Big Data for 21st Century Economic Statistics

Katharine G. Abraham, University of Maryland ESCoE Conference on Economic Measurement May 12, 2021



Infrastructure and methods for U.S. economic statistics developed in years after WWII

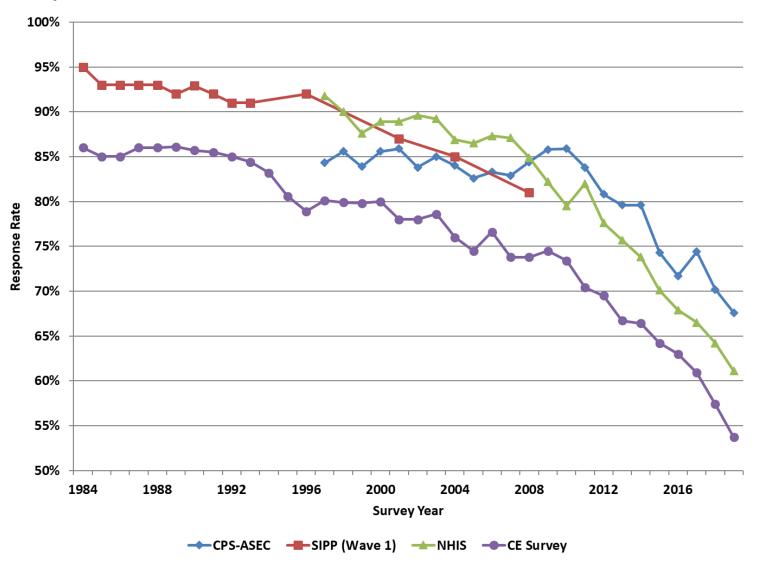
- For decades, surveys based on probability samples have provided reliable estimates at lower cost than complete enumerations
 - Surveys underlie estimates of employment, unemployment, earnings, labor turnover, job openings, production, sales, prices, ...
 - Samples designed to represent the population of interest
 - Questionnaires designed to collect desired information
- Periodic censuses and administrative data used to benchmark
- Tasks allocated across several statistical agencies
 - Bureau of Labor Statistics (BLS), U.S. Census Bureau and Bureau of Economic Analysis (BEA) primary responsibility for economic data



Increasing difficulty of obtaining survey responses

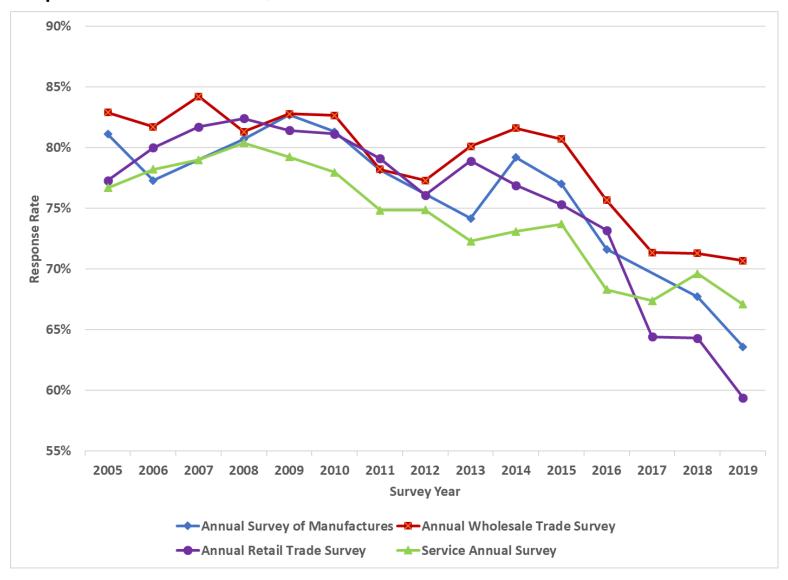


Unit response rates, selected household surveys



Source: Meyer, Mok and Sullivan (2015), adapted and updated

Unit response rates, selected annual business surveys



Source: U.S. Census Bureau

- Increasing difficulty of obtaining survey responses
- Increasing demand for more timely data



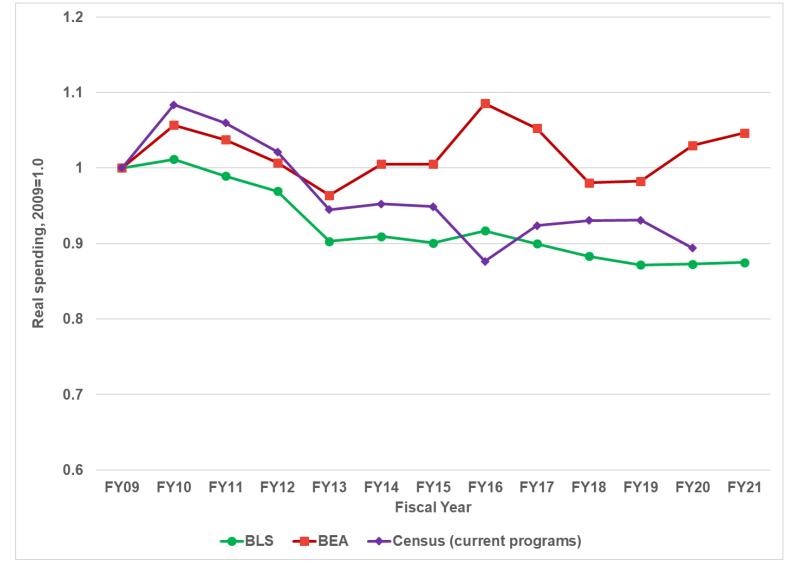
- Increasing difficulty of obtaining survey responses
- Increasing demand for more timely data
- Increasing demand for more disaggregated data



- Increasing difficulty of obtaining survey responses
- Increasing demand for more timely data
- Increasing demand for more disaggregated data
- Stagnant or declining agency budgets



Economic statistics agency real funding trends (2009=1)



Source: American Statistical Association; Statistical Programs of the U.S. Government, various years

Big data to the rescue?

- Natively digital data have proliferated in recent years
- Agencies considerable experience with administrative big data
- New frontier: Using big data from private sources in the production of economic statistics to:
 - Reduce respondent burden
 - Increase timeliness and/or reduce revisions of published data
 - Increase the granularity of published data
 - Lower statistical agency costs?



Wealth of private big data

- Scanner data from retail outlets
- Prices, product characteristics and other information on the Web
- Credit card transactions data (e.g., JP Morgan Chase data, Spending Pulse MasterCard data)
- Payroll processing and scheduling data
- Sensor data (e.g., satellite imaging, traffic cameras)
- GPS tracking data (e.g., tractors, trucks)



Considerations in incorporating naturally occurring data into official statistics

Survey data

- Small but representative share of target population observed directly
- Data elements selected to meet statistical needs
- Quality control central to survey process, though errors in measurement may arise

Naturally occurring data

- Large but not necessarily representative convenience samples
- Data elements reflect needs and constraints of business processes
- Data elements relevant to business processes most likely to be accurate



Considerations in incorporating naturally occurring data into official statistics (continued)

Survey data

- Comparability of data over time controlled by survey statistician
- Data records designed for statistical analysis; typically well documented
- Data "owned" by statistical agency, typically collected from respondents under a pledge of confidentiality

Naturally occurring data

- Comparability of data over time may be disrupted by changes in business requirements
- Data records reflect business purposes; may or may not be well documented
- Data "owned" by business where it was generated; obtaining data may be expensive or raise legal, business or other concerns (including concerns about relying on a monopoly provider)



Considerations in incorporating naturally occurring data into official statistics (continued)

Survey data

 Agencies' physical and human infrastructure developed for collection and processing of survey data

Naturally occurring data

 Naturally occurring data sets require enhancements to computing capacity and additional staff skills



Criteria for deciding when to adopt big data for official statistics

- Collecting data using current methods has become difficult or is proving inadequate to meet users' demands
- Data a good fit for the intended purpose
- Quality of estimates of similar or better quality
- Costs are lower or added cost can be justified based on improvements to estimates
- Risk of relying on 3rd-party data suppliers can be mitigated



BLS: Using big data to improve the Consumer Price Index (CPI)

- CPI price data collected by surveying businesses and rental units
 - Commodities and Services Survey: ~94,000 prices per month
 - Housing Survey: ~8,000 rental housing unit quotes per month
 - Majority of data collected by personal visit
- Program underway to substitute data from alternative sources where feasible and cost-effective
 - Similar programs underway at ONS and Statistics Canada



CPI Alternative Data Pipeline

April 2021

Beginning Phase Research Phase Implementation = 22% 13% 7% 2% 5. Approved for implementation 4% 3. Developing methodology 13% 1. Identify new sources New vehicles **Medical services** 2. Collecting data 2% Purchased data (JD Power), full item Purchased data, insurance payments to **Apparel** coverage, targeted deployment 2022 physician's services, hospital services Web-scraping, one footwear retailer Airline Wireless phone service **General merchandise** Corporate data, one airline Purchased web-scraped data, offer Web-scraping, many item categories prices for new plans Food away from home Residential telecommunication service Corporate data, one fast food company 6. In development 3% Purchased web-scraped data, offer **Motor fuel** prices for new plans Corporate data, full item coverage, Airfare targeted deployment June 2021 Web-scrape aggregator site, near full item coverage **Vehicle leasing** 7. Parallel testing 0% Purchased data (JD Power), near full item coverage 8. In production 3% **Hotels** Web-scrape aggregator site, near full **Used cars** Purchased data, longtime source item coverage

HUD administrative data, government

subsidized rental properties

4. Seeking approval 0%

Housing

Source: Fleck, Murphy and Paplomatas (2021)

Corporate data, March 2018

Publicly available data, longtime source

CorpX

CorpY

Postage

Corporate data, March 2019

BEA: Using big data to improve early GDP estimates

- GDP "advance" estimates released one month after end of quarter
 - Data from Census Quarterly Services Survey (QSS) not yet available
 - Estimated services spending for that release extrapolated from past data
- BEA researchers explored methods for "nowcasting" QSS estimates (Chen et al. forthcoming)
 - Tested models based on different algorithms and different predictor variables

$$y_{it} = f_m[g_k(X_t, Y_{i,t-p})]$$

- y_{it} quarterly growth in industry i, f_m algorithm, and g_k variable selection operator
- Candidate X's include traditional data (employment, prices) and nontraditional data (credit card transactions, Google search queries)
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BEA: Using big data to improve early GDP estimates (continued)

- Goal: Reduce revisions between advance and 3rd estimates
 - Greatest reductions: Ensemble models using employment, credit card data
 - Significant improvements in predictions for a number of sectors
- Nowcasting has been incorporated into GDP production process
 - Most often used for health care services and software investment



Census Bureau: Using big data to produce state-level retail sales estimates

- Monthly retail sales data collected from a survey sample of ~13,000 retail and food services businesses
 - Data collected at company level; no geographic component to design
- Census has explored use of point-of-sale data from 3rd party vendor NPD to reduce respondent burden and improve national estimates (Hutchinson forthcoming)
- NPD data key input to new experimental monthly state-level estimates
 - Estimates for total retail excluding non-store retailers, 11 specific sectors
 - Top-down estimates: National sales allocated based on share of industry's annual payroll in each state
 - Bottom-up estimates: Sum of sales for pre-selected multi-unit businesses from NPD, survey reporters operating in a single state, and imputed values for other retailers
 - Composite estimates: Weighted sum of two estimates; weights based on relative variances
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Census Bureau: Using big data to produce data on residential construction

- Building Permits Survey (BPS) tracks residential building permits
 - Monthly data for states and metropolitan areas
- Will be replaced with 3rd party data (~70% of single family units) and online permit information (Studds and Abriatis 2021)
 - Beginning January 2022, monthly census of all jurisdictions
 - Studying possibility of weekly estimates
 - Studying possibility of estimates for smaller geographic areas (e.g., zip codes)



Census Bureau: Using big data to produce data on residential construction (continued)

- Survey of Construction tracks residential building starts, completions, sales and unit characteristics
 - Monthly data on housing starts, completions and sales for Census regions
 - Annual data on housing characteristics for Census regions
- Work in progress on using satellite images to measure building starts, completions, and selected unit characteristics



Image Categorization

Pre-constructions

90 days before permit authorization date or earlier







Ground untouched and no major delimitations or excavations.

Construction Starts

Between 30 days after permit authorization date and 125 days after it.







Visible excavation or foundation

Construction Completions

270 days after permit authorization date.







Completed roof covering the area where there was a foundation or excavation previously.

Shape your future



Source: Studds and Abriatis (2021)

Census Bureau: Using big data to produce data on residential construction (continued)

- Survey of Construction tracks residential building starts, completions, sales and unit characteristics
 - Monthly data on starts, completions and sales, annual data on unit characteristics
 - Published for Census regions
- Work in progress on using satellite images to identify building starts completions, and selected unit characteristics
 - Field collection of supplemental information not obtainable via satellite will continue
 - Reduced collection costs will allow for larger sample, additional geographic and type of construction detail
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Examples cover a spectrum of uses for big data in production of official statistics

- Substituting for selected survey observations
- Improving early estimates
- Producing more disaggregated estimates
- Replacing survey data entirely!



Pandemic accelerated growth in interest in nontraditional data

Intense demand for real-time data

Intense demand for local area data

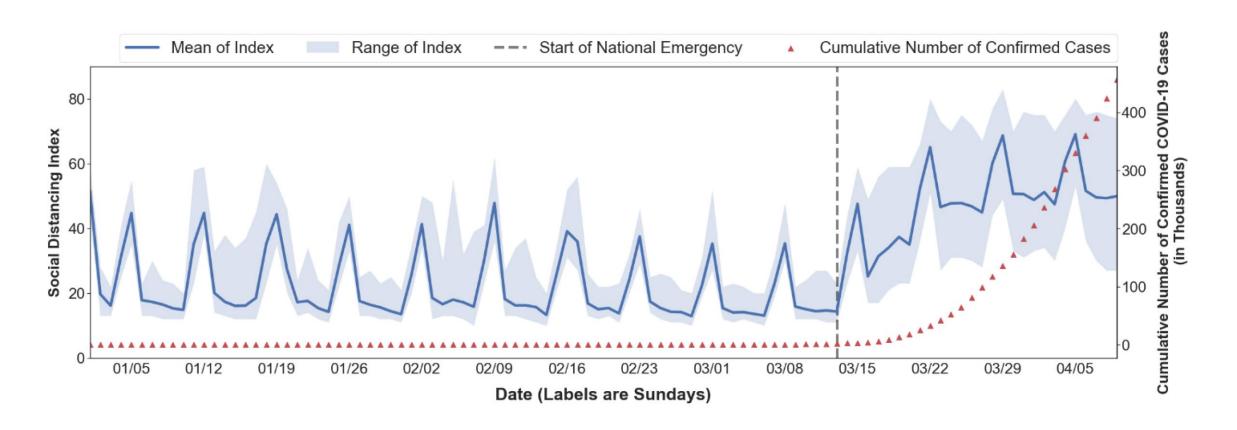


Tsunami of U.S. academic research on the impact of the crisis!

- To consider a few examples:
 - Mobile phone data showed early increase in social distancing, but it was far from uniform (Pan et al. 2020)



Evolution of social distancing index over time



Source: Pan et al. (2020)

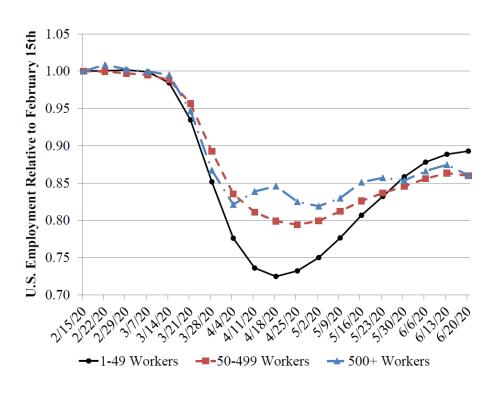
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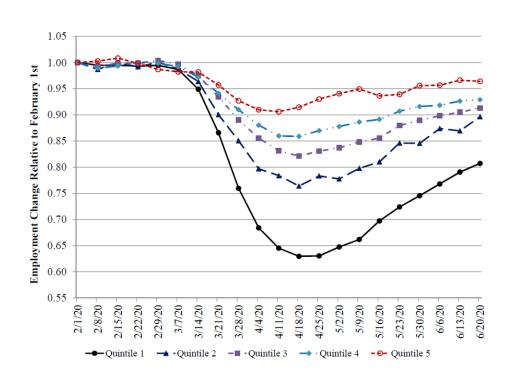


Pandemic impacts on employment

By firm size



By wage quintile



Source: Cajner et al. (2020)

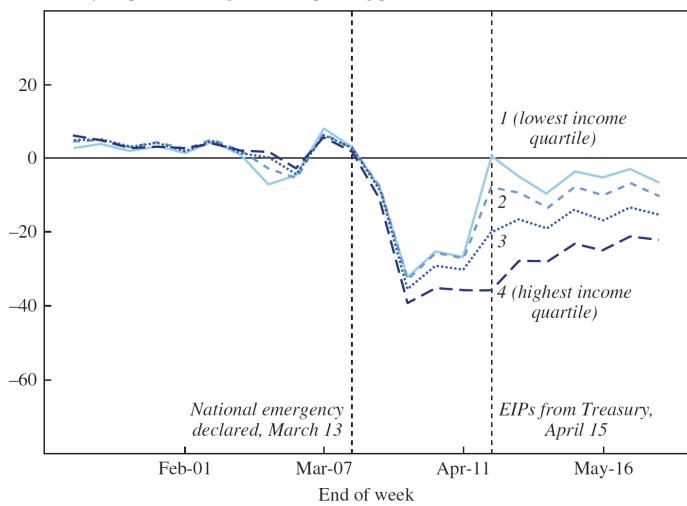
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Pandemic impacts on household spending

Year-over-year percent change in total spending per household



Source: Cox et al. (2020)

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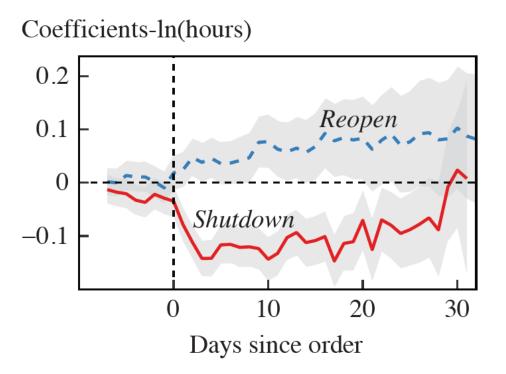
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 - Data from Homebase, a small business scheduling software company, showed modest shutdown order effects on hours (Bartik et al. 2020)

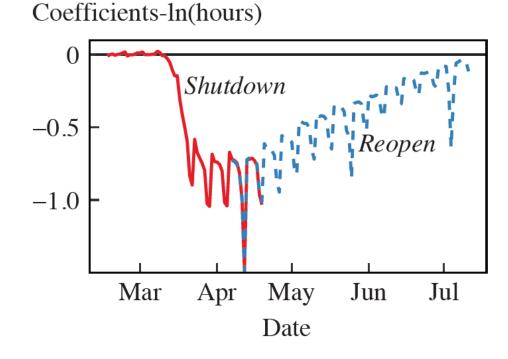


Shutdown order and common time effects on hours early in pandemic

Event study estimates of shutdown order effects on hours

Calendar time effects on hours





Source: Bartik et al. (2020)

The Rich Cut Their Spending. That Has Hurt All the Workers Who Count on It.

June 17, 2020



The Dark Side Of The Recovery Revealed In Big Data

October 27, 2020 · 6:31 AM ET

The Washington Post The recession is over for the rich, but the working class is far from recovered August 13, 2020 at 5:55 p.m. EDT

Smartphone data shows out-of-state Visitors flocked to Georgia as restaurants and other businesses . The Mashington Post

May 7, 2020 at 6:00 a.m. EDT reopened

U.S. statistical agencies responded too... though largely in traditional ways

- U.S. Census Bureau launched new Household Pulse and Small Business Pulse surveys in April 2020
- BLS added questions to Current Population Survey in May 2020 and fielded new Business Response Survey in July-September 2020
- Since June 2020, Bureau of Economic Analysis has used credit card transaction data to produce weekly estimates of retail spending (exclusive of non-store retailers)



Some national statistical offices have done more with information from nontraditional data

- Statistics Netherlands created Center for Big Data Statistics in 2016
 - Focus on satellite data, social media data and sensor data
- ONS Data Science Campus established in 2017
 - Faster Indicators program seeks to use real-time big data to provide more timely and more granular economic insights
 - Projects undertaken during pandemic have included
 - Using data from Google Community Mobility reports to produce usable information on mobility patterns (e.g., travel to work, travel to retail establishments)
 - Using Barclaycard data to produce near-real-time information on consumer spending
 - Using text extracted from business websites to learn about how they are responding to the pandemic



What is the role of a national statistics office?

- Traditional view: Produce portfolio of high-quality official statistics with well-documented properties published on a regular schedule
 - Private big data incorporated into existing structure in cases where it can be shown to be preferable on grounds of respondent burden, quality or cost
- Expanded vision: Serve as a source of credible information that can best inform important current national, state and local policy decisions
 - Larger role for national statistics office in using private naturally occurring data to shed light on questions official statistics cannot answer



What is the role of a national statistics office? (cont'd)

- Agencies that do not adapt to demand for more timely and more granular information risk being perceived as less relevant
 - Statistical offices no longer have a monopoly on data provision
- Re-invention will pose challenges
 - Appropriate resources
 - Leadership and buy-in
 - In the United States, perhaps a different agency structure
- Viewed in a positive light, it's an exciting time to be an economic statistician!



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References

Bartik, Alexander W., Marianne Bertrand, Feng Lin, Jesse Rothstein, and Matthew Unrath. 2020. "Measuring the Labor Market at the Onset of the COVID-19 Crisis," *Brookings Papers on Economic Activity*, June, 239-268.

Cajner, Tomaz, Leland Crane, Ryan Decker, John Grigsby, Adrian Hamins Puertolas, Erik Hurst, Christopher Kurz, and Ahu Yildirmaz. 2020. "The U.S. Labor Market During the Beginning of the Pandemic Recession," *Brookings Papers on Economic Activity*, June, 3-33.

Chen, Jeffrey C., Abe Dunn, Kyle Hood, Alex Driessen and Andrea Batch. Forthcoming. "Off to the Races: A Comparison of Machine Learning and Alternative Data for Predicting Economic Indicators," in Katharine G. Abraham, Ron Jarmin, Brian Moyer and Matthew Shapiro, eds., *Big Data for 21st Century Economic Statistics*, Chicago: University of Chicago Press.

Cox, Natalie, Peter Ganong, Pascal Noel, Joseph Vavra, Arlene Wong, Diana Farrell, Fiona Greig, and Erica Deadman. 2020. "Initial Impacts of the Pandemic on Consumer Behavior: Evidence from Linked Income, Spending, and Savings Data," *Brookings Papers on Economic Activity*, June, 35-69.

Hutchinson, Rebecca. J. Forthcoming. "Investigating Alternative Data Sources to Reduce Respondent Burden in United States Census Bureau Retail Economic Data Products," in Katharine G. Abraham, Ron Jarmin, Brian Moyer and Matthew Shapiro, eds., *Big Data for 21st Century Economic Statistics*, Chicago: University of Chicago Press.

References (continued)

Meyer, Bruce D., Wallace K. C. Mok, and James X. Sullivan. 2015. "Household Surveys in Crisis," *Journal of Economic Perspectives*, 29 (4): 199-226.

Pan, Yixuan, Aref Darzi, Aliakbar Kabiri, Guangchen Zhao, Weiyu Luo, Chenfeng Xiong, and Lei Zhang. 2020. "Quantifying Human Mobility Behavior Changes in Response to Non-Pharmaceutical Interventions During the COVID-19 outbreak in the United States," unpublished working paper.

Fleck, Susan, Bonnie Murphy, and Alexander Paplomatas. 2021. "Recent and Upcoming Advancements Using Alternative Collection Methods for BLS Price Programs," presentation to BLS Price Index Users Conference, April 21.

Studds, Stephanie L. and William Abriatis. 2021. "Construction Reengineering," presentation to Census Scientific Advisory Committee, March 19.