

# Measuring AI's direct contribution to US GDP



Tina Highfill | Bureau of Economic Analysis  
David Wasshausen | Bureau of Economic Analysis  
Gregory Prunchak | Bureau of Economic Analysis

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## Abstract

Much of the current literature on the economic impact of artificial intelligence (AI) focuses on the uses of AI, but little is known about the production of AI and its impact on the economy. Accurate and reliable macroeconomic statistics describing AI production are necessary to understand the contribution of AI to economic growth, the industries involved in production, and the number of workers employed in the AI production process. This note summarizes a recent paper by the US Bureau of Economic Analysis (BEA) that outlines the concepts and challenges related to measuring production of AI using standard national accounting practices employed to calculate official statistics like gross domestic product (GDP). As with many new or transitioning areas of the economy, AI measurement issues relate generally to defining the scope of production and source data availability.

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Disclaimer: This policy brief is based on "[Concepts and Challenges of Measuring Production of Artificial Intelligence in the U.S. Economy](#)". The views expressed are those of the authors and not necessarily those of the institutions the authors are affiliated with.

## Measuring Economic Phenomena

Macroeconomic statistics like GDP are often published by type of industry, such as manufacturing and construction. For areas of the economy that are not easily identifiable under standard industry classifications, or that span multiple industries, national accountants sometimes develop “satellite accounts” that allow for an analysis of selected slices of the economy. For example, **BEA has developed satellite accounts** for outdoor recreation, travel and tourism, the marine economy, and the digital economy. Satellite accounts are developed using data and methods consistent with official economic statistics, so they can be used to identify what share of the economy is attributable to the satellite account area. For example, recent BEA reports show that the **marine economy** represented 1.8 percent of GDP in 2022, while the **digital economy** represented 10.0 percent.

The US Supply and Use Tables (SUTs) provide the basis for BEA's GDP by Industry statistics and its satellite accounts. The SUTs provide a comprehensive accounting of the value of products purchased by consumers, businesses, and government, plus the value that is imported and exported, all delineated by the producing industry. Developing a satellite account involves three main steps:

1. Identify relevant product categories within the SUTs
2. Isolate relevant shares of economic activity within product categories, when necessary
3. Use the SUT framework to determine economic activity by industry, including contribution to GDP, gross output, employment, and compensation.

Identifying relevant product categories to include in a satellite account (step 1) requires a solid definition of the subject being measured.

## Defining the Scope of AI Production

Defining the subject of interest is often the most important phase of a satellite account because it provides the overall framework for the new statistics and shapes which goods and services are chosen to be part of the account. AI has been defined in many ways, though a common underlying theme is technologies that perform tasks by mimicking human-like senses, learning, and actions (table 1).

For the purposes of developing an AI satellite account, an operationalizable definition must clearly delineate what products and industries are considered part of AI production. In general, existing AI definitions tend to mention, at least implicitly, both hardware and software components needed for computing. Three industries in particular appear to be significant: manufacturing (including accelerator chips), information (including software publishing), and professional and business services (including R&D and computer and data services).

Table 1. Examples of Definitions for Artificial Intelligence	
<b>US Census Bureau</b>	“Artificial Intelligence is computer systems and software that are able to perform tasks normally requiring human intelligence, such as decision-making, visual perception, speech recognition, and language processing. Types or applications of AI include machine learning, natural language processing, virtual agents, predictive analytics, machine vision, voice recognition, decision making systems, data analytics, text analytics, image processing, etc.”
<b>National Institute of Standards and Technology</b>	AI technologies and systems “comprise software and/or hardware that can learn to solve complex problems, make predictions or undertake tasks that require human-like sensing (such as vision, speech, and touch), perception, cognition, planning, learning, communication, or physical action.”
<b>Organisation for Economic Co-operation and Development (OECD)</b>	“An AI system is a machine-based system that, for explicit or implicit objectives, infers, from the input it receives, how to generate outputs such as predictions, content, recommendations, or decisions that can influence physical or virtual environments. Different AI systems vary in their levels of autonomy and adaptiveness after deployment.”
<b>United States Patent and Trademark Office</b>	“For patent applications and grants, we define AI as comprising one or more of eight component technologies [knowledge processing, speech, AI hardware, evolutionary computation, natural language processing, machine learning, vision, planning/control]. These components span software, hardware, and applications, and a single patent document may contain multiple AI component technologies.”

## Identifying AI Production in the Supply and Use Tables

The SUTs are comprised of thousands of product categories that are often very specific, but this is not always the case depending on the original source of the data. For example, the SUTs separately track production of three different types of powered circular saw blades for woodworking (solid tooth, inserted tooth, and other), but there is only one general category for engineering services. Given the general nature of many of the SUTs product categories, all BEA satellite accounts use external data sources to isolate relevant production in certain areas, and this would be true for an AI satellite account.

Though it depends on how the economic production of AI is ultimately operationalized, one approach is to focus on the foundational areas present in existing definitions, specifically, manufacturing of chips, software publishing, computer and data services, and R&D. Within BEA’s SUTs, each of these production areas are included in categories that include non-AI production since there are currently no product categories related solely to AI.

## AI Manufacturing Data

Within government statistics, the International Trade Administration (ITA) has trade data that shows dozens of categories of semiconductor manufacturing that could be useful in identifying the imported and exported value of AI-related chips. Similarly, the US Trade in Advanced Technology Products by Technology Group and Country data from the Census have AI products embedded in certain categories that could potentially be useful. However, neither of these

datasets could be used directly to quantify AI chip production, since the data categories commingle AI chips with non-AI chips.

Private data sources may be necessary to measure the production of AI chip manufacturing. The International Data Corporation (**IDC**), a private data firm, sells an “Artificial Intelligence Infrastructure Tracker” that contains revenue information for AI hardware, software, and various related services. Since purchasing data is sometimes expensive and necessitates special funding, this data source may not be feasible. An alternative is to use revenue information from public financial reports for major chip producers, such as 10-K filings required by the Security and Exchange Commission or publicly traded companies like **NVIDIA**.

## AI Software Data

As with manufacturing, since no standalone categories exist in the SUTs for the production of AI software and relevant computer and data services, private vendor data or public financial information for key companies would likely be needed to measure production in this area. Another possible resource is the Occupational Employment and Wage Statistics (**OEWS**) data from the US Bureau of Labor Statistics. The OEWS provides employment and wage information for occupations within industries and has been used in previous satellite accounts as a proxy to isolate production within general product categories. Within the software publishing and computer systems design and related services industries, the OEWS data provide information on a few occupations that are AI-related, such as data scientists, software quality assurance analysts and testers, and database architects. Although these occupations include non-AI work, the data could still be used for trend analysis and perhaps to serve as a comparison for potential estimates.

## AI R&D Data

BEA currently uses spending data from National Science Foundation (NSF) surveys to estimate R&D production in the SUTs. In 2019, the NSF’s Business Enterprise Research and Development survey asked about AI-related R&D for the first time. In that [survey](#), 2,590 domestic companies noted positive AI R&D spending for that year. While questions related to AI have not been asked consistently since 2019, the survey shows the possibility for these questions to be added to existing NSF surveys. Likewise, it may be possible to update data collections from the National Center for Education Statistics to measure R&D in academia.

AI R&D occurs in both the private and government sectors. In 2018 alone, the Defense Advanced Research Projects Agency (**DARPA**) launched a \$2 billion initiative to continue AI R&D, an area that the agency has been engaged in for over five decades. Information on AI R&D spending by the federal government is available via public budget documents for each agency.

## AI Prices Data

The price indexes currently used to deflate components within the three categories mentioned above come from multiple sources, including the BLS Producer Price Index and Consumer Price Index, the Federal Reserve Board’s communication equipment prices, and several BEA-derived indexes that utilize private data sources. Until government agencies collect data specifically on AI products, private data sources would likely be needed to estimate price trends. The impact of prices on growth could be substantial in many areas, especially manufacturing of chips and servers.

## Next Steps for Measuring Production of AI

As with many new or transitioning areas of the economy, AI measurement issues generally relate to defining scope and availability of source data. Although many AI definitions appear similar in concept, extensive research and outreach to subject matter experts would be needed to ensure a comprehensive and operational definition to anchor the development of macroeconomic statistics like GDP. There are often legitimate reasons to include or exclude certain types of economic production in satellite accounts. In the case of AI production, one could potentially argue that construction of data centers should also be included in a satellite account. But perhaps a larger issue is data availability. Since AI production is commingled with non-AI production within the SUTs, external data sources would be needed to isolate economic production of AI to develop satellite account statistics. While there is always the potential to change or add government data collections to better isolate AI production in the SUTs, that would take many years to implement and become available for the uses described in this Policy Brief.

### About the author(s)

**Tina Highfill** is a senior research economist at the Bureau of Economic Analysis, part of the US Department of Commerce. Tina has expertise leading research and development for many prominent economic measurement initiatives, including several that originated from Executive Orders and the U.S. Congress. Tina conducts foundational research involving areas of the economy that are new or under transition, including artificial intelligence, health care, global value chains, domestic energy production, and the bioeconomy. Dr. Highfill holds a bachelor's degree from Virginia Tech, a master's degree from The Johns Hopkins University, and a PhD from Virginia Commonwealth University.

**David Wasshausen** is Associate Director for National Economic Accounts at the Bureau of Economic Analysis. In that role, David oversees the calculation of official economic statistics that track the performance of the U.S. economy. These include BEA's flagship economic measure, gross domestic product (GDP), as well as its major components such as consumer spending and business investment. He also oversees industry statistics, including GDP by industry, gross output, and BEA's input-output accounts. With more than 30 years of experience in national economic accounting, David has played a prominent role in BEA's efforts to provide a fuller and more timely picture of the U.S. economy. Mr. Wasshausen holds a master's degree in economics from American University and a bachelor's degree in economics from Miami University.

**Greg Prunchak** is an Economist with the Bureau of Economic Analysis (BEA) in the U.S. Department of Commerce since 2015. Greg oversees measures of private fixed investment, foreign trade, and change in private inventories within U.S. GDP. Prior to joining BEA, Greg worked on the retail and wholesale economic indicator programs at the U.S. Census Bureau measuring economic activity in those sectors. Greg has spent his career working to improve measures of economic activity in the U.S. by utilizing new and alternative collection methods, data sources, and measurement techniques. Greg holds an MA in Applied Economics from Johns Hopkins University.

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Designed by the Information Management and Services Division of the Oesterreichische Nationalbank (OeNB)

SUERF Secretariat

c/o OeNB, Otto-Wagner-Platz 3A-1090 Vienna, Austria

Phone: +43 1 40 420 7206

E-Mail: [suerf@oenb.at](mailto:suerf@oenb.at)

Website: <https://www.suerf.org/>