federal, non-federal, higher education and other non-profit organization.

Figure 6: The intensity of investment in GERD versus agGERD, 1950 – 2015

Panel a: GERD, BERD, and PERD as a share of GDP



Panel b: AgGERD, AgBERD, and AgPERD as a share of AgGDP



Sources: GERD (total), BERD (business) and PERD (Public) R&D spending data are from NSF (2018); AgGERD, AgBERD and AgPERD data are from InStePP International Innovation Accounts: Research and Development Spending (2019); GDP data are authors' compilation based on data from UN (2017), World Bank (2017) and Johnston and Williamson (2017); AgGDP data are authors' compilation based on data from UN (2017), and BEA (2017).

Table 2: Company sales and R&D intensity, 2014/15

	Sales	R&D Expenditures	R&D Intensity
	(millions dollars)		(percentage)
Manufacturing Industries	5,358,542	235,776	4.4
Chemicals	1,023,512	68,575	6.7
Pharmaceuticals and medicines	456,424	58,879	12.9
Others	567,088	9,640	1.7
Machinery	360,719	13,347	3.7
Automobiles, trailers and parts	795,662	19,096	2.4
Others	1,901,685	28,525	1.5
Non-manufacturing industries	3,691,358	118,123	3.2
Information	1,105,520	65,226	5.9
Finance and insurance	709,990	5,680	0.8
Others	1,453,882	10,177	0.7
All industries	9,049,901	352,946	3.9
Food and Agriculture	767,857	9,447	1.2
Food	554,237	3,821	0.7
Agriculture (machinery, chemicals and biology)	213,620	5,626	2.6

Sources: Manufacturing, non-manufacturing and all industries data are from National Science Board (2018, Table 4.10); Food and Agriculture data are from Lee et al. (2020) taken from InSTePP International Innovation Accounts: Research and Development Spending (2019)

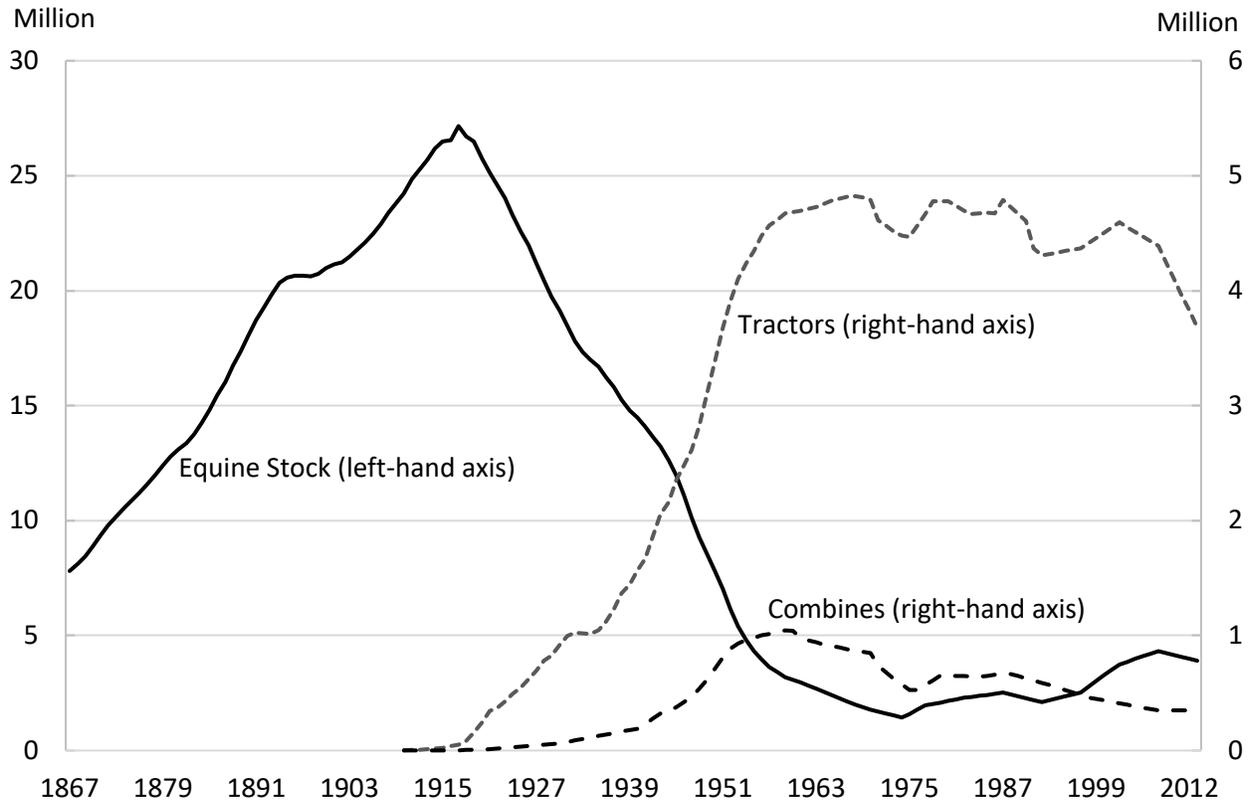
Notes: Sales from manufacturing and non-manufacturing industries includes domestic net sales of companies that perform or fund R&D, transfers to foreign subsidiaries, and export sales to foreign companies; excludes intracompany transfers and sales by foreign subsidiaries. R&D intensity from manufacturing and non-manufacturing industries represent domestic R&D paid for by the company and others and performed by the company divided by domestic net sales.

Table 3: Returns to agricultural R&D

	Units	Number of		Values			
		Studies	Estimates	10th percentile	Mean	Median	90% percentile
Reported Data							
United States							
Internal rate of return	% per year	80	986	12.46	63.03	31.85	101.2
Benefit-cost ratio	Ratio	21	312	0.52	19.33	11.98	45
Rest of world							
Internal rate of return	% per year	381	1,641	16	57.58	41.2	101
Benefit-cost ratio	Ratio	147	487	1.65	31.38	12.12	69.6
Imputed Data (delta = 5%)							
United States							
Benefit-cost ratio	Ratio	66	796	2.11	34.38	7.46	44.61
Modified internal rate of return	% per year	66	796	7.65	11.37	12.28	19.17
Rest of world							
Benefit-cost ratio	Ratio	345	1,330	2.56	25.65	8.95	57.27
Modified internal rate of return	% per year	345	1,330	8.35	13.6	12.96	20.17

Sources: Derived from InSTePP compilation used by Rao et al. (2019 a and b).

Figure 7: American agricultural mechanization, 1867–2012

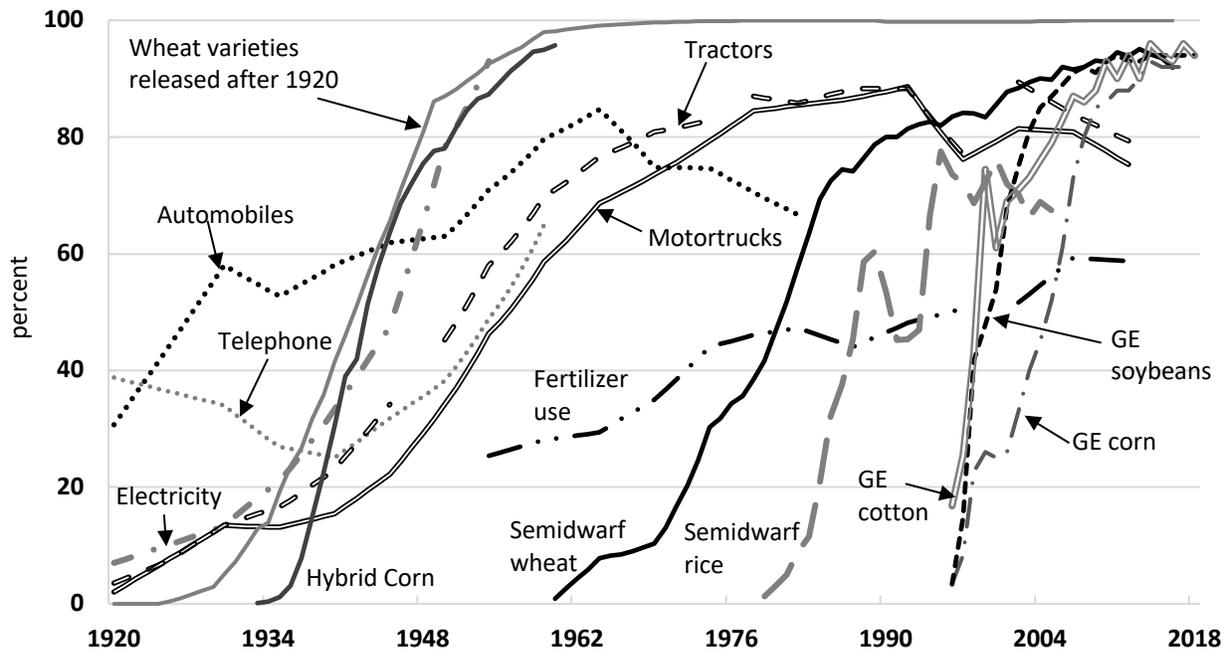


Sources: Pardey and Alston (2019, Figure 5)

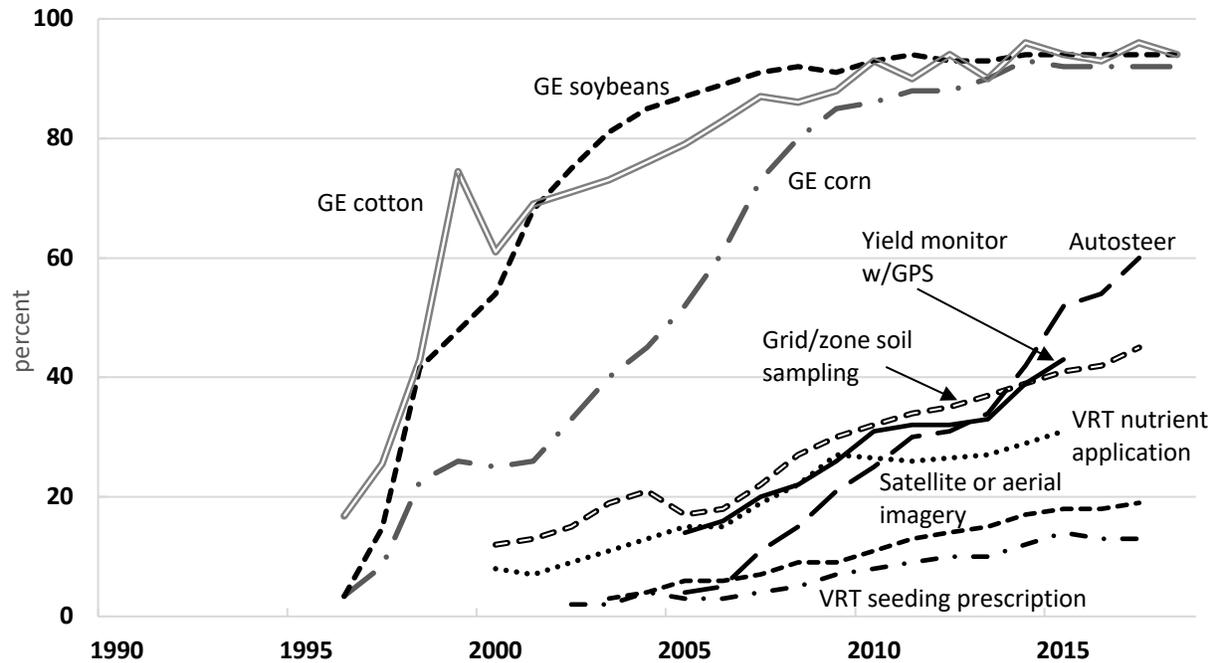
Notes: See Alston et al. (2010, page 29) for notes on equine stock.

Figure 8: Waves of technological adoption in U.S. agriculture

Panel a: Mechanical, chemical and conventional genetic improvement technologies



Panel b: Modern genetic transformation technologies and precision agriculture technologies



Sources: Mechanical, infrastructure and fertilizer data were developed by authors based on estimates from the U.S. Census of Agriculture (Bureau of Census and USDA-NASS, various years). Data for intercensus years were linearly interpolated. Data on hybrid corn are from Alston et al. (2010); GE soybeans, GE corn, and GE cotton data are from Alston et al. (2010) and USDA-NASS “June Agricultural Survey” available at www.ers.usda.gov/data-products/adoption-of-genetically-engineered-crops-in-the-us/. Semidwarf wheat and rice areas and area of wheat varieties released after 1920 are unpublished data from InSTePP. Estimated precision agriculture technologies cropland area shares are from Erickson et al. (2017).

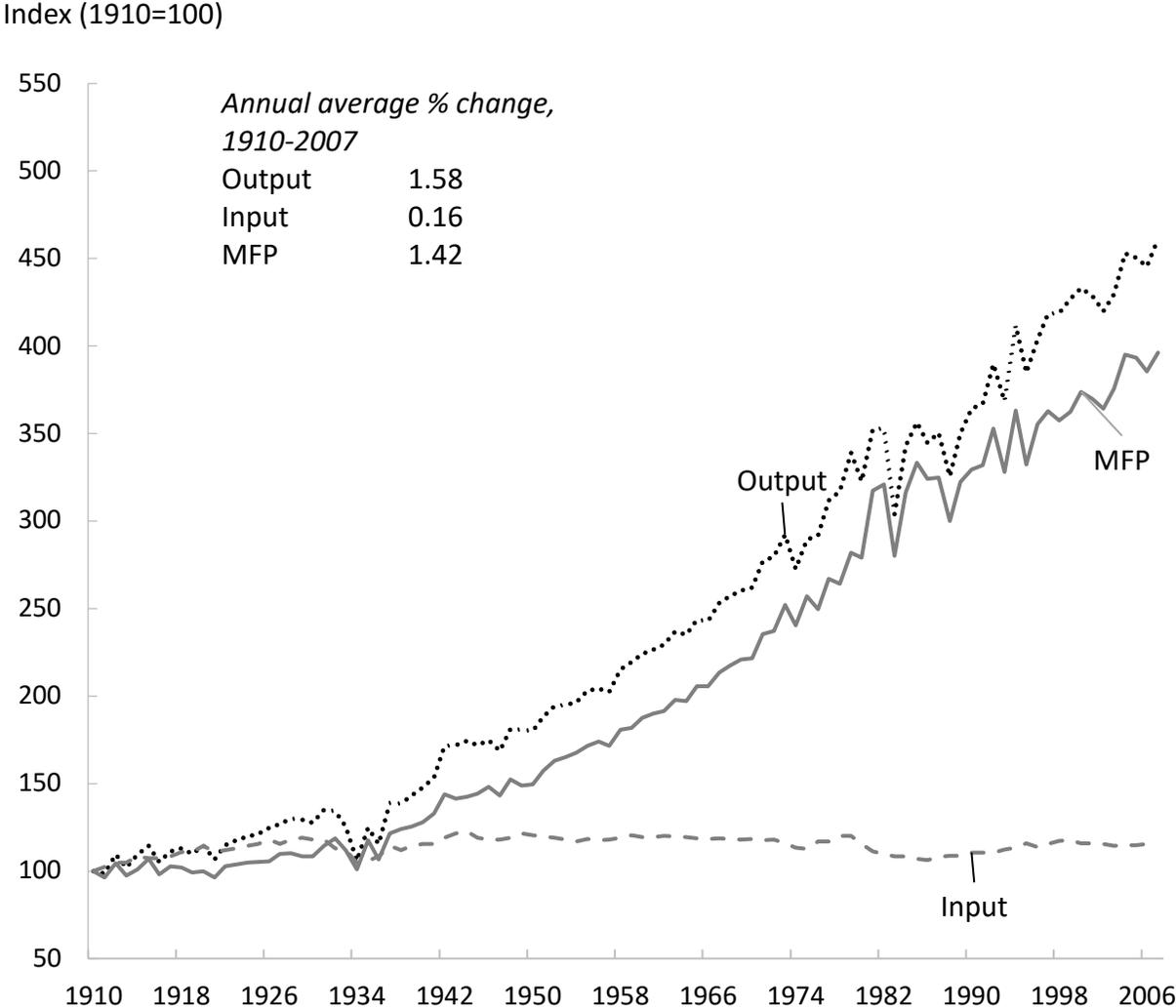
Notes: For automobiles, motor trucks, tractors, electricity, and telephone, the data represent the shares of farms using the designated technology. For hybrid corn, semidwarf rice, semidwarf wheat, GE soybeans and GE corn, the data represent the shares of area planted to the designated technology. For fertilizer, the data represent the share of cropland with fertilizer application. For precision agriculture technologies (autosteer, yield monitor with GPS, grid/zone soil sampling, satellite or areal imagery, VRT nutrient application, and VRT seeding prescription) data represent the share of the market area of various precision ag technologies used by farmers.

Tractors: From 1920 to 1945, tractors include wheel, crawler, and garden; from 1950 to 1969, tractors include wheel and crawler; from 1978 to 1997, tractors include wheel tractor only; from 2002 to 2012, tractors include wheel and crawler.

Automobiles: Details concerning the sudden and sustained drop in the number of automobiles in 1969 are reported in US Department of Commerce, Bureau of Census (1973, p. 11).

Fertilizer: From 1954 to 1997 fertilizer do not include lime whereas from 2002 to 2012 lime is included. Manure is excluded in all years. From 1978 to 2012, acres on which fertilizers were applied are reported for cropland only, pastureland, and total (i.e., cropland and pastureland). In 1959, 1964, 1969 and 1974 however, data were reported for pastureland and total acres fertilized. Thus, acres of cropland fertilized were estimated for those years by subtracting acres of pastureland fertilized from total acres fertilized. In 1954, only data on total acres fertilized was available. Thus, we estimated the area of cropland fertilized in 1954 by applying the share of cropland fertilized in 1959 to the 1954 cropland area.

Appendix Figure A-1: Quantity indexes of output, input, and MFP, U.S. agriculture, 1910–2007



Sources: Pardey and Alston (2019, Figure 1).

Appendix Table A-1: Food expenditures at home and away from home, 1988 and 2018

	1988 (Nominal)	1988 (Inflated)*	2018
FAH			
Grocery Stores	177.85	385.93	460.01
Warehouse clubs and supercenters	5.13	11.13	167.96
Other	68.37	148.35	152.96
Total	251.34	545.41	780.93
FAFH			
Full-service restaurants	73.13	158.70	337.82
Limited-service restaurants	72.23	156.73	340.17
Other	76.71	166.46	252.64
Total	222.07	481.90	930.62

*Calculated using CPI data from the U.S. Bureau of Labor Statistics: www.bls.gov/data/inflation_calculator.htm

Source: Calculated by USDA, Economic Research Service, from various sources. See Okrent et al. (2018) for details.

Appendix Table A-2: Regional growth in U.S. agricultural inputs, outputs, and productivity

	United States	Pacific	Mountain	N. Plains	S. Plains	Central	Southeast	Northeast
	<i>annual average growth rate</i>							
All Inputs	-0.11	0.82	0.45	0.16	-0.12	-0.27	-0.41	-0.84
Land	-0.10	0.59	0.41	-0.13	0.21	-0.51	0.29	-0.67
Labor	-1.74	-0.47	-0.90	-1.64	-1.80	-1.92	-2.15	-2.26
Capital	-0.07	0.08	0.02	0.25	0.13	-0.06	-0.74	-1.16
Materials	1.88	2.58	2.35	2.09	2.05	1.57	2.29	0.65
All Outputs	1.68	2.50	1.97	2.12	1.66	1.44	1.63	0.84
Livestock	1.39	2.32	2.03	1.44	2.16	0.44	2.34	0.78
Field Crops	1.65	0.76	1.00	2.60	0.95	2.29	0.27	0.60
Specialty Crops	2.21	2.82	2.60	1.20	1.55	1.62	2.36	0.60
Multi-Factor Productivity								
1949–2002	1.78	1.82	1.59	1.89	1.88	1.61	2.09	1.64
1949–1990	2.02	2.02	1.89	2.31	2.01	1.70	2.49	2.16
1990–2002	0.97	1.15	0.57	0.43	1.47	1.30	0.72	-0.14

Sources: InSTePP production accounts.

Notes: Average annual growth rates for inputs and multi-factor productivity span the period 1949–2002; for outputs they span the period 1949–2006.