Total Survey Error through the Lens of Statistical Product First

Sallie Ann Keller, Ph.D., Associate Director and Chief Scientist Research and Methodology Directorate U.S. Census Bureau **September 18, 2024** The views expressed in this perspective are those of the presenter and not the Census Bureau.



Acknowledgements:

Thanks to John Eltinge, Cass Dorius, Michael Hawes, and Erika Becker-Medina for their help in developing this presentation.

Also, thanks to Census Bureau staff and researchers for their contributions that led up to this effort.



Forces Driving Transformation Today

rapidly changing new data use of increased demand technology for information sources declining new data challenges to response tools traditional rates artificial data collection intelligence methods



The Statistical Last Mile



20th Century

Federal government the dominant user

Statistical system a near monopoly

Output mostly cross-tabs

Published in books, mostly deposited in libraries, then electronically

Source data acquisition difficult and costly

Privacy and confidentiality risks were small

Computation expensive and limited



The

Statistical

Last Mile

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21st Century

Many diverse users

Many more organizations produce similar statistical products

Output varied and complex featuring visualizations and analysis tools

Data accessed electronically online or in secure enclaves

Source data more abundant, available, and less costly—but not necessarily designed for statistical use

Privacy and confidentiality risks are much greater

Computation vastly improved



The

Statistical

Last Mile

The Statistical Last Mile Challenge

How does the Census Bureau deliver the right statistical information at the right time in actionable formats to address diverse data user needs?





Flipping the Focus

Determine what information stakeholders need to reach their objectives

From there, shape the statistical products to be developed





Statistical Product First Approach

Moving from managing surveys to managing *statistics*



Statistical Product First Approach



Ensures Data Support Purposes and Uses





What are some purposes and uses?

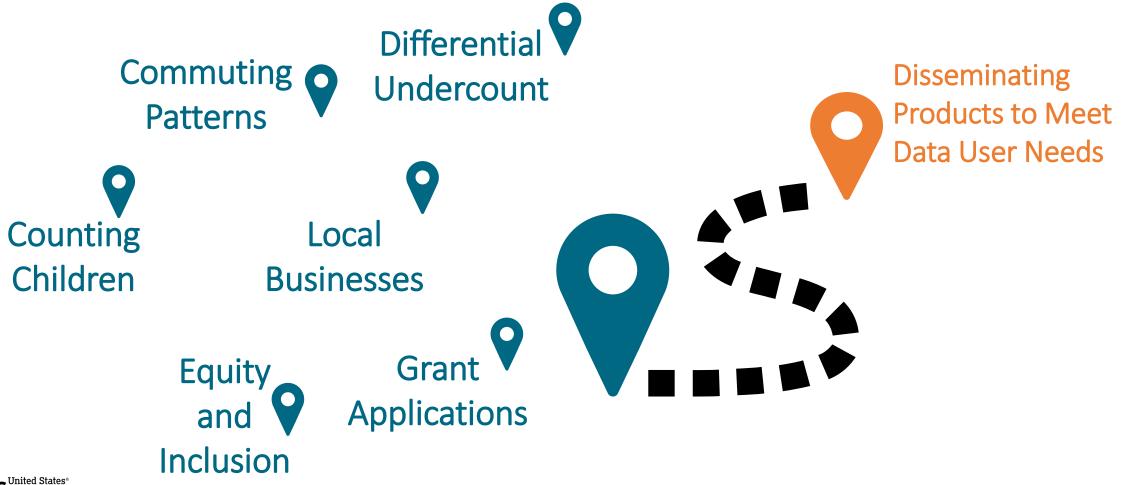




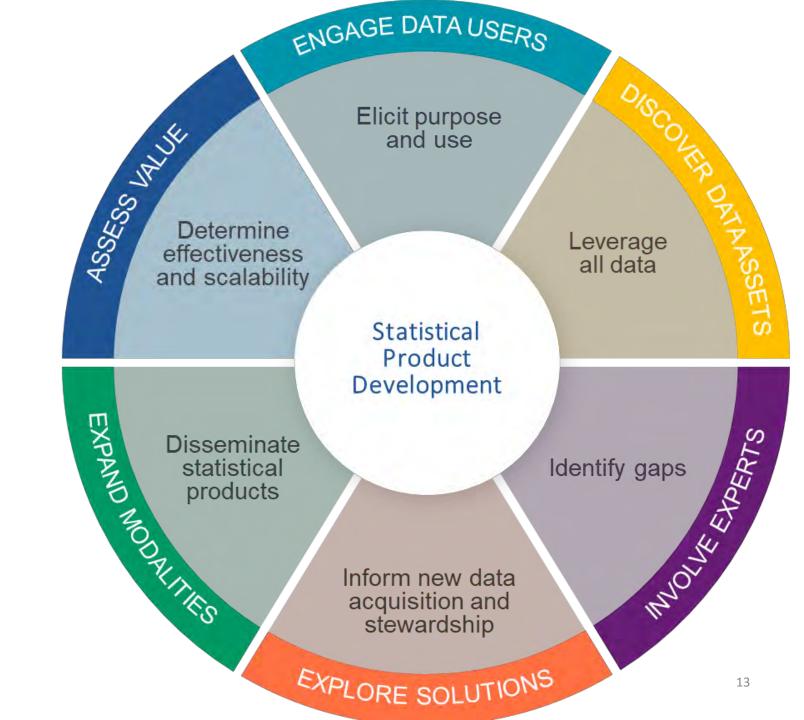


What are some purposes and uses?





Statistical Product First Innovation Cycle





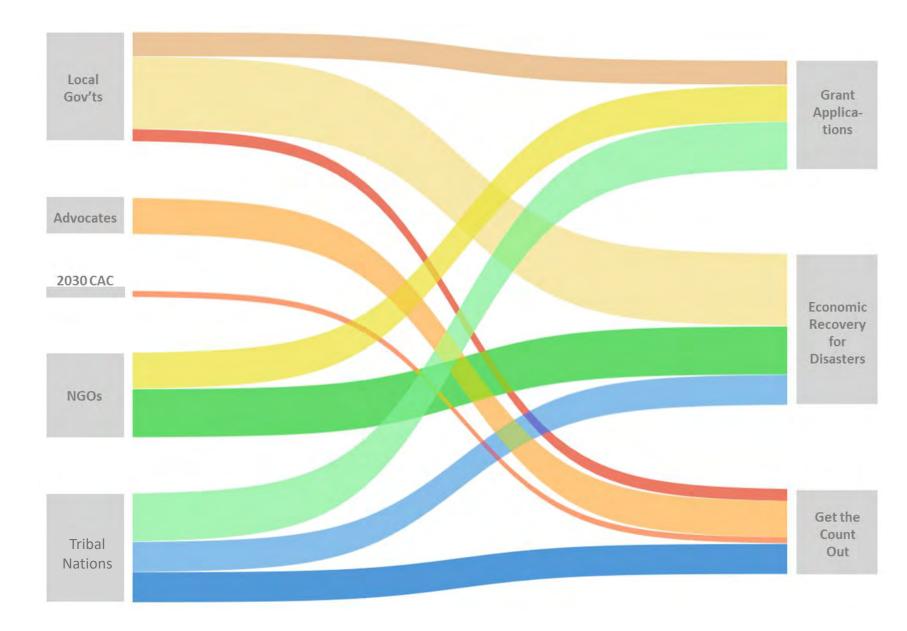
Illuminating the Ecosystem of Data Users to Elicit Purpose and Use





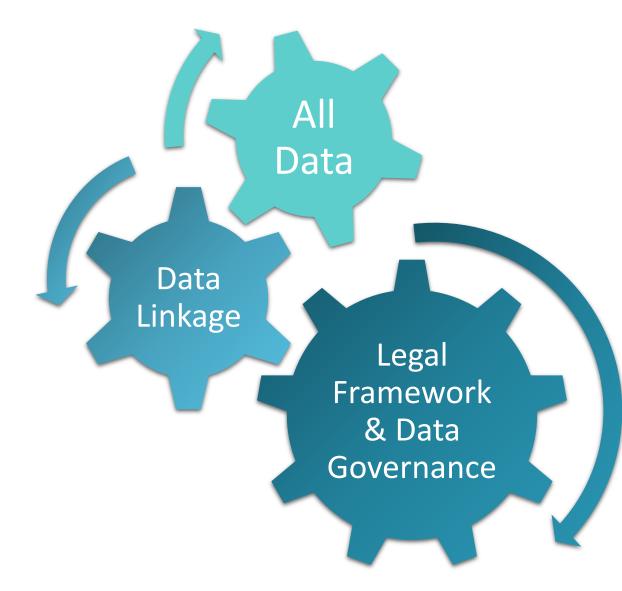
Develop User Segmentations

Currently developing a methodology to identify cross-cuts of the stakeholder group segmentation focused on common purpose and use needs.





Leverage Data Infrastructure



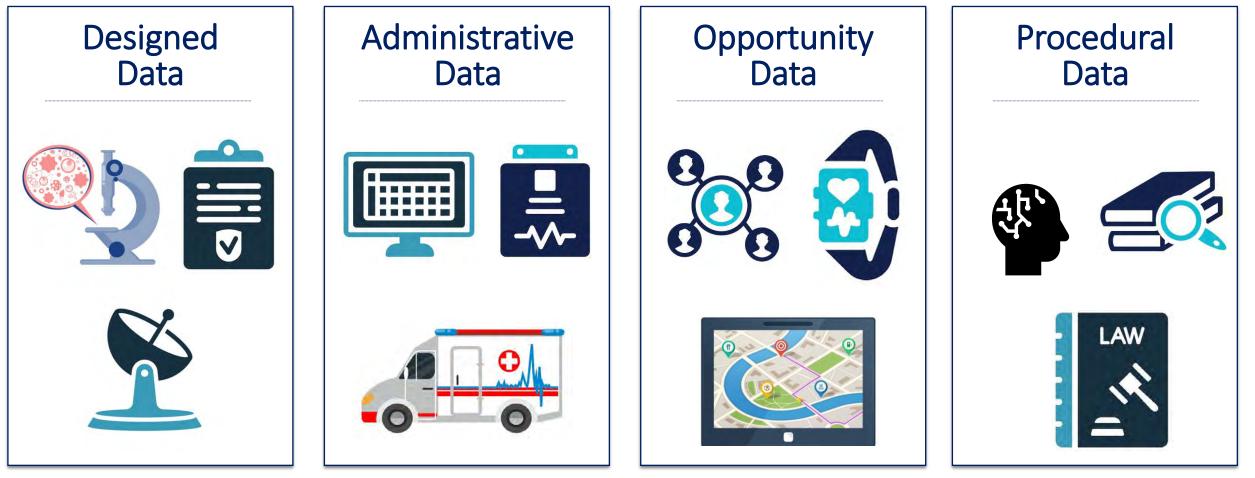


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Use ALL Data Assets

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Going beyond the survey data we collect





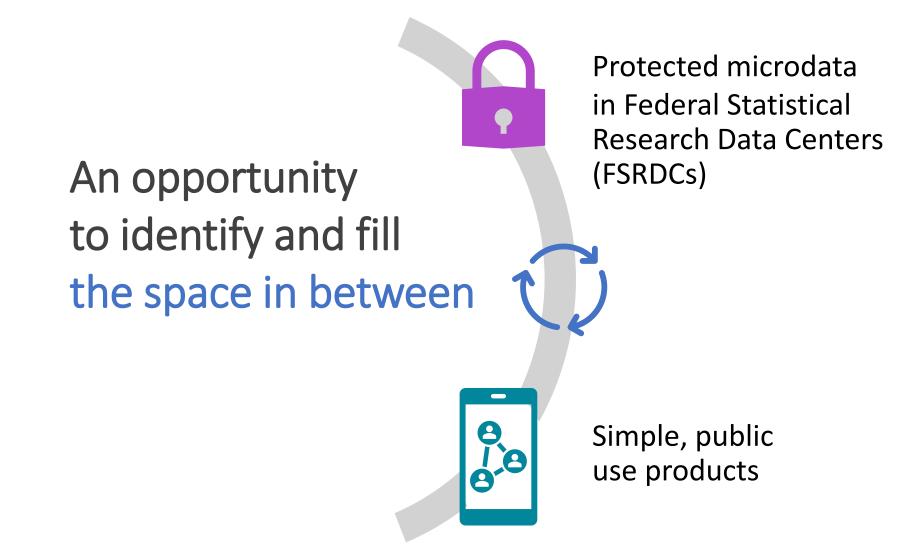


We collect a variety of linkage fields and tailor linkage methodology to each linkage type – people, places, jobs, and organizations.





Statistical Product Spectrum for ALL Users





Adhere to Legal Framework and Data Governance

Title 13 directs us to acquire and use external data records for statistical purposes.

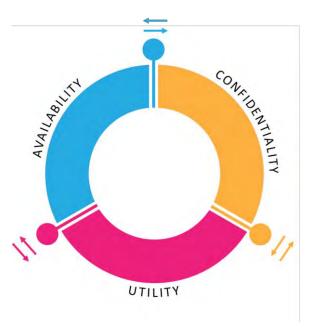
Foundations for Evidence-Based Policymaking Act promotes and encourages data sharing.

Confidential Information Protection and Statistical Efficiency Act sets forth functional separation of statistical versus administrative uses of data.

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Ethical Data Stewardship

Responsible development and dissemination of Census Bureau products requires confidentially protection of data subjects' information while exclusively using this information for statistical purposes.





Enabling Technologies

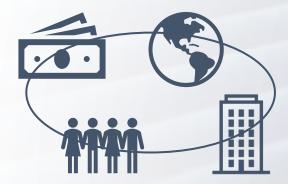
Data Ingest and Collection for the Enterprise (all data assets)



Enterprise Data Lake



Enterprise Linked Frames



Enterprise Dissemination Services





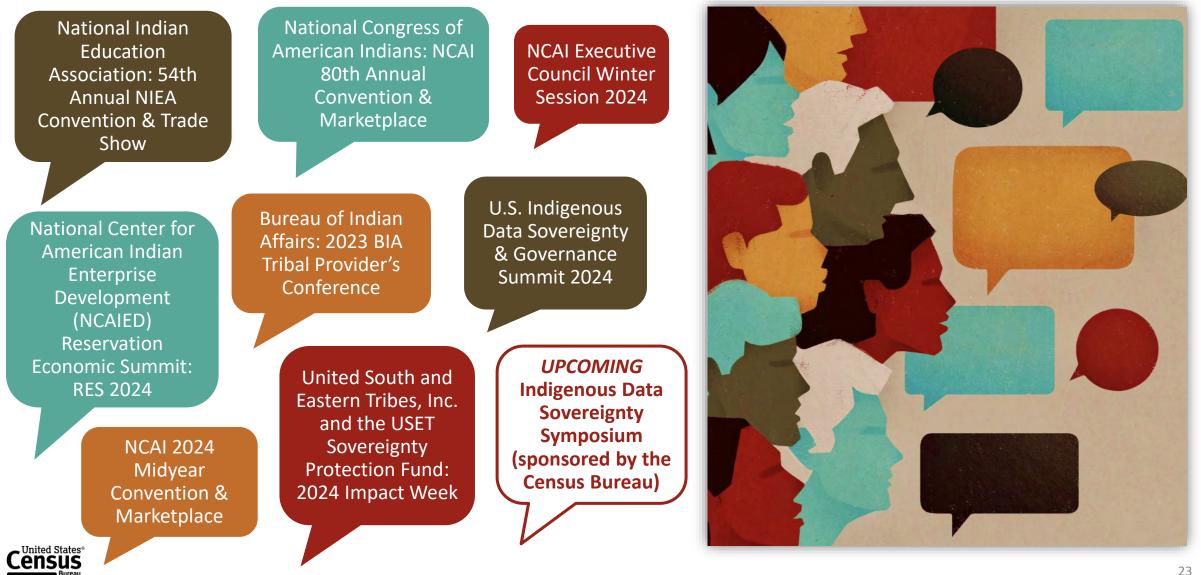


Statistical Product First

Selected Demonstrations



Engagement with Indigenous Communities



Collaboration through Engagement



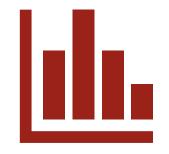
Improving Access to Tribal Data

Repackaging statistics across a variety of tribal and native boundaries through an infographic decision tree of current Census products. Collaborating on the dissemination with "My Tribal Area".

Facilitating Administrative Records

Working with AD REC pilot project to help uncover tribal incentives to encourage and facilitate the provisioning of data to the Census Bureau.



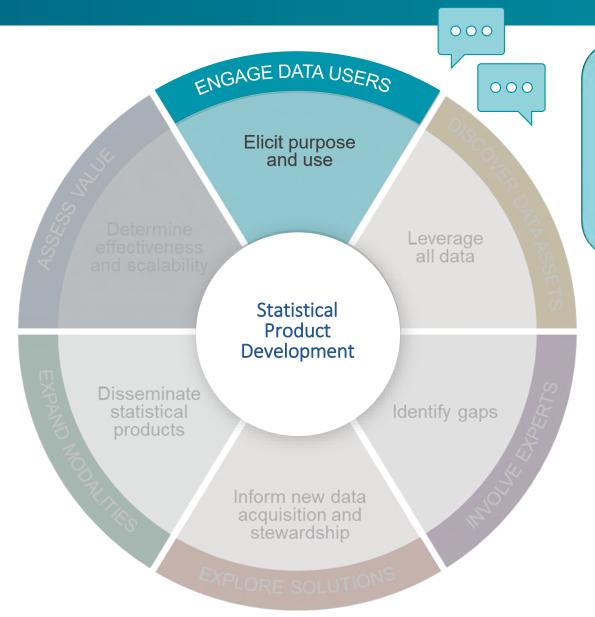


Exploring GDP for Tribal Areas

Exploring user feedback on the creation and value of GDP-like measures for tribal regions.



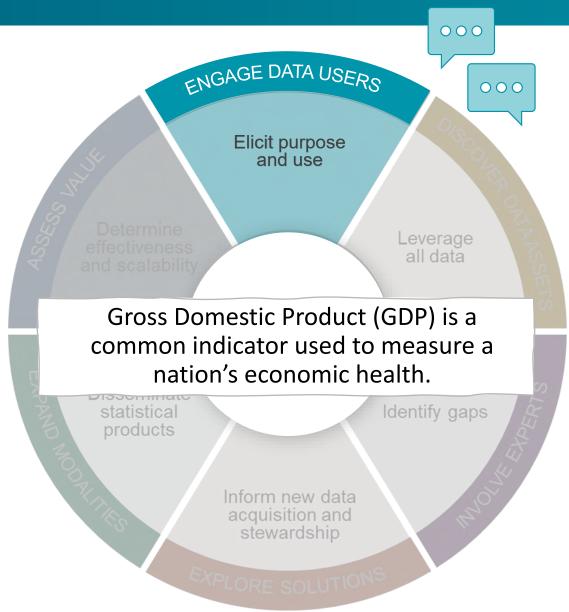




Data user engagements led to powerful purpose and use needs:

We need a GDP-like product to measure the economic health of tribal regions.



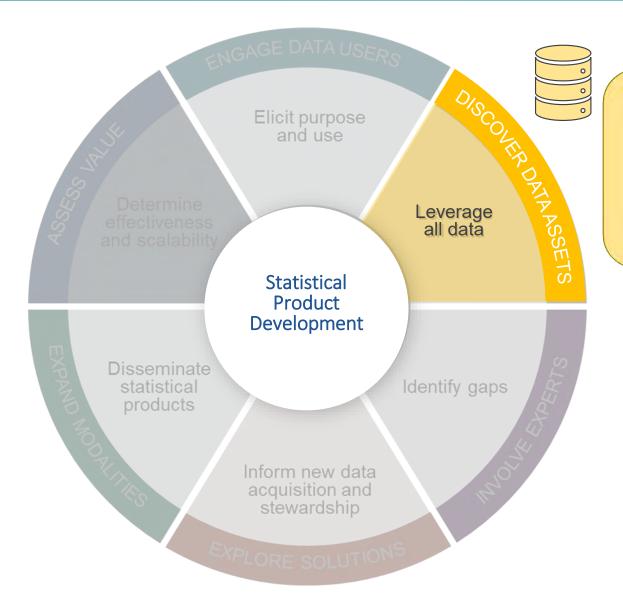


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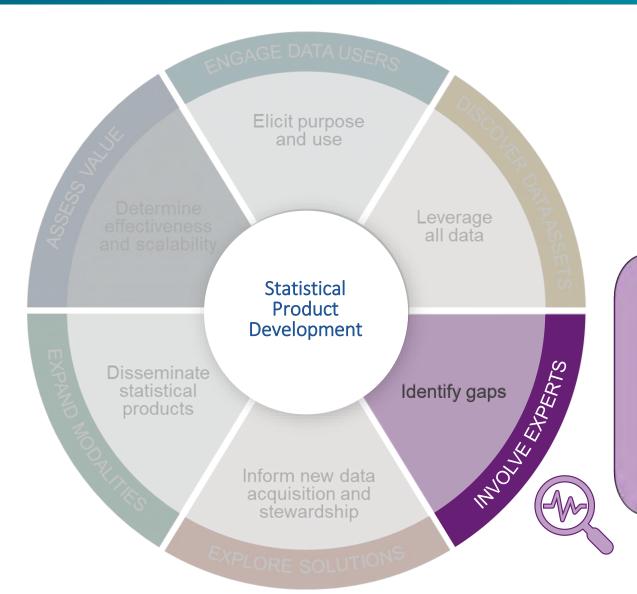




The Census Bureau collects vast amounts of economic data and then provides the estimates to the Bureau of Economic Analysis, where official GDP measures are produced.



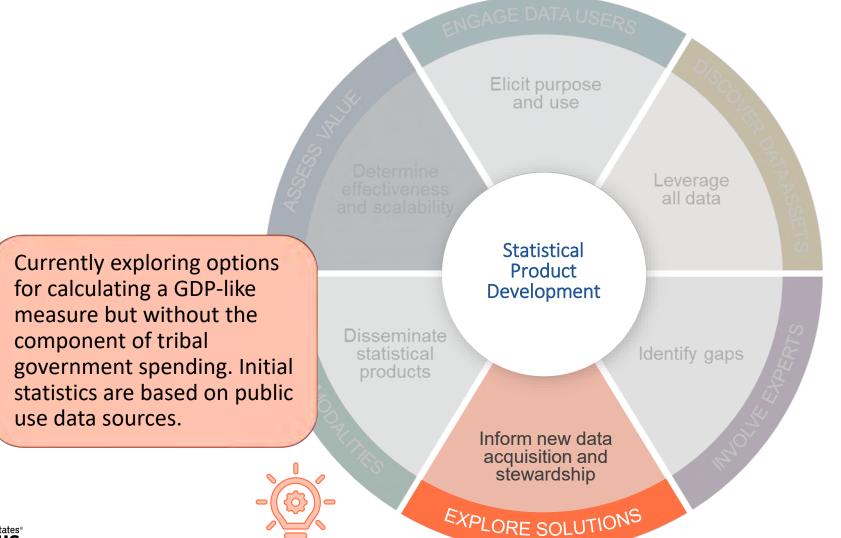
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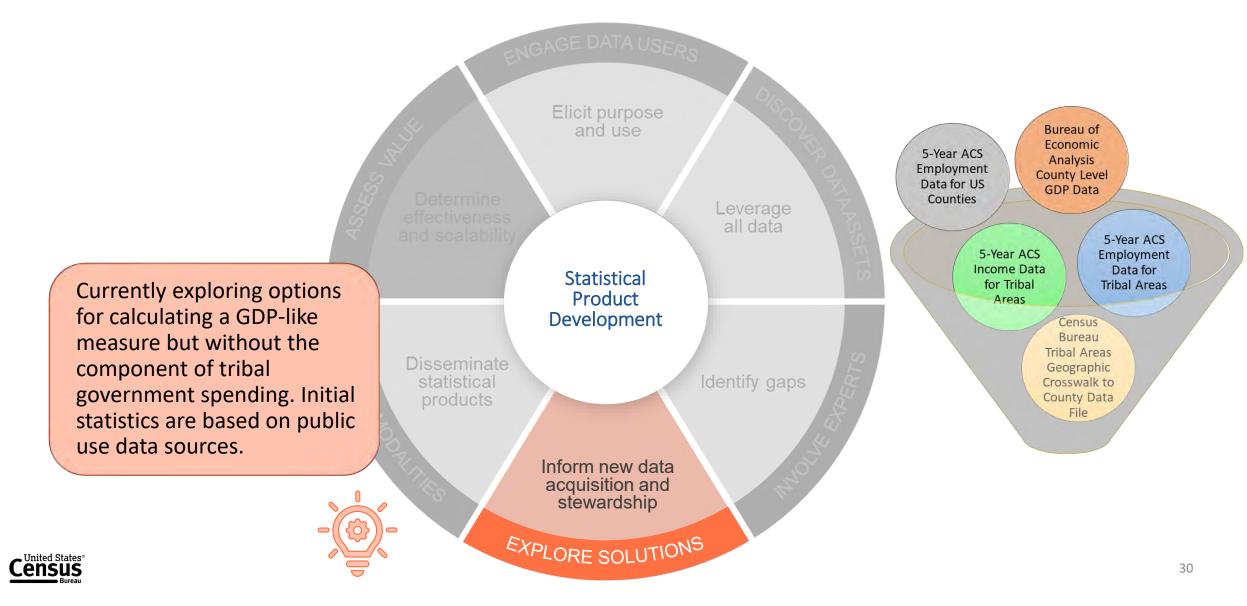
The GDP formula includes government spending. The Census Bureau collects this information as part of its Census of Governments. However, tribal nations are not included in that data collection.

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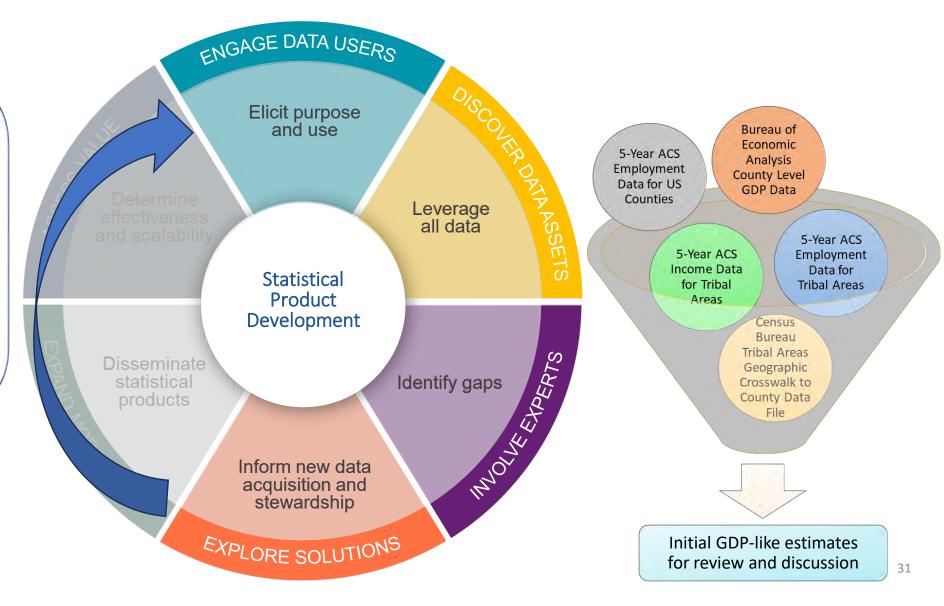


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Using early estimates to ground more engagements and discussions around purpose and use needs. These will inform more data discovery of confidential data assets and the refinement of new tribal GDP-like statistical products.





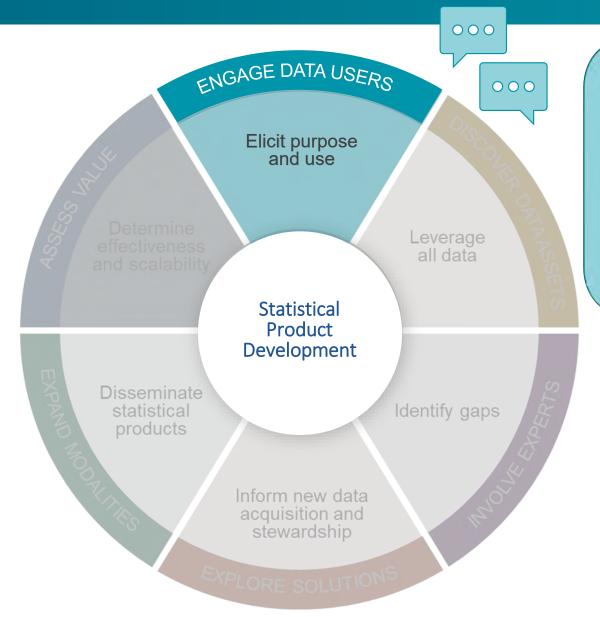
Seeking to support the needs of grant applicants

State and local governments, non-profits, and tribal organization members (SLaNT) have all shared the need for easier access to Census data to support funding applications on Grants.gov.





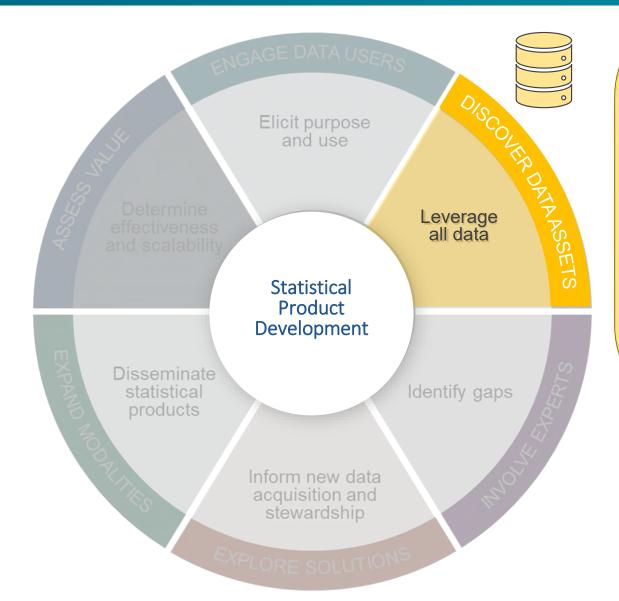




We compiled information from listening sessions to create four personas that reflected stakeholders' needs and could be used to help support the goals of state and local governments, nonprofits, and tribal organization members (SLaNT).





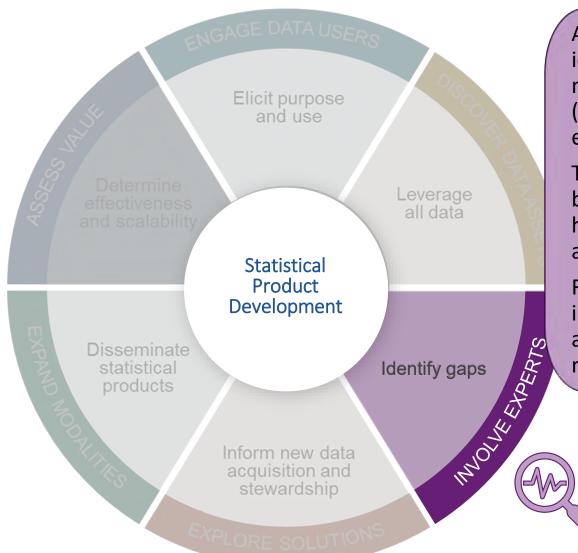


We analyzed all grants currently available on Grants.gov and identified the frequency and availability of funding opportunities by data domain.

Then, we identified potential Census data sources that support applications for the known funding opportunities.







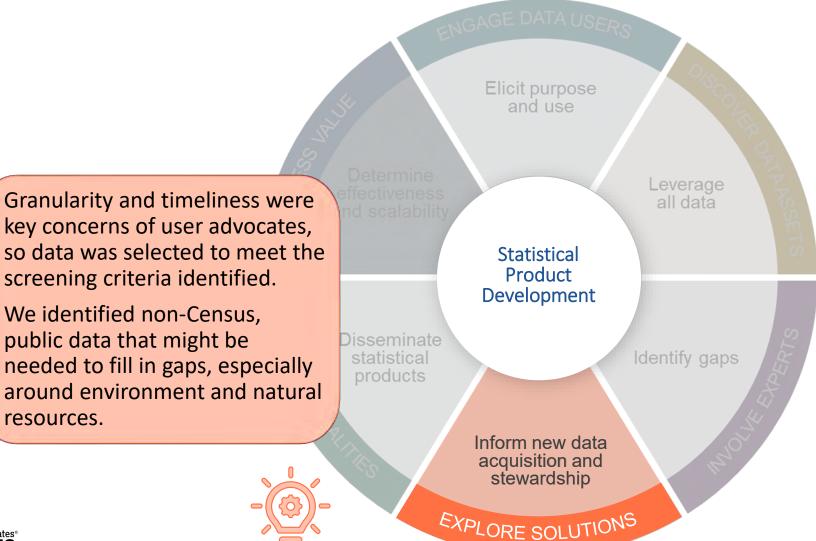
A criteria matrix was used to identify which data domains matter most to our stakeholders (e.g., health and nutrition, education, transportation, etc.).

Teams identified a list of 100 basic data elements that would help diverse SLaNT data users apply for grants.

Results were shared back with internal experts and user advocates to identify what was missing.



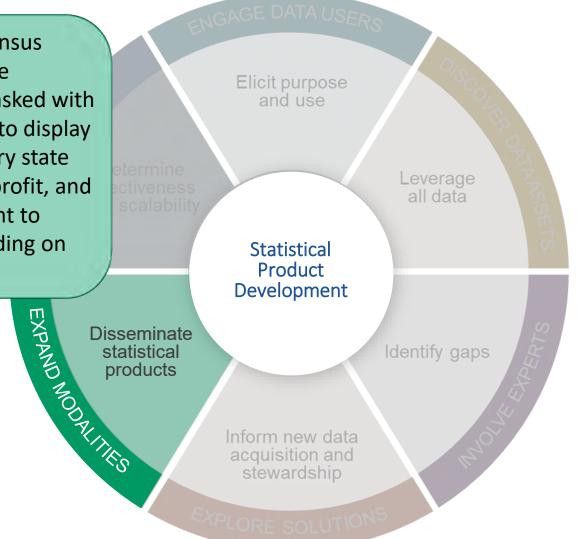
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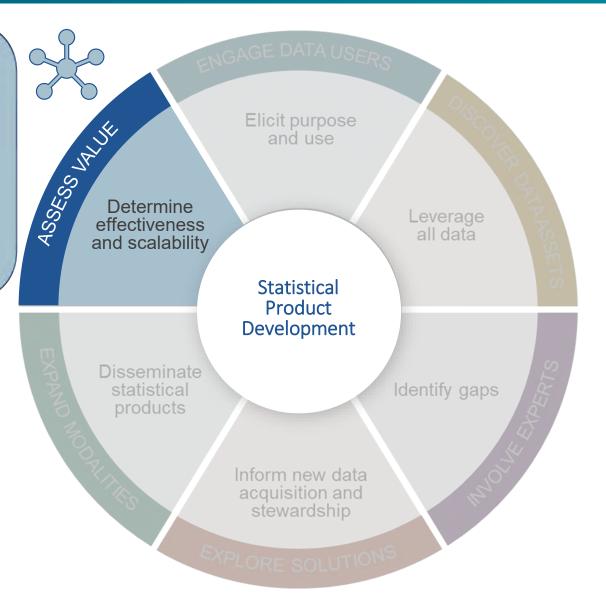
Capstone teams from the Census Bureau, Georgetown, and the University of Virginia were tasked with developing a prototype tool to display the 'Top 100' data items every state and local government, non-profit, and tribal organization might want to know when applying for funding on Grants.gov.





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Plans are underway to share the 'Top 100' items and tool prototype with stakeholders and user advocates to identify what works and what doesn't before this tool goes through a second innovation cycle on its way to becoming a public facing statistical product.

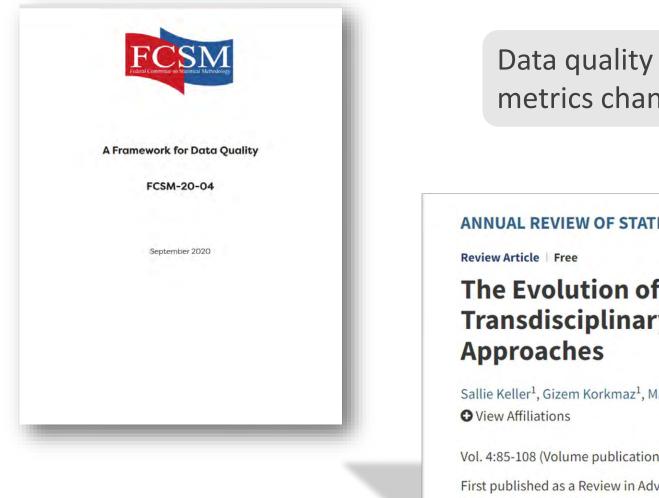




How do we ensure quality specifically, data **and** statistical product quality—is captured throughout the SPF Innovation Cycle?



Reflecting on the Evolution of Data Quality



© Annual Reviews



Data quality is a relative concept, consequently metrics change depending on use.

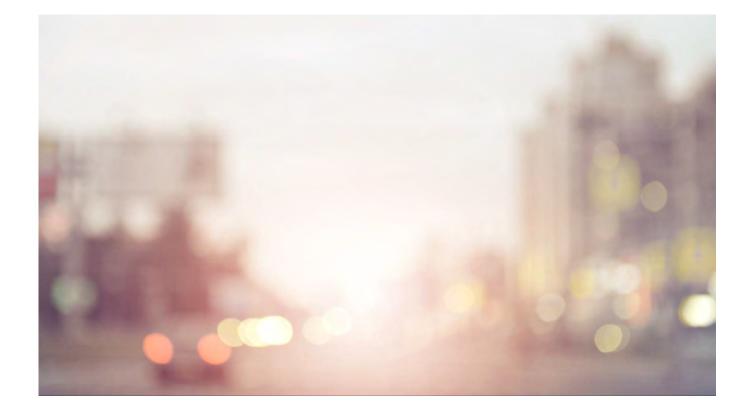
ANNUAL REVIEW OF STATISTICS AND ITS APPLICATION Volume 4, 2017

The Evolution of Data Quality: Understanding the **Transdisciplinary Origins of Data Quality Concepts and**

Sallie Keller¹, Gizem Korkmaz¹, Mark Orr¹, Aaron Schroeder¹, and Stephanie Shipp¹

Vol. 4:85-108 (Volume publication date March 2017) https://doi.org/10.1146/annurev-statistics-060116-054114

First published as a Review in Advance on January 06, 2017



"Far better an approximate answer to the right questions, which is often vague, than an exact answer to the wrong questions, which can always be made precise."

~ John Tukey



Reflecting on the Evolution of Data Quality

Statistics—as the science of uncertainty—has been central to data quality across all fields of endeavor



Historically, data quality was addressed by controlling measurement and data collection processes and through data ownership Deming (1940s) focused on measuring inputs and processes using statistical process control to minimize product inspections after the product was built—Total Quality Management

Tukey (1960s) introduced exploratory data analysis

1980s and 1990s: Total **Data** Quality Management (TDQM) was also introduced

Survey research has focused on data quality from the perspective of data collection process, starting with the research questions and level of acceptable variability in the findings

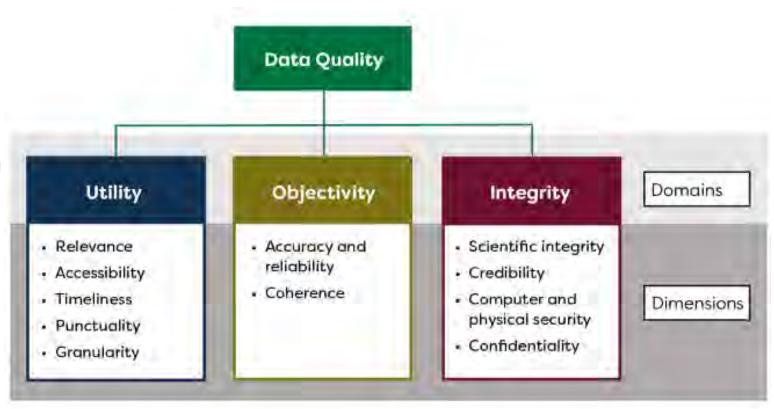
By 2004, Total Survey Error framework introduced and put into practice (Groves, Fowler, Couper, Lepkowski, Singer, Tourangeau)

Data Quality Today

- Traditionally data collection—including data discovery processes—is designed to minimize bias and maximize information content and verify quality of data after it is collected.
- This tradition is being challenged as we consider diverse uses, iterative processes, and building statistical products based on blended data through massive data repurposing.

2020 FCSM took on the challenge and proposed a data quality framework to support statistical products.

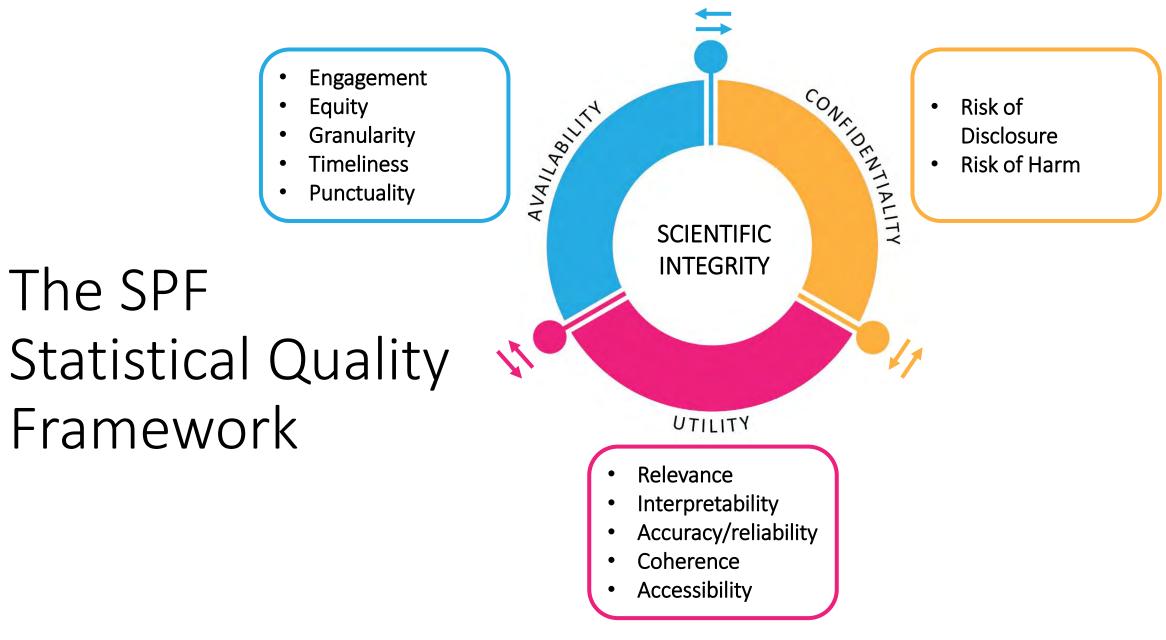






Building on the FCSM Framework to Operationalize a Statistical Quality Framework for the SPF Innovation Cycle

- Not a linear process.
- Simultaneously consider stakeholder engagement, data discovery, statistical methods, product design, while factoring in the needs of confidentiality.
- Scientific integrity—integral to the statistical quality framework—must be maintained throughout the SPF innovation cycle.

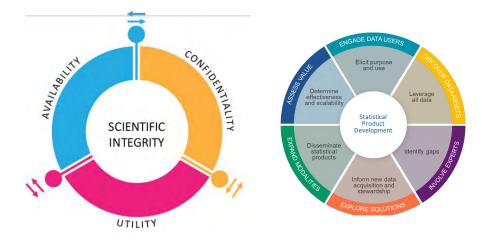


| | UTILITY | RELEVANCE | What purposes and uses do Census statistics (i.e., statistical products) need to support? |
|----|---------|--------------------------|---------------------------------------------------------------------------------------------------------------------------------------------------------|
| | | INTERPRETABILITY | Will the statistics be readily understandable and interpretable by the intended data users? |
| | | ACCURACY/ RELIABILITY | Will the statistics be of sufficient accuracy and reliability to support valid statistical inference and decision-making for their intended uses? |
| Q. | | COHERENCE | Will the statistics align sufficiently with other relevant statistical assets to permit broader purposes and uses? |
| Ċ | | ACCESSIBILITY | Can data users easily obtain the statistical products and documentation in forms and formats that are understandable to them? |

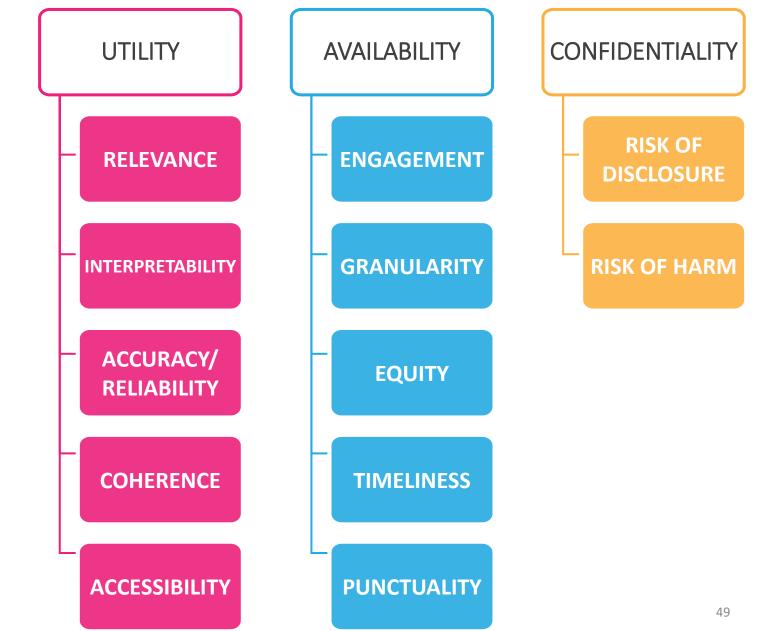
| AVAILABILITY | ENGAGEMENT | How can we iteratively engage with stakeholders to understand their purpose and use needs and verify the products developed meet their needs? |
|------------------|-------------|-----------------------------------------------------------------------------------------------------------------------------------------------------|
| | GRANULARITY | Are the statistics sufficiently disaggregated (by subgroups, geographies, and characteristics) to support purpose and use needs? |
| | EQUITY | How well will the statistics support important uses by diverse groups of data users—whether demographic or various levels of data acumen? |
| | TIMELINESS | How current do the statistics need to be to support their intended uses? |
| | PUNCTUALITY | Are user expectations, through the development and dissemination of statistical products, being managed appropriately? |
| Census Bureau | | 47 |

| CONFIDENTIALITY | | |
|------------------|-----------------------|------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| | RISK OF DISCLOSURE | What is the likelihood the statistics could lead to a disclosure that would reveal (or substantially improve inferences about) the identity or characteristics of a data subject? How reliable (certain) would those inference be? |
| | | How strong is the confidentiality guarantee being given to data subjects? |
| | RISK OF HARM | What is the risk of harm to a data subject were a disclosure to occur? How damaging might those harms be? |
| \dot{O} | | What impact might this have on the agency's reputation? |
| Census Bureau | | 48 |

The Statistical Quality Framework—and all its dimensions—guides development

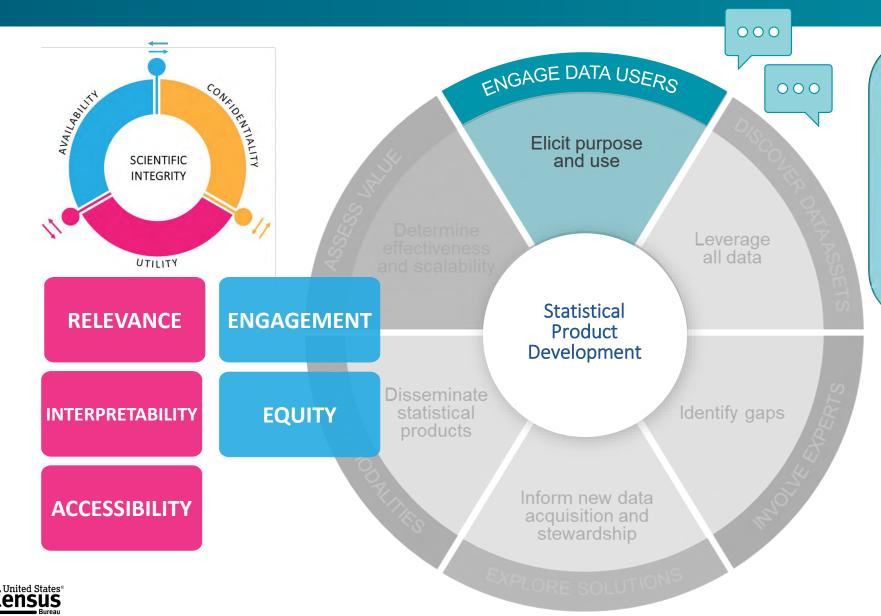


- Scientific integrity—integral to the statistical quality framework undergirds the SPF innovation cycle.
- These 12 dimensions take on different nuances depending on where we are in the cycle.
- The ability to quantitatively vs. qualitatively assess development will also differ depending on where we are in the SPF innovation cycle.



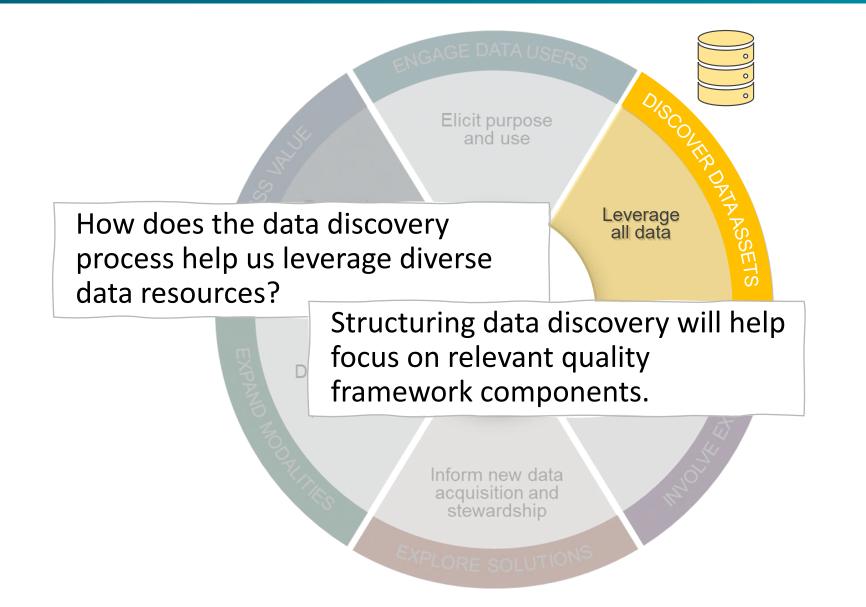


Exemplar: Grants.gov



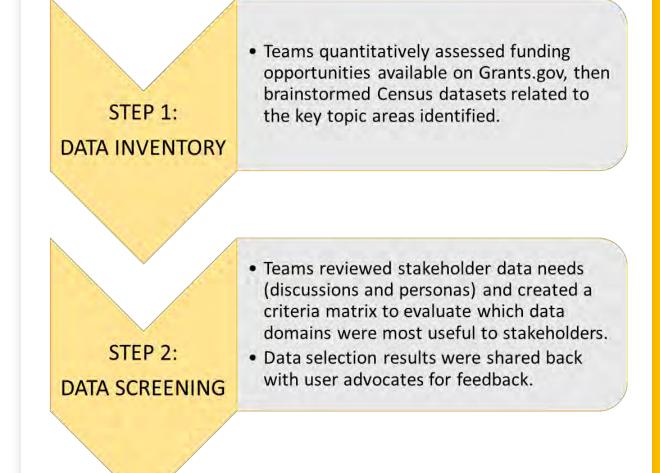
We compiled information from listening sessions to create four personas that reflected stakeholders' needs and could be used to help support the goals of state and local governments, nonprofits, and tribal organization members (SLaNT).

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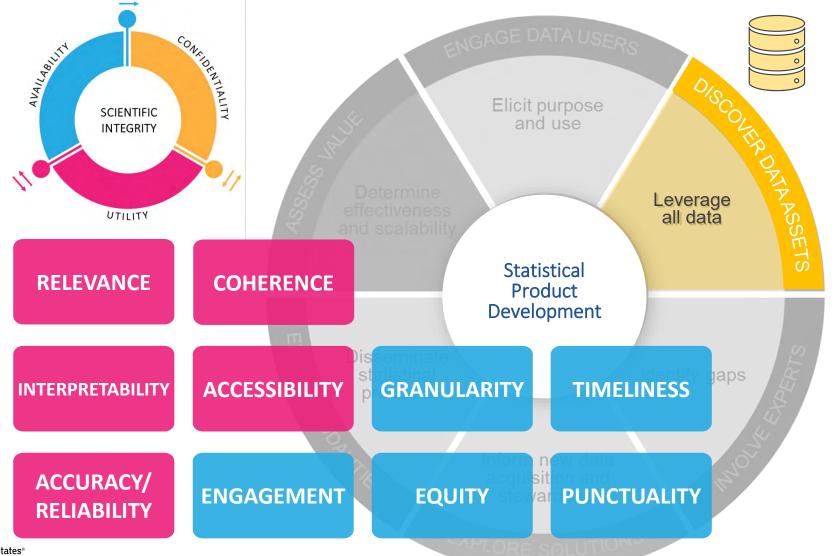


How can we help organizations and applicants better utilize Census data when applying for funding via Grants.gov?









We analyzed all grants currently available on Grants.gov and identified the frequency and availability of funding opportunities by data domain.

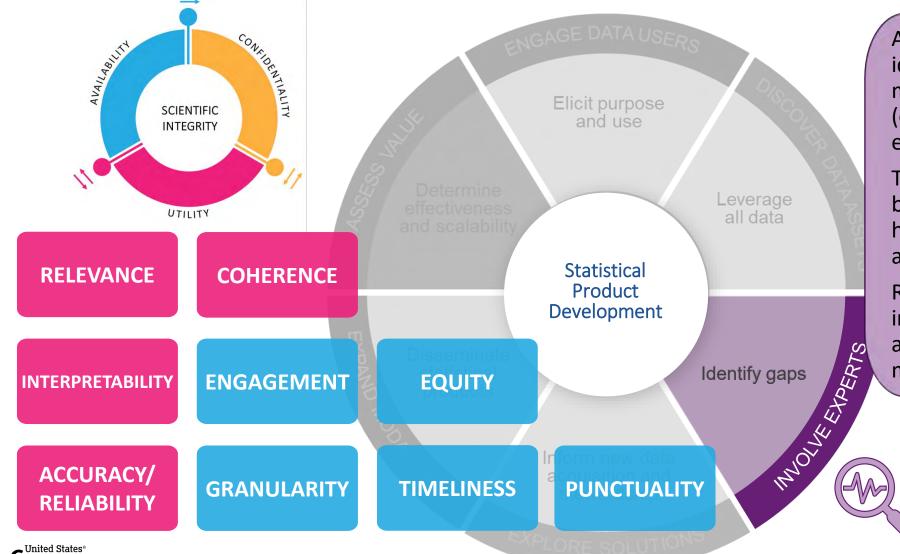
Then, we identified potential Census data sources that support applications for the known funding opportunities.

| Data | USER PERSONAS | INTERESTS | FUNDING CATEGORIES | | | | |
|-----------------------------------------|--------------------------------------------------------------------------------|------------------------------------------------------------------------------------------------------------------------------------------------------|-------------------------------------------------------------------------------------------------------------------------------------------------------------------|--|--|--|--|
| Data Screening Criteria Matrix | Divya—a state public health official in Missouri | Education, Human Rights, Human Services, Transportation, Workforce Development | Education, Humanities, Transportation, Employment, Labor and Training, Income Security Social Services, Law Justice and Legal Services | | | | |
| | Marcus—a city manager in Maine | Public Safety, Transportation, Parks and Recreation, Social Services, Job Creation, Business Growth, and Prosperity within his community | Transportation, Community Development, Humanities, Employment, Labor and Training, Business and Commerce | | | | |
| | Sarah—a grant writer for a local community action agency in Pennsylvania | Poverty, Homelessness, Food Insecurity, Heating in Winter, and Other Social Challenges | Humanities, Food and Nutrition, Natural Resources, Income Security Social Services, Housing, Law Justice and Legal Services, Community Development | | | | |
| United States | Charles—a community organizer in California | Community | Community Development | | | | |



| | USER PERSO | NAS | INTERESTS | FUNDING CATEGORIES | | | | |
|-----------------------------------------|-------------------------------|-------------------------------------------------|----------------------------------------------------------------------------------------------|-------------------------------------------------------------------------------------------------------------------------------------------------------------------|--|--|--|--|
| Data Screening Criteria Matrix | Divya—a sta official in Mi | te public health ssouri | Education, Human Rights, Human Services, Transportation, Workforce Development | Education, Humanities, Transportation, Employment, Labor and Training, Income Security Social Services, Law Justice and Legal Services | | | | |
| | Marcus—a c Maine | ity manager in | Public Safety, Transportation, Parks and Recreation, Social | Transportation, Community Development, Humanities, | | | | |
| | | | sts influenced the data measurement selectio | | | | | |
| | - | int writer for a local action agency in a | Poverty, Homelessness, Food Insecurity, Heating in Winter, and Other Social Challenges | Humanities, Food and Nutrition, Natural Resources, Income Security Social Services, Housing, Law Justice and Legal Services, Community Development | | | | |
| | Charles—a c in California | ommunity organizer | Community | Community Development | | | | |
| Census Bureau | | | | 55 | | | | |





A criteria matrix was used to identify which data domains matter most to our stakeholders (e.g., health and nutrition, education, transportation, etc.).

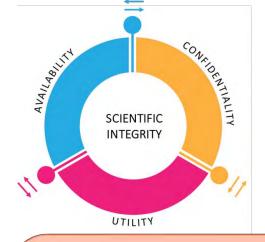
Teams identified a list of 100 basic data elements that would help diverse SLaNT data users apply for grants.

Results were shared back with internal experts and user advocates to identify what was missing.

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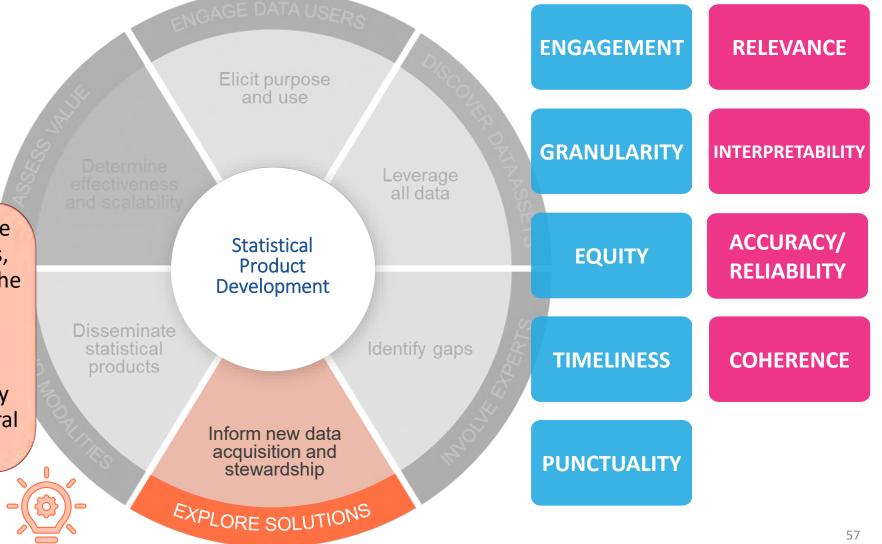


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Granularity and timeliness were key concerns of user advocates, so data was selected to meet the screening criteria identified.

We identified non-Census, public data that might be needed to fill in gaps, especially around environment and natural resources.





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Capstone teams from the Census Bureau, Georgetown, and the University of Virginia were tasked with developing a prototype tool to display the 'Top 100' data items every state and local government, non-profit, and tribal organization might want to know when applying for funding on Grants.gov.

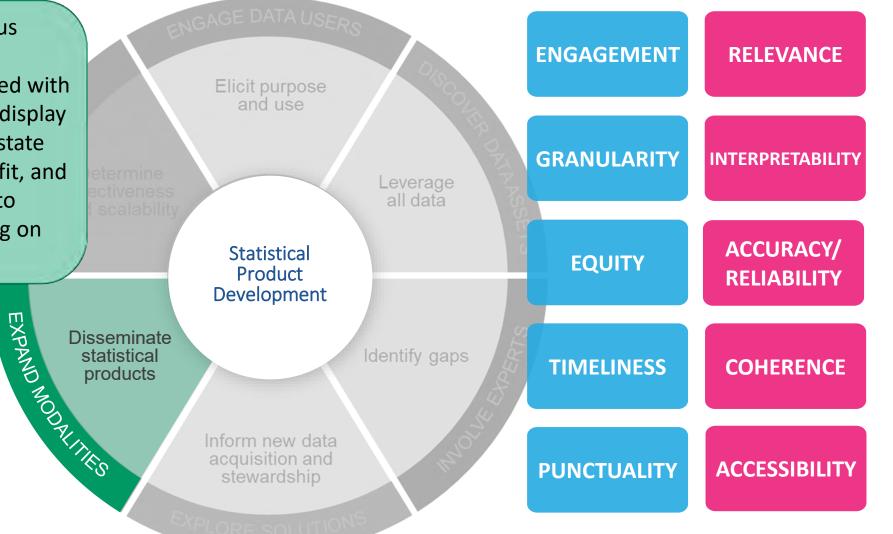
CONFIDENTIALITY

SCIENTIFIC INTEGRITY

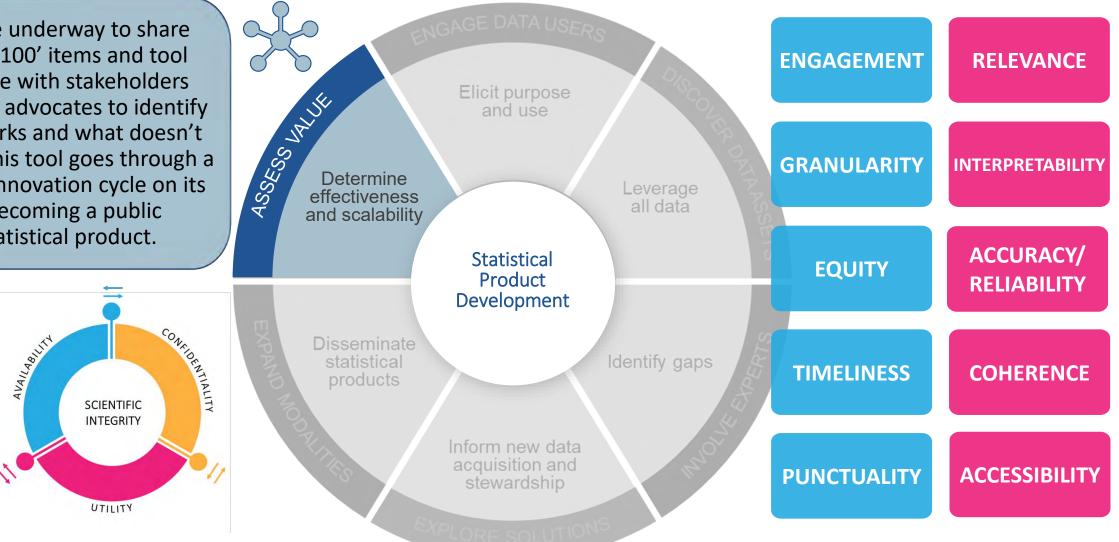
UTILITY

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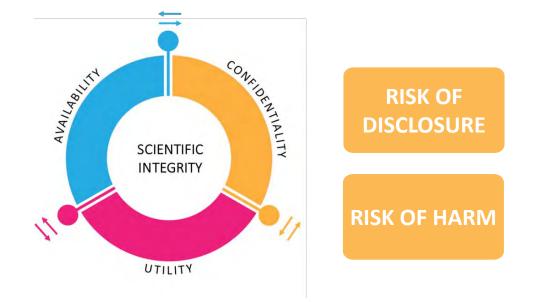
Plans are underway to share the 'Top 100' items and tool prototype with stakeholders and user advocates to identify what works and what doesn't before this tool goes through a second innovation cycle on its way to becoming a public facing statistical product.





What about confidentiality?

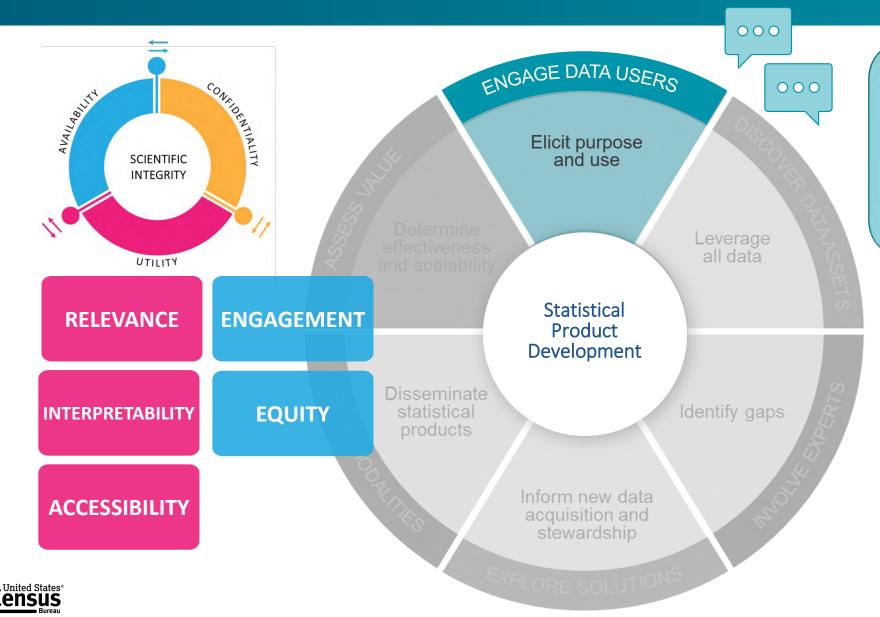




If new statistical products are a repurposing of public use, already disseminated statistics, there is no need to incorporate new decisions on confidentiality.

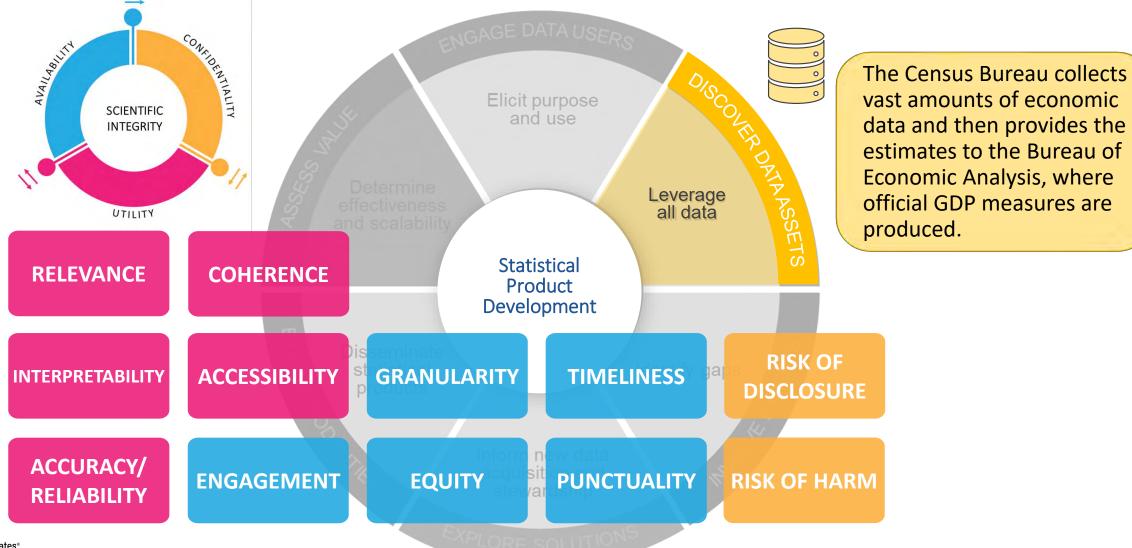
As we get user feedback and identify more gaps, new data acquisition choices and uses may introduce a new focus on confidentiality.





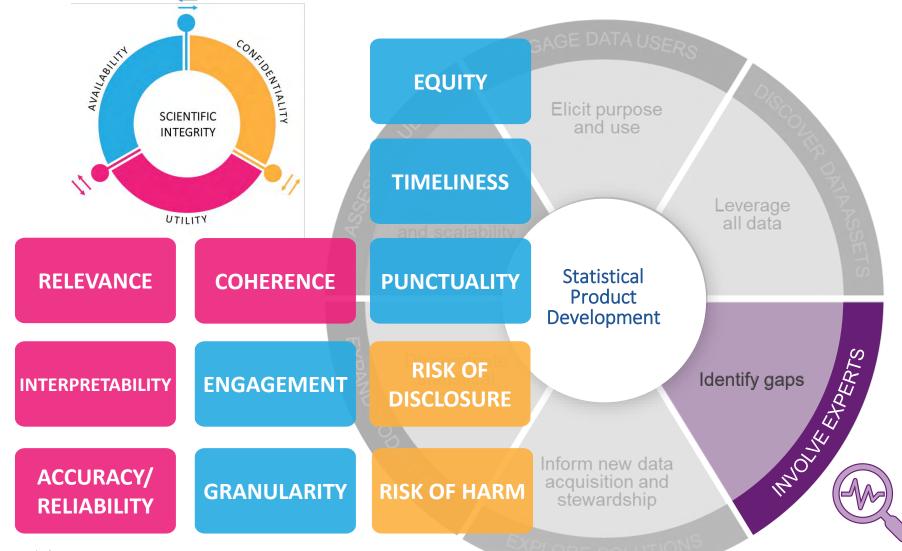
Data user engagements led to powerful purpose and use needs:

We need a GDP-like product to measure the economic health of tribal regions.





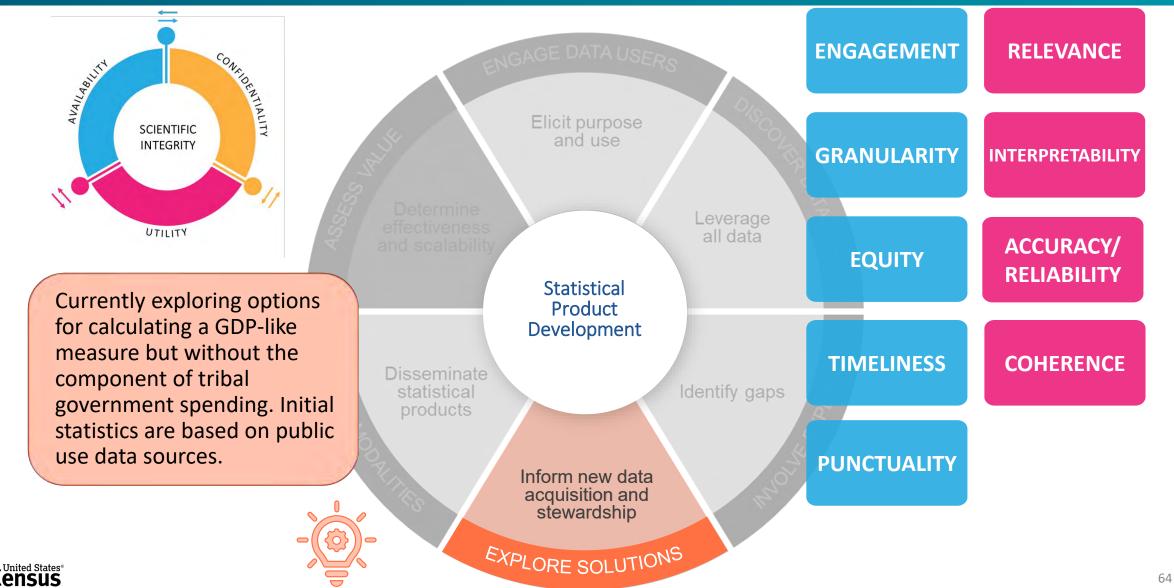
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The GDP formula includes government spending. The Census Bureau collects this information as part of its Census of Governments. However, tribal nations are not included in that data collection.



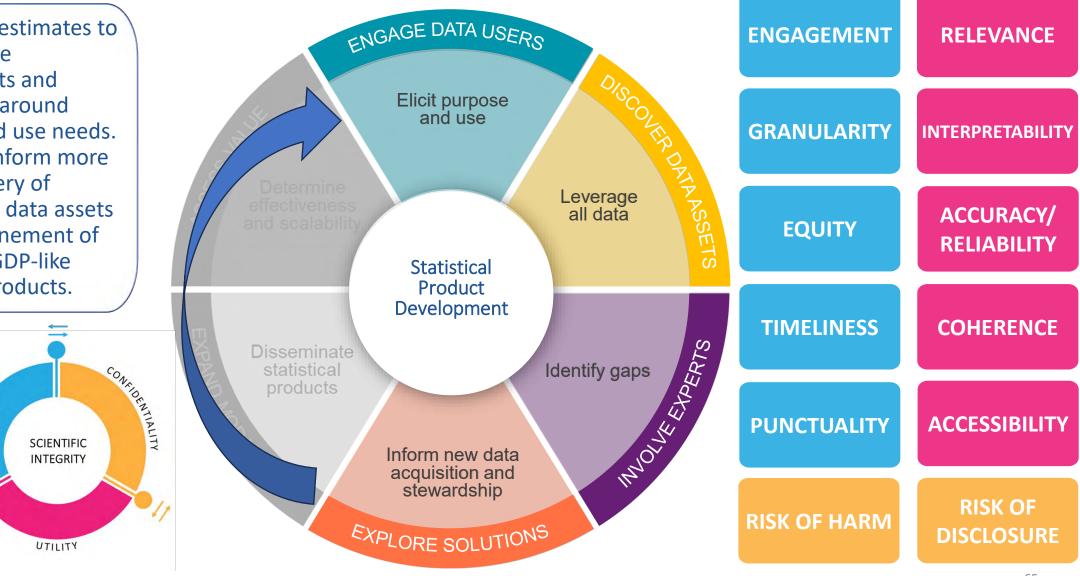




Using early estimates to ground more engagements and discussions around purpose and use needs. These will inform more data discovery of confidential data assets and the refinement of new tribal GDP-like statistical products.

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Imagine the Art of the Possible



Thank You

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