Marketing, Other Intangibles, and Output Growth in 61 United States Industries

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Abstract

The System of National Accounts is considering including marketing as an intangible asset in the national accounts. This paper contributes to that discussion by developing macroeconomic measures of marketing investment and stocks for the U.S. We also construct and analyze measures of how marketing and other intangibles contribute to output growth in the 61 industries that comprise the U.S. private business sector.

We find that marketing contributes approximately as much to output growth as software and R&D do. From 1987 to 2020, our preferred estimates of the annual contribution to output growth are 0.15 percent for R&D, 0.19 percent for software, and 0.18 percent for marketing. Marketing's larger factor share offsets the more rapid growth of software. Marketing contributes even more to output if quality is adjusted to allow for better targeting associated with digital advertising. There is a close relationship between data flows, software, and digital marketing, and national accountants must allocate expenditures among these categories.

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In recent years, intangibles have played an increasing role in discussions of economic growth. An early study by Corrado, Hulten, and Sichel (2005) (CHS) was especially influential because it established the framework within which economists typically examine the importance of intangibles. Subsequent work has improved understanding of many of the intangible assets that CHS first suggested should be treated as capital. Corrado, Hulten, and Sichel (2009) concluded that incorporating intangibles substantially increased measures of capital deepening and somewhat raised labor productivity growth.

Empirical research has shown that marketing often increases purchases for several years and therefore qualifies as investment. An experiment demonstrated that random exposure to cable advertising increased household purchases of products for at least two years (Lodish et al. 1995). Recent research identified natural experiments where advertising influenced behavior for years (Bursztyn and Cantoni 2016; Bronnenberg et al. 2012).

Corrado and Hao (2014) prepared comprehensive estimates of marketing investment for the U.S. macroeconomy, combining estimates of purchased advertising, several additional types of purchased marketing services, and own-account marketing. Heys and Fotopoulou (2022) consider investment in design, organizational capital, firm-specific training, branding, and financial product innovation. Corrado, Haskel, Jona-Lasinio, and Iommi (2022) conclude that economic researchers should include the full complement of intangibles.

Statistical agencies have been slower to bring intangibles into official statistics. The System of National Accounts (SNA) now includes software, R&D, and entertainment originals as investment.² The SNA is currently considering including Marketing Assets as an additional type of intangible investment (IMF, 2022).³ As part of that discussion, the IMF requested comments on capitalization of Marketing Assets and, in response, we identify and discuss a number of relevant issues.

This paper develops macroeconomic measures of marketing assets broadly similar to Corrado and Hao (2014) and Heys and Fotopoulou (2022). We also construct and analyze measures of marketing investment for each of our 61 industries that jointly comprise the U.S. private business sector.

Our marketing measures are based on input-output (IO) tables and occupational information. First, we obtain data on each industry's purchases of advertising from the U.S. IO tables; purchased advertising is defined as the commodity associated with NAICS industry 5418 (advertising, public relations, and related services). Second, we measure each industry's purchases of other marketing services by its purchases from selected portions of NAICS industries 5182, 5415, 5416, and 5419, again from the IO tables. Third, we develop stocks of own-account marketing from occupational data collected by the Bureau of Labor Statistics (BLS) Occupational Employment and Wages Statistics (OEWS). We follow Corrado and Hao (2014) and Heys and Fotopoulou (2022) in converting measures of occupations into own-account stocks. By combining data from these three sources, we develop measures of Marketing Assets for the U.S. economy and for each industry. The rest of this paper refers to these joint measures of purchased advertising, purchased other marketing services, and own-account marketing as marketing.

² "Entertainment, literary, and artistic originals" is a SNA category. The Bureau of Economic Analysis refers to this category as "entertainment originals." We use that terminology throughout this paper.

³ This IMF document is available at <u>https://unstats.un.org/unsd/nationalaccount/SNAUpdate/GZTT.asp</u>. The draft cited here is from row G.9 of that site in the column marked Endorsed.

Our work contributes to ongoing discussion along two main lines. First, we develop prototype measures of the extent and impact of marketing investment in the United States. These measures cover the U.S. private economy and each industry. Our analysis shows that it is feasible to develop reasonable measures of marketing assets for the United States. The paper also considers several potential difficulties that statistical agencies will have to address as they measure marketing. Second, we use information on marketing and other intangibles to examine sources of growth in various industries.

1. Overview and Theoretical Framework

As in many studies of intangibles, we measure output by value added in constant dollars. Capital services are measured by quantities of assets weighted by their corresponding rental prices. Labor is measured in hours. We begin with a production function, expressed in growth rates:

$$v_{j,t} = (\alpha_k)_{j,t} k_{j,t} + (\alpha_l)_{j,t} l_{j,t} + tf p_{j,t}$$
(1)

where $v_{j,t}$ is the rate of growth of real value added in industry j in year t, $k_{j,t}$ is the rate of growth of capital service input, and $l_{j,t}$ is the rate of growth of labor input. $tfp_{j,t}$ is the corresponding growth of total factor productivity, typically calculated as a residual. $(\alpha_k)_{j,t}$ and $(\alpha_l)_{j,t}$ are the cost shares for capital and labor, each calculated as averages for years t and t - 1.⁴

The effect that any specific capital service, *i*, has on output growth follows the framework implied in expression (1). Specifically:

$$v_{i,j,t} = (\alpha_k)_{i,j,t} k_{i,j,t}$$
⁽²⁾

where $(\alpha_k)_{i,j,t}$ is the share of asset *i* in the value added of industry *j* in year *t*. $k_{i,j,t}$ is correspondingly the growth of service *i* in that same industry and year. The longer-term contribution of any capital service to output growth for the 33 years from 1987 to 2020, *LTCON*_{*i*,*j*}, is similarly:

$$LTCON_{i,j} = \left\{ \left[\prod_{t=1988}^{2020} (v_{i,j,t} + 1.00) \right]^{\left(\frac{1.0}{33.0}\right)} \right\} - 1$$
(3)

as calculated from the geometric mean of one plus the annual contributions.⁵

Our study considers 7 different types of intangibles: R&D, entertainment originals, own-account software, custom software, pre-packaged software, purchased marketing, and own-account marketing. Each intangible is studied in 61 industries over the 1987-200 period. To remove the effects of business cycles, we present results for the 1990-2000, 2000-2007, and 2007-2020 subperiods.⁶ We often measure

⁴ Our measures of capital stocks and services are calculated as Tornqvist indexes, using BLS Productivity Program methods. Labor composition indexes are also prepared with Tornqvist aggregation. The BLS obtains value added output data from the Bureau of Economic Analysis; these data are based on Fisher indexes.

⁵ Stiroh (2002, equation (7), page 1172) analyzes how industries contribute to national labor productivity growth in a value-added framework. In contrast, this paper concentrates on contributions to growth within industries.

⁶ United States recessions begin in July 1990, March 2001, December 2007, and February 2020. We select 1990 and 2007 as initial points in which the economy was still normal for a considerable part of the year in question, and 2000 as the last normal year prior to March 2001. We expand the 2007-2019 period to include 2020 because the COVID-19 recession is currently classified as lasting for only 2 months.

the relative importance of different forms of capital through their shares of capital services and their contributions to output.

Section II below describes how we develop measures of purchased and own-account marketing, which are the central ingredients of our study. Section 3 considers how these new measures of marketing investment affect United States macroeconomic growth. This section also compares the macroeconomic contribution of marketing with the impact of other sources of growth. Section IV uses detailed industry data to examine several specific hypotheses about marketing. Section V examines the relationships between marketing, other intangibles, and additional sources of growth within data for individual industries. Section VI concludes. The Appendices provide further information on how we calculate stocks of purchased and own-account marketing and measure their impact on the economy.

2. Stocks of Purchased and Own-Account Marketing

2.A Stocks of Purchased Advertising.

As Corrado, Hulten, and Sichel (2009, page 670) remark, "Expenditures for advertising are a large part of the investments in brand equity." Purchased advertising is the largest single element of marketing that we consider in this study. We measure how much advertising each industry acquires by its purchases of the commodity "advertising." ⁷ This includes advertising purchased from NAICS industry 5418, "Advertising, public relations, and related services," as well as advertising purchased from other industries such as print media, radio and TV, and the Internet. We work with the commodity version of purchased advertising because commodity data include all advertising that each industry purchases regardless of its source.

We use the IO tables to estimate industry purchases of advertising and other sources of purchased marketing services. For 1997 to 2020, we use the annual IO use tables developed by the Employment Projections program of the BLS.⁸ For 1982 to 1996, we use the Bureau of Economic Analysis (BEA) Historical Input-Output Tables, which offer less industry detail. We calculate the ratio of "advertising, public relations, and related services" to "miscellaneous professional, scientific, and technical services" in each industry in 1997 and use each industry-specific ratio to approximate advertising expenditures from 1982 to 1996. Our assumptions concerning depreciation imply that investments made prior to 1982 have fully depreciated by the time our analysis starts in 1987.

There has been some controversy about the usefulness of IO information to measure advertising, both at the individual industry level (Rogers and Tokle, 1995) and at the aggregate level (Silk and Berndt, 2020). To illustrate how the IO commodity data measure aggregate advertising, consider data for the year 2012. Silk and Berndt (2020, p. 47) suggest suggests that in 2012 expenditures on advertising and marketing were approximately \$90 billion, and expenditures on media and internet services were about an additional \$180 billion, implying total expenditures of approximately \$270 billion. The graph in Figure

Deleted:

⁷ Corrado and Hao (2014, pages 21-24) present several reasons why they believe that advertising has an effect at the industry level and is not simply dissipated by marketing expenditures of rivals. In practice, the official guidelines for national accounting, System of National Accounts 2008, exclude externalities from the national accounts (United Nations 2008, section 3.92). Accordingly, long-lived marketing would be capitalized even if it were dissipated by marketing expenditures of rivals. ⁸ The Employment Projections IO Tables are based on BEA Input Output Use tables.

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3 of their paper suggests that advertising expenditures reported to the IRS were perhaps a little closer to \$280 billion. The data used in this paper imply that at the commodity level advertising expenditures in the private economy were approximately \$305 billion in 2012.

To deflate advertising expenditures, for 1997 to 2020 we use the BEA price index for the gross output price of commodities in NAICS industry 5418 ("advertising, public relations, and related services") as our deflator for advertising.⁹ These BEA prices incorporate Producer Price Indexes (PPIs) for internet publishers, newspapers, radio, and TV, and other industries that produce advertising and also reflect certain other costs. For years prior to 1997 we prepare a new measure of the price of the commodity advertising that also reflects PPIs and certain costs.¹⁰ Appendix C briefly describes how 1982 to 1997 prices were prepared. We use the price index for advertising to measure output prices for all forms of marketing.

There is some question as to how well existing price deflators measure the output price of marketing. Mandel (2019) argues that the quality-adjusted price of advertising has declined rapidly in recent years because digital advertising is more effective than previous marketing methods. In particular, digital advertising can target potential customers more precisely than print or broadcast advertising can. Section 4.C considers Mandel's important hypothesis in more detail.

The question of what percentage of advertising expenditures represents investment is a central issue on which there is little conclusive evidence. We therefore adopt the same investment ratios used in other studies. The U.K. Office of National Statistics (ONS) has been a leader in the analysis of intangibles. Heys and Fotopoulou (2022), of the ONS, assume that 60 percent of purchased advertising services and 80 percent of purchased marketing services represent investment. We adopt these percentages in our baseline measures. Our alternative measure follows Corrado, Hulten, and Sichel (2005; 2009) and Corrado and Hao (2014) and assumes that 60 percent of purchased advertising services and 95 percent of purchased marketing services represents investment.

On the basis of Corrado and Hao (2014), Villalonga (2004), and Corrado, Hulten, and Sichel (2009) we select 45% as the central rate of depreciation. We choose 65% as an alternative depreciation rate. These rates imply service lives of 4 and 2 years, respectively.¹¹ We use these same rates of depreciation for all forms of marketing. Once we have determined nominal expenditures, the deflator, the proportion of expenditures that is investment, and depreciation, we measure stocks of each asset through standard perpetual inventory calculations.

2.B Purchases of Other Marketing Services

⁹ The BEA deflator for advertising is from the "Gross Output by Industry" files, under Underlying Detail, tab UGO 204-A, table line 144. We do not directly use the Producer Price Index data on the price of advertising as a commodity because that begins only in 2010.

¹⁰ Prior to 1997 the PPI provides much less information about the price of services.

¹¹ Corrado, Hulten, and Sichel (2005; 2009) initially assumed 60% annual depreciation. Corrado and Hao (2014, page 10) recommend lengthening the service life of investments in brand from three to about four years, which approximately corresponds to 45% depreciation. Consequently, we use 45% percent as our primary rate of depreciation and 65% as an alternative rate. Vitorino (2014, page 20) selects a 20 percent depreciation rate for advertising. However, Bagwell (2005, page 44), using similar information, suggests a greater rate of depreciation. We believe that, on balance, the overall evidence indicates that advertising depreciates more rapidly than 20 percent per year.

Firms purchase marketing services from industries other than advertising (NAICS 5418). Corrado and Hao (2014) include purchases from marketing consulting (NAICS 541613) and market research (NAICS 541961). We also include website design and hosting purchased from NAICS industries 5182 and 5415.¹² To the best of our knowledge, our study is the first work to include web design and hosting as marketing investment. For NAICS industries 5182, 5415, 5418, and 5419, we first calculate the proportion of output from that industry which represents marketing services; we estimate the presence of marketing services from data in the quinquennial Economic Census and make adjustments for underand mis-reporting. Between Census years, we use the Services Annual Survey (SAS) to interpolate and extrapolate. Such data provide reasonable information on the overall purchases of marketing services, but as Appendix C explains, it is a challenge to assign these amounts to specific purchasing industries. IO tables do not provide sufficient detail to track purchases of very detailed goods. We are therefore forced to allocate purchased marketing services to the industries that use them through data for the next higher level IO sector. Since we are considering purchases of marketing from additional industries, our estimates of purchased marketing are generally larger than those in Corrado and Hao. Appendix A shows how much each type of marketing service contributes to investment in marketing in each year.

Nakamura, Samuels, and Soloveichik (2017) suggest that each of these estimates of marketing should be priced at the price of overall advertising. They find that advertising viewership costs are more closely associated with each other than with measures of content creation. Their Figure 9 shows that the viewership cost of digital media is correlated with viewership costs in other media and that the correlation increased in the 2010s as digital media became more prevalent. For this reason, we use the BEA advertising price index, instead of a cloud price deflator or other content creation costs, to price all portions of marketing purchases. Section 4.C emphasizes that the topic of adjusting marketing output prices for unmeasured quality change requires further consideration.

Stocks of Own-Account Marketing.

The literature typically draws a sharp distinction between purchased marketing and own-account marketing expenditures. While it is useful to know the approximate magnitudes of each of these two types of expenditures, we caution that these expenditures are inevitably closely interrelated. Internal marketing personnel are highly involved in external marketing campaigns. From this perspective, estimates of total marketing are more reliable than separate estimates of purchased and own-account resources. In the final analysis, the total marketing effort is what really counts.¹³

Own-account marketing expenditures are generally measured based on occupational employment.¹⁴ We use the presence of certain occupations in each industry to measure the quantity of

¹² Section 2.A above used the commodity version of the IO tables to measure advertising services because the overall production of advertising is far greater than output in the advertising industry (5418). In other industries that supply marketing services, commodity output is much closer to industry output, and we use standard IO tables to measure output there.

¹³ It is also difficult to distinguish between own-account production and output that is sold to customers.
¹⁴ Nakamura, Samuels, and Soloveichik (2017) also argue that measures of purchased advertising should be supplemented by data on own-account marketing. For example, a television network might use unsold advertising slots to promote an upcoming show. Following Soloveichik (2013) and Nakamura, Samuels, and Soloveichik (2017), we also include radio, TV, and other media expenditures that advertise their own product as part of own-account expenditures.

own-account expenditures. We do not distinguish between own-account advertising and marketing but instead define an overall own-account category which we call own-account marketing.

We obtain each industry's occupational employment for 2002 to 2020 from the OEWS.¹⁵ The OEWS is collected over a rotating three-year cycle, in which a third of the sample is collected each year. We prepare data on relevant employment over a three-year period and only later assign the observed patterns to individual years. Appendix B lists the occupations that we assigned to marketing and describes how occupational information on occupations is converted into own-account marketing stocks. Before 2002, we extrapolate own-account marketing in each industry with data on aggregate occupational employment from the OEWS, overall output for each industry, and purchased marketing services.

Estimates of the time that each occupation spends on long-term investment would ideally depend on careful time studies. Unfortunately, this type of conclusive evidence does not appear to exist. Our baseline measure follows Heys and Fotopoulou (2022) and assumes that 30 percent of own-account expenditures are investment. Our alternative measure follows Corrado and Hao (2014) and assumes that 60 percent of own-account expenditures are investment. We further assume that the same rates of depreciation selected for purchased advertising, 45 percent as a base rate and 65 percent as an alternative, also hold for own-account marketing. Once our assumptions about expenditures, deflators, the investment portion, and depreciation are set, we construct perpetual inventory stocks of ownaccount advertising for each industry.

Existing work on own-account marketing (Corrado and Hao, 2014; Heys and Fotopoulou, 2022) uses a relatively narrow list of relevant occupations. We think it is possible that a wider range of occupations, especially in sales, may also contribute to the value of marketing assets. Many sales workers develop continuing relationships with their customers that eventually lead to greater long-term sales. We do not know of any empirical studies that document how much time sales workers spend investing in longer term relationships. However, because sales workers are such a large group, even a small proportion of their time could substantially increase measures of marketing investment. We think that this is a potentially important topic that should be carefully considered before marketing is included in the Accounts.

The IMF discussion of marketing assets frequently refers to the value of trademarks and logos. Dosi, Gavrilova, Silva, and Soares (2022) recently began work on the value of trademarks, determining how much a new trademark, in itself, adds to the value of a firm. However, we believe that the value of a trademark more fundamentally reflects a firm's underlying assets, including its marketing, R&D, and organizational capabilities. We think that future work that integrates the value of a trademark with these underlying capabilities will strengthen the usefulness of measures of marketing assets.

Adapting Existing Data to Include Marketing Assets as an Additional Intangible.

The BLS Productivity Database contains many data elements that are useful in measuring the impact of marketing. This includes gross output and value added, in both current and constant dollars, and measures of K (capital), L (labor), E (energy), M (materials), and S (purchased services). The data on

¹⁵ The OEWS began collecting some initial data as early as the 1990's. However, the initial data used different industry codes and sometimes different occupations codes. Therefore, it cannot always be consistently linked.

E, M, and S together provide measures of intermediate input. This subsection describes how we measure output and input from the BLS data and how we modify existing BLS data to allow for purchased and own-account marketing as additional intangibles.

We begin with the existing measures of gross output and purchased services and the new measures of marketing investment described above. For each of these series we have prices in current dollars and chain-type quantity indexes. In addition, we also have measures of value added developed by the BEA. The BEA prepares value added by double deflation, deflating both gross output and intermediate inputs (Moyer, Planting, Fahim-Nader, and Lum, 2004). The investment portion of marketing must be removed from each industry's purchased services and transferred to capital investment. By construction, a smaller quantity of purchased services requires that intermediate prices be recalculated for each industry. This new price is then used to compute adjusted quantities of intermediate inputs. We use the double deflation method described in Moyer et al. (2004) to remove marketing from purchased services, recalculate intermediates, and recalculate value added output by removing our new measures of intermediate input from gross output.

Gross output and value added both increase when portions of marketing are treated as investment. It is necessary to decide where to allocate the extra value-added income. Previous work on intangibles in the U.S. Accounts, such as studies of R&D and software,¹⁶ has assumed that the added income from capitalization all goes to capital. To be consistent with those studies, we also assume that the added income from capitalization goes to capital, and that there is no effect on employee compensation.¹⁷

The decision to assign all additional income from capitalization of intangibles to capital has important implications. Koh, Santaeulalia, and Zheng (2020) show that the decline in the labor share observed in the U.S. occurs solely because all the additional income from intangibles is assigned to capital. They argue that such an allocation is "extreme", and that a portion of the new value added created should instead be assigned to labor; they recommend detailed micro analysis to determine where extra output should be assigned. The Koh et al. study is insightful and thought provoking. If further work supports their interpretation, some of the value created by capitalization, for marketing and other intangibles, may eventually be assigned to labor. If such adjustments turn out to be appropriate, existing estimates of property income and the associated rental prices are probably too high.

Once we have constructed stocks of purchased and own-account marketing and estimated the increase in property income associated with these investments, we are ready to value these stocks. To determine rental prices, we treat purchased and own-account marketing just like any other capital asset. As is standard procedure, we begin with data on property income in each industry and year, determine an internal rate of return for each industry/year, and then calculate rental prices that reflect asset price changes, rates of depreciation, and tax parameters.

¹⁶ For example, Fraumeni and Okubo (2005) describe how the additional income that arises from capitalization of R&D is assigned to capital.

¹⁷ Our estimates assume that the increased income from marketing investment goes to an increase in the gross operating surplus. The increased surplus goes to corporate profits or to the capital or labor income of proprietors. We use ratios from the existing National Accounts to assign the new surplus to each of these three categories in each industry. Almost all the additional surplus is assigned to capital, since proprietor labor income is typically a small share of total gross operating surplus.

3. The Macroeconomic Impact of Marketing Assets

This section analyzes how the new measures of marketing assets affect macroeconomic growth in the private sector. First, we measure the effect that purchased and own-account marketing, other intangibles, other inputs, and TFP make to output growth. Second, we look at the flow of capital services to goods and services industries.

3.A The Effect of Marketing on Output Growth

Panel A of Figure 1 shows how intangibles, which now include the new purchased and own-account marketing assets, have consistently grown more rapidly than tangibles. Panel B shows that intangibles, which originally were less influential than information and communication technology capital or other assets, are now more important. This occurred because other forms of capital made less of a contribution, not because the contribution of intangibles increased.

<Figure 1, Panels A and B go here>

<Table 1 goes here>

Table 1 shows our central results using the basic assumptions summarized in Table C-1. Of the presently recognized intangibles R&D and software have the greatest impact on macroeconomic growth. Over the entire 1987 to 2020 period, R&D contributed 0.15 percent a year to output growth and the three types of software together added 0.19 percent a year. The two types of marketing contributed 0.18 percent a year to output growth. This evidence makes the important point that marketing contributes about as much to output growth as R&D or software do. Appendix C explains how we obtain these central estimates.

Background information helps to clarify the effects of both R&D and software. Table 1 includes only the direct effects of R&D—the immediate returns to firms that initially conduct research. Evidence shows that R&D spillovers account for more than half of the total returns to R&D and that the spillover portion of total returns has increased in recent years (Bloom, Schankerman, and Van Reenen, 2013; Lucking, Bloom, and Van Reenen, 2019; Sveikauskas, 2007). These well-documented spillovers show that social returns to R&D are much greater than the private returns shown in Table 1. It has so far been difficult to assign R&D spillovers to specific industries. However, Martin et al. (2022) recently developed a framework that may be able to assign R&D spillovers to specific industries.

On software, Table 1 shows that pre-packaged software affects growth most strongly. That might seem to contradict Bessen (2020; 2022), who has shown that large firms often dominate their industries by developing highly productive proprietary computer systems; these powerful proprietary systems might seem to be own-account software. However, the BEA classifies software-as-a-service (SaaS) as pre-packaged software, and this category has grown rapidly, so U.S. data show that pre-packaged software contributes strongly to growth.¹⁸

¹⁸ BEA specialists mention that some countries prefer to think of software as a purchased service. However, BEA plans to continue classifying SAAS as an investment in pre-packaged software. They also comment that the increased reliance on cloud computing has altered the way that firms pay for software and that counting SAAS as an investment in software helps to describe the new payment patterns.

Despite the importance of marketing, inclusion of marketing as an additional intangible does not greatly increase measured economic growth. Table 2 shows that capitalization of purchased and ownaccount marketing increases output growth by less than 0.1 percent a year. This growth increase is similar to the increase associated with capitalization of R&D (Ribarsky 2022).

Table 1 reports our best judgment about the role of marketing, and its importance relative to R&D and software, in the United States economy. However, it is also useful to present supplementary information under a variety of different assumptions. Tables C-1 and C-2 of Appendix C report corresponding results under several alternative assumptions. These sensitivity tests show that changes in the percentage of marketing expenditures that is investment have a considerable impact on the implied contribution of marketing. Since advertising expenditures are substantial, changes in this investment proportion are particularly influential. In contrast, changes in the rate of depreciation, within the values that the literature suggests, have little effect on the implied role of marketing. These results suggest that, as further work on marketing proceeds, researchers could usefully concentrate on measuring the proportion of expenditures that represents long-term investment. Time diaries of the activities of marketing workers might be able to provide more conclusive evidence.¹⁹

<Table 2 goes here.>

Figure 2 shows the relative importance of purchased and own-account marketing over time, as measured by the flow of capital services for marketing as a percentage of nominal value added. Figure A-2 in Appendix A shows that investment in many categories of marketing, as a percentage of GDP, has turned upwards in recent years.

Figure 2 shows that purchased marketing accounts for a considerably larger proportion of total marketing than own-account marketing does. In part, these patterns arise because our baseline estimates assume that only 30 percent of own-account marketing expenditures represent investment. If we instead assume, as in our alternative set of assumptions, that 60 percent of own-account expenditures is investment, then the lower orange line in Figure 2 would be twice as high and much closer to the blue line (purchased marketing).

<Figure 2 goes about here>

Table 3 shows the rate of growth of investment for various types of capital assets in different time periods. Investment growth slowed over time for most asset categories. Prepackaged software grew rapidly in each time period. It might seem surprising that prepackaged software has grown so quickly (Table 3), whereas purchased marketing contributed more to output growth (Table 1). Table 4, which shows the annual rates of growth, factor shares, and contributions to output growth for each of these two types of assets, explains these different patterns. Purchased marketing's larger factor share offsets the more rapid growth of pre-packaged software and allows purchased marketing to contribute more to growth.

<Tables 3 and 4 go here>

¹⁹ If we adopt 65 percent depreciation instead of 45 percent, marketing stocks are 41% lower than reported in our baseline estimates. With 65 percent depreciation, the 1987-2020 contributions of purchased and own-account marketing fall to 0.13 and 0.04 percent, compared to 0.14 and 0.06 percent in the baseline case with 45% depreciation. Investment proportions are held at their baseline values in these calculations.

3.B The Flow of Capital Services in Goods and Services

The goods sector consists of agriculture, mining, utilities, construction, and manufacturing. Services are the rest of the private economy. The private economy represents about three-quarters of GDP, and excludes general government, government enterprises, most nonprofits, and households.

Figure 3-A shows the flow of capital services to the goods sector (in blue) and to the services sector (in orange) over time. We calculate the annual flow of capital services in each industry and add them for all goods and for all services. Capital services were slightly less in goods from 1987 to 1990. However, by 2020 only 30 percent of capital services occurred in goods.

<Figures 3-A and 3-B go about here>

Figure 3-B shows that the intangible share of capital services was originally greater in goods than in services. It was not until about 2001 that the intangible share in services surpassed the share in goods, so that the expanding role of services started to increase the overall amount of intangibles. R&D represents a large portion of the intangibles in manufacturing. Figure 3-B shows that, at the end of the technology boom in 2000, the share of capital payments spent on intangibles began to decline for goods but continued to increase in services.

4. Influences on Marketing Assets

4.A The Impact of Consumer or Business Markets on the Level and Growth of Marketing

IO tables provide information on how much output each industry delivers to intermediate products, consumption, and investment. Deliveries to consumption tell us how important the consumer market is, and deliveries to intermediate products and investment show how important the business (B2B) market is in each of our 61 industries.²⁰ We use data from BEA's detailed 2012 IO table.

We seek to understand how marketing practices differ between consumer and B2B industries. In the cross-section we measure the importance of marketing in each industry by the flow of capital services to marketing as a proportion of that industry's value added. We examine the growth of marketing investment and marketing's influence on labor productivity growth.

We find no evidence that the intensity or rate of growth of marketing activities differs between industries oriented to consumer or business markets. The shares of purchased and own-account marketing similarly do not differ between consumer or business industries. Defining marketing intensity in industry *i* as the flow of capital services to marketing divided by the value added observed in that industry, we estimate the following regression:

Marketing Intensity_i = $\alpha + \beta Consumer Share_i + \gamma Business Share_i$ (4)

These regressions show no sign that the consumer or business orientation characteristic of an industry affects observed marketing intensity. The type of customer similarly does not affect the intensity of purchased and own-account marketing or our measures of time-series effects.

²⁰ Since some output is delivered to government, the shares of consumer and business are not perfectly collinear.

We had expected to find more marketing in consumer-oriented industries. The World Advertising Research Center (WARC) occasionally reports the U.S. industries in which advertising expenditures are the greatest. Their 2022 report lists these industries, in order, as retail, media and publications, business and industrial, financial services, technology and electronics, pharma and healthcare, technology and utilities, automotive, and amusement and leisure.²¹ That WARC list appears to be heavily weighted towards consumer goods.

A possible explanation is that national income procedures assign a firm's advertising to each of its establishments, which are often classified in different industries. U.S. NIPA data report heavy advertising expenditures in wholesale trade, financial functions, and management of companies. Such procedures probably assign advertising to economic functions well. These national income conventions may explain why we cannot establish a relationship between the customer type and observed marketing.

4.B The Effect of the Presence of ICT on the Future Growth of Marketing

We hypothesized that the presence of ICT would lead to a more rapid growth of investment in marketing, and that the link between ICT and the subsequent growth of marketing became stronger in more recent years, as digital marketing became more prevalent. We measured the presence of ICT in each industry in any year as the share of ICT assets, including software, in current value added.²²

We did not find any clear impact of ICT on marketing in our U.S. industry data. With more detailed data, such as information on many firms in the same industry, or data for the same sector in different countries (Chen, Niebel, and Saam 2016), the effects of ICT might be clearer.

4.C The Effect of Advertising if Digital Advertising is Substantially More Effective

Mandel (2019) emphasizes that digital advertising, viewed on personal computers or mobile phones, is inherently more effective than print media advertising. Digital advertisers know more about the interests and concerns of potential customers and can target or customize ads towards likely buyers. This is a quality change, in the same sense that cars with more horsepower and houses with more square footage are of higher quality and represent more output. Consistent with that hypothesis, advertisers are shifting to digital advertising very rapidly. The Services Annual Survey shows that the digital share of the advertising market increased from 0.9 percent in 2002 to 38.2 percent in 2015 and 58.3 percent in 2020.²³ Growth of this magnitude suggests that digital advertising offers important advantages to advertisers, most notably the targeting of specific consumers. As Mandel states (2019, page 4) "In the economic sense, digital advertising is more productive than print advertising." Also (page 12) "The simplest explanation for all these observations is that advertisers are finding that they can get a bigger bang for their buck by spending their money online rather than in print."

²¹ https://www.marketingcharts.com/advertising-trends/spending-and-spenders-227936/7?et blog.

The website Zippia has a somewhat similar list of the industries, but their list is restricted to digital advertising. ²² We consider investment in marketing from 1990, 2000, and 2007, the beginning of each our three subperiods. We also consider growth since the expansion of digital marketing in 2012. For each of these four time periods, we cannot establish any effect of ICT intensity on the future growth of marketing investment.

²³ This digital share refers the industry 5418 only. Purchases of the commodity advertising from other industries, purchases of other marketing services, or own-account marketing may have different digital shares.

Mandel (2019) suggests that digital advertising is five/thirds as effective as print advertising. That is, every dollar spent on digital advertising brings a bonus of .67 cents of extra output due to the greater effectiveness of digital ads. With a 60 percent increase in the digital share over the years, that would imply 60 * 2/3 or a 40 percent increase in the effective amount of advertising just from the switch to the Internet. That seems to be a remarkable amount of additional advertising output, even allowing for the overwhelming success of firms like Google, Facebook, and TikTok. Perhaps these magnitudes arise because Mandel was comparing digital advertising with print media, which is a particularly stagnant advertising category.

Even if the quality differences are not so large as Mandel suggests, it is plausible that typical deflators do not adequately adjust for quality improvements in advertising. To examine these possibilities, Table 5 considers effectiveness bonuses of .10 percent or .20 percent for every 1 percent increase in the digital share. In these cases, the long-term 60 percent increase in the digital share would be associated with a 6 or 12 percent gain in the real amount of advertising. These increases in output are strongest since 2015 when the digital share of advertising increased from 38 to 58 percent.

The first column of Table 5 shows the Internet share of the advertising market from 2002 to 2020, from the SAS. The second and third columns report the extra bonus of advertising output if each additional dollar spent on digital advertising brings a bonus of 10 or 20 cents of additional output.

<Table 5 goes about here.>

Table 5 shows that if digital advertising brings even modest productivity advantages, advertising output increases 6 to 12 percent by 2020 solely because of the shift to the Internet. Equivalently, the price per unit of advertising output would decline by 6 or 12 percent by 2020 just because of the output expansion due to digital advertising. In 2020, the present official estimate of advertising output price, 103.696 are 10% lower, to 98.01 (103.696/1.058) or 92.92 (103.696/1.116).

These calculations show that the implied effect on the price of advertising is substantial even if digital ads are only slightly more effective than other forms of advertising. We do not at present know exactly how much more effective digital ads are. However, this exercise has shown that, even if digital ads are only slightly more effective, that is sufficient to lower the implied price of advertising substantially. Lower prices would in turn show that advertising has increased output growth more rapidly. We think further evaluations of the productivity advantage of digital advertising and other productivity improvements associated with information and communication technology, such as better targeting of direct mail, ought to be explored to understand how marketing has affected economic growth.²⁴

²⁴ If digital marketing is an incompletely-measured quality improvement, then marketing prices would have increased less than currently available price indices indicate, so that the amount of output and the productivity gains associated with marketing are greater than existing prices indicate.

Another crucial question about advertising or marketing is whether expenditures by one firm cancel expenditures by a rival firm, so that the net effect of expenditures is reduced. In our judgment, this topic cannot be understood solely from industry data. As in the analysis of R&D (Bloom, Schankerman, and Van Reenen (2013)), this issue requires both firm and industry data, so that it is possible to evaluate the presence of positive or negative spillovers.

5. Distribution of Assets Across Industries and Their Effect on Growth

5.A Stocks of Asset Types in Different Industries

We now consider the importance of asset types at industry level. The sectors considered are manufacturing, other goods, trade, finance, and other services. Table 6 shows the importance of each type of capital as a percentage of total capital stocks in each of these five sectors. Panel A of Table 6 reports tangible assets and Panel B shows intangible assets for 2012. Equipment accounted for 31 percent of capital stocks in manufacturing. Similarly, inventories were 24 percent of all stocks and entertainment originals are 7 percent of total stocks in other services. Purchased marketing is most important in trade and other services, accounting for 3 percent of total stocks.

<Table 6 panels A and B go about here>

5.B Correlations between Sources of Growth

We have developed, for 1987 to 2020, measures of how much each source of growth, every intangible, each tangible form of capital, the composition of labor, labor input, and TFP, contributes to output growth in each of the 61 industries. These measures are all expressed as average annual contributions.

Table 7 shows how the average annual contributions to growth are correlated across industries. We highlight correlations of special interest in yellow. The high correlations between various forms of software show that industries which use one form of software tend to use others as well. Similarly, the two forms of marketing are highly correlated. In addition, both forms of marketing are also highly correlated with the impact of software. That is not surprising for own-account marketing, which is strongly driven by occupations that are closely associated with data and computer personnel. However, purchased marketing, which consists largely of advertising, is also closely linked to the presence of software.

Other connections are less strong. There is some support for the well-established connection between R&D and the presence of marketing, especially for own-account marketing. We had thought that potential drivers of economic growth such as ICT, improvements in the composition of labor, or TFP might be associated with a more rapid growth of intangibles. There is some evidence that ICT (which here excludes software) may have some effect on the growth of intangibles, but measures of labor composition and TFP appear to have little connection to intangibles growth. Measures of TFP may be subject to measurement error, partially because they are based on value added rather than gross output.²⁵

²⁵ Berndt and Wood (1975) suggest that measures of productivity growth based on gross output are preferable to estimates obtained from value added. De Loecker and Scott (2016) and Gandhi, Navarro, and Rivers (2020) show the usefulness of gross output methods and describe how they can be implemented. We initially thought that connections between TFP and the presence of intangibles might provide some clues about which intangibles have spillover effects. Indeed, we find a mild positive connection between R&D and TFP. However, industries differ in so many important ways that it is preferable to search for spillovers within data for firms in a single industry.

<Table 7 goes about here.>

The economics literature, such as Bessen, Denk, Kim, and Righi (2020), discusses how intangibles have altered the nature of production. This literature typically concentrates on firm data because intangibles frequently affect firms in the same industry differently. However, Table 7 shows that differences between industries can also provide some useful information.

5.C Industry Concentration of Intangibles

This subsection provides evidence on the extent to which the use of each intangible is concentrated in a few leading industries. Table 8 reports industry concentration for each intangible in 1987, 2002, and 2020, as measured by the percentage of the total stock of that intangible observed in the top 10 of our 61 industries.

<Table 8 goes about here>

Entertainment originals always have concentration of 100%, since only 5 industries hold this asset. Concentration of R&D declines modestly. However, concentration of software and marketing increases markedly, especially after 2002. Bessen (2022) describes how software has become more proprietary since 2000, as firms develop their own computer systems. Much of the concentration of software has occurred within industries, as firms with effective digital systems displace their competitors. However, Table 8 shows that, since 2002, software also become more concentrated across industries; each type of software has also become more concentrated. Both forms of marketing have similarly become more concentrated since 2002. The same internal data systems that are known to make software more effective for leading firms are likely to make the same firms' marketing more successful and concentrated.

The second portion of this subsection lists the five industries with the largest stocks of purchased and own-account marketing, in order, in 1987, 2002, and 2020.

<Table 9 goes about here>

The lists of leading industries are generally reasonable. Conventional lists of leading advertisers are likely to emphasize consumer industries, such as retail trade, pharmaceuticals, electronics, automotive, food, and finance. National accounts methods frequently assign advertising expenditures to different functions of a firm, such as retail or wholesale trade, finance, or the management of companies, rather than to the final product eventually sold. That probably explains why relatively few consumer industries appear on the list of the largest advertisers.²⁶

6. Conclusions

The IMF report on Marketing Assets urged examination of the feasibility of incorporating these assets into the national accounts. The summary of the IMF report on Marketing Assets (2022, page 2) states "As part of the global consultation, it is proposed to enquire to what extent economies still face

²⁶ It is probably an anomaly that construction is the third largest purchaser of advertising in 1987. Recall that, prior to 1997, we measure advertising from data on "miscellaneous professional, scientific, and technical services". This category includes items such as accounting, architectural, engineering, and design services as well as advertising. Shifts among these categories, between 1982 and 1997, could overstate advertising in construction in 1987.

measurement challenges, which prevent capitalizing marketing assets." This paper shows that, for the United States, building on the approach of Corrado, Hulten, and Sichel (2005; 2009) and Corrado and Hao (2014), this is feasible.

It would be useful to develop an understanding of how sales workers affect the value of marketing assets and to integrate emerging work on the value of trademarks (Dosi et al., 2022) with a more general view of firm assets and capabilities. Two trends emerging in the literature are likely to have a strong impact on how marketing and other intangibles are understood. Mandel (2019) has argued that it is difficult to develop measures of the output price of marketing because recent shifts to digital advertising, such as through the Internet or smart phones, make marketing far more productive than prior advertising media. Koh, Santaeulalia, and Zheng (2022) propose that a portion of the added income produced by capitalization of intangibles should be credited to labor rather than to capital. These two lines of thought could have a strong influence on how marketing is eventually understood. In addition, time diaries could help estimate what proportion of time advertising and own-account marketing workers devote to long-term investment. Finally, there are also further issues to consider, such as the appropriate treatment of licensing and franchising.

Despite these topics that require further attention, the clear message of our paper is that a remarkably lot can be done to develop a comprehensive treatment of marketing for the United States. It is possible to construct broad measures of purchased advertising, other purchases of marketing services, and parallel data on own-account marketing. These measures all rest on solid and highly detailed data. We think that, overall, the results from our study strongly support the idea that the countries of this world will be able to capitalize marketing effectively.

Our most central empirical result is that marketing contributes as much to overall output growth as R&D or software. Table 1 shows that the contribution of marketing to output growth increased over time, whereas the contributions of R&D and software tended to stabilize. Own-account marketing grew more quickly than purchased marketing, steadily over the entire period. Purchases from web design and hosting and from marketing services, together with increased own-account employment of technical and marketing skills, all helped to drive marketing investment. However, capitalization of both forms of marketing has only a modest effect on the growth of output.

These estimates of the impact of marketing largely occur because we have classified certain elements of the revolution in computers and data as contributors to marketing. Investments in web design and hosting are certainly a central element of marketing investment. Similarly, as in Corrado and Hao (2014), own-account investments in computer and marketing occupations are crucial in developing the internal capabilities of the firms that compose each industry. It will be important to determine how these investments should be allocated between the data revolution and marketing.

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Appendix A. Overview of Investment in Marketing Assets

Appendix A provides a condensed view of our basic data set. The figures below show investment in each form of marketing as a percentage of total investment in marketing in every year. Figure A-1a reports investment shares under our basic assumptions, which treat 60 percent of expenditures on purchased advertising, 80 percent of expenditures on other marketing purchases, and 30 percent of own-account marketing expenditures as investment.

Figure A-1a shows that in 2020 purchases of advertising accounted for 52 percent of total investment in marketing, which represented a sharp decline from 71 percent in 1987.²⁷ In contrast, web design and hosting did not exist at all in 1987, but by 2020 represented 9 percent of marketing investment. The share of investment in marketing services increased moderately after 1997, and the share of marketing analysis also increased modestly. As a result of changes in these components, especially the decline in advertising, the purchased marketing share of investment declined from 84 percent of investment in 1987 to 78 percent by 2020. Conversely, the own-account share of marketing investment increased from 16 percent in 1987, to 19 percent in 2002, and 22 percent by 2020. Table C-1a, in the online Appendix, lists the specific investment shares that underly Figure A-1a.

Figure A-1b shows the investment share based on our alternative assumptions in which 60 percent of expenditures on advertising, 95 percent of expenditures on other purchased marketing services, and 60 percent of own-account marketing are treated as investment. Since the alternative assumptions weight own-account expenditures more heavily, the own-account share of investment is greater in this scenario. The own-account share of total investment then increases from 27 percent in 1987 to 35 percent in 2020. Table C-1b in online Appendix C reports the actual data used in Figure A-1b.

Throughout Appendix A, "Ad agency services" refers to NAICS 51, 5418, and a portion of NAICS 3231. We remind the reader that this study deals with advertising as a commodity, so that "Ad agency services" here includes large amounts of advertising provided by the print media, radio and TV, and the Internet. Web design and hosting refers to NAICS 5415 and 5182. Marketing consulting is NAICS 5416, and marketing surveys refers to NAICS 5419.²⁸ The own-account share represents own-account investment by all our 61 industries.

<Figure A-1a goes about here>

The baseline assumption is that 60 percent of advertising expenditures, 80 percent of other marketing expenditures, and 30 percent of own-account expenditures represent investment.

Figure A-1b relies on the same data on expenditures, but this alternative assumes that 60 percent of advertising, 95 percent of other purchased marketing services, and 60 percent of own-account expenditures represent investment. As a consequence, own-account marketing accounts for a substantially larger share of total investment. Table C-1b in Appendix C provides the data that underly Figure A-1b.

²⁷ The data on purchased advertising are commodity data, so they reflect the sharp decline in print media.
²⁸ Our estimates of the marketing consulting and marketing research purchased in 2012 are somewhat greater than the 2007 to 2011 estimates in Corrado and Hao (2014).

<Figure A-1b goes about here>

Figure A-2 shows marketing investments from each of the providers as a percentage of GDP. These data show that, since 2010, investment in most forms of marketing increased as a percentage of GDP, partially because the growth of GDP slowed. The shares of GDP devoted to advertising and to own-account marketing increased substantially, but web services and marketing analysis also increased.

<Figure A-2 goes about here>

Appendix B. Measures of Own-Account Marketing

Table B-1 shows the occupations selected to represent own-account marketing in the United States. For each occupation, we list its Standard Occupational Code (SOC), and 2021 national employment in that occupation in the OEWS data.

<Table B-1 goes about here>

These occupational definitions changed over time, and the OEWS is available only for years since the late 1990s. Because less occupation detail is available for earlier years, in certain cases we assume that the distribution of detailed occupations observed in a later year also holds true in earlier years.

We considered including several other occupational categories. We decided not to include advertising sales agents (occupational code 41-3011) because their work is typically sold as purchased advertising. Similarly, we do not include employees working in NAICS industry 541800 because their output is already counted in purchased advertising, so that their inclusion would represent double counting. We exclude editors (27-3041) and writers and authors (27-3043) because, in our judgment, most of these workers are not involved in marketing.

Once we have developed measures of labor compensation for the selected occupations, we estimate total expenditures as twice compensation, following Corrado and Hao (2014). Because the category system changes, the occupations we use are not exactly the same ones that they use. Our primary estimates assume that own-account workers spend 30 percent of their time on long-term investment, following Heys and Fotopoulou (2022). Our alternative measures assume that own-account workers spend .6 of their time on investment, a compromise between the .8 for managers and .5 for computer personnel and media workers assumed in Corrado and Hao (2013).

Table B-2 presents the amounts of own-account expenditures developed from the presence of each specific occupation. Marketing analysts, advertising, public relations, and related occupations, and web specialists account for most of the observed growth in expenditures. Marketing analysts have grown especially rapidly, probably largely because of the new capabilities permitted by improved methods and tools of data analysis. These three categories of occupations account for 145.9/161.8, or over 90 percent, of the total increase in expenditures observed between 2002 and 2020.

<Table B-2 goes about here>

Appendix C. Detailed information on how several of our main estimates are determined.

Appendix C discusses several issues that help explain how we obtain the results reported.

Assumptions concerning the investment proportions and depreciation rates used in the baseline and alternative cases: Sensitivity tests.

Table C-1 reports the investment proportions used for advertising, other purchased marketing, and own-account marketing under our baseline and alternative sets of assumptions. Table C-2 shows how sensitive our results are to other assumptions concerning the investment proportion or the rate depreciation.

 Table C-1.
 Summary of Assumptions Concerning Construction of Capital Stocks

| | Advertising expenditures | Other purchased marketing expenditures | Own-account expenditures | Depreciation Rate |
|---------------------------------------|--------------------------|--|-----------------------------|----------------------|
| Baseline assumption | 60% | 80% | 30% | 45% |
| Alternative assumption | 60% | 95% | 60% | 45% |
| Sensitivity test for advertising 1 | 40% | 80% | 30% | 45% |
| Sensitivity test for advertising 2 | 80% | 80% | 30% | 45% |

Table C-2. Output Contribution of Marketing Under Alternative Assumptions, 1987-2020

| | Total marketing | Purchased marketing | Own-account marketing | Software effect | R&D effect |
|-------------------------------|--------------------|---------------------|--------------------------|--------------------|---------------|
| Basic assumptions | .18 | .15 | .03 | .19 | .15 |
| Alternative assumptions | .23 | .16 | .07 | .19 | .15 |
| Advertising 40% investment | .15 | .12 | .03 | .19 | .15 |
| Advertising 80% investment | .22 | .19 | .03 | .19 | .15 |
| Depreciation rate of 65% | .17 | .14 | .03 | .19 | .15 |

Table C-2 shows how the estimated impact of marketing varies with the percentage of expenditures that is treated as investment. The alternative set of assumptions suggests that the investment percentage is greater for expenditures on purchased marketing and, especially, for own-account marketing, so that their implied effect on output growth is stronger. If the investment percentage of advertising is greater, or lower, the implied impact of advertising on output growth similarly increases or decreases. In contrast, a change in depreciation, within the range that most estimates say is realistic, has little impact on output growth. The effect of software and of R&D does not vary across these different possibilities, so the relative importance of marketing varies with the specific parameters used to measure the marketing stock, especially the proportion of expenditures that is regarded as investment.

How we measure the impact of capital input on output growth.

The second portion of this appendix explains how we measure the effect of different forms of capital on output growth. Tables 1 and 4 of the main text discuss the impact that several types of intangibles have on output growth. This subsection explains how we determine these estimates.

Analysis of the effect of the influence of intangibles begins with measures of the stock and rental price of each asset in each year and industry. The BLS Capital Database currently contains information on 101 assets. We add further information on purchased marketing and own-account marketing, making a total of 103 assets. We remind the reader that in the BLS Database most rental prices are designed to exhaust property income. However, in a sizable number of industries, approaching 30 percent, the BLS uses external information to develop more plausible internal rates of return.

For private business contributions to growth, we first Tornqvist aggregate each asset across the 61 industries for each of the 103 assets in every year. We calculate the share of each asset as the share of (stocks * rental prices)/value added. Finally, the contribution of each asset to output growth is calculated from the percentage change in that stock multiplied by that stock's associated share of value added.

The contribution of each intangible to output growth in a particular industry is measured in a very similar way. The percentage increase in the quantity of a specific intangible in that industry is then weighted by that intangible's share of value added in that industry.

Estimates of the price of marketing prior to 1997.

The third subsection describes how we prepare a price index for advertising in the years from 1982 to 1997. The present study measures the price of marketing from BEA estimates of the price of the commodity advertising from 1997 to 2020. These BEA prices depend largely on PPI data. PPI data provide much less information on the prices of services prior to 1997. Consequently, it is more difficult to price advertising before 1997.

Conceptually, the underlying product is a cognitive impression by a viewer of the ads, weighted more heavily for those viewers who are likely to buy or consume something in the product category being advertised. This is challenging to construct as the technologies have changed so much over time and our information about the experiences of the population are limited.

Conceptually, the underlying product is a viewer impression combined with a marketing message. Each impression should then be weighted by the likelihood that the marketing message persuades the viewer to buy or consume the product. (An impression doesn't have value if the viewer has zero chance of buying a product, or if the viewer is certain to buy the product, regardless of the message.) This is challenging to construct as the technologies for targeting viewers and tailoring marketing messages to each viewer have changed so much over time, and because our information about the degree of targeting and tailoring is limited.

We looked carefully at this issue. The deflator selected for 1982 to 1997 makes a difference in that different deflators can have a considerable effect on the implied growth of marketing during these early years. We would have preferred to rely heavily on the available PPIs. However, some tests showed that the BEA deflators for 1997 to 2020 reflect output costs for advertising as well as PPI information. We consequently prepared 1982 to 1997 deflators that weight the PPI evidence one-half, and advertising costs also one-half.

Hourly advertising costs are calculated by dividing nominal expenditures in a particular category with an estimate of the time spent by users reading/listening/viewing advertising. Nominal expenditures for each advertising category are calculated from the same datasets used earlier to calculate total advertising. Data on user time is mostly drawn from Nakamura et al., which studied free content supported by advertising and marketing for the 1929 to 2017 period. This paper supplements that data with estimates of user time after 2018 and estimates of direct mail reading time are taken from the Post Office's Household Diary Survey

(https://www.prc.gov/docs/119/119244/Final%20HDS%202020%20Annual%20report.pdf).

Like other advertising price indexes (e.g., the PPIs), our advertising price index declines in the period where ads start to be digitally targeted over the Internet, and dips in recessions.

Estimates of expenditures on web services, marketing consulting services, and purchased marketing services in each industry.

We prepare annual data on expenditures on web services, on marketing consulting services, and on purchased marketing services from the Census of Services and the Services Annual Survey (SAS). We believe that our estimates of the national totals for each of these categories are reasonably reliable. Unfortunately, these same sources cannot provide reliable information on how much each industry spends on each of these categories, especially since it is difficult to track expenditures on secondary products.

We therefore allocate the national totals for each of these categories to individual industries based on the IO tables. However, the IO tables do not cover the detailed industries that produce these products. For example, the specific industry that provides marketing consulting services is NAICS 541613, but the most detailed input-output industry for which data are available is NAICS 541610, management consulting services in general. Consequently, our allocation of purchased marketing services to individual industries is not precise. Further research on how much marketing services each industry buys would improve national accounts.

Appendix D. Marketing services purchased from various categories of suppliers

Appendix D is to be published online only. It provides information on the marketing services purchased from each type of supplier.

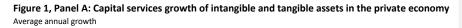
Table D-1a. Providers of Marketing Services, in Percentages With baseline investment proportions

| Year Ad Agency Services Own-Account Share Marketing Consulting Marketing Surveys Web Design and Hosting 1987 76.8% 13.5% 2.3% 7.4% 0.0% 1988 66.0% 23.3% 2.6% 8.0% 0.0% 1989 65.8% 23.5% 2.8% 7.9% 0.0% 1990 65.5% 23.7% 3.0% 7.8% 0.0% 1991 65.1% 23.9% 3.1% 7.8% 0.0% 1992 65.0% 23.9% 3.2% 8.0% 0.0% 1993 64.9% 24.0% 3.3% 7.8% 0.0% 1994 65.2% 24.0% 3.5% 7.3% 0.0% 1995 65.0% 24.3% 3.9% 6.7% 0.1% 1996 64.1% 24.4% 4.1% 7.2% 0.2% 1997 63.5% 24.6% 4.8% 7.2% 1.2% 1999 62.1% 24.6% 4.8% 7.2% 1.2 | | | with baseline in | | | | | | | |
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| 2013 50.2% 28.3% 9.5% 7.0% 5.0% 2014 49.4% 28.7% 9.8% 7.3% 5.0% 2015 49.2% 29.3% 9.5% 7.0% 5.1% 2016 49.1% 29.4% 9.5% 7.0% 5.0% 2017 48.6% 29.9% 9.5% 6.6% 5.5% 2018 48.4% 30.4% 9.2% 6.4% 5.6% 2019 48.4% 30.0% 9.1% 6.7% 5.8% | | | | | | | | | | |
| 2014 49.4% 28.7% 9.8% 7.3% 5.0% 2015 49.2% 29.3% 9.5% 7.0% 5.1% 2016 49.1% 29.4% 9.5% 7.0% 5.0% 2017 48.6% 29.9% 9.5% 6.6% 5.5% 2018 48.4% 30.4% 9.2% 6.4% 5.6% 2019 48.4% 30.0% 9.1% 6.7% 5.8% | | | | | | | | | | |
| 2015 49.2% 29.3% 9.5% 7.0% 5.1% 2016 49.1% 29.4% 9.5% 7.0% 5.0% 2017 48.6% 29.9% 9.5% 6.6% 5.5% 2018 48.4% 30.4% 9.2% 6.4% 5.6% 2019 48.4% 30.0% 9.1% 6.7% 5.8% | | | | | | | | | | |
| 2016 49.1% 29.4% 9.5% 7.0% 5.0% 2017 48.6% 29.9% 9.5% 6.6% 5.5% 2018 48.4% 30.4% 9.2% 6.4% 5.6% 2019 48.4% 30.0% 9.1% 6.7% 5.8% | | | | | | | | | | |
| 2017 48.6% 29.9% 9.5% 6.6% 5.5% 2018 48.4% 30.4% 9.2% 6.4% 5.6% 2019 48.4% 30.0% 9.1% 6.7% 5.8% | | | | | | | | | | |
| 2018 48.4% 30.4% 9.2% 6.4% 5.6% 2019 48.4% 30.0% 9.1% 6.7% 5.8% | 2016 | | | | | | | | | |
| 2019 48.4% 30.0% 9.1% 6.7% 5.8% | 2017 | | | | | | | | | |
| | | | | | | | | | | |
| 2020 48.0% 30.8% 8.9% 6.3% 6.0% | 2019 | | 30.0% | | | | | | | |
| | 2020 | 48.0% | 30.8% | 8.9% | 6.3% | 6.0% | | | | |

| Table D-1b. Providers of Marketing Services, in Percentages | |
|---|--|
| With alternative investment proportions. | |

| Year | Ad Agency | Own-Account | Marketing | Marketing | Web Design and |
|------|-----------|-------------|------------|-----------|----------------|
| | Services | Share | Consulting | Surveys | Hosting |
| 1987 | 66.6% | 23.3% | 2.4% | 7.6% | 0.0% |
| 1988 | 66.0% | 23.3% | 2.6% | 8.0% | 0.0% |
| 1989 | 65.8% | 23.5% | 2.8% | 7.9% | 0.0% |
| 1990 | 65.5% | 23.7% | 3.0% | 7.8% | 0.0% |
| 1991 | 65.1% | 23.9% | 3.1% | 7.8% | 0.0% |
| 1992 | 65.0% | 23.9% | 3.2% | 8.0% | 0.0% |
| 1993 | 64.9% | 24.0% | 3.3% | 7.8% | 0.0% |
| 1994 | 65.2% | 24.0% | 3.5% | 7.3% | 0.0% |
| 1995 | 65.0% | 24.3% | 3.9% | 6.7% | 0.1% |
| 1996 | 64.1% | 24.4% | 4.1% | 7.2% | 0.2% |
| 1997 | 63.5% | 24.6% | 4.3% | 7.0% | 0.6% |
| 1998 | 62.6% | 24.5% | 4.8% | 6.9% | 1.2% |
| 1999 | 62.1% | 24.6% | 4.8% | 7.2% | 1.3% |
| 2000 | 62.5% | 24.1% | 5.0% | 7.2% | 1.2% |
| 2001 | 60.3% | 25.3% | 5.7% | 7.4% | 1.3% |
| 2002 | 59.5% | 25.7% | 6.1% | 7.3% | 1.3% |
| 2003 | 58.7% | 26.5% | 6.0% | 7.3% | 1.5% |
| 2004 | 59.2% | 25.3% | 6.3% | 7.4% | 1.7% |
| 2005 | 58.8% | 25.1% | 6.8% | 7.4% | 1.9% |
| 2006 | 58.2% | 25.2% | 6.9% | 7.5% | 2.2% |
| 2007 | 57.9% | 24.7% | 7.2% | 7.7% | 2.4% |
| 2008 | 56.3% | 24.9% | 7.9% | 8.0% | 2.9% |
| 2009 | 53.2% | 27.1% | 8.1% | 8.2% | 3.4% |
| 2010 | 52.1% | 27.1% | 8.6% | 8.3% | 3.8% |
| 2011 | 50.9% | 27.7% | 9.0% | 8.1% | 4.3% |
| 2012 | 51.3% | 27.4% | 9.2% | 7.2% | 4.8% |
| 2013 | 50.2% | 28.3% | 9.5% | 7.0% | 5.0% |
| 2014 | 49.4% | 28.7% | 9.8% | 7.3% | 5.0% |
| 2015 | 49.2% | 29.3% | 9.5% | 7.0% | 5.1% |
| 2016 | 49.1% | 29.4% | 9.5% | 7.0% | 5.0% |
| 2017 | 48.6% | 29.9% | 9.5% | 6.6% | 5.5% |
| 2018 | 48.4% | 30.4% | 9.2% | 6.4% | 5.6% |
| 2019 | 48.4% | 30.0% | 9.1% | 6.7% | 5.8% |
| 2020 | 48.0% | 30.8% | 8.9% | 6.3% | 6.0% |
| | | | | | |





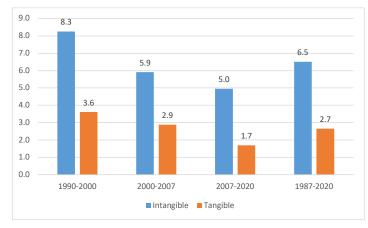
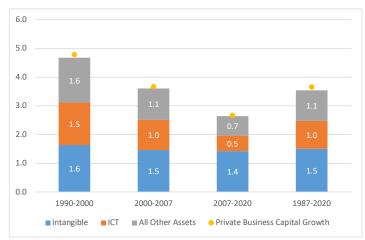


Figure 1, Panel B: Contributions of different types of assets to private capital growth Percentage point contribution, average annual growth



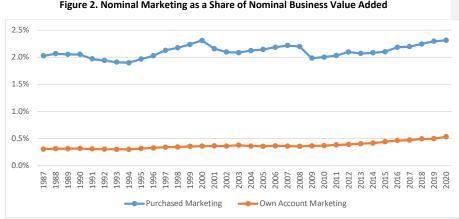
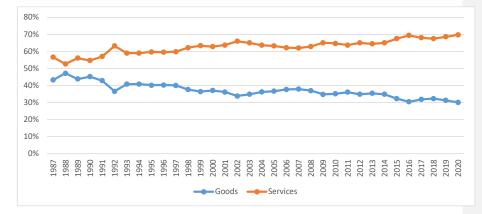


Figure 2. Nominal Marketing as a Share of Nominal Business Value Added

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Figure 3-A. Flow of Capital Input to the Production of Goods and Services, 1987-2020



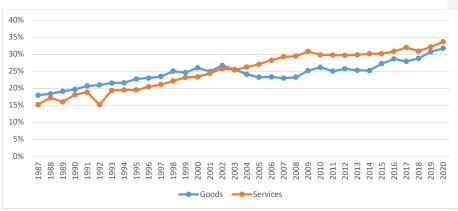
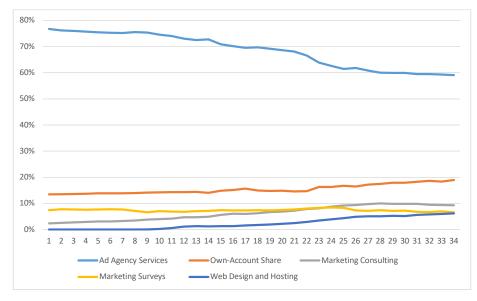
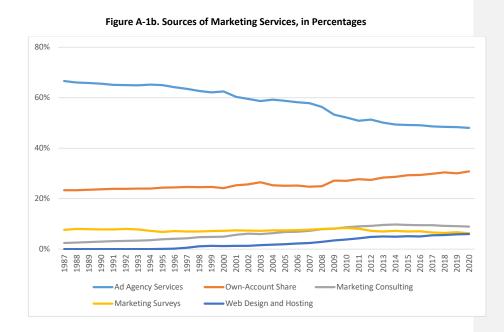


Figure 3-B. The Intangible Share of Capital Services, in Goods and in Services, 1987-2020

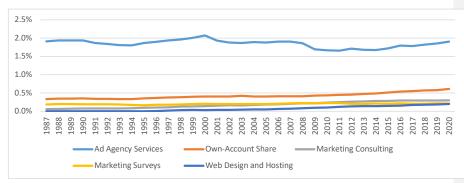
29

Figure A-1a. Sources of Marketing Services, in Percentages









Tables

Table 1. Input and TFP contributions to private business output growth

| | 1990-2000 | 2000-2007 | 2007-2020 | 1987-2020 |
|-------------------------|-----------|-----------|-----------|-----------|
| Purchased Marketing | 0.14 | 0.16 | 0.17 | 0.15 |
| Own Account Marketing | 0.03 | 0.03 | 0.05 | 0.03 |
| Entertainment Originals | 0.04 | 0.04 | 0.02 | 0.03 |
| R&D | 0.17 | 0.13 | 0.14 | 0.15 |
| Software, Custom | 0.07 | 0.05 | 0.07 | 0.06 |
| Software, Own Account | 0.03 | 0.04 | 0.03 | 0.03 |
| Software, Prepackaged | 0.11 | 0.10 | 0.11 | 0.10 |
| ICT | 0.54 | 0.39 | 0.22 | 0.37 |
| All Other Assets | 0.56 | 0.41 | 0.28 | 0.42 |
| Labor Input | 1.43 | 0.30 | 0.31 | 0.75 |
| TFP Residual | 0.93 | 1.10 | 0.45 | 0.71 |
| Private Business Output | 3.9 | 2.5 | 1.7 | 2.7 |

Table 2. Purchased and Own-Account Marketing Long-Run Effects

| 1987-2020 | TFP | Output | Capital | Labor |
|--|-------|--------|---------|-------|
| Private Industry Without Any Marketing | 0.752 | 2.707 | 3.388 | 1.174 |
| Private Industry With Only Purchased Marketing | 0.703 | 2.762 | 3.599 | 1.174 |
| Private Industry With All Marketing | 0.699 | 2.784 | 3.655 | 1.174 |

Table 3. Real Investment Growth by Asset Category

| Average annual change | 1990-2000 | 2000-2007 | 2007-2020 | 1987-2020 |
|------------------------------------|-----------|-----------|-----------|-----------|
| Purchased Advertising | 10.7 | 6.4 | 4.0 | 7.3 |
| Own Account Advertising | 10.8 | 7.1 | 6.8 | 8.6 |
| Entertainment Originals | 4.3 | 2.5 | 1.0 | 2.5 |
| R&D | 5.5 | 2.7 | 3.6 | 4.2 |
| Software, Custom | 16.2 | 4.3 | 6.1 | 9.3 |
| Software, Own Account | 7.4 | 3.2 | 4.2 | 5.3 |
| Software, Prepackaged | 24.0 | 9.8 | 11.0 | 17.8 |
| ICT | 8.4 | 6.4 | 4.5 | 6.0 |
| All Other Assets | 3.2 | 1.3 | 0.8 | 1.7 |
| Private Business All Assets Growth | 4.5 | 2.4 | 2.5 | 3.1 |

| | | owth, percent Factor Share, percent hange change | | | Contribution percenta | |
|-----------|---------------------------|---|--------------------------|------------------------|--------------------------|---------------------|
| | Software, pre-packaged | Purchased Marketing | Software, prepackaged | Purchased marketing | Software, prepackaged | Purchased marketing |
| 2008 | 6.5% | 8.0% | 0.9% | 3.1% | 0.06 | 0.25 |
| 2009 | 6.2% | 2.7% | 0.9% | 3.4% | 0.06 | 0.09 |
| 2010 | 4.0% | -0.5% | 1.0% | 3.6% | 0.04 | -0.02 |
| 2011 | 4.6% | 0.9% | 0.9% | 3.5% | 0.04 | 0.03 |
| 2012 | 9.9% | 3.4% | 0.9% | 3.4% | 0.09 | 0.12 |
| 2013 | 12.3% | 5.1% | 0.9% | 3.5% | 0.12 | 0.18 |
| 2014 | 11.5% | 5.4% | 1.0% | 3.6% | 0.11 | 0.19 |
| 2015 | 10.7% | 5.4% | 1.1% | 3.6% | 0.12 | 0.20 |
| 2016 | 11.2% | 5.7% | 1.1% | 3.7% | 0.13 | 0.21 |
| 2017 | 11.9% | 6.0% | 1.2% | 3.9% | 0.14 | 0.23 |
| 2018 | 12.9% | 6.3% | 1.2% | 3.7% | 0.16 | 0.23 |
| 2019 | 13.3% | 6.8% | 1.3% | 3.5% | 0.18 | 0.24 |
| 2020 | 12.8% | 5.6% | 1.4% | 3.7% | 0.19 | 0.21 |
| 1987-2020 | 17.9% | 7.5% | 0.8% | 2.4% | 0.10 | 0.15 |
| 1990-2000 | 26.9% | 9.6% | 0.5% | 1.5% | 0.11 | 0.14 |
| 2000-2007 | 11.6% | 6.7% | 0.9% | 2.4% | 0.10 | 0.16 |
| 2007-2020 | 10.3% | 4.8% | 1.1% | 3.6% | 0.11 | 0.17 |

Table 4. The Effect of Prepackaged Software and Purchased Marketing on Output Growth

| | (1) | (2) = .1 * (1) | (3) = .2 * (1) | (4) |
|------|----------------------|-------------------|-------------------|-----------|
| Year | Digital share (1) | Bonus, model 1 | Bonus, model 2 | BEA price |
| 2002 | 0.9 | 0.1 | 0.2 | 82.636 |
| 2003 | 1.5 | 0.2 | 0.3 | 83.778 |
| 2004 | 2.5 | 0.3 | 0.5 | 84.609 |
| 2005 | 4.1 | 0.4 | 0.8 | 89.700 |
| 2006 | 7.1 | 0.7 | 1.4 | 95.709 |
| 2007 | 11.4 | 1.1 | 2.3 | 97.537 |
| 2008 | 13.6 | 1.4 | 2.7 | 98.088 |
| 2009 | 16.0 | 1.6 | 3.2 | 97.696 |
| 2010 | 18.5 | 1.9 | 3.7 | 97.612 |
| 2011 | 22.3 | 2.2 | 4.4 | 98.723 |
| 2012 | 26.0 | 2.6 | 5.2 | 100.000 |
| 2013 | 29.1 | 2.9 | 5.8 | 101.210 |
| 2014 | 30.8 | 3.1 | 6.2 | 102.127 |
| 2015 | 38.2 | 3.8 | 7.6 | 102.775 |
| 2016 | 42.1 | 4.2 | 8.4 | 103.337 |
| 2017 | 47.1 | 4.7 | 9.4 | 103.809 |
| 2018 | 50.7 | 5.1 | 10.0 | 103.952 |
| 2019 | 55.1 | 5.5 | 11.0 | 104.915 |
| 2020 | 58.3 | 5.8 | 11.6 | 103.696 |

 Table 5.
 The Digital Share of the Advertising Market and the Digital Bonus, 2002-2020

| Table 6. Panel A: Shares of Tangibles in | Total Capital Stock Values by Sector, 2012 |
|--|--|
|--|--|

| | Equipment | Structures | Inventories | Land | Total Tangible |
|---------------------------------|-----------|------------|-------------|-------|-------------------|
| Manufacturing | 30.6% | 12.4% | 5.5% | 25.0% | 73.5% |
| Other Goods | 16.8% | 3.7% | 23.8% | 54.9% | 99.3% |
| Trade | 14.8% | 23.9% | 22.3% | 32.9% | 93.8% |
| Finance, Insurance, Real Estate | 14.5% | 0.2% | 21.0% | 61.0% | 96.7% |
| Other Services | 22.1% | 1.5% | 13.2% | 44.1% | 80.8% |

| Table 6. Panel B: Shares of Intangibles in Total Capital Stock Values by Sector, 201 | Table 6. | Panel B: | Shares of Intangibl | es in Total Ca | pital Stock Value | es by Sector | ·. 2012 |
|--|----------|----------|---------------------|----------------|-------------------|--------------|---------|
|--|----------|----------|---------------------|----------------|-------------------|--------------|---------|

| | Purchased Marketing | Own Account Marketing | Entertainment Originals | R&D | Software, Custom | Software, Own Account | Software, Prepackaged | Total Intangible |
|----------------|------------------------|-----------------------------|----------------------------|-------|---------------------|-----------------------------|--------------------------|---------------------|
| Manufacturing | 1.0% | 0.2% | 0.0% | 23.6% | 0.9% | 0.3% | 0.4% | 26.5% |
| Other Goods | 0.2% | 0.0% | 0.0% | 0.3% | 0.1% | 0.0% | 0.1% | 0.7% |
| Trade | 3.2% | 0.4% | 0.0% | 1.0% | 0.8% | 0.4% | 0.5% | 6.2% |
| FIRE | 1.2% | 0.2% | 0.0% | 0.3% | 0.7% | 0.2% | 0.7% | 3.3% |
| Other Services | 3.2% | 0.7% | 6.7% | 5.0% | 2.0% | 1.0% | 0.7% | 19.2% |

| Table 7. Correlations between sources of growth, measured as contributions to annual growth, across 61 | |
|--|--|
| industries, 1987-2020 | |

Note: These correlations use estimates of contributions from a slightly earlier version of the data.

| | ICT | Sw, P | Sw, C | Sw, O | Struct | Land | Inven | R&D | Originals | Purch Ads | OA Ads | Labor Input | Labor hours | Labor comp | TFP residual |
|------------------------|-------|--------------------|--------------------|--------------------|--------|--------|--------|--------|-----------|--------------------|--------------------|----------------|----------------|---------------|-----------------|
| Equipment | 0.420 | 0.104 | 0.003 | 0.001 | 0.141 | -0.038 | 0.043 | -0.029 | -0.056 | 0.129 | -0.047 | 0.007 | 0.034 | -0.238 | -0.240 |
| ICT | | <mark>0.617</mark> | <mark>0.645</mark> | <mark>0.642</mark> | 0.218 | 0.375 | -0.057 | 0.297 | 0.082 | <mark>0.659</mark> | <mark>0.665</mark> | 0.088 | 0.103 | -0.154 | -0.134 |
| Software, Purchased | | | <mark>0.824</mark> | <mark>0.741</mark> | 0.191 | 0.349 | -0.015 | 0.343 | -0.040 | <mark>0.684</mark> | <mark>0.721</mark> | 0.109 | 0.102 | 0.040 | -0.053 |
| Software, Custom | | | | <mark>0.966</mark> | 0.154 | 0.484 | 0.003 | 0.493 | -0.031 | <mark>0.719</mark> | <mark>0.911</mark> | 0.097 | 0.093 | 0.016 | -0.013 |
| Software, Own Acct | | | | | 0.167 | 0.547 | -0.009 | 0.432 | -0.026 | <mark>0.700</mark> | <mark>0.892</mark> | 0.105 | 0.106 | -0.033 | -0.091 |
| Structures | | | | | | 0.198 | -0.096 | -0.008 | -0.009 | 0.116 | 0.149 | 0.118 | 0.141 | -0.228 | -0.111 |
| Land | | | | | | | -0.049 | 0.234 | -0.017 | 0.402 | 0.516 | 0.002 | -0.006 | 0.074 | -0.034 |
| Inventories | | | | | | | | 0.220 | -0.171 | 0.114 | -0.027 | -0.108 | -0.108 | 0.019 | 0.137 |
| R&D | | | | | | | | | -0.039 | <mark>0.302</mark> | <mark>0.483</mark> | -0.039 | -0.051 | 0.115 | 0.161 |
| Originals | | | | | | | | | | 0.166 | 0.102 | -0.011 | -0.002 | -0.078 | -0.041 |
| Purch Ads | | | | | | | | | | | <mark>0.746</mark> | 0.145 | 0.143 | -0.016 | -0.141 |
| OA Ads | | | | | | | | | | | | 0.192 | 0.177 | 0.086 | -0.015 |
| Labor input | | | | | | | | | | | | | 0.518 | -0.232 | -0.062 |
| Labor hours | | | | | | | | | | | | | | -0.268 | -0.197 |
| Labor composition | | | | | | | | | | | | | | | 0.090 |

 Table 8. Stocks of each intangible in the top 10 Industries, as a percentage of all stocks of that intangible in the private sector, 1987, 2012, and 2020.

| Intangible | 1987 Industry concentration | 2002 Industry concentration | 2020 Industry concentration |
|-------------------------|-----------------------------|-----------------------------|-----------------------------|
| Purchased Marketing | 66% | 62% | 70% |
| Own Account Marketing | 69% | 70% | 74% |
| Entertainment Originals | 100% | 100% | 100% |
| R&D | 82% | 76% | 79% |
| Software, Custom | 66% | 67% | 80% |
| Software, Own Account | 66% | 72% | 84% |
| Software, Prepackaged | 63% | 63% | 70% |
| Total marketing | 63% | 62% | 69% |
| Total software | 66% | 66% | 75% |

 Table 9. Industries with the largest stocks of purchased or own-account marketing, 1987, 2012, and

 2020.

Purchased marketing

| 1987 | 2002 | 2020 |
|---|--|--|
| Retail trade | Retail trade | Retail trade |
| Wholesale trade | Wholesale trade | Wholesale trade |
| Construction | Broadcasting and telecommunications | Federal Reserve banks, credit intermediation, and related activities |
| Management of companies and enterprises | Food services and drinking places | Miscellaneous professional, scientific, and technical services |
| Food services and drinking places | Miscellaneous professional, scientific, and technical services | Management of companies and enterprises |

Own-account marketing

| 1987 | 2002 | 2020 | | |
|--|--|--|--|--|
| Broadcasting and telecommunications | Broadcasting and telecommunications | Management of companies and enterprises | | |
| Publishing industries, except internet (includes software) | xcept internet (includes companies and enterprises | | | |
| Management of companies and enterprises | Wholesale trade | Wholesale trade | | |
| Wholesale trade | Publishing industries, except internet (includes software) | Broadcasting and telecommunications | | |
| Computer and electronic products | Computer and electronic products | Publishing industries, except internet (includes software) | | |

| Table B-1. Oc | cupations that | Produce Own | Account Ma | rketing |
|---------------|----------------|-------------|--------------------------------|---------|
|---------------|----------------|-------------|--------------------------------|---------|

| Occupation title | SOC occupation code | 2021 employment |
|---|---------------------|-----------------|
| Advertising and promotion managers | 11-2011 | 22,520 |
| Marketing managers | 11-2021 | 278,690 |
| Public relations managers and fund-raising managers | 11-2030 | 83,040 |
| Market research analysts and marketing specialists | 13-1161 | 727,540 |
| Web developers | 15-1254 | 84,820 |
| Web and digital interface designers | 15-1255 | 82,380 |
| Other computer occupations | 15-0000, part | 224,375 |
| Survey researchers | 19-3022 | 8,850 |
| Art directors | 27-1011 | 42,080 |
| Merchandise displayers and window trimmers | 27-1026 | 159,790 |
| Public relations specialists | 27-3031 | 242,710 |
| Media and communications workers, all other | 27-3099 | 16,540 |

Table B-2. The Contributions of Specific Occupations to Own-Account Marketing Expenditures

| Year | Web design and hosting | Marketing analysis | Advertising, public relations, and related | Marketing research | Other computer |
|------|---------------------------|-----------------------|---|-----------------------|-------------------|
| | (1) | (2) | (3) | (4) | (5) |
| 2002 | 4.0 | 13.2 | 61.7 | 0.6 | 11.5 |
| 2012 | 7.7 | 44.1 | 80.7 | 1.0 | 18.3 |
| 2020 | 13.8 | 79.5 | 131.5 | 0.7 | 27.3 |

Note: SOC codes for each column:

Column (1): 15-1254, 15-1255.

Column (2): 13-1161.

Column (3):11-2011, 11-2021, 11-2032, 27-1011, 27-1026, 27-3031, and 27-3099.

Column (4): 19-3022. Column (5): 15-0000 other.

Column (5): 15-0000 other, calculated as 5% of computer workers outside of web design and ting

hosting